

MACHINE LEARNING

EXERCISES

Elements of non-parametric techniques

All the course material is available on the web site

Course web site: <http://pralab.diee.unica.it/MachineLearning>

➤ **And here: <https://github.com/unica-ml/ml/tree/master/exercises>**

Exercise 1

Given the following patterns belonging to three different classes A, B, and C

A	1.1	1.7	1.2	1.6
	1.3	1.4	2.0	1.9
B	2.7	2.6	2.2	2.2
	1.4	1.2	2.0	1.3
C	1.4	1.2	1.8	1.5
	2.5	2.4	2.6	2.9

We want to classify the unknown pattern:

$$x_t = (2; 2)'$$

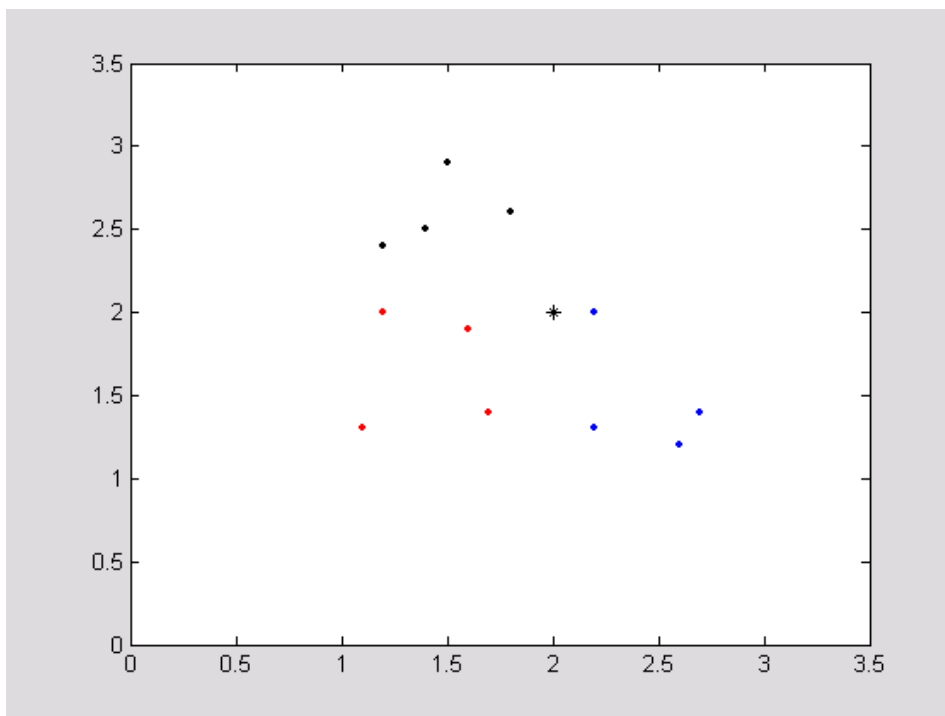
but we do not know from which probability distribution the pattern has been generated. Then, we can use a non-parametric method like the k -nn pattern classifier.

A) Classify the pattern x_t with values of $k=1, \dots, 4$ using the *Euclidean* and the *Manhattan* distance.

Manhattan distance: $|x_1 - x_2| + |y_1 - y_2|$.

B) Use the “*leave-one-out*” method to select the best value of the “ k ” parameter between $k=1$ and $k=4$, using the *Euclidean* distance. The “*leave-one-out*” method works as follows:

- 1) Given the training set D with n patterns (12 patterns in this exercise)
 - 2) for $i=1, \dots, n$, use the training set $\{D - \{x_i\}\}$ and then classify the pattern x_i left out.
 - 3) Repeat the point (2)
 - 4) Compute the error probability (number of errors for the classifications of the n patterns left out)
- You should use the above “*leave-one-out*” method for $k=1$ and $k=4$ and then select the value of the k parameter that provides the minimum error.



