

Duality Model of TCP/AQM



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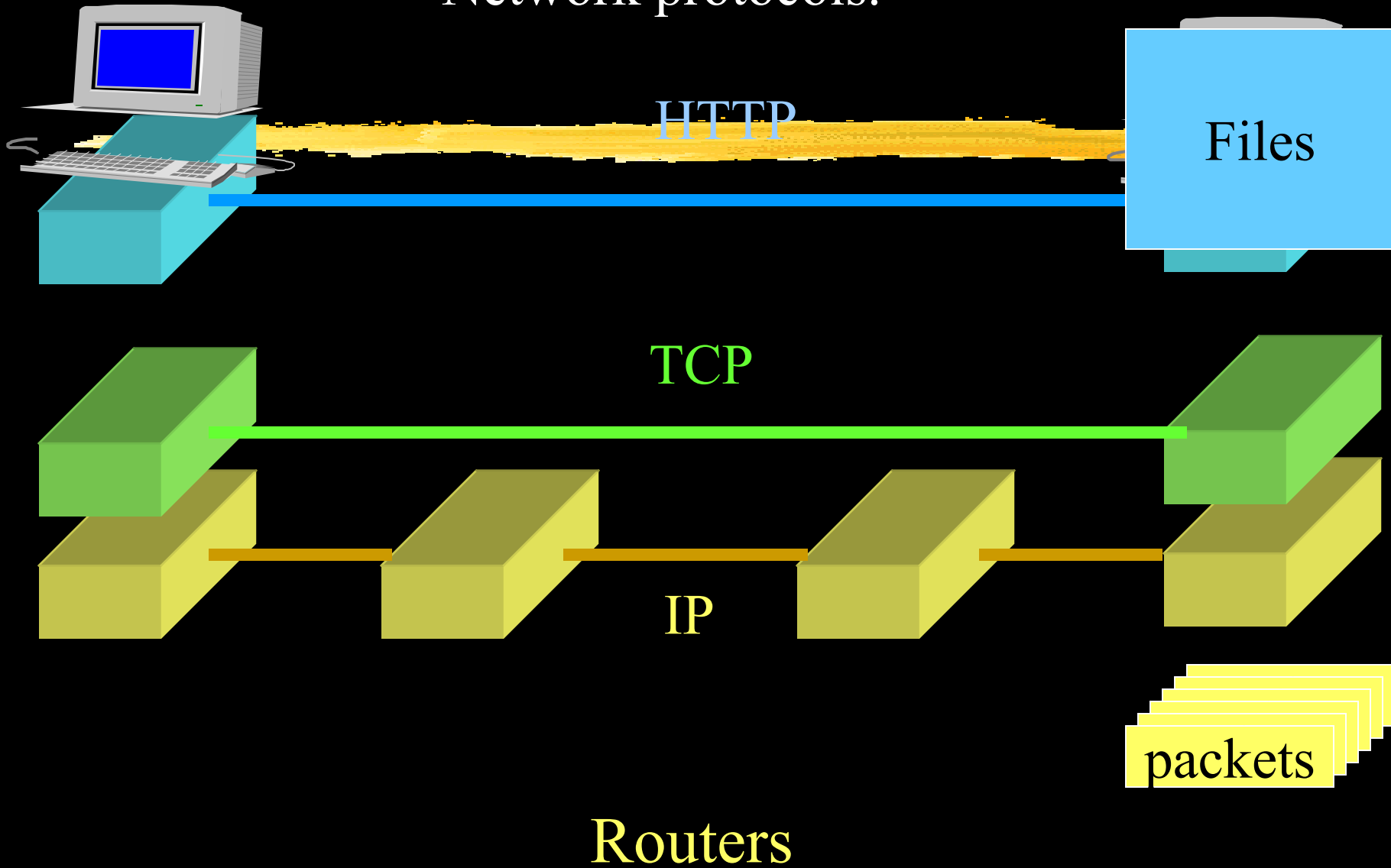
2002

