

LECTURE 4 (13)

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MATHEMATICS

Plan:

- ① Eigenvalues and eigenvectors
- ② Diagonalization

Reading:

[ME] 23.1-23.4, 23.6-23.7,
23.9

① Eigenvalues and eigenvectors

A : $n \times n$ -matrix

Definition

A number λ is called an eigenvalue for A if the linear system

$$A \cdot \underline{v} = \lambda \cdot \underline{v}$$

has non-trivial solutions $\underline{v} \neq \underline{0}$. In that case, all solutions \underline{v} of this linear system are called eigenvectors for A with eigenvalue λ .

Ex:

$$A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}: \quad A \cdot \underline{v} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \cdot \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2x + y \\ x + 2y \end{pmatrix}$$

$$\lambda \cdot \underline{v} = \lambda \cdot \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \lambda x \\ \lambda y \end{pmatrix}$$

$$\begin{pmatrix} 2x+y \\ x+2y \end{pmatrix} = \begin{pmatrix} \lambda x \\ \lambda y \end{pmatrix}$$

$$\begin{aligned} 2x+y &= \lambda x \\ x+2y &= \lambda y \end{aligned}$$

$$\begin{aligned} (2-\lambda)x + y &= 0 \\ x + (2-\lambda)y &= 0 \end{aligned}$$

$$\begin{aligned} 2x - \lambda x + y &= 0 \\ x + 2y - \lambda y &= 0 \end{aligned}$$

Nontrivial solutions
 $(x, y) \neq (0, 0)$

$$\Leftrightarrow \begin{vmatrix} 2-\lambda & 1 \\ 1 & 2-\lambda \end{vmatrix} = 0$$

Conclusion:

The eigenvalues of

$$A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \text{ are}$$

$$\lambda_1 = 3, \lambda_2 = 1$$

$$(2-\lambda) \cdot (2-\lambda) - 1^2 = 0$$

$$\lambda^2 - 4\lambda + (4-1) = 0$$

$$\lambda = \frac{4 \pm \sqrt{4^2 - 4 \cdot 1 \cdot 3}}{2}$$

$$= \frac{4 \pm 2}{2}$$

$$\lambda = 3, \lambda = 1$$

Eigen vectors:

$$\lambda_1 = 3: \begin{aligned} -x + y &= 0 \\ \cancel{x - y} &= 0 \end{aligned}$$

$$y = x, x \text{ free}$$

$$\underline{v} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ x \end{pmatrix} = x \cdot \underline{\underline{\begin{pmatrix} 1 \\ 1 \end{pmatrix}}}$$

$$\lambda_2 = 1: \begin{aligned} x + y &= 0 \\ \cancel{x + y} &= 0 \end{aligned}$$

$$y = -x, x \text{ free}$$

$$\underline{v} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ -x \end{pmatrix} = x \cdot \underline{\underline{\begin{pmatrix} 1 \\ -1 \end{pmatrix}}}$$

Conclusion:

Eigenvectors with $\lambda = 3$: Multiples of $\underline{v}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

Eigenvectors with $\lambda = 1$: Multiples of $\underline{v}_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$

If you choose y free
 $x = -y, y$ free

$$\underline{v} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -y \\ y \end{pmatrix} = -y \cdot \underline{\underline{\begin{pmatrix} -1 \\ 1 \end{pmatrix}}}$$

General method: A $n \times n$ -matrix

$$A \cdot \underline{v} = \lambda \cdot \underline{v}$$

$$A \cdot \underline{v} - \lambda \cdot \underline{v} = \underline{0}$$

$$A \cdot \underline{v} - \lambda I \cdot \underline{v} = \underline{0}$$

$$(A - \lambda I) \cdot \underline{v} = \underline{0}$$

$$A - \lambda I = \begin{pmatrix} a_{11} & a_{12} & \dots \\ a_{21} & a_{22} & \dots \\ \vdots & \vdots & \ddots \end{pmatrix} - \begin{pmatrix} \lambda & 0 & \dots \\ 0 & \lambda & \dots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

Characteristic equation:

$$|A - \lambda I| = 0$$

$$= \begin{pmatrix} a_{11} - \lambda & a_{12} & \dots \\ a_{21} & a_{22} - \lambda & \dots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

(a) Eigenvalues of A = Solutions of the characteristic equation

(b) Eigenvectors of A with given eigenvalue λ = Solutions of the linear system $(A - \lambda I) \cdot \underline{v} = \underline{0}$

Comment:

Part a): Solve polynomial equation of degree n
→ difficult

Part b): Solve linear system
→ easy (Gaussian elimination)

If there are k free variables s_1, \dots, s_k , then the solutions (= eigenvectors) can be written

$$\underline{v} = s_1 \underline{v}_1 + s_2 \underline{v}_2 + \dots + s_k \underline{v}_k$$

and $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$ are linearly independent.

