

Assignment No - A6

1 TITLE :

Select an Industrial sector and write an application using BIA tool for maximizing the profit.

- Design an application using tool for clustering the iris flowers dataset.
- Design an application using tool to predict whether to play tennis or not.

2 MATHEMATICAL MODEL :

Mathematical model is given as below,

S=fs,e,X,Y,Fme,DD,NDD,Mem_shared

Where,

s = Initial State

e = End State

X=Input

Datasets of iris flowers and tennis.

Y=Ooutput

Clustering of iris flowers and prediction whether to play tennis or not,
using the interactive table, scatterplot, histogram, etc.

Fme=Algorithms used (K-means and Nave-Bayes)

- K-means Algorithm : (Used for iris dataset)
 1. Randomly select c cluster centers.
 2. Calculate the distance between each data point and cluster centers.
 3. Assign the data point to the cluster center whose distance from the cluster center is minimum of all the cluster centers..
 4. Recalculate the new cluster center using:

$$v_i = (1/c_i) \sum_{j=1}^{c_i} x_i$$

where, c_i represents the number of data points in i th cluster.

5. Recalculate the distance between each data point and new obtained cluster centres.
6. If no data point was reassigned then stop, otherwise repeat from step 3.

- Naive-Bayes Algorithm : (Used for tennis dataset)

$$P(c|x) = \frac{P(x|c)P(c)}{P(x)}$$

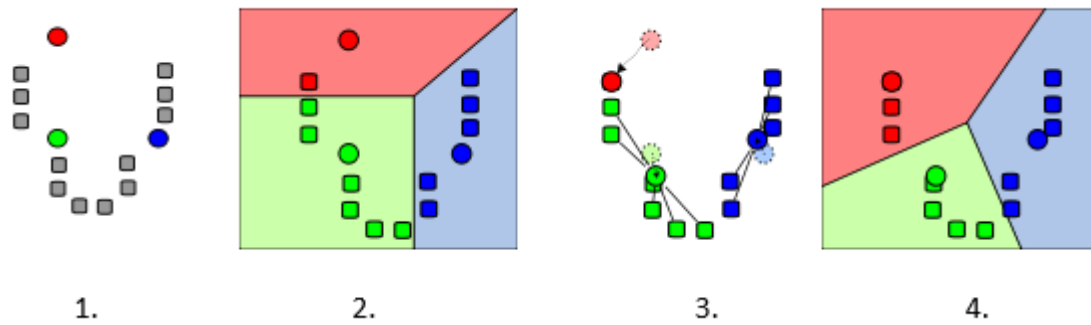
Likelihood
Class Prior Probability

Posterior Probability
Predictor Prior Probability

$$P(c|X) = P(x_1|c) \times P(x_2|c) \times \dots \times P(x_n|c) \times P(c)$$

Fig : Bayes Rule

K-means example :



1. k initial "means" (in this case k=3) are randomly generated within the data domain (shown in color)
2. k clusters are created by associating every observation with the nearest mean. The partitions here represent the diagram generated by the means
3. The centroid of each of the k clusters becomes the new mean.
4. Steps 2 and 3 are repeated until convergence has been reached.

Naive-Bayes Example :

Outlook	Temperature	Humidity	Windy	Class
sunny	hot	high	false	N
sunny	hot	high	true	N
overcast	hot	high	false	P
rain	mild	high	false	P
rain	cool	normal	false	P
rain	cool	normal	true	N
overcast	cool	normal	true	P
sunny	mild	high	false	N
sunny	cool	normal	false	P
rain	mild	normal	false	P
sunny	mild	normal	true	P
overcast	mild	high	true	P
overcast	hot	normal	false	P
rain	mild	high	true	N

Figure 1: Given a training set, we can compute the probabilities

Outlook	P	N		Humidity	P	N
sunny	2/9	3/5		high	3/9	4/5
overcast	4/9	0		normal	6/9	1/5
rain	3/9	2/5				
Temperature				Windy		
hot	2/9	2/5		true	3/9	3/5
mild	4/9	2/5		false	6/9	2/5
cool	3/9	1/5				

$X = \langle \text{sunny}, \text{mild}, \text{high}, \text{true} \rangle$

$Pr(X|no).Pr(no) = (3/5.2/5.4/5.3/5).5/14 = 0.04$

$Pr(X|yes).Pr(yes) = (2/9.4/9.3/9.3/9).9/14 = 0.007$

DD=Deterministic Data

Iris dataset, Tennis Dataset.

