

Introduction to Stochastic Processes

“EXAME DE ÉPOCA ESPECIAL”

2nd. Semester — 2013/14

Duration: 3h00m

2014/07/21 — 9AM, Room C01

- Please justify all your answers.
- This exam has THREE PAGES and SIX GROUPS. The total of points is 40.0.

Group 1 — Introduction to Stochastic Processes

2.5 points

Consider a stochastic process $\{X(t) : t \geq 0\}$ — with stationary and independent increments — and assume that $X(0) = 0$.

- (a) Show that the mean function is equal to $E[X(t)] = \mu t$, where $\mu = E[X(1)]$. (1.0)

Hint: The only solution to the functional equation $f(t+s) = f(t) + f(s)$ is $f(t) = ct$, where $c = f(1)$. This result is relevant to solve lines (a) and (b).

- **Stochastic process**

$\{X(t) : t \geq 0\}$ with stationary and independent increments

- **Initial condition**

$$X(0) = 0$$

- **A property of the mean function**

Let

$$f(t) = E[X(t)]$$

$$\stackrel{X(0)=0}{=} E[X(t) - X(0)].$$

Then, by capitalizing on the stationary increments of this process, we get

$$f(t+s) = E[X(t+s)]$$

$$= E[X(t+s) - X(0)]$$

$$= E\{[X(t+s) - X(s)] + [X(s) - X(0)]\}$$

$$= E[X(t+s) - X(s)] + E[X(s) - X(0)]$$

$$\stackrel{stat. inc.}{=} E[X(t) - X(0)] + E[X(s) - X(0)]$$

$$= f(t) + f(s),$$

for $t, s \geq 0$.

- **Deriving the mean function**

As mentioned in the hint, the only solution to the functional equation $f(t+s) = f(t) + f(s)$ is

$$f(t) = \mu t,$$

where $\mu = f(1) = E[X(1)]$.

QED

- (b) Prove that the variance function is given by $V[X(t)] = \sigma^2 t$, where $\sigma^2 = V[X(1)]$. (1.5)

- **A property of the variance function**

Let

$$g(t) = V[X(t)]$$

$$\stackrel{X(0)=0}{=} V[X(t) - X(0)].$$

Then, by capitalizing on both the stationary and independent increments of this process, we obtain

$$g(t+s) = V[X(t+s)]$$

$$= V[X(t+s) - X(0)]$$

$$= V\{[X(t+s) - X(s)] + [X(s) - X(0)]\}$$

$$\stackrel{indep. inc.}{=} V[X(t+s) - X(s)] + V[X(s) - X(0)]$$

$$\stackrel{stat. inc.}{=} V[X(t) - X(0)] + V[X(s) - X(0)]$$

$$= g(t) + g(s),$$

for $t, s \geq 0$.

- **Deriving the variance function**

Once again, the only solution to the functional equation $g(t+s) = g(t) + g(s)$ is

$$g(t) = \sigma^2 t,$$

where $\sigma^2 = g(1) = V[X(1)]$.

QED

Group 2 — Poisson Processes

9.5 points

1. Arrivals of customers at a supermarket are modeled by a Poisson process with a rate of $\lambda = 10$ customers per minute.

- (a) Let M (resp. N) be the number of customers arriving between 9:00 and 9:10 (resp. 9:30 and 9:35). (1.0)

What is the distribution of $M + N$?

- **Stochastic process**

$$\{N(t) : t \geq 0\} \sim PP(\lambda = 10)$$

$N(t)$ = number of arrivals by time t (time in minutes)

- **Relevant facts**

$$N(t) \sim \text{Poisson}(\lambda t)$$

$\{N(t) : t \geq 0\}$ has stationary and independent increments

- **R.v.**

M = number of customers arriving between 9:00 and 9:10

N = number of customers arriving between 9:30 and 9:35

- **Distributions of M and N**

Due to the stationary increments of the process $\{N(t) : t \geq 0\}$ and the fact that $N(t) \sim \text{Poisson}(10t)$, we can add that:

$$\begin{aligned}
M &= N(9 \times 60 + 10) - N(9 \times 60) \\
&\sim N(9 \times 60 + 10 - 9 \times 60) \\
&\sim N(10) \\
&\sim \text{Poisson}(10 \times 10 = 100) \\
N &= N(9 \times 60 + 35) - N(9 \times 60 + 30) \\
&\sim N(9 \times 60 + 35 - 9 \times 60 - 30) \\
&\sim N(5) \\
&\sim \text{Poisson}(10 \times 5 = 50).
\end{aligned}$$

• **Distribution of $M + N$**

Since M and N refer to the number of arrivals in two non-overlapping time intervals, we can invoke the fact that the process has independent increments to conclude that M and N are independent r.v.

Moreover, since the sum of two independent Poisson r.v. with parameters λ_i , $i = 1, 2$, has a Poisson distribution with parameter $(\lambda_1 + \lambda_2)$, we get

$$M + N \sim \text{Poisson}(100 + 50 = 150).$$

- (b) Suppose that 300 customers arrived during the first 30 minutes. (1.0)

Obtain an approximate value to the probability that at most 200 customers arrived during the first 20 minutes?

• **R.v.**

$$(N(s) \mid N(t) = n) \stackrel{form.}{\sim} \text{Binomial}(n, s/t), \quad 0 < s < t$$

• **Requested probability** (approximate value)

Using the normal approximation to the binomial c.d.f., we obtain

$$\begin{aligned}
P[N(20) \leq 200 \mid N(30) = 300] &= F_{\text{Binomial}(n=300, s/t=20/30)}(200) \\
&\simeq \Phi \left[\frac{200 - 300 \times \frac{2}{3}}{\sqrt{300 \times \frac{2}{3} \times (1 - \frac{2}{3})}} \right] \\
&= \Phi(0) \\
&= 0.5.
\end{aligned}$$

[According to *Mathematica*, $F_{\text{Binomial}(n=300, s/t=20/30)}(200) = 0.521703$.]

- (c) Admit any customer spends a random time (in minutes) in the supermarket with a Weibull distribution with scale parameter $\alpha = 5\sqrt{2}$ (resp. shape parameter $\beta = 2$). (2.0)

Find the probability that there are at least 50 customers still in the supermarket 5 minutes after it opened.

