

# Tutorial 4

Machine Learning and Big Data for Economics and Finance

## List of activities

- I. Complete **Section 4.6 Lab: Logistic Regression, LDA, QDA, and KNN**, subsections 4.6.1, 4.6.2 and 4.6.5.
- II. Complete the list of exercises in this tutorial.

### Exercise 1. Logistic and logit transformations

- Show step by step that the inverse of

$$f(x) = \log\left(\frac{x}{1-x}\right)$$

is given by

$$f^{-1}(x) = \frac{1}{1 + e^{-x}}.$$

- Show that  $f^{-1}$  is increasing.
- Show that as  $x \rightarrow -\infty$ ,  $f^{-1}(x) \rightarrow 0$  and as  $x \rightarrow \infty$ ,  $f^{-1}(x) \rightarrow 1$ .

**Solution.**

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$$\begin{aligned} u &=_{(i)} \log\left(\frac{x}{1-x}\right) \\ -u &=_{(ii)} \log\left(\frac{1-x}{x}\right) \\ e^{-u} &= \frac{1-x}{x} \end{aligned}$$

where (i) follows from  $\log\left(\frac{a}{b}\right) = \log(a) - \log(b) = -(-\log(a) + \log(b)) = -\log\left(\frac{b}{a}\right)$ .

(ii) follows from taking exponentials on both sides.

Finally, some algebra shows that  $xe^{-u} = 1 - x \Rightarrow x(1 + e^{-u}) = 1$  which finally implies that

$$x = \frac{1}{1 + e^{-u}}$$

which is the inverse function

- If  $x$  increases then  $-x \searrow$  then  $e^{-x} \searrow$  then  $1 + e^{-x} \searrow$  then  $\frac{1}{1 + e^{-x}}$  increases.
- $x \rightarrow -\infty$  then  $e^{-x} \rightarrow \infty$  that is  $f^{-1}(x) \rightarrow 0$ .  
 $x \rightarrow \infty$  then  $e^{-x} \rightarrow 0$  then  $f^{-1}(x) \rightarrow 1$ .

**Exercise 2.** Write an R function `loglik_logit` that takes data and a parameter  $\beta$  as input and that outputs the logarithm of the likelihood of the logistic regression model.

Test your function on the dataset in the file `LR1.csv` where the model is

$$\text{logit}(\Pr\{Y=1|X=x\}) = -5 + x\beta$$

Maximize the likelihood and compare to the function `glm`.

**Solution.**

Let  $p(x; \beta) = \frac{e^{-5+x\beta}}{1 + e^{-5+x\beta}}$ , then we need to write the logarithm of the function

$$L(\beta) = \prod_{i: y_i=1} p(x_i; \beta) \prod_{i': y_{i'}=0} (1 - p(x_{i'}; \beta))$$

Taking logarithms

$$\text{Log}(L(\beta)) = \sum_{i: y_i=1} \log(p(x_i; \beta)) + \sum_{i': y_{i'}=0} \log(1 - p(x_{i'}; \beta))$$

This is the function we need to code in R. Look at the code file attached to the solutions.