

Simple Linear Regression

- Useful for Predicting quantitative [Response](#)
- Predicting Y on the basis of a single predictor variable X
- Assuming that there is a linear relationship between X and Y

$$Y \approx \beta_0 + \beta_1 X$$

- This can be read as regressing Y on X
- or Y onto X

Example :

- TV ads $\rightarrow X$
- Sales $\rightarrow Y$
- Sales $\approx \beta_0 + \beta_1 TV$
 - β_0, β_1 two unknown constants
 - $\beta_0 \rightarrow$ Slop of X
 - $\beta_1 \rightarrow$ intercept of Y
 - They are called model **Coefficients** or **Parameters**

After Using Training data to estimate $\hat{\beta}_0, \hat{\beta}_1$

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$$

Estimating the Coefficients

- In practice β_0, β_1 are unknown, we usually use [Training Data](#) to estimate them : $(x_1, y_1), \dots, (x_n, y_n)$
- we try to estimate β_0, β_1 as close as possible to the data points so:
 - $y_i \approx \hat{\beta}_0 + \hat{\beta}_1 x_i$
- To minimize as much as possible \rightarrow we use Least squares criterion

Let $\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x_i \rightarrow e_i = y_i - \hat{y}_i$

- y_i observed response
- \hat{y}_i predicted response

With that we have [Residual Sum of Squares](#)