An Introduction to the WEKA Data Mining System

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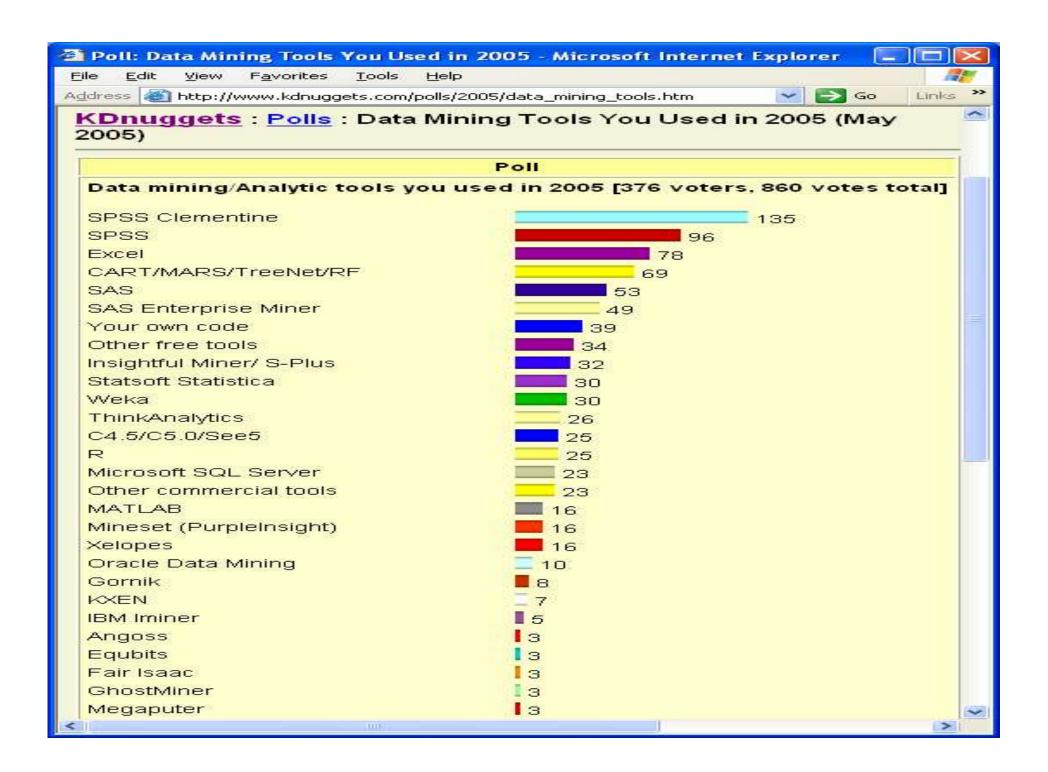
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Data Mining

- "Drowning in Data yet Starving for Knowledge" ???
- "Computers have promised us a fountain of wisdom but delivered a flood of data" William J. Frawley, Gregory Piatetsky-Shapiro, and Christopher J. Matheus
- Data Mining: "The non trivial extraction of implicit, previously unknown, and potentially useful information from data"
 William J Frawley, Gregory Piatetsky-Shapiro and Christopher J Matheus
- Data mining finds valuable information hidden in large volumes of data.
- Data mining is the analysis of data and the use of software techniques for finding patterns and regularities in sets of data.
- Data Mining is an interdisciplinary field involving:
 - Databases
 - Statistics
 - Machine Learning
 - High Performance Computing
 - Visualization
 - Mathematics

Data Mining Software

- **KDnuggets**: Polls : Data Mining Tools You Used in 2005 (May 2005) PollData mining/Analytic tools you used in 2005 [376 voters, 860 votes total]
- Enterprise-level: (US \$10,000 and more)
 Fair Isaac, IBM, Insightful, KXEN, Oracle, SAS, and SPSS
- Department-level: (from \$1,000 to \$9,999)
 Angoss, CART/MARS/TreeNet/Random Forests,
 Equbits, GhostMiner, Gornik, Mineset, MATLAB,
 Megaputer, Microsoft SQL Server, Statsoft Statistica,
 ThinkAnalytics
- Personal-level: (from \$1 to \$999): Excel, See5
- Free: C4.5, R, Weka, Xelopes



Weka Data Mining Software

KDnuggets: News: 2005: n13: item2

SIGKDD Service Award is the highest service award in the field of data mining and knowledge discovery. It is is given to one individual or one group who has performed significant service to the data mining and knowledge discovery field, including professional volunteer services in disseminating technical information to the field, education, and research funding.

The **2005 ACM SIGKDD Service Award** is presented to **the Weka team** for their development of the freely-available Weka Data Mining Software, including the accompanying book Data Mining: Practical Machine Learning Tools and Techniques (now in second edition) and much other documentation.

The Weka team includes **Ian H. Witten** and **Eibe Frank**, and the following major contributors (in alphabetical order of last names): Remco R. Bouckaert, John G. Cleary, Sally Jo Cunningham, Andrew Donkin, Dale Fletcher, Steve Garner, Mark A. Hall, Geoffrey Holmes, Matt Humphrey, Lyn Hunt, Stuart Inglis, Ashraf M. Kibriya, Richard Kirkby, Brent Martin, Bob McQueen, Craig G. Nevill-Manning, Bernhard Pfahringer, Peter Reutemann, Gabi Schmidberger, Lloyd A. Smith, Tony C. Smith, Kai Ming Ting, Leonard E. Trigg, Yong Wang, Malcolm Ware, and Xin Xu.

The Weka team has put a tremendous amount of effort into continuously developing and maintaining the system **since 1994**. The development of Weka was funded by a grant from the New Zealand Government's Foundation for Research, Science and Technology.

The **key features** responsible for Weka's success are:

- it provides many different algorithms for data mining and machine learning
- is is open source and freely available
- it is platform-independent
- it is easily useable by people who are not data mining specialists
- it provides flexible facilities for scripting experiments
- it has kept up-to-date, with new algorithms being added as they appear in the research literature.

Weka Data Mining Software

KDnuggets: News: 2005: n13: item2 (cont.)

- The Weka Data Mining Software has been downloaded **200,000 times** since it was put on SourceForge in April 2000, and is currently downloaded at a rate of 10,000/month. The Weka mailing list has over **1100** subscribers in **50** countries, including subscribers from many major companies.
- There are **15 well-documented substantial projects** that incorporate, wrap or extend Weka, and no doubt many more that have not been reported on Sourceforge.
- Ian H. Witten and Eibe Frank also wrote a **very popular book "Data Mining: Practical Machine Learning Tools and Techniques"** (now in the second edition), that seamlessly integrates Weka system into teaching of data mining and machine learning. In addition, they provided **excellent teaching material** on the book website.
- This book became one of the most popular textbooks for data mining and machine learning, and is **very frequently cited in scientific publications**.
- Weka is a **landmark system in the history of the data mining and machine learning** research communities, because it is the only toolkit that has gained such widespread adoption and survived for an extended period of time (the first version of Weka was released 11 years ago). Other data mining and machine learning systems that have achieved this are individual systems, such as C4.5, not toolkits.
- Since Weka is freely available for download and offers many powerful features (sometimes not found in commercial data mining software), it has become one of the most widely used data mining systems. Weka also became one of the favorite vehicles for data mining research and helped to advance it by making many powerful features available to all.

In sum, the Weka team has made an outstanding contribution to the data mining field.



Interested in doing an MSc or PhD in Machine Learning here at Waikato and spending some time overseas while working on your project? Then check this out.

Weka Machine Learning Project

An exciting and potentially far-reaching development in computer science is the invention and application of methods of machine learning. These enable a computer program to automatically analyse a large body of data and decide what information is most relevant. This crystallised information can then be used to automatically make predictions or to help people make decisions faster and more accurately.

The overall goal of our project is to build a state-of-the-art facility for developing machine learning (ML) techniques and to apply them to real-world data mining problems. Our team has incorporated several standard ML techniques into a software "workbench" called WEKA, for Waikato Environment for Knowledge Analysis. With it, a specialist in a particular field is able to use ML to derive useful knowledge from databases that are far too large to be analysed by hand. WEKA's users are ML researchers and industrial scientists, but it is also widely used for teaching.

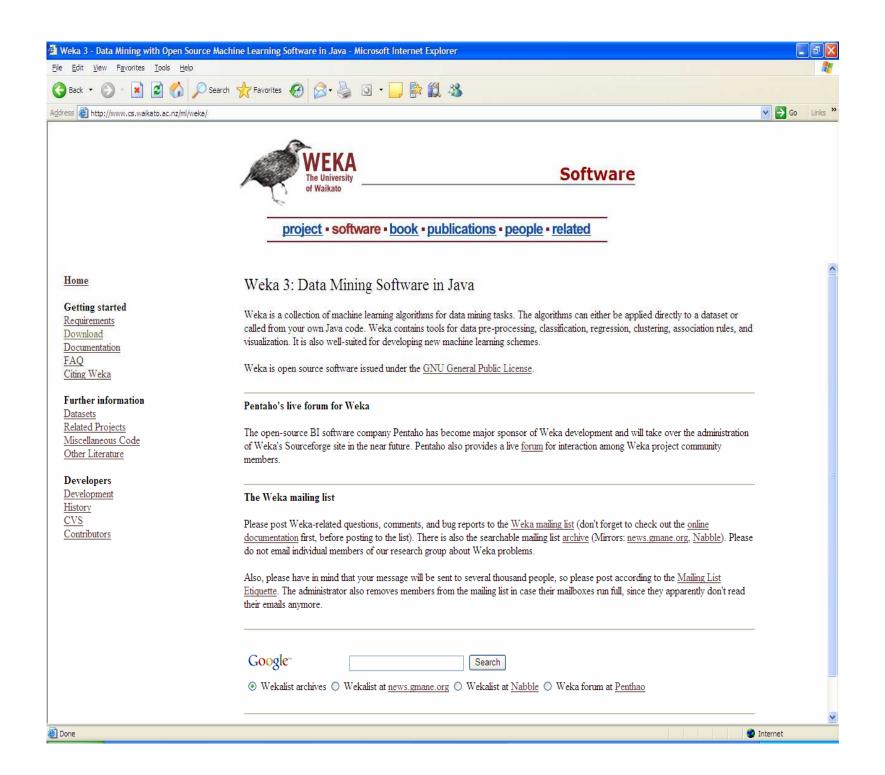
Our objectives are to

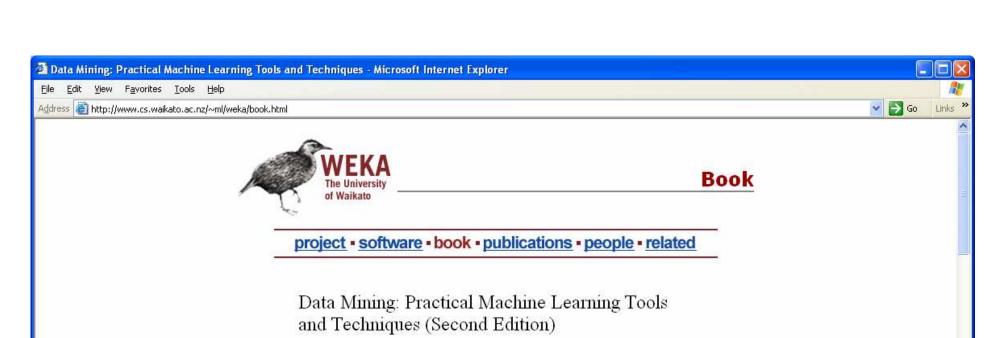
- make ML techniques generally available;
- apply them to practical problems that matter to New Zealand industry;
- develop new machine learning algorithms and give them to the world;
- contribute to a theoretical framework for the field.

Our machine learning package is publically available and presents a collection of algorithms for solving real-world data mining problems. The software is written entirely in Java and includes a uniform interface to a number of standard ML techniques. Please feel free to browse around.

Found only on the islands of New Zealand, the weka is a flightless bird with an inquistive nature. (How should you pronounce <u>WEKA</u>? What does the weka <u>sound</u> like.)





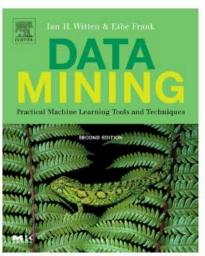


Ian H. Witten, Eibe Frank

Morgan Kaufmann June 2005 525 pages Paper ISBN 0-12-088407-0



Eibe Frank and Ian Witten



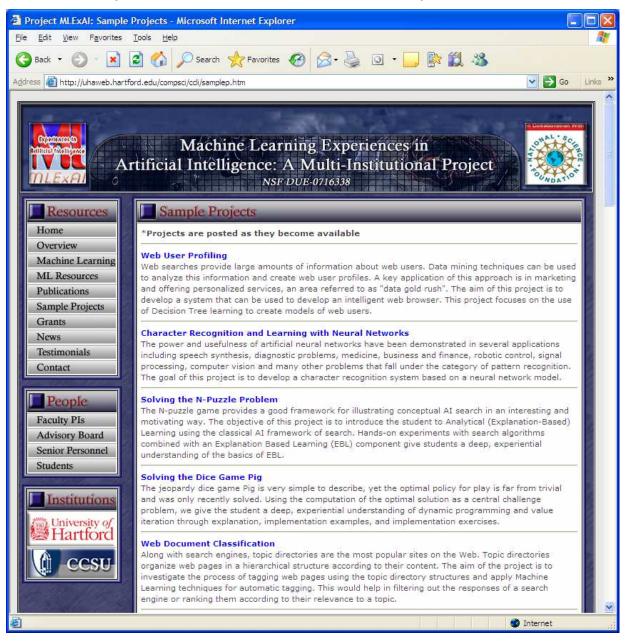
Click here to order from Amazon.com

Comments

"If you have data that you want to analyze and understand, this book and the associated Weka toolkit are an excellent way to start."

