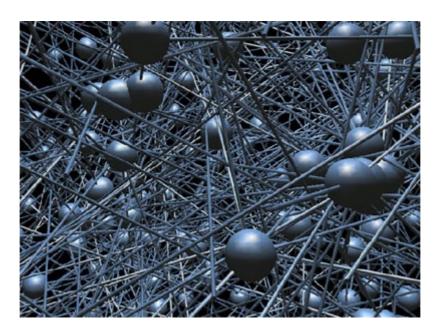
KB – Neural Data Mining with Python sources

© Roberto Bello (March 2013)

Introduction

The aim of this work is to present and describe in detail the algorithms to extract the knowledge hidden inside data using Python language, which allows us to read and easily understand the nature and the characteristics of the rules of the computing utilized, as opposed to what happens in commercial applications, which are available only in the form of running codes, which remain impossible to modify.

The algorithms of computing contained within the work, are minutely described, documented and available in the Python source format, and serve to extract the hidden knowledge within the data whether they are textual or numerical kinds. There are also various examples of usage, underlining the characteristics, method of execution and providing comments on the obtained results.



The KB application consists of three programs of computing:

- KB_CAT: for the extraction of knowledge from the data and the cataloging of records in homogeneous groups within them
- KB_STA: for the statistical analysis of the homogeneity of the groups between them and in the groups within them in order to identify the groups most significant and the most important variables that characterize each group
- KB_CLA: for the almost instantaneous classification of new records in catalogued groups before found by the program KB_CAT

The programs have been written in Python language using the most easily understood commands, instructions and functions, and those most similar to those of other languages (e.g. C, C++, PHP, Java, Ruby); however, the programs are full of comments and explanations.

The KB application to acquire hidden knowledge in data is the result of almost five years of study, programming and testing, also of other languages (Clipper, Fortran,

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 1 di 112

Ruby, C e C++).

The cost of the book is low considering the importance of the included algorithms of computing and the hard work in its programming and in the subsequent repeated and thorough testing of the input data of files containing thousands of records; the files used arrived from various sectors of interest (medicine, pharmacology, food industry, political polls, sport performances, market researches, etc.).

The author has chosen to use simple forms of communication, avoiding complicated mathematical formulas, observing that the important concepts can be expressed in an easy but not simplistic way.

Albert Einstein said: Do things in the easiest way possible, but without simplifying.

The author hopes to give a small contribution to encouraging young people to regain a love for maths, but above all hopes they will regain the desire to run programs on their computers, and therefore avoid using them just to consume their fingers surfing facebook or downloading music and films from Internet.

In his professional experience, firstly in computer fields and then as an industrial manager, he has repeatedly realized programs with mathematical contents, statistics and operations research which have considerably contributed to the economic results and good management of the companies that have seen him as a significant protagonist in important projects.

Business Intelligence and the bad use of statistics

Statistics are often wrong, or rather, the people who use them make mistakes. They make mistakes when they apply statistical aggregation instruments to pieces of information from sources coming from completely different objects or situations. First of all they cut, then mix and finally put them together. And to finish off they expect to pass judgement on this.

In this way the researcher in political trends break up the opinions of the people interviewed, mixing the single answers, joining them, crossing them and finally passing judgement with certainties that can only be attributed to *virtual people interviewed* that they have created, subjects that do not exist in real life and certainly are not traceable to individual people or to homogeneous groups of people who have been interviewed.

Similarly the *Business Intelligence* makes available the tools of data analysis that are able to cut the data and then reassembling them into multidimensional structures in which the peculiarities of information starting positions were destroyed. So Business Intelligence mixes companies from different sectors with turnovers not compatible, with very different sizes, belonging to different markets, etc., thereby abusing the will to change from time to time variables for data mining.

Which decisions on subjects (or situations) could be applied to, having destroyed the global informative world of the original subjects (or situations)?

To give an example, if we had a file of mammals where men and primates were included, we could obtain, as a result, that mammals, on average, have three legs. Where can I find a mammal that has an average of three legs?

To have real statistics we need to conserve, as much as possible, intact the informative property of the starting data of the subject or the situation .

Techniques derived from neural networks use an analysis approach to data which respect the informative properties of the starting data.

In fact they do not ask the user to define the variables to cross, and therefore do not allow to occur absurd crossed values.

Quite simply they require that the maximum number of groups that the algorithm has to create is inserted

The original informative contents are not destroyed, the subject's data are processed in relationship to the data of other subjects (or situations).

Retain all the information attributable to the subject and create the categories of membership of the subjects (or situations) in which the subjects (or situations) will be similar to each other.

Other techniques are able to point out what are the significant variables of aggregation and aggregate values which are important for each group created.

Also indicate what are the variables that are not influential in cataloging.

More sophisticated techniques can process any kind of data set highlighting if there is information in the file or if they contain only numbers or characters not related to each other by internal relations: the model must follow the data and not vice versa (JB Benzecri).

Learning by induction and the neural networks

Induction is a very important method of learning for living creatures.

One of the first philosophers to resort to this concept was Aristotle, who attributed the merit of having discovered it to Socrates, who maintained that induction was in fact, "the process of the particular that leads to the universal" (Top., I, 12, 105 a 11). Still according to Aristotle it is neither the senses through induction nor rationality through deduction that gives a guarantee of truth, but only intellectual intuition: this allows to collect the essence of reality, forming valid and universal principles, from which syllogistic reasoning will draw coherent conclusions with premises.

Learning, life and evolution are linked together.

In fact life is evolution and evolution is learning what is necessary for survival. Learning is the capacity to elaborate information with critical intelligence. Therefore, critical elaboration of information is life. (Roberto Bello).

A simple example can illustrate how one learns by induction.

Let's imagine a person who had never seen containers such as glasses, bottles, jars, cups, vases, boxes, flagons, jugs, chalices, tetra pack and so on.

Without saying anything I will show him real examples of objects that belong to the above mentioned categories.

The person can look at, smell, touch and weigh the objects shown to him.

After having examined a sufficient number of objects the person will easily be able to put the objects into categories containing the objects which on the whole are similar to each other, favouring some characteristics rather than others which are not considered relevant.

When the learning has taken place, I could show another object in the shape of a glass, which is of a different colour, made of a different material and of a different weight, still obtaining the cataloging of the object in the category of glasses.

With the help of induction, the person in training could make two categories of glasses: one with handles (beer mugs) and the other without handles.

Learning has allowed the person to recognize the distinctive aspects of the object to go from the specific to the universal ignoring the non relevant aspects.

The algorithms based on the neural networks, and in particular referring to the map of Kohonen (SOM Self Organizing Map), are based on the principals which have just been illustrated in this example.

Such a model of neural networks demonstrates in an important way the biological

mechanisms of the central nervous system; many studies have demonstrated that precise zones exist on the surface of the cranial cortex, each of which respond to a precise sensory or muscular function.

Each neuron specializes in responding to precise stimulus through a continual interaction with the neighbouring neurons.

We have zones reserved for hearing, sight, muscular activity etc., and the spacial demarcation between the different groups is so clear that we talk of the formation of bubbles of activity.

The neural networks model presented by Kohonen imitates the behaviour described above.

The architecture is quite simple; the network is formed by a rectangular grate, also known as Kohonen's layer, made up of neurons from the output level, each one occupying a precise position and connected to all the entry units.

The weight of the connections between the input and output levels are kept up to date thanks to the process of learning, where the connections between the neurons of the output level have weights which produce excitement among the surrounding neurons and inhibition in distant ones.

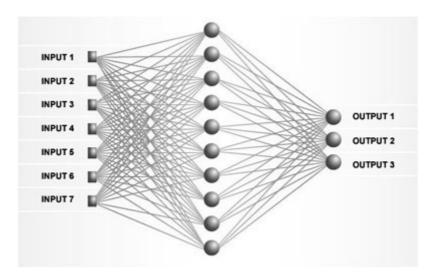


Diagram of a neural network

The SOM networks are applied to many practical problems; they are able to discover important properties autonomously in input data and therefore they are especially useful in the process of Data Mining, above all for problems of cataloging.

The algorithms of learning of the Kohonen network begin from the start-up phase of the synapse weights, which must have casual values in space (0.0 - 0.99999) and be different for each neuron.

Subsequently the weights are presented to the network as input values and the algorithm allows the network to self-organize and correct the weights after each data input, until a state of equilibrium is reached. Kohonen's network is also known as a competitive network since it is based on the principle of competition between neurons to win and to remain active; only the weight of the active units are updated. The winning unit i* is that which possesses the potential for major activation; the more a unit is active for a certain pattern of input data, the more the vector of the

synapse weight is similar inside the pattern.

On the basis of this idea it is possible to find the winning unit by calculating the euclidean distance between the input vector and the relevant vector of synapse weight. At this point is selected the neuron i* that corresponds to the minimum distance.

Once the winning neuron has been determined, is carried out an automatic learning of the weight of the neuron itself and of those which are part of its neighbourhood, based on a rule of *hebbian* type.

In particular, a formula of modification of the weights which derives from the original rule of Hebb is used; considering that this would increase the weight to infinity, so is introduced a factor of forgetfulness, pushing the weights toward the input vectors to which the unit responds more.

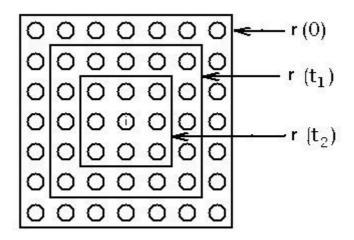
In this way a relative map of the characteristics of input is created where the neighbouring units respond to precise stimulus of admission thanks to the similarity of the synapse weights.

For this aim it is also necessary to introduce the concept of the function of proximity, that determines the area of size r around i^* where the units are active.

The less the dimension of the proximity, the lower the number of the units of the layer of Kohonen whose weights are modified significantly, therefore the higher the capacity of the neurons to differentiate and to acquire details but also to increase the complexity of the system of learning.

According to Kohonen the size of the function of proximity must be varied, initially choosing it to cover all of the units of the layer and gradually reducing it.

In this way you will go from learning the main features up to learning the details of the specialized areas in responding to particular stimuli.



Representation of the gradual reduction of proximity

Once the learning phase has been completed the network is able to supply answers in relation to the new input presented. The property of generalization derives from the fact that even the neurons near to those selected are modified.

The network must therefore self-organize in areas that are composed of a large set of values around the input from which the algorithm learns, this will ensure that if there is an input never seen before, but with similar characteristics, the network will be able to classify properly.

Besides this, compared to supervised algorithms, the self-organized processes of learning (SOM) result to be efficient even if are used incomplete or incorrect input data, a characteristic that makes this neural network particularly suitable to be used in the process of *Data Mining*.

In fact Kohonen algorithms, at the end of the phase of non supervised training, produces a three-dimensional matrix that can be used to classify new records in groups with the most similar characteristics.

While the training phase can require a lot of time to run, that of classifying new records in the groups with the most similarities is almost instantaneous, making this function especially useful for processes with real time reactions (e.g. quality control, productions in a continuous cycle, automation in industrial processes, control systems, monitoring the messages on the Net, etc.).

The algorithms of the neural networks have, as a common aspect, the inability to explain the characteristics of the groups obtained.

It is possible, using the information contained in the training matrix and resorting to other technical statistics, to provide information on the characteristics of every group helping the researcher to deepen the analysis of the results to gather better results of their research.

It is also possible to determine if the overall view of the records used in the training phase has knowledge contents or, on the contrary, it is made up of data which have little connection between them and therefore not suitable for the use of research: in fact it is possible to compute the global index of homogeneity of the groups on the whole (*Knowledge Index*), informing the researcher of the suitability of the output files to achieve the expected goals.

KB

Python for KB

The Python program language is a language that can be freely downloaded from Internet.

Python is compatible with Windows, Linux/Unix, Mac OS X, OS/2, Amiga and Smart-phones /Tablets.

Python is distributed on license Open-Source: its use is free of charge also for commercial products.

The site from where the Python language can be downloaded is www.python.org/download, choosing the compatible version for your computer.

Installing Python in Windows involves choosing the extended file *msi* to download from Internet.

To install Python in Linux (and in particular in Linux Ubuntu) use the *Software Manager* of the Linux distribution, which automatically connects to the official site (*repository*), downloading what is necessary for a safe, complete and automatic installation; Linux distributions usually already contain the Python language pre-installed.

Whatever the operating system for the installation of Python may be, the programs can only be used in command mode option by opening the file containing the Python program (for example: program.py), typing python program.py:

- in a DOS window (with execute) in Windows
- in a terminal window in Linux

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 6 di 112

Details

The KB application is a system which extracts knowledge from data based on algorithms of the map of Kohonen revised and modified by the author.

KB can elaborate any table of numeric data and/or text, tables where the first line of the table is destined to the description of the columns / variables and the first column of the table is destined to the codes (arbitrary) of identification of the record / case.

In KB, functions are included with the aim of:

- normalizing the numeric data comparing it to the standard deviation or to the maximum value of the variable / column according to the user's choice
- transforming the alphanumeric data into numeric data conveniently returned equidistant between them
- inserting statistical functions able to judge the quality of the results of the cataloging for each group and globally
- writing different output files:
 - records / cases arranged by group code according to the chosen category
 - synthetic statistical information on the characteristics of the different groups also in relation to statistical indexes with reference to entire populations of records / cases
 - the final training matrix having the minimum error

The neural networks have the known defect of being *black boxes* in that they are able to catalog but don't *explain*:

- what the characteristics of each group are
- what the columns/variables are important in each group for the cataloging
- · what the most homogeneous groups are on the inside
- if, in its global sense, the input table contains information in relation between the variables or if the table is purely a group of numbers and letters without any inductive value.

Appendixes contain the programs written in Python KB_CAT (for the cataloging), KB_STA (for the analysis of the results) and KB_CLA (for the classification).

They have to be converted into file in text form using *cut and paste*; the programs have to be stored with names:

- kb cat.py
- kb_sta.py
- kb_cla.py

The name of the programs can also be different from kb_cat, kb_sta, kb_cla, as long as the *extension* is ".py" to allow the Python language to recognize the programs and run them.

Some of the test files are also reproduced and the results obtained are shown in the DOS window (Windows) or in the Terminal Window (Linux), results are contained in files in text format.

Collecting and arranging input data

The use of programs based on the algorithms of Kohonen require data to be KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 7 di 112

prepared and normalized.

To begin with it is important to carefully choose the data to be analysed.

Information from which the user intends to extract knowledge must be contained in tables that have the following characteristics:

- the format must be text (txt, csv)
- the fields must be separated by tabulation (tab)
- the first column is destined to identify each line with the identification code of every record (e.g. Client's code, product name, production lot, etc.)
- the first line must contain the descriptions of the columns separated by the tabulation (tab)
- values are contained in the cells from the second column to the last column and from the second line to the last line
- all the values of all the columns and all the lines must be separated by tabulation (tab)
- empty fields or those not containing anything cannot exist
- a column which contain numerical data cannot contain data with text

To convert tables into text you can resort to programs xls (Excel) or OpenOfficeCalc (ods) which are able to read the input formats and convert them into (csv) format, choosing the tabulation field (tab) and space (empty) to delimit the text.

For the quality of the results, the famous saying *garbage in, garbage out* is always valid; it is fundamental to collect good quality data that allows the research to be described and explained in the most complete way possible.

You also need to decide what size of the data is to be used as an input file (see the following suggestions).

The neural networks give the same weight to all of the variables inserted; if a variable oscillates in an interval (1000 - 10000) and the other in an interval (0 - 1), the variations of the first tend to reduce the importance of the second, even if the latter could be more significant in determining the results of the classification.

To do this transformation techniques exist which make the variables compatible among them, making them fall inside a certain interval (*range*).

The KB_CAT program can apply different techniques of normalization of the numeric values and text data.

Numeric values can be normalized through two methods:

- Normalization with the maximum: the new values of the column are obtained dividing the original values for the maximum value of the column, in this way the new values vary between zero and 1
- Standardization: the new values are obtained subtracting from the original value the mean of the column and dividing the result of the difference for the standard deviation of the column.

From the columns containing strings of characters are extracted the value of those containing different strings, they are sorted, counted and then are used to determine the attribution step of a numeric value between 0 and 1.

The KB_CAT program does not foresee the automatic transformation of the date or the time. The date must be transformed by the user in pseudo continue numeric variables assigning the value 0 to the most remote date and increasing by a unit every subsequent date, or expressing the 365 days of the year in thousandths, according to the formula: 0,9999* days a year/365.

The year could also be indicated using another variable. The pair of variables

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 8 di 112

should preferably be expressed as a ratio, through a single value which will offer information which is clearer and more immediate; in this way the derived variables can be calculated starting from the input variables.

Let us imagine that two variables are present: the weight and height of a person.

Considered separately they have little meaning, it would be better to obtain the coefficient of the body mass which is definitely a good synthetic index of obesity (body weight expressed in kilogrammes divided by height in meters squared).

Another important step in preliminary elaboration of data is to try and simplify the problem you want to resolve.

To do this it may be useful to reorganize the space of the input values, space which grows essentially as grows the size of data.

A technique to reduce the number of variables and improve the ability of learning of the neural networks, which is often used, is the *principal component analysis*, that try to identify a sub-space *m* size which is the most significant possible in respect to the input space *n* size.

The m final variables are called *principal components* and are linear combinations of n initial variables.

Other methods used to reduce the size of the problem to resolve are the elimination of the variables which are strongly linked between them and not useful to achieve the desired result.

In the first case it is important to consider that the connection does not always imply a cause/effect relationship, therefore eliminating some variables must be done with extreme care by the user.

It is very common to reorganize file of input data that needs to be cataloged by examining the results of the first processing runs which often indicate that certain variables/columns are worthless: their elimination in subsequent processing often contribute to improve the cataloging having put an end to the noise of the useless variables/columns.

In the processing of data relating to clinical trials, it was verified that the personal data of gender, nationality, residence, education, etc., not giving in those cases no contribution to cataloging, could be omitted improving the quality of new learning.

A very important aspect to consider is related to the number of records contained in an input file to catalog.

Often better results are obtained with smaller files which are able to generalize better and produce training matrices which are more predictive.

On the contrary, a file containing a large number of records could produce an invalid training of *overfitting* causing a photographic effect which can only classify new records which are almost identical to those used in the phase of cataloging.

As scientists at Cern have already discovered, it's more important to properly analyse the fraction of the data that is important ("of interest") than to process all the data. TomHCAnderson

In statistics we talk about overfitting (excessive adaptation) when a statistics model fits the observed data (the sample) using an excessive number of parameters.

An absurd and wrong model converges perfectly if it is complex enough to adapt to the quantity of data available.

It is impossible to prove at first glance the best number of records to be contained in a file to catalog: too much depends on the number of variables and the informative contents of the variables for all of the records present in the file.

The best suggestion is to carry out distinct runs with the original file and with other

files obtained with a lesser number of records

To obtain a smaller sized file you can extract records from the original file by random choice, you can use the small program KB_RND which is present in appendix 4.

Indicate in **InputFile** the file from which you want to extract the smaller sized output file (**OutputFile**).

Indicate in **Out number of cells** the number of cells (lines x columns) of the output file.

Other times it is convenient to remove from the initial input file, the records which clearly contain values contradictory, absurd or missing: in so doing you reduce the size of the file and improve the quality by reducing the noise.

General warnings for using the KB programs

It is important that the files that are in input and output, while processing the three programs kb_cat, kb_sta, kb_cla are not open in other *windows* for reading or writing: if this happens kb_cat, kb_sta, kb_cla would go into error.

Processing the three programs can be interrupted by pressing ctrl and the c keys.

KB_CAT Knowledge Data Mining and cataloging into homogeneous groups

Generality, aims and functions

KB_CAT is the first of the three programs to use and it is the most important. Its purpose is to analyse any kind of textual file structured in two-dimensional table containing numeric values and/or text data.

KB_CAT:

- controls that the table to process does not contain errors of format
- · normalizes the numeric values and the text data
- starts the training phase searching for the minimum error which decreases during the processing until it reaches the minimum value of the alpha chosen by the user.

Once the processing has been completed, the program will write the output file containing the results which can also be used by the other two programs KB_STA and KB_CLA.

Source of KB_CAT (see attachment 1)

Test Input file (copy and paste then save with name *vessels.txt*); fields separated by tabulation

Description	Shape	material	height	colour	weight	haft	plug
glass_1	cut_cone	pewter	10	pewter	20	no	no
glass_2	cut_cone	plastic	9	white	4	no	no
glass_3	cut_cone	terracotta	8	grey	20	no	no
beer_jug	cut_cone	porcelain	18	severals	25	no	no
dessert_glass	cut_cone	glass	17	transparent	17	no	no
wine_glass	cut_cone	glass	15	transparent	15	no	no
jug	cylinder	terracotta	25	white	40	yes	no
bottle_1	cylinder_cone	glass	40	green	120	no	cork
bottle_2	cylinder_cone	glass	40	transparent	125	no	cork
bottle_3	cylinder_cone	glass	45	opaque	125	no	plastic
bottle_4	cylinder_cone	glass	35	green	125	no	metal
magnum_bottle	cylinder_cone	glass	50	green	170	no	metal
carboy	ball_cone	glass	80	green	15000	no	cork
ancient_bottle	ball_cone	glass	40	green	150	no	cork
champagne_glass	cut_cone	crystal	17	transparent	17	no	no
cup_1	cut_cone	ceramic	10	white	30	yes	no
milk_cup	cut_cone	terracotta	15	blue	35	yes	no
tea_cup	cut_cone	terracotta	7	white	30	yes	no
cup_2	cut_cone	glass	20	transparent	35	yes	no
coffee_cup	cut_cone	ceramic	6	white	20	yes	no
tetrapack1	parallelepiped	mixed	40	severals	20	no	plastic
tetrapack2	parallelepiped	plastic	40	severals	21	no	plastic
tetrapack3	parallelepiped	millboard	40	severals	22	no	no
cleaning_1	parall_cone	plastic	30	white	50	yes	plastic
cleaning_2	cylinder_cone	plastic	30	blue	60	yes	plastic

Description	Shape	material	height	colour	weight	haft	plug
tuna_can	cylinder	metal	10	severals	10	no	no
tuna_tube	cylinder	plastic	15	severals	7	no	plastic
perfume	parallelepiped	glass	7	transparent	15	no	plastic
cleaning_3	Cone	plastic	100	severals	110	yes	plastic
visage_cream	cylinder	metal	15	white	7	no	no
cd	parallelepiped	plastic	1	transparent	4	no	no
trousse	cylinder	plastic	1	silver	7	no	yes
watering_can	Irregular	plastic	50	green	400	yes	no
umbrella_stand	cylinder	metal	100	grey	3000	no	no
pot_1	cylinder	metal	40	grey	500	two	yes
pot_2	cut_cone	metal	7	grey	200	yes	yes
toothpaste	cylinder	plastic	15	severals	7	no	plastic
pyrex	parallelepiped	glass	10	transparent	300	two	glass
plant_pot	cut_cone	terracotta	30	brown	200	no	no
pasta_case	parallelepiped	glass	35	transparent	150	no	metal

How to run

Being positioned in the file containing kb_cat.py and the input file to process, start KB_CAT typing in the commands window of DOS Windows (or in the Terminal of Linux), the command:

python kb_cat.py

python runs the program (with python language) kb_cat.py.

The program will start subsequently asking

Input File = vessels.txt

vessels.txt is the file in format txt containing the table of the records / cases to catalog, shown above.

If you want to give more importance to one particular variable/column, all you have to do is to duplicate the value, one or more times, in additional variables/columns: if you want to make the variable important for three times its original weight, create another two variables/columns calling them for example *shape1* and *shape2* with values which are identical to the original variable.

Number of Groups (3-20) = 3

The value 3 is the square root of the maximum number of groups to catalog (in this KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 12 di 112

case 9); since the training matrix has a cube form base; the maximum number of training groups can only be the square of the value that has been entered.

There are no useful rules for fixing the best number of the parameter Number of Groups: it is advisable to initially try with low values and gradually carry out other processing with higher values if are obtained groups containing too many records, and on the other hand, reduce the value of the parameter Number of Groups if are obtained groups containing few records.

Sometimes though, the researcher is interested in analysing groups with few numbers of records but with rare and singular characteristics: in this case groups containing few records are welcome.

Normalization (Max, Std, None) = m

The value m (M) indicates the request to normalize numerical data dividing them by the maximum value of the column / variable.

The value s (S) indicates the request to normalize numerical data subtracting from each input value the average of the variable / column and dividing the result by the *standard deviation* of the variable / column.

It is not advisable to insert the value N (None) above all in the presence of variables which are very different among them with a large difference between the minimum and maximum value (range).

Start Value of alpha (from 1.8 to 0.9) = 0.9

KB_CAT, like all algorithms of the neural networks, runs cycles making loops which consider all the input.

In these loops the alpha parameter plays an important role from its initial value (Start Value) to its final value (End Value) also considering the value of the decreasing step.

Occasionally an excessive length of time for the processing can be noted having chosen a large number of groups for a file containing a lot of records and with distant start and end values of alpha and with a very small decreasing step of alpha; usually in these cases you will notice the minimum error remains the same in many loops.

It is advisable to stop the processing, by pressing the two keys *ctrl* e *c*, together and repeat it using more suitable parameter values.

End Value of alpha (from 0.5 to 0.0001) = 0.001

The *alpha* parameter used by the KB_CAT to refine the cataloging of records into different groups: a low *alpha* value involves a longer cycle time of the computing with the possibility of obtain a lower final minimum error but also a hypothetical greater chance of *over fitting (photo effect)*.

Decreasing step of alpha (from 0.1 to 0.001) = 0.001

Choose the value of the step of decreasing alpha to be applied to each loop.

Forced shut down of processing

In the case of wanting to shut down the processing while it is running, you just

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 13 di 112

need to press the two keys *ctrl* and *c* at the same time. Obviously the files that were in writing will not be valid.

