

AI 669: Machine Learning

Assignment 0

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# I. Linear Algebra

1.

$$a) \begin{bmatrix} 1 & 2 \\ 4 & 5 \\ 7 & 8 \end{bmatrix}_{[3 \times 2]} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix}_{[3 \times 3]}$$

The dimension of the first matrix is  $3 \times 2$

The dimension of the second matrix is  $3 \times 3$

Hence, we cannot compute this matrix product

$$b) \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}_{[3 \times 3]} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix}_{[3 \times 3]} = \begin{bmatrix} 4 & 3 & 5 \\ 10 & 9 & 11 \\ 16 & 15 & 17 \end{bmatrix}_{[3 \times 3]}$$

c)

$$\begin{bmatrix} 1 & 2 & 1 & 2 \\ 4 & 1 & -1 & -4 \end{bmatrix}_{[2 \times 4]} \begin{bmatrix} 0 & 3 \\ 1 & -1 \\ 2 & 1 \\ 5 & 2 \end{bmatrix}_{[4 \times 2]} = \begin{bmatrix} 14 & 6 \\ -21 & 2 \end{bmatrix}_{[2 \times 2]}$$

$$d) \quad \underline{x} = [1 \ 0 \ 1]^T = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

$$\bullet \quad |\underline{x}|_2 = \sqrt{1^2 + 0^2 + 1^2} = \sqrt{2}$$

$$\bullet \quad \underline{x}^T \underline{x} = [1 \ 0 \ 1] \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = (1)(1) + (0)(0) + (1)(1) \\ = 2$$

$$\bullet \quad \underline{x} \underline{x}^T = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} [1 \ 0 \ 1] = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix}$$

2.

$$\begin{cases} x_1 + x_2 - x_3 - x_4 = 1 \\ 2x_1 + 5x_2 - 7x_3 - 5x_4 = -2 \\ 2x_1 - x_2 + x_3 + 3x_4 = 4 \\ 5x_1 + 2x_2 - 4x_3 - 2x_4 = 6 \end{cases}$$

$$a) \quad \underline{\tilde{A}} \underline{x} = \underline{b}$$

$$\Rightarrow \begin{bmatrix} 1 & 1 & -1 & -1 \\ 2 & 5 & -7 & -5 \\ 2 & -1 & 1 & 3 \\ 5 & 2 & -4 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 4 \\ 6 \end{bmatrix}$$

$$\Rightarrow \underline{\underline{A}} = \begin{bmatrix} 1 & 1 & -1 & -1 \\ 2 & 5 & -7 & -5 \\ 2 & -1 & 1 & 3 \\ 5 & 2 & -4 & -2 \end{bmatrix}; \quad \underline{\underline{x}} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}; \quad \underline{\underline{b}} = \begin{bmatrix} 1 \\ -2 \\ 4 \\ 6 \end{bmatrix}$$

b)

$$\underline{\underline{A}} \underline{\underline{x}} = \underline{\underline{b}}$$

Since  $\underline{\underline{A}}$  is invertible matrix, we have

$$\underline{\underline{x}} = \underline{\underline{A}}^{-1} \underline{\underline{b}} = \begin{bmatrix} 1 & 1 & -1 & -1 \\ 2 & 5 & -7 & -5 \\ 2 & -1 & 1 & 3 \\ 5 & 2 & -4 & -2 \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ -2 \\ 4 \\ 6 \end{bmatrix}$$

$$= \begin{bmatrix} 1.83 \\ -0.33 \\ 0.75 \\ -0.25 \end{bmatrix}$$

## II. Vector Calculus

1.