-Jim Gray, Microsoft Research

Using Weka to teach Machine Learning, Data and Web Mining http://uhaweb.hartford.edu/compsci/ccli/



Machine Learning, Data and Web Mining by Example ("learning by doing" approach)

- Data preprocessing and visualization
- Attribute selection
- Classification (OneR, Decision trees)
- Prediction (Nearest neighbor)
- Model evaluation
- Clustering (K-means, Cobweb)
- Association rules

Initial Data Preparation (Weka data input)

- Raw data (Japanese loan data)
- Web/Text documents (Department data)

Japanese loan data (a sample from a loan history database of a Japanese bank)

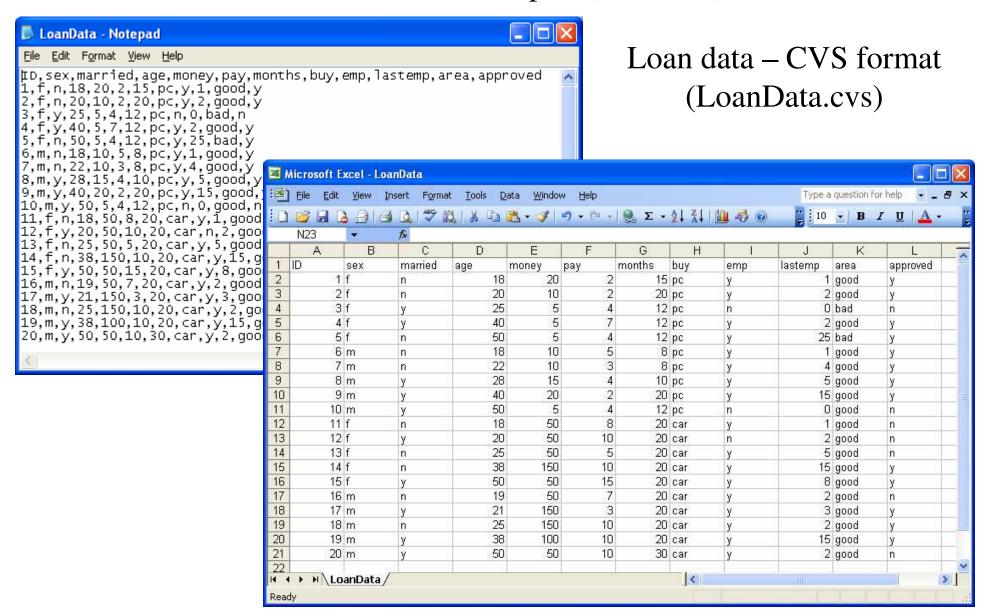
Clients: s1,..., s20

- Approved loan: s1, s2, s4, s5, s6, s7, s8, s9, s14, s15, s17, s18, s19
- Rejected loan: s3, s10, s11, s12, s13, s16, s20

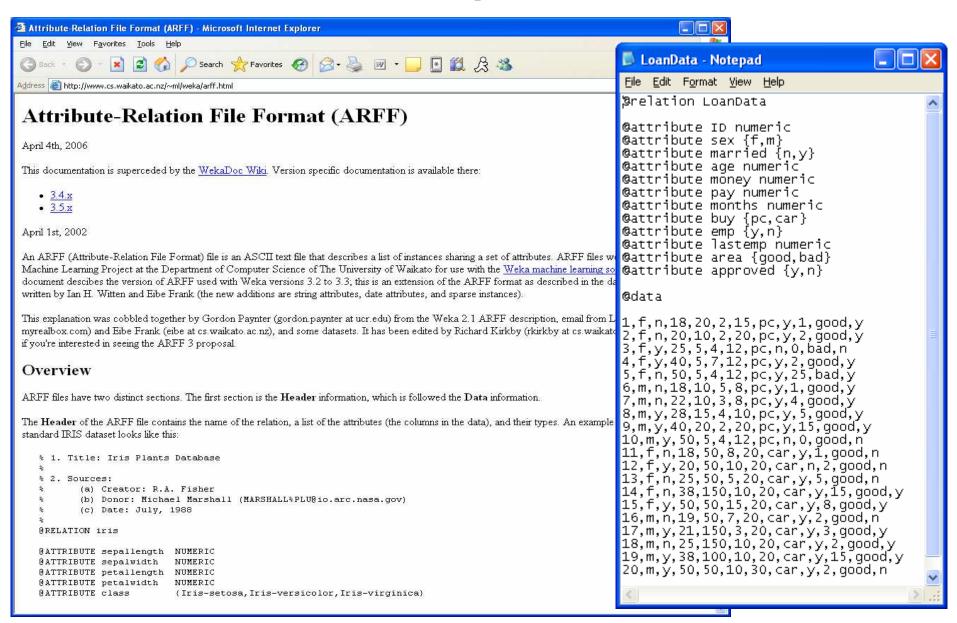
Clients data:

- unemployed clients: s3, s10, s12
- loan is to buy a personal computer: s1, s2, s3, s4, s5, s6, s7, s8, s9, s10
- loan is to buy a car: s11, s12, s13, s14, s15, s16, s17, s18, s19, s20
- male clients: s6, s7, s8, s9, s10, s16, s17, s18, s19, s20
- not married: s1, s2, s5, s6, s7, s11, s13, s14, s16, s18
- live in problematic area: s3, s5
- age: s1=18, s2=20, s3=25, s4=40, s5=50, s6=18, s7=22, s8=28, s9=40, s10=50, s11=18, s12=20, s13=25, s14=38, s15=50, s16=19, s17=21, s18=25, s19=38, s20=50
- money in a bank (x10000 yen): s1=20, s2=10, s3=5, s4=5, s5=5, s6=10, s7=10, s8=15, s9=20, s10=5, s11=50, s12=50, s13=50, s14=150, s15=50, s16=50, s17=150, s18=150, s19=100, s20=50
- monthly pay (x10000 yen): s1=2, s2=2, s3=4, s4=7, s5=4, s6=5, s7=3, s8=4, s9=2, s10=4, s11=8, s12=10, s13=5, s14=10, s15=15, s16=7, s17=3, s18=10, s19=10, s20=10
- months for the loan: s1=15, s2=20, s3=12, s4=12, s5=12, s6=8, s7=8, s8=10, s9=20, s10=12, s11=20, s12=20, s13=20, s14=20, s15=20, s16=20, s17=20, s18=20, s19=20, s20=30
- years with the last employer: s1=1, s2=2, s3=0, s4=2, s5=25, s6=1, s7=4, s8=5, s9=15, s10=0, s11=1, s12=2, s13=5, s14=15, s15=8, s16=2, s17=3, s18=2, s19=15, s20=2

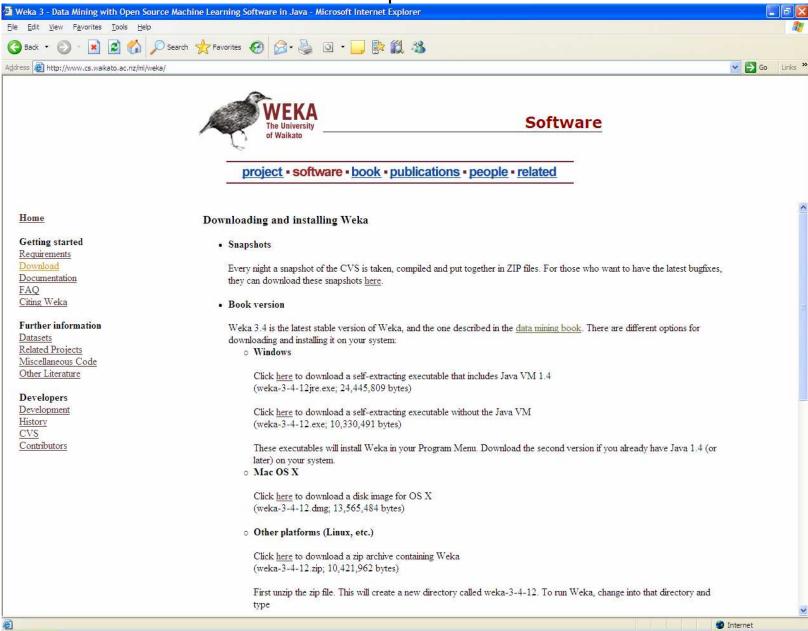
Relations, attributes, tuples (instances)



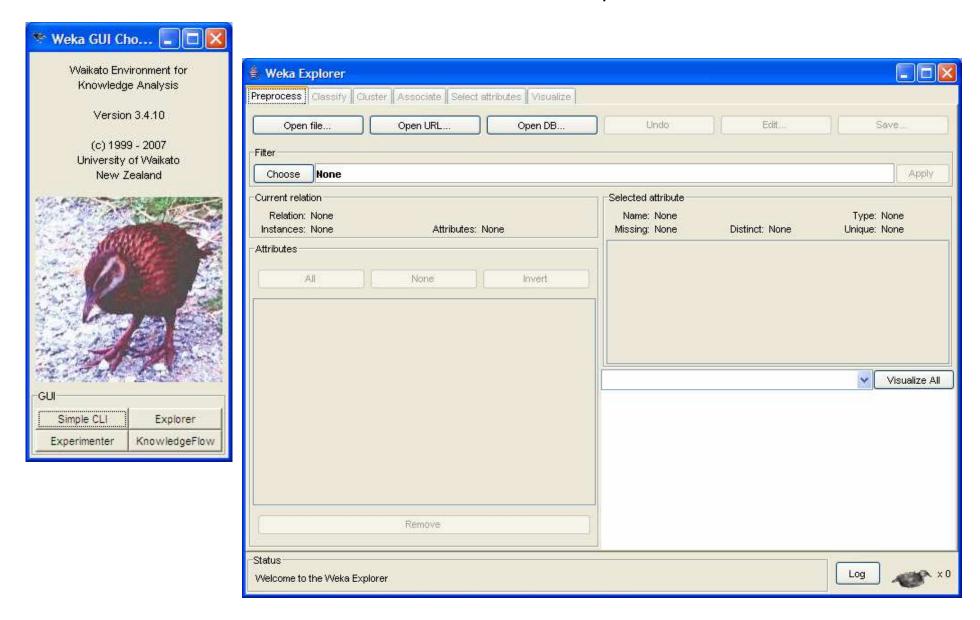
Attribute-Relation File Format (ARFF) - http://www.cs.waikato.ac.nz/~ml/weka/arff.html



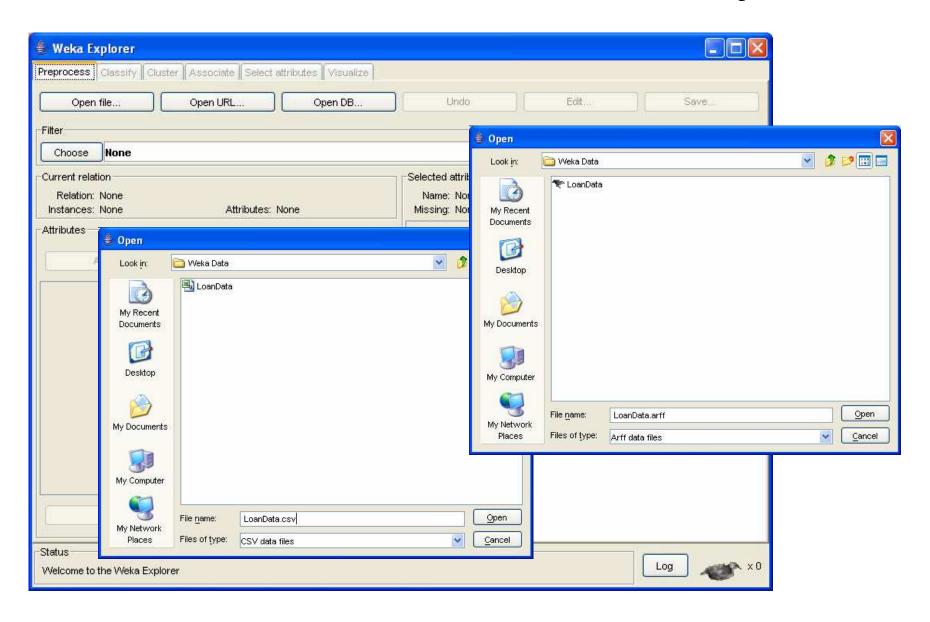
Download and install Weka - http://www.cs.waikato.ac.nz/~ml/weka/



