

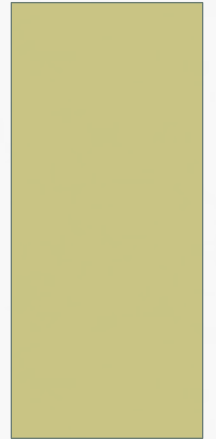


Ricardo Aler Mur

- In this lecture, the Machine Learning subject is introduced by using a classification task example, where sky objects have to be classified, that illustrates the main processes that must be followed in other classification tasks. Also, some application examples are used to illustrate the possible domains of application of Machine Learning.
- Then, three main concepts are introduced:
 - What can be done (tasks)
 - What kind of models can be learned to solve those tasks
 - Each type of model can be generated by several different algorithms
- Finally, each kind of task is illustrated by giving an example, showing what the input data looks like and how the obtained models can be interpreted.

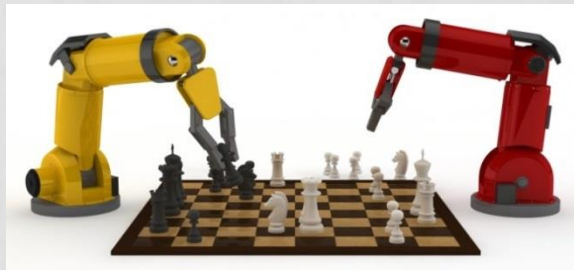
PRESENTATION
MACHINE LEARNING I
MASTER IN BIG DATA ANALYTICS

RICARDO ALER MUR (aler@inf.uc3m.es). 2.2B29



MACHINE LEARNING

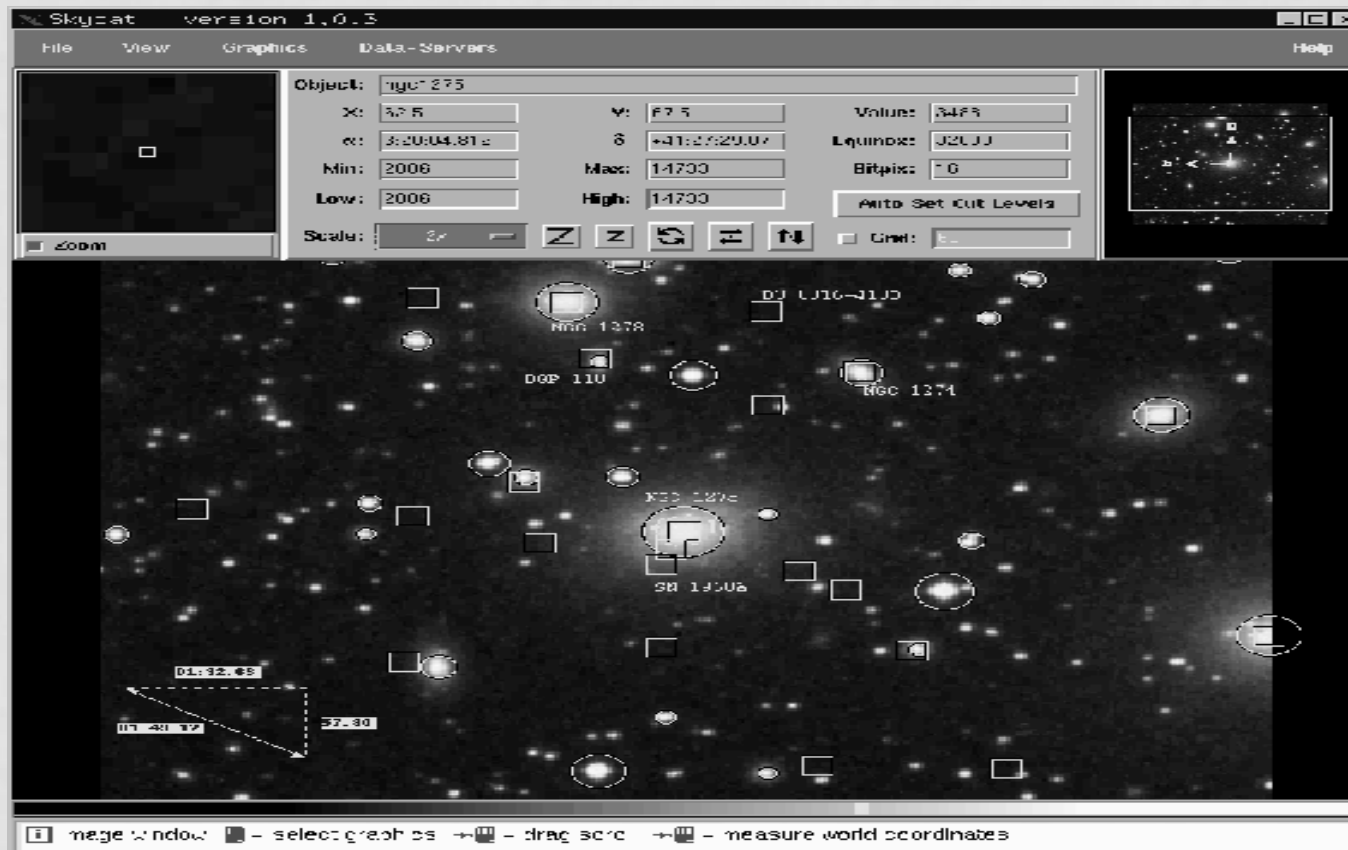
- In general, it's a subfield of **Artificial Intelligence** that tries to make computers and machines learn



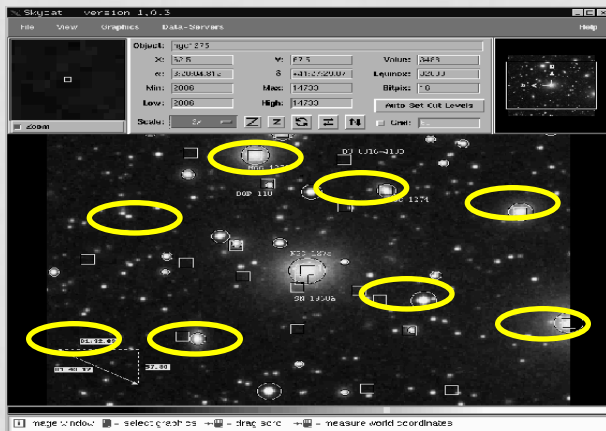
- In practice, it tries to create models from data and thus is closely related to statistics. This is the point of view we will follow in this course

WHAT IS MACHINE LEARNING

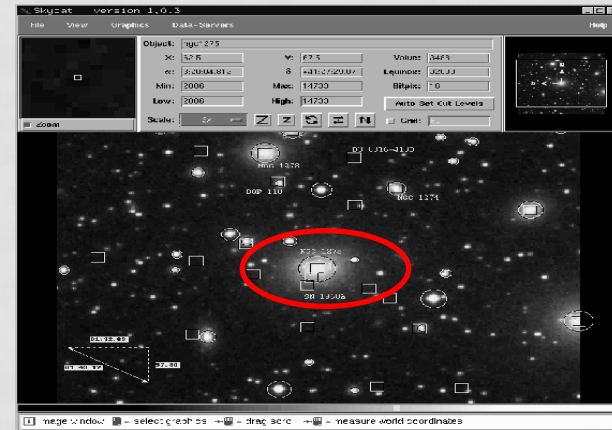
- Example: Skycat: AUTOMATIC CLASSIFICATION OF OBJECTS IN THE SKY



Training data (labeled pictures of sky objects: galaxies, stars, nebulae, ...)



Pictures in the catalog have been labeled by a human expert (astronomer)



?

