

Laplacian:

$$L_G = D_G - A_G \quad \text{— adjacency matrix.}$$

$$D_G = \text{diag}(\deg(v_1), \dots, \deg(v_n))$$

(1) $L_G \cdot \underset{\substack{\uparrow \\ \text{all } 1\text{'s vector}}}{\mathbf{1}} = \mathbf{0}.$

In each row... \rightarrow $\begin{matrix} v_1 & \sim & v_2 \\ & \sim & v_3 \\ & & v_6 \end{matrix}$
 so sum of row is 0.
 $[3 \ -1 \ -1 \ 0 \ 0 \ -1]$

Multiplying by $\frac{1}{\text{sum of row}}$ gives $\mathbf{0}.$

(2) L_G is positive semidefinite.
 In particular, $x^T L_G x = \sum_{i \sim j} (x_i - x_j)^2$
 ~ connected by an edge.

Directed incidence matrix $B(G) = |V| \times |E|$ matrix.
 rows labeled by v
 columns labeled by e .
 entry $e = (v_{i_1}, v_{i_2})$
 $+1$ at row v_{i_1}
 -1 at row v_{i_2}
 0 elsewhere.

NTS:

$$B(G) B(G)^T = L_G$$

$$V \begin{bmatrix} \text{---} \uparrow \text{---} \end{bmatrix} E \begin{bmatrix} \downarrow \\ \vdots \\ \downarrow \end{bmatrix}$$

i, i : number of
edges (1-1 or -1-1)

\Rightarrow deg of i along diags.

$i, j, i \neq j$: only -1 if connected by same
edge. (a 1 and -1 pair)

$$\therefore B(G) B(G)^T = L_G$$

$$\text{Now } x^T B(G) B(G)^T x = (B(G)^T x)^T (B(G)^T x)$$

$$y^T y$$

$$\sum_{i \sim j} (x_i - x_j)^2$$

$B(G)^T x$ is $x_i - x_j$

in each row, where

i is the location of
+1 and j is location of -1.

$\therefore L_G$ is
positive
semidefinite.

$(B(G)^T x)^T (B(G)^T x)$ is dot product of each row of
 $(B(G)^T x)$ w/ itself.

Gram Matrix

$$v_1, \dots, v_k \in V$$

$$G(v_1, \dots, v_k) = (\langle v_i, v_j \rangle) = G$$

(1) G is positive semidefinite.

$$M_{k \times n} = \begin{bmatrix} -v_1- \\ -v_2- \\ \vdots \\ -v_k- \end{bmatrix}$$

$$MM^T = G$$

(row \times transpose col = inner product)

$$x^T MM^T x$$

$$(M^T x)^T (M^T x) \rightarrow \text{by same logic as last}$$

prob, positive semidefinite.

(2) nonsingular $\Leftrightarrow v_1, \dots, v_k$ linearly independent

In \mathbb{R} , $\text{rk}(A^T A) = \text{rk}(A)$, so if M has full row rank, G does too. (Note if M has less cols than rows then rows must be lin. dep and G is singular)

$$\text{Vol}_k(\text{Para}(v_1, \dots, v_k)) = \pm \det \begin{bmatrix} -v_1- \\ -v_2- \\ \vdots \\ -v_k- \end{bmatrix}_{k \times k}$$

where $v_i \in \mathbb{R}^k$.

$$\det G = \det(M) \det(M^T) \\ = \det(M)^2$$

↓
only holds for
equal dim.

we know $\text{Vol}_k(\text{Para}(v_1, \dots, v_k)) = \pm \det M,$
 \parallel
 $\sqrt{\det G}.$

what if $v_1, \dots, v_k \in V$? (arbitrary Euclidean space).

Pick ONB of $\text{span}(v_1, \dots, v_k),$

$$\tilde{M} = [v_1 \dots v_k]_{\text{ONB}}$$

\tilde{M} is $k \times k$ by 1st miracle

$$\text{Is } G(v_1, \dots, v_k) = \tilde{M} \tilde{M}^T?$$

Recall: $\langle v_i, v_j \rangle$ ($v_i \neq v_j$)

$$\text{ONB}(u_1, \dots, u_k) = \sum_{m=1}^k (v_i \cdot u_m)(v_j \cdot u_m)$$

$$= \sum_{m=1}^k \alpha_{im} \alpha_{jm} = [v_i]_{\text{ONB}} \cdot [v_j]_{\text{ONB}}.$$

If some of

v_1, \dots, v_k

lin. dep, vol

0.

