

Orthogonal Vectors and Matrices

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix} \quad \begin{array}{c} \text{transpose} \\ A^T = \begin{bmatrix} a_{11} & a_{21} & a_{31} \\ a_{12} & a_{22} & a_{32} \end{bmatrix} \end{array}$$

$$\text{adj} \quad A^* = \begin{bmatrix} \overline{a_{11}} & \overline{a_{21}} & \overline{a_{31}} \\ \overline{a_{12}} & \overline{a_{22}} & \overline{a_{32}} \end{bmatrix}$$

Hermitian

$$A = A^*$$

adjoint

Symmetric

$$A \in \mathbb{R}^{n \times n}, \quad A = A^T$$

Real case:

orthogonal vectors, \vec{x}, \vec{y}

$$\vec{x}^T \vec{y} = 0$$

orthonormal

orthogonal & normalized

orthonormal set $\{\vec{q}_1, \dots, \vec{q}_n\}$ $q_i \in \mathbb{R}^m$

Let $\vec{v} \in \mathbb{R}^m$

$$\vec{r} = \vec{v} - (q_1^T \vec{v}) q_1 - (q_2^T \vec{v}) q_2 - \dots - (q_n^T \vec{v}) q_n$$

$$\vec{q}_i^T \vec{r} = 0 \quad (\vec{r} \text{ orthogonal to } q_i)$$

for $m=n$ decompose \vec{v} as

$$\vec{v} = \sum_{i=1}^n (q_i^T \vec{v}) q_i = \sum_{i=1}^n (q_i \vec{q}_i^T) \vec{v}$$

Orthogonal Matrix

$Q \in \mathbb{R}^{n \times n}$ is orthogonal if

$$Q^T Q = I$$

$\left(\begin{array}{l} Q \in \mathbb{C}^{n \times n}, \text{ unitary} \\ Q^* Q = I \end{array} \right)$

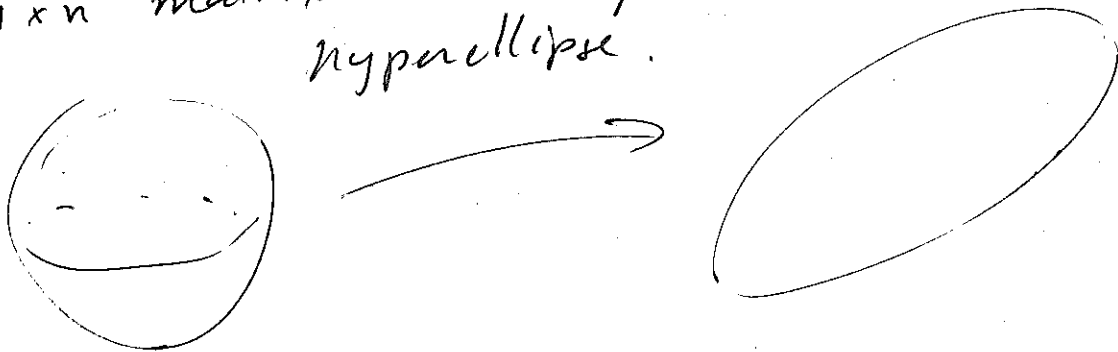
$$\left[\begin{array}{c} \text{---} \\ \text{---} \\ \text{---} \end{array} \begin{array}{c} \vec{q}_1^T \\ \vec{q}_2^T \\ \vdots \\ \vec{q}_n^T \end{array} \text{---} \right] \left[\begin{array}{c} \downarrow \\ \vec{q}_1 \\ | \\ | \\ | \\ \vec{q}_2 \\ \dots \\ \vec{q}_n \\ | \\ | \\ | \end{array} \right] = \left[\begin{array}{c} 1 \\ \dots \\ 1 \end{array} \right]$$

mult by Q : rotation ($\det = 1$), or reflection ($\det = -1$)

SVD

T&B: "Many problems of linear algebra can be better understood if we first ask the question: what if we take the SVD?"

