

PRACTICE: BINARY CLASSIFICATION

- From the Bayes Classifier, predict the class for each test data and compute the error.

Training sample								
x_1	a	a	b	a	a	b	b	b
x_2	b	a	a	a	a	b	b	b
y	1	1	1	1	-1	-1	-1	-1

Test sample				
x_1	a	a	b	b
x_2	a	b	a	b
prediction	?	?	?	?
real	1	-1	1	1

- Suppose that $\pi_1 = \pi_0 = 0,5$ and the densities are $g_1 = \mathcal{N}(0, 1)$ and $g_0 = 0,7\mathcal{N}(0, 1) + 0,3\mathcal{N}(-1, 2)$.
 - Assuming equal cost find:
 - Plot the densities and write the Bayes rule for this classification task.
 - Write the Bayes decision boundary and find its solutions.
 - Assume that $C(1, 0) = 2$ and $C(0, 1) = 6$. Repeat questions above.
- Generate 100 observations from a bivariate Gaussian distribution $\mathcal{N}(\mu_1, \Sigma_1)$ with $\mu_1 = (3, 1)'$ and $\Sigma_1 = I$ (identity matrix and label them as 1. Generate another 100 observations from a bivariate Gaussian distribution $\mathcal{N}(\mu_2, \Sigma_2)$ with $\mu_2 = (1, 3)'$ and $\Sigma_2 = I$ and label them as 0. Together, these 200 observations constitute the training set.
 - Write an R code to generate this data set.
 - Plot this data using different colors for the two classes.
 - Assuming that priors are equals, find the Bayes Classifier.
 - Compute the training error.
 - Train a linear regression model, using the function $\text{lm}(y \sim x)$, with the training set.
 - Plot the boundary decision of Bayes Classifier and the line obtained by the linear regression model.
 - Generate a test set of 50 observations and compute the test error of Bayes Classifier and the linear model.

x_1	x_2	x_3	y
0	3	0	Red
2	0	0	Red
0	1	3	Red
0	1	2	Green
-1	0	1	Green
1	1	1	Red

- Consider the following table.

- With the euclidean distance, what is the prediction with $k = 1$ and with $k = 3$ for the test observation $(0, 0, 0)$?
- If the Bayes decision boundary in this problem is highly non-linear, then would we expect the best value for k to be large or small? Why?