Week 12 Thursday Recull.  $\vec{X}_{k+1} = A \vec{X}_k$ If A = M2x2 11R) and has two different eigen value  $\lambda_1 \neq \lambda_2$  and eigenvector  $\vec{v}_i$  and  $\vec{v}_z$ then. for any vector  $\vec{x}_0 = x, \vec{v}_1 + x_2 \vec{v}_2$  $A^{k}\vec{\lambda}_{0} = \chi_{1} \cdot \lambda_{1}^{k} \vec{u}_{1} + \chi_{2} \cdot \lambda_{2}^{k} \vec{u}_{2}$ What if A commot be digonalized? eg.  $A = \begin{pmatrix} 0.8 & 1 \\ 0 & 0.8 \end{pmatrix}$   $\det(A - \lambda \overline{1}) = \begin{pmatrix} 0.8 - \lambda & 1 \\ 0 & 0.8 - \lambda \end{pmatrix} = (\lambda - 0.8)$  $\Rightarrow \lambda_1 = \lambda_2 = 0.8 \qquad (A - 0.8I) \cdot \vec{x} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0$  $\Rightarrow \vec{v} = \begin{pmatrix} x_1 \\ y \end{pmatrix} = x_1 \cdot \begin{pmatrix} y \\ y \end{pmatrix}$ so if  $\vec{x}_0 = \vec{e}$ , then  $\vec{x}_k \rightarrow \vec{o}$ .  $A^{k} \cdot \binom{\prime}{o} = \lambda^{k} \cdot \binom{\prime}{o}$ A ei= 2 ei A ez = \ dz + e,

 $A \vec{e}_1 = \lambda \vec{e}_1$   $A \vec{e}_2 = \lambda \vec{e}_2 + \vec{e}_1$   $A^2 \vec{e}_3 = \lambda \cdot A \vec{e}_2 + A \vec{e}_1 = \lambda \cdot (\lambda \vec{e}_2 + \vec{e}_1) + \lambda \vec{e}_1$   $= \lambda^2 \vec{e}_2 + 2\lambda \vec{e}_1$   $A^3 \vec{e}_2 = \lambda^2 \cdot A \vec{e}_2 + 2\lambda A \vec{e}_1 = \lambda^2 \cdot (\lambda \vec{e}_2 + \vec{e}_1) + 2\lambda \cdot \lambda \vec{e}_1$   $= \lambda^3 \vec{e}_1 + 3\lambda^2 \vec{e}_1$ 

$$\lambda^{3} (\lambda \vec{e}_{2} + \vec{e}_{1}) + 3\lambda^{2} \lambda e_{1}$$

$$= \lambda^{4} \vec{e}_{2} + 4\lambda^{3} e_{1}$$

$$A^{k}\vec{e}_{2} = \lambda^{k}\vec{e}_{2} + k \cdot \lambda^{k-1}\vec{e}_{i}$$

$$\lambda^{=0.8} \lim_{k \to \infty} \lambda^{k} = 0 \quad \lim_{k \to \infty} \lambda^{k-1} = 0$$

2. Complex eigenvalue.

$$A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \qquad \det (A - \lambda I) = \begin{pmatrix} -\lambda & -1 \\ 1 & -\lambda \end{pmatrix} = \lambda^2 + 1$$

$$\lambda_{1} = i$$

$$\begin{pmatrix} -i & -1 \\ -i & -i \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} = 0$$

$$\begin{pmatrix} i & -1 \\ -i & i \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} = 0$$

$$\Rightarrow \vec{X} = \begin{pmatrix} i & x_{2} \\ x_{2} \end{pmatrix} = x_{2} \cdot \begin{pmatrix} i \\ i \end{pmatrix}$$

$$\Rightarrow \vec{X} = x_{2} \begin{pmatrix} -i \\ x_{2} \end{pmatrix}$$

$$P = \begin{pmatrix} i & -i \\ i & i \end{pmatrix} \qquad \text{then} \qquad P' \cdot A \cdot P = \begin{pmatrix} i \\ -i \end{pmatrix}$$

$$Way 1: 1 \qquad -1 \qquad -\frac{1}{2} = -\frac{1}{2}$$

$$\begin{pmatrix} \lambda & -\lambda & | \lambda & 0 \\ \lambda & \lambda & | \lambda & 0 \\ \lambda & \lambda & \lambda & 1 \end{pmatrix}$$

