

Chap 8: Inner products

$$F = \mathbb{R}, \mathbb{C}$$

\odot on \mathbb{R}^n : dot product

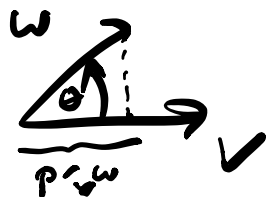
- standard inner product

$$V = (a_1, \dots, a_n), W = (b_1, \dots, b_n)$$

$$V \cdot W = \sum a_i b_i \in \mathbb{R}$$

Geometrically:

V, W (& O): lie in a plane.



In this configuration:
projection of W onto V

$$\begin{aligned} V \cdot W &= (\text{length } V) (\text{length } p_{V,W}) \\ &= (\text{length } V) (\text{length } W) \cos \theta \\ &= (\text{length } V) (\text{length } p_{W,V}) \end{aligned}$$



Use signed length of proj. as sign.

Length in terms of dot product.

$$V, \text{ length } \|V\| \geq 0 \quad \|V\|^2 = V \cdot V = \sum a_i^2$$

$$\|v\| = \sqrt{\underbrace{\sum a_i^2}_{\geq 0}} \geq 0; \quad = 0 \Leftrightarrow v=0$$

$$V = \mathbb{R}^n$$

$$V \times V \rightarrow \mathbb{R} \quad (v, w) \mapsto v \cdot w$$

Properties:

- real valued

- linear as function in 1st vbl.

$$(v_1 + v_2) \cdot w = v_1 \cdot w + v_2 \cdot w$$

$$(c v) \cdot w = c (v \cdot w)$$

- Symmetric;

$$v \cdot w = w \cdot v$$

linear as fn of 2^o vbl

- \therefore bilinear

- positive definite

$$\underbrace{v \cdot v}_{\geq 0}, \quad \> 0 \text{ unless } v=0.$$

if just have this: pos. semi-def.

In gen, if we have v.s. V over \mathbb{R} ,
 $\star \quad V \times V \rightarrow \mathbb{R}$

↑ called an inner product
 if it has these properties.

Ex 2 p 271 H&K \mathbb{R}^2
 $(v|w)$

$$v = (a_1, a_2), \quad w = (b_1, b_2)$$

$$(v|w) = a_1 b_1 - \underset{\substack{\uparrow \\ -1}}{a_2} b_1 - \underset{\substack{\uparrow \\ -1}}{a_1} b_2 + 4 a_2 b_2$$

Symmetric
 bilinear

pos def - need to check

$$\begin{aligned} (v|v) &= a_1^2 - 2a_1 a_2 + 4a_2^2 \\ &= \underbrace{(a_1 - a_2)^2}_{\geq 0} + \underbrace{3a_2^2}_{\geq 0} \end{aligned}$$

$\geq 0 \Leftrightarrow a_1 = a_2 = 0$
This is an inner product.

On \mathbb{C} : V v.s. / \mathbb{C}

Variants

\mathbb{C}^n dot product on \mathbb{C}^n
 standard hermitian inner product

$$V = (a_1, \dots, a_n) \in \mathbb{C}^n$$

$$W = (b_1, \dots, b_n) \in \mathbb{C}^n$$

$$\text{Defn } V \cdot W = \sum a_i \bar{b}_i$$

E.g. if $a_i \in \mathbb{R}, b_i \in \mathbb{R}$

$$V \in \mathbb{R}^n \subset \mathbb{C}^n, \quad W \in \mathbb{R}^n \subset \mathbb{C}^n$$

$$V \cdot W = \sum a_i \bar{b}_i = \sum a_i b_i = V \cdot W$$

↑
in \mathbb{C}^n sense

↑
in \mathbb{R}^n sense.

Properties of \mathbb{C}^n dot product;
 → \mathbb{C}^n valued $V \times V \rightarrow \mathbb{C}$
 ($V = \mathbb{C}^n$)

— linear in 1st vbr

$$V \cdot W = \sum a_i \bar{b}_i$$

$$W \cdot V = \sum b_i \bar{a}_i$$

$$b_i \bar{a}_i = \overline{a_i \bar{b}_i} \\ = \bar{a}_i \bar{\bar{b}_i}$$

$$W \cdot V = \overline{V \cdot W}$$

— conjugate symmetric

- ? in \mathbb{C} vbl

$$V \cdot (w_1 + w_2) = V \cdot w_1 + V \cdot w_2$$

$$V \cdot (c w) = \bar{c} (V \cdot w)$$

Conjugate linear in \mathbb{C} vbl.

$$V \times V \rightarrow \mathbb{C}$$

Sesquilinear ($\bar{\cdot}$ linear)

Pos def?

$$V = (a_1, \dots, a_n) \in \mathbb{C}^n \quad a_i \in \mathbb{C}$$

$$V \cdot V = a_1 \bar{a}_1 + \dots + a_n \bar{a}_n$$

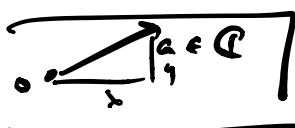
$a \in \mathbb{C} \quad a = x + iy \quad x, y \in \mathbb{R}$

$a \bar{a} = (x + iy)(x - iy) = x^2 + y^2 \geq 0$

$|a| = \sqrt{x^2 + y^2}$

$= 0 \Leftrightarrow x, y = 0 \quad |a|^2$

$\Leftrightarrow a = 0$



$$V \cdot V = a_1 \bar{a}_1 + \dots + a_n \bar{a}_n = |a_1|^2 + \dots + |a_n|^2 \geq 0$$

