

STAT 531: Assignment # 3

Due on November 5, 2020

Dr. Lu TTh 11:00 am

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Problem 1

(a) We want to calculate the 2-step transition probability matrix.

$$P^2 = \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} = \begin{pmatrix} 0.52 & 0.48 \\ 0.48 & 0.52 \end{pmatrix}$$

(b) We want to find $P(X_3 = I | X_0 = I)$.

$$P^3 = \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} = \begin{pmatrix} 0.496 & 0.504 \\ 0.504 & 0.496 \end{pmatrix}$$

And so $P(X_3 = I | X_0 = I) = 0.496$

Problem 2

(a) We want to find $P(X_2 = 2 | X_0 = 1)$.

$$P^2 = \begin{pmatrix} 0.8 & 0.2 \\ 0.4 & 0.6 \end{pmatrix} \begin{pmatrix} 0.8 & 0.2 \\ 0.4 & 0.6 \end{pmatrix} = \begin{pmatrix} 0.72 & 0.28 \\ 0.56 & 0.44 \end{pmatrix}$$

And so $P(X_2 = 2 | X_0 = 1) = 0.28$.

(b) We want to find π_2 in $\boldsymbol{\pi} = \lim_{n \rightarrow \infty} \boldsymbol{\pi}^{(n)} = (\pi_1 \ \pi_2)$. We start with

$$\begin{aligned} \begin{pmatrix} \pi_1 & \pi_2 \end{pmatrix} \begin{pmatrix} 0.8 & 0.2 \\ 0.4 & 0.6 \end{pmatrix} &= \begin{pmatrix} \pi_1 & \pi_2 \end{pmatrix} \\ &\rightarrow 0.8\pi_1 + 0.4\pi_2 = \pi_1 \\ &\quad 0.2\pi_1 + 0.6\pi_2 = \pi_2 \\ &\quad \pi_1 + \pi_2 = 1 \\ &\rightarrow \underbrace{\begin{pmatrix} -0.2 & 0.4 \\ 1 & 1 \end{pmatrix}}_A \begin{pmatrix} \pi_1 \\ \pi_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ &\quad \det(A) = -0.2 - 0.4 = -0.6 \end{aligned}$$

We calculate each part as

$$\begin{aligned} \pi_1 &= \frac{1}{-0.6} \det \begin{pmatrix} 0 & 0.4 \\ 1 & 1 \end{pmatrix} \\ &= \frac{-0.4}{-0.6} \\ &= \frac{2}{3} \end{aligned}$$

$$\begin{aligned} \pi_2 &= \frac{1}{-0.6} \det \begin{pmatrix} -0.2 & 0 \\ 1 & 1 \end{pmatrix} \\ &= \frac{-0.2}{-0.6} \\ &= \frac{1}{3} \end{aligned}$$

Therefore, the probability that Nov. 1 next year Rainbow will rain is $\pi_2 = \frac{1}{3}$.

Problem 3

(a) We want to find the distribution π_2 of X_2 . We find

$$\begin{aligned} P^{(2)} &= (0 \ 1)P^2 \\ \rightarrow P^{(2)} &= (0 \ 1) \begin{pmatrix} 0.8 & 0.2 \\ 0.1 & 0.9 \end{pmatrix} \begin{pmatrix} 0.8 & 0.2 \\ 0.1 & 0.9 \end{pmatrix} \\ &= (0 \ 1) \begin{pmatrix} 0.66 & 0.34 \\ 0.17 & 0.83 \end{pmatrix} \\ &= (0.17 \ 0.83) \end{aligned}$$

Hence, the distribution of X_2 is given by $X_2 \sim (0.17 \ 0.83)$.

(b) We want to find the steady-state distribution of X_n .

$$\begin{aligned} (\pi_1 \ \pi_2) \begin{pmatrix} 0.8 & 0.2 \\ 0.1 & 0.9 \end{pmatrix} &= (\pi_1 \ \pi_2) \\ \rightarrow 0.8\pi_1 + 0.1\pi_2 &= \pi_1 \\ 0.2\pi_1 + 0.9\pi_2 &= \pi_2 \\ \pi_1 + \pi_2 &= 1 \\ \rightarrow \underbrace{\begin{pmatrix} -0.2 & 0.1 \\ 1 & 1 \end{pmatrix}}_A \begin{pmatrix} \pi_1 \\ \pi_2 \end{pmatrix} &= \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ \det(A) &= -0.2 - 0.1 = -0.3 \end{aligned}$$