ML
Algorithm

Model

Spiral galaxy

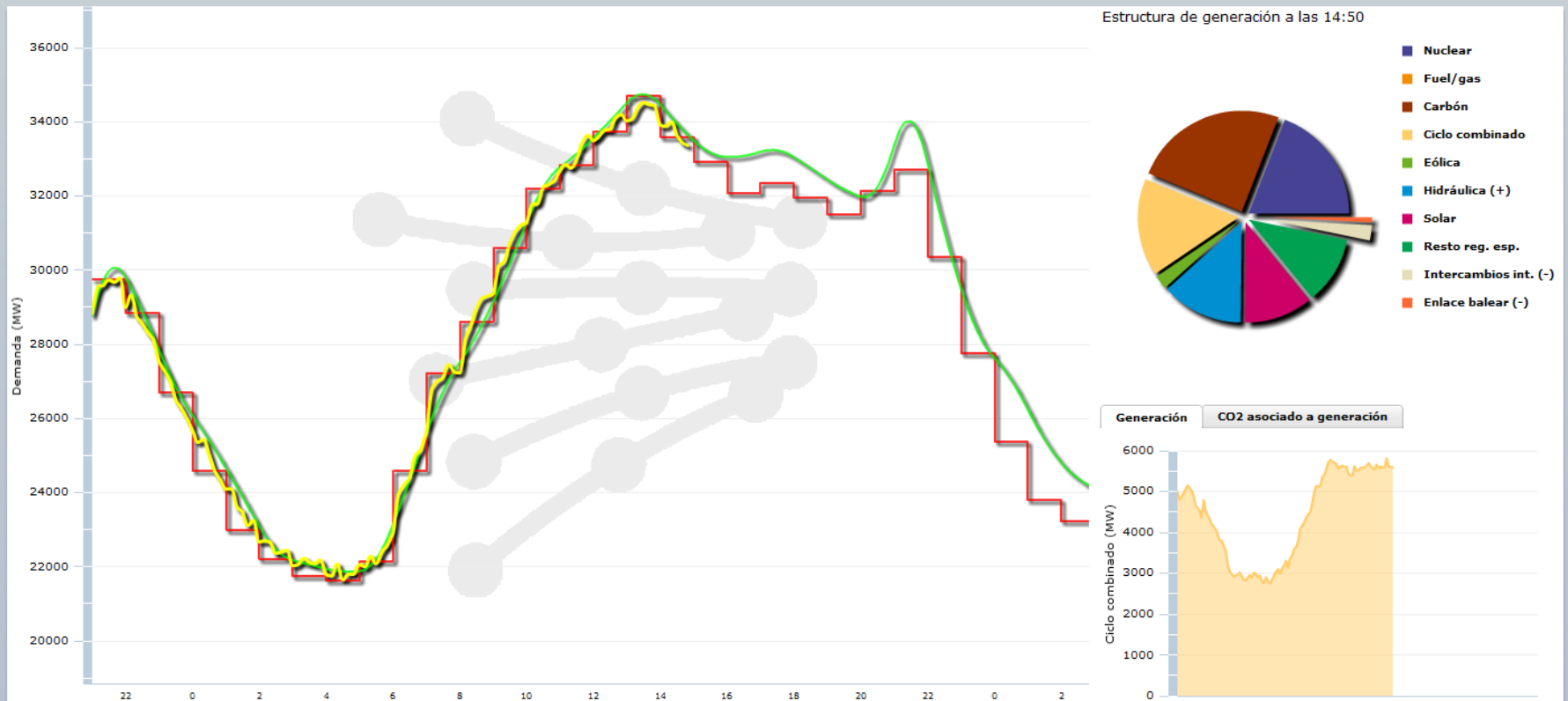
APPLICATIONS

- Finances and banking
 - Credit card fraud detection
 - Credit default prediction
- Market analysis:
 - Market basket analysis
 - Market segmentation
- Insurance:
 - Expensive clients
- Education:
 - Prediction of school dropouts
- Industry:
 - Electric (energy) load forecasting
 - Solar / wind energy forecasting

<https://demanda.ree.es/demanda.html>

ELECTRIC LOAD FORECASTING

Demanda de energía eléctrica en tiempo real, estructura de generación y emisiones de CO2



Demanda (MW) a las 14:50 de 08/09/2014 Real = 33357 Prevista = 33648 Emisiones CO2 (t/h) = 11500

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APPLICATIONS II

- Medicine:
 - Illness diagnosis
- Science:
 - Illness prediction from DNA analysis
 - Prediction if a new substance causes cancer
 - SKYCAT
- Internet:
 - Spam detection (SpamAssassin)
 - Web: book recommendation (amazon.com)



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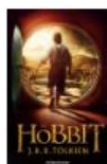
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Descripción del producto

Descripción del producto

Concebida en un primer momento como una continuación de El Hobbit, acabó por convertirse en una historia independiente por derecho propio de mucho más alcance y extensión. En 1999 la trilogía de El Señor de los Anillos fue elegida como «Libro del Milenio» por los participantes de una encuesta de Amazon.com. En la adormecida e idílica Comarca, un joven hobbit recibe un encargo: custodiar el Anillo Único y emprender el viaje para su destrucción en las Grietas del Destino. Consciente de la importancia de su misión, Frodo abandona la Comarca e inicia el camino hacia Mordor con la compañía inesperada de Sam, Pippin y Merry. Pero sólo con la ayuda de Aragorn conseguirán vencer a los Jinetes Negros y alcanzar el refugio de la Casa de Elrond en Rivendel.

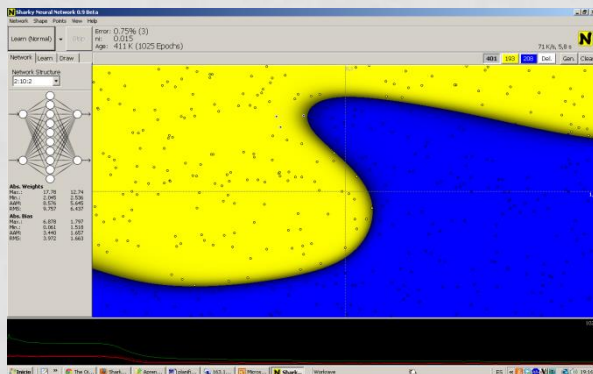
SYLLABUS

1. Overview and introduction to Machine Learning: tasks and models.
2. Predictive models:
 - Decision trees, regression trees
 - K Nearest Neighbour (KNN)
 - Machine Learning pipeline: training, => ML algorithm => model => test / evaluation . Preprocessing, hyperparameter tuning, ...
3. Ensemble methods: bagging, boosting, stacking
4. Preprocessing: selection of attributes and methods of dimensionality reduction
5. Machine learning software for Big Data:
 1. Python: scikit-learn, numpy
 2. Mapreduce
 3. Spark: pyspark, MLLIB
6. Other topics:
 1. Online learning
 2. Metaheuristics: genetic algorithms, genetic programming, ...

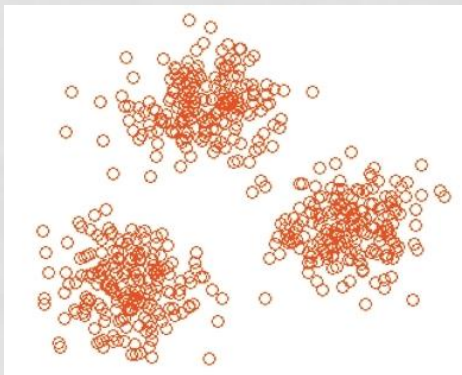
TASKS AND ALGORITHMS

- What can be done?

