Recitation 6

Markov Chains

Definition (Markov chain)

A sequence of random variables (X_0, X_1, \dots) is a Markov chain with state space E and "transition matrix" P if for all $t \ge 0$,

$$\mathbb{P}(X_{t+1} = y \mid X_0 = x_0, \dots, X_t = x_t) = P(x_t, y)$$

for all x_0, \ldots, x_t such that $\mathbb{P}(X_0 = x_0, \ldots, X_t = x_t) > 0$.

Stochastic matrix: $P_{ij} \ge 0$, $\sum_{i=1}^{n} P_{ij} = 1$ for all $1 \le j \le n$.

Definition (Invariant measure)

A vector $\mu \in \Delta_n$ is called an invariant measure for the transition matrix P if $\mu = P\mu$, i.e. if μ is an eigenvector of P associated with the eigenvalue 1.

Perron-Frobenius theorem

Theorem (Perron-Frobenius, stochastic case)

Let P be a stochastic matrix such that there exists $k \geq 1$ such that all the entries of P^k are strictly positive. Then the following holds:

- 1. 1 is an eigenvalue of P and there exists an eigenvector $\mu \in \Delta_n$ associated to 1.
- 2. The eigenvectors associated to 1 are unique up to scalar multiple (i.e. $Ker(P Id) = Span(\mu)$).
- 3. For all $x \in \Delta_n$, $P^t x \xrightarrow[t \to \infty]{} \mu$.

Is the condition "there exists $k \ge 1$ such that all the entries of P^k are strictly positive" necessary? Let's see!

Definition (Irreducible Markov chain)

If for all $1 \le i, j \le n$, there exists $k \ge 1$ such that $P_{ij}^k > 0$, we say that the Markov chain is irreducible.

Definition (Aperiodic Markov chain)

If for all $1 \le i \le n$, we have $\gcd(\{k|P_{ii}^k>0\})=1$, we say that the Markov chain is aperiodic.

- 1. Show that if "there exists $k \geq 1$ such that all the entries of P^k are strictly positive", then the Markov chain is irreducible and aperiodic. The converse is also true but harder to prove (come to office hours if you want to know!).
- 2. Show that irreducible non-aperiodic Markov chains have no invariant measure.
- 3. Show that non-irreducible aperiodic Markov chains have several invariant measures.

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3. Show an example of a non-irreducible aperiodic Markov chains that has several invariant measures.

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4. Remember from the lecture that the PageRank algorithm actually computes the invariant measure of the transition matrix

$$G = \alpha P + \frac{1 - \alpha}{N} \begin{bmatrix} 1 & \cdots & 1 \\ \vdots & \cdots & \vdots \\ 1 & \cdots & 1 \end{bmatrix}$$

with $\alpha \approx 0.85$. Given the previous questions, what would be the problems in taking $\alpha = 1$?

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Spectral theorem

Theorem (Spectral theorem)

Let $A \in \mathbb{R}^{n \times n}$ be a symmetric matrix. Then, A has n orthogonal eigenvectors q_1, \ldots, q_n and we can write $A = Q\Lambda Q^\top$, where $Q = \begin{bmatrix} q_1 & \cdots & q_n \end{bmatrix}$ and Λ is diagonal.

Remember that a matrix A is diagonalizable iff it has n linearly independent eigenvectors (equivalently $A=V\Lambda V^{-1}$). Thus, the spectral theorem says that symmetric matrices are diagonalizable in an orthogonal basis.

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Theorem (Courant-Fischer principle)

The eigenvalues $\lambda_1 \geq \cdots \geq \lambda_n$ of a symmetric matrix A are given by

$$\lambda_k = \max_{\substack{S \subset \mathbb{R}^n \\ \dim(S) = k}} \min_{\substack{x \in S \\ x \neq 0}} \frac{x^\top A x}{x^\top x} = \min_{\substack{S' \subset \mathbb{R}^n \\ \dim(S') = n - k + 1}} \max_{\substack{x \in S' \\ x \neq 0}} \frac{x^\top A x}{x^\top x}$$

We will show this theorem. Seems like a lot, but we'll go step by step! We will only show the first equality as the argument for the second one is analogous.

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Let v_1, \dots, v_n be the orthogonal basis of eigenvectors (resp.).

- (i) Show that for $S_k = \operatorname{span}(v_1, \ldots, v_k)$, $\min_{x \in S, x \neq 0} \frac{x^\top Ax}{x^\top x} \ge \lambda_k$.
- (ii) Show that for all $x \in S_k' = \operatorname{span}(v_k, \dots, v_n)$ such that $x \neq 0$, we have $\frac{x^\top Ax}{x^\top x} \leq \lambda_k$.
- (iii) Check that for any subspace S with $\dim(S) = k$, $S \cup S'_k \neq \{0\}$.
- (iv) Conclude.

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Extra question: Spectral theorem

For a symmetric matrix, show how you can use a modification of Gaussian elimination to find a decomposition $A = V\Lambda V^{\top}$ for some diagonal matrix Λ .