

Matrix norms

$$\|\cdot\| : \mathbb{C}^{m \times n} \rightarrow \mathbb{R}$$

1. $\|A\| \geq 0$, $\|A\| = 0 \Leftrightarrow A = 0$
2. $\|sA\| = |s| \|A\|$
3. $\|A+B\| \leq \|A\| + \|B\|$

$$\vec{x} \in \mathbb{C}^n \rightarrow \vec{y} = A\vec{x} \in \mathbb{C}^m$$

$$\|\vec{x}\|_{(n)} \xrightarrow{?} \|\vec{y}\|_{(m)}$$

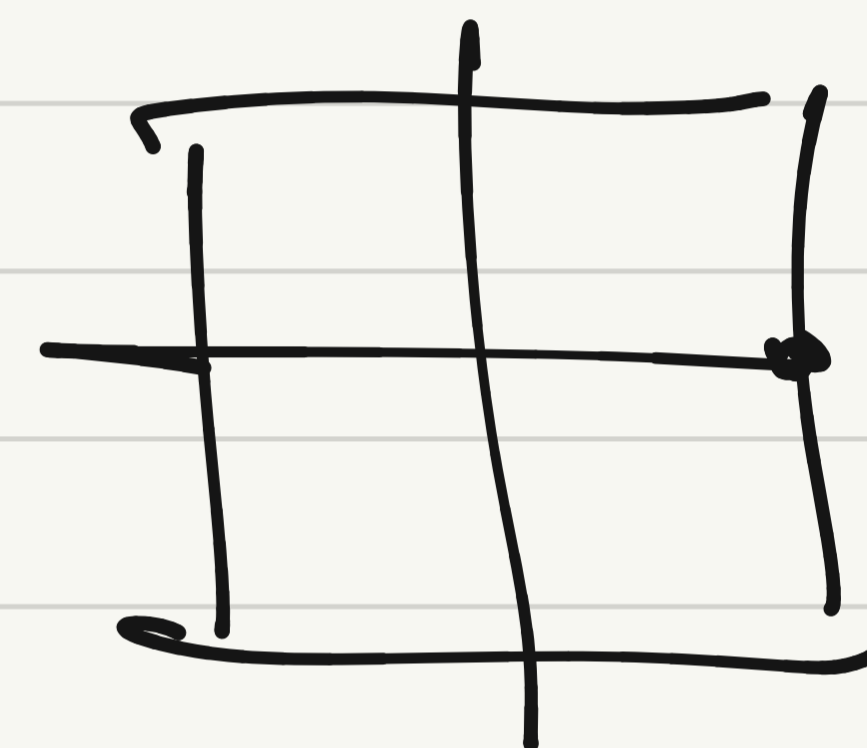
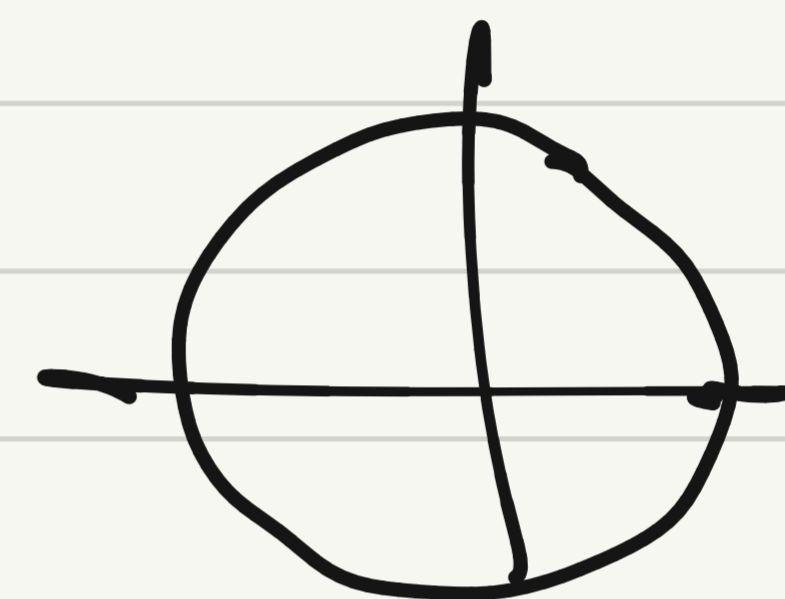
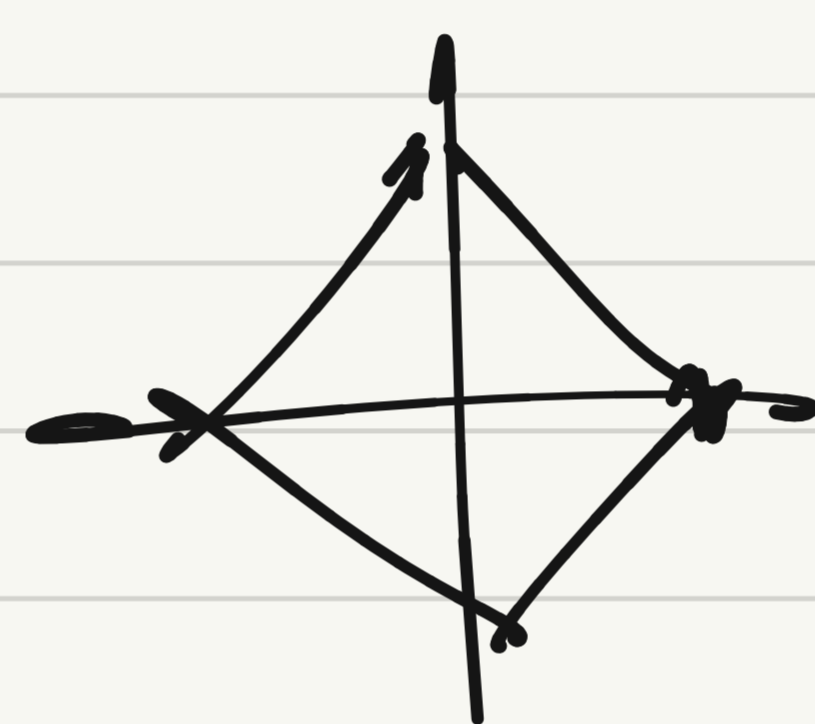
$$\|A\|_{(m,n)} = \sup_{\vec{x} \neq \vec{0}} \frac{\|A\vec{x}\|_{(m)}}{\|\vec{x}\|_{(n)}}$$

