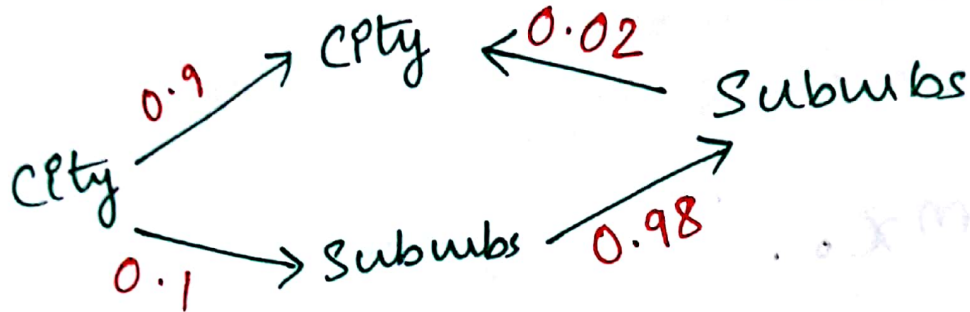


30th October

DIAGONALIZABILITY AND APPLICATIONS

- 1) Matrix limits
- 2) Matrix functions (Matrix polynomials / exponentials)

MARKOV CHAINS



	Currently living in the city	Currently living in the suburb
living in the city next year	0.9	0.02
living in the suburb next year	0.1	0.98

2x2

① Column sums should be 1

② No entry should be negative.

Transition State Matrix

$$P = \begin{pmatrix} 0.7 \\ 0.3 \end{pmatrix} \begin{array}{l} \leftarrow \text{Proportion of city dwellers} \\ \leftarrow \text{Proportion of suburban dwellers} \end{array}$$

$$AP = \begin{pmatrix} 0.9 & 0.02 \\ 0.1 & 0.98 \end{pmatrix} \begin{pmatrix} 0.7 \\ 0.3 \end{pmatrix}$$

$$= \begin{pmatrix} 0.636 \\ 0.364 \end{pmatrix}$$

$$A(AP) = A^2P = \begin{pmatrix} 0.5796 \\ 0.4203 \end{pmatrix}$$

$$\lim_{m \rightarrow \infty} A^m P$$

Is A diagonalizable?

$$Q = \begin{pmatrix} 1/6 & -1/6 \\ 5/6 & 1/6 \end{pmatrix}; D = \begin{pmatrix} 1 & 0 \\ 0 & 0.88 \end{pmatrix}$$

$$Q^{-1}AQ = D$$

$$\therefore L = \lim_{m \rightarrow \infty} A^m = \lim_{m \rightarrow \infty} Q D^m Q^{-1}$$

$$= Q \left[\lim_{m \rightarrow \infty} D^m \right] Q^{-1}$$

$$\text{See } \lim_{m \rightarrow \infty} D^m = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

$$L = \begin{bmatrix} 1/6 & 1/6 \\ 5/6 & 5/6 \end{bmatrix} \rightarrow \text{stationary distribution of stochastic process.}$$

$$\therefore \lim_{m \rightarrow \infty} A^m P = LP = \begin{pmatrix} 1/6 \\ 5/6 \end{pmatrix}$$

Eventually, $1/6^{\text{th}}$ of the population will stay in the city and rest in the suburbs.

Can you always have an existential limit?

Definition: Let $A \in M_{n \times n}(\mathbb{C})$. For, $1 \leq i, j \leq n$ define $\rho_i(A)$ to be the sum of absolute value of entries in the i^{th} row of A and define $\gamma_j(A)$ to be the sum of absolute value of entries in the j^{th} column.

$$\rho_i(A) = \sum_{j=1}^n |A_{ij}| \quad \text{for } i = 1, 2, \dots, n$$

$$\gamma_j(A) = \sum_{i=1}^n |A_{ij}| \quad \text{for } j = 1, 2, \dots, n.$$

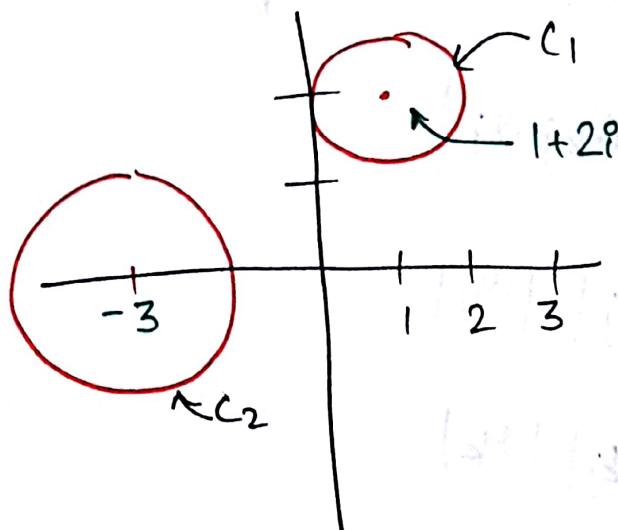
Gerschgorin Disk. C_i :

