

SA402 Fall 2023 · Final Exam · List of Formulas

Markov chains

- n -step transition probabilities: $\mathbf{P}^{(n)} = \mathbf{P}^n$
- n -step state probabilities: $\mathbf{q}^{(n)\top} = \mathbf{q}^\top \mathbf{P}^n$
- First-passage probabilities, starting in \mathcal{A} and ending in \mathcal{B} in the n th step: $\mathbf{F}_{\mathcal{A}\mathcal{B}}^{(n)} = \mathbf{P}_{\mathcal{A}\mathcal{A}}^{n-1} \mathbf{P}_{\mathcal{A}\mathcal{B}}$
- Steady-state probabilities for recurrent class \mathcal{R} :

$$\begin{aligned} \pi_{\mathcal{R}}^\top \mathbf{P}_{\mathcal{R}\mathcal{R}} &= \pi_{\mathcal{R}}^\top \\ \pi_{\mathcal{R}}^\top \mathbf{1} &= 1 \end{aligned}$$

- Absorption probabilities for transient states \mathcal{T} and absorbing state $\mathcal{R} = \{j\}$:

$$\alpha_{\mathcal{T}\mathcal{R}} = \mathbf{N} \mathbf{P}_{\mathcal{T}\mathcal{R}} \quad \text{where} \quad \mathbf{N} = (\mathbf{I} - \mathbf{P}_{\mathcal{T}\mathcal{T}})^{-1}$$

- Expected time to absorption for transient states \mathcal{T} : $\boldsymbol{\mu}_{\mathcal{T}} = \mathbf{N} \mathbf{1}$

Poisson processes

Useful distributions:

	$X \sim \text{Poisson}(\mu)$	$X \sim \text{Exponential}(\lambda)$	$X \sim \text{Erlang}(n, \lambda)$
pmf / pdf	$p_X(a) = \begin{cases} \frac{e^{-\mu} \mu^a}{a!} & \text{if } a = 0, 1, 2, \dots \\ 0 & \text{o/w} \end{cases}$	$f_X(a) = \begin{cases} \lambda e^{-\lambda a} & \text{if } a \geq 0 \\ 0 & \text{o/w} \end{cases}$	$f_X(a) = \begin{cases} \frac{\lambda (\lambda a)^{n-1} e^{-\lambda a}}{(n-1)!} & \text{if } a \geq 0 \\ 0 & \text{o/w} \end{cases}$
cdf	$F_X(a) = \sum_{k=0}^{\lfloor a \rfloor} \frac{e^{-\mu} \mu^k}{k!}$	$F_X(a) = \begin{cases} 1 - e^{-\lambda a} & \text{if } a \geq 0 \\ 0 & \text{o/w} \end{cases}$	$F_X(a) = \begin{cases} 1 - \sum_{k=0}^{n-1} \frac{e^{-\lambda a} (\lambda a)^k}{k!} & \text{if } a \geq 0 \\ 0 & \text{o/w} \end{cases}$
expected value	$E[X] = \mu$	$E[X] = \frac{1}{\lambda}$	$E[X] = \frac{n}{\lambda}$
variance	$\text{Var}(x) = \mu$	$\text{Var}(X) = \frac{1}{\lambda^2}$	$\text{Var}(X) = \frac{n}{\lambda^2}$

Markov processes

- Steady-state probabilities:

$$\begin{aligned} \pi^\top \mathbf{G} &= \mathbf{0}^\top \\ \pi^\top \mathbf{1} &= 1 \end{aligned}$$

Birth-death processes

- Steady-state probabilities:

$$d_0 = 1 \quad d_j = \frac{\lambda_0 \lambda_1 \cdots \lambda_{j-1}}{\mu_1 \mu_2 \cdots \mu_j} \quad \text{for } j = 1, 2, \dots \quad D = \sum_{i=0}^{\infty} d_i \quad \pi_j = \frac{d_j}{D} \quad \text{for } j = 0, 1, 2, \dots$$

- Expected number of customers in the system: $\ell = \sum_{n=0}^{\infty} n \pi_n$

- Expected number of customers in the queue, s parallel servers: $\ell_q = \sum_{n=s+1}^{\infty} (n-s) \pi_n$

- Effective arrival rate: $\lambda_{\text{eff}} = \sum_{i=0}^{\infty} \lambda_i \pi_i$

- Little's law (system-wide): $\ell = \lambda_{\text{eff}} w$

- Little's law (queue only): $\ell_q = \lambda_{\text{eff}} w_q$

Standard queueing models

M/M/ ∞ :

- Steady-state probabilities: $\pi_j = \Pr\{L = j\}$ where $L \sim \text{Poisson}(\lambda/\mu)$

M/M/ s :

- Steady-state probabilities:

$$\rho = \frac{\lambda}{s\mu} \quad \pi_0 = \left[\left(\sum_{j=0}^s \frac{(s\rho)^j}{j!} \right) + \frac{s^s \rho^{s+1}}{s!(1-\rho)} \right]^{-1} \quad \pi_j = \begin{cases} \frac{(\lambda/\mu)^j}{j!} \pi_0 & \text{for } j = 1, 2, \dots, s \\ \frac{(\lambda/\mu)^j}{s! s^{j-s}} \pi_0 & \text{for } j = s+1, s+2, \dots \end{cases}$$

- Expected number of customers in queue: $\ell_q = \frac{\pi_s \rho}{(1-\rho)^2}$

- Expected number of customers in the system: $\ell = \ell_q + \frac{\lambda}{\mu}$

G/G/ s :

- Whitt's approximation:

G = generic interarrival time random variable with rate $\lambda = \frac{1}{E[G]}$

X = generic service time random variable with rate $\mu = \frac{1}{E[X]}$

$$\varepsilon_G = \frac{\text{Var}[G]}{E[G]^2} \quad \varepsilon_X = \frac{\text{Var}[X]}{E[X]^2} \quad \hat{w}_q \approx \frac{\varepsilon_G + \varepsilon_X}{2} w_q$$