$\neq 0 \Leftrightarrow \text{all } a_i = 0 \Leftrightarrow V = 0.$

Norm: $\|V\| \geq 0, \quad \|V\|^2 = V \cdot V \in \mathbb{R}, \quad V \cdot V \geq 0.$

$$V \in \mathbb{C}^n \mapsto \tilde{V} \in \mathbb{R}^{2n}$$

"

$$(a_1, \dots, a_n) \mapsto (x_1, y_1, x_2, y_2, \dots, x_n, y_n)$$

$$a_j = x_j + iy_j$$

$$\|V\| = \sqrt{|a_1|^2 + \dots + |a_n|^2}$$

in \mathbb{C}^n

$$= \sqrt{x_1^2 + y_1^2 + \dots + x_n^2 + y_n^2}$$

$$= \|\tilde{V}\|$$

in \mathbb{R}^{2n}

Norm agrees:

\mathbb{C}^n vs. \mathbb{R}^{2n} .

Caution: $V, W \in \mathbb{C}^n$

$$\mapsto \tilde{V}, \tilde{W} \in \mathbb{R}^{2n}$$

$$\text{Usually, } v \cdot w \neq \tilde{V} \cdot \tilde{W}$$

$$\begin{array}{ccc} \uparrow & & \uparrow \\ \mathbb{C}^n & & \mathbb{R}^{2n} \\ \uparrow & & \uparrow \\ \mathbb{R} & & \mathbb{R} \end{array}$$

usually

$\notin \mathbb{R}$.

In gen, if have $V \times V \rightarrow \mathbb{C}$

\uparrow vs \mathbb{C}

satisfying above properties
(ex val, sesquiline, pos def):
(conj sym, lin 1st val)

ex (hermitian inner product on V .)

Ex. of inf dim'd vs with
an inner product. (1-upt in analysis)

\mathbb{R} , $V = \{\text{cont. fns: } [0,1] \rightarrow \mathbb{R}\}$

inner prod: $\langle f, g \rangle = \int_0^1 f(x)g(x) dx$

\mathbb{C} $V = \{\text{cont fns: } [0,1] \rightarrow \mathbb{C}\}$

inner prod: $\langle f, g \rangle = \int_0^1 f(x)\overline{g(x)} dx$

V f.d.v.s \mathbb{R} or \mathbb{C} .

Basis $B: v_1, \dots, v_n$.

We can write any vector in terms of B

$$v = a_1 v_1 + \dots + a_n v_n \quad \begin{array}{l} a_i \text{': coord's of} \\ v \text{ wrt } B. \end{array}$$

$$v = \sum a_i v_i$$

$$w = \sum b_i v_i$$

$$\text{Def'n } \langle v, w \rangle = \sum_{i=1}^n a_i b_i \text{ if } \mathbb{R}$$

$$\text{or } \sum_{i=1}^n a_i \overline{b_i} \text{ if } \mathbb{C}.$$

Generalizes dot prod: can $V = \mathbb{R}^n$ or \mathbb{C}^n
& $B: e_1, \dots, e_n$.

V vs \mathbb{R} or \mathbb{C}

$\langle \cdot, \cdot \rangle$ inner product

$\hookrightarrow \|\cdot\|$ norm.

$$\|v\|^2 = \langle v, v \rangle.$$

Converse? From $\|\cdot\|$,
we can get $\langle \cdot, \cdot \rangle$:

Polarization identities \mathbb{R} , & \mathbb{C} :

\mathbb{R} (8-3)

$$\langle v, w \rangle = \frac{1}{4} \|v+w\|^2 - \frac{1}{4} \|v-w\|^2$$

p 274

\mathbb{C} (8-4)

$$\begin{aligned} \langle v, w \rangle &= \frac{1}{4} \|v+w\|^2 + \frac{i}{4} \|v+iw\|^2 \\ &\quad - \frac{1}{4} \|v-w\|^2 - \frac{i}{4} \|v-iw\|^2 \end{aligned}$$

Can verify by writing out each

$$\|z\|^2 = \langle z, z \rangle$$

& use bilinearity (or sesquilinearity)
& expand.

\therefore to know $\langle \cdot, \cdot \rangle$, it suffices to know $\|\cdot\|$.

To know $\langle v, w \rangle$ for
 all $v, w \in V$,
 suffices to know this on the
 basis vectors v_1, \dots, v_n

$$a_{ij} = \langle v_i, v_j \rangle \in \mathbb{R}, \dots \in \mathbb{C} \quad (\text{if } \mathbb{C})$$

$$\text{Reason: } v = \sum_{i=1}^n b_i v_i$$

$$w = \sum_{j=1}^n c_j v_j$$

$$\begin{aligned} \mathbb{R} \\ (*) \quad \underline{\underline{\langle v, w \rangle}} &= \sum_{i,j=1}^n b_i c_j \langle v_i, v_j \rangle = \sum_{i,j} b_i c_j a_{ij} \\ &\quad \uparrow \\ &\quad c_j \in \mathbb{C} \end{aligned}$$

$$A = (a_{ij}) \quad n \times n$$

mx of the inner product $\langle \cdot, \cdot \rangle$
 w.r.t this basis

Ex. \mathbb{R}^n , std basis, dot product

$$a_{ij} = \langle e_i, e_j \rangle = \delta_{ij}$$

$$\therefore A = I$$

Ex. \mathbb{R}^2 , std basis,

inner prod: $v = (a_1, a_2), w = (b_1, b_2)$

$$\langle v, w \rangle = a_1 b_1 - a_2 b_1 - a_1 b_2 + 4 a_2 b_2$$

↑ ↑ ↑ ↑
1 -1 -1 4

$$A = \begin{pmatrix} 1 & -1 \\ -1 & 4 \end{pmatrix}$$

In gen, $\forall \mathbb{R}$: A is symmetric.

