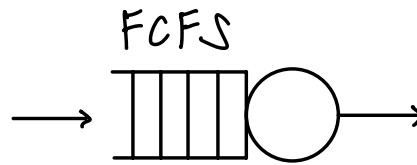


2.2 slowdown

(a) $\lambda = \frac{1}{2}$ job/sec

$$S = \begin{cases} 1 & \text{with probability } \frac{3}{4} \\ 2 & \text{otherwise } 1 - \frac{3}{4} = \frac{1}{4} \end{cases}$$



$$E[T] = \frac{29}{12}$$

$$E[\text{slowdown}] = ?$$

$$\text{slowdown}(j) = \frac{\tau(j)}{s(j)}$$

→ resp. t.
→ serv. t.

$$E[S] = 1 \cdot \frac{3}{4} + 2 \cdot \frac{1}{4} = \frac{3}{4} + \frac{2}{4} = \frac{5}{4}$$

$$\rho = \lambda \cdot E[S] = \frac{1}{2} \times \frac{5}{4} = \frac{5}{8}, \quad \frac{5}{8} < 1$$

$$E[\text{slowdown}] = \frac{E[T]}{E[S]} = \frac{\frac{29}{12}}{\frac{5}{4}} = \frac{29}{12} \times \frac{4}{5} = \frac{29}{15} \#$$

(b) If the service order were SJF, this method wouldn't work directly. As SJF prioritizes shorter job, altering the response time distribution. Thus, using the simple ratio of mean response time to mean service time would be inaccurate.

6.2 Simplified Power Usage in Server Farms

We have 10 r.v., each representing the time for which a server is on for a specific job. X_1, X_2, \dots, X_{10} .

The total time that all server are on is $X = X_1 + X_2 + \dots + X_{10}$.

Each variable X_i is uniformly distributed between 1 and 9 sec.

PDF for X_i is:

$$p(x) = \frac{1}{8} \text{ for } 1 \leq x \leq 9$$

the range of the distribution is 8 sec.
 $p(x) = \frac{1}{b-a}$ for $a \leq x \leq b$

The expected value of uniform distributed r.v. X_i over an interval $[a, b]$ is given by $E(X_i) = \frac{a+b}{2}$.

$$E(X_i) = \frac{1+9}{2} = 5 \text{ seconds}$$

Since there are 10 jobs, each with an expected time of 5 seconds, the total time expected $E(X)$ for all 10 server is:

$$E(X) = 10 \times 5 = 50 \text{ seconds}$$

Each consumes power at a rate of $P = 240$ watts when it is on. The average power consumed over time can be calculated by

$$\text{Average Power consumption} = 240 \times 50 = \underline{12000 \text{ watt.}}$$

8.1 Solving for Limiting Distributions

CPU instruction (C)

Memory instruction (M)

User interaction instruction (U)

transition matrix

$$P = \begin{matrix} & \begin{matrix} C & M & U \end{matrix} \\ \begin{matrix} C \\ M \\ U \end{matrix} & \begin{pmatrix} 0.7 & 0.2 & 0.1 \\ 0.8 & 0.1 & 0.1 \\ 0.9 & 0.1 & 0 \end{pmatrix} \end{matrix}$$

The stationary distribution (π_C, π_M, π_U) satisfies the equations:

$$\pi_C = \pi_C(0.7) + \pi_M(0.8) + \pi_U(0.9)$$

$$\pi_M = \pi_C(0.2) + \pi_M(0.1) + \pi_U(0.1)$$

$$\pi_U = \pi_C(0.1) + \pi_M(0.1) + \pi_U(0)$$

We have the normalization equation: $1 = \pi_C + \pi_M + \pi_U$

$$\pi_C = 0.7\pi_C + 0.8\pi_M + (0.1\pi_C + 0.1\pi_M)(0.9) = 0.79\pi_C + 0.89\pi_M$$

$$\Rightarrow 0.21\pi_C - 0.89\pi_M = 0 \Rightarrow \pi_M = \frac{0.21}{0.89}\pi_C$$

$$\pi_M = 0.2\pi_C + 0.1\pi_M + (0.1\pi_C + 0.1\pi_M)(0.1) = 0.21\pi_C + 0.11\pi_M$$

$$\Rightarrow 0.89\pi_M - 0.21\pi_C = 0 \Rightarrow \pi_M = \frac{0.21}{0.89}\pi_C$$

$$\pi_C + \frac{0.21}{0.89}\pi_C + (0.1\pi_C + 0.1 \cdot \frac{0.21}{0.89}\pi_C) = 1$$

$$\pi_C = 0.74, \pi_M = 0.17, \pi_U = 0.09$$

This means the program spends approximately 74% of the time in C, 17% in M, and 9% in U.

