

Homework 1

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Let $x = [x_1, x_2]^T$ be the input / feature vector and y the output variable in a regression task.

Consider the following scenarios that relate y with x .

i) $y = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1 x_2 + x_1^2$

ii) $y = \text{sign}(\theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1 x_2 + \theta_4 x_1^2)$

iii) $y = 2x_1 + \text{sign}(3 - T)x_2 + \text{ReLU}(3)x_1 x_2$

iv) $y = \theta + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1 x_2$

a) How many and which are the parameters involved in each scenario?

b) In which of the above scenarios y has linear dependence with the involved parameters?

• Scenario i) $y = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1 x_2 + x_1^2$

a) There are 4 parameters $\theta_0, \theta_1, \theta_2, \theta_3$

b) y has linear dependence on the parameters $\theta_0, \theta_1, \theta_2, \theta_3$ as each term appears in linear form with respect to parameters. Although the dependence regarding the parameters is linear, due to the term x_1^2 y is not linear to the input x 's.

• Scenario ii) $y = \text{sign}(\theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1 x_2 + \theta_4 x_1^2)$

a) There are 5 parameters $\theta_0, \theta_1, \theta_2, \theta_3, \theta_4$

b) y is not linearly dependent on the parameters as, sign function is not a linear function

• Scenario iii) $y = 2x_1 + \text{sign}(3 - T)x_2 + \text{ReLU}(3)x_1 x_2$

a) There are no parameters.

b) y has no linear dependence on parameters as there are no parameters.

• Scenario iv) $y = \beta + \beta_1 x_1 + \beta_2 x_2 + \beta_{12} x_1 x_2$

- a) The parameter here is one and it is β .
- b) y is linearly dependent on β

Exercise 2.

Consider different Machine Learning tasks where the input x is two-dimensional if $x = [x_1, x_2]^T$ and the output y is scalar. Assume also that in each case, a data set that consists of pairs $(y_i, x_i) \equiv (y_i, [x_{i1}, x_{i2}]^T)$, $i=1, \dots, N$ is available. Which of the following models are parametric and which are non-parametric.

- Model 1: $y = \theta_0 + \theta_1 x_1 + \theta_2 x_2$

This model is parametric including parameters $\theta_0, \theta_1, \theta_2$

- Model 2: $y = \text{min}(x_1, x_2)$

This model is non-parametric as it depends only on the x_1, x_2 inputs.

- Model 3: $\text{ReLU}(\theta_0 + \theta_1 x_1)$

This model is parametric as it contains parameters θ_0, θ_1

- Model 4: $\sum_{i=1}^N \theta_i (x_{i1}x_1 - x_{i2}x_2)$

This model is parametric as it contains θ_i parameters depending on the summation.

Exercise 3.

a) Define the parametric set of the quadratic functions $f_\theta: \mathbb{R} \rightarrow \mathbb{R}$ and give two instances of it. What is the dimensionality of θ ?

The function has the following form:

$$f_\theta(x) = \theta_0 + \theta_1 x_1 + \theta_2 x^2$$

Instances are: $f_\theta(x) = 5 + 6x + 7x^2$

$$f_\theta(x) = -3 + 2x + 8x^2$$

The dimensionality of θ is 3 ($\theta_0, \theta_1, \theta_2$)

b) Define the parametric set of the 3rd degree polynomials $f_\theta: \mathbb{R}^2 \rightarrow \mathbb{R}$ and give two instances of it. What is the dimensionality of θ ?

The function has the following form:

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

Instances are:

$$f_\theta(x_1, x_2) = 1 + 2x_1 + 3x_2 + x_1^2 + x_1 x_2 + x_2^2 + x_1^3 + x_1^2 x_2 + x_1 x_2^2 + x_2^3$$

$$f_\theta(x_1, x_2) = 8 + 12x_1 + 3x_2 + x_1^2 + 7x_1 x_2 + x_2^2 + 2x_1^3 + 6x_1^2 x_2 + 4x_1 x_2^2 + 8x_2^3$$

The dimensionality of θ is 10. (θ_0 to θ_9)

c) Define the parametric set of the 3rd degree polynomial $f_0: \mathbb{R}^3 \rightarrow \mathbb{R}$ and give two instances of it. What is the dimensionality of θ ?