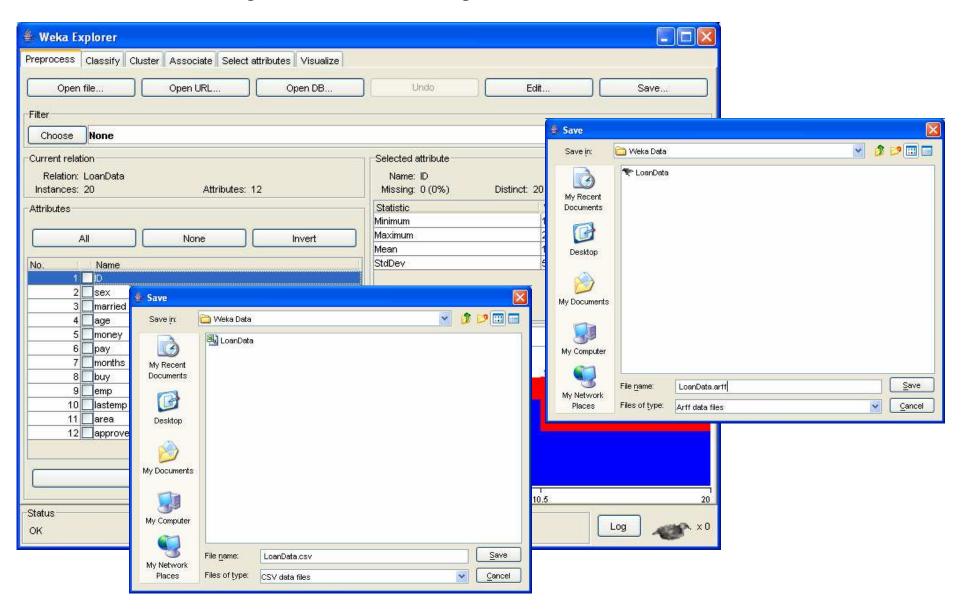
Run Weka and select the Explorer



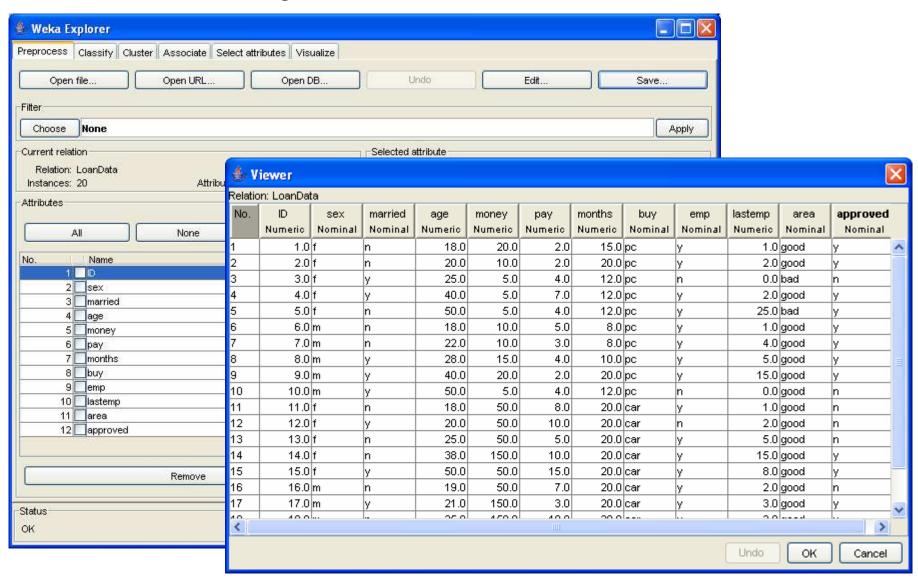
Load data into Weka – ARFF format or CVS format (click on "Open file...")

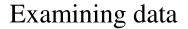


Converting data formats through Weka (click on "Save...")



Editing data in Weka (click on "Edit...")





Weka Explorer

Open file.

Choose None

Relation: LoanData

All

6 pay 7 months 8 buy 9 ■emp 10 astemp 11 area 12 approved

Name 2 sex 3 married 4 age 5 money

Current relation

Instances: 20

Attributes

Status OK

• Attribute type and properties

Preprocess Classify Cluster Associate Select attributes Visualize

Open URL.

