

COURSE NAME:  
**DATA WAREHOUSING & DATA MINING**

---

# LECTURE 16

## TOPICS TO BE COVERED:

---

- ✘ What is classification?
- ✘ What is prediction?
- ✘ Issues regarding classification and prediction

# CLASSIFICATION & PREDICTION

- ✘ **Databases are rich with hidden information that can be used for intelligent decision making.**
- ✘ Classification and prediction are two forms of data analysis that can be used to extract models describing important data classes or to predict future data trends.

# CLASSIFICATION VS. PREDICTION

---

## × Classification:

- + predicts categorical class labels
- + classifies data (constructs a model) based on the training set and the values (**class labels**) in a classifying attribute and uses it in classifying new data

## × Prediction:

- + models continuous-valued functions, i.e., predicts unknown or missing values

# APPLICATIONS

---

## × Typical Applications

- + Credit Approval
- + Target Marketing
- + Medical Diagnosis
- + Treatment Effectiveness Analysis
- + Performance Prediction

# EXAMPLE

---

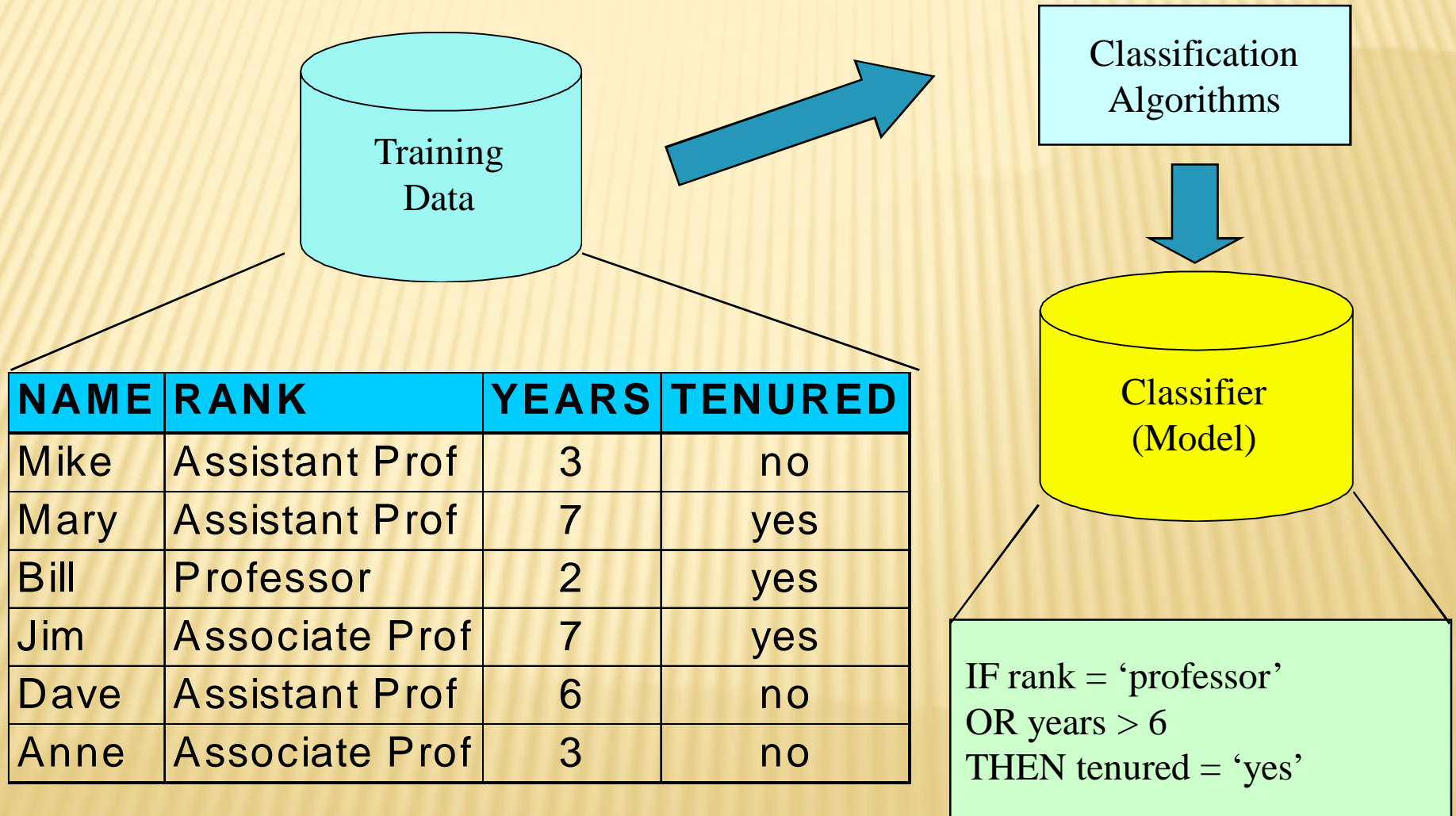
- ✘ **A bank loans officer needs analysis of her data in order to learn which loan applicants are “safe” and which are “risky” for the bank**
- ✘ A marketing manager at *AllElectronics* needs data analysis to help guess whether a customer with a given profile will buy a new computer.
- ✘ A medical researcher wants to analyze breast cancer data in order to predict which one of three specific treatments a patient should receive.
- ✘ In each of these examples, the data analysis task is classification, where a model or classifier is constructed to predict *categorical labels*, such as “safe” or “risky” for the loan application data; “yes” or “no” for the marketing data; or “treatment A,” “treatment B,” or “treatment C” for the medical data. These categories can be represented by discrete values, where the ordering among values has no meaning.
- ✘ For example, the values 1, 2, and 3 may be used to represent treatments A, B, and C, where there is no ordering implied among this group of treatment regimes.