• **R.v.**

S = time spent in the supermarket by a customer

$$S \sim \text{Weibull}(\alpha = 5\sqrt{2}, \beta = 2)$$

• **Non-homogenous Bernoulli splitting**

A customer, who arrived at time s ($0 < s < t$), will be still in the supermarket at time t with probability

$$\begin{aligned}
p(s) &= P(S > t - s) \\
&= \int_{t-s}^{+\infty} f_S(u) du \\
&= \int_{t-s}^{+\infty} \frac{\beta}{\alpha} \left(\frac{x}{\alpha}\right)^{\beta-1} e^{-\left(\frac{x}{\alpha}\right)^\beta} dx \\
&= -e^{-\left(\frac{x}{\alpha}\right)^\beta} \Big|_{t-s}^{+\infty} \\
&= e^{-\left(\frac{t-s}{\alpha}\right)^\beta}.
\end{aligned}$$

Furthermore, the number of customers in this supermarket at time t , $N_{sup}(t)$, results from a non-homogenous Bernoulli splitting of $\{N(t) : t \geq 0\}$. Consequently,

$$N_{sup}(t) \stackrel{form.}{\sim} \text{Poisson} \left(\lambda \int_0^t p(s) ds \right),$$

where

$$\begin{aligned}
\int_0^t p(s) ds &\stackrel{\beta=2}{=} \int_0^t e^{-\left(\frac{t-s}{\alpha}\right)^2} ds \\
&= \sqrt{2\pi} \times \alpha/\sqrt{2} \times \int_0^t \frac{1}{\sqrt{2\pi} \times \alpha/\sqrt{2}} e^{-\frac{(s-t)^2}{2 \times (\alpha/\sqrt{2})^2}} ds \\
&= \sqrt{2\pi} \times \alpha/\sqrt{2} \times \left[F_{N(0, (\alpha/\sqrt{2})^2)}(t) - F_{N(0, (\alpha/\sqrt{2})^2)}(0) \right] \\
&= \sqrt{2\pi} \times \alpha/\sqrt{2} \times \left[\Phi \left(\frac{t-t}{\alpha/\sqrt{2}} \right) - \Phi \left(\frac{0-t}{\alpha/\sqrt{2}} \right) \right] \\
&= \sqrt{2\pi} \times \alpha/\sqrt{2} \times \left[\Phi(0) - \Phi \left(-\frac{t}{\alpha/\sqrt{2}} \right) \right] \\
&\stackrel{\alpha=5\sqrt{2}, t=5}{=} \sqrt{2\pi} \times 5 \times [0.5 - \Phi(-1)] \\
&= \sqrt{2\pi} \times 5 \times [0.5 - 1 + \Phi(1)] \\
&\stackrel{tables}{=} \sqrt{2\pi} \times 5 \times [0.5 - 1 + 0.8413] \\
&\simeq 4.277561.
\end{aligned}$$

• **Requested probability**

Using the normal approximation to the Poisson c.d.f., we obtain

$$\begin{aligned}
P[N_{sup}(t) \geq 50] &\simeq 1 - F_{\text{Poisson}(10 \times 4.277561)}(50 - 1) \\
&\simeq 1 - \Phi \left[\frac{(50 - 1) - 10 \times 4.277561}{\sqrt{10 \times 4.277561}} \right] \\
&\simeq 1 - \Phi(0.95) \\
&\stackrel{tables}{=} 1 - 0.8289 \\
&= 0.1711.
\end{aligned}$$

2. Suppose that the emissions of very rare particles are governed by a non-homogeneous Poisson process with intensity function $\lambda(t) = e^{-t}$, $t \geq 0$.

- (a) Find the probability that no particles were emitted in the first 2 hours and exactly one particle was emitted after those first 2 hours. (1.5)

• **Stochastic process**

$$\{N(t) : t \geq 0\} \sim NHPP$$

$N(t)$ = number of particle emissions by time t

• **Intensity function**

$$\lambda(t) = e^{-t}, t \geq 0$$

• **Mean value function**

$$\begin{aligned} m(t) &= E[N(t)] \\ &= \int_0^t \lambda(s) ds \\ &= \int_0^t e^{-s} ds \\ &= -e^{-s} \Big|_0^t \\ &= 1 - e^{-t}, t \geq 0 \end{aligned}$$

• **Relevant facts**

$$\begin{aligned} N(t) &\sim \text{Poisson}(m(t)) \\ N(t+s) - N(s) &\sim \text{Poisson}(m(t+s) - m(s)) \\ \{N(t) : t \geq 0\} &\text{ has independent increments} \end{aligned}$$

• **Requested probability**

Since $m(t) = 1 - e^{-t} \in [0, 1]$, for $t \geq 0$, we can devise the distribution of $N(+\infty) - N(2)$, the total number of particles emitted after the first 2 hours:

$$N(+\infty) - N(2) \sim \text{Poisson}(m(+\infty) - m(2) = (1 - e^{-\infty}) - (1 - e^{-2}) = e^{-2}).$$

Thus, the requested probability:

$$\begin{aligned} P[N(2) = 0, N(\infty) - N(2) = 1] &\stackrel{\text{indep. inc.}}{=} P[N(2) = 0] \times P[N(\infty) - N(2) = 1] \\ &= \frac{e^{-m(2)} [m(2)]^0}{0!} \\ &\quad \times \frac{e^{-[m(+\infty) - m(2)]} [m(+\infty) - m(2)]^1}{1!} \\ &= e^{-(1 - e^{-2})} \times e^{-e^{-2}} e^{-2} \\ &= e^{-3}. \end{aligned}$$

- (b) Obtain $E[S_1 | N(2) = 0, N(\infty) - N(2) = 1]$. (2.0)

Hint: Recall that $E(X) = \int_0^{+\infty} [1 - F_X(x)] dx$ for any non-negative r.v. X .

• **R.v.**

S_1 = time of the emission of the first particle

• **C.d.f. of $[S_1 | N(2) = 0, N(\infty) - N(2) = 1]$**

For $0 < t < 2$,

$$\begin{aligned} F_{S_1 | N(2)=0, N(\infty)-N(2)=1}(t) &= P[S_1 \leq t | N(2) = 0, N(\infty) - N(2) = 1] \\ &= 0. \end{aligned}$$

