

Naive Bayes Classifier

1. Bayes' law

Among all patients in the haematology section 1% is infected with the HIV virus, 97% of which test positive. Among non-infected patients this is 4%. Calculate the probability that a patient who is tested HIV positive really is infected with the HIV virus.

C_1 = "HIV+ patients"

C_2 = "HIV- patients"

X = "test result"

$$P(C_1) = 0.01 \quad P(X = + \mid C_1) = 0.97$$

$$P(C_2) = 0.99 \quad P(X = + \mid C_2) = 0.04$$

$$P(C_1 \mid X = +) =$$

$$\frac{P(C_1) \cdot P(X = + \mid C_1)}{P(C_1) \cdot P(X = + \mid C_1) + P(C_2) \cdot P(X = + \mid C_2)}$$

Terminology

- $P(C_1)$, $P(C_2)$: prior probability
- $P(C_1 \mid X = +)$, $P(C_2 \mid X = +)$: posterior probability

In general

- n classes: C_1, C_2, \dots, C_n
- k predictor variables X_1, X_2, \dots, X_k

Question

If $X_1 = x_1, X_2 = x_2, \dots, X_k = x_k$, than what is the most probable class?

$$\begin{aligned} & P(C_i \mid X_1 = x_1, X_2 = x_2, \dots, X_k = x_k) ? \\ = & P(C_i \mid x_1, x_2, \dots, x_k) ? \end{aligned}$$

Bayes' law:

Wet van Bayes:

$$\begin{aligned} & P(C_i \mid x_1, x_2, \dots, x_k) \\ = & \frac{P(C_i) \cdot P(x_1, x_2, \dots, x_k \mid C_i)}{P(x_1, x_2, \dots, x_k)} \end{aligned}$$

Example

- Titanic dataset
- 4 variables, 2201 observations
- X1: class
('0' = personnel, '1' = most expensive class, '2' = middle class, '3' = cheapest class)
- X2: age
('1' = adult, '0' = child)
- X3: gender
('1' = male, '0' = female)
- C: rescued or not
('1' = rescued, '0' = not rescued)

Example question: Will a girl in the cheapest class probably be rescued or not?

$P(\text{rescued} \mid \text{cheapest, child, female}) = ?$

$P(\text{not rescued} \mid \text{cheapest, child, female}) = ?$

$$\begin{aligned}
 &P(\text{rescued} \mid \text{cheapest, child, female}) \\
 &= \frac{P(\text{rescued}) \cdot P(\text{cheapest, child, female} \mid \text{rescued})}{P(\text{cheapest, child, female})}
 \end{aligned}$$

$$\begin{aligned}
 &P(\text{not rescued} \mid \text{cheapest, child, female}) \\
 &= \frac{P(\text{not rescued}) \cdot P(\text{cheapest, child, female} \mid \text{not rescued})}{P(\text{cheapest, child, female})}
 \end{aligned}$$

Remarks:

- denominator $P(\text{cheapest, child, female})$ no need to compare.
- To answer all questions, the following needs to be known:

$P(\text{rescued})$ and $P(\text{not rescued})$
 $P(x_1, x_2, x_3 \mid \text{rescued})$ and
 $P(x_1, x_2, x_3 \mid \text{not rescued})$
 for all combinations of x_1, x_2, x_3
 $\rightarrow 2 \cdot 4 \cdot 2 \cdot 2 = 32$ possibilities
 Estimate probabilities via frequencies in training set

Problem:

- large amount of combinations x_1, x_2, x_3
- certain combinations may not occur in a class C_i in the training set

→ estimate: $P(x_1, x_2, x_3 \mid C_i) = 0$.

→ $P(C_i \mid x_1, x_2, x_3) = 0$

2. Solution: Naive Bayes

- Assumption: X_1, X_2, X_3 not interdependent in every class C_i

$$\begin{aligned} \rightarrow P(x_1; x_2; x_3 \mid C_i) \\ = P(x_1 \mid C_i) \cdot P(x_2 \mid C_i) \cdot P(x_3 \mid C_i) \end{aligned}$$

- $P(x_1 \mid C_i), P(x_2 \mid C_i), P(x_3 \mid C_i)$
Estimate via frequencies in training set
- less quickly zero; fewer chances: $2 \cdot (4+2+2) = 16$