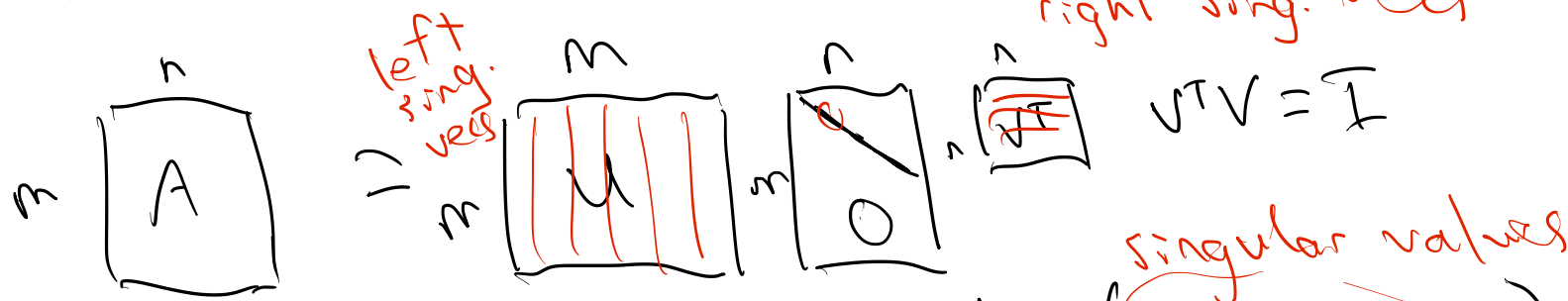


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The Singular Value Decomposition (SVD)

$A \in \mathbb{R}^{m \times n}$ $m \geq n$. We can write $A = U \Sigma V^T$



$$\Sigma = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n)$$
$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$$

$$\|A\|_2 = \|\cancel{U} \Sigma \cancel{V}^T\|_2 = \|\Sigma\|_2 = \sup_{\|x\|_2=1} \|\Sigma x\|_2 = \sqrt{\sum_{i=1}^n x_i^2 \sigma_i^2} = \sigma_1$$

$$\|A\|_F = \|\Sigma\|_F = \sqrt{\sum \sigma_i^2}$$

SVD existence (sketch)

$$\|A\|_2 = \sigma_1 = \sup_{\|x\|_2=1} \|Ax\|_2 \quad \text{Let } \|Av_1\|_2 = \sigma_1, \quad \|v_1\|_2 = 1$$

$$u_1 = Av_1 / \sigma_1, \quad \|u_1\|_2 = 1$$

$$\bar{U} = [u_1 \quad u_2] \quad \bar{V} = \begin{bmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{bmatrix}$$

$$\bar{U}^T A \bar{V} = \begin{bmatrix} u_1^T A v_1 & \dots \\ \vdots & \vdots \\ u_{m-1}^T A v_1 & \dots \\ & \vdots \\ & & B \end{bmatrix} \quad \begin{matrix} u_1^T A v_1 = \sigma_1, u_1^T u_1 = \sigma_1 \\ u_2^T A v_1 = \sigma_1, u_2^T u_1 = 0 \end{matrix}$$

$$\left\| \begin{pmatrix} \sigma_1 & w^T \\ 0 & B \end{pmatrix} \begin{pmatrix} \sigma_1 \\ w \end{pmatrix} \right\|_2 = \left\| \begin{pmatrix} \sigma_1^2 + w^T w \\ Bw \end{pmatrix} \right\|_2 \approx \sigma_1^2 + w^T w$$

$$\|S\|_2 = \|\bar{U}^T A \bar{V}\|_2 = \|A\|_2 = \sigma_1 \implies w^T w = \|w\|_2^2 = 0 \implies w = 0$$

Assume $B = \underline{U} \underline{\Sigma} \underline{V}^T$ (induction)

$$A = \bar{U} \begin{pmatrix} \sigma_1 & 0 \\ 0 & B \end{pmatrix} \bar{V}^T = \bar{U} \begin{bmatrix} 1 & 0 \\ 0 & \underline{U} \end{bmatrix} \begin{bmatrix} \sigma_1 & 0 \\ 0 & \underline{\Sigma} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & \underline{V}^T \end{bmatrix} \bar{V}^T$$

unique?