Ex 1:

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

$$\begin{vmatrix} 0-\lambda & 1 \\ -1 & 0-\lambda \end{vmatrix} = 0$$

$$\lambda^2 + 1 = 0$$

$$\lambda^2 = -1 \quad \text{no solutions} \\ \text{(among real numbers)}$$

∴

No (real) eigenvalues

Ex 2:

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

$$\begin{vmatrix} 1-\lambda & 1 \\ 0 & 1-\lambda \end{vmatrix} = 0$$

$$(1-\lambda)^2 - 0 = 0$$

$$\rightarrow (1-\lambda)^2 = 0$$

only one (double) root

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

$$\lambda_1 = 1$$

$$\lambda_2 = 1$$

(this is called multiplicity two)

Ex:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$|A - \lambda I| = 0 \rightarrow \begin{vmatrix} a-\lambda & b \\ c & d-\lambda \end{vmatrix} = 0$$

$$\text{trace} \rightarrow \text{tr}(A) = a+d \\ \det(A) = ad-bc$$

$$(a-\lambda)(d-\lambda) - bc = 0$$

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

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

$$\lambda^2 - \text{tr}(A) \cdot \lambda + \det(A) = 0$$

Ex 3:

$$A = \begin{pmatrix} 3 & 1 \\ -1 & 7 \end{pmatrix}$$

$$\lambda^2 - 10\lambda + 22 = 0$$

$$\lambda = \frac{10 \pm \sqrt{10^2 - 4 \cdot 22}}{2}$$

$$= \frac{10 \pm \sqrt{12}}{2} = 5 \pm \sqrt{3}$$

$$\lambda_1 = 5 + \sqrt{3}$$

$$\lambda_2 = 5 - \sqrt{3}$$

Characteristic equation:

$$|A - \lambda I| = \begin{vmatrix} a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} - \lambda \end{vmatrix} = 0$$

$$= (-\lambda)^n + \dots \text{ (terms of lower deg.)} = 0$$

Fact: $|A - \lambda I| = (\lambda - \lambda_1)(\lambda - \lambda_2) \dots (\lambda - \lambda_r) \cdot Q(\lambda) = 0$
where $Q(\lambda) = 0$ has no solutions

Possibilities:

- i) $Q(\lambda)$ is a non-zero constant ($r = n$)
 \Rightarrow there are n solutions
- ii) $Q(\lambda)$ is not a constant ($r < n$)
 \Rightarrow there are $r < n$ solutions

Fact: 1) When A is symmetric then $r = n$.

$$\left. \begin{array}{l} 2) \det(A) = \lambda_1 \dots \lambda_n \\ \operatorname{tr}(A) = \lambda_1 + \dots + \lambda_n \end{array} \right\} \text{ when there are } n \text{ solutions}$$

Ex of possibility i) and ii):

i) $|A - \lambda I| = -(\lambda - 2)(\lambda - 4)(\lambda - 10) \Rightarrow$ Eigenvalues 2, 4, 10

ii) $|A - \lambda I| = -(\lambda - 1) \cdot (\lambda^2 + 4)$
(no solution at $\lambda^2 + 4 = 0$)

Ex: $A = \begin{pmatrix} 7 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 0 & 7 \end{pmatrix}$ $n=3$

$$\begin{vmatrix} \lambda-7 & 0 & 3 \\ 0 & 2-\lambda & 0 \\ 3 & 0 & 7-\lambda \end{vmatrix} = 0$$

$$+ (2-\lambda) \cdot \begin{vmatrix} 7-\lambda & 3 \\ 3 & 7-\lambda \end{vmatrix} = 0$$

$$(2-\lambda) \cdot (\lambda^2 - 14\lambda + 40) = 0$$

$$\lambda = 2 \text{ or } \lambda^2 - 14\lambda + 40 = 0$$

$$\lambda = \frac{14 \pm \sqrt{14^2 - 4 \cdot 40}}{2} = \frac{14 \pm 6}{2} = 7 \pm 3$$

$$\underline{\underline{\lambda_1 = 2}}$$

$$\underline{\underline{\lambda_2 = 10}}$$

$$\underline{\underline{\lambda_3 = 4}}$$

Factorization: $-(\lambda-2) \cdot (\lambda-10)(\lambda-4) = 0$

Eigenvectors:

