Exercise 1

<u>Parametric f(.)</u>: Linear Regression; where f(x) = a + bx, x being the independent variable. The parameters are a and b and are calculated using the data of the problem.

Non-parametric f(.): Decision Trees; in the sense that the only parameter is the size of the tree which depends on the size of the dataset but not the actual values of the data.

Exercise 2

a. Parametric set of the quadratic functions f_{ϑ} : $R \rightarrow R$

That is, for a given $x = [x_1]^T \in R$ it is:

$$f_{\theta}(x_1) = \theta_0 + \theta_1 x_1 + \theta_2 x_1^2, \qquad \theta = [\theta_0, \theta_1, \theta_2]^T$$

and
$$F := \{ f_{\theta}(\cdot) : \theta \in \mathbb{R}^3 \}$$

eg.a.1
$$\theta = [4,5,3]^T$$
 then the instance of F is $f_{\theta}(x_1) = 4 + 5x_1 + 3x_1^2$

eg.a.2
$$\theta = [2,1,0]^T$$
 then the instance of F is $f_{\theta}(x_1) = 2 + x_1$

b. Parametric set of the 3rd degree polynomials functions f_{ϑ} : R2 \rightarrow R

That is, for a given $x = [x_1, x_2]^T \in \mathbb{R}^2$ it is:

$$f_{\theta}(x_1, x_2) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1 x_2 + \theta_4 x_1^2 + \theta_5 x_2^2 + \theta_6 x_1^2 x_2 + \theta_7 x_2^2 x_1 + \theta_8 x_1^3 + \theta_9 x_2^2,$$

$$\theta = \left[\theta_0, \theta_1, \theta_2, \theta_3, \theta_y, \theta_5, \theta_6, \theta_7, \theta_8, \theta_9\right]^T$$

and
$$F := \{ f_{\theta}(\cdot) : \theta \in R^{10} \}$$

 $\underline{\text{eg.b.1}} \theta = [1,4,3,2,5,6,7,1,2,3]^T$ then the instance of F is

$$f_{\theta}(x_1, x_2) = 1 + 4x_1 + 3x_2 + 2x_1x_2 + 5x_1^2 + 6x_2^2 + 7x_1^2x_2 + x_2^2x_1 + 2x_1^3 + 3x_2^2$$

 $\underline{\text{eg.b.2}} \ \theta = [0,1,34,0,7,0,1,2,0,1]^T$ then the instance of F is

$$f_{\theta}(x_1, x_2) = x_1 + 34x_2 + 7x_1^2 + x_1^2x_2 + 2x_2^2x_1 + x_2^2$$

c. Parametric set of the 3rd degree polynomials functions f_{θ} : R3 \rightarrow R

That is, for a given $x = [x_1, x_2, x_3]^T \in \mathbb{R}^3$ it is:

$$f_{\theta}(x_{1}, x_{2}, x_{3}) = \theta_{0} + \theta_{1}x_{1} + \theta_{2}x_{2} + \theta_{3}x_{3} + \theta_{4}x_{1}^{2} + \theta_{5}x_{2}^{2} + \theta_{6}x_{3}^{2} + \theta_{7}x_{1}x_{2} + \theta_{8}x_{1}x_{3} + \theta_{9}x_{2}x_{3} + \theta_{10}x_{1}^{3} + \theta_{11}x_{2}^{3} + \theta_{12}x_{3}^{3} + \theta_{13}x_{1}^{2}x_{2} + \theta_{14}x_{1}^{2}x_{3} + \theta_{15}x_{2}^{2}x_{1} + \theta_{16}x_{2}^{2}x_{3} + \theta_{17}x_{3}^{2}x_{1} + \theta_{18}x_{3}^{2}x_{2} + \theta_{19}x_{1}x_{2}x_{3}$$

$$\boldsymbol{\theta} = \left[\theta_0, \theta_1, \theta_{2, \dots}, \theta_{19}\right]^T$$

