

REGRESSION& REGULARIZATION EXERCISE

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Developing the Ridge coefficient. Let $L(\cdot)$ be:

$$L(y, \hat{y}) = \sum_{i=1}^N \left(y^{(i)} - \hat{y}^{(i)} \right)^2 + \lambda \|w\|_2^2$$

we need to show that $w_{Ridge} = (X^T X + \lambda I)^{-1} X^T y$

$$\begin{aligned} L(y, \hat{y}) &= \sum_{i=1}^N \left(y^{(i)} - \hat{y}^{(i)} \right)^2 + \lambda \|w\|_2^2 \\ &= (y - \hat{y})^T (y - \hat{y}) + \lambda w^T w \\ &= (y - X^T w)^T (y - X^T w) + \lambda w^T w \\ &= y^T y - 2w^T X^T y + w^T X^T X w + \lambda w^T w \end{aligned}$$

Taking the derivative by w and using FOC:

$$\begin{aligned} \frac{\partial L}{\partial w} &= 0 - 2X^T y + 2X^T X w + 2\lambda w \stackrel{FOC}{=} 0 \\ \iff (X^T X + \lambda I) w &= X^T y \\ \implies w_{Ridge} &= (X^T X + \lambda I)^{-1} X^T y \end{aligned}$$

Bonus question: Let $G \sim N(1, \sigma^2)$ and $X' = X \cdot G$ s.t $EX' = E[X \cdot G] = EX$
First notice the followig property about G

$$Var(G) = \sigma^2 = E(G - EG)^2 = EG^2 - E^2 G \stackrel{EG=1}{=} EG^2 - 1 \iff EG^2 = \sigma^2 + 1$$

Now looking back on the loss of OLS we can write:

$$\begin{aligned} L &= \|y - w^T X'\|^2 \\ &= (y - Gw^T X)^T (y - Gw^T X) \\ &= [y^T y - 2Gw^T X^T y + G^2 w^T X^T X w] \\ &= [y^T y - G2w^T X^T y + G^2 w^T X^T X w] \end{aligned}$$

$$\begin{aligned} EL &= E[y^T y - G2w^T X^T y + G^2 w^T X^T X w] \\ &= E[y^T y] - 2EGEw^T X^T y + EG^2 E[w^T X^T X w] \\ &= E[y^T y] - 2Ew^T X^T y + (1 + \sigma^2) E[w^T X^T X w] \\ &= MSE(L) + \sigma^2 E[w^T X^T X w] \\ &= MSE(L) + \sigma^2 w^T E[X^T X] w \\ &= MSE(L) + \sigma^2 \cdot \sigma_X^2 w^T w \\ &= MSE(L) + \sigma^2 \cdot w^T w \end{aligned}$$

* if X is noramilized and $\forall l, m \in \{1, 2, \dots, p\}$ s.t $l \neq m$ $Cov(x_l, x_m) = 0$ then $\sigma_X^2 = I_p$ and $X \sim N(0, I_p)$
From the above we can see that we get the Ridge Loss with $\lambda = \sigma^2$