KB_CAT produce the following output:

In the window DOS Windows (or the Terminal Linux)

# KB_CAT KN	NOWLEDGE DISCO	VERY	IN DATA MI	N]	ING (CATA	LOG PROGRAM) #	
# by ROBERT	O BELLO (COPY	RIGHT	r MARCH 201	1	ALL RIGH	IS RESERVED) #	
# Language	used: PYTHON						#	
#########	+ # # # # # # # # # # # #	####	##########	# #	#######	#########	##################	
InputFile				:	vessels.	txt		
Number of 0	Groups (3 - 20)		:	3			
Normalizati	ion(Max, Std,	None)	:	m			
Start value	e of alpha (1.	8 - 0	0.9)	:	0.9			
End value o	of alpha (0.5	- 0.0	0001)	:	0.001			
Decreasing	step of alpha	(0.1	1 - 0.001)	:	0.001			
Record 40 Columns 7								
**** Epoch	15 WITH	MIN	ERROR		3.616	alpha	0.88650	
**** Epoch	39 WITH	MIN	ERROR		3.612	alpha	0.86490	
**** Epoch	41 WITH	MIN	ERROR		3.608	alpha	0.86310	
**** Epoch	44 WITH	MIN	ERROR		3.460	alpha	0.86040	
**** Epoch	46 WITH	MIN	ERROR		3.456	alpha	0.85860	
**** Epoch	48 WITH	MIN	ERROR		3.451	alpha	0.85680	
**** Epoch	50 WITH	MIN	ERROR		3.447	alpha	0.85500	
**** Epoch	52 WITH	MIN	ERROR		3.443	alpha	0.85320	
**** Epoch	54 WITH	MIN	ERROR		3.439	alpha	0.85140	
**** Epoch	56 WITH	MIN	ERROR		3.435	alpha	0.84960	
**** Epoch	58 WITH	MIN	ERROR		3.431	alpha	0.84780	
**** Epoch	60 WITH	MIN	ERROR		3.426	alpha	0.84600	
**** Epoch	62 WITH	MIN	ERROR		3.422	alpha	0.84420	
**** Epoch	64 WITH	MIN	ERROR		3.418	alpha	0.84240	
**** Epoch	66 WITH	MIN	ERROR		3.414	alpha	0.84060	
**** Epoch	68 WITH	MIN	ERROR		3.410	alpha	0.83880	
**** Epoch	70 WITH	MIN	ERROR		3.371	alpha	0.83700	
**** Epoch	72 WITH	MIN	ERROR		3.366	alpha	0.83520	
**** Epoch	74 WITH	MIN	ERROR		3.362	alpha	0.83340	
**** Epoch	76 WITH	MIN	ERROR		3.358	alpha	0.83160	
**** Epoch	78 WITH	MIN	ERROR		3.353	alpha	0.82980	
**** Epoch	WITH 08	MIN	ERROR		3.349	alpha	0.82800	
**** Epoch	82 WITH	MIN	ERROR		3.345	alpha	0.82620	
**** Epoch	84 WITH	MIN	ERROR		3.341	alpha	0.82440	
**** Epoch	86 WITH	MIN	ERROR		3.336	alpha	0.82260	

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 14 di 112

**** Epoc	ch 88	WITH MIN ERROR	3.332 alpha	0.82080
**** Epoc	ch 90	WITH MIN ERROR	3.328 alpha	0.81900
**** Epoc	ch 92	WITH MIN ERROR	3.324 alpha	0.81720
**** Epoc	ch 94	WITH MIN ERROR	3.320 alpha	0.81540
**** Epoc	ch 96	WITH MIN ERROR	3.316 alpha	0.81360
**** Epoc	ch 98	WITH MIN ERROR	3.311 alpha	0.81180
**** Epoc	ch 102	WITH MIN ERROR	3.229 alpha	0.80820
**** Epoc	ch 107	WITH MIN ERROR	3.229 alpha	0.80370
**** Epoc	ch 109	WITH MIN ERROR	3.225 alpha	0.80190
**** Epoc	ch 111	WITH MIN ERROR	3.222 alpha	0.80010
**** Epoc	ch 113	WITH MIN ERROR	3.218 alpha	0.79830
Epoch 125	min err	3.21823	min epoch 113 alpha	0.78840
**** Epoc	ch 126	WITH MIN ERROR	3.218 alpha	0.78660
**** Epoc	ch 128	WITH MIN ERROR	3.214 alpha	0.78480
**** Epoc	ch 130	WITH MIN ERROR	3.211 alpha	0.78300
**** Epoc	ch 133	WITH MIN ERROR	3.206 alpha	0.78030
**** Epoc	ch 136	WITH MIN ERROR	3.201 alpha	0.77760
**** Epoc	ch 139	WITH MIN ERROR	3.196 alpha	0.77490
**** Epoc	ch 142	WITH MIN ERROR	3.191 alpha	0.77220
**** Epoc	ch 146	WITH MIN ERROR	3.065 alpha	0.76860
**** Epoc	ch 149	WITH MIN ERROR	3.060 alpha	0.76590
**** Epoc	ch 165	WITH MIN ERROR	3.024 alpha	0.75150
**** Epoc	ch 167	WITH MIN ERROR	3.008 alpha	0.74970
**** Epoc	ch 169	WITH MIN ERROR	3.004 alpha	0.74790
**** Epoc	ch 171	WITH MIN ERROR	3.000 alpha	0.74610
**** Epoc	ch 173	WITH MIN ERROR	2.996 alpha	0.74430
**** Epoc	ch 175	WITH MIN ERROR	2.993 alpha	0.74250
**** Epoc		WITH MIN ERROR	2.989 alpha	0.74070
**** Epoc	ch 179	WITH MIN ERROR	2.985 alpha	0.73890
**** Epoc	ch 181	WITH MIN ERROR	2.982 alpha	0.73710
**** Epoc		WITH MIN ERROR	2.978 alpha	0.73530
**** Epoc		WITH MIN ERROR	2.974 alpha	0.73350
**** Epoc		WITH MIN ERROR	2.971 alpha	0.73170
**** Epoc		WITH MIN ERROR	2.967 alpha	0.72990
**** Epoc		WITH MIN ERROR	2.964 alpha	0.72810
**** Epoc		WITH MIN ERROR	2.960 alpha	0.72630
**** Epoc		WITH MIN ERROR	2.957 alpha	0.72450
**** Epoc		WITH MIN ERROR	2.953 alpha	
**** Epoc		WITH MIN ERROR	2.950 alpha	
**** Epoc		WITH MIN ERROR	2.946 alpha	0.71910
**** Epoc		WITH MIN ERROR	2.943 alpha	0.71730
**** Epoc	ch 205	WITH MIN ERROR	2.940 alpha	0.71550

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 15 di 112

**** Epoch	207	WITH MIN ERROR	2.936	alpha	0.71370
**** Epoch	209	WITH MIN ERROR	2.933	alpha	0.71190
**** Epoch	211	WITH MIN ERROR	2.921	alpha	0.71010
**** Epoch	213	WITH MIN ERROR	2.918	alpha	0.70830
**** Epoch	215	WITH MIN ERROR	2.915	alpha	0.70650
**** Epoch	217	WITH MIN ERROR	2.912	alpha	0.70470
**** Epoch	219	WITH MIN ERROR	2.909	alpha	0.70290
**** Epoch	221	WITH MIN ERROR	2.906	alpha	0.70110
**** Epoch	223	WITH MIN ERROR	2.903	alpha	0.69930
**** Epoch	225	WITH MIN ERROR	2.863	alpha	0.69750
**** Epoch	227	WITH MIN ERROR	2.861	alpha	0.69570
**** Epoch	229	WITH MIN ERROR	2.858	alpha	0.69390
**** Epoch	231	WITH MIN ERROR	2.855	alpha	0.69210
**** Epoch	233	WITH MIN ERROR	2.852	alpha	0.69030
**** Epoch	235	WITH MIN ERROR	2.849	alpha	0.68850
**** Epoch	241	WITH MIN ERROR	2.843	alpha	0.68310
**** Epoch	243	WITH MIN ERROR	2.840	alpha	0.68130
Epoch 250	min err	2.83977	min epoch 243	alpha	0.67590
**** Epoch	281	WITH MIN ERROR	2.783	alpha	0.64710
**** Epoch	283	WITH MIN ERROR	2.780	alpha	0.64530
**** Epoch	285	WITH MIN ERROR	2.777	alpha	0.64350
**** Epoch	287	WITH MIN ERROR	2.774	alpha	0.64170
**** Epoch	289	WITH MIN ERROR	2.772	alpha	0.63990
**** Epoch	291	WITH MIN ERROR	2.769	alpha	0.63810
**** Epoch	293	WITH MIN ERROR	2.766	alpha	0.63630
**** Epoch	295	WITH MIN ERROR	2.764	alpha	0.63450
**** Epoch	297	WITH MIN ERROR	2.761	alpha	0.63270
**** Epoch	299	WITH MIN ERROR	2.758	alpha	0.63090
**** Epoch	301	WITH MIN ERROR	2.756	alpha	0.62910
**** Epoch	303	WITH MIN ERROR	2.753	alpha	0.62730
**** Epoch	305	WITH MIN ERROR	2.751	alpha	0.62550
**** Epoch	307	WITH MIN ERROR	2.748	alpha	0.62370
**** Epoch	309	WITH MIN ERROR	2.746	alpha	0.62190
**** Epoch	311	WITH MIN ERROR	2.687	alpha	0.62010
**** Epoch	320	WITH MIN ERROR	2.636	alpha	0.61200
**** Epoch	323	WITH MIN ERROR	2.632	alpha	0.60930
**** Epoch	326	WITH MIN ERROR	2.628	alpha	0.60660
Epoch 375	min err	2.62765	min epoch 326	alpha	0.56340
**** Epoch	485	WITH MIN ERROR	2.558	alpha	0.46350
Epoch 500	min err	2.55849	min epoch 485	alpha	0.45090
**** Epoch	539	WITH MIN ERROR	2.554	alpha	0.41490
**** Epoch	551	WITH MIN ERROR	2.394	alpha	0.40410

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 16 di 112

**** Epoch 621	1 WITH	MIN ERROR	2.362	alpha	0.34110
Epoch 625 m	in err	2.36245	min epoch 621	alpha	0.33840
**** Epoch 702	2 WITH	MIN ERROR	2.186	alpha	0.26820
**** Epoch 744	4 WITH	MIN ERROR	2.160	alpha	0.23040
Epoch 750 m	in err	2.15974	min epoch 744	alpha	0.22590
Epoch 875 m	in err	2.15974	min epoch 744	alpha	0.11340
**** Epoch 941	1 WITH	MIN ERROR	1.859	alpha	0.05310
Epoch 1000 r	min err	1.85912	min epoch 941	l alpha	0.00100

Min alpha reached

```
# KB CAT KNOWLEDGE DISCOVERY IN DATA MINING (CATALOG PROGRAM)
# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
# Language used: PYTHON
EPOCH 941 WITH MIN ERROR
                             1.859 starting alpha
                                                     0.90000
                                                             ending
        0.00100 Iterations 39960 Total Epochs 999
Output File Catalog.original vessels M g3 out.txt
Output File Catalog.sort
                         vessels M g3 outsrt.txt
Output File Summary sort vessels_M_g3_sort.txt
Output File Matrix Catal.
                         vessels M g3 catal.txt
Output File Means, STD, CV. vessels M g3 medsd.txt
Output File CV of the Groups vessels M g3 cv.txt
Output File Training Grid
                         vessels M g3 grid.txt
Output File Run Parameters
                         vessels M g3 log.txt
Elapsed time (seconds) :
                         15
*KIndex* 0.8438
```

As you can see, during the processing, the minimum error decreases from 3.616 (epoch 15) to 1.859 epoch 941).

The processing was completed at the epoch 1000, when the parameter value *alpha* reaches a minimum value of 0.001.

References to output files are also listed:

- Catalog.original = input file cataloged, NOT in order of groups and with original values (NOT normalized)
- Catalog.sort = input file cataloged, IN ORDER of groups and with original values (NOT normalized)
- Summary.sort = input file cataloged, IN ORDER of groups and with NORMALIZED values.
- Matrix Catal. = files with three columns (progressive number of records, group codes and subgroup codes)
- Means, STD, CV = files with a column for every variable and with three lines (mean, standard deviation and coefficient of variation)
- CV of the Groups = files of the coefficient of variations of the groups and
 KB Neural Data Mining with Python sources Roberto Bello Pag. 17 di 112

- of the variables / columns with totals of the records classified into groups
- Training Grid = files containing the values of the training matrix with minimum error
- Run Parameters = files containing references to input files, parameters of computing and output files
- KIndex (Knowledge Index) is a KB index that measures how much knowledge is contained in the cataloged groups: if KIndex reached its maximum value of 1, every group would be made up of records with constant values in all variables / columns and each group would different from the other groups.

KIndex is calculated using means of CV of the variables / columns of the groups of input files before cataloging: see the source program KB_CAT for the computing details.

In the case under examination, the Kindex value, not particularly high (0.8438), suggests to run a new processing increasing, for example, the number of groups from 3 to 4 obtaining a certain improvement of Kindex.

File - Output/Catalog.original (vessels_M_g3_out.txt)

It is identical to the input file with the addition of the column for the input of the code of the group it belongs to.

The Output/Catalog.sort file is more interesting, in that it shows the classified records that each group belong to.

File of Output/Catalog.sort (vessels_*M*_*g*3_*outsrt.txt*)

This is identical to the previous file but the records / cases are in order according to the code of the group it belongs to.

Group	description	shape	material	height	colour	weight	haft	plug
G_00_00	ancient_bottle	ball_cone	glass	40	Green	150	no	cork
G_00_00	bottle_1	cylinder_cone	glass	40	Green	120	no	cork
G_00_00	bottle_4	cylinder_cone	glass	35	Green	125	no	metal
G_00_00	carboy	ball_cone	glass	80	Green	15000	no	cork
G_00_00	magnum_bottle	cylinder_cone	glass	50	Green	170	no	metal
G_00_00	plant_pot	cut_cone	terracotta	30	Brown	200	no	no
G_00_00	umbrella_stand	cylinder	metal	100	Grey	3000	no	no
G_00_01	pot_1	cylinder	metal	40	Grey	500	two	yes
G_00_02	coffee_cup	cut_cone	ceramic	6	White	20	yes	no
G_00_02	cup_1	cut_cone	ceramic	10	White	30	yes	no
G_00_02	cup_2	cut_cone	glass	20	transparent	35	yes	no
G_00_02	pot_2	cut_cone	metal	7	Grey	200	yes	yes
G_01_00	beer_jug	cut_cone	porcelain	18	severals	25	no	no
G_01_00	bottle_2	cylinder_cone	glass	40	transparent	125	no	cork
G_01_00	bottle_3	cylinder_cone	glass	45	opaque	125	no	plastic
G_01_00	glass_1	cut_cone	pewter	10	pewter	20	no	no
G_01_00	glass_3	cut_cone	terracotta	8	Grey	20	no	no
G_01_00	tuna_can	cylinder	metal	10	severals	10	no	no
G_02_00	cd	parallelepiped	plastic	1	transparent	4	no	no
G_02_00	champagne_glass	cut_cone	crystal	17	transparent	17	no	no

Group	description	shape	material	height	colour	weight	haft	plug
G_02_00	dessert_glass	cut_cone	glass	17	transparent	17	no	no
G_02_00	glass_2	cut_cone	plastic	9	White	4	no	no
G_02_00	pasta_case	parallelepiped	glass	35	transparent	150	no	metal
G_02_00	perfume	parallelepiped	glass	7	transparent	15	<mark>no</mark>	plastic
G_02_00	tetrapack1	parallelepiped	mixed	40	severals	20	<mark>no</mark>	plastic
G_02_00	tetrapack2	parallelepiped	plastic	40	severals	21	<mark>no</mark>	plastic
G_02_00	tetrapack3	parallelepiped	millboard	40	severals	22	<mark>no</mark>	no
G_02_00	toothpaste	cylinder	plastic	15	severals	7	no	plastic
G_02_00	trousse	cylinder	plastic	1	silver	7	no	yes
G_02_00	tuna_tube	cylinder	plastic	15	severals	7	<mark>no</mark>	plastic
G_02_00	visage_cream	cylinder	metal	15	White	7	no	no
G_02_00	wine_glass	cut_cone	glass	15	transparent	15	no	no
G_02_01	pyrex	parallelepiped	glass	10	transparent	300	two	glass
G_02_02	cleaning_1	parall_cone	plastic	30	White	50	yes	plastic
G_02_02	cleaning_2	cylinder_cone	plastic	30	Blue	60	yes	plastic
G_02_02	cleaning_3	cone	plastic	100	severals	110	yes	plastic
G_02_02	jug	cylinder	terracotta	25	White	40	yes	no
G_02_02	milk_cup	cut_cone	terracotta	15	Blue	35	yes	no
G_02_02	tea_cup	cut_cone	terracotta	7	White	30	yes	no
G_02_02	watering_can	irregular	plastic	50	Green	400	yes	no

On first sight you can see that the program KB_CAT is able to catalog records in homogeneous groups for content.

It is important to note that the vessels.txt files are formed by just 40 records which are all quite different.

For example:

- the group G_00_00 is characterised by objects that are primarily of a green colour, and with haft
- the group G_00_02 is primarily formed by objects of a cut_cone shape, with haft and without a plug
- the group G_02_00 is characterised by objects that are parallelepiped / cylinder / cut cone shape and without haft
- the group G 02 02 is made up of plastic and terracotta objects with haft

If the processed input file had been formed with numerous records and with many variables / columns, it would not have been so easy to draw conclusions on the results of the cataloging only visually examining the files.

The KB_STA program is dedicated to resolving the problem which has just been highlighted.

Output/Means, Std, CV (vessels_M_g3_medsd.txt)

File containing the Means, the Maximums, the Std and the CV with normalized values of the whole population.

Low values of the CV (coefficient of variation) indicate that the values of the variables / columns are not dispersed.

shape	material	height	colour	weight	haft	plug
Mean1	Mean2	Mean3	Mean4	Mean5	Mean6	Mean7

shape	material	height	colour	weight	haft	plug
0.4892	0.5000	28.075	0.6222	530.32	0.3000	0.5900
Max1	Max2	Max3	Max4	Max5	Max6	Max7
1.0000	1.0000	100.0	1.0000	15000.	1.0000	1.0000
Std1	Std2	Std3	Std4	Std5	Std6	Std7
0.7371	0.8164	60.592	0.8210	6103.1	1.1474	0.6526
CV_1	CV_2	CV_3	CV_4	CV_5	CV_6	CV_7
1.5066	1.6329	2.1582	1.3194	11.508	3.8248	1.1062

Output/CV files (vessels_M_g3_cv.txt)

Groups	shape	material	height	colour	weight	haft	plug	Means	N_recs
G_00_00	0.69	0.77	0.45	0.27	1.91	0	0.91	0.71	7
G_00_01	0	0	0	0	0	0	0	0	1
G_00_02	0	1.04	0.52	0.34	1.05	0	0.25	0.46	4
G_01_00	0.32	0.57	0.69	0.30	0.93	0	0.47	0.47	6
G_02_00	0.51	0.52	0.71	0.15	1.61	0	0.21	0.53	14
G_02_01	0	0	0	0	0	0	0	0	1
G_02_02	0.51	0.13	0.78	0.79	1.19	0	0.14	0.51	7
Means	0.44	0.53	0.62	0.32	1.35	0	0.35	0.51	40
Total	1.51	1.63	2.16	1.32	11.51	3.82	1.11	3.29	40

The file contains information relevant for measuring the quality of the cataloging.

The value contained in every cell represents the importance of the values of the variables / columns in the group: the more the value is close to zero, the more the variable / column is important in the cataloging.

If the value is equal to zero, the variable / column in that group will have an identical value: for example all groups have identical values in the variable *haft*.

The values in the cells of the penultimate column (Means) indicate if the groups are internally homogeneous considering all the variables / columns: the higher the value is close to zero, the greater the similarity of the record / cases to each other within the group under consideration.

The groups G_00_02 and G_01_00 are homogeneous, while the group G_00_00 is not, due to the important CV values of the variables *weight* and *plug*.

It is also important to compare the values contained in every line / column with the value contained in the last two lines: *Means* and *Total* (referring to the all records before the cataloging).

Output/Training Grid (vessels_M_g3_grid.txt)

The file contains the values of the three-dimensional training matrix with minimum error; this matrix is used by the KB_CLA program used to classify new records /

cases that can be recognised and classified according to what has previously been learnt by the program KB_CAT.

Group	SubGroup	Variable/Column	Values
0	0	0	0,3976159
0	0	1	0,4249143
0	0	2	0,4221095
0	0	3	0,3706712
0	0	4	0,1070639
0	0	5	0,0721792
0	0	6	0,4288610
0	1	0	0,3760895
0	1	1	0,3555886
0	1	2	0,3351283
0	1	3	0,4836650
0	1	4	0,0767009
0	1	5	0,3319249
0	1	6	0,5141450
0	2	0	0,3522021
0	2	1	0,1886213
0	2	2	0,1638941
0	2	3	0,6998640
0	2	4	0,0115530
0	2	5	0,8734927
0	2	6	0,7434203
1	0	0	0,5722823
1	0	1	0,4691723
1	0	2	0,2864130
1	0	3	0,6216960
1	0	4	0,0428225
1	0	5	0,0569301
1	0	6	0,5809196
1	1	0	0,5466298
1	1	1	0,5135355
1	1	2	0,2899887
1	1	3	0,6104640

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 21 di 112

Group	SubGroup	Variable/Column	Values
1	1	4	0,0296109
1	1	5	0,3230673
1	1	6	0,6348858
1	2	0	0,4737209
1	2	1	0,5358513
1	2	2	0,2805610
1	2	3	0,6148486
1	2	4	0,0108941
1	2	5	0,9004934
1	2	6	0,6831299
2	0	0	0,6283160
2	0	1	0,4785080
2	0	2	0,2024570
2	0	3	0,7459708
2	0	4	0,0055453
2	0	5	0,0683992
2	0	6	0,6433004
2	1	0	0,6078937
2	1	1	0,5633861
2	1	2	0,2537548
2	1	3	0,6914334
2	1	4	0,0067944
2	1	5	0,2961828
2	1	6	0,6576649
2	2	0	0,5420435
2	2	1	0,7055653
2	2	2	0,3505488
2	2	3	0,5606647
2	2	4	0,0126543
2	2	5	0,8661729
2	2	6	0,6630445

Statistical analysis of the results of the cataloging

The file contains the results of the processing of KB_CAT statistically analysed running the program KB_STA, using the parameters below listed.

```
# KB STA KNOWLEDGE DISCOVERY IN DATA MINING (STATISTICAL PROGRAM)
# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
# Language used: PYTHON
INPUT - Catalogued Records File ( outsrt.txt) ->
vessels M q3 outsrt.txt INPUT - Groups / CV File ( cv.txt)
                                                                                                   ->
vessels M g3 cv.txt
Group Consistency (% from 0 to 100)
                                                                  -> 0
Variable Consistency (% from 0 to 100)
                                                                  -> 0
Select groups containing records >=
                                                                  -> 4
                                                                  -> 1000
Select groups containing records <=
Summary / Detail report (S / D)
                                                                   -> D
Display Input Records (Y / N)
-> vessels M_g3_sta.txt
Report File
KB_STA - Statistical Analysis from: vessels M_g3_outsrt.txt
and from: vessels M q3 cv.txt
Min Perc. of group Consistency: 0 Min Perc. of variable Consistency: 0
Min Number of records: 4 Max Number of records: 1000
by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
______
G 00 00 Consistency 0.7140 %Consistency 0.0 Records 7 %Records 17.50
*** shape Consistency 0.6910 %Consistency 3.22
G_00_00 ID record ancient_bottle Value ball_cone
G_00_00 ID record bottle_1 Value cylinder_cone
G_00_00 ID record bottle_4 Value cylinder_cone
G_00_00 ID record carboy Value ball_cone
G_00_00 ID record magnum_bottle Value cylinder_cone
G_00_00 ID record plant_pot Value cut_cone
G_00_00 ID record umbrella_stand Value cylinder
Value cylinder_cone Frequency 3 Percentage 42.00
Value ball_cone Frequency 2 Percentage 28.00
Value cylinder Frequency 1 Percentage 14.00
Value cut_cone Frequency 1 Percentage 14.00
*** material Consistency 0.7687 %Consistency
                                                                                              0.00
G_00_00 ID record ancient_bottle Value glass
G_00_00 ID record bottle_1 Value glass
G_00_00 ID record bottle_4 Value glass
G_00_00 ID record carboy Value glass
G_00_00 ID record magnum_bottle Value glass
G_00_00 ID record plant_pot Value terracotta
G_00_00 ID record umbrella_stand Value metal
Value glass Frequency 5 Percentage 71.00
Value terracotta Frequency 1 Percentage 14.00
Value metal Frequency 1 Percentage 14.00
*** height Consistency 0.4537 %Consistency
                                                                                           36.46
G_00_00 ID record ancient_bottle Value 40.0
G_00_00 ID record bottle_1 Value 40.0 G_00_00 ID record bottle_4 Value 35.0 G_00_00 ID record carboy Value 80.0 G_00_00 ID record magnum_bottle Value 50.0
G_00_00 ID record plant_pot Value 30.0
G_00_00 ID record umbrella_stand Value 100.0

Min 30.00 Max 100.00 Step 17.50

First Quartile (end) 47.50 Frequency % 57.14

Second Quartile (end) 65.00 Frequency % 14.29

Third Quartile (end) 82.50 Frequency % 14.29

Fourth Quartile (end) 100.00 Frequency % 14.29
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 23 di 112

	r	Consist	ency 0.267	73	%Consistency		62.56
G 00 00		record	-		_		
G 00 00	ID	record	bottle 1		green		
G 00 00	ID	record	bottle 4		green		
G_00_00	ID	record	carboy	Value	green		
G_00_00	ID	record	magnum_bottle	Value	green		
G_00_00	ID	record	plant_pot	Value	brown		
G_00_00		record	umbrella_stand				
Value gree			Frequency	5		71.00	
Value grey			Frequency	1	,	14.00	
Value brow			Frequency	1	Percentage	14.00	0 00
*** weight			_		%Consistency		0.00
G_00_00		record	<pre>ancient_bottle bottle_1</pre>	Value Value			
G_00_00 G_00_00		record record	bottle 4	Value			
G_00_00 G_00_00		record	carboy		15000.0		
G_00_00		record	magnum bottle				
G 00 00		record	-	Value			
G 00 00		record	umbrella stand				
Min 120.		Max	15000.00 Step				
			3840.00 Fr			<mark>71</mark>	
Fourth Quar			15000.00 Fr				
		sistenc			sistency	100.00	
G_00_00		record		Value	no		
G_00_00	ID	record	bottle_1	Value	no		
G_00_00	ID	record	bottle_4	Value	no		
G_00_00	ID	record	carboy	Value	no		
G_00_00	ID	record	magnum_bottle	Value	no		
G_00_00	ID	record	plant_pot	Value	no		
G_00_00	ID	record	umbrella_stand	Value			
Value no			Frequency	7	Percentage	100.00	
*** plug			-		sistency	0.00	
c = 00 = 00		record	ancient bottle	Va⊥ue	cork		
G_00_00			-				
G_00_00	ID	record	bottle_1	Value	cork		
G_00_00 G_00_00	ID ID	record record	bottle_1 bottle_4	Value Value	cork metal		
G_00_00 G_00_00 G_00_00	ID ID ID	record record record	bottle_1 bottle_4 carboy	Value Value Value	cork metal cork		
G_00_00 G_00_00 G_00_00	ID ID ID	record record record	bottle_1 bottle_4 carboy	Value Value Value	cork metal cork		
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00	ID ID ID ID	record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot	Value Value Value Value Value	cork metal cork metal no		
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00	ID ID ID ID ID ID	record record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand	Value Value Value Value Value Value	cork metal cork metal no	42.00	
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork	ID ID ID ID ID ID	record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency	Value Value Value Value Value 3	cork metal cork metal no no Percentage	42.00 28.00	
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork	ID ID ID ID ID ID	record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency	Value Value Value Value Value 3 2	cork metal cork metal no no Percentage Percentage	28.00	
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork	ID ID ID ID ID ID ID ID	record record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency	Value Value Value Value Value Value 2 2	cork metal cork metal no no Percentage Percentage Percentage	28.00 28.00	
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta	ID ID ID ID ID ID A	record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency	Value Value Value Value Value Value 2 ======	cork metal cork metal no no Percentage Percentage Percentage	28.00 28.00	
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID ID ID ID ID C al Consi	record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency 0.4559 %Consiste	Value Value Value Value Value 3 2 2 ======= ncy 1	cork metal cork metal no no Percentage Percentage Percentage 2 Records 4 %Consistency	28.00 28.00 %Records	
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID ID ID ID ID C al Consi	record record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency 0.4559 %Consisteed	Value Value Value Value Value 3 2 2 ======= ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone	28.00 28.00 %Records	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID ID ID ID ID C al Consi	record record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency Frequency 0.4559 %Consiste cocffee_cup	Value Value Value Value 3 2 2 ======= ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone	28.00 28.00 ===================================	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID ID ID ID ID C al Consi	record record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency Frequency 0.4559 %Consiste cocffee_cup	Value Value Value Value 3 2 2 ======= ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone	28.00 28.00 ===================================	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID ID ID ID ID C al Consi	record record record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency Frequency 0.4559 %Consiste cocffee_cup	Value Value Value Value 3 2 2 ======= ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone	28.00 28.00 ===================================	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency	Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value Value Value Value Value Value	cork metal cork metal no no Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone cut_cone cut_cone cut_cone Percentage	28.00 28.00 *Records	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ======== G_00_02 *** shape G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 Value cut *** materi	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency 0.4559 %Consiste ency coffee_cup cup_1 cup_2 pot_2 Frequency tency 1.0392	Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value Value Value Value Value Value Value Value Value	cork metal cork metal no no Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone cut_cone cut_cone cut_cone cut_cone	28.00 28.00 ===================================	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ======== G_00_02 *** shape G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 Value cut_ *** materi G_00_02	ID I	record record record record record record record record consiste record record record record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency	Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone	28.00 28.00 *Records	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency ***Trequency ***Trequency ***Trequency ***Trequency ***Trequency ***Trequency coffee_cup cup_1 cup_2 pot_2 Frequency ***Trequency **Trequency ***Trequency ***Trequency ***Trequency ***Trequency **	Value Value Value Value Value 3 2 2 ======= ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone	28.00 28.00 *Records	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency **Tequency **Tequency **Tequency **Tequency **Tequency **Tequency **Tequency coffee_cup cup_1 cup_2 pot_2 **Tequency **Tequen	Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone gistency ceramic ceramic glass	28.00 28.00 *Records	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency	Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone gistency ceramic ceramic glass metal	28.00 28.00 *Records	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency	Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone gistency ceramic ceramic glass metal Percentage	28.00 28.00 *Records	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ======== G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 Value cut_ *** materi G_00_02 G_00_02 G_00_02 Value cut_ Value cera Value meta	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency 0.4559 %Consiste ency 0.000 coffee_cup cup_1 cup_2 pot_2 Frequency tency 1.0392 coffee_cup cup_1 cup_2 pot_2 Frequency Frequency Frequency Frequency Frequency Frequency Frequency Frequency	Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone percentage sistency ceramic ceramic glass metal Percentage Percentage	28.00 28.00 ===================================	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ======== G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 G_00_02 Value cut *** materi G_00_02 G_00_02 G_00_02 Value cut Value meta Value meta Value glas	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency 0.4559 %Consiste ncy coffee_cup cup_1 cup_2 pot_2 Frequency tency 1.0392 coffee_cup cup_1 cup_2 pot_2 Frequency Frequency Frequency Frequency Frequency Frequency Frequency Frequency	Value Value Value Value Value Value Value O Value 1	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone percentage glass metal Percentage Percentage Percentage Percentage	28.00 28.00 ===================================	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency	Value Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone Percentage sistency ceramic ceramic glass metal Percentage Percentage Percentage Percentage %Consistency	28.00 28.00 ===================================	10.00
G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 G_00_00 Value cork Value no Value meta ====================================	ID I	record	bottle_1 bottle_4 carboy magnum_bottle plant_pot umbrella_stand Frequency Frequency Frequency 0.4559 %Consiste ncy coffee_cup cup_1 cup_2 pot_2 Frequency tency 1.0392 coffee_cup cup_1 cup_2 pot_2 Frequency Frequency Frequency Frequency Frequency Frequency Frequency Frequency	Value Value Value Value Value Value 3 2 2 ====== ncy 1 00 Value	cork metal cork metal no no Percentage Percentage Percentage Percentage 2 Records 4 %Consistency cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone cut_cone Percentage Sistency ceramic ceramic glass metal Percentage Percentage Percentage Percentage Percentage %Consistency 6.0	28.00 28.00 *Records 100.00 0.00	10.00