Moreover, for $t \geq 2$,

$$\begin{aligned} F_{S_1 | N(2)=0, N(\infty)-N(2)=1}(t) &= P[S_1 \leq t | N(2) = 0, N(\infty) - N(2) = 1] \\ &= P[N(t) \geq 1 | N(2) = 0, N(\infty) - N(2) = 1] \\ &= \frac{P[N(t) \geq 1, N(2) = 0, N(\infty) - N(2) = 1]}{P[N(2) = 0, N(\infty) - N(2) = 1]} \\ &\stackrel{\text{indep. inc.}}{=} \{P[N(2) = 0] \times P[N(t) - N(2) = 1] \\ &\quad \times P[N(\infty) - N(t) = 0]\} \\ &\quad \div \{P[N(2) = 0] \times P[N(\infty) - N(2) = 1]\} \\ &= \frac{P[N(t) - N(2) = 1] \times P[N(\infty) - N(t) = 0]}{P[N(\infty) - N(2) = 1]}, \end{aligned}$$

where

$$\begin{aligned} P[N(t) - N(2) = 1] &= \frac{e^{-[m(t) - m(2)]} [m(t) - m(2)]^1}{1!} \\ &= e^{-[(1 - e^{-t}) - (1 - e^{-2})]} \times (e^{-2} - e^{-t}) \\ &= e^{-(e^{-2} - e^{-t})} \times (e^{-2} - e^{-t}) \\ P[N(\infty) - N(t) = 0] &= \frac{e^{-[m(+\infty) - m(t)]} [m(+\infty) - m(t)]^0}{0!} \\ &= e^{-[1 - (1 - e^{-t})]} \\ &= e^{-e^{-t}} \\ P[N(\infty) - N(2) = 1] &= \frac{e^{-[m(+\infty) - m(2)]} [m(+\infty) - m(2)]^1}{1!} \\ &= e^{-[1 - (1 - e^{-2})]} e^{-2} \\ &= e^{-2 - e^{-2}}. \end{aligned}$$

Consequently, for $t \geq 2$,

$$\begin{aligned} F_{S_1 | N(2)=0, N(\infty)-N(2)=1}(t) &= \frac{e^{-(e^{-2} - e^{-t})} (e^{-2} - e^{-t}) \times e^{-e^{-t}}}{e^{-2 - e^{-2}}} \\ &= 1 - e^{2-t}. \end{aligned}$$

• **Requested conditional expected value**

Since we are dealing with a non-negative r.v.,

$$\begin{aligned} E[S_1 | N(2) = 0, N(\infty) - N(2) = 1] &= \int_0^{+\infty} [1 - F_{S_1 | N(2)=0, N(\infty)-N(2)=1}(t)] dt \\ &= \int_0^2 dt + \int_2^{+\infty} e^{2-t} dt \\ &= 2 + (-e^{2-t}) \Big|_2^{+\infty} \\ &= 2 + 1 \\ &= 3. \end{aligned}$$

3. Suppose the number of claims generated by a portfolio of insurance policies is governed by a conditional Poisson process with random rate Λ (claims per month).

Obtain the autocovariance function of this stochastic process. (2.0)

- **Relevant stochastic process**

$\{N(t) : t \geq 0\} \sim \text{ConditionalPP}(\Lambda)$

$N(t)$ = number of claims up to month t

- **Important**

$\{N(t) : t \geq 0\}$ has stationary increments.

$\{N(t) \mid \Lambda = \lambda\} : t \geq 0\} \sim PP(\lambda)$ and therefore, conditionally on $\{\Lambda = \lambda\}$, we deal with stationary and independent increments. Furthermore,

$$\begin{aligned} (N(t) \mid \Lambda = \lambda) &\sim \text{Poisson}(\lambda t) \\ E[N(t) \mid \Lambda = \lambda] &= \lambda t \\ V[N(t) \mid \Lambda = \lambda] &= \lambda t. \end{aligned}$$

- **Mean value function**

$$\begin{aligned} E[N(t)] &= E\{E[N(t) \mid \Lambda = \lambda]\} \\ &= E(\Lambda t) \\ &= E(\Lambda) \times t \end{aligned}$$

- **Variance function**

$$\begin{aligned} V[N(t)] &= V\{E[N(t) \mid \Lambda = \lambda]\} + E\{V[N(t) \mid \Lambda = \lambda]\} \\ &= V(\Lambda t) + E(\Lambda t) \\ &= V(\Lambda) \times t^2 + E(\Lambda) \times t \end{aligned}$$

- **Autocovariance function**

Please note that, for $0 \leq s < t$,

$$\begin{aligned} E[N(s) \times N(t)] &= E\{N(s) \times [N(t) - N(s) + N(s)]\} \\ &= E\{N(s) \times [N(t) - N(s)]\} + E[N^2(s)] \\ &= E\{N(s) \times [N(t) - N(s)]\} + E[N^2(s)] \\ &= E(E\{N(s) \times [N(t) - N(s)] \mid \Lambda\}) + E[N^2(s)], \end{aligned}$$

where

$$\begin{aligned} E[N^2(s)] &= V[N(s)] + E^2[N(s)] \\ &= V(\Lambda) s^2 + [E(\Lambda) s]^2 \\ &= E(\Lambda^2) s^2 \end{aligned}$$

and the r.v. $E\{N(s) \times [N(t) - N(s)] \mid \Lambda\}$ takes value

$$\begin{aligned} E\{N(s) \times [N(t) - N(s)] \mid \Lambda = \lambda\} &\stackrel{\text{cond. indep. inc}}{=} E[N(s) \mid \Lambda = \lambda] \times E[N(t) - N(s) \mid \Lambda = \lambda] \\ &\stackrel{\text{cond. stat. inc}}{=} E[N(s) \mid \Lambda = \lambda] \times E[N(t - s) \mid \Lambda = \lambda] \\ &= \lambda s \times \lambda (t - s) \\ &= \lambda^2 s(t - s), \end{aligned}$$

with associated p.(d.)f. $f_\Lambda(\lambda)$. Therefore

$$\begin{aligned} E\{N(s) \times [N(t) - N(s)]\} &= E(E\{N(s) \times [N(t) - N(s)] \mid \Lambda\}) \\ &= E[\Lambda^2 s(t - s)] \\ &= E(\Lambda^2) s(t - s) \\ E[N(s) \times N(t)] &= E(\Lambda^2) s(t - s) + E(\Lambda^2) s^2 \\ &= E(\Lambda^2) s t. \end{aligned}$$

Finally, for $0 \leq s < t$,

$$\begin{aligned} \text{cov}(N(s), N(t)) &= E[N(s) \times N(t)] - E[N(s)] \times E[N(t)] \\ &= E(\Lambda^2) s t - E^2(\Lambda) s t \\ &= V(\Lambda) s t. \end{aligned}$$

Group 3 — Renewal Processes

8.0 points

1. Airplanes land at a small airport according to a Poisson process with rate λ (airplanes per hour).

(a) Derive the renewal function $m(t)$ of the renewal process consisting of counting just EVEN (2.5) landings (i.e., the 2nd., 4th., 6th., etc. landings).