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

$$\Rightarrow f'(x) = \frac{-(1 + e^{-x})'}{(1 + e^{-x})^2} = \frac{-(e^{-x})(-x)'}{(1 + e^{-x})^2}$$

$$= \frac{e^{-x}}{(1 + e^{-x})^2}$$

$$b) \quad f(x) = e^{\left(-\frac{1}{2b^2}(x-\mu)^2\right)}$$

$$\Rightarrow f'(x) = e^{\left(-\frac{1}{2b^2}(x-\mu)^2\right)} \cdot \left(-\frac{1}{2b^2}(x-\mu)^2\right)'$$
$$= e^{\left(-\frac{1}{2b^2}(x-\mu)^2\right)} \cdot \left(-\frac{1}{2b^2} \cdot 2(x-\mu)\right)$$

$$= \frac{-(x-\mu)}{b^2} \cdot e^{\left(-\frac{1}{2b^2}(x-\mu)^2\right)}$$

2.

$$a) \quad f(z) = \log(1+z) \quad , \quad z = \underline{x}^T \underline{x} \quad , \quad \underline{x} \in \mathbb{R}^D$$

$$= \log(1 + \underline{x}^T \underline{x})$$

$$\Rightarrow \nabla_{\underline{x}} f = \frac{1}{1 + \underline{x}^T \underline{x}} \nabla_{\underline{x}} (1 + \underline{x}^T \underline{x}) = \frac{2 \underline{x}}{1 + \underline{x}^T \underline{x}}$$

$$b) \quad f(z) = e^{(-\frac{1}{2} z)}$$

$$z = g(\underline{y}) = \underline{y}^T \underline{S}^{-1} \underline{y} ;$$

$$\underline{y} = h(\underline{x}) = \underline{x} - \underline{\mu}$$

$$\begin{cases} \underline{x}, \underline{\mu} \in \mathbb{R}^D \\ \underline{S} \in \mathbb{R}^{D \times D} \\ \underline{S} \text{ is symmetric} \end{cases}$$

$$\Rightarrow f(z) = e^{(-\frac{1}{2} \cdot \underline{y}^T \underline{S}^{-1} \underline{y})}$$

$$= e^{(-\frac{1}{2} \cdot (\underline{x} - \underline{\mu})^T \underline{S}^{-1} (\underline{x} - \underline{\mu}))}$$

$$\Rightarrow \nabla_{\underline{x}} f = e^{(-\frac{1}{2} (\underline{x} - \underline{\mu})^T \underline{S}^{-1} (\underline{x} - \underline{\mu}))}$$

$$\cdot \nabla_{\underline{x}} \left( -\frac{1}{2} (\underline{x} - \underline{\mu})^T \underline{S}^{-1} (\underline{x} - \underline{\mu}) \right)$$

$$= e^{(-\frac{1}{2} (\underline{x} - \underline{\mu})^T \underline{S}^{-1} (\underline{x} - \underline{\mu}))}$$

$$\cdot \left( -\frac{1}{2} \right) 2 \underline{S}^{-1} (\underline{x} - \underline{\mu})$$



$$= - \Sigma^{-1} (\underline{x} - \underline{\mu}) e^{\left( -\frac{1}{2} (\underline{x} - \underline{\mu})^T \Sigma^{-1} (\underline{x} - \underline{\mu}) \right)}$$

### III. Probability

1.

a) The marginal distribution,

$$P(X) =$$

$$\begin{aligned} \bullet \quad P(X = x_1) &= P(X = x_1, Y = y_1) + P(X = x_1, Y = y_2) \\ &\quad + P(X = x_1, Y = y_3) \\ &= \sum_{j=1}^3 P(X = x_1, Y = y_j) = 0.01 + 0.05 + 0.1 \\ &= 0.16 \end{aligned}$$

Similarly,

$$P(X = x_2) = 0.02 + 0.1 + 0.05 = 0.17$$

$$P(X = x_3) = 0.03 + 0.05 + 0.03 = 0.11$$

$$P(X = x_4) = 0.1 + 0.07 + 0.05 = 0.22$$

$$P(X = x_5) = 0.1 + 0.2 + 0.04 = 0.34$$

$$\begin{aligned} \bullet \quad P(Y = y_1) &= \sum_{i=1}^5 P(X = x_i, Y = y_1) \\ &= 0.01 + 0.02 + 0.03 + 0.1 + 0.1 \\ &= 0.26 \end{aligned}$$