```
G_00_02 ID record cup_1 Value 10.0
G_00_02 ID record cup_2 Value 20.0
G_00_02 ID record pot_2 Value 7.0
Min 6.00 Max 20.00 Step 3.50
First Quartile (end) 9.50 Frequency $ 50.00
Second Quartile (end) 13.00 Frequency $ 25.00
Fourth Quartile (end) 20.00 Frequency $ 25.00
*** colour Consistency 0.3431 %Consistency 24.74
G_00_02 ID record coffee_cup Value white
G_00_02 ID record cup_1 Value white
G_00_02 ID record cup_2 Value transparent
G_00_02 ID record pot_2 Value grey
Value white Frequency 2 Percentage 50.00
Value transparent Frequency 1 Percentage 25.00
Value grey Frequency 1 Percentage 25.00
*** weight Consistency 1.0460 %Consistency
                                                                                                                                                          0.00
G_00_02 ID record cup_1 Value 30.0

G_00_02 ID record cup_1 Value 35.0

G_00_02 ID record cup_2 Value 200.0

Min 20.00 Max 200.00 Step 45.00

First Quartile (end) 65.00 Frequency $ 75.00

Fourth Quartile (end) 200.00 Frequency $ 25.00

*** haft Consistency 0.0000 %Consistency 100.00
*** haft Consistency 0.0000 %Consistency 1

G_00_02 ID record coffee_cup Value yes

G_00_02 ID record cup_1 Value yes

G_00_02 ID record cup_2 Value yes

G_00_02 ID record pot_2 Value yes

Value no

G_00_02 ID record coffee_cup Value no

G_00_02 ID record cup_1 Value no

G_00_02 ID record cup_2 Value no

G_00_02 ID record cup_2 Value yes

Value no

G_00_02 ID record pot_2 Value yes

Value yes

Value yes

Frequency 3 Percentage 75.00

Value yes

Frequency 1 Percentage 25.00
value yes
 ______
 G_01_00 Consistency 0.4666 %Consistency 10 Records 6 %Records 15.00
 *** shape Consistency 0.3168 %Consistency 32.10
*** material Consistency 0.5657 %Consistency 0.00
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 25 di 112

```
*** height Consistency 0.6877 %Consistency
                                                                                                                                                                                                                                                                                                                                                                     0.00
  *** colour Consistency 0.2997 %Consistency
                                                                                                                                                                                                                                                                                                                                                                                                 35.77
  G_01_00 ID record beer_jug Value severals
G_01_00 ID record bottle_2 Value transparent
G_01_00 ID record bottle_3 Value opaque
G_01_00 ID record glass_1 Value pewter
G_01_00 ID record glass_3 Value grey
G_01_00 ID record tuna_can Value severals
  Value severals

Value severals

Frequency

Value transparent

Frequency

Value pewter

Frequency

Value opaque

Frequency

Value opaque

Frequency

Value grey

Frequency

Value grey

Frequency

Value grey

Frequency

Value opaque

Frequency

Value opaque

Frequency

Value opaque

V
*** weight Consistency 0.9283 %Consistency

G_01_00 ID record beer_jug Value 25.0

G_01_00 ID record bottle_2 Value 125.0

G_01_00 ID record bottle_3 Value 125.0

G_01_00 ID record glass_1 Value 20.0

G_01_00 ID record glass_3 Value 20.0

G_01_00 ID record tuna_can Value 10.0

Min 10.00 Max 125.00 Step 28.75

First Quartile (end) 38.75 Frequency % 66.67

Fourth Quartile (end) 125.00 Frequency % 33.33

*** haft Consistency 0.0000 %Consistency

G_01_00 ID record beer_jug Value no

G_01_00 ID record bottle_2 Value no

G_01_00 ID record bottle_3 Value no

G_01_00 ID record glass_1 Value no

G_01_00 ID record glass_1 Value no

G_01_00 ID record glass_3 Value no

G_01_00 ID record glass_3 Value no

G_01_00 ID record glass_3 Value no

G_01_00 ID record tuna_can Value no

Value no Frequency 6 Percentage 100.00

*** plug Consistency 0.4677 %Consistency
                                                                                                                                                                                                                                                                                                                                                                                                         0.00
                                                                                                                                                                                                                                                                                                                                                              100.00
  ValuenoFrequency6Percentage100.00*** plugConsistency0.4677%ConsistencyG_01_00ID recordbeer_jugValue noG_01_00ID recordbottle_2Value corkG_01_00ID recordbottle_3Value plasticG_01_00ID recordglass_1Value noG_01_00ID recordglass_3Value noG_01_00ID recordtuna_canValue noValuenoFrequency4PercentageValueplasticFrequency1PercentageValuecorkFrequency1PercentageValuecorkFrequency1PercentageValuecorkFrequency1PercentageValuecorkFrequency1PercentageValuecorkFrequency1PercentageValuecorkFrequency1PercentageValuecorkFrequency1PercentageValuecorkFrequency1Percentage
                                                                                                                                                                                                                                                                                                                                                                 0.00
    ______
   G_02_00 ID record champagne_glass Value cut_cone
G_02_00 ID record dessert_glass Value cut_cone
G_02_00 ID record glass_2 Value cut_cone
G_02_00 ID record pasta_case Value parallelepiped
G_02_00 ID record perfume Value parallelepiped
                                 KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 26 di 112
```

```
G 02 00
          ID record tetrapack1
                                     Value parallelepiped
G 02 00
                                     Value parallelepiped
          ID record
                      tetrapack2
G 02 00
          ID record
                      tetrapack3
                                     Value parallelepiped
          ID record toothpaste
G_02_00
                                     Value cylinder
                                     Value cylinder
G_02_00
          ID record trousse
G 02 00
          ID record tuna tube
                                     Value cylinder
G 02 00
          ID record
                      visage cream
                                     Value cylinder
                    wine_glass
          ID record
                                     Value cut cone
G 02 00
Value parallelepiped Frequency
                                       6 Percentage
                                                       42.00
Value cylinder
                                                       28.00
                     Frequency
                                           Percentage
Value cut_cone
                                                       28.00
                     Frequency
                                          Percentage
*** material Consistency
                                0.5228
                                           %Consistency
                                                                  1.36
G 02 00
       ID record cd
                                     Value plastic
          ID record champagne glass Value crystal
G 02 00
G 02 00
          ID record dessert glass
                                     Value glass
          ID record glass_2
ID record pasta_case
G 02 00
                                     Value plastic
                      glass 2
G_02_00
                                     Value glass
G_02_00
          ID record perfume
                                     Value glass
G 02 00
          ID record tetrapack1
                                     Value mixed
G_02_00
          ID record tetrapack2
                                     Value plastic
G 02 00
          ID record tetrapack3
                                     Value millboard
                                     Value plastic
G 02 00
          ID record toothpaste
G 02 00
          ID record trousse
                                     Value plastic
G 02 00
          ID record tuna_tube
                                     Value plastic
          ID record
                      visage_cream
G 02 00
                                     Value metal
G_02_00
          ID record wine_glass
                                     Value glass
Value plastic Frequency
                                     6 Percentage
                                                       42.00
                   Frequency
Value glass
                                       4 Percentage
                                                       28.00
                                       1 Percentage
1 Percentage
Value mixed
                                                       7.00
                     Frequency
Value millboard
                     Frequency
                                                       7.00
Value metal
                                       1 Percentage 7.00
                     Frequency
Value crystal
                                           Percentage 7.00
                   Frequency
*** height Consistency
                                          %Consistency
                                0.7067
                                                                  0.00
G 02 00
          ID record
                      cd
                                     Value 1.0
G 02 00
          ID record
                      champagne_glass Value 17.0
G 02 00
          ID record dessert glass Value 17.0
                      glass_2
G 02 00
          ID record
                                     Value 9.0
G 02 00
          ID record pasta_case
                                     Value 35.0
          ID record
G_02_00
                      perfume
                                     Value 7.0
          ID record tetrapack1
G 02 00
                                     Value 40.0
          ID record tetrapack2
                                    Value 40.0
G 02 00
G_02_00
          ID record tetrapack3
                                     Value 40.0
          ID record toothpaste
G 02 00
                                     Value 15.0
          ID record trousse
                                     Value 1.0
G 02 00
G 02 00
          ID record tuna tube
                                     Value 15.0
G 02 00
          ID record
                                     Value 15.0
                      visage_cream
                                     Value 15.0
G 02 00
          ID record
                      wine glass
                                     9.75
Min 1.00
              Max
                      40.00 Step
First Quartile (end)
                        10.75
                                  Frequency %
                                                    28.57
                          20.50
Second Quartile (end)
                                  Frequency %
                                                    42.86
                          40.00 Frequency %
Fourth Quartile (end)
                                                    28.57
*** colour Consistency
                                0.1507
                                           %Consistency
                                                                 71.57
G 02 00
          ID record cd
                                     Value transparent
G 02 00
          ID record
                      champagne glass Value transparent
G_02 00
          ID record
                      dessert glass Value transparent
G 02 00
          ID record
                      glass 2
                                     Value white
          ID record
G_02_00
                      pasta_case
                                     Value transparent
G 02 00
          ID record
                      perfume
                                     Value transparent
G 02 00
          ID record
                                     Value severals
                      tetrapack1
                      tetrapack2
G 02 00
          ID record
                                     Value severals
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 27 di 112

```
G 02 00
           ID record tetrapack3
                                    Value severals
           ID record toothpaste
G 02 00
                                      Value severals
           ID record trousse
ID record tuna_tube
G 02 00
                                      Value silver
G_02_00
                                      Value severals
           ID record visage_cream Value white
G_02_00
G 02 00 ID record wine_glass
                                      Value transparent
                                     6 Percentage
5 Percentage
2 Percentage
Value transparent Frequency
Value severals Frequency
                                                         42.00
Value severals
                     Frequency
                                                         35.00
               Frequency
Frequency
Value white
                                                         14.00
Value silver
                                       1 Percentage 7.00
                                1.6075
*** weight Consistency
                                                                     0.00
                                            %Consistency
G_02_00 ID record cd
                                      Value 4.0
G 02 00
           ID record champagne_glass Value 17.0
         ID record dessert_glass Value 17.0
G 02 00
          ID record glass_2 Value 4.0
ID record pasta_case Value 150.0
G 02 00
G_02_00 ID record pasta_case
G_02_00 ID record perfume
G_02_00 ID record tetrapack1
                                      Value 15.0
                                    Value 20.0
Value 21.0
          ID record tetrapack2
G 02 00
G_02_00
           ID record tetrapack3
                                     Value 22.0
           ID record toothpaste
G 02 00
                                      Value 7.0
           ID record trousse
                                      Value 7.0
G 02 00
           ID record tuna_tube Value 7.0
G 02 00
           ID record visage_cream Value 7.0
G_02_00
G_02_00
           ID record wine_glass Value 15.0
             Max
                       150.00 Step 36.50
Min 4.00
                                                     92.86
First Quartile (end) 40.50 Frequency %
Fourth Quartile (end) 150.00 Frequency %

*** haft Consistency 0.0000 %Consistency C.02.00
                                                     7.14
                                      %Consistency
                                                             100.00
G 02 00
           ID record cd
                                      Value no
G 02 00
           ID record champagne_glass Value no
G 02 00
           ID record dessert glass Value no
           ID record glass_2
ID record pasta_case
G_02_00
                                      Value no
G 02 00
                                      Value no
G 02 00
           ID record perfume
                                      Value no
                                    Value no
G 02 00
           ID record tetrapack1
           ID record tetrapack2
G 02 00
                                     Value no
           ID record tetrapack3
ID record toothpaste
G 02 00
                                      Value no
G_02_00
                                      Value no
G 02 00
           ID record trousse
                                     Value no
           ID record tuna_tube Value no
G 02 00
           ID record visage_cream Value no
G 02 00
           ID record wine_glass
G 02 00
                                      Value no
Value no Frequency 14 Percentage 100.00
*** plug
           Consistency 0.2125
                                      %Consistency
                                                              59.91
G 02 00
                                      Value no
           ID record
                       cd
           ID record
G 02 00
                       champagne_glass Value no
G 02 00
           ID record dessert_glass Value no
G 02 00
           ID record glass 2
                                      Value no
           ID record pasta_case
                                     Value metal
G 02 00
           ID record perfume
ID record tetrapack1
G 02 00
                                      Value plastic
G 02 00
                                      Value plastic
G 02 00
                                      Value plastic
           ID record tetrapack2
G 02 00
           ID record tetrapack3
                                      Value no
G_02_00
           ID record toothpaste
                                      Value plastic
           ID record
G 02 00
                       trousse
                                      Value yes
           ID record trousse
ID record tuna_tube
G 02 00
                                      Value plastic
G 02 00
           ID record visage cream
                                      Value no
G 02 00
           ID record wine_glass
                                      Value no
Value no
                      Frequency
                                        7 Percentage
                                                        50.00
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 28 di 112

```
ValueplasticFrequency5Percentage35.00ValueyesFrequency1Percentage7.00ValuemetalFrequency1Percentage7.00
  ______
  G_02_02 Consistency 0.5053 %Consistency 2 Records 7 %Records 17.50
*** shape Consistency 0.5070 %Consistency
G_02_02 ID record cleaning_1 Value parall_cone
G_02_02 ID record cleaning_2 Value cylinder_cone
G_02_02 ID record cleaning_3 Value cone
G_02_02 ID record jug Value cylinder
G_02_02 ID record milk_cup Value cut_cone
G_02_02 ID record tea_cup Value cut_cone
G_02_02 ID record watering_can Value irregular
Value cut_cone Frequency 2 Percentage 28.00
Value parall_cone Frequency 1 Percentage 14.00
Value irregular Frequency 1 Percentage 14.00
Value cylinder_cone Frequency 1 Percentage 14.00
Value cylinder Frequency 1 Percentage 14.00
Value cylinder Frequency 1 Percentage 14.00
Value cone Frequency 1 Percentage 14.00
*** material Consistency 0.1260 %Consistency
G_02_02 ID record cleaning_1 Value plastic
  *** shape Consistency 0.5070 %Consistency 0.00
                                                                                                                                                                                                                                                                                         75.06
*** material Consistency 0.1260 %Consistency

G_02_02 ID record cleaning_1 Value plastic

G_02_02 ID record cleaning_2 Value plastic

G_02_02 ID record cleaning_3 Value plastic

G_02_02 ID record jug Value terracotta

G_02_02 ID record milk_cup Value terracotta

G_02_02 ID record tea_cup Value terracotta

G_02_02 ID record watering_can Value plastic

Value plastic Frequency 4 Percentage 57.00

Value terracotta Frequency 3 Percentage 42.00

*** height Consistency 0.7815 %Consistency

G_02_02 ID record cleaning 1 Value 30.0
                                                                                                                                                                                                                                                                                                 0.00
G_02_02 ID record cleaning_1 Value 30.0
G_02_02 ID record cleaning_2 Value 30.0
G_02_02 ID record cleaning_3 Value 100.0
G_02_02 ID record jug Value 25.0
G_02_02 ID record milk_cup Value 15.0
G_02_02 ID record tea_cup Value 7.0
G_02_02 ID record watering_can Value 50.0
 Min 7.00 Max 100.00 Step 23.25

First Quartile (end) 30.25 Frequency % 71.43

Second Quartile (end) 53.50 Frequency % 14.29

Fourth Quartile (end) 100.00 Frequency % 14.29

*** colour Consistency 0.7856 %Consistency
                                                                                                                                                                                                                                                                                                 0.00
 G_02_02 ID record cleaning_1 Value white
G_02_02 ID record cleaning_2 Value blue
G_02_02 ID record cleaning_3 Value severals
G_02_02 ID record cleaning_3 Value severals
G_02_02 ID record jug Value white
G_02_02 ID record milk_cup Value blue
G_02_02 ID record tea_cup Value white
G_02_02 ID record tea_cup Value green
Value white Frequency 3 Percentage 42.00
Value blue Frequency 2 Percentage 28.00
Value severals Frequency 1 Percentage 14.00
Value green Frequency 1 Percentage 14.00
*** weight Consistency 1.1928 %Consistency 0.00
G_02_02 ID record cleaning_1 Value 50.0
G_02_02 ID record cleaning_2 Value 60.0
G_02_02 ID record cleaning_3 Value 110.0
G_02_02 ID record jug Value 40.0
G_02_02 ID record milk_cup Value 35.0
G_02_02 ID record tea_cup Value 30.0

KB - Neural Data Mining with Python sources - Roberto Bello - Pag. 29 d
                        KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 29 di 112
```

G_02	_		record	watering_	_					
Min	30.0		Max	400.00		p 92.5				
			e (end)			requency		85.7		
			e (end)			requency		14.2		
			onsister	-		000	%Consis	tency	-	100.00
G_02			record	cleaning_	_		-			
G_02			record			Value	_			
G_02	_		record	cleaning_	_3	Value	_			
G_02	_	ID	record	jug		Value	yes			
G_02	_	ID	record	milk_cup		Value	yes			
G_02	_02	ID	record			Value	yes			
G_02	_02	ID	record	watering_	can	Value	yes			
<mark>Valu</mark>	e yes			Frequency		7	Percent	age	100.00	
***	plug	Co	nsistend	cy 0.144	13	%Cons	istency		71.44	
G_02	_02	ID	record	cleaning_	_1	Value	plastic			
G_02	_02	ID	record	cleaning_	_2	Value	plastic			
G_02	_02	ID	record	cleaning_	_3	Value	plastic			
G_02	_02	ID	record	jug		Value	no			
G_02	_02	ID	record	milk_cup		Value	no			
G_02	_02	ID	record	tea_cup		Value	no			
G 02	02	ID	record	watering	can	Value	no			
Value	e no			Frequency	_	4	Percent	age	57.00	
Value	e plas	stic		Frequency		3	Percent	age	42.00	
====	======		======		=====			=====	=======	======
Mea	ns Cor	nsis	tency 0.	.5145 %Consi	stenc	y 0 Reco	rds	40	%Records	100.00
***	shape		Consi	istency	0.43	357	%Consis	tency		15.32
***	materi	ial	Consi	istency	0.52	183	%Consis	tency		0.00
***	height	;	Consi	istency	0.61	.82	%Consis	tency		0.00
***	colour	<u>-</u>	Cons	sistency	0.31	.63	%Consis	tency		38.52
***	weight	:	Consi	istency	1.34	98	%Consis	tency		0.00
***	haft		Consi	istency	0.00	000	%Consis	tency		100.00
***	plug		Consi	istency	0.35	30	%Consis	tency		31.39

Other input files to KB_CAT (animals.txt)

The *animals.txt* file is formed by 84 records with 15 variables / columns (Fur, Feather, Eggs, Milk, Flying, Aquatic, Predatory, Teeth, Invertebrate, Lungs, Poisonous, Flippers, Legs, Tail, Domestic).

ANIMAL	FUR	FEATHER	EGGS	MILK	FLYING	AQUATIC	PREDATORY	TEETH	INVERT.	LUNGS	POIS.	FLIP.	LEGS	TAIL	DOM.
SKYLARK	0	1	1	0	1	0	0	0	1	1	0	0	2	1	0
DUCK	0	1	1	0	1	1	0	0	1	1	0	0	2	1	0
ANTELOPE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
BEE	1	0	1	0	1	0	0	0	0	1	1	0	6	0	1
LOBSTER	0	0	1	0	0	1	1	0	0	0	0	0	6	0	0
HERRING	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
FIELD_MOUSE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
HAWK	0	1	1	0	1	0	1	0	1	1	0	0	2	1	0
BUFFALO	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
KANGAROO	1	0	0	1	0	0	0	1	1	1	0	0	2	1	0
GOAT	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
CARP	0	0	1	0	0	1	0	1	1	0	0	1	0	1	1
CHUB	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
CAVY	1	0	0	1	0	0	0	1	1	1	0	0	4	0	1
DEER	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
SWAN	0	1	1	0	1	1	0	0	1	1	0	0	2	1	0
BOAR	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 30 di 112

	FUR	FEATHER	EGGS	MILK	FLYING	AQUATIC	PREDATORY	TEETH	INVERT.	LUNGS	POIS.	FLIP.	LEGS	TAIL	DOM.
LADYBIRD	0	0	1	0	1	0	1	0	0	1	0	0	6	0	0
DOVE	0	1	1	0	1	0	0	0	1	1	0	0	2	1	1
CROW	0	1	1	0	1	0	1	0	1	1	0	0	2	1	0
HAMSTER	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
DOLPHIN	0	0	0	1	0	1	1	1	1	1	0	1	0	1	0
CODFISH	0	0	1	0	0	1	0	1	1	0	0	1	0	1	0
ELEPHANT	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
PHEASANT	0	1	1	0	1	0	0	0	1	1	0	0	2	1	0
FALCON	0	1	1	0	1	0	1	0	1	1	0	0	2	1	0
мотн	1	0	1	0	1	0	0	0	0	1	0	0	6	0	0
FLAMINGO	0	1	1	0	1	0	0	0	1	1	0	0	2	1	0
SEAL	1	0	0	1	0	1	1	1	1	1	0	1	0	0	0
GULL	0	1	1	0	1	1	1	0	1	1	0	0	2	1	0
PRAWN	0	0	1	0	0	1	1	0	0	0	0	0	6	0	0
СНЕЕТАН	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
GIRAFFE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
GORILLA	1	0	0	1	0	0	0	1	1	1	0	0	2	0	0
CRAB	0	0	1	0	0	1	1	0	0	0	0	0	4	0	0
SEAHORSE	0	0	1	0	0	1	0	1	1	0	0	1	0	1	0
KIWI	0	1	1	0	0	0	1	0	1	1	0	0	2	1	0
LION	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
SEA_LION	1	0	0	1	0	1	1	1	1	1	0	1	2	1	0
LEOPARD	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
HARE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
SNAIL	0	0	1	0	0	0	0	0	0	1	0	0	0	0	0
LYNX	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
PIKE	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
WOLF	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
MONGOOSE	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
CAT	1	0	0	1	0	0	1	1	1	1	0	0	4	1	1
MOLLUSK	0	0	1	0	0	0	1	0	0	0	0	0	0	0	0
FLY	1	0	1	0	1	0	0	0	0	1	0	0	6	0	0
MIDGE	0	0	1	0	1	0	0	0	0	1	0	0	6	0	0
	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
	1				0										
	1	0	0	1	0	0	1	1	1	1	0	0	4	0	0
						0		0							
SPARROW	0	1	1	0	1		0		1	1	0	0	2	1	0
STURGEON	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
PERCH	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
SHARK	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
PENGUIN	0	1	1	0	0	1	1	0	1	1	0	0	2	1	0
PIRANHA	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
POLYP	0	0	1	0	0	1	1	0	0	0	0	0	8	0	0
CHICKEN	0	1	1	0	1	0	0	0	1	1	0	0	2	1	1
PONY	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
FLEA	0	0	1	0	0	0	0	0	0	1	0	0	6	0	0
	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
PUMA	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
	0	0	1	0	0	1	1	1	1	1	0	0	4	0	0
					l .	l .	0	1	1	1	0	0	4	1	1
POLECAT	1	0	0	1	0	0	0	-	_	_	_		-		
POLECAT	1 0	0	1	0	0	1	0	1	1	1	0	0	4	0	0
FOLECAT FROG REINDEER TOAD															0

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 31 di 112

ANIMAL	FUR	FEATHER	EGGS	MILK	FLYING	AQUATIC	PREDATORY	TEETH	INVERT.	LUNGS	POIS.	FLIP.	LEGS	TAIL	DOM.
SEA_SNAKE	0	0	0	0	0	1	1	1	1	0	1	0	0	1	0
SOLE	0	0	1	0	0	1	0	1	1	0	0	1	0	1	0
STARFISH	0	0	1	0	0	1	1	0	0	0	0	0	5	0	0
OSTRICH	0	1	1	0	0	0	0	0	1	1	0	0	2	1	0
MOLE	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
TORTOISE	0	0	1	0	0	0	0	0	1	1	0	0	4	1	0
TERMITE	0	0	1	0	0	0	0	0	0	1	0	0	6	0	0
TUNA	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
TRITON	0	0	1	0	0	1	1	1	1	1	0	0	4	1	0
VAMPIRE	1	0	0	1	1	0	0	1	1	1	0	0	2	1	0
WORM	0	0	1	0	0	0	0	0	0	1	0	0	0	0	0
WASP	1	0	1	0	1	0	0	0	0	1	1	0	6	0	0
MINK	1	0	0	1	0	1	1	1	1	1	0	0	4	1	0
CALF	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1

Processing of animals.txt file with KB_CAT

The *animals.txt* file has been processed with the following parameters:

Input File -> animals.txt

Number of Groups (3 - 20) -> 4
Normalization (Max, Std, None) -> M
Start Value of alpha (from 1.8 to 0.9) -> 1.8
End Value of alpha (from 0.5 to 0.0001) -> 0.0001
Decreasing step of alpha (from 0.1 to 0.001) -> 0.001

The processing ended with a Minimum Error of 5.255 at the time 971 producing the results in the files:

Output File Catalog.original animals M g4 out.txt Output File Catalog.sort animals M q4 outsrt.txt Output File Summary sort animals_M_g4_sort.txt Output File Matrix Catal. animals_M_g4_catal.txt Output File Means, STD, CV. animals M g4 medsd.txt Output File CV of the Groups animals M g4 cv.txt **Output File Training Grid** animals_M_g4_grid.txt Output File Run Parameters animals M q4 log.txt

Output file/Catalog.sort ordered by group using animals.txt

Group	ANIMAL	FUR	FEATH.	EGGS	MILK	FLYING	AQUAT.	PRED.	TEETH	VERT.	LUNGS	POIS.	FLIP.	LEGS	TAIL	DOM.
G_00_00	ANTELOPE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
G_00_00	BUFFALO	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
G_00_00	CALF	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
G_00_00	CAT	1	0	0	1	0	0	1	1	1	1	0	0	4	1	1
G_00_00	DEER	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
G_00_00	ELEPHANT	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
G_00_00	FLD_MOUSE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
G_00_00	GIRAFFE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
G_00_00	GOAT	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
G_00_00	HAMSTER	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
G_00_00	HARE	1	0	0	1	0	0	0	1	1	1	0	0	4	1	0
G_00_00	KANGAROO	1	0	0	1	0	0	0	1	1	1	0	0	2	1	0

Group	ANIMAL	FUR	FEATH.	EGGS	MILK	FLYING	AQUAT.	PRED.	TEETH	VERT.	LUNGS	POIS.	FLIP.	LEGS	TAIL	DOM.
G 00 00	PONY	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
	REINDEER	1	0	0	1	0	0	0	1	1	1	0	0	4	1	1
	SQUIRREL	1	0	0	1	0	0	0	1	1	1	0	0	2	1	0
G_00_00		1	0	0	1	1	0	0	1	1	1	0	0	2	1	0
G_00_01	CAVY	1	0	0	1	0	0	0	1	1	1	0	0	4	0	1
G_00_01		1	0	0	1	0	0	0	1	1	1	0	0	2	0	0
	BEE	1	0	1	0	1	0	0	0	0	1	1	0	6	0	1
G_00_03		0	0	1	0	0	1	1	0	0	0	0	0	4	0	0
	FLY	1	0	1	0	1	0	0	0	0	1	0	0	6	0	0
	LADYBIRD	0	0	1	0	1	0	1	0	0	1	0	0	6	0	0
G_00_03		0	0	1	0	0	1	1	0	0	0	0	0	6	0	0
G_00_03		0	0	1	0	1	0	0	0	0	1	0	0	6	0	0
G_00_03		0	0	1	0	0	0	1	0	0	0	0	0	0	0	0
G_00_03		1	0	1	0	1	0	0	0	0	1	0	0	6	0	0
G_00_03		0	0	1	0	0	1	1	0	0	0	0	0	8	0	0
G_00_03		0	0	1	0	0	1	1	0	0	0	0	0	6	0	0
	STARFISH	0	0	1	0	0	1	1	0	0	0	0	0	5	0	0
G_00_03		1	0	1	0	1	0	0	0	0	1	1	0	6	0	0
G_01_00	BEAR	1	0	0	1	0	0	1	1	1	1	0	0	4	0	0
G_01_00		1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00		1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00		1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00		1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00	LYNX	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
	MINK	1	0	0	1	0	1	1	1	1	1	0	0	4	1	0
G_01_00	MOLE	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
		1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
	MONGOOSE	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00		1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00	PUMA	1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00		1	0	0	1	0	0	1	1	1	1	0	0	4	1	0
G_01_00		0	0	0	0	0	0	1	0	0	1	1	0	8	1	0
G_01_02	SCORPION	0	0	1	0	0	0	0	0	0	1	0	0	6	0	0
G_01_03		0	0	1	0	0	0	0	0	0	1	0	0	0	0	0
		0	0	1	0	0	0	0	0	0	1	0	0	6	0	0
G_01_03		0	0	1	0	0	0	0	0	0	1	0	0	0	0	0
G_01_03																
G_02_00		1	0	0	1	0	1	1	1	1	1	0	1	0	0	0
G_02_00		1	0	0	1	0	1	1	1	1	1	0	1	2	1	0
	SEA_LION						1				1			4	1	0
	DUCKBILL	0	0	1	0	0	1	0	1	1	1	0	0	4	0	0
G_02_02		0		1	0	0		0	0	1	1	0	0	4	1	0
	TORTOISE	0	0				1			1	0	0		0	1	
G_03_00			0	1	0	0		0	1				1			1
G_03_00		0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
G_03_00		0	0	1	0	0	1	0	1	1	0	0	1		1	0
G_03_00		0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
G_03_00		0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
G_03_00		0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
G_03_00		0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
	SEAHORSE	0	0	1	0	0	1	0	1	1	0	0	1	0	1	0
	SEA_SNAKE	0	0	0	0	0	1	1	1	1	0	1	0	0	1	0
G_03_00		0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
G_03_00	SOLE	0	0	1	0	0	1	0	1	1	0	0	1	0	1	0