induced norm

or operator norm

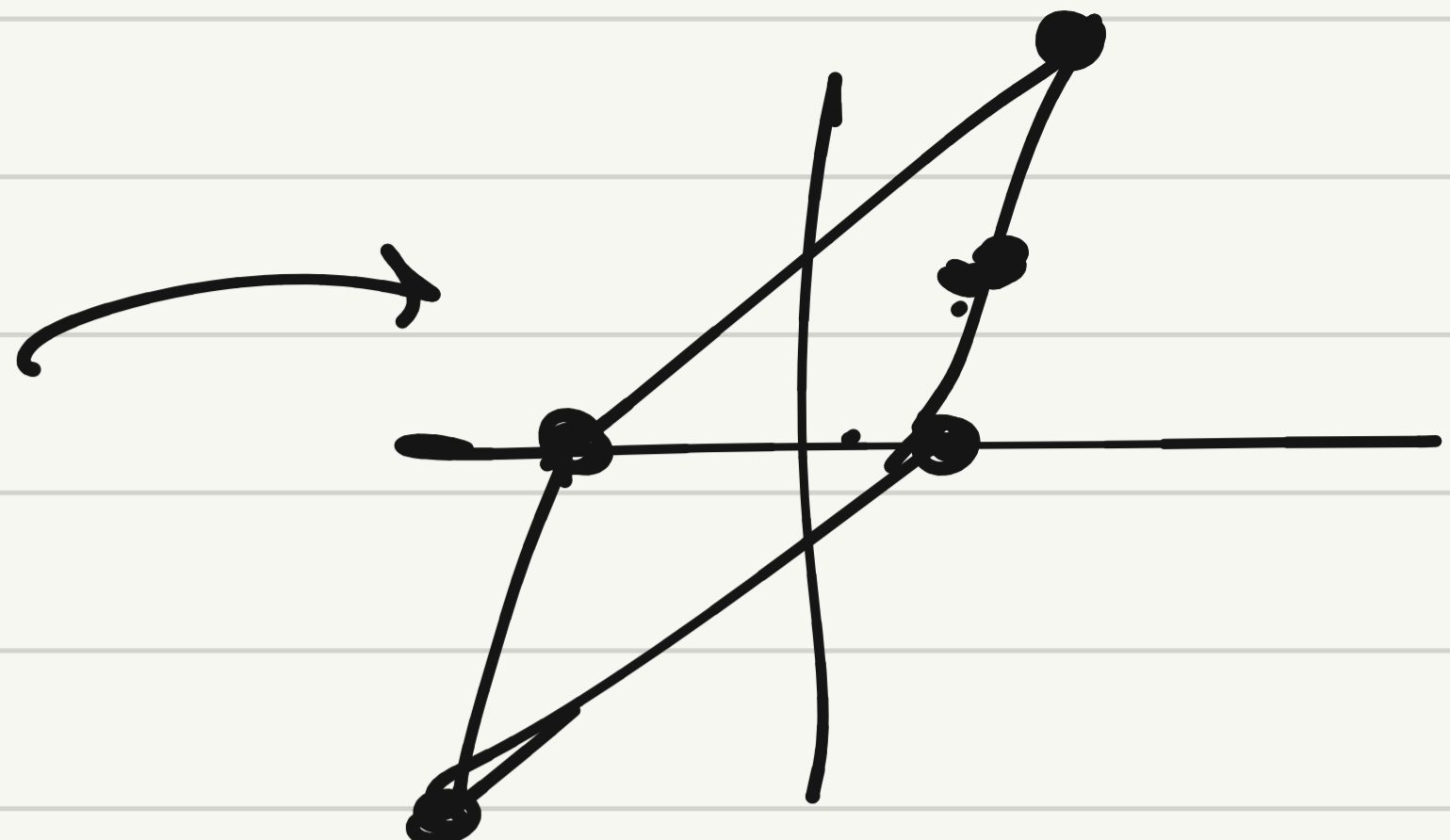
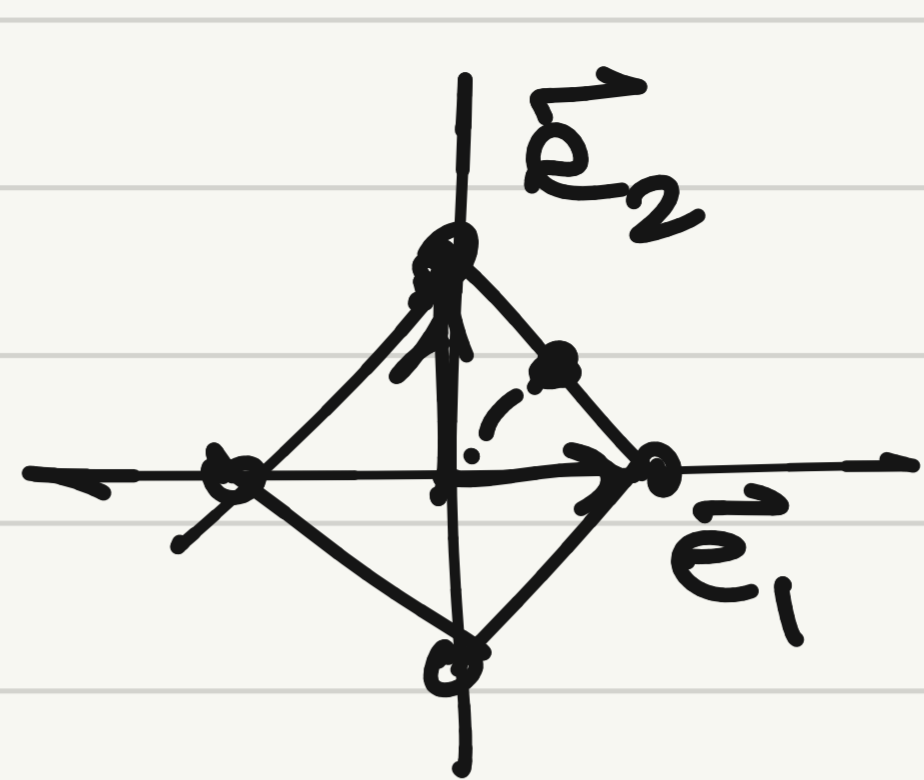
$$= \sup_{\|\vec{x}\|=1} \|A\vec{x}\|$$

$$\frac{\|A \cdot (s\vec{x})\|}{\|s\vec{x}\|} = \frac{\cancel{|s|} \|A\vec{x}\|}{\cancel{|s|} \|\vec{x}\|}$$



$$\|A\|_p = \sup \frac{\|A\vec{x}\|_p}{\|\vec{x}\|_p}$$

$$\left\{ \begin{aligned} \|A\|_1 &= \sup_{\|\vec{x}\|_1=1} \|A\vec{x}\|_1 \\ &= \max_j \|A\vec{e}_j\|_1 \\ &= \max_j \|\vec{a}_j\|_1 = \max_j \sum_i |a_{ij}| \end{aligned} \right.$$