# CLASSIFICATION—A TWO-STEP PROCESS

---

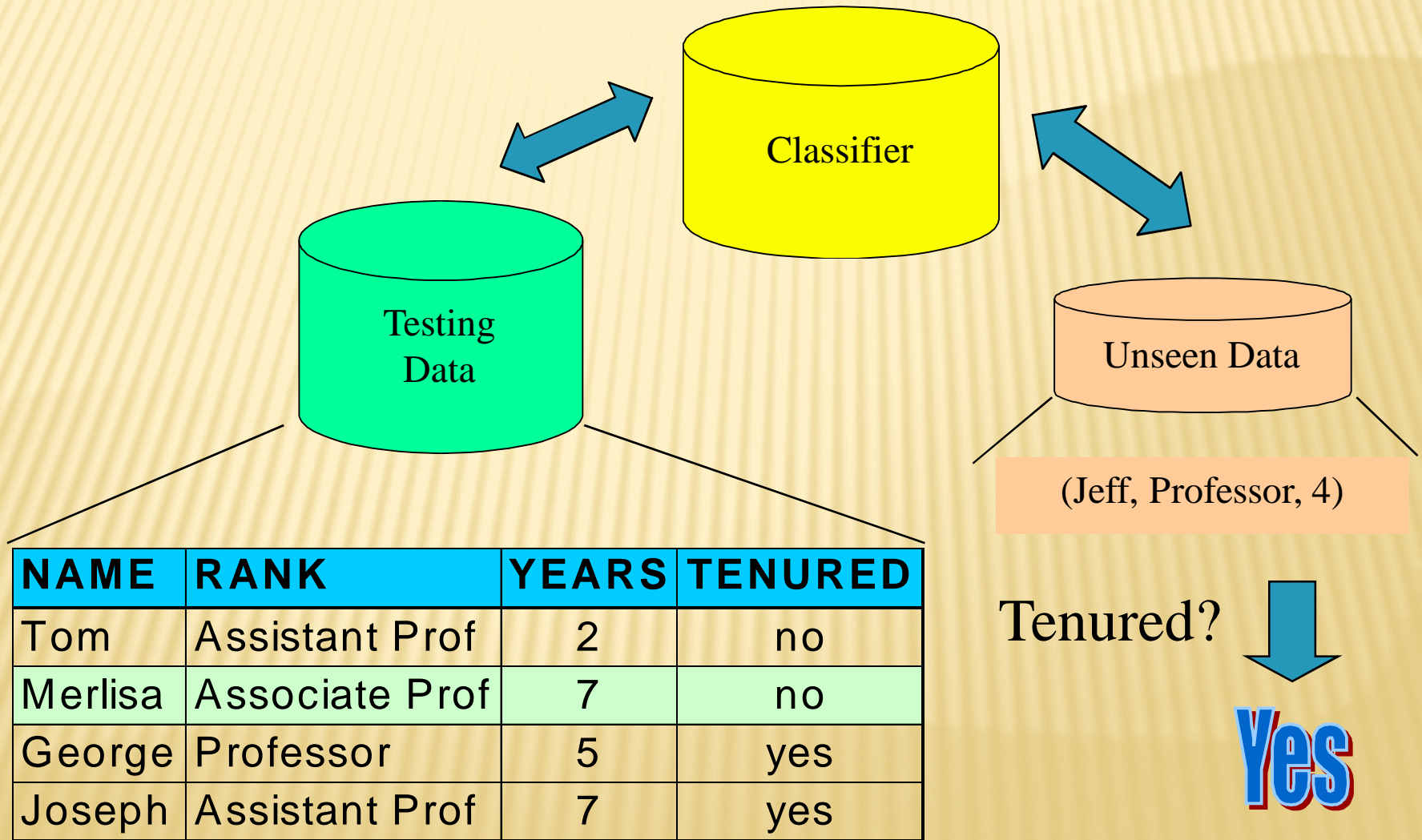
- ✘ Model construction: describing a set of predetermined classes
  - + Each tuple/sample is assumed to belong to a predefined class, as determined by the class label attribute
  - + The set of tuples used for model construction: training set
  - + The model is represented as classification rules, decision trees, or mathematical formulae
- ✘ Model usage: for classifying future or unknown objects
  - + Estimate accuracy of the model
    - ✘ The known label of test sample is compared with the classified 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 over-fitting will occur

