

## Topic 7.5 Regression Analysis and Learning

- ✓ **Regression analysis** is a set of statistical processes for [estimating](#) the relationships among variables.
- ✓ From biology: 'Heights of descendants of tall ancestors tend to regress down towards a normal average ([regression toward the mean](#))'.
- ✓ Most commonly, estimates the [average value](#) of the dependent variable when the independent variables are fixed.
- ✓ Widely used in statistics for [prediction](#) and [forecasting](#), and substantially overlap with the field of [machine learning](#).

### ❖ Linear Regression:

- Simplest form of Regression
- Data are modeled using a **straight line** [Fitting a straight line]
- Bivariate Linear Regression (dependent and independent variables) is similar to Univariate function,  $y = f(x) = ax + b$ , where y-output, x-input(variable)

■ Linear Regression Learning Problem:

$$Y = \alpha X + \beta$$

$Y$  – random variable (response, dependent)

$X$  – random variable (predictor, independent)

$\alpha, \beta$  - regression coefficients, that are **to be learned**

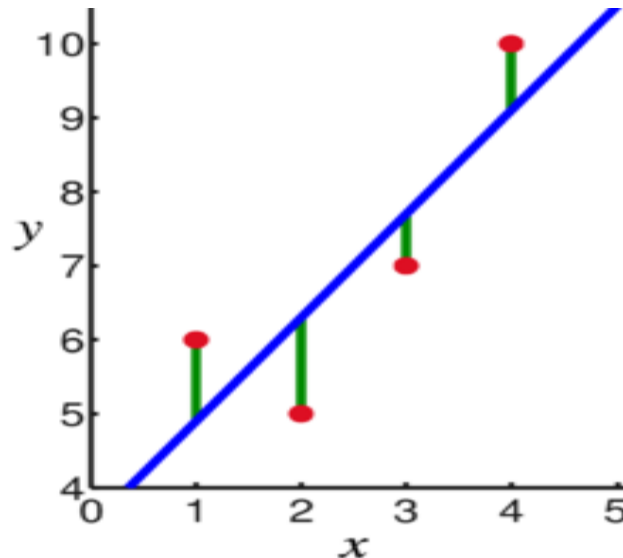
- To solve means to find estimated values of  $\alpha$  and  $\beta$  that best describes the field data
- Methods of **least squares** can be used to find  $\alpha$  and  $\beta$  minimizing error between the actual data and the estimate of the line.
- Traditionally, from Gauss, the squared loss function values, summed over all training examples are minimized, yielding
$$\alpha = \frac{\sum_{i=1:s} (x_i - x') (y_i - y')}{\sum_{i=1:s} (x_i - x')^2}, \quad \beta = y' - \alpha x',$$
where  $x'$  - average of  $x_1, x_2, \dots, x_s$ ,  $y'$  - of  $y_1, y_2, \dots, y_s$ ,  
given sample data points  $(x_1, y_1), (x_2, y_2), \dots, (x_s, y_s)$ .
- The line thus obtained can be used to predict an appropriate value of  $y$ , given an unknown  $x$ .

- Mean Absolute Error (MAE, L1 loss) is sometimes used to assess performance of a model that does not consider the direction of the outliers.
- For a data point  $y_i$  and its predicted value  $\hat{y}_i$ , where  $n$  is the total number of data points in the dataset:

$$\text{MAE} = \sum_{i=1:n} |y_i - \hat{y}_i| / n$$

- Mean Squared Error (MSE, L2 loss) is also used which is computed as follows:

$$\text{MSE} = \sum_{i=1:n} (y_i - \hat{y}_i)^2 / n$$



[Source: Internet]

➤ **Example.** Sample data (Salary data)

Serial	X (Year of experience)	Y (Salary in 1000 Taka)
1	3	30
2	8	57
3	9	64
4	13	72
5	3	36
6	6	43
7	11	59
8	21	90
9	1	20
10	16	83

✓ We get,  $Y = 3.5X + 23.6$ , and from it predict 58.6K salary after 10 years of experience.

➤ We can think of multiple regression like the one below:

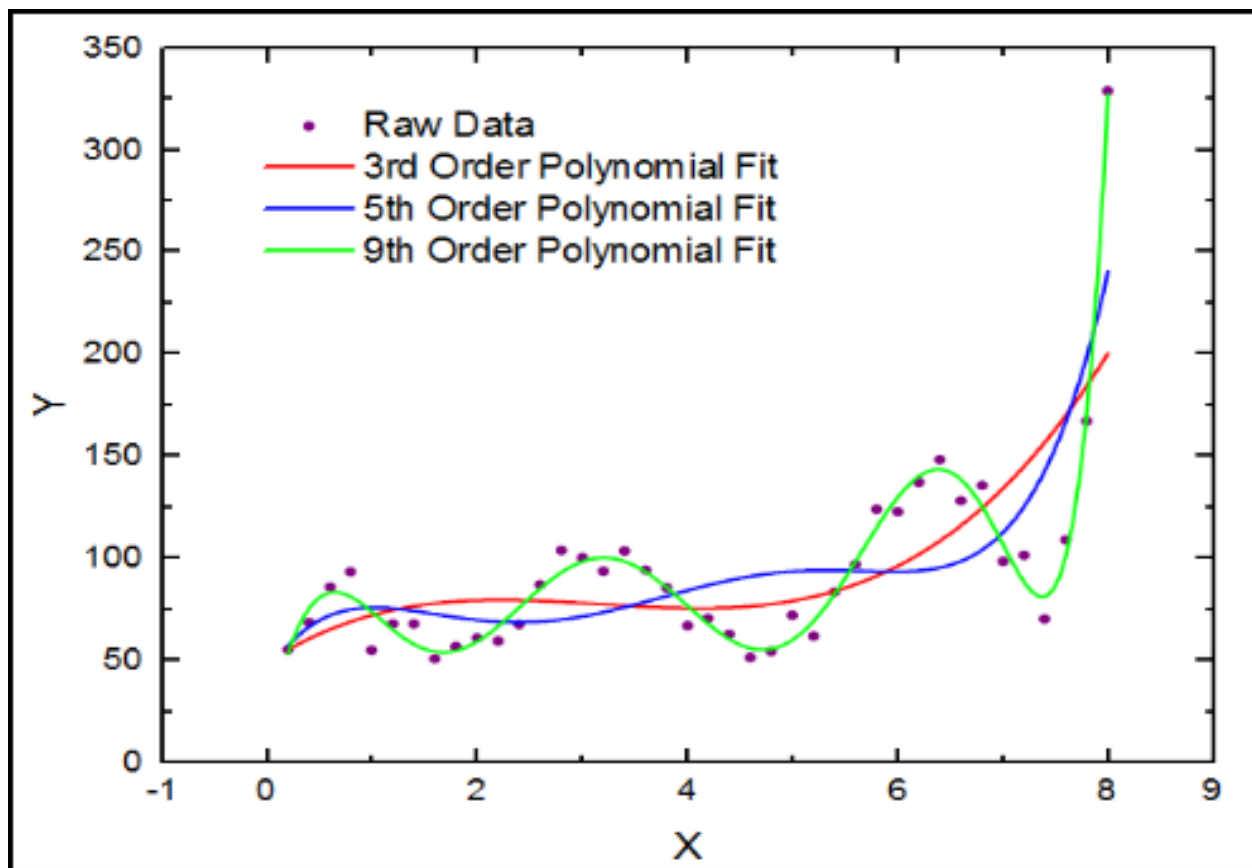
$$Y = \alpha_1 X_1 + \alpha_2 X_2 + \beta,$$

which can also be solved using least squares method.

➤ And nonlinear regression (polynomial) like the one below:

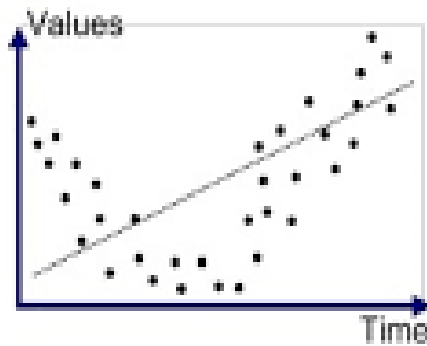
$$Y = \alpha_3 X^3 + \alpha_2 X^2 + \alpha_1 X + \beta,$$

transforming it, and applying least squares method.

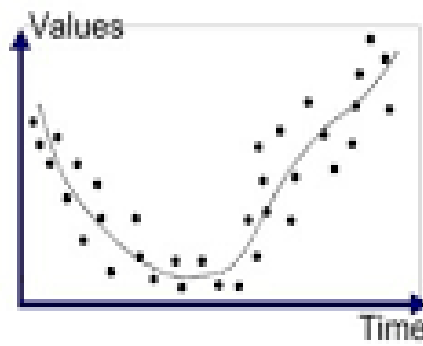


[Source: Internet]

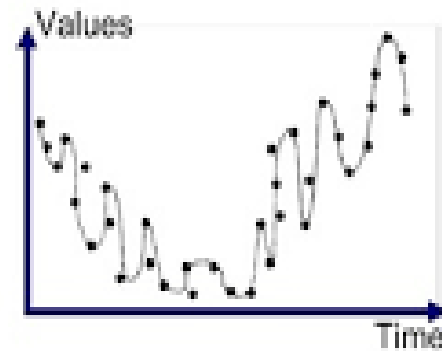
- Mind **overfitting and underfitting models** with data



Underfitted



Good Fit/Robust



Overfitted

[Source: Internet]