

## An Introduction to WEKA 3.9.x

Some of the slides are taken from presentation by Yizhou Sun

1

---

---

---

---

---

---

---

---

## Content

- What is WEKA?
- The Explorer:
  - Preprocess data
  - Classification
  - Clustering
  - Association Rules
  - Attribute Selection
  - Data Visualization
- References and Resources

2

---

---

---

---

---

---

---

---

## What is WEKA?

- **Waikato Environment for Knowledge Analysis**
  - It's a data mining/machine learning tool developed by Department of Computer Science, University of Waikato, New Zealand.
  - Weka is also a bird found only on the islands of New Zealand.
  - <https://www.youtube.com/watch?v=1vgA3CN2PH0>
  - Weka software is developed in Java.



3

---

---

---

---

---

---

---

---

## Download and Install WEKA

- Website:  
<http://www.cs.waikato.ac.nz/~ml/weka/index.html>
- Support multiple platforms (written in java):
  - Windows, Mac OS X and Linux

4

4

## Main Features

- 49+ data preprocessing tools
- 76+ classification/regression algorithms
- 8+ clustering algorithms
- 3+ algorithms for finding association rules
- 15+ attribute/subset evaluators + 10+ search algorithms for feature selection

5

5

## Main GUI

- Four graphical user interfaces
  - “The Explorer” (exploratory data analysis)
  - “The Experimenter” (experimental environment)
  - “The KnowledgeFlow” (new process model inspired interface)
  - “Workbench” (unified GUI that combines above three)
- One old fashioned Command Line Interface (CLI)

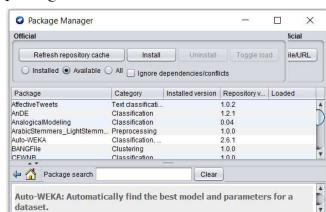


6

6

## The package management system

- Weka community keeps adding new algorithms and features.
- These are placed into plugin packages.
- A package management system allows the user to browse and install packages of interest.



7

7

## Content

- What is WEKA?
- **The Explorer:**
  - Preprocess data
  - Classification
  - Clustering
  - Association Rules
  - Attribute Selection
  - Data Visualization
- References and Resources

8

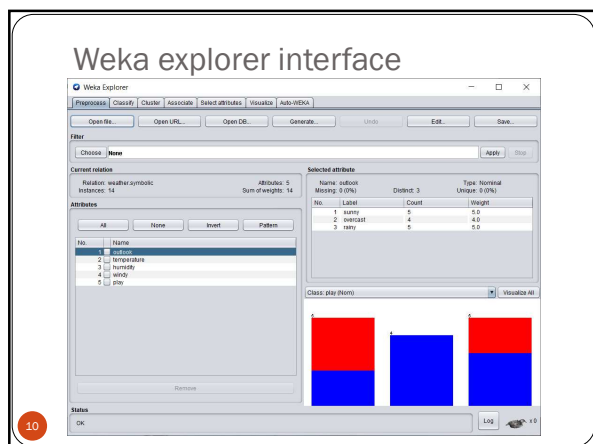
8

## Explorer: pre-processing the data

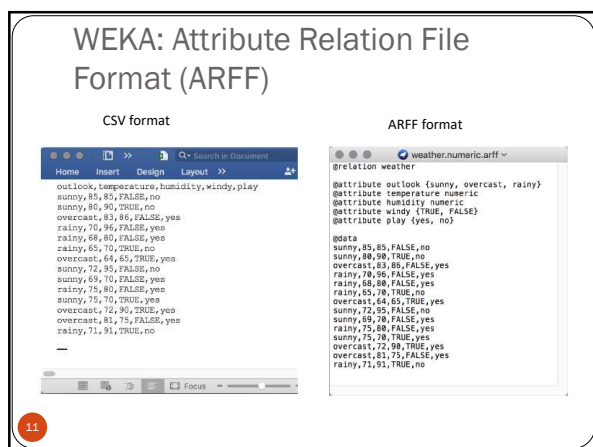
- Data can be imported from a file in various formats: ARFF, CSV, C4.5, binary
- Data can also be read from a URL or from an SQL database (using JDBC)
- Pre-processing tools in WEKA are called "filters"
- WEKA contains filters for:
  - Discretization, normalization, resampling, attribute selection, transforming and combining attributes, ...