Acknowledgments

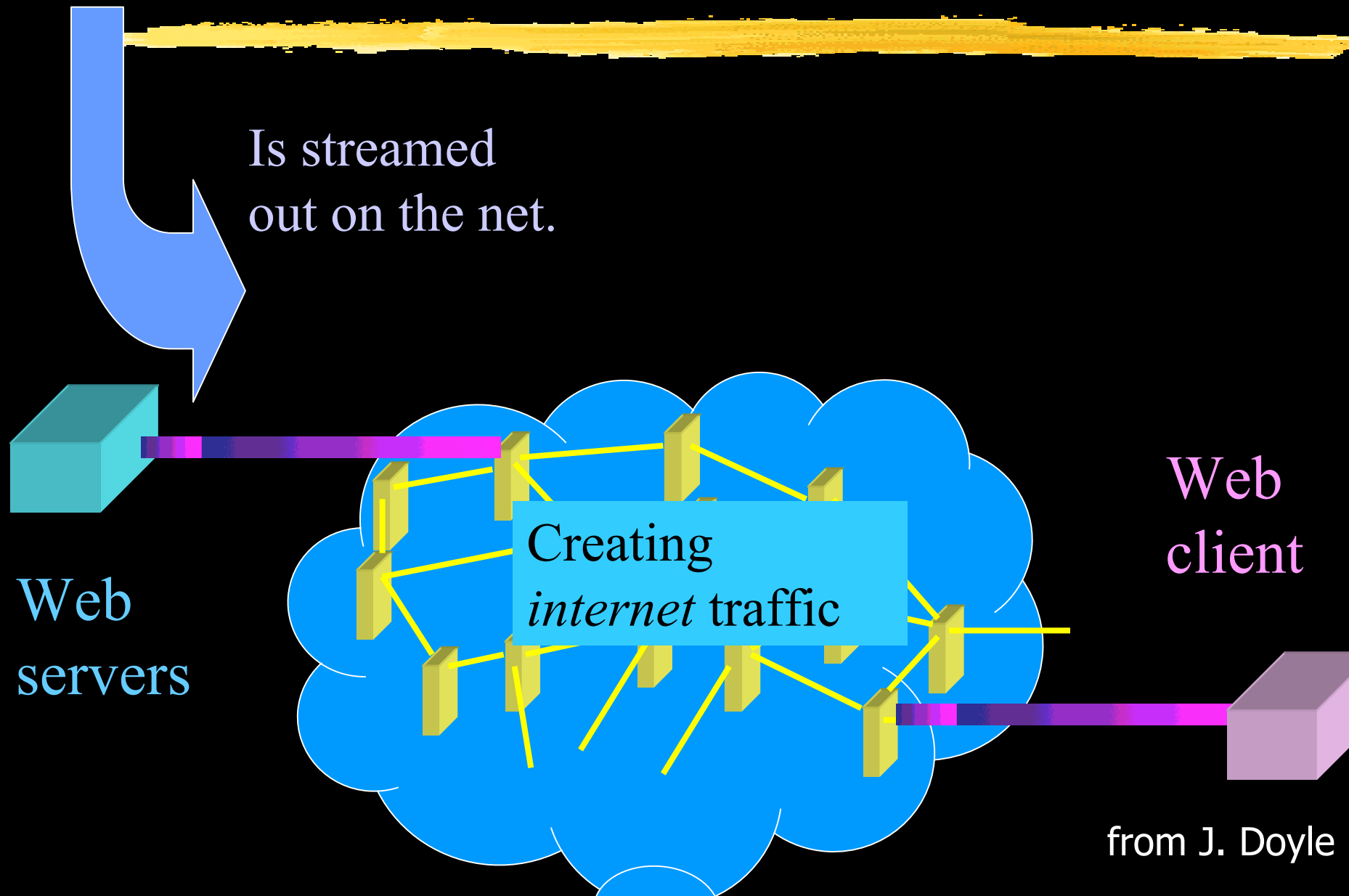


- S. Athuraliya, D. Lapsley, V. Li, Q. Yin (UMelb)
- S. Adlakha (UCLA), D. Choe (Postech/Caltech), J. Doyle (Caltech), K. Kim (SNU/Caltech), L. Li (Caltech), F. Paganini (UCLA), J. Wang (Caltech)
- L. Peterson, L. Wang (Princeton)

Network protocols.

