Quiz Section 9

Review

- 1) An estimator $\hat{\theta}$ for a parameter θ of a probability distribution is **unbiased** iff $\mathbb{E}[\hat{\theta}(X_1,\ldots,X_n)] = \theta$
- 2) discrete-time stochastic process (DTSP) is a sequence of random variables $X^{(0)}, X^{(1)}, X^{(2)}, ...$, where $X^{(t)}$ is the value at time t. For example, the temperature in Seattle or stock price of TESLA each day, or which node you are at after each time step on a random walk on a graph.
- 3) Markov Chain is a DTSP, with the additional following three properties:
 - (a) ...has a finite (or countably infinite) state space $S = \{s_1, \dots, s_n\}$ which it bounces between, so each $X^{(t)} \in S$.
 - (b) ...satisfies the **Markov property**. A DTSP satisfies the Markov property if the future is (conditionally) independent of the past given the present. Mathematically, it means,

$$\mathbb{P}\left(X^{(t+1)} = x_{t+1} \mid X^{(0)} = x_0, X^{(1)} = x_1, \dots, X^{(t-1)} = x_{t-1}, X^{(t)} = x_t\right) = \mathbb{P}\left(X^{(t+1)} = x_{t+1} \mid X^{(t)} = x_t\right).$$

- (c) ...has **fixed transition probabilities**. Meaning, if we are at some state s_i , we transition to another state s_j with probability *independent* of the current time. Due to this property and the previous, the transitions are governed by n^2 probabilities: the probability of transitioning from one of n current states to one of n next states. These are stored in a square $n \times n$ transition probability matrix (TPM) M, where $M_{ij} = \mathbb{P}\left(X^{(t+1)} = s_j \mid X^{(t)} = s_i\right)$ is the probability of transitioning from state s_i to state s_j for any/every value of t.
- **4)** A **stationary distribution** of a Markov chain is a probability distribution on states that is unchanged by taking one step of the Markov chain.

Task 1 – A biased estimator

In class, we showed that the maximum likelihood estimate of the variance θ_2 of a normal distribution (when both the true mean μ and true variance σ^2 are unknown) is what's called the *population variance*. That is

$$\hat{\theta}_2 = \left(\frac{1}{n} \sum_{i=1}^n (x_i - \hat{\theta}_1)^2\right)$$

where $\hat{\theta}_1 = \frac{1}{n} \sum_{i=1}^n x_i$ is the MLE of the mean. Is $\hat{\theta}_2$ unbiased?

Task 2 – Weather Forecast

A weather forecaster predicts sun with probability θ_1 , clouds with probability $\theta_2 - \theta_1$, rain with probability $\frac{1}{2}$ and snow with probability $\frac{1}{2} - \theta_2$. This year, there have been 55 sunny days, 100 cloudy days, 160 rainy days and 50 snowy days. What is the maximum likelihood estimator for θ_1 and θ_2 ?

Task 3 – Faulty Machines

You are trying to use a machine that only works on some days. If on a given day, the machine is working it will break down the next day with probability 0 < b < 1, and works on the next day with probability 1 - b. If it is not working on a given day, it will work on the next day with probability 0 < r < 1 and not work the next day with probability 1 - r.

- a) In this problem we will formulate this process as a Markov chain. First, let $X^{(t)}$ be a variable that denotes the state of the machine at time t. Then, define a state space $\mathcal S$ that includes all the possible states that the machine can be in. Lastly, for all $A,B\in\mathcal S$ find $\mathbb P(X^{(t+1)}=A\mid X^{(t)}=B)$ (A and B can be the same state).
- b) Suppose that on day 1, the machine is working. What is the probability that it is working on day 3?
- c) As $n \to \infty$, what does the probability that the machine is working on day n converge to? To get the answer, solve for the *stationary distribution*.

Task 4 - Another Markov Chain

Suppose that the following is the transition probability matrix for a 4 state Markov chain (states 1,2,3,4).

$$M = \begin{bmatrix} 0 & 1/2 & 1/2 & 0 \\ 1/3 & 0 & 0 & 2/3 \\ 1/3 & 1/3 & 0 & 1/3 \\ 1/5 & 2/5 & 2/5 & 0 \end{bmatrix}$$

- a) What is the probability that $X^{(2)} = 4$ given that $X^{(0)} = 4$?
- b) Write down the system of equations that the stationary distribution must satisfy and solve them.

Task 5 - Three Tails

You flip a fair coin until you see three tails in a row. Model this as a Markov chain with the following states:

- S: start state, which we are only in before flipping any coins.
- H: We see a heads, which means no streak of tails currently exists.
- T: We've seen exactly one tail in a row so far.
- TT: We've seen exactly two tails in a row so far.
- TTT: We've accomplished our goal of seeing three tails in a row, stop flipping, and stay there.
- a) Write down the transition probability matrix.
- b) Write down the system of equations whose variables are D(s) for each state $s \in \{S, H, T, TT, TTT\}$, where D(s) is the expected number of steps until state TTT is reached starting from state s. Solve this system of equations to find D(S).
- c) Write down the system of equations whose variables are $\gamma(s)$ for each state $s \in \{S, H, T, TT, TTT\}$, where $\gamma(s)$ is the expected number of heads seen before state TTT is reached. Solve this system to find $\gamma(S)$, the expected number of heads seen overall until getting three tails in a row.