9

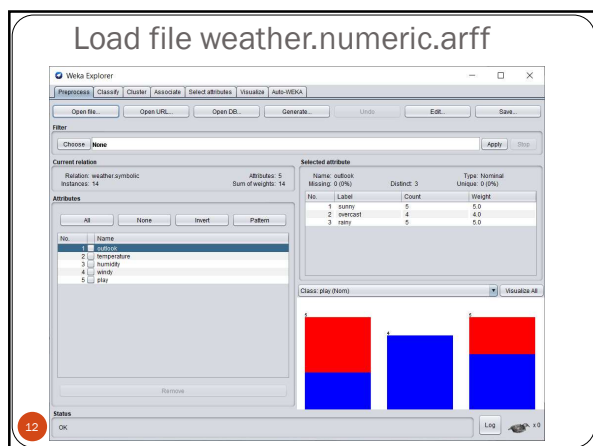
9



10

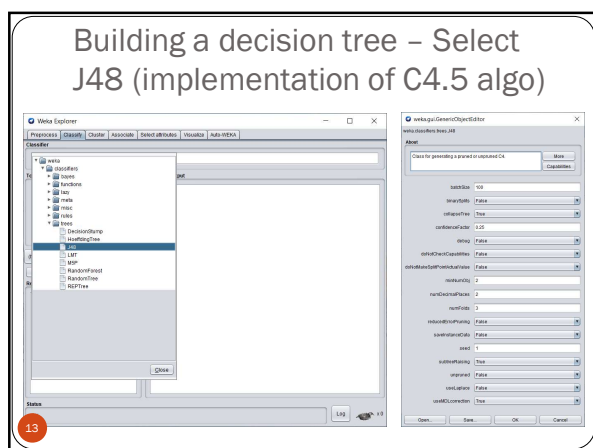


11



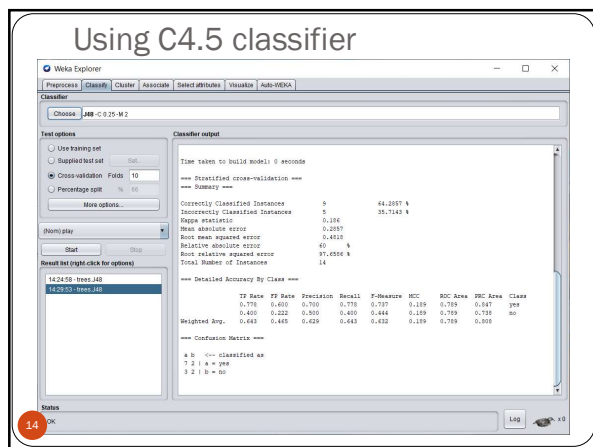
12

## Building a decision tree – Select J48 (implementation of C4.5 algo)



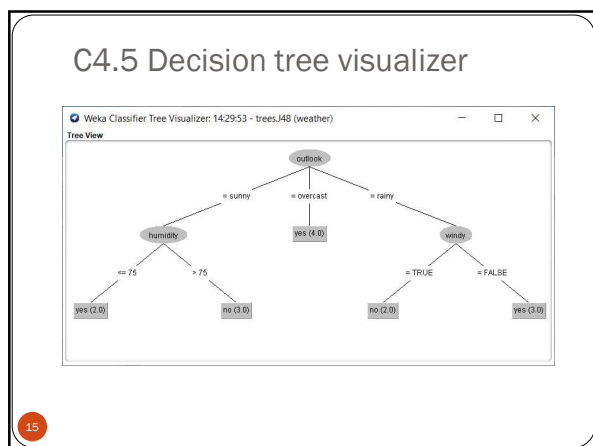
13

## Using C4.5 classifier

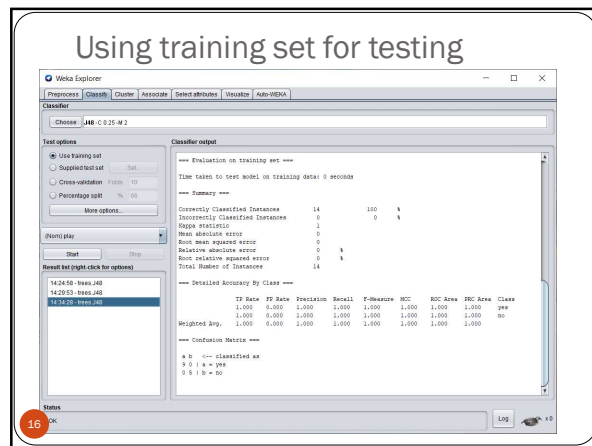


14

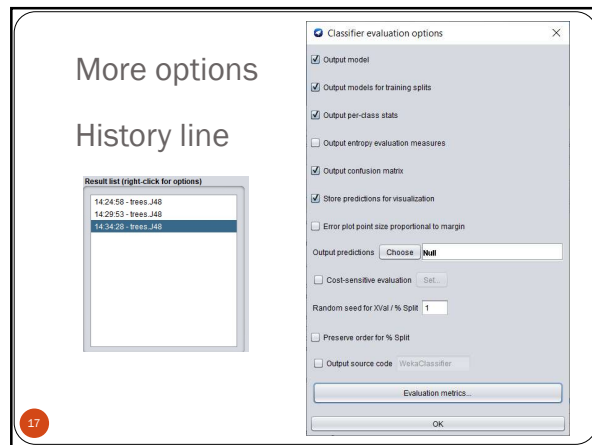
## C4.5 Decision tree visualizer



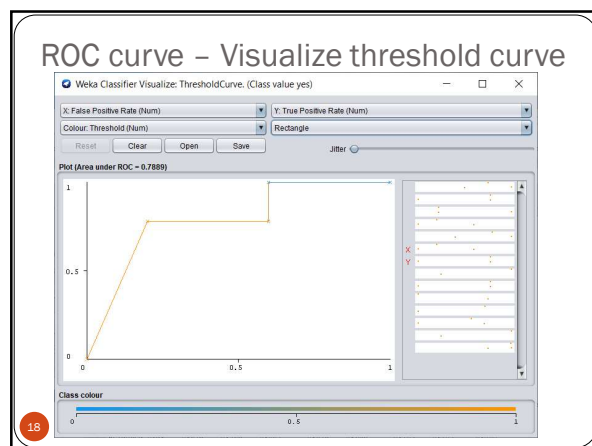
15



16

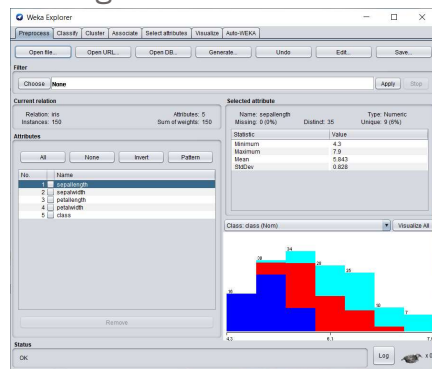


17



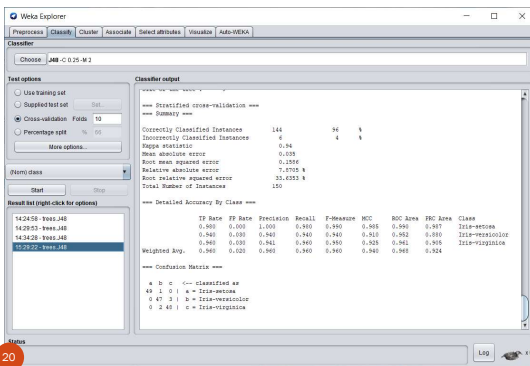
18

## Working with iris data – iris.arff



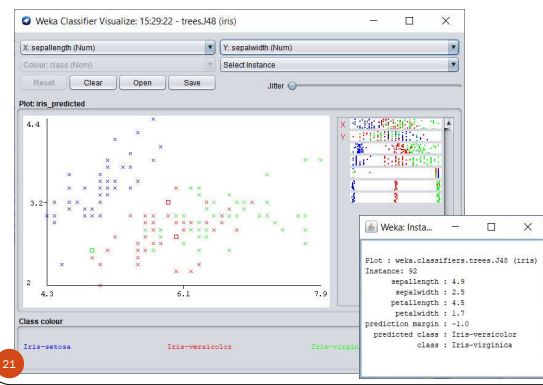
19

## Use J48 with cross validation



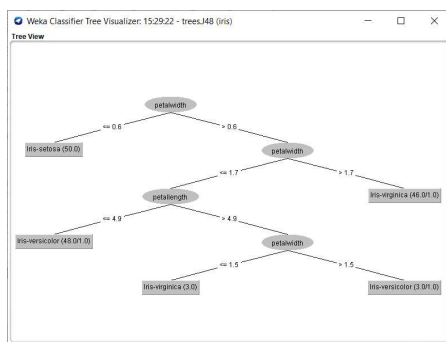
20

## Iris classifier visualization



21

## Iris data – J48 tree



22

22

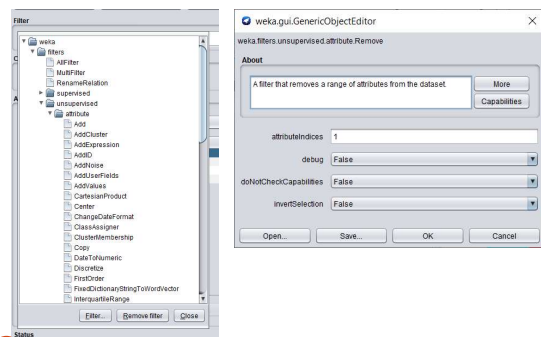