None

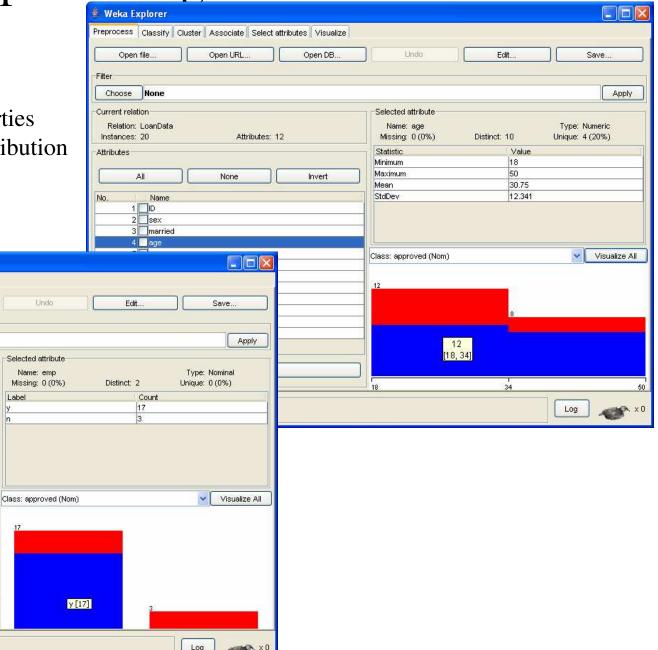
Remove

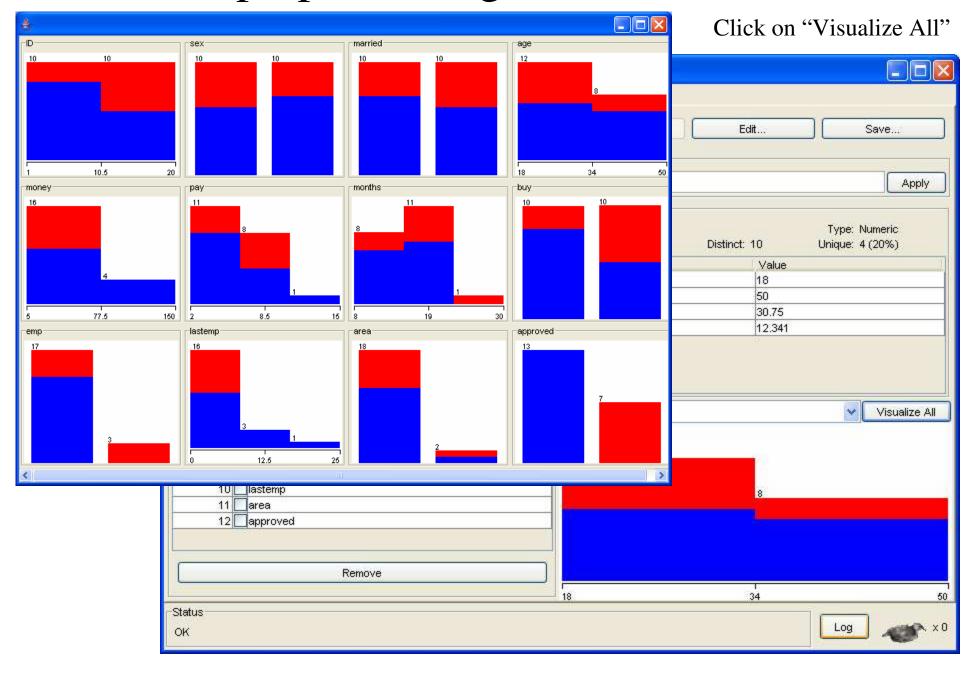
Attributes: 12

Invert

Label

• Class (last attribute) distribution

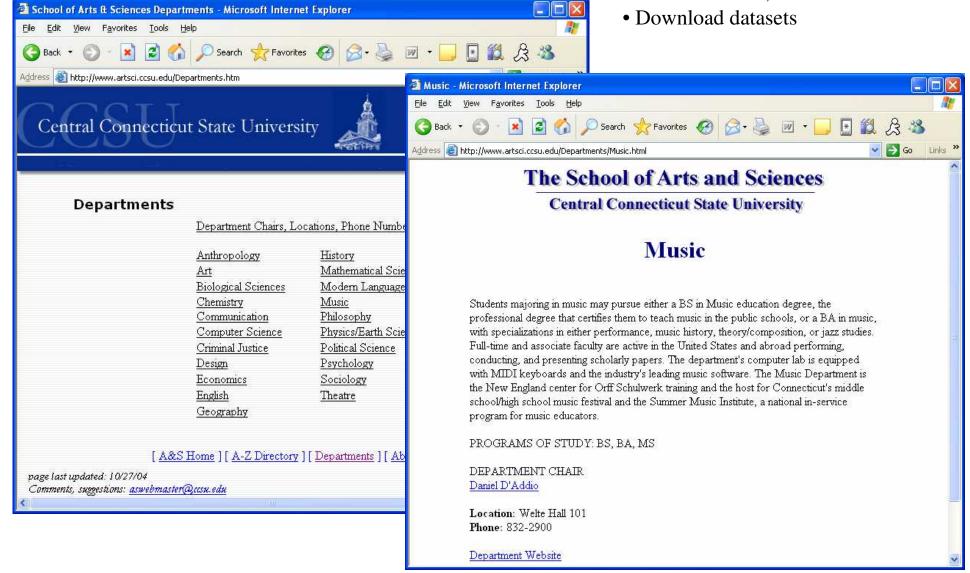


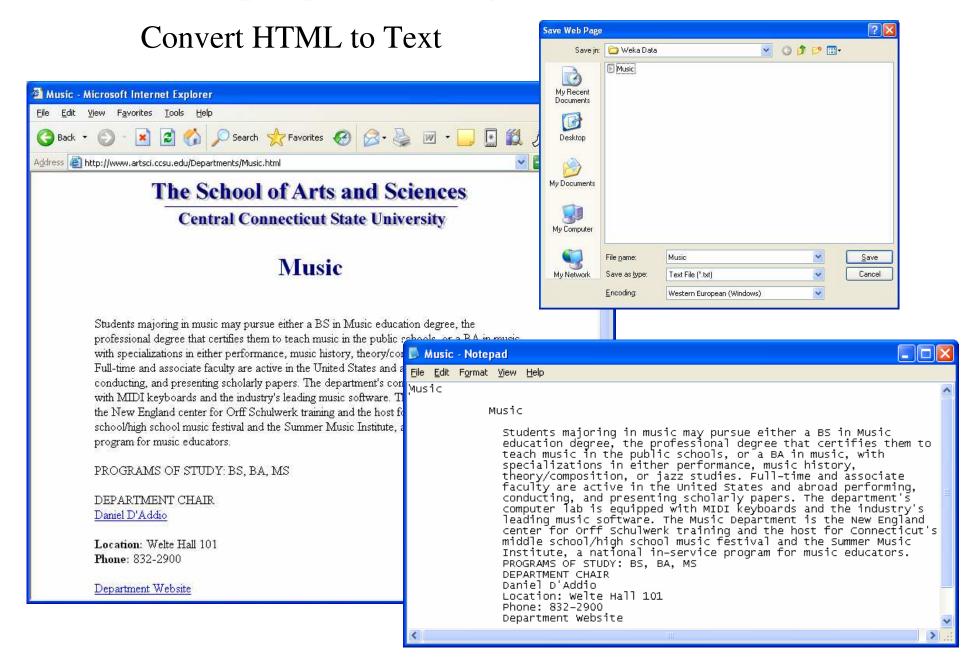


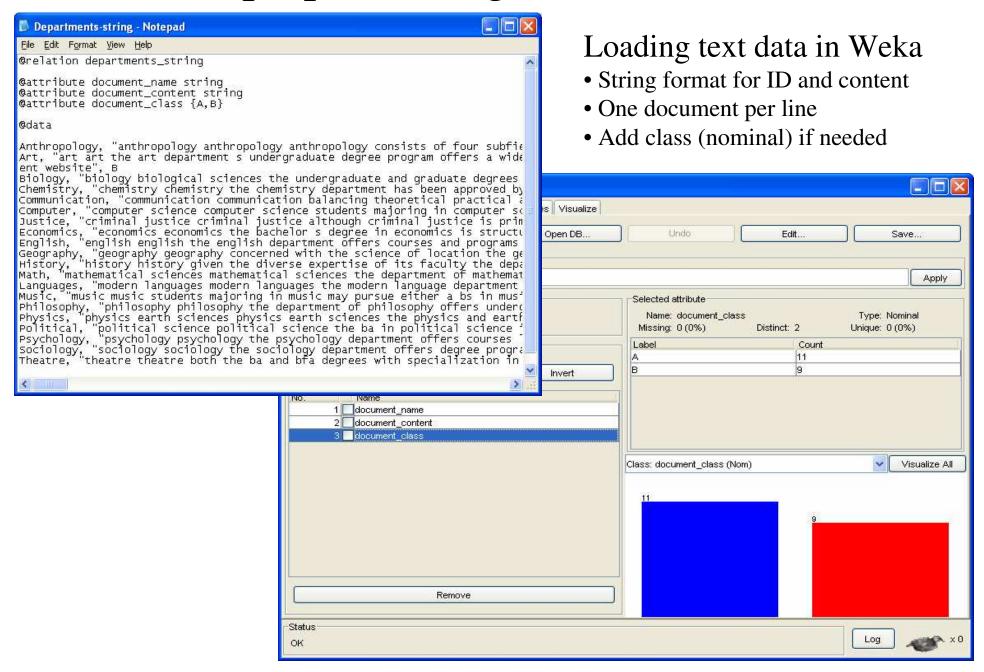
Web/Text documents - Department data

http://www.cs.ccsu.edu/~markov/

• Download Ch1, DMW Book

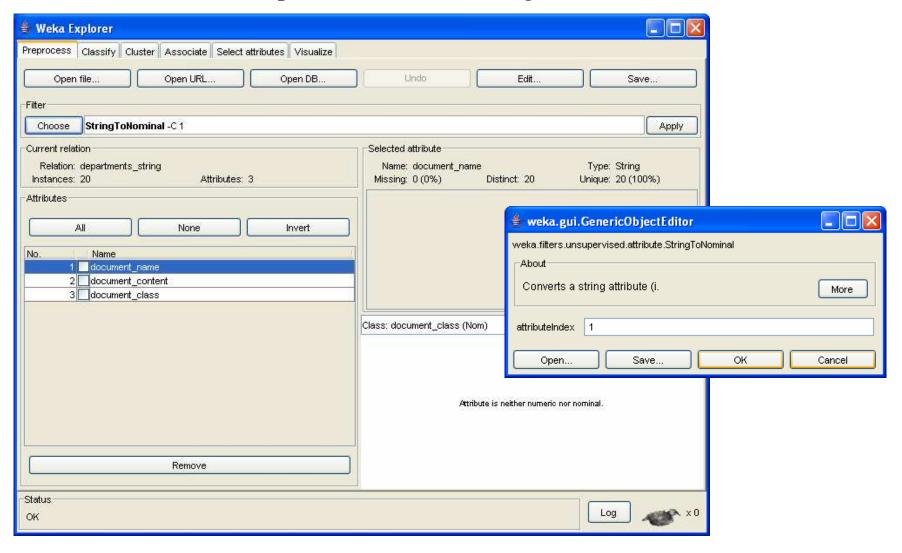






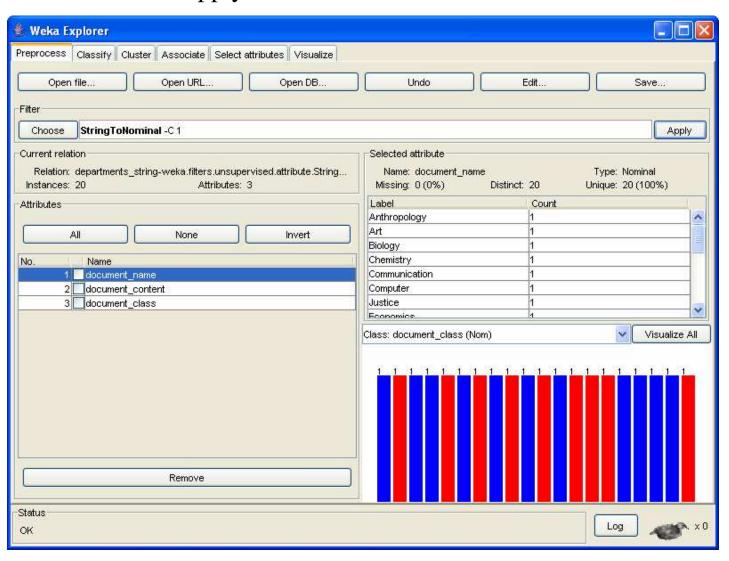
Converting a string attribute into nominal

Choose filters/unsupervised/attribute/StringToNominaland and set the index to 1

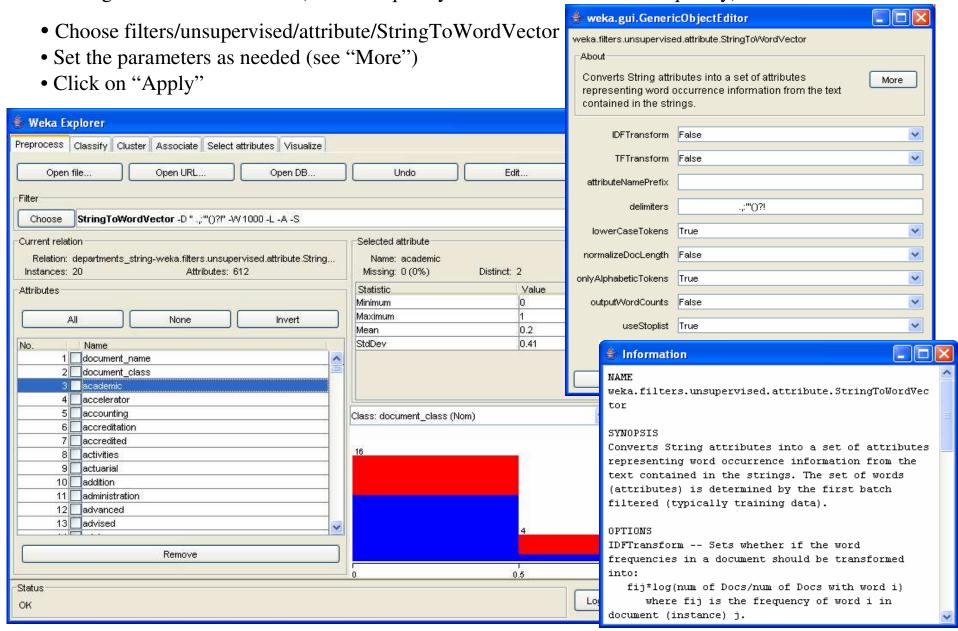


Converting a string attribute into nominal

Click on Apply – document_name is now nominal

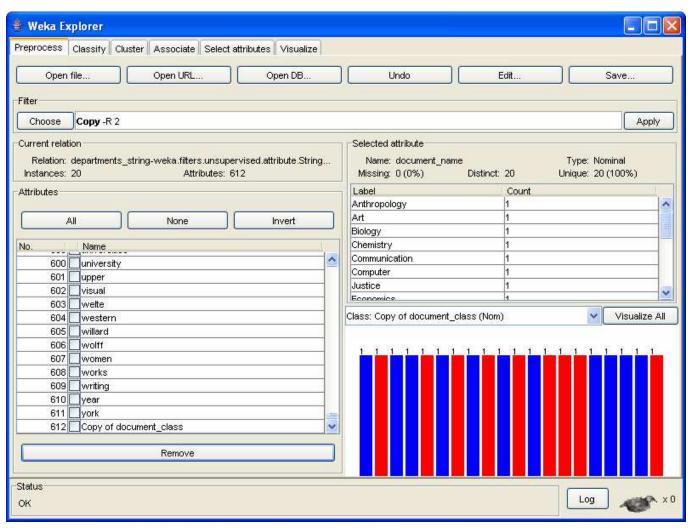


Converting text data into TFIDF (Term Frequency – Inverted Document Frequency) attribute format

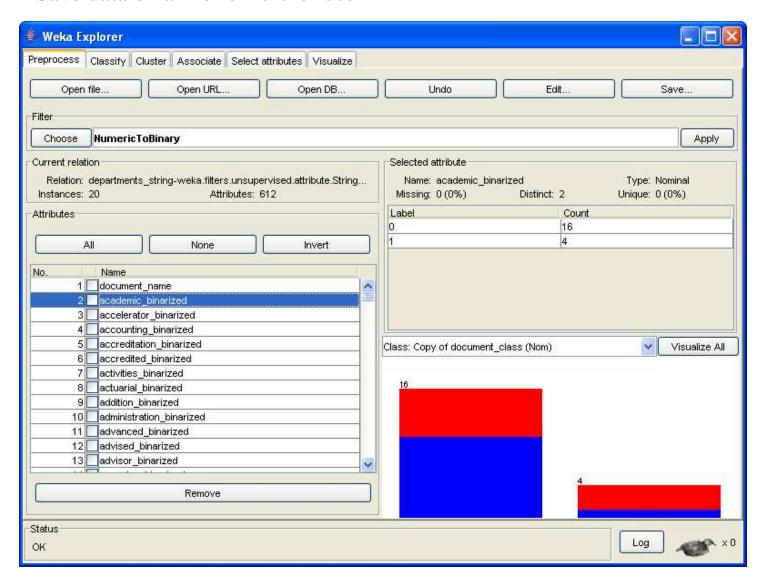


Make the class attribute last

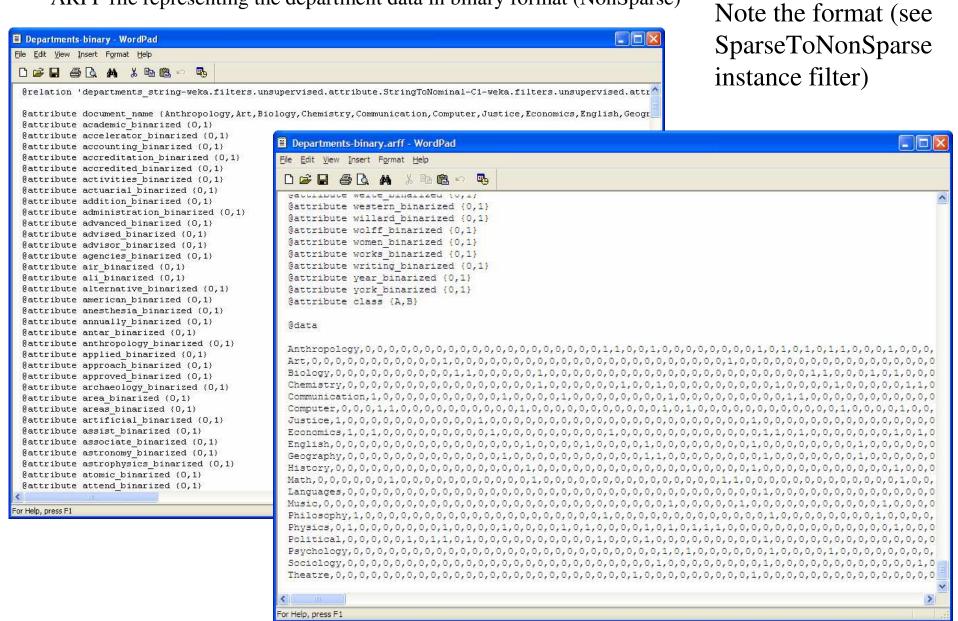
- Choose filters/unsupervised/attribute/Copy
- Set the index to 2 and click on Apply
- Remove attribute 2



- Change the attributes to nominal (use NumericToBinary filter)
- Save data on a file for further use



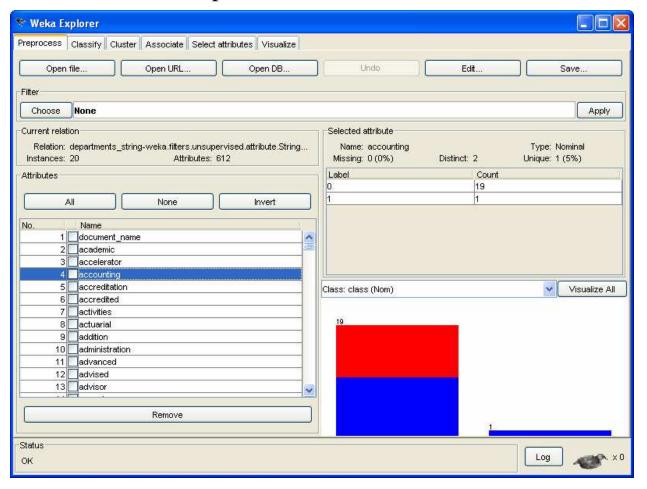
ARFF file representing the department data in binary format (NonSparse)