# CLASSIFICATION PROCESS (1): MODEL CONSTRUCTION





# CLASSIFICATION PROCESS (2): USE THE MODEL IN PREDICTION



# 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 the training set

## × Unsupervised learning (clustering)

- + The class labels of training data is unknown
- + Given a set of measurements, observations, etc. with the aim of establishing the existence of classes or clusters in the data

# WHAT IS PREDICTION?

---

- ✘ Prediction is similar to classification
  - + First, construct a model
  - + Second, use model to predict unknown value
    - ✘ Major method for prediction is regression
      - ✘ Linear and multiple regression
      - ✘ Non-linear regression
- ✘ Prediction is different from classification
  - + Classification refers to predict categorical class label
  - + Prediction models continuous-valued functions

# PREDICTIVE MODELING IN DATABASES

---

- ✘ Predictive modeling: Predict data values or construct generalized linear models based on the database data.
- ✘ One can only predict value ranges or category distributions
- ✘ Method outline:
  - + Minimal generalization
  - + Attribute relevance analysis
  - + Generalized linear model construction
  - + Prediction
- ✘ Determine the major factors which influence the prediction
  - + Data relevance analysis: uncertainty measurement, entropy analysis, expert judgement, etc.
- ✘ Multi-level prediction: drill-down and roll-up analysis

# REGRESS ANALYSIS AND LOG-LINEAR MODELS IN PREDICTION

---

- × Linear regression:  $Y = \alpha + \beta X$ 
  - + Two parameters ,  $\alpha$  and  $\beta$  specify the line and are to be estimated by using the data at hand.
  - + using the least squares criterion to the known values of  $Y_1, Y_2, \dots, X_1, X_2, \dots$
- × Multiple regression:  $Y = b_0 + b_1 X_1 + b_2 X_2$ .
  - + Many nonlinear functions can be transformed into the above.

---

# ISSUES REGARDING CLASSIFICATION AND PREDICTION

# ISSUES (1): DATA PREPARATION

---

## ✘ Data cleaning

- + Preprocess data in order to reduce noise and handle missing values

## ✘ Relevance analysis

- + Remove the irrelevant or redundant attributes.
- + Correlation analysis can be used to identify whether any two given attributes are statistically related.
- + Attribute subset selection can be used in these cases to find a reduced set of attributes such that resulting probability distribution of the data classes is as close as possible to the original distribution obtained using all attributes.

# ISSUES (1): DATA PREPARATION

- + Hence, relevance analysis, in the form of correlation analysis and attribute subset selection, can be used to detect attributes that do not contribute to the classification or prediction task
- × Data transformation
  - + Generalize and/or normalize data
  - + Normalization involves scaling all values for a given attribute so that they fall within a small specified range, such as -1:0 to 1:0, or 0:0 to 1:0. In methods that use distance measurements.
  - + The data can also be transformed by *generalizing it to higher-level concepts*. This is particularly useful for continuous valued attributes. For example, numeric values for the attribute *income* can be *generalized* to discrete ranges, such as *low, medium, and high*.



# COMPARING CLASSIFICATION AND PREDICTION METHODS

---

Classification & Prediction methods can be compared and Evaluated according following criteria:

- ✘ Predictive accuracy
  - + The accuracy of a classifier refers to the ability of a given classifier to correctly predict the class label of new or previously unseen data (i.e., tuples without class label information).
  - + the accuracy of a predictor refers to how well a given predictor can guess the value of the predicted attribute for new or previously unseen data.
- ✘ Speed and scalability
  - + time to construct the model
  - + time to use the model

# COMPARING CLASSIFICATION AND PREDICTION METHODS

---

- × Robustness
  - + handling noise and missing values
- × Scalability
  - + efficiency in disk-resident databases
- × Interpretability:
  - + understanding and insight provided by the model
- × Goodness of rules
  - + decision tree size
  - + compactness of classification rules