


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

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

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

Claim: $A v \cdot w = v \cdot A w \quad \forall v, w$

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

We can use this formula to λ is real.

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

Then

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

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

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

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.

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

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

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

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 $v_{1,\lambda} \dots v_{k,\lambda} \in V_\lambda$

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

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 $\{v_1, \dots, v_k, w_1, \dots, w_r\}$
are mutually orthogonal!

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

So it suffices to show that
A is diagonalizable, then G-S on
all the V_λ individually, will get 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 $w \in W$, $\underline{w \cdot v = 0}$.

$$\begin{aligned} \text{Then } \underline{Aw \cdot v} &= w \cdot Av \\ &= w \cdot \lambda v = \lambda (w \cdot v) \\ &= \lambda \cdot 0 = \underline{0} \end{aligned}$$

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

codomain is W since it's invariant

$$A : \mathbb{R}^n \longrightarrow \mathbb{R}^n$$

A symm




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

u_1, \dots, u_n


$$A : W \longrightarrow W$$


A symm

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


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

u_1, \dots, u_{n-1}



$$A : W' \longrightarrow W'$$


$$\dim = n-2$$

u_1, \dots, u_{n-2}





u_1, u_2

$\uparrow (v)$

$$A : \mathbb{R}^1 \longrightarrow \mathbb{R}^1$$

$$\dim = 1$$

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 on \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

$$A = Q \Lambda Q^{-1} = Q \Lambda Q^T. \quad \square$$

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) says that these are real, which they are.

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

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

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

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 v_2, v_4 individually

$$v_4 \downarrow \\ \text{span} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\downarrow \text{G-S} \\ \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$v_2 \downarrow \\ \text{span} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$\downarrow \text{G-S} \\ \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

part (c)

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

diagonalizing

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

called the spectral decomposition

$$B_{\text{new coord}} = S^{-1}AS$$

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

$$A = SBS^{-1}$$

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

word netes

If A , the adjoint A^*

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

is the unique matrix

s.t.

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

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

K L

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

$$\langle v, w \rangle = 3v_1w_1 + 6v_2w_2$$

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

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

$$(x \ y) \begin{pmatrix} \overset{x^2}{1} & 2 \\ 2\sqrt{1} & \underset{y^2}{1} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

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

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

$$\begin{array}{ccc} m \times m & m \times n & n \times n \\ K^{-1} & A & L \end{array}$$

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

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

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

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

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

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

$$\begin{bmatrix} 0 & 1 & & \\ & 0 & 1 & \\ & & 0 & 1 \\ & & & 0 \end{bmatrix}$$

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

Unitary
Hermitian

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

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

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

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

$$e^N = \underbrace{I + N + \frac{1}{2!} N^2 + \frac{1}{3!} N^3 + \frac{1}{4!} N^4}_{N^5 = 0} \left. \vphantom{e^N} \right\}$$

~~$+ \frac{1}{5!} 0 + \dots$~~

y_1, \dots, y_n $y_2(t)$

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

\vdots

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

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

$$f' = Af$$

$$f = e^{Ax}$$

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

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

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

$$\lim = 1 - r^{n+1} = 1$$

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

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

$$\sum_n \frac{\chi(G_n)}{n}$$

Dirichlet L-function

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

prime ideals

$(2), (3), (5), (7)$

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

$(6) \ni 6$

$2 \cdot 3 = 6$

$2, 3 \notin (6)$

\mathbb{C} as a ring

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

prime ideals = $\{(f)$

f is an irreducible polynomial

$= \{(x-a) \mid \} \Rightarrow a$

max ideals of \mathbb{C}

\Downarrow

elements of \mathbb{C}

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

\rightsquigarrow prime ideals



$$(x^2 - y)$$

irreducible
 $f(x, y)$

$$\text{and } \frac{x=a, y=b}{x^2 - y = 0}$$



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



$$\frac{(a, b)}{x^2 - y = 0}$$

$$\underline{x^2 - y = 0}$$