Class A: red points; Class B: blue points; Class C: black points;

A) Classify the pattern x_t with values of $k=1, \dots, 4$ using the *Euclidean* and the *Manhattan* distance.

Squared *Euclidean* distances

	a1	a2	a3	a4
x_t	1.3000	0.4500	0.6400	0.1700

	b1	b2	b3	b4
x_t	0.8500	1.0000	0.0400	0.5300

	c1	c2	c3	c4
x_t	0.6100	0.8000	0.4000	1.0600

Classification result

k	A	B	C	Classification
1		1		B
2	1	1		A-B*
3	1	1	1	A-B-C*
4	2	1	1	A

* -> you can do a random choice among the classes

Manhattan distances

	a1	a2	a3	a4
x_t	1.60	0.90	0.80	0.50

	b1	b2	b3	b4
x_t	1.30	1.40	0.20	0.90

	c1	c2	c3	c4
x_t	1.10	1.20	0.80	1.40

k	A	B	C	Classification
1		1		B
2	1	1		A-B *
3	2	1		A**
4	2	1	1	A

* -> you can do a random choice among the classes

**-> you can do a random choice among the classes A, B, C. You have B-A-A or B-A-C, so the classification is A or one random choice among A, B, C

B) Use the “*leave-one-out*” method to select the best value of the “k” parameter between k=1 and k=4, using the *Euclidean* distance.

In order to use the “*leave-one-out*” method, you should compute the Euclidean distances between each pattern and all the other ones, so that, for each value of k, k=1 or k=4, you can compute the classification of the pattern x_i left out ($\{D - \{x_i\}\}$).

Here the Euclidean distances between each pattern and all the other ones:

	b1	b2	b3	b4
a1	2.57	2.26	1.70	1.21
a2	1.0	0.85	0.61	0.26
a3	2.61	2.60	1.0	1.49
a4	1.46	1.49	0.37	0.72

	c1	c2	c3	c4
a1	1.53	1.22	2.18	2.72
a2	1.30	1.25	1.45	2.29
a3	0.29	0.16	0.72	0.90
a4	0.40	0.41	0.53	1.01

	c1	c2	c3	c4
b1	2.90	3.25	2.25	3.69
b2	3.13	3.40	2.60	4.10
b3	0.89	1.16	0.52	1.30
b4	2.08	2.21	1.85	3.05

	a1	a2	a3	a4
a1	0	0.37	0.50	0.61
a2	0.37	0	0.61	0.26
a3	0.50	0.61	0	0.17
a4	0.61	0.26	0.17	0

	b1	b2	b3	b4
b1	0	0.05	0.61	0.26
b2	0.05	0	0.80	0.17
b3	0.61	0.80	0	0.49
b4	0.26	0.17	0.49	0

	c1	c2	c3	c4
c1	0	0.05	0.17	0.17
c2	0.05	0	0.40	0.34
c3	0.17	0.40	0	0.18
c4	0.17	0.34	0.18	0

In the following tables, for each value of k ($k=1$ or $k=4$), we compute the classification of the pattern x_i left out ($\{D - \{x_i\}\}$).

- $k=1$

pattern	a1	a2	a3	a4	b1	b2	b3	b4	c1	c2	c3	c4
One nearest neighbor	A	A	C	A	B	B	A	B	C	C	C	C

Error = 2/12

- $k=4$

pattern	a1	a2	a3	a4	b1	b2	b3	b4	c1	c2	c3	c4
Classes of the 4 nearest neighbors	AA AB	AA AB	AA CC	AA BC	AB BB	AB BB	AB BC	AB BB	AC CC	AC CC	BC CC	AC CC
Classification	A	A	A-C*	A	B	B	B	B	C	C	C	C

Error = between 0/12 and 1/12 depending on the random choice for “tie breaks” highlighted in yellow.

The best value of the parameter “ k ” is therefore $k=4$ (minimum error).