Group	ANIMAL	FUR	FEATH.	EGGS	MILK	FLYING	AQUAT.	PRED.	TEETH	VERT.	LUNGS	POIS.	FLIP.	LEGS	TAIL	DOM.
G_03_00	STURGEON	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
G_03_00	TUNA	0	0	1	0	0	1	1	1	1	0	0	1	0	1	0
G_03_01	FROG	0	0	1	0	0	1	1	1	1	1	0	0	4	0	0
G_03_01	TRITON	0	0	1	0	0	1	1	1	1	1	0	0	4	1	0
G_03_02	GULL	0	1	1	0	1	1	1	0	1	1	0	0	2	1	0
G_03_02	KIWI	0	1	1	0	0	0	1	0	1	1	0	0	2	1	0
G_03_02	PENGUIN	0	1	1	0	0	1	1	0	1	1	0	0	2	1	0
G_03_03	CHICKEN	0	1	1	0	1	0	0	0	1	1	0	0	2	1	1
G_03_03	CROW	0	1	1	0	1	0	1	0	1	1	0	0	2	1	0
G_03_03	DOVE	0	1	1	0	1	0	0	0	1	1	0	0	2	1	1
G_03_03	DUCK	0	1	1	0	1	1	0	0	1	1	0	0	2	1	0
G_03_03	FALCON	0	1	1	0	1	0	1	0	1	1	0	0	2	1	0
G_03_03	FLAMINGO	0	1	1	0	1	0	0	0	1	1	0	0	2	1	0
G_03_03	HAWK	0	1	1	0	1	0	1	0	1	1	0	0	2	1	0
G_03_03	OSTRICH	0	1	1	0	0	0	0	0	1	1	0	0	2	1	0
G_03_03	PHEASANT	0	1	1	0	1	0	0	0	1	1	0	0	2	1	0
G_03_03	SKYLARK	0	1	1	0	1	0	0	0	1	1	0	0	2	1	0
G_03_03	SPARROW	0	1	1	0	1	0	0	0	1	1	0	0	2	1	0
G_03_03	SWAN	0	1	1	0	1	1	0	0	1	1	0	0	2	1	0

Output file/CV ordered by group using animals.txt

Groups	FUR	FEATHER	EGGS	MILK	FLYING	AQUATIC	PRED.	TEETH	VERT.	LUNGS	POIS.	FLIP.	LEGS	TAIL	DOMEST.	*Mean*	N_recs
G_00_00	0	0	0	0	3,87	0	3,87	0	0	0	0	0	0,22	0	1,29	0,62	16
G_00_01	0	0	0	0	0	0	0	0	0	0	0	0	0,33	0	1	0,09	2
G_00_02	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
G_00_03	1,63	0	0	0	1,1	1,1	0,76	0	0	1,1	3,16	0	0,36	0	0	0,61	11
G_01_00	0	0	0	0	0	3,46	0	0	0	0	0	0	0	0,29	0	0,25	13
G_01_02	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
G_01_03	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0,07	4
G_02_00	0,71	0	0	0	0	0	0	0	0	0	0	0	1,41	0,71	0	0,19	3
G_02_01	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
G_02_02	0	0	0	0	0	1	0	1	0	0	0	0	0	1	0	0,2	2
G_03_00	0	0	0,29	0	0	0	0,67	0	0	0	3,46	0,29	0	0	3,46	0,54	13
G_03_01	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0,07	2
G_03_02	0	0	0	0	1,41	0,71	0	0	0	0	0	0	0	0	0	0,14	3
G_03_03	0	0	0	0	0,3	2,24	1,73	0	0	0	0	0	0	0	2,24	0,43	12
Means	0,24	0	0,04	0	0,97	1,05	1,19	0,02	0	0,14	0,95	0,04	0,19	0,12	1,13	0,41	84
Total	2,69	5,25	2,12	2,9	4,53	3,29	2,39	2,02	1,23	1,32	10,95	5,25	1,68	1,46	6,31	3,57	84

In this example you can see the importance of the difference of the CV of the whole file (CV_Pop 3,57) before the cataloging as opposed to the means of the CV of the groups (CV_Med 0,41).

The cataloging has produced an improvement of 88,5%.

Further confirmation derives from the presence of many zero values in the cells in the table, values which indicate that in that variable / column of the group there is just one value that is constantly repeated: the variable / column (for that constant value) is certainly important for the cataloging of records in that group.

Verify the validity of a manual cataloging

Sometimes it can be useful to certify the validity of a manual cataloging using KB_CAT.

In the example described below the variable / column D1 contains the input code of KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 34 di 112

the animal specie (bird, mammal, insect, fish, ...); the variable / column was then duplicated in another two columns (D2 e D3).

The modification of the original file has two different consequences:

- it reinforces the importance of the code of the animal specie in respect to all the other variables / columns
- the new structure of the file allows KB_CAT to process the data in a similar way to algorithms typical of supervised networks with different input variables / columns and a variable / column objective (in our case D1, D2, D3 considered globally)

Input file to KB_CAT (animals d.txt)

FEATHE EGG											
ANIMAL	FUR	R	S	MIL	K XXX	D1	D2	D3			
SKYLARK		0	1	1	0	bird	bird	Bird			
DUCK		0	1	1	0	bird	bird	Bird			
ANTELOPE		1	0	0	1	mammal	mammal	Mammal			
BEE		1	0	1	0	insect	insect	Insect			
LOBSTER		0	0	1	0	shellfish	shellfish	Shellfish			
HERRING		0	0	1	0	fish	fish	Fish			
FIELD_MOUS											
E		1	0	0	1	mammal	mammal	Mammal			
								bird_of_pre			
HAWK		0	1	1	0	bird_of_prey	bird_of_prey	у — —			
BUFFALO		1	0	0	1	mammal	mammal	Mammal			
KANGAROO		1	0	0	1	mammal	mammal	Mammal			
GOAT		1	0	0	1	mammal	mammal	Mammal			
CARP		0	0	1	0	fish	fish	Fish			
CHUB		0	0	1	0	fish	fish	Fish			
CAVY		1	0	0	1	mammal	mammal	Mammal			
DEER		1	0	0	1	mammal	mammal	Mammal			
SWAN		0	1	1	0	bird	bird	Bird			
BOAR		1	0	0	1	mammal	mammal	Mammal			
LADYBIRD		0	0	1	0	insect	insect	Insect			
DOVE		0	1	1	0	bird	bird	Bird			
CROW		0	1	1	0	bird	bird	Bird			
HAMSTER		1	0	0	1	mammal	mammal	Mammal			
DOLPHIN		0	0	0	1	mammal	mammal	Mammal			
CODFISH		0	0	1	0	fish	fish	Fish			
ELEPHANT		1	0	0	1	mammal	mammal	Mammal			
PHEASANT		0	1	1	0	bird	bird	Bird			
								bird_of_pre			
FALCON		0	1	1	0	bird_of_prey	bird_of_prey	у			
MOTH		1	0	1	0	insect	insect	Insect			
FLAMINGO		0	1	1	0	bird	bird	Bird			
SEAL		1	0	0	1	mammal	mammal	Mammal			
GULL		0	1	1	0	bird	bird	Bird			
PRAWN		0	0	1	0	shellfish	shellfish	Shellfish			
CHEETAH		1	0	0	1	mammal	mammal	Mammal			
GIRAFFE		1	0	0	1	mammal	mammal	Mammal			
GORILLA		1	0	0	1	mammal	mammal	Mammal			
CRAB		0	0	1	0	shellfish	shellfish	Shellfish			
SEAHORSE		0	0	1	0	fish	fish	Fish			

FEATHE EGG MILK XXX **ANIMAL FUR** R S D1 D₂ D3 **KIWI** 0 1 1 0 bird bird Bird 1 0 1 LION 0 Mammal mammal mammal SEA LION 1 0 0 1 Mammal mammal mammal **LEOPARD** 1 0 0 1 Mammal mammal mammal **HARE** 1 0 0 1 Mammal mammal mammal Invertebrat 0 0 **SNAIL** 1 0 invertebrate invertebrate 0 0 LYNX 1 1 mammal mammal Mammal 0 0 PIKE 1 0 fish fish Fish 0 0 WOLF 1 1 Mammal mammal mammal **MONGOOSE** 1 0 0 1 Mammal mammal mammal CAT 1 0 0 1 mammal mammal Mammal Invertebrat **MOLLUSK** 0 0 1 0 invertebrate invertebrate e FLY 1 0 1 0 insect insect Insect 0 0 1 0 MIDGE insect insect Insect **OPOSSUM** 1 0 0 1 mammal mammal Mammal 1 0 1 1 **DUCKBILL** mammal Mammal mammal **BEAR** 1 0 0 1 Mammal mammal mammal **SPARROW** 0 1 1 0 bird bird Bird 0 0 1 0 Fish STURGEON fish fish **PERCH** 0 0 1 0 fish fish Fish SHARK 0 0 1 0 Fish fish fish 0 0 PENGUIN 1 1 bird bird Bird **PIRANHA** 0 0 1 0 fish Fish fish Invertebrat **POLYP** 0 0 1 0 invertebrate invertebrate е **CHICKEN** 0 1 1 0 bird bird Bird **PONY** 1 0 0 1 mammal mammal Mammal 0 0 1 **FLEA** 0 insect insect Insect 0 0 **PUMA** 1 1 mammal mammal Mammal **POLECAT** 1 0 0 1 Mammal mammal mammal 0 0 1 0 **FROG** amphibian amphibian Amphibian 1 0 0 1 REINDEER mammal mammal Mammal 0 0 0 TOAD 1 amphibian amphibian Amphibian **SQUIRREL** 1 0 0 1 mammal Mammal mammal **SCORPION** 0 0 0 0 arachinida arachinida Arachinida SEA SNAKE 0 0 0 0 Reptiles reptiles reptiles SOLE 0 0 1 0 fish fish Fish **Echinoder STARFISH** 0 0 1 0 echinoderm echinoderm m **OSTRICH** 0 1 1 0 bird bird Bird 1 0 0 1 Mammal MOLE mammal mammal **TORTOISE** 0 0 1 0 reptiles reptiles Reptiles **TERMITE** 0 0 1 0 insect insect Insect 0 0 1 0 **TUNA** fish Fish fish **TRITON** 0 0 1 0 amphibian amphibian **Amphibian VAMPIRE** 1 0 0 1 mammal mammal Mammal Invertebrat **WORM** 0 0 1 0 invertebrate invertebrate е **WASP** 1 0 1 0 insect Insect insect MINK 1 0 0 1 mammal mammal Mammal

FEATHE EGG

ANIMAL FUR R S MILK XXX D1 D2 D3 CALF 1 0 0 1 mammal mammal Mammal

The processing was carried out with the following parameters:

```
animals_d.txt
Input File
Number of Groups (3 - 20,
Normalization (Max, Std, None)
Start Value of alpha (from 1.8 to 0.9)
End Value of alpha (from 0.5 to 0.001)
Decreasing step of alpha (from 0.1 to 0.001)
Number of Groups (3 - 20)
                                                                                  ->
                                                                                        4
                                                                                 ->
                                                                                       M
                                                                                 ->
                                                                                       1.8
                                                                                       0.0001
                                                                                       0.001
                                                      animals d M g4 out.txt
animals d M g4 outsrt.txt
animals d M g4 sort.txt
animals d M g4 catal.txt
animals d M g4 medsd.txt
animals d M g4 cv.txt
animals d M g4 grid.txt
animals d M g4 grid.txt
Output File Catalog.original Output File Catalog.sort
Output File Summary sort
Output File Matrix Catal.
Output File Means, STD, CV.
Output File CV of the Groups
Output File Training Grid
Output File Run Parameters
                                                       animals_d_M_g4_log.txt
```

Results obtained processing animals_d.txt (Output/Catalog.sort)

Group	ANIMAL	FUR	FEATHER		MILK	XXX	D1	D2	D3
G_00_00	BEE	1.0	0.0	1.0	0.0		insect	insect	Insect
G_00_00	CRAB	0.0	0.0	1.0	0.0		shellfish	shellfish	Shellfish
G_00_00	FLY	1.0	0.0	1.0	0.0		insect	insect	Insect
G_00_00	LADYBIRD	0.0	0.0	1.0	0.0		insect	insect	Insect
G_00_00	LOBSTER	0.0	0.0	1.0	0.0		shellfish	shellfish	Shellfish
G_00_00	MIDGE	0.0	0.0	1.0	0.0		insect	insect	Insect
G_00_00	MOLLUSK Processing the contract of the contrac	0.0	0.0	1.0	0.0		invertebrate	invertebrate	Invertebrate
G_00_00	MOTH	1.0	0.0	1.0	0.0		insect	insect	Insect
G_00_00	POLYP	0.0	0.0	1.0	0.0		invertebrate	invertebrate	Invertebrate
G_00_00	PRAWN	0.0	0.0	1.0	0.0		shellfish	shellfish	Shellfish
G_00_00	SNAIL SNAIL	0.0	0.0	1.0	0.0		invertebrate	invertebrate	Invertebrate
G_00_00	WASP	1.0	0.0	1.0	0.0		insect	insect	Insect
G_00_00	WORM	0.0	0.0	1.0	0.0		invertebrate	invertebrate	Invertebrate
G_00_01	FLEA	0.0	0.0	1.0	0.0		insect	insect	Insect
G_00_01	STARFISH	0.0	0.0	1.0	0.0		echinoderm	echinoderm	Echinoderm
G_00_01	TERMITE	0.0	0.0	1.0	0.0		insect	insect	Insect
G_00_03	CHICKEN	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	CROW	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	DOVE	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	DUCK	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	FALCON	0.0	1.0	1.0	0.0		bird_of_prey	bird_of_prey	bird_of_prey
G_00_03	FLAMINGO	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	HAWK	0.0	1.0	1.0	0.0		bird_of_prey	bird_of_prey	bird_of_prey
G_00_03	OSTRICH	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	PHEASANT	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	SKYLARK	0.0	1.0	1.0	0.0		bird	bird	bird
G_00_03	SPARROW	0.0	1.0	1.0	0.0		bird	bird	bird
G 00 03	SWAN	0.0	1.0	1.0	0.0		bird	bird	bird
G_01_01	TORTOISE	0.0	0.0	1.0	0.0		reptiles	reptiles	reptiles
G 01 02	SCORPION	0.0	0.0	0.0	0.0		arachinida	arachinida	arachinida
G_01_02		0.0	0.0	1.0	0.0		amphibian	amphibian	amphibian
G_01_03		0.0	1.0	1.0	0.0		bird	bird	bird
G_01_03		0.0	1.0	1.0	0.0		bird	bird	bird
G 01 03		0.0	1.0	1.0	0.0		bird	bird	bird

Group ANIMAL	FUR	FEATHER	EGGS	MILK	XXX	D1	D2	D3
G 02 00 ANTELOPE	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G 02 00 BUFFALO	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G 02 00 DEER	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G 02 00 ELEPHANT	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_02_00 FIELD_MOUSE	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G 02 00 GIRAFFE	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_02_00 GORILLA	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_02_00 HARE	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_02_00 KANGAROO	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_02_00 SQUIRREL	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_02_00 VAMPIRE	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_02_02 DUCKBILL	1.0	0.0	1.0	1.0		mammal	mammal	mammal
G_02_03 FROG	0.0	0.0	1.0	0.0		amphibian	amphibian	amphibian
G_02_03 TRITON	0.0	0.0	1.0	0.0		amphibian	amphibian	amphibian
G_03_00 CALF	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_00 CAT	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_00 CAVY	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_00 GOAT	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_00 HAMSTER	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_00 PONY	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_00 REINDEER	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 BEAR	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 BOAR	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 CHEETAH	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 LEOPARD	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 LION	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 LYNX	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 MINK	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 MOLE	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 MONGOOSE	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 OPOSSUM	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 POLECAT	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 PUMA	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_01 WOLF	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_02 DOLPHIN	0.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_02 SEAL	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_02 SEA_LION	1.0	0.0	0.0	1.0		mammal	mammal	mammal
G_03_02 SEA_SNAKE	0.0	0.0	0.0	0.0		reptiles	reptiles	reptiles
G_03_03 CARP	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 CHUB	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 CODFISH	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 HERRING	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 PERCH	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 PIKE	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 PIRANHA	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 SEAHORSE	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 SHARK G 03 03 SOLE	0.0	0.0 0.0	1.0 1.0	0.0		fish fish	fish fish	fish fish
G_03_03 STURGEON	0.0	0.0	1.0	0.0		fish	fish fish	fish
G_03_03 TUNA	0.0	0.0	1.0	0.0		fish	fish	fish
G_03_03 TONA	0.0	0.0	1.0	0.0		11011	11011	11011

The manual cataloging is confirmed except for the coloured records of the groups G_00_00, G_00_01, G_01_02 and G_03_02.

The questions that the researchers could ask in analogue cases, but which are much more complicated and important in real life of companies and organisations, could be:

- do errors occur during the gathering of data?
- do errors exist while inserting data?
- do mutations occur inside the groups?
- are there defects in production / working?
- Is there a lack in the design of the manual classification?
- Is it necessary to introduce other variables / columns?

Comparison of the results of the automatic cataloging of *iris.txt* to those recognized by botanists

The reliability of the algorithms of neural networks is often appreciated by the coincidence between the automatic cataloging of files *iris.txt* (150 records) and that carried out by botanists.

KB_CAT with a cataloging into 3 groups was made with the following results.

Group RecId	Sepal_Length	Sepal_Width	Petal_Length	Petal_Width
G_00_00 versicolor100	5.7	2.8	4.1	1.3
G_00_00 versicolor54	5.5	2.3	4.0	1.3
G_00_00 versicolor56	5.7	2.8	4.5	1.3
G_00_00 versicolor60	5.2	2.7	3.9	1.4
G_00_00 versicolor63	6.0	2.2	4.0	1.0
G 00 00 versicolor69	6.2	2.2	4.5	1.5
G 00 00 versicolor70	5.6	2.5	3.9	1.1
G 00 00 versicolor72	6.1	2.8	4.0	1.3
G 00 00 versicolor73	6.3	2.5	4.9	1.5
G 00 00 versicolor81	5.5	2.4	3.8	1.1
G 00 00 versicolor83	5.8	2.7	3.9	1.2
G 00 00 versicolor85	5.4	3.0	4.5	1.5
G 00 00 versicolor88	6.3	2.3	4.4	1.3
G 00 00 versicolor90	5.5	2.5	4.0	1.3
G 00 00 versicolor91	5.5	2.6	4.4	1.2
G 00 00 versicolor93	5.8	2.6	4.0	1.2
G 00 00 versicolor95	5.6	2.7	4.2	1.3
G 00 00 versicolor97	5.7	2.9	4.2	1.3
G 00 00 virginica107	4.9	2.5	4.5	1.7
G 00 00 virginica120	6.0	2.2	5.0	1.5
G 00 01 versicolor84	6.0	2.7	5.1	1.6
G 00 01 virginica102	5.8	2.7	5.1	1.9
G 00 01 virginical12	6.4	2.7	5.3	1.9
G 00 01 virginical14	5.7	2.5	5.0	2.0
G 00 01 virginica122	5.6	2.8	4.9	2.0
G 00 01 virginica124	6.3	2.7	4.9	1.8
G 00 01 virginica127	6.2	2.8	4.8	1.8
G 00 01 virginica128	6.1	3.0	4.9	1.8
G 00 01 virginica135	6.1	2.6	5.6	1.4
G 00 01 virginica139	6.0	3.0	4.8	1.8
G 00 01 virginica143	5.8	2.7	5.1	1.9
G 00 01 virginica147	6.3	2.5	5.0	1.9
G 00 01 virginica150	5.9	3.0	5.1	1.8
G 00 02 virginica101	6.3	3.3	6.0	2.5
G 00 02 virginica103	7.1	3.0	5.9	2.1
G 00 02 virginica105	6.5	3.0	5.8	2.2
G 00 02 virginica106	7.6	3.0	6.6	2.1
G 00 02 virginica108	7.3	2.9	6.3	1.8
	•	= · •	-	- · •

Group RecId	Conal Longth	Sonal Width	Petal Length	Dotal Width
G 00 02 virginica109	6.7	2.5	5.8	1.8
G 00 02 virginical10	7.2	3.6	6.1	2.5
G 00 02 virginicall3	6.8	3.0	5.5	2.1
_ _			5.1	
G_00_02 virginica115	5.8	2.8		2.4
G_00_02 virginica116	6.4	3.2	5.3	2.3
G_00_02 virginical18	7.7	3.8	6.7	2.2
G_00_02 virginical19	7.7	2.6	6.9	2.3
G_00_02 virginica121	6.9	3.2	5.7	2.3
G_00_02 virginica123	7.7	2.8	6.7	2.0
G_00_02 virginica125	6.7	3.3	5.7	2.1
G_00_02 virginica126	7.2	3.2	6.0	1.8
G_00_02 virginica129	6.4	2.8	5.6	2.1
G_00_02 virginica131	7.4	2.8	6.1	1.9
G 00 02 virginica132	7.9	3.8	6.4	2.0
G 00 02 virginica133	6.4	2.8	5.6	2.2
G 00 02 virginica136	7.7	3.0	6.1	2.3
G 00 02 virginica137	6.3	3.4	5.6	2.4
G 00 02 virginica140	6.9	3.1	5.4	2.1
G 00 02 virginica141	6.7	3.1	5.6	2.4
G 00 02 virginica142	6.9	3.1	5.1	2.3
G 00 02 virginica144	6.8	3.2	5.9	2.3
G 00 02 virginica145	6.7	3.3	5.7	2.5
	6.7		5.2	2.3
G_00_02 virginica146		3.0		
G_00_02 virginica148	6.5	3.0	5.2	2.0
G_00_02 virginica149	6.2	3.4	5.4	2.3
G_01_00 versicolor58	4.9	2.4	3.3	1.0
G_01_00 versicolor61	5.0	2.0	3.5	1.0
G_01_00 versicolor65	5.6	2.9	3.6	1.3
G_01_00 versicolor68	5.8	2.7	4.1	1.0
G_01_00 versicolor80	5.7	2.6	3.5	1.0
G_01_00 versicolor82	5.5	2.4	3.7	1.0
G_01_00 versicolor89	5.6	3.0	4.1	1.3
G_01_00 versicolor94	5.0	2.3	3.3	1.0
G 01 00 versicolor96	5.7	3.0	4.2	1.2
G 01 00 versicolor99	5.1	2.5	3.0	1.1
G 01 01 versicolor62	5.9	3.0	4.2	1.5
G 01 01 versicolor64	6.1	2.9	4.7	1.4
G 01 01 versicolor67		3.0	4.5	1.5
G 01 01 versicolor71		3.2	4.8	1.8
G 01 01 versicolor79		2.9	4.5	1.5
G 01 01 versicolor86		3.4	4.5	1.6
G 01 01 versicolor92	6.1	3.0	4.6	1.4
G 01 02 versicolor78	6.7	3.0	5.0	1.7
G 01 02 virginica104		2.9	5.6	1.8
G 01 02 virginical11	6.5	3.2	5.1	2.0
G 01 02 virginical17		3.0	5.5	1.8
— —				
G_01_02 virginica130		3.0	5.8	1.6
G_01_02 virginica138		3.1	5.5	1.8
G_02_00 setosa1	5.1	3.5	1.4	0.2
G_02_00 setosa10	4.9	3.1	1.5	0.1
G_02_00 setosa11	5.4	3.7	1.5	0.2
G_02_00 setosa12	4.8	3.4	1.6	0.2
G_02_00 setosa13	4.8	3.0	1.4	0.1
G_02_00 setosa14	4.3	3.0	1.1	0.1
G_02_00 setosa15	5.8	4.0	1.2	0.2
G_02_00 setosa16	5.7	4.4	1.5	0.4
G_02_00 setosa17	5.4	3.9	1.3	0.4

Group RecId	Sepal Length	Sepal Width	Petal Length	Petal Width
G 02 00 setosa18	5.1	3.5	1.4	0.3
G 02 00 setosa19	5.7	3.8	1.7	0.3
G 02 00 setosa2	4.9	3.0	1.4	0.2
G 02 00 setosa20	5.1	3.8	1.5	0.3
G 02 00 setosa21	5.4	3.4	1.7	0.2
G 02 00 setosa22	5.1	3.7	1.5	0.4
G 02 00 setosa23	4.6	3.6	1.0	0.2
 G_02_00 setosa24	5.1	3.3	1.7	0.5
G 02 00 setosa25	4.8	3.4	1.9	0.2
G 02 00 setosa26	5.0	3.0	1.6	0.2
G 02 00 setosa27	5.0	3.4	1.6	0.4
G 02 00 setosa28	5.2	3.5	1.5	0.2
G 02 00 setosa29	5.2	3.4	1.4	0.2
G 02 00 setosa3	4.7	3.2	1.3	0.2
G 02 00 setosa30	4.7	3.2	1.6	0.2
G 02 00 setosa31	4.8	3.1	1.6	0.2
G 02 00 setosa32	5.4	3.4	1.5	0.4
G 02 00 setosa33	5.2	4.1	1.5	0.1
G 02 00 setosa34	5.5	4.2	1.4	0.2
G 02 00 setosa35	4.9	3.1	1.5	0.2
G 02 00 setosa36	5.0	3.2	1.2	0.2
G 02 00 setosa37	5.5	3.5	1.3	0.2
G 02 00 setosa38	4.9	3.6	1.4	0.1
G_02_00 setosa39	4.4	3.0	1.3	0.2
G 02 00 setosa4	4.6	3.1	1.5	0.2
G_02_00 setosa40	5.1	3.4	1.5	0.2
G_02_00 setosa40 G_02_00 setosa41	5.0	3.5	1.3	0.3
G_02_00 setosa41 G_02_00 setosa42	4.5	2.3	1.3	0.3
G_02_00 setosa42 G_02_00 setosa43	4.4	3.2	1.3	0.2
G_02_00 setosa43 G_02_00 setosa44	5.0			
G_02_00 setosa44 G 02 00 setosa45		3.5	1.6	0.6
G_02_00 setosa45 G 02 00 setosa46	5.1	3.8	1.9	0.4
	4.8	3.0	1.4	0.3
G_02_00 setosa47	5.1	3.8	1.6	0.2
G_02_00 setosa48	4.6	3.2	1.4	0.2
G_02_00 setosa49	5.3	3.7	1.5	0.2
G_02_00 setosa5	5.0	3.6	1.4	0.2
G_02_00 setosa50	5.0	3.3	1.4	0.2
G_02_00 setosa6	5.4	3.9	1.7	0.4
G_02_00 setosa7	4.6	3.4	1.4	0.3
G_02_00 setosa8	5.0	3.4	1.5	0.2
G_02_00 setosa9	4.4	2.9	1.4	0.2
G_02_02 versicolor51		3.2	4.7	1.4
G_02_02 versicolor52		3.2	4.5	1.5
G_02_02 versicolor53		3.1	4.9	1.5
G_02_02 versicolor55		2.8	4.6	1.5
G_02_02 versicolor57		3.3	4.7	1.6
G_02_02 versicolor59		2.9	4.6	1.3
G_02_02 versicolor66		3.1	4.4	1.4
G_02_02 versicolor74		2.8	4.7	1.2
G_02_02 versicolor75		2.9	4.3	1.3
G_02_02 versicolor76		3.0	4.4	1.4
G_02_02 versicolor77		2.8	4.8	1.4
G_02_02 versicolor87		3.1	4.7	1.5
G_02_02 versicolor98		2.9	4.3	1.3
G_02_02 virginica134	6.3	2.8	5.1	1.5

With the exception of records highlighted in yellow, the automatic KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 41 di 112

cataloging has confirmed the botanists's opinion (inserted in the column *RecId*), reaching a high value of the *Knowledge Index* (0.9311).

