

Lab 13

In this Lab:

- simple linear regression
- multiple linear regression

Simple Linear Regression

Problem: a pair of correlated rvs (X, Y)
we want to estimate Y based on
observation X .

Solution:

estimator: $\mathbb{E}[Y|X] = g(X)$
→ a function of X

where

$$g(u) = \mathbb{E}[Y|X=u] = \int v f_{Y|X}(v|u) dv$$

linear estimator:

$$\hat{\mathbb{E}}[Y|X] = L^*(X)$$

where

$$L^*(u) = \mu_y + \frac{\text{cov}(Y, X)}{\text{Var}(X)} (u - \mu_x)$$

Calculating μ_x , μ_y , $\text{var}(x)$ and $\text{cov}(y, x)$ requires knowing the distributions of X , Y and their joint distribution, which often is not available to us.

often we have some samples (x_1, y_1)

(x_2, y_2)

\vdots

(x_n, y_n)

$$\mu_x \approx \frac{x_1 + \dots + x_n}{n}$$

$$\mu_y \approx \frac{y_1 + \dots + y_n}{n}$$

$$\text{Var}(X) = \mathbb{E}[(X - \mu_x)^2] \approx \frac{(x_1 - \mu_x)^2 + \dots + (x_n - \mu_x)^2}{n}$$

$$\text{cov}(X, Y) = \mathbb{E}[(X - \mu_x)(Y - \mu_y)] \approx \frac{(x_1 - \mu_x)(y_1 - \mu_y) + \dots + (x_n - \mu_x)(y_n - \mu_y)}{n}$$

Another name: linear regression

(of y values based on x values)

Example 1

X_1 = mid-term 1 score y = final exam score

Example 2

X_2 = mid-term 2 score y = final exam score

Example 3

X_3 = quiz score y = final exam score

Think:

we can estimate y based on X_1 , or X_2 , or X_3 .
But would we be better off by estimating based on all X_1 , X_2 and X_3 ?

Multiple Linear Regression

Problem: same as simple linear regression,
but X is a vector instead of a scalar.
 \downarrow
 dim- n

Before:

$$\begin{aligned}\hat{E}[Y|X] &= E[Y] + \frac{\text{Cov}(Y, X)}{\text{Var}(X)} (X - E[X]) \\ &= E[Y] + \text{Cov}(Y, X) \text{Cov}(X, X)^{-1} (X - E[X])\end{aligned}$$

Now

$$\hat{E}[Y|X] = E[Y] + \underset{1 \times 1}{\text{Cov}(Y, X)} \underset{1 \times n}{\text{Cov}(X, X)^{-1}} \underset{n \times n}{(X - E[X])} \underset{n \times 1}{}$$

$$\text{Cov}(Y, X) = [\text{Cov}(Y, X_1), \dots, \text{Cov}(Y, X_n)]$$

$$\text{Cov}(X, X) = \begin{bmatrix} \text{Cov}(X_1, X_1) & \dots & \text{Cov}(X_1, X_n) \\ \text{Cov}(X_2, X_1) & \dots & \text{Cov}(X_2, X_n) \\ \vdots & & \\ \text{Cov}(X_n, X_1) & \dots & \text{Cov}(X_n, X_n) \end{bmatrix}$$

$\hat{E}[Y/X_1, X_2, X_3]$ should be in the form:

$$\alpha_1 X_1 + \alpha_2 X_2 + \alpha_3 X_3 + \beta$$