Exercise 2

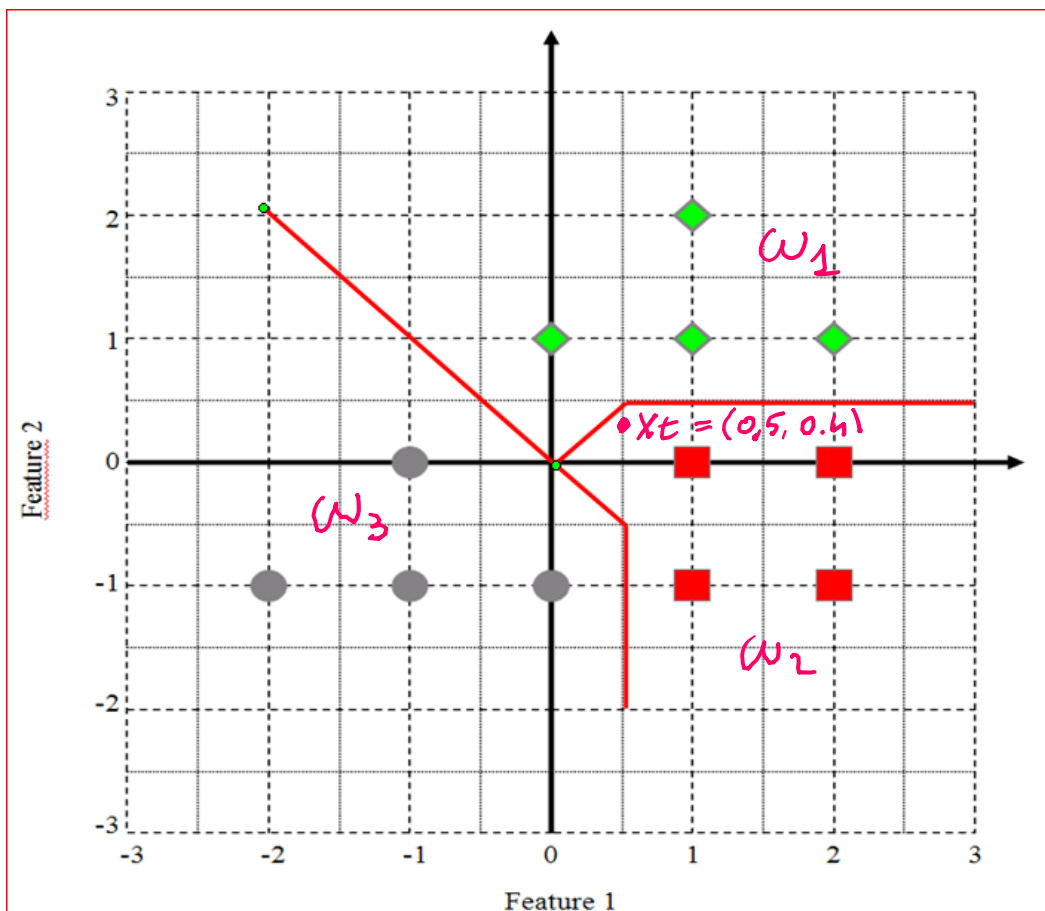
Given the following patterns belonging to three classes:

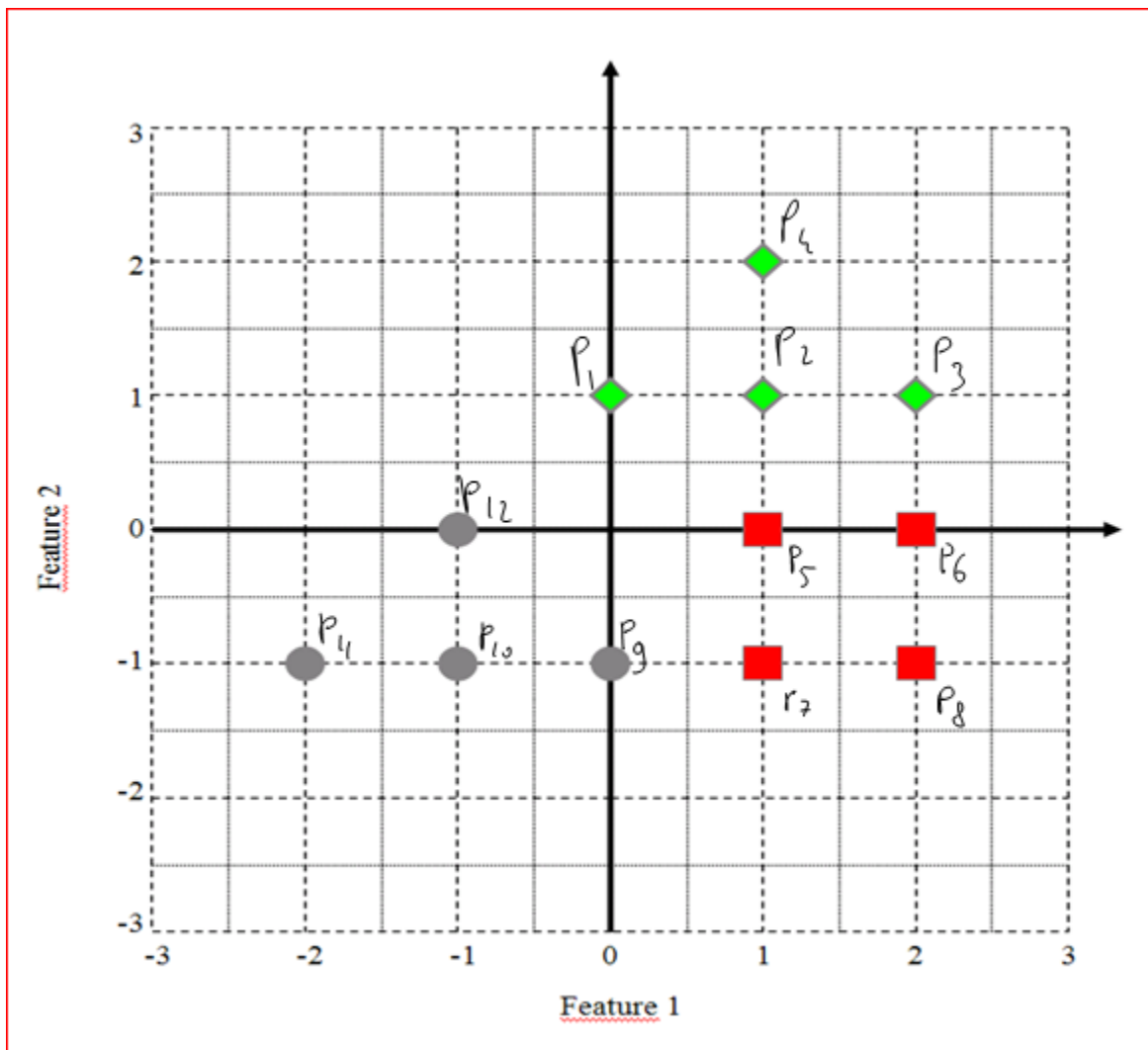
Class	ω_1				ω_2				ω_3			
	p1	p2	p3	p4	p5	p6	p7	p8	p9	p10	p11	p12
	0.0	1	2	1	1	2	1	2	0	-1	-2	-1
	1.0	1	1	2	0	0	-1	-1	-1	-1	-1	0

- 1) Classify the unknown pattern $X_t = (0.5, 0.4)^t$ using the k-nn classifier and the Euclidean distance for $k=1$ and $k=3$.
- 2) Estimate the optimal value of the parameter “k”, between $k=1$ and $k=3$, using the Manhattan distance. Explain the used method.

