

So FAR :

Defined adjoints of operators

$$T: V \rightarrow W$$

$$T^*: W \rightarrow V$$

\rightsquigarrow

generalized
matrix transpose

self-adjoint operators

$$T: V \rightarrow V$$

$$T = T^*$$

generalized symmetric
matrices

normal operators :

$$T^* T = T T^*$$

spectral thm.

proved in class only for

$$F = \mathbb{C}.$$

"There is an orthonormal basis for which T is diagonal (T is normal)."

Consider the following problem: Find an operator T_k , whose image has dimension equal to k that best approximates T .

e.g. Suppose T has matrix $\begin{bmatrix} 5 & 0 \\ 0 & 1 \end{bmatrix} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$

w.r.t. the standard basis. What is a good candidate for T_1 ?

$T_1 : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ whose image has dimension 1.

Given by matrix $\begin{bmatrix} 5 & 0 \\ 0 & 0 \end{bmatrix}$.

e.g. $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ with matrix $\begin{bmatrix} 700 & 0 & 0 \\ 0 & 100 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ w.r.t. the standard basis.

Candidate for T_1 ? $\begin{bmatrix} 700 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$.

Candidate for T_2 ? $\begin{bmatrix} 700 & 0 & 0 \\ 0 & 100 & 0 \\ 0 & 0 & 0 \end{bmatrix}$.

"Best approximation": want a notion of "norm".

So can talk about $\|T - T_k\|$: the "size" of $T - T_k$.

Lots of options! See, for example, Wiki "Matrix norms".

Option 1 Treat a matrix like a vector.

Recall: $M_{n \times m}(\mathbb{R}) \xrightarrow{\sim} \mathbb{R}^{n \cdot m}$

Define $\left\| \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} \right\| = \sqrt{\sum |a_{ij}|^2}$

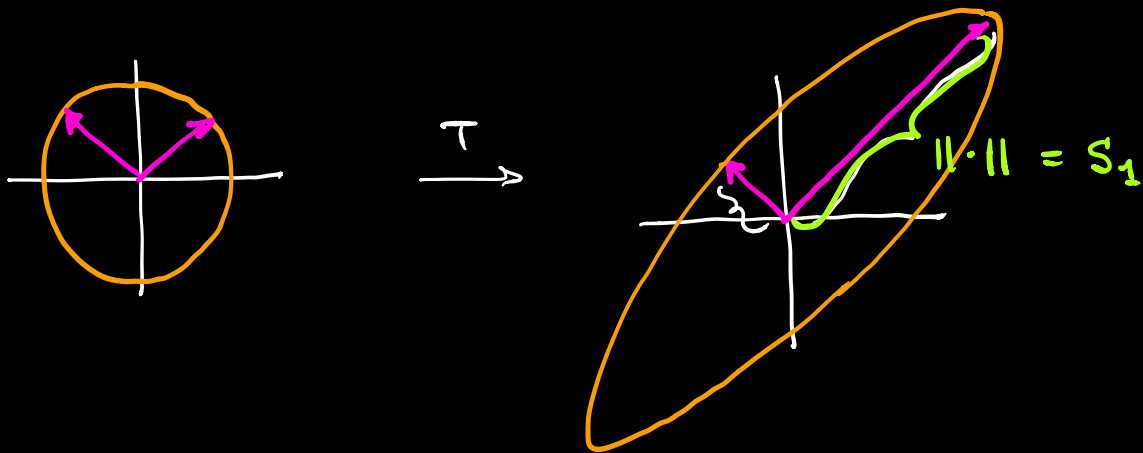
FROBENIUS NORM

$$\text{e.g. } \left\| \begin{bmatrix} 700 & & 0 \\ & 100 & \\ 0 & & 1 \end{bmatrix} - \begin{bmatrix} 700 & & 0 \\ & 0 & 0 \\ 0 & & 0 \end{bmatrix} \right\| =$$

$$\left\| \begin{bmatrix} 0 & & 0 \\ & 100 & \\ 0 & & 1 \end{bmatrix} \right\| = \sqrt{100^2 + 1^2} \approx 100 = \text{operator norm}$$

Exercise: Frobenius norm is independent of choice of orthonormal basis.

2. "Better norm": Operator norm



Define $\|T\| = S_1 = \max_{\|v\|=1} \|T(v)\|$.

Can rephrase this: $\|T\| = \max_{v \in V} \frac{\|T(v)\|}{\|v\|}$.

