

MATH 22 LECTURE 29 CLASSWORK

AUGUST 23, 2017

(1) Let $A = \begin{bmatrix} -1 & 0 \\ -1 & 0 \end{bmatrix}$.

- (a) Maximize $\|Ax\|$ subject to the constraint that $\|\mathbf{x}\| = 1$.

$$A^T A = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}, \text{ so } \lambda_1 = 2, \lambda_2 = 0 \Rightarrow \sigma_1 = \sqrt{2}$$

$\sqrt{2}$ is the largest $\|A\mathbf{x}\|$ can be for \mathbf{x} on the unit circle in \mathbb{R}^2 .

- (b) Compute the SVD of A and A^T .

$$A = \underbrace{\begin{bmatrix} -\sqrt{2} & -\sqrt{2} \\ -\sqrt{2} & \sqrt{2} \end{bmatrix}}_{U} \underbrace{\begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix}}_{\Sigma} \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}}_{V^T}$$

$$A^T = V \Sigma^T U^T$$

- (c) Find orthonormal bases for as many fundamental spaces as possible!

$\{\vec{u}_1\}$ is an o.n.b. for $\text{Col } A = (\text{Nul } (A^T))^{\perp}$

$\{\vec{u}_2\}$ is an o.n.b. for $((\text{Col } A)^{\perp})^{\perp} = \text{Nul } (A^T)$

$\{\vec{v}_2\}$ is an o.n.b. for $\text{Nul } A = (\text{Row } A)^{\perp}$

$\{\vec{v}_1\}$ is an o.n.b. for $(\text{Nul } A)^{\perp} = \text{Row } A$

- (d) What is the best rank 1 approximation of A ? A is rank 1, so A .

$$(2) \text{ Let } A = \begin{bmatrix} 1 & 2 & 0 \\ 1 & 0 & 2 \\ 0 & 1 & -1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 2 \\ 0 \\ -5 \end{bmatrix}.$$

(a) Find all least-squares solutions to $A\mathbf{x} = \mathbf{b}$.

$$A^T A = \begin{bmatrix} 2 & 2 & 2 \\ 2 & 5 & -1 \\ 2 & -1 & 5 \end{bmatrix}, \quad A^T \vec{b} = \begin{bmatrix} 2 \\ -1 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 2 & 2 & 2 \\ 2 & 5 & -1 & -1 \\ 2 & -1 & 5 & 5 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 2 & 2 \\ 0 & 1 & -1 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\Rightarrow \hat{\mathbf{x}} = \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}, \text{ where } x_3 \in \mathbb{R}.$$

(b) Let $W = \text{Col}A$. Decompose $\mathbf{b} = \hat{\mathbf{b}} + \mathbf{z}$ with $\hat{\mathbf{b}} \in W$ and $\mathbf{z} \in W^\perp$. Prove that this decomposition is unique.

$$\hat{\mathbf{b}} = A \hat{\mathbf{x}} = A \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} + x_3 A \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 2 \\ -1 \end{bmatrix}$$

$$\vec{z} = \vec{b} - \hat{\vec{b}}. \quad \text{Suppose } \vec{b} = \hat{\vec{b}} + \vec{z} = \hat{\vec{b}}_1 + \vec{z}_1,$$

with $\hat{\vec{b}}_1 \in W$ and $\vec{z}_1 \in W^\perp$. Then

$$\underbrace{\hat{\vec{b}} - \hat{\vec{b}}_1}_{\in W} = \underbrace{\vec{z}_1 - \vec{z}}_{\in W^\perp} \quad \text{which implies}$$

$$\left. \begin{aligned} (\hat{\vec{b}} - \hat{\vec{b}}_1) \cdot (\hat{\vec{b}} - \hat{\vec{b}}_1) &= 0 \\ (\vec{z}_1 - \vec{z}) \cdot (\vec{z}_1 - \vec{z}) &= 0 \end{aligned} \right\} \Rightarrow \begin{aligned} \hat{\vec{b}} - \hat{\vec{b}}_1 &= \vec{0} \\ \vec{z}_1 - \vec{z} &= \vec{0} \end{aligned}$$

(c) Can we compute the QR -factorization of A ?

