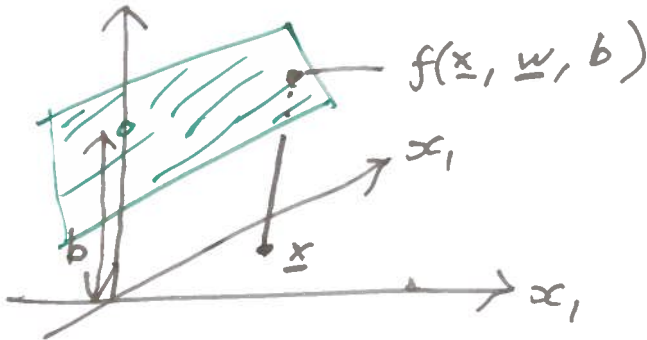
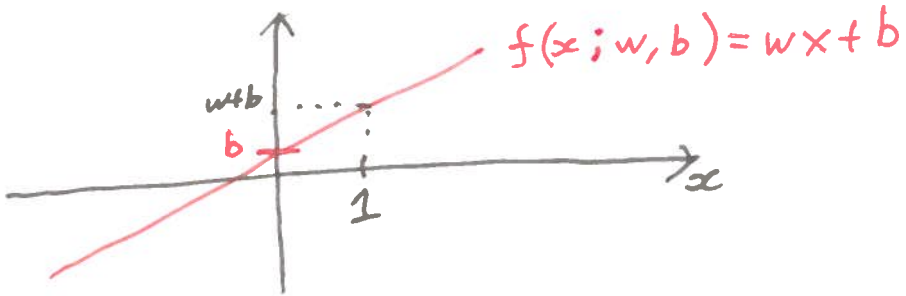


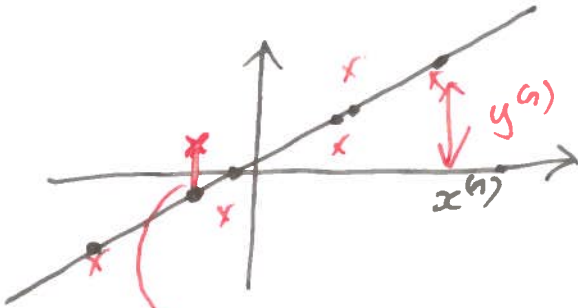
Linear Functions

L2 2017 ①



$$\begin{aligned} f(\underline{x}; \underline{w}, b) &= w_1 x_1 + w_2 x_2 + b \\ &= \underline{w}^T \underline{x} + b \end{aligned}$$

Data $\{ (\underline{x}^{(n)}, y^{(n)}) \}_{n=1}^N$



Residual $y^{(n)} - f(x^{(n)}; w, b)$

In vector and matrix format:

$$y = \begin{bmatrix} y^{(1)} \\ y^{(2)} \\ \vdots \\ y^{(N)} \end{bmatrix}$$

$$X = \begin{bmatrix} - \underline{x}^{(1)T} - \\ - \underline{x}^{(2)T} - \\ \vdots \\ x_1^{(N)} \dots x_D^{(N)} \end{bmatrix}$$

$N \times 1$ matrix

D -dimensional regression.

Python: y is $(N, 1)$ vector

Then $y[:, \text{None}]$ is $(N, 1)$ array

↳ (Really NumPy)

$$\underline{f} = \begin{bmatrix} f(\underline{x}^{(1)}; \underline{w}, b) \\ \vdots \\ f(\underline{x}^{(N)}; \underline{w}, b) \end{bmatrix}$$

Least squares fitting

$$\text{Minimize } \sum_{n=1}^N (y^{(n)} - f(x^{(n)}; \underline{w}, b))^2$$

$$\text{Minimize } (\underline{y} - \underline{f})^T (\underline{y} - \underline{f})$$

Models with zero intercept ($b=0$)

$$f(\underline{x}; \underline{w}) = \underline{w}^T \underline{x} = \underline{x}^T \underline{w}$$

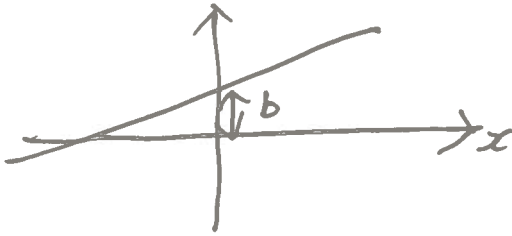
$$\underline{f} = \underset{N \times 1}{X} \underset{N \times D}{w} \approx \underset{D \times 1}{y}$$

"Linear map"

$$\left. \begin{aligned} g(\underline{x} + \underline{z}) &= g(\underline{x}) + g(\underline{z}) \\ g(c\underline{x}) &= c g(\underline{x}) \end{aligned} \right\}$$

Matlab: $w_fit = X \setminus y;$

Python $= np.linalg.lstsq(X, y)[0]$



$$X' = \begin{bmatrix} 1 & -x^{(1)T} \\ 1 & -x^{(2)T} \\ \vdots & \vdots \\ 1 & -x^{(N)T} \end{bmatrix}$$

$$\underline{w}' = \operatorname{argmin} \| \underline{y} - X' \underline{w}' \|^2$$

$$\begin{aligned} \text{Fits } \underline{y} \text{ with } \underline{f} = X' \underline{w}' &= w_1 + X w_{2:D+1} \\ &= b + X \underline{w} \end{aligned}$$

$$\underline{f} = \underline{\Phi} \underline{w}$$

↑ Any representation we like
Each row is a datapoint

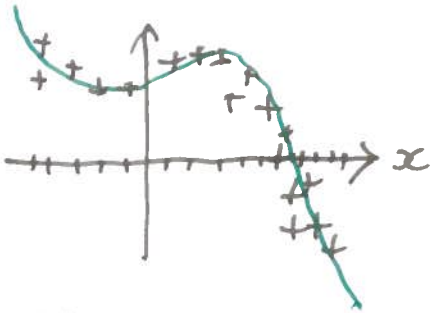
$$\underline{\Phi} = \begin{bmatrix} \underline{\phi}(x^{(1)})^T \\ \vdots \\ \underline{\phi}(x^{(n)})^T \end{bmatrix}$$

Example

$$\underline{\phi}(x) = [1 \ x \ x^2 \ x^3]^T$$

Fit $y \approx \underline{f} = \underline{\Phi} \underline{w}$ (by "linear regression")

$$f(x) = w_1 + w_2 x + w_3 x^2 + w_4 x^3$$



cubic polynomial fit