NDD=Non -Deterministic Data

Number of clusters, etc.

Mem-shared=memory shared by the applications.

3 THEORY :

3.1 Introduction to Business Analytics and Intelligence :

Business Analytics :

Business analytics (BA) is the practice of iterative, methodical exploration of an organization's data with emphasis on statistical analysis.

Business analytics is used by companies committed to data-driven decision making.

Business Intelligence is querying, reporting, OLAP, and alert tools can answer questions such as what happened, how many, how often, where the problem is, and what actions are needed. Business analytics can answer questions like why is this happening, what if these trends continue, what will happen next (that is, predict), what is the best that can happen (that is, optimize).

Some examples of BA are :

- Exploring data to find new patterns and relationships (data mining)
- Forecasting future results (predictive modeling, predictive analytics)
- Explaining why a certain result occurred (statistical analysis, quantitative analysis)

Business Intelligence :

Business intelligence (BI) is often described as "the set of techniques and tools for the transformation of raw data into meaningful and useful information for business analysis purposes".

BI technologies are capable of handling large amounts of unstructured data to help identify, develop and otherwise create new strategic business opportunities. The goal of BI is to allow for the easy interpretation of these large volumes of data. Identifying new opportunities and implementing an effective strategy based on insights can provide businesses with a competitive market advantage and long-term stability.

BI technologies provide historical, current and predictive views of business operations. Common functions of business intelligence technologies are reporting, online analytical processing, analytics, data mining, process mining, complex event processing, business performance management, benchmarking, text mining, predictive analytics and prescriptive analytics.

BI can be used to support a wide range of business decisions ranging from operational to strategic.

KNIME tool :

KNIME, the Konstanz Information Miner, is an open source data analytics, reporting and integration platform. KNIME integrates various components for machine learning and data mining through its modular data pipelining concept. A graphical user interface allows assembly of nodes for data pre-processing *ETL : Extraction, Transformation, Loading*, for modelling and data analysis and visualization. Since 2006, KNIME has been used in pharmaceutical research,[2] but is also used in other areas like CRM customer data analysis, business intelligence and financial data analysis.

Innovative organizations use our open-source, enterprise-grade analytics platform to discover the potential hidden in their data, mine for fresh insights or predict new futures. It is quick to deploy, easy to scale, and intuitive, KNIME is used in over 60 countries on data of every kind: from numbers to images, molecules to humans, signals to complex networks, from kilo- to petabytes, or simple reports to complex analyses.

4 Conclusion :

Hence, we have designed an application using tool for categorizing the iris flowers dataset and an application to predict whether to play tennis or not.

5 References

1. Data Mining Concepts and Techniques, Second Edition, Jiawei Han, Micheline Kamber, Jian Pei.
2. <https://www.knime.org/knime>
3. <http://searchdatamanagement.techtarget.com/definition/business-intelligence>
4. http://www.webopedia.com/TERM/B/Business_Intelligence.html
5. <http://searchbusinessanalytics.techtarget.com/definition/business-analytics-BA>

- Input:
- Training and Input Data:

Table View - 35 Interactive Table

File: Hiltite Navigation View: Output

Row ID	D Col0	D Col1	D Col2	D Col3	S Col4	S Cluster
Row0	5.1	3.5	1.4	0.2	Iris-setosa	cluster_2
Row1	4.9	3	1.4	0.2	Iris-setosa	cluster_2
Row2	4.7	3.2	1.3	0.2	Iris-setosa	cluster_2
Row3	4.6	3.1	1.5	0.2	Iris-setosa	cluster_2
Row4	5	3.6	1.4	0.2	Iris-setosa	cluster_2
Row5	5.4	3.9	1.7	0.4	Iris-setosa	cluster_2
Row6	4.6	3.4	1.4	0.3	Iris-setosa	cluster_2
Row7	5	3.4	1.5	0.2	Iris-setosa	cluster_2
Row8	4.4	2.9	1.4	0.2	Iris-setosa	cluster_2
Row9	4.9	3.1	1.5	0.1	Iris-setosa	cluster_2
Row10	5.4	3.7	1.5	0.2	Iris-setosa	cluster_2
Row11	4.8	3.4	1.6	0.2	Iris-setosa	cluster_2
Row12	4.8	3	1.4	0.1	Iris-setosa	cluster_2
Row13	4.2	3	1.1	0.1	Iris-setosa	cluster_2
Row14	5.9	4	1.2	0.2	Iris-setosa	cluster_2
Row15	5.7	4.4	1.5	0.4	Iris-setosa	cluster_2
Row16	5.4	3.9	1.3	0.4	Iris-setosa	cluster_2
Row17	5.1	3.5	1.4	0.3	Iris-setosa	cluster_2
Row18	5.7	3.8	1.7	0.3	Iris-setosa	cluster_2
Row19	5.1	3.8	1.5	0.3	Iris-setosa	cluster_2
Row20	5.4	3.4	1.7	0.2	Iris-setosa	cluster_2
Row21	5.1	3.7	1.5	0.4	Iris-setosa	cluster_2
Row22	4.6	3.6	1	0.2	Iris-setosa	cluster_2
Row23	5.1	3.5	1.7	0.5	Iris-setosa	cluster_2
Row24	4.8	3.4	1.9	0.2	Iris-setosa	cluster_2
Row25	7	3.2	4.7	1.4	Iris-versicol	cluster_1
Row26	6.4	3.2	4.5	1.5	Iris-versicol	cluster_1
Row27	6.9	5.1	4.9	1.5	Iris-versicol	cluster_0
Row28	5.5	3.2	4	1.3	Iris-versicol	cluster_1
Row29	6.5	2.8	4.6	1.5	Iris-versicol	cluster_1
Row30	5.7	2.8	4.5	1.3	Iris-versicol	cluster_1
Row31	6.3	3.3	4.7	1.6	Iris-versicol	cluster_1
Row32	4.9	2.4	3.3	1	Iris-versicol	cluster_1
Row33	5.6	2.9	4.6	1.3	Iris-versicol	cluster_1
Row34	5.2	2.7	3.9	1.4	Iris-versicol	cluster_1
Row35	5	2	3.5	1	Iris-versicol	cluster_1
Row36	5.5	3	4.2	1.5	Iris-versicol	cluster_1
Row37	6	2.2	4	1	Iris-versicol	cluster_1
Row38	6.1	2.9	4.7	1.4	Iris-versicol	cluster_1
Row39	5.6	2.9	3.6	1.3	Iris-versicol	cluster_1
Row40	6.7	3.1	4.4	1.4	Iris-versicol	cluster_1