Hint: Capitalize on the fact that $\frac{\lambda^2}{s(s+2\lambda)} = \frac{\lambda}{2s} - \frac{\lambda}{2(s+2\lambda)}$.

- **Original stochastic process**

$\{N^*(t) : t \geq 0\} \sim PP(\lambda)$

$N^*(t)$ = number of landings until time t

- **Original inter-renewal times**

$X_i^* \stackrel{i.i.d.}{\sim} X^*, i \in \mathbb{N}$

$X^* \sim \text{Exponential}(\lambda)$

- **Renewal process**

$\{N(t) : t \geq 0\}$

$N(t)$ = number of EVEN landings until time t

- **Inter-renewal times**

$X_i \stackrel{i.i.d.}{\sim} X, i \in \mathbb{N}$

$X \sim \text{Gamma}(2, \lambda)$ (convolution of two indep. exponentially distributed r.v.)

- **Deriving the renewal function**

Since the $X \sim \text{Gamma}(2, \lambda)$, its LST is given by

$$\begin{aligned} \bar{F}(s) &= \int_{0^-}^{+\infty} e^{-sx} dF(x) \\ &= M_X(-s) \\ &\stackrel{\text{form.}}{=} \left(\frac{\lambda}{\lambda + s} \right)^2. \end{aligned}$$

Moreover, the LST of the renewal function can be obtained in terms of \tilde{F} :

$$\begin{aligned}\tilde{m}(s) &\stackrel{\text{form.}}{=} \frac{\tilde{F}(s)}{1 - \tilde{F}(s)} \\ &= \frac{\left(\frac{\lambda}{\lambda+s}\right)^2}{1 - \left(\frac{\lambda}{\lambda+s}\right)^2} \\ &= \frac{\lambda^2}{s(s+2\lambda)} \\ &= \frac{\lambda}{2s} - \frac{\lambda}{2(s+2\lambda)}.\end{aligned}$$

Taking advantage of the LT in the formulae, we obtain

$$\begin{aligned}\frac{dm(t)}{dt} &= LT^{-1}[\tilde{m}(s), t] \\ &= LT^{-1}\left[\frac{\lambda}{2s} - \frac{\lambda}{2(s+2\lambda)}, t\right] \\ &= \frac{\lambda}{2} \times LT^{-1}\left[\frac{1}{s}, t\right] + \frac{\lambda}{2} \times LT^{-1}\left[\frac{1}{(s+2\lambda)^1}, t\right] \\ &= \frac{\lambda}{2} \times 1 + \frac{\lambda}{2} \times \frac{t^{1-1} e^{-2\lambda t}}{(1-1)!} \\ &= \frac{\lambda}{2} + \frac{\lambda e^{-2\lambda t}}{2} \\ m(t) &= \int_0^t \left(\frac{\lambda}{2} + \frac{\lambda e^{-2\lambda x}}{2}\right) dx \\ &= \left(\frac{\lambda x}{2} - \frac{e^{-2\lambda x}}{4}\right)\Big|_0^t \\ &= \frac{\lambda t}{2} - \frac{1 - e^{-2\lambda t}}{4}, t \geq 0.\end{aligned}$$

(b) Show that the renewal function obtained in (a) verifies the elementary renewal theorem. (1.0)

• **Verification of the elementary renewal theorem (ERT)**

Let $\mu = E(X) = E[\text{Gamma}(2, \lambda)] = \frac{2}{\lambda}$. Then

$$\begin{aligned}\lim_{t \rightarrow +\infty} \frac{m(t)}{t} &= \lim_{t \rightarrow +\infty} \left(\frac{\frac{\lambda t}{2} - \frac{1 - e^{-2\lambda t}}{4}}{t}\right) \\ &= \frac{\lambda}{2} - \frac{1}{+\infty} \\ &= \frac{1}{\mu},\end{aligned}$$

hence the ERT is verified.

(c) Obtain an approximate value to the probability that the number of EVEN landings exceeds 10 in the first day, when $\lambda = 1$. (1.5)

• **Inter-renewal times**

$$\begin{aligned}X_i &\stackrel{i.i.d.}{\sim} X, i \in \mathbb{N} \\ X &\sim \text{Gamma}(2, \lambda = 1) \\ \mu &= E(X) \stackrel{\text{form. 2}}{=} 2\end{aligned}$$

$$\sigma^2 = V(X) \stackrel{\text{form. 2}}{=} 2$$

• **Requested approximate probability**

$$\begin{aligned}P[N(t) > n] &= 1 - P[N(t) < n + 1] \\ &\stackrel{\text{form}}{\simeq} 1 - \Phi\left[\frac{(n+1) - t/\mu}{\sqrt{t\sigma^2/\mu^3}}\right] \\ &\stackrel{t=24h, n=10}{=} 1 - \Phi\left[\frac{(10+1) - 24/2}{\sqrt{24 \times 2/2^3}}\right] \\ &\simeq 1 - \Phi(-0.29) \\ &= \Phi(0.29) \\ &\stackrel{\text{tables}}{=} 0.6141.\end{aligned}$$

• **Obs.** — This a rough approximation of the exact value of the requested probability:

$$1 - P[N(t) \leq 9] = 1 - P[N^*(t) \leq 2 \times 9 + 1] \stackrel{\text{table}}{=} 1 - 0.1803 = 0.8197.$$

2. The time (in minutes) Clotilde takes to get to the top of a sky piste is a r.v. with c.d.f. F (3.0) and expected value μ_F , whereas the duration of the descent (also in minutes) is uniformly distributed in the interval $(0,7)$. Admit she decides to rent a new pair of skis for c monetary units whenever the descent lasts more than τ minutes.

Find the value of τ that minimizes the money spent in ski rentals per time unit in the long-run. Comment this result.