Attribute Selection

Finding a minimal set of attributes that preserve the class distribution

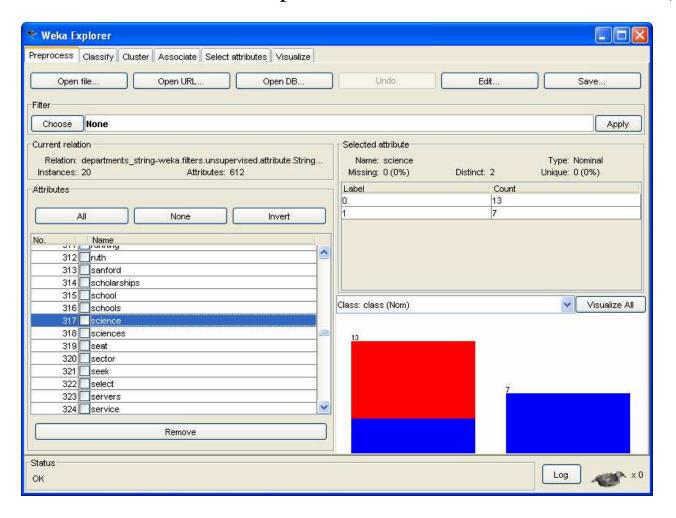
Attribute relevance with respect to the class – not relevant attribute (*accounting*)



IF accounting=1 THEN class=A (Error=0, Coverage = 1 instance → **overfitting**)
IF accounting=0 THEN class=B (Error=10/19, Coverage = 19 instances → **low accuracy**)

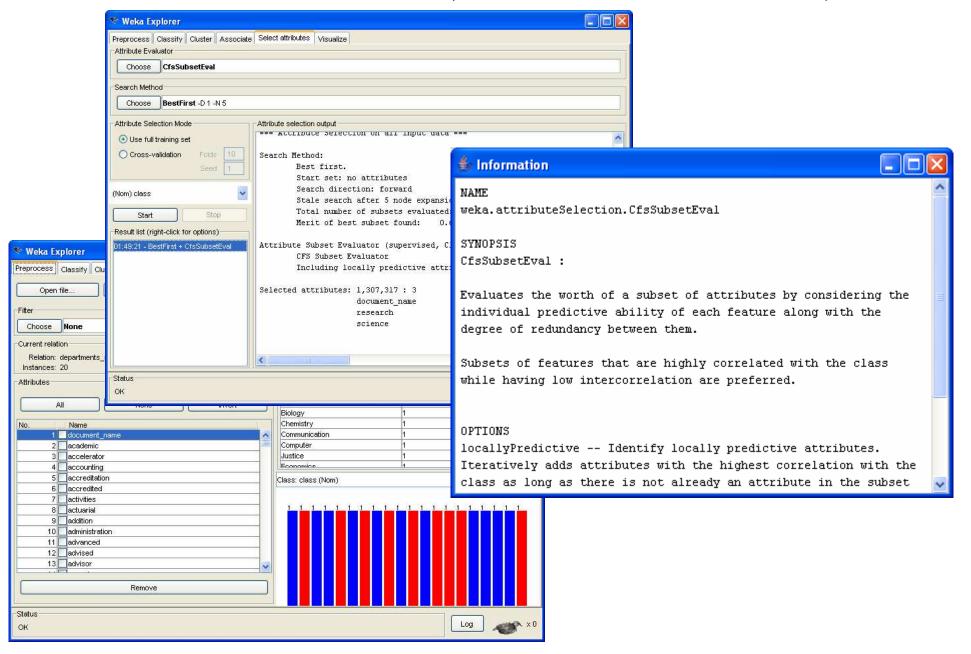
Attribute Selection

Attribute relevance with respect to the class – relevant attribute (*science*)

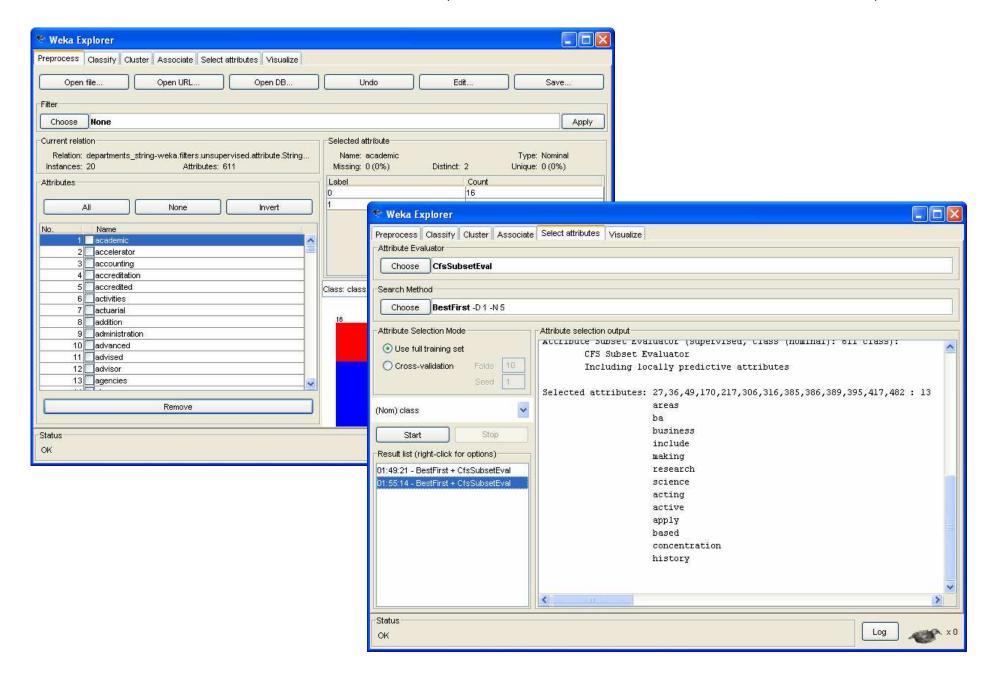


IF accounting=1 THEN class=A (Error=0, Coverage = 7 instance)
IF accounting=0 THEN class=B (Error=4/13, Coverage = 13 instances)

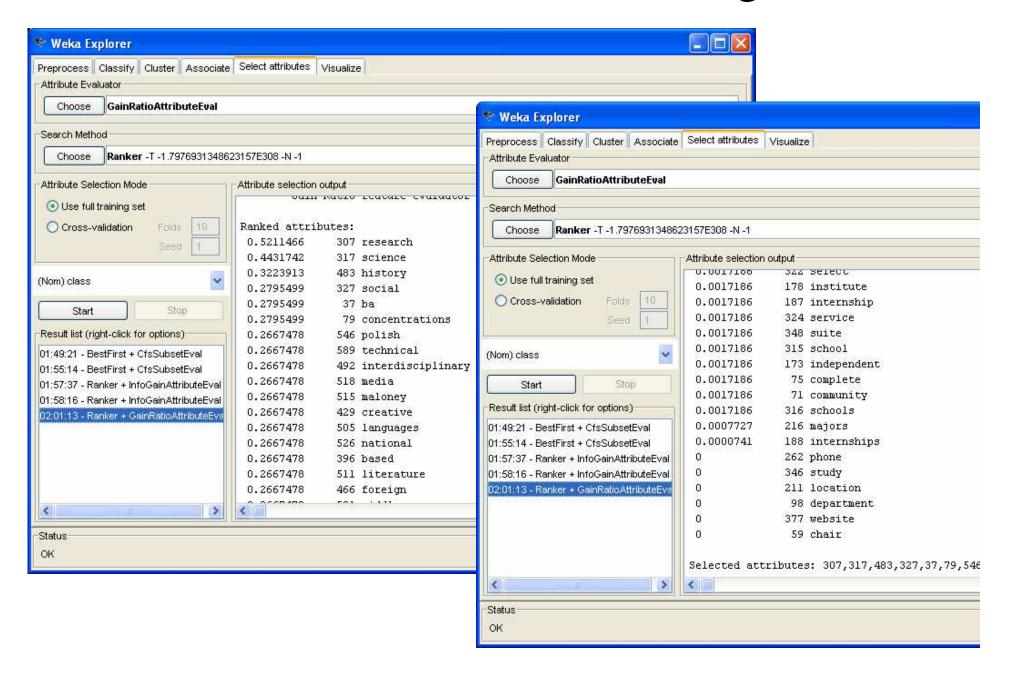
Attribute Selection (with document_name)



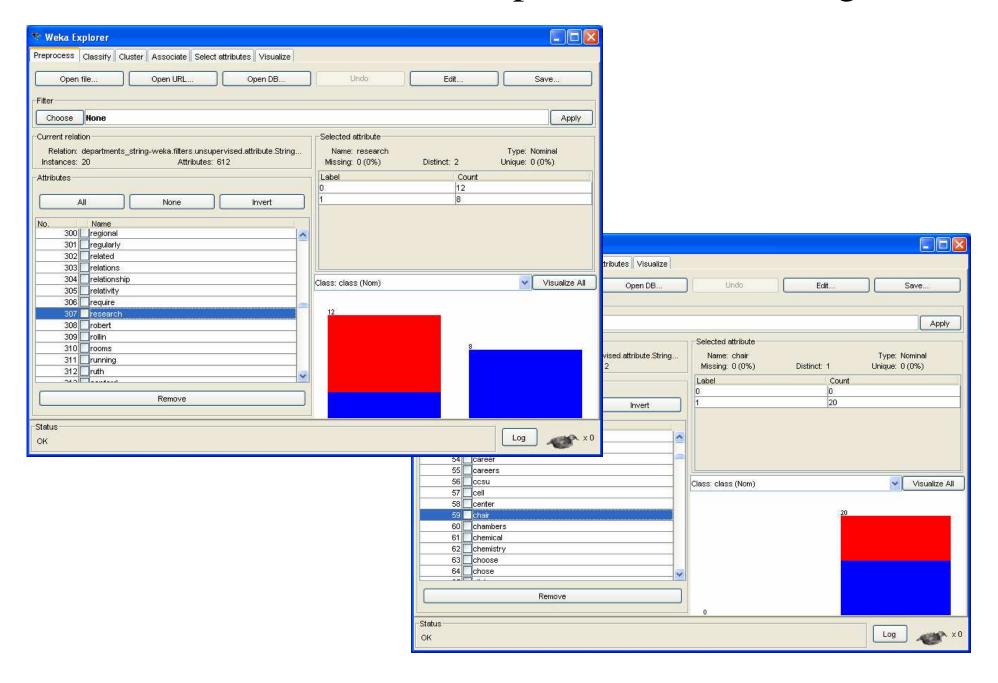
Attribute Selection (without document_name)



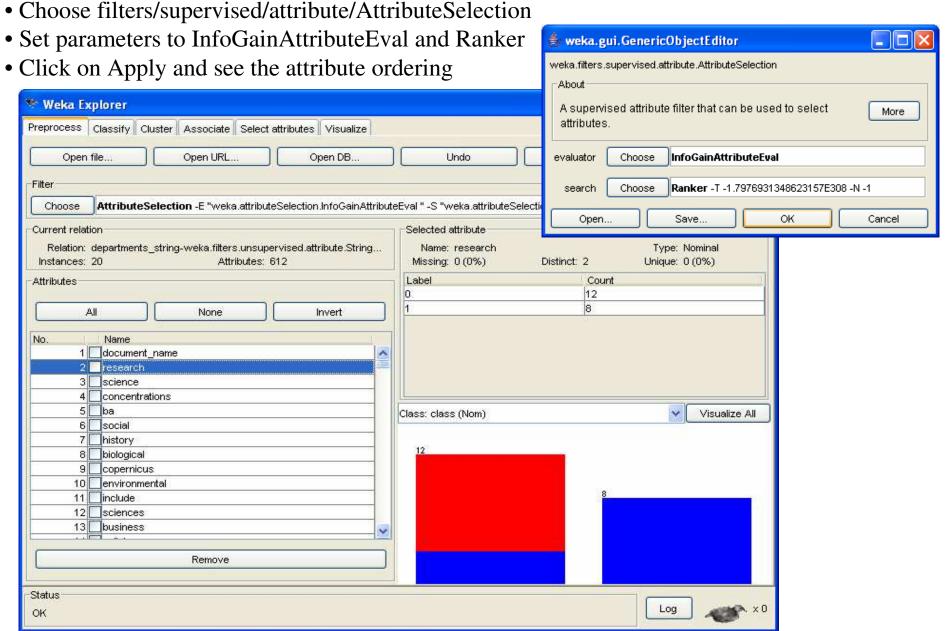
Attribute Selection (ranking)