(all lower-dim
subspaces have

vol. 0)

$$v_i = \begin{bmatrix} \alpha_{i1} \\ \vdots \\ \alpha_{ik} \end{bmatrix} = [v_i]_{\text{ONB}}$$

$$A^* A = I$$

↳ unitary

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

$$A^*:$$

conjugate
transpose

$$A \in O(n) \quad \text{orthogonal matrix}$$

$$\hookrightarrow A \in M_n(\mathbb{R})$$

$$A^T A = I$$

x is a complex vector:

$$x \in \mathbb{C}^n.$$

$$Ax = \lambda x$$

WTS:

$$|\lambda| = 1.$$

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \quad x^* = [\overline{x_1} \quad \dots \quad \overline{x_n}]$$

$$x^* A^* A x = (Ax)^* (Ax)$$

$$= (\lambda x)^* (\lambda x)$$

$$= \lambda^2 x^* x$$

$$= \lambda^2 \|x\|^2$$

$$x^* x = \|x\|^2$$

b/c each entry is $\overline{x_i} x_i$

$$= |x_i|^2 \text{ and}$$

$$\text{but } x^* A^* A x =$$

$$x^* x = \|x\|^2$$

$$\text{so } |\lambda|^2 \|x\|^2 = \|x\|^2$$

$$\text{and } |\lambda|^2 = 1$$

$$\text{so } \boxed{|\lambda| = 1}$$

□

(Note since x is eigenvector that $\|x\| \neq 0$).

Lecture.

Hermitian dot product in \mathbb{C}^n

$$\underline{x}, \underline{y} \in \mathbb{C}^n$$

$$\underline{x} \cdot \underline{y} = \underline{x}^* \underline{y} = \sum \bar{x}_i y_i$$

$$\underline{x}^* \cdot \underline{x} = \sum |x_i|^2 > 0 \quad \text{unless } \underline{x} = \underline{0}$$

$$= \|\underline{x}\|^2 \quad (\text{definition})$$

Note $\|\underline{x}\|^2 \in \mathbb{R}$.

$$\underline{x} \perp \underline{y} \quad \text{if} \quad \underline{x}^* \underline{y} = 0.$$

Hermitian space: \mathbb{V} over \mathbb{C} with Hermitian

inner product:

positive definite sesquilinear Hermitian inner product

$$A = A^* \rightarrow \text{Hermitian matrix} \quad \downarrow$$

Thm All eigenvalues are real.

$f: V \times V \rightarrow \mathbb{C}$ is sesquilinear if

(1) linear in second variable:

$$(1a) f(x, y_1 + y_2) = f(x, y_1) + f(x, y_2)$$

$$(1b) f(x, \lambda y) = \lambda f(x, y)$$

(2) $\frac{1}{2}$ -linear in first variable

$$(2a) f(x_1 + x_2, y) = f(x_1, y) + f(x_2, y)$$

$$(2b) f(\lambda x, y) = \bar{\lambda} f(x, y) \quad \star$$

Hermitian \Rightarrow

$$f(x, y) = \overline{f(y, x)}.$$

If f Hermitian \Rightarrow quadratic form $f(x, x)$ always real.

$$f(x, x) = \overline{f(x, x)}.$$

Positive definite: $f(x, x) > 0$ unless $x = 0$.

Ex. $f: [0, 1] \rightarrow \mathbb{C}$

$$\langle f, g \rangle = \int_0^1 \bar{f} \cdot g \cdot dt$$

Unitary matrix: columns are ONB of \mathbb{C}^n
 \hookrightarrow w/ Hermitian dot prod

$$A^* A = I.$$

V, W complex Hermitian spaces.

$\varphi: V \rightarrow W$ $(\exists!)(\varphi^*: W \rightarrow V)$ "conjugate-transpose" =
 s.t. $(\forall x \in V)(\forall y \in W)(\langle \varphi(x), y \rangle_W = \langle x, \varphi^*(y) \rangle_V)$ "adjoint"

(Thm. - DO - proof sim to real.)

DO Gram-Schmidt for Hermitian spaces.

