

# P03 Probability Distributions and Bayesian Classification

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## 1 Probability and Distributions

1) Consider the following registry where an experiment is repeated six times and four events (A, B, C and D) are detected. Considering frequentist estimates for probabilities, compute:

	D	C	B	A
1	0	0	1	1
2	0	1	1	1
3	1	0	0	0
4	1	0	0	0
5	0	0	0	0
6	0	0	0	0

- $p(A)$
- $p(A, B)$
- $p(B | A)$
- $p(A, B, C)$
- $p(A | B, C)$
- $p(A, B, C, D)$
- $p(D | A, B, C)$

2) Consider the following set of height measures in centimeters of a group of people:

$$X | 180 \ 160 \ 200 \ 171 \ 159 \ 150$$

What are the maximum likelihood parameters of a gaussian distribution for this set of points? Plot it approximately.

3) Consider the following set of two dimensional measures:

$$\begin{array}{l|l} X_1 & -2 \ -1 \ 0 \ -2 \\ X_2 & 2 \ \ 3 \ 1 \ 1 \end{array}$$

What are the maximum likelihood parameters of a Gaussian distribution for this set of points? What is the shape of the Gaussian? Draw it approximately using a contour map.

4) Consider the following set of two dimensional measures:

$$\begin{array}{c|cccc} X_1 & 2 & 1 & 0 & 2 \\ X_2 & -2 & 3 & -1 & 1 \end{array}$$

What are the maximum likelihood parameters of a Gaussian distribution for this set of points? What is the shape of the Gaussian? Draw it approximately using a contour map.

## 2 Simple Bayesian Learning

1) Assuming that 1 means *True* and 0 means *False*, consider the following features and class:

- $X_1$ : “Fast processing”
- $X_2$ : “Decent Battery”
- $X_3$ : “Good Camera”
- $X_4$ : “Good Look and Feel”
- $X_5$ : “Easiness of Use”
- *Class*: “iPhone”

You are given the following training set:

$X_1$	$X_2$	$X_3$	$X_4$	$X_5$	<i>Class</i>
1	1	0	1	0	1
1	1	1	0	0	0
0	1	1	1	0	0
0	0	0	1	1	0
1	0	1	1	1	1
0	0	1	0	0	1
0	0	0	0	1	1

And the query vector  $\mathbf{x} = [1 \ 1 \ 1 \ 1 \ 1]^T$ .

- a) Using Bayes’ rule, without making any assumptions, compute the class for the query vector.
- b) What is the problem of working with this data set without assumptions?
- c) Compute the class for the same query vector under the Naive Bayes assumption.

2) From the following training set:

$X_1$	$X_2$	$X_3$	$X_4$	$X_5$	<i>C</i>
1	1	0	1	0	<i>a</i>
1	0	0	1	1	<i>a</i>
1	0	0	1	1	<i>a</i>
1	1	1	0	1	<i>b</i>
0	0	1	1	1	<i>b</i>
1	0	0	0	0	<i>c</i>

- a) Compute the class for the pattern  $\mathbf{x} = [1 \ 0 \ 1 \ 0 \ 1]^T$  under the Naive Bayes assumption.
- b) What is the posterior probability  $p(b \mid \mathbf{x})$ ?
- c) What do you do if we have missing features? More specifically, under the Naive Bayes assumption, to what class does  $\mathbf{x}_{\text{missing}} = [1 \ ? \ 1 \ ? \ 1]^T$  belong to?

**3)** So far we have been dealing always with discrete feature domains. In this exercise, we will work with continuous values for features like Height and Weight. Assuming that 1 means *True* and 0 means *False*, consider the following features and class:

- $X_1$ : “Weight (Kg)”
- $X_2$ : “Height (Cm)”
- $Class$ : “NBA Player”

You are given the following training set:

$X_1$	$X_2$	$Class$
170	160	0
80	220	1
90	200	1
60	160	0
50	150	0
70	190	1

And the query vector  $\mathbf{x} = [100 \ 225]^T$ .

- a) Compute the most probable class for the query vector assuming that the likelihoods are 2-dimensional Gaussians.
- b) Compute the most probable class for the query vector, under the Naive Bayes assumption, using 1-dimensional Gaussians to model the likelihoods.

### 3 Thinking Questions

- a) Assuming training examples with  $d$  boolean features, how many parameters do you have to estimate if you make no assumptions about how the data is distributed? What about if you make the Naive Bayes assumption?
- b) Is Naive Bayes a linear classifier?