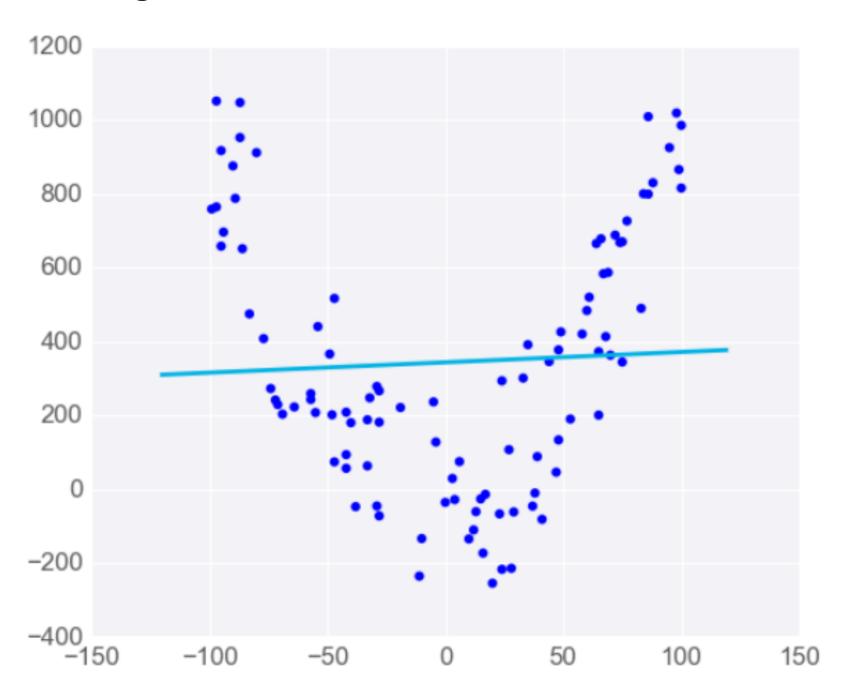
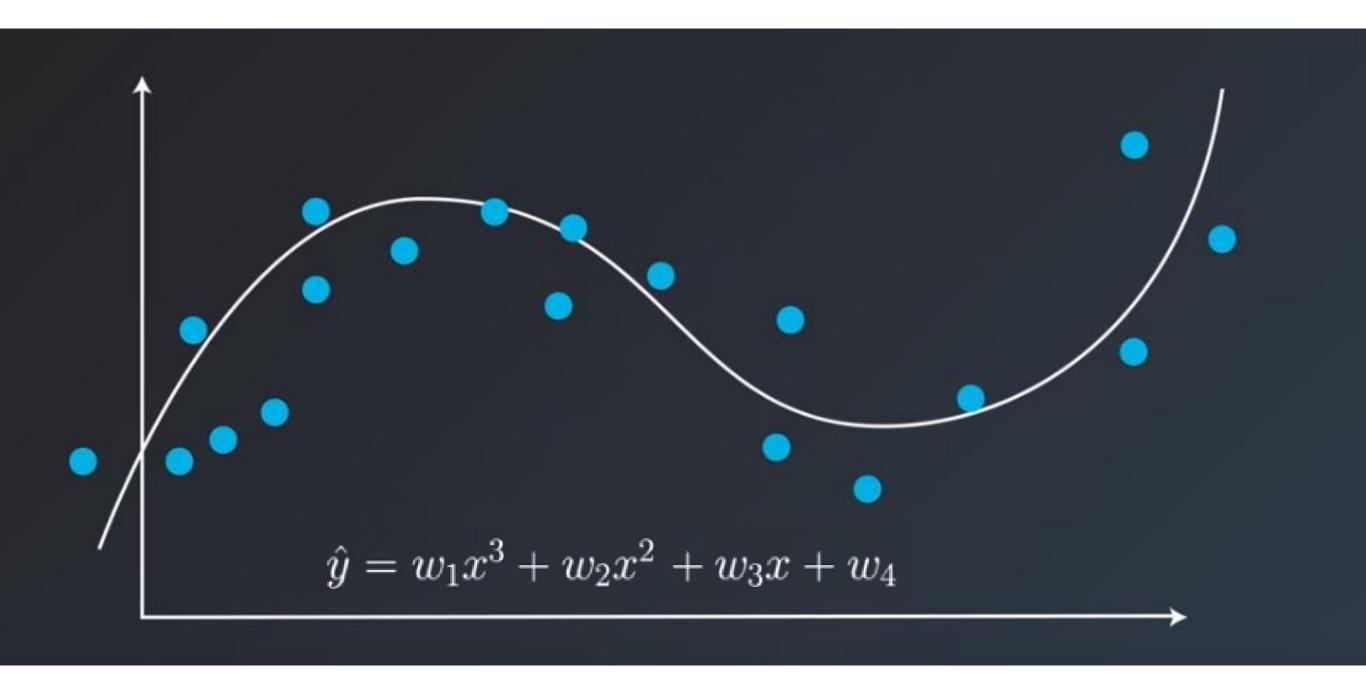
# Polynomial Regression