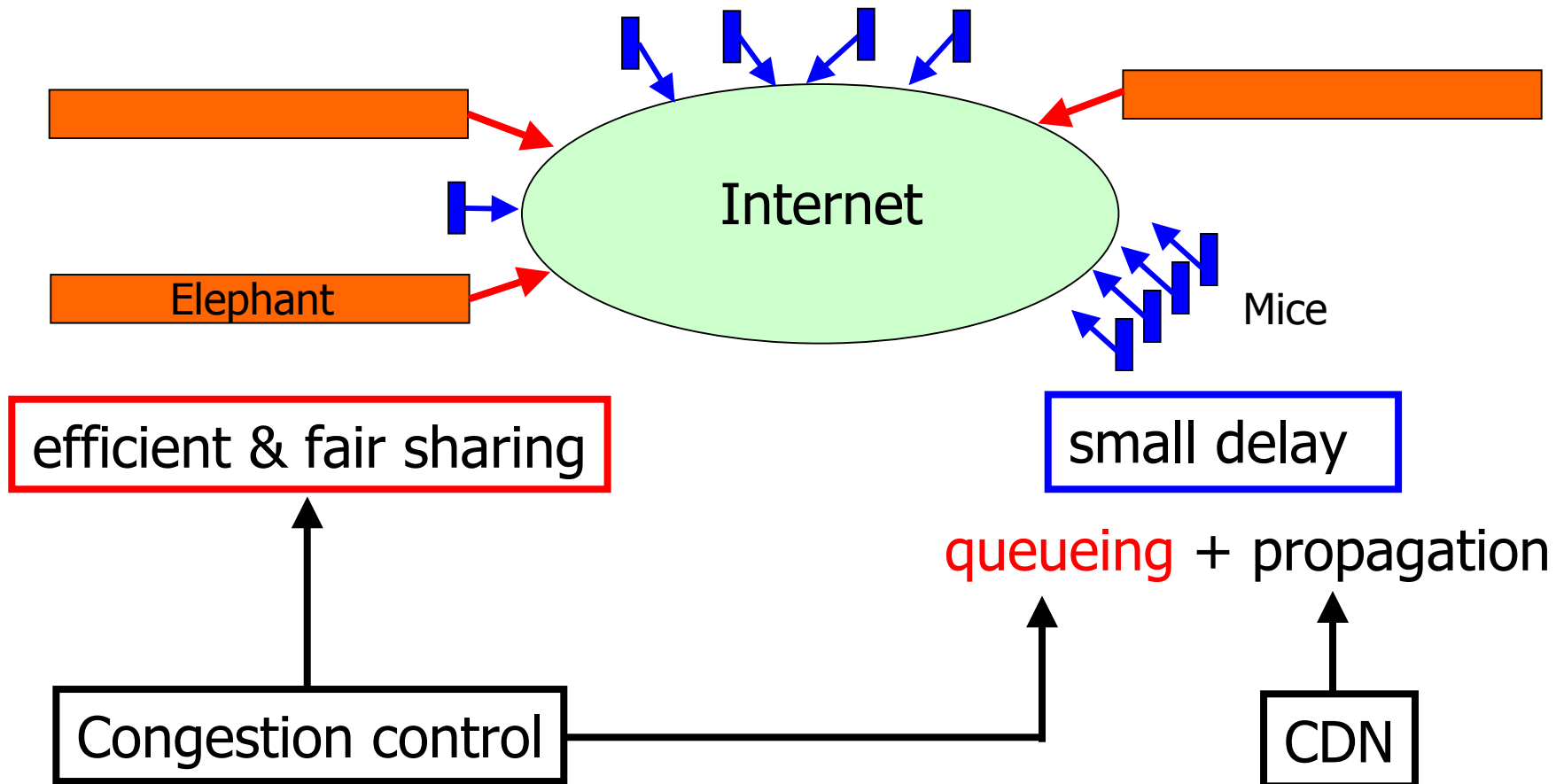


web traffic

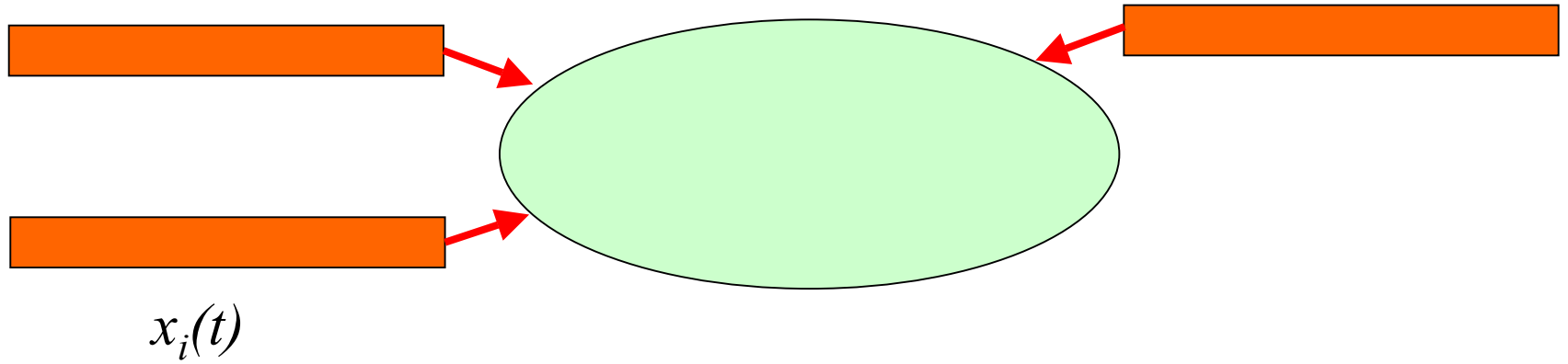


Congestion Control

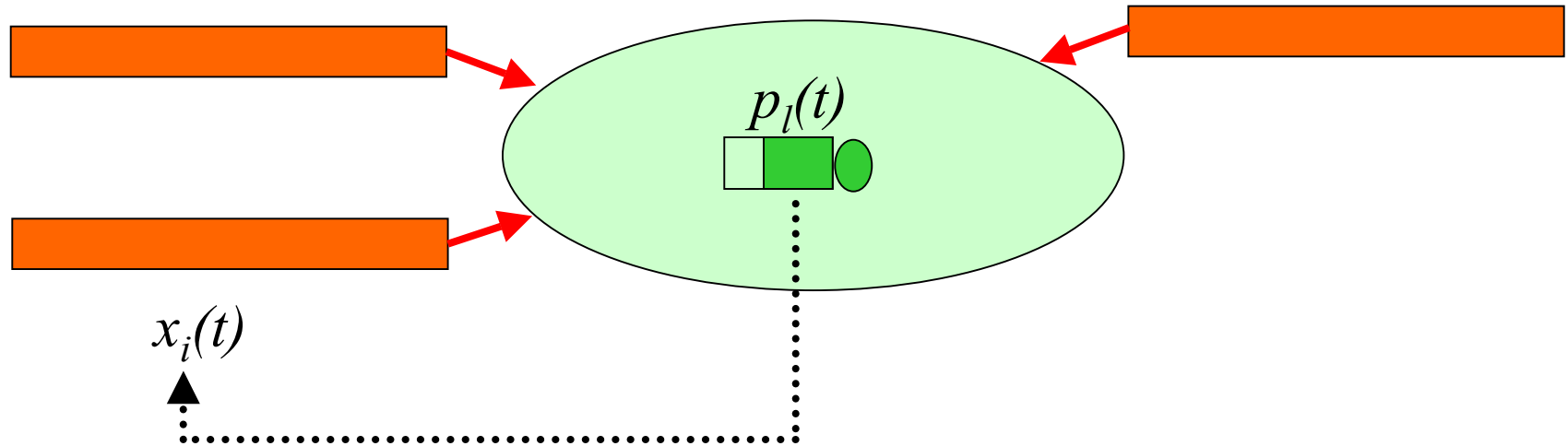
Heavy tail → Mice-elephants



TCP



TCP & AQM



Example congestion measure $p_l(t)$

- Loss (Reno)
- Queueing delay (Vegas)

Outline

Protocol (Reno, Vegas, RED, REM/PI...)

$$\begin{aligned}x(t+1) &= F(p(t), x(t)) \\ p(t+1) &= G(p(t), x(t))\end{aligned}$$