The function has the following form:

$$f_0(x_1, x_2, x_3) = \delta_0 + \delta_1 x_1 + \delta_2 x_2 + \delta_3 x_3 + \delta_4 x_1^2 + \delta_5 x_2 x_3 + \dots + \delta_9 x_3^3$$

Instances are:

$$f_0(x_1, x_2, x_3) = 1 + x_1 + 2x_2 + 3x_3 + x_1^2 + 5x_2 x_3 + 7x_1 x_2^2 + \dots + 10x_3^3$$

$$f_0(x_1, x_2, x_3) = 4 + x_1 + x_2 + x_3 + \dots + x_3^3.$$

The dimensionality of θ is 20: (δ_0 to δ_9)

d) Consider the function $f_0(x): \mathbb{R}^5 \rightarrow \mathbb{R}, f_0(x) = \frac{1}{1 + \exp(-\delta^T x)}$.

Define the associated parametric set and give two instances of it. What is the dimensionality of θ ?

Parametric Set: consists of all functions of the form,

$$f_0(x) = \frac{1}{1 + \exp(-(\delta_0 - \delta_1 x_1 - \delta_2 x_2 - \delta_3 x_3 - \delta_4 x_4 - \delta_5 x_5))}$$

Two instances are:

$$f_0(x) = \frac{1}{1 + \exp(-(1 + 2x_1 + 3x_2 + 4x_3 + 5x_4 + 6x_5))}$$

$$\text{Where } \delta = [1, 2, 3, 4, 5, 6]$$

$$f_0(x) = \frac{1}{1 + \exp(-(0.3 + 2x_1 + 0.1x_2 + 0.7x_3 + 0.8x_4 + 1.2x_5))}$$

$$\text{Where } \delta \text{ is } [0.3, 2, 0.1, 0.7, 0.8, 1.2]$$

The dimensionality of θ is 6: $[\delta_0, \delta_1, \delta_2, \delta_3, \delta_4, \delta_5]$

e) In which of the above cases f_θ is linear with respect to θ ?

f_θ is linear with respect to θ in cases

a) $f_\theta(x) = \theta_0 + \theta_1 x + \theta_2 x^2$

b) $f_\theta(x_1, x_2) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_1^2 + \theta_4 x_1 x_2 + \dots + \theta_g x_2^2$

c) $f_\theta(x_1, x_2, x_3) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3 + \dots + \theta_{19} x_3^3$

In all above cases the function f_θ is linear with respect to θ as each θ parameter appears linearly despite products and powers of x s.

Exercise 4

Verify that for two l -dimensional column vectors $\delta = (\delta_1, \delta_2, \dots, \delta_l)$ ⁷ and $x = [x_1, x_2, \dots, x_l]^T$ it holds $(\delta^T x) x = (x \cdot x^T) \delta$.

a) Left part of the equation is: $(\delta^T x) x$

The product $\delta^T x$ is computed as

$$\delta^T x = \delta_1 x_1 + \delta_2 x_2 + \dots + \delta_l x_l = \sum_{i=1}^l \delta_i x_i$$

$$(\delta^T x) x = \left(\sum_{i=1}^l \delta_i x_i \right) \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_l \end{bmatrix} = \begin{bmatrix} \left(\sum_{i=1}^l \delta_i x_i \right) x_1 \\ \left(\sum_{i=1}^l \delta_i x_i \right) x_2 \\ \vdots \\ \left(\sum_{i=1}^l \delta_i x_i \right) x_l \end{bmatrix}$$

b) The right part of the equation is: $(x \cdot x^T) \delta$:

$$x \cdot x^T = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_l \end{bmatrix} [x_1 \ x_2 \ \dots \ x_l] = \begin{bmatrix} x_1^2 & x_1 x_2 & \dots & x_1 x_l \\ x_1 x_2 & x_2^2 & \dots & x_2 x_l \\ \vdots & \vdots & \ddots & \vdots \\ x_l x_1 & x_l x_2 & \dots & x_l^2 \end{bmatrix}$$

$$(x \cdot x^T) \cdot \delta = \begin{bmatrix} x_1^2 & x_1 x_2 & \dots & x_1 x_l \\ x_1 x_2 & x_2^2 & \dots & x_2 x_l \\ \vdots & \vdots & \ddots & \vdots \\ x_l x_1 & x_l x_2 & \dots & x_l^2 \end{bmatrix} \begin{bmatrix} \delta_1 \\ \delta_2 \\ \vdots \\ \delta_l \end{bmatrix} = \begin{bmatrix} x_1(x_1 \delta_1 + \dots + x_l \delta_l) \\ x_2(x_1 \delta_1 + \dots + x_l \delta_l) \\ \vdots \\ x_l(x_1 \delta_1 + \dots + x_l \delta_l) \end{bmatrix}$$

$$= \begin{bmatrix} x_1 \left(\sum_{i=1}^l (x_i \delta_i) \right) \\ x_2 \left(\sum_{i=1}^l (x_i \delta_i) \right) \\ \vdots \\ x_l \left(\sum_{i=1}^l (x_i \delta_i) \right) \end{bmatrix} = (\delta^T x) \cdot x.$$

So we proved that $(\delta^T x) \cdot x = (x \cdot x^T) \delta$.

Exercise 5

a) Verify the following identities

$$X^T X = \sum_{u=1}^N X_u X_u^T \text{ and } X^T y = \sum_{u=1}^N y_u X_u$$

i) $X^T X = \sum_{u=1}^N X_u X_u^T$:

$$X = \begin{bmatrix} X_1^T \\ X_2^T \\ \vdots \\ X_N^T \end{bmatrix} = \begin{bmatrix} X_{11} & X_{12} & \cdots & X_{1l} \\ X_{21} & X_{22} & \cdots & X_{2l} \\ \vdots & \vdots & \ddots & \vdots \\ X_{N1} & X_{N2} & \cdots & X_{Nl} \end{bmatrix}$$

$$X^T X = \begin{bmatrix} \sum_{u=1}^N X_u^2 & \sum_{u=1}^N X_{u1} X_{u2} & \cdots & \sum_{u=1}^N X_{u1} X_{ul} \\ \sum_{u=1}^N X_{u2} X_{u1} & \sum_{u=1}^N X_{u2}^2 & \cdots & \sum_{u=1}^N X_{u2} X_{ul} \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{u=1}^N X_{ul} X_{u1} & \sum_{u=1}^N X_{ul} X_{u2} & \cdots & \sum_{u=1}^N X_{ul}^2 \end{bmatrix}$$

$$\text{So, } X^T X = \sum_{u=1}^N X_u X_u^T$$

ii) $X^T y = \sum_{u=1}^N y_u X_u$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_N \end{bmatrix}, \quad X^T y = \begin{bmatrix} \sum_{u=1}^N X_{u1} y_u \\ \sum_{u=1}^N X_{u2} y_u \\ \vdots \\ \sum_{u=1}^N X_{ul} y_u \end{bmatrix} = \sum_{u=1}^N y_u X_u$$

b) What is the size of $X, y, X^T X$ and $X^T y$

- 1) Size of X is $N \times l$ (samples:rows, features:columns)
- 2) Size of y is $N \times 1$
- 3) Size of $X^T X$ is $l \times l$ ($X^T: l \times N, X: N \times l$)
- 4) Size of $X^T y$ is $l \times 1$

c) Assume that a column vector of 1's is added in front of the column $\begin{bmatrix} X_{11} \\ X_{21} \\ \vdots \\ X_{N1} \end{bmatrix}$ of matrix X

i) What will be the changes in the dimensionality of the quantities in (b)?