$$C_i = \{ z \in \mathbb{C} \mid |z - A_{ii}| < r_i \}$$

$$\text{where } r_i = \rho_i(A) - |A_{ii}|$$

Example

$$A = \begin{bmatrix} 1+2i & 1 \\ 2i & -3 \end{bmatrix}$$



Theorem Gerschgorin's disk theorem
Let $A \in M_{n \times n}(\mathbb{C})$. Then every
eigenvalue of A is contained in a
Gerschgorin's disk.

Proof: Let λ be an eigenvalue of A
and the corresponding eigenvector is

$$\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} = v$$

$$Av = \lambda v$$

$$\sum_{j=1}^n A_{ij} v_j = \lambda v_i \quad i=1, 2, \dots, n$$

Let v_k be the coordinate of v having
the largest absolute value.

We will show $\lambda \in C_K$.

$$\begin{aligned} & |\lambda v_k - A v_k| \\ &= \left| \sum_{j=1}^n A_{kj} v_j - A_{kk} v_k \right| \\ &= \left| \sum_{j \neq k} A_{kj} v_j \right| \\ &\leq \sum_{j \neq k} |A_{kj}| |v_j| \\ &\leq \sum_{j \neq k} |A_{kj}| |v_k| \\ &= |v_k| \sum_{j \neq k} |A_{kj}| \\ &= |v_k| r_k. \end{aligned}$$

Thus,

$$|v_k| |\lambda - A_{kk}| \leq |v_k| r_k$$
$$\Rightarrow |\lambda - A_{kk}| \leq r_k.$$

Corollary: let λ be an eigenvalue of $A \in M_{n \times n}(\mathbb{C})$. Then $|\lambda| < \rho(A)$

Corollary: let λ be an eigenvalue of $A \in M_{n \times n}(\mathbb{C})$. Then $|\lambda| \leq \min(\rho(A), \gamma(A))$

Invariant Subspaces and Cayley Hamilton Theorem

DEFINITION: Let T be a linear operator on a vector space V . A subspace W of V is called a T invariant subspace of V if $T(W) \subseteq W$, that is, if $v \in W$, $T(v) \in W$.

Examples

V be a vector space. $T: V \rightarrow V$.

1. $\{0\}$
2. V
3. $R(T)$
4. $N(T)$
5. E_λ for any eigenvalue of T .

Examples:

$$T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

$$T \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} a+b \\ b+c \\ 0 \end{pmatrix}$$

$$W = \left\{ \begin{pmatrix} t \\ s \\ 0 \end{pmatrix} : t, s \in \mathbb{R} \right\}$$

$\hookrightarrow T$ invariant subspace.

Let T be a linear operator on $V = \mathbb{R}^2$
a vector space V . Let x be a non zero
vector

$$x \neq 0, x \in V.$$

$$W = \text{span} \{x, T(x), T^2(x), \dots\}$$

$$w \in W$$

$$w = \sum_{i=0}^k a_i T^i(x)$$

T -cyclic subspace of V generated by x .

Ex W : cyclic subspace of V generated by x
 W is the smallest T -invariant subspace
containing x

Ex

$$T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

$$T \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} -b+c \\ a+c \\ 3c \end{pmatrix}$$

$$e_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \in \mathbb{R}^3$$

~~$W = \text{span}\{e_1\}$~~

$$W = \text{span} \{e_1, T(e_1), T^2(e_1), \dots\}$$

$$T(e_1) = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$

$$T(T(e_1)) = \begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix} = -e_1$$

$$W = \text{span} \{e_1, e_2\}$$

Theorem: Let T be a linear operator on a finite dimensional vector space V . Let W be a T invariant subspace of V . Then the characteristic polynomial of T_W divides the characteristic polynomial of T .