$$P^{-1} = \begin{pmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

Way 2: Cramer's 
$$P'=\frac{1}{2i}\begin{pmatrix}1&i\\-1&i\end{pmatrix}$$

$$=\frac{1}{2}\cdot\begin{pmatrix}-i&1\\i&1\end{pmatrix}$$

$$P^{-1}AP = \frac{1}{2}\begin{pmatrix} -i & 1 \\ i & l \end{pmatrix} \cdot \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} i & -i \\ l & l \end{pmatrix}$$

$$=\frac{1}{2}\cdot\begin{pmatrix}1&i\\1&-i\end{pmatrix}\begin{pmatrix}i&-i\\i&1\end{pmatrix}=\frac{1}{2}\cdot\begin{pmatrix}2i&0\\0&-2i\end{pmatrix}$$

$$= \begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix} : \neq )$$

$$A_{loo} = ?$$

$$A = P - D \cdot P'$$
 then  $A'' = (PDP') \cdot (PDP') \cdot \cdots \cdot (PDP')$ 

$$= P \cdot D^{100} \cdot P^{-1}$$

$$= P \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \cdot P^{-1} = I.$$

eg. 
$$A = \begin{pmatrix} -1 & -4 \\ 2 & 3 \end{pmatrix}$$

1). 
$$\det(A-\lambda I) = \begin{vmatrix} -1-\lambda & -4 \\ 2 & 3-\lambda \end{vmatrix} = (\lambda-3)(\lambda+1) + 8$$
  
=  $\lambda^2 - 2\lambda + 5$ 

$$\Delta = 4 - 20 = -16$$

$$\lambda = \frac{2 \pm \lambda \Delta}{2} = 1 \pm 2i$$

$$\overline{\lambda} = 4 \cdot i \qquad \sum_{X = \frac{1}{2} \pm \frac{1}{2}}^{\infty} \frac{2}{2}$$

Rmk: if A is Max(R) and  $\lambda$  is a couplex eigenful at the eigenvector  $\overrightarrow{v}$ , i.e.  $\overrightarrow{A} \overrightarrow{v} = \lambda \overrightarrow{v}$ , then.

$$\overline{A \cdot \overrightarrow{v}} = A \cdot \overrightarrow{V} = \overline{\lambda} \overrightarrow{v} = \frac{\overline{\lambda}}{2} \cdot \frac{\overrightarrow{v}}{2}$$

so  $\overline{\lambda}$  is also an eigenvalue of A with eigenvector  $\overrightarrow{v}$ .

$$\lambda_1 = 1 + 2i \qquad \begin{pmatrix} -1 - (1/\tau)i \\ -2 - 2i \\ 2 & 3 - (1/\tau)i \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \ge 0$$

$$(1 - i)(2 + 2i) = 2 \cdot (1 - i^2) = 4 \qquad = x_2 \cdot \begin{pmatrix} i - 1 \\ 1 \end{pmatrix}$$

$$\lambda_2 = 1 - 2i \qquad \text{by the Ymk.} \qquad \overrightarrow{x} = x_2 \cdot \begin{pmatrix} -i - 1 \\ 1 \end{pmatrix} \qquad \frac{1}{4 + 6i} = \alpha - 6i$$

$$\lambda = PDP^T \qquad P = \begin{pmatrix} 1 - 1 & -i - 1 \\ 1 & 1 \end{pmatrix} \qquad \det(P) = \begin{pmatrix} 1 - i - 1 \\ 1 & 1 \end{pmatrix}$$

$$(D = P^T A P)$$

 $A^{100} = P \cdot D^{100} \cdot P^{-1} = \frac{1}{2i} \binom{i-1}{i} - \frac{i-1}{i} \binom{(1+2i)^{100}}{0} \binom{(1-2i)^{100}}{(1-2i)^{100}} \binom{1}{i+1}$ 

$$\begin{aligned}
| 1+2i = \beta \cdot e^{i\theta} \\
| (1+2i)^{(150)} = \beta^{(150)} \cdot e^{(150)} \\
&= \frac{1}{2i} \cdot \left( \frac{(i-1) \cdot (1+2i)^{(150)} \cdot (-i-1) \cdot (1-2i)}{(1+2i)^{(150)} \cdot (1-2i)^{(150)}} \right) \begin{pmatrix} -1 & i+1 \\ 1 & i-1 \end{pmatrix} \\
&= \begin{pmatrix} (1+2i)^{(150)} \cdot (1-i) \cdot & -2 \cdot (1+2i)^{(150)} + 2 \cdot (1-2i)^{(150)} \\ -(1+2i)^{(150)} \cdot (1+i) / 2i & (1-2i)^{(150)} \cdot (i+1) + / \\ -(1+2i)^{(150)} + (1-2i)^{(150)} \cdot (i-1) / 2i \end{pmatrix}
\end{aligned}$$

