HES-SO Machine Learning

# $\begin{array}{c} {\rm Practical\ work\ 02-26th\ of\ septembre\ 2017} \\ {\rm Linear\ Regression} \end{array}$

#### Summary for the organisation:

- Submit the solutions of the practical work before Monday 12h00 next week in Moodle.
- Preferred modality: iPython notebook.
- Alternative modality : pdf report.
- The file name must contain the number of the practical work, followed by the names of the team members by alphabetical order, for example O2\_dupont\_muller\_smith.pdf.
- Put also the name of the team members in the body of the notebook (or report).
- Only one submission per team.

#### Exercice 1 Get the data

- Get 'lausanne-appart.xlsx' from moodle
- Open a new iPython notebook
- Read columns 0 (x) and column 2 (y) from the excel file 'lausanne-appart.xlsx'
- Visualize the data
  - a) histogram to visualize the distribution of the renting price
  - b) histogram to visualize the distribution of the living area
  - c) scatter plot of living area as a function of renting price

## Exercice 2 Normal equations for linear regression

In the case of a monovariable linear regression, we want to discover the parameters  $\theta$  of a simple linear model defined with

$$h_{\theta}(\mathbf{x}) = \theta_0 + \theta_1 x \tag{1}$$

The "best"  $\theta$ 's are the ones that will minimise the squared difference between the gotten outputs and the target outputs defined with

$$J(\theta) = \frac{1}{2N} \sum_{n=1}^{N} (h_{\theta}(\mathbf{x}_n) - y_n)^2$$
(2)

A closed form solution to this problem is given by the following normal equation:

$$\theta = (X^T X)^{-1} X^T \vec{y} \tag{3}$$

With X and  $\vec{y}$  defined as

$$X = \begin{pmatrix} 1 & x_{1,1} & \dots & x_{1,D} \\ 1 & \ddots & & & \\ 1 & \vdots & x_{n,d} & \vdots \\ 1 & & \ddots & \\ 1 & x_{N,1} & \dots & x_{N,D} \end{pmatrix}$$
(4)

$$y = \begin{pmatrix} y_1 \\ \vdots \\ y_N \end{pmatrix} \tag{5}$$

- a) Implement Equation 3 assuming that x is the living area and y is the renting price. Use numpy for the vector operations. Plot the computed line on top of the scatter plot of exercise 1.
- b) Compute the overall cost value according to Equation 2.

#### Exercice 3 Batch gradient descent for linear regression

Implement the batch gradient descent algorithm for the previous problem. As seen in the theory, the update rules are

$$\theta_0 \leftarrow \theta_0 - \alpha \frac{1}{N} \sum_{n=1}^{N} (h_{\theta}(\mathbf{x}_n) - y_n)$$
 (6)

$$\theta_1 \leftarrow \theta_1 - \alpha \frac{1}{N} \sum_{n=1}^{N} (h_{\theta}(\mathbf{x}_n) - y_n) x_{n,1}$$
 (7)

**Remark** You need to iterate several times over the training set. If you have problems of convergence, you need to use a smaller value of  $\alpha$ . Values such as 0.000001 are common.

- a) Plot the cost value (Equation 2) as a function of the iterations. What do you observe?
- b) Imagine a stopping criterion, i.e. when do we stop iterating on the training set?

- c) Plot the computed line  $h_{\theta}(\mathbf{x})$  on top of the scatter plot of exercise 1.
- d) Compute the final cost value according to Equation 2 and compare it to the one of exercise 2. What can you conclude?

#### Exercice 4 Stochastic gradient descent for linear regression

Implement the stochastic gradient descent algorithm for the previous problem. As seen in the theory, the update rules are

$$\theta_i \leftarrow \theta_i - \alpha (h_\theta(\mathbf{x}_n) - y_n) x_{n,i}$$
 (8)

- a) Plot the computed line  $h_{\theta}(\mathbf{x})$  on top of the scatter plot of exercise 1.
- b) How many samples do you need to visit for reaching the convergence?
- c) What kind of stopping criterion could we use here?
- d) Compute the final cost value according to Equation 2 and compare it to the one of exercise 2 and 3. What can you conclude?

# Exercice 5 Optional – Mini-batch gradient descent for linear regression

Implement the mini-batch gradient descent algorithm for the previous problem, adding a parameter B defining the size of the mini-batch. Check that when B = N, you fall back on the batch gradient descent solution, and when B = 1, you get the behaviour of stochastic gradient descent.

## Exercice 6 Optional – multi-variable linear regression

- a) Implement one of the gradient descent algorithm (ex. 3-5) for the multi-variable linear regression assuming  $x_1$  being the living area and  $x_2$  the square of the living area. Plot the computed curve (second order) on top of the scatter plot of exercise 1.
- b) Implement one of the gradient descent algorithm (ex. 3-5) for the multi-variable linear regression assuming  $x_1$  being the living area and  $x_2$  the number of bedrooms.