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Regression

a statistical tool for the investigation of relationships between variables.

- finding the relationship between dependent and independent variables

Regression Analysis

a form of predictive modeling which investigates the relationship between a dependent (target) and independent variable(s). independent variables are also known as predictors.

- indicates the strength of impact of multiple independent variables on a dependent variable

Simple Linear Regression

- measures how strong the relationship is between 2 variables
- uses a straight line (of best fit) to model the relationship

Nonlinear regression models

- representing the relationship with a curved line
Example:

$$Y = a + bx + cx^2$$

Examples of Linear Regression

- predicting house prices
 - given some dataset of bedroom numbers, square footage, location, etc.

- stock price prediction
 - using historical stock data, predict the future price of a stock
- employee performance prediction
 - given data on employee characteristics, predict their future job performance or productivity

Linear Regression Model

$$Y = a + bX$$

- X = independent variable
- Y = dependent variable

where...

- a - constant which equals the value of Y when X=0
- X - independent variable that is predicting Y
- Y - value of dependent variable that is being predicted
- b - slope of regression line u.e. how much y changes for each one unit change in X

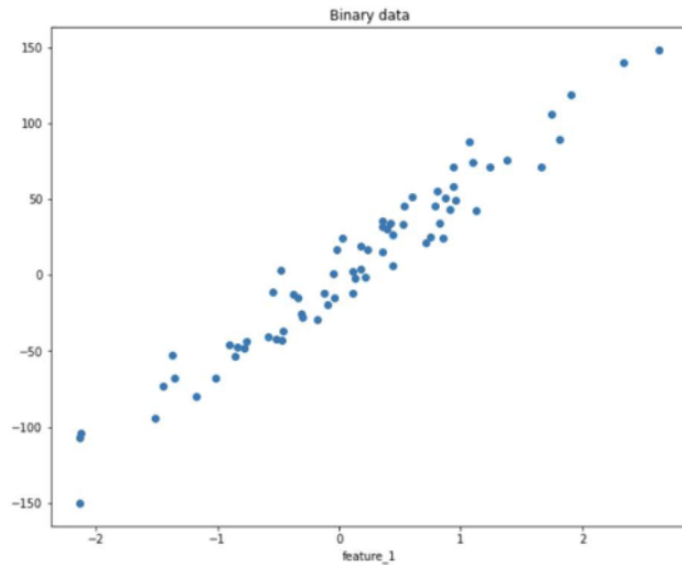
Line of Best Fit

a straight line that best represents the data on a scatter plot

- best linear approximation for some set of data

How to obtain best fit line (Value of a and b)?

- Ordinary Least Square
 - Find a model whose line fits best with respect to the given data



<https://nandeshwar.in/100-days-of-deep-learning/what-is-linear-regression-with-derivation/>

How to obtain best fit value

we define some function to represent the error between our prediction and the actual data. We then minimize this error function. Its common to use the ordinary least square method.

- Ordinary Least Square

- Find parameters which minimize the sum of the square of the distance (d) of the actual point from the model fit line $\hat{y}_i = \beta_0 + \beta_1 x_i$

$$E = \sum_{i=0}^n (y_{(actual)} - y_{(predicted)})^2$$

Minimize the sum of the square of error distance

$$E = \sum_{i=0}^n (y_i - \hat{y}_i)^2$$

$$E = \sum_{i=0}^n (y_i - \beta_0 - \beta_1 x_i)^2$$

$$\frac{\partial E}{\partial \beta_0} = \sum_{i=1}^n -2 (y_i - \beta_0 - \beta_1 x_i) = 0$$

$$\frac{\partial E}{\partial \beta_1} = \sum_{i=1}^n -2 x_i (y_i - \beta_0 - \beta_1 x_i) = 0$$

Finding the value of Beta_0

- Ordinary Least Square

- Find β_0

$$\frac{\partial E}{\partial \beta_0} = \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i) = 0$$

$$\sum_{i=1}^N y_i = N\bar{y}$$

$$n\beta_0 = n\bar{y} - n\beta_1\bar{x}$$

$$\beta_0 = \bar{y} - \beta_1\bar{x}$$

- note that this value depends on some average values from the data set

Finding Beta_1

- Ordinary Least Square

- Find β_1

$$\frac{\partial E}{\partial \beta_1} = \sum_{i=1}^n -2x_i (y_i - \beta_0 - \beta_1 x_i) = 0$$

$$\sum_{i=1}^n x_i y_i - \beta_0 x_i - \beta_1 x_i^2 = 0$$

$$\sum_{i=1}^n x_i y_i - (\bar{y} - \beta_1 \bar{x}) x_i - \beta_1 x_i^2 = 0$$

$$\sum_{i=1}^n x_i y_i - \bar{y} \sum_{i=1}^n x_i + \beta_1 \bar{x} \sum_{i=1}^n x_i - \beta_1 \sum_{i=1}^n x_i^2 = 0$$

$$\beta_1 = \frac{\sum_{i=1}^n x_i y_i - n \bar{x} \bar{y}}{\sum_{i=1}^n x_i^2 - n \bar{x}^2} \xrightarrow{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \sum_{i=1}^n x_i y_i - n \bar{x} \bar{y}} \beta_1 = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

- in the bottom part, we meaningfully represent this information to be useful for our model
- note that the $\sum (x_i - \bar{x})(y_i - \bar{y})$ is the covariance between x and y
- we then normalize this value to x (dividing by sum of squares of x)
- putting it together, we divide the covariance of x and y by the covariance of just x

R-Squared (goodness of fit)

a number that indicates how well data fits a statistical model

- defined between [0,1]
- higher value indicates a better fit

equation: $1 - ((\text{sum of squares between y actual and y predicted}) / (\text{sum of squares between actual y values and their mean}))$