


When does a matrix have enough eigenvectors to form a basis?



When is a matrix diagonalizable?

Thm let A be a real symmetric matrix.

- (a) All eigenvalues are real. ✓
- (b) Eigenvectors for distinct eigenvalues are orthogonal. ✓
- (c) There is an orthonormal basis of eigenvectors of A for \mathbb{R}^n . ✓
- (d) All symmetric matrices are diagonalizable by an orthogonal matrix in \mathbb{R}^n . ✓

Pf: (a) Since A is real and symmetric, $\bar{A} = A$, and $A^T = A$.

So all real matrices have possibly complex eigenvalues / eigenvectors

$A : \mathbb{C}^n \rightarrow \mathbb{C}$, so we consider complex scalars and complex dot product.

$$\left(z \cdot w = z^T \bar{w} \right)$$

Claim: $Av \cdot w = v \cdot Aw \quad \forall v, w$

$$\begin{aligned} \text{Pf } Av \cdot w &= (Av)^T \bar{w} = v^T A^T \bar{w} \\ &= v^T \bar{A} \bar{w} = v^T (\bar{A} \bar{w}) \\ &= v \cdot Aw \end{aligned}$$

We can use this formula to λ is real.

Let λ be an eigenvalue of A , let
 v be a corresponding eigenvector.

Then

$$Av \cdot v = \lambda v \cdot v = \lambda \|v\|^2$$

$$\begin{aligned} Av \cdot v &= * \\ &= v \cdot Av = v \cdot (\lambda v) \\ &= \bar{\lambda} v \cdot v = \bar{\lambda} \|v\|^2 \end{aligned}$$

$$\Rightarrow \lambda \|v\|^2 = \bar{\lambda} \|v\|^2$$

In fact $\|v\|^2 \neq 0$ because
 v is an eigenvector, $v \neq 0$.

$$\Rightarrow \lambda = \bar{\lambda} \Rightarrow \lambda \in \mathbb{R}.$$

∴ all eigenvalues of symmetric
matrices are real!

(b) Let λ, μ be distinct eigenvalues.

Let $v \in V_\lambda$, $w \in V_\mu$. $v, w \neq 0$.

WTS $v \cdot w = 0$.

Consider $Av \cdot w$. By part (a),

we no longer have any reason to consider complex scalars.

λ, μ, v, w are all real. We're back to the real dot product.

$$Av \cdot w = \lambda v \cdot w$$

$$\begin{aligned} Av \cdot w &\stackrel{*!!}{=} v \cdot Aw = v \cdot (\mu w) \\ &= \mu v \cdot w \end{aligned}$$

$$\lambda(v \cdot w) = \mu(v \cdot w)$$

We know that
 $\lambda \neq \mu$.
 $\lambda - \mu \neq 0$.

$$\implies (\lambda - \mu)(v \cdot w) = 0 \quad \lambda - \mu \neq 0$$

$$v \cdot w = 0.$$

So eigenvectors for distinct eigenvalues
are orthogonal.

(c) We need to show that \mathbb{R}^n
has an orthonormal basis of eigenvectors
of A .

so far, (b) says that $\lambda \neq \mu$

$$v_\lambda \perp v_\mu.$$

Oh, since v_λ are orthogonal to
each other, then you can find
a basis of v_λ and G-S each
one individually. Then you'll
get an ~~orthonormal basis~~
only mutually orth.

If $v_{1,\lambda} \dots v_{n,\lambda} \in V_\lambda$

$w_{1,\mu}, \dots, w_{r,\mu} \in V_\mu$

by G-S $v_{i,\lambda} \cdot v_{j,\lambda} = 0, w_{i,\mu} \cdot w_{j,\mu} = 0$

by (b) $v_{i,\lambda} \cdot w_{j,\mu} = 0$

All together $\{v_1, \dots, v_n, w_1, \dots, w_r\}$
are mutually orthogonal!

But doesn't say that this set
spans...

So it suffices to show that
 A is diagonalizable, then G-S or
all the V_λ individually, will give you
an orthonormal basis of
eigenvectors.

Induction proof ..

Let v be an eigenvector for λ .

Let $W = \underline{\text{Span}(v)^\perp}$.

Claim: W is an invariant subspace
for A .

- If $w \in W$, then $Aw \in W$.
(definition of invariant yesterday --)

Let $w \in W$, $\underline{w \cdot v = 0}$.

Then $\underline{Aw \cdot v} = w \cdot Av$

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

$$= \lambda \cdot 0 = \underline{0}$$

So $Aw \in \text{Span}(v)^\perp = W$.

So $\text{Span}(v)^\perp$ is invariant under A .