## Open CSV file in Weka

No.	1: sepalwidth	2: sepalwidth	3: petalwidth	4: petalwidth	5: class
	Number	Number	Number	Number	Nominal
1	4.9	3.1	1.5	0.1	Iris-setosa
2	4.8	3.0	1.4	0.1	Iris-setosa
3	4.3	3.0	1.1	0.1	Iris-setosa
4	5.2	4.1	1.5	0.1	Iris-setosa
5	4.9	3.1	1.5	0.1	Iris-setosa
6	4.9	3.1	1.5	0.1	Iris-setosa
7	5.1	3.5	1.4	0.2	Iris-setosa
8	4.9	2.0	1.4	0.2	Iris-setosa
9	4.7	2.2	1.3	0.2	Iris-setosa
10	4.6	3.1	1.5	0.2	Iris-setosa
11	5.0	3.5	1.4	0.2	Iris-setosa
12	5.0	3.4	1.5	0.2	Iris-setosa
13	4.4	2.9	1.4	0.2	Iris-setosa
14	5.4	3.7	1.5	0.2	Iris-setosa
15	4.8	3.4	1.6	0.2	Iris-setosa
16	5.8	4.0	1.2	0.2	Iris-setosa
17	5.4	3.4	1.7	0.2	Iris-setosa
18	4.6	3.6	1.0	0.2	Iris-setosa
19	4.8	3.4	1.9	0.2	Iris-setosa
20	5.0	3.0	1.6	0.2	Iris-setosa
21	5.2	3.5	1.5	0.2	Iris-setosa
22	5.2	3.4	1.4	0.2	Iris-setosa
23	4.7	3.2	1.6	0.2	Iris-setosa
24	4.9	3.1	1.6	0.2	Iris-setosa

