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# Probability and Statistics

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Linear regression Model



# Outline I



## Example I



Hooke's law states that the force applied to a spring is proportional to the distance that the spring is stretched. Thus, if  $F$  is the force applied and  $x$  is the distance that the spring has been stretched, then  $F = kx$ . The proportionality constant  $k$  is called the spring constant. Some physics students want to determine the spring constant for a given spring. They apply forces of 2, 5, and 7 pounds, which have the effect of stretching the spring 5, 8, and 10 inches, respectively.

$$5k = 2 \implies k = 0.4$$

$$8k = 5 \implies k = 0.625$$

$$10k = 7 \implies k = 0.7.$$

First guess:

$$k = \frac{0.4 + 0.625 + 0.7}{3} = \frac{1}{3} \cdot 0.4 + \frac{1}{3} \cdot 0.625 + \frac{1}{3} \cdot 0.7.$$

## Example II



First guess:

$$k = \frac{0.4 + 0.625 + 0.7}{3} = \frac{1}{3} \cdot 0.4 + \frac{1}{3} \cdot 0.625 + \frac{1}{3} \cdot 0.7.$$

Optimal value:

$$\begin{aligned} k &= \frac{5 \cdot 2 + 8 \cdot 5 + 10 \cdot 7}{5^2 + 8^2 + 10^2} \\ &= \frac{5^2 \cdot 0.4 + 8^2 \cdot 0.625 + 10^2 \cdot 0.7}{5^2 + 8^2 + 10^2} \\ &= \frac{25}{189} \cdot 0.4 + \frac{64}{189} \cdot 0.625 + \frac{100}{189} \cdot 0.7. \end{aligned}$$

Also

$$\begin{bmatrix} 5 & 8 & 10 \end{bmatrix} \begin{bmatrix} 5 \\ 8 \\ 10 \end{bmatrix} k = \begin{bmatrix} 5 & 8 & 10 \end{bmatrix} \begin{bmatrix} 2 \\ 5 \\ 7 \end{bmatrix}$$

Thus  $189k = 120$ .

Many problems in engineering and science involve exploring the relationships between two or more variables -> Regression analysis.

- In a chemical process: the yield of the product is related to the process-operating temperature.
- Regression analysis can be used to build a model to predict yield at a given temperature level.

- The simple linear regression considers a single regressor or predictor  $x$  and a dependent or response variable  $Y$ .
- The expected value of  $Y$  at each level of  $x$  is a random variable.

$$E(Y|x) = \alpha + \beta x$$

- We assume that each observation,  $Y$ , can be described by the model.

$$Y = \alpha + \beta x + \epsilon$$

That is

$$Y_1 = \alpha + \beta X_1 + \epsilon_1$$

$$Y_2 = \alpha + \beta X_2 + \epsilon_2$$

...

$$Y_n = \alpha + \beta X_n + \epsilon_n$$

The least-squares estimates of the intercept and slope in the simple linear regression model are

$$b = \frac{S_{xy}}{S_{xx}}$$

and

$$a = \bar{y} - b\bar{x}.$$

with  $S_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2$  and  $S_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$ , where

$$S_{xx} = \sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - \frac{1}{n} \left( \sum_{i=1}^n x_i \right)^2$$

$$S_{xy} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \sum_{i=1}^n x_i y_i - \frac{1}{n} \left( \sum_{i=1}^n x_i \right) \left( \sum_{i=1}^n y_i \right)$$

$$S_{yy} = \sum_{i=1}^n (y_i - \bar{y})^2 = \sum_{i=1}^n y_i^2 - \frac{1}{n} \left( \sum_{i=1}^n y_i \right)^2.$$

The fitted or estimated regression line is therefore

$$\hat{y}_i = a + bx_i$$

Note that each pair of observations satisfies the relationship

$$y_i = a + bx_i + \epsilon_i, \quad i = 1, \dots, n$$

where  $e_i = y_i - \hat{y}_i$  is called the residual. The residual describes the error in the fit of the model to the  $i$ -th observation  $y_i$ .