Clinical trials on hepatitis B virus

In 2006 KB_CAT was used to process data from an important research on clinical trials regarding 1414 subjects with 33 variables / columns.

The research concerned the hepatitis B virus, the characteristics of typical carriers, of symptomless carriers, those with low viral repetition, the possible evolution of the virus in other pathologies, the identification *marker*, the diagnosis and the treatment.

The problem of identification consists in determining whether a person has the characteristics enough to associate him with a group of carriers of the virus.

The variables / columns that concern the generalities of the subjects, such as age, race, weight, height, the area of birth, and residency were omitted as they resulted to be of little importance in previous runs.

The weight and height of the subjects would be misleading if adopted separately. For this reason it was calculated the body mass index (*BMI*) that connects the two attributes, through the relationship

$$BMI = kg/(m)^2$$

In addition, it was also calculated an index, particularly significant that correlates weight, height, gender and age of the subject. The real index of weight, which also takes account of the muscles and the body of the subjects is in fact the percentage of fat mass (*FAT*). There are different formulas, very similar to each other, to calculate this index. In this case it was considered the formula of *Deurenberg* that, most of the other, takes into account the age of the subject:

Regarding the variables related to the potus, indicating the number of glasses of wine / beer / spirits drunk daily and how long the subjects were drinking, were calculated the alcohol units as the product of the two values. Consequently, it was eliminated field indicating whether or not the subject was a teetotaler.

The input file, after the revision, it is shown in the following table:

Generalities	Age, Gender, BMI, FAT, Case
Potus	Total-UA (Total alcoholic Units)
Diagnosis	Diagnosis, Steatohepatitis, Steatosis
Therapy	PreInterferone, PreLamivudina, PreAdefovir,
	PrePegInterferone, PreEntecavir, PreTecnofovir,
	Interferone, Peginterferone, Lamivudina, Adefovir,
	Tecnofovir

Laboratory trials	AST, ALT, HBeAg, AntiHBe, AntiHBclgM, HBVDNAgualitative, HBVDNAguantitative,
	GentipoHBV, AntiDelta, HIV, AntiHCV,
	HCVRNAQualitative, GentipoHCV

The most significant characteristics of the more numerous groups are contained in the following table.

Group	M/F	FAT %	Case	UA	Diagnosis	Adefo-vir	AST/ALT	HBeAg	AntiHBclgM
1_01	М	11-43	Prevalent	73000	Chronic Hepatitis	No	high	Negative	Negative
1_03	М	1743	Prevalent	29200	Chronic Hepatitis	No	medium	Negative	Not researched
2_01	М	1636	Prevalent	45000	Carrier in non repeating phase	No	normal	Negative	Negative
2_08	F	2356	Prevalent	9000	Chronic Hepatitis and carrier in non repeating phase	No	normal	Negative	Negative
3_01	М	2837	Prevalent	282800	HCC and Cirrhosis	No	high	Negative	Negative
3_04	М	1434	Incidental	64000	Chronic Hepatitis and carrier in non repeating phase	No	high	Negative	Not researched
4_01	M	2838	Prevalent	73000	HCC and Cirrhosis	No	high		Positive / not researched
8_01	М	01/12/47	Prevalent	45600	Chronic Hepatitis	No	high	Positive	Negative
8_08	М	1737	Prevalent	73000	Cirrhosis	yes	medium	Negative	Negative

From a comprehensive analysis of the results obtained, not published in detail, the following conclusions have emerged:

- women drink less than men, and consequently suffer from hepatitis rather than from cirrhosis
- only men over the age of 50 years show a diagnosis of HCC, such persons do not have steatosis and the steatohepatitis positive does not take a very high level
- carriers of the virus in the *non repetitive phase* show normal values in AST and ALT laboratory exams
- HIV positive are almost exclusively men with chronic hepatitis
- the diagnosis of *cirrhosis* is only present in subjects who are over the age of 40 years
- the carriers in *non repetitive phase* have a percentage of *incident cases* greater than the other diagnoses
- subjects who have been diagnosed HCC are nearly always prevalent cases
- the highest values ALT and AST regard the subjects who have been diagnosed with cirrhosis or HCC
- nearly all the women in the *non repetitive phase* result teetotaler: this is not the case for the men
- the lowest values of FAT concerns men who have been diagnosed with chronic hepatitis and show values of units of alcohol reduced

Not good, but better! (2006 mail)

Dear Roberto.

it really does seem that your software is a winner.

As I anticipated, I have sent the conclusions of the E. to P. A. (hepatologist

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 43 di 112

and top-publisher, who is native of Milan but works in Palermo) also because he was one of the owners of the database which was used and especially because it was him, with his data who carried out the more detailed clinical/statistical analysis.

He phoned me this morning asking which doctor had contributed with analysing the database and writing the report. When I explained that neither you nor E. Are doctors and that the conclusions were drawn with your software, he didn't want to believe me.

All (and I mean ALL) the conclusions are correct and coincide with those taken from statistical analysis and from the clinical observations.

I explained that the software is in the process of being verified and that his collaboration would be useful in this.

As it would be a case of verifying that the conclusions of the software are parallel to those of the clinical/statistical analysis and in other databases and once this has been done, an informative/clinical publication could be released to verify your application at least in this field.

He would be pleased to collaborate with you in this way and will start by sending you one or two clinical databases which have already been analysed (from 800 to 8.000 cases) on which the small KB will be tested. A big pat on the back and congratulations!

KB_STA - the statistical analysis of the result of the cataloging

Generalities, aims and functions

The aim of KB_STA is to help researchers in the analysis of the results of processing.

KB_STA has proven to be indispensable when the input file is of a large size, either in the amount of records or the number of variables / columns.

A purely visual exam of records in each group would be difficult resulting in a lot of hard work highlighting the need to subject the groups to costly external analysis, complex and with questionable results.

KB_STA resolves the problem of the black box which is typical of algorithms of neural networks.

KB STA:

- submit the file of CV groups to statistical analysis
- evaluates the degree of homogeneity of the groups within them
- evaluates the importance of the variables / columns in cataloging the records in the groups
- groups the records in each group for each variable / column in quartiles (if numeric) or frequency tables (if text values)
- if required, shows for each group and for each variable / column the original value of input records

Source of KB_STA (see attachment 2)

How to use

Having the kb_sta.py program and the input file to process already in the folder, run KB STA by typing in the window:

python kb_sta.py

where with **python** you ask the **kb sta.py** to be run (in python language)

The program begins the processing by asking in succession:

Catalogued Records File (_outsrt.txt) : vessels_M_g3_outsrt.txt

vessels_M_g3_outsrt.txt is the file in txt format containing the table of records / cases cataloged and arranged in group_code sequence.

The file *vessels_M_g3_outsrt.txt* is one of the results of the previous processing with the program KB CAT.

Groups / CV File (_cv.txt) : vessels_M_g3_cv.txt

vessels_M_g3_cv.txt is the file in *txt* format containing the table of the CV of the groups.

The file *vessels_M_g3_cv.txt* is one of the results from the previous processing with the KB CAT program.

It is important that this file and the previous one come from the same KB_CAT processing.

Report File (output) : vessels_M_g3_report.txt

vessels_M_g3_report.txt is the output file that will contain the statistical analysis of the results obtained from the previous program of cataloging.

It is useful, for clarity, that the name of the report file beginnings as the two previous, as just exemplified above, in the case of statistical analysis with different parameters, the names may change in the final part of the name (example _r1, _r2, _r3).

Display Input Records (Y / N) : n

Group Consistency (% from 0 to 100) : 0

Parameter to request the display of the groups with a percentage of homogeneity inside them not less than that indicated, it is advisable to carry out the initial processing with the parameter set equal to zero, which would show all groups, and then use a different parameter in relation to the results achieved.

Too high a value of this parameter may produce an empty list.

Variable Consistency (% from 0 to 100) : 0

Parameter to request the display of variables, within groups with a percentage of homogeneity of the variable is not less than that indicated, it is advisable to carry out the initial processing with the parameter set equal to zero, which would show all the variables of the groups, and then use a different parameter in relation to the results

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 45 di 112

obtained.

Too high a value of this parameter may produce an empty list.

Select groups containing records >= : 2

The parameter to request the visualization of the groups composed of at least x records. The groups formed by a single record are automatically homogeneous at 100% for all the variables / columns.

Select groups containing records <= : 1000

The parameter to request the display of groups composed of a number of records less than x. The parameter can be useful for examining the groups containing only a few records.

Summary / Detail report (S / D) : d

If the parameter has the value of s/S, the report will contain the values of homogeneity (consistency), the total number of records and the percentage of records cataloged in the group.

If the parameter has a value of d/D, the report will contain numerical values for each quartile, while for each text variable the report will contain the frequency distribution of the different text values.

Display Input Records (Y / N) : n

If the parameter has the value of n/N the input records belonging to the groups will not be visualized, on the contrary (y/Y) will be visualized.

KB_STA running

Analysis of the results of the cataloging of *vessels.txt* Example for the group G_00_00

```
G_00_00 Consistency 0.7140 %Consistency 79 Records 7 %Records 17.50 *** shape Consistency 3.22 Value cylinder_cone Frequency 3 Percentage 42.00 Value ball_cone Frequency 2 Percentage 28.00 Value cylinder Frequency 1 Percentage 14.00 Value cut_cone Frequency 1 Percentage 14.00 *** material Consistency 0.7687 %Consistency 0.00 Value glass Frequency 5 Percentage 71.00 Value terracotta Frequency 1 Percentage 14.00 Value metal Frequency 1 Percentage 14.00
```

*** height	Consistency	0.4537	%Cons	istency	36.46
Mean 53.57 Mi	n 30.00 Max	100.00	Step 17.5	0	
<mark>First Quartile</mark>	(end) 47.5	0 Frequency	% 57	<mark>.14</mark>	
Second Quartile	(end) 65.0	0 Frequency	% 14	<mark>.29</mark>	
Third Quartile	(end) 82.5	0 Frequency	% 14	.29	
Fourth Quartile	(end) 100.0	0 Frequency	% 14	.29	
*** colour	Consistency	0.2673	%Cons	istency	62.56
Value green	Frequency	5	Percentage	71.00	
Value grey	Frequency	1	Percentage	14.00	
Value brown	Frequency	1	Percentage	14.00	
*** weight	Consistency	1.9116	%Cons	istency	0.00
Mean 2680.71	Min 120.00	Max 15000	0.00 Step	3720.00	
<mark>First Quartile</mark>	(end) 3840.0		85	<mark>.71</mark>	
Fourth Quartile	(end) 15000.0	0 Frequency	% 14	.29	
*** haft	Consistency			istency	100.00
Value no	Frequency	7	Percentage	100.00	
*** plug	Consistency	0.9055	%Cons	istency	0.00
Value cork	Frequency	3	Percentage	42.00	
Value no	Frequency	2	Percentage	28.00	
Value metal	Frequency	2	Percentage	28.00	
==========		=========	========		

The group G_00_00 is composed predominantly of *glass* containers, with a *height* from 30 to 65 cm, *green* color, with a *weight* up to 3840 and without handle (*haft*).

Analysis of the results of a political poll of 2007

The analysed case takes into consideration a political poll, carried out on 22nd and 23rd November 2007 by Prof. Paolo Natale of the State University of Milan, Department of Political Sciences. After an initial processing of the data which gave no evident results, the database was updated eliminating the fields which were evidently not relevant and grouping some variables.

Fields regarding the size of the area of residence, the judgement on the weight of democracy and politics were eliminated. The field regarding the province of residence was changed, by grouping together the provinces among the north, the centre and the south.

The new starting database contains 982 records relating to persons who have participated in the political poll by answering the questions that follow:

- gender (men / women)
- coalition of confidence regardless of the vote (ES extreme left, left SS, CS center-left, center CC, CD center-right, right DD, ES extreme right, ** does not answer)
- profession
- believer (yes/no)
- religion
- expectations for the economic situation for the next 6 months
- judgment on the current state of country's economy
- protection (who you can trust)
- security (perception)
- prediction of the winner of a possible short-term election
- opinion of the government's actions

- opinion of the opposition's actions
- interest in politics
- confidence in coalition
- short term vote
- party voted in 2006
- party you intend to vote in the next election
- PDL party
- age
- region
- qualifications
- attendance at religious functions

The processing with the program KB_CAT was made with 4 groups.

The results of the cataloging obtained by KB_CAT are been then processed by KB_STA.

From an analysis of the results obtained we can draw the following observations:

- supporters of the *left* believe, with the exception of those in the group G_04_02, that the eventual winner will be the *center right*, or do not respond to who will be the future winners (group G_02_04)
- supporters of the *center left / left* defend the government and considering his work on *average*
- supporters of the centre left / left on average give a positive opinion on the opposition
- group G_04_04 is formed of apolitical people, agnostic (or extremely reserved), people who prefer not to express an opinion, they are pensioners and unemployed, over 50 years of age
- does not exist, as in the past, a relationship between the profession, the economic condition and the trusted party
- the category of pensioners is divided between those who imagine a victory by the centre right (group G_02_03) and those who prefer not to reply (Group G 02 04 e G 04 04)
- age does not affect cataloging
- in all groups people say they do not want to give the vote to the PDL, even in groups in which the same people speculate that the winner will be the *center* right (G 01 04, G 02 03, G 04 01)

A large part of the observations expressed above were confirmation of the loss of ideological values linked to the *hard core* of belonging to a social class, age group, level of education, area of residency, etc., important characteristics in the past for political tendencies for the voters.

In 2007 the idea of *liquid* voters came into use, that is, the people who are no longer a *supporter* of a party or an alliance, but able to evaluate the results of government and opposition actions and decide whether and how to vote.

It 'obvious that if you define detailed profiles of voters you can then formulate specific election programs and not only based on ideologies almost meaningless.

KB_CLA – Classification of new records

Generalities, aims and functions

The KB_CAT program produces the file containing the training matrix (for example vessels_M_g3_grid.txt) which can be immediately used to classify records that are similar to records that have been previously cataloged.

The use of classification programs on the fly are very useful when you must act quickly taking into consideration the knowledge already acquired.

Classifications running in real time can be found, for example:

- in banking / insurance fields for the prevention of illegal activity
- in the business of mobile phones to identify customers in preparing to move to competition
- in the quality control of industrial processes and products
- in companies to avoid cases of insolvency with clients

Source of KB_CLA (attachment 3)

How to run

KB_CLA requires that the file of the new records / cases to classify has the same structure and a similar content as the file used in the previous KB_CAT processing.

For the same structure we mean that the file of the new records / cases must have the same number of variables / columns with an identical format of data (numerical / textual).

For similar content means that the file of new records / cases should contain records from samples of the same universe.

Acquired knowledge for the cataloging of animals, can not be used to classify new containers!

Input files

= n_vessels.txt

Contents of the file n_vessels.txt

The records / cases to be classify are reported in the following table and are identified by the first character N in the description.

description	shape	material	height colour	weight haft	plug
n_glass	cut_cone	terracotta	6 transparent	22 No	no
n_bottle	cylinder_cone	glass	37 brown	120 No	metal
n_tea_cup	cut_cone	ceramic	7 white	28 Yes	no
n_cup	cut_cone	glass	22 transparent	36 Yes	no
n_coffee_cup	cut_cone	glass	6 transparent	19 Yes	no
n_perfume	cylinder	glass	7 transparent	12 No	plastic
n_trousse	cylinder	plastic	1 blue	6 No	yes
n_plant_pot	cut_cone	terracotta	40 brown	180 No	no
n_pasta_case	cylinder	glass	30 transparent	130 No	metal

Number of Groups (3 - 20) = 3

Normalization(Max, Std, None) = m

File Training Grid = vessels_M_g3_grid.txt

KB_CLA running

```
# KB CLA KNOWLEDGE DISCOVERY IN DATA MINING (CLASSIFY PROGRAM)
# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
# Language used: PYTHON
InputFile
                               : n vessels.txt
Number of Groups (3 - 20)
                              : 3
Normalization(Max, Std, None)
File Training Grid
                               : vessels_M_g3_grid.txt
Output File Classify.original n vessels CM g3 out.txt
                     n_vessels_CM_g3_outsrt.txt
Output File Classify.sort
Output File Summary sort
                      n vessels CM g3 sort.txt
Output File Matrix Catal. n_vessels_CM_g3_catal.txt
Output File Means, STD, CV. n_vessels_CM_g3_medsd.txt
Output File CV of the Groups n vessels CM g3 cv.txt
Output File Training Grid vessels_M_g3_grid.txt
Output File Run Parameters n vessels CM g3 log.txt
Elapsed time (seconds) : 0.16115
```

Analysis of the results of the classification of n_vessels.txt

The new records, recognisable by the first letter *N*, have been inserted into the previous table obtained by KB_CAT.

Group description	shape	material	height	colour	Weight	haft	plug
G_00_00 ancient_bottle	ball_cone	glass	40.0	green	150.0	no	cork
G_00_00 bottle_1	cylinder_cone	glass	40.0	green	120.0	no	cork
G_00_00 bottle_4	cylinder_cone	glass	35.0	green	125.0	no	metal
G_00_00 carboy	ball_cone	glass	0.08	green	15000.0	no	cork
G_00_00 magnum_bottle	cylinder_cone	glass	50.0	green	170.0	no	metal
G_00_00 plant_pot	cut_cone	terracotta	30.0	brown	200.0	no	no
G_00_00 umbrella_stand	cylinder	metal	100.0	grey	3000.0	no	no
G_00_00 <mark>n_bottle</mark>	cylinder_cone	glass	37.0	brown	120.0	no	metal
G_00_00	cut_cone	terracotta	6.0	transparent	22.0	no	no
G_00_00 <mark>n_pasta_case</mark>	cylinder	glass	30.0	transparent	130.0	no	metal
G_00_00 <mark>n_plant_pot</mark>	cut_cone	terracotta	40.0	brown	180.0	no	no
G_00_01 pot_1	cylinder	metal	40.0	grey	500.0	two	yes
G_00_02 coffee_cup	cut_cone	ceramic	6.0	white	20.0	yes	no
G_00_02 cup_1	cut_cone	ceramic	10.0	white	30.0	yes	no
G_00_02 cup_2	cut_cone	glass	20.0	transparent	35.0	yes	no

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 50 di 112

	description	shape	material	height 7.0	colour	Weight 200.0		plug
G_00_02		cut_cone	metal	6.0	grey	19.0	yes	yes
	n_coffee_cup	cut_cone	glass	7.0	transparent white	28.0	yes	no
G_00_02 G_01_00	n_tea_cup	cut_cone	ceramic	7.0 18.0		25.0	yes	no
		cut_cone	porcelain	40.0	severals		no	no
G_01_00		cylinder_cone	•		transparent	125.0	no	cork
G_01_00	_	cylinder_cone	•	45.0	opaque	125.0	no	plastic
G_01_00	<u> </u>	cut_cone	pewter	10.0	pewter	20.0	no	no
G_01_00	-	cut_cone	terracotta		grey	20.0	no	no
	tuna_can	cylinder	metal	10.0	severals	10.0	no	no
	n_perfume	cylinder	glass	7.0	transparent	12.0	no	plastic
	n_trousse	cylinder	plastic	1.0	blue	6.0	no	yes
G_01_02		cut_cone	glass	22.0	transparent	36.0	yes	no
G_02_00		parallelepiped	•	1.0	transparent	4.0	no	no
	champagne_glass		crystal	17.0	transparent	17.0	no	no
	dessert_glass	cut_cone	glass	17.0	transparent	17.0	no	no
G_02_00	–	cut_cone	plastic	9.0	white	4.0	no	no
	pasta_case	parallelepiped	•	35.0	transparent	150.0	no	metal
G_02_00	•	parallelepiped	•	7.0	transparent	15.0	no	plastic
	tetrapack1	parallelepiped		40.0	severals	20.0	no	plastic
	tetrapack2	parallelepiped	•	40.0	severals	21.0	no	plastic
	tetrapack3	parallelepiped		40.0	severals	22.0	no	no
	toothpaste	cylinder	plastic	15.0	severals	7.0	no	plastic
G_02_00		cylinder	plastic	1.0	silver	7.0	no	yes
	tuna_tube	cylinder	plastic	15.0	severals	7.0	no	plastic
G_02_00	visage_cream	cylinder	metal	15.0	white	7.0	no	no
G_02_00	wine_glass	cut_cone	glass	15.0	transparent	15.0	no	no
G_02_01		parallelepiped	glass	10.0	transparent	300.0	two	glass
G_02_02	cleaning_1	parall_cone	plastic	30.0	white	50.0	yes	plastic
G_02_02	cleaning_2	cylinder_cone	plastic	30.0	blue	60.0	yes	plastic
G_02_02	cleaning_3	cone	plastic	100.0	severals	110.0	yes	plastic
G_02_02	jug	cylinder	terracotta	25.0	white	40.0	yes	no
G_02_02	milk_cup	cut_cone	terracotta	15.0	blue	35.0	yes	no
G_02_02	tea_cup	cut_cone	terracotta	7.0	white	30.0	yes	no
G_02_02	watering_can	irregular	plastic	50.0	green	400.0	yes	no

The new records have been classified, almost completely, in the correct way except for the two records highlighted in pink colour.

The *n_glass* record has been classified in the group G_00_00 with the variables *colour* e *weight* with values that are not present in other records of the group.

Political opinions in Facebook (January 2013)

A sample of 1070 political opinions present in 14 different groups of discussion was examined: fb_casini, fb_fini, fb_bonino, fb_di_pietro, fb_corsera, fb_fanpage, fb_brambilla, fb_storace, fb_maroni, fb_bersani, fb_meloni, fb_grillo, fb_termometro_politico, fb_fattoquotidiano.

For every post are taken into consideration:

- the group of discussion
- the topic of the discussion (e.g. gay marriage and adoption, interview by TG3, Rai 3 news, the moral drift)
- the political leader involved (Bersani, Casini, Berlusconi, etc.)
- the attitude inferred from the judgment expressed by the author of the post using a 5-point scale (1 = an abusive opinion, 2 = a negative opinion, 3 = an indifferent opinion, 4 = a positive opinion, 5 = a laudatory judgment).
 - KB Neural Data Mining with Python sources Roberto Bello Pag. 51 di 112

The research aims to explore the possible relationships existing between the groups, the arguments, the politicians and the opinions expressed. KB_CAT has cataloged the 1070 opinions in 25 groups, 15 of which contain a significant number of views.

Group 00 Records 132

groups: fb_corsera, fb_fanpage

politicians: Berlusconi, Dell'Utri, Bersani

topics: always-candidated, moral drift, euro 100000 (cheque for Veronica)

abusive opinion, negative opinion

comment: groups not aligned, and especially the fb corsera group in which the

insults abound

Group 04 Records 115

groups: fb_maroni, fb_storace, fb_meloni

politicians: Maroni, Storace, Meloni

topics: diary

positive, laudatory opinions comment: groups aligned

Group 24 Records 85

groups: fb_fanpage, fb_grillo, fb_fattoquotidiano

politicians: Grillo, Ingroia

topics: various positive opinion

comment: 2 of the groups not aligned but Grillo and Ingroia are new so they attract

people

Group 02 Records 69

groups: fb_brambilla, fb_casini, fb_bersani politicians: Brambilla, Casini, Bersani

topics: various

positive, laudatory opinions

comment: aligned groups, where the animal rights mission "pays"

Group 40 Records 69

groups: fb_termometro_politico politicians: Berlusconi, Ingroia

topics: public services, succession, taxes

offensive, negative opinions

comment: not aligned, politicians and topics that are "hot"

Group 44 Records 66

groups: fb_corsera, fb_meloni, fb_fanpage politicians: Pannella, Meloni, Vendola

topics: alliances, diary attack offensive, negative opinions

comment: group fb corsera not aligned and the unpopular alliance of Pannella with

Storace

Group 43 Records 66 groups: fb_fanpage politician: Monti

topics: no_marr_adop_gay, monti_su_fb

offensive, negative opinions

comment; not aligned and disagreement on no_marr_adop_gay

.....

Group 12 Records 58

groups: fb_bonino, fb_brambilla, fb_casini politicians: Bonino, Brambilla, Casini topics: president republic, animalist bill

positive opinion

comment: aligned groups

Group 11 Records 51

groups: fb_casini, fb_bersani politicians: Casini, Bersani

topics: the 11 lies, interview with TG3 (national news), interview with TG5 (national

news)

negative opinion

comment: criticism for aligned groups on non shared opinions

Group 23 Records 44

groups: fb_dipietro, fb_casini politicians: Di Pietro, Casini, Grillo topics: tv_adverts, last_word positive, laudatory opinion comment: aligned groups

.....

Group 42 Records 43

groups: fb_fanpage, fb_corsera

politicians: Monti, Grillo

topics: piper, profile_fb, monti_exorcist

offensive opinion

comment: not aligned groups

Group 14 Records 40

groups: fb_fanpage, fb_grillo politicians: Monti, Grillo

topics: no_marr_adop_gay, meeting_lecce

positive, laudatory opinion

comment: laudatory opinions on the two topics of Monti and Grillo

Group 21 Records 39

groups: fb_bonino, fb_casini politicians: Bonino, Casini

topics: regional, pact_monti_bersani, president_republic

positive opinion

comment: aligned groups

Group 20 Records 33 groups: fb_fanpage politician: Bersani

topics: pact_monti_bersani offensive negative opinion comment: not aligned group

Group 31 Records 32 groups: fb_di_pietro politician: Di Pietro

topics: rivoluzione civile, tv advert

offensive negative opinion

comment: aligned group and disagreement on "rivoluzione" (civil revolution)

Summary

There are close relationships between the typology of the groups, the politicians, the topics and the opinions.

In the groups "aligned":

- positive and laudatory opinions are plentiful
- any possible disagreements arise from sympathizers who do not share any political positions or from opponents who were immediately marginalised
- obscene language is rare and the syntactical and grammatical forms are proper

In the groups "not aligned":

- · prevails much dissent on consensus
- bad language is the norm and the syntactic and grammar forms are poor
- persons are aware not to suffer criticism
- open discussions on issues banned in the groups "aligned"

Know4Business (Cloud version in Google App Engine)

Giulio Beltrami, software engineer and expert in innovative architectures ICT of the social type, has transferred the KB in the field of Cloud Computing of the Google App Engine with the name of Know4Business.

Know4Business is usable in *pay to use* mode and is obtainable in Internet at the link http://know4business.appspot.com/.

Know4Business adds to Google-Apps a powerful general-purpose discovering of the hidden knowledge, in your business data, enabled by a well-known neural-network self-learning data-mining algorithms.

Know4Business provides 5 tools, to fulfill the knowledge discovery:

•SOURCE for preparing (checking, normalizing, cleaning, filtering and encoding) the input data.

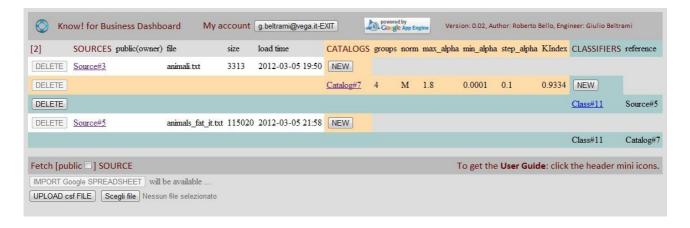
- •CATALOGUE for discovering groups in the sample data, that are in some way or another "similar", without using known structures.
- •STATISTICS to evaluate the success of the catalogue clustering.
- •CLASSIFIER for generalizing known structures, to apply to new data.
- •AGENCY to return some diagnosis and suggestions, about the user data management, checking the success of the classification. plus an interactive CONSOLE to help the data-analyst to:
- •run the catalogue and the other tools
- •Report and graph the results of the tools

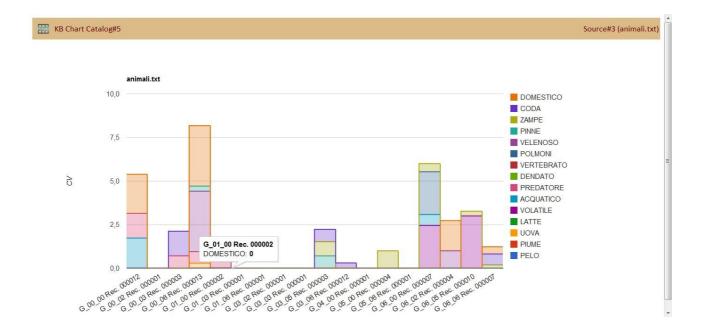
Know4Business end-to-end process of knowledge discovery, provides a simple work-flow, with some useful feedback capabilities:

- •Classification operates forward to the catalogue of the sample data.
- •Statistics of the catalogue clustering can suggest some filtering rules, both on cardinality and dimensions, on the sample data.
- •Agency can also influence some source filtering rules and/or put something to data management, depending on the circumstances. which enables a kind of data knowledge "auto-poiesis", minimizing human intervention.