By rank-nullity 1

$$\dim(\text{span}(v)) + \dim(\text{span}(v)^\perp) = n = \dim(\mathbb{R}^n).$$

$$1 + \dim(\text{span}(v)^\perp) = n$$

$$\dim(\text{span}(v)^\perp) = n-1.$$

$$\dim(W) = n-1$$

W is invariant and $\dim(W) = n-1$
 $\dim(\mathbb{R}^n) = n$

$A : \mathbb{R}^n \rightarrow \mathbb{R}^n$ but since W is invariant

$A|_W : W \rightarrow W$

restrict domain to W

wdomain is W since it's invariant

$$A : \mathbb{R}^n \longrightarrow \mathbb{R}^n \quad \dim(\mathbb{R}^n) = n$$

A symm

u_1, \dots, u_n

$$A : W \longrightarrow W \quad \dim(W) = n-1$$

A symm

↑

(Induction on $\dim(\mathbb{R}^n) = n$)

u_1, \dots, u_{n-1}

$$A : W^1 \longrightarrow W^1 \quad \dim = n-2 \quad u_1, \dots, u_{n-2}$$

↓

⋮

↑

u_1, u_2

↑ (b)

$$A : \mathbb{R}^1 \longrightarrow \mathbb{R}^1 \quad \dim = 1 \quad u$$

so $A|_W: W \rightarrow W$

by recursion, we can

make $\{u_1, \dots, u_{n-1}\}$ an orthonorm.

basis of eigenvectors in $\text{span}(v)^\perp$



$$\left\{ u_1, \dots, u_{n-1}, \frac{v}{\|v\|} \right\}$$

is an orthonormal basis of
eigenvectors in \mathbb{R}^n

(d)

Since $\{u_1, \dots, u_n\}$ is a orth. basis
of eigenvectors, we can diagonalize
by a orthog matrix

$$Q = (u_1, \dots, u_n) \longrightarrow Q^T = Q^{-1}$$

Q orthogonal

$$A = Q \Delta Q^{-1} = Q \Delta Q^T.$$

□

Ex Let $A = \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix}$.

$$\begin{aligned}P_A(\lambda) &= (3-\lambda)^2 - 1 \\&= \lambda^2 - 6\lambda + 8 \\&= (\lambda-4)(\lambda-2)\end{aligned}$$

$\lambda = 4, 2$ (a) say that
these are real,
which they are.

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

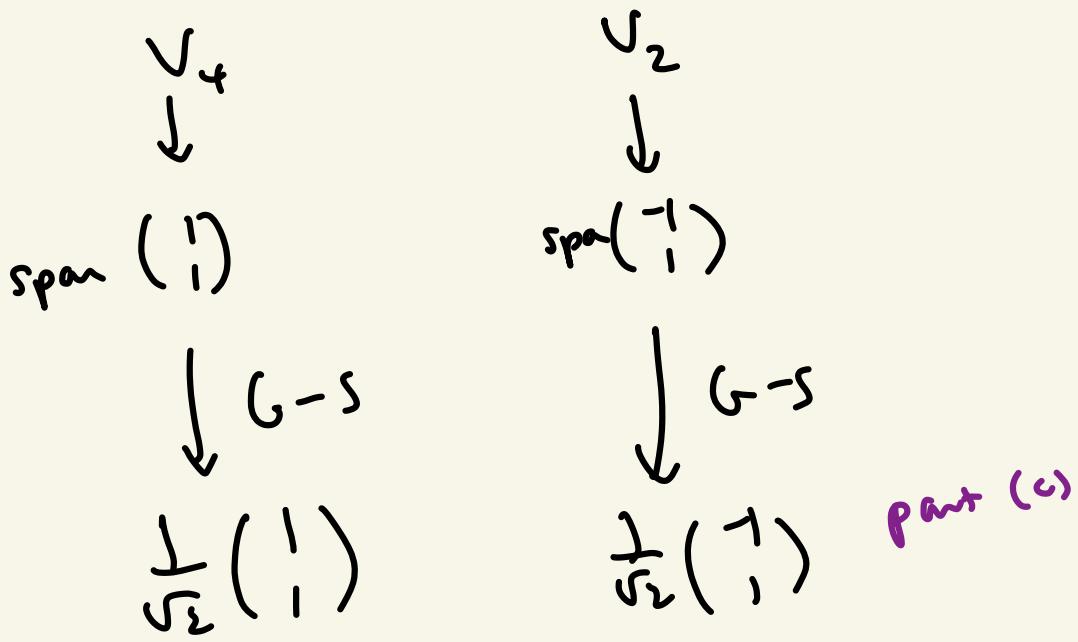
$$V_4 = \ker(A - 4I) = \text{span}(\begin{pmatrix} 1 \\ 1 \end{pmatrix})$$
$$v = \begin{pmatrix} 1 \\ 1 \end{pmatrix} *$$

$$V_2 = \ker(A - 2I) = \text{ker}(\begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix})$$
$$= \text{span}(\begin{pmatrix} -1 \\ 1 \end{pmatrix}). v = \begin{pmatrix} -1 \\ 1 \end{pmatrix} *$$