table1 kmeans.png

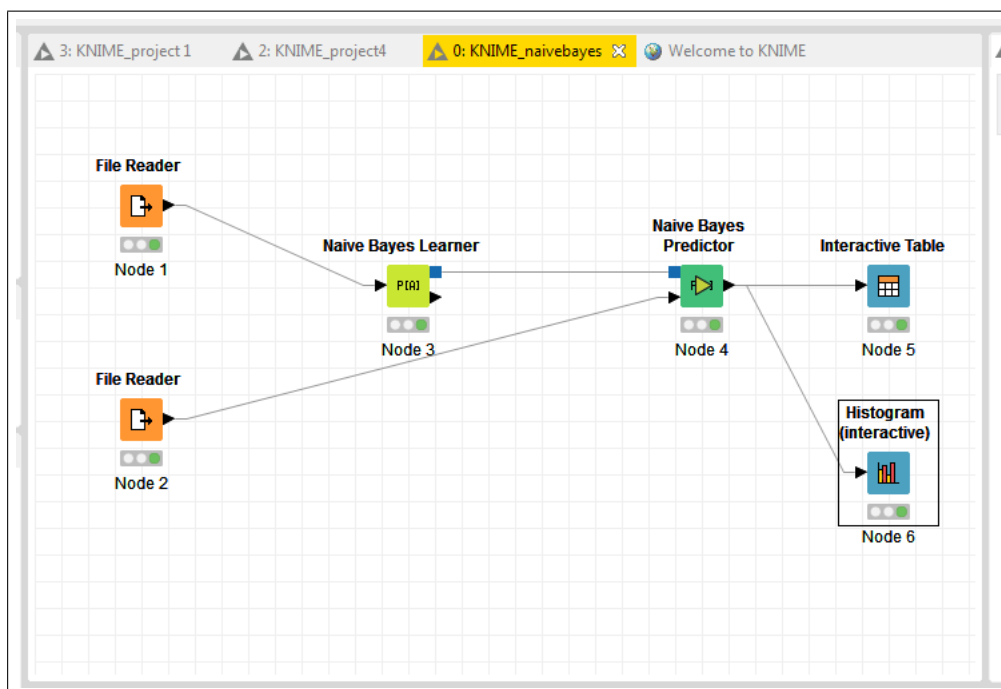
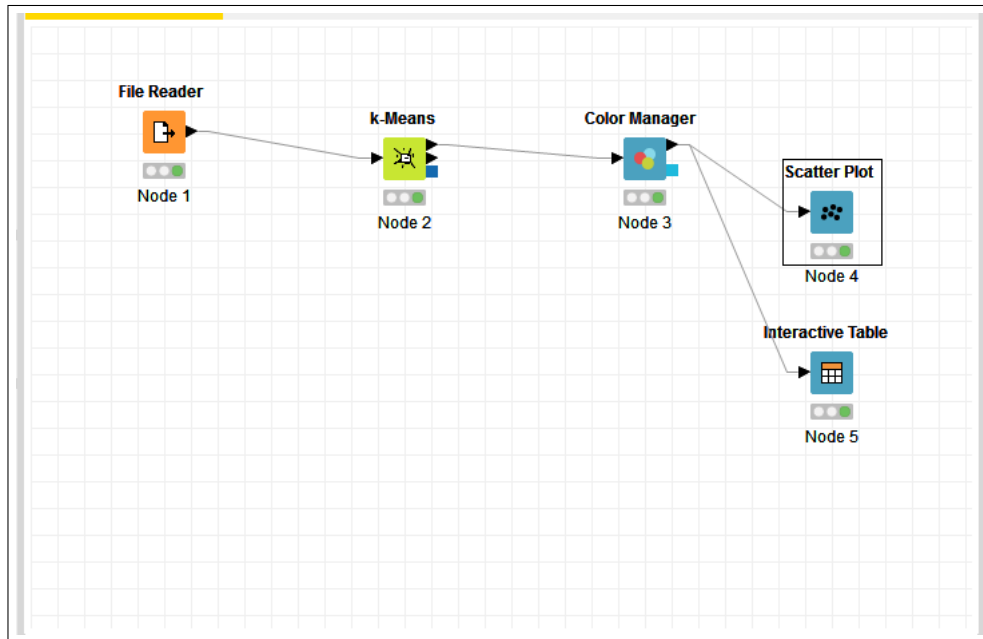
Table View - 35 Interactive Table