Proof $[\varphi^*]_{\text{ONB}} := [\varphi]_{\text{ONB}}^*$

Def. φ is a unitary transformation of V if φ preserves the inner product. \sim "congruence" (preserve distance)

Real Euclidean space

$\varphi: V \rightarrow V$ is an orthogonal transformation if

it preserves the inner product

i.e. $(\forall x, y \in V)(\langle x, y \rangle = \langle \varphi(x), \varphi(y) \rangle).$

but $\langle x, y \rangle = \langle \varphi x, \varphi y \rangle = \langle x, \varphi^* \varphi y \rangle \quad \forall x, y$

so $\varphi^* \varphi = \text{id}$ (identity transformation)

$\forall x, y \quad x^T A y = x^T B y \Rightarrow A = B.$

(DO) $(\forall x, y) (\langle x, \varphi y \rangle = \langle x, \psi y \rangle) \Rightarrow \varphi = \psi.$

Thus, $\varphi^* = \varphi^{-1}$. (In real: $\varphi^T = \varphi^{-1}$)

If $\varphi^* = \varphi^{-1}$ then $\forall x \quad \|\varphi x\| = \|x\|$, so

if $\varphi x = \lambda x$ then

$\|x\| = \|\varphi x\| = |\lambda| \|x\|$ so $|\lambda| = 1.$

Complex Gram matrix

$v_1, \dots, v_n \in V$ (Hermitian space)

$G(v_1, \dots, v_k) = (\langle v_i, v_j \rangle)_{k \times k}$

(DO) $G^* = G$ (G is Hermitian), positive semidefinite, nonsingular $\Leftrightarrow v_1, \dots, v_k$ lin. indep.

Thm. V complex Hermitian space,

$$\underline{e} = (e_1, \dots, e_n) \text{ ONB}$$

$$\Rightarrow \langle \underline{x}, \underline{y} \rangle = [\underline{x}]_{\underline{e}}^* [\underline{y}]_{\underline{e}}$$

$\underline{x} \in \mathbb{F}^n$ b_1, \dots, b_n : basis of \mathbb{F}^n

$$[\underline{x}]_{\underline{b}} = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix}$$

$$\underline{x} = \sum \alpha_i b_i$$

\uparrow
system of lin. equations.

\mathbb{R} or \mathbb{C}
 \uparrow \uparrow
Euclidean Hermitian

e_1, \dots, e_n ONB of V

$$\underline{x} = \sum_{i=1}^n \alpha_i e_i$$

$$\boxed{\alpha_i = \langle e_i, \underline{x} \rangle}$$

(take dot product of \underline{x} w/ any basis vector to get that coefficient.)

\uparrow
coordinates wrt ONB

"Fourier coefficients"

$$\alpha_i = \int f(x) \underline{\cos(2x)} dx$$

\uparrow
ONB.

(DO) $f: [0, 2\pi] \rightarrow \mathbb{R}$ cont

$$\langle f, g \rangle = \int_0^{2\pi} f \cdot g \, dt$$

then $1, \cos t, \sin t, \cos 2t, \sin 2t, \dots$ are orthogonal

Question: characterize those $\varphi: V \rightarrow V$ (V is a complex Hermitian space) for which there exists an orthonormal eigenbasis.

over \mathbb{R} ?

Suppose $A \in M_n(\mathbb{R})$ has ON eigenbasis e_1, \dots, e_n .

$$Ae_i = \lambda_i e_i.$$

$$\underline{x} = \sum \alpha_i e_i$$

$$A\underline{x} = \sum \alpha_i \lambda_i e_i$$

Eigensubspaces U_λ

$$U_\lambda = \{x \in \mathbb{R}^n \mid Ax = \lambda x\}$$

If \exists eigenbasis then

$$\mathbb{R}^n = \bigoplus_{\lambda} U_\lambda$$

If \exists ON eigenbasis then

$\Rightarrow U_\lambda \perp U_\mu$ if $\lambda \neq \mu$ are eigenvalues.

Orthogonal projection

V is finite-dim Euclidean space.

$$\forall v \in V,$$

$$u \in U$$

$$v = u + w \text{ where } u \in U \text{ and } w \in U^\perp.$$

$$U^\perp \oplus U = V \rightarrow$$

Define linear map $\pi_U : v \mapsto u$.

$$u = \pi_U(v). \quad (\pi_U : V \rightarrow U),$$

Cor. If $\varphi : V \rightarrow V$ symmetric (\mathbb{R})

$$\text{then } \varphi = \sum_{\lambda} \lambda \cdot \pi_{U_{\lambda}}.$$

Proof: $x = \sum_{\lambda} u_{\lambda} \text{ where } u_{\lambda} \in U_{\lambda} \forall \lambda.$

$$\text{so } u_{\lambda} = \pi_{U_{\lambda}}(x) \text{ b/c } \sum_{\mu \neq \lambda} u_{\mu} \perp u_{\lambda}.$$

$$\varphi x = \sum_{\lambda} \varphi(u_{\lambda}) = \sum_{\lambda} \lambda u_{\lambda} = \sum_{\lambda} \lambda \pi_{U_{\lambda}}(x).$$

□

This is an equivalent form of the Spectral Thm.