3. Differential Egnetion.

$$\vec{X}_{k+1} = A \cdot \vec{X}_{k} \qquad \vec{X}_{k+1} - \vec{X}_{k} = (A - I) \cdot \vec{X}_{k}$$

$$\vec{X}(t) = \begin{pmatrix} X_{i}(t) \\ \vdots \\ X_{i}(t) : R \rightarrow IR.$$

$$\vec{x}'(t) = A \cdot \vec{x}(t)$$

If 
$$\vec{x}(t) = (x_1(t))$$
 then  $x_1'(t) = a_1 x_1(t)$ 

$$\frac{dx_1(t)}{dt} = a_1 \cdot x_1(t)$$

$$\int \frac{dx_1(t)}{x_1(t)} = \int a_1 \cdot dt \qquad \text{in} |x_1(t)| = a_1 t + C$$

$$x_1(t) = C \cdot e^{a_1 t}$$

It. A is diagonal matrix.

$$\begin{pmatrix} x_i'(t) \\ x_2'(t) \\ \vdots \\ x_n'(t) \end{pmatrix} = \begin{pmatrix} a_i \\ a_2 \\ \vdots \\ a_n \end{pmatrix} \begin{pmatrix} x_i(t) \\ \vdots \\ x_n(t) \end{pmatrix}$$

$$= \begin{pmatrix} a_i \\ a_2 \\ \vdots \\ x_n(t) \end{pmatrix}$$

$$= \begin{pmatrix} x_i(t) \\ \vdots \\ x_n(t) \\ \vdots \\ x_n(t) \end{pmatrix}$$

If A is diagonalizable. say 
$$A = PDP^{-1}$$
 then.

take linear transformation on variables y(t)= P. x(t)

then 
$$\vec{y}(t) = \vec{D} \cdot \vec{y}(t)$$
 which is diagonal.

eg. 
$$A = \begin{pmatrix} 2 & 3 \\ -1 & -2 \end{pmatrix}$$
  $\vec{x}(t) = A \cdot \vec{x}(t)$ 

$$det(A - \lambda I) = \begin{vmatrix} 2-\lambda & 3 \\ -1 & -2-\lambda \end{vmatrix} = \lambda^2 - 4 + 3 = \lambda^2 - 1 = 0$$

$$\lambda = 1 \qquad \left(\begin{array}{cc} 1 & 3 \\ -1 & -3 \end{array}\right) \left(\begin{array}{c} \chi_1 \\ \chi_2 \end{array}\right) = 0 \quad (=) \quad \vec{X} = \begin{pmatrix} -3\chi_2 \\ \chi_2 \end{pmatrix} = \chi_2 \cdot \begin{pmatrix} -3 \\ 1 \end{pmatrix}$$

$$\lambda = -1 \qquad \left(\begin{array}{ccc} 3 & 3 \\ -1 & -1 \end{array}\right) \begin{pmatrix} \chi_1 \\ \chi_{\sim} \end{pmatrix} \Rightarrow \lambda = \lambda_{\sim} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

so 
$$P = \begin{pmatrix} -3 & -1 \\ 1 & 1 \end{pmatrix}$$
  $P^{-1} = \frac{1}{-2} \begin{pmatrix} 1 & 1 \\ -1 & -3 \end{pmatrix}$ 

$$\vec{y}(t) = \begin{pmatrix} -\frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{3}{2} \end{pmatrix} \vec{x}(t) \quad i.e. \quad y_{i}(t) = \frac{-1}{2} x_{i}(t) - \frac{1}{2} x_{i}(t) \\ y_{i}(t) = \frac{1}{2} x_{i}(t) + \frac{3}{2} x_{i}(t)$$

then we get 
$$\vec{y}(t) = \begin{pmatrix} 1 \\ -1 \end{pmatrix} \vec{y}(t)$$