• **Renewal process**

$$\{N(t) : t \geq 0\}$$

$N(t)$ = number of completed cycles of ascent/descent by time t

• **R.v.**

A_n = duration of the n^{th} ascent to the top of the piste

$$A_n \stackrel{i.i.d.}{\sim} A \sim F, n \in \mathbb{N}$$

D_n = duration of the n^{th} descent from the top of the piste

$$D_n \stackrel{i.i.d.}{\sim} D \sim \text{Uniform}(0, 7), n \in \mathbb{N}$$

• **Inter-renewal times**

$$X_n \stackrel{i.i.d.}{\sim} X, n \in \mathbb{N}$$

$$X = A + D$$

• **Reward renewal process**

$$\{R(t) = \sum_{n=1}^{N(t)} R_n : t \geq 0\}$$

$R(t)$ = total spent in ski rentals until time t

$$R_n = \begin{cases} c, & \text{if } D_n > \tau \text{ (i.e., if the } n^{\text{th}} \text{ descent lasted more than } \tau \text{ minutes)} \\ 0, & \text{otherwise} \end{cases}$$

$$(X_n, R_n) \stackrel{i.i.d.}{\sim} (X, R), n \in \mathbb{N}$$

- **Expected inter-renewal time**

$$E(X) = E(A) + E(D) \stackrel{\text{form.}}{=} \mu_F + \frac{0+7}{2} = \mu_F + \frac{7}{2}$$

- **Expected amount spent (per completed cycle of ascent/descent)**

For $0 < \tau \leq 7$, we have

$$\begin{aligned} E(R) &= c \times P(D > \tau) + 0 \times P(D \leq \tau) \\ &= c \times \int_{\tau}^{+\infty} f_D(u) du \\ &= c \times \int_{\tau}^7 \frac{1}{7} du \\ &= c \times \frac{7 - \tau}{7} \end{aligned}$$

- **Amount spent (per completed cycle of ascent/descent) per time unit in the long-run**

Since $E(X), E(R) < +\infty$, we can add that

$$\frac{R(t)}{t} \xrightarrow{w.p.1} \frac{E(R)}{E(X)},$$

where

$$\begin{aligned} \frac{E(R)}{E(X)} &= h(\tau) \\ &= \frac{c(7-\tau)}{\mu_F + \frac{7}{2}} \\ &= \frac{2c(7-\tau)}{14\mu_F + 49}, 0 < \tau \leq 7. \end{aligned}$$

- **Minimizing the amount spent per time unit in the long-run**

Since $h(\tau)$, $0 < \tau \leq 7$, is a decreasing function of τ in the interval $(0, 7]$, the value of τ that minimizes $h(\tau)$ is equal to $\tau^* = 7$.

- **Comment**

Not only τ^* does not depend on the ski rental (c) or on μ_F , but also its value means that Clotilde should never replace her skis if she is willing to minimize the amount spent per time unit in the long-run in ski rentals.

Group 4 — Renewal Processes (cont'd)

1.5 points

The number of inspections by a supervisor to an industrial plant is governed by a delayed renewal process such that:

- the first inspection time (in years) follows an exponential distribution with unit mean;
- the subsequent inter-inspection times follow an exponential distribution with expected value equal to 0.5.

Derive the c.d.f. of S_2 , the time of the second inspection.

- **Delayed renewal process**

$$\{N_D(t) : t \geq 0\}$$

$N_D(t)$ = number of inspections done by time t

- **Inter-renewal times**

X_i independent r.v., $i \in \mathbb{N}$

$X_1 \sim \text{Exponential}(1)$

$X_i \stackrel{i.i.d.}{\sim} \text{Exponential}(0.5^{-1} = 2)$

- **Important**

$G(x) = P(X_1 \leq x) = 1 - e^{-x}$, for $x \geq 0$ (0, otherwise)

$F(x) = P(X_i \leq x) = 1 - e^{-2x}$, for $x \geq 0$ and $i \in \mathbb{N} \setminus \{1\}$ (0, otherwise)

- **R.v.**

S_2 = time of the second inspection

- **C.d.f. of S_2**

$$\begin{aligned} P(S_n \leq t) &\stackrel{\text{form.}}{=} (G \star F_{n-1})(t) \\ &\stackrel{n=2}{=} \int_0^t G(t-x) dF(x) \\ &= \int_0^t [1 - e^{-(t-x)}] \times 2e^{-2x} dx \\ &= \int_0^t 2e^{-2x} dx - 2e^{-t} \int_0^t e^{-x} dx \\ &= -e^{-2x} \Big|_0^t + 2e^{-t} \times (-e^{-x}) \Big|_0^t \\ &= (1 - e^{-2t}) + 2e^{-t}(e^{-t} - 1) \\ &= 1 + e^{-2t} - 2e^{-t}. \end{aligned}$$

Group 5 — Discrete time Markov chains

9.5 points

1. A study of occupational mobility of families across generations was conducted after WWII. Three occupation levels were identified:

- *upper* level (executive, managerial, high administrative, professional) — state 1;
- *middle* level (high grade supervisor, non-manual, skilled manual) — state 2;
- *lower* level (semi-skilled or unskilled) — state 3.

Transition probabilities from generation to generation were estimated to be

$$\mathbf{P} = \begin{bmatrix} 0.45 & 0.48 & 0.07 \\ 0.05 & 0.70 & 0.25 \\ 0.01 & 0.50 & 0.49 \end{bmatrix}.$$

(a) Determine $f_{ij}^n = P(X_n = j, X_{n-1} \neq j, \dots, X_1 \neq j \mid X_0 = i)$, for $i, n = 1, 2, 3$ and $j = 1$. (2.0)

- **DTMC**

$\{X_n : n \in \mathbb{N}_0\}$

$X_n =$ level of the family at the n^{th} generation

- **State space**

$\mathcal{S} = \{1, 2, 3\}$

1 = *upper* level

2 = *middle* level

3 = *lower* level

- **TPM**

$$\mathbf{P} = \begin{bmatrix} 0.45 & 0.48 & 0.07 \\ 0.05 & 0.70 & 0.25 \\ 0.01 & 0.50 & 0.49 \end{bmatrix}$$

- **Requested probabilities**

Let:

(i) $f_{ij}^n = P(X_n = j, X_{n-1} \neq j, \dots, X_1 \neq j \mid X_0 = i)$ be the probability of reaching state j for the first time starting from state i , for $i, j \in \mathcal{S}$ and $n \in \mathbb{N}$;

(ii) $\underline{f}_j^n = [f_{ij}^n]_{i \in \mathcal{S}}$ be the associated vector for fixed $j \in \mathcal{S}$ and $n \in \mathbb{N}$.

According to the formulae,

$$\underline{f}_j^n = \begin{cases} \underline{f}_j^1 = [P_{ij}]_{i \in \mathcal{S}}, & n = 1 \\ {}^{(j)}\mathbf{P} \times \underline{f}_j^{n-1} = [{}^{(j)}\mathbf{P}]^{n-1} \times \underline{f}_j^1, & n = 2, 3, \dots, \end{cases}$$

where ${}^{(j)}\mathbf{P}$ is obtained by setting all the entries of the j^{th} column of \mathbf{P} equal to 0.