[When the unknown pattern is equidistant from different training patterns, we select first training patterns with a lower class index]

Geometrical representation of the pattern distribution:





Solution:

Question (1):

Evaluate the euclidean distance between the unknown pattern and the training patterns

Class	ω_1				ω_2				ω_3			
	p1	p2	p3	p4	p5	p6	p7	p8	p9	p10	p11	p12
X_t	0.61	0.61	2.61	2.81	0.41	2.41	2.21	4.21	2.21	4.21	8.21	2.41
	W1	W1			W2							

- $K=1 \quad x \rightarrow \omega_2$
- $K=3 \quad x \rightarrow \omega_1$

Question (2):

The optimal value of K can be estimated using the leave-one-out method. See the previous exercise.

Let us evaluate the distances between all the training patterns using the given metric:

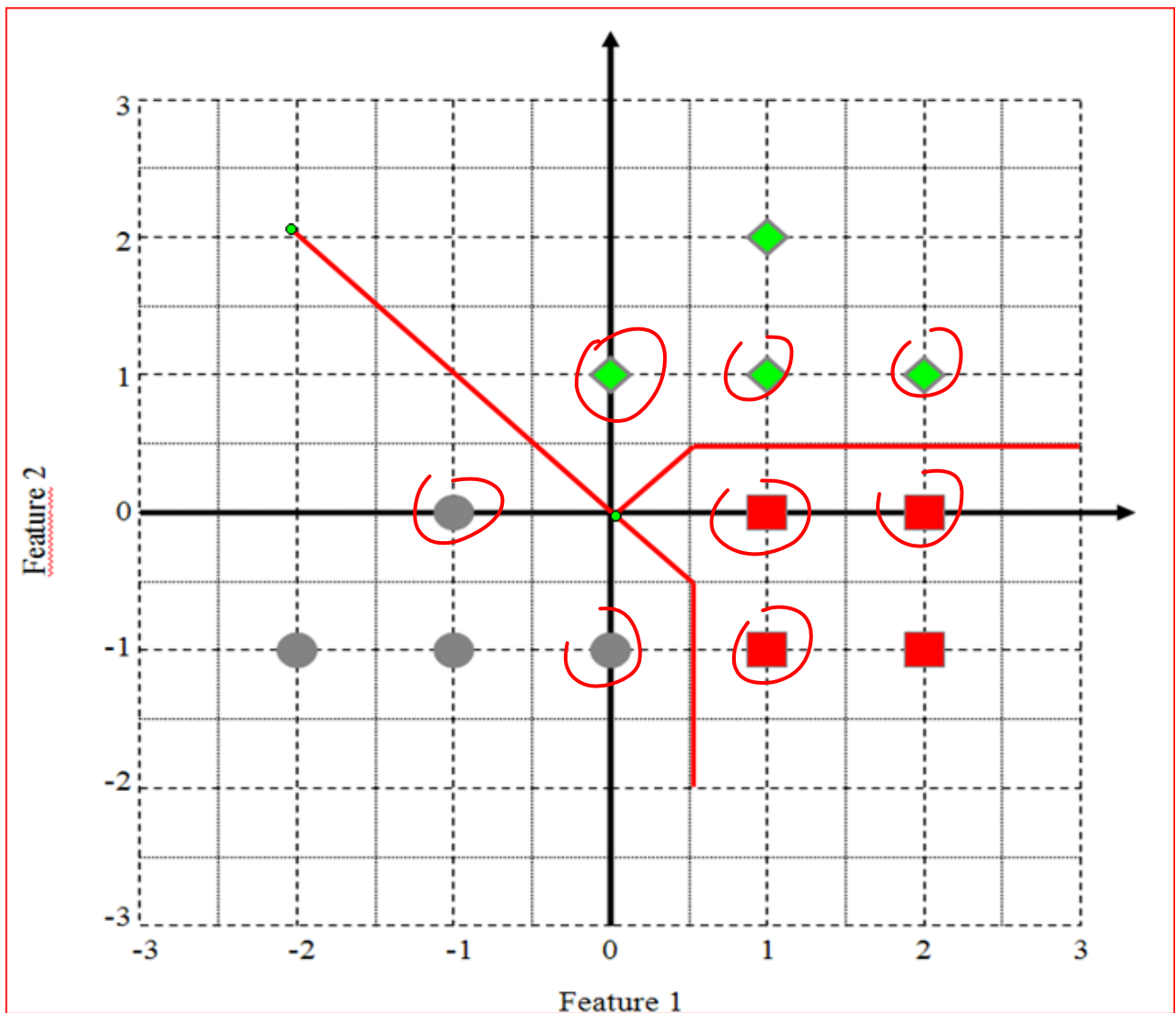
$$d(p_a, p_b) = |a_1 - b_1| + |a_2 - b_2|$$