8.6 Threshold Queue

(a) transition matrix P given:

$$P = \begin{array}{c|ccccccc} & j & 0 & 1 & 2 & \dots & T & T+1 \\ \hline i & & & & & & & \\ 0 & & 0.4 & 0.6 & & & & \\ 1 & & 0.4 & & 0.6 & & & \\ 2 & & & 0.4 & & & & \\ \vdots & & & & 0.4 & & & \\ & & & & & \ddots & & \\ & & & & & & 0.6 & \\ T & & & & & & 0.2 & 0.4 \\ T+1 & & & & & & 0.6 & 0.4 \\ & & & & & & & 0.6 \\ & & & & & & & 0.6 \end{array}$$

$N < T$:

$$P(N \rightarrow N+1) = 0.6$$

$$P(N \rightarrow N-1) = 0.4$$

$N \geq T$:

$$P(N \rightarrow N+1) = 0.4$$

$$P(N \rightarrow N-1) = 0.6$$

transit probability

$$\alpha_n = P(N=n \rightarrow N=n+1)$$

$$\beta_n = P(N=n \rightarrow N=n-1)$$

$n < T$:

$$\alpha_n = 0.6$$

$$\beta_n = 0.4$$

$n \geq T$:

$$\alpha_n = 0.4$$

$$\beta_n = 0.6$$

$$n \leq T, \quad \frac{0.6}{0.4} = \frac{3}{2}, \quad \pi_n = \left(\frac{3}{2}\right)^n \pi_0$$

$$\pi_T = \pi_0 \left(\frac{3}{2}\right)^T$$

$$n > T, \quad \frac{0.4}{0.6} = \frac{2}{3}, \quad \pi_n = \left(\frac{2}{3}\right)^{n-T} \pi_T = \left(\frac{2}{3}\right)^{n-T} \pi_0 \left(\frac{3}{2}\right)^T$$

$$\sum_{n=0}^{\infty} \pi_n = 1, \quad S_0: n=0 \sim T \quad S_2: n=T+1 \sim \infty$$

$$S_1 = \sum_{n=0}^T \pi_n = \pi_0 \sum_{n=0}^T \left(\frac{3}{2}\right)^n = \pi_0 \frac{\left(\frac{3}{2}\right)^{T+1} - 1}{\frac{3}{2} - 1} = 2\pi_0 \left[\left(\frac{3}{2}\right)^{T+1} - 1\right] \quad \boxed{\frac{r^{n+1} - 1}{r - 1}}$$

$$S_2 = \sum_{n=T+1}^{\infty} \pi_n = \pi_0 \left(\frac{3}{2}\right)^T \sum_{m=1}^{\infty} \left(\frac{2}{3}\right)^m = \pi_0 \left(\frac{3}{2}\right)^T \left(\frac{\frac{2}{3}}{1 - \frac{2}{3}}\right) = 2\pi_0 \left(\frac{3}{2}\right)^T \quad \boxed{\frac{r}{1-r}}$$

$$S = S_1 + S_2 = 2\pi_0 \left[\left(\frac{3}{2}\right)^{T+1} - 1 + \left(\frac{3}{2}\right)^T\right]$$

$$2\pi_0 \left[\left(\frac{3}{2}\right)^{T+1} - 1 + \left(\frac{3}{2}\right)^T\right] = 1$$

$$2\pi_0 \left[\frac{3}{2} \cdot \left(\frac{3}{2}\right)^T - 1 + \left(\frac{3}{2}\right)^T\right] = 1$$

