

Week – 1I

Intro to ML

ML Bootcamp 2021



Careera Analytics Lab

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One of the simplest
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One of the
simplest algorithms
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Lets start
practicing!

Quick Refresher on ML Concepts

- **Datasets:** The data that you process, transform and use to train your ML models
- **Features:** The individual columns(if it is a table) in your data
- **Supervised/Unsupervised:** Either with or without answers
 - **Classification:** Predicting a limited set of outputs Eg: Pos/Neg sentiment, Cat/Dog, Bearish/Bullish
 - **Regression:** Output has an infinite possibilities Eg: Housing prices, stock prices, time etc.
- **Machine Learning:** A set of algorithms or a method to learn through experience using data
- **Training and Testing:** We usually split our data into two parts, one to train the model and another to test the model's performance
- **Decision Boundary:** The line separating the classes for a given ML problem
- **Hypothesis:** The model that you would like to determine
- **Weights/Parameters:** This is what defines your model as these are the values that are adjusted to train your model to make accurate predictions
- **Ground-Truth:** The labels/answers that are provided along with the dataset

Labels/Ground-Truth
provided



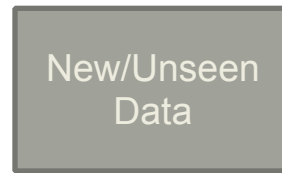
Learns to mimic the
dataset by understanding
the features and correcting
its predictions

STEP 1

Supervised Learning



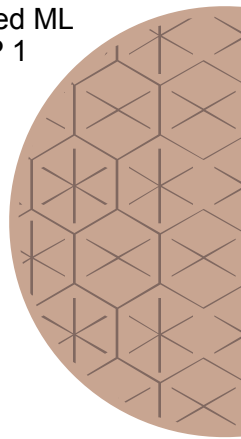
STEP 2



Input to the Trained ML
Model from STEP 1



Predictions



Unsupervised Learning

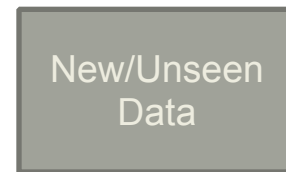
No Labels/Ground-Truth
this time



Learns the patterns
underlying the data to
group them: Clustering

STEP 1

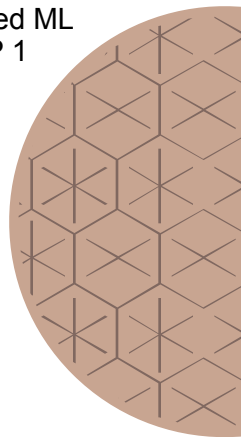
STEP 2



Input to the Trained ML
Model from STEP 1



Predictions/Clusters



Classification Vs Regression

Problem Definition	Classification or Regression?
Predicting the sentiment of a given Amazon Review. Output/ Ground-Truth: 1 → Positive 0 → Negative	
Weather Forecasting Output/ Ground-Truth: The temperature in Celsius or Farenheit	
Stock Market Price prediction Output/ Ground-Truth: The next day's stock price of a given company	
Next word prediction in E-mail: Output/ Ground-Truth: The next word based on the previous words	

Linear Regression

1.

$$y = \beta_0 + \beta_1 * x_1 + \beta_2 * x_2 + \beta_3 * x_3 \dots$$



Dependent Variable/
Output/ Prediction



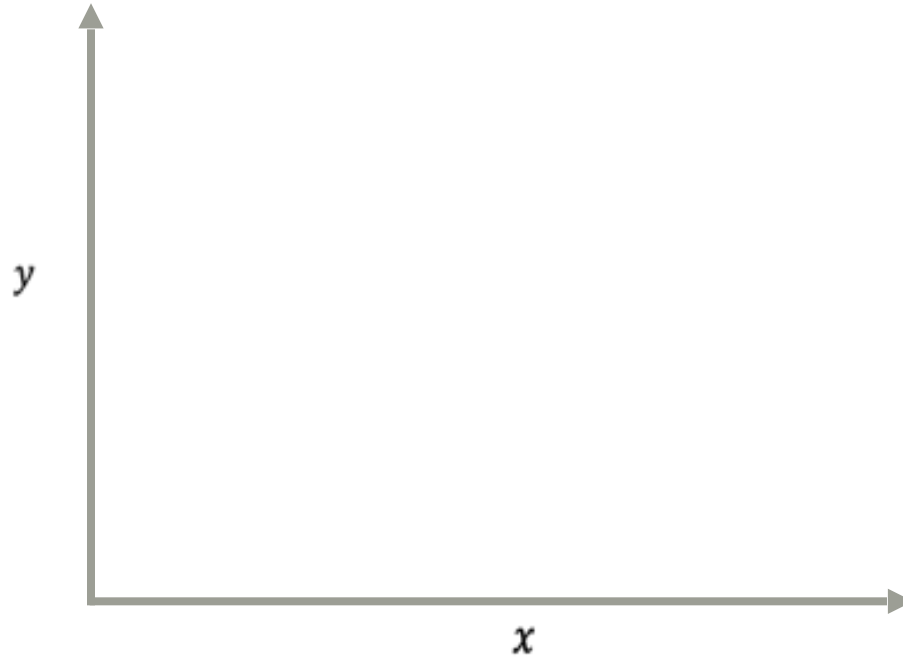
Coefficients / Weights



Independent Variable/
Input/ Predictors/
Features

This is an equation that is “Linear” in the input parameters β
Goal here is to find the best β that will get you close to the
ground-truth y values

How do we model Linear Regression?



How to find the best β ?

1. Linear Algebra, for example by using matrix inversion to find β directly
- Ordinary Least Squares Method (which follows the Maximum Likelihood Estimation Framework)
 - Gradient Descent

Gradient Descent

1.