$$a_{ij} = \overline{a_{ji}}$$

" "

$$\langle v_i, v_j \rangle = \langle v_j, v_i \rangle$$

$$\langle v, w \rangle = v A [w] \in \mathbb{R} \text{ by } (*).$$

↑ 1xn nxn nx1

$\forall \mathbb{C}$: $a_{ji} = \overline{a_{ij}}$, Conjugate sym. mx.
"Hermitian mx"

$$A^t = \overline{A}$$

$$\langle v, w \rangle = v A [\overline{w}]$$

§ 8.2 V vs \mathbb{R} or \mathbb{C} ,

together w inner prod: inner product space

$\langle \cdot, \cdot \rangle \mapsto \|\cdot\|$ norm. $\left(\begin{array}{l} \langle \cdot, \cdot \rangle \text{ is dot prod.} \\ \Downarrow \\ \|\cdot\|: \text{use } \langle \cdot, \cdot \rangle \end{array} \right)$

$$\|v\| = \sqrt{v \cdot v}$$

Properties of norm: $\|\cdot\|: V \rightarrow \mathbb{R}$

- ① $\|c\alpha\| = |c| \|\alpha\|$
- ② $\|\alpha\| \geq 0, \quad \alpha = 0 \Leftrightarrow \alpha = 0$
- ③ $|\langle \alpha, \beta \rangle| \leq \|\alpha\| \|\beta\|$
Cauchy-Schwarz inequality
- ④ $\|\alpha + \beta\| \leq \|\alpha\| + \|\beta\|$
Triangle inequality

Follow from def of $\|\cdot\|$ in terms of $\langle \cdot, \cdot \rangle$ + properties of inner prod.

(Th 1, p 277, H + K). ①, ②, ④

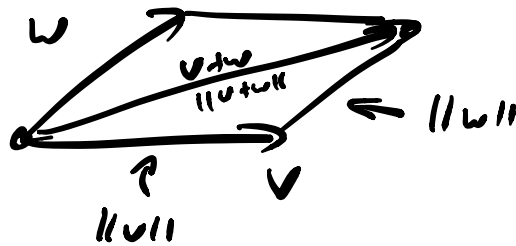
Caution: Not every norm comes from an inner product.

Ex. $\mathbb{R}^n, v = (a_1, \dots, a_n)$
 $\|v\| = \sum_{i=1}^n |a_i|$. l_1 -norm.

For dot product, + assoc $\|\cdot\|$:

$$\text{C-S: } |\alpha \cdot \beta| = \|\alpha\| \|\beta\| \underbrace{|\cos \theta|}_{\leq 1} \\ \leq \|\alpha\| \|\beta\|.$$

Δ -ins:



Ex. of C-S:

dot prod on $V = \mathbb{R}^n$

$$V = (a_1, \dots, a_n) \\ W = (b_1, \dots, b_n)$$

$$\left| \sum_{i=1}^n a_i b_i \right| \leq \left| \sum_{i=1}^n a_i^2 \right|^{1/2} \left| \sum_{i=1}^n b_i^2 \right|^{1/2}$$

for all $a_1, \dots, a_n, b_1, \dots, b_n \in \mathbb{R}$.

Ex. of C-S:

$$V = \{ \text{cont } f_n: [0, 1] \rightarrow \mathbb{R} \}$$

$$\langle f, g \rangle = \int_0^1 f(x)g(x) dx$$

$$\left| \int_0^1 f(x)g(x) dx \right| \leq \left| \int_0^1 f(x)^2 dx \right|^{1/2} \left| \int_0^1 g(x)^2 dx \right|^{1/2}$$

In \mathbb{R}^n , with dot product:

If $v, w \neq 0$, making an θ of Θ ,

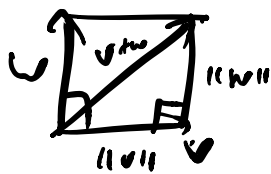
$$v \perp w \Leftrightarrow \Theta = \pm 90^\circ \Leftrightarrow \cos \Theta = 0 \\ \Leftrightarrow v \cdot w = 0$$

View $0 \perp v$, for all v .

More generally, in any inner prod. sp.,
define $v \perp w$ if $\langle v, w \rangle = 0$
 v, w are orthogonal

As to \mathbb{R}^n , we have

$$v \perp w \Leftrightarrow \|v+w\|^2 = \|v\|^2 + \|w\|^2$$



Pythagorean theorem

for all inner product spaces
(by bilinearity)

$S \subset V$ subset
inn. pr. sp.

$$\text{Let } S^\perp = \{v \in V \mid \forall w \in S, v \perp w\}$$

Orthogonal complement ("S perp")

Then: S^\perp is a subspace of V
(easy to check)

Ex: $V = \mathbb{R}^3$, $S = \{e_1\}$ dot product

$S^\perp = \text{span of } \{e_2, e_3\}$
 $= yz \text{ plane.}$

$W = \text{span } S$
 $= x \text{-axis}$

$W^\perp = S^\perp$

Ex. $V = \mathbb{R}^2$, $S = \{(1, 2)\}$

$(-2, 1) \in S^\perp$

$W = \text{span } S$

$W^\perp = S^\perp$

$S^\perp = \text{span of } \{(-2, 1)\}$.

In general, given $S \subset V$,

let $W = \text{span } S$.

