

CSCI 6751 V1 | Artificial Intelligence

Midterm Examination

Oct 14, 2025

Total 50 points

Time: 15 minutes

GOOD LUCK

Group 2

Student Name & ID _____

MCQ: Select only the most appropriate option.

Question 1. (5 points)

A model is trained with Gradient Descent. The cost function decreases very slowly and erratically, requiring a huge number of iterations. The learning rate is suspect. What is the most likely scenario?

- a) The learning rate is too high; it should be decreased to prevent overshooting.
- b) The learning rate is too low; it should be increased to navigate flat regions and saddle points more effectively.
- c) The learning rate is optimal; the problem is a non-convex cost function.
- d) The model has converged; no action is needed.

Question 2. (5 points)

When using an iterative algorithm like Gradient Descent, stopping the training process before full convergence can act as a form of regularization. This is because:

- a) It explicitly adds a penalty term to the loss function, similar to L2 regularization.
- b) It limits the effective complexity of the model by preventing it from perfectly fitting the training noise, thus reducing variance.
- c) It ensures the model remains linear, thus increasing bias but improving interpretability.
- d) It forces the parameters to be exactly zero, mimicking L1 regularization.

Question 3. (5 points)

In the multivariate linear regression normal equation $\theta = (X^T X)^{-1} X^T Y$, if the design matrix X has dimensions $m \times (n+1)$ (m examples, n features plus intercept), and Y is $m \times 1$, what are the dimensions of the resulting parameter vector θ ?

- a) $m * 1$ b) $(n+1) * 1$ c) $m * (n+1)$ d) $(n+1) * m$

Question 4. (5 points)

A model has high error on both training and test data. Increasing model complexity reduces training error to near zero, but test error remains high. This sequence describes the transition from a model suffering primarily from _____ to one suffering primarily from _____.

- a) High variance; High bias b) High bias; High variance
c) Underfitting; Optimal fitting d) High bias; Low bias

Question 5. (5 points)

Given a rule "If X is A and Y is B then Z is C," with input membership values $\mu_A(x)=0.7$ and $\mu_B(y)=0.8$, the firing strength is:

- a) 1.5 b) 0.56 c) 0.7 d) 0.8

Question 6. (5 points)

The expected prediction error for a new test point x_0 can be decomposed as $E[(y_0 - \hat{y}_0)^2] = \text{Var}(\hat{y}_0) + [\text{Bias}(\hat{y}_0)]^2 + \text{Var}(\varepsilon)$.

If you take a model with high variance and add a very strong L2 regularization penalty (large λ in Ridge), what is the primary effect on this decomposition?

- a) Variance increases and Bias decreases.
b) Variance decreases and Bias increases.
c) Both Variance and Bias decrease.
d) The irreducible error, $\text{Var}(\varepsilon)$, is reduced.

Question 7. (5 points)

$$A = \{(x_1, 0.4), (x_2, 0.7)\}$$

$$B = \{(x_1, 0.8), (x_2, 0.2)\}$$

What is the membership value for element x_1 in the union $A \cup B$, defined as $\mu_{A \cup B}(x) = \max[\mu_A(x), \mu_B(x)]$?

- a) 0.4 b) 0.8 c) 0.32 d) 1.2

Question 8. (5 points)

Given a rule "If X is A and Y is B then Z is C," with input membership values $\mu_A(x)=0.7$ and $\mu_B(y)=0.8$, the firing strength is:

Question 9. (5 points)

What is the key difference between the "support" and the "core" of a fuzzy set?

- a) The support has $\mu(x) > 0$, the core has $\mu(x) = 1$.
 - b) The support has $\mu(x) = 1$, the core has $\mu(x) > 0$.
 - c) The support is always a subset of the core.
 - d) They are two terms for the same concept.

Question 10. (5 points)

During Gradient Descent, if the learning rate is set too high, what is a likely consequence?

- a) The algorithm will converge too slowly.
 - b) The algorithm may overshoot the minimum.**
 - c) The algorithm is guaranteed to find the global minimum.
 - d) The model will inevitably underfit.