## Linear Regression Works Best When the Data is Linear





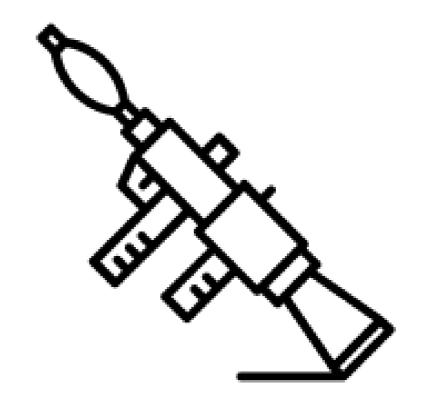
# Polynomial Regression





## **Model Selection**





**Simple Problem** 

**Complex Solution** 

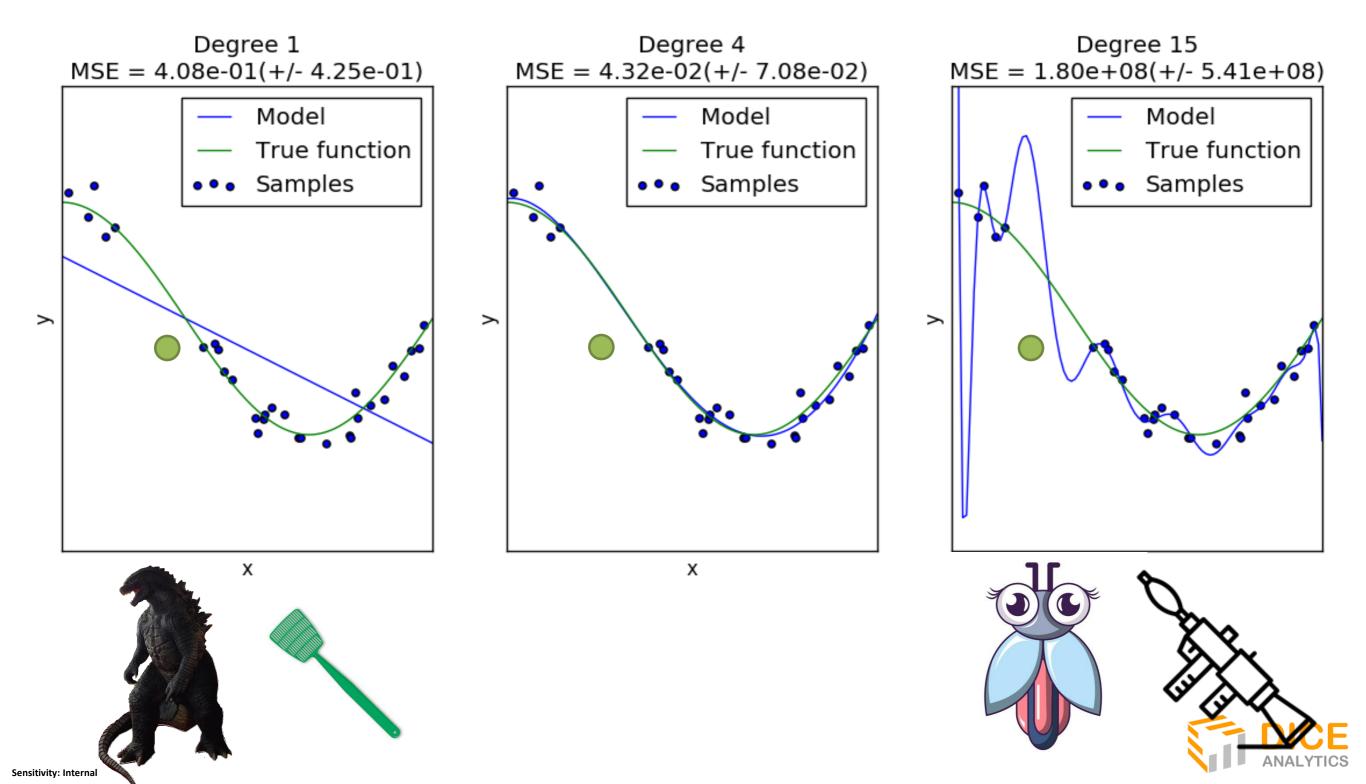