23

23

## Filter example – Remove attribute

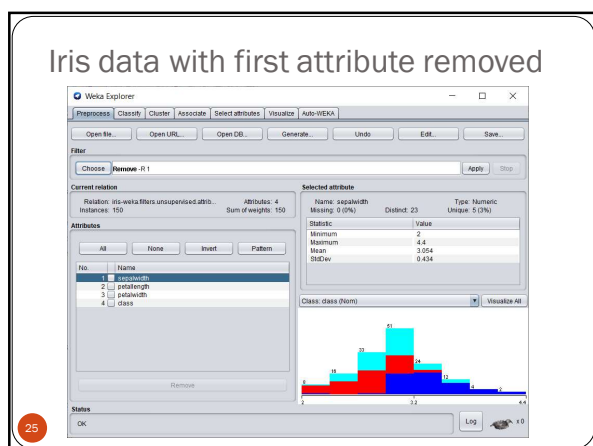


24

24

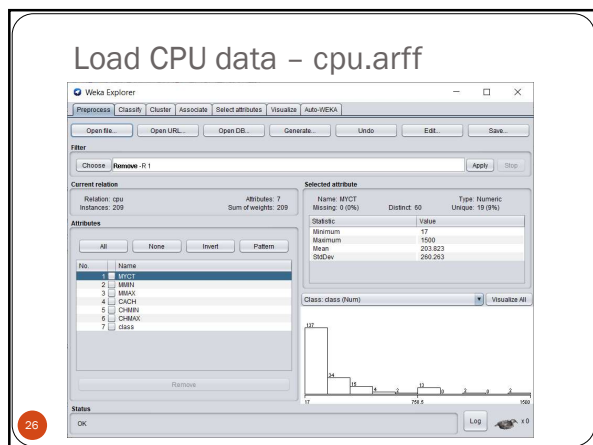


## Iris data with first attribute removed

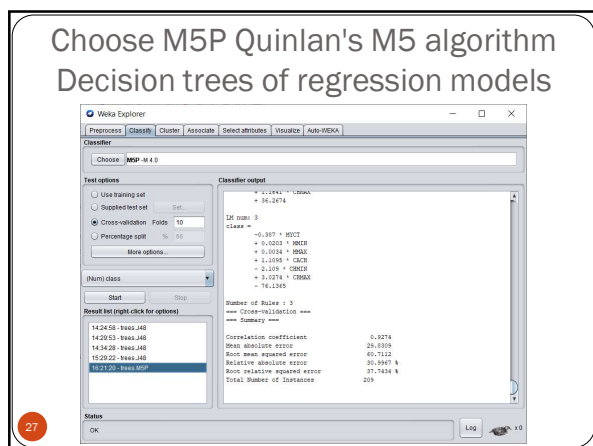


25

## Load CPU data – cpu.arff

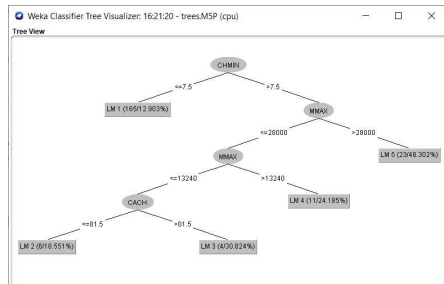


26

Choose M5P Quinlan's M5 algorithm  
Decision trees of regression models

27

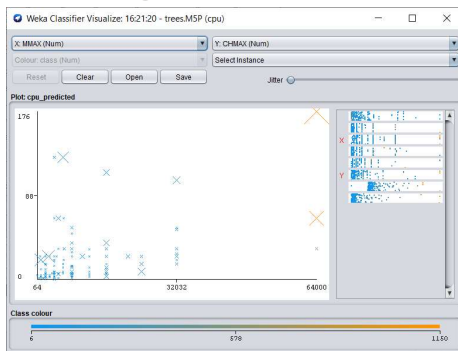
## M5P linear regression models tree



- The first number at each leaf is the number of instances that reach it
- The second is the root mean squared error of the predictions expressed as a percentage of the standard deviation

