# Hands-on Machine Learning Project



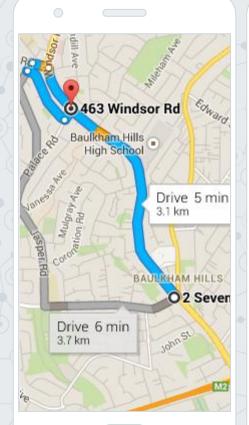
### **Main Idea of ML**

Let's start with the explaining what is Machine learning.

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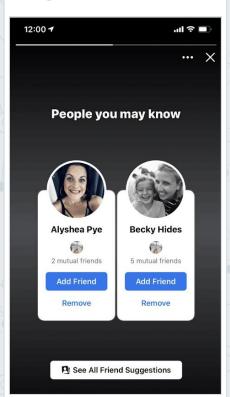
"Field of study that gives computers the ability to learn without being explicitly programmed"

- Arthur Samuel (1959)

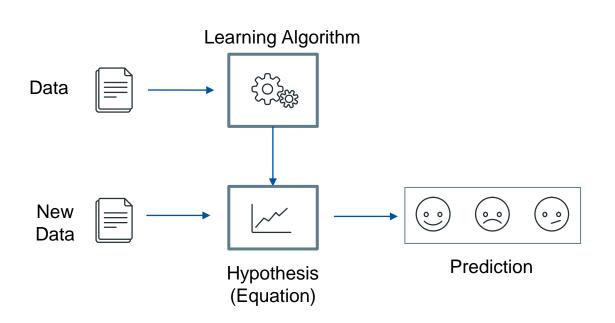








#### Machine Learning – Main Idea



#### Data:

- Labeled Data
- Unlabeled Data

#### Prediction:

- Category
- Number
- Cluster

C2 General

#### Machine Learning – Problems' Types

#### **Supervised ML**

Learn to predict target values from labelled data.

- Classification (target values are discrete classes)
- Regression (target values are continuous values)

#### **Unsupervised ML**

Find structure in unlabeled data.

- Clustering (find groups of similar instances in the data)
- Outlier Detection (finding unusual patterns)



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## Split the Data

In order to mimic the old/ new data we discussed, our number one step in any ML project is splitting the data.



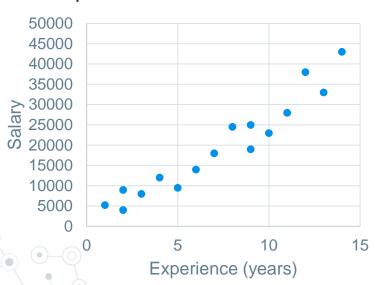
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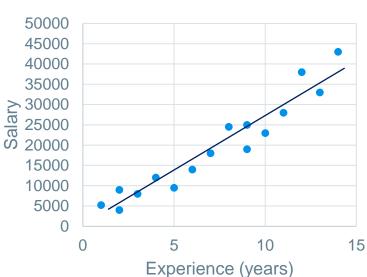
## **ML Algorithms**

Now it's time to learn some ML algorithms.

#### **Linear Regression**

- It's a regression Algorithm used in supervised ML problems.
- the objective is to find the **regression line** that best fits the data. Experience





#### Linear Regression - Equation

The regression line has an equation of the form

"
$$Y = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \cdots$$

**X** is the features

Y is the predicted value

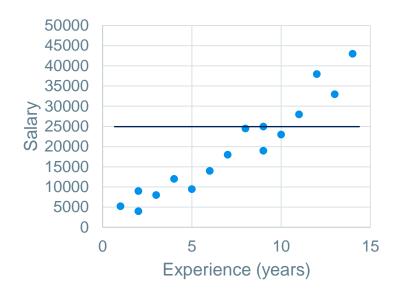
 $\theta_0$  is the bias term ( $\boldsymbol{y}$  when  $\boldsymbol{x} = 0$ )

Training Phase "
$$Y = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \cdots$$
"

Testing Phase 
$$"Y = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \cdots"$$