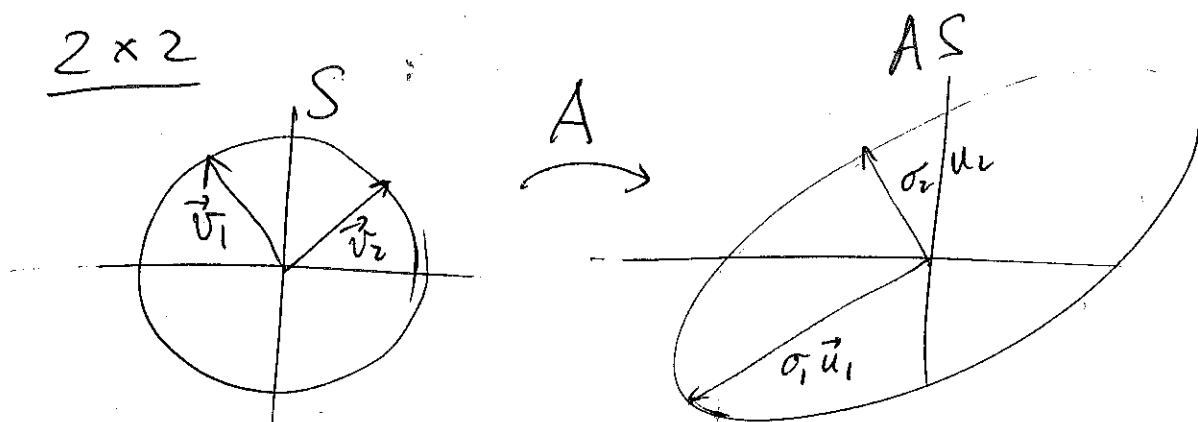
$n \times n$ matrix A maps unit sphere to hyperellipse.



Hyperellipse: stretch sphere by factors $\sigma_1, \dots, \sigma_n$ in some orthogonal directions $\vec{u}_1, \dots, \vec{u}_n$

$\sigma_i \vec{u}_i$: principal semi-axis of the hyperellipse. lengths $\sigma_1, \dots, \sigma_n$

2x2



Singular Values the lengths $\sigma_1, \dots, \sigma_n$ of the n principal semi-axes of ellipse. AS

Left Singular Vectors $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n$ in directions of principal semi-axes of AS

Right Singular Vectors $\vec{v}_1, \dots, \vec{v}_n$ preimages of $\sigma_j \vec{u}_j$
 $A \vec{v}_j = \sigma_j \vec{u}_j$

Reduced SVD

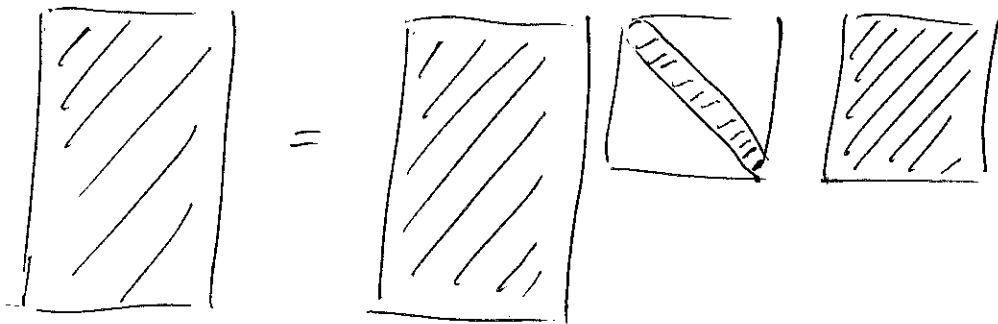
A $m \times n$

$$A v_j = \sigma_j u_j \quad 1 \leq j \leq n$$

or

$$\begin{bmatrix} A \\ \vdots \\ \vdots \\ \vdots \end{bmatrix}_{m \times n} \begin{bmatrix} | & | & | \\ \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_n \\ | & | & | \end{bmatrix}_{n \times n} = \begin{bmatrix} | & | & | \\ \vec{u}_1 & \vec{u}_2 & \dots & \vec{u}_n \\ | & | & | \end{bmatrix}_{m \times n} \begin{bmatrix} \sigma_1 & & & \\ & \sigma_2 & & \\ & & \ddots & \\ & & & \sigma_n \end{bmatrix}_{n \times n}$$
$$AV = \hat{U} \hat{\Sigma}$$