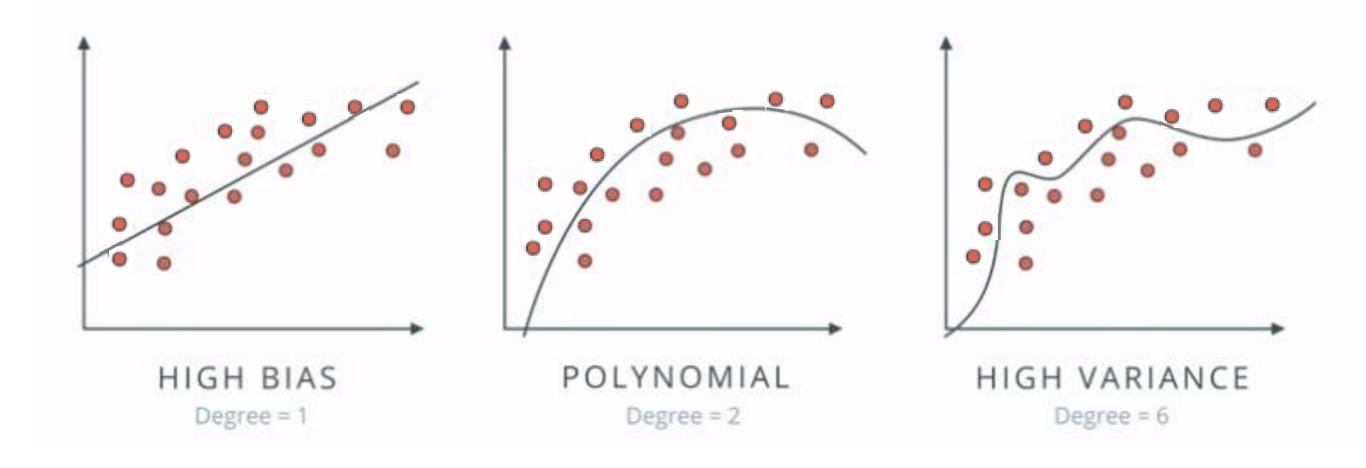
## **Model Selection**



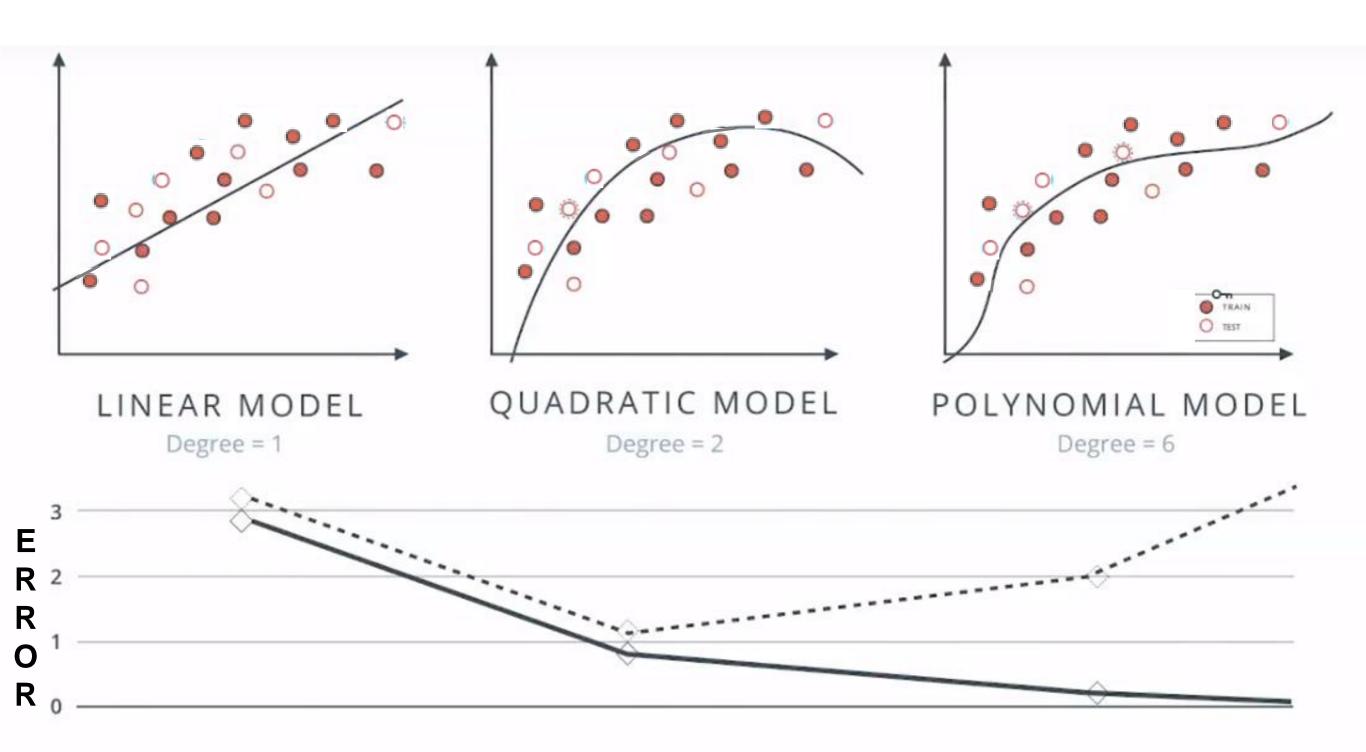


# Under-fitting & Over-fitting

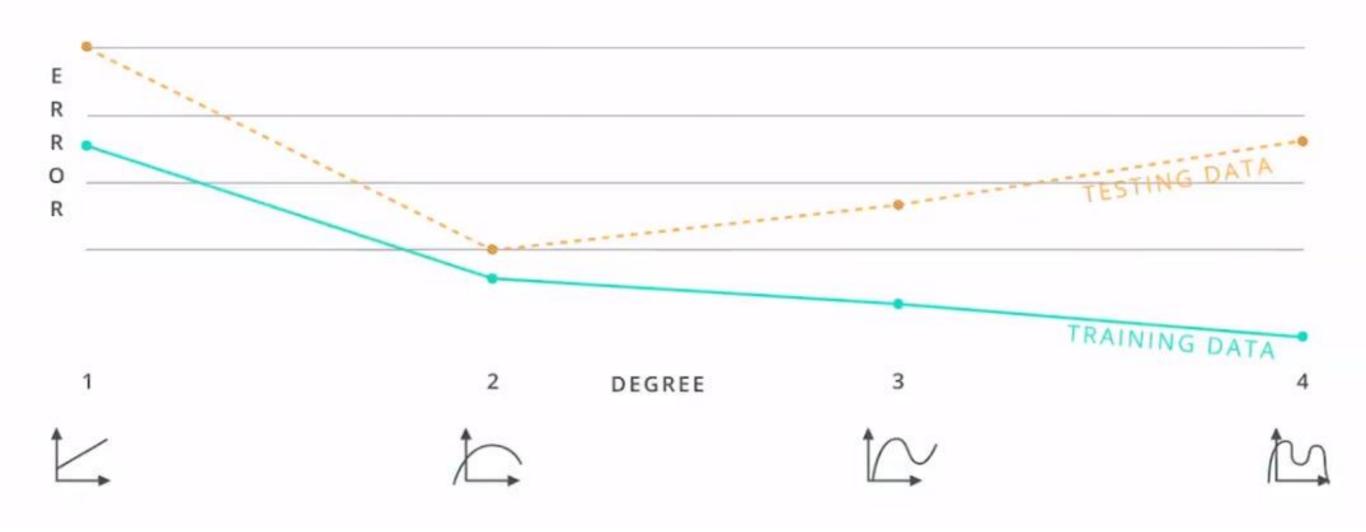