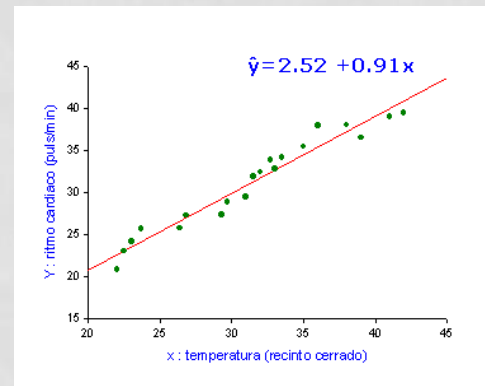
Classification



Clustering



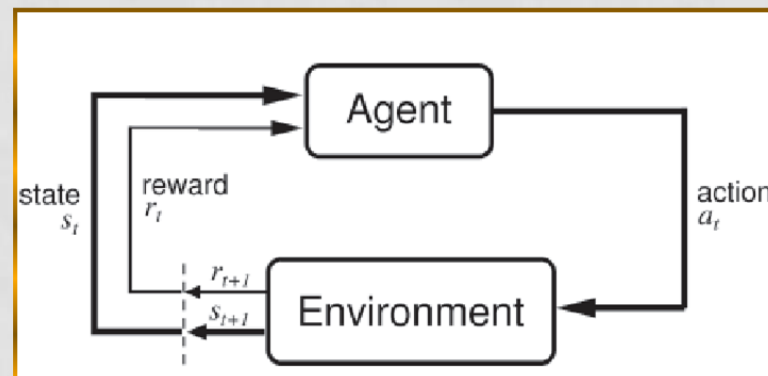
Regression



Market basket analysis



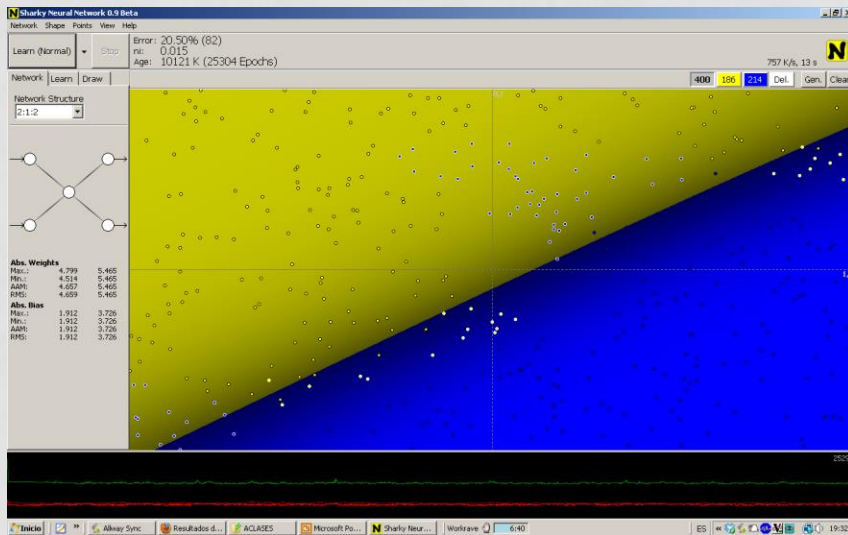
Reinforcement learning



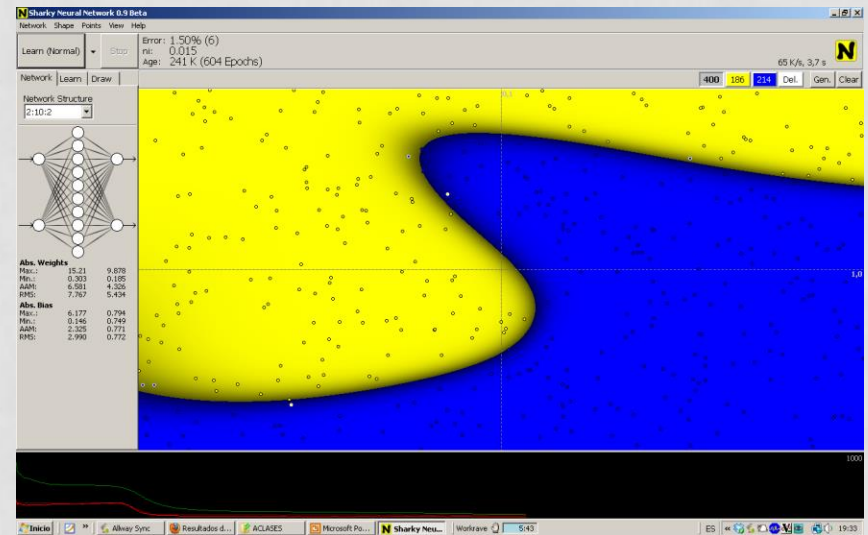
MODELS

- What models can be obtained?

Linear



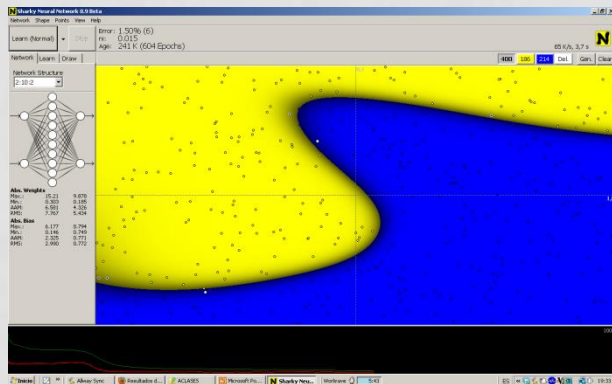
Non linear



MODELS

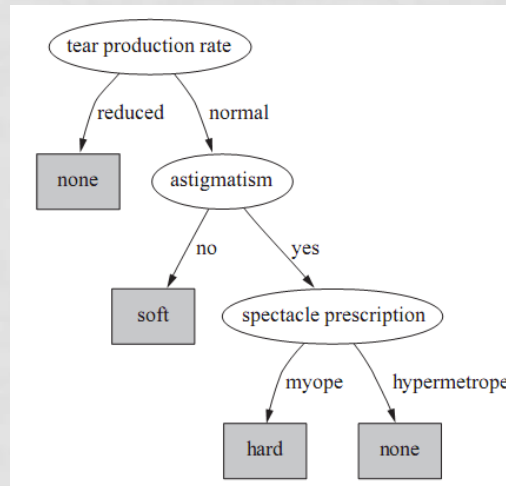
- What models can be obtained?

Functions: $y = 3x^3 + 2$

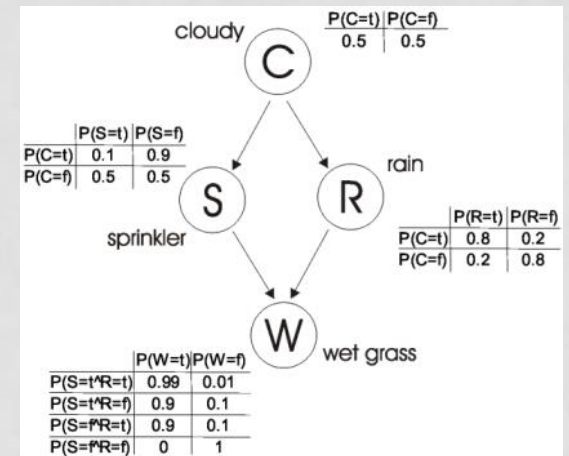


Rules

Decision trees



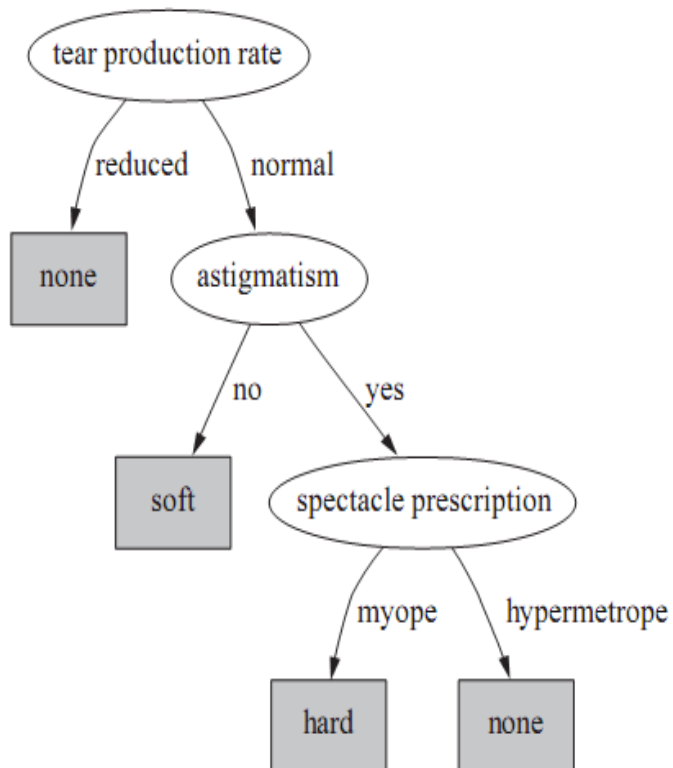
Bayesian networks



If humidity = normal and windy = false then play = yes

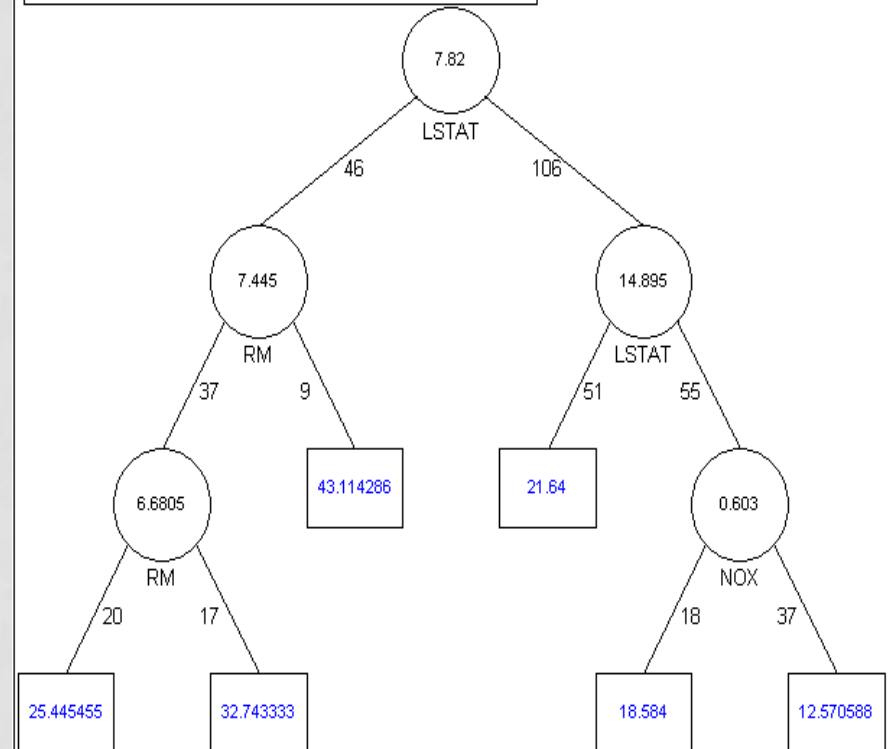
And many more: neural networks, nearest neighbor, ...