By adding a column vector of 1's in matrix X we get the following changes in sizes:

- Matrix X has new size of $l+1$ columns therefore size of X is now $N \times (l+1)$.
- Vector y remains same.
- Matrix $X^T X$ is the product of the new X^T with dimensions $(l+1) \times N$ and the new X with $(N \times l+1)$. So the $X^T X$ will have $(l+1) \times (l+1)$ dimensions.
- Vector $X^T y$ will be $(l+1) \times 1$ as the columns were increased by 1.

ii) Do the identities given in (a) still hold?

$$1. X^T X = \sum_{u=1}^N X_u X_u^T$$

The matrix X' is: $X' = \begin{bmatrix} 1 & x_{11} & x_{12} & \dots & x_{1l} \\ 1 & x_{21} & x_{22} & \dots & x_{2l} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{N1} & x_{N2} & \dots & x_{Nl} \end{bmatrix}$

$$X'^T = \begin{bmatrix} 1 & 1 & \dots & 1 \\ x_{11} & x_{21} & \dots & x_{N1} \\ x_{12} & x_{22} & \dots & x_{N2} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1l} & x_{2l} & \dots & x_{Nl} \end{bmatrix}$$

$$X'^T X = \begin{bmatrix} 1 & 1 & \dots & 1 \\ x_{11} & x_{21} & \dots & x_{N1} \\ x_{12} & x_{22} & \dots & x_{N2} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1l} & x_{2l} & \dots & x_{Nl} \end{bmatrix} \cdot \begin{bmatrix} 1 & x_{11} & x_{12} & \dots & x_{1l} \\ 1 & x_{21} & x_{22} & \dots & x_{2l} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{N1} & x_{N2} & \dots & x_{Nl} \end{bmatrix}$$

$$X'^T X' = \begin{bmatrix} N & \sum_{u=1}^N x_{u1} & \dots & \sum_{u=1}^N x_{ul} \\ \sum_{u=1}^N x_{u1} & \sum_{u=1}^N x_{u1}^2 & \dots & \sum_{n=1}^N x_{n1} x_{nl} \\ \sum_{u=1}^N x_{u2} & \sum_{u=1}^N x_{u1} x_{u2} & \dots & \sum_{n=1}^N x_{n2} x_{nl} \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{u=1}^N x_{ul} & \sum_{n=1}^N x_{n1} x_{nl} & \dots & \sum_{n=1}^N x_{nl}^2 \end{bmatrix}$$

$$x^T u = \begin{bmatrix} 1 \\ x_{u1} \\ x_{u2} \\ \vdots \\ x_{up} \end{bmatrix}, \quad x_n^T x_n^T = \begin{bmatrix} 1 & x_{u1} & x_{u2} & \cdots & x_{up} \\ x_{u1} & x_{u1}^2 & x_{u1}x_{u2} & \cdots & x_{u1}x_{up} \\ x_{u2} & x_{u1}x_{u2} & x_{u2}^2 & \cdots & x_{u2}x_{up} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_{up} & x_{u1}x_{up} & x_{u2}x_{up} & \cdots & x_{up}^2 \end{bmatrix}$$

$$\sum_{u=1}^N x_u^T x_n^T = \begin{bmatrix} N & \sum_{n=1}^N x_{u1} & \cdots & \sum_{n=1}^N x_{up} \\ \sum_{u=1}^N x_{u1} & \sum_{n=1}^N x_{u1}^2 & \cdots & \sum_{n=1}^N x_{u1}x_{up} \\ \sum_{u=1}^N x_{u2} & \sum_{n=1}^N x_{u1}x_{u2} & \cdots & \sum_{n=1}^N x_{u2}x_{up} \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{u=1}^N x_{up} & \sum_{n=1}^N x_{u1}x_{up} & \cdots & \sum_{n=1}^N x_{up}^2 \end{bmatrix}$$

So it still holds.

