# Key Machine Learning Concepts

Start Simple First

#### Don't Start with Neural Networks!

- Complexity
- Maintenance
- Explainability
- Slow and costly
- Turing's Rule of Least Power

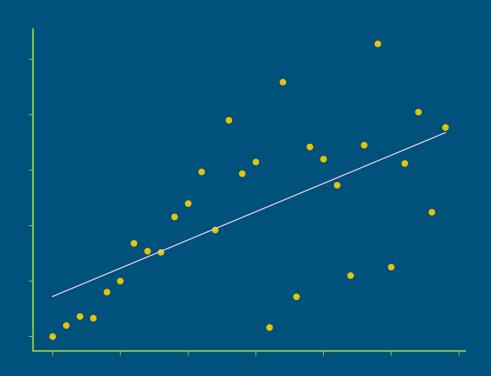
### Linear Regression

$$Model: y = mx + b$$

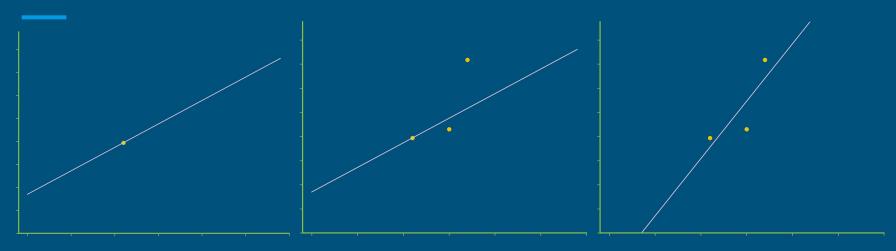
- Supervised
- High Bias
- Low Variance

$$\mathbf{m} = \sum (\mathbf{x}_{i} - \overline{\mathbf{x}})(\mathbf{y}_{i} - \overline{\mathbf{y}}) / \sum (\mathbf{x}_{i} - \overline{\mathbf{x}})^{2}$$

$$b = \overline{y} - m\overline{x}$$



# Judging Goodness of Fit



Loss Function: loss = 
$$\frac{1}{2n} \sum ((Model(x) - y)^2)$$

$$(9^2 + 0^2) / 4 = 81 / 4 = 20.25$$

$$(4.5^2 + 4.5^2) / 4 = 40.5 / 4 = 10.125$$

#### **Gradient Descent**

Find partial derivative of each parameter

$$m_1 := m_0 - (Learning Rate * D_m)$$

Optimizer: Minimizes Loss Function by updating Model parameters

Stochastic Gradient Descent (SGD)

$$D_{m} = \frac{1}{2n} \sum (\frac{d}{dm} (y - (mx + b))^{2})$$

$$u = y - (mx + b)$$

$$D_{m} = \frac{1}{2n} \sum (\frac{d}{dm} (u^{2}))$$

$$\frac{d}{dm} (u^{2}) = 2u * \frac{du}{dm}$$

$$\frac{du}{dm} = \frac{d}{dm} (y - (mx + b)) = -x$$

$$\frac{d}{dm} = 2(y - (mx + b)) * -x$$

$$D_{m} = \frac{1}{2n} \sum (2(y - (mx + b))(-x))$$

$$D_{m} = \frac{1}{2n} \sum ((y - (mx + b))(-x))$$

### Multiple Linear Regression

$$y_0 = b + m_0 x_{00} + m_1 x_{01} ... m_n x_{0n}$$
  
 $Y = BX$ 

$$\begin{bmatrix} y_0 \\ \vdots \\ y_p \end{bmatrix} = \begin{bmatrix} b \\ m_0 \\ \vdots \\ m_n \end{bmatrix} \times \begin{bmatrix} 1, x_{00} \dots x_{0n} \\ 1, \vdots & \vdots \\ 1, x_{p0} \dots x_{pn} \end{bmatrix}$$

## Multiple LR - Gradient Descent

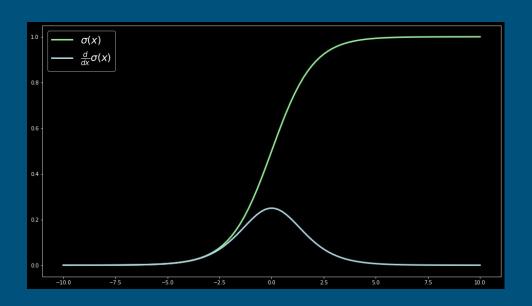
Loss = 
$${}^{1}/{}_{2n} (Y - BX)^{T}(Y - BX)$$
  
 $D_{B} = {}^{1}/{}_{n} X^{T}(BX - Y)$   
B1 := B0 -  $\alpha D_{R}$ 

## Logistic Regression

Sigmoid Activation Function: Condenses output to be between 0 and 1

Sigmoid = 1 / (1 + 
$$e^{-x}$$
) =  $\sigma(x)$ 

$$y = 1 / 1 + e^{-(mx + b)} = \sigma(mx + b)$$



#### Neural Networks

3 layers, 3 neurons per layer, sigmoid activation function

$$Y = \sigma(\sigma(\sigma(M_1X + b_1)M_2 + b_2)M_3 + b_3)$$

$$M_q \in \mathbb{R}^{3\times n}$$

$$b_q \in \mathbb{R}^{3\times 1}$$

#### Principles for Machine Learning

- Model: Describes variable relationships: Linear Regression
- Loss Function: Describes goodness of fit: Mean Squared Error
- Optimizer: Minimizes loss function: Stochastic Gradient Descent (SGD)
- Hyperparameters:
  - Learning Rate: ~10<sup>-3</sup> 10<sup>-6</sup>
  - Batch Size: n
- Evaluation Metrics: Tells human how well model is working: RMSE

# Happy Hunting!