$$2\pi_0 \left[\frac{5}{2} \cdot \left(\frac{3}{2}\right)^T - 1\right] = 1$$

$$\pi_0 = \frac{1}{2 \left[\frac{5}{2} \cdot \left(\frac{3}{2}\right)^T - 1\right]}$$

$$= \frac{1}{5 \left(\frac{3}{2}\right)^T - 2}$$

$$n \leq T: \quad \pi_n = \frac{\left(\frac{3}{2}\right)^n}{2 \left[\frac{5}{2} \cdot \left(\frac{3}{2}\right)^T - 1\right]}$$

$$n > T: \quad \pi_n = \frac{\left(\frac{3}{2}\right)^T \cdot \left(\frac{2}{3}\right)^{n-T}}{2 \left[\frac{5}{2} \cdot \left(\frac{3}{2}\right)^T - 1\right]}$$

$$(b) \quad E[N] = \sum_{n=0}^{\infty} n \pi_n = \sum_{n=0}^T n \pi_n + \sum_{n=T+1}^{\infty} n \pi_n = E_1 + E_2$$

$$E_1 = \sum_{n=0}^T n \pi_n = \pi_0 \sum_{n=0}^T n \left(\frac{3}{2}\right)^n$$

$$S_1 = \sum_{n=0}^{\infty} n \left(\frac{3}{2}\right)^n = \frac{\frac{3}{2} [T \cdot \left(\frac{3}{2}\right)^{T+1} - (T+1) \cdot \left(\frac{3}{2}\right)^T + 1]}{\left(\frac{3}{2} - 1\right)^2 \approx 0.25}$$

$$= 6 [T \cdot \left(\frac{3}{2}\right)^{T+1} - (T+1) \cdot \left(\frac{3}{2}\right)^T + 1]$$

$$E_1 = \pi_0 \times S_1 = 6 \pi_0 [T \cdot \left(\frac{3}{2}\right)^{T+1} - (T+1) \cdot \left(\frac{3}{2}\right)^T + 1]$$

$$n > T, \quad k = n - T$$

$$E_2 = \sum_{n=T+1}^{\infty} n \pi_n = \pi_0 \left(\frac{3}{2}\right)^T \sum_{k=1}^{\infty} (k+T) \left(\frac{2}{3}\right)^k$$

$$= \pi_0 \left(\frac{3}{2}\right)^T \left[\sum_{k=1}^{\infty} k \left(\frac{2}{3}\right)^k + T \sum_{k=1}^{\infty} \left(\frac{2}{3}\right)^k \right]$$

$$\sum_{k=1}^{\infty} k \left(\frac{2}{3}\right)^k = \frac{\frac{2}{3}}{\left(1 - \frac{2}{3}\right)^2} = \frac{\frac{2}{3}}{\left(\frac{1}{3}\right)^2} = \frac{\frac{2}{3}}{\frac{1}{9}} = 6$$

$$\sum_{k=1}^{\infty} \left(\frac{2}{3}\right)^k = \frac{\frac{2}{3}}{1 - \frac{2}{3}} = 2$$

$$E_2 = \pi_0 \left(\frac{3}{2}\right)^T (6 + 2T)$$

$$E[N] = E_1 + E_2$$

$$= 6\pi_0 \left[T \cdot \left(\frac{3}{2}\right)^{T+1} - (T+1) \cdot \left(\frac{3}{2}\right)^T + 1 \right] + \pi_0 \left(\frac{3}{2}\right)^T (6+2T)$$

$$6\pi_0 \left[T \cdot \left(\frac{3}{2}\right)^{T+1} - (T+1) \cdot \left(\frac{3}{2}\right)^T + 1 \right] = 6\pi_0 \left[T \cdot \frac{3}{2} \cdot \left(\frac{3}{2}\right)^T - (T+1) \cdot \left(\frac{3}{2}\right)^T + 1 \right]$$

$$= 6\pi_0 \left(\frac{3}{2}\right)^T \left[\boxed{\frac{3}{2}T - (T+1)} + \frac{1}{\left(\frac{3}{2}\right)^T} \right]$$

