

## Flight Delay

1. Professor Norouzi is building a flight delay linear regression model using a dataset of historical flight information. Suppose the flight information dataset contains 400 flight records and 10 columns.

Of the 10 columns, 1 contains the observed flight delays in minutes, and 9 columns have non-constant numerical features related to the flight (quantitative variables).

Assume that we find the least squares estimate for parameter vector  $\theta$  (with intercept) as  $\hat{\theta} = (\mathbb{X}^T \mathbb{X})^{-1} \mathbb{X}^T \mathbb{Y}$  for a design matrix  $\mathbb{X}$  and true output vector  $\mathbb{Y}$ .

- (a) What is the shape of the design matrix  $\mathbb{X}$ ?

- (b) What is the shape of the parameter vector  $\theta$ ?

- (c) What is the shape of the true output vector  $\mathbb{Y}$ ?

- (d) What is the shape of the predicted output vector  $\hat{\mathbb{Y}}$ ?

- (e) What is the residual vector? Write down the equation for calculating the residual vector using  $\mathbb{Y}$ ,  $\mathbb{X}$ , and  $\hat{\theta}$ .

- (f) What is always true about the residuals in any Ordinary Least Squares (OLS) regression? Select all that apply.
- ☐ A. They are orthogonal to the column space of the design matrix ( $\mathbb{X}$ ).
  - ☐ B. Their sum is equal to the mean squared error.
  - ☐ C. Their sum is equal to zero.
  - ☐ D. They are orthogonal to the vector of predictions ( $\hat{\mathbb{Y}}$ )

## Linear Regression Fundamentals

2. In this problem, we will review some of the core concepts of linear regression. Suppose we create a linear model with parameters  $\hat{\theta} = [\hat{\theta}_0, \dots, \hat{\theta}_p]$ . Given an observation  $\vec{x}$ , this model predicts  $\hat{y} = \hat{\theta} \cdot \vec{x} = \hat{\theta}_0 + \hat{\theta}_1 x_1 + \hat{\theta}_2 x_2 + \dots + \hat{\theta}_p x_p$ . The design matrix and observations used to construct  $\hat{\theta}$  are  $\mathbb{X}$  and  $\mathbb{Y}$ .

- (a) Suppose  $\hat{\theta} = [2, 0, 1]$  and we receive an observation  $\vec{x}_1 = [1, 2, 5]$ . What  $\hat{y}_1$  value will this model predict for the given observation?

- (b) Suppose the true  $y_1$  was 3. What will be the  $L_2$  loss for our prediction  $\hat{y}_1$  from the previous part?

- (c) Which of the following statements are always true? **Select all that apply**

- ☐ A.  $\vec{x}_1^T \hat{\theta} = 0$   
☐ B.  $\vec{x}_1^T \hat{\theta} = \hat{y}_1$   
☐ C.  $\vec{x}_1^T \hat{\theta} - y_1 = 0$   
☐ D.  $\mathbb{X} \hat{\theta} - \mathbb{Y} = 0$   
☐ E.  $\mathbb{X}^T (\mathbb{X} \hat{\theta} - \mathbb{Y}) = 0$

- (d) (T/F) Define the residuals of this model as  $e_i = y_i - \hat{y}_i$ . For all data points  $x_i$  and  $y_i$  in  $\mathbb{X}$ ,  $\mathbb{Y}$ , the sum of residuals  $\sum_i e_i = 0$ . Justify why.

- (e) Suppose we arbitrarily removed a feature from the design matrix. Which of the following could happen to the new optimal loss compared to the old optimal loss?

- ☐ A. The optimal loss must decrease.  
☐ B. The optimal loss may decrease, but it must not increase.  
☐ C. The optimal loss may increase, but it must not decrease.  
☐ D. The optimal loss must increase.