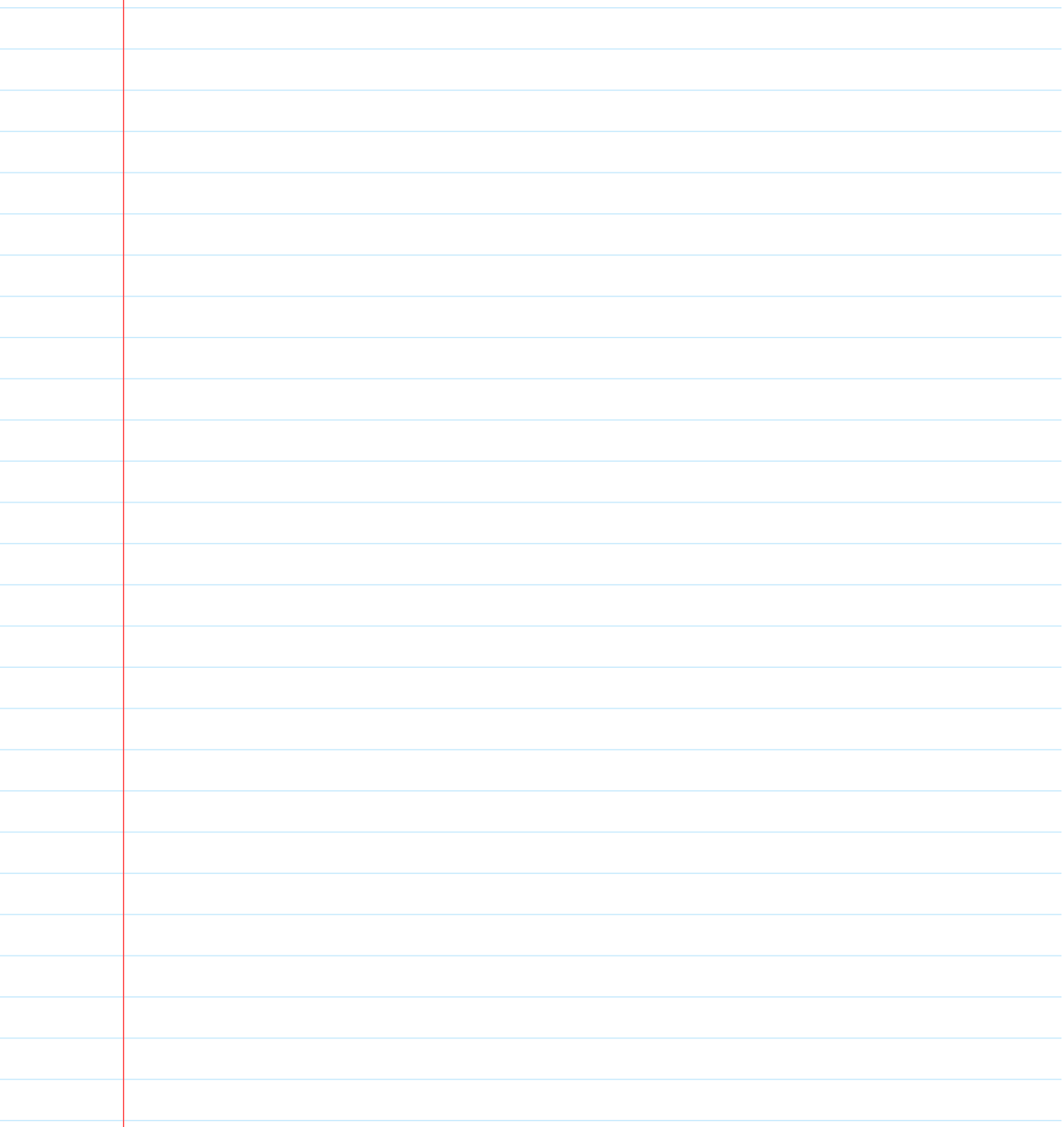


Lec23-mwf

Monday, March 09, 2009 5:38 PM



Deriving AIMD window-based control

Wednesday, March 11, 2009 3:07 PM

- Take $\alpha = 1$

- Approximate the differential equation by a difference equation

$$\frac{x_s(t+\delta) - x_s(t)}{\delta} = K_s [\gamma_s - \beta x_s \cdot z_s]$$

- Recall that

$$x_s(t) = \frac{w_s(t)}{d_s} \quad \begin{array}{l} \leftarrow \text{window size} \\ \leftarrow \text{RTT} \end{array}$$