When $j = 1$, we successively get

$${}^{(1)}\mathbf{P} = \begin{bmatrix} 0 & 0.48 & 0.07 \\ 0 & 0.70 & 0.25 \\ 0 & 0.50 & 0.49 \end{bmatrix}$$

$$\underline{f}_1^1 = [P_{i1}]_{i \in \mathcal{S}}$$

$$= \begin{bmatrix} 0.45 \\ 0.05 \\ 0.01 \end{bmatrix}$$

$$\underline{f}_1^2 = {}^{(1)}\mathbf{P} \times \underline{f}_1^1$$

$$= \begin{bmatrix} 0 & 0.48 & 0.07 \\ 0 & 0.70 & 0.25 \\ 0 & 0.50 & 0.49 \end{bmatrix} \times \begin{bmatrix} 0.45 \\ 0.05 \\ 0.01 \end{bmatrix}$$

$$= \begin{bmatrix} 0.0247 \\ 0.0375 \\ 0.0299 \end{bmatrix}$$

$$\begin{aligned} \underline{f}_1^3 &= {}^{(1)}\mathbf{P} \times \underline{f}_1^2 \\ &= \begin{bmatrix} 0 & 0.48 & 0.07 \\ 0 & 0.70 & 0.25 \\ 0 & 0.50 & 0.49 \end{bmatrix} \times \begin{bmatrix} 0.0247 \\ 0.0375 \\ 0.0299 \end{bmatrix} \\ &= \begin{bmatrix} 0.020093 \\ 0.033725 \\ 0.033401 \end{bmatrix}. \end{aligned}$$

(b) What is the long-run percentage of generations that a family spends in state 3?¹ (2.0)

- **Important**

We are dealing with an irreducible DTMC with finite state space. Hence, all states are positive recurrent[, by Prop. 3.55]. Furthermore, the DTMC seems aperiodic.

- **Stationary distribution**

Since the DTMC is irreducible positive recurrent and aperiodic we can add that

$$\lim_{n \rightarrow +\infty} P_{ij}^n = \pi_j > 0, \quad i, j \in \mathcal{S},$$

where $\{\pi_j : j \in \mathcal{S}\}$ is the unique stationary distribution and satisfies the following system of equations:

$$\begin{cases} \pi_j = \sum_{i \in \mathcal{S}} \pi_i P_{ij}, \quad j \in \mathcal{S} \\ \sum_{j \in \mathcal{S}} \pi_j = 1. \end{cases}$$

Equivalently [(see Prop. 3.68)], the row vector denoting the stationary distribution, $\underline{\pi} = [\pi_j]_{j \in \mathcal{S}}$, is given by

$$\underline{\pi} = \underline{1} \times (\mathbf{I} - \mathbf{P} + \mathbf{ONE})^{-1},$$

where:

$\underline{1} = [1 \quad \dots \quad 1]$ a row vector with $\#\mathcal{S}$ ones;

\mathbf{I} = identity matrix with rank $\#\mathcal{S}$;

$\mathbf{P} = [P_{ij}]_{i, j \in \mathcal{S}}$ is the TPM;

\mathbf{ONE} is the $\#\mathcal{S} \times \#\mathcal{S}$ matrix all of whose entries are equal to 1.

By capitalizing on the inverse in the footnote, we obtain

$$\begin{aligned} \underline{\pi} &= \underline{1} \times (\mathbf{I} - \mathbf{P} + \mathbf{ONE})^{-1} \\ &= \underline{1} \times \left(\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} - \begin{bmatrix} 0.45 & 0.48 & 0.07 \\ 0.05 & 0.70 & 0.25 \\ 0.01 & 0.50 & 0.49 \end{bmatrix} + \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \right)^{-1} \\ &= \underline{1} \times \begin{bmatrix} 1.55 & 0.52 & 0.93 \\ 0.95 & 1.3 & 0.75 \\ 0.99 & 0.5 & 1.51 \end{bmatrix}^{-1} \end{aligned}$$

¹The following result may be useful: $\begin{bmatrix} 1.55 & 0.52 & 0.93 \\ 0.95 & 1.3 & 0.75 \\ 0.99 & 0.5 & 1.51 \end{bmatrix}^{-1} \simeq \begin{bmatrix} 1.179441 & -0.237819 & -0.608289 \\ -0.513963 & 1.054516 & -0.207219 \\ -0.603090 & -0.193256 & 1.129679 \end{bmatrix}$.

$$\begin{aligned}\underline{\pi} &\simeq [1 \ 1 \ 1] \times \begin{bmatrix} 1.179441 & -0.237819 & -0.608289 \\ -0.513963 & 1.054516 & -0.207219 \\ -0.603090 & -0.193256 & 1.129679 \end{bmatrix} \\ &= [0.062389 \ 0.623440 \ 0.314171].\end{aligned}$$

Thus, the long-run percentage of generations that a family spends in state 3 is equal to [the sum of the entries of the 3rd. column of $(\mathbf{I} - \mathbf{P} + \mathbf{ONE})^{-1}$]:

$$\pi_3 \simeq 0.314171.$$

- (c) Determine the expected number of generations it takes a family to reach state 1, starting from state 3. **(2.0)**

• **Initial/present state**

$$X_0 = i$$

• **Important**

To obtain the expected number of generations until a family reaches state 1, given $X_0 = i$, we have to consider another DTMC where state 1 is absorbing. The associated TPM is

$$\mathbf{P}' = \begin{bmatrix} 1 & 0 & 0 \\ 0.05 & 0.70 & 0.25 \\ 0.01 & 0.50 & 0.49 \end{bmatrix}.$$

• **Requested expected value**

Let

$$\mathbf{Q} = \begin{bmatrix} 0.70 & 0.25 \\ 0.50 & 0.49 \end{bmatrix}$$

be the substochastic matrix governing the transitions between the states in $T = \{2, 3\}$, the class of transient states of this new DTMC, and

$$\tau = \inf\{n \in \mathbb{N}_0 : X_n \notin T\}$$

be the number of generations until a family reaches state 1. Then, by capitalizing on the fact that

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix},$$

we obtain

$$\begin{aligned}[E(\tau | X_0 = i)]_{i \in T} &= (\mathbf{I} - \mathbf{Q})^{-1} \times \underline{\mathbf{1}} \\ &= \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 0.70 & 0.25 \\ 0.50 & 0.49 \end{bmatrix} \right)^{-1} \times \underline{\mathbf{1}} \\ &= \begin{bmatrix} 0.3 & -0.25 \\ -0.5 & 0.51 \end{bmatrix}^{-1} \times \underline{\mathbf{1}}\end{aligned}$$

$$\begin{aligned}[E(\tau | X_0 = i)]_{i \in T} &= \frac{1}{0.3 \times 0.51 - (-0.25) \times (-0.5)} \begin{bmatrix} 0.51 & 0.25 \\ 0.5 & 0.3 \end{bmatrix} \times \underline{\mathbf{1}} \\ &= \frac{1}{0.028} \begin{bmatrix} 0.76 \\ 0.8 \end{bmatrix} \\ &= \begin{bmatrix} 27.142857 \\ 28.571429 \end{bmatrix}.\end{aligned}$$

Thus, the requested expected value equals

$$E(\tau | X_0 = 3) = 28.571429.$$

2. The following model can be used to describe the number of women (mothers and daughters) in a given area. The number of mothers is a r.v. $X \sim \text{Poisson}(\lambda)$. Independently of the others, every mother gives birth to a $\text{Poisson}(\mu)$ -distributed number of daughters.

