

When u multiply a ~~trans~~ eigen vector
by its transformation, it becomes
 λ times eigen vector.

$$Ax = \lambda x, \text{ also } A^2x = \lambda(Ax) \\ A^2x = \lambda^2 x$$

if the transformation squashes onto a lower dimension
null space of transformation are eigen vectors,
with eigen values 0.

$$\begin{vmatrix} a-\lambda & b \\ c & d-\lambda \end{vmatrix} = 0 \Rightarrow (a-\lambda)(d-\lambda) - bc = 0$$

$$\lambda^2 - \lambda(a+d) - bc = 0$$

$$+ ad$$

$$\lambda^2 - \lambda(a+d) + (ad-bc) = 0$$

$\lambda \rightarrow$ eigen values

Sum of eigen values = $+(a+d) = \text{Trace}$

product of eigen vals = $(ad-bc) = \text{Det.}$

Triangular matrix eigen values

$$\begin{vmatrix} 1 & 3 \\ 0 & 0 \end{vmatrix} \Rightarrow \text{upper } \Delta^k \text{ matrix}$$

eigen values = 1, 0

reason: Trace \rightarrow sum of diagonal entries = sum of eigen values

~~product~~ = prod of diagonal entries = prod of eigen values

2) from above diagonal entries
are eigen values

if A and B share same eigen ~~val~~
vector x , and A scales or squishes
 x by λ , B scales or squishes x by β

$$ABx = (\beta\lambda)x$$

proof

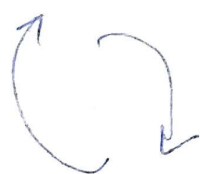
$$\begin{aligned} ABx &= A(Bx) \\ &= A(\beta x) \\ &= A\beta x \\ &= \beta(Ax) \\ &= \beta\lambda x \\ &= \underline{\underline{\beta\lambda x}} \end{aligned}$$

$$\begin{aligned} BAx &= B(Ax) \\ &= B\lambda x \\ &= \lambda(Bx) \\ &= \lambda\beta x \\ &= \underline{\underline{\lambda\beta x}} \end{aligned}$$

we know $\beta\lambda = \lambda\beta$
(scalars)

$$\Rightarrow ABx = BAx$$

if $AB = BA$ then



A, B share same independent
eigen vectors

A has eigen values $\rightarrow \lambda_1, \lambda_2$

$A + nI$ has eigen values $\rightarrow \lambda_1 + n, \lambda_2 + n$

intuition

they are just added a bias terms for each vector.



$$\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 3 & 0 \\ 0 & 4 \end{bmatrix}$$

Q) u is eigen vector of uv^T ($|uv^T| = 0$
rank = 1)

$$u = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}_{2 \times 1} \quad v = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}_{2 \times 1}$$

$$\begin{bmatrix} u_1 \\ u_2 \end{bmatrix}_{2 \times 1} \begin{bmatrix} v_1 & v_2 \end{bmatrix}_{1 \times 2} = \begin{bmatrix} u_1 v_1 & u_1 v_2 \\ u_2 v_1 & u_2 v_2 \end{bmatrix}$$

u, v^T

is rank 1 \Rightarrow it squashes 2D plane into a 1D line

$\Rightarrow \det = 0 \Rightarrow \lambda = 0$ is a eigen value

λ_1 be other eigen value
 $\lambda_2 = 0$

$$\lambda_1 + \lambda_2 = \text{Trace}$$

$$\lambda_1 + 0 = u_1 v_1 + u_2 v_2$$

$$\boxed{\lambda_1 = u_1 v_1 + u_2 v_2}$$

challenging prob

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Heisenberg eqn.

$$\|Ax\| \|Bx\| \geq \frac{1}{2} \|x\|^2$$

$$AB - BA = I$$

$$ABx - BAx = x$$

$$x^T ABx - x^T BAx = x^T x$$

$$x^T A(Bx) - x^T B(Ax) = x^T x$$

$$x^T A^T (Bx) + x^T (B^T) (Ax) = x^T x$$

$$(Ax)^T (Bx) + (Bx)^T (Ax) = x^T x$$

$$|(Ax)^T Bx| \leq \|Ax\| \|Bx\|$$

$$|(Bx)^T (Ax)| \leq \|Bx\| \|Ax\|$$

$$\|Ax\| \|Bx\| + \|Ax\| \|Bx\| \geq x^T x$$

$$2 \|Ax\| \|Bx\| \geq x^T x$$

$$|A+B| \geq |A| + |B| \quad (\text{from Wilkinson's det thm})$$

$$|x^T x| = \text{len}(x)^2$$

$$\det(A+B) \geq (\det A)^{\frac{1}{n}} (\det B)^{\frac{1}{n}}$$

$$(Ax)^T (Bx) + (Bx)^T (Ax) = x^T x = \text{len}(x)^2$$

$$|(Ax)^T (Bx) + (Bx)^T (Ax)| = |x^T x|$$

$$|(Ax)^T (Bx)| + |(Bx)^T (Ax)| \geq (\text{len}(x))^2$$

max value
 $\hookrightarrow \|Ax\| \|Bx\|$

$$2 \|Ax\| \|Bx\| \geq (\text{len}(x))^2$$

$$2 \|Ax\| \|Bx\| \geq |x^T x|$$

$$2 \|Ax\| \cdot \|Bx\| \geq |x^T x|$$

$$\frac{\|Ax\|}{|x|} \cdot \frac{\|Bx\|}{|x|} \geq \frac{1}{2}$$

not wrong