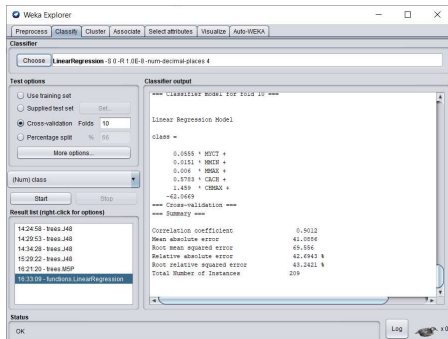
28

## Visualization of errors – larger the cross, larger the error – M5P



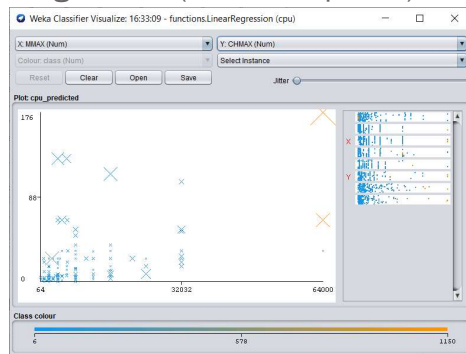
29

## Choose functions -> Linear regression for CPU data



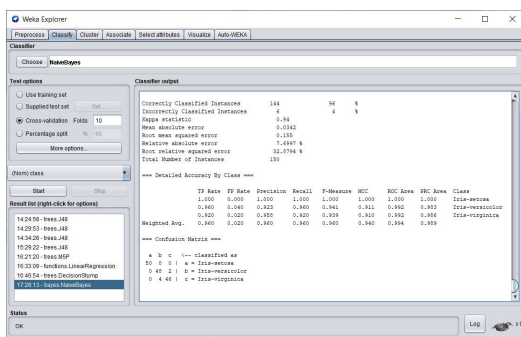
30

## Visualization of errors –linear regression (M5P is superior)



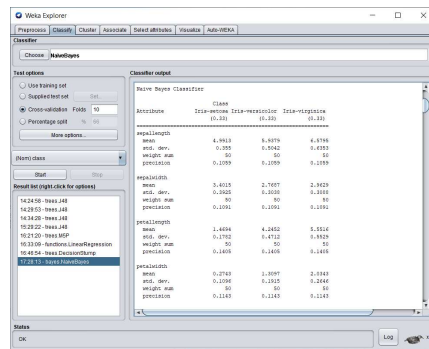
31

## Naïve Bayes classifier (Iris data)



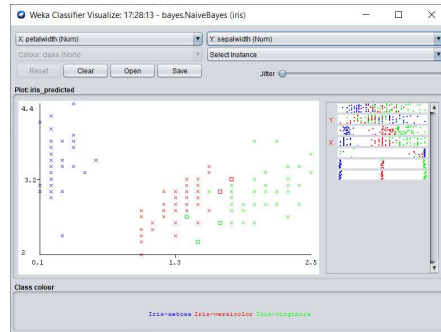
32

## Naïve Bayes classifier uses normal distribution to model numeric attributes



33

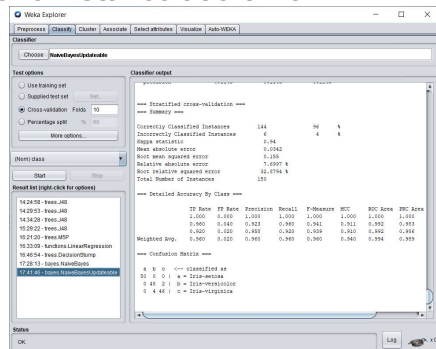
## Naïve Bayes classifier visualization



34

34

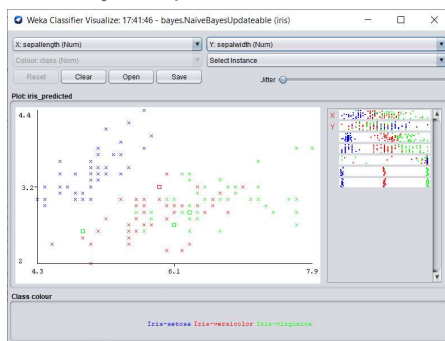
## Naïve Bayes updatable – process one instance at a time



35

35

## Naïve Bayes updatable visualization



36

36

## Explorer: clustering data

- WEKA contains “clusterers” for finding groups of similar instances in a dataset
- Implemented schemes are:
  - k*-Means, EM, Cobweb, *X*-means, FarthestFirst
- Clusters can be visualized and compared to “true” clusters (if given)
- Evaluation based on loglikelihood if clustering scheme produces a probability distribution

37

37

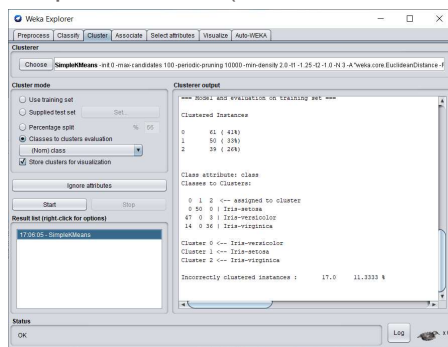
## The K-Means Clustering Method

- Given  $k$ , the *k*-means algorithm is implemented in four steps:
  - Partition objects into  $k$  nonempty subsets
  - Compute seed points as the centroids of the clusters of the current partition (the centroid is the center, i.e., *mean point*, of the cluster)
  - Assign each object to the cluster with the nearest seed point
  - Go back to Step 2, stop when no more new assignment

38

38

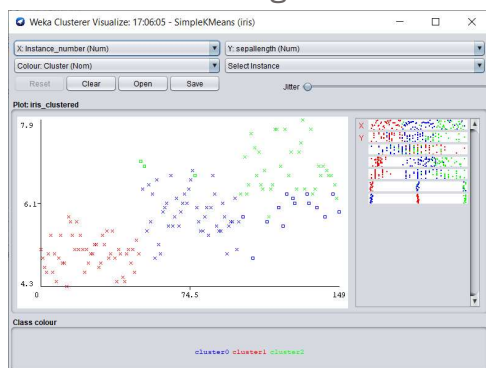
## Clustering – open iris.arff, select Simple Kmeans (numClusters = 3)



