# **Probability Review** Identities

$$P(A \cap B) = P(A)P(B)$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(A) = P(A|B)P(B) + P(A|\bar{B})P(\bar{B})$$

$$f(x) = \frac{dF(x)}{dx}, \int_{-\infty}^{\infty} f(x) = 1$$

$$P(a < X < b) = \int_{a}^{b} f(x)dx$$

$$F(y) = \int_{-\infty}^{y} f(x)dx$$

$$P(X < x) = F(X)$$

$$\mathbb{E}[X^{n}] = \int_{-\infty}^{\infty} x^{n}f(x)dx$$

$$Var(X) = \mathbb{E}[X^{2}] - \mathbb{E}[X]$$

#### Uniform Distribution: $X \sim uniform(a, b)$

$$f(x) = \begin{cases} \frac{1}{b-a}, & \text{if } a < x < b\\ 0, & \text{otherwise} \end{cases}$$

$$F(x) = \begin{cases} 0, & \text{if } x < a\\ \frac{x-a}{b-a}, & \text{if } a \le x \le b\\ 1, & \text{if } x > b \end{cases}$$

$$\mathbb{E}[X] = \frac{a+b}{2} \qquad \qquad Var[X] = \frac{(b-a)}{1}$$

**Exponential Distribution:**  $X \sim exp(\lambda)$  $f(x) = \lambda e^{-\lambda x}$  $F(x) = 1 - \lambda e^{-\lambda x}$  $Var[X] = \frac{1}{\lambda^2}$  $\mathbb{E}[X] = \frac{1}{\lambda}$ P(X > s + t | X > s) = P(X > t)[Memoryless]

#### **Poisson Distribution:** $X \sim pois(\lambda)$

 $P(N(t) = n) = \frac{(\lambda t)^t}{1} e^{-\lambda t}$  $\mathbb{E}[X] = Var(X)$ 

## **Poisson Process**

$$N(0) = 0$$
  

$$f(x) = \lambda e^{-\lambda x}$$
  

$$P(X > s + t | X > s) = P(X > t)$$
  

$$\lambda = \sum_{i=1}^{n} \lambda_i, Y = \left(\sum_{i=1}^{n} X_i\right) \sim \text{pois}(\lambda)$$
  

$$X \sim pois(\lambda), X = [X_1, X_2]$$
  

$$\implies X_{1,2} \sim pois(\frac{\lambda}{2})$$

# **Performance Analysis**

## Identities

 $A_i(t)$  [Number of arrivals] Ē)  $C_i(t)$  [Completions]  $B_i(t)$  [Busy time]  $S_i(t) = \frac{B_i(t)}{C_i(t)}$  [Avg process time]  $D_i$  [Processing time of cycle]  $V_i(t)$  [Visits to device]  $\lim_{t \to \infty} \frac{A_i(t)}{t} = \lim_{t \to \infty} \frac{C_i(t)}{t}$ pdf) N(t) = A(t) - C(t)(cdf) $R(t) \approx \int_0^t \frac{A(s) - C(s)}{A(t)} ds$  $\frac{(-a)^2}{(2)}$   $\bar{N}(t) \approx \int_0^t \frac{A(s) - C(s)}{t} ds$  $\bar{N}(t) = \frac{R(t)A(t)}{t}$ Z $(pdf) \quad \mathbb{E}[N] = N, \lambda = X, R = R + Z$ (cdf) **Operation Laws**  $\mathbb{E}[N] = \lambda \mathbb{E}[R]$  $\rho_i = \mathbb{E}[S_i]X_i = \frac{\lambda_i}{\rho_i}$  $\rho_i = \mathbb{E}[S_i]\mathbb{E}[Vi]X = \mathbb{E}[D_i]X$  $X_i = \mathbb{E}[V_i]X$ [pmf]  $\mathbb{E}[R] = \frac{N}{N} - \mathbb{E}[Z]$ 

# **Bottleneck Analysis**

 $D_{max}$  [Bottleneck Device] [Arrival See Time Average]  $\mathbb{E}[R] \ge D$ [Memoryless] [Merge Poisson Processes]  $\mathbb{E}[R] \ge max(D, ND_{max} - \mathbb{E}[Z])$   $X \le min(\frac{1}{D_{max}}, \frac{N}{D + \mathbb{E}[Z]})$  $N^* = \frac{D + \mathbb{E}[Z]}{D_{max}}$ [Split Poisson Processes]

# **Queuing Models**

(Arrivals / Service Times / Number of servers / Room in queue)

# M/M/1

$$\lambda_{i}(t) = \frac{A_{i}(t)}{t} [\text{Arrival Rate}] \qquad \mu > \lambda [\text{Stability condition}]$$

$$X_{i}(t) = \frac{C_{i}(t)}{t} [\text{Throughput}] \qquad \mu > \lambda [\text{Stability condition}]$$

$$\pi_{0} = 1 - \frac{\lambda}{\mu} = 1 - \rho \qquad \pi_{i} = \pi_{0}(\frac{\lambda}{\mu})^{i} = (1 - \rho)\rho^{i}$$

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$$\mathbb{E}[N_{i}] = \mathbb{E}[S_{i}]$$

$$\mathbb{E}[N] = \frac{\lambda}{\mu - \lambda} = \frac{\rho}{1 - \rho} \qquad \mathbb{E}[N_{i}] = \mathbb{E}[N_{i}] - \rho$$

$$\mathbb{E}[R_{i}] = \mathbb{E}[S_{i}]\mathbb{E}[V_{i}]$$

$$\mathbb{E}[R_{i}] = \mathbb{E}[S_{i}]\mathbb{E}[V_{i}]$$

$$V_{user} = V_{0} = 1$$

$$\lambda_{i} = X_{i} [\text{Steady state}]$$

$$N/M/C$$

$$[\text{Avg response time}] \qquad \rho = \frac{\lambda}{c\mu} \qquad c\mu > \lambda [\text{Stability condition}]$$

$$\pi_{0} = (\frac{\lambda}{\mu})^{c} \frac{1}{1 - \rho} \qquad \pi_{i} = \begin{cases} \frac{\lambda^{i}}{i!\mu^{i}}\pi_{0}, & \text{if } \\ \frac{\lambda^{i}}{i!\mu^{i}}\pi_{0}, & \text{if } \end{cases}$$

$$\mathbb{E}[N] = \lambda \mathbb{E}[R]$$
[Think time]  
[Closed System]
$$\mathbb{E}[R] = \mathbb{E}[R_Q] + \mathbb{E}[S] = \mathbb{E}[R_Q] + \frac{1}{\mu}$$

$$\mathbb{E}[R_Q] = \frac{(\frac{\lambda}{\mu})^c \mu}{(c-1)!(c\mu-\lambda)^2}$$

 $V_{user}$ 

[Little's Law] [Utilization Law] [Bottleneck Law] [Forced Flow Law]

[Closed System Response Time Law]

 $\implies$  optimal X and  $\mathbb{E}[R]$ 

# $M/M/\infty$ $\rho = \lambda/\mu$

# $\mu > \lambda$ [Always Stable] $\pi_i = \frac{(\lambda/\mu)^i}{i!} e^{-\frac{\lambda}{\mu}} = \frac{\rho^i}{i!} e^{-\rho}$

$$\begin{aligned}
& \mu \neq \lambda, \mu \\
& \mu \neq \lambda, \mu \\
& \mu \neq \lambda, \mu, \mu \\
& \mu \neq \lambda, \mu, \mu \\
& \mu \neq \lambda, \mu \end{pmatrix}$$

 $\begin{pmatrix} 2 \\ 2 \end{pmatrix}$   $\begin{pmatrix} 3 \\ 3 \end{pmatrix}$   $\begin{pmatrix} \cdots \\ c \\ c \end{pmatrix}$   $\begin{pmatrix} c \\ c \end{pmatrix}$ 

$$\frac{D_{max}}{r}, \frac{N}{D + \mathbb{E}[Z]}) \quad \mathbb{E}[R] = \frac{1}{\mu} = \mathbb{E}[S] \qquad \mathbb{E}[R_Q] = 0$$

 $P(\text{job is queued}) = \sum_{i=0}^{\infty} \pi = \frac{1}{c!} (\frac{\lambda}{\mu})^c \frac{1}{1-\rho} \pi_0$ 

ondition] if i < c $\int \frac{1}{c!\mu^i c^{i-c}} \pi_0, \quad \text{if } i \ge c$ 

$$\mathbb{E}[N_Q] = \lambda \mathbb{E}[R_Q]$$

[Erlang C Formula]

#### **Birth-Death Process**

CTMC where state transitions increase or decrease by a constant factor.

$$\pi_0 = \frac{1}{1 + \sum_{k=1}^{\infty} \prod_{i=1}^k \frac{\lambda_{i-1}}{\mu_i}}$$
$$\pi_i = \frac{\prod_{j=0}^{i-1} \lambda_j}{\prod_{j=1}^i \mu_j} \pi_0$$

#### **Threshold System**

T>0, Arrival rate s, processing rate s. If  $r>s,N\rightarrow 0.$  If  $s>r,N\rightarrow \infty.$ 

 $\pi_0 = \frac{1}{1 - \frac{r}{s}} (\frac{s}{r})^T - 1$  $\pi_i = \begin{cases} (\frac{s}{r})^i \pi_0, & \text{if } i < T\\ (\frac{s}{r})^{i-T} (\frac{r}{s})^2 \pi_0, & \text{if } i \ge T \end{cases}$ 

#### **Jackson Networks**

- 1. External arrivals form a Poisson process
- 2. All service times are exponentially distributed and the service discipline at all queues is first-come, first-served
- 3. internal routing of jobs between servers is probabilistic
- 4. The utilization of all of the queues is less than one

#### Solved via markov model

- 1. Markov Chain: We may solve the corresponding Discrete Time Markov Chain to find its steady state distribution,  $\mathbb{E}[N]$ , and  $\mathbb{E}[R]$ . If there are N jobs and k nodes, we will have a lower bound of  $\Omega(\binom{N+k-1}{k-1}^2)$  when solving the system of equations.
- 2. **Product form:** Using a temporary value for each node's arrival rate,  $\bar{\lambda}$ , determine the ratios between the balance equations and then recover the real values using the actual arrival rate,  $\lambda$ , finding the steady-state distribution,  $\mathbb{E}[N]$ , and  $\mathbb{E}[R]$ . Still suffers from a combinatorial explosion in complexity with a lower bound of  $\Omega(\binom{N+k-1}{k-1})$ .
- 3. Mean Value Analysis: Uses the Arrival Theorem in a recursive algorithm to analyse specific nodes when there are N jobs in the system. We only have access to expectations and utilization of specific nodes, i.e.  $\mathbb{E}[R_i], \mathbb{E}[N_i], \rho_i$  but is more performant with an upper bound of  $\mathcal{O}(Nk)$ .

#### M/G/1

- Markovian (modulated by a Poisson process), service times have a General distribution and there is a single server
- $\mathbb{E}[S] = \frac{1}{\mu}$
- high variance in service distribution  $\implies$  high response time
- Has equal  $\mathbb{E}[N]$  for all blind non-pre-emptive service policies

#### Pollaczek–Khinchine formula

 $\mathbb{E}[N] = \rho + \frac{\rho^2 + \lambda^2 \sigma_s^2}{2(1-\rho)}$ 

# Service Policies

Blind and non-blind policy relates to knowledge of job size on arrival. If service times that jobs require are known, then the optimal scheduling policy is shortest remaining processing time (SRPT).

- first-come, first-served (FCFS)
- processor sharing (PS) where all jobs in the queue share the service capacity between them equally
- last-come, first served (LCFS) with/without preemption where a job in service may or may not be interrupted with work being conserved
- generalized foreground-background (FB) scheduling also known as least-attained-service where the jobs which have received least processing time so far are served first and jobs which have received equal service time share service capacity using processor sharing
- shortest job first (SJF) with/without preemption, where the job with the smallest size receives service
- shortest remaining processing time (SRPT) where the next job to serve is that with the smallest remaining processing requirement

## Failure/Hazard Rate

- Increasing Failure Rate (IFR): h(t) is non-decreasing in t, the expected remaining work is decreasing, non pre-emptive is preferable.
- Decreasing Failure Rate (DFR): h(t) is non-increasing in t, the expected remaining work is increasing, pre-emptive policy is preferable.

$$\begin{split} h(t) &= \frac{f(t)}{1 - F(t)} & \mathbb{E}[\text{Remaining time}] = \frac{1}{h(t)} \\ X &\sim uniform(a, b) \text{ (IFR)} \implies & h(t) = \frac{1}{b - t} \\ X &\sim exp(\lambda) \text{ (IFR and DFR)} \implies & h(t) = \frac{\lambda e^{-\lambda t}}{1 - (1 - e^{-\lambda t})} \\ \text{Time average Excess} &= \mathbb{E}[S_c] = \frac{\mathbb{E}[S_c]}{2\mathbb{E}[S]} \\ \mathbb{E}[R_Q] &= \frac{\rho}{1 - \rho} \mathbb{E}[S_c] \end{split}$$

#### Pareto Distribution

- popular DFR, "80-20 rule", pre-emptive policy is preferable
- 50% of the load on the system comes from 1% of the jobs
- $\alpha$  shape parameter,  $\alpha = 1, X > t \implies P(X > 2t) = \frac{1}{2}$
- $0 < alpha < 1, Var(X) = \infty, \mathbb{E}[X] = \infty$
- Survival Function:

$$\overline{F}(x) = \Pr(X > x) = \begin{cases} \left(\frac{x_{\mathrm{m}}}{x}\right)^{\alpha} & x \ge x_{\mathrm{m}}, \\ 1 & x < x_{\mathrm{m}}, \end{cases}$$

#### Misc

- $\sum_{i=0}^{\infty} \alpha^i = \frac{1}{1-\alpha}, |\alpha| < 1.$ •  $h = \frac{f}{a} \implies h' = \frac{f'g - fg'}{g^2}$
- Max system utilization  $\implies$  only bottleneck utilization is 100%
- Want to minimize  $\mathbb{E}[R]$  and maximize X.
- Operation Laws work regardless of distributions of random variables
- exponential distributions are a very good assumption for modeling arrivals, but only moderately good for modelling processing times