$\|A\|_1 = \max$ abs. col. sum

$\|A\|_\infty = \max$ abs. row sum $= \max_i \sum_j |a_{ij}|$

4. $\|AB\| \leq \|A\| \|B\|$: submultiplicativity

induced \Rightarrow submult.

submult but not induced:

Frobenius norm

Hilbert-Schmidt norm

$$\|A\|_F = \sum_i \sum_j |a_{ij}|^2$$

$$\|A\|_F = \sqrt{\sum_i \sum_j |a_{ij}|^2} = \sqrt{\sum_j \|\vec{a}_j\|^2} = \sqrt{\text{tr}(A^*A)} = \sqrt{\text{tr}(AA^*)}$$

$$\|AB\|_F \leq \|A\|_F \|B\|_F$$

$$\|A\vec{x}\|_2 \leq \|A\|_F \|\vec{x}\|_2$$

$$\text{tr} M = \sum_i m_{ii}$$

unitarily invariant : A , unitary Q, R

$$\|A\|_F = \|QA\|_F = \|AR\|_F$$

$$\|A\|_2 = \|QA\|_2 = \|AR\|_2$$

not true for
1-norm or
 ∞ -norm

$$\begin{aligned} \text{Condition number of mat-vec mult} &= \sup_{\delta x} \frac{\|A \delta x\| / \|A x\|}{\|\delta x\| / \|x\|} = \frac{\|x\|}{\|A x\|} \sup_{\delta x} \frac{\|A \delta x\|}{\|\delta x\|} \\ &= \frac{\|x\|}{\|A x\|} \cdot \|A\| \end{aligned}$$

$$\text{Cond. num of matrix} = \sup_{\vec{x} \neq 0} \frac{\|A\vec{x}\|}{\|\vec{x}\|} \cdot \|A\| = \|A\| \sup_{\vec{x} \neq 0} \frac{\|\vec{x}\|}{\|A\vec{x}\|}$$

