

Quiz 1 - sample

Note Title

9/3/2007

1. For the following "widget" dataset, compute

a) decision rule

b) error rate on training set

using the IR algorithm.

2. For the same dataset, but using the naive Bayes algorithm, compute

a) decision rule only for "normal, loud" widgets

b) what is the probability that a "normal, loud" widget is "bad"?

c) clearly identify the prior probabilities, likelihoods, and posterior probabilities

you used for (a) & (b).

d) what specific independence assumption did you use to answer (a) & (b)?

Data set for widgets :

appearance	sound	class
normal	quiet	good
normal	quiet	good
deformed	quiet	bad
deformed	loud	bad
normal	quiet	bad
deformed	quiet	good
normal	quiet	good
normal	quiet	good
normal	loud	bad
normal	loud	bad
normal	loud	bad
deformed	quiet	bad
normal	loud	good
normal	quiet	good
normal	quiet	good
normal	quiet	good

Solution:

1. IR for appearance:

<u>appearance</u>	<u>good</u>	<u>bad</u>	<u>decision</u>	<u>errors</u>
normal	8	4	good	4/12
deformed	1	3	bad	1/4
				total errors: 5/16

IR for sound:

<u>sound</u>	<u>good</u>	<u>bad</u>	<u>decision</u>	<u>errors</u>
quiet	8	3	good	3/11
loud	1	4	bad	1/5
				total errors: 4/16

- a) decision rule: attribute 'sound' has lowest error rate, so choose that.
Rule is: quiet \rightarrow good
loud \rightarrow bad

- b) error rate on training set: 4/16 (ie. 1/4).

2. Need to compute $P(A|N,L)$ and $P(B|N,L)$ using Bayes' Rule; assuming independence of features:

$$\begin{aligned} P(A|N,L) &= P(N,L|A) \times P(A) / P(N,L) \\ &= P(N|A) \times P(L|A) \times P(A) / P(N,L) \end{aligned}$$

$$\begin{aligned} P(B|N,L) &= P(N,L|B) \times P(B) / P(N,L) \\ &= P(N|B) \times P(L|B) \times P(B) / P(N,L) \end{aligned}$$

summary statistics we will need :

instances :	16
good :	9
bad :	7
normal + good :	8
normal + bad :	4
loud + good :	1
loud + bad :	4

probability estimates :

$$P(A) = 9/16$$
$$P(B) = 7/16$$
$$P(N|A) = 8/9$$
$$P(N|B) = 4/7$$
$$P(L|A) = 1/9$$
$$P(L|B) = 4/7$$

plugging in:

$$\begin{aligned} P(A|N,L) &= \frac{8}{9} \times \frac{1}{9} \times \frac{9}{16} \bigg/ P(N,L) \\ &= \frac{1}{18} \bigg/ P(N,L) = 0.055 \bigg/ P(N,L) \end{aligned}$$

$$\begin{aligned} P(B|N,L) &= \frac{4}{7} \times \frac{4}{7} \times \frac{7}{16} \bigg/ P(N,L) \\ &= \frac{1}{28} \bigg/ P(N,L) = 0.036 \bigg/ P(N,L) \end{aligned}$$

These 2 probabilities must sum to 1, so we know immediately that $P(N,L) = \frac{1}{18} + \frac{1}{28} = 0.091$

So,

$$P(A|N,L) = \frac{0.055}{0.091} = 0.61$$

$$P(B|N,L) = \frac{0.036}{0.091} = 0.39$$

So 'Good' has higher posterior probability for normal, loud widgets

i.e. decision rule is

noisy, loud \rightarrow good.

b) prob that normal, low budget is bad = $P(B|N,L) = 0.39$

c) Prior probs - $P(A), P(B)$

Likelihoods - $P(N|A), P(N|B), P(L|A), P(L|B)$

Posteriors - $P(A|N,L), P(B|N,L)$

d) 2 independence assumptions:

$$P(N,L|A) = P(N|A) P(L|A)$$

$$P(N,L|B) = P(N|B) P(L|B)$$