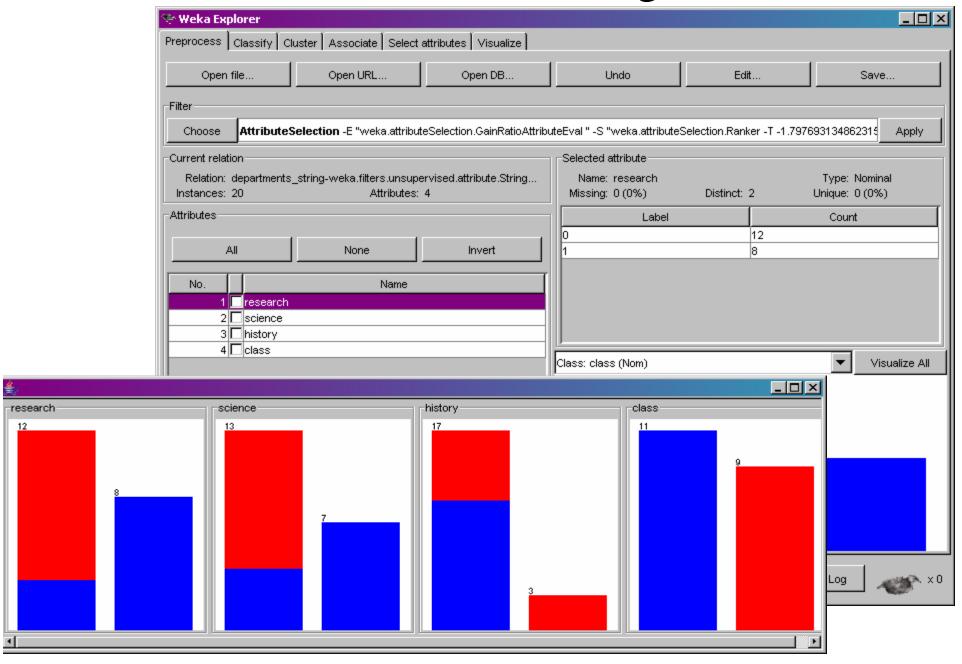
Attribute Selection (explanation of ranking)



Attribute Selection (using filters)



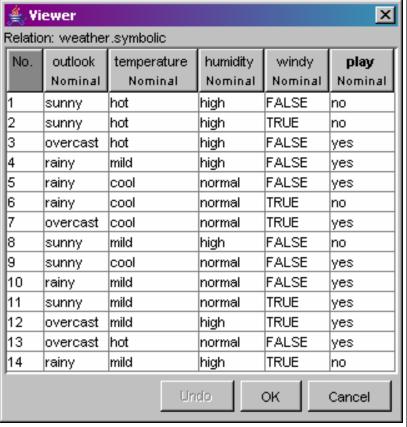
Attribute Selection (using filters)



Classification – creating models (hypotheses) *Mapping (independent attributes -> class)*

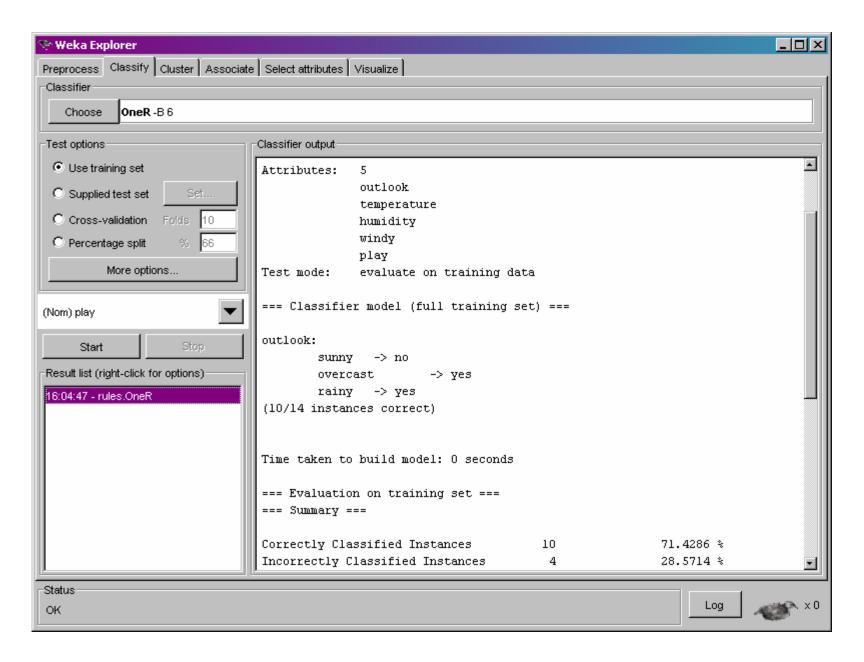
Inferring rudimentary rules - OneR

Weather data (weather.nominal.arff)



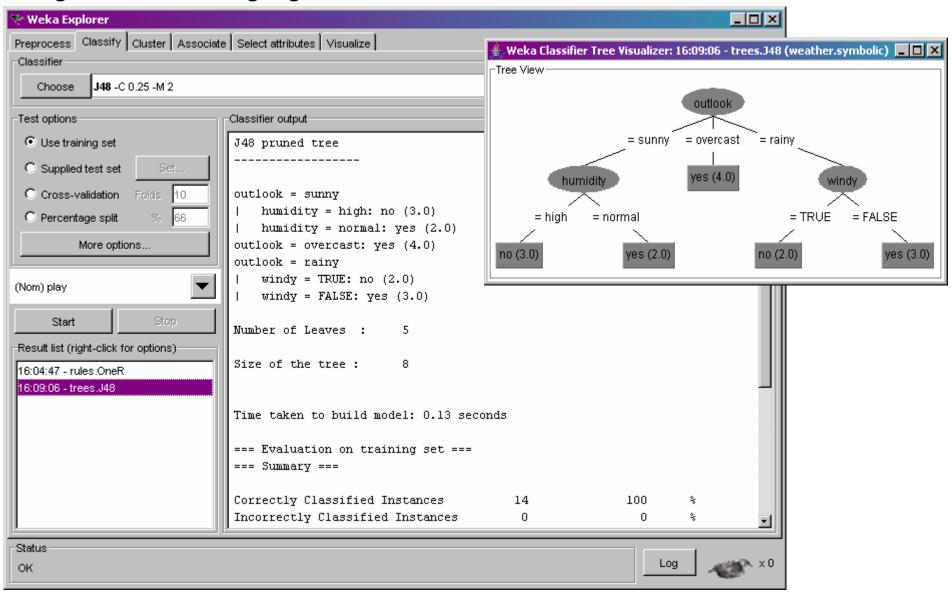
Attribute	Rules	Errors	Total	
			error	
outlook	sunny -> no	2/5	4/14	
	overcast -> yes	0/4		
	rainy -> yes	2/5		
temperature	hot -> no	2/4	5/14	
	mild -> yes	2/6		
	cool -> yes	1/4		
humidity	high -> no	3/7	4/14	
	normal -> yes	1/7		
windy	false -> yes	2/8	5/14	
	true -> no	3/5		

Classification – OneR



Classification – decision tree

Right click on the highlighted line in Result list and choose Visualize tree



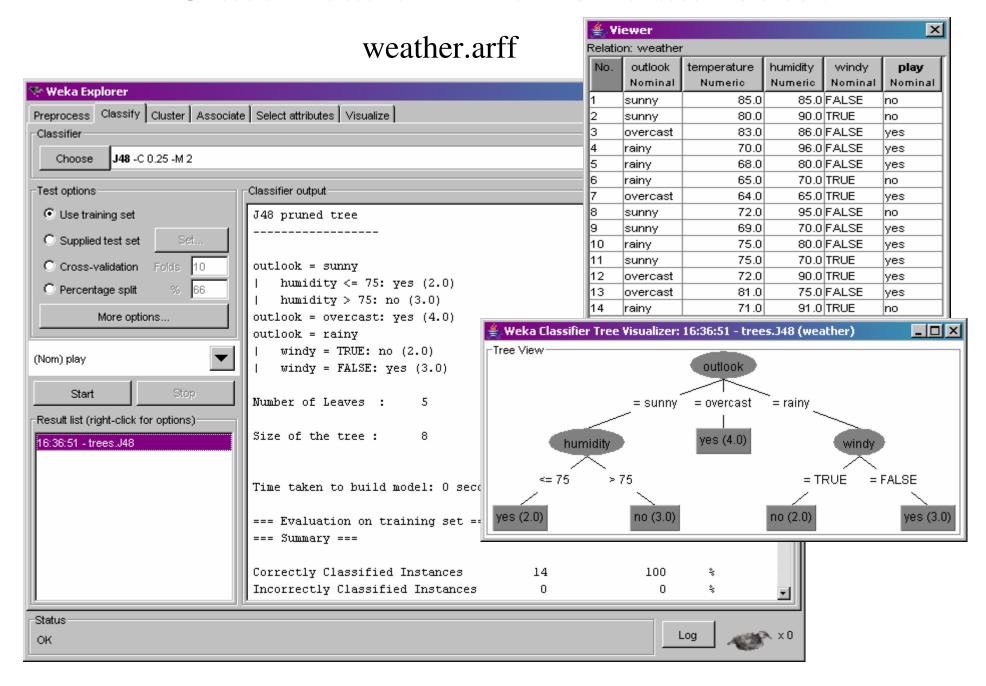
Classification – decision tree

Top-down induction of decision trees (TDIDT, old approach know from pattern recognition):

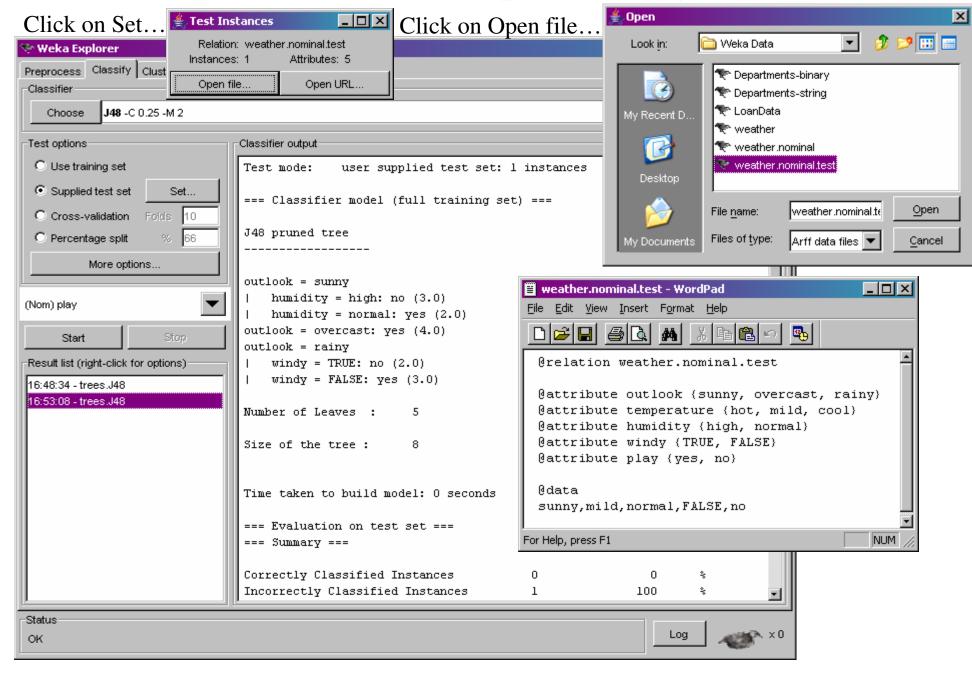
- Select an attribute for root node and create a branch for each possible attribute value.
- Split the instances into subsets (one for each branch extending from the node).
- Repeat the procedure recursively for each branch, using only instances that reach the branch (those that satisfy the conditions along the path from the root to the branch).
- Stop if all instances have the same class.

ID3, C4.5, J48 (Weka): Select the attribute that minimizes the class entropy in the split.