$$\Downarrow$$

$$\frac{3}{2}T - T - 1 = \frac{1}{2}T - 1$$

$$E[N] = 6\pi_0 \left(\frac{3}{2}\right)^T \left(\frac{1}{2}T - 1 + \frac{1}{\left(\frac{3}{2}\right)^T} \right) + \pi_0 \left(\frac{3}{2}\right)^T (6+2T)$$

$$= \pi_0 \left(\frac{3}{2}\right)^T \left[6 \left(\frac{1}{2}T - 1 + \frac{1}{\left(\frac{3}{2}\right)^T} \right) + 6 + 2T \right]$$

$$6 \left(\frac{1}{2}T - 1 + \frac{1}{\left(\frac{3}{2}\right)^T} \right) + 6 + 2T = 3T - 6 + \frac{6}{\left(\frac{3}{2}\right)^T} + 6 + 2T$$

$$= 5T + \frac{6}{\left(\frac{3}{2}\right)^T}$$

$$E[N] = \pi_0 \left(\frac{3}{2}\right)^T \left(5T + \frac{6}{\left(\frac{3}{2}\right)^T} \right) = \pi_0 \left(5T \left(\frac{3}{2}\right)^T + 6 \right)$$

$$E[N] = \frac{1}{5 \left(\frac{3}{2}\right)^T - 2} \left(5T \left(\frac{3}{2}\right)^T + 6 \right)$$

#

(c)

$$T=0$$

$$E[N] = \frac{1}{5\left(\frac{3}{2}\right)^T - 2} \left(5T\left(\frac{3}{2}\right)^T + 6 \right)$$

↓

$$5 - 2 = 3$$

$$\underline{E[N] = 2} \#$$

10.3 Time to empty

(a)

1. Increase by 1 packet with a $p = 0.4$
2. Decrease by 1 packet with a $p = 0.6$

In Markov chain, the system has *memoryless property*, meaning the expected time from state 2 to state 1 is same as the expected time from state 1 to state 0.

$$T_{2,1} = E[T_{1,0}]$$



$$E[T_{1,0}] = 0.6 \cdot \overset{\text{Time}}{1} + 0.4 \cdot (\overset{\text{Time}}{1} + E[T_{1,0} + T_{2,1}])$$

$$\Rightarrow E[T_{1,0}] = 0.6 \cdot 1 + 0.4 \cdot (1 + E[T_{1,0}])$$

$$\Rightarrow E[T_{1,0}] = 0.6 + 0.4 + 0.4 \times E[T_{1,0}]$$

$$\Rightarrow E[T_{1,0}] - 0.4 \times E[T_{1,0}] = 1$$

$$\Rightarrow 0.6 E[T_{1,0}] = 1$$

$$\Rightarrow \underline{E[T_{1,0}] = \frac{1}{0.6} = \frac{5}{3}} \#$$

the expected time from state 1 to state 0 is $\frac{5}{3}$.

11.3 Double Exponential

$$X \sim \text{Exp}(\mu)$$

$$Y = 2X$$

$$\text{p.d.f. } f_X(x) = \begin{cases} \mu e^{-\mu x} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

$$\text{c.d.f. } F_X(x) = \begin{cases} 1 - e^{-\mu x} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

$$F_Y(y) = P(Y \leq y) = P(2X \leq y) = P(X \leq \frac{y}{2}) = F_X(\frac{y}{2})$$

$$F_Y(y) = 1 - e^{-\mu(\frac{y}{2})} = 1 - e^{-y(\frac{\mu}{2})}$$

$$\underline{Y \sim \text{Exp}(\frac{\mu}{2})} \#$$



Doubling all the sizes changes the exponential distribution from rate μ to rate μ divided by 2 ($\frac{\mu}{2}$), the new job sizes still follow an exponential distribution but with half the original rate.

11.5 Practice with Definition of Poisson Process

(a) (i)

$$E[\text{Green}] = \lambda \times p = 50 \times 0.05 = 2.5 \text{ packets/sec}$$

$$E[\text{Yellow}] = \lambda \times q = 50 \times 0.95 = 47.5 \text{ packets/sec}$$

In a poisson process, the arrivals of different types of events are independent. Therefore, even if an unusually high number of green packets is observed, the expected number of yellow packets remain at 47.5 #.

