## Classification

## Machine Learning 2021-2022 - UMONS Souhaib Ben Taieb

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Suppose we collect data for a group of students in a statistics class with variables:

- $X_1$  = hours studied.
- $X_2$  = undergrad GPA.
- Y = receive an A.

We fit a logistic regression and produce estimated coefficients:

- $\hat{\beta}_0 = -6$
- $\hat{\beta}_1 = 0.05$
- $\hat{\beta}_2 = 1$
- a) Estimate the probability that a student who studies for 40h and has an undergrad GPA of 3.5 gets an A in the class.
- b) How many hours would the above student need to study to have a 50% chance of getting an A in the class?

Suppose that we wish to predict whether a given stock will issue a dividend this year ("Yes" or "No") based on X, last year's percent profit. We examine a large number of companies and discover that the mean value of X for companies that issued a dividend was X = 10, while the mean for those that didn't was X = 0. In addition, the variance of X for these two sets of companies was  $\sigma^2 = 36$ . Finally, 80% of companies issued dividends. Assuming that X follows a normal distribution, predict the probability that a company will issue a dividend this year given that its percentage profit was X = 4 last year.

Hint: Recall that the density function for a normal random variable is:

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right)$$

You will need to use Bayes' theorem.

Consider the following dataset with n=8 observations, three binary input features and a binary response.

$X_1$	$X_2$	$X_3$	Y
1	0	1	1
1	1	1	1
0	1	1	0
1	1	0	0
1	0	1	0
0	0	0	1
0	0	0	1
0	0	1	0

Assume we are using a naive Bayes classifier to predict the value of Y from the values of the other variables.

• What is 
$$P(Y = 1 | X_1 = 1, X_2 = 1, X_3 = 0)$$
?

• What is 
$$P(Y = 0|X_1 = 1, X_2 = 1)$$
?

Now, suppose that we are using a joint Bayes classifier to predict the value of Y from the values of the other variables.

• What is 
$$P(P(Y=1|X_1=1,X_2=1,X_3=0))$$
?

• What is 
$$P(Y = 0|X_1 = 1, X_2 = 1)$$
?

Suppose that we take a data set, divide it into equally-sized training and test sets, and then try out two different classification procedures. First we use logistic regression and get an error rate of 20% on the training data and 30% on the test data. Next we use 1-nearest neighbors (i.e. K = 1) and get an average error rate (averaged over both test and training data sets) of 18%. Based on these results, which method should we prefer to use for classification of new observations? Why?

This problem relates to the QDA model, in which the observations within each class are drawn from a normal distribution with a class specific mean vector and a class specific covariance matrix. We consider the simple case where p = 1; i.e. there is only one feature.

Suppose that we have K classes, and that if an observation belongs to the  $k^{th}$  class, then X comes from a one-dimensional normal distribution,  $X \sim \mathcal{N}(\mu_k, \sigma_k^2)$ . Prove that, in that case, the Bayes' classifier is not linear. Argue that it is in fact quadratic.