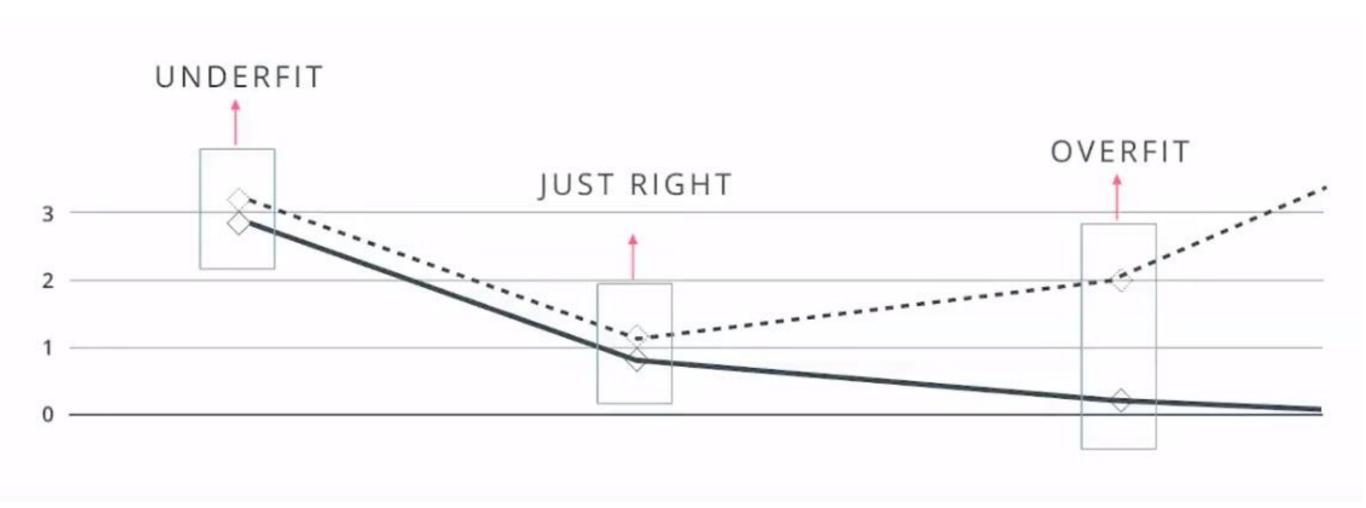




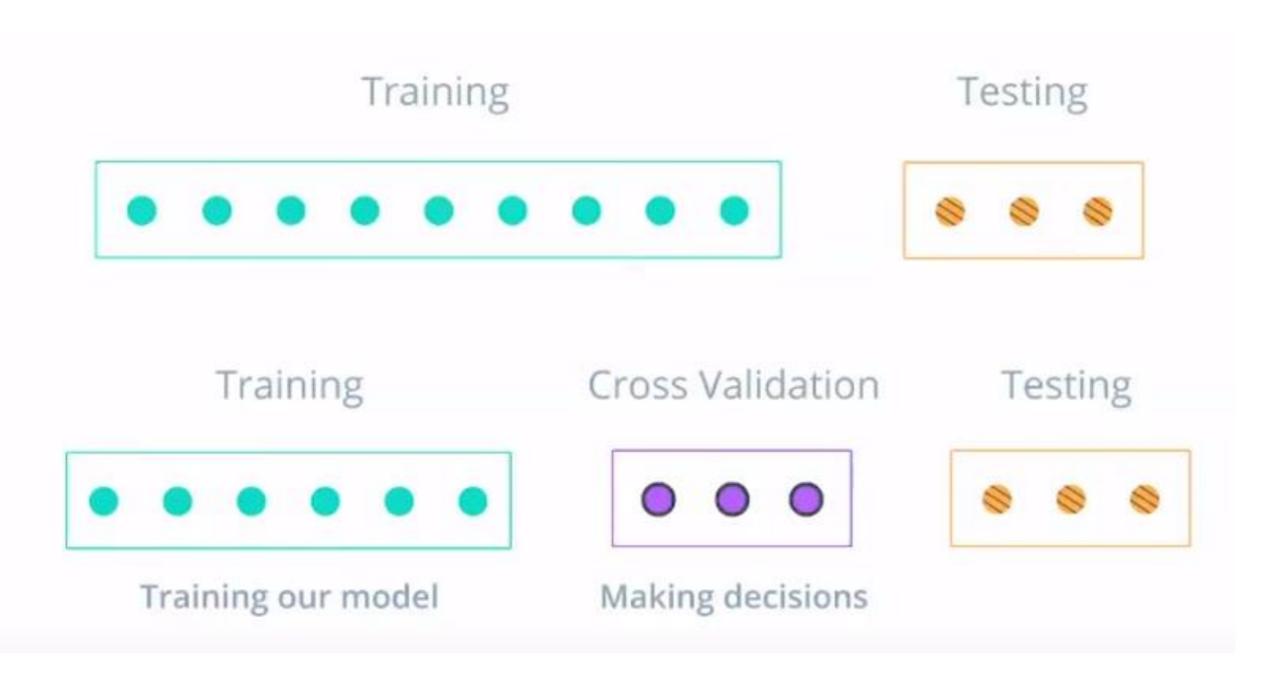




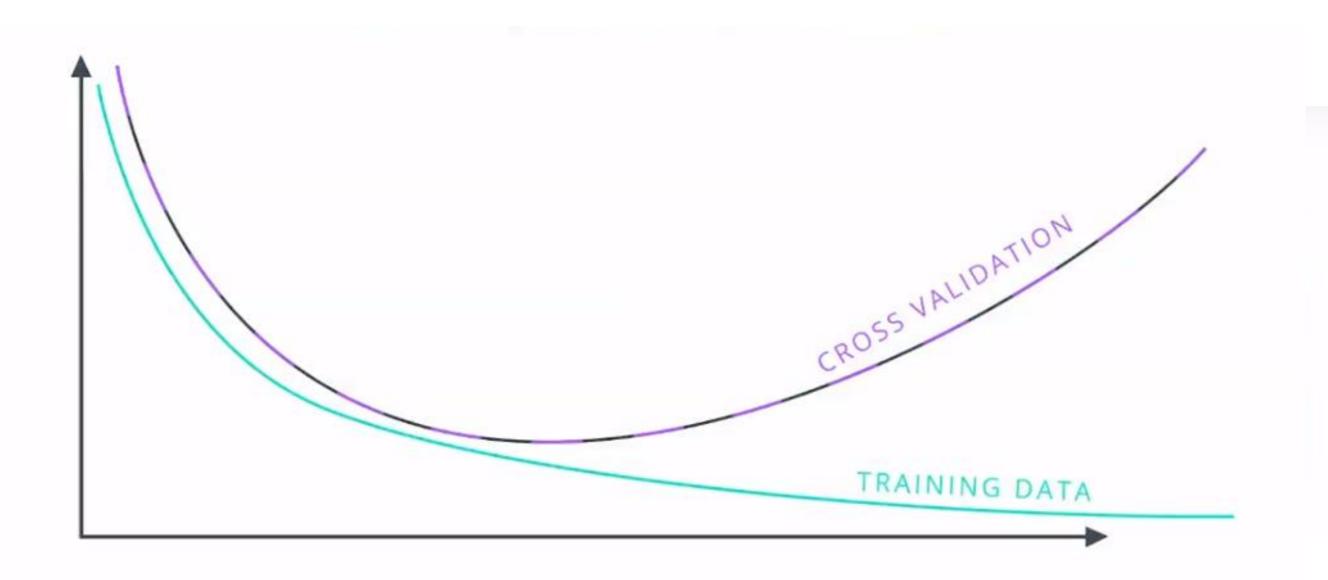






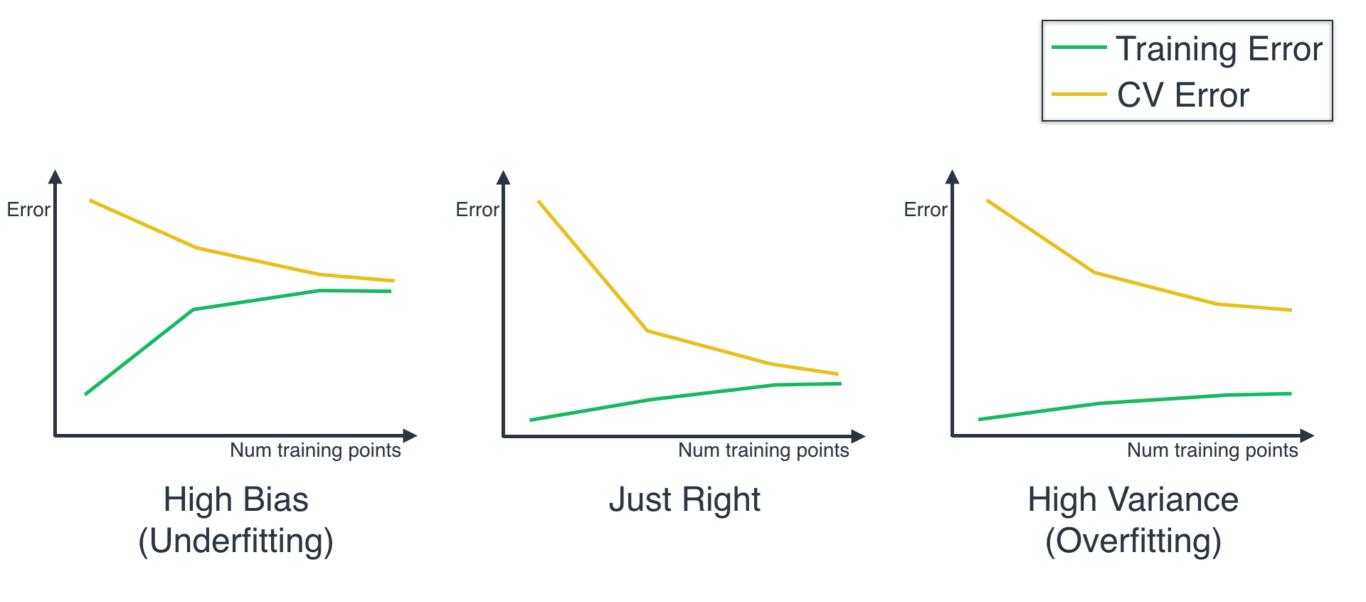






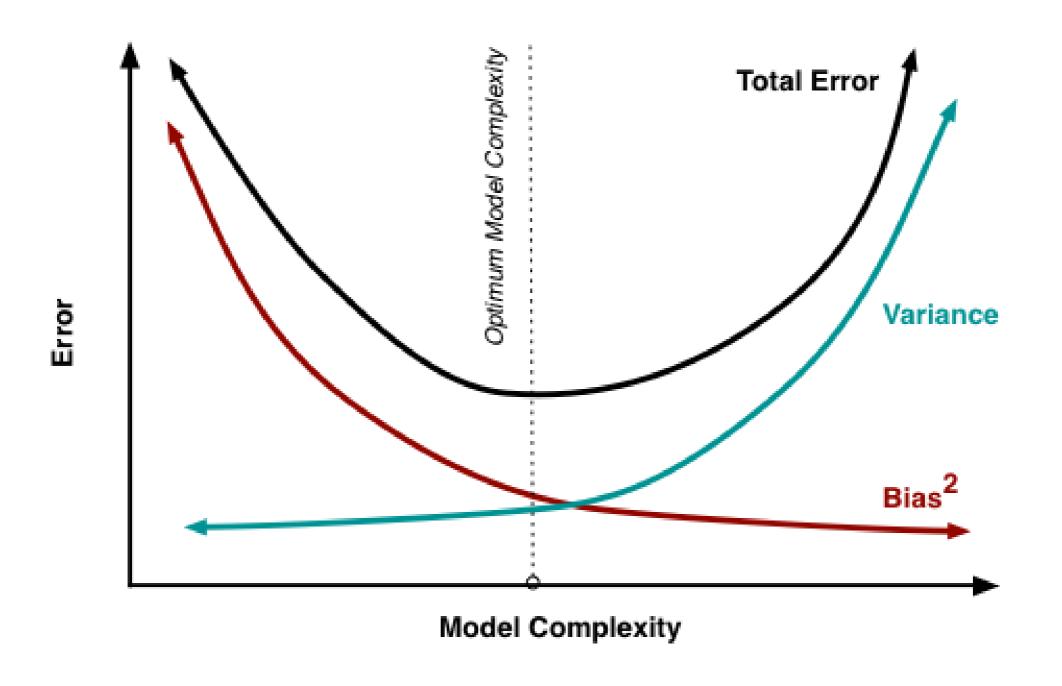