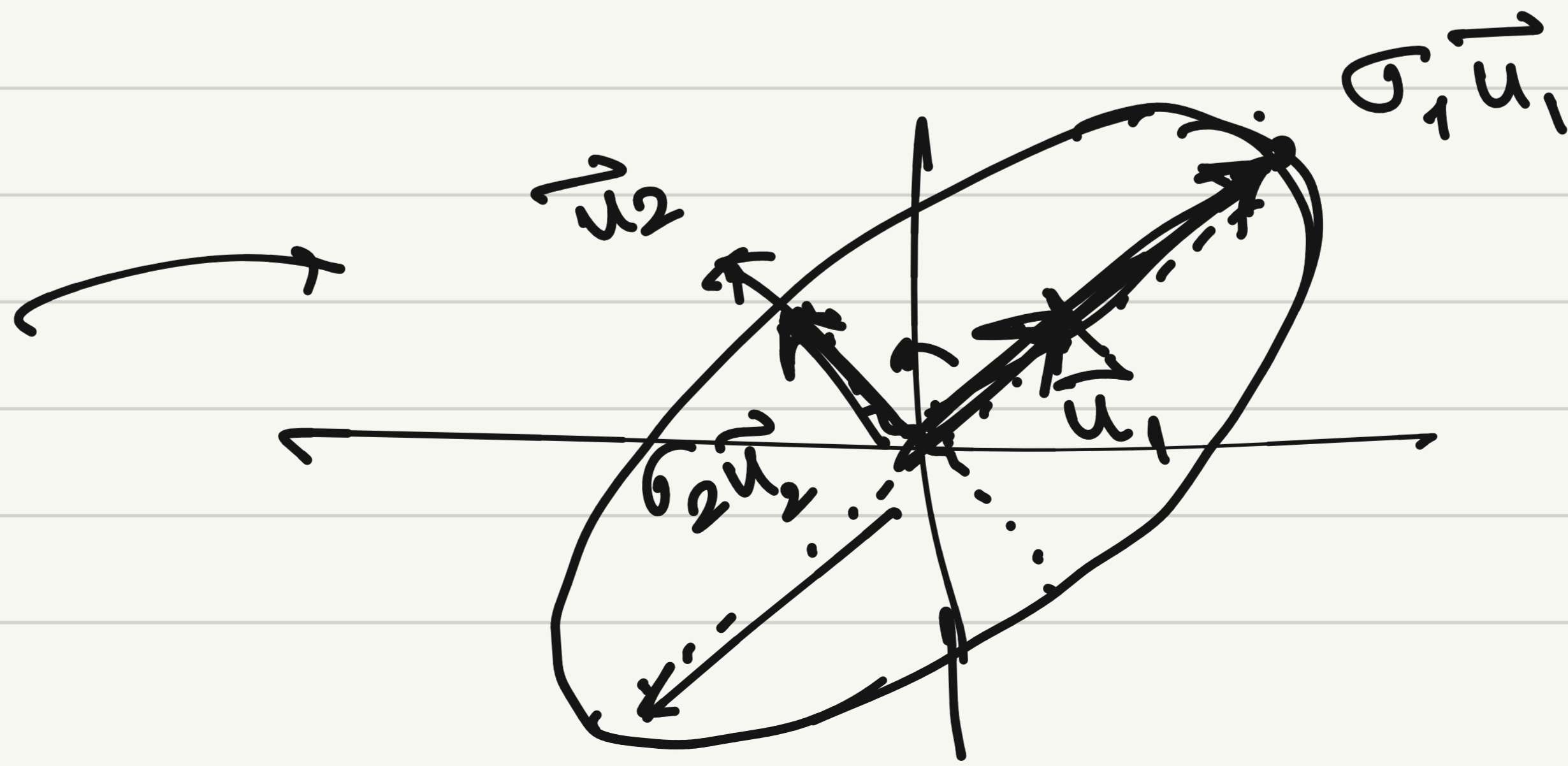