2

No, we need A to have linearly independent columns to compute QR .

$$(3) \text{ Let } A = \begin{bmatrix} 4 & -2 \\ 1 & 1 \end{bmatrix}.$$

(a) Check that A is diagonalizable and diagonalize it.

$$\text{charpoly}(A) = (3-\lambda)(2-\lambda), \text{ so we know } A \text{ diagonalizable.}$$

$$\text{we find } \vec{x}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \in \text{Nul}(A - \lambda_1 I_2) \quad \lambda_1 := 3$$

$$\vec{x}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \in \text{Nul}(A - \lambda_2 I_2) \quad \lambda_2 := 2$$

$$\text{Then } A = \underbrace{\begin{bmatrix} 2 & -1 \\ 1 & -1 \end{bmatrix}}_P \underbrace{\begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}}_D \underbrace{\begin{bmatrix} 1 & -1 \\ 1 & -2 \end{bmatrix}}_{P^{-1}}$$

(b) Is the matrix P unique? If so, prove it. If not, provide an example.

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 2 & -2 \\ -1 & 2 \end{bmatrix}, \text{ so No.}$$

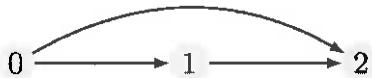
Bases are not unique.

$$(c) \text{ Compute } A^{100} \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad P := \begin{bmatrix} 2 & -1 \\ 1 & -1 \end{bmatrix}$$

$$P^{-1} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}, \quad P \cdot D^{100} = \begin{bmatrix} 2 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 3^{100} & 0 \\ 0 & 2^{100} \end{bmatrix} = \begin{bmatrix} 2 \cdot 3^{100} (-1) 2^{100} \\ 3^{100} (-1) 2^{100} \end{bmatrix}$$

$$\Rightarrow \underbrace{P \cdot D^{100} \cdot P^{-1} \begin{bmatrix} 1 \\ -1 \end{bmatrix}}_{A^{100} \begin{bmatrix} 1 \\ -1 \end{bmatrix}} = \underbrace{\begin{bmatrix} 4 \cdot 3^{100} - 3 \cdot 2^{100} \\ 2 \cdot 3^{100} - 3 \cdot 2^{100} \end{bmatrix}}_{\in \mathbb{R}^2}$$

(4) Consider the "web" given below:



- (a) Use the PageRank algorithm with $\alpha = 1$ to find a probability vector that measures the importance of each node. Explain how the algorithm changes if we instead let $\alpha = 0.85$.

$$S = \begin{bmatrix} 0 & 0 & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{3} \\ \frac{1}{2} & 1 & \frac{1}{3} \end{bmatrix}, \text{ charpoly } \xrightarrow{\text{why?}} (\lambda - 1) \cdot \text{quadratic}$$

$$S - I_3 = \begin{bmatrix} -1 & 0 & \frac{1}{3} \\ \frac{1}{2} & -1 & \frac{1}{3} \\ \frac{1}{2} & 1 & -\frac{2}{3} \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & -\frac{1}{3} \\ 0 & 1 & -\frac{1}{2} \\ 0 & 0 & 0 \end{bmatrix}$$

$$\Rightarrow \lambda = 1 \text{ eigenvector is } \begin{bmatrix} \frac{1}{3} \\ \frac{1}{2} \\ 1 \end{bmatrix}.$$

Rescaling we get the desired probability vector

$$\vec{q} = \left(\frac{1}{11/6}\right) \begin{bmatrix} \frac{1}{3} \\ \frac{1}{2} \\ 1 \end{bmatrix} = \begin{bmatrix} 2/11 \\ 3/11 \\ 6/11 \end{bmatrix}.$$

If $\alpha = 0.85$, then take $G_0 = \alpha S + (1-\alpha)\left(\frac{1}{2}\right) \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ and repeat this process with G_1 .

- (b) How do you know that the steady-state vector (of the dynamical system defined by A) is unique?

S^2 has all strictly positive values (i.e. S is regular) which (by a black box) guarantees a unique steady state vector.