vv214: Markov Chains. Matrix Norms. Ranking Problem.

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- 1. Markov Chains.
- 2. Matrix Norms.
- 3. Ranking Problem.

Markov Chains

- We are given a set of states and initial probability distribution over states.
- A Markov chain is a memory-less stochastic process where $X_t = i$ is the state at time t and the probability of reaching the state $X_{t+1} = j$ at time t+1 is

$$\mathbb{P}(X_{t+1}=j|X_t=i)=p_{ij}$$

- It depends on the current state but on time or other variables.
- $ightharpoonup T = (p)_{ij}$ is the transition matrix $p_{ij} \geq 0$

$$\sum_{i=1}^{n} p_{ij} = 1 \qquad \qquad \sum_{j=1}^{n} p_{ij} = 1$$

left stochastic matrix right stochastic matrix

Let $q_t \in \mathbb{R}^n$ be the distribution of the process at time t then

$$q_{t+1} = Tq_t$$
 OR $q_{t+1}^T = q_t^T T$

Markov Chains

Let $A_{n \times n}$ be a right (left) stochastic matrix.

Definition: A non-zero vector $\bar{x} \in \mathbb{R}^n$ is called the right eigenvector of the matrix A if

$$A\bar{x} = \lambda_R \bar{x} \Rightarrow \det(A - \lambda_R I) = 0$$

Definition: A non-zero vector $\bar{x} \in \mathbb{R}^n$ is called the left eigenvector of the matrix A if

$$\bar{y}^T A = \lambda_L \bar{y}^T \Rightarrow \det(A^T - \lambda_L I) = 0$$

Are left and right eigenvalues the same?

$$0 = \det(A^T - \lambda_L I) = \det(A^T - \lambda_L I^T) = \det(A - \lambda_L I)^T = \det(A - \lambda_L I)$$

$$\lambda_L = \lambda_R$$

Definition: A norm of a linear operator $T: X \to Y$ is

$$||T|| = \max_{x \in X: ||x|| = 1} ||Tx|| \quad \text{OR} \quad ||T|| = \max_{x \in X: x \neq 0} \frac{||Tx||}{||x||}$$

Exercise: Show that all the properties of a norm are satisfied.

Definition: A linear operator is said to be bounded if $||T|| < +\infty$

Lemma: T is bounded iff $\exists C > 0 \quad ||Tx|| \le C \cdot ||x||$

Show that

$$||A||_1 = \max_{j=1,n} \sum_{i=1}^n |a_{ij}| \qquad ||A||_{\infty} = \max_{i=1,n} \sum_{j=1}^n |a_{ij}|$$

Define

$$||A||_F = \sqrt{\sum_{i,j} a_{ij}^2} = \sqrt{trace(A^T A)}$$

Show that

$$||A||_2 = \max_{||ar{x}||_2 = 1} ||Aar{x}||_2 = \sqrt{\lambda_{max}}, \ \lambda_{max} \ \text{is the largest eigenvalue of} \ A^T A$$

$$A = \begin{pmatrix} 0 & 1 \\ 1 & 1 \\ 1 & 0 \end{pmatrix} \quad \text{Find} \quad ||A||_1, ||A||_{\infty}, ||A||_F, ||A|_2$$

Exercise: Show that $\forall \bar{x} \in \mathbb{R}^n \quad ||\bar{x}||_{\infty} \leq ||\bar{x}||_2 \leq ||\bar{x}||_1$

 $\forall \bar{x} \in \mathbb{R}^n \quad ||\bar{x}||_i \leq \alpha ||\bar{x}||_j \text{ where } \alpha \text{ is the } (i,j) \text{ of the matrix}$

$$\left(\begin{array}{ccc}
\star & \sqrt{n} & n \\
1 & \star & \sqrt{n} \\
1 & 1 & \star
\end{array}\right)$$

 $\forall A_{n \times n} \quad ||A||_i \le \alpha ||A||_j$ where α is the (i,j) of the matrix

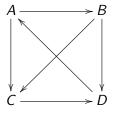
$$\begin{pmatrix}
\star & \sqrt{n} & n & \sqrt{n} \\
\sqrt{n} & \star & \sqrt{n} & 1 \\
n & \sqrt{n} & \star & \sqrt{n} \\
\sqrt{n} & \sqrt{n} & \sqrt{n} & \star
\end{pmatrix}$$

MATLAB

- \triangleright norm(A,2) returns $||A||_2$
- ▶ norm(A) is the same as norm(X,2)
- ▶ norm(A,'fro') returns ||A||_F
- ▶ norm(V,P) returns the p-norm of V
- ▶ norm(V,Inf) returns $||V||_{\infty}$

Eigenvalues of a Stochastic matrix

* Consider the results of a tournament



where " $A \rightarrow B$ " means A defeated B.

How to rank the players fairly?

* The Page-Brin idea:it should be worth more to defeat a better player \Leftarrow based on Perron (1907)-Frobenius (1912) Theorem

How do we know who is better before ranking them?

* Define recursion!

Give everyone the initial score of 1

$$ar{x}_0 = \left(egin{array}{c} 1 \\ 1 \\ 1 \\ 1 \end{array}
ight) egin{array}{c} A \\ B \\ C \\ D \end{array}$$

* Define for all $n \ge 0$

$$\bar{x}_{n+1} = A\bar{x}_n$$

where

$$A = \begin{array}{cccc} & A & B & C & D \\ A & & & & & \\ C & & & & \\ D & & & & \\ \end{array}$$

$$\bar{x}_{1} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 2 \\ 2 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} A \\ B \\ C \\ D \end{pmatrix}$$

$$\bar{x}_{2} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 2 \\ 2 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \\ 1 \\ 2 \end{pmatrix} \begin{pmatrix} A \\ B \\ C \\ D \end{pmatrix}$$

$$\bar{x}_{3} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 3 \\ 3 \\ 3 \\ 2 \\ 3 \end{pmatrix} \begin{pmatrix} A \\ B \\ C \\ D \end{pmatrix}$$

The (n+1)th score of a player A is the sum of the nth scores of the players that the player A defeated.

$$\bar{x}_5 = \begin{pmatrix} 8 \\ 6 \\ 3 \\ 5 \end{pmatrix}, \ \bar{x}_{10} = \begin{pmatrix} 35 \\ 34 \\ 21 \\ 26 \end{pmatrix}, \ \bar{x}_{100} = \begin{pmatrix} 1037 \\ 933 \\ 547 \\ 731 \end{pmatrix}$$

Is
$$A > B > D > C$$
? \Rightarrow analyze $\bar{x}_n = A^n \bar{x}_0$ as $n \to \infty$

Theorem (Perron-Frobenius): There exists a *largest positive* eigenvalue λ_{PF} for a nonnegative matrix A such that the rescaled system

$$\bar{x}_n = \left(\frac{1}{\lambda_{PF}}A\right)^n \bar{x}_0$$

converges to an equilibrium state \bar{x}_{∞} .

$$ar{x}_{\infty} = ar{x}_{\infty+1} = rac{1}{\lambda_{PF}} A ar{x}_{\infty} \Rightarrow A ar{x}_{\infty} = \lambda_{PF} ar{x}_{\infty}$$

The equilibrium state is the eigenvector associated with $\lambda_{PF}!!!$

The largest positive eigenvalue is

$$\lambda_{PF} = 1.3953369...$$

and

$$ar{x}_{\infty} = \left(egin{array}{c} 0.321... \\ 0.288... \\ 0.165... \\ 0.230... \end{array}
ight) \left. egin{array}{c} A \\ B \\ C \\ D \end{array}
ight.$$