
Theorem 1. Let A be $n \times m$ matrix over \mathbb{R} and let h be a vector. Then the solution to the problem

$$\min_h \|Ah\|_2 \quad \text{such that} \quad \|h\|_2 = 1,$$

is the right singular vector corresponding to the least singular value in the singular value decomposition of the matrix A .

Proof. Let $A = U\Sigma V^T$, where $U \in \mathbb{R}^{n \times n}$, $\Sigma \in \mathbb{R}^{n \times m}$ and $V \in \mathbb{R}^{m \times m}$ be the singular value decomposition of A . $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_m$ be the singular values of A ordered in decreasing manner in Σ .

$$\begin{aligned} \min_h \|Ah\|_2 &= \min_h \|U\Sigma V^T h\|_2 \\ &= \min_h \|\Sigma V^T h\|_2 \quad \because \|Ux\|_2 = \|x\|_2 \text{ if } U \text{ is orthogonal} \end{aligned}$$

Let $V^T h = y$ ($y \in \mathbb{R}^{m \times 1}$) $\implies Vy = h$ and also $\|h\|_2 = \|Vy\|_2 = \|y\|_2$. Therefore minimizing over h is same as minimizing over y .

$$\min_h \|Ah\|_2 = \min_y \|\Sigma y\|_2$$

Now, $\Sigma y = (\sigma_1 y_1, \sigma_2 y_2, \dots, \sigma_m y_m, 0, \dots, 0)$.

Now,

$$\begin{aligned} \|\Sigma y\|_2^2 &= \sum_{i=1}^n |\sigma_i y_i|^2 \\ &= \sum_{i=1}^m |\sigma_i y_i|^2 \\ &\geq \sigma_m^2 \sum_{i=1}^m |y_i|^2 \\ &\geq \sigma_m^2 \quad \|\|y\|_2 = 1 \end{aligned}$$

The above minimum is achieved for $y = (0, 0, \dots, 1)$ (all zeros except 1 at m^{th} location).

$\therefore \min_y \|\Sigma y\|_2 = \sigma_m$ for $y = (0, 0, \dots, 1)$ where 1 is at m^{th} location.

Since $h = Vy$, the solution h is the m^{th} column (last column) of orthogonal matrix V i.e, solution h is the right singular vector corresponding to the least singular value. ■