



MACHINE LEARNING

DAY 2

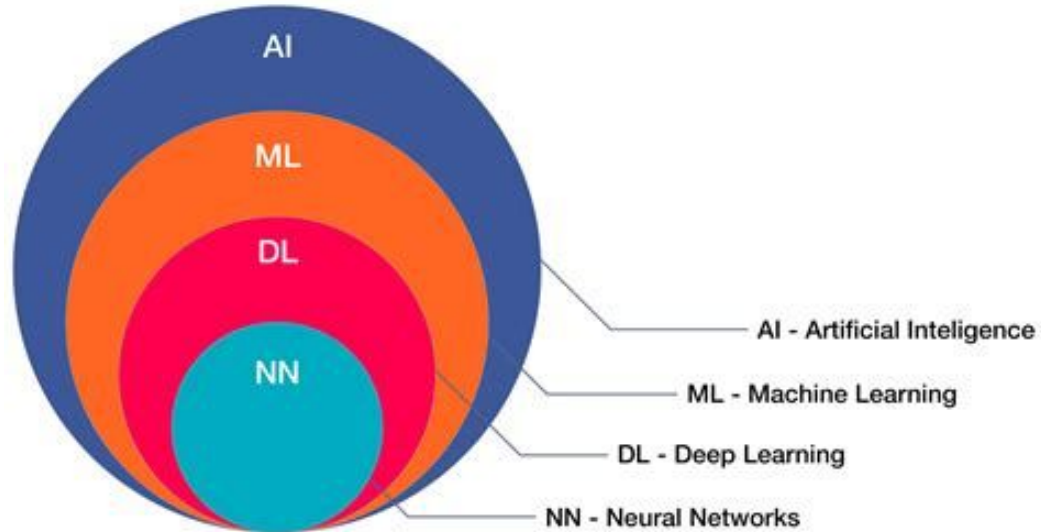
Lets go buy some chocolates!