and $F := \{ f_{\theta}(\cdot) : \theta \in \mathbb{R}^{20} \}$

 $\underline{\text{eg.c.1}} \theta = [1,4,3,2,5,6,7,1,2,3,0,0,0,1,12,9,54,0,1,0]^T$ then the instance of F is

$$f_{\theta}(x_1, x_2, x_3) = 1 + 4x_1 + 3x_2 + 2x_3 + 5x_1^2 + 6x_2^2 + 7x_3^2 + x_1x_2 + 2x_1x_3 + 3x_2x_3 + x_1^2x_2 + 12x_1^2x_3 + 9x_2^2x_1 + 54x_2^2x_3 + x_3^2x_2$$

 $\underline{\text{eg.c.2}} \theta = [0,0,0,0,1,0,0,0,0,0,0,0,0,0,0,0,0,0]^T$ then the instance of F is

$$f_{\theta}(x_1, x_2, x_3) = x_1^2 + x_2^2 x_3$$

d.
$$f_{\theta}(x): R^5 \to R, f_{\theta}(x) = \frac{1}{1 + exp(-\theta^T x)}$$

That is, for a given $\mathbf{x} = [\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4, \mathbf{x}_5]^\mathsf{T} \in \mathbb{R}^3$ and $\theta = [\theta_0, \theta_1, \theta_2, \theta_3, \theta_4, \theta_5]^T$ it is:

$$f_{\theta}(\mathbf{x}_{1}, \mathbf{x}_{2}, \mathbf{x}_{3}, \mathbf{x}_{4}, \mathbf{x}_{5}) = \frac{1}{1 + exp(-\theta_{0} - \theta_{1}x_{1} - \theta_{2}x_{2} - \theta_{3}x_{3} - \theta_{4}x_{4} - \theta_{5}x_{5})}$$

and $F \coloneqq \{f_{\theta}(\cdot) : \theta \in R^6\}$

 $\underline{\text{eg.d.1}} \theta = [-1,1,-1,1,1,-1]^T$ then the instance of F is

$$f_{\theta}(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4, \mathbf{x}_5) = \frac{1}{1 + exp(1 - x_1 + x_2 - x_3 - x_4 + x_5)}$$

eg.d.2 $\theta = [0,0,0,0,-1]^T$ then the instance of F is

$$f_{\theta}(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4, \mathbf{x}_5) = \frac{1}{1 + exp(x_5)}$$

e. f is linear with respect to θ in a, b, c but not in d

Exercise 3

$$(\theta^T x) x = ([\theta_1 \, \theta_2 ... \theta_l] [x_1 \, x_2 ... x_l]^T) [x_1 \, x_2 ... x_l]^T = (x_1 \theta_1 + \theta_2 \, x_2 + ... + \theta_l x_l) [x_1 \, x_2 ... x_l]^T =$$

$$\begin{bmatrix} \theta_1 x_1^2 + \theta_2 x_2 x_1 + & \cdots & + \theta_1 x_l x_1 \\ \theta_1 x_1 x_2 + \theta_2 x_2^2 + & \cdots & + \theta_1 x_l x_2 \\ \vdots & & & & \\ \theta_1 x_1 x_l + \theta_2 x_2 x_l + & \cdots & + \theta_l x_l^2 \end{bmatrix} = \begin{bmatrix} x_1^2 & x_2 x_1 & \cdots & x_l x_1 \\ x_1 & x_2 x_2^2 & \cdots & x_l x_2 \\ \vdots & & & & \\ x_1 x_l & x_2 x_l & \cdots & x_l^2 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \vdots \\ \theta_l \end{bmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_l \end{bmatrix} [x_1 x_2 \dots x_l]) \theta^{\mathsf{T}} = (x x^T) \theta$$