Classification – numeric attributes

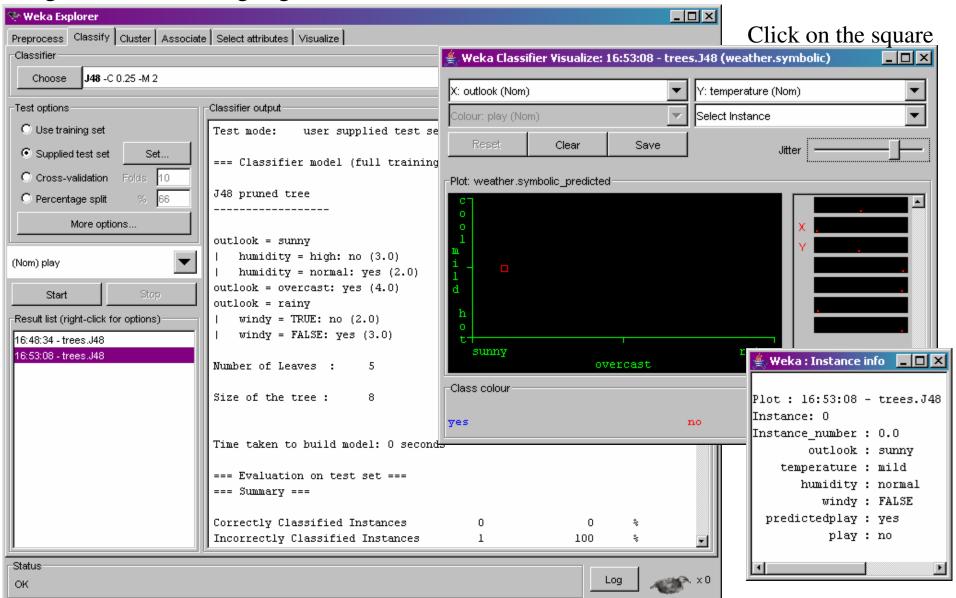


Classification – predicting class

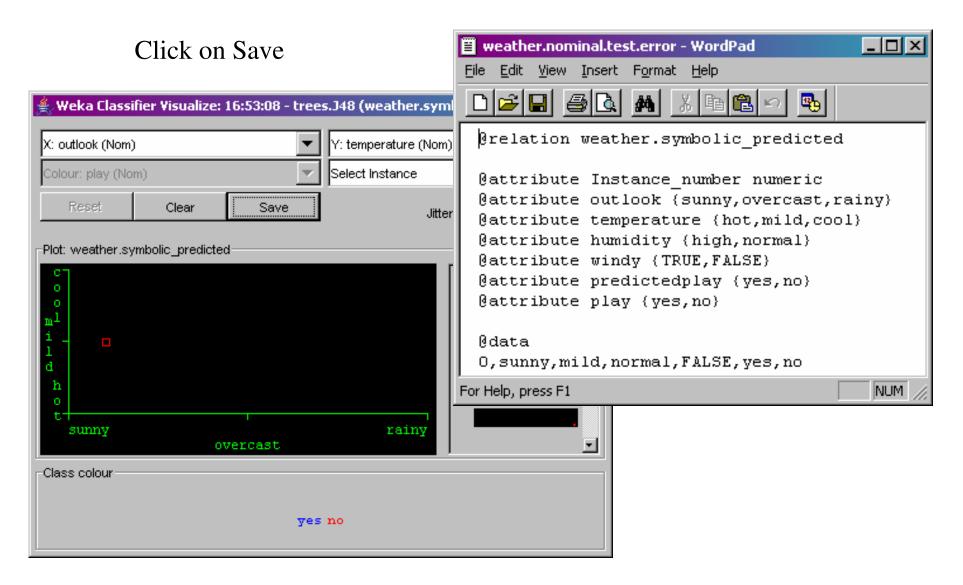


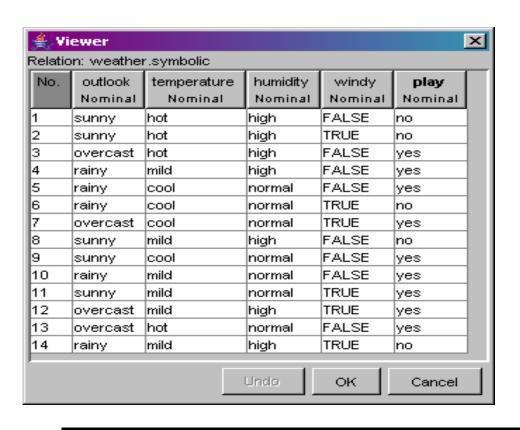
Classification – predicting class

Right click on the highlighted line in Result list and choose Visualize classifier errors



Classification – predicting class





test: (sunny, cool, high, TRUE, ?)

• K-nearest neighbor (KNN, IBk)

Take the class of the nearest neighbor or the majority class among K neighbors

K=1 -> no

 $K=3 \rightarrow no$

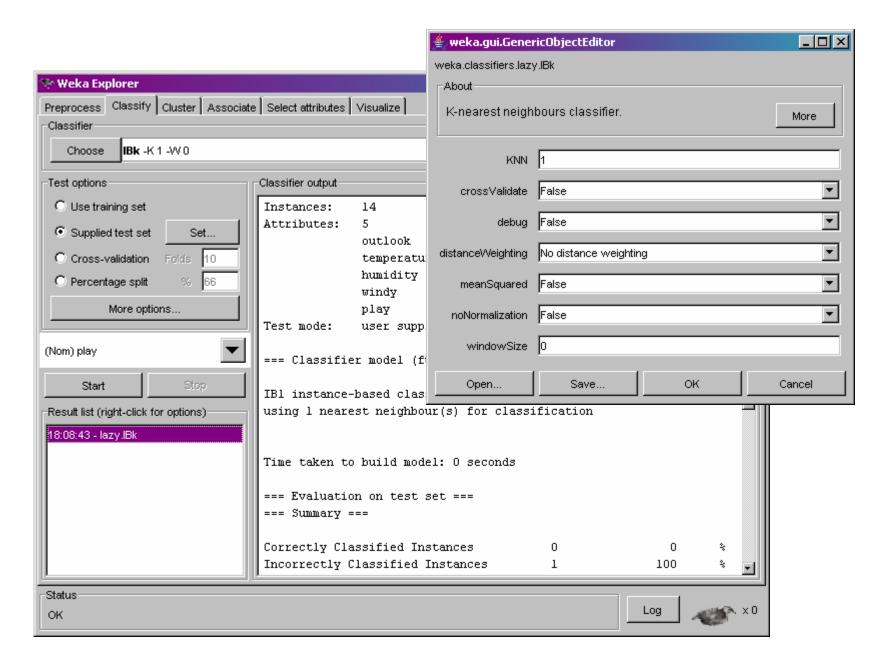
 $K=5 \rightarrow yes$

K=14 -> yes (Majority predictor, ZeroR)

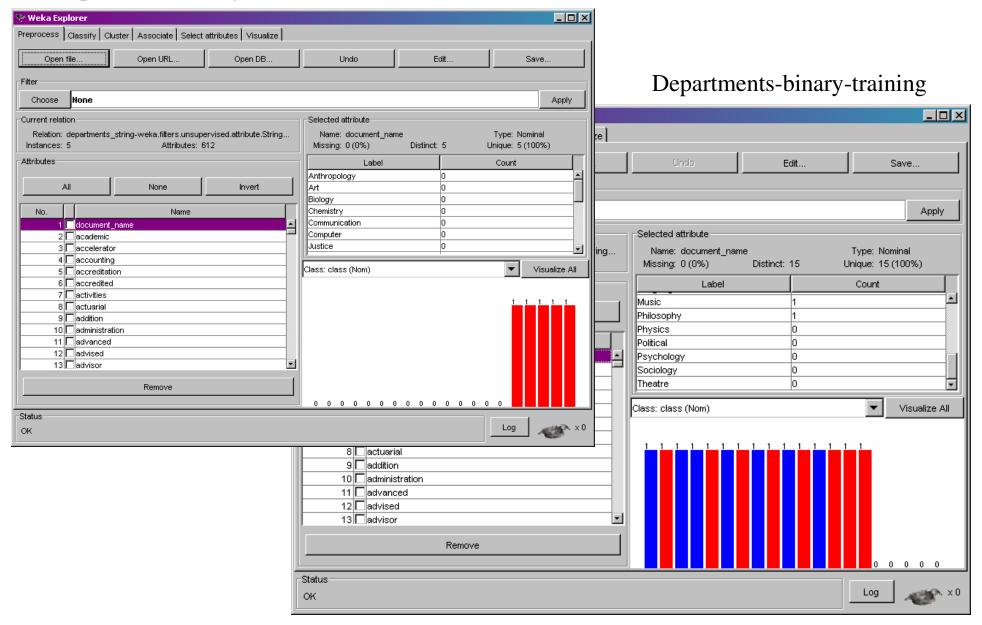
Weighted K-nearest neighbor K=5 -> undecided no=1/1+1/2=1.5 yes=1/2+1/2+1/2=1.5

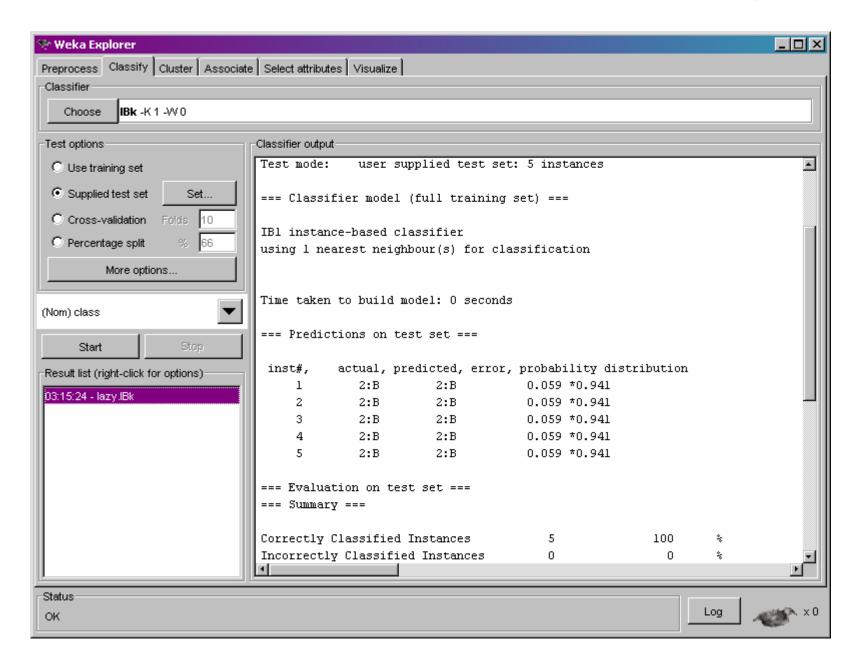
X	2	8	9	11	12	•••	10
Distance(test,X)	1	2	2	2	2	•••	4
play	no	no	yes	yes	yes	• • •	yes

- Distance is calculated as the number of different attribute values
- Euclidean distance for numeric attributes

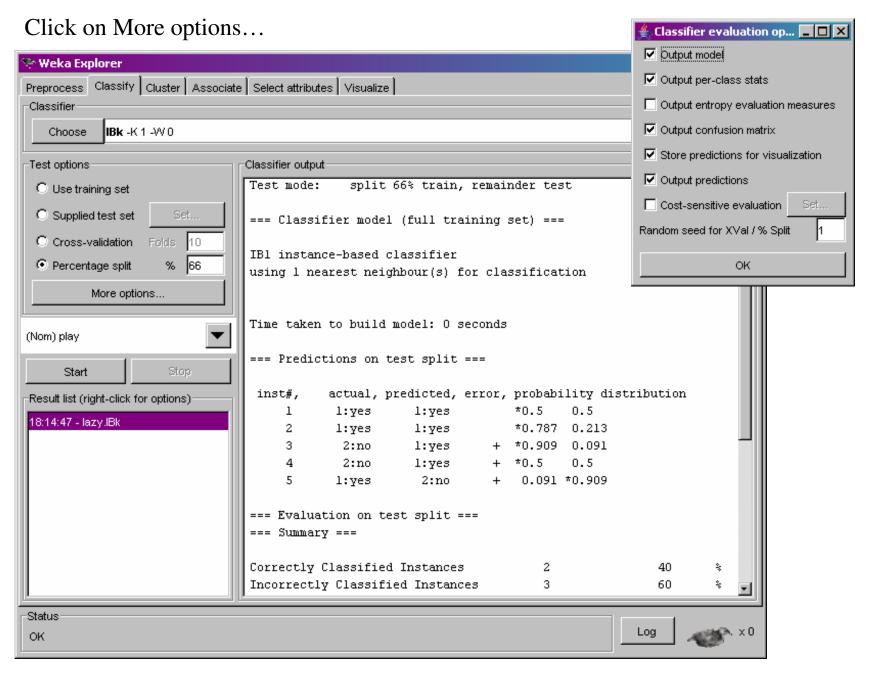


Departments-binary-test.arff

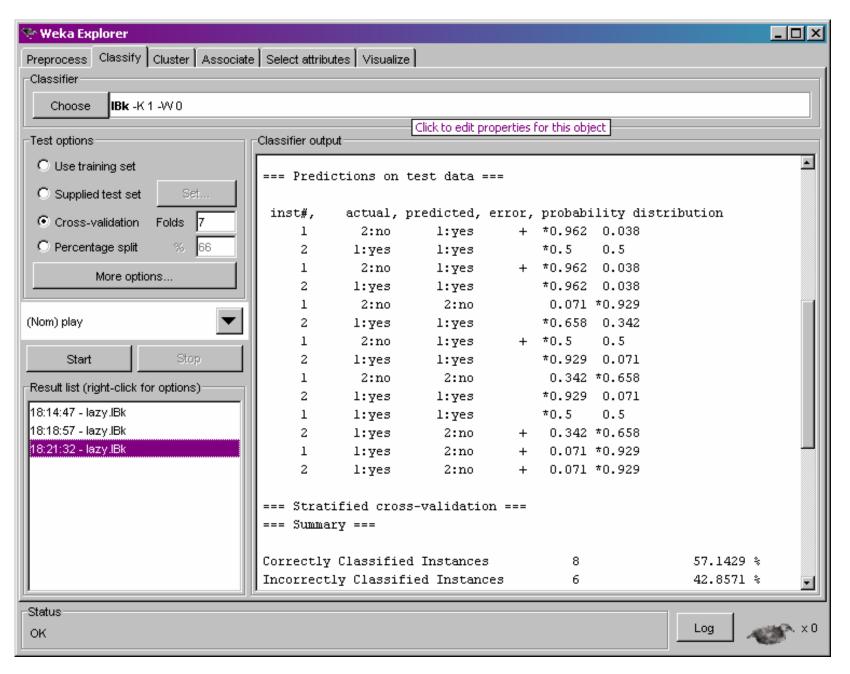




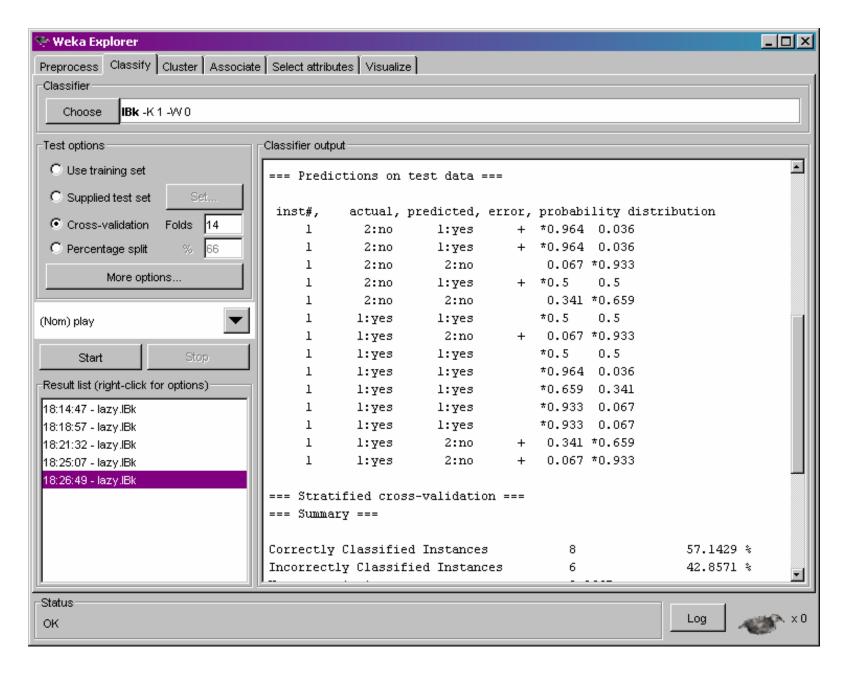
Model evaluation – holdout (percentage split)