In fact $\begin{pmatrix} 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} -1 \\ 1 \end{pmatrix} = 0$ (b) predicted this

(c) \mathbb{R}^2 should have an orthonormal basis of eigenvectors

$G-S \quad V_2, V_4 \quad$ individually



$$Q = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} \quad \text{orthonormalizing}$$

$$\begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{-1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix}$$

called the spectral decomposition

$$B = S^{-1} A S$$

new coord

$$S = (v_1 \dots v_n)$$

$$A = S B S^{-1}$$

$$B = \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix}, \text{ is } \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}$$

word notes

If A_1 the adjoint A^*
 $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$
 is the unique matrix

s.t.

$$\langle Av, w \rangle = \langle v, A^*w \rangle \quad \forall v, w$$

$$A : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

K L

$$A^* = K^{-1} A L$$

$$\langle v, w \rangle = 3v_1v_1 + 6v_2w_2$$

$$\rightarrow (v_1, v_2) K \begin{pmatrix} w_1 \\ w_2 \end{pmatrix} =$$

$$\text{If } K = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$(x \ y) \begin{pmatrix} x^2 \\ \frac{1}{2}y^2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

$$x^2 + 4xy + y^2$$

$$\rightarrow (v_1, v_2) \begin{pmatrix} 3 & 0 \\ 0 & 6 \end{pmatrix} \begin{pmatrix} w_1 \\ w_2 \end{pmatrix}$$

$$\begin{matrix} m \times m & m \times n & h \times h \\ K & A & L \end{matrix}$$

$$P^2 = P. \quad \text{Let } v \in \text{img}(P).$$

$$\text{then } Pv = v.$$

$$v \in \text{img}(P) \Rightarrow v = Pv$$

$$Pv = P \cdot Pv = P^2 v = Pv = v$$

$$Pv = Iv \quad \forall v \in \text{img}(P).$$

$$N^k = 0 \quad \text{for some } k.$$

$$\begin{bmatrix} 0 & & & \\ 0 & 0 & & \\ 0 & 0 & 0 & \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$e^{i\theta}$ = $\cos\theta + i\sin\theta$

Unitary

Hermitian

$$\frac{e^A}{1} = I + A + \frac{1}{2!}A^2 + \frac{1}{3!}A^3 + \frac{1}{4!}A^4 + \dots$$

$$\underline{e}^D = e^{D+N} = \underline{\underline{e}^D e^N}$$

$$\begin{bmatrix} * & * \\ x & x \\ x & x \end{bmatrix} = \begin{bmatrix} * & * \\ x & x \\ x & x \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & x \end{bmatrix}$$

$$\underline{e}^D = e^{\begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}} = \begin{bmatrix} e^{\lambda_1} & & \\ & \ddots & \\ & & e^{\lambda_n} \end{bmatrix}$$

$$N^5 = 0$$

$$e^N = I + N + \frac{1}{1!} N^2 + \frac{1}{2!} N^3 + \frac{1}{3!} N^4 + \dots$$

~~$$+ \frac{1}{5!} N^5 + \dots$$~~

$$y_1 \dots y_n \quad y_n(t)$$

$$y_1' = a_{11}y_1 + \dots + a_{1n}y_n$$

⋮
⋮

$$y_n' = a_{n1}y_1 + \dots + a_{nn}y_n$$

$$\begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}' = A \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$f' = af$$

↓

$$f = e^{ax}$$

$$\begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} = e^{At}$$

$$1 + r + r^2 + r^3 + \dots = \frac{1}{1-r}$$

$$\lim_{n \rightarrow \infty} (1-r)(1+r + \dots + r^n) = (1-r)\sum$$

$$\lim_{n \rightarrow \infty} 1 - r^{n+1} = 1$$

$$\sum = \frac{1}{1-r}$$

$$\chi : G \rightarrow M \xrightarrow{\times r} \mathbb{R}$$

$$\sum_n \frac{\chi(n)}{n} \quad \text{Dirichlet L-function}$$

$\mathbb{Z} \Rightarrow (1), (2), (3), (4) \dots$

prime ideals
(2), (3), (5), (7)

$\mathbb{Z}/(p)$ \Rightarrow simple

$$(6) \rightarrow 6 \quad 2 \cdot 3 = 6$$

$2;3 \notin (6)$

C as a ring

$\mathbb{C}[x]$ is a ring

prime ideals = $\{ (f) \mid f \text{ is an irreducible polynomial} \}$

$$= \{ (\pi - \omega) \mid \} \Rightarrow a$$

$\mathbb{C}[x,y]$

→ prime ideals



$$x=a, y=b$$

(x^2-y) irreducible
 $f(x,y)$ and $\underline{(x-a, y-b)}$



graph of
 $f(x,y) = 0$

(a,b)

$x^2-y=0$