We calculate each part as

$$\begin{aligned} \pi_1 &= \frac{1}{-0.3} \det \begin{pmatrix} 0 & 0.1 \\ 1 & 1 \end{pmatrix} \\ &= \frac{-0.1}{-0.3} \\ &= \frac{1}{3} \end{aligned}$$

$$\begin{aligned} \pi_2 &= \frac{1}{-0.3} \det \begin{pmatrix} -0.2 & 0 \\ 1 & 1 \end{pmatrix} \\ &= \frac{-0.2}{-0.3} \\ &= \frac{2}{3} \end{aligned}$$

Therefore, the distribution of X_n is given by $X_n \sim (\frac{1}{3} \ \frac{2}{3})$

Problem 5

- (a) In order to show that P is a regular, there must exist some power n such that all entries in the P^n are greater than but not equal to 0.

$$P = \begin{pmatrix} 0.8 & 0.14 & 0.04 & 0.02 \\ 0 & 0.6 & 0.3 & 0.1 \\ 0 & 0 & 0.65 & 0.35 \\ 0.9 & 0 & 0 & 0.1 \end{pmatrix}$$

$$\rightarrow P^3 = \begin{pmatrix} 0.5678 & 0.20972 & 0.15012 & 0.07236 \\ 0.2295 & 0.2286 & 0.35535 & 0.18655 \\ 0.48825 & 0.0441 & 0.287225 & 0.180425 \\ 0.6732 & 0.189 & 0.0936 & 0.0442 \end{pmatrix}$$

Since all entries of P^3 greater than but not equal to 0, P is a regular Markov chain.

- (b) We want to find the steady state distribution π .

$$\begin{pmatrix} \pi_1 & \pi_2 & \pi_3 & \pi_4 \end{pmatrix} \begin{pmatrix} 0.8 & 0.14 & 0.04 & 0.02 \\ 0 & 0.6 & 0.3 & 0.1 \\ 0 & 0 & 0.65 & 0.35 \\ 0.9 & 0 & 0 & 0.1 \end{pmatrix} = \begin{pmatrix} \pi_1 & \pi_2 & \pi_3 & \pi_4 \end{pmatrix}$$

$$\begin{aligned} &\rightarrow 0.8\pi_1 + 0.9\pi_4 = \pi_1 \\ &\quad 0.14\pi_1 + 0.6\pi_2 = \pi_2 \\ &\quad 0.04\pi_1 + 0.3\pi_2 + 0.65\pi_3 = \pi_3 \\ &\quad 0.02\pi_1 + 0.1\pi_2 + 0.35\pi_3 + 0.1\pi_4 = \pi_4 \\ &\quad \pi_1 + \pi_2 + \pi_3 + \pi_4 = 1 \end{aligned}$$

$$\rightarrow \underbrace{\begin{pmatrix} -0.2 & 0 & 0 & 0.9 \\ 0.14 & -0.4 & 0 & 0 \\ 0.04 & 0.3 & -0.35 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix}}_A \begin{pmatrix} \pi_1 \\ \pi_2 \\ \pi_3 \\ \pi_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\det(A) = -0.2503$$

We calculate each part as

$$\begin{aligned} \pi_1 &= \frac{1}{-0.2503} \det \begin{pmatrix} 0 & 0 & 0 & 0.9 \\ 0 & -0.4 & 0 & 0 \\ 0 & 0.3 & -0.35 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix} \\ &= 0.503 \end{aligned}$$

$$\begin{aligned} \pi_2 &= \frac{1}{-0.2503} \det \begin{pmatrix} -0.2 & 0 & 0 & 0.9 \\ 0.14 & 0 & 0 & 0 \\ 0.04 & 0 & -0.35 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix} \\ &= 0.176 \end{aligned}$$

$$\begin{aligned}\pi_3 &= \frac{1}{-0.2503} \det \begin{pmatrix} -0.2 & 0 & 0 & 0.9 \\ 0.14 & -0.4 & 0 & 0 \\ 0.04 & 0.3 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix} \\ &= 0.209\end{aligned}$$

$$\begin{aligned}\pi_4 &= \frac{1}{-0.2503} \det \begin{pmatrix} -0.2 & 0 & 0 & 0 \\ 0.14 & -0.4 & 0 & 0 \\ 0.04 & 0.3 & -0.35 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix} \\ &= 0.112\end{aligned}$$

Therefore, the steady state distribution of X_n is given by $\boldsymbol{\pi} = (\ 0.503 \quad 0.176 \quad 0.209 \quad 0.112 \)$

Problem 8

The Markov Chain is given by the figure below.