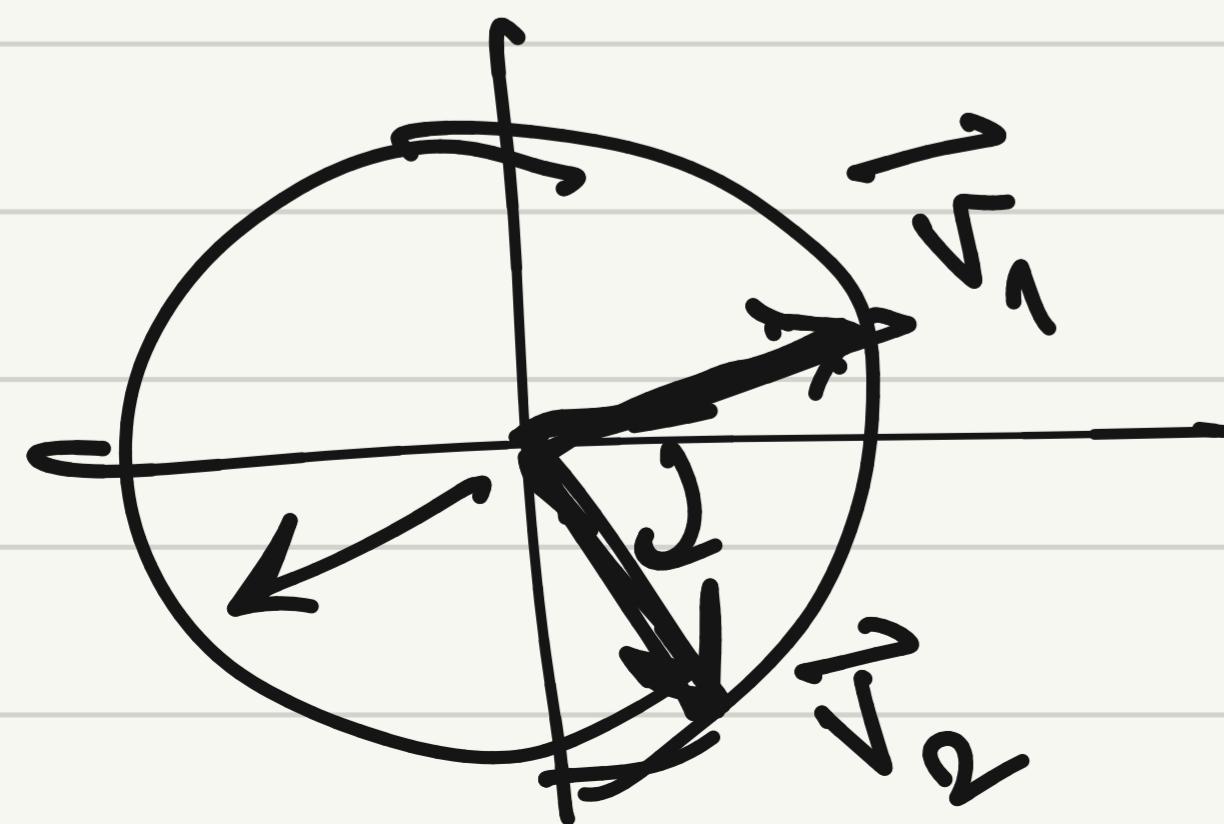
if A is singular, $\kappa(A) = \infty$

