

ML WEEK3

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We have a data set of observations $\mathbf{x} = (x_1, x_2, \dots, x_N)^T$, representing N observations of the scalar variable x and their corresponding target values $\mathbf{t} = (t_1, t_2, \dots, t_N)^T$ make predictions for some new value of the input variable x

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

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

By maximum likelihood we have : $A = p(t|x, w, \beta) = \prod(N(t|y(x, w), \beta^{-1}))$

$$\Rightarrow A = \left(\frac{1}{\sqrt{2\pi\beta^{-1}}}\right)^N \exp\left\{-\frac{\beta}{2}(t_n - y(x, w))^2\right\}$$

$$\Rightarrow \ln A = -\frac{N}{2} \ln(2\pi) + \frac{N}{2} \ln(\beta^{-1}) - \frac{1}{2} \beta \sum (t_n - y(x_n, w))^2$$

$\ln A$ max when $B = -\frac{1}{2} \beta \sum (t_n - y(x_n, w))^2$ gets max or we'll say $-B$ gets min

Suppose: x is matrix of all observations : $X = [1, x]$, w is a matrix with entries are w_0 and w_1 then $-B = \|Xw - t\|_2^2$

Apply matrix calculus to find the derivative of $-B$, we get:

$$(-B)' = 2X^T(Xw - t)$$

$$\text{Let } (B)' = 0 \Rightarrow w = (X^T X)^{-1} X^T t$$

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Proof $A^T A$ is invertible if and only A has full rank(A 's columns are linear independent)

Suppose A is matrix with linear independent columns($a_1, a_2, a_3 \dots a_n$), that means $a_1 x_1 + a_2 x_2 + a_3 x_3 + \dots a_n x_n = 0$ and $x_1 = x_2 = x_3 = \dots x_n = 0$ or we'll say $Ax=0$ then nullspace of A is full of vector 0 : $N\{A\} = \{0\}$

We obviously don't know if A is invertible or not but if we can proof that $A^T A$ is invertible then what we suppose about A is correct.

We know that if a matrix is square and linearly dependent columns then it is invertible

Suppose vector v such that $v \in N\{A^T A\}$

$$\Rightarrow A^T A v = 0 \Rightarrow v^t A^T A v = v^t 0 \Rightarrow (A v)^T (A v) = 0 \Rightarrow \|A v\|_2^2 = 0$$

$\Rightarrow A v = 0$ then v is also in the nullspace of A , for what we suppose is that A has linear independent columns therefore v has to be zeros vector

or $N\{A^T A\} = N\{A\} \Rightarrow$ the only solution for the equation $A^T A v = 0$ is $v_1 = v_2 = \dots v_n = 0$ or we'll say $A^T A$ has linear dependent columns moreover, $A^T A$ is square matrix $\Rightarrow A^T A$ is invertible under the condition that A 's columns are linear independent or A has full rank