39

39

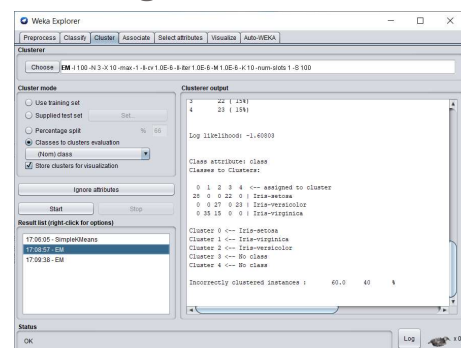
## K-means clustering visualization



40

40

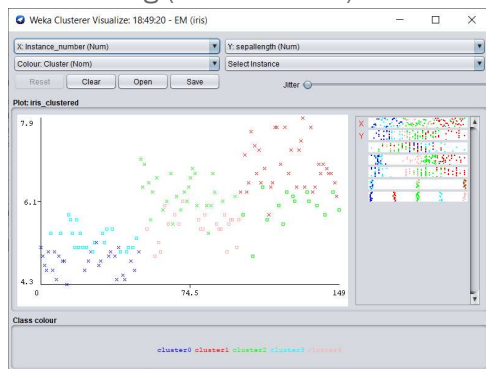
## EM (Expectation-Maximization) clustering – out of the box



41

41

## EM clustering (out of the box) visualization



42

42

### EM clustering – numClusters = 3

Weka Explorer

Preprocess Classify **Cluster** Associate Select attributes Visualize Auto-WEKA

Cluster

Choose EM ( 100-A 2.3-10-max-1-4-on-1.0E-6-Iter-1.0E-6-W 1.0E-6-K 10-num-opts 1-S 100)

Cluster mode

☐ Use training set  
☐ Supplied test set  
☐ Percentage split  
☒ Classes to clusters evaluation  
☒ Store clusters for visualization  
☐ Ignore attributes

Start Stop

Result list (right-click for options)

17:05:05 - SimpleKMeans  
 17:08:07 - EM  
 17:09:38 - EM

Status

OK Log

Cluster output

0 44 ( 43%)  
 1 50 ( 33%)  
 2 34 ( 24%)

log likelihood: -2.055

Class attribute: class  
 Classes to Clusters:  
 0 1 2 <== assigned to cluster  
 0 50 0 1 Iris-setosa  
 0 0 50 1 Iris-versicolour  
 14 0 34 1 Iris-virginica  
 Cluster 0 <== Iris-setosa  
 Cluster 1 <== Iris-versicolour  
 Cluster 2 <== Iris-virginica  
 Incorrectly clustered instances : 14.0 9.333 %

43

### EM (numClusters = 3) visualization

Weka Clusterer Visualize: 17:09:38 - EM (iris)

X: Instance\_number (Num) Y: Sepallength (Num)

Colour: Cluster (Nom) Select Instance

Reset Clear Open Save Jitter

Plot iris\_clustered

Class colour

cluster0 cluster1 cluster2

44

### Hierarchical clustering (numClusters = 3)

Weka Explorer

Preprocess Classify **Cluster** Associate Select attributes Visualize Auto-WEKA

Cluster

Choose HierarchicalCluster (N 3-L 50-K 0.1-P 0-A "Weka core EuclideanDistance-R Fast-Inf")

Cluster mode

☐ Use training set  
☐ Supplied test set  
☐ Percentage split  
☒ Classes to clusters evaluation  
☐ Print class  
☒ Store clusters for visualization  
☐ Ignore attributes

Start Stop

Result list (right-click for options)

17:05:05 - SimpleKMeans  
 17:08:07 - EM  
 17:09:38 - EM  
 17:20:55 - HierarchicalCluster  
 17:21:07 - HierarchicalCluster  
 17:21:21 - HierarchicalCluster  
 17:21:37 - HierarchicalCluster  
 17:22:07 - HierarchicalCluster

Status

OK Log

Cluster output

Model and evaluation on training set ---

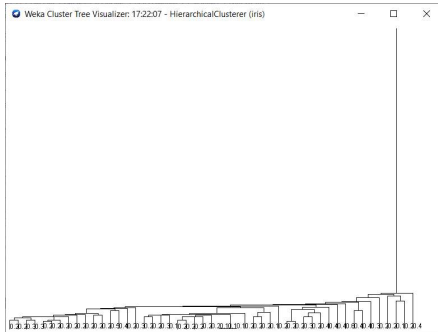
Clustered Instances

0 49 ( 33%)  
 1 5 ( 3%)  
 2 100 ( 67%)

Class attribute: class  
 Classes to Clusters:  
 0 2 2 <== assigned to cluster  
 49 1 0 1 Iris-setosa  
 0 0 50 1 Iris-versicolour  
 0 0 50 1 Iris-virginica  
 Cluster 0 <== Iris-setosa  
 Cluster 1 <== Iris-versicolour  
 Cluster 2 <== Iris-virginica  
 Incorrectly clustered instances : 51.0 34 %