$$\text{if } A \text{ is invertible, let } \vec{x} = A^{-1}\vec{y} \rightarrow \|A\| \sup_{\vec{y} \neq 0} \frac{\|A^{-1}\vec{y}\|}{\|\vec{y}\|}$$

$$\kappa(A) = \|A\|_2 \|A^{-1}\|_2$$

Singular Value Decomposition (SVD)

$$A \in \mathbb{R}^{2 \times 2}$$



$$A\vec{v}_1 = \sigma_1\vec{u}_1$$

$$A\vec{v}_2 = \sigma_2\vec{u}_2$$

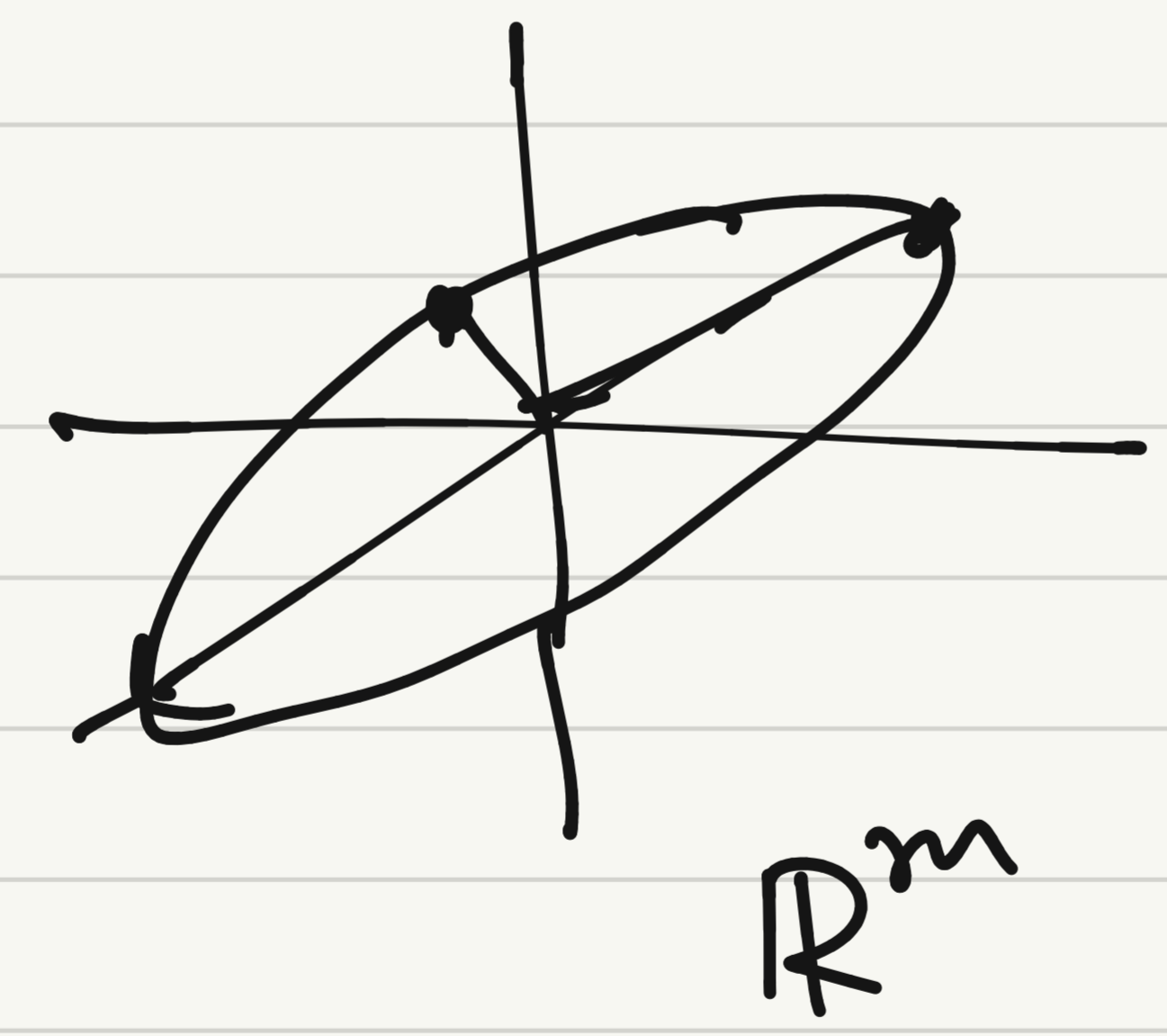
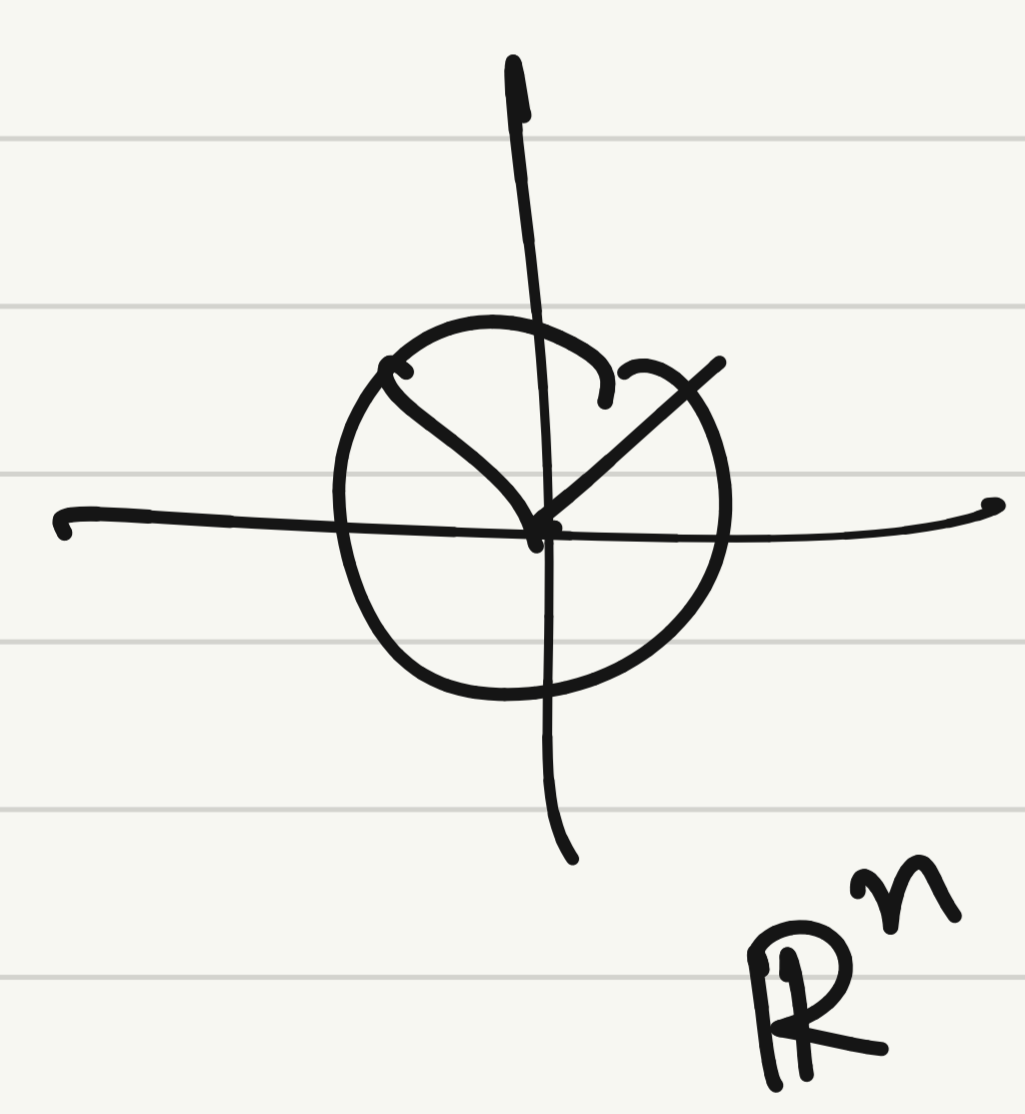
$$A \begin{bmatrix} \vec{v}_1 & \vec{v}_2 \end{bmatrix} = \begin{bmatrix} \sigma_1\vec{u}_1 & \sigma_2\vec{u}_2 \end{bmatrix}$$