	P1	P2	P3	P4	P5	P6	P7	P8	P9	P10	P11	P12
P1	0	1	2	2	2	3	3	4	2	3	4	2
P2		0	1	1	1	2	2	3	3	4	5	3
P3			0	2	2	1	3	2	4	5	6	4
P4				0	2	3	3	4	4	5	6	4
P5					0	1	1	2	2	3	4	2
P6						0	2	1	3	4	5	3
P7							0	1	1	2	3	3
P8								0	2	3	4	5
P9									0	1	2	2
P10										0	1	1
P11											0	2
P12												0

Supposing that, in the case of equal distances or equal number of classification, we choose first the training patterns with a lower class index, the result is:

True Class	Pattern	K=1 Nearest pattern	K=1 Assigned Class	K=3 Nearest patterns	K=3 Assigned Class
ω_1	P1	P2(ω_1)	ω_1	P2(ω_1), P3(ω_1), P4(ω_1)	ω_1
ω_1	P2	P1(ω_1)	ω_1	P1(ω_1), P4(ω_1), P5(ω_2)	ω_1
ω_1	P3	P2(ω_1)	ω_1	P2(ω_1) P6(ω_2) P1(ω_1)	ω_1
ω_1	P4	P2(ω_1)	ω_1	P2(ω_1) P1(ω_1), P3(ω_1)	ω_1
ω_2	P5	P2(ω_1)	ω_1	P2(ω_1) P6(ω_2) P7(ω_2)	ω_2
ω_2	P6	P3(ω_1)	ω_1	P3(ω_1) P5(ω_2) P8(ω_2),	ω_2
ω_2	P7	P5(ω_2)	ω_2	P5(ω_2) P8(ω_2) P9(ω_3)	ω_2
ω_2	P8	P6(ω_2)	ω_2	P6(ω_2) P7(ω_2) P3(ω_1),	ω_2
ω_3	P9	P7(ω_2)	ω_2	P7(ω_2) P10(ω_3) P1(ω_1)	ω_1
ω_3	P10	P9(ω_3)	ω_3	P9(ω_3) P11(ω_3) P12(ω_3)	ω_3
ω_3	P11	P10(ω_3)	ω_3	P10(ω_3) P9(ω_3) P12(ω_3)	ω_3
ω_3	P12	P10(ω_3)	ω_3	P10(ω_3) P1(ω_1), P5(ω_2)	ω_1
	ERROR		3/12		2/12

The 'leave-one-out' method suggests to choose $k=3$



TABLES EXERCISE 3

$$X_{A1}=(0.4 \ 0.5 \ 12 \ 10)^T; \ X_{A2}=(5 \ 11 \ 10 \ 10)^T$$

$$X_{B1}=(1 \ 1 \ 0 \ 6)^T; \ X_{B2}=(5 \ 2 \ 0.6 \ 0.6)^T$$

$$X_t=(0.4, 0.5, 0.4, 0.5)^T$$

Distances between X_t and all the other patterns.

	$d(x_{A1},x_t)^2$	$d(x_{A2},x_t)^2$	$d(x_{B1},x_t)^2$	$d(x_{B2},x_t)^2$
Feature = 1	0	21.16	0.36	21.16
Feature = 1,2	0	131.41	0.61	23.41
Feature = 1,2,3	134.56	223.57	0.77	23.45
Feature = 1,2,3,4	224.81	313.82	31.02	23.46