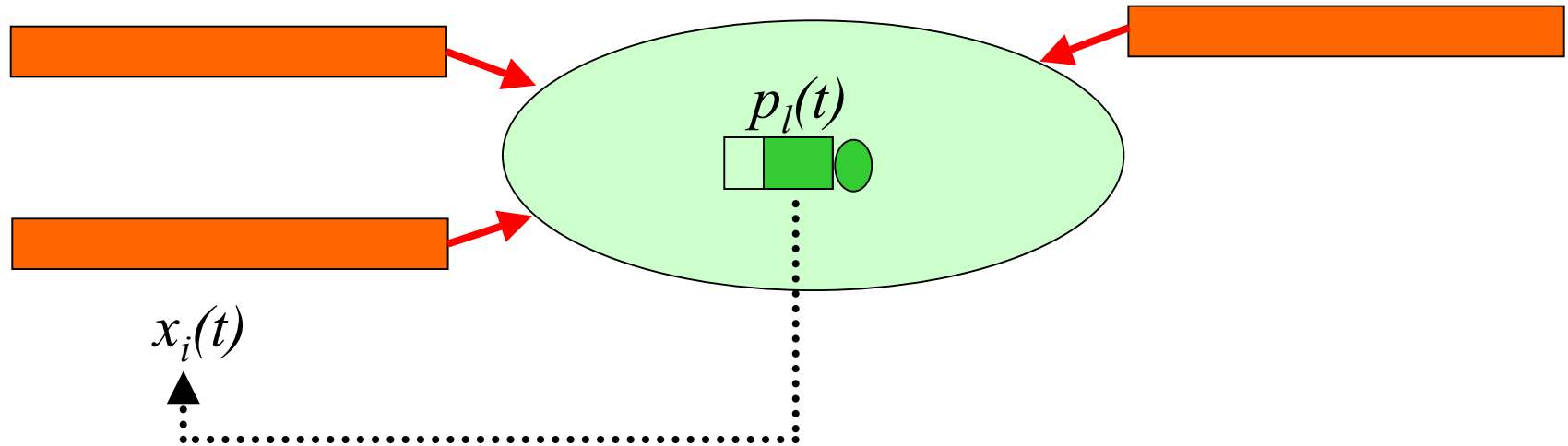
Equilibrium

- Performance
 - Throughput, loss, delay
- Fairness
- Utility

Dynamics

- Local stability
- Cost of stabilization

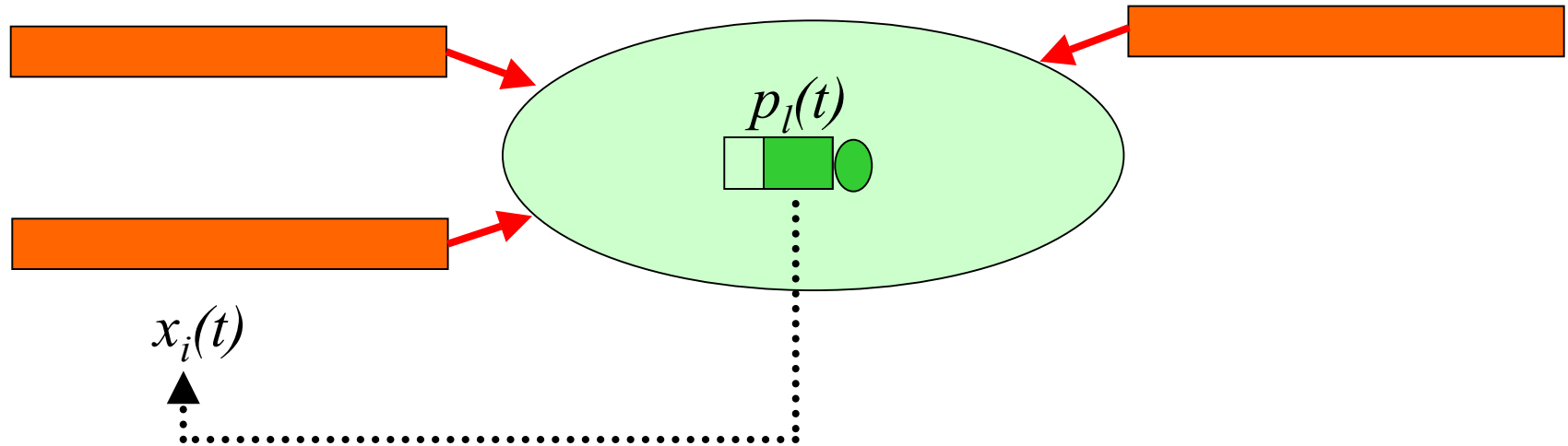
TCP & AQM



Duality theory $\rightarrow$ equilibrium

- Source rates $x_i(t)$ are primal variables
- Congestion measures $p_l(t)$ are dual variables
- Flow control is optimization process

TCP & AQM



Control theory → stability

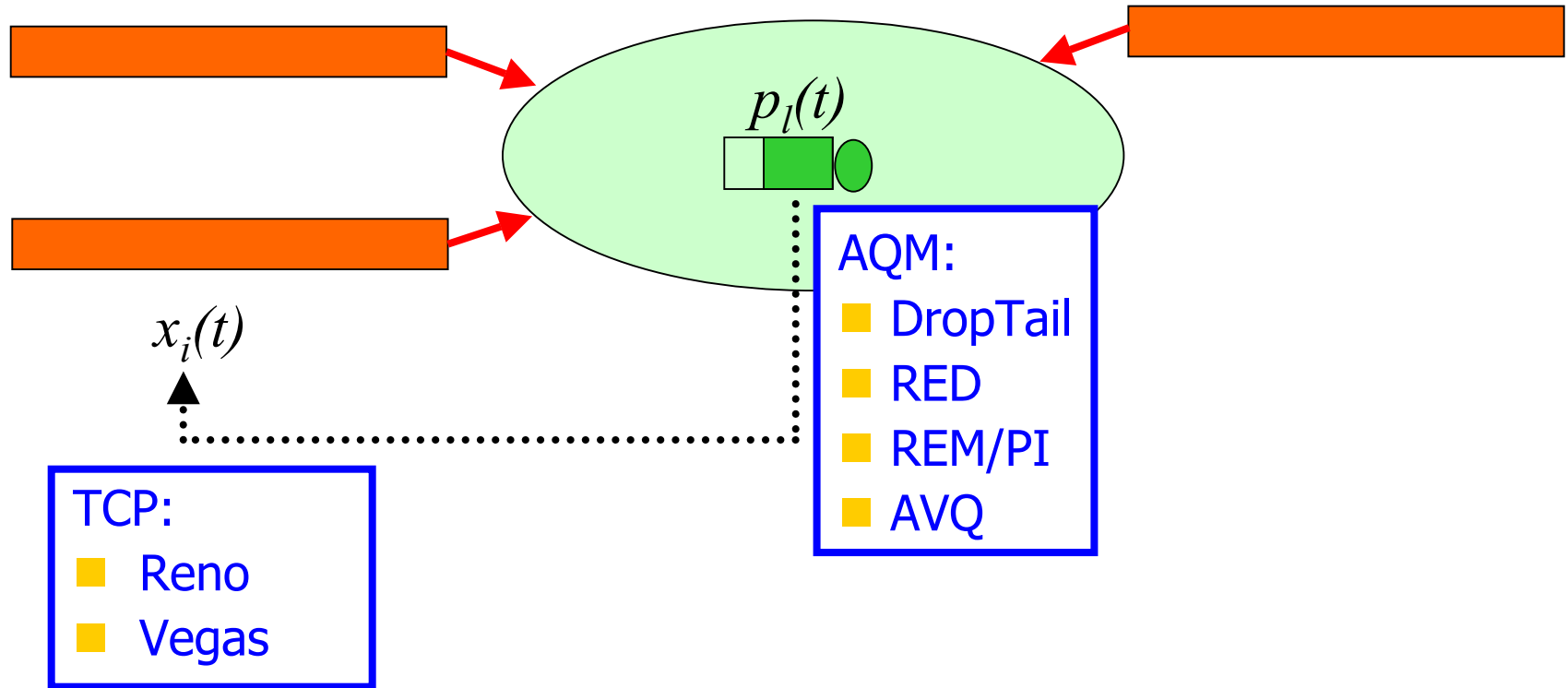
- Internet as a feedback system
- Distributed & delayed