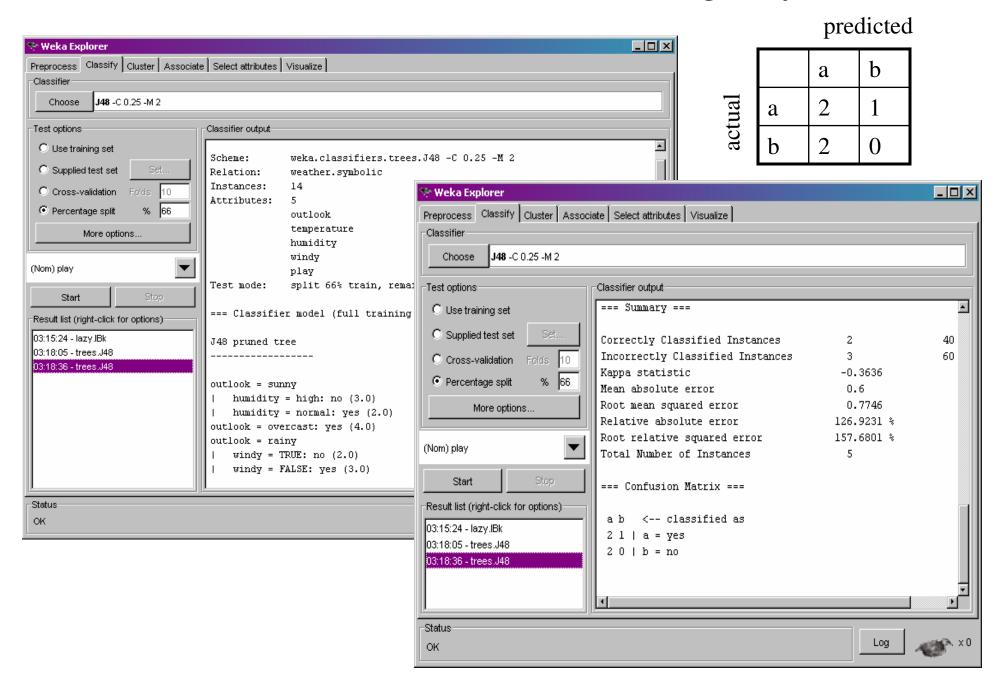
Model evaluation – cross validation



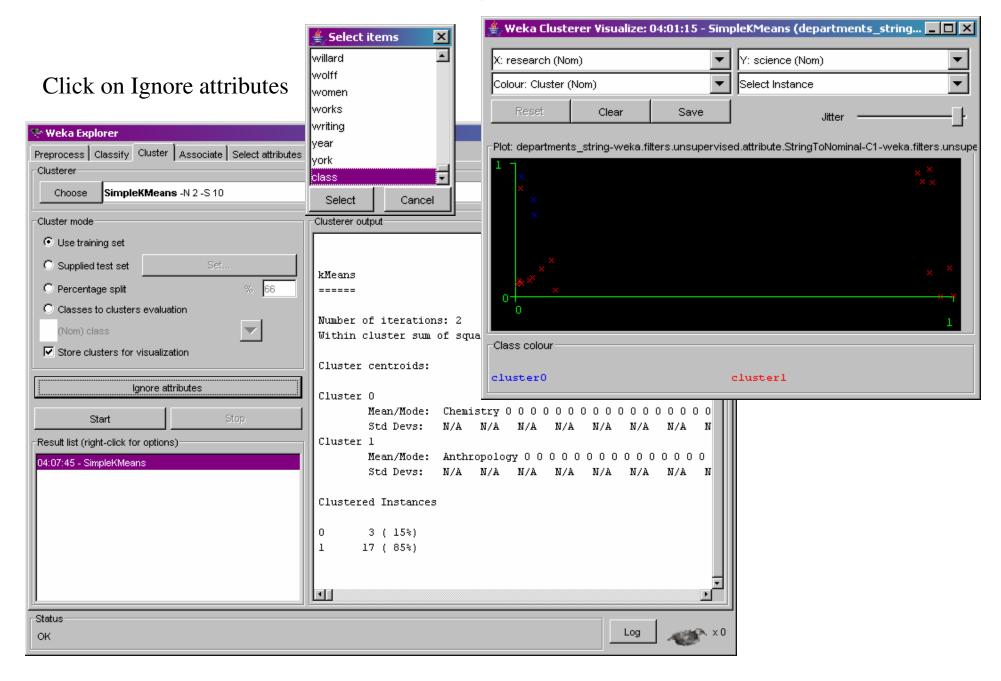
Model evaluation – leave one out cross validation



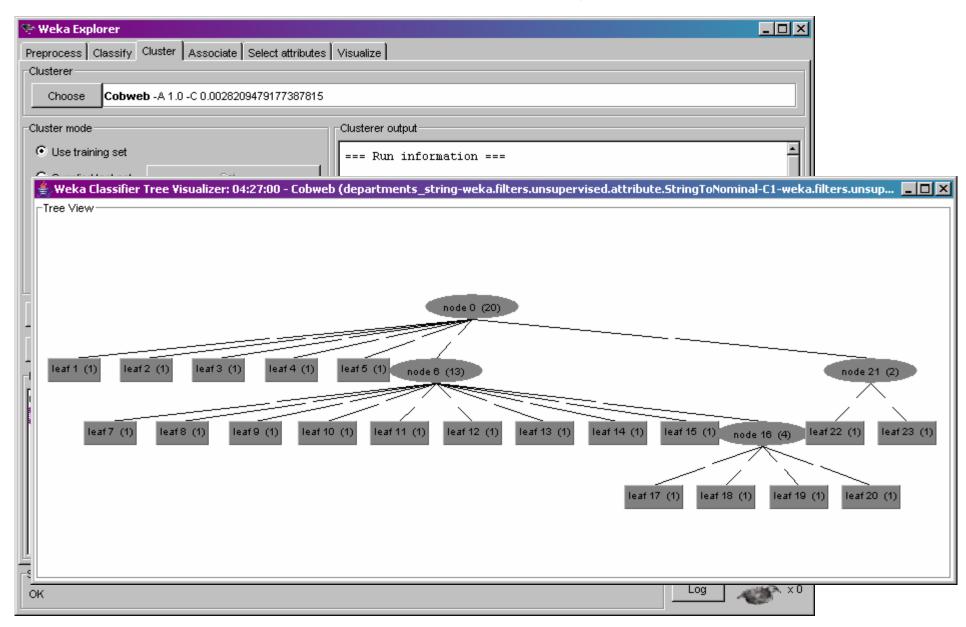
Model evaluation – confusion (contingency) matrix



Clustering – k-means

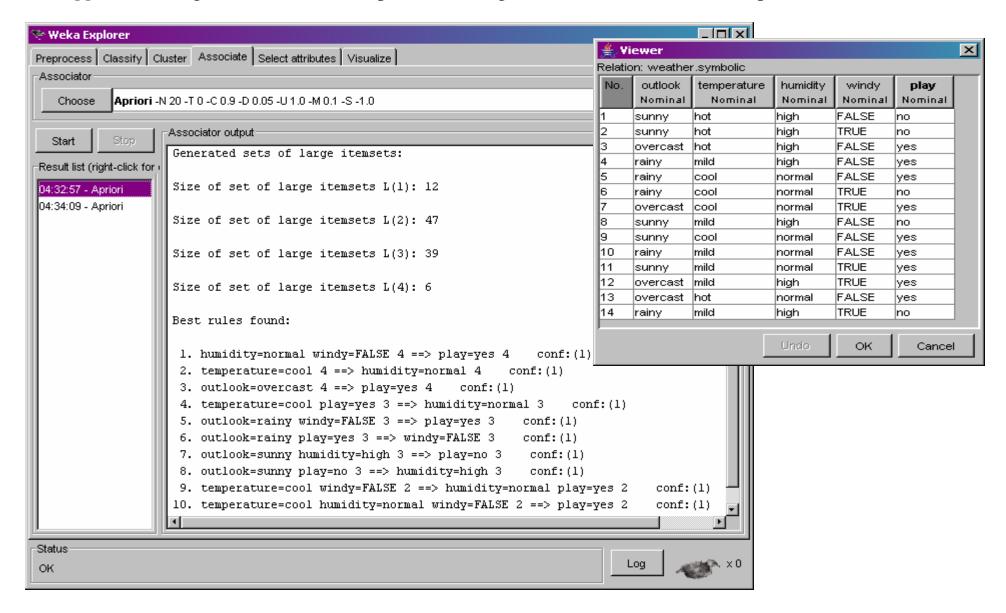


Hierarchical Clustering – Cobweb

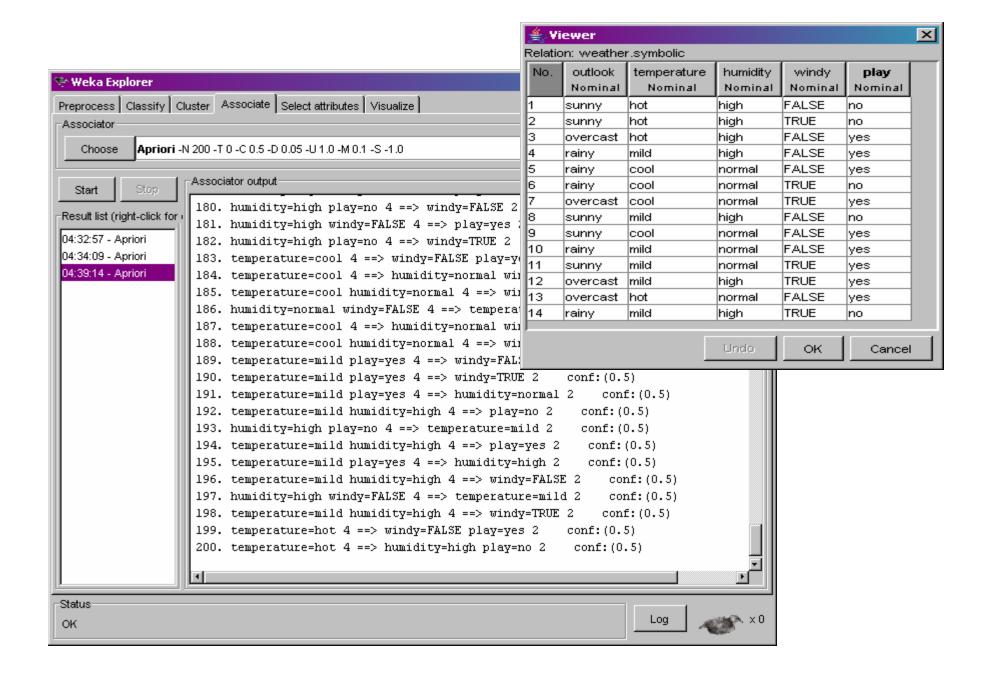


Association Rules $(A \Rightarrow B)$

- Confidence (accuracy): P(B|A) = (# of tuples containing both A and B) / (# of tuples containing A).
- Support (coverage): P(A,B) = (# of tuples containing both A and B) / (total # of tuples)



Association Rules



And many more ...

Thank you!