Let $N_s(t, t+\delta)$ denote the # of packets (or time-outs) received by user s in the interval $(t, t+\delta)$.

$$\text{Then } z_s \approx \frac{N_s(t, t+\delta)}{\delta}$$

- Hence

$$\frac{w_s(t+\delta) - w_s(t)}{\delta d_s} = K_s \left[\gamma_s - \beta \frac{w_s(t) N_s(t, t+\delta)}{d_s} \right]$$

$$\Rightarrow w_s(t+\delta) - w_s(t) = K_s [\gamma_s \delta d_s - \beta w_s(t) N_s(t, t+\delta)]$$

- Let $A_s(t, t+\delta)$ denote the # of acks received

by w_s in the time-interval $(t, t+\delta)$.

$$X_s(t) = \frac{W_s(t)}{d_s} \approx \frac{A_s(t, t+\delta)}{\delta}$$

— Using $\delta = \frac{A_s(t, t+\delta)}{X_s(t)} = \frac{A_s(t, t+\delta)}{W_s(t)} \cdot d_s$

$$\Rightarrow W_s(t+\delta) - W_s(t) = K_s \left[\frac{r_s d_s^2}{W_s(t)} A_s(t, t+\delta) \right.$$

$$\left. - \beta W_s(t) N_s(t, t+\delta) \right]$$

Interpretation: if $K_s r_s d_s^2 = 1$ & $\beta K_s = \frac{1}{2}$

— For each ack, window-size increases by $\frac{1}{W_s}$

— For each nack, window-size decreases by $\frac{1}{2} W_s$

— A2MD

(30)

- Take $U_s(x_s) = \gamma_s \log x_s$

- Approximate the differential equation by a difference equation

$$\frac{x_s(t+\delta) - x_s(t)}{\delta} = K_s [\gamma_s - \beta \cdot z_s]$$

- Recall that

$$x_s(t) = \frac{w_s(t)}{d_s} \quad \begin{array}{l} \leftarrow \text{window size} \\ \leftarrow \text{RTT} \end{array}$$

Let $N_s(t, t+\delta)$ denote the # of nacks (or time-outs) received by user s in the interval $(t, t+\delta)$.

$$\text{Then } z_s \approx \frac{N_s(t, t+\delta)}{\delta}$$

- Hence

$$\frac{w_s(t+\delta) - w_s(t)}{\delta d_s} = K_s \left[\gamma_s - \beta \frac{N_s(t, t+\delta)}{\delta} \right]$$

$$\Rightarrow w_s(t+\delta) - w_s(t) = K_s [\gamma_s \delta d_s - \beta d_s N_s(t, t+\delta)]$$

- Let $A_s(t, t+\delta)$ denote the # of acks received

by w_s in the time-interval $(t, t+\delta)$.

$$X_s(t) \approx \frac{A_s(t, t+\delta)}{\delta}$$

$$\text{— Using } \delta = \frac{A_s(t, t+\delta)}{X_s(t)} = \frac{A_s(t, t+\delta)}{W_s(t)} \cdot ds$$

$$\Rightarrow W_s(t+\delta) - W_s(t) = K_s \left[\frac{r_s ds^2}{W_s(t)} A_s(t, t+\delta) - \beta ds N_s(t, t+\delta) \right]$$

Interpretation: if $K_s r_s ds^2 = 1$ & $\beta K_s ds = 1$,

- For each ack, window-size increases by $\frac{1}{W_s}$
- For each nack, window-size decreases by 1.

— A2AD

(30)

Packet loss rate

Wednesday, March 11, 2009 4:22 PM

- One consequence of this interpretation is that we can derive an equation that relates the TCP throughput with the packet loss rate.

- Recall the differential equation

$$\dot{x}_s = k_s (\alpha \delta_s - \beta x_s^\alpha \cdot z_s)$$

- Using the parameters corresponding to AIMD:

$$\alpha = 1, \quad k_s \delta_s = \frac{1}{d_s^2} \quad - \quad k_s \beta = 0.5,$$

$$\Rightarrow \dot{x}_s = \left(\frac{1}{d_s^2} - 0.5 x_s \cdot z_s \right)$$

At equilibrium, $\dot{x}_s = 0$, we then have

$$\frac{1}{d_s^2} - 0.5 x_s \cdot x_s \cdot p_s = 0$$

where p_s is the packet-loss probability experienced by user s . We then have

$$x_s = \frac{1}{d_s \sqrt{\frac{p_s}{2}}}$$

Interpretation:

① TCP throughput is inversely proportional to RTT.

— long flows suffer

② TCP throughput is inversely proportional to the square root of P_s

Both have been observed empirically.

④

Critique

Wednesday, March 11, 2009 4:28 PM

- ① Penalty function approach does not produce exact solution to the original problem.
 - packet drops will occur at equilibrium
- ② When buffer size is large, delay will increase
 - TCP will always drive the buffer full.

Later we will investigate other approaches:

- duality, primal-dual
- AQM: RED, REM, AVQ.

Duality approach to congestion control

Thursday, March 19, 2009 4:45 PM

- We now want to develop other solutions to the congestion control problem

$$\begin{aligned} \max \quad & \sum_s U_s(x_s) \\ \text{Sub to} \quad & \sum_j H_j^i x_s \leq R_i \quad \forall i \\ & m_s \leq x_s \leq M_s \end{aligned}$$

(9.1)

- Note that this formulation incorporates a number of objectives
 - high utilization
 - low congestion
 - fairness
- We want our new solution to also retain the distributed nature of TCP
 - source only needs a congestion signal from the links that it passes through (e.g. packet loss)
 - no information of other user's utility
 - link only needs the aggregate rate
 - no information of each individual user.
 - no per-flow information.

Now let us see if duality helps.

Associate a Lagrangian multiplier q_i for each link.
The Lagrangian is given by

$$\begin{aligned} L(\vec{x}, \vec{q}) &= - \sum_s U_s(x_s) + \sum_i q_i \left[\sum_s H_s^i x_s - R_i \right] \\ &= - \sum_s \left[U_s(x_s) - x_s \sum_i q_i H_s^i \right] - \sum_i q_i R_i \end{aligned}$$

↑
separable.

The dual objective function is given by

$$\begin{aligned} g(\vec{q}) &= \min_{\vec{x}} L(\vec{x}, \vec{q}) \\ &= - \sum_s \max_{x_s} \left[U_s(x_s) - x_s \sum_i q_i H_s^i \right] - \sum_i q_i R_i \end{aligned}$$

Hence, each user should maximize

$$\begin{aligned} U_s(x_s) - x_s \cdot \sum_i q_i H_s^i \\ \Rightarrow U_s'(x_s) = \sum_i q_i H_s^i \end{aligned}$$

$$x_s^* = \left[U_s'^{-1} \left(\sum_i q_i H_s^i \right) \right]_{[m_s, M_s]}^+$$

Further,

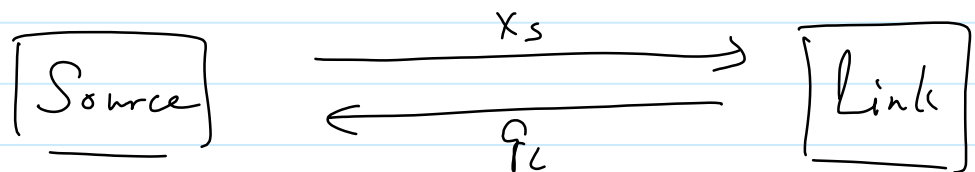
$$\frac{\partial}{\partial q_i} g = \sum_s H_s^i x_s - R_i$$

In order to maximize g

$$q_i(t+1) = \left[q_i(t) + \delta \left(\sum_s H_s^i x_s^*(t) - R_i \right) \right]^+$$

To summarize:

- The dual variable q_L can be viewed as the "price" of capacity at link L
 - Source algorithm: Knowing the price of all links passed by the user, each user computes $x_s(t)$ that maximizes the net utility $U_s(x_s) - x_s \sum_L q_L H_s^L$
 - Link algorithm: Each link update q_L to balance supply & demand.
-



Optimality of dual congestion controller - skip

Wednesday, March 25, 2009 9:36 AM

Theorem: Assume that

- ① On the interval $[m_s, M_s]$, the utility function U_s is increasing, strictly concave, and twice continuously differentiable
- ② For feasibility, assume $\sum_s H_s m_s < R^L \forall L$
- ③ The curvature of $U_s(\cdot)$ is bounded away from zero on $[m_s, M_s]$, i.e.,

$$-U_s''(x_s) \geq \frac{1}{\alpha_s} \text{ for all } x_s \in [m_s, M_s].$$

If the stepsize δ satisfies

$$0 < \delta < \frac{2}{\bar{\alpha} \bar{L} \bar{S}},$$

where $\bar{\alpha} = \max_s \alpha_s$, $\bar{L} = \max_s \sum_L H_s^L$ is the length of the longest-path used by any user, and $\bar{S} = \max_L \sum_s H_s^L$ is the # of users sharing the most-congested link, then starting from any $\bar{q}(0)$, the sequence $(x(t), \bar{q}(t))$ converges, and the limit is primal-dual optimal.

Note: Assume perfect feedback of \bar{q}_i .

Proof: Show that the dual objective function satisfies a Lipschitz condition.
 (see hand-out window-inverse.pdf).

Note that $\frac{\partial}{\partial \vec{r}_L} g = \sum_s H_s^L X_s^* - R_L$. Let \vec{r}^1, \vec{r}^2 be two price vectors, and let $X_s(\vec{r}^1)$ & $X_s(\vec{r}^2)$ be the corresponding rate control decisions (that maximizes net-utility). If g satisfies a Lipschitz condition, i.e.

$$\|\nabla g(\vec{r}^1) - \nabla g(\vec{r}^2)\| \leq \beta \|\vec{r}^1 - \vec{r}^2\|$$

which implies

$$\begin{aligned} & - \left(\nabla g(\vec{r}^1) - \nabla g(\vec{r}^2) \right) \cdot (\vec{r}^1 - \vec{r}^2) \\ & \leq \frac{1}{\beta} \|\nabla g(\vec{r}^1) - \nabla g(\vec{r}^2)\|^2 \end{aligned}$$

The latter is equivalent to.

$$\begin{aligned} & - \sum_L \left[\left(\sum_s H_s^L X_s(\vec{r}^1) - R_L \right) - \left(\sum_s H_s^L X_s(\vec{r}^2) - R_L \right) \right] \\ & \quad \times (\vec{r}_L^1 - \vec{r}_L^2) \\ & \geq \frac{1}{\beta} \sum_L \left[\left(\sum_s H_s^L X_s(\vec{r}^1) - R_L \right) - \left(\sum_s H_s^L X_s(\vec{r}^2) - R_L \right) \right]^2 \end{aligned}$$

See hand out for the detailed proof + show that $\beta = \sum_s \bar{L}$

Refs :

- S. H. Low and D. E. Lapsley, "Optimization Flow Control-I: Basic Algorithm and Convergence," IEEE/ACM Transactions on Networking, vol. 7, no. 6, pp. 861-874, December 1999.
- S. Low, "A Duality Model of TCP and Queue Management Algorithms," IEEE/ACM Trans. on Networking, 11(4):525-536, August 2003

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