

## COMP441/541 Deep Learning, Fall 2025

### SELF-ASSESSMENT QUIZ (THEORY)

Due Date: 23:59 Wednesday, October 8, 2025

*Each student enrolled to COMP441/541 must complete this quiz on prerequisite math knowledge. The purpose is to self-assess whether you have the right background for the course. The topics covered in this problem set are very crucial so if you are having trouble with solving a problem, this indicates that you should spend a considerable amount of time to study that topic in its entirety.*

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#### Points and Vectors

- Given two vectors  $x = [a_1, a_2, a_3]$  and  $y = [a_1, -a_2, a_3]$ . Write down the equation for calculating the angle between  $x$  and  $y$ . When is  $x$  orthogonal to  $y$ ?

#### Planes

- Consider a hyperplane described by the  $d$ -dimensional normal vector  $[\theta_1, \dots, \theta_d]$  and offset  $\theta_0$ . Derive the equation for the *signed distance* of a point  $x$  from the hyperplane, which is defined as the perpendicular distance between  $x$  and the hyperplane, multiplied by +1 if  $x$  lies on the same side of the plane as the vector  $\theta$  points and by -1 if  $x$  lies on the opposite side  $x$  from the hyperplane.

#### Matrices

- Suppose that  $A^T(AB - C) = 0$ , where  $0$  is an  $m \times 1$  vector of zeros, derive an expression for  $B$ . Assume that all relevant matrices needed for this calculation are invertible.
- Find the eigenvalues and eigenvectors of the matrix  $A = \begin{bmatrix} 13 & 5 \\ 2 & 4 \end{bmatrix}$ .

#### Probability

- Let

$$p(X_1 = x_1) = \alpha_1 e^{-\frac{(x_1 - \mu_1)^2}{2\sigma_1^2}}$$

$$p(X_2 = x_2 | X_1 = x_1) = \alpha e^{-\frac{(x_2 - x_1)^2}{2\sigma^2}}$$

where  $X_1$  and  $X_2$  are continuous random variables. Show that

$$p(X_2 = x_2) = \alpha_2 e^{-\frac{(x_2 - \mu_2)^2}{2\sigma_2^2}}$$

by explicitly calculating the values of  $\alpha_2$ ,  $\mu_2$  and  $\sigma_2$ .

#### MLE and MAP

- Let  $p$  be the probability of landing head of a coin. You flip the coin 3 times and note that it landed 2 times on tails and 1 time on heads. Suppose  $p$  can only take two values: 0.3 or 0.6. Find the Maximum Likelihood Estimate of  $p$  over the set of possible values  $\{0.3, 0.6\}$
- Suppose that you have the following prior on the parameter  $p$ :  $P(p = 0.3) = 0.3$  and  $P(p = 0.6) = 0.7$ . Given that you flipped the coin 3 times with the observations described above, find the MAP estimate of  $p$  over the set  $\{0.3, 0.6\}$ , using the prior.

**Optimization**

Gradient ascent/descent methods are typical tools for maximizing/minimizing functions. Let  $L(x, \theta)$  be a function of two vector arguments,  $x = [x_1, x_2]^T$  and  $\theta = [\theta_1, \theta_2]^T$ . We would like to find the optimum value of vector  $\theta$  which maximizes/minimizes  $L(x, \theta)$  where  $x$  is assumed to be given.

The gradient  $\nabla_{\theta} L(x, \theta)$  is a vector with two components corresponding to partial derivatives

$$\frac{\partial}{\partial \theta_j} L(x, \theta), \quad j = 1, 2$$

8. Evaluate the gradient when  $L(x, \theta) = \log(1 + \exp(-\theta^T x))$ .
9. Into which direction does the gradient (viewed as a vector) point? Is the value of  $L(x, \theta)$  larger or smaller if we evaluate it at  $\theta' = \theta + \epsilon \nabla_{\theta} L(x, \theta)$  where  $\epsilon$  is a small real number?