Know4Business - Main advantages:

- •The ease of use, based upon a simple HTML 5 GUI, to use the tools and to look to the results.
- •The clear implementation, based upon an object oriented paradigm and an authentic SaaS, for the cloud computing, architecture.





APPENDIXES

Appendix 1 - KB_CAT source

```
# -*- coding: utf-8 -*-
# KB CAT KNOWLEDGE DISCOVERY IN DATA MINING (CATALOG PROGRAM)
# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
# Language used: PYTHON
import os
import random
import copy
import datetime
def mean(x):
               # mean
 n = len(x)
 mean = sum(x) / n
 return mean
def sd(x):
               # standard deviattion
 n = len(x)
 mean = sum(x) / n
 sd = (sum((x-mean)**2 for x in x) / n) ** 0.5
 return sd
class ndim:
                   # from 3D array to flat array
   def __init__(self,x,y,z,d):
      self.dimensions=[x,y,z]
      self.numdimensions=d
      self.gridsize=x*y*z
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 56 di 112

```
def getcellindex(self, location):
       cindex = 0
       cdrop = self.gridsize
       for index in xrange(self.numdimensions):
          cdrop /= self.dimensions[index]
          cindex += cdrop * location[index]
       return cindex
   def getlocation(self, cellindex):
       res = []
       for size in reversed(self.dimensions):
          res.append(cellindex % size)
          cellindex /= size
       return res[::-1]
""" how to use ndim class
n=ndim(4,4,5,3)
print n.getcellindex((0,0,0))
print n.getcellindex((0,0,1))
print n.getcellindex((0,1,0))
print n.getcellindex((1,0,0))
print n.getlocation(20)
print n.getlocation(5)
print n.getlocation(1)
print n.getlocation(0)
print("# KB CAT KNOWLEDGE DISCOVERY IN DATA MINING (CATALOG PROGRAM)
                                                                          #")
print("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                                          #")
print("# Language used: PYTHON
                                                                          #")
# input and run parameters
error = 0
while True:
 arch input = raw input('InputFile
                                                        : ')
 if not os.path.isfile(arch input):
   print("Oops! File does not exist. Try again... or CTR/C to exit")
 else:
   break
while True:
     KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 57 di 112
```

```
try:
    num gruppi = int(raw input('Number of Groups (3 - 20)
                                                                     : '))
  except ValueError:
    print("Oops! That was no valid number. Try again...")
  else:
    if(num gruppi < 3):</pre>
      print("Oops! Number of Groups too low. Try again...")
    else:
      if(num gruppi > 20):
        print("Oops! Number of Groups too big. Try again...")
      else:
        break
while True:
  normaliz = raw input('Normalization(Max, Std, None)
                                                        : ')
  normaliz = normaliz.upper()
  normaliz = normaliz[0]
  if(normaliz <> 'M' and normaliz <> 'S' and normaliz <> 'N'):
    print("Oops! Input M, S or N. Try again...")
  else:
   break
while True:
  try:
    max_alpha = float(raw_input('Start value of alpha (1.8 - 0.9) : '))
  except ValueError:
    print("Oops! That was no valid number. Try again...")
  else:
    if(max alpha > 1.8):
     print("Oops! Start value of alpha too big. Try again...")
    else:
      if(max alpha < 0.9):
        print("Oops! Start value of alpha too low. Try again...")
      else:
        break
while True:
  try:
    min alpha = float(raw input('End value of alpha (0.5 - 0.0001) : '))
  except ValueError:
    print("Oops! That was no valid number. Try again...")
  else:
    if(min alpha > 0.5):
      print("Oops! alpha too big. Try again...")
    else:
      if(min alpha < 0.0001):
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 58 di 112
```

```
print("Oops! alpha too low. Try again...")
      else:
       break
while True:
  try:
   step alpha = float(raw input('Decreasing step of alpha (0.1 - 0.001) : '))
  except ValueError:
   print("Oops! That was no valid number. Try again...")
  else:
    if(step alpha > 0.1):
      print("Oops! Decreasing step of alpha too big. Try again...")
   else:
      if(step_alpha < 0.001):</pre>
       print("Oops! Decreasing step of alpha too low. Try again...")
      else:
       break
file_input
            = arch_input
gruppi num = num gruppi
tipo_norm = normaliz
alpha min = min alpha
alpha max = max alpha
alpha step = step alpha
# outputs files
file_input = arch_input
tipo norm = normaliz
gruppi num = num gruppi
nome input = file input.split(".")
arch_output = nome_input[0] + "_" + tipo_norm + "_g" + str(gruppi_num) + "_out.txt"
arch outsrt = nome input[0] + " " + tipo norm + " g" + str(gruppi num) + " outsrt.txt"
            = nome_input[0] + "_" + tipo_norm + "_g" + str(gruppi_num) + "_sort.txt"
arch_sort
arch catal = nome_input[0] + "_" + tipo_norm + "_g" + str(gruppi_num) + "_catal.txt"
            = nome_input[0] + "_" + tipo_norm + "_g" + str(gruppi_num) + "_medsd.txt"
arch medsd
arch cv
            = nome input[0] + " " + tipo norm + " g" + str(gruppi num) + " cv.txt"
arch grid
            = nome input[0] + " " + tipo norm + " g" + str(gruppi num) + " grid.txt"
            = nome_input[0] + "_" + tipo_norm + "_g" + str(gruppi_num) + "_log.txt"
arch log
# start time
t0 = datetime.datetime.now()
# read input file
arr_r = []
arr orig = []
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 59 di 112
```

```
arr c = []
mtchx = []
mtchy = []
txt_col = []
xnomi
       = []
# the numbers of variables / columns in all record must be the same
n rows = 0
n cols = 0
err cols = 0
index = 0
for line in open(file_input).readlines():
  linea = line.split()
 if(index == 0):
   xnomi.append(linea)
   n_cols = len(linea)
  else:
    arr r.append(linea)
   if(len(linea) != n_cols):
      err cols = 1
     print("Different numbers of variables / columns in the record " + str(index)
        + " cols " + str(len(linea)))
  index += 1
if(err cols == 1):
  print("File " + file input + " contains errors. Exit ")
 quit()
index = 0
while index < len(arr_r):</pre>
  linea = arr_r[index]
  index c = 0
 while index c < len(linea):</pre>
    if linea[index c].isdigit():
      linea[index c] = float(linea[index c])
    index_c += 1
  arr r[index] = linea
  index += 1
arr orig = copy.deepcopy(arr r)  # original input file
testata cat = copy.deepcopy(xnomi[0]) # original header row
# finding columns containing strings and columns containing numbers
testata = xnomi[0]
testata_orig = copy.deepcopy(xnomi[0])
n cols = len(testata) - 1
n_rows = len(arr_r)
ind c = 1
err_type = 0
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 60 di 112
```

```
while ind c < len(testata):</pre>
  ind r = 1
  tipo_num = 0
  tipo_txt = 0
  while ind_r < len(arr_r):</pre>
    arr_c = arr_r[ind_r]
    if isinstance(arr c[ind c],basestring):
     tipo_txt = 1
    else:
     tipo_num = 1
    ind r += 1
  if tipo_num == 1 and tipo_txt == 1:
    print "The columns / variables " + testata[ind c] + " contains both strings and
numbers."
    print arr c
    err type = 1
  ind c += 1
if err type == 1:
  print "Oops! The columns / variables contains both strings and numbers. Exit. "
  quit()
index_c = 1
while index_c <= n_cols:</pre>
  txt col = []
  index = 0
  while index < len(arr r):</pre>
    arr_c = arr_r[index]
    if(isinstance(arr_c[index_c],str)):
      txt_col.append(arr_c[index_c])
    index += 1
  set txt col = set(txt col)
                                          # remove duplicates
  txt col = list(set(set txt col))
  txt_col.sort()
  # from strings to numbers
  if(len(txt col) > 0):
    if(len(txt col) > 1):
      passol = 1.0 / (len(txt col) - 1)
    else:
      passo1 = 0.0
    index = 0
    while index < len(arr_r):</pre>
      arr_c = arr_r[index]
      campo1 = arr_c[index_c]
      indice1 = txt_col.index(campo1)
      if(len(txt col) == 1): # same values in the column
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 61 di 112

```
else:
       val_num1 = float(passo1 * indice1)
     arr_c[index_c] = val_num1 + 0.000000001  # to avoid zero values in means
                                             # (to prevent zero divide in CV)
     index += 1
  index c += 1
# means, max & std
xmeans = []
xmaxs = []
xmins = []
                     ### aggiunto Roberto 4/03/2012
xsds = []
xcv = []
index c = 0
while index_c <= n_cols:</pre>
  xmeans.append(0.0)
  xmaxs.append(-9999999999999999))
  xsds.append(0.0)
  xcv.append(0.0)
  index_c += 1
# means & max
index = 0
while index < n_rows:
  arr_c = arr_r[index]
 index_c = 1
 while index_c <= n_cols:
   xmeans[index_c] += arr_c[index_c]
   if(arr_c[index_c] > xmaxs[index_c]):
     xmaxs[index_c] = arr_c[index_c]
   index_c += 1
  index += 1
index c = 1
while index_c <= n_cols:</pre>
  xmeans[index_c] = xmeans[index_c] / n_rows
  index c += 1
# std
index = 0
while index < n_rows:</pre>
  arr_c = arr_r[index]
  index c = 1
  while index_c <= n_cols:
   xsds[index_c] += (arr_c[index_c] - xmeans[index_c])**2
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 62 di 112
```

val num1 = float(1)

```
index c += 1
  index += 1
index_c = 1
while index_c <= n_cols:</pre>
  xsds[index_c] = (xsds[index_c] / (n_cols - 1)) ** 0.5
  index c += 1
# Means, Max, Std, CV output file
medsd_file = open(arch_medsd, 'w')
# columns names
medsd file.write('%s %s ' % ('Function' , "\t"))
index_c = 1
while index c <= n cols:
 medsd_file.write('%s %s ' % (testata[index_c], "\t"))
  index c += 1
medsd_file.write('%s' % ('\n'))
# means
medsd file.write('%s %s ' % ('Mean' , "\t"))
index_c = 1
while index c <= n cols:
 valore = str(xmeans[index c])
 valore = valore[0:6]
 medsd_file.write('%s %s ' % (valore, "\t"))
  index c += 1
medsd_file.write('%s' % ('\n'))
medsd file.write('%s %s ' % ('Max' , "\t"))
index c = 1
while index c <= n cols:
 valore = str(xmaxs[index_c])
 valore = valore[0:6]
 medsd_file.write('%s %s ' % (valore, "\t"))
  index c += 1
medsd file.write('%s' % ('\n'))
# std
medsd file.write('%s %s ' % ('Std' , "\t"))
index_c = 1
while index_c <= n_cols:</pre>
  valore = str(xsds[index c])
 valore = valore[0:6]
  medsd_file.write('%s %s ' % (valore, "\t"))
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 63 di 112
```

```
index c += 1
medsd_file.write('%s' % ('\n'))
medsd_file.write('%s %s ' % ('CV' , "\t"))
index c = 1
med cv gen = 0.0
                         # cv average of all columns / variables
while index_c <= n_cols:</pre>
  if xmeans[index c] == 0:
   media1 = 0.000001
  else:
    media1 = xmeans[index_c]
  xcv[index c] = xsds[index c] / abs(media1)
  valore = str(xcv[index_c])
 med cv gen += xcv[index c]
  valore = valore[0:6]
  medsd_file.write('%s %s ' % (valore, "\t"))
  index c += 1
med_cv_gen = med_cv_gen / n_cols
str_med_cv_gen = str(med_cv_gen)
str_med_cv_gen = str_med_cv_gen[0:6]
medsd_file.write('%s' % ('\n'))
medsd file.close()
# input standardization
# standardization on max
if tipo_norm == 'M':
  index = 0
  while index < n rows:
    arr_c = arr_r[index]
    index c = 1
    while index_c <= n_cols:</pre>
                                ## aggiornare anche kb_cla.py
      if xmaxs[index_c] == 0.0:
        xmaxs[index_c] = 0.00001
      arr_c[index_c] = arr_c[index_c] / xmaxs[index_c]
      index c += 1
    index += 1
# standardization on std
if tipo norm == 'S':
  index = 0
  while index < n rows:
    arr_c = arr_r[index]
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 64 di 112
```

```
index c = 1
    while index_c <= n_cols:
      if xsds[index c] == 0.0:
        xsds[index_c] = 0.00001
      arr_c[index_c] = (arr_c[index_c] - xmeans[index_c]) / xsds[index_c]
      if arr_c[index_c] < xmins[index_c]:</pre>
                                             ### aggiunto Roberto 4/03/2012
        xmins[index c] = arr c[index c]
                                                     ### aggiunto Roberto 4/03/2012
      index c += 1
    index += 1
  # aggiungo xmins per eliminare i valori negativi (aggiunto da Roberto 4/03/2012)
  index = 0
  while index < n_rows:</pre>
    arr c = arr r[index]
    index_c = 1
    while index c <= n cols:
      arr_c[index_c] = arr_c[index_c] - xmins[index_c]
      print arr c[index c]
      index c += 1
    index += 1
  # fine aggiunta da Roberto 4/03/2012
# start of kohonen algorithm
# min and max vectors
vmaxs = []
vmins = []
index_c = 0
while index c <= n cols:
  vmaxs.append(-1000000000000000)
  vmins.append( 10000000000000.0)
  index_c += 1
# columns min & max
index = 0
while index < n rows:
  arr c = arr r[index]
  index c = 1
  while index c <= n cols:
    if arr_c[index_c] > vmaxs[index_c]:
      vmaxs[index_c] = arr_c[index_c]
    if arr_c[index_c] < vmins[index_c]:</pre>
      vmins[index_c] = arr_c[index_c]
    index c += 1
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 65 di 112
```

```
# run parameters and temp arrays
n = n_rows
m = n_{cols}
nx = gruppi num
ny = gruppi num
ix = 950041
                                    # integer as random seed
nsteps = int(10000 * nx * ny)
                                    # number of steps
nepoks = int(nsteps / n ** 0.5)
                                    # number of epochs
unit calc = int(n * m * nx * ny) # running units
passo = int(5000 / n)
                                    # step of visualization on monitor
rmax = nx - 1
rmin = 1.0
if passo < 1:
 passo = 1
grid = []
                              # training grid
index = 0
while index < nx * ny * m:
  grid.append(0.0)
 index += 1
n=ndim(nx,ny,m,3)
random.seed(ix)
                                            # initial value of random seed to obtain the
same sequences in new runs
index = 0
while index < nx:
  index_c = 0
 while index c < ny:
    index k = 0
   while index k < m:
      ig = n.getcellindex((index,index_c,index_k))
      grid[ig] = random.random()
      index_k += 1
    index c += 1
  index += 1
gridp = copy.deepcopy(grid)
                              # initial previous grid = current grid
gridm = copy.deepcopy(grid)
                              # initial min grid = current grid
# for each record in each epoch
iter
discrea = 100000000000.0
                                 # current error
discrep = 0.0
                          # previous error
if nepoks < 20:
  nepoks = 20
                                 # min epochs = 20
```

index += 1

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 66 di 112

```
nepokx = 0
min epok = 0
                                # epoch with min error
min_err = 1000000000.0
                              # min error
alpha = float(alpha_max)
                              # initial value of alpha parameter
                             # initial value of ir parameter ir
        = 0.0
        = 1
print " "
print 'Record ' + str(n rows) + ' Columns ' + str(n cols)
# main loop
try:
 while ne <= nepoks:
    if (ne % passo == 0): # print running message when modulo division = zero
     min err txt = "%14.5f" % min err  # format 8 integers and 3 decimals
     alpha txt = "%12.5f" % alpha  # format 6 integers and 5 decimals
     print ('Epoch ' + str(ne) + ' min err ' + min_err_txt + ' min epoch ' +
          str(min_epok - 1) + " alpha " + alpha_txt)
    if min err < 1000000000.0:</pre>
     nepokx += 1
    if min err > discrea and discrep > discrea and discrea > 0.0:
     min_epok = ne
                                 # current epoch (min)
     min err = discrea
      # copy current grid to min grid
     gridm = copy.deepcopy(grid)
     min_err_txt = "%12.3f" % min_err # format 8 integers and 3 decimals
     alpha txt = "%12.5f" % alpha  # format 6 integer and 5 decimals
     print ('**** Epoch ' + str(ne - 1) + ' WITH MIN ERROR ' + min_err_txt +
           alpha " + alpha_txt)
    # cheking the current value of alpha
    if alpha > alpha min:
     discrea = discrep
     discrep = 0.0
      # copy current grid to previous grid
     gridp = copy.deepcopy(grid)
      # from the starting row to the ending row
     i = 0
     while i < n rows:
       iter += 1
        # find the best grid coefficient
       ihit = 0
        jhit = 0
       dhit = 100000.0
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 67 di 112
```

```
igy = 0
while igx < nx:
  iqy = 0
  while igy < ny:
    d = 0.0
    neff = 0
    k = 0
    arr c = arr r[i]
    while k < m:
                   # update the sum of squared deviation of input
                    # value from the grid coefficient
      ig = n.getcellindex((igx,igy,k))
      d = d + (arr c[k+1] - grid[ig]) ** 2
      k += 1
    d = d / float(m)
    \# d = d / m
    if d < dhit:
      dhit = d
      ihit = int(igx)
      jhit = int(igy)
    igy += 1
  iqx += 1
# update iteration error
discrep = discrep + dhit
# now we have the coordinates of the best grid coefficient
ir = max(rmax * float(1001 - iter) / 1000.0 + 0.9999999999 , 1)
ir = int(ir)
# new alpha value to increase the radius of groups proximity
alpha = max(alpha_max * float(1 - ne * alpha_step) , alpha_min)
# update the grid coefficients applying alpha parameter
inn0 = int(ihit) - int(ir)
inn9 = int(ihit) + int(ir)
jnn0 = int(jhit) - int(ir)
jnn9 = int(jhit) + int(ir)
while inn0 <= inn9:
  jnn0 = int(jhit) - int(ir)
 while jnn0 <= jnn9:
    if not (inn0 < 0 \text{ or } inn0 >= nx):
      if not (jnn0 < 0 \text{ or } jnn0 >= ny):
        arr c = arr r[i]
        k = 0
        while k < m:
          ig = n.getcellindex((inn0,jnn0,k))
          grid[ig] += alpha * (arr_c[k+1] - grid[ig])
          k += 1
    jnn0 += 1
  inn0 += 1
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 68 di 112

```
i += 1
    else:
      print
     print "Min alpha reached "
     print
     break
   ne += 1
except KeyboardInterrupt:
 print
  print "KeyboardInterrupt (Ctrl/C) "
 print
  pass
# computing results
# grid = grid min
grid = copy.deepcopy(gridm)
# write min grid file
arch grid file = open(arch grid, 'w')
ii = 0
while ii < nx:
  j = 0
 while j < ny:
   k = 0
   while k < m:
      ig = n.getcellindex((ii,j,k))
      arch_grid_file.write('%6i %s %.6i %s %.6i %s %14.7f %s' % (ii,' ', j ,' ', k,' ',
grid[ig], "\n"))
      k += 1
    j += 1
  ii += 1
arch grid file.close()
# catalog input by min grid
ii = 0
while ii < n rows:
  ihit = 0
  jhit = 0
  dhit = 100000.0
  # from 1 to numbers of groups
  ir = 0
  while ir < nx:
                         # from 1 to numbers of groups
    jc = 0
   while jc < ny:
                       # from 1 to numbers of groups
      d = 0.0
     neff = 0
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 69 di 112

```
k = 0
      while k < n cols: # update the sum of squared deviation of input
                         # value from the grid coefficient
        arr_c = arr_r[ii]
        ig = n.getcellindex((ir,jc,k))
        d = d + (arr_c[k+1] - grid[ig]) ** 2
        k += 1
      d = d / m
      if d < dhit:
                         # save the coordinates of the best coefficient
        dhit = d
       ihit = ir
        jhit = jc
      jc += 1
    ir += 1
  mtchx.append(ihit)
 mtchy.append(jhit)
  ii += 1
# write arch catal file
arch_catal_file = open(arch_catal, 'w')
ii = 0
while ii < n_rows:
 arch catal file.write("%.6i %s %.6i %s %.6i %s" % (ii, ' ', mtchx[ii], ' ', mtchy[ii],
"\n"))
 ii += 1
arch catal file.close()
# matrix of statistics
arr_cv = []
                           # CV array of the Groups and Total
                           # means array of the Groups
arr med = []
riga cv = []
                                  # CV row in arr cv
arr col = []
                                  # group temporary array
                                  # input data array (normalized)
arr_grsg = []
                               # copy of arr grsg (for file out sort)
arr_grsg_c = []
# input matrix sort in group sequence
ii = 0
ix = 0
while ii < n_rows:
 ix += 1
 gr1 = str(mtchx[ii])
  if mtchx[ii] < 10:
    gr1 = '0' + str(mtchx[ii])
  sg1 = str(mtchy[ii])
  if mtchy[ii] < 10:</pre>
    sg1 = '0' + str(mtchy[ii])
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 70 di 112

```
riga norm = arr r[ii]
  im = 0
  riga_norm1 = []
  while im <= m:
   riga norm1.append(str(riga norm[im]))
   im += 1
  riga norm2 = " ".join(riga norm1)
  gr_sg_txt = "G_" + gr1 + "_" + sg1 + " " + str(ix) + " " + riga_norm2
  arr grsg.append(gr sg txt)
 ii += 1
arr grsg.sort()
ii = 0
while ii < n_rows:
  arr_grsg_c.append(arr_grsg[ii])
 ii += 1
# setup of arr cv matrix
num gr = 0
gruppo0 = ""
ir = 0
while ir < n_rows:
  grsg_key = arr_grsg_c[ir].split()
 if not grsg_key[0] == gruppo0:
   gruppo0 = grsg_key[0]
   num gr +=1
   ic = 1
   riga1 = []
   rigal.append(grsg_key[0])
   while ic \leq m + 2:
                            # adding new columns for row mean and n° of records
     rigal.append(0.0)
     ic += 1
                              # cv row
   arr cv.append(rigal)
  ir += 1
riga1 = []
rigal.append("*Means*")  # adding new row for cv mean
ic = 1
while ic <= m + 2:
                                 \# adding new column for row mean and n^{\circ} of records
 rigal.append(0.0)
 ic += 1
arr_cv.append(riga1)
def found(x):
  ir = 0
 while ir < len(arr_cv):</pre>
   linea cv = arr cv[ir]
    key cv = linea cv[0]
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 71 di 112
```

```
return ir
    ir += 1
ir = 0
irx = len(arr_grsg_c)
ic = 3
linea_cv = arr_cv[0]
icx = len(linea cv)
val\_col = []
while ic < icx:
 ir = 0
  gruppo = ""
 val col = []
 while ir < irx:
    linea = arr_grsg_c[ir].split()
    if linea[0] == gruppo or gruppo == "":
      gruppo = linea[0]
      val_col.append(float(linea[ic]))
    else:
      i_gruppo = found(gruppo)
      linea_cv = arr_cv[i_gruppo]
      media_v = abs(mean(val_col))
      if media v == 0.0:
         media_v = 0.000000001
      std_v = sd(val_col)
      cv_v = std_v / media_v
      linea_cv[ic-2] = cv_v
                                                 # cv value
      linea_cv[len(linea_cv)-1] = len(val_col) # number of records
      val col = []
      val col.append(float(linea[ic]))
      gruppo = linea[0]
    ir += 1
  i gruppo = found(gruppo)
  linea_cv = arr_cv[i_gruppo]
  media v = abs(mean(val col))
  if media v == 0.0:
   media v = 0.0000000001
  std_v = sd(val_col)
  cv v = std v / media v
  linea_cv[ic-2] = cv_v
                                                 # cv value
  linea_cv[len(linea_cv)-1] = len(val_col)
                                                 # number of records
  ic += 1
ir = 0
irx = len(arr cv)
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 72 di 112
```

if key cv == x:

```
linea cv = arr cv[0]
icx = len(linea_cv) - 2
ic = 1
num rec1 = 0
while ir < irx:
                                                 # rows mean
 media riga = 0.0
  ic = 1
 num col1 = 0
  linea_cv = arr_cv[ir]
  while ic < icx:
   media_riga += float(linea_cv[ic])
   num col1 += 1
   ic += 1
 linea cv[icx] = media riga / num col1
 num_rec1 += linea_cv[icx + 1]
 ir += 1
ir = 0
ic = 1
while ic < icx:
                                 # weighted mean of columns
 media_col = 0.0
 ir = 0
 num rec1 = 0
 while ir < irx - 1:
   linea_cv = arr_cv[ir]
   media_col = media_col + linea_cv[ic] * linea_cv[icx+1] # linea_cv[icx+1] = number
of records
   num_rec1 = num_rec1 + linea_cv[icx+1]
   ir += 1
  linea cv = arr cv[irx - 1]
  linea cv[ic] = media col / num rec1
  ic += 1
# updating mean of the row
linea cv = arr cv[irx - 1]
linea_means = linea_cv[1:icx]
media riga = mean(linea means)
linea_cv[icx] = media_riga  # Total mean
linea cv[icx + 1] = num rec1
                                  # n° of records
cv_media_gen_after = str(media_riga)
cv_media_gen_after = cv_media_gen_after[0:6]
# write cv file
testata_cv = testata
testata cv[0] = "*Groups*"
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 73 di 112

```
testata cv.append("*Mean*")
testata cv.append("N recs")
arch cv file = open(arch cv, 'w')
ic = 0
while ic <= icx + 1:
  arch_cv_file.write('%s %s ' % (testata_cv[ic], " "*(9-len(testata_cv[ic]))))
 ic += 1
arch cv file.write('%s' % ('\n'))
ir = 0
while ir < irx:
 ic = 0
  linea cv = arr cv[ir]
  while ic <= icx + 1:
    if ic == 0:
      arch cv file.write('%s %s ' % (linea cv[0], " "))
   else:
      if ic <= icx:
        arch_cv_file.write('%7.4f %s ' % (linea_cv[ic], " "))
        arch_cv_file.write('%6i %s ' % (linea_cv[ic], " "))
   ic += 1
  arch_cv_file.write('%s' % ("\n"))
 ir += 1
ic = 0
media_xcv = mean(xcv[1:icx])
while ic <= icx : # print CV input (before catalogue)</pre>
  if ic == 0:
    arch_cv_file.write('%s %s ' % ("*CVinp*", " "))
  else:
    if ic < icx:
      arch cv file.write('%7.4f %s ' % (xcv[ic], " "))
    else:
      arch_cv_file.write('%7.4f %s ' % (media_xcv, " "))
      arch_cv_file.write('%6i %s ' % (linea_cv[ic+1], " "))
  ic += 1
arch cv file.write('%s' % ("\n"))
#======istruzioni aggiunte Roberto Bello 29/02/2012============================
#know_index = str(1.0 - float(cv_media_gen_after) / float(str_med_cv_gen))
#know index = know index[0:6]
#arch_cv_file.write('%s %s %s' % ('*KIndex* ', know_index, '\n'))
#======fine istruzioni aggiunte da Roberto Bello 29/02/2012===========
arch_cv_file.close()
# writing out catalog file
```