Let W be the total number of women (mothers and daughters) in the area. Show that:

- (a) the p.g.f. of W is given by $e^{-\lambda[1-s e^{-\mu(1-s)}]}$; **(2.5)**

• **Auxiliary r.v.**

X = number of mothers

$X \sim \text{Poisson}(\lambda)$

Z_l = number of daughters from mother l

$Z_l \stackrel{i.i.d.}{\sim} Z, l \in \mathbb{N}$

• **Important r.v.**

$W = X + \sum_{i=1}^X Z_i$

= total number of women (mothers and daughters)

• **Requested p.g.f.**

$$\begin{aligned}P_W(s) &= E(s^W) \\ &= E\left(s^{X + \sum_{i=1}^X Z_i}\right) \\ &= E\left[E\left(s^{X + \sum_{i=1}^X Z_i} \mid X\right)\right],\end{aligned}$$

where the r.v. $E\left(s^{X + \sum_{i=1}^X Z_i} \mid X\right)$ takes value

$$\begin{aligned}E\left(s^{X + \sum_{i=1}^X Z_i} \mid X = x\right) &\stackrel{X \perp\!\!\!\perp Z_i}{=} s^x E\left(s^{\sum_{i=1}^x Z_i}\right) \\ &\stackrel{Z_i \stackrel{i.i.d.}{\sim} Z}{=} s^x [E(s^Z)]^x \\ &= [s P_Z(s)]^x,\end{aligned}$$

with probability $P(X = x)$. Consequently,

$$\begin{aligned}P_W(s) &= E\left\{[s P_Z(s)]^X\right\} \\ &\stackrel{form.}{=} e^{-\lambda[1-s P_Z(s)]} \\ &\stackrel{form.}{=} e^{-\lambda[1-s e^{-\mu(1-s)}]}.\end{aligned}$$

(b) $E(Z) = \lambda(1 + \mu)$. (1.0)

• Requested expected value

$$\begin{aligned} E(W) &\stackrel{\text{form.}}{=} \left. \frac{dP_W(s)}{ds} \right|_{s=1} \\ &= \left. \frac{d e^{-\lambda[1-s] e^{-\mu(1-s)}}}{ds} \right|_{s=1} \\ &= \left. -\frac{d \lambda[1-s] e^{-\mu(1-s)}}{ds} \right|_{s=1} \times e^{-\lambda[1-s] e^{-\mu(1-s)}} \Big|_{s=1} \\ &= \lambda \times \left. \frac{d s e^{-\mu(1-s)}}{ds} \right|_{s=1} \times 1 \\ &= \lambda \times [e^{-\mu(1-s)} + s \mu e^{-\mu(1-s)}] \Big|_{s=1} \\ &= \lambda(1 + \mu). \end{aligned}$$

QED

Group 6 — Continuous time Markov chains

9.0 points

1. Passengers arrive at (resp. trains depart from) a train station according to a Poisson process with rate equal to 20 passengers per minute (resp. 12 trains per hour). Let $X(t)$ be the number of passengers at the station at time t waiting for the next train to depart.

(a) Draw the rate diagram and derive the infinitesimal generator \mathbf{R} of the CTMC $\{X(t) : t \geq 0\}$. (1.5)

Hint: Even though $\{X(t) : t \geq 0\}$ is not a birth-death process, it might be useful to interpret an arrival of a passenger as a *birth* and note that a departure of a train implies the “*death*” of all passengers at the train station.

• CTMC

$\{X(t) : t \geq 0\}$

$X(t)$ = no. of passengers at the train station at time t waiting for the next train to depart

• Auxiliary r.v.

B = time (in minutes) until the arrival of the next passenger

$B \sim \text{Exponential}(\lambda = 20)$

D = time (in minutes) until the departure of the next train

$D \sim \text{Exponential}(\mu = 12/60 = 0.2)$

• State space

$\mathcal{S} = \mathbb{N}_0$

• Possible transitions (embedded DTMC)

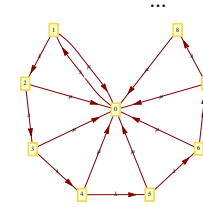
If we interpret an arrival of a passenger as a *birth* and note that a departure of a train implies the “*death*” of all passengers at the train station, the embedded DTMC transitions from:

– state i to state 0 ($i \in \mathbb{N}$) — if a train departs before the next passenger arrives;

– state i to state $i + 1$ ($i \in \mathbb{N}_0$) — if a passenger arrives before the next train departs.

• Rate diagram

Recall that the rate diagram of a CTMC is a directed graph — with no loops — in which each state is represented by a node and there is an arc going from node i to node j (if $q_{ij} > 0$) with q_{ij} written on it.