Outline

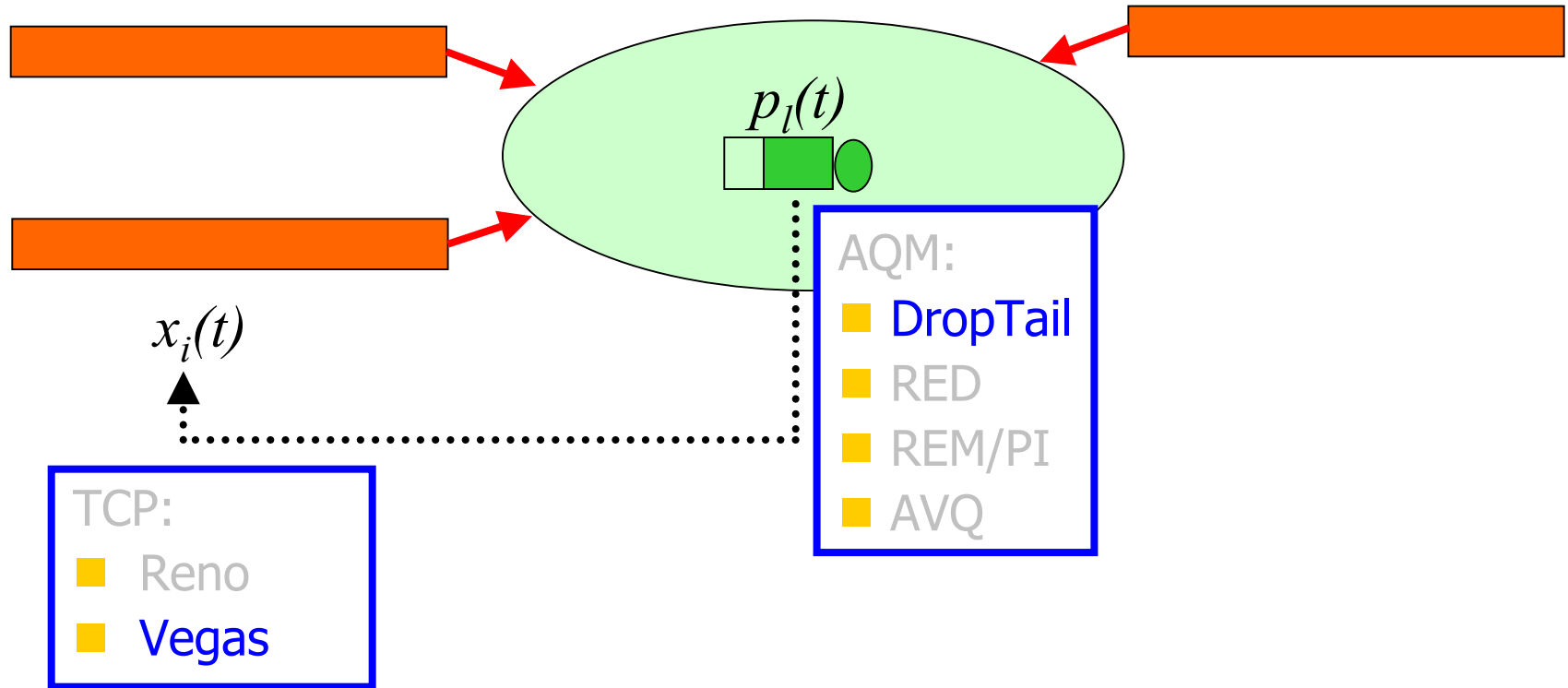


- TCP/AQM
 - Reno/RED
- Equilibrium
 - Duality model
- Stability & optimal control
 - Linearized model
- A scalable control