Get $S^\perp = W^\perp$

Ex. $V^\perp = 0$

(bc if $v \in V^\perp$ then
 $\langle v, v \rangle = 0$, so $v = 0$. ✓)

Ex. $0 = \{0\}$. $0^\perp = V$

$S \subset V$ subset

Say S is orthogonal if \mathbb{R}^n , dot prod.

any two vectors in S are orthog. Ex $S = \{(1, 2), (2, 1)\}$

Say S is orthonormal if

orthogonal & all the vectors in S
have unit length ("normalized")

Ex. $\{e_1, \dots, e_k\} \subset \mathbb{R}^n$ $k \leq n$
orthonormal

Ex $V = \{ \text{cont fns } [0, 1] \rightarrow \mathbb{R} \}$
Infinite set

$\{1, \cos 2\pi x, \sin 2\pi x, \cos 4\pi x, \sin 4\pi x, \dots\}$
is orthogonal.

Thm (H&K, Th 2, p 279)

An orthogonal set of non-0 vectors
is lin. independent.

Pf. $S = \{v_1, \dots, v_n\}$ $v_i \neq 0$. orthog.

If $\sum_{i=1}^n a_i v_i = 0$, wts all $a_i = 0$.

$$0 = \langle 0, v_2 \rangle = \langle \sum_{i=1}^n a_i v_i, v_2 \rangle$$

$$= \sum_{i=1}^n a_i \langle v_i, v_2 \rangle = a_2 \|v_2\|^2$$

$$a_2 \|v_2\|^2 = 0 \quad v_2 \neq 0, \therefore \|v_2\|^2 \neq 0.$$

$\therefore a_2 = 0.$ ✓

$S = \{v_1, \dots, v_k\} \subset V$, orthog. set
of non-0 vectors

\therefore lin. ind.

If $v \in \text{span } S$ then $v = \sum_{i=1}^k a_i v_i$,
uniquely. (lin. ind.) $\left\{ \begin{array}{l} \text{scalars} \end{array} \right.$

There is a formula for a_i 's:

$$a_i = \frac{\langle v, v_i \rangle}{\|v_i\|^2}$$

Pf. $\langle v, v_i \rangle = \langle \sum_j a_j v_j, v_i \rangle$
 $= \sum_j a_j \langle v_j, v_i \rangle = a_i \|v_i\|^2$
orthog. ✓

In particular, if the orthog. set $\{v_1, \dots, v_n\}$
spans V , then it's a basis, &

we get a formula for coords of any vector.
 $v = \sum a_i v_i$

Even better: if have o.n.b.

then $\|v_i\|^2 = 1$, so

$$a_i = \langle v, v_i \rangle. \quad (\text{Simpler})$$

How to construct an orthogonal basis?

Gram-Schmidt orthogonalization
— uses above formula

Given β_1, \dots, β_n lin. ind.
in an inner prod. space,

We can find $\alpha_1, \dots, \alpha_n$ orthogonal
(+ lin. ind.) with the same
span as β_1, \dots, β_n .

In particular, if β 's are a basis for V ,
then α 's are an orthog. basis for V .

Then: $\frac{\alpha_1}{\|\alpha_1\|}, \dots, \frac{\alpha_n}{\|\alpha_n\|}$ is o.n.b. for V .

Procedure: Start with β_1, \dots, β_n .

Construct $\alpha_1, \dots, \alpha_n$ inductively

$$\alpha_1 = \beta_1.$$

$$\alpha_2 = \beta_2 - \frac{\langle \beta_2, \alpha_1 \rangle}{\|\alpha_1\|^2} \alpha_1$$

$$\alpha_3 = \beta_3 - \frac{\langle \beta_3, \alpha_1 \rangle}{\|\alpha_1\|^2} \alpha_1 - \frac{\langle \beta_3, \alpha_2 \rangle}{\|\alpha_2\|^2} \alpha_2$$

etc.

Easy to check that each α_i is

\perp to the previous ones,

\dagger α 's have same span as β 's.

(see H & K, pp 200 - 281.

see Example 12, p 282)

Ex. Gram-Schmidt to get an orthog. basis
of some given subspace of \mathbb{R}^n
(w.r.t dot product)

e.g. plane in \mathbb{R}^3 (or \mathbb{C}^3)

- see prob 6 on supplementary study
problems for Exam 2

- see #4 on PS14.

Once we have an orthog. basis,

\mapsto get o.n.b. by normalization.

Can also use G-S to extend a
 given orthog. set of non-0 vectors
 to an orthog. basis:

Orthog. set $\alpha_1, \dots, \alpha_k$ ($\neq 0$)

Extend to a basis: add $\beta_{k+1}, \dots, \beta_n$

Apply G-S to $\alpha_1, \dots, \alpha_k, \beta_{k+1}, \dots, \beta_n$

1st k steps do nothing

Remaining steps replace $\beta_{k+1}, \dots, \beta_n$

by $\alpha_{k+1}, \dots, \alpha_n$
 s.t. $\{\alpha_1, \dots, \alpha_k, \dots, \alpha_n\}$ is orthog.,
 same span, so basis.
 orthogonal.

V vs with \langle, \rangle .