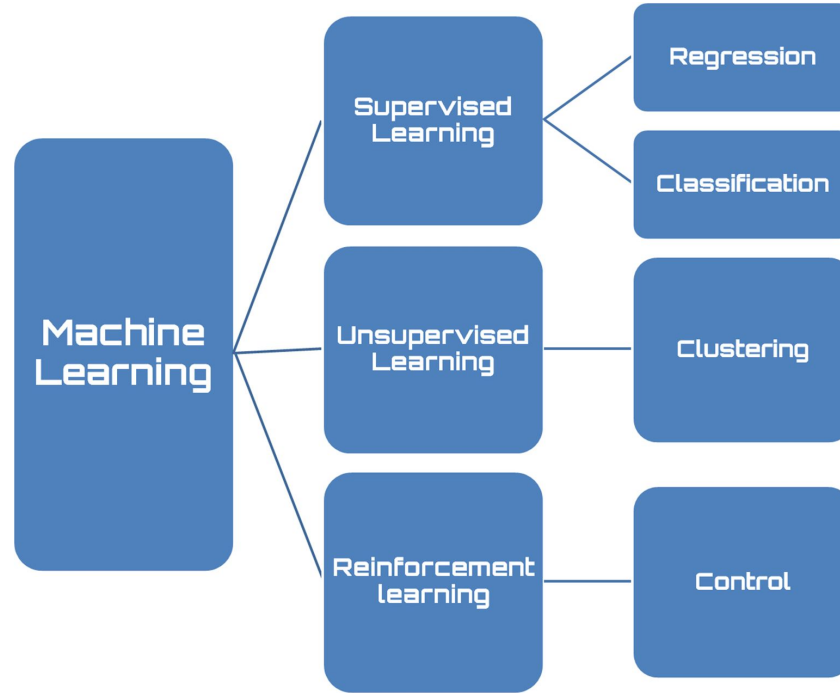
NO.	PRICE
1	10
2	20
4	40

3	?
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BIG PICTURE

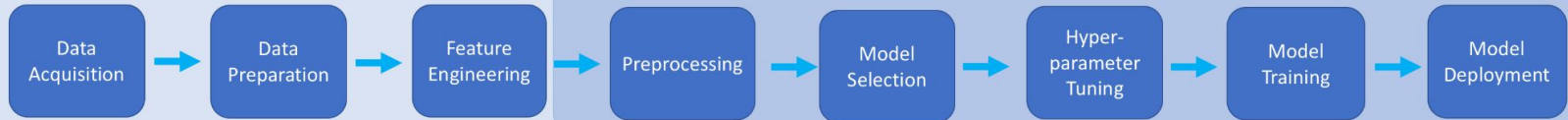


CATEGORIES

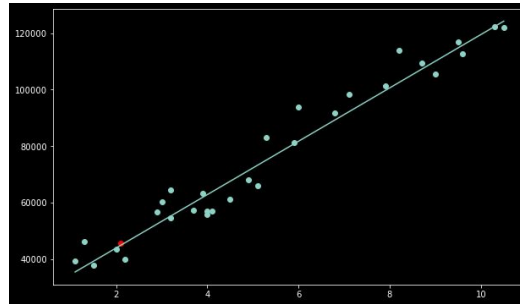
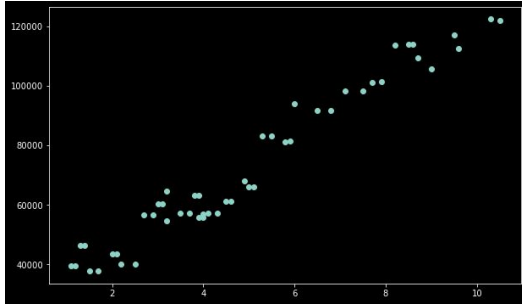


Machine Learning Pipeline

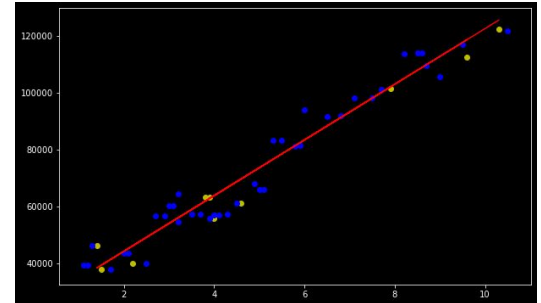
Prescience Automated Pipeline



REGRESSION



$$\begin{aligned} Y &= m \times X + c \\ &= c + m \times X \\ h_{\theta}(X) &= \Theta_0 + \Theta_1 \times X \end{aligned}$$



Hypothesis: $h_{\theta}(x) = \theta_0 + \theta_1 x$

Parameters: θ_0, θ_1

Cost Function: $J(\theta_0, \theta_1) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})^2$

Goal: $\underset{\theta_0, \theta_1}{\text{minimize}} J(\theta_0, \theta_1)$

Gradient Descent

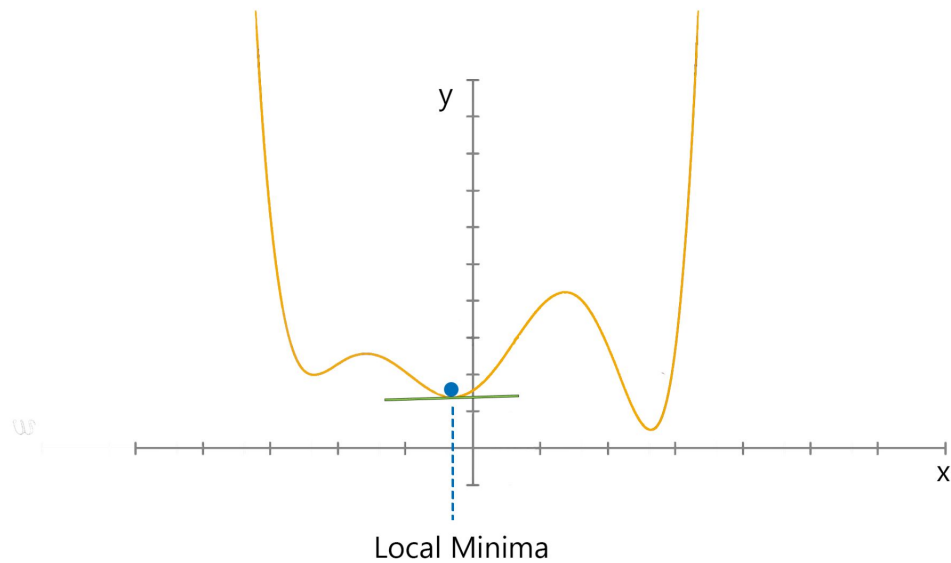
Remember that the general form of gradient descent is:

$$\begin{aligned} & \textit{Repeat} \{ \\ & \quad \theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta) \\ & \} \end{aligned}$$

We can work out the derivative part using calculus to get:

$$\begin{aligned} & \textit{Repeat} \{ \\ & \quad \theta_j := \theta_j - \frac{\alpha}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) x_j^{(i)} \\ & \} \end{aligned}$$

Step in Linear Regression



repeat until convergence {

$$\theta_0 := \theta_0 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})$$

$$\theta_1 := \theta_1 - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) \cdot x^{(i)}$$

}