KB - Neural Data Mining with Python sources - Roberto Bello - Pag. 74 di 112

```
testata cat1 = []
testata_cat1.append("*Group*")
arch_output_file = open(arch_output, 'w')
ic=0
while ic < icx:
  testata_cat1.append(testata_cat[ic])
  ic += 1
ic=0
while ic < len(testata cat1):</pre>
  arch_output_file.write('%s %s ' % (testata_cat1[ic], " "*(15-len(testata_cat1[ic]))))
  ic += 1
arch output file.write('%s ' % ("\n"))
index = 0
while index < len(arr_orig):</pre>
 riga orig = arr orig[index]
  ic = 0
  while ic < len(riga_orig):</pre>
    if not(isinstance(riga_orig[ic],str)):
      riga_orig[ic] = str(riga_orig[ic])
    ic += 1
  # place before 0 if gr / sg < 10</pre>
  gr1 = str(mtchx[index])
  if mtchx[index] < 10:</pre>
    gr1 = '0' + str(mtchx[index])
  sg1 = str(mtchy[index])
  if mtchy[index] < 10:</pre>
    sg1 = '0' + str(mtchy[index])
  arr_rig0 = "G_" + gr1 + "_" + sg1 + " "*8
  arch_output_file.write('%s ' % (arr_rig0))
  ic=0
  while ic < len(riga orig):
    arch_output_file.write('%s %s ' % (riga_orig[ic], " "*(15-len(riga_orig[ic]))))
    ic += 1
  arch_output_file.write('%s ' % ("\n"))
  index += 1
testata_cat1 = []
testata cat1.append("*Group*")
testata cat1.append("*RecNum*")
arch sort file = open(arch sort, 'w')
ic=0
while ic < icx:
 testata_cat1.append(testata_cat[ic])
ic=0
while ic < len(testata cat1):</pre>
  arch_sort_file.write('%s %s ' % (testata_cat1[ic], " "*(15-len(testata_cat1[ic]))))
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 75 di 112
```

```
arch sort file.write('%s ' % ("\n"))
index = 0
while index < len(arr_grsg_c):</pre>
  riga_grsg = arr_grsg_c[index].split()
  ic = 0
  while ic < len(riga grsg):
    val_txt = riga_grsg[ic]
   val txt = val txt[0:13]
    arch_sort_file.write('%s %s ' % (val_txt, " "*(15-len(val_txt))))
    ic += 1
  if index < len(arr_grsg_c) - 1:</pre>
    arch sort file.write('%s ' % ("\n"))
  index += 1
arch sort file.close()
# writing out catalog and sorted file
arr outsrt = []
index = 0
while index < len(arr_orig):</pre>
  riga_sort = []
  # place before 0 if gr / sg < 10</pre>
  gr1 = str(mtchx[index])
  if mtchx[index] < 10:</pre>
    gr1 = '0' + str(mtchx[index])
  sg1 = str(mtchy[index])
  if mtchy[index] < 10:</pre>
    sg1 = '0' + str(mtchy[index])
  riga_sort.append("G_" + gr1 + "_" + sg1)
  ic = 0
  riga orig = arr orig[index]
  while ic < len(riga orig):
    val riga = riga orig[ic]
   riga_sort.append(val_riga)
    ic += 1
  arr_outsrt.append(riga_sort)
  index += 1
for line in arr outsrt:
  line = "".join(line)
arr_outsrt.sort()
testata_srt = []
testata srt.append("*Group*")
arch_outsrt_file = open(arch_outsrt, 'w')
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 76 di 112
```

ic += 1

```
ic=0
while ic < icx:
 testata srt.append(testata orig[ic])
 ic += 1
ic=0
while ic < len(testata srt):</pre>
 arch outsrt file.write('%s %s' % (testata_srt[ic], " "*(15-len(testata_srt[ic]))))
 ic += 1
arch outsrt file.write('%s' % ("\n"))
index = 0
key_gruppo = ""
while index < len(arr outsrt):</pre>
 riga sort = arr outsrt[index]
 index c = 0
 while index c < len(riga sort):
   if index c == 0:
     if riga sort[0] != key gruppo:
      # arch outsrt file.write('%s ' % ("\n"))
      key gruppo = riga sort[0]
   valore = riga sort[index c]
   arch outsrt file.write('%s %s' % (valore, " "*(15-len(valore))))
   index_c += 1
 if index < len(arr grsg c) - 1:
   arch outsrt file.write('%s' % ("\n"))
 index += 1
arch_outsrt_file.close()
print("# KB CAT KNOWLEDGE DISCOVERY IN DATA MINING (CATALOG PROGRAM)
                                                                      #")
print("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                                      #")
print("# Language used: PYTHON
                                                                      #")
arch_log_file = open(arch_log, 'w')
arch log file.write("%s %s" %
arch log file.write("%s %s" % ("# KB CAT KNOWLEDGE DISCOVERY IN DATA MINING (CATALOG
PROGRAM)
                  #", "\n"))
arch log file.write("%s %s" % ("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS
RESERVED)
                   #", "\n"))
arch log file.write("%s %s" % ("# Language used: PYTHON
#", "\n"))
arch log file.write("%s %s" %
arch log file.write("%s %s %s" % ("Input File
", file input, "\n"))
arch_log_file.write("%s %s %s" % ("Numer of Groups (3 - 20)
                                                                     ->
", str(gruppi_num), "\n"))
```

```
arch log file.write("%s %s %s" % ("Normalization (Max, Std, None)
                                                                                  ->
", tipo norm, "\n"))
arch_log_file.write("%s %s %s" % ("Start Value of alpha (from 1.8 to 0.9)
", str(alpha max), "\n"))
arch log file.write("%s %s %s" % ("End Value of alpha (from 0.5 to 0.0001)
                                                                                  ->
", str(alpha min), "\n"))
arch_log_file.write("%s %s %s" % ("Decreasing step of alpha (from 0.1 to 0.001)
", str(alpha_step), "\n"))
arch log file.write("%s"
\n"))
arch log file.write("%s %s %s" % ("Output File Catalog.original
                                                                 ", arch output,
"\n"))
arch_log_file.write("%s %s %s" % ("Output File Catalog.sort
                                                                 ", arch outsrt,
"\n"))
arch log file.write("%s %s %s" % ("Output File Summary sort
                                                                 ", arch sort, "\n"))
arch log file.write("%s %s %s" % ("Output File Matrix Catal.
                                                                 ", arch_catal,
arch log file.write("%s %s %s" % ("Output File Means, STD, CV.
                                                                 ", arch medsd,
"\n"))
arch_log_file.write("%s %s %s" % ("Output File CV of the Groups
                                                                 ", arch cv, "\n"))
arch log file.write("%s %s %s" % ("Output File Training Grid
                                                                 ", arch grid, "\n"))
arch log file.write("%s %s %s" % ("Output File Run Parameters
                                                                 ", arch log, "\n"))
#======istruzioni aggiunte Roberto Bello 29/02/2012============================
know_index = str(1.0 - float(cv_media_gen_after) / float(str_med_cv_gen))
know_index = know_index[0:6]
arch log file.write('%s %s %s' % ('*KIndex* ', know index, '\n'))
#======fine istruzioni aggiunte da Roberto Bello 29/02/2012==========
min err txt = "%12.3f" % min err
                                   # format 8 integer and 3 decimals
alpha txt = "%12.5f" % alpha
                                    # format 6 integer and 5 decimals
alpha min txt = "%12.5f" % alpha min # format 6 integer and 5 decimals
print
if min err == 100000000.000:
  print("Oops! No result. Try again with new alpha parameters")
print
print ("EPOCH " + str(min epok -1) + " WITH MIN ERROR " + min err txt +
  " starting alpha " + alpha_min_txt + " ending alpha " + alpha_txt +
  " Iterations " + str(iter) + " Total Epochs " + str(ne - 1))
print
print 'Output File Catalog.original ' + arch output
                                 ' + arch_outsrt
print 'Output File Catalog.sort
print 'Output File Summary sort
                                  ' + arch sort
print 'Output File Matrix Catal.
                                  ' + arch catal
print 'Output File Means, STD, CV. ' + arch medsd
print 'Output File CV of the Groups ' + arch cv
print 'Output File Training Grid
                                  ' + arch grid
print 'Output File Run Parameters
                                   ' + arch log
```

```
print 'CV before Catalog
                           ' + str_med_cv_gen
                           ' + cv_media_gen_after
print 'CV after Catalog
know_index = str(1.0 - float(cv_media_gen_after) / float(str_med_cv_gen))
know_index = know_index[0:6]
print 'Knowledge Index
                           ' + know index
print
# Elapsed time
t1 = datetime.datetime.now()
elapsed time = t1 - t0
print "Elapsed time (seconds) : " + str(elapsed time.seconds)
print
Appendix 2 – KB_STA source
# -*- coding: utf-8 -*-
# KB STA KNOWLEDGE DISCOVERY IN DATA MINING (STATISTICAL PROGRAM)
# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                            #
# Language used: PYTHON
import os
import random
import copy
import datetime
def fp_conversion(value):  # from string containing number to float
  try:
     return float(value)
  except ValueError:
      return (value)
def count(s, e):
                  # frequencies count
    return len([x for x in s if (x == e)])
print("# KB STA KNOWLEDGE DISCOVERY IN DATA MINING (STATISTICAL PROGRAM)
                                                                 #")
print("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                                 #")
print("# Language used: PYTHON
                                                                 #")
# input / output files and run parameters
error = 0
while True:
 file input = raw input('Cataloged Records File ( outsrt.txt) : ')
```

```
if not os.path.isfile(file input):
    print("Oops! File does not exist. Try again... or CTR/C to exit")
  else:
    break
while True:
                                                             : ')
  file gruppi = raw input('Groups / CV File ( cv.txt)
  if not os.path.isfile(file_gruppi):
    print("Oops! File does not exist. Try again... or CTR/C to exit")
  else:
   break
while True:
  file_rappor = raw_input('Report File (output)
                                                                      : ')
  if os.path.isfile(file rappor):
   print("Oops! File exist. Try again... or CTR/C to exit")
  else:
    break
while True:
  try:
    omog_perc = int(raw_input("Group Consistency (% from 0 to 100) : "))
  except ValueError:
    print("Oops! That was no valid number. Try again...")
  else:
    if(omog_perc < 0):</pre>
      print("Oops! Group Consistency too low. Try again...")
      if(omog perc > 100):
        print("Oops! Group Consistency too big. Try again...")
      else:
        break
while True:
  try:
    omog_vari = int(raw_input("Variable Consistency (% from 0 to 100) : "))
  except ValueError:
    print("Oops! That was no valid number. Try again...")
  else:
    if(omog vari < 0):</pre>
      print("Oops! Variable Consistency too low. Try again...")
    else:
      if(omog vari > 100):
        print("Oops! Variable Consistency too big. Try again...")
      else:
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 80 di 112
```

```
while True:
 try:
   rec_min = int(raw_input("Select groups containing records >= : "))
  except ValueError:
   print("Oops! That was no valid number. Try again...")
  else:
   if(rec min < 1):</pre>
     print("Oops! Number of records too low. Try again...")
   else:
     break
while True:
  try:
   rec max = int(raw input("Select groups containing records <= : "))</pre>
  except ValueError:
   print("Oops! That was no valid number. Try again...")
  else:
   if(rec max < 1):
      print("Oops! Number of records too low. Try again...")
   if (rec max < rec min):</pre>
      print("Oops! Number of records must be >= " + str(rec_min) + " Try again...")
   else:
      break
while True:
  est rapp = raw input("Summary / Detail report (S / D)
                                                           : ")
  est_rapp = est_rapp.upper()
  est_rapp = est_rapp[0]
  if(est_rapp <> 'S' and est_rapp <> 'D'):
   print("Oops! Input S, D. Try again...")
  else:
   break
inp rapp = "N"
if est_rapp == "D" or est_rapp == "d":
  while True:
    inp rapp = raw input("Display Input Records (Y / N)
                                                              : ")
   inp rapp = inp rapp.upper()
   inp_rapp = inp_rapp[0]
   if(inp rapp <> 'Y' and inp rapp <> 'N'):
      print("Oops! Input Y, N. Try again...")
    else:
     break
# start time
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 81 di 112

```
t0 = datetime.datetime.now()
# initial setup
                         # input rows
arr_r = []
arr c = []
                         # row list of arr_c
xnomi
        = []
                          # headings row
len var = []  # max string lenght of variable
# the numbers of variables / columns in all record must be the same
n \text{ rows} = 0
n cols = 0
err_cols = 0
index = 0
file_log = file_input + "_log.txt"
for line in open(file input).readlines():
  linea = line.split()
  if(index == 0):
    xnomi.append(linea)
   n cols = len(linea)
  else:
   arr_r.append(linea)
   if(len(linea) != n_cols):
     err cols = 1
     print("Different numbers of variables / columns in the record " + str(index)
        + " cols " + str(len(linea)))
  index += 1
if(err cols == 1):
  print("File " + file_input + " contains errors. Exit ")
 quit()
index = 0
while index < len(arr r):</pre>
  linea = arr r[index]
  index c = 0
 while index_c < len(linea):</pre>
                                     # converting strings containing numbers to float
   linea[index_c] = fp_conversion(linea[index_c])
   index_c += 1
  arr r[index] = linea
  index += 1
testata = xnomi[0]
n_{cols} = len(testata) - 1
n rows = len(arr r)
index = 0
while index < len(testata):</pre>
                                              finding max string len (for the report)
  len_var.append(len(testata[index]))
 index += 1
index = 0
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 82 di 112
```

```
while index < len(arr r):</pre>
  linea = arr_r[index]
  index c = 0
  while index_c < len(linea):</pre>
    if isinstance(linea[index c],basestring): # text
      len_campo = len(linea[index_c])
                                        # number
    else:
      len_campo = len(str(linea[index_c]))
    if len campo > len var[index c]:
      len_var[index_c] = len_campo
    index c += 1
  index += 1
max len = max(len var)
arr_cv = []
testata cv = []
index = 0
# reading Groups / CV file
for line in open(file gruppi).readlines():
  linea = line.split()
  if(index == 0):
    n_cols = len(linea)
    testata cv.append(linea)
    arr cv.append(linea)
    if(len(linea) != n_cols):
      err cols = 1
      print("Different numbers of variables / columns in the record " + str(index)
        + " cols " + str(len(linea)))
  index += 1
if(err cols == 1):
  print("File " + file gruppi + " contains errors. Exit ")
  quit()
for line in arr_cv:
  index c = 0
  linea = line
  while index c < len(linea):</pre>
                                  # converting strings containing numbers to float
    linea[index_c] = fp_conversion(linea[index_c])
    index c += 1
ind_fine = len(arr_cv)
                                  # last value of arr cv
ind_fine = ind_fine - 1
arr c = arr cv[ind fine]
                                  # row of totals
tot_omogen = float(arr_c[-2])
                                 # consistency totals
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 83 di 112
```

```
tot record = float(arr c[-1])  # total of records
arch rappor = open(file rappor, 'w')
index = 0
ind fine = len(arr cv)
ind fine = ind fine -2
                               # row of CV Totals
testata cv = testata cv[0]
testata cv = testata cv[:-2]
                                # removing the last two columns
arch rappor.write("%s %s %s %s %s %s %s" % ("KB STA - Statistical Analysis from: ",
file input, " and from: ", file gruppi, "\[ \overline{\} n"))
arch rappor.write("%s %s "% (("Min Perc. of group Consistency: ",
str(omog_perc), " Min Perc. of variable Consistency: ",
  str(omog vari), "\nMin Number of records: " , str(rec min), " Max Number of records: "
, str(rec max),"\n")))
arch rappor.write("%s " % ("by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS
RESERVED) \n"))
while index < len(arr cv):</pre>
   arr c = arr cv[index]
    intero = int(arr_c[-1])  # totals of records
   perc omogen = 0.0
   if not tot omogen == 0:
     perc omogen = (arr c[-2] * 100.0) / tot omogen
   perc omog = 100 - int(perc omogen)
   if perc omog < 0.0:
     perc omog = 0.0
   perc rec = intero * 100.0 / tot record
    if (perc omog >= omog perc and rec min <= intero and rec max >= intero) or arr c[0]
== "*Means*":
     arch_rappor.write("%s " %
( "-------
      arch rappor.write(("%s %s %.4f %s %3s %s %5s %s %7.2f %s " % (arr c[0], "
Consistency ", arr_c[-2], " %Consistency ",
       str(perc_omog), " Records ", str(intero), " %Records ", perc_rec, "\n")))
     ind c = 0
     cod gruppo = arr c[0]
     while ind c < len(arr c) - 3:
       omogen perc = 0.0
       if float(arr c[ind c +1]) == 0.0:
         arr c[ind c] = "0.00000"
        if not arr c[-2] \le 0.0:
         omogen perc = 100.0 - arr c[ind c +1] * 100.0 / arr c[-2] # CV of group
variabile divided by CV of the group
       else:
         omogen perc = 100.0
       if omogen perc <= 0.0:
         omogen perc = 0.0
        if omogen perc >= omog vari and (est rapp == "d" or est rapp == "D"): #
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 84 di 112
```

```
consistency value >= min parameter
          arch rappor.write("%s %s %s %10.4f %s %10.2f %s" % (("*** ", testata cv[ind c
+ 11 +
           " " * (max len - len(testata cv[ind c + 1])) , "Consistency\t",
           float(arr_c[ind_c + 1]),"\t%Consistency\t",omogen_perc, "\n")))
          # computing variables frequencies and quartiles
          # 1) variables frequencies
          ind sort = 0
          arr temp = []
                          # variable array of records included in the group
          ind temp = 0
          while ind_sort < len(arr_r): # list of variable values in the group</pre>
            linea = arr r[ind sort]
            if linea[0].strip() == cod gruppo:
              arr temp.append(linea[ind c + 2])
              if (est_rapp == "d" or est_rapp) == "D" and (inp_rapp == "y" or inp_rapp
== "Y"):
               arch rappor.write(("%s %s %s %s %s %s %s" % (linea[0], "\tID record\t",
str(linea[1]) +
                " " * (max_len - len(str(linea[1]))), "Value", linea[ind_c + 2], "\n")))
              ind temp += 1
            ind sort += 1
          # 2) converting strings containing numbers to float
          ind temp = 0
          tipo num = 0
          tipo txt = 0
          while ind_temp < len(arr_temp): # texts or numbers</pre>
            arr temp[ind temp] = fp conversion(arr temp[ind temp])
            if isinstance(arr_temp[ind_temp],basestring):
               tipo txt = 1
            else:
               tipo num = 1
            ind temp += 1
          if tipo num == 1 and tipo txt == 1:
            print "The columns / variable " + testata[ind_c] + " contains both strings
and numbers. Exit. "
            quit()
          if tipo num == 0: # the variable is a text
            arr temp.sort()
            # 3) computing frequencies
            key1 = ""
            key freq = []
                                 # keys and frequencies
          arr t index = 0
            while arr t index < len(arr temp):</pre>
```

```
if arr temp[arr t index] <> key1:
                kf valore = []
                kf_valore.append(arr_temp.count(arr_temp[arr_t_index]))
                kf valore.append(arr_temp[arr_t_index])
                key_freq.append(kf_valore)
                key1 = arr_temp[arr_t_index]
              arr_t_index += 1
            key_freq.sort()
                            # frequencies ascending sort
            key_freq.reverse() # frequencies descending sort
            ris out index = 0
            while ris out index < len(key freq) and (est rapp == "d" or est rapp ==
"D"):
              kf_valore = key_freq[ris_out_index]
              arch rappor.write("%s %s %s %7i %s %.2f %s" % (("Value\t", kf valore[1] +
             " " * (max len - len(kf valore[1])),
                "Frequency\t", kf valore[0],"\tPercentage\t",
kf_valore[0]*100/len(arr_temp), "
n")))
              ris_out_index += 1
          if tipo txt == 0:
                               # the variabile is a number
            # computing means
            if len(arr temp) > 0:
              mean arr = sum(arr temp)/len(arr temp)
            # computing the step of quartiles
            arr_temp.sort()
            if len(arr temp) > 0:
              minimo = arr_temp[0]
              massimo = arr temp[len(arr temp) - 1]
              passo = (float(massimo) - float(minimo)) / 4.0
              q1 = minimo + passo
              q2 = q1 + passo
              q3 = q2 + passo
              q4 = q3 + passo
              fr1 = 0.0 # first quartile
              fr2 = 0.0
                           # second quartile
              fr3 = 0.0
                           # third quartile
              fr4 = 0.0  # fourth quartile
              arr index = 0
              while arr index < len(arr temp):</pre>
                if arr temp[arr index] <= q1:</pre>
                  fr1 += 1
                elif arr temp[arr index] <= q2:</pre>
                  fr2 += 1
                elif arr temp[arr index] <= q3:</pre>
                  fr3 += 1
```

```
else:
                fr4 += 1
              arr index += 1
            records = len(arr temp)
            p1 = fr1 * 100 / records
             p2 = fr2 * 100 / records
            p3 = fr3 * 100 / records
            p4 = fr4 * 100 / records
            if (est rapp == "d" or est rapp == "D"):
              arch rappor.write("%s %.2f %s %.2f %s %.2f %s %.2f %s" % ("Mean\t",
mean arr,"Min\t", minimo, "\tMax\t", massimo,"\tStep\t", passo,
              if p1 > 0.0:
                arch rappor.write(("%s %10.2f %s %7.2f %s" % ("First Quartile (end)
", q1,
                  " Frequency %\t", p1, "\n")))
              if p2 > 0.0:
                arch rappor.write(("%s %10.2f %s %7.2f %s" % ("Second Quartile (end)
", q2,
                  " Frequency %\t", p2, "\n")))
              if p3 > 0.0:
                arch rappor.write(("%s %10.2f %s %7.2f %s" % ("Third Quartile (end)
", q3,
                  " Frequency %\t", p3, "\n")))
              if p4 > 0.0:
                arch rappor.write(("%s %10.2f %s %7.2f %s" % ("Fourth Quartile (end)
", q4,
                  " Frequency %\t", p4, "\n")))
       ind c += 1
   index += 1
arch_rappor.close()
arch_log = open(file_log, 'w')
arch log.write("%s %s" %
"\n"))
arch log.write("%s %s" % ("# KB_STA KNOWLEDGE DISCOVERY IN DATA MINING (STATISTICAL
PROGRAM)
                    #", "\n"))
arch_log.write("%s %s" % ("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
#", "\n"))
arch_log.write("%s %s" % ("# Language used: PYTHON
#", "\n"))
arch log.write("%s %s" %
"\n"))
arch_log.write("%s %s %s" % ("INPUT - Cataloged Records File ( outsrt.txt)
-> ", file input, "\n"))
arch log.write("%s %s %s" % ("INPUT - Groups / CV File ( cv.txt)
-> ", file_gruppi, "\n"))
arch log.write("%s %s %s" % ("Group Consistency (% from 0 to 100)
-> ", str(omog perc), "\n"))
arch_log.write("%s %s %s" % ("Variable Consistency (% from 0 to 100)
```

```
-> ", str(omog vari), "\n"))
arch log.write("%s %s %s" % ("Select groups containing records >=
-> ", str(rec_min), "\n"))
arch log.write("%s %s %s" % ("Select groups containing records <=
-> ", str(rec max), "\n"))
arch log.write("%s %s %s" % ("Summary / Detail report (S / D)
-> ", est_rapp, "\n"))
arch log.write("%s %s %s" % ("Display Input Records (Y / N)
-> ", inp rapp, "\n"))
arch_log.write("%s %s" %
"\n"))
arch log.write("%s %s %s" % ("Report File
-> ", file_rappor, "\n"))
arch log.close()
# Elapsed time
t1 = datetime.datetime.now()
elapsed time = t1 - t0
print "Elapsed time (seconds)
                         : " + str(elapsed time.seconds) + "." +
str(elapsed_time.microseconds)
print
Appendix 3 – KB_CLA source
# -*- coding: utf-8 -*-
# KB CLA KNOWLEDGE DISCOVERY IN DATA MINING (CLASSIFY PROGRAM)
```

```
# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                                   #
# Language used: PYTHON
import os
import random
import copy
import datetime
def mean(x):
                # mean
 n = len(x)
 mean = sum(x) / n
 return mean
def sd(x):
                # standard deviattion
 n = len(x)
 mean = sum(x) / n
 sd = (sum((x-mean)**2 for x in x) / n) ** 0.5
 return sd
class ndim:
                    # from 3D array to flat array
   def init (self,x,y,z,d):
      self.dimensions=[x,y,z]
```

```
self.numdimensions=d
       self.gridsize=x*y*z
   def getcellindex(self, location):
       cindex = 0
       cdrop = self.gridsize
       for index in xrange(self.numdimensions):
          cdrop /= self.dimensions[index]
          cindex += cdrop * location[index]
       return cindex
   def getlocation(self, cellindex):
       res = []
       for size in reversed(self.dimensions):
          res.append(cellindex % size)
          cellindex /= size
       return res[::-1]
""" how to use ndim class
n=ndim(4,4,5,3)
print n.getcellindex((0,0,0))
print n.getcellindex((0,0,1))
print n.getcellindex((0,1,0))
print n.getcellindex((1,0,0))
print n.getlocation(20)
print n.getlocation(5)
print n.getlocation(1)
print n.getlocation(0)
....
print("# KB CLA KNOWLEDGE DISCOVERY IN DATA MINING (CLASSIFY PROGRAM)
                                                                           #")
print("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                                           #")
print("# Language used: PYTHON
                                                                           #")
# input and run parameters
error = 0
while True:
 arch_input = raw_input('InputFile
                                                         : ')
 if not os.path.isfile(arch input):
   print("Oops! File does not exist. Try again... or CTR/C to exit")
 else:
   break
     KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 89 di 112
```

```
while True:
  try:
    num_gruppi = int(raw_input('Number of Groups (3 - 20)
                                                                        : '))
  except ValueError:
    print("Oops! That was no valid number. Try again...")
  else:
    if(num gruppi < 3):</pre>
      print("Oops! Number of Groups too low. Try again...")
    else:
      if(num gruppi > 20):
        print("Oops! Number of Groups too big. Try again...")
      else:
        break
while True:
  normaliz
             = raw input('Normalization(Max, Std, None)
                                                                : ')
  normaliz
             = normaliz.upper()
  normaliz = normaliz[0]
  if(normaliz <> 'M' and normaliz <> 'S' and normaliz <> 'N'):
    print("Oops! Input M, S or N. Try again...")
  else:
    break
while True:
  arch_grid = raw_input('File Training Grid
                                                                 : ')
  if not os.path.isfile(arch grid):
    print("Oops! File does not exist. Try again... or CTR/C to exit")
  else:
    break
file input
             = arch input
gruppi num
             = num gruppi
             = normaliz
tipo_norm
# outputs files
file input
             = arch input
tipo norm
             = normaliz
gruppi num
            = num gruppi
             = file input.split(".")
nome input
arch output = nome input[0] + " " + "C" + tipo norm + " g" + str(gruppi num) +
" out.txt"
arch_outsrt = nome_input[0] + "_" + "C" + tipo_norm + "_g" + str(gruppi_num) +
" outsrt.txt"
             = nome_input[0] + "_" + "C" + tipo_norm + "_g" + str(gruppi_num) +
arch sort
" sort.txt"
             = nome input[0] + " " + "C" + tipo norm + " g" + str(gruppi num) +
arch catal
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 90 di 112
```

```
" catal.txt"
arch_medsd = nome_input[0] + "_" + "C" + tipo_norm + "_g" + str(gruppi_num) +
"_medsd.txt"
             = nome_input[0] + "_" + "C" + tipo_norm + "_g" + str(gruppi_num) +
arch cv
" cv.txt"
             = nome_input[0] + "_" + "C" + tipo_norm + "_g" + str(gruppi_num) +
arch log
"_log.txt"
# start time
t0 = datetime.datetime.now()
# read input file
arr_r
      = []
arr_orig = []
arr c
       = []
mtchx
        = []
mtchy
       = []
txt col = []
xnomi
        = []
# the numbers of variables / columns in all record must be the same
n rows = 0
n cols = 0
err cols = 0
index = 0
for line in open(file_input).readlines():
  linea = line.split()
  if(index == 0):
    xnomi.append(linea)
   n cols = len(linea)
  else:
    arr r.append(linea)
   if(len(linea) != n_cols):
     err cols = 1
      print("Different numbers of variables / columns in the record " + str(index)
        + " cols " + str(len(linea)))
  index += 1
if(err cols == 1):
  print("File " + file_input + " contains errors. Exit ")
  quit()
index = 0
while index < len(arr r):</pre>
  linea = arr_r[index]
  index c = 0
  while index_c < len(linea):</pre>
    if linea[index c].isdigit():
      linea[index_c] = float(linea[index_c])
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 91 di 112
```

```
index c += 1
  arr r[index] = linea
  index += 1
arr_orig = copy.deepcopy(arr_r)  # original input file
testata cat = copy.deepcopy(xnomi[0]) # original header row
# finding columns containing strings and columns containing numbers
testata = xnomi[0]
testata orig = copy.deepcopy(xnomi[0])
n_{cols} = len(testata) - 1
n_rows = len(arr_r)
ind c = 1
err type = 0
while ind_c < len(testata):</pre>
 ind r = 1
 tipo_num = 0
  tipo txt = 0
  while ind_r < len(arr_r):</pre>
    arr_c = arr_r[ind_r]
    if isinstance(arr_c[ind_c],basestring):
     tipo_txt = 1
    else:
     tipo num = 1
    ind r += 1
  if tipo num == 1 and tipo txt == 1:
    print "The columns / variables " + testata[ind_c] + " contains both strings and
numbers."
   err_type = 1
  ind_c += 1
if err type == 1:
  print "Oops! The columns / variables contains both strings and numbers. Exit. "
  quit()
index c = 1
while index_c <= n_cols:</pre>
 txt col = []
  index = 0
  while index < len(arr_r):</pre>
    arr_c = arr_r[index]
    if(isinstance(arr c[index c],str)):
      txt_col.append(arr_c[index_c])
    index += 1
  set_txt_col = set(txt_col)
                                          # remove duplicates
  txt col = list(set(set txt col))
  txt_col.sort()
```

