

# Pattern recognition - 9th lab

## Naive Bayes

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# Bayes equation

## Bayes equation

We will use the Bayes equation:

$$P(\omega_i|\vec{x}) = \frac{P(\vec{x}|\omega_i)P(\omega_i)}{P(\vec{x})} \quad (1)$$

## Naive Bayes

Our classifier is naive and assumes that features are independent:

$$P(\vec{x}|\omega_i) = \prod_k P(x_k|\omega_i) \quad (2)$$

# Classifier

## Classifier

We classify by finding the class with the highest probability:

$$pred_i = \arg \max_i \left( \frac{P(\vec{x}|\omega_i)P(\omega_i)}{P(\vec{x})} \right) \quad (3)$$

$$= \arg \max_i (P(\vec{x}|\omega_i)P(\omega_i)) \quad (4)$$

$$= \arg \max_i \left( P(\omega_i) \prod_k P(x_k|\omega_i) \right) \quad (5)$$

# Classifier

## Calculating values

We assume categorical features. Therefore for each  $k$  the  $x_k$  can only have finitely many values. Let us have a training set of size  $N$ . We denote the amount of elements in the category  $\omega_i$  as  $N_i$  and the amount of elements which are in the category  $\omega_i$  for the  $k$ -th feature which has value of  $v$  as  $N_{i,k,v}$ . We can then define:

$$P(\omega_i) = \frac{N_i}{N} \quad (6)$$

$$P(x_k = v | \omega_i) = \frac{N_{i,k,v}}{N_i} \quad (7)$$

# Classifier

age	income	student	credit rating	buys computer
<=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	yes	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

## Exercise

Calculate which class would a client with random features belong to according to naive bayes.

# Classifier

## Numerical features

In case that the feature is numeric we cannot apply the calculation from the previous slide. We will therefore need to estimate the probability  $P(x_k|\omega_i)$  by some distribution function.

## Parametric methods

If we use a parametric method we select a distribution and fit its parameters to the data.

## Non-parametric methods

We can also create a distribution function which is calculated based on the points in the training set in the neighborhood of the point we are interested in.

# Matlab

## fitcnb

`Mdl = fitcnb(T,'column_name')` - returns a naive Bayes classifier for the table `T` and classification target in the column `column_name`.

## Malab - Table type

Working with tables:

<https://www.mathworks.com/help/matlab/tables.html>

This is the most important part for us:

[https://www.mathworks.com/help/matlab/matlab\\_prog/access-data-in-a-table.html](https://www.mathworks.com/help/matlab/matlab_prog/access-data-in-a-table.html)

# Naive Bayes on table data

## Na dátach

```
load census1994  
Mdl = fitcnb(adulddata, 'salary');
```

## Exercise

Determine the accuracy of the classifier by using `Mdl.predict` on the table `adulttest` and compare the results.



# Matlab

## fitcnb

$Mdl = \text{fitcnb}(X,y)$  - returns naive Bayes classifier for non-table data.

## Exercise

Test the naive Bayes classifier on the fisheriris database.

## Exercise

Use the data from the 6th lab and display the classification boundary using your modified version of showSVM from the same lab.