File: Hiltite Navigation View: Output

Row ID	D Col0	D Col1	D Col2	D Col3	S Col4	S Cluster
Row34	5.2	2.7	3.9	1.4	Iris-versicol	cluster_1
Row35	5	2	3.5	1	Iris-versicol	cluster_1
Row36	5.5	3	4.2	1.5	Iris-versicol	cluster_1
Row37	6	2.2	4	1	Iris-versicol	cluster_1
Row38	6.1	2.9	4.7	1.4	Iris-versicol	cluster_1
Row39	5.6	2.9	3.6	1.3	Iris-versicol	cluster_1
Row40	6.7	3.1	4.4	1.4	Iris-versicol	cluster_1
Row41	5.6	3	4.5	1.5	Iris-versicol	cluster_1
Row42	5.8	2.7	4.1	1	Iris-versicol	cluster_1
Row43	6.2	3.2	4.5	1.5	Iris-versicol	cluster_1
Row44	5.6	2.5	3.9	1.1	Iris-versicol	cluster_1
Row45	5.9	2.2	4.8	1.8	Iris-versicol	cluster_1
Row46	6.1	2.8	4	1.3	Iris-versicol	cluster_1
Row47	6.3	2.5	4.9	1.5	Iris-versicol	cluster_1
Row48	6.1	2.9	4.7	1.2	Iris-versicol	cluster_1
Row49	6.4	2.9	4.3	1.3	Iris-versicol	cluster_1
Row50	6.3	2.3	8	2.5	Iris-virginica	cluster_0
Row51	5.8	2.7	6.1	1.9	Iris-virginica	cluster_1
Row52	7.1	3	5.9	2.1	Iris-virginica	cluster_0
Row53	6.3	3.9	5.6	1.8	Iris-virginica	cluster_0
Row54	6.5	3	5.8	2.2	Iris-virginica	cluster_0
Row55	7.6	3	6.6	2.1	Iris-virginica	cluster_0
Row56	4.9	2.5	4.5	1.7	Iris-virginica	cluster_1
Row57	7.3	2.9	6.3	1.8	Iris-virginica	cluster_0
Row58	6.7	2.5	5.8	1.8	Iris-virginica	cluster_0
Row59	7.2	3.6	6.1	2.5	Iris-virginica	cluster_0
Row60	6.5	3.2	5.1	2	Iris-virginica	cluster_0
Row61	6.4	2.7	6.3	1.9	Iris-virginica	cluster_0
Row62	6.8	3	5.5	2.1	Iris-virginica	cluster_0
Row63	5.7	3.5	5	2	Iris-virginica	cluster_1
Row64	5.8	2.8	5.1	2.4	Iris-virginica	cluster_1
Row65	6.4	3.2	5.3	2.3	Iris-virginica	cluster_0
Row66	6.5	3	5.5	1.8	Iris-virginica	cluster_0
Row67	7.7	3.8	6.7	2.2	Iris-virginica	cluster_0
Row68	7.7	3.6	6.9	2.3	Iris-virginica	cluster_0
Row69	6	2.2	5	1.5	Iris-virginica	cluster_1
Row70	6.9	3.2	5.7	2.3	Iris-virginica	cluster_0
Row71	5.6	2.8	4.9	2	Iris-virginica	cluster_1
Row72	7.7	2.8	6.7	2	Iris-virginica	cluster_0
Row73	6.3	3.7	4.9	1.8	Iris-virginica	cluster_1
Row74	6.7	3.3	5.7	2.1	Iris-virginica	cluster_0

table2 kmeans.png

- Knime Design:
Training and Input Data:



- Output

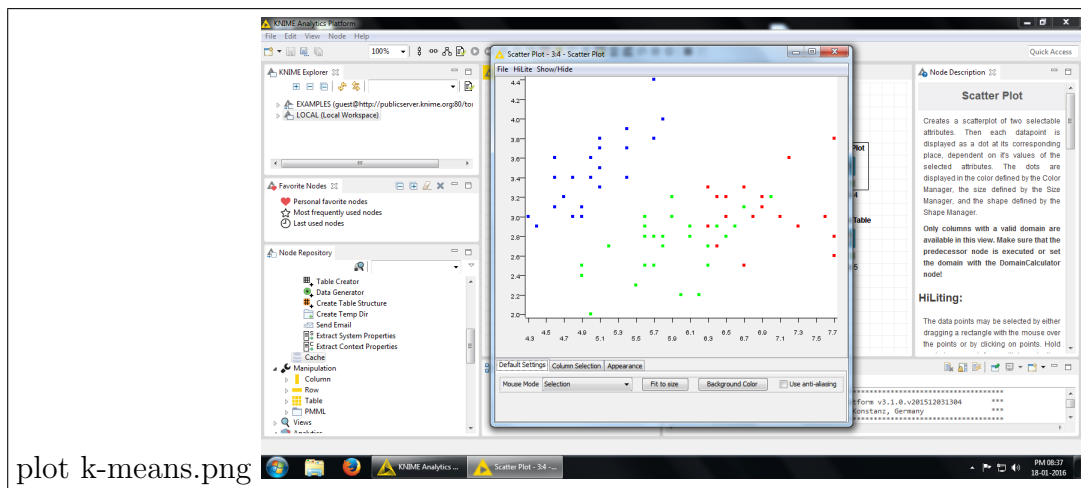
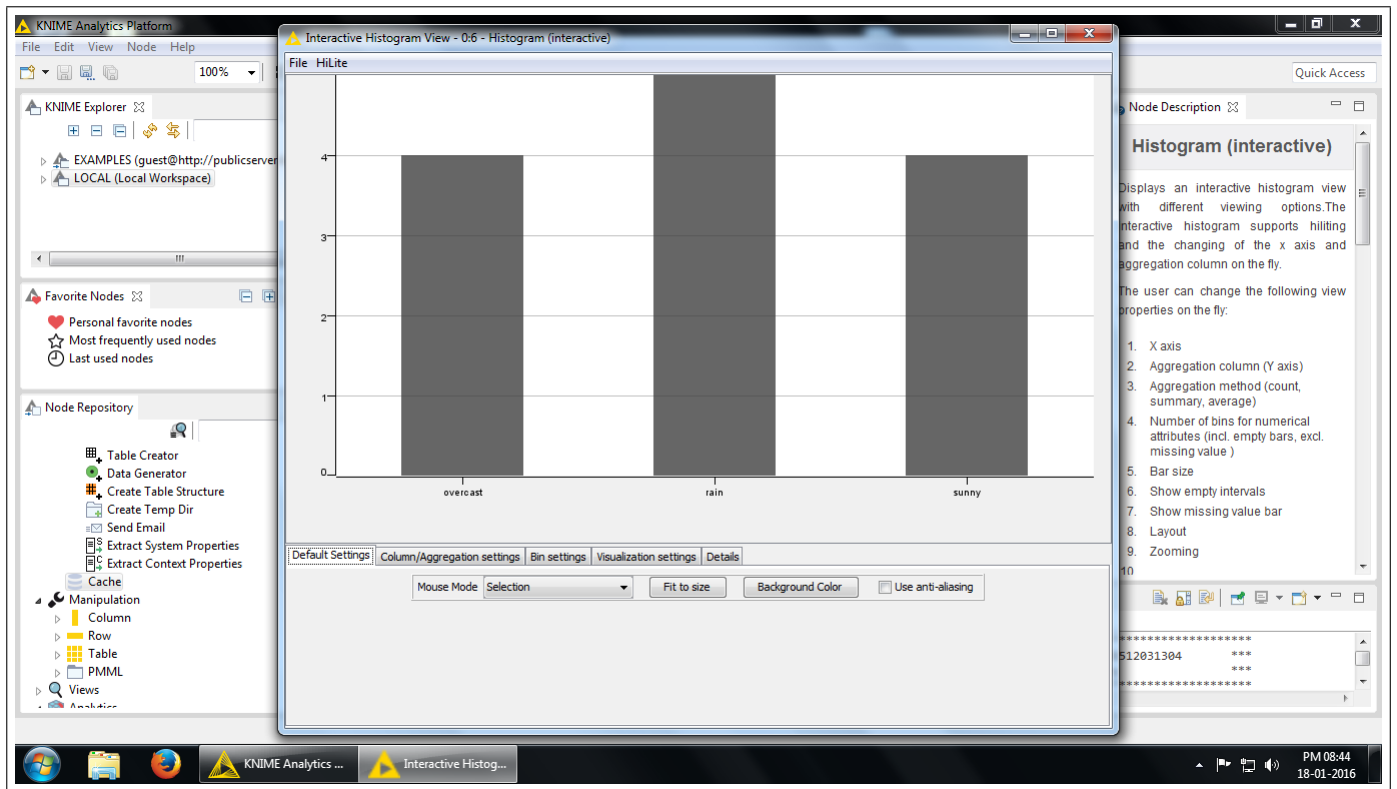


Table View - 0:5 - Interactive Table					
File Hilite Navigation View Output					
Row ID	S sunny	S hot	S high	S weak	S Predict...
D2	sunny	hot	high	strong	yes
D3	overcast	hot	high	weak	yes
D4	rain	mild	high	weak	yes
D5	rain	cool	normal	weak	yes
D6	rain	cool	normal	strong	yes
D7	overcast	cool	normal	strong	yes
D8	sunny	mild	high	weak	yes
D9	sunny	cool	normal	weak	yes
D10	rain	mild	normal	weak	yes
D11	sunny	mild	normal	strong	yes
D12	overcast	mild	high	strong	yes
D13	overcast	hot	normal	weak	yes
D14	rain	mild	high	strong	yes



Roll No.	Name of Student	Date of Performance	Date of Submission	Sign.
302	Abhinav Bakshi	15/12/2015	5/01/2016	