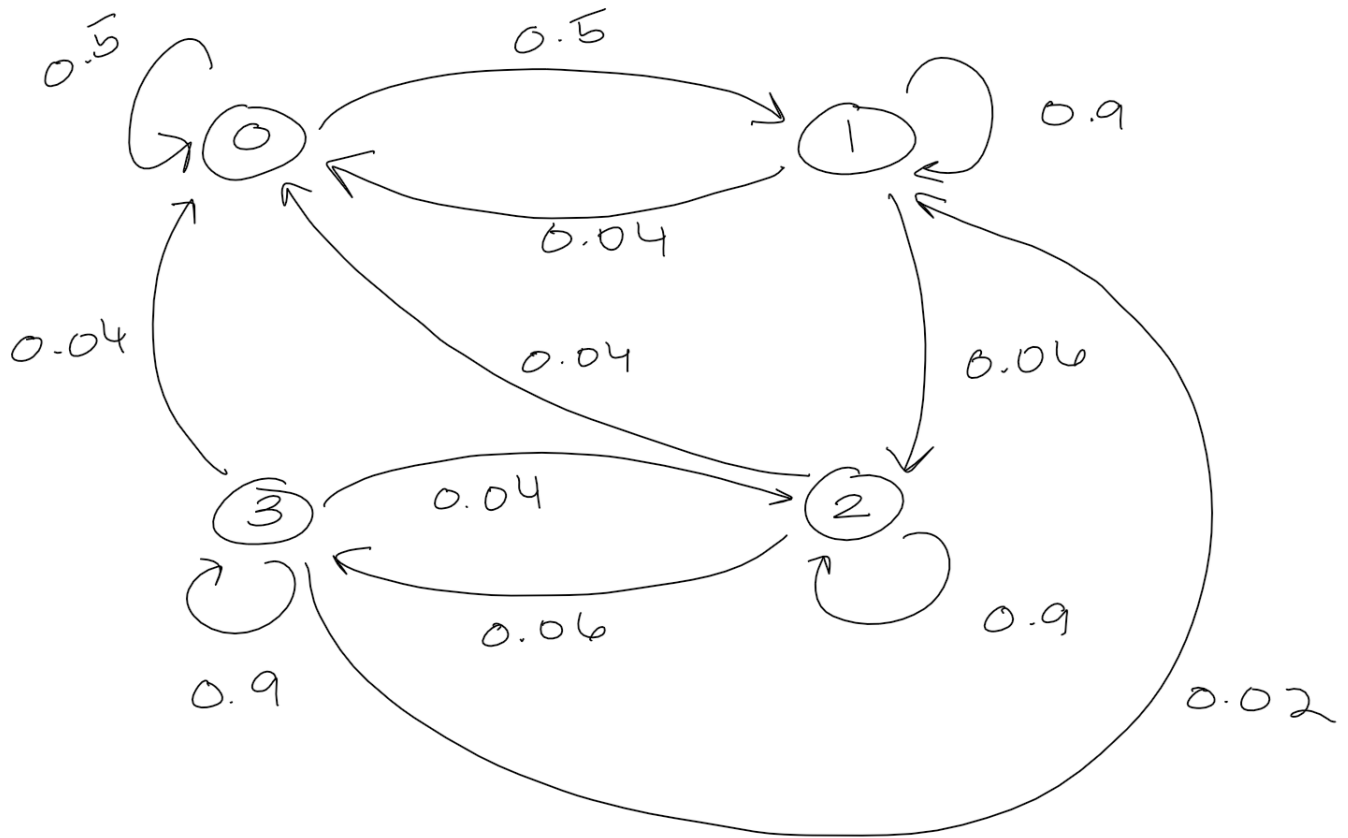


Figure 1: Markov Chain for Lan car reports

The transition matrix P is given below:

$$P = \begin{pmatrix} 0.5 & 0.5 & 0 & 0 \\ 0.04 & 0.9 & 0.06 & 0 \\ 0.04 & 0 & 0.9 & 0.06 \\ 0.04 & 0.02 & 0.04 & 0.9 \end{pmatrix}$$

Problem 9

The Markov Chain is given by the figure below.

