

Difference between

$$\lim_{n \rightarrow \infty} P_i \{X_n = a\} = 0.56 \quad \checkmark \quad \text{jump times tally at jump time}$$

$$\lim_{t \rightarrow \infty} P_i \{Y(t) = a\} = 0.39 \quad \leftarrow \begin{array}{l} \text{arbitrary time} \\ \text{time weighted avg} \\ \text{state statistic} \end{array}$$

M/G/1 • at arrival pts same
G/M/1 ← at arrivals different

Lifetimes of components: s_1, s_2, \dots, s_n

$K = j$ if j component caused failure

$$T = \min \{s_1, \dots, s_n\}$$

Let Z_j be R.V. such that $P\{Z_j \leq t\} = \varphi(j, t)$

$$P\{T > t\} = P\{s_1 > t, s_2 > t, \dots, s_n > t\} = e^{-(\lambda_1 + \lambda_2 + \dots + \lambda_n)t}$$

$$\text{Let } \lambda = \lambda_1 + \dots + \lambda_n \Rightarrow P\{T > t\} = 1 - e^{-\lambda t} \quad \text{for } t \geq 0$$

$P\{K = j\}$ — Assume s_1 and s_2

$$P\{K = 1\} = P\{s_1 \leq s_2\} = E[P\{s_1 \leq s_2 | s_2\}]$$

$$P\{s_1 \leq s_2 | s_2 = a\} = \underbrace{1 - e^{-\lambda_1 a}}_{\text{for } a \geq 0}$$

$$E[P\{s_1 \leq s_2 | s_2\}] = \int_0^\infty (1 - e^{-\lambda_1 a}) \lambda_2 e^{-\lambda_2 a} da = \dots = \frac{\lambda_1}{\lambda}$$

$$P\{K = 1, T \leq t\} = \frac{\lambda_1}{\lambda} (1 - e^{-\lambda t})$$

$$P\{K=j, T \leq t\} = \frac{\lambda_j}{\lambda} \underbrace{P\{T \leq t\}}_{\text{pdf}} \quad 1 - e^{-\lambda t} \rightarrow \text{pdf } \lambda e^{-\lambda t}$$

$$P_i\{X_1=j, T_1 \leq t\} = \left(\frac{\lambda_j}{\lambda}\right) \underbrace{P\{Z_i + T \leq t\}}$$

$$Q(z, j, t) = \int_0^t \lambda_j e^{-\lambda s} \underbrace{\varphi(z, t-s)}_{\text{pdf}} ds$$

$$P(i, j) = \frac{\lambda_j}{\lambda} \rightarrow \text{all rows are the same}$$

$v = (\lambda_1, \lambda_2, \dots, \lambda_n)$ is invariant

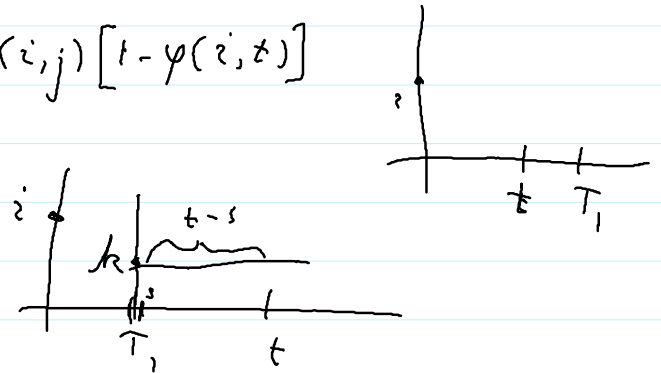
$$\mu(i) = E_i[T_1] = E_i[Z_i + T] = E[Z_i] + \frac{1}{\lambda}$$

$$v \cdot \mu = \sum_j \lambda_j E[Z_j] + \sum_j \frac{\lambda_j}{\lambda} = 1 + \sum_j \lambda_j E[Z_j]$$

$$h(i, t) = P_i \{ Y(t) = j, w(t) = 0 \} \quad \text{for a fixed } j$$

$$P_i \{ Y(t) = j, w(t) = 0, T_1 > t \} = I(i, j) [1 - \varphi(i, t)]$$

$$P_i \{ Y(t) = j, w(t) = 0, T_1 \leq t \}$$



$$P_i \{ Y(t) = j, w(t) = 0 \} = I(i, j) [1 - \varphi(i, t)] + \sum_k \int_{[0, t]} Q(i, k, ds) P_k \{ Y(t-s) = j, w(t-s) = 0 \}$$

$$\begin{aligned} P_i \{ Y(t) = j, w(t) = 0 \} &= \sum_k \int_{[0, t]} R(i, k, ds) I(k, j) [1 - \varphi(k, t-s)] \\ &= \int_{[0, t]} R(i, j, ds) [1 - \varphi(j, t-s)] \end{aligned}$$

$$\lim_{t \rightarrow \infty} P_i \{ Y(t) = j, w(t) = 0 \} = \frac{\nu(j)}{\nu \cdot \mu} \int_0^\infty [1 - \varphi(j, t)] dt$$

$$\lim_{t \rightarrow \infty} P_i \{ Y(t) = j, w(t) = 0 \} = \frac{\lambda_j E[Z_j]}{1 + \sum_k \lambda_k E[Z_k]}$$

Lognormal

$$X \rightarrow \text{Normal} \quad \text{mean} = \mu_N \quad \text{var} = \sigma_N^2$$

$$Y = e^X \quad \text{with} \quad E[Y] = \mu_L \quad \text{and} \quad \text{Var}(Y) = \sigma_L^2$$

Y is called lognormal, $\ln(Y)$ is normal

Assume know μ_L and σ_L^2

$$\sigma_N^2 = \ln\left(\frac{\sigma_L^2}{\mu_L^2} + 1\right) \quad \mu_N = \ln(\mu_L) - \frac{1}{2}\sigma_N^2$$

Assume know μ_N and σ_N^2

$$\mu_L = \exp\left(\mu_N + \frac{1}{2}\sigma_N^2\right) \quad \text{and} \quad \sigma_L^2 = \mu_L^2 \times (\exp(\sigma_N^2) - 1)$$