TCP & AQM



TCP & AQM



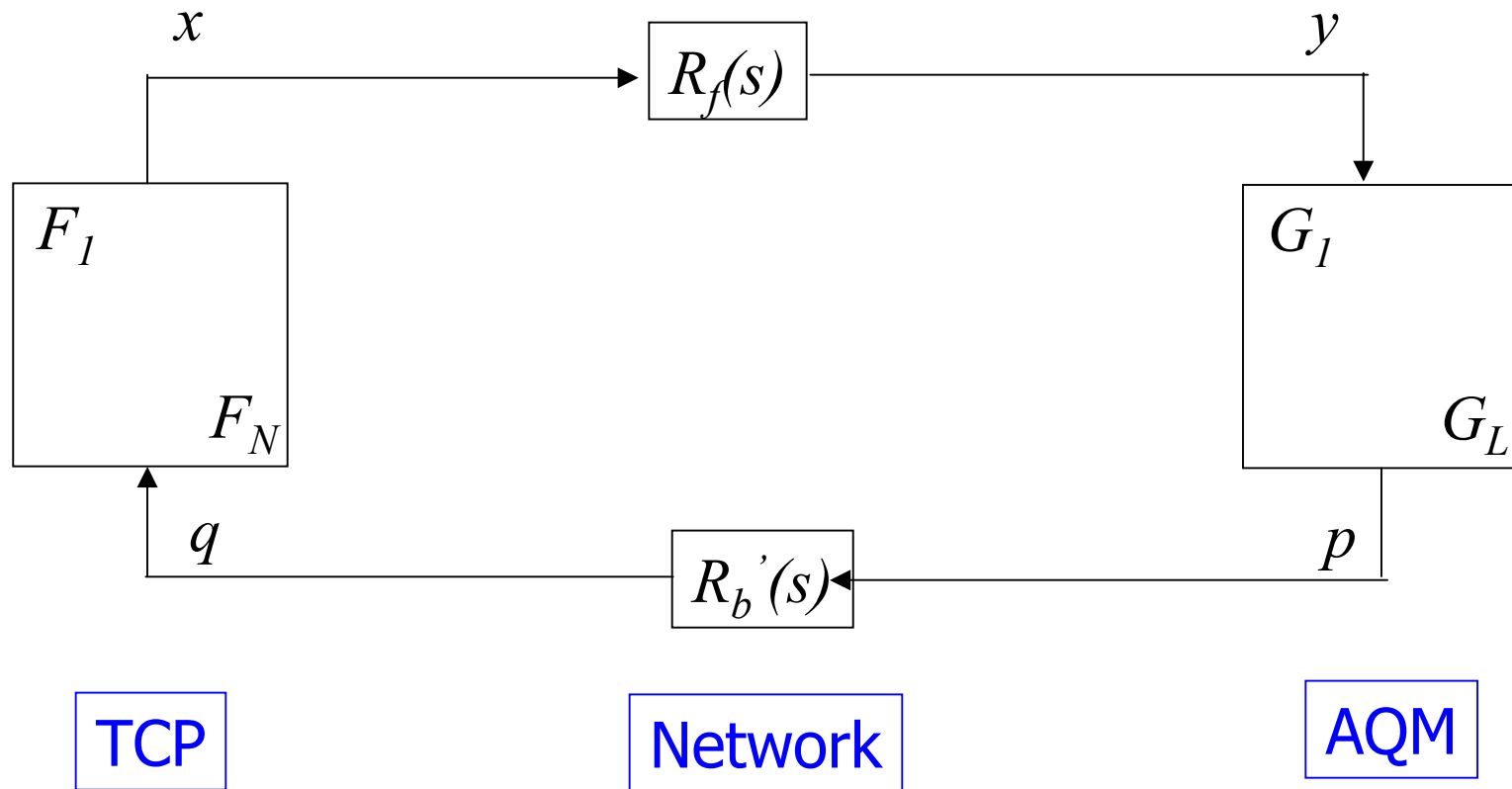
TCP/AQM



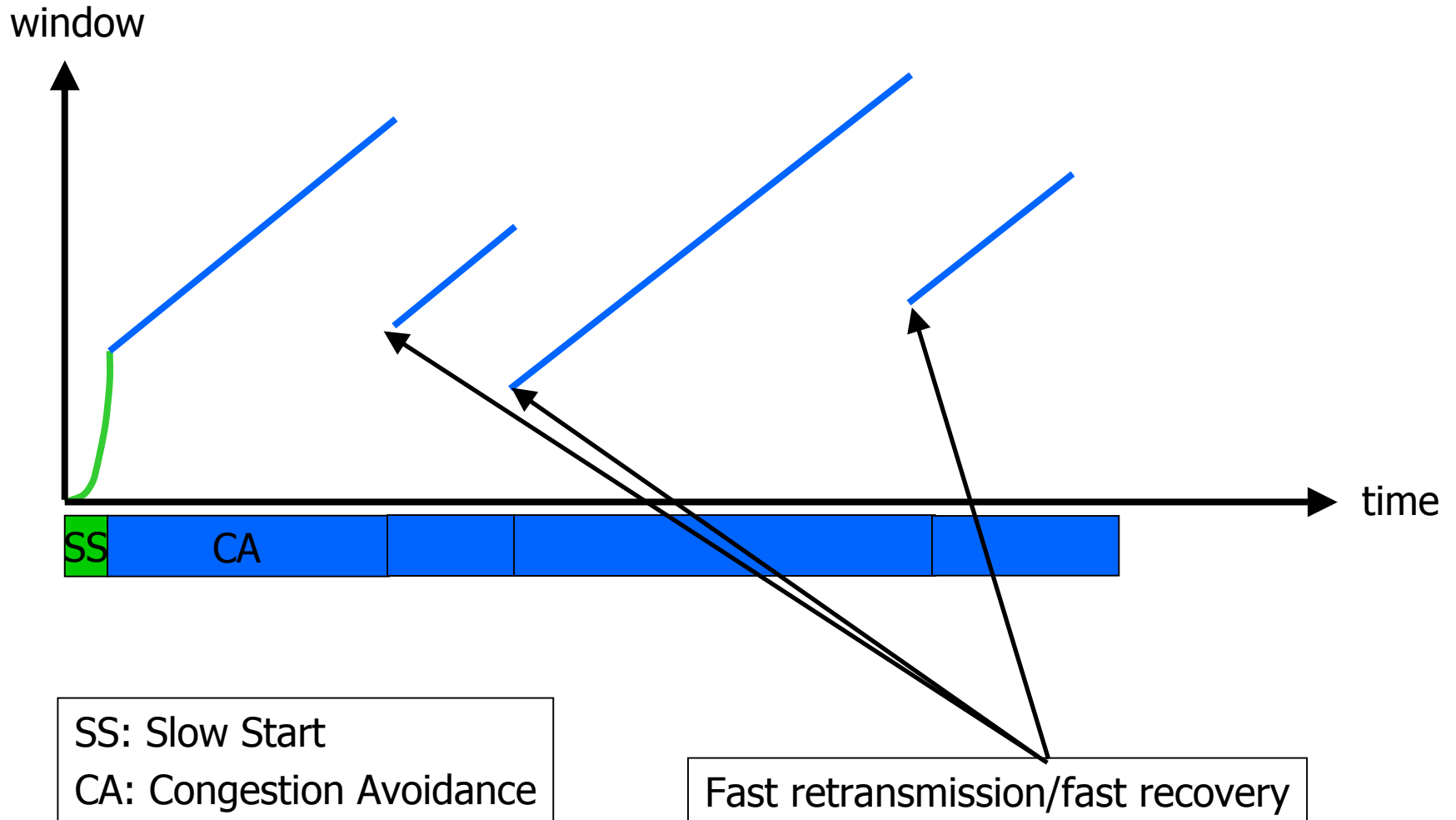
- Tahoe (Jacobson 1988)
 - Slow Start
 - Congestion Avoidance
 - Fast Retransmit
- Reno (Jacobson 1990)
 - Fast Recovery
- Vegas (Brakmo & Peterson 1994)
 - New Congestion Avoidance
- RED (Floyd & Jacobson 1993)
- REM/PI (Athuraliya et al 2000, Hollot et al 2001)
- AVQ (Kunniyur & Srikant 2001)

Model structure

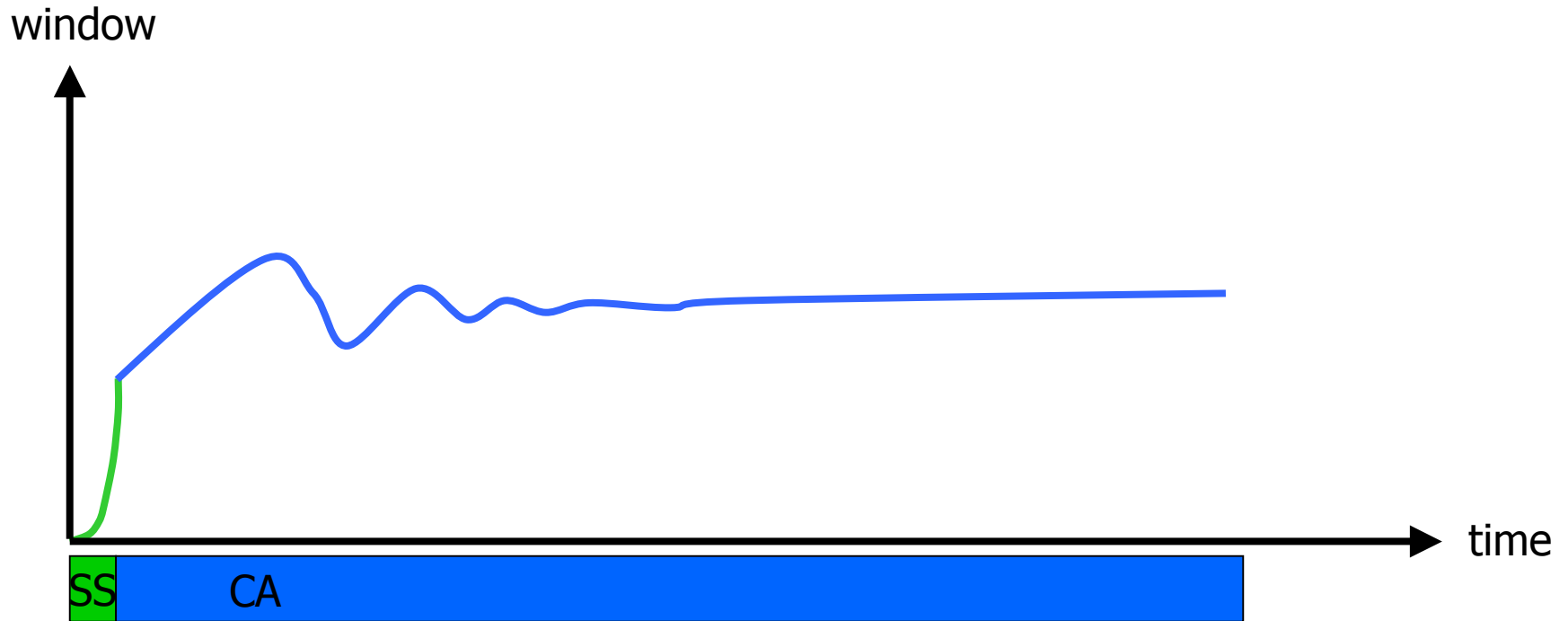
Multi-link multi-source network



TCP Reno (Jacobson 1990)