Claim. If $T \in \mathcal{L}(V)$ has an **O.E.B.** with eigenvalues $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n > 0$

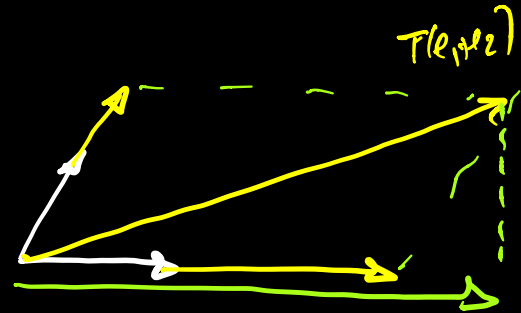
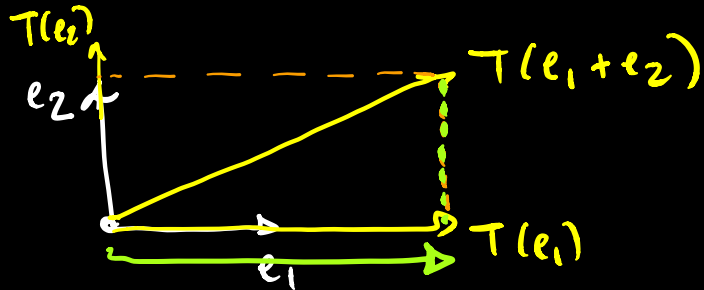
then the operator T_k whose matrix w.r.t.

the O.E.B. is $\begin{bmatrix} \lambda_1 & & & 0 \\ & \ddots & & \\ & & \lambda_k & \\ 0 & & & 0 \dots 0 \end{bmatrix}$, then

$\|T_k - T\| \leq \|S - T\| \quad \forall S \in \mathcal{L}(V)$ whose image has dimension = k .

We are writing $T(v) = \sum \lambda_i \langle v, e_i \rangle e_i$

where $\{e_1, \dots, e_n\}$ is an orthonormal basis.



What can we say if T is not self-adjoint or normal?

For example, $T: V \rightarrow W$, where $V \cong W$?

Clever idea: work with two different orthonormal bases!

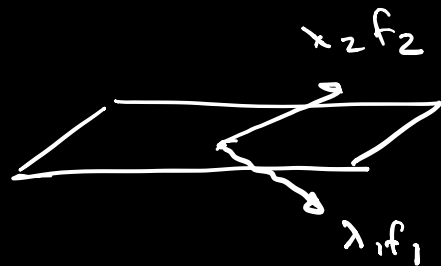
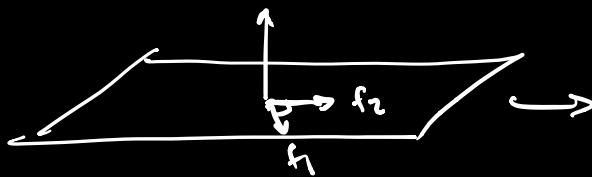
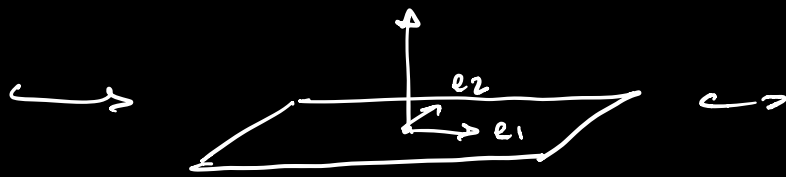
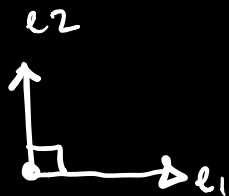
In other other words,

$$T_k(v) = \sum_{i=1}^k \sigma_i \langle v, e_i \rangle f_i,$$

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_k$$

$$T(v) = \sum_{i=1}^{\min(n,m)} \sigma_i \langle v, e_i \rangle f_i.$$

T looks like the composition of a projection or inclusion, a rotation, and a stretching.



Pf of SVD:

NOTE: $T^*T: V \rightarrow V$

Lemma (~~To be proven~~) T^*T is self-adjoint.

$$\begin{aligned} \langle T^*T v_1, v_2 \rangle &= \langle T v_1, T v_2 \rangle \\ &= \langle v_1, \underbrace{T^*T v_2} \rangle. \end{aligned}$$

defn. of adj, since $(T^)^* = T$*

We conclude that $(T^*T)^* = T^*T$.

\exists an O.E.B. e_1, \dots, e_n of T^*T .

Set $f_i' = T(e_i)$.

Note: $\langle f_i', f_j' \rangle = \langle T e_i, T e_j \rangle$
 $= \langle e_i, T^* T e_j \rangle$
 $\rightarrow \alpha \langle e_i, e_j \rangle$
 $= 0.$

Set: $f_i = \frac{f_i'}{\|f_i'\|}$

Then: $T(e_i) = \underbrace{\|f_i'\|}_{\sigma_i} f_i$

Note: keeping takes care of subtleties. □