

Classification

Classification

- Given a collection of records (*training set*)
 - Each record contains a set of *attributes*, one of the attributes is the *class*.
- Find a *model* for class attribute as a function of the values of other attributes.
- Goal: previously unseen records should be assigned a class as accurately as possible.
 - A *test set* is used to determine the accuracy of the model.
Usually, the given data set is divided into training and test sets, with training set used to build the model and test set used to validate it.
- Classification models predict categorical class labels.

Naive Bayes Classification

- Let D be a training set of tuples and their associated class labels.
- Tuple is represented by an n -dimensional attribute vector, $X = (x_1, x_2, \dots, x_n)$,
- Measurements made on the tuple from n attributes, respectively, A_1, A_2, \dots, A_n

Naive Bayes Classification

- Naïve Bayesian classifier predicts that tuple X belongs to the class C_i , if and only if

$$P(C_i|X) > P(C_j|X) \quad \text{for } 1 \leq j \leq m, j \neq i.$$

Bayes' theorem

$$P(C_i|X) = \frac{P(X|C_i)P(C_i)}{P(X)}.$$

$$\begin{aligned} P(X|C_i) &= \prod_{k=1}^n P(x_k|C_i) \\ &= P(x_1|C_i) \times P(x_2|C_i) \times \cdots \times P(x_n|C_i). \end{aligned}$$

Naive Bayes - Example

| <i>RID</i> | <i>age</i> | <i>income</i> | <i>student</i> | <i>credit_rating</i> | <i>Class: buys_computer</i> |
|------------|-------------|---------------|----------------|----------------------|-----------------------------|
| 1 | youth | high | no | fair | no |
| 2 | youth | high | no | excellent | no |
| 3 | middle_aged | high | no | fair | yes |
| 4 | senior | medium | no | fair | yes |
| 5 | senior | low | yes | fair | yes |
| 6 | senior | low | yes | excellent | no |
| 7 | middle_aged | low | yes | excellent | yes |
| 8 | youth | medium | no | fair | no |
| 9 | youth | low | yes | fair | yes |
| 10 | senior | medium | yes | fair | yes |
| 11 | youth | medium | yes | excellent | yes |
| 12 | middle_aged | medium | no | excellent | yes |
| 13 | middle_aged | high | yes | fair | yes |
| 14 | senior | medium | no | excellent | no |

- Classify the following x ,

$X = (age = youth, income = medium, student = yes, credit_rating = fair)$

- Prior probability

$$P(\text{buys computer} = \text{yes}) = 9/14 = 0.643$$

$$P(\text{buys computer} = \text{no}) = 5/14 = 0.357$$

Conditional Probability

- $P(\text{age} = \text{youth} \mid \text{buys computer} = \text{yes}) = 2/9 = 0.222$
- $P(\text{age} = \text{youth} \mid \text{buys computer} = \text{no}) = 3/5 = 0.600$
- $P(\text{income} = \text{medium} \mid \text{buys computer} = \text{yes}) = 4/9 = 0.444$
- $P(\text{income} = \text{medium} \mid \text{buys computer} = \text{no}) = 2/5 = 0.400$
- $P(\text{student} = \text{yes} \mid \text{buys computer} = \text{yes}) = 6/9 = 0.667$
- $P(\text{student} = \text{yes} \mid \text{buys computer} = \text{no}) = 1/5 = 0.200$
- $P(\text{credit rating} = \text{fair} \mid \text{buys computer} = \text{yes}) = 6/9 = 0.667$
- $P(\text{credit rating} = \text{fair} \mid \text{buys computer} = \text{no}) = 2/5 = 0.400$

- $P(X|\text{buys computer} = \text{yes}) =$
 $P(\text{age} = \text{youth} \mid \text{buys computer} = \text{yes}) \times$
 $P(\text{income} = \text{medium} \mid \text{buys computer} = \text{yes}) \times$
 $P(\text{student} = \text{yes} \mid \text{buys computer} = \text{yes}) \times$
 $P(\text{credit rating} = \text{fair} \mid \text{buys computer} = \text{yes})$

 $= 0.222 \times 0.444 \times 0.667 \times 0.667 = 0.044.$

- $P(X|\text{buys computer} = \text{no}) = 0.600 \times 0.400 \times 0.200 \times 0.400 = 0.019.$
- To find the class, C_i , that maximizes $P(X|C_i)P(C_i)$, Compute
- $P(X|\text{buys computer} = \text{yes})P(\text{buys computer} = \text{yes}) = 0.044 \times 0.643$
 $= 0.028$
- $P(X|\text{buys computer} = \text{no})P(\text{buys computer} = \text{no}) = 0.019 \times 0.357$
 $= 0.007$

Naïve Bayesian classifier predicts buys computer = yes for tuple X.

Practice Question???

play tennis?

Naive Bayesian Classifier Example

| Outlook | Temperature | Humidity | Windy | Class |
|----------|-------------|----------|-------|-------|
| sunny | hot | high | false | N |
| sunny | hot | high | true | N |
| overcast | hot | high | false | P |
| rain | mild | high | false | P |
| rain | cool | normal | false | P |
| rain | cool | normal | true | N |
| overcast | cool | normal | true | P |
| sunny | mild | high | false | N |
| sunny | cool | normal | false | P |
| rain | mild | normal | false | P |
| sunny | mild | normal | true | P |
| overcast | mild | high | true | P |
| overcast | hot | normal | false | P |
| rain | mild | high | true | N |

Classify a new sample X

X =

- „outlook = sunny
- „temperature = cool
- „humidity = high
- „windy = false
- Play Tennis = ?

Reference:

Data Mining: Concepts and Techniques , Jiawei Han and
Micheline Kamber