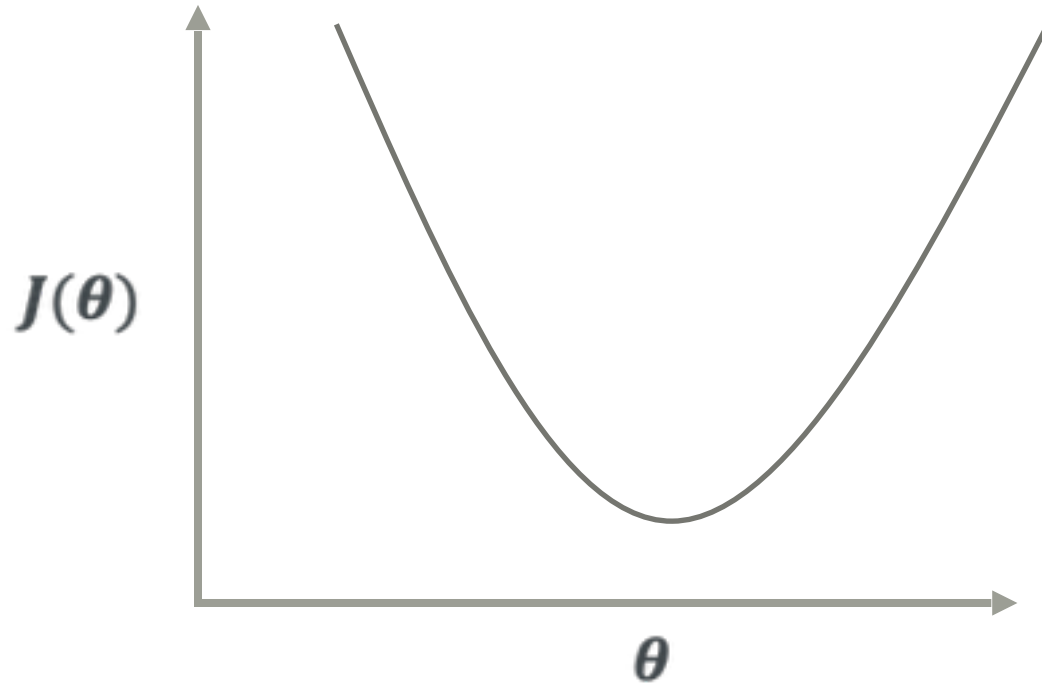
$$\text{Cost Function } J(\theta) = \frac{1}{2} \sum_{i=1}^N (y_i - \hat{y}_i)^2$$

These are the
parameters like
 β earlier

The ground-truth
value

The model's
predicted value

Convex Optimization Problem



Logistic Regression

1. The name is contradicting but this algorithm is meant to solve classification problems
 - Let's look at our previous model for Linear Regression

$$y = \beta_0 + \beta_1 * x_1 + \beta_2 * x_2 + \beta_3 * x_3 \dots$$

- How can we transform the above model such that we can address the problem of Binary Classification?

$$y = f(\beta_0 + \beta_1 * x_1 + \beta_2 * x_2 + \beta_3 * x_3 \dots)$$

where,

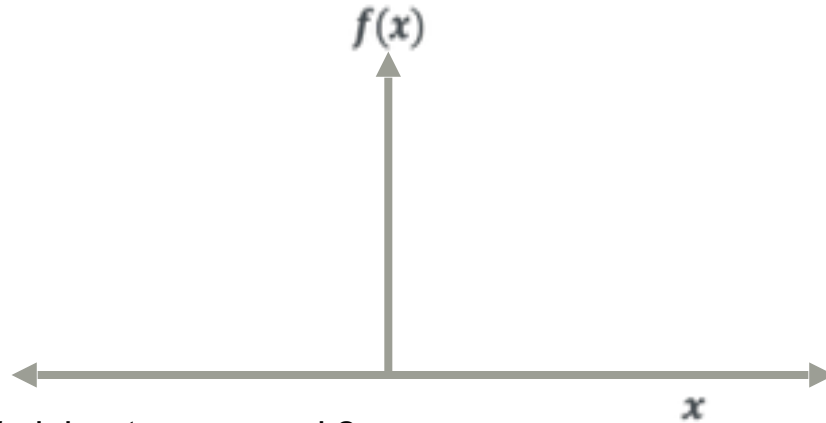
$$f(x) = \frac{1}{1 + e^{-x}}$$

—————> Logit Function

Expanding a bit further

1.

$$y = f(\beta_0 + \beta_1 * x_1 + \beta_2 * x_2 + \beta_3 * x_3 \dots)$$
$$= \frac{1}{1 + e^{-(\beta_0 + \beta_1 * x_1 + \beta_2 * x_2 + \beta_3 * x_3 \dots)}} = \begin{cases} 1 & \text{if } y \geq 0.5 \\ 0 & \text{if } y < 0.5 \end{cases}$$



Summary

- **Machine Learning** is a set of algorithms that learn from the data
- **Supervised learning** is a class of problems where the ground-truth/labels are provided
- **Unsupervised learning** problems do not have any explicit labels and the algorithms explore the underlying structure
- **Classification** problems have a finite set of outputs whereas **Regression** problems deal with the space of real numbers as outputs
- **Linear Regression**: An algorithm originating from Linear Algebra that is used for Regression problems. Eg: Weather Forecasting, House price prediction and Stock market trend analysis
- **Logistic Regression**: An algorithm that converts Linear Regression formulation into a classification problem. Eg: Sentiment Analysis, Titanic survival prediction etc.



Lets start Practicing!



Thanks

Do you have any questions?

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