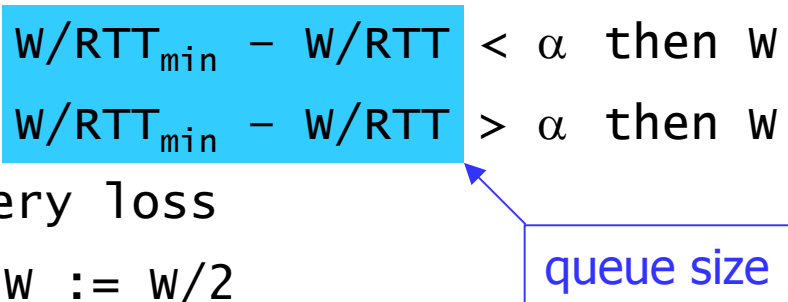
TCP Vegas (Brakmo & Peterson 1994)



- Converges, no retransmission
- ... provided buffer is large enough

Vegas model

```
for every RTT
{
  if  $W/RTT_{\min} - W/RTT < \alpha$  then  $W++$ 
  if  $W/RTT_{\min} - W/RTT > \alpha$  then  $W--$ 
}
for every loss
   $W := W/2$ 
```



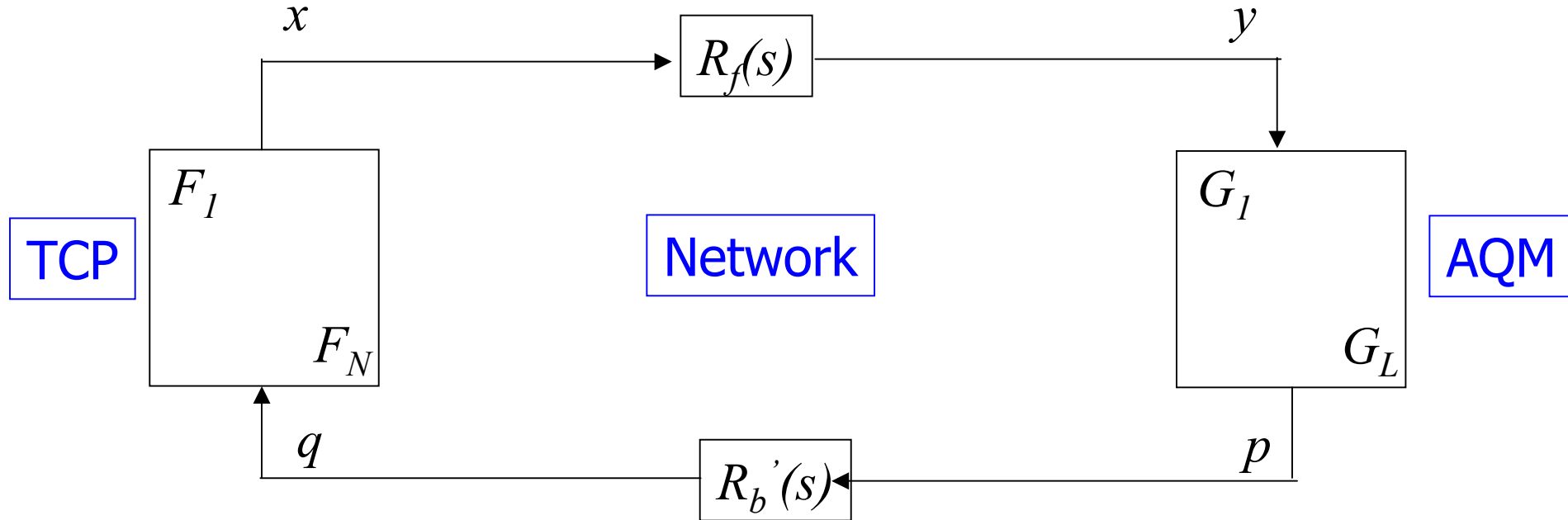
$$F_i: \quad x_i(t+1) = \begin{cases} x_i(t) + \frac{1}{D_i^2} & \text{if } x_i(t)q_i(t) < \alpha_i d_i \end{cases}$$

$$x_i(t+1) = \begin{cases} x_i(t) - \frac{1}{D_i^2} & \text{if } x_i(t)q_i(t) > \alpha_i d_i \end{cases}$$

$$x_s(t+1) = x_s(t) \quad \text{else}$$

$$G_l: \quad p_l(t+1) = [p_l(t) + y_l(t)/c_l - 1]^+$$

Vegas model



$$F_i = x_i(t) + \frac{1}{(d_i + q_i(t))^2} \operatorname{sgn}(\alpha_i d_i - x_i(t) q_i(t))$$

$$G_l = \left(p_l(t) + \frac{y_l(t)}{c_l} - 1 \right)^+$$

Overview

Protocol (Reno, Vegas, RED, REM/PI...)

$$\begin{aligned}x(t+1) &= F(p(t), x(t)) \\ p(t+1) &= G(p(t), x(t))\end{aligned}$$

Equilibrium

- Performance
 - Throughput, loss, delay
- Fairness
- Utility

Dynamics

- Local stability
- Cost of stabilization

Outline



- TCP/AQM
 - Reno/RED
- Equilibrium
 - Duality model
- Stability
 - Linearized model
- A scalable control

Flow control

- Interaction of **source rates** $x_s(t)$ and **congestion measures** $p_l(t)$
- Duality theory
 - They are **primal** and **dual** variables
 - Flow control is optimization process
- Example congestion measure
 - Loss (Reno)
 - Queueing delay (Vegas)