$$2) x^{1T} y = \sum_{u=1}^N y_u x_u^1$$

$$x^{1T} = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_{11} & x_{21} & \cdots & x_{N1} \\ x_{12} & x_{22} & \cdots & x_{N2} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1p} & x_{2p} & \cdots & x_{Np} \end{bmatrix}, \quad x^{1T} y = \begin{bmatrix} \sum_{u=1}^N y_u \\ \sum_{u=1}^N x_{u1} y_u \\ \sum_{u=1}^N x_{u2} y_u \\ \vdots \\ \sum_{u=1}^N x_{up} y_u \end{bmatrix}$$

$$x_u^1 = \begin{bmatrix} 1 \\ x_{u1} \\ x_{u2} \\ \vdots \\ x_{up} \end{bmatrix} \Rightarrow \sum_{u=1}^N y_u x_u^1 = \sum_{u=1}^N y_u \begin{bmatrix} 1 \\ x_{u1} \\ x_{u2} \\ \vdots \\ x_{up} \end{bmatrix} = \begin{bmatrix} \sum_{u=1}^N y_u \\ \sum_{u=1}^N y_u x_{u1} \\ \vdots \\ \sum_{u=1}^N y_u x_{up} \end{bmatrix}$$

$$\text{So } x^{1T} y = \sum_{u=1}^N y_u x_u^1.$$

Exercise 6

A body moves on straight line and performs a smoothly accelerating motion. In the following table is given the velocity at certain time instances.

t (sec)	1	2	3	4	5
v (m/sec)	5.1	6.8	9.2	10.9	13.1

By using the given formula $v = v_0 + at$ we get that v = velocity, v_0 = initial velocity, acceleration is a and t is time.

So we get the matrix X by adding one column with 1's for v_0 and t values in second column.

The vector y is for velocity instances.

$$X = \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \\ 1 & 5 \end{bmatrix}, \quad y = \begin{bmatrix} 5.1 \\ 6.8 \\ 9.2 \\ 10.9 \\ 13.1 \end{bmatrix}$$

The Least Squares (LS) estimator is

$$\hat{\theta} = (X^T X)^{-1} X^T y$$

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

$$X^T \cdot y = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 & 5 \end{bmatrix} \begin{bmatrix} 5.1 \\ 6.8 \\ 9.2 \\ 10.9 \\ 13.1 \end{bmatrix} = \begin{bmatrix} 45.1 \\ 155.4 \end{bmatrix}$$

$$(X^T \cdot X)^{-1} : (X^T \cdot X)^{-1} = \frac{1}{\det(X^T \cdot X)} \cdot \text{adj}(X^T \cdot X)$$

$$\det(X^T \cdot X) = 5 \cdot 55 - 15 \cdot 15 = 275 - 225 = 50.$$

$$\text{adj}(X^T \cdot X) = \begin{bmatrix} 55 & -15 \\ -15 & 5 \end{bmatrix}$$

$$(X^T \cdot X^{-1}) = \frac{1}{50} \begin{bmatrix} 55 & -15 \\ -15 & 5 \end{bmatrix} = \begin{bmatrix} 1.1 & -0.3 \\ -0.3 & 0.1 \end{bmatrix}$$

$$\text{So } \mathcal{D} = (X^T \cdot X)^{-1} \cdot X^T y$$

$$= \begin{bmatrix} 1.1 & -0.3 \\ -0.3 & 0.1 \end{bmatrix} \cdot \begin{bmatrix} 45.1 \\ 155.4 \end{bmatrix} = \begin{bmatrix} 2.99 \\ 2.01 \end{bmatrix}$$

$$\text{So } \mathcal{D} = \begin{bmatrix} v_0 \\ \alpha \end{bmatrix} = \begin{bmatrix} 2.99 \\ 2.01 \end{bmatrix}$$

Therefore $v_0 = 2.99 \text{ m/s}$
 $\alpha = 2.01 \text{ m/s}^2$

b) Equation of velocity of the body as function of t.

$$v(t) = v_0 + \alpha t$$

$$v(t) = 2.99 + 2.01 t.$$

c) Estimate the velocity of the body at $t = 2.3$

$$v(2.3) = 2.99 + 2.01 \cdot 2.3 = 7.61 \text{ m/s}$$