

# Recitation 6

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# Stochastic Processes Rabbit Hole

- ▶ Markov Chains are a topic in *stochastic processes*
- ▶ Stochastic Processes (&)
  - ▶ Finite Markov Chains (Discrete time, Discrete space)
  - ▶ Infinite Markov Chains (Discrete time, inf. discrete space)
  - ▶ Poisson Process (Continuous time, Discrete space)
  - ▶ Brownian Motion (Continuous time, continuous space)
- ▶ Main Assumption: Markov Property (Memoryless)
- ▶ Lots of Linear Algebra!

# Questions: Stochastic Matrices

Let  $A, B \in \mathbb{R}^{n \times n}$  be stochastic matrices. True or False for 1,2,3.

1.  $A$  is always invertible
2. The eigenvector corresponding to the largest eigenvalue of  $A$  is unique
3.  $A$  cannot have zero as its eigenvalue
4. Prove that  $AB$  is a stochastic matrix.

Hint for 4. Is there a way express the "sum of each column is 1" property as a matrix multiplication?

# Solutions 1: Stochastic Matrices

Let  $A$  be a stochastic matrix. True or False for 1,2,3.

## Solution

1.  $A$  is always invertible

False  $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$

2. The eigenvector corresponding to the largest eigenvalue of  $A$  is

unique. False  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

3.  $A$  cannot have zero as its eigenvalue. False:  $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$

## Solutions 2: Stochastic Matrices

4. Prove that the product of two stochastic matrices is a stochastic matrix.

### Solution

*Let  $A, B$  be stochastic matrices in  $\mathbb{R}^n$ .*

*Each entry is non-negative:*

$$AB_{i,j} = \sum_{k=1}^n A_{i,k} B_{k,j}$$

*This summation is a sum of non-negative products, hence it is also non-negative.*

*Sum of each columns is 1:*

*Note that the property of each column summing to 1 can be seen as:*

*A matrix  $A$  is stochastic when*

$$\begin{bmatrix} 1 & \dots & 1 \end{bmatrix} A = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix}.$$

*Then*

$$\begin{bmatrix} 1 & \dots & 1 \end{bmatrix} AB = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix} A = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix}.$$

*So  $AB$  is stochastic.*

# Change of Basis

- ▶ Sometimes, a matrix  $A$  ‘prefers’ certain directions (*eigenvectors*)
- ▶ (!!!) These directions act as *anchors* for understanding the action of a matrix.
- ▶ These are directions that we will *orient* or *change our basis* to.
- ▶ This is related to  $P^{-1}DP$ , or diagonalization (Lec 7).
  - ▶ Defined as  $P^{-1}DP$  or  $PDP^{-1}$ , depending on which text/notes you reference.

## Question: Change of Basis

Let  $A$  have eigenvectors  $v_1, \dots, v_n$  with corresponding eigenvalues  $\lambda_1, \dots, \lambda_n$ . Let  $\mathcal{B} = \{v_1, \dots, v_n\}$  be a basis of  $\mathbb{R}^n$ . Let  $x = \sum_{i=1}^n \alpha_i v_i$

1. Let  $P$  be a linear transformation that maps (canonical basis vectors)  $e_i$  to  $v_i$ , for all  $i \in 1, \dots, n$ . Write the matrix  $P$ .
2. What is  $PDP^{-1}x$ ?
3. Let  $k \in \mathbb{N}$ . What is  $(PDP^{-1})^k x$ ?
4. If  $A = PDP^{-1}$ , give an interpretation for the action of  $A$ .

# Solutions 1: Change of Basis

Let  $A$  have eigenvectors  $v_1, \dots, v_n$  with corresponding eigenvalues  $\lambda_1, \dots, \lambda_n$ . Let  $\mathcal{B} = \{v_1, \dots, v_n\}$  be a basis of  $\mathbb{R}^n$ . Let  $x = \sum_{i=1}^n \alpha_i v_i$

## Solution

1. Let  $P$  be a linear transformation that maps (canonical basis vectors)  $e_i$  to  $v_i$ , for all  $i \in 1, \dots, n$ . Write the matrix  $P$ .

$$P = \begin{bmatrix} | & & | \\ v_1 & \dots & v_n \\ | & & | \end{bmatrix}$$

2. Let  $D = \text{Diag}(\lambda_1, \dots, \lambda_n)$ . What is  $PDP^{-1}x$ ?

$$P^{-1}x = P^{-1} \sum_{i=1}^n \alpha_i v_i = \sum_{i=1}^n \alpha_i e_i$$

$$DP^{-1}x = D(\sum_{i=1}^n \alpha_i e_i) = \sum_{i=1}^n \lambda_i \alpha_i e_i$$

$$PDP^{-1}x = P(\sum_{i=1}^n \lambda_i \alpha_i e_i) = \sum_{i=1}^n \lambda_i \alpha_i v_i$$