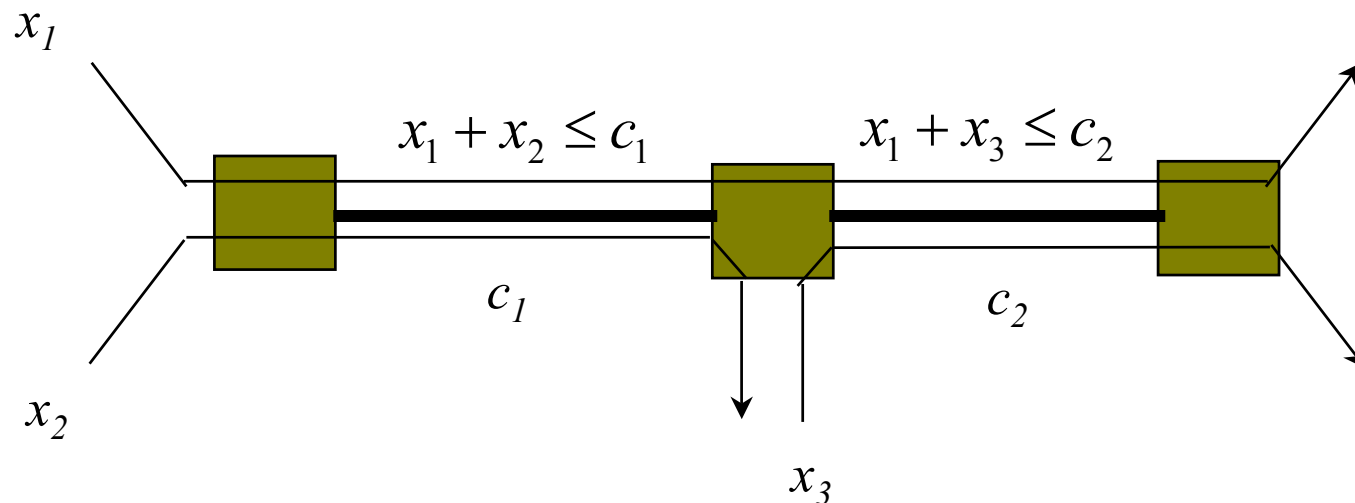
Model

■ Network

- Links l of capacities c_l

■ Sources i

- $L(s)$ - links used by source i
- $U_i(x_i)$ - utility if source rate = x_i



Primal problem



$$\begin{array}{ll} \max_{x_i \geq 0} & \sum_i U_i(x_i) \\ \text{subject to} & y_l \leq c_l, \quad \forall l \in L \end{array}$$

- Assumptions
 - Strictly concave increasing U_i
- Unique optimal rates x_i exist
- Direct solution impractical

Related Work



- Formulation
 - Kelly 1997
- Penalty function approach
 - Kelly, Maulloo & Tan 1998
 - Kunniyur & Srikant 2000
- Duality approach
 - Low & Lapsley 1999
 - Athuraliya & Low 2000
- Extensions
 - Mo & Walrand 2000
 - La & Anantharam 2000
- Dynamics
 - Johari & Tan 2000, Massoulié 2000, Vinnicombe 2000, ...
 - Hollot, Misra, Towsley & Gong 2001
 - Paganini 2000, Paganini, Doyle, Low 2001, ...

Duality Approach

$$\text{Primal: } \max_{x_s \geq 0} \sum_s U_s(x_s) \quad \text{subject to } x^l \leq c_l, \quad \forall l \in L$$

$$\text{Dual: } \min_{p \geq 0} D(p) = \left(\max_{x_s \geq 0} \sum_s U_s(x_s) + \sum_l p_l (c_l - x^l) \right)$$

Primal-dual algorithm:

$$x(t+1) = F(p(t), x(t))$$

$$p(t+1) = G(p(t), x(t))$$

Duality Model of TCP

- Source algorithm iterates on rates
- Link algorithm iterates on prices
- With **different** utility functions

Primal-dual algorithm:

$$x(t+1) = F(p(t), x(t)) \quad \leftarrow \text{Reno, Vegas}$$

$$p(t+1) = G(p(t), x(t)) \quad \leftarrow \text{DropTail, RED, REM}$$

Duality Model of TCP

(x^*, p^*) primal-dual optimal if and only if

$$y_l^* \leq c_l \quad \text{with equality if } p_l^* > 0$$

Primal-dual algorithm:

$$x(t+1) = F(p(t), x(t)) \quad \leftarrow \text{Reno, Vegas}$$

$$p(t+1) = G(p(t), x(t)) \quad \leftarrow \text{DropTail, RED, REM}$$

Duality Model of TCP

Any link algorithm that stabilizes queue

- generates Lagrange multipliers
- solves dual problem

Primal-dual algorithm:

$$x(t+1) = F(p(t), x(t)) \leftarrow \text{Reno, Vegas}$$

$$p(t+1) = G(p(t), x(t)) \leftarrow \text{DropTail, RED, REM}$$

Gradient algorithm

■ Gradient algorithm

source: $x_i(t+1) = U_i'^{-1}(q_i(t))$

link: $p_l(t+1) = [p_l(t) + \gamma_l(y_l(t) - c_l)]^+$

Theorem (Low & Lapsley '99)

Converge to optimal rates in distributed
asynchronous environment

Gradient algorithm

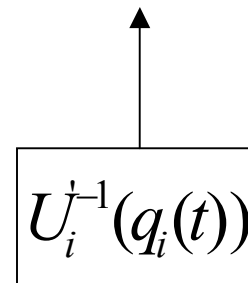
■ Gradient algorithm

source: $x_i(t+1) = U_i^{-1}(q_i(t))$

link: $p_l(t+1) = [p_l(t) + \gamma_l(y_l(t) - c_l)]^+$

■ Vegas: **approximate** gradient algorithm

$$F_i = x_i(t) + \frac{1}{(d_i + q_i(t))^2} \operatorname{sgn}(\bar{x}_i(t) - x_i(t))$$


$$U_i^{-1}(q_i(t))$$

Summary: equilibrium

■ Flow control problem

$$\begin{aligned} \max_{x_s \geq 0} \quad & \sum_s U_s(x_s) \\ \text{subject to} \quad & x^l \leq c_l, \quad \forall l \in L \end{aligned}$$

■ Primal-dual algorithm

$$x(t+1) = F(p(t), x(t)) \quad \leftarrow \text{Reno, Vegas}$$

$$p(t+1) = G(p(t), x(t)) \quad \leftarrow \text{DropTail, RED, REM}$$

■ TCP/AQM

- Maximize aggregate source utility
- With **different** utility functions

Implications



- Performance
 - Rate, delay, queue, loss
- Fairness
 - Utility function
- Persistent congestion

Performance

■ Delay

- Congestion measures: end to end *queueing* delay

- Sets rate $x_s(t) = \alpha_s \frac{d_s}{q_s(t)}$

- Equilibrium condition: Little's Law

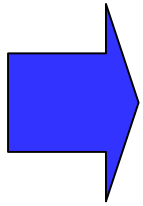
■ Loss

- No loss if converge (with sufficient buffer)

