

03/05/20

Q - R Decomposition

$$X = QR$$

- ① Use Gram - Schmit to derive Q from X.
- ② ~~Compute~~ <sup>Compute</sup> entries of R from X & Q

$$\text{Let } X = [\vec{x}_1 | \vec{x}_2 | \vec{x}_3 | \dots]$$

We first want an orthogonal basis from  $\text{Colsp}[X]$

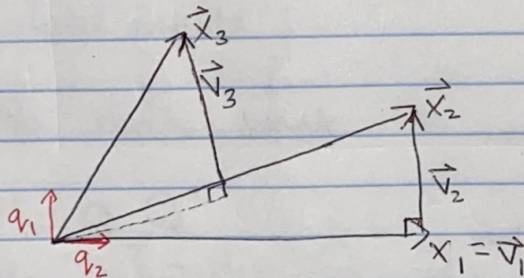
$$\vec{v}_1 = \vec{x}_1$$

$$\vec{v}_2 = \vec{x}_2 - \text{proj}_{\vec{x}_1}(\vec{x}_2)$$

$$\vec{v}_3 = \vec{x}_3 - \text{proj}_{[\vec{x}_1, \vec{x}_2]}(\vec{x}_3) = \vec{x}_3 - \text{proj}_{[\vec{x}_1, \vec{x}_2]}(\vec{x}_3) = \vec{x}_3 - \text{proj}_{\vec{v}_1}(\vec{x}_3) - \text{proj}_{\vec{v}_2}(\vec{x}_3)$$

$$\vdots$$

$$\vec{v}_j = \vec{x}_j - \sum_{i=1}^{j-1} \text{proj}_{\vec{v}_i}(\vec{x}_j)$$



$$\vec{q}_1 = \frac{1}{\|\vec{v}_1\|} \vec{v}_1$$

$$\vec{q}_2 = \frac{1}{\|\vec{v}_2\|} \vec{v}_2 \Rightarrow Q = [\vec{q}_1 | \vec{q}_2 | \dots]$$

$$\vdots$$

$$\vec{q}_j = \frac{1}{\|\vec{v}_j\|} \vec{v}_j$$

$$X = QR$$

$$[\vec{x}_1 | \vec{x}_2 | \vec{x}_3 | \dots] = [\vec{q}_1 | \vec{q}_2 | \vec{q}_3 | \dots]$$

$$\vec{x}_1 = a\vec{q}_1 = \|\vec{x}_1\| \vec{q}_1$$

$$\vec{x}_2 = b\vec{q}_1 + d\vec{q}_2$$

$$\vec{q}_i^T \vec{q}_j \vec{x}_i = \vec{q}_i^T \frac{1}{\|\vec{x}_i\|} \vec{x}_i \vec{x}_2$$

$$= \vec{q}_i^T \frac{\|\vec{x}_i\|}{\|\vec{x}_i\|} \vec{x}_2$$

upper triangular

$$\begin{bmatrix} a & b & c & \dots \\ 0 & d & e & \dots \\ 0 & 0 & f & \dots \\ 0 & 0 & 0 & \dots \end{bmatrix}$$

$$\begin{bmatrix} \vec{q}_1^T \vec{x}_1 & \vec{q}_1^T \vec{x}_2 \\ 0 & \vec{q}_2^T \vec{x}_2 \\ \vdots & \vdots \\ 0 & 0 \end{bmatrix}$$



$$\vec{x}_2 \in \text{Span} \{ \vec{q}_1, \vec{q}_2 \}$$

$$\begin{aligned} \vec{x}_3 = \text{proj}_{[\vec{q}_1, \vec{q}_2]}(\vec{x}) &= \text{proj}_{\vec{q}_1}(\vec{x}_2) + \text{proj}_{\vec{q}_2}(\vec{x}_2) \\ &= H_1 \vec{x}_2 + H_2 \vec{x}_2 \\ &= \vec{q}_1 \vec{q}_1^T \vec{x}_2 + \vec{q}_2 \vec{q}_2^T \vec{x}_2 \\ &= b \vec{q}_1 + d \vec{q}_2 \end{aligned}$$

Recall:

$$H = X(X^T X)^{-1} X^T$$

$$H = Q Q^T$$

$$\vec{\hat{y}} = H \vec{y}$$

not equal to each other

$$\begin{aligned} \vec{b} &= (X^T X)^{-1} X^T \vec{y} \\ \vec{b}' &= (Q^T Q)^{-1} Q^T \vec{y} \\ \vec{b}' &= Q^T \vec{y} \end{aligned}$$

$$SSR = \| \text{proj}_Q(\vec{y}) \|^2 - n \bar{y}^2$$

No correlation means they are orthogonal

If two vectors are orthogonal then you get back exactly 0

Generalization Error:

error when using your model on  $\vec{x}'_Q, \notin \mathcal{D}$   
i.e. in the future.

Overfitting monotonically increases your error

We learned that  $R^2$ , RMSE can't be trusted by looking at a demo.