If $Q = Q^T$ in a finite dim Euclidean space
then $V =$ direct sum of orthogonal subspaces

$$u_i \text{ s.t. } Q = \sum \lambda_i \pi_{u_i}.$$

DO $\pi_u^\perp = \pi_u.$

Cor. A symmetric $(R) \Leftrightarrow \exists$ ON eigenbasis.
 \Rightarrow spectral Thm.
 \Leftarrow what we just did

$\pi_u^2 = \pi_u \Leftarrow$ idempotent $x^2 = x$

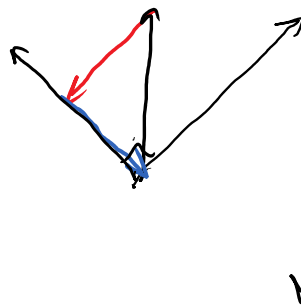
DO If $u_1 \perp u_2$, then $\pi_{u_1} \pi_{u_2} = 0.$

Eigenvalues of projections:

$$\lambda = \{0, 1\}.$$

$$\lambda^2 = \lambda$$

$$\lambda(\lambda - 1) = 0$$



For complex -- the answer is not the same

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

Hermitian : $A^* = A$

Unitary : $A^* = A^{-1}$

Def A is normal if $AA^* = A^*A$.

Thm (Complex Spectral Theorem)

A has an orthonormal eigenbasis $\Leftrightarrow A$ is normal.

Thm $\varphi: V \rightarrow V$ (Hermitian space, finite-dim)

$\Rightarrow \exists$ max. chain of invariant subspaces.

$$\{0\} = U_0 \subset U_1 \subset \dots \subset U_n = V \quad \dim U_i = i$$

$$\varphi(U_i) \subseteq U_i \quad \forall i \in [n]$$


Eq. to saying if $A \in M_n(\mathbb{C})$, then $A \sim \nabla$.

i.e. $(\exists B \in M_n(\mathbb{C})) (\exists B^{-1} \text{ and } B^{-1}AB \text{ is diagonal})$

Thm. $\exists B \in U(n) = \{n \times n \text{ unitary matrices}\}$
 \hookrightarrow orthonormal basis change.

Eq. to finding ONB v_i s.t. $e_i \in U_i$,
 e_1, \dots, e_n .

Proof. Such a basis b_1, \dots, b_n exists \rightarrow Gram-Schmidt + normalize. \square

(DO) If $A \in M_n(\mathbb{R})$ and all eigenvalues of A are real $\Rightarrow A \sim_{\text{ortho.}}$ 

$A, B \in M_n(\mathbb{C})$.

A is unitarily similar to B ($A \sim_u B$) if

$$\exists S \in U(n) \text{ s.t. } B = S^{-1}AS$$

$$\parallel$$

$$S^*AS.$$

(DO) If $A \sim_u B$ and A is normal then B is normal.

(DO) Diag. matrices are normal.

$$D = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}$$


$$D^* = \begin{pmatrix} \bar{\lambda}_1 & & 0 \\ & \ddots & \\ 0 & & \bar{\lambda}_n \end{pmatrix}$$

$$D^*D = \begin{pmatrix} |\lambda_1|^2 & & 0 \\ & \ddots & \\ 0 & & |\lambda_n|^2 \end{pmatrix} = DD^*$$

This shows

\exists ON eigenbasis \Rightarrow normal.

(\Rightarrow)

⇐ May assume A is , why?

Every complex matrix is unitarily similar to a triangular matrix.

(Do) $U(n)$ is a group (closed under multiplication + inverses)

(Do) If a triangular matrix is normal, then it is diagonal

(Prove by induction on \dim -) This shows ⇐.

Spectral Theorem, restated.

If $A \in M_n(\mathbb{R})$ and $A = A^T$ then $\exists S \in O(n)$ s.t. $S^{-1}AS$ is diagonal

(Every symmetric real matrix is orthogonally similar to a diagonal matrix -)

Singular Value Decomposition

$$A \in \mathbb{R}^{k \times l} \Rightarrow$$

$$\exists S \in O(k), \\ T \in O(l)$$

s.t.

$$S^{-1}AT = \begin{bmatrix} \sigma_1 & & & & 0 \\ & \ddots & & & \\ & & \sigma_r & & \\ 0 & & & \ddots & \\ & & & & 0 \end{bmatrix}$$

where $r = \text{rk}(A)$

$$\sigma_1 \geq \dots \geq \sigma_r > 0$$

"singular values of A " -

unique.

