

1) The linear regression can be expressed as to the

$$E(w) = \frac{1}{N} \|Xw - y\|^2$$

Minimizing the function with respect to w leads to the optimal w as $(X^T X)^{-1} X^T y$. This solution holds only when $X^T X$ is non singular.

The objective function is commonly minimized instead

$$E_2(w) = \|Xw - y\|^2 + \lambda \|w\|^2$$

where $\lambda > 0$

$$\begin{aligned} E_2(w) &= (Xw - y)^T (Xw - y) + \lambda (w^T w) \\ &= w^T X^T X w - 2w^T X^T y + y^T y + \lambda (w^T w) \end{aligned}$$

Differentiating w to minimize $E_2(w)$, we get

$$2X^T X w - 2X^T y + \lambda w = 0$$

$$\Rightarrow (2X^T X + \lambda)w = 2X^T y$$

$$\Rightarrow w = \text{inv}(2X^T X + \lambda) + 2X^T y$$

Earlier the problem was that $X^T X$ could be singular. As $\lambda > 0$ and user defined we can use λ such that $2X^T X + \lambda$ is not singular and hence invertible. Hence a solution to w can be found as $\text{inv}(X^T X + \lambda)$ exists.

2) The hat matrix $H = X(X^T X)^{-1} X^T$, given $X^T X$ is invertible.

a) H is symmetric if $H = H^T$ we have

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

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

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

$$= X(X^T X)^{-1} X^T \quad (\text{Here } X^T X \text{ is invertible})$$

$$H^T = H \rightarrow \text{hence } H \text{ is symmetric.}$$

b) Show that $H^k = H$, for any positive integer k .

Let $k=2$

H is a $n \times n$ square matrix and it is idempotent which can be verified as follows $\Rightarrow H^2 = H \cdot H$

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

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

Here $(X^T X)(X^T X)^{-1} = I$ because $A \cdot A^{-1} = I$

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

Hence for any positive integer k , $H^k = H$. Hence proved.

c) Show that $(I-H)^k = I-H$ (for any positive integer k)

Here I is any identity matrix of size N .

Similarly, as in part b,

Let $k=2$

$$(I-H)^2 = (I-H)(I-H) = I \cdot I - 2HI + H \cdot H = I - 2H + H = I - H$$

($H \cdot H = H \rightarrow$ proved in part b)

Hence for any positive integer k , $(I-H)^k = I-H$. Hence proved.

d) Show that $\text{trace}(H) = d+1$, where the trace is the sum of diagonal elements.

It is known that $\text{trace}(AB) = \text{trace}(BA)$ — (1)

Now,

$$\begin{aligned} \text{trace}(H) &= \text{trace}\left(X(X^T X)^{-1} X^T\right) \\ &= \text{trace}\left((X^T X)^{-1} X^T X\right) \text{ — from (1)} \\ &= \text{trace}(I) \end{aligned}$$

↑
because $A^{-1}A = I$
 $(X^T X)^{-1}(X^T X) = I$

Then $\text{trace}(H) = d+1$. Hence proved.