



8. Example: Assessing the
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8. Example: Assessing the performance of planes

Setup:

An aerospace manufacturing company would like to assess the performance of its existing planes for its latest design. Based on a sample size of $n = 1000$ flights, each with an identically designed plane, it collects data of the form $(x_1, y_1), \dots, (x_{1000}, y_{1000})$, where x represents the distance traveled and y represents liters of fuel consumed.

You, as a statistician hired by the company, decide to perform linear regression on the model $y = a + bx$ to predict the efficiency of the design. In the context of linear regression, recall that the mathematical model calls for:

$$\mathbf{Y} = \begin{pmatrix} y_1 \\ \vdots \\ y_{1000} \end{pmatrix} \in \mathbb{R}^{1000}, \quad \boldsymbol{\varepsilon} \in \mathbb{R}^{1000}, \quad \mathbf{X} = \begin{pmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_{1000} \end{pmatrix} \in \mathbb{R}^{1000 \times 2}, \quad \boldsymbol{\beta} = \begin{pmatrix} a \\ b \end{pmatrix} \in \mathbb{R}^2.$$

Assume that $\boldsymbol{\varepsilon} \sim \mathcal{N}(0, \sigma^2 I_{1000})$ for some fixed σ^2 , so that $\mathbf{Y} \sim \mathcal{N}(\mathbf{X}\boldsymbol{\beta}, \sigma^2 I_{1000})$.

Prediction using Regression

1/1 point (graded)

Using the setup as above, you compute the LSE, which comes out to

$$\hat{\beta} = \begin{pmatrix} \hat{a} \\ \hat{b} \end{pmatrix} = \begin{pmatrix} 0.8 \text{ liters} \\ 15.0 \text{ liters / km} \end{pmatrix}.$$

Just from $\hat{\beta}$, what is a reasonable prediction for the total amount of fuel a plane (in liters) consumes after 200 kilometers?

✓ Answer: 3000.8

Solution:

The LSE gives the “best-fitting” model $y = 0.8 + 15x$. Plugging in $x = 200$ gives $y = 3000.8$.

You have used 1 of 3 attempts

❗ Answers are displayed within the problem

The Maximum Likelihood Estimator

2/2 points (graded)

Using the same setup as the previous problem:

$$\hat{\beta} = \begin{pmatrix} \hat{a} \\ \hat{b} \end{pmatrix} = \begin{pmatrix} 0.8 \text{ liters} \\ 15.0 \text{ liters / km} \end{pmatrix}.$$

Using $n = 1000$ samples, by thinking of \mathbf{Y} as the vector of observations, we might also consider the Maximum Likelihood Estimator β_{MLE} . As a reminder, β_{MLE} maximizes, over all choices of β , the likelihood (or the log-likelihood) of $\mathbf{Y} \sim \mathcal{N}(\mathbb{X}\beta, \sigma^2 I_{1000})$.

Numerically, $\beta_{MLE} = \begin{pmatrix} a_{MLE} \\ b_{MLE} \end{pmatrix}$, where:

$a_{MLE} =$

✓ Answer: 0.8

 $b_{MLE} =$

✓ Answer: 15.0

Solution:

This scenario is the homoscedastic gaussian case, **so the MLE coincides with the LSE**. Therefore, $a_{MLE} = \hat{a} = 0.8$ and $b_{MLE} = \hat{b} = 15.0$. To see why, recall that $\mathcal{N}(\mathbb{X}\beta, \sigma^2 I_{1000})$ follows the density

$$f(\mathbf{Y}) = \frac{\exp\left(-\frac{1}{2}(\mathbf{Y} - \mathbb{X}\beta)^T (\sigma^2 I)^{-1} (\mathbf{Y} - \mathbb{X}\beta)\right)}{\sqrt{(2\pi)^{1000} \det(\sigma^2 I)}}$$

so the log-likelihood function can be written


$$\log f(\mathbf{Y}) = -500 \log(2\pi\sigma^2) - \frac{1}{2\sigma^2}(\mathbf{Y} - \mathbb{X}\beta)^T (\mathbf{Y} - \mathbb{X}\beta)$$

The vector β only appears in the second term, so maximizing this expression is the same as minimizing

$$(\mathbf{Y} - \mathbb{X}\beta)^T (\mathbf{Y} - \mathbb{X}\beta) = \|\mathbf{Y} - \mathbb{X}\beta\|^2$$

which, by definition, is attained by the least-squares estimator $\hat{\beta}$.

You have used 1 of 3 attempts

 Answers are displayed within the problem

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