# Impact of training points



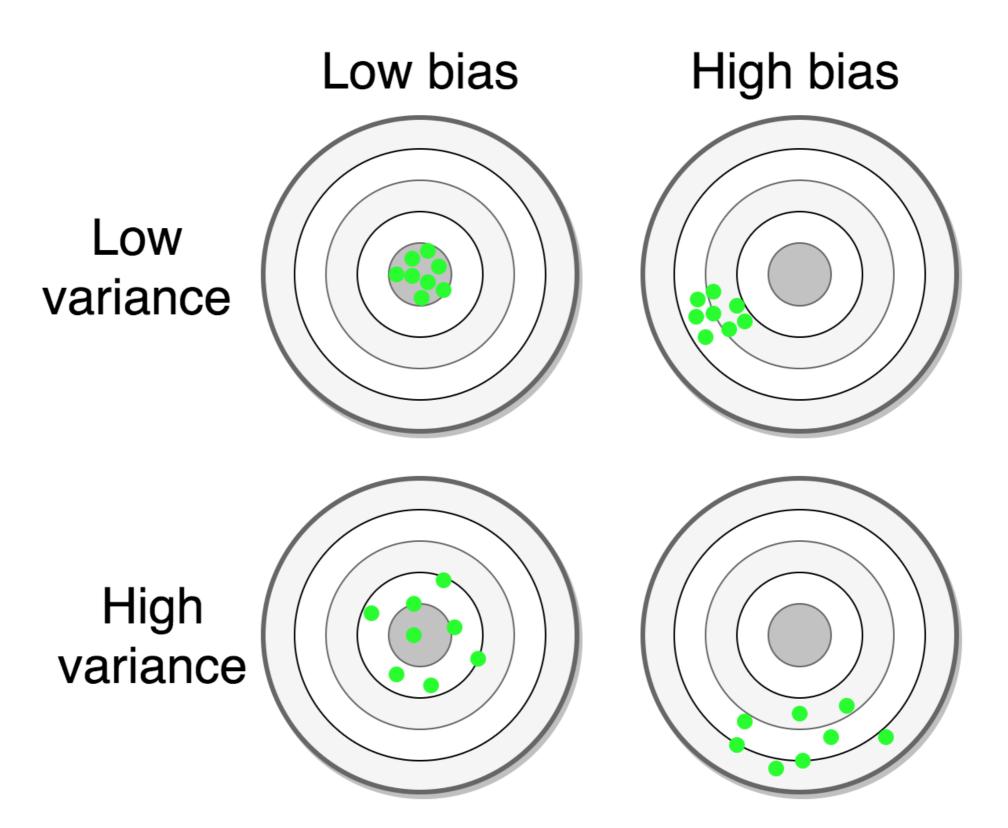


# Bias Variance Trade-off





## Bias Variance Trade-off



https://elitedatascience.com/bias-variance-tradeoff



