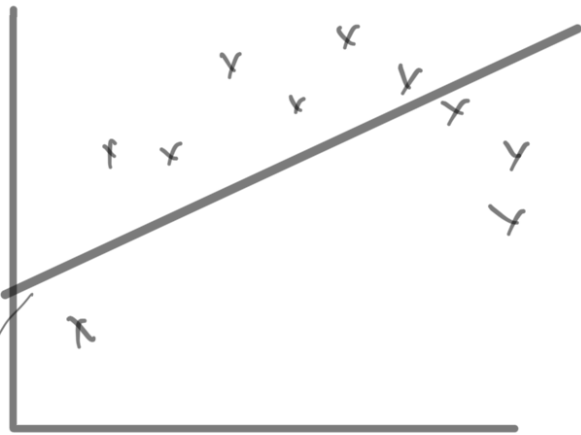


Stanford CS229 Lecture-7

Bias :- Introduced by approximating a complex problem

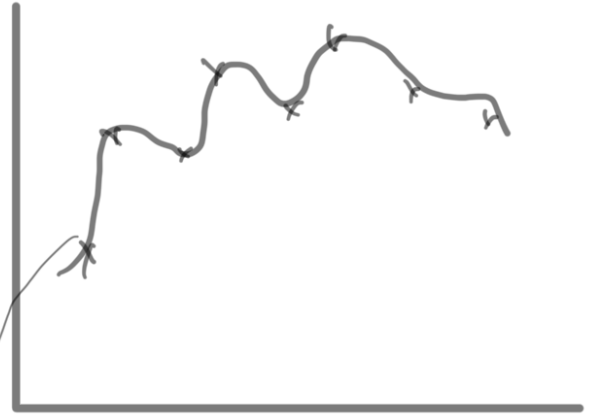
↳ reflects how far is our model from true values

Variance :- Reflects how predictions vary from dataset.



High bias (Underfitting)

→ $\theta_0 + \theta_1 x$ (linear regression)



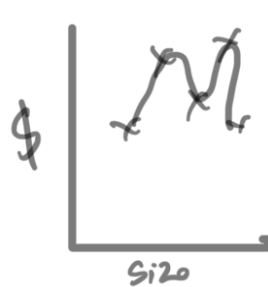
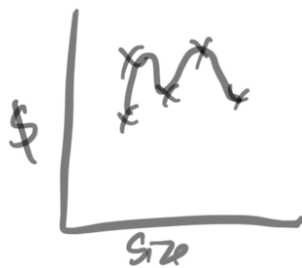
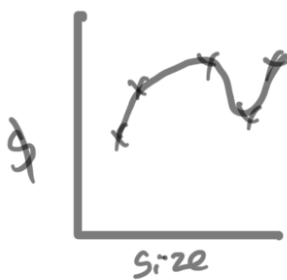
High Variance (Overfitting)

→ $\theta_0 + \theta_1 x + \theta_2 x^2 + \dots$
(Polynomial regression)

High Bias

↳ Here the dataset was pre-conceived to fit a straight line. i.e Bias

High Variance



3 random draws from same dataset

High variance from all the 3 samples as an average (overfitting)

Regularization :-

↳ Generalizing a model to prevent overfitting.

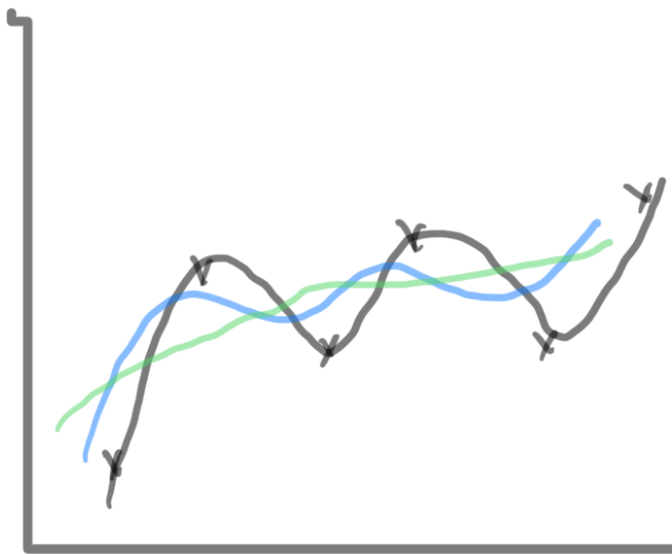
Ex:-

$$\min_{\theta} \frac{1}{2} \sum_{i=1}^m \|y^{(i)} - \theta^T x^{(i)}\|^2 + \lambda \|\theta\|^2$$

↳ Linear Regression optimization objective

↳ Regularization term

Ex:-



$\lambda = 0$
 $\lambda = 1$
 $\lambda = 2$

as you can see
as we increase
 λ (regularisation),
we decrease overfitting