45

## Hierarchical clustering – tree visualization



46

---

---

---

---

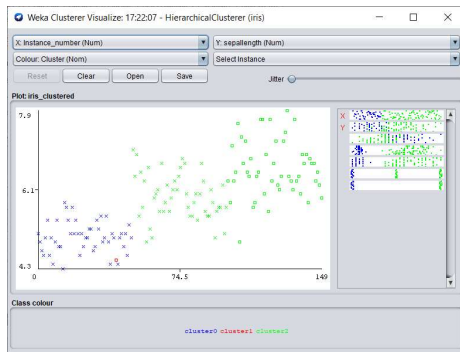
---

---

---

---

## Hierarchical clustering - visualization



47

---

---

---

---

---

---

---

---

## Explorer: finding associations

- WEKA contains an implementation of the Apriori algorithm for learning association rules
  - Works only with discrete data
- Can identify statistical dependencies between groups of attributes:
  - milk, butter  $\Rightarrow$  bread, eggs (with confidence 0.9 and support 2000)
- Apriori can compute all rules that have a given minimum support and exceed a given confidence

48

48

---

---

---

---

---

---

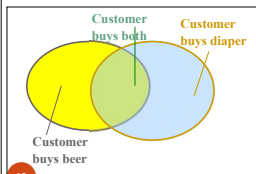
---

---



## Basic Concepts: Frequent Patterns

Tid	Items bought
10	Beer, Nuts, Diaper
20	Beer, Coffee, Diaper
30	Beer, Diaper, Eggs
40	Nuts, Eggs, Milk
50	Nuts, Coffee, Diaper, Eggs, Milk

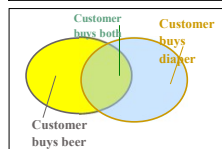


- **itemset**: A set of one or more items
- **k-itemset**  $X = \{x_1, \dots, x_k\}$
- **(absolute) support**, or, **support count** of  $X$ : Frequency or occurrence of an itemset  $X$
- **(relative) support**,  $s$ , is the fraction of transactions that contains  $X$  (i.e., the probability that a transaction contains  $X$ )
- An itemset  $X$  is **frequent** if  $X$ 's support is no less than a *minsup* threshold

49

## Basic Concepts: Association Rules

Tid	Items bought
10	Beer, Nuts, Diaper
20	Beer, Coffee, Diaper
30	Beer, Diaper, Eggs
40	Nuts, Eggs, Milk
50	Nuts, Coffee, Diaper, Eggs, Milk



- Find all the rules  $X \rightarrow Y$  with minimum support and confidence
- **support**,  $s$ , probability that a transaction contains  $X \cup Y$
- **confidence**,  $c$ , conditional probability that a transaction having  $X$  also contains  $Y$

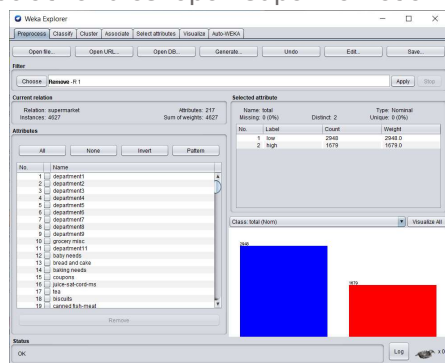
Let *minsup* = 50%, *minconf* = 50%

Freq. Pat.: Beer:3, Nuts:3, Diaper:4, Eggs:3, {Beer, Diaper}:3

- Association rules: (many more!)
  - $\text{Beer} \rightarrow \text{Diaper}$  (60%, 100%)
  - $\text{Diaper} \rightarrow \text{Beer}$  (60%, 75%)