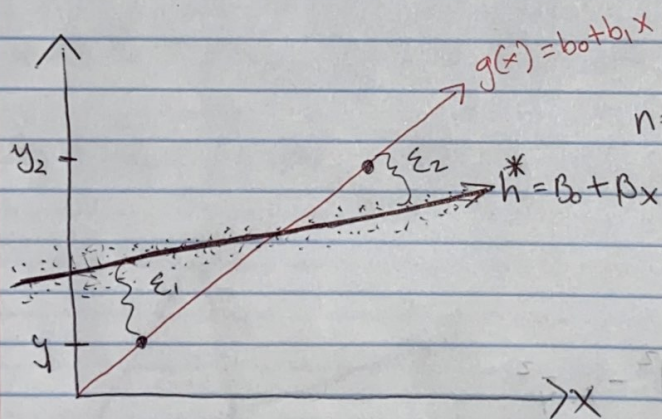


$D \xrightarrow{A} g \xrightarrow{D} R^2, RMSE, \text{etc not trusted}$

$D^* \rightarrow R^2, RMSE, \text{etc are trusted}$

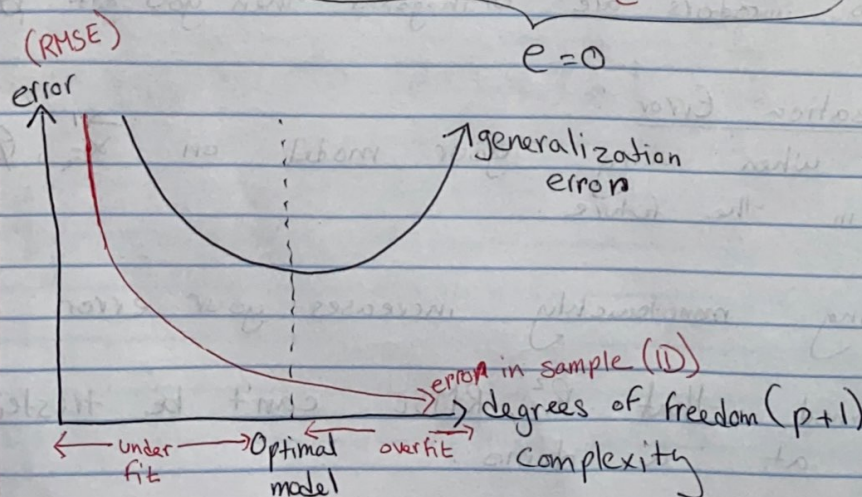
$D$  past  
 $D^*$  future  
time

Assume "Stationarity" which mean  $z's, t, f, h^*$ , measure protocols of  $\vec{x}'s$  and  $y$  do not change with time.



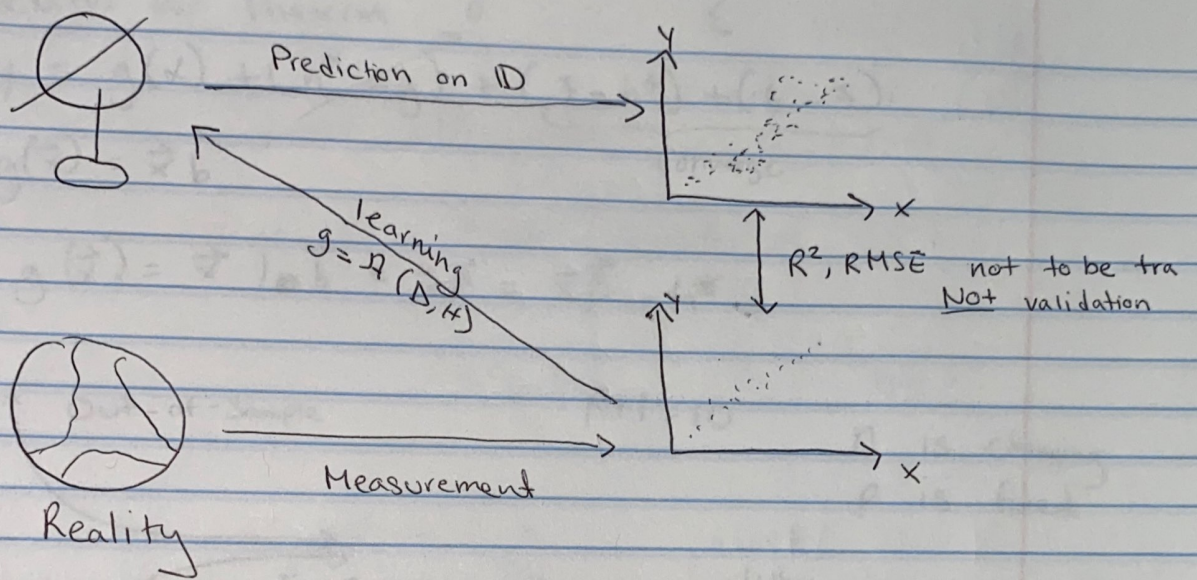
$$y = g + (h^* - g) + (f - h^*) + (t - f)$$

where  $h^*, f, t$  are fixed  
 $\epsilon = 0$



\* Overfitting increases generalization error





$$D = \underbrace{D_{\text{tran}}}_{g = H(D_{\text{tran}}, H)} \cup D_{\text{test}}$$

Simulating the future  
and disjoint

don't let  $A$   
peek in this set.  
Use it to "validate"

↓  
Estimates generalization  
error