## Solutions 2: Change of Basis

Let  $A$  have eigenvectors  $v_1, \dots, v_n$  with corresponding eigenvalues  $\lambda_1, \dots, \lambda_n$ . Let  $\mathcal{B} = \{v_1, \dots, v_n\}$  be a basis of  $\mathbb{R}^n$ . Let  $x = \sum_{i=1}^n \alpha_i v_i$

### Solution

3. Let  $k \in \mathbb{N}$  What is  $(PDP^{-1})^k x$ ?

$$(PDP^{-1})^2 = PDP^{-1}PDP^{-1} = PD^2P^{-1}$$

Likewise,

$$(PDP^{-1})^k = PD^kP^{-1}$$

4. If  $A = PDP^{-1}$ , give an interpretation for the action of  $A$  on a vector  $x$ .

*First view  $x$  as a vector in the coordinates of basis  $v_1, \dots, v_n$ .*

*$P^{-1}$  transforms these coordinates into the standard basis.*

*$D$  stretches the new coordinate  $e_i$  by the eigenvalue  $\lambda_i$ .*

*$P$  transforms these coordinates back into the basis of  $v_1, \dots, v_n$*

# Questions: Spectral Theorem

1. Let  $A \in \mathbb{R}^{n \times n}$  be a symmetric matrix. Give a vector  $v$  with  $\|v\| = 1$  such that  $\|Av\|$  is maximized.
2. Let  $A \in \mathbb{R}^{n \times m}$ ,  $B \in \mathbb{R}^{n \times n}$ , and  $C \in \mathbb{R}^{m \times m}$ . Prove that if  $B$  and  $C$  are invertible, then  $\text{rank}(A) = \text{rank}(BAC)$ .
3. Let  $A = P^{-1}DP$ , where  $P$  is invertible and  $D$  is diagonal, and  $A, D, P \in \mathbb{R}^{n \times n}$ . Prove that  $\text{Tr}(A) = \text{Tr}(D)$ .

# Solutions 1: Spectral Theorem

1. Let  $A \in \mathbb{R}^{n \times n}$  be a symmetric matrix. Give a vector  $v$  with  $\|v\| = 1$  such that  $\|Av\|$  is maximized.

## Solution

Let  $A = U\Lambda U^T$ , where  $U$  is orthogonal with columns  $u_1, \dots, u_n$ , and  $\Lambda$  is diagonal with entries  $\lambda_1, \dots, \lambda_n$ .

Let  $v = \sum_{i=1}^n \alpha_i u_i$ , where  $\sum_{i=1}^n \alpha_i^2 = 1$

Then,

$$U^T v = \sum_{i=1}^n \alpha_i e_i.$$

$$\Lambda U^T v = \sum_{i=1}^n \lambda_i \alpha_i e_i$$

$$U\Lambda U^T v = \sum_{i=1}^n \lambda_i \alpha_i u_i$$

$$\|U\Lambda U^T v\| = \left\| \sum_{i=1}^n \lambda_i \alpha_i u_i \right\|$$

$$\|U\Lambda U^T v\| = \sum_{i=1}^n \lambda_i^2 \alpha_i^2 \|u_i\| \quad \text{by orthonormality}$$

$$\|U\Lambda U^T v\| = \sum_{i=1}^n \lambda_i^2 \alpha_i^2$$

Maximize this quantity by setting  $\alpha_j = 1$ , where  $j$  is the index with the largest magnitude eigenvalue.

## Solutions 2 : Spectral Theorem

2. Let  $A \in \mathbb{R}^{n \times m}$ ,  $B \in \mathbb{R}^{n \times n}$ , and  $C \in \mathbb{R}^{m \times m}$ . Prove that if  $B$  and  $C$  are invertible, then  $\text{rank}(A) = \text{rank}(BAC)$ .

### Solution

*From Hw 3:*

*If  $M$  is invertible, then*

*$\text{rank}(MA) = \text{rank}(A)$ , and  $\text{rank}(AM) = \text{rank}(A)$ .*

*Then,*

*$\text{rank}(BAC) = \text{rank}((B)(AC)) = \text{rank}(AC) = \text{rank}(A)$ .*

## Solutions 3 : Spectral Theorem

3. Let  $A = P^{-1}DP$ , where  $P$  is invertible and  $D$  is diagonal, and  $A, D, P \in \mathbb{R}^{n \times n}$ . Prove that  $\text{Tr}(A) = \text{Tr}(D)$ .

### Solution

*From Hw3:  $\text{Tr}(AB) = \text{Tr}(BA)$ .*

*Then*

$$\text{Tr}(A) = \text{Tr}(P^{-1}DP) = \text{Tr}((P^{-1}D)P) = \text{Tr}(P(P^{-1}D)) = \text{Tr}(D).$$