

CS/ECE/ME532 Period 11 Activity

Estimated time: 45 mins for Q1 and 20 mins for Q2.

1. See `period_11.ipynb`.

2. Let a 4-by-2 matrix \mathbf{X} have SVD $\mathbf{X} = \mathbf{U}\mathbf{S}\mathbf{V}^T$ where $\mathbf{U} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix}$, $\mathbf{S} = \begin{bmatrix} 1 & 0 \\ 0 & \gamma \end{bmatrix}$, and $\mathbf{V} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$.

a) Express the solution to the least-squares problem $\arg \min_{\mathbf{w}} \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ as a function of \mathbf{U} , \mathbf{S} , \mathbf{V} , and \mathbf{y} .

b) Let $\mathbf{y} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$. Find the weights \mathbf{w} that minimize $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ as a function of γ . Calculate $\|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2$ and $\|\mathbf{w}\|_2^2$ as a function of γ for this value of \mathbf{w} . What happens to $\|\mathbf{w}\|_2^2$ as $\gamma \rightarrow 0$?

c) Now consider a “low-rank” inverse. Instead of writing

$$(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T = \sum_{i=1}^p \frac{1}{\sigma_i} \mathbf{v}_i \mathbf{u}_i^T$$

where p is the number of columns of \mathbf{X} (assumed less than the number of rows), we approximate

$$(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \approx \sum_{i=1}^r \frac{1}{\sigma_i} \mathbf{v}_i \mathbf{u}_i^T$$

In this approximation we only invert the largest r singular values, and ignore all of them smaller than σ_r . If $r = 1$, use the low-rank inverse to find \mathbf{w} , $\|\mathbf{y} - \mathbf{X}\mathbf{w}\|_2^2$,

and $\|\mathbf{w}\|_2^2$ when $\mathbf{y} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$ as in part b). Compare $\|\mathbf{y} - \mathbf{X}\mathbf{w}\|_2^2$, and $\|\mathbf{w}\|_2^2$ to the results for part b).