

# EE16A - Lecture 27 Notes

Name: Felix Su SID: 25794773

Spring 2016 GSI: Ena Hariyoshi

## Review

1. **Eigenvalue-Eigenvector Eq.:**  $A\vec{v} = \lambda\vec{v}$

## Diagonalization

- Set of eigenvectors ( $\vec{v}_i$ ) and eigenvalues ( $\lambda_i$ ) for  $i = 1, \dots, N$  s.t.  $[A\vec{v}_1 \ \dots \ A\vec{v}_N] = [\lambda\vec{v}_1 \ \dots \ \lambda\vec{v}_N]$ 
  - Left side: Take out  $A$
  - Right side: Take out diagonal eigenvalue matrix ( $\lambda$ 's)
  - $\implies A [\vec{v}_1 \ \dots \ \vec{v}_N] = [\vec{v}_1 \ \dots \ \vec{v}_N] \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$
  - So,  $AV = V\Lambda$
  - Postmultiply by  $V^{-1}$  on both sides to get  $A = V\Lambda V^{-1}$

**Summary Equation for Eigen Decomposition of Matrix A:**

$$AV = V\Lambda \text{ where } V = [\vec{v}_1 \ \dots \ \vec{v}_N] \text{ and } \Lambda = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \quad (1)$$

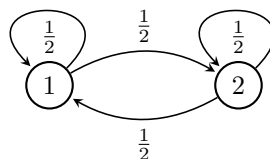
$$A = V\Lambda V^{-1} \quad (2)$$

## Pagerank

- $A = V\Lambda V^{-1}$
- $s[n] = A^n s[0]$
- $A^n = V\Lambda^n V^{-1}$ 
  - If  $|\lambda_i| > 1 \rightarrow \lambda_i^n$  keeps growing
  - If  $|\lambda_i| < 1 \rightarrow \lambda_i^n$  decays towards 0
  - If  $|\lambda_i| = 1 \rightarrow \lambda_i^n$  stays the same

### Example:

Pagerank Example



$$A = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

1 linearly independent col/row  $\implies \text{rank}(A) = 1$

One of the eigenvalues has to be zero:  $\lambda_1 = 0$ , the other is non-zero:  $\lambda_2 = n$

0 eigenvalue corresponds to singular matrix

A is a projection matrix onto vector  $\vec{r} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

Projection matrix  $P = \frac{\vec{r}\vec{r}^T}{\vec{r}^T\vec{r}}$

A is symmetric

A is column and row-stochastic (doubly-stochastic)

column-stochastic matrix = Markov matrix  $A \geq 0$  (non-negative)

$$A - \lambda I = \begin{bmatrix} \frac{1}{2} - \lambda & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} - \lambda \end{bmatrix}$$

$$\det(A - \lambda I) = (\frac{1}{2} - \lambda)^2 - (\frac{1}{2})^2 = 0 \implies \lambda(1 - \lambda) = 0, \text{ so, } \lambda = 0, 1$$

Markov matrices always have 1 as an eigenvalue (All other eigenvalues are smaller), so, as the system continues to run, the other eigenvalues decay to 0 and the eigenvector corresponding to the  $\lambda_i = 1$  will be the **important score** for the website

$$\lambda_1 = 0, \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = 0 \implies \vec{v}_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\lambda_2 = 1, \begin{bmatrix} \frac{1}{2} - 1 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} - 1 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{bmatrix} = 0 \implies \vec{v}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$V = [\vec{v}_1 \quad \vec{v}_2] = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

$$\Lambda = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

$$V^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$A = V\Lambda V^{-1} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = [\vec{v}_1 \quad \vec{v}_2] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \begin{bmatrix} \vec{w}_1^T \\ \vec{w}_2^T \end{bmatrix} = \lambda_1 \vec{v}_1 \vec{w}_1^T + \lambda_2 \vec{v}_2 \vec{w}_2^T$$

$$A^n = V\Lambda^n V^{-1} = \sum_{i=1}^N \lambda_i^n \vec{v}_i \vec{w}_i^T$$

### General Form of nth Power of a Matrix

$$A^n = V\Lambda^n V^{-1} = \sum_{i=1}^N \lambda_i^n \vec{v}_i \vec{w}_i^T \quad (3)$$

### Importance Score for Pagerank

#### Pagerank Importance Score

Converges to the value  $A^n s[0]$  where:

$$\lim_{n \rightarrow \infty} A^n = \lambda_i \vec{v}_i \vec{w}_i^T = \vec{v}_i \vec{w}_i^T \text{ where } \lambda_i = 1 \text{ if } |\lambda_i| < 1 \forall i = 1, \dots, N-1 \quad (4)$$

### Preview of Next Lecture

- If  $s[n+1] = As[n]$ , and  $\vec{q}[n] = V^{-1}s[n] \implies \vec{s}[n] = V\vec{q}[n]$
- Then,  $V\vec{q}[n+1] = AV\vec{q}[n] \implies \vec{q}[n+1] = V^{-1}AV\vec{q}[n]$

- Because  $A = V\Lambda V^{-1} \implies V^{-1}\Lambda V = \Lambda$
- $\therefore \bar{q}[n+1] = \Lambda \bar{q}[n]$ 
  - Decouples dynamic system