$$A \begin{bmatrix} \vec{v}_1 \\ \vec{v}_2 \end{bmatrix} = \begin{bmatrix} \sigma_1 \vec{u}_1 & \sigma_2 \vec{u}_2 \end{bmatrix} = \begin{bmatrix} \vec{u}_1 & \vec{u}_2 \end{bmatrix} \begin{bmatrix} \sigma_1 \\ \sigma_2 \end{bmatrix}$$

left singular vectors
singular values

right singular vectors
U
Σ

$A^{m \times n}$
 $m \geq n, \text{rank}(A) = n$



$$A \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix} = \begin{bmatrix} \vec{u}_1 & \dots & \vec{u}_n \end{bmatrix} \begin{bmatrix} \sigma_1 & & \\ & \dots & \\ & & \sigma_n \end{bmatrix}$$

orthogonal
orthogonal
+ve reals

Square
U
Σ

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

$$\Rightarrow \boxed{A = \hat{U} \hat{\Sigma} \hat{V}^*}$$

orthogonal

$A \in \mathbb{R}^{m \times n}$, $m \geq n$, full rank

$$A = \hat{U} \hat{\Sigma} V^* \quad \therefore \boxed{\text{reduced SVD}}$$

\hat{U} : $m \times n$, orthonormal cols.

$\hat{\Sigma}$: $n \times n$, diagonal

V : $n \times n$, orthogonal

$$\begin{bmatrix} A \end{bmatrix} = \underbrace{\begin{bmatrix} | & | & | \end{bmatrix}}_{\hat{U}} \begin{bmatrix} \diagdown \end{bmatrix} \begin{bmatrix} | & | & | \end{bmatrix}_{V^*}$$

n orthonormal vectors in \mathbb{R}^m

$\Rightarrow \exists m-n$ more vectors

$$\begin{bmatrix} | & | & | \end{bmatrix} \rightarrow \begin{bmatrix} \hat{U} & | & | & | \end{bmatrix}$$

$\hat{U} : m \times n$ $U : m \times m$

$$A = \begin{bmatrix} | & | & | & \vdots \end{bmatrix} \begin{bmatrix} \diagdown \end{bmatrix} \begin{bmatrix} | & | & | \end{bmatrix}$$

orthonormal basis

$U : m \times m$ $\Sigma : m \times n$ $V^* : n \times n$

$\therefore \boxed{\text{full SVD}}$

If $A \in \mathbb{C}^{m \times n}$, a Singular value decomposition is $A = U \Sigma V^*$ s.t.

U : $m \times m$, unitary

Σ : $m \times n$, diagonal with nonnegative real entries

V : $n \times n$, unitary

Existence & uniqueness

Thm. (i) Every matrix $A \in \mathbb{C}^{m \times n}$ has an SVD,

(ii) $\{\sigma_1, \dots, \sigma_p\}$ ($p = \min(m, n)$) is uniquely determined,

by convention
 $\sigma_1 \geq \sigma_2 \geq \sigma_3 \geq \dots \geq \sigma_p$

(iii) If σ_j 's are distinct, then $\{\vec{u}_1, \dots, \vec{u}_p\}$, $\{\vec{v}_1, \dots, \vec{v}_p\}$ are uniquely determined (up to complex sign),
Complex scalar \mathbb{C} with $|z|=1$