```
# from strings to numbers
  if(len(txt_col) > 0):
   if(len(txt_col) > 1):
      passol = 1.0 / (len(txt_col) - 1)
      passo1 = 0.0
   index = 0
   while index < len(arr_r):</pre>
      arr c = arr r[index]
     campo1 = arr_c[index_c]
      indice1 = txt col.index(campo1)
      if(len(txt_col) == 1): # same values in the column
       val num1 = float(1)
      else:
       val num1 = float(passo1 * indice1)
      arr_c[index_c] = val_num1 + 0.00000001
                                            # to avoid zero values in means
                                              # (to prevent zero divide in CV)
      index += 1
  index c += 1
# means, max & std
xmeans = []
xmaxs = []
                      ### aggiunto Roberto 4/03/2012
xmins = []
xsds = []
xcv = []
index c = 0
while index_c <= n_cols:</pre>
 xmeans.append(0.0)
 xmins.append(9999999999999999)
                                   ### aggiunto Roberto 4/03/2012
  xsds.append(0.0)
  xcv.append(0.0)
  index_c += 1
# means & max
index = 0
while index < n rows:
 arr c = arr r[index]
  index c = 1
 while index c <= n cols:
   xmeans[index_c] += arr_c[index_c]
   if(arr_c[index_c] > xmaxs[index_c]):
      xmaxs[index_c] = arr_c[index_c]
   index c += 1
  index += 1
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 93 di 112

```
index c = 1
while index_c <= n_cols:
  xmeans[index_c] = xmeans[index_c] / n_rows
  index_c += 1
# std
index = 0
while index < n_rows:</pre>
  arr c = arr r[index]
 index c = 1
  while index_c <= n_cols:
    xsds[index_c] += (arr_c[index_c] - xmeans[index_c])**2
    index c += 1
  index += 1
index c = 1
while index c <= n cols:
  xsds[index c] = (xsds[index c] / (n cols - 1)) ** 0.5
  index c += 1
# Means, Max, Std, CV output file
medsd_file = open(arch_medsd, 'w')
# columns names
index c = 1
while index_c <= n_cols:
 medsd_file.write('%s %s %s ' % ('Col' + str(index_c), testata[index_c], "\t"))
  index_c += 1
medsd_file.write('%s' % ('\n'))
# means
index c = 1
while index c <= n cols:
 valore = str(xmeans[index_c])
 valore = valore[0:6]
 medsd_file.write('%s %s %s ' % ('Mean' + str(index_c), valore, "\t"))
 index c += 1
medsd file.write('%s' % ('\n'))
# max
index c = 1
while index_c <= n_cols:</pre>
  valore = str(xmaxs[index_c])
  valore = valore[0:6]
  medsd_file.write('%s %s %s ' % ('Max' + str(index_c), valore, "\t"))
  index c += 1
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 94 di 112
```

```
medsd file.write('%s' % ('\n'))
# std
index_c = 1
while index c <= n cols:
 valore = str(xsds[index_c])
 valore = valore[0:6]
 medsd_file.write('%s %s %s ' % ('Std' + str(index_c), valore, "\t"))
 index c += 1
medsd_file.write('%s' % ('\n'))
# CV
index c = 1
med cv gen = 0.0
                           # cv average of all columns / variables
while index_c <= n_cols:
 if xmeans[index c] == 0:
   media1 = 0.000001
  else:
   media1 = xmeans[index c]
  xcv[index c] = xsds[index c] / abs(media1)
  valore = str(xcv[index_c])
 med_cv_gen += xcv[index_c]
 valore = valore[0:6]
 medsd file.write('%s %s %s ' % ('CV ' + str(index c), valore, "\t"))
  index c += 1
med cv gen = med cv gen / n cols
str_med_cv_gen = str(med_cv_gen)
str_med_cv_gen = str_med_cv_gen[0:6]
medsd_file.write('%s' % ('\n'))
medsd file.close()
# input standardization
# standardization on max
if tipo_norm == 'M':
 index = 0
 while index < n rows:
   arr_c = arr_r[index]
   index c = 1
   while index_c <= n_cols:</pre>
     if xmaxs[index c] == 0.0:
        xmaxs[index_c] = 0.00001
      arr_c[index_c] = arr_c[index_c] / xmaxs[index_c]
      index c += 1
    index += 1
```

```
# standardization on std
if tipo_norm == 'S':
  index = 0
  while index < n rows:
    arr_c = arr_r[index]
   index c = 1
   while index_c <= n_cols:</pre>
      if xsds[index c] == 0.0:
       xsds[index_c] = 0.00001
      arr_c[index_c] = (arr_c[index_c] - xmeans[index_c]) / xsds[index_c]
      if arr_c[index_c] < xmins[index_c]:</pre>
                                             ### aggiunto Roberto 4/03/2012
                                                     ### aggiunto Roberto 4/03/2012
        xmins[index_c] = arr_c[index_c]
      index_c += 1
    index += 1
  # aggiungo xmins per eliminare i valori negativi (aggiunto da Roberto 4/03/2012)
  while index < n rows:
    arr_c = arr_r[index]
   index_c = 1
   while index_c <= n_cols:
      arr_c[index_c] = arr_c[index_c] - xmins[index_c]
     index c += 1
    index += 1
  # fine aggiunta da Roberto 4/03/2012
# start of kohonen algorithm
n = len(arr_r) - 1
m = len(arr c) - 1
nx = gruppi_num
ny = gruppi num
rmax = nx
rmin = 1.0
grid = []
                              # training grid
index = 0
while index < nx * ny * m:
 grid.append(0.0)
 index += 1
n=ndim(nx,ny,m,3)
# carico la Grid di addestramento da arch grid
for line in open(arch_grid).readlines():
  linea = line.split()
  index = int(linea[0])
  index c = int(linea[1])
  index k = int(linea[2])
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 96 di 112
```

```
valore = float(linea[3])
  ig = n.getcellindex((index,index_c,index_k))
  grid[ig] = valore
# from the starting row to the ending row
i = 0
while i < n rows:
  # find the best grid coefficient
  ihit = 0
  jhit = 0
  dhit = 100000.0
  igx = 0
  igy = 0
 while igx < nx:
   igy = 0
   while igy < ny:
     d = 0.0
     neff = 0
     k = 0
     arr_c = arr_r[i]
     while k < m:
                    # update the sum of squared deviation of input
                     # value from the grid coefficient
        ig = n.getcellindex((igx,igy,k))
        d = d + (arr_c[k+1] - grid[ig]) ** 2
       k += 1
      d = d / float(m)
      \# d = d / m
      if d < dhit:
        dhit = d
        ihit = int(igx)
        jhit = int(igy)
      igy += 1
    igx += 1
  i += 1
# computing results
# catalog input by min grid
ii = 0
while ii < n_rows:
 ihit = 0
  jhit = 0
  dhit = 100000.0
  # from 1 to numbers of groups
  ir = 0
  while ir < nx:
                         # from 1 to numbers of groups
      KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 97 di 112
```

```
jc = 0
    while jc < ny: # from 1 to numbers of groups
      d = 0.0
      neff = 0
     k = 0
      while k < n_{cols}: # update the sum of squared deviation of input
                         # value from the grid coefficient
        arr_c = arr_r[ii]
       ig = n.getcellindex((ir,jc,k))
       d = d + (arr_c[k+1] - grid[ig]) ** 2
       k += 1
      d = d / m
      if d < dhit:
                       # save the coordinates of the best coefficient
       dhit = d
       ihit = ir
       jhit = jc
      jc += 1
    ir += 1
  mtchx.append(ihit)
  mtchy.append(jhit)
  ii += 1
# write arch catal file
arch_catal_file = open(arch_catal, 'w')
ii = 0
while ii < n_rows:
  arch catal file.write("%.6i %s %.6i %s %.6i %s" % (ii, ' ', mtchx[ii], ' ', mtchy[ii],
"\n"))
 ii += 1
arch catal file.close()
# matrix of statistics
                          # CV array of the Groups and Total
arr_cv = []
                          # means array of the Groups
arr_med = []
riga_cv = []
                                 # CV row in arr_cv
arr_col = []
                                 # group temporary array
arr_grsg = []
                                 # input data array (normalized)
                               # copy of arr grsg (for file out sort)
arr grsg c = []
# input matrix sort in group sequence
ii = 0
ix = 0
while ii < n rows:
 ix += 1
 gr1 = str(mtchx[ii])
  if mtchx[ii] < 10:</pre>
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 98 di 112

```
gr1 = '0' + str(mtchx[ii])
  sg1 = str(mtchy[ii])
  if mtchy[ii] < 10:</pre>
   sg1 = '0' + str(mtchy[ii])
  riga_norm = arr_r[ii]
  im = 0
  riga norm1 = []
  while im <= m:
   riga norm1.append(str(riga norm[im]))
   im += 1
 riga norm2 = " ".join(riga norm1)
  gr_sg_txt = "G_" + gr1 + "_" + sg1 + " " + str(ix) + " " + riga_norm2
  arr grsg.append(gr sg txt)
 ii += 1
arr grsg.sort()
ii = 0
while ii < n rows:
  arr_grsg_c.append(arr_grsg[ii])
 ii += 1
# setup of arr_cv matrix
num_gr = 0
gruppo0 = ""
ir = 0
while ir < n rows:
  grsg_key = arr_grsg_c[ir].split()
 if not grsg_key[0] == gruppo0:
   gruppo0 = grsg_key[0]
   num_gr +=1
   ic = 1
   riga1 = []
   rigal.append(grsg_key[0])
   while ic <= m + 2:
                         # adding new columns for row mean and n° of records
     rigal.append(0.0)
     ic += 1
   arr_cv.append(rigal) # cv row
  ir += 1
riga1 = []
rigal.append("*Means*")  # adding new row for cv mean
ic = 1
while ic <= m + 2:
                                # adding new column for row mean and n° of records
 rigal.append(0.0)
  ic += 1
arr cv.append(riga1)
def found(x):
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 99 di 112

```
ir = 0
  while ir < len(arr_cv):
   linea_cv = arr_cv[ir]
   key_cv = linea_cv[0]
   if key_cv == x:
     return ir
    ir += 1
ir = 0
irx = len(arr_grsg_c)
ic = 3
linea_cv = arr_cv[0]
icx = len(linea cv)
val_col = []
while ic < icx:
  ir = 0
  gruppo = ""
 val col = []
 while ir < irx:
    linea = arr_grsg_c[ir].split()
    if linea[0] == gruppo or gruppo == "":
      gruppo = linea[0]
     val col.append(float(linea[ic]))
    else:
      i_gruppo = found(gruppo)
      linea_cv = arr_cv[i_gruppo]
     media_v = abs(mean(val_col))
      if media_v == 0.0:
         media v = 0.0000000001
      std v = sd(val col)
      cv_v = std_v / media_v
                                                 # cv value
      linea cv[ic-2] = cv v
      linea_cv[len(linea_cv)-1] = len(val_col) # number of records
      val_col = []
      val_col.append(float(linea[ic]))
      gruppo = linea[0]
    ir += 1
  i gruppo = found(gruppo)
  linea_cv = arr_cv[i_gruppo]
  media v = abs(mean(val col))
  if media_v == 0.0:
   media_v = 0.000000001
  std_v = sd(val_col)
  cv_v = std_v / media_v
  linea cv[ic-2] = cv v
                                                  # cv value
     KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 100 di 112
```

```
linea cv[len(linea cv)-1] = len(val col)
                                               # number of records
 ic += 1
ir = 0
irx = len(arr_cv)
linea_cv = arr_cv[0]
icx = len(linea_cv) - 2
ic = 1
num rec1 = 0
while ir < irx:
                                                 # rows mean
 media riga = 0.0
 ic = 1
 num col1 = 0
  linea_cv = arr_cv[ir]
 while ic < icx:
   media_riga += float(linea_cv[ic])
   num col1 += 1
   ic += 1
  linea_cv[icx] = media_riga / num_col1
  num_rec1 += linea_cv[icx + 1]
 ir += 1
ir = 0
ic = 1
while ic < icx:
                                 # weighted mean of columns
 media_col = 0.0
 ir = 0
 num rec1 = 0
 while ir < irx - 1:
    linea cv = arr cv[ir]
   media col = media col + linea cv[ic] * linea cv[icx+1] # linea cv[icx+1] = number
of records
   num_rec1 = num_rec1 + linea_cv[icx+1]
   ir += 1
  linea_cv = arr_cv[irx - 1]
  linea cv[ic] = media col / num rec1
  ic += 1
# updating mean of the row
linea cv = arr cv[irx - 1]
linea_means = linea_cv[1:icx]
media riga = mean(linea means)
linea_cv[icx] = media_riga
                              # Total mean
                                  # n° of records
linea_cv[icx + 1] = num_rec1
cv_media_gen_after = str(media_riga)
cv_media_gen_after = cv_media_gen_after[0:6]
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 101 di 112

```
# write cv file
testata cv = testata
testata cv[0] = "*Groups*"
testata cv.append("*Mean*")
testata cv.append("N recs")
arch cv file = open(arch cv, 'w')
ic = 0
while ic <= icx + 1:
 arch_cv_file.write('%s %s ' % (testata_cv[ic], " "*(9-len(testata_cv[ic]))))
arch cv file.write('%s' % ('\n'))
ir = 0
while ir < irx:
  ic = 0
 linea cv = arr cv[ir]
 while ic <= icx + 1:
    if ic == 0:
      arch_cv_file.write('%s %s ' % (linea_cv[0], " "))
   else:
      if ic <= icx:
        arch cv file.write('%7.4f %s ' % (linea cv[ic], " "))
        arch cv file.write('%6i %s ' % (linea cv[ic], " "))
   ic += 1
  arch cv file.write('%s' % ("\n"))
  ir += 1
ic = 0
media xcv = mean(xcv[1:icx])
while ic <= icx : # print CV input (before catalogue)</pre>
  if ic == 0:
    arch cv file.write('%s %s ' % ("*CVinp*", " "))
  else:
    if ic < icx:
      arch cv file.write('%7.4f %s ' % (xcv[ic], " "))
      arch cv file.write('%7.4f %s ' % (media xcv, " "))
      arch cv file.write('%6i %s ' % (linea cv[ic+1], " "))
  ic += 1
arch cv file.write('%s' % ("\n"))
#======istruzioni aggiunte Roberto Bello 29/02/2012============================
#know_index = str(1.0 - float(cv_media_gen_after) / float(str_med_cv_gen))
     KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 102 di 112
```

```
#know index = know index[0:6]
#arch cv file.write('%s %s %s' % ('*KIndex* ', know index, '\n'))
#======fine istruzioni aggiunte da Roberto Bello 29/02/2012==========
arch_cv_file.close()
# writing out catalog file
testata cat1 = []
testata cat1.append("*Group*")
arch output file = open(arch output, 'w')
ic=0
while ic < icx:
 testata cat1.append(testata_cat[ic])
 ic += 1
ic=0
while ic < len(testata cat1):</pre>
  arch_output_file.write('%s %s ' % (testata_cat1[ic], " "*(15-len(testata_cat1[ic]))))
  ic += 1
arch output file.write('%s ' % ("\n"))
index = 0
while index < len(arr_orig):</pre>
  riga_orig = arr_orig[index]
 ic = 0
  while ic < len(riga orig):
    if not(isinstance(riga_orig[ic],str)):
      riga orig[ic] = str(riga orig[ic])
    ic += 1
  # place before 0 if gr / sg < 10</pre>
  gr1 = str(mtchx[index])
  if mtchx[index] < 10:</pre>
    gr1 = '0' + str(mtchx[index])
  sg1 = str(mtchy[index])
  if mtchy[index] < 10:
    sg1 = '0' + str(mtchy[index])
  arr_rig0 = "G_" + gr1 + "_" + sg1 + " "*8
  arch output file.write('%s ' % (arr rig0))
  ic=0
  while ic < len(riga orig):
    arch output file.write('%s %s ' % (riga orig[ic], " "*(15-len(riga orig[ic]))))
  arch output file.write('%s ' % ("\n"))
  index += 1
testata_cat1 = []
testata cat1.append("*Group*")
testata cat1.append("*RecNum*")
arch sort file = open(arch sort, 'w')
ic=0
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 103 di 112

```
while ic < icx:
  testata_cat1.append(testata_cat[ic])
  ic += 1
ic=0
while ic < len(testata cat1):</pre>
  arch_sort_file.write('%s %s ' % (testata_cat1[ic], " "*(15-len(testata_cat1[ic]))))
  ic += 1
arch_sort_file.write('%s ' % ("\n"))
index = 0
while index < len(arr_grsg_c):</pre>
  riga grsg = arr grsg c[index].split()
  while ic < len(riga grsg):
   val_txt = riga_grsg[ic]
    val txt = val txt[0:13]
    arch_sort_file.write('%s %s ' % (val_txt, " "*(15-len(val_txt))))
  if index < len(arr_grsg_c) - 1:</pre>
    arch sort file.write('%s ' % ("\n"))
  index += 1
arch sort file.close()
# writing out catalog and sorted file
arr outsrt = []
index = 0
while index < len(arr_orig):</pre>
  riga sort = []
  \# place before 0 if gr / sg < 10
  gr1 = str(mtchx[index])
  if mtchx[index] < 10:</pre>
    gr1 = '0' + str(mtchx[index])
  sg1 = str(mtchy[index])
  if mtchy[index] < 10:</pre>
    sg1 = '0' + str(mtchy[index])
  riga_sort.append("G_" + gr1 + "_" + sg1)
  ic = 0
  riga orig = arr orig[index]
  while ic < len(riga_orig):</pre>
    val riga = riga orig[ic]
    riga_sort.append(val_riga)
    ic += 1
  arr_outsrt.append(riga_sort)
  index += 1
for line in arr outsrt:
  line = "".join(line)
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 104 di 112

```
arr_outsrt.sort()
testata_srt = []
testata srt.append("*Group*")
arch_outsrt_file = open(arch_outsrt, 'w')
ic=0
while ic < icx:
 testata srt.append(testata orig[ic])
 ic += 1
ic=0
while ic < len(testata_srt):</pre>
 arch outsrt file.write('%s %s' % (testata srt[ic], " "*(15-len(testata srt[ic]))))
 ic += 1
arch outsrt file.write('%s' % ("\n"))
index = 0
key_gruppo = ""
while index < len(arr_outsrt):</pre>
 riga sort = arr outsrt[index]
 index_c = 0
 while index_c < len(riga_sort):</pre>
   if index_c == 0:
     if riga sort[0] != key gruppo:
       # arch_outsrt_file.write('%s ' % ("\n"))
      key gruppo = riga sort[0]
   valore = riga_sort[index_c]
   arch_outsrt_file.write('%s %s' % (valore, " "*(15-len(valore))))
   index_c += 1
 if index < len(arr_grsg_c) - 1:</pre>
   arch_outsrt_file.write('%s' % ("\n"))
 index += 1
arch outsrt file.close()
print("# KB CLA KNOWLEDGE DISCOVERY IN DATA MINING (CLASSIFY PROGRAM)
                                                                          #")
print("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                                          #")
print("# Language used: PYTHON
                                                                          #")
arch_log_file = open(arch_log, 'w')
arch_log_file.write("%s %s" %
arch log file.write("%s %s" % ("# KB CLA KNOWLEDGE DISCOVERY IN DATA MINING (CLASSIFY
PROGRAM)
                  #", "\n"))
arch_log_file.write("%s %s" % ("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS
RESERVED)
                    #", "\n"))
arch log file.write("%s %s" % ("# Language used: PYTHON
    KB - Neural Data Mining with Python sources - Roberto Bello - Pag. 105 di 112
```

```
#", "\n"))
arch_log_file.write("%s %s" %
arch log file.write("%s %s %s" % ("Input File
", file input, "\n"))
arch_log_file.write("%s %s %s" % ("Numer of Groups (3 - 20)
                                                                               _>
", str(gruppi num), "\n"))
arch log file.write("%s %s %s" % ("Normalization (Max, Std, None)
", tipo norm, "\n"))
arch log file.write("%s %s %s" % ("File Training Grid
                                                                              ->
", arch_grid, "\n"))
arch_log_file.write("%s"
\n"))
arch log file.write("%s %s %s" % ("Output File Classify.original
                                                               ", arch_output,
"\n"))
arch log file.write("%s %s %s" % ("Output File Classify.sort
                                                               ", arch outsrt,
arch log file.write("%s %s %s" % ("Output File Summary sort
                                                               ", arch sort, "\n"))
arch_log_file.write("%s %s %s" % ("Output File Matrix Catal.
                                                               ", arch_catal,
"\n"))
arch log file.write("%s %s %s" % ("Output File Means, STD, CV.
                                                               ", arch medsd,
"\n"))
arch log file.write("%s %s %s" % ("Output File CV of the Groups
                                                               ", arch_cv, "\n"))
arch log file.write("%s %s %s" % ("Output File Training Grid
                                                               ", arch grid, "\n"))
arch log file.write("%s %s %s" % ("Output File Run Parameters
                                                               ", arch log, "\n"))
#======istruzioni aggiunte Roberto Bello 29/02/2012============================
know index = str(1.0 - float(cv media gen after) / float(str med cv gen))
know index = know index[0:6]
arch log file.write('%s %s %s' % ('*KIndex* ', know index, '\n'))
#=======fine istruzioni aggiunte da Roberto Bello 29/02/2012============
print
print 'Output File Classify.original ' + arch output
print 'Output File Classify.sort
                                 ' + arch outsrt
print 'Output File Summary sort
                                 ' + arch sort
print 'Output File Matrix Catal.
                                 ' + arch catal
print 'Output File Means, STD, CV.
                                  ' + arch medsd
print 'Output File CV of the Groups ' + arch cv
print 'Output File Training Grid
                                  ' + arch grid
print 'Output File Run Parameters
                                 ' + arch log
print 'CV after Catalog
                                 ' + cv media gen after
know index = str(1.0 - float(cv media gen after) / float(str med cv gen))
know_index = know_index[0:6]
print 'Knowledge Index
                                ' + know index
print
# Elapsed time
t1 = datetime.datetime.now()
elapsed_time = t1 - t0
```

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 106 di 112

```
print "Elapsed time (seconds) : " + str(elapsed_time.seconds) + "." +
str(elapsed_time.microseconds)
print
```

Appendix 4 – KB_RND source

```
# -*- coding: utf-8 -*-
import os
import random
import copy
import datetime
print("# KB RND KNOWLEDGE DISCOVERY IN DATA MINING (RANDOM FILE SIZE REDUCE)
print("# by ROBERTO BELLO (COPYRIGHT MARCH 2011 ALL RIGHTS RESERVED)
                                                                         #")
print("# Language used: PYTHON
                                                                         #")
# input and run parameters
error = 0
while True:
                                                       : ')
 arch input = raw input('InputFile
 if not os.path.isfile(arch input):
   print("Oops! File does not exist. Try again... or CTR/C to exit")
 else:
   break
while True:
                                                        : ')
 arch_output = raw_input('OutputFile
 if os.path.isfile(arch output):
   print("Oops! File does exist. Try again... or CTR/C to exit")
 else:
   break
while True:
 try:
                                                             : '))
   num_cells_out = int(raw_input('Out number of cells (<= 90000)</pre>
 except ValueError:
   print("Oops! That was no valid number. Try again...")
 else:
   if(num_cells_out > 90000):
     print("Oops! Number of Cells too big. Try again...")
   else:
     break
# start time
```

```
t0 = datetime.datetime.now()
# read input file
arr_r
      = []
arr_rnd = []
arr_out = []
xnomi
       = []
index
      = 0
for line in open(arch_input).readlines():
  linea = line.split()
  if(index == 0):
    xnomi.append(linea)
  else:
   arr r.append(linea)
  index += 1
rec input = index - 1
num_cols = len(linea)
num_records_out = int(num_cells_out / num_cols)
print "Nun. Records Input " + str(rec_input)
if rec_input < num_records_out:</pre>
 num_records_out = rec_input
# random values sequence
ix = 950041
                                     # integer as random seed
                                            # initial value of random seed to obtain the
random.seed(ix)
same sequences in new runs
index = 0
while index < num records out:
  val rnd = int(random.random()*rec input)
  doppio = 0
  for index_rnd, item in enumerate(arr_rnd): # check for duplicates
    if item == val rnd:
      doppio = 1
  if doppio == 0:
    arr rnd.append(val rnd)
    index += 1
# arr_out writing
index = 0
arr_out.append(xnomi[0])
                                      # header
while index < len(arr_rnd):</pre>
```

```
key1 = arr rnd[index]
                                           # from source to random output
  arr_out.append(arr_r[key1])
  index += 1
# write arch out file
arch out file = open(arch output, 'w')
index = 0
while index < len(arr out):</pre>
  line = arr out[index]
  ncol = 0
  while ncol < len(line):
    field = line[ncol].strip()
    if ncol < len(line) - 1:
      arch out file.write('%s%s' % (field, "\t"))
    else:
      arch_out_file.write('%s%s' % (field, "\n"))
    ncol += 1
  index += 1
arch out file.close()
# Elapsed time
t1 = datetime.datetime.now()
elapsed\_time = t1 - t0
print "Elapsed time (microseconds) : " + str(elapsed time.microseconds)
print
```

KB – Guarantee and copyright

The author guarantees that the work is without viruses and malicious codes also considering:

- the text is in pdf format
- the python programs are in txt format and do not contain malicious code, as easily verifiable by a simple reading of their
- the test files are in a txt format
- the language for the processing of the programs (python) is of Open Source type and is universally recognised as reliable and safe.

As regards the copyright, the author does not intend to renounce his legal rights on the algorithms and on the method of computing and analysis contained in the KB programs.

Roberto Bello

Graduate in Economics and Commerce with specialization in Operations Research Data Scientist, expert in Knowledge Mining, in Data Mining and in Open Source ICT Strategist of the ClubTI of Milan (www.clubtimilano.net) Researcher of the AISF (www.accademiascienzeforensi.it) Expert (CTP) and ex CTU (Technical Office Consultant) of the Court of Milan

KB – Neural Data Mining with Python sources – Roberto Bello - Pag. 109 di 112

Author of professional publications available on www.lulu.com/spotlight/robertobb Founding associate of AIPI (Italian Professional Computer Association) In the past CIO of Plasmon, of Wrangler in Italy and consultant for the most important Italian food companies

Linkedin: it.linkedin.com/pub/roberto-bello/4/1a5/677

Table of contents

KB – Neural Data Mining with Python sources.	1
Introduction	1
Business Intelligence and the bad use of statistics.	2
Learning by induction and the neural networks	3
KB	6
Python for KB	6
Details	7
Collecting and arranging input data	7
General warnings for using the KB programs	10
KB_CAT Knowledge Data Mining and cataloging into homogeneous groups	10
Generality, aims and functions.	10
Source of KB_CAT (see attachment 1)	11
Test Input file (copy and paste then save with name vessels.txt); fields separated by	
tabulation	
How to run	
Input File = vessels.txt	
Number of Groups $(3-20) = 3$	12
Normalization (Max, Std, None) = m	
Start Value of alpha (from 1.8 to 0.9) = 0.9	
End Value of alpha (from 0.5 to 0.0001) = 0.001	
Decreasing step of alpha (from 0.1 to 0.001) = 0.001	
Forced shut down of processing.	
KB_CAT produce the following output:	
In the window DOS Windows (or the Terminal Linux)	
File - Output/Catalog.original (vessels_M_g3_out.txt)	
File of Output/Catalog.sort (vessels_M_g3_outsrt.txt)	
Output/Means, Std, CV (vessels_M_g3_medsd.txt)	
Output/CV files (vessels_M_g3_cv.txt)	
Output/Training Grid (vessels_M_g3_grid.txt)	
Statistical analysis of the results of the cataloging	
Other input files to KB_CAT (animals.txt)	
Processing of animals.txt file with KB_CAT	
Output file/Catalog.sort ordered by group using animals.txt	
Verify the validity of a manual cataloging.	
Input file to KB_CAT (animals_d.txt)	
The processing was carried out with the following parameters:	
Results obtained processing animals_d.txt (Output/Catalog.sort)	37
Comparison of the results of the automatic cataloging of iris.txt to those recognized by	20
botanists	
Clinical trials on hepatitis B virus	
Not good, but better! (2006 mail)	
KB_STA – the statistical analysis of the result of the cataloging	
Generalities, aims and functions	
Source of KB_STA (see attachment 2)	
How to use	
KB_STA running.	
Analysis of the results of the cataloging of vessels.txt	
Analysis of the results of a political poll of 2007	
_	
Generalities, aims and functions	
Source of KB_CLA (attachment 3)	
110W tO 1UII	49

Input files	= n_vessels.txt	49
Contents of the file n_vessels.txt		49
Number of Groups $(3-20)$	= 3	50
Normalization(Max, Std, None)	= m	50
File Training Grid	= vessels_M_g3_grid.txt	50
KB_CLA running		50
Analysis of the results of the classification of n vessels.txt		
Political opinions in Facebook (January 2013)		51
Know4Business (Cloud version in Google App Engine)		
APPENDIXES		56
Appendix 1 – KB_CAT source	56	
Appendix 2 – KB STA source		
Appendix 3 – KB CLA source		88
Appendix 4 – KB RND source		107
KB – Guarantee and copyright.		
Roberto Bello		109