Linear Regression

- 1. What are the assumptions of linear regression?
- 2. What is the functional form of the loss function that is minimized in linear regression?
- 3. When does the ordinary least square estimate of parameters become equal to the maximum likelihood estimate of parameters in linear regression?
- 4. Given a set of predictions from linear regression model and the ground truth as follows, can you find the value of the loss function

n_observation	Predicted	Actual
1	10	9
2	12	12
3	16	18
4	21	19
5	27	29
6	33	28
7	45	43
8	54	56

- 5. What is meant by the p-value of a distribution? Should you always take those coefficients whose p-values are less than 0.05?
- 6. In a linear regression output, you get a table as follows.

coefficient	mean	std deviation	p-value
pressure	0.01	0.002	0.02
temperature	1.5	0.13	0.017
humidity	2.5	1.91	0.23
wind_speed	0.2	0.17	0.34

- a. Which distribution does the p-values depicted above correspond to?
- b. Does the data need to be scaled before fitting a linear regression model? Justify
- c. What checks do you need to perform after fitting the linear regression model?
- 7. What is meant by R-squared value of a model? How is it different from adjusted R-squared?
- 8. Define AIC, BIC.
 - a. How are these used to select the best models?
 - b. Why are these quantities needed?
 - c. How can you calculate the AIC or BIC of a decision tree?

Logistic Regression

- 1. Which probability distribution can be used to model a random variable that has only two outcomes? How is the probability distribution parameterized?
- 2. What do you mean by odds of an event? What is log-odds?

- 3. Which assumptions of linear regression are violated if we want to model a binary classification problem with it?
- 4. What is the sigmoid function? Mention its parameters
- 5. What is the shape of the decision boundary for a logistic regression?
- 6. What is the loss function for logistic regression?
- 7. Can you use regularization in logistic regression? How does the loss function change in that case?
- 8. How can you solve a multiclass classification problem using logistic regression?

Decision Trees

- 1. What is entropy? What is information gain in context of decision trees? What is Gini impurity?
- 2. What are the different types of algorithms that are used to train decision trees?
- 3. Briefly explain any one decision tree algorithm in the context of a classification problem
- 4. How does the decision tree treat categorical variables?
- 5. What is inductive bias in the context of different algorithms for fitting decision trees?
- 6. Why are decision trees not affected by outliers?
- 7. Why is pruning necessary for decision trees?
- 8. Is feature scaling necessary for decision trees?
- 9. What are the limitations of decision trees?

Random Forest

- 1. Why do we need to have random forests on top of decision trees?
- 2. What are the different ensemble methods that random forest uses to improve upon the decision tree algorithm?
- 3. What do you mean by bagging? What is feature subsampling?
- 4. Prove that in bagging, nearly 2/3rd of the original dataset appears in the training bagged samples
- 5. What are the hyperparameters that you should tune for while fitting RF? Are there any heuristics to choosing them?

Clustering

- 1. Describe the algorithm of KMeans. How do you determine k?
- 2. What is the loss function for KMeans?
- 3. What are the drawbacks of KMeans? What are the workarounds?
- 4. Describe the algorithm of Hierarchical Agglomerative Clustering. Can categorical and continuous variables both be given as input? What about feature scaling?
- 5. What disadvantages does DBSCAN and GMM overcome with respect to clustering as compared to KMeans?