V, \hat{U} orthonormal columns
 V orthogonal matrix



$$A = \hat{U} \hat{\Sigma} V^T$$

or $(A = \hat{U} \hat{\Sigma} V^*)$

Full SVD

Columns of U are n orthonormal vectors in m -dim space. If $n < m$, find $m-n$ additional orthonormal vectors to complete basis for \mathbb{R}^m extend \hat{U} to orthogonal matrix U .

$$\begin{matrix} A & = & U & \Sigma & V^T \\ m \times n & & m \times m & m \times n & n \times n \end{matrix}$$

Rank-deficient A

A $m \times n$, rank $r < n$

$$\underbrace{\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r}_{\text{non-zero}} > \underbrace{\sigma_{r+1} = \dots = \sigma_n}_{\text{zero}} = 0$$

$$A = \sum_{i=1}^r \sigma_i \vec{u}_i \vec{v}_i^T$$

$$\begin{matrix} A & = & U & \Sigma & V^* \\ m \times n & & m \times m & m \times n & n \times n \end{matrix}$$

$$A \in \mathbb{C}^{m \times n}$$

Every A has svd. σ_i uniquely determined

(A square, σ_i distinct $\Rightarrow u_i, v_i$ unique up to complex sign)

Projector

projector: square matrix that satisfies

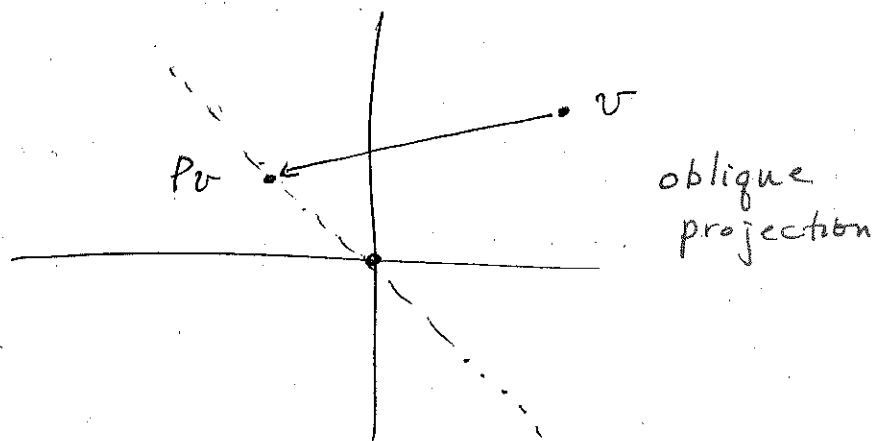
$$P^2 = P \quad (\text{idempotent})$$

orthogonal projector + non-orthogonal or oblique projector

$$v \in \text{Range}(P)$$

$$v = Px \Rightarrow Pv = P^2x = Px = v$$

$$P(Pv - v) = P^2v - Pv = Pv - Pv = 0$$



Complementary Projector

P projector $\Rightarrow (I - P)$ also projector

$$\begin{aligned} (I - P)^2 &= I - 2P + P^2 = I - 2P + P \\ &= I - P \quad \checkmark \end{aligned}$$

Orthogonal Projector

(not an orthogonal matrix)

$\text{range}(P) \perp \text{range}(I-P)$

Theorem Projector P is orthogonal projector iff $P = P^T$

Proof show $Px \perp (I-P)y \quad \forall x, y \iff P = P^T$ (2)

$$\textcircled{1} (Px)^T [(I-P)y] = x^T P^T (y - Py) = x^T P^T y - x^T P^T P y = 0$$