Similarly,

$$P(Y=y_2) = 0.05 + 0.1 + 0.05 + 0.07 + 0.2 = 0.47$$

$$P(Y=y_3) = 0.1 + 0.05 + 0.03 + 0.05 + 0.04 = 0.27$$

b) The conditional distribution,

$$\begin{aligned} P(X=x_1 | Y=y_1) &= P(X=x_1 | Y=y_1) + P(X=x_2 | Y=y_1) \\ &\quad + P(X=x_3 | Y=y_1) + P(X=x_4 | Y=y_1) \\ &\quad + P(X=x_5 | Y=y_1) \end{aligned}$$

$$\begin{aligned} &= \frac{P(X=x_1, Y=y_1)}{P(Y=y_1)} + \frac{P(X=x_2, Y=y_1)}{P(Y=y_1)} + \frac{P(X=x_3, Y=y_1)}{P(Y=y_1)} \\ &\quad + \frac{P(X=x_4, Y=y_1)}{P(Y=y_1)} + \frac{P(X=x_5, Y=y_1)}{P(Y=y_1)} \end{aligned}$$

$$= \frac{1}{0.26} (0.01 + 0.02 + 0.03 + 0.1 + 0.1)$$

$$= 1.$$

$$P(Y=y_j | X=x_3) = \frac{\sum_{j=1}^3 P(Y=y_j, X=x_3)}{P(X=x_3)} = 1.$$



2.

First coin is fair  $\left\{ \begin{array}{l} P_{1, \text{head}} = \frac{1}{2} \\ P_{1, \text{tail}} = \frac{1}{2} \end{array} \right.$

Second coin is biased  $\left\{ \begin{array}{l} P_{2, \text{head}} = \frac{1}{10} \\ P_{2, \text{tail}} = \frac{9}{10} \end{array} \right.$

- a) The probability that we picked the fair coin is  $\frac{1}{2}$   
b) The probability of the first toss being head is

$$\left(\frac{1}{2}\right)\left(\frac{1}{2}\right) + \left(\frac{1}{2}\right)\left(\frac{1}{10}\right) = 0.3$$

c)

We have,

$$P(\text{fair coin} \mid \text{both tosses are heads}) = \frac{P(\text{both tosses are heads} \mid \text{fair coin}) \cdot P(\text{fair coin})}{P(\text{both tosses are heads})}$$

$$P(\text{fair coin}) = \frac{1}{2}$$

$$P(\text{both toss are heads} \mid \text{fair coin}) = \left(\frac{1}{2}\right)\left(\frac{1}{2}\right) = \left(\frac{1}{4}\right)$$

$$P(\text{both toss are heads}) = \left(\frac{1}{2}\right)\left(\frac{1}{4}\right) + \left(\frac{1}{2}\right)\left(\frac{1}{100}\right) = 0.13$$

$$\Rightarrow P(\text{fair coin} \mid \text{both tosses are heads}) = \frac{(1/4)(1/2)}{0.13} = 0.9615$$

d)

• Similar to section c, we have

$$\begin{aligned} & P(\text{biased coin} \mid \text{both tosses are heads}) \\ &= \frac{P(\text{both tosses are heads} \mid \text{biased coin}) P(\text{biased coin})}{P(\text{both tosses are heads})} \\ &= \frac{\left(\frac{1}{10}\right) \left(\frac{1}{10}\right) \cdot \left(\frac{1}{2}\right)}{0.13} = 0.0385 \end{aligned}$$

• If both tosses are heads, the probability that the third coin toss is head, is

$$\begin{aligned} & P(\text{fair coin} \mid \text{both tosses are heads}) (112) \\ &+ P(\text{biased coin} \mid \text{both tosses are heads}) (1110) \\ &= (0.9615) \cdot (112) + (0.0385) (1110) = 0.4846 \end{aligned}$$