(a) (ii)

$$Y \sim \text{Poisson}(47.5)$$

$$P\{N(t+s) - N(s) = n\} = \frac{e^{-\lambda t} (\lambda t)^n}{n!}, \quad n = 0, 1, 2, \dots$$

$$P\{N(t) = n\} = \frac{e^{-\lambda t} (\lambda t)^n}{n!}, \quad n = 0, 1, 2, \dots$$

$$= \frac{e^{-47.5 \times 1} (47.5 \times 1)^{200}}{200!}$$

The probability is extremely low, almost zero. #

(b) red: $\lambda_1 = 30$ packets/sec
black: $\lambda_2 = 10$ packets/sec

Assume total of $N = 60$ packets arrive in one second.

$$\lambda = \lambda_1 + \lambda_2 = 40 \text{ packets/sec}$$

In a poisson process, given N events occurring within time t , these event is uniformly distributed in time.

red packet success Prob.: green packet success Prob.:

$$p = \frac{\lambda_1}{\lambda} = \frac{30}{40} = 0.75, \quad q = 1 - p = 0.25$$

$$X \sim \text{Binomial}(60, 0.75)$$

$$P\{X = k\} = \binom{n}{k} p^k q^{n-k}$$

$$= \binom{60}{40} (0.75)^{40} (0.25)^{20}$$

$$= \underline{0.03834} \#$$

(c)

100 packets have arrive within 30 sec.

Given that $N=10$ events occur within time T , these events are uniformly distributed in time.

$$p = \frac{t}{T} = \frac{10}{30} = \frac{1}{3}, \quad q = 1 - \frac{1}{3} = \frac{2}{3}$$

$$N(10) \sim \text{Binomial}(100, \frac{1}{3})$$

$$\begin{aligned} P\{S_n = k\} &= \binom{n}{k} p^k q^{n-k} \\ &= \binom{100}{20} \left(\frac{1}{3}\right)^{20} \left(\frac{2}{3}\right)^{80} \\ &= \underline{0.0013} \# \end{aligned}$$

11.7 Malware and Honeypots

(a) $X \sim \text{Exp}(\mu)$, $0 < s < t$

$$\begin{aligned} \text{c.d.f. : } F_X(x) &= P(X \leq x) = 1 - e^{-\mu x} \\ &= P(X \leq t-s) = 1 - e^{-\mu(t-s)} \end{aligned}$$

$$\begin{aligned} \text{(b) } P &= \frac{1}{t} \int_0^t (1 - e^{-\mu(t-s)}) ds \quad u = t-s, du = -ds \\ &= \frac{1}{t} \int_t^0 (1 - e^{-\mu u}) (-du) = \frac{1}{t} \int_0^t (1 - e^{-\mu u}) du \end{aligned}$$

$$\int_0^t (1 - e^{-\mu u}) du = \int_0^t 1 du - \int_0^t e^{-\mu u} du$$

$$= t - \left[-\frac{1}{\mu} e^{-\mu u} \right]_0^t = t - \left(-\frac{1}{\mu} e^{-\mu t} + \frac{1}{\mu} e^0 \right)$$

$$= t + \frac{1}{\mu} e^{-\mu t} - \frac{1}{\mu}$$

$$P = \frac{1}{t} \left(t + \frac{1}{\mu} e^{-\mu t} - \frac{1}{\mu} \right)$$

$$= 1 - \frac{1}{\mu t} + \frac{1}{\mu t} e^{-\mu t}$$

$$= 1 - \frac{1}{\mu t} (1 - e^{-\mu t})$$

$$= 1 - \frac{1 - e^{-\mu t}}{\mu t}$$

(c)