If λ is an eigenvalue of A of multiplicity m , then the linear system

$$(A - \lambda I) \cdot \underline{v} = \underline{0}$$

has at most m degrees of freedom and at least one degree of freedom.

If you multiply out, you get $-\lambda^3 + 16\lambda^2 - 68\lambda + 80 = 0$. This is difficult to solve! Try to keep $\det(A - \lambda I)$ factorized if possible!

Ex:

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

$$\lambda^2 - 6\lambda + 9 = 0$$

$$(\lambda - 3)^2 = 0$$

$$\lambda_1 = \underline{3}, \lambda_2 = \underline{3}$$

Eigenvectors:

$$\begin{pmatrix} 4-3 & -1 \\ 1 & 2-3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$x - y = 0$$

$$\cancel{x - y} = 0$$

$x = y,$
 y free

degrees of freedom
is at least 1 and at
most 2 (since $\lambda = 3$
has mult. 2).

Computations show
that # degrees of freedom

(= 1) in this case

$$\begin{cases} \underline{v} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ y \end{pmatrix} = y \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} \\ = y \cdot \underline{v}_1, \underline{v}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \\ \text{Eigenvectors } (\lambda = 3) \\ = \text{all multiples of } \underline{v}_1. \end{cases}$$

② Diagonalization

Defn: An $n \times n$ -matrix A is diagonalizable if there is a diagonal matrix D and an invertible matrix P such that

$$P^{-1}AP = D$$

Fact: If A has n eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ (counted with multiplicity) and if there are n linearly independent eigenvectors for A , $\underline{v}_1, \underline{v}_2, \dots, \underline{v}_n$, then A is diagonalizable and we may take

Order of eigenvalues correspond to order of eigenvectors

$$A \cdot \underline{v}_i = \lambda_i \cdot \underline{v}_i$$

$$P = \left(\begin{array}{c|c|c} \underline{v}_1 & \underline{v}_2 & \dots & \underline{v}_n \end{array} \right) \quad D = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{pmatrix}$$

Ex: $A = \begin{pmatrix} 7 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 0 & 7 \end{pmatrix}$

Eigenvalues: $\lambda_1 = 2 \quad \lambda_2 = 4 \quad \lambda_3 = 6$
(we found these before)

Concl: $D = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 6 \end{pmatrix}$

A is diagonalizable

(find $\underline{v}_2, \underline{v}_3$ in similar way)

$$P = \left(\begin{array}{c|c|c} \underline{v}_1 & \underline{v}_2 & \underline{v}_3 \end{array} \right)$$

$$\underline{v}_1 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \underline{v}_2 \quad \underline{v}_3$$

Eigenvectors:

$$\lambda = 2: \begin{pmatrix} 5 & 0 & 3 \\ 0 & 0 & 0 \\ 3 & 0 & 5 \end{pmatrix} \underline{x} = \underline{0}$$

$$\begin{pmatrix} 5 & 0 & 3 \\ 0 & 0 & 16/5 \\ 0 & 0 & 0 \end{pmatrix}$$

one free var: x_2

$$\underline{v} = \begin{pmatrix} 0 \\ x_2 \\ 0 \end{pmatrix} = x_2 \cdot \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \begin{array}{l} x_1 = 0 \\ x_2 \text{ free} \\ x_3 = 0 \end{array}$$

How to check if A is diagonalizable:

A: $n \times n$ - matrix

i) Compute eigenvalues of A: $\lambda_1, \lambda_2, \dots, \lambda_r$

Use multiplicity (ie. if $\lambda=1$ has multiplicity 2, then $\lambda_1=1, \lambda_2=1$ is included twice)

If $r=n$ (there are n eigenvalues), then we take

$$D = \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{pmatrix}$$

and check ii), if $r < n$ then A is not diagonalizable.