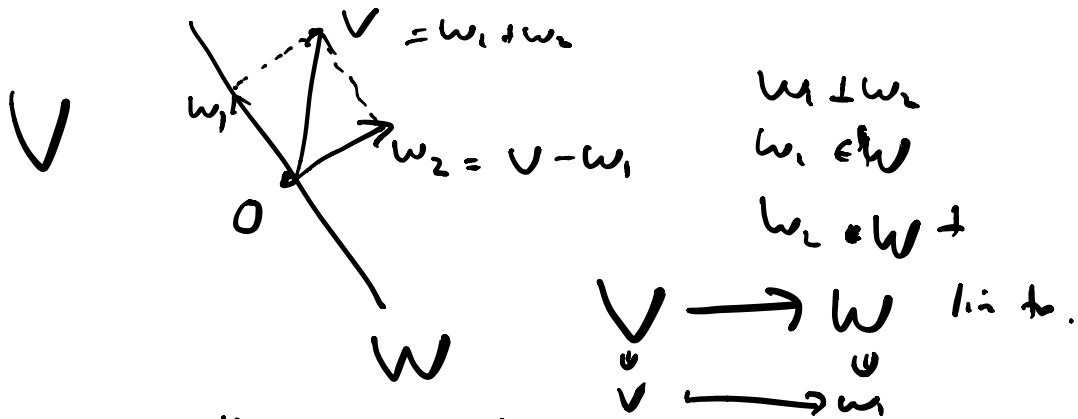
$$\text{span}\{\alpha_1, \dots, \alpha_k\} = W$$

Full basis $\underbrace{\alpha_1, \dots, \alpha_k}_{\text{orthog. span} = W}, \underbrace{\alpha_{k+1}, \dots, \alpha_n}_{\text{span} = W'}$

$$W' = W^\perp$$

$$V = W \oplus W^\perp \quad \text{b/c } W \cap W^\perp = 0$$

Sometimes written $W \oplus W^\perp$
 - orthogonal direct sum.



\circ orthogonal projection

$Im = W \quad Ker = W^\perp$

w_1 is the point in W that is closest to $v \in V$

"best approx to v in W "

SP.3 Adjoints

V fld in prod space / $F = \mathbb{R} \text{ or } \mathbb{C}$
 $\hookrightarrow \dim = n.$

Take $\beta \in V$

$f_\beta: V \rightarrow F$ lin. map.

$\alpha \mapsto \langle \alpha, \beta \rangle$

$f_\beta \in V^*$, dual space.
 f_β linear functional on V

Take o.n.b. of V : v_1, \dots, v_n

$$\beta = \sum_{i=1}^n b_i v_i \quad b_i = \langle \beta, v_i \rangle.$$

$$f_\beta(\alpha) = \langle \alpha, \beta \rangle = \langle \alpha, \sum b_i v_i \rangle$$

$$= \sum b_i \langle \alpha, v_i \rangle = \sum a_i b_i \text{ if } \mathbb{R}$$

$\alpha = \sum a_i v_i$

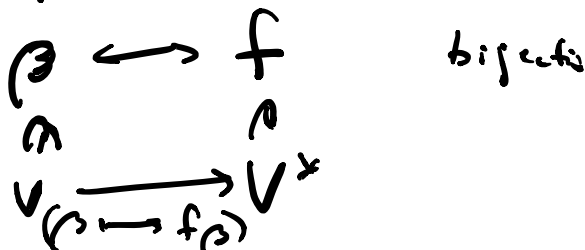
$$\dots = \sum \bar{b}_i \langle \alpha, v_i \rangle = \sum a_i \bar{b}_i \text{ if } \mathbb{C}$$

M_x of $f_\beta: V \rightarrow F$

$1 \times n$ is $(b_1, \dots, b_n) / \mathbb{R}$
 $(\bar{b}_1, \dots, \bar{b}_n) / \mathbb{C}$

Conversely every $f \in V^*$

$(f: V \rightarrow F)$ is the form f_β
for some β .



$$\Phi: V \xrightarrow{\text{bij}} V^* \quad \beta \mapsto f_\beta$$

easy to check: Φ is a lin. transf
 \therefore isomorphism.

$$\langle \cdot, \cdot \rangle \mapsto \Phi$$

(v.s. $V \xrightarrow{\sim} V^{**}$
 even w/o $\langle \cdot, \cdot \rangle$)

Adjoint of a lin transf:

Thm (H & K, Th 7, p 293
 + Cor., p 294)

Let $T: V \rightarrow V$ lin op.

on a fin dim inn. p.s.p

w o.n.b. v_1, \dots, v_n (wrt $\langle \cdot, \cdot \rangle$).

Let A be the mtrx of T (wrt this basis)

$$\text{Let } A^* = \begin{cases} A^t & \text{if } \mathbb{R} \\ \bar{A}^t & \text{if } \mathbb{C} \end{cases}$$

Let $T^*: V \rightarrow V$ be the
 corresp lin tr. (wrt this basis)

Then $\forall \alpha, \beta \in V$:

$$(*) \quad \langle T\alpha, \beta \rangle = \langle \alpha, T^*\beta \rangle$$

(*) Adjoint property.

Pf is by direct computation.

T^* is the adjoint of T .

(this is uniquely determined by (*)).

Q: What is the relationship
between above $T^*: V \rightarrow V$
(adjoint)

to the $T^t: V^* \rightarrow V^*$
also called T^* ?

$T \leftrightarrow A$ w.r.t. basis of V

$T^* \leftrightarrow A^*$... (= A^t if \mathbb{R})

$T^t \leftrightarrow A^t$... dual basis of V^*

If \mathbb{R} ,

$T^*: V \rightarrow V$

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

both have matrix A^t

$\langle, \rangle \rightsquigarrow \mathcal{I}: V \xrightarrow{\sim} V^*$

If identify V, V^* via \mathcal{I} : T^t becomes T^* .