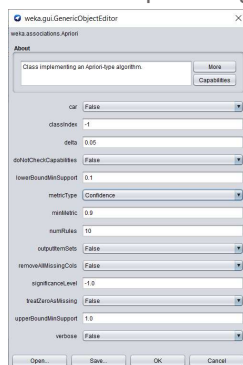
50

## Associative rules: open supermarket.arff



51

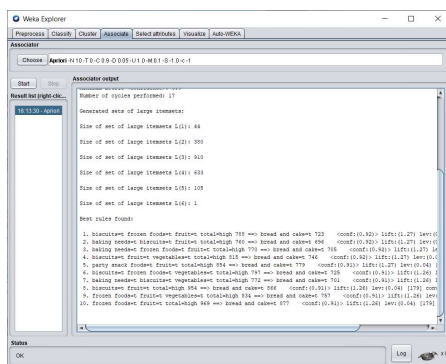
## Associative rules: Apriori algorithm



52

52

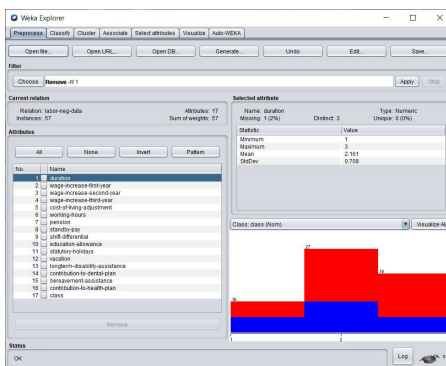
## Associative rules – Apriori output



53

53

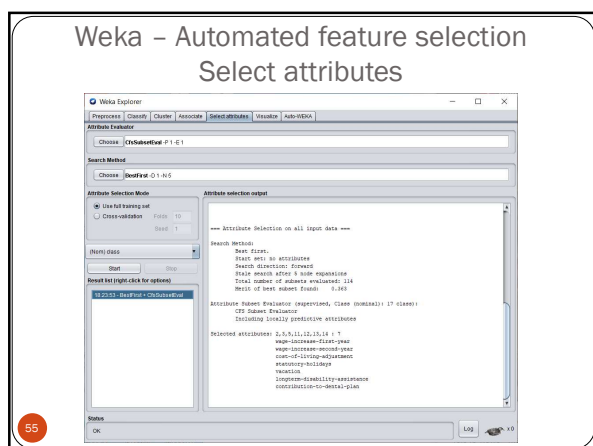
## Weka – Automated feature selection – labor.arff



54

54

Weka – Automated feature selection  
Select attributes



55

---

---

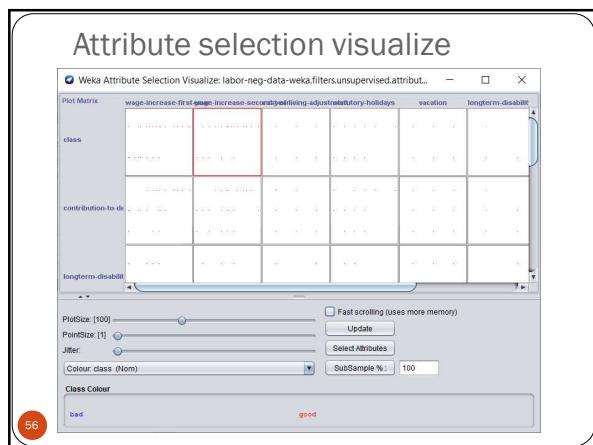
---

---

---

---

## Attribute selection visualize



56

---

---

---

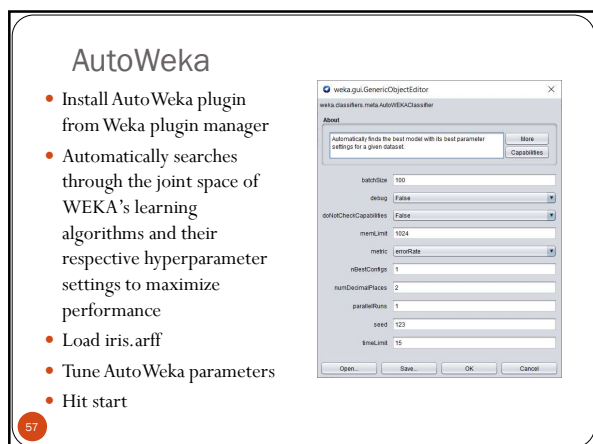
---

---

---

## AutoWeka

- Install AutoWeka plugin from Weka plugin manager
- Automatically searches through the joint space of WEKA's learning algorithms and their respective hyperparameter settings to maximize performance
- Load iris.arff
- Tune AutoWeka parameters
- Hit start



57

---

---

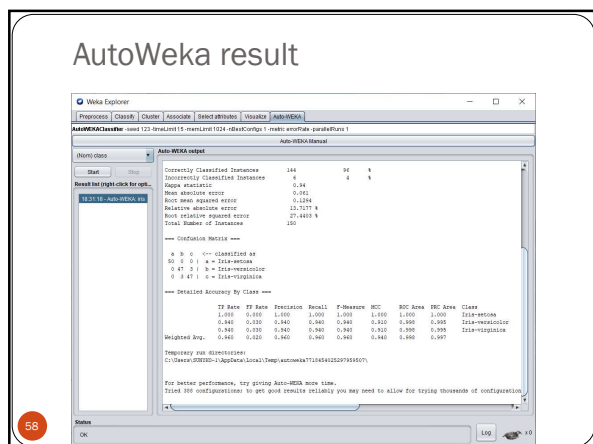
---

---

---

---

## AutoWeka result



58

## References and Resources

- References:
  - WEKA website: <http://www.cs.waikato.ac.nz/~ml/weka/index.html>
  - WEKA Tutorial:
    - Machine Learning with WEKA: A [presentation](#) demonstrating all graphical user interfaces (GUI) in Weka.
    - A [presentation](#) which explains how to use Weka for exploratory data mining.
  - WEKA Data Mining Book:
    - Ian H. Witten and Eibe Frank, Data Mining: Practical Machine Learning Tools and Techniques (Fourth Edition)
  - WEKA Wiki: [http://weka.sourceforge.net/wiki/index.php/Main\\_Page](http://weka.sourceforge.net/wiki/index.php/Main_Page)
  - AutoWeka Software:
    - <http://www.cs.ubc.ca/labs/beta/Projects/autoweka/#software>
  - Others:
    - Jiawei Han and Micheline Kamber, Data Mining: Concepts and Techniques, 2nd ed.

59