Real Euclidean spaces

(DO)

Show

SVD



this result

$$\varphi: \underset{k}{V} \rightarrow \underset{l}{W}$$

$$\Rightarrow \exists \text{ ONB } e_1, \dots, e_k \text{ in } V \text{ and} \\ \text{ONB } f_1, \dots, f_l \text{ in } W \text{ and}$$

$$\sigma_1 \geq \dots \geq \sigma_r > 0 \quad \text{unique}$$

$$\text{s.t. for } 1 \leq i \leq r \quad \varphi(e_i) = \sigma_i f_i \\ \varphi^T(f_i) = \sigma_i e_i$$

for $j > r$

$$\varphi(e_j) = 0$$

$$\varphi^T(f_j) = 0$$

Follows from Spectral Thm

$$\underbrace{\varphi^T \varphi}_{\substack{\hat{i} = 1, \dots, r}}(e_i) = \varphi^T(\sigma_i f_i) = \sigma_i^2 e_i$$

$\varphi^T \varphi$ is pos. semidefinite symm.

$$\lambda_1 \geq \dots \geq \lambda_n \geq 0$$

$$\lambda_1 \geq \dots \geq \lambda_r > 0 = \lambda_{r+1} = \dots = \lambda_n.$$

so $\sigma_i = \sqrt{\lambda_i}$ for $i = 1, \dots, r$

(e_i) ON eigenbasis of $\varphi^T \varphi$ (DO) verify this.

(f_i) ON eigenbasis of $\varphi \varphi^T$

This has many applications in modern math —

"Netflix Problem"

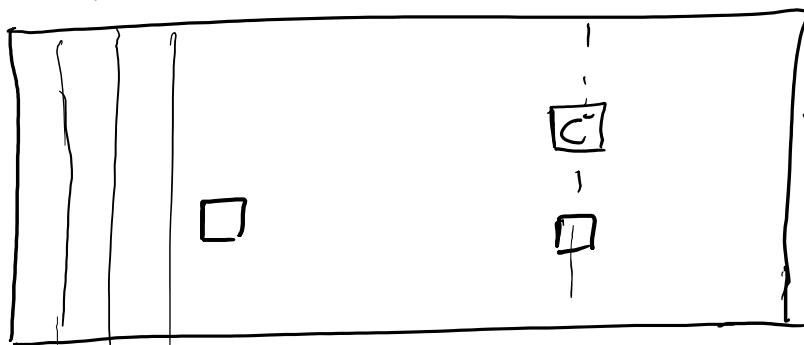
users

n

$$A = \underbrace{B}_{\text{rows}} \cdot \underbrace{C}_{\text{cols}}$$

$(n+k)r$

$n \cdot k$ dof



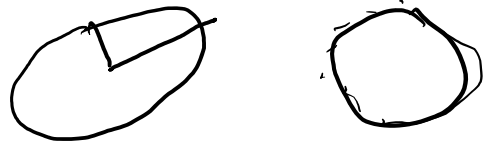
= moves
k

sparsely populated — Netflix wants to guess empty entries and see which movies you might like.

Suppose there existed an ideal matrix with everyone's true rating over every matrix.

How can we see this matrix with our imperfect observations?

Occam's razor: simplicity = truth
(e.g. Kepler's laws)



Note the "Netflix matrix" is low rank
(relative to # of entries)

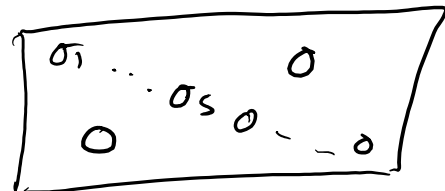
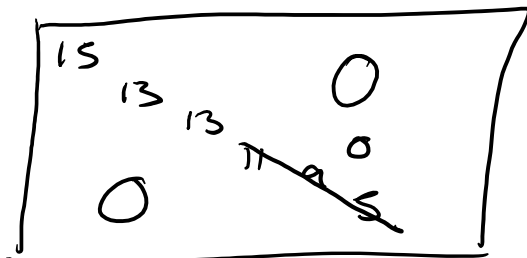
- low rank approximation

Human preferences

follow low rank approx.?

$$\text{rk}(A) = S$$

$$r < S$$



transform back -
closest rank r
approx.