Find w 1

Observations: $x = (x_1, x_2, ..., x_n)^T$ Target: $t = (t_1, t_2, ..., t_n)^T = y(x, W) + \epsilon$

Model: $y(x, w) = w_0 + x_1 w_1 + x_2 w_2 + ... + x_n w_n$

$$t = y(x, w) + N(0, \beta^{-1})t = N(y(x, w), \beta^{-1}) \text{ with } \beta = \sigma^{-1}$$

$$p(t) = N(t|y(x, w), \beta^{-1})$$

$$\to p(t_n) = N(t_n|y(x_n, w), \beta^{-1})$$

$$\to p(t|x, w, \beta) = \prod_{n=1}^{N} N(t_n|y(x_n, w), \beta^{-1})$$

$$N(t_n|y(x, w), \beta^{-1}) = \frac{1}{\sqrt{2\pi\beta^{-1}}} e^{-(t_n - y(x, w))^2\beta/2}$$

Using Maximum Likelihood, we have:

$$p(t|x,w,\beta) \to max$$

$$log(\prod_{n=1}^{N} e^{-(t_n - y(x,w))^2 \beta/2}) \to max$$

$$\sum_{n=1}^{N} (t_n - y(x,w))^2 \to min$$
 We have: $\mathbf{x} = \begin{bmatrix} 1 & x_1 \\ \dots \\ 1 & x_n \end{bmatrix}, \ \mathbf{t} = \begin{bmatrix} t_1 \\ \dots \\ t_n \end{bmatrix}, \ \mathbf{w} = \begin{bmatrix} w_0 \\ w_1 \end{bmatrix} \to \mathbf{y} = \begin{bmatrix} y_1 \\ \dots \\ y_n \end{bmatrix} = \begin{bmatrix} w_1 x_1 \\ \dots \\ w_n x_n \end{bmatrix} = xw$
$$\to t - \mathbf{y} = \begin{bmatrix} t_1 - y_1 \\ \dots \\ t_n - y_n \end{bmatrix} \to ||t - \mathbf{y}||_2^2 = \sum_{n=1}^{N} (t_n - y_n)^2$$

$$L = ||t - xw||_2^2$$

$$\frac{\sigma L}{\sigma w} = -2x^T (t - xw) = 0$$

$$x^T t = x^T xw$$

$$w = (x^T x)^{-1} x^T t$$

Machine Learning

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$\mathbf{X}^T X$ inverse when X full rank 2

When X full rank, Xa = 0 has the trivial solution a = 0

- \rightarrow The reduced row echelon form of X is I
- $\rightarrow \prod_{i=1}^{N} E_i A = I$ with E_i is the matrix that A reduced row $\rightarrow A = \prod_{i=1}^{i=1} E_i^{-1} I \rightarrow A \ inverse$