#### Linear Regression - Initialize theta

We start by initializing the theta values to any random value include zeros



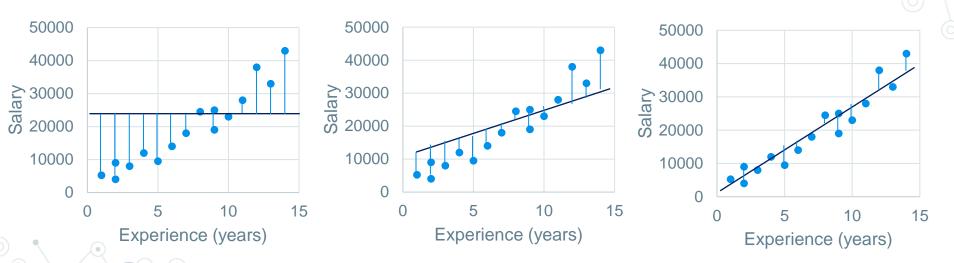
We find that this value doesn't fit our data, so we need to find the best fit values.

#### **Linear Regression - Cost function**

we use the MSE as a cost function.

Error = 
$$\sum_{i=1}^{n} (actual\_output - predicted\_output) ** 2$$

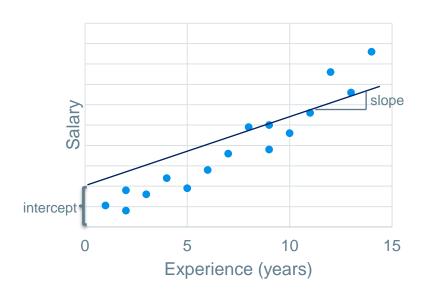
Where the error is the distance between the actual point to the regression line.



To find the best line we need to minimize the cost function, so we use the gradient descent algorithm.

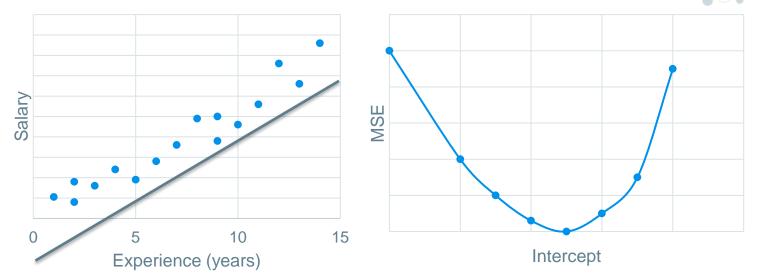
#### **Gradient Descent**

Each line represented using two information: intercept, slop.



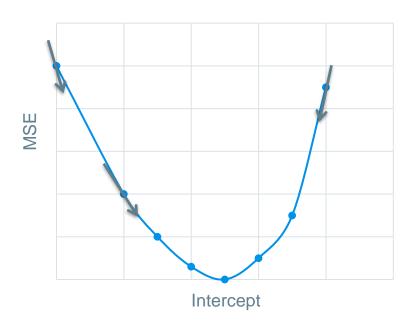
#### **Gradient Descent**

O To simplify the algorithm let's keep the same slop and only change the intercept.



#### **Gradient descent**

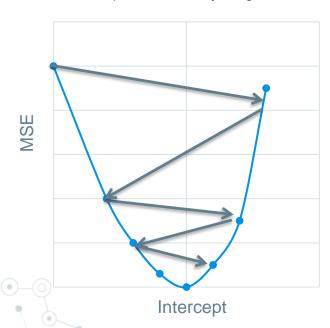
At each step, we calculate the slope of the cost function, we get the direction we need to move towards, in order to reach the local minimum.

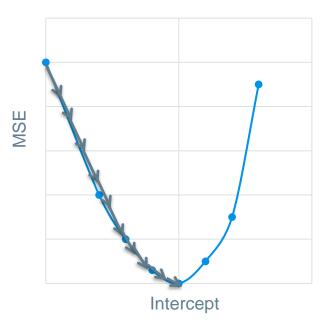


#### **Gradient descent**

To choose the step size we use the hyperparameter  $\alpha$ , which is the learning rate.

If this step size, alpha, is too large, we will overshoot the minimum, If alpha is too small, we will take too many iterations to get to the minimum. So, alpha needs to be *just right*.





### Simple Linear Regression

### Multi Linear Regression

Experience	Salary		

Experience	Title	Major	 Salary