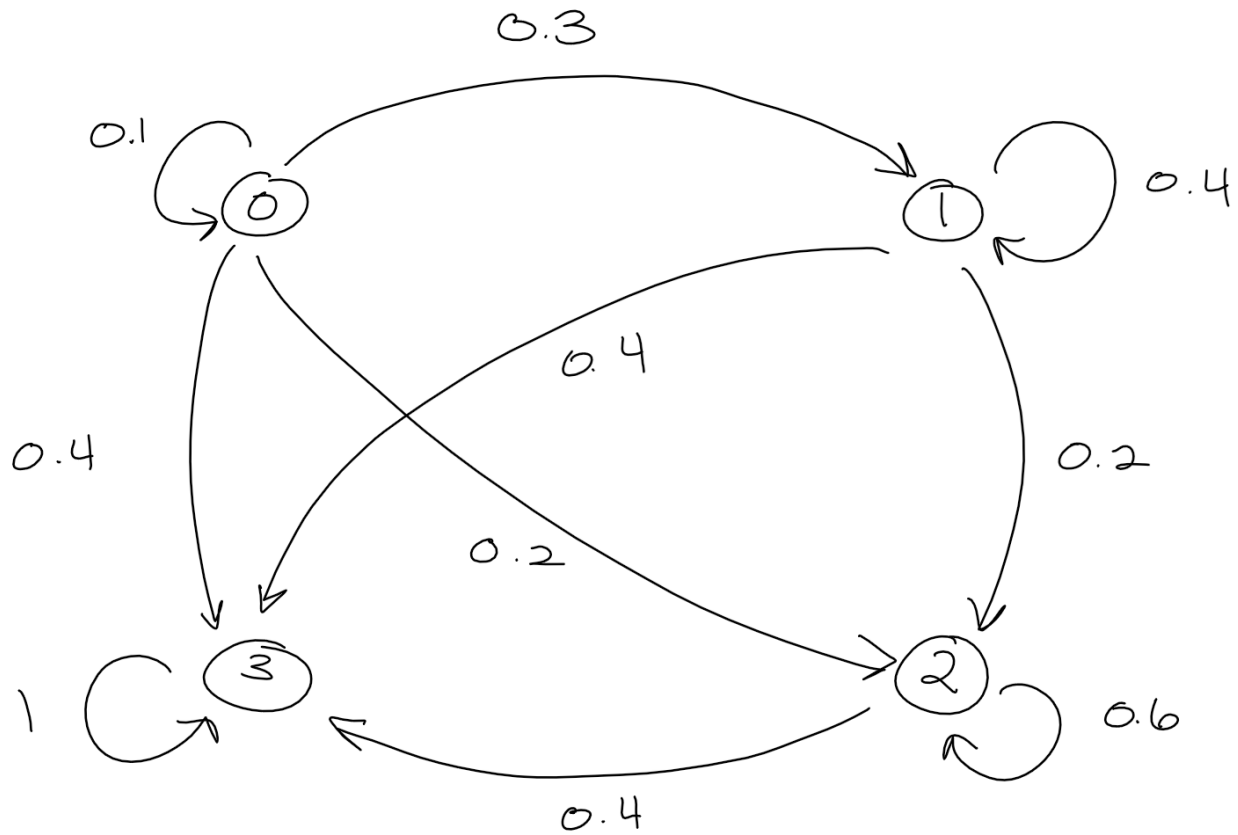


Figure 2: Markov Chain for Problem 9

The transition matrix P is given below:

$$P = \begin{pmatrix} 0.1 & 0.3 & 0.2 & 0.4 \\ 0 & 0.4 & 0.2 & 0.4 \\ 0 & 0 & 0.6 & 0.4 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

Problem 13

- (a) We want to show that the complete conditional for μ is normal and the complete conditional for σ^2 is inverse gamma. We also want to find the corresponding parameters.

First note that,

$$\begin{aligned} p(\mu, \phi | x) &= \frac{p(x, \mu, \phi)}{p(x)} \\ &\propto p(x, \mu, \phi) \end{aligned} \quad (*)$$

since denominator does not involve μ or ϕ .