ii) Compute eigen vectors for A:

If λ_i has multiplicity m_i , then there are two possibilities:

i) $(A - \lambda_i I)\underline{v} = \underline{0}$ has m_i degrees of freedom

Solutions: $\underline{v} = s_1 \underline{v}_1 + s_2 \underline{v}_2 + \dots + s_{m_i} \underline{v}_{m_i}$

where s_1, \dots, s_{m_i} are free var's,

$\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_{m_i}\}$ eigenvectors
(automatically lin. independent)

\Rightarrow A diagonalizable (if this is the case for all eigenvalues)

$$P = (\underline{v}_1 | \underline{v}_2 | \dots | \underline{v}_n)$$

ii) $(A - \lambda_i I)\underline{v} = \underline{0}$ has less than m_i degrees of freedom:

A not diag.

(not enough eigenvectors)

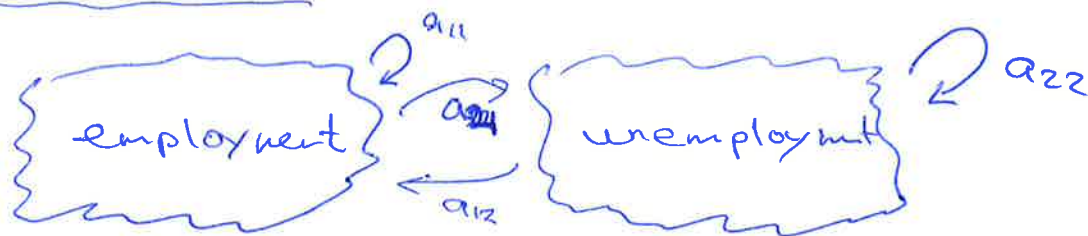
Two eigenvectors for A corresponding to different eigenvalues are always linearly independent.

Facts:

- 1) If A is symmetric, then it is diagonalizable.
- 2) If A has n different (i.e. all multiplicity 1) eigenvalues, then it is diagonalizable.

A diagonalizable \iff $\begin{cases} A \text{ has } n \text{ eigenvalues} \\ + \\ A \text{ has } n \text{ linearly} \\ \text{independent eigenvectors} \end{cases}$

Application: Markov chains



state vector: $\begin{pmatrix} e \\ u \end{pmatrix}$

$e =$ share of employed
 $u =$ ———— unemployed

transition matrix:

$$\begin{pmatrix} e_{t+1} \\ u_{t+1} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} e_t \\ u_t \end{pmatrix}$$

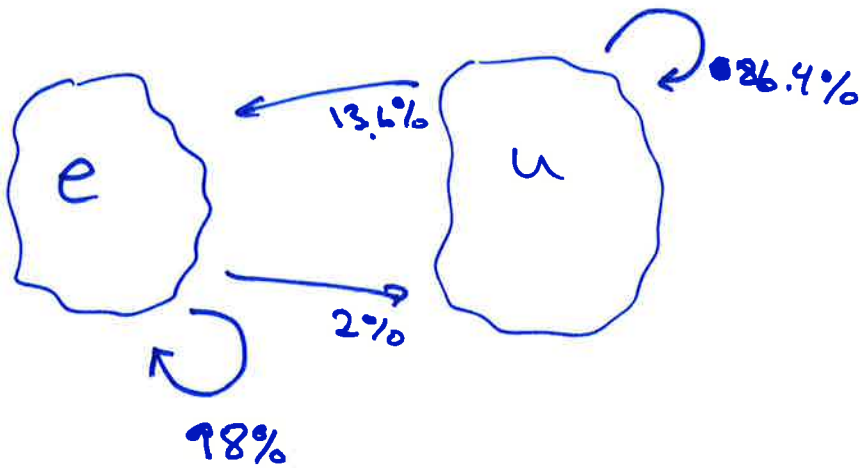
$$e_{t+1} = a_{11} e_t + a_{12} u_t$$

$$u_{t+1} = a_{21} e_t + a_{22} u_t$$

Ex: $A = \begin{pmatrix} 0.98 & 0.136 \\ 0.02 & 0.864 \end{pmatrix}$

$\underline{v}_0 = \begin{pmatrix} 0.9 \\ 0.1 \end{pmatrix}$

initial state
= 10% unemploy.