• Infinitesimal generator

This matrix has entries

$$r_{ij} = \begin{cases} q_{ij}, & i \neq j \\ -\nu_i = -\sum_{m \in \mathcal{S}} q_{im}, & j = i \end{cases}$$

and in this case it is equal to

$$\mathbf{R} = [r_{ij}]_{i,j \in \mathcal{S}} = \begin{bmatrix} -\lambda & \lambda & 0 & 0 & 0 & 0 & \cdots \\ \mu & -(\lambda + \mu) & \lambda & 0 & 0 & 0 & \cdots \\ \mu & 0 & -(\lambda + \mu) & \lambda & 0 & 0 & \cdots \\ \mu & 0 & 0 & -(\lambda + \mu) & \lambda & 0 & \cdots \\ \vdots & \vdots & \vdots & \ddots & \ddots & \ddots & \ddots \end{bmatrix}$$

(b) Write the Kolmogorov’s forward differential equations in terms of $P_{0j}(t) = P[X(t) = j \mid X(0) = 0]$, for $j \in \mathbb{N}_0$. (1.0)

• Kolmogorov’s forward differential equations

These can be written in matrix form:

$$\frac{d\mathbf{P}(t)}{dt} = \left[\frac{dP_{ij}(t)}{dt} \right]_{i,j \in \mathcal{S}} \stackrel{\text{form.}}{=} \mathbf{P}(t) \times \mathbf{R}.$$

Since $i = 0$, we are only interested in the first row of the previous matrix

$$\left[\frac{dP_{0j}(t)}{dt} \right]_{j \in \mathcal{S}} = [P_{0j}(t)]_{j \in \mathcal{S}} \times \mathbf{R}.$$

Hence the following Kolmogorov’s forward differential equations:

$$\begin{aligned} \frac{dP_{00}(t)}{dt} &= -\lambda P_{00}(t) + \mu \sum_{m=1}^{+\infty} P_{0m}(t) \\ &= -\lambda P_{00}(t) + \mu [1 - P_{00}(t)] \\ &= -(\lambda + \mu) P_{00}(t) + \mu \end{aligned}$$

$$\begin{aligned}
\frac{dP_{01}(t)}{dt} &= \lambda P_{00}(t) - (\lambda + \mu) P_{01}(t) \\
\frac{dP_{02}(t)}{dt} &= \lambda P_{01}(t) - (\lambda + \mu) P_{02}(t) \\
&\vdots \\
\frac{dP_{0j}(t)}{dt} &= \lambda P_{0j-1}(t) - (\lambda + \mu) P_{0j}(t), j \in \mathbb{N}.
\end{aligned}$$

(c) Show that the equilibrium probabilities $P_j = \lim_{t \rightarrow +\infty} P_{0j}(t) = (1/101) \times (100/101)^j$, $j \in \mathbb{N}_0$. (2.5)

Hint: Recall that $\underline{P} \times \mathbf{R} = \underline{0}$ and $\sum_{j=0}^{+\infty} P_j = 1$, where $\underline{P} = [P_j]_{j \in \mathbb{N}_0}$ is the row vector of the equilibrium probabilities.

• **Equilibrium probabilities** $P_j = \lim_{t \rightarrow +\infty} P_{0j}(t)$

Let $\underline{P} = [P_j]_{j \in \mathbb{N}_0}$ be the row vector of the equilibrium probabilities. Then these probabilities can be obtained by solving

$$\underline{P} \times \mathbf{R} = \underline{0},$$

subjected to $\sum_{j=0}^{+\infty} P_j = 1$. For instance, the first set of equations (corresponding to $i = 0$) lead to

$$\begin{cases}
-\lambda P_0 + \mu \sum_{m=1}^{+\infty} P_m = 0 \\
\lambda P_0 - (\lambda + \mu) P_1 = 0 \\
\lambda P_1 - (\lambda + \mu) P_2 = 0 \\
\vdots \\
\lambda P_{j-1} - (\lambda + \mu) P_j = 0, j \in \mathbb{N}
\end{cases}$$

$$\begin{cases}
-\lambda P_0 + \mu(1 - P_0) = 0 \\
P_1 = \frac{\lambda}{\lambda + \mu} P_0 \\
P_2 = \frac{\lambda}{\lambda + \mu} P_1 \\
\vdots \\
P_j = \frac{\lambda}{\lambda + \mu} P_{j-1}, j \in \mathbb{N}.
\end{cases}$$

$$\begin{cases}
P_0 = \frac{\mu}{\lambda + \mu} \\
P_1 = \frac{\lambda}{\lambda + \mu} \times \frac{\mu}{\lambda + \mu} \\
P_2 = \left(\frac{\lambda}{\lambda + \mu}\right)^2 \times \frac{\mu}{\lambda + \mu} \\
\vdots \\
P_j = \left(\frac{\lambda}{\lambda + \mu}\right)^j \times \frac{\mu}{\lambda + \mu}, j \in \mathbb{N}.
\end{cases}$$

Equivalently,

$$P_j \stackrel{\lambda=20, \mu=0.2}{=} (1/101) \times (100/101)^j, j \in \mathbb{N}_0.$$

2. An average of 80 jobs are submitted to a university computer center per hour. Assuming that the computer service is modeled as an $M/M/1$ queueing system:

(a) what should be the service rate if the average turnaround time (period from the submission a job until getting this job done) is to be smaller than 10 minutes? (1.5)

• **Birth-death queueing system**

$M/M/1$ with $\lambda = 80$ (jobs per hour)

• **Traffic intensity/ergodicity condition**

$$\rho = \frac{\lambda}{\mu} = \frac{80}{\mu} < 1$$

• **Performance measure (in the long-run)**

W_s = turnaround time (in hours)

$$E(W_s) \stackrel{form.}{=} \frac{1}{\mu(1-\rho)}$$

• **Requested service rate**

We have to deal with $\mu > 80$ and

$$\mu : E(W_s) < \frac{10}{60} \text{ (i.e., 10 minutes)}$$

$$\frac{1}{\mu \left(1 - \frac{\lambda}{\mu}\right)} < \frac{1}{6}$$

$$\frac{1}{\mu - \lambda} < \frac{1}{6}$$

$$\mu > \frac{6}{1} + \lambda$$

$$\mu > 86 \text{ (jobs per hour).}$$

(b) find the average number of jobs found in the system, when the service rate is equal to $\mu = 1.5$ jobs per minute; (1.0)

• **Traffic intensity/ergodicity condition**

$$\rho = \frac{\lambda}{\mu} = \frac{80}{1.5 \times 60} = \frac{8}{9} < 1$$

• **Performance measure (in the long-run)**

L_s = number of customers in the drive-in banking service

• **Requested expected value**

$$\begin{aligned}
E(L_s) &\stackrel{form.}{=} \frac{\rho}{1 - \rho} \\
&= \frac{\frac{8}{9}}{1 - \frac{8}{9}} \\
&= 8.
\end{aligned}$$

(c) calculate the probability that the turnaround time exceeds 10 minutes, considering the same service rate as in (b). (1.5)

• **Performance measure (in the long-run)**

W_s = turnaround time (in hours)

$W_s \sim \text{Exponential}(\mu(1 - \rho))$

• **Requested probability**

$$\begin{aligned}
P(W_s > t) &= e^{-\mu(1-\rho)t} \\
&\stackrel{t=\frac{10}{60}, etc.}{=} e^{-90(1-8/9)t} \\
&= e^{-\frac{10}{3}}.
\end{aligned}$$