V, \langle, \rangle

$$T: V \rightarrow V$$

T lin. op.

$$T \in \text{End } V$$

\uparrow

A , matrix w.r.t. o.n.b.

$\text{Hom}(V, V)$

\rightsquigarrow adjoint $T^*: V \rightarrow V$

$$T^* \in \text{End } V$$

$$Q: \text{End } V \rightarrow \text{End } V$$

$$\begin{array}{ccc} \psi & & \psi \\ T & \longmapsto & T^* \end{array}$$

corresp

$$M_n(F) \rightarrow M_n(F)$$

$$\begin{array}{ccc} \psi & & \psi \\ A & \longmapsto & A^* = \begin{cases} A^t & \text{if } F/\mathbb{R} \\ \bar{A}^t & \text{if } F/\mathbb{C} \end{cases} \end{array}$$

If F/\mathbb{R} , lin transp $A \mapsto A^t$

If F/\mathbb{C} , conj lin $A \mapsto \bar{A}^t$

$$(TU)^* = U^* T^*$$

since true for transpos.

$$T^{**} = T$$

similar F/\mathbb{C} , with conj
b/c ok for matrix.

If $T^* = T$ we say
 T is self-adjoint.

Which lin transf's are self-adjoint?

$T \leftrightarrow A$ wrt orthonormal basis?

$$T = T^* \Leftrightarrow A = A^*$$

If \mathbb{R} : $A = A^t$: A is symmetric:
 $a_{ji} = a_{ij}$

If \mathbb{C} : $A = \bar{A}^t$, $A^t = \bar{A}$

A is cong sym (Hermitian)

$$a_{ji} = \overline{a_{ij}}$$

V inn. prod. sp $\langle \cdot, \cdot \rangle$

lin op $T: V \rightarrow V$, iso of v.s.s
ignores $\langle \cdot, \cdot \rangle$.

Under what circumstances does T
preserve $\langle \cdot, \cdot \rangle$? i.e. $\langle v, w \rangle = \langle Tv, Tw \rangle$

— iso of inner product spaces

T preserves inner product \Leftrightarrow ?
 $\Leftrightarrow T$ preserves norm $\| \cdot \|$
 \uparrow
 $\|T(v)\| = \|v\|$ (PS13 #5)

Q: Which T preserve norm?

Ans: Nec + Suf condition:

$$T^* = T^{-1}$$

Reason: $\langle v, v' \rangle \stackrel{?}{=} \langle Tv, Tv' \rangle$
 \uparrow
 $\stackrel{?}{=} \langle v, T^*Tv' \rangle$ (Ass: invertible)

$$\langle v, v' \rangle = \langle v, T^*Tv' \rangle \quad \forall v, v'$$

$$\langle v, T^*Tv' - v' \rangle = 0 \Rightarrow v \perp T^*Tv' - v'$$

$$\Rightarrow v \perp 0$$

$$T^*Tv' - v' = 0 \quad \forall v' \Rightarrow T^*Tv' = v'$$

$$\Leftrightarrow T^*T = I \Leftrightarrow T^* = T^{-1}$$

Impt class of lin op's T :
 pres. inner prod; pres. norm; $T^* = T^{-1}$

These are called unitary.

Reason for terminology:

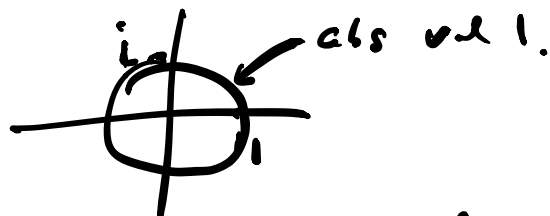
T preserves length.

So if $\|v\|=1$ then $\|T(v)\|=1$.

Preserves set of vectors of unit length

Every eigenvalue has abs. value 1.

Eigenvalues lie on unit circle in \mathbb{C}



Ex. 1×1 mx. $A = (a)$

$$\text{Unitary} \Leftrightarrow \bar{a} = a^{-1} \Leftrightarrow a\bar{a} = 1 \Leftrightarrow |a| = 1$$

\Downarrow $|eigenvalue| = 1$ $|a|^2$ on unit circle

$$\text{Ex. } A = \begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix}$$

$$A^* = \bar{A}^t = \begin{pmatrix} -i & 0 \\ 0 & i \end{pmatrix} = A^{-1}$$

Real mx $A \in M_n(\mathbb{R})$;

unitary condition is $A^t = A^{-1}$.

Real unitary matrix: Orthogonal matrix
(list)

name: preserve inner prod.

\Rightarrow " orthogonality

$$V \perp W \Rightarrow TV \perp TW.$$

Real matrix: rotations, reflections.

Unitary matrix: $A^* = A^{-1}$
invertible.

Under \cdot :

closed

assoc

id

inverse

group

$$\text{Unitary sp } \left\{ \begin{array}{l} n \times n \text{ } \mathbb{C} \\ \text{Unitary matrix} \end{array} \right\} = U(n) \subset GL_n(\mathbb{C})$$

$$\text{Orthog. sp } \left\{ \begin{array}{l} n \times n \text{ } \mathbb{R} \\ \text{Orthog. matrix} \end{array} \right\} = O(n) \subset GL_n(\mathbb{R})$$

[Caution: Sometimes refer to a

$n \times n$ matrix as "orthogonal" if $A^t = A^{-1}$.

See H&K, p 304. Different from unitary]

Thm (H&K, Thm 15, p 312)

Let $T: V \rightarrow V$ be self adjoint operator

on an inner prod space V . (so hermitic,
or symmetric
if real)

a) The every cx root of char. poly.
 $P_T(x)$ is a real #.