Decision trees and regression trees

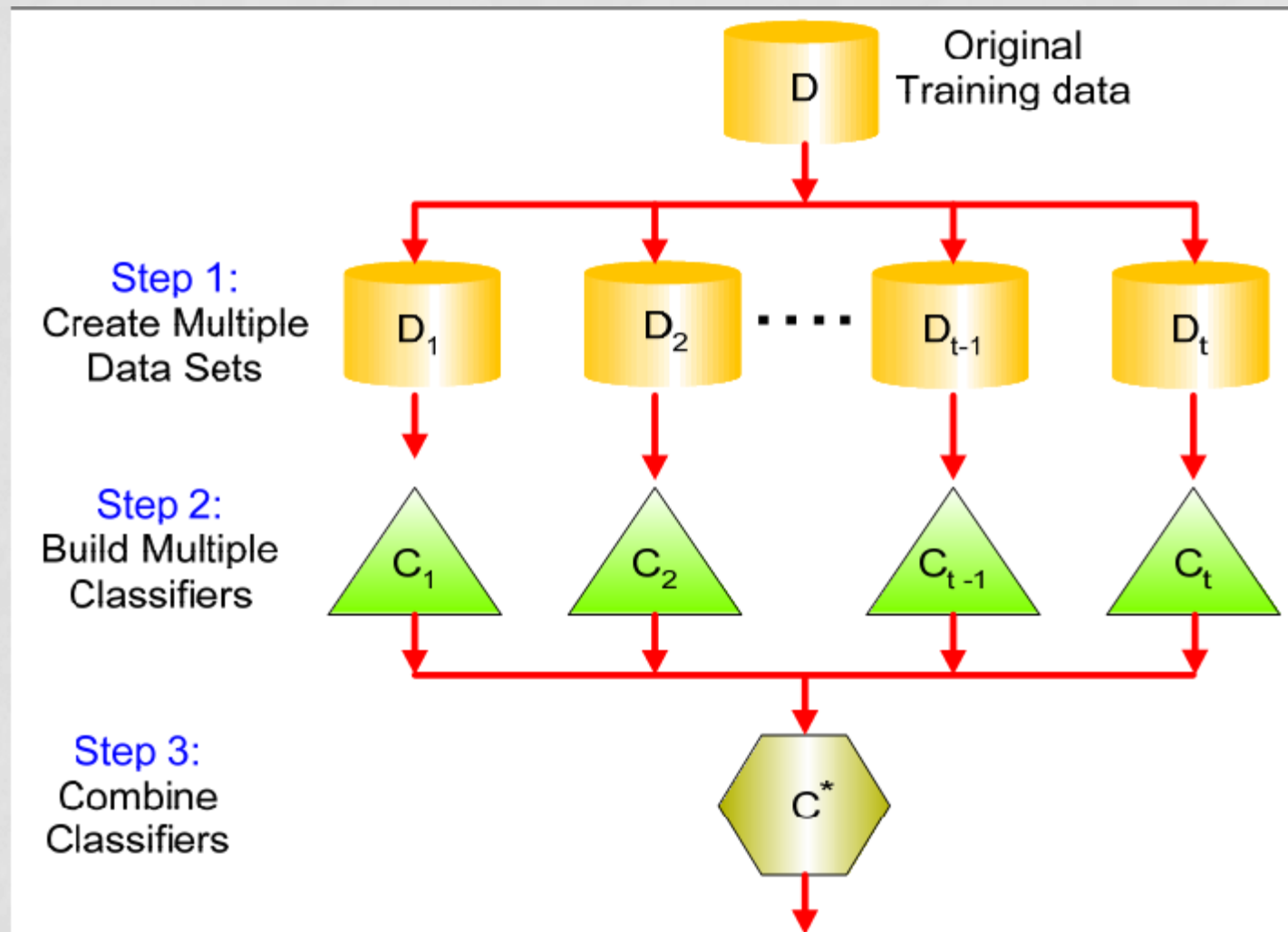


XLMiner : Regression Tree - Pruned Tree (Using Validation Data)

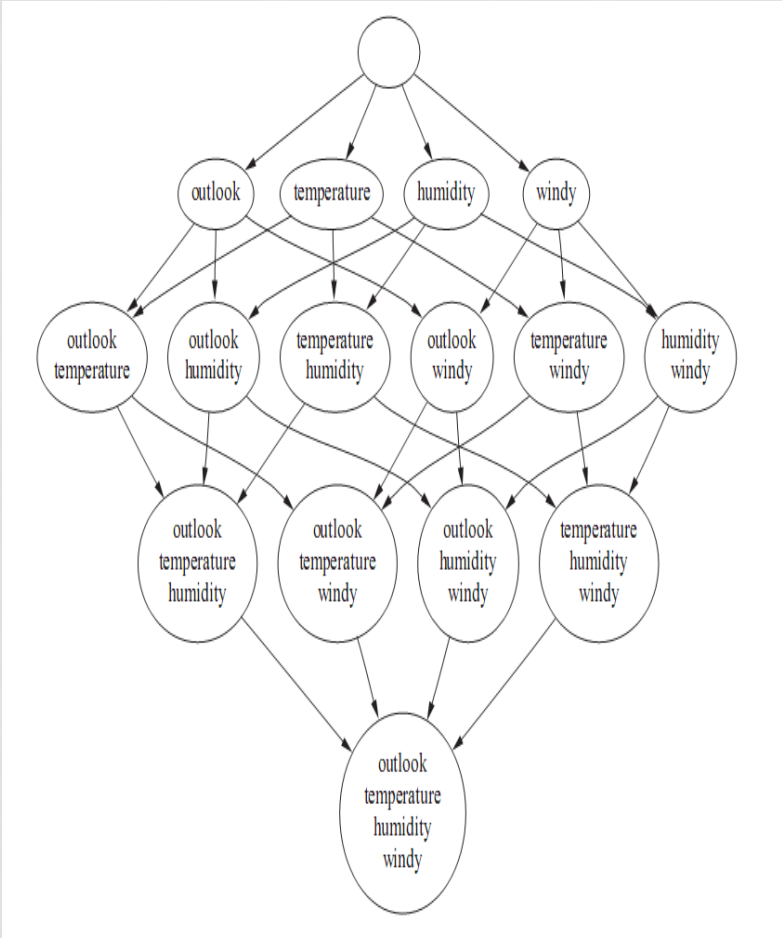
Place the cursor on a Decision Node to read the decision rule