# Regularization / Shrinkage

## Ridge / L2

$$\sum_{i=1}^{n} \left( y_i - \beta_0 - \sum_{j=1}^{p} \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^{p} \beta_j^2 = RSS + \lambda \sum_{j=1}^{p} \beta_j^2$$

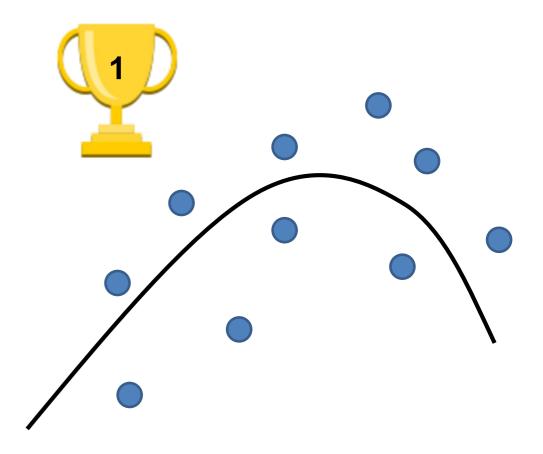
### Lasso / L1

$$\sum_{i=1}^{n} \left( y_i - \beta_0 - \sum_{j=1}^{p} \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^{p} |\beta_j| = RSS + \lambda \sum_{j=1}^{p} |\beta_j|.$$