Q) What is T_W ?

$$T: V \rightarrow V$$

W : T -invariant subspace of V .

$$T(W) \subseteq W.$$

$T(W)$: restriction of T on W .

$$T_W: W \rightarrow W.$$

Proof: Let $\{\gamma_1, \dots, \gamma_k\}$ be an ordered basis of W .

$\gamma_2 = \{\gamma_1, \dots, \gamma_k, \gamma_{k+1}, \dots, \gamma_n\}$ extended ordered basis of V .

$$\text{let } B_1 = [T_W]_{\gamma_1}$$

$$A = [T]_{\gamma_2}$$

So, it is clear, that

$$A = \begin{bmatrix} B_1 & B_2 \\ 0 & B_3 \end{bmatrix}$$

$$\det(A - tI) = \det \left[\begin{bmatrix} B_1 & B_2 \\ 0 & B_3 \end{bmatrix} - t \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix} \right]$$

sample

Characteristic
polynomial of
 T_w .

Theorem

② $\{u, T(u), T^2(u), \dots, T^{k-1}(u)\}$

(b) ^{IF} $a_0 + a_1(T(\varphi)) + \dots + a_{k-1}T^{k-1}(\varphi) = 0$

then the characteristic polynomial of

Proof: Let $v \neq 0$. $\{v\}$ is L.I.

$$B = \{u, \tau(u), \dots, \tau^{j-1}(u)\}$$

a linearly independent set where j is the largest positive integer.

Let $Z = \text{span}(B)$

Also, $T^j(v) \in \mathbb{Z}$.

$$b) W = \text{span} \{v, T(v), \dots, T^{(k-1)}(v)\}$$

$$T^k(v) = -a_0 v + a_1 T(v) + \dots - a_{k-1} T^{k-1}(v).$$

$$[Tw]_B = \begin{bmatrix} 0 & 0 & \dots & 0 & -a_0 \\ 1 & 0 & \dots & 0 & -a_1 \\ 0 & 1 & \dots & 0 & -a_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & -a_{k-1} \end{bmatrix}$$

$$\det([Tw]_B - tI_k) = (-1)^k (a_0 + a_1 t + \dots + a_{k-1} t^{k-1} + t^k)$$

controllable canonical form.

The Cayley Hamilton Theorem

Let T be a linear operator on a finite dimensional vector space V . Let $f(t)$ be the characteristic polynomial of T . Then $f(T) = 0$, the zero transformation. That is, T "satisfies" its characteristic polynomial.

Proof: We will show that $f(T)(v) = 0 \quad \forall v \in V$.

Let $v \neq 0$,

W is a T -cyclic subspace generated by v .

Let $(\text{dimension of } W) \dim(W) = k$.

\exists scalars a_0, a_1, \dots, a_{k-1} such that $a_0 v + a_1 T(v) + \dots + a_{k-1} T^{k-1}(v) + T^k(v) = 0$.

This means that

$$g(t) = (-1)^k (a_0 + a_1 t + a_2 t^2 + \dots + a_{k-1} t^{k-1} + t^k)$$

is the char. Polynomial of T_W .

$$g(T) = (-1)^k (a_0 I + a_1 T + \dots + a_{k-1} T^{k-1} + T^k)(v)$$

$$= 0.$$

Since $g(t) \mid f(t)$ → characteristic polynomial of T .
 $f(T)(v) = 0 \quad \forall v \in V$

$\Rightarrow f(T)$ is a zero linear operator

$\Rightarrow T$ satisfies its characteristic polynomial

Corollary: Let A be an $n \times n$ matrix with $f(T)$ as its characteristic polynomial then $f(A) = 0$

Ex 1

$$A = \begin{bmatrix} 1 & 1 \\ 0 & -1 \end{bmatrix}$$

$$f(t) = t^2 - 1$$

$$\text{C.H.} \Rightarrow f(A) = 0$$

$$A^2 - I = 0$$

$$\Rightarrow A^2 = I$$

$$\Rightarrow \underline{\underline{A = A^{-1}}}$$

Ex

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

$$f(t) = (t-1)^2$$

$$f(t) = t^2 - 2t + 1$$

$$\Rightarrow f(A) = 0$$

$$\Rightarrow A^2 - 2A + I = 0$$

$$\Rightarrow \boxed{A^2 = 2A - I} \quad (*)$$

$$A^3 = 2A^2 - A = 2(2A - I) - A = 3A - 2I$$

$$\boxed{A^{-1} = 2I - A}$$

(if A^{-1} exists)