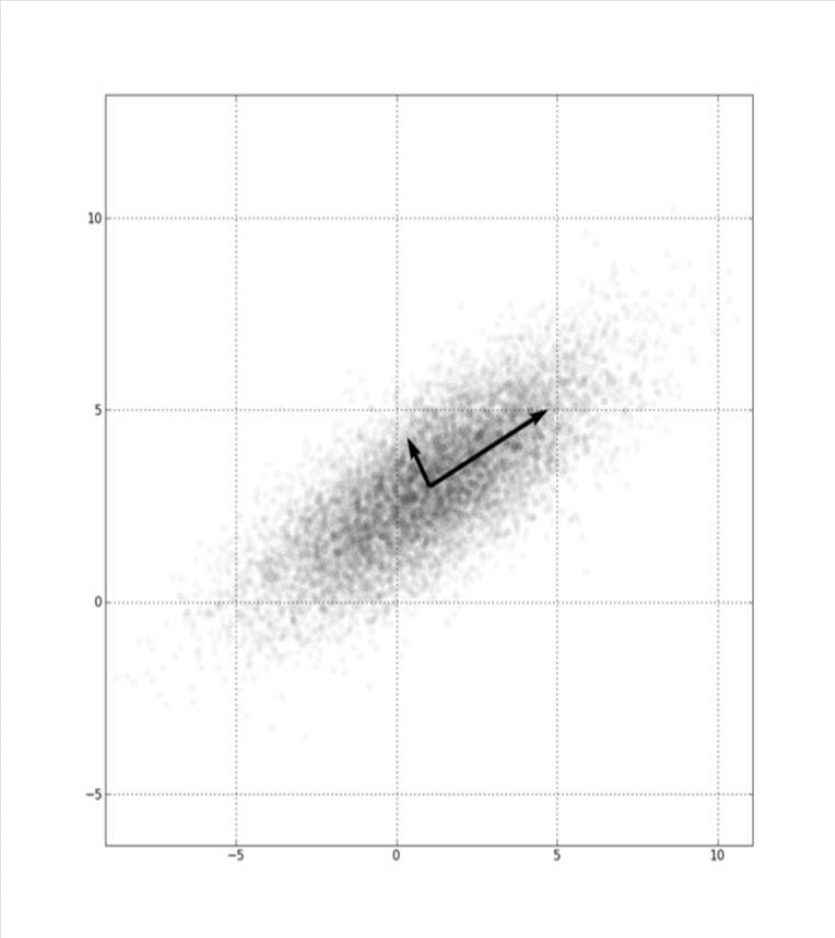
Ensembles of classifiers



ATTRIBUTE SELECTION AND TRANSFORMATION

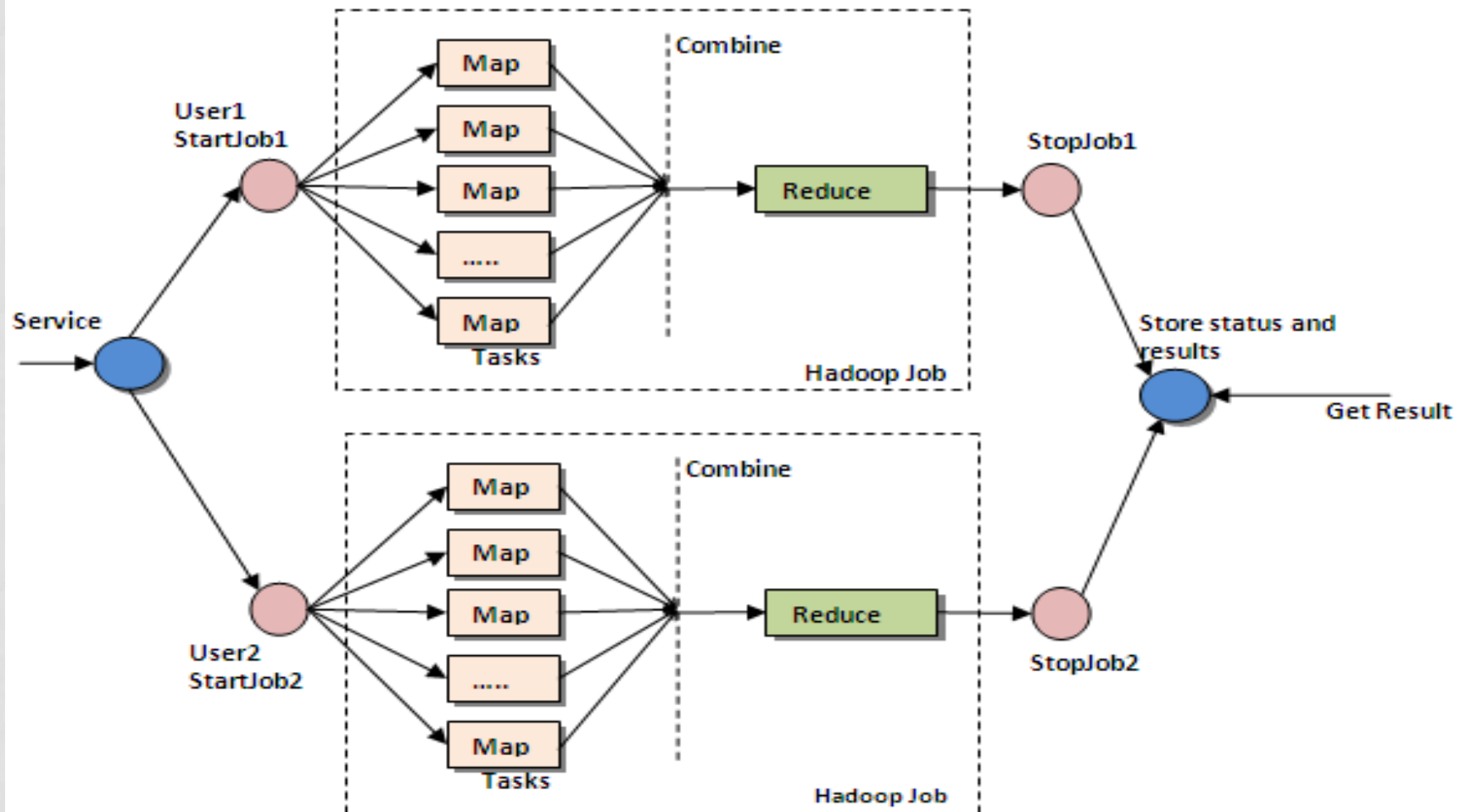


Attribute selection



Principal Component Analysis and Random Projections

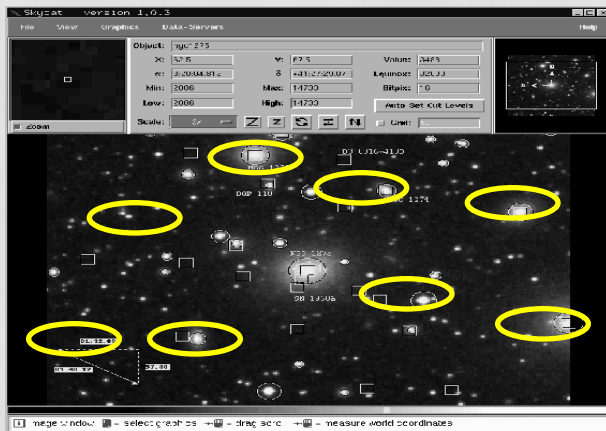
BIG DATA / MAP-REDUCE, SPARK (MLLIB)



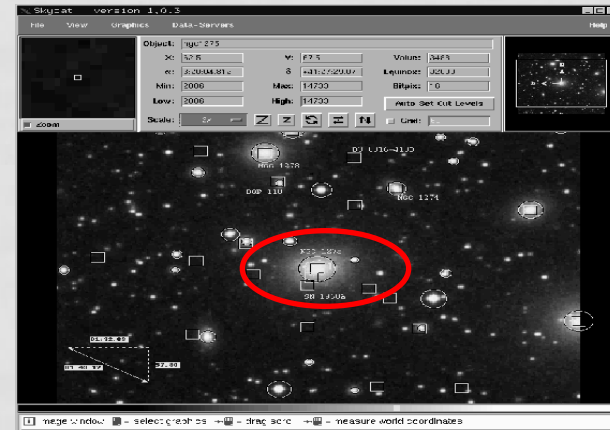
TASKS / MODELS / ALGORITHMS

- What can be done? Tasks:
 - Supervised ML: classification, regression, ...
 - Unsupervised ML: clustering, association, ...
 - Semi-supervised ML
 - Reinforcement learning
- What kind of models can be learned?
 - Attribute-value:
 - Trees
 - Nearest neighbor
 - Functions: neural networks, support vector machines, ...
 - Bayesian networks
 - Ensembles (bagging, boosting, stacking, ...)
 - Relational
- How can models be learned? Algorithms:
 - Linear models: linear regression, simple perceptron, naive bayes, SVM with linear kernel, ...
 - Neural networks: backpropagation, rprop, ...
 - Decision trees: ID3, C4.5, C5.0, ...
 - Nearest neighbour: IB1, ...

Training data (labeled pictures of sky objects: galaxies, stars, nebulae, ...)



Pictures in the catalog have been labeled by a human expert (astronomer)



?

ML
Algorithm

Model

- Trees
- Nearest neighbor
- Functions: neural networks, support vector machines, ...
- Bayesian networks
- Ensembles (bagging, boosting, stacking, ...)

Spiral galaxy

TASKS

- **Inductive learning**(from instances)
 - **Supervised learning:**
 - **Classification:**
 - Regression
 - Semi-supervised learning
 - Unsupervised learning:
 - Clustering
 - Association
 - Reinforcement learning

CLASSIFICATION TASK. AN EXAMPLE:

- Bank credit approval:
 - An Internet bank owns a large data base with information about clients whose credits were approved or rejected
 - The banks requires a model to determine if a new customer will repay the loan or not
 - Instances (client records in the database):
 - Input attributes : credit time-length (years), amount, overdue accounts?, own house?
 - Class: yes/no
 - Rule-based model:
 - **IF** (overdue accounts > 0) **THEN** repay loan = no
 - **IF** (overdue accounts = 0) **AND** ((salary > 2500) **OR** (years > 10)) **THEN** repay loan = yes

CLASSIFICATION TASK. AN EXAMPLE:

T = training set (instances)

Years	Amount	Salary	Own house?	Overdue accounts?	Repay loan
15	60000	1900	Yes	2	No
2	30000	3500	Yes	0	Yes
9	9000	1700	Yes	1	No
15	18000	3000	No	0	Yes
10	24000	2100	No	0	No
...

x (or input attributes)

y (class, or
output attribute)

test set

Years	Amount	Salary	Own house?	Overdue accounts?	Repay loan
10	50000	3000	Yes	0	??