Estimate  $E(Y|X = x^*)$ :  $\hat{y}^* = a + bx^*$

The sum of squares of the residuals

$$SSE = \sum (y_i - \hat{y}_i)^2.$$

The total sum of squares of the response variable

$$SST = \sum (y_i - \bar{y})^2.$$

The sum of squares for regression

$$SSR = \sum (\hat{y}_i - \bar{y})^2$$

Fundamental identity

$$SST = SSE + SSR$$

Computational formula

$$SST = S_{yy}$$

$$SSR = bS_{xy}$$

$$SSE = S_{yy} - bS_{xy}$$

It can be shown that

$$E(SSE) = (n - 2)\sigma^2.$$

An unbiased estimator of  $\sigma^2$

$$s^2 = \frac{SSE}{n - 2}$$

Coefficient of determination

$$r^2 = 1 - \frac{SSE}{SST}.$$

- Slope Properties

$$E(b) = \beta, \quad V(b) = \frac{\sigma^2}{S_{xx}} \equiv \sigma_b^2$$

The estimated standard error of the slope

$$S_b = \sqrt{\frac{S^2}{S_{xx}}}.$$

Moreover

$$T = \frac{b - \beta}{S_b} \sim \mathbf{t}(n - 2).$$

- Intercept Properties

$$E(a) = \alpha, \quad V(a) = \frac{\sigma^2 \mu_{xx}}{S_{xx}} \equiv \sigma_a^2.$$

with  $\mu_{xx} = \frac{1}{n} \sum x_i^2$ . The estimated standard error of the slope

$$s_a = \sqrt{\frac{S^2 \mu_{xx}}{S_{xx}}} = \sqrt{S_b^2 \mu_{xx}}.$$

Moreover

$$T = \frac{a - \alpha}{s_a} \sim \mathbf{t}(n - 2).$$

# Confidence interval



Confidence interval for the slope

$$b \pm t_{v/2, n-2} \cdot S_b$$

Confidence interval for the intercept

$$a \pm t_{v/2, n-2} \cdot S_a$$

- Slope:  $T = \frac{b - \beta_0}{S_b}$  with

$$H_1 : \beta \neq \beta_0, \quad H_1 : \beta < \beta_0, \quad H_1 : \beta > \beta_0$$

- Intercept:  $T = \frac{a - \alpha_0}{S_a}$  with

$$H_1 : \alpha \neq \alpha_0, \quad H_1 : \alpha < \alpha_0, \quad H_1 : \alpha > \alpha_0$$

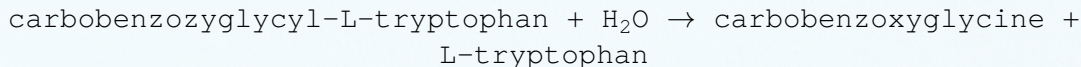
Observation Number	Hydrocarbon Level $x(\%)$	Purity $y(\%)$
1	0.99	90.01
2	1.02	89.05
3	1.15	91.43
4	1.29	93.74
5	1.46	96.73



# Michaelis-Menten Equation



The hydrolysis of carbobenzozyglycyl-L-tryptophan catalyzed by pancreatic carboxypeptidase occurs according to the reaction:



The following data on the rate of formation of L-tryptophan at 25°C and pH = 7.5 was obtained:

Substrate Concentration (mM)	Rate
2.5	0.026
5.0	0.037
10.0	0.054
15.0	0.061
20.0	0.065

- ① Plot the initial rate as a function of substrate concentration. What shape does the data follow? 95% confidence intervals for  $\beta_1$  and  $\beta_2$ , and find the  $r^2$  value for the fit. Are they what you expect?
- ② Now linearize the model using the Lineweaver-Burk method and solve for  $V_{max}$  and  $K_M$ . Find the 95% confidence intervals for the slope and intercept of your Lineweaver-Burk plot and determine the  $r^2$  value.
- ③ Make a residual plot to assess the fit from part d.