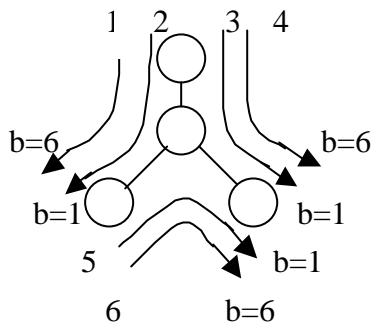


MDP for multirate loss network



Exact analysis

State space:

$$(X_1, X_2, X_3, X_4, X_5, X_6)$$

$$\sum_{k \in C_n} b_k X_k \leq m_n$$

m_n : n-th Link's capacity

C_n : class that traverse Link n

Action

Revenue

For each state

$$\sum_{i=1}^6 (1 - q_i) r_i \lambda_i$$

$$X = (X_1, \dots, X_6)$$

$$A_x = \{a \mid (a_1, \dots, a_6)\}$$

$$a_i = \begin{cases} 0 & \text{block class } i \\ 1 & \text{accept class } i \end{cases}$$

Decomposed MDP

- look at a single Link only
- Assume initial policy is CS(complete share) policy
- Form a MDP at each link
- classify calls according to bandwidth requirement
- e.g. $X=(X_1, X_2)$ $b_1=6$ $b_2=1$
- modified link reward

for original class i calls, $i=1, \dots, 6$

$$\text{modified link reward} = \frac{r_i}{|R_i|} \quad |R_i|: \text{length of routing path}$$

$$\text{for aggregated class } r_i' = \sum_j \frac{\lambda_j}{\sum_j \lambda_j} \times \frac{r_i}{|R_i|}$$

$$\begin{cases} \lambda_1 = 1, \lambda_2 = 1, \lambda_3 = 2, \lambda_4 = 2 \\ \gamma_1 = 6, \gamma_2 = 1, \gamma_3 = 12, \gamma_4 = 2 \end{cases}$$

$$mr_1 = \frac{6}{2} = 3 \quad mr_2 = \frac{1}{2} \quad mr_3 = 6 \quad mr_4 = 1$$

$$\gamma_1^1 = \frac{1}{1+2} 3 + \frac{2}{1+2} 6$$

$$\gamma_2^1 = \frac{1}{1+2} \times \frac{1}{2} + \frac{2}{1+2} \times 1$$

state descriptor:

$$(X_1, X_2)$$

Action:

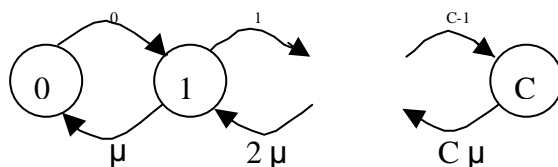
$$(a_1, a_2), \quad a_i = \begin{cases} 0 \\ 1 \end{cases}$$

Reward (loss):

$$\sum (1 - a_i) \cdot \gamma_i^1 \cdot \lambda_i^1$$

Decomposed MDP: Link indep. + Approximation

Reduce to one-dimension CTMC state descriptor: (# of busy circuits)



$$\lambda_i = \frac{2}{2} + i(1 - \frac{1}{2})$$

$$i = \frac{\lambda_i}{2}$$

$$= \sum_{i=1}^k b_i \quad i \quad 2 = \sum_{i=1}^k b_i^2 \quad i$$

$$g = 0 + \bar{\lambda}_0(V_1 - V_0)$$

$$g = 0 + \bar{\lambda}_1(V_2 - V_1) - (V_1 - V_0)$$

$$g = 0 + \bar{\lambda}_{C-1}(V_C - V_{C-1}) - (C-1)(V_{C-1} - V_{C-2})$$

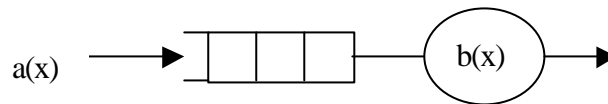
$$g = \sum_{i=1}^k \gamma_i \lambda_i - C(V_C - V_{C-1})$$

$$\Rightarrow \left\{ \begin{array}{l} W_i = V_{i+1} - V_i = \frac{g}{\lambda_i E(\lambda, i)} = (V_{i+bk} - V_{i+bk-1}) + \dots = W_{i+bk-1} + \dots + W_i \\ g = \sum_{i=1}^k \gamma_i \lambda_i - C W_{C-1} \\ W_{C-1} = \frac{\sum_{i=1}^k \gamma_i \lambda_i E(\bar{\lambda}, C)}{\lambda_{C-1} E(\lambda, C-1)} \end{array} \right.$$

Policy improvement:

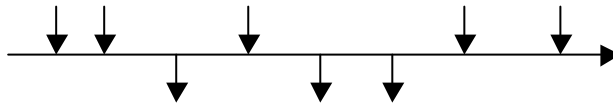
State i : accept class k if $\gamma_k > (V_{i+bk} - V_i) / \lambda_k$

M/G/1 Queue



- service time are no longer exponentially distribution
 $b(x)$: service time pdf
 $a(x)$: $\lambda e^{-\lambda x}$ Poisson process
- model this system as a CTMC need two state variable
(# in queue, remaining service time of the customer in service)

solution: Embedded Markov



state descriptor : (# in queues)

track the system state only after m states of departure

$P_K^D = \text{Prob}(\text{departure sees } k \text{ customers left behind})$

$P_K^A = \text{Prob}(\text{arriving customer sees } k \text{ customers in queue})$

$P_K^R = \text{Prob}(\text{random observer sees } k \text{ customers in queue})$

$$P_K^R = P_K^A = P_K^D$$

Two arrows point upwards from the text 'PASTA' to the second and third terms of the equation above.

PASTA

Define:

$a_n(T)$: # of occurrences of arrivals that changes state to n from (n-1)

$d_n(T)$: # of occurrences of departure that changes state to n from (n+1)

Lemma 1 : $|a_n(T) - d_n(T)| \leq 1$

Lemma 2 : $d(T) = a(T) + n(0) - n(T)$

$$\begin{aligned} P_K^D &= \lim_{T \rightarrow \infty} \frac{d_k(T)}{d(T)} \\ &= \lim_{T \rightarrow \infty} \frac{a_k(T) + d_k(T) - a_k(T)}{a(T) + n(0) - n(T)} \\ &= \lim_{T \rightarrow \infty} \frac{a_k(T)}{a(T)} = p_K^A \end{aligned}$$

Constrain

single server → single departure

Poisson arrival → PASTA