

# **Linear Regression**

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### **Questions to discuss**



- 1. What is the relationship between Machine Learning (ML) and Supervised Learning (SL)?
- 2. What is Linear Regression and how does it work?
- 3. What are the evaluation metrics for regression?
- 4. What are the pros and cons of linear regression?

# What is the relationship between ML and SL?



#### Machine Learning (ML)

- Machine Learning is the ability of a computer to do some task without being explicitly programmed.
- The ability to do the tasks comes from the underlying model which is the result of the learning process.
- The model is generated by learning from huge volumes (both in breadth and depth) of historical data reflecting the real world in which the processes are performed.

#### Examples of what machine learning algorithms can do

- Search through the data to look for patterns in the form of trends, cycles, associations, etc.
- Express these patterns as mathematical structures (model)
- Using those patterns to test the unseen data

#### Supervised Learning (SL)

- It builds a mathematical model using data that contains both the inputs and the desired output (labels or ground truth)
- There are basically two types of supervised learning:
  - Regression where the desired output is in the form of continuous values
    - e.g. predicting the house prices based on some features like area, the number of rooms, etc.
  - Classification where the desired output is in the form of categories
    - e.g. predicting if the person is likely to default on a loan based on the features like age, past transactions, etc.
- The model learns from the training data using these 'target variables' as reference variables.
- The model thus generated is used to make predictions

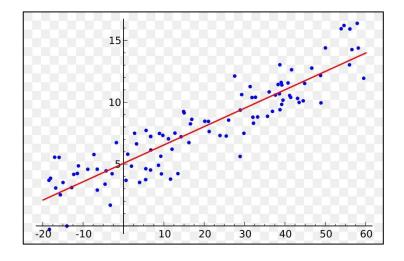
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# What is Linear Regression and how does it work?



- Linear regression is a way to identify a relationship between the independent variable(s) and dependent variable
- We can use these relationships to predict values for one variable for the given value(s) of other variable(s)
- It assumes the relationship between variables can be modeled through a linear equation or an equation of a line.
- The variable, which is used in prediction is termed as independent/explanatory/regressor where the predicted variable is termed as dependent/target/response variable.
- In the case of linear regression with a single explanatory variable, the linear combination can be expressed as:

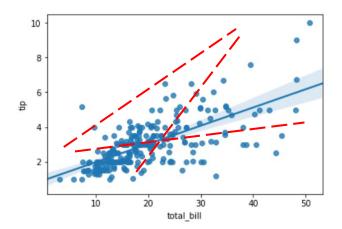
response = intercept + constant \* explanatory variable



# What is the best fit line in linear regression?



- Learning from the data, the model generates a line that fits the data.
- Our aim is to find a regression line that best fits the data
- By best fit, it means that the line will be such that the cumulative distance of all the points from the line is minimized
- Mathematically, the line that minimizes the sum of squared error of residuals is called Regression Line or the Best Fit Line.



- In the example here, you can see a scatter plot between the *tip* amount and the *total bill* amount
- We can see that there is a positive correlation between these two as the bill amount increases, the tip increases
- The line in blue that you see is the 'best fit' line those in red are some examples of all other lines that are not the 'best fit'

## What is Multiple Linear Regression?



- This is just the extension of the concept of simple linear regression with one variable
- In the real world, any phenomenon or outcomes could be driven by many different independent variables
- Therefore the need to have a mathematical model that can capture this relationship
  - Ex: predicting the price of a house, we need to consider various attributes related to the house, such as area, number of rooms, number of kitchens, etc.
- Such a regression problem is an example of multiple regression.
- It can be represented by :

target = intercept + constant1\*feature1 + constant2\*feature2 + constant3\*feature3 + .....

• The model aims to find the constants and intercept such that this line is the best fit.

#### What are evaluation metrics?



- Evaluating a model is very important as it helps us understand the model performance.
- Evaluation metrics allow us to quantify our model's performance using a single number.
- Comparing the metric values for train and test sets helps us get an idea about the fit of the model.
  - If the model performance is low on the train and test sets, then the model is said to underfit the data
  - o If the model performance is high on the train set but low on the test set, then the model is said to overfit the data.
- The aim is to find the model which best fits our data.

# What are the evaluation metrics for regression?



R-squared	Adjusted R-squared	Mean Absolute Error	Root Mean Square Error
<ul> <li>Measure of the % of variance in the target variable explained by the model</li> <li>Generally the first metric to look at for linear regression model performance</li> <li>Higher the better</li> </ul>	<ul> <li>Conceptually, very similar to R-squared but penalizes for the addition of too many variables</li> <li>Generally used when you have too many variables as adding more variables always increases R^2 but not Adjusted R^2</li> <li>Higher the better</li> </ul>	<ul> <li>Simplest metric to check prediction accuracy</li> <li>Same unit as the dependent variable</li> <li>Not sensitive to outliers i.e. errors doesn't increase too much if there are outliers</li> <li>Difficult to optimize from a mathematical point of view (pure maths logic)</li> <li>Lower the better</li> </ul>	<ul> <li>Another metric to measure the accuracy of prediction</li> <li>Same unit as the dependent variable</li> <li>Sensitive to outliers - errors will be magnified due to the square function</li> <li>But has other mathematical advantages that will be covered later</li> <li>Lower the better</li> </ul>

# What are the pros and cons of linear regression?



#### Pros:

- Simple to implement and easier to interpret the outputs coefficient.
- Helpful if the relationship between the independent and dependent variable is linear

#### Cons:

- Has a lot of statistical assumptions which are not always true for real-world data
- Outliers can have huge effects on regression
- Assumes that the input variables are independent. It gets highly affected by multicollinearity.

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Metric	Formula	
R-squared	$R^{2} = 1 - \frac{\frac{1}{n} \sum_{i=1}^{n} (y_{i} - \widehat{y}_{i})^{2}}{\frac{1}{n} \sum_{i=1}^{n} (y_{i} - \overline{y})^{2}}$	
Adjusted R-squared	$Adj. R^2 = 1 - \left[ \frac{(1 - R^2)(n - 1)}{n - k - 1} \right]$	
Mean Absolute Error	$MAE = \frac{1}{n} \sum_{i=1}^{n}  y_i - \widehat{y}_i $	
Root Mean Square Error	$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (y_i - \widehat{y}_i)^2}$	

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