Thus,
$$(\theta^T x)x = (xx^T)\theta$$

Exercise 4

First, we will show that $X^T = [x_1 \quad x_2 \quad \cdots \quad x_N]$

Since
$$X = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1l} \\ x_{21} & x_{22} & \cdots & x_{2l} \\ \vdots & \vdots & \ddots & \vdots \\ x_{N1} & x_{N2} & \cdots & x_{Nl} \end{bmatrix}$$
, then $X^T = \begin{bmatrix} x_{11} & x_{21} & \cdots & x_{N1} \\ x_{12} & x_{22} & \cdots & x_{N2} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1l} & x_{2l} & \cdots & x_{Nl} \end{bmatrix} = \begin{bmatrix} x_1 & x_2 & \cdots & x_N \end{bmatrix}$

a.
$$X^TX = \begin{bmatrix} x_1 & x_2 & \cdots & x_N \end{bmatrix} \begin{bmatrix} x_1^T \\ x_2^T \\ \vdots \\ x_N^T \end{bmatrix} = x_1x_1^T + x_2x_2^T + \cdots + x_Nx_N^T = \sum_{n=1}^N x_nx_n^T$$

b. $X^Ty = \begin{bmatrix} x_1 & x_2 & \cdots & x_N \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_N \end{bmatrix} = x_1y_1 + x_1y_1 + \cdots + x_Ny_N = \sum_{n=1}^N x_ny_n$

But $\sum_{n=1}^{N} x_n y_n = \sum_{n=1}^{N} y_n x_n$, since y_n is just a number.

Exercise 6

The velocity of a body performing smoothly accelerating motion is given by the equation:

$$v = v_0 + at$$

Where t is time, v_0 is the initial speed and α is the acceleration.

$$y = v = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \end{bmatrix} = \begin{bmatrix} 5.1 \\ 6.8 \\ 9.2 \\ 10.9 \\ 13.1 \end{bmatrix}, \qquad X = \begin{bmatrix} 1 & t_1 \\ 1 & t_2 \\ 1 & t_3 \\ 1 & t_4 \\ 1 & t_5 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \\ 1 & 5 \end{bmatrix}, \qquad \theta^T = \begin{bmatrix} v_0 & \alpha \end{bmatrix}$$

Taking advantage of the least squares error criterion $\hat{\theta} = (X^T X)^{-1} X^T y$, we will find the initial speed and the acceleration.

First, we need to prove that the inverse matrix of X^TX exists.

$$X^{T}X = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 & 5 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \\ 1 & 5 \end{bmatrix} = \begin{bmatrix} 5 & 15 \\ 15 & 55 \end{bmatrix}$$

Thus, det=275-225=50 and $(X^TX)^{-1} = \begin{bmatrix} 1.1 & -0.3 \\ -0.3 & 0.1 \end{bmatrix}$

$$(X^TX)^{-1}X^T = \begin{bmatrix} 1.1 & -0.3 \\ -0.3 & 0.1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 & 5 \end{bmatrix} = \begin{bmatrix} 0.8 & 0.5 & 0.2 & -0.1 & -0.4 \\ -0.2 & -0.1 & 0 & 0.1 & 0.2 \end{bmatrix}$$

$$(X^TX)^{-1}X^Ty = \begin{bmatrix} 0.8 & 0.5 & 0.2 & -0.1 & -0.4 \\ -0.2 & -0.1 & 0 & 0.1 & 0.2 \end{bmatrix} \begin{bmatrix} 5.1 \\ 6.8 \\ 9.2 \\ 10.9 \\ 13.1 \end{bmatrix} = \begin{bmatrix} 2.99 \\ 2.01 \end{bmatrix}$$

In conclusion, $\boldsymbol{\hat{\theta}} = \begin{bmatrix} 2.99 \\ 2.01 \end{bmatrix}$

Initial speed is 2.99, acceleration is 2.01 and velocity at any time point t>0 is given by the equation

$$v = 2.99 + 2.01t$$

At t = 2.3 velocity of the body is 7.613.