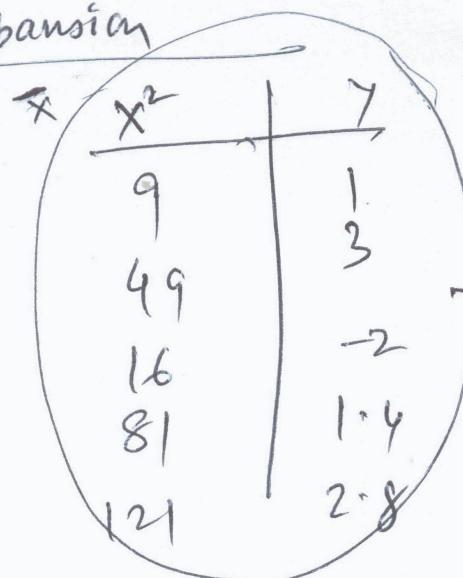


Non-linear Models

- ① Stay with a linear model, but transform our data.
 ↳ kernel methods.
- ② Make the model non-linear
 ↳ neural networks.

Basis function expansion

x	y
3	1
7	3
4	-2
9	1.4
11	2.8



$$\cancel{w^T x = y}$$

$$\cancel{w^T x^2 = y}$$

Regularization

Simple

↓
larger training error

Complex

↓
low training error

Occam's Razor

≠
low test error

How do we force regularization?

Smaller values for the weights \Rightarrow less complex models.

	x	x^2	x^3	
w_0	2	7	9	w_3
	1	3.5	4.5	1.5

→ more complex

linear regression

$$J(w) = \frac{1}{2} (y - Xw)^T (y - Xw)$$

error on the training data.

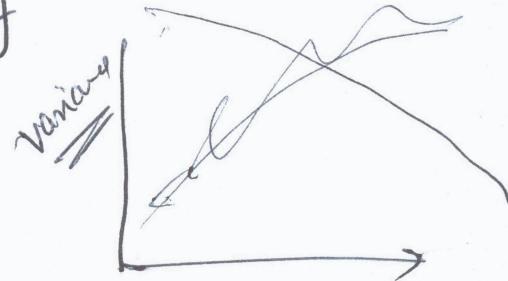
$$\tilde{J}(w) = J(w) + \frac{\lambda}{2} \|w\|_2^2$$

$\|w\|_2^2 \equiv w_1^2 + w_2^2 + \dots$

Regularization
Penalty

a scalar value

$$0 \leq \lambda \dots$$



Ridge Regression

$$\tilde{J}(w) = J(w) + \lambda \|w\|$$

LASSO

$$\frac{d}{dw} \tilde{J}(w) = \frac{d}{dw} \left[\frac{1}{2} (y - Xw)^T (y - Xw) \right]$$

$$\|w\|^2 \equiv w^T w$$

$$+ \frac{\lambda}{2} \frac{d}{dw} \|w\|^2$$

$$= X^T X w - X^T y + \lambda w$$

$$\frac{d}{dw} \tilde{J}(w) = 0$$

$$X^T X w - X^T y + \lambda w = 0$$

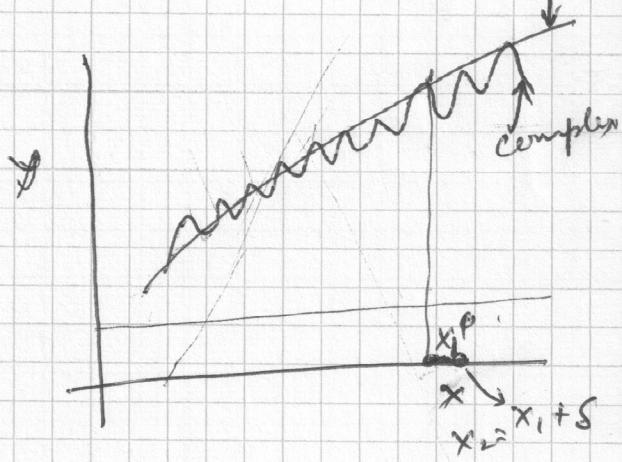
$$w = (X^T X + \lambda I)^{-1} X^T y$$

Identity matrix
 $D \times D$

$$\begin{bmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{bmatrix}$$

Why does a small $\|w\|_2^2$ ensure a simpler model?

Essentially means that all w_i 's are small. simple



$$y = w_0 + w_1 x$$

$$y_1 = w_0 + w_1 x_1$$

$$\begin{aligned} y_2 &= w_0 + w_1 x_2 \\ &= w_0 + w_1 (x_1 + s) \end{aligned}$$

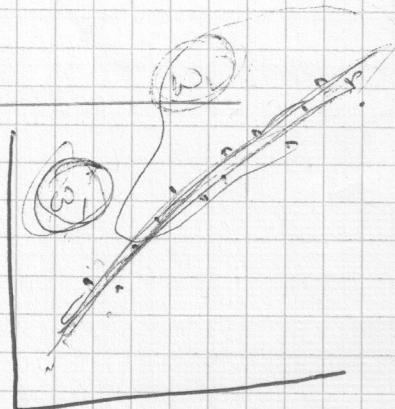
$$y_2 - y_1 = w_1 s$$

$$y = w_0 + w_1 x + w_2 x^2$$

$$y_1 = w_0 + w_1 x_1 + w_2 x_1^2$$

$$y_2 = w_0 + w_1 (x_1 + s) + w_2 (x_1 + s)^2$$

$$y_2 - y_1 = w_1 s + (2w_2 x_1 s) + w_2 s^2$$



Ridge

$$\|w\|_2^2 \rightarrow (w_1^2 + w_2^2 + w_3^2)$$

$$\begin{pmatrix} 0 & 0 & \vdots \end{pmatrix}$$

LASSO

$$\|w\|_1 \rightarrow (|w_1| + |w_2| + |w_3|)$$

$$\begin{pmatrix} 0 & 0 & \ddots \end{pmatrix}$$