Algorithm

Model

IF OA > 0 THEN NO

**IF OA == 0 AND
S > 2500 THEN Yes**

Repay loan =
yes

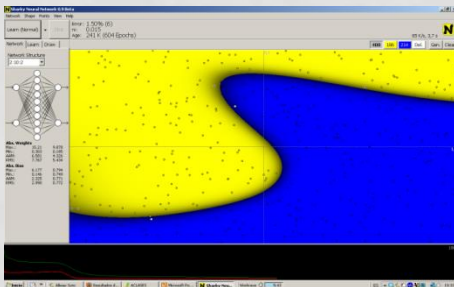
IMPORTANT: MODELS

- In the previous slide, the model built from training data is a set of rules:

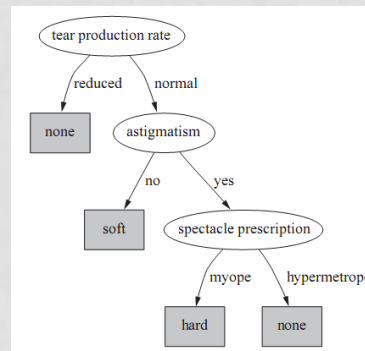
IF OA >0 **THEN** NO **ELSEIF** OA==0 **AND** S>2500 **THEN** Yes

- But there are many more that can be learned:

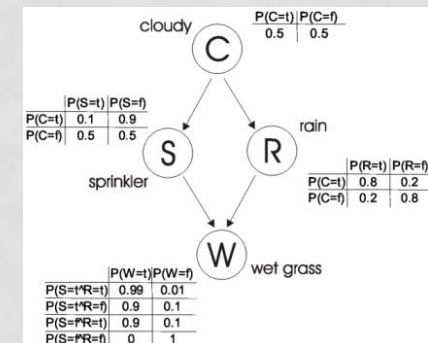
Functions: $y = 3x^3 + 2$



Decision trees



Bayesian networks



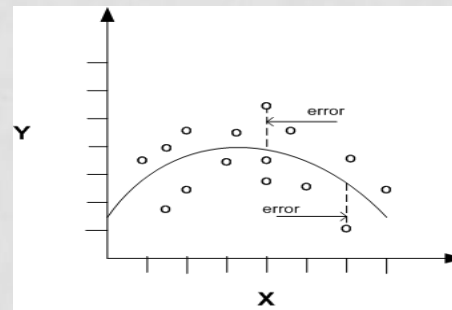
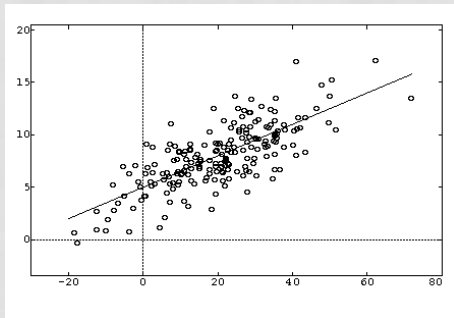
And many more: neural networks, nearest neighbor, support vector machines (SVMs).

TASKS

- **Inductive learning**(from instances)
 - **Supervised learning:**
 - Classification
 - **Regression**
 - Semi-supervised learning
 - Unsupervised learning:
 - Clustering
 - Association
 - Reinforcement learning

REGRESSION

- If the class is continuous, it is a **regression** problem
- Models are typically mathematical functions $y=g(x)$
 - Linear: $y = ax+b$
 - Non linear: $y = a*x^2+bx+c$ / $y = \log(\sin(x))$





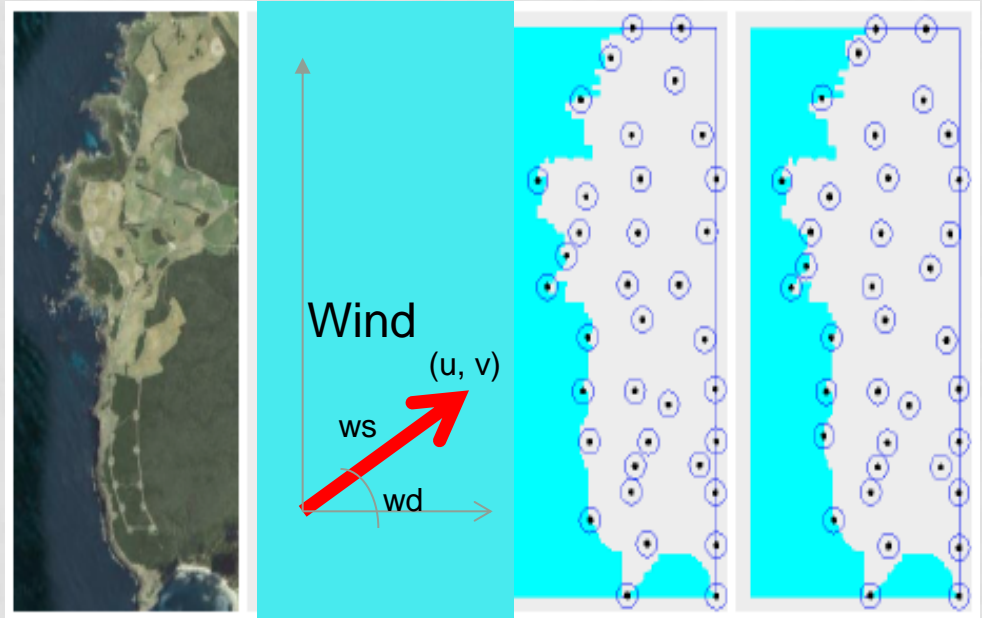
REGRESSION EXAMPLE

- A wind power forecasting problem: predicting hourly power generation at 7 wind farms



Some input variables:

- ws: wind speed
- wd: wind direction
- (u,v): wind direction vector



Model to estimate electricity production from ws, wd, u, v?
 $wp = f(ws, wd, u, v, \dots)$

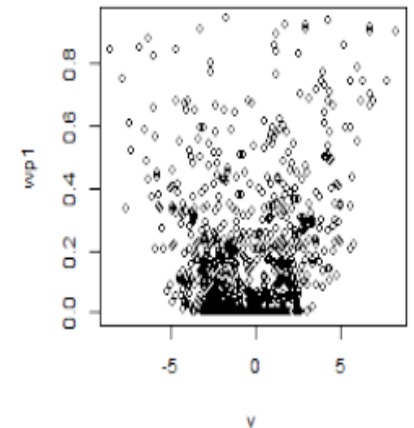
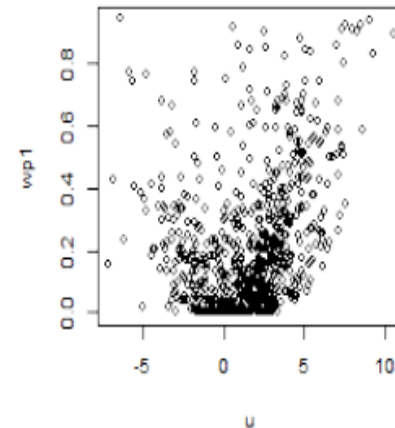
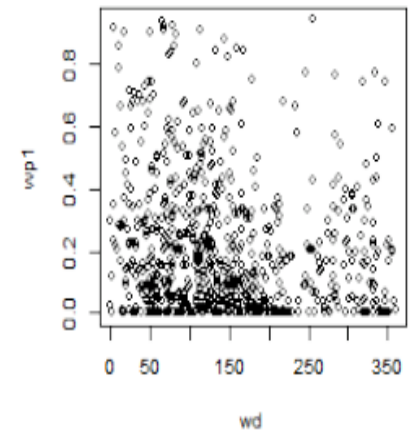
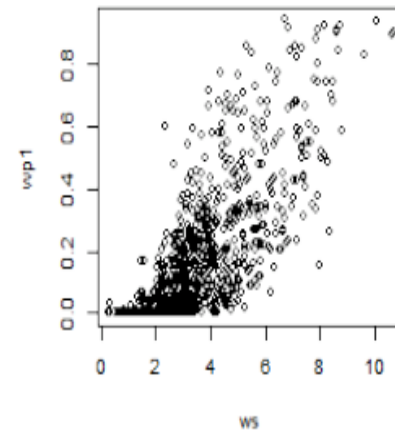
REGRESSION EXAMPLE

DATA

	date	hors	u	v	ws	wd	dateB	wp1	
1	2009-07-01	01:00:00	1	2.34	-0.79	2.47	108.68	2009-07-01	0.085
2	2009-07-01	02:00:00	2	2.18	-0.99	2.40	114.31	2009-07-01	0.020
3	2009-07-01	03:00:00	3	2.20	-1.21	2.51	118.71	2009-07-01	0.060
4	2009-07-01	04:00:00	4	2.35	-1.40	2.73	120.86	2009-07-01	0.045
5	2009-07-01	05:00:00	5	2.53	-1.47	2.93	120.13	2009-07-01	0.035
6	2009-07-01	06:00:00	6	2.66	-1.29	2.96	115.79	2009-07-01	0.005

Some input variables:

- ws: wind speed
- wd: wind direction
- (u,v): wind direction vector



REGRESSION EXAMPLE

DATA

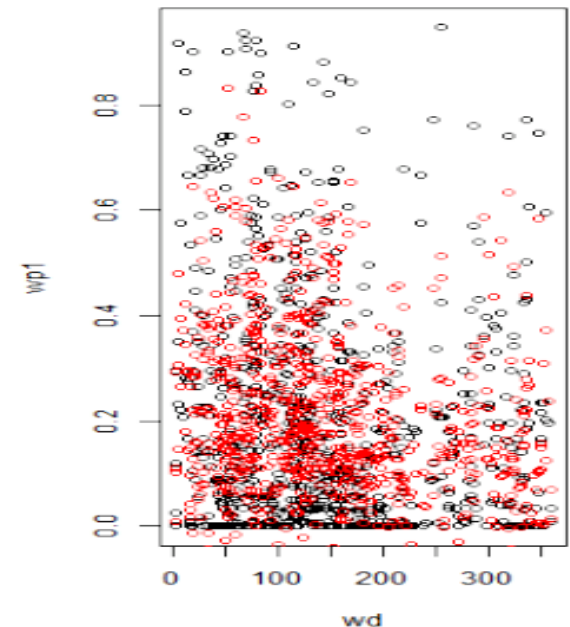
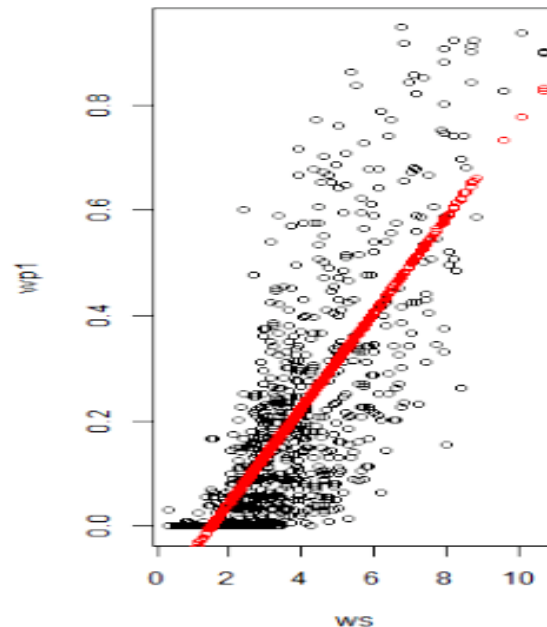
		date	hors	u	v	ws	wd	dateB	wp1
1	2009-07-01	01:00:00	1	2.34	-0.79	2.47	108.68	2009-07-01	0.085
2	2009-07-01	02:00:00	2	2.18	-0.99	2.40	114.31	2009-07-01	0.020
3	2009-07-01	03:00:00	3	2.20	-1.21	2.51	118.71	2009-07-01	0.060
4	20-axis	04:00:00	4	2.35	-1.40	2.73	120.86	2009-07-01	0.045
5	2009-07-01	05:00:00	5	2.53	-1.47	2.93	120.13	2009-07-01	0.035
6	2009-07-01	06:00:00	6	2.66	-1.29	2.96	115.79	2009-07-01	0.005

Linear model:

$$wp = f(ws, wd, u, v)$$

$$wp = a_1 * ws + a_2 * wd + a_3 * u + a_4 * v + b$$

Obviously, a non-linear model could do better

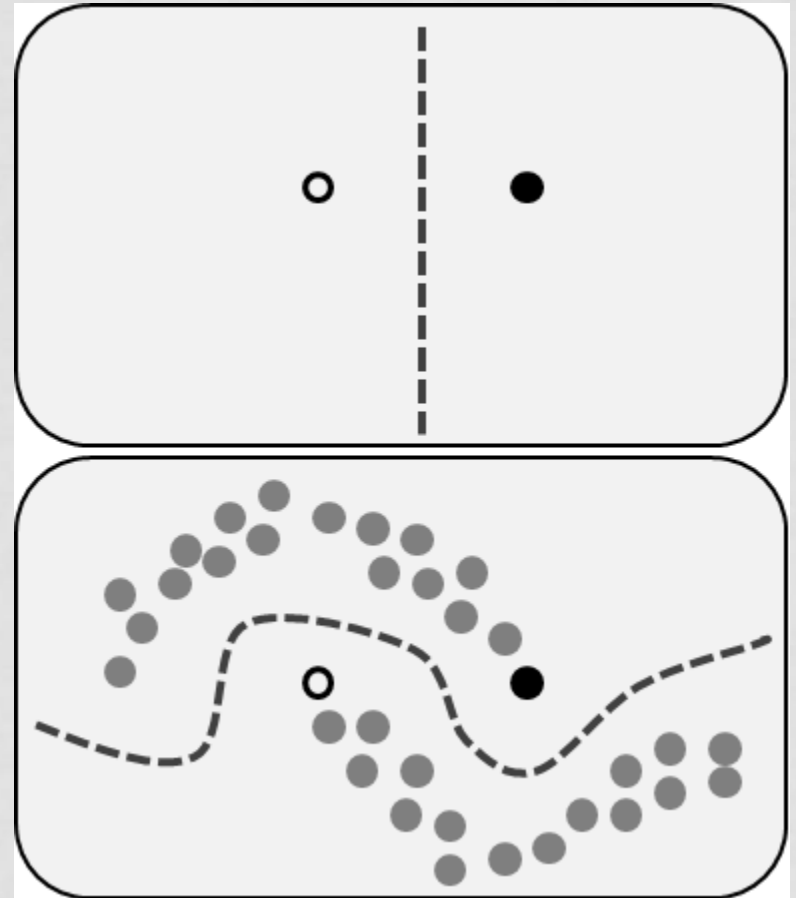


TASKS

- **Inductive learning**(from instances)
 - Supervised learning:
 - Classification
 - Regression
 - **Semi-supervised learning**
 - Unsupervised learning:
 - Clustering
 - Association
 - Reinforcement learning

SEMISUPERVISED LEARNING

- When both labelled and unlabelled instances are available
- Why: labelling instances may be costly (ex: to perform a biopsy to determine if a person has cancer)



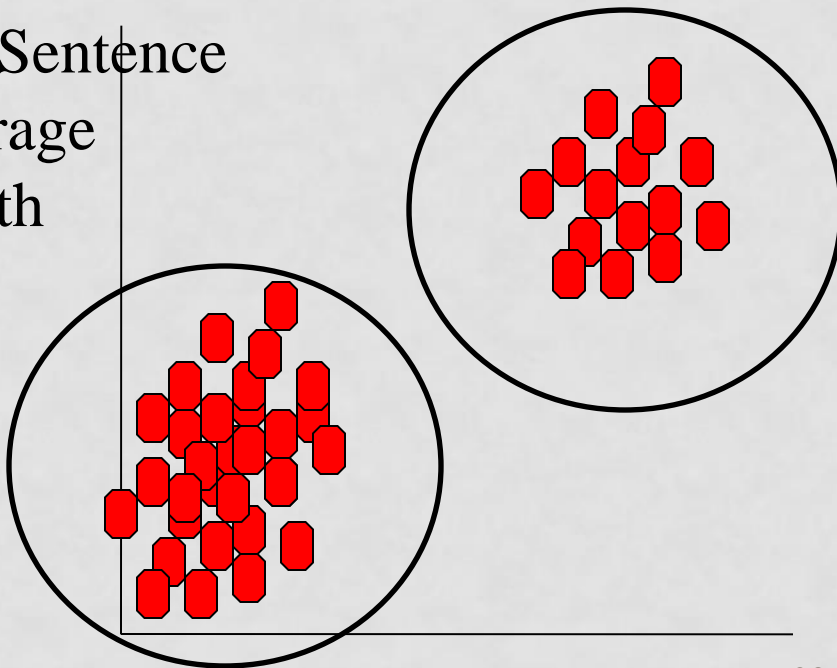
TASKS

- **Inductive learning**(from instances)
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 - **Unsupervised learning:**
 - **Clustering**
 - Association
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UNSUPERVISED LEARNING (NO LABELS): CLUSTERING

- To determine natural clusterings in instance space, based on the input attributes (no labels)

X2: Sentence
Average
length



X1: Word average³² length

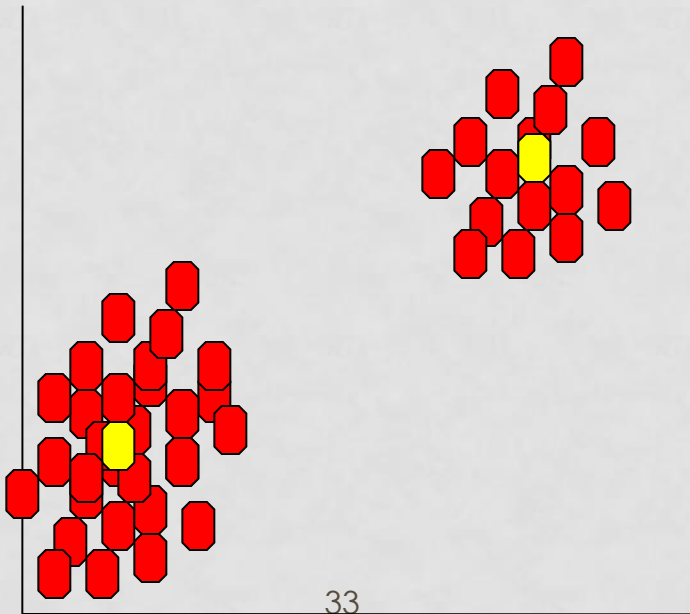
Example: each point is a different book. 2 groups:

- * Long words and sentences (philosophy?)
- * Short words and sentences (best-sellers?)