(ie all the eigenvalues are real)

b) Any two eigenvectors corresp
to distinct eigenvalues are
orthogonal. (not just lin ind)

Pf is an easy computation.

In 1×1 case:

$A = (a)$ is self adj.

$$\Leftrightarrow A^* = A \Leftrightarrow \bar{a} = a \Leftrightarrow a \in \mathbb{R}$$

the one eigenvalue is $a \in \mathbb{R}$.

Thm (Thm 18, p 314, H + K)

(Spectral thm for self-adjoint operators)

Let $T: V \rightarrow V$ be a self-adjoint
lin. op. on a fin dim inner prod sp V .

Then V has a only consisting of
eigenvectors of T .

Pf is by induction on dim,
using prev result.

So: For a self-adjoint linear
 $T: V \rightarrow V$ (\leftrightarrow real sym mx,
or cx Her m)

Can find a basis
that's good in two ways:

1) o.n.b. (\therefore basis is easy to
work with)

2) puts T in diagonal form.

(Since basis of eigenvectors for T)

($\therefore T$ is easy to work with in this basis)

In particular:

Every real sym mx is
diagonalizable over \mathbb{R} .

(More: not o.n.b.)

A real sym mx (or cx Her m)

$$\rightarrow P_A(x) = \prod_{i=1}^n (x - a_i) \quad a_i \in \mathbb{R}$$

$$A \sim D = \begin{pmatrix} a_1 & & 0 \\ & \ddots & \\ 0 & & a_n \end{pmatrix} \in GL_n(\mathbb{R})$$

Just chg of basis from given
 one (eg e_1, \dots, e_n) to a new one.

Takes one to one; \therefore preserves $<, >$.

Chg of basis matrix C is unitary.
 So is C^{-1} .

$$D = C^{-1} A C$$

$$C \text{ unitary; } C^{-1} = C^*$$

$$\boxed{D = C^* A C}$$

$\underbrace{\hspace{10em}}_{\text{diag}} \quad \underbrace{\hspace{10em}}_{\text{unitary}}$

(easier to compute than C^{-1})

Summary: A, D are unitarily equivalent.

A non $C \times$ herm $n \times n$

$\Rightarrow A$ is unitarily equiv to a real diag. $n \times n$.

A non real sym $n \times n \Rightarrow A$ is orthogonally equiv to $\dots D$.

Does this carry over to other T, A ?

Not in gen. — might not be diagonalizable
— eigenvectors might not be \perp .

But — carries over into \mathbb{R} part of essential to a class of T, A : normal.

$$AA^* = A^*A \quad (\text{Similarly for } T)$$

Thm (Spectral Thm for normal operators)

(H & K, p 317, Th 22 + Cor)

$A \in M_n(\mathbb{C})$ normal

$\iff A$ is unitarily equiv to a diag $\text{diag} \lambda_i \in \mathbb{C}$.

(P f: \implies : Every lin tr. / \mathbb{C}

is Δ 'ble w/out orb;

normal + Δ 's \implies diag;

\impliedby : UDU^{-1}
normal

Ex. A hermitian (and real sym)
 \Rightarrow normal ($A = A^*$)

Ex. A unitary (and real orthogonal)
($A^* = A^{-1}$)

S: if A is unitary / \mathbb{C}

The \exists of eigenvectors,
i.e. diagonalizable over \mathbb{C} .

Ex. $A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \in O(2) \subset U(2)$

$A \sim D = \begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix}$ ↑ eigenvectors: $i, -i$
unitary ↑ not real max.

Caution: A is not real diagonalizable.
(no real eigenvectors)

What if we work over other fields?

No notion of pos. pos def.

Chap 10, H & K.

V / F

Bilinear form $f: V \times V \rightarrow F$

$f(v, w) \in F$ generalized inner prod
 $\mathcal{B} = \{v_1, \dots, v_n\}$ basis of V
 $a_{ij} = f(v_i, v_j)$ $A = (a_{ij})$
 $n \times n$ of f w.r.t \mathcal{B} .
 rank of A ("rank of f ")

So f is non-degenerate if
 $\text{rank} = \dim = n$.

Eg: $\forall v \neq 0 \exists w \neq 0 : f(v, w) \neq 0$

True for inn prod on real v.s:
 $v \neq 0 \Rightarrow v^\perp \neq V$. Take w not
 orthog to v .

Symmetric bilinear form:

$f(v, w) = f(w, v)$
 $\Leftrightarrow A$ is sym ($A = A^t$)
 (true for real inn. prod.)

Say $v \perp w$ (w.r.t f)
 if $f(v, w) = 0$.

f sym bil form
(gen'l'ly real in prod)

Defn $q(v) = f(v, v)$
(gen'l'ly $\|v\|^2$) Quadratic form

Ex. f on F^n be "dot product"

$$v = (a_1, \dots, a_n)$$

$$w = (b_1, \dots, b_n)$$

$$f(v, w) = \sum a_i b_i$$

$$q(v) = a_1^2 + \dots + a_n^2$$

$f \rightsquigarrow q$

Sym bil
form

quad
form

$f \leftrightarrow q$

2 vbls

1 vbl

✓

Conversely: can recover f from q :

Polarization identity: (gen'l'ly P.I. \mathbb{R})

$$f(v, w) = \frac{1}{4} q(v+w) - \frac{1}{4} q(v-w)$$

Unless $2=0$ in F ("characteristic 2")

V f.d.s / F , $\text{char } F \neq 2$

f sym bilinear form on V .