Second, note that the likelihood function is given by

$$L(\mu, \phi) = (2\pi)^{-n/2} \phi^{n/2} \exp\left(-\frac{\phi}{2} \sum_{i=1}^n (x_i - \mu)^2\right) \quad (**)$$

Third, the joint density of (x, μ, ϕ) is given by

$$p(x, \mu, \phi) = (\text{constant}) \times \phi^{n/2} \exp\left(-\frac{\phi}{2} \sum_{i=1}^n (x_i - \mu)^2\right) \times \exp\left(-\frac{1}{2s}(\mu - m)^2\right) \times \phi^{a-1} \exp(-b\phi) \quad (***)$$

(1) Complete conditional for μ :

$$\begin{aligned} p(\mu | \phi, x) &= \frac{p(\mu, \phi | x)}{p(\phi | x)} \\ &\propto \frac{p(x, \mu, \phi)}{p(\phi | x)} \text{ according to } (*) \\ &\propto p(x, \mu, \phi) \text{ since denominator does not involve } \mu \end{aligned}$$

Hence, we perform some algebra to find the parameters

$$\begin{aligned} p(\mu | \phi, x) &\propto p(x, \mu, \phi) \\ &\propto \phi^{n/2} \exp\left(-\frac{\phi}{2} \sum_{i=1}^n (x_i - \mu)^2\right) \times \exp\left(-\frac{1}{2s}(\mu - m)^2\right) \times \phi^{a-1} \exp(-b\phi) \text{ from } (***) \\ &\propto \exp\left(-\frac{\phi}{2} \sum_{i=1}^n (x_i - \mu)^2\right) \times \exp\left(-\frac{1}{2s}(\mu - m)^2\right) \\ &\propto \exp\left(-\frac{1}{2} \left(\phi \sum x^2 - \phi 2 \sum x\mu + \phi n\mu^2 + \frac{1}{s}\mu^2 - \frac{1}{s}2\mu m + \frac{1}{s}m^2 \right)\right) \\ &\propto \exp\left(-\frac{1}{2} \left(\mu^2 \left(n\phi + \frac{1}{s} \right) + \phi \sum x^2 - \phi 2 \sum x\mu - \frac{1}{s}2\mu m + \frac{1}{s}m^2 \right)\right) \\ &\propto \exp\left(-\frac{1}{2} \left(n\phi + \frac{1}{s} \right) \left(\mu - \frac{1}{\phi + s^{-1}} \left(\phi \sum x + \frac{1}{s}m \right) \right)^2\right) \end{aligned}$$

Hence, $(\mu | \phi, x) \sim N\left(\frac{1}{n\phi + s^{-1}} \left(\phi \sum_{i=1}^n x_i + \frac{1}{s}m \right), n\phi + \frac{1}{s}\right)$

(2) Complete conditional for σ^2 :

$$\begin{aligned}
 p(\phi|\mu, x) &= \frac{p(\mu, \phi|x)}{p(\mu|x)} \\
 &\propto \frac{p(x, \mu, \phi)}{p(\mu|x)} \text{ according to } (*) \\
 &\propto p(x, \mu, \phi) \text{ since denominator does not involve } \phi
 \end{aligned}$$

Hence, we perform some algebra to find the parameters

$$\begin{aligned}
 p(\phi|\mu, x) &\propto p(x, \mu, \phi) \\
 &\propto \phi^{n/2} \exp\left(-\frac{\phi}{2} \sum_{i=1}^n (x_i - \mu)^2\right) \times \exp\left(-\frac{1}{2s}(\mu - m)^2\right) \times \phi^{a-1} \exp(-b\phi) \text{ from } (***) \\
 &\propto \phi^{n/2} \exp\left(-\frac{\phi}{2} \sum_{i=1}^n (x_i - \mu)^2\right) \times \phi^{a-1} \exp(-b\phi) \\
 &\propto \phi^{n/2+a-1} \exp\left(-\phi \left(b + \frac{1}{2} \sum_{i=1}^n (x_i - \mu)^2\right)\right)
 \end{aligned}$$

And so, $(\phi|\mu, x) \sim \text{Gamma}\left(\frac{n}{2} + a - 1, b + \frac{1}{2} \sum_{i=1}^n (x_i - \mu)^2\right)$.

Therefore, $(\sigma^2|\mu, x) \sim \text{InvGamma}\left(\frac{n}{2} + a - 1, \frac{1}{b + \frac{1}{2} \sum_{i=1}^n (x_i - \mu)^2}\right)$

(b)

(c)

(d)