#### **SIMPLE MODEL**

#### **COMPLEX MODEL**



## ERROR:

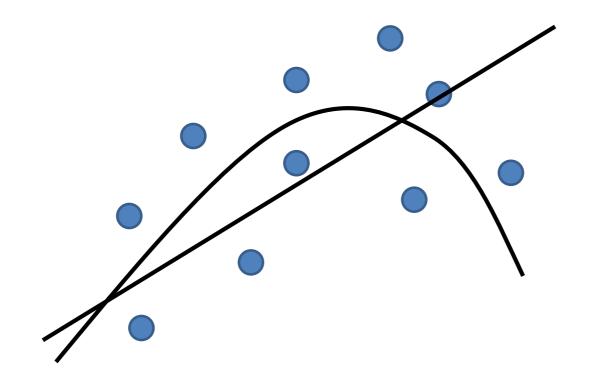
$$3x_1 + 4x_2 + 5$$

### **ERROR:**

$$(2x_1^3 - 2x_1^2x_2 - 4x_2^3 + 3x_1^2 + 6x_1x_2 + 4x_2^2 + 5$$



# L1 (Lasso) Regularization

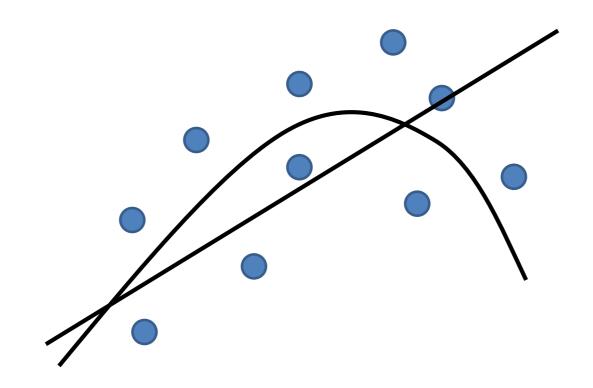


$$2x_1^3 - 2x_1^2x_2 - 3x_1^2 + 4x_2^3 + 4x_2^2 + 5$$

$$|2| + |-2| + |3| + |4| + |4| + |4| = 21$$



# L2 (Ridge) Regularization

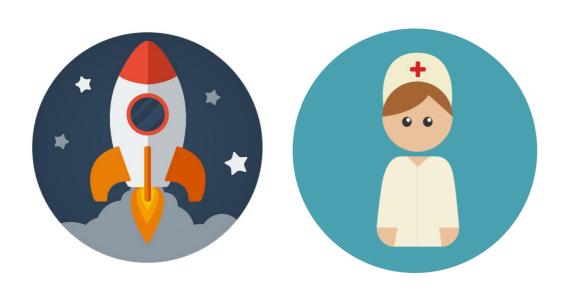


$$2x_1^3 - 2x_1^2x_2 - 34x_1^2 + 48x_1^2 = 6x_1x_2 + 4x_2^2 + 5$$

$$2^2 + (-2)^2 + (-4)^2 + 3^2 + 6^2 + 4^2 = 85$$



# Simple vs Complex Models



Requires LOW ERROR OK if it's COMPLEX

PUNISHMENT on COMPLEXITY should be SMALL

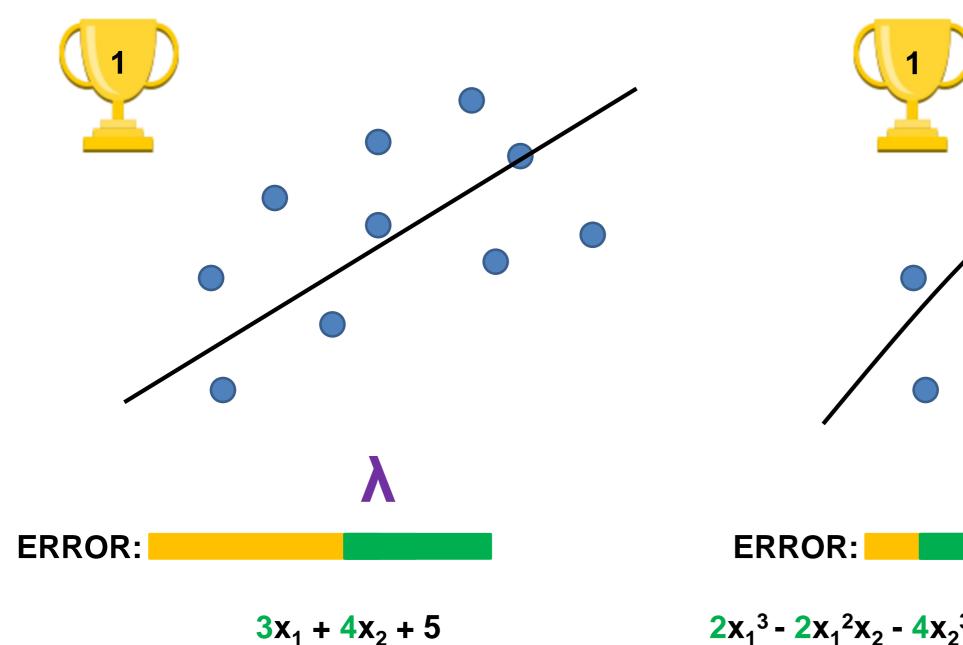


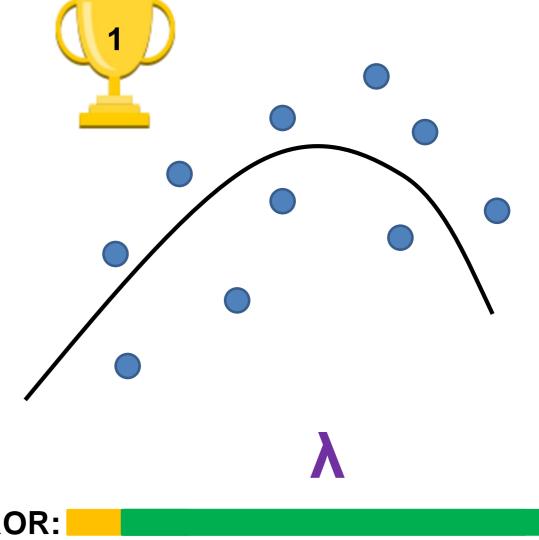
Requires SIMPLICITY OK with ERRORs

PUNISHMENT on COMPLEXITY should be BIG



## The λ Parameter





$$2x_1^3 - 2x_1^2x_2 - 4x_2^3 + 3x_1^2 + 6x_1x_2 + 4x_2^2 + 5$$

SAFAGE A

