

## **Chapter VI: Classification**

## Knowledge Discovery in Databases

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## **Chapter VI: Classification**

#### Classification: basic concepts.

Decision-tree induction.

Bayes classification methods.

Rule-based classification.

Model evaluation and selection.

Techniques to improve classification accuracy: ensemble methods.

Summary.



## Supervised vs. unsupervised learning

#### Supervised learning (classification).

## Supervision:

The **training data** (observations, measurements, etc.) are accompanied by **labels** indicating the **class** of the observations.

New data is classified based on a **model** created from the training data.

#### **Unsupervised learning (clustering).**

The class labels of training data are unknown.

Or rather, there are no training data.

Given a set of measurements, observations, etc., the goal is to find classes or clusters in the data.

See next chapter.



## Prediction problems: classification vs. numerical prediction

#### Classification:

Predicts categorical class labels (discrete, nominal).

Constructs a model based on the training set and the values (class labels) in a classifying attribute and uses it in classifying new data.

#### **Numerical prediction:**

Models continuous-valued functions.

I.e. predicts missing or unknown (future) values.

#### Typical applications of classification:

Credit/loan approval: Will it be paid back?

Medical diagnosis: Is a tumor cancerous or benign? Fraud detection: Is a transaction fraudulent or not? Web-page categorization: Which category is it?



## Classification – a two-step process

#### Model construction: describing a set of predetermined classes:

Each tuple/sample is assumed to belong to a predefined class, as determined by the class-label attribute.

The set of tuples used for model construction is the **training set**.

The model is represented as classification rules, decision trees, or mathematical formulae.

#### Model usage, for classifying future or unknown objects:

Estimate accuracy of the model:

The known label of **test samples** is compared with the result from the model.

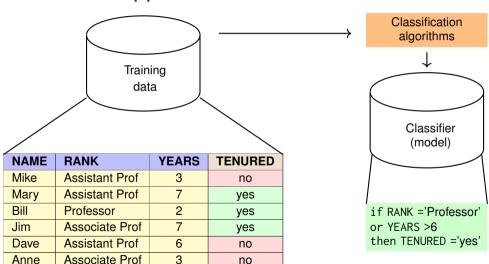
**Accuracy rate** is the percentage of test-set samples that are correctly classified by the model.

Test set is independent of training set (otherwise overfitting).

If the accuracy is acceptable, use the model to classify data tuples whose class labels are not known.

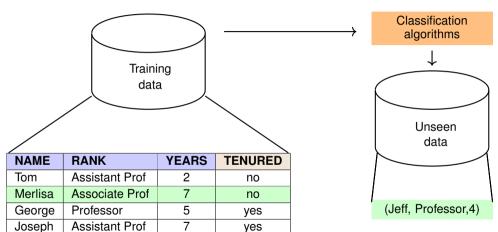


## Classification – a two-step process





## Process (II): using the model in prediction





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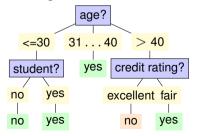


## Decision-tree induction: an example

#### Training dataset: buys computer.

The dataset follows an example of Quinlan's ID3 (playing tennis).

#### Resulting tree:



| age   | income | student | credit_rating | buys_coputer |
|-------|--------|---------|---------------|--------------|
| <= 30 | high   | no      | fair          | no           |
| <= 30 | high   | no      | excellent     | no           |
| 31 40 | high   | no      | fair          | yes          |
| > 40  | medium | no      | fair          | yes          |
| > 40  | low    | yes     | fair          | yes          |
| > 40  | low    | yes     | excellent     | no           |
| 31 40 | low    | yes     | excellent     | yes          |
| <= 30 | medium | no      | fair          | no           |
| <= 30 | low    | no      | fair          | yes          |
| > 40  | medium | yes     | fair          | yes          |
| <= 30 | medium | yes     | excellent     | yes          |
| 31 40 | medium | no      | excellent     | yes          |
| 31 40 | high   | yes     | fair          | yes          |
| > 40  | medium | no      | excellent     | no           |



## Algorithm for decision-tree induction

#### Basic algorithm (a greedy algorithm):

Tree is constructed in a top-down recursive divide-and-conquer manner.

Attributes are categorical.

If not: discretize in advance.

At start, all the training examples are at the root.

Examples are partitioned recursively based on selected attributes.

Test attributes are selected on the basis of a heuristic or statistical measure.

E.g. information gain – see on the next slide.

#### Conditions for stopping partitioning:

All samples for a given node belong to the same class.

There are no remaining attributes for further partitioning.

Majority voting is employed for classifying the leaf.

There are no samples left (i.e. partition for particular value is empty).



# Thank you for your attention. Any questions about the sixt chapter?

Ask them now, or again, drop me a line: \$\frac{1}{4}\$ luciano.melodia@fau.de.