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

$$E[N_1(t)] \times p = \lambda t p$$

$$\lambda = \frac{E[N_1(t)]}{t_p} \approx \frac{N_1(t)}{t_p} \quad \#$$

(d)

$$N(t) = N_1(t) + N_2(t)$$

$$E[N_2(t)] = E[N(t)] - E[N_1(t)]$$

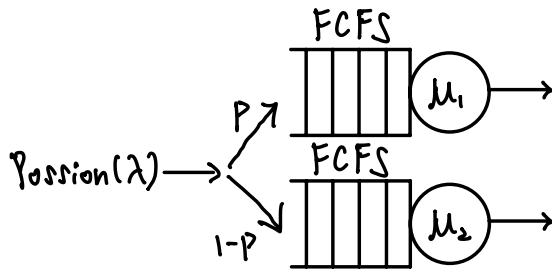
$$= \lambda t - \lambda t p$$

$$= \lambda t (1 - p)$$

$$E[N_2(t)] = \left(\frac{N_1(t)}{t_p} \right) t (1 - p)$$

$$= N_1(t) \frac{1 - p}{p} \quad \#$$

13.2 Server Farm



A fraction p of the jobs go to Server 1, which has a service rate of μ_1 . And the other fraction $1-p$ go to Server 2, which has a service rate of μ_2 .

mean response time formula for an M/M/1 queue:

$$E[T] = \frac{1}{\mu - \lambda}$$

$$\text{Server 1} \Rightarrow \lambda_1 = p\lambda$$

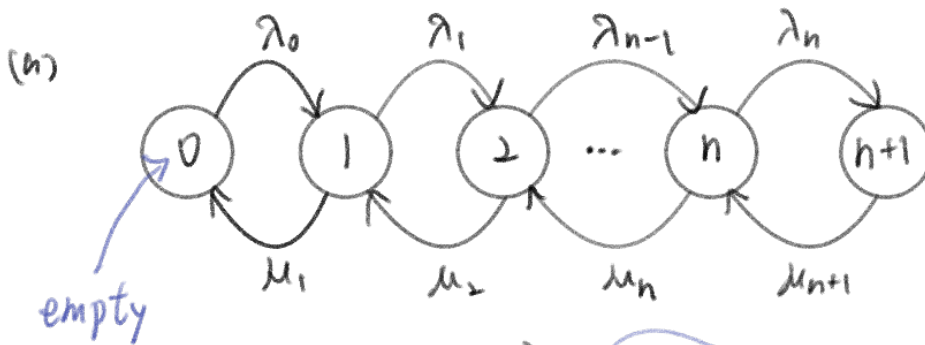
$$\text{Server 2} \Rightarrow \lambda_2 = (1-p)\lambda$$

$$E[T_1] = \frac{1}{\mu_1 - p\lambda}, \quad E[T_2] = \frac{1}{\mu_2 - (1-p)\lambda}$$

$$E[T] = p \cdot E[T_1] + (1-p) \cdot E[T_2]$$

$$E[T] = \frac{p}{\mu_1 - p\lambda} + \frac{1-p}{\mu_2 - (1-p)\lambda} \quad \#$$

13.10 Busy Period in $M/M/1$



(b)
$$E[B] = \frac{1}{\lambda + \mu} + \left(\frac{\lambda}{\lambda + \mu} \right) E[B_2]$$

\downarrow If it is arrived, the remaining busy period is expected.
 The expected time from State 1 to next event.

(c)
$$E[B_2] = \frac{1}{\mu} + E[B]$$

$$E[B] = E[S] + E[N] \times E[B]$$

$$\Rightarrow E[B] = \frac{1}{\mu} + \left(\frac{\lambda}{\mu} \right) E[B]$$

$$\Rightarrow E[B] - \left(\frac{\lambda}{\mu} \right) E[B] = \frac{1}{\mu}$$

$$\Rightarrow E[B] \left(1 - \frac{\lambda}{\mu} \right) = \frac{1}{\mu}$$

$$\Rightarrow E[B] \left(\frac{\mu - \lambda}{\mu} \right) = \frac{1}{\mu}$$

$$\Rightarrow E[B] = \frac{1}{\mu} \times \frac{\mu}{\mu - \lambda} = \frac{1}{\mu - \lambda}$$

$E[S] = \frac{1}{\mu}$
 $E[N] = \lambda E[S] = \frac{\lambda}{\mu}$

This result show that when $\lambda < \mu$, the expected value of the busy period is similar to the exponential distribution of the parameter $\mu - \lambda$.