Assume (2). Then

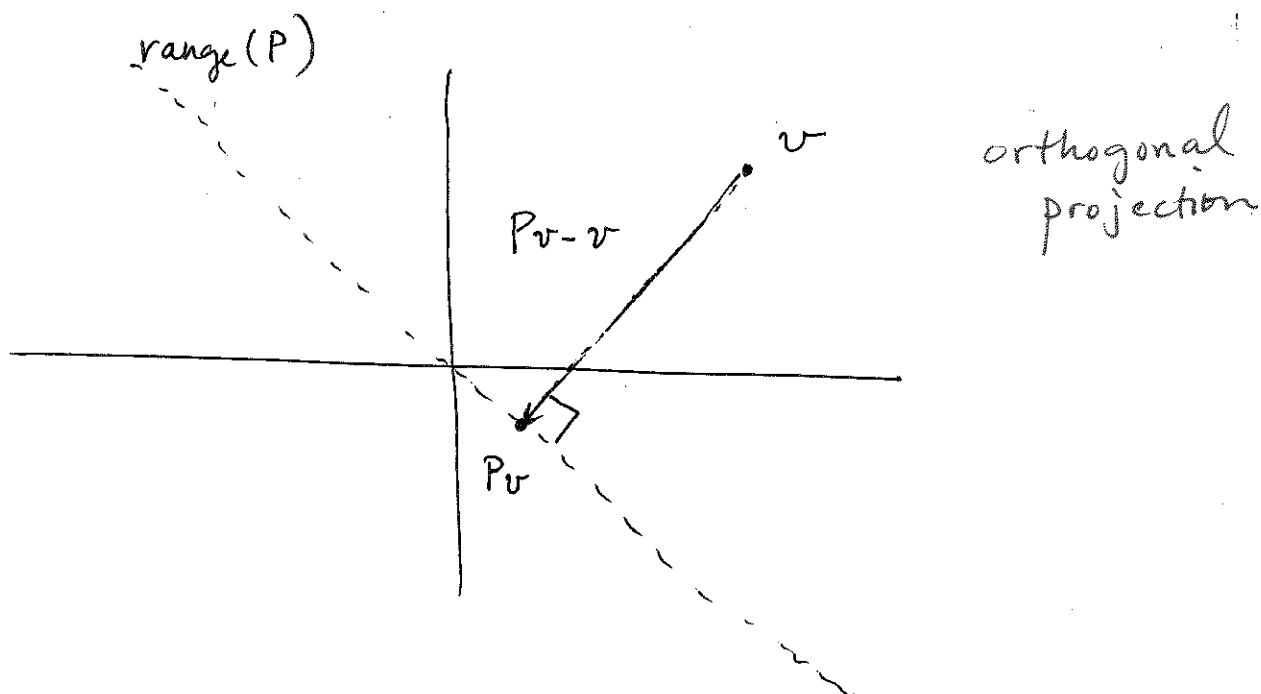
$$\begin{aligned} x^T P^T y - x^T P^T P y &= x^T P y - x^T P^2 y \\ &= x^T P y - x^T P y = 0 \Rightarrow \textcircled{1} \checkmark \end{aligned}$$

Assume (1). Then $x^T P^T y = x^T P y \quad \forall x, y$

In particular $e_i^T P^T e_j = e_i^T P e_j \quad \forall i, j \in \{1, \dots, n\}$

$\Rightarrow P^T = P$

taking transpose $P = P^T P \left\{ \Rightarrow P^T = P \Rightarrow \textcircled{2} \checkmark \right.$



Projection with an orthonormal basis.

orthogonal projector P ($= P^T$)

$$P = U \Sigma U^T$$

$$\text{and } P^2 = U \Sigma U^T U \Sigma U^T = U \Sigma^2 U^T$$

$$\Rightarrow \Sigma = \Sigma^2$$

Singular values are all 0 or 1.

$$(r = \text{rank } P \leq n) \quad P = \sum_{i=1}^r \vec{u}_i \vec{u}_i^T = \hat{U} \hat{U}^T, \text{ where } \hat{U} = \begin{bmatrix} | & & | \\ u_1 & \dots & u_r \\ | & & | \end{bmatrix}$$

For any \hat{Q} w/ orthonormal columns

$\hat{Q} \hat{Q}^T$ is an orthogonal projector

onto columnspace of \hat{Q} .

Complementary projector $I - \hat{Q} \hat{Q}^T$ is also orthogonal projector

rank 1

$$P = \left(\frac{\vec{g} \vec{g}^T}{\vec{g}^T \vec{g}} \right)$$

\vec{g} unit vector

rank $n-1$

$$P^\perp = \left(I - \frac{\vec{g} \vec{g}^T}{\vec{g}^T \vec{g}} \right)$$

or

$$P = \frac{\vec{v} \vec{v}^T}{\vec{v}^T \vec{v}}$$

\vec{v} arbitrary vector

normalize