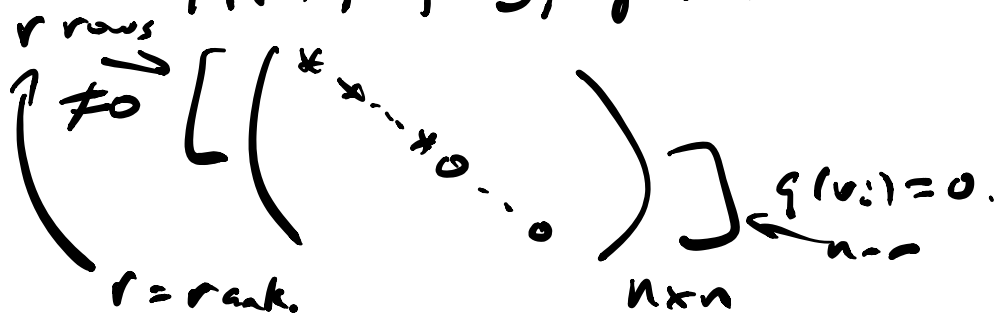
Gram-Schmidt works as before.

\exists orthog basis v_1, \dots, v_n of V :

$$f(v_i, v_j) = 0 \text{ for } i \neq j.$$

Es: $A = (a_{ij})$ $a_{ij} = f(v_i, v_j)$
 diagonal max.

H + K, Th 3, p 369



If f non-deg (ok: no 0's).

Chk $\mathcal{E}: q(v_i) = f(v_i, v_i)$

If $q(v_i) \neq 0$, replace v_i by $\frac{v_i}{\sqrt{q(v_i)}}$
 as $\sqrt{\quad}$

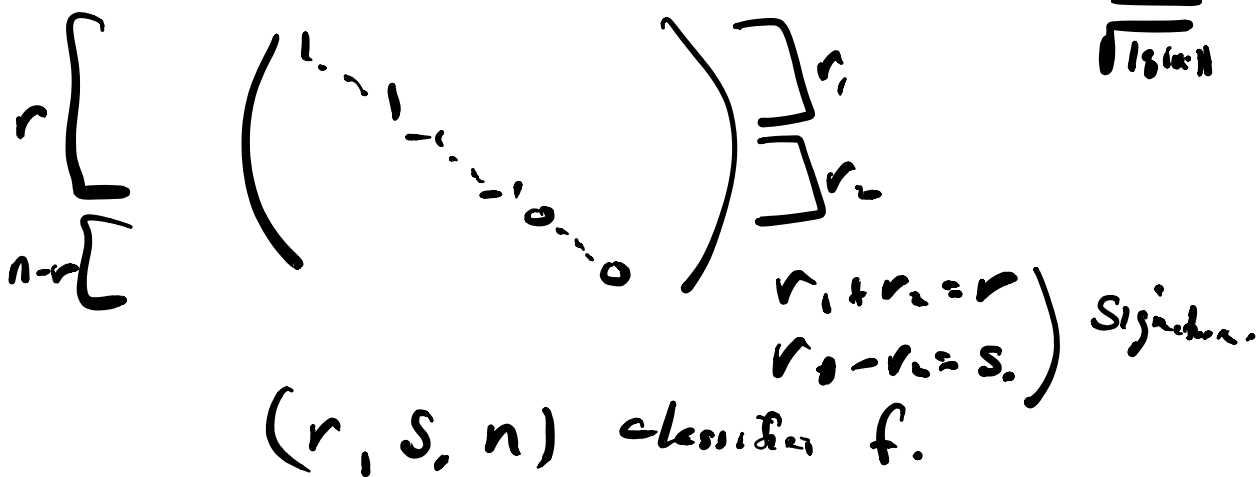


Classify f 's / \mathcal{E}

Over \mathbb{R} : $g(v_i) = f(v_i, v_i)$

If $g(v_i) > 0$, replace v_i by $\frac{v_i}{\sqrt{g(v_i)}}$

If $g(v_i) < 0$, replace v_i by $\frac{v_i}{\sqrt{|g(v_i)|}}$



H & K, p 370, Th 4 & 5.

Some applications

1) Rule for finding max/min
for $f: \mathbb{R}^n \rightarrow \mathbb{R}$

— via Hessian mx

$$A = \left(\frac{\partial^2 F}{\partial x_i \partial x_j} \right)$$

At a crit pt: real sym. mx
is diagonizable

Trace cost pt to \mathcal{O} .

$$A \sim D = \begin{pmatrix} a_1 & & 0 \\ & \ddots & \\ 0 & & a_n \end{pmatrix}$$

a_i 's are eigenvalues. \uparrow

$$f = C + a_1 x_1^2 + \dots + a_n x_n^2 + \text{h.o.t.}$$

If all $a_i > 0$: min.
 \uparrow eigenvalues

If all $a_i < 0$: max

If mixed: saddle pt.

If some $a_i = 0$: This test fails.

Special case $n=2$.

use rule for max/min of
a f of 2 vls.

2) Page Rank algorithm.
in Google
— using eigenvalues
Wikipedia

3) Singular Value decomposition

- statistics,
- Signal processing,
- Weather forecasting, --

^a "Diagonal" $n \times n$ $n \times n$

$$A \quad n \times n \quad \text{or } T: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$\text{" } (a_{ij}) : a_{ij} = 0 \text{ unless } i=j.$$

Make ch of vls on $\mathbb{R}^n \leftarrow$ (can be diff
- - - - $\mathbb{R}^m \leftarrow$ even if not

Then can do this

- via algorithm that uses
pos def nx (pos def f)
+ adjoints.

- Wikipedia