CLUSTER REPRESENTATION

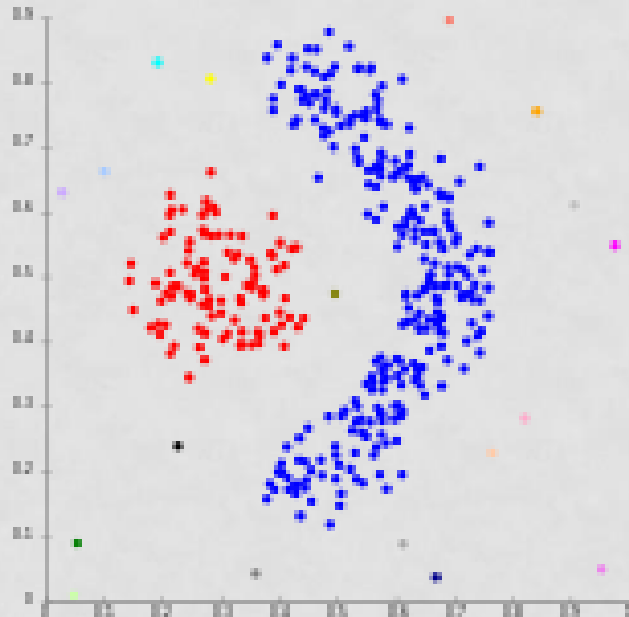
- Most commonly: centroids (ex: k-means algorithm)

K-MEANS: <http://www.youtube.com/watch?v=74rv4snLI70>



CLUSTERING

- Clustering is not so well defined as classification: clustering based on neighbourhood or connectivity?



CLUSTERING EXAMPLE

- Human resources department would like to cluster employees in order to understand the different types of employee and treat them accordingly (fire problematic workers? 😊).

CLUSTERING EXAMPLE.

TRAINING DATA

Id	Salary	Married	Car	Offspring	Own-house	Syndicate	Sick leave	Years working	Sex
1	1000	Yes	No	0	No	No	7	15	M
2	2000	No	Yes	1	No	Yes	3	3	F
3	1500	Yes	Yes	2	Yes	Yes	5	10	M
4	3000	Yes	Yes	1	No	No	15	7	F
5	1000	Yes	Yes	0	Yes	Yes	1	6	F
...

MODEL (CLUSTERS)

	GROUP 1	GROUP 2	GROUP 3
Salary	1535	1428	1233
Married (No/Yes)	77%/22%	98%/2%	0%/100%
Car	82%/18%	1%/99%	5%/95%
Offspring	0.05	0.3	2.3
Own-house	99%/1%	75%/25%	17%/83%
Syndicated	80%/20%	0%/100%	67%/33%
Sick leave	8.3	2.3	5.1
Years working	8.7	8	8.1
Sex (M/W)	61%/39%	25%/75%	83%/17%

MODEL (CLUSTERS)

- Cluster 1: No offspring and rented house. Low level of syndication. Lots of sick leaves
- Cluster 2: No offspring and own-car. High syndication level. Few sick leaves. Typically women living in rented houses
- Cluster 3: Married men with children and own-car and own-houses. Low syndication level

TASKS

- **Inductive learning**(from instances)
 - Supervised learning:
 - Classification
 - Regression
 - Semi-supervised learning
 - **Unsupervised learning:**
 - Clustering
 - **Association**
 - Reinforcement learning

MARKET BASKET ANALYSIS (ASSOCIATION)

- A supermarket needs to know customer behavior.
 - Ex: if customer buys X then s/he also buys Y
- Service might be improved (putting together products bought together, etc.)

TRAINING DATA (CUSTOMER BASKETS)

Id	Eggs	Oil	Napies	Wine	Milk	Butter	Salmon	Lettuce	...
1	Yes	No	No	Yes	No	Yes	Yes	Yes	...
2	No	Yes	No	No	Yes	No	No	Yes	...
3	No	No	Yes	No	Yes	No	No	No	...
4	No	Yes	Yes	No	Yes	No	No	No	...
5	Yes	Yes	No	No	No	Yes	No	Yes	...
6	Yes	No	No	Yes	Yes	Yes	Yes	No	...
7	No	No	No	No	No	No	No	No	...
8	Yes	Yes	Yes	Yes	Yes	Yes	Yes	No	...
...

MODEL

- Rules **IF** $At_1=a$ **AND** $At_2=b$ y ... **THEN** $At_n=c$
 - **IF** nappies=Yes **THEN** milk=Yes
 - **IF** butter = Yes **AND** salmon = Yes **THEN** wine = Yes
- Also: **IF** $At_1=a$ **AND** $At_2=b$ **THEN** $At_n=c$, **$At_4=D$**

Service might be improved (putting together nappies and milk, etc.)

ASSOCIATION

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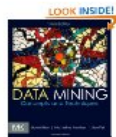
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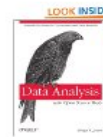
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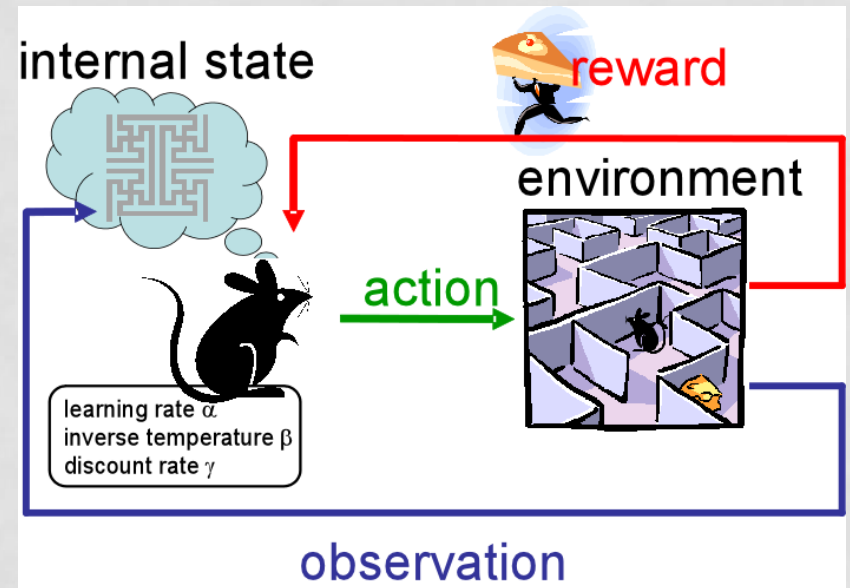
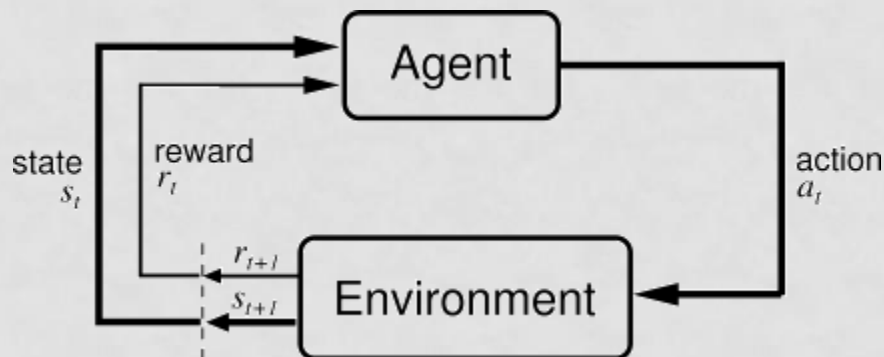
Editorial Reviews

TASKS

- **Inductive learning**(from instances)
 - Supervised learning:
 - Classification
 - Regression
 - Semi-supervised learning
 - Unsupervised learning:
 - Clustering
 - Association
 - **Reinforcement learning**

TASK: REINFORCEMENT LEARNING

- The goal of learning is a “policy” π so that the agent (mouse) knows what to do at each situation (in the case of the mouse, a situation is a particular location within the maze). Robotics.
- Actions:
 - forward
 - turn left
 - turn right



TASKS

- **Inductive learning**(from instances)
 - Attribute-value models
 - Supervised learning:
 - Semi-supervised learning
 - Unsupervised learning:
 - Reinforcement learning
 - **Relational learning**

Relational Learning

- For instance, learn the concept of “being a daughter”
- **IF** X is female **AND** Y is the mother of X **THEN** X is a daughter of Y
- Compare this rule with:
IF Overdue Accounts ==0 **AND** Salary >2500 **THEN** Repay loan = Yes
- Relational rules use variables (X, Y) and relations

Relational Learning: ILP (inductive logic programming)

<i>Training examples</i>		<i>Background knowledge</i>	
<i>daughter(mary, ann).</i>	\oplus	<i>parent(ann, mary).</i>	<i>female(ann).</i>
<i>daughter(eve, tom).</i>	\oplus	<i>parent(ann, tom).</i>	<i>female(mary).</i>
<i>daughter(tom, ann).</i>	\ominus	<i>parent(tom, eve).</i>	<i>female(eve).</i>
<i>daughter(eve, ann).</i>	\ominus	<i>parent(tom, ian).</i>	

Learned Knowledge:

daughter(X, Y) \leftarrow female(X), mother(Y, X).

daughter(X, Y) \leftarrow female(X), father(Y, X).

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