Name:		
ranic.		

Roll Number:

Quiz-3

Max. Time: 20 mins Max. Marks: 20

Note: Solve all parts. Limit your written responses to the provided space.

- Q.1. $[5 \times 1 = 5]$ Choose by putting a check mark on the most appropriate option or encircling it. Note: No cutting/overwriting is allowed.
- 1. Fitting a given dataset with a higher degree polynomial is always a good strategy for regression if noise is suspected in the data.
- (A) True (B) False
- 2. Logistic regression for binary classification can only yield linear decision boundaries in the feature space.
- (A) True (B) False
- 3. Stochastic gradient descent is more efficient for very large datasets compared to the batch gradient descent.
- (A) True (B) False
- 4. Mean squared error can be used as a hypothesis for linear regression.
- (A) True (B) False
- 5. Loss function for logistic regression measures the misclassification penalty.
- (A) True (B) False

Q.2. [3+4+3+5]

a) Given a training dataset \mathcal{D} with the feature matrix $\mathcal{X} \in \mathbb{R}^{n \times d}$ and a target vector $\mathcal{Y} \in \mathbb{R}^{n \times 1}$, write the expression for optimal parameters to predict the target value using closed form (normal equations) solution.

$$\Theta = (\mathcal{X}^T \mathcal{X})^{-1} \mathcal{X}^T \mathcal{Y}$$

b) Compute the cost function when the following training vector is presented to the LMS algorithm for linear regression.

$$< \mathbf{x}, \mathbf{t} \ge < [2.5 \ 1.6 \ 2.0 \ -1.4]^T, 5 >$$

Assume the parameter vector is: $\mathbf{\theta} = [0.02 - 1.2 \ 0.05 \ 0.001 \ 0.3]^T$

Solution:

$$h_{\theta}(\mathbf{x}) = \theta^T \mathbf{x} = -1.4714$$

$$\therefore J(\theta) = 0.5 \times (5 - (-1.4714))^2 = 20.94$$

Name:	Roll Number:

c) Why is the mean squared error cost function a bad choice for logistic regression?

<mark>Ans</mark>.

This is because the sigmoid (logistic) function is non-linear in the parameters of the model. The mean squared error cost function based on this hypothesis (i.e., logistic function) becomes non-convex and, therefore, would be a bad choice to find optimal parameter values with gradient descent learning.

d) Write down the cost function and the parameter update rule for logistic regression.

Cost function:

$$\ell(\theta) = \sum_{i=1}^{m} y^{(i)} \log h(x^{(i)}) + (1 - y^{(i)}) \log(1 - h(x^{(i)}))$$

Parameter update rule (using stochastic gradient descent):

$$\theta_j := \theta_j + \alpha \left(y^{(i)} - h_{\theta}(x^{(i)}) \right) x_j^{(i)}$$
 (for every j)

1

Name:	Roll Number: