

$$E(Y(t)) = \lambda \int_0^t E_A(h(t, \tau, A)) d\tau.$$

$$E(h(t, \tau, A)) = 1 \cdot P(h(t, \tau, A) = 1)$$

$$= P(A \geq t - \tau) = P(A \geq \max(t - \tau, 0)) = \int_{\max(t - \tau, 0)}^{\infty} f(x) dx$$

$$E(Y(t)) = \lambda \int_0^t \left( \int_{\max(t - \tau, 0)}^{\infty} f(x) dx \right) d\tau \quad f(x) = \mu \exp(-\mu x)$$

$$= \lambda \mu \int_0^t \left( \int_{\max(t - \tau, 0)}^{\infty} \exp(-\mu x) dx \right) d\tau = \lambda \mu \int_0^t \left( \int_{t - \tau}^{\infty} \exp(-\mu x) dx \right) d\tau$$

Queueing Problem. Kleirock (UCLA) 1970's

Income, Outcome, Number of Resource. 010010001

Poisson. Poisson (Exponential)  $M/G/K$  <sup>↑</sup> Kendall

$M/G/\infty$   $f(x)$  Queue Length

$$Y(t) = \sum_{k=1}^{N(t)} h(t, \tau_k, A_k)$$

$$h(t, \tau, A) = \begin{cases} 1 & \tau \leq t \leq \tau + A \\ 0 & \text{others} \end{cases}$$

$A \geq 0.$

$$P(\bar{x}_k | \bar{x}_{k-1}, \dots, \bar{x}_1) = P(\bar{x}_k | \bar{x}_{k-1}) \quad \text{Markov Property.}$$

$\downarrow$  Future     $\downarrow$  Now     $\downarrow$  Past  
 C        B        A

$$P(C|BA) = P(C|B)$$

$$P(CA|B) = P(C|B)P(A|B)$$

$$P(CA|B) = \underline{P(C|AB)}P(A|B) = P(C|B)P(A|B)$$

$$P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2, \bar{x}_1 = x_1) = P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2)$$

$$P(\bar{x}_3 \in A | \bar{x}_2 = x_2, \bar{x}_1 = x_1) \neq P(\bar{x}_3 \in A | \bar{x}_2 = x_2)$$

$$= \lambda \int_0^t \exp(-\mu(t-\tau)) d\tau = \lambda \exp(-\mu t) \int_0^t \exp(\mu\tau) d\tau$$

$$= \frac{\lambda}{\mu} \exp(-\mu t) (\exp(\mu t) - 1) = \boxed{\frac{\lambda}{\mu}} (1 - \exp(-\mu t)) \quad \text{Steady State}$$

Markov Chains, 1905  $\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n$   $P(\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)$

$$= P(\bar{x}_n | \bar{x}_{n-1}, \dots, \bar{x}_1) P(\bar{x}_{n-1}, \dots, \bar{x}_1) = P(\bar{x}_n | \bar{x}_{n-1}, \dots, \bar{x}_1) P(\bar{x}_{n-1} | \bar{x}_{n-2}, \dots, \bar{x}_1)$$

$$= \prod_{k=1}^n P(\bar{x}_k | \bar{x}_{k-1}) \quad \text{Constraint Nearest Neighborhood } P(\bar{x}_{n-2}, \dots, \bar{x}_1)$$



$$P(\bar{x}_3 \in A | \bar{x}_2 = x_2, \bar{x}_1 = x_1) = \sum_{x_3 \in A} P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2, \bar{x}_1 = x_1).$$

$$= \sum_{x_3 \in A} P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2) = P(\bar{x}_3 \in A | \bar{x}_2 = x_2).$$

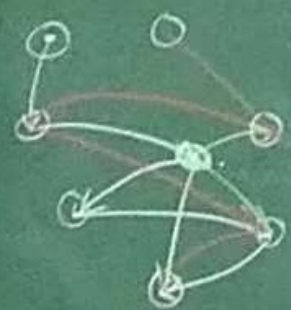
$$P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2, \bar{x}_1 \in A) = P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2)$$

$$\Leftrightarrow P(\bar{x}_3 = x_3, \bar{x}_1 \in A | \bar{x}_2 = x_2) = P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2) P(\bar{x}_1 \in A | \bar{x}_2 = x_2)$$

$$P(\bar{x}_3 = x_3 | \bar{x}_2 \in A, \bar{x}_1 = x_1) = P(\bar{x}_3 = x_3 | \bar{x}_2 \in A)$$

$$A = \Omega. \quad P(\bar{x}_3 = x_3 | \bar{x}_2 \in \Omega, \bar{x}_1 = x_1) = P(\bar{x}_3 = x_3 | \bar{x}_1 = x_1)$$

$$= P(\bar{x}_3 = x_3 | \bar{x}_2 \in \Omega) = P(\bar{x}_3 = x_3)$$



Jump!

Time Discrete Continuous  
state Discrete Markov Chains, Poisson.

Continuous Time Series Gaussian (Brown Motion)

$$X_t = f(X_{t-1}, W_t)$$





Chapman-Kolmogorov  
(C-K)  
Matrix  
Multiplication

$$\sum_{x_1} P(\bar{x}_n = j, \bar{x}_{n-1} = x_{n-1} | \bar{x}_1 = x_1, \bar{x}_0 = i)$$

$$P(\bar{x}_n = j | \bar{x}_0 = i) = \sum_k P(\bar{x}_n = j, \bar{x}_m = k | \bar{x}_0 = i)$$

$$= \sum_k P(\bar{x}_n = j | \bar{x}_m = k, \bar{x}_0 = i) P(\bar{x}_m = k | \bar{x}_0 = i)$$

$$= \sum_k P(\bar{x}_n = j | \bar{x}_m = k) P(\bar{x}_m = k | \bar{x}_0 = i)$$

$$P_{ij}(h) = \sum_k P_{ik}(m) P_{kj}(h-m)$$

$P(\bar{x}_n = x_j | \bar{x}_m = x_i)$ .  $n > m$ . Transition Probability.

$$P(\bar{x}_1 = x_1, \bar{x}_2 = x_2, \dots, \bar{x}_n = x_n) = \prod_{k=1}^{n-1} P(\bar{x}_{k+1} = x_{k+1} | \bar{x}_k = x_k)$$

Local  $\Rightarrow$  Global.  $P(\bar{x}_n = x_n | \bar{x}_{n-1} = x_{n-1})$  One-Step TP.

$$P(\bar{x}_n = x_j | \bar{x}_m = x_i) = P_{ij}(n, m) = P_{ij}(n-m).$$

Stationary  
Transition Probability

$$P(\bar{x}_n = j | \bar{x}_0 = i) = P(\bar{x}_n = j, \bar{x}_0 = i) / P(\bar{x}_0 = i).$$

$$P(n) = (P_{ij}(n))_{ij} \Rightarrow P(n) = P(m) \cdot P(n-m) \quad \underline{m \leq n}$$

$$P(n) = P(n-1) \cdot P(1) = P(n-2)(P(1))^2 = \dots = (P(1))^n$$

$$P(1) = P = \begin{pmatrix} \alpha & 1-\alpha \\ 1-\beta & \beta \end{pmatrix} \quad 0 \leq \alpha, \beta \leq 1 \quad P_{ij} \geq 0, \quad \sum_j P_{ij} = 1$$

Jordan Canonical Form.  $P = B^{-1} \Lambda B \Rightarrow P^n = B^{-1} \Lambda^n B \quad (\underline{P_{ii} > 0})$

$$\det(\lambda I - P) = 0 \Rightarrow \lambda_1, \lambda_2$$

$$(\lambda_1 I - P)x = 0, \lambda_1, \det(\lambda I - P) = (\lambda - \lambda_1)^k \in \mathbb{C}.$$

$$(\lambda_1 I - P)^2 x = 0, \lambda_1, \lambda_1$$

$$(\lambda_1 I - P)^3 x = 0, \dots, \begin{pmatrix} \lambda_1 & 1 & & \\ & \lambda_1 & 1 & \\ & & \lambda_1 & 1 \\ & & & \lambda_1 \end{pmatrix}$$

$$P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2, \bar{x}_1 \in A) = P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2)$$

$$\Leftrightarrow P(\bar{x}_3 = x_3, \bar{x}_1 \in A | \bar{x}_2 = x_2) = P(\bar{x}_3 = x_3 | \bar{x}_2 = x_2) P(\bar{x}_1 \in A | \bar{x}_2 = x_2)$$

$$P(\bar{x}_3 = x_3 | \bar{x}_2 \in A, \bar{x}_1 = x_1) = P(\bar{x}_3 = x_3 | \bar{x}_2 \in A)$$