Long term:

$\underline{v}_1 = A \cdot \underline{v}_0$ (after one week)

$\underline{v}_2 = A \cdot \underline{v}_1 = A \cdot (A \underline{v}_0) = A^2 \cdot \underline{v}_0$ (after two weeks)

⋮

$\underline{v}_n = A^n \cdot \underline{v}_0$ (after n weeks)

What happens when $n \rightarrow \infty$ (long term)?

Eigenvalues and eigenvectors:

$$\begin{vmatrix} 0.98 - \lambda & 0.136 \\ 0.02 & 0.864 - \lambda \end{vmatrix} = 0$$

$$\lambda^2 - 0.844\lambda + 0.844 = 0$$

$$\lambda = \frac{0.844 \pm \sqrt{0.844^2 - 4 \cdot 0.844}}{2}$$

$\underline{\lambda}_1 = 1$

$\underline{\lambda}_2 = 0.844$

$\underline{\lambda} = 1: \begin{pmatrix} -0.02 & 0.136 \\ 0.02 & -0.136 \end{pmatrix}$

$-0.02x + 0.136y = 0$

$x = \frac{0.136}{0.02} y = 6.8y$
(y free)

$\underline{v} = \begin{pmatrix} 6.8y \\ y \end{pmatrix} = y \cdot \begin{pmatrix} 6.8 \\ 1 \end{pmatrix}$

$\Rightarrow \underline{v}_1 = \begin{pmatrix} 6.8 \\ 1 \end{pmatrix}$

$\underline{\lambda} = 0.844:$

$\begin{pmatrix} 0.136 & 0.136 \\ 0.02 & 0.02 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

$y = -x, x \text{ free}$

$\underline{v} = \begin{pmatrix} x \\ -x \end{pmatrix} = x \cdot \begin{pmatrix} 1 \\ -1 \end{pmatrix} \quad \underline{v}_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$

Conclusion:

A has eigen values $\lambda_1=1, \lambda_2=0.844$

and eigenvectors $\underline{v}_1 = \begin{pmatrix} 6.8 \\ 1 \end{pmatrix}$ $\underline{v}_2 = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$

$$D = \begin{pmatrix} 1 & 0 \\ 0 & 0.844 \end{pmatrix} \quad P = \begin{pmatrix} 6.8 & 1 \\ 1 & -1 \end{pmatrix} \quad P^{-1} = \frac{1}{-7.8} \begin{pmatrix} -1 & -1 \\ -1 & 6.8 \end{pmatrix} \\ = \frac{1}{7.8} \begin{pmatrix} 1 & 1 \\ 1 & -6.8 \end{pmatrix}$$

Can compute A^n :

$$P^{-1}AP = D \Rightarrow A = PDP^{-1} \Rightarrow A^n = (PDP^{-1})(PDP^{-1}) \dots (PDP^{-1}) \\ = P \cdot D^n \cdot P^{-1}$$

$$A^n = \begin{pmatrix} 6.8 & 1 \\ 1 & -1 \end{pmatrix} \cdot \begin{pmatrix} 1^n & 0 \\ 0 & 0.844^n \end{pmatrix} \cdot \frac{1}{7.8} \begin{pmatrix} 1 & 1 \\ 1 & -6.8 \end{pmatrix}$$

When $n \rightarrow \infty$, $D^n \rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ and

$$A^n \rightarrow \begin{pmatrix} 6.8 & 1 \\ 1 & -1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \cdot \frac{1}{7.8} \begin{pmatrix} 1 & 1 \\ 1 & -6.8 \end{pmatrix}$$

$$= \begin{pmatrix} 6.8 & 0 \\ 1 & 0 \end{pmatrix} \cdot \frac{1}{7.8} \begin{pmatrix} 1 & 1 \\ 1 & -6.8 \end{pmatrix}$$

$$= \frac{1}{7.8} \begin{pmatrix} 6.8 & 6.8 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 6.8/7.8 & 6.8/7.8 \\ 1/7.8 & 1/7.8 \end{pmatrix}$$

$$\approx \underline{\underline{\begin{pmatrix} 0.872 & 0.872 \\ 0.128 & 0.128 \end{pmatrix}}}$$

Conclusion: In the long run ($n \rightarrow \infty$), the state vector is

$$A^n \cdot \underline{v}_0 \rightarrow \begin{pmatrix} 0.872 & 0.872 \\ 0.128 & 0.128 \end{pmatrix} \cdot \begin{pmatrix} 0.9 \\ 0.1 \end{pmatrix} = \underline{\underline{\begin{pmatrix} 0.872 \\ 0.128 \end{pmatrix}}}$$

That is, unemployment is 12.8%

Explanation: Why is $P^{-1}AP = D$ when

$$D = \begin{pmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{pmatrix}, P = (\underline{v}_1 | \underline{v}_2 | \dots | \underline{v}_n)?$$

$$\begin{aligned} A \cdot P &= A \cdot (\underline{v}_1 | \underline{v}_2 | \dots | \underline{v}_n) = (A\underline{v}_1 | A\underline{v}_2 | \dots | A\underline{v}_n) \\ &= (\lambda_1 \underline{v}_1 | \lambda_2 \underline{v}_2 | \dots | \lambda_n \underline{v}_n) \end{aligned}$$

~~$$D \cdot P = \begin{pmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ 0 & 0 & \lambda_3 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_n \end{pmatrix} A =$$~~

$$P \cdot D = (\underline{v}_1 | \underline{v}_2 | \dots | \underline{v}_n) \cdot \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \lambda_n \end{pmatrix} = (\lambda_1 \underline{v}_1 | \lambda_2 \underline{v}_2 | \dots)$$

This means that $AP = PD$. If P is invertible, left multiplication with P^{-1} gives

$$AP = PD$$

$$P^{-1}AP = P^{-1}PD = D$$

$$P^{-1}AP = D$$

Ex: $A = \begin{pmatrix} 7 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 0 & 2 \end{pmatrix}$

$$\begin{vmatrix} 7-\lambda & 0 & 3 \\ 0 & 2-\lambda & 0 \\ 3 & 0 & 2-\lambda \end{vmatrix} = 0$$

$$(2-\lambda) \cdot \begin{vmatrix} 7-\lambda & 3 \\ 3 & 2-\lambda \end{vmatrix} = 0$$

$$(2-\lambda) \cdot (\lambda^2 - 14\lambda + 40) = 0$$

$$\lambda_1 = 2 \quad \lambda_2 = 4, \lambda_3 = 10$$

choose (random)
order of
eigenvalues

$$D = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 10 \end{pmatrix}$$

$\lambda_1 = 2$: $\begin{pmatrix} 5 & 0 & 3 \\ 0 & 0 & 0 \\ 3 & 0 & 5 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$

$$\begin{aligned} 5x + 3z &= 0 \\ 3x + 5z &= 0 \end{aligned} \quad \left. \begin{array}{l} \\ \end{array} \right\} \cdot (-1/5)$$

$$5x + 3z = 0 \quad x = 0$$

$$\frac{16}{5}z = 0 \quad z = 0$$

y = free

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ y \\ 0 \end{pmatrix} = y \cdot \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$

$$\underline{v_1} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$

$\lambda_2 = 4$: $\begin{pmatrix} 3 & 0 & 3 \\ 0 & -2 & 0 \\ 3 & 0 & 3 \end{pmatrix}$

$$3x + 3z = 0 \quad x = -z$$

$$-2y = 0 \quad y = 0$$

$$z = \text{free}$$

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} -z \\ 0 \\ z \end{pmatrix} = z \cdot \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \Rightarrow \underline{v_2} = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}$$

$\lambda_3 = 10$: $\begin{pmatrix} -3 & 0 & 3 \\ 0 & -8 & 0 \\ 3 & 0 & -3 \end{pmatrix}$

$$-3x + 3z = 0 \quad x = z$$

$$-8y = 0 \quad y = 0$$

$$z = \text{free}$$

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} z \\ 0 \\ z \end{pmatrix} = z \cdot \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \Rightarrow \underline{v_3} = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

choose
correspondingly
order of
 $\underline{v}_1, \underline{v}_2, \underline{v}_3$

$$P = (\underline{v}_1 | \underline{v}_2 | \underline{v}_3) = \begin{pmatrix} 0 & -1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{pmatrix}$$
$$D = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 10 \end{pmatrix}$$

This means:

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

know this from
theory, not
necessary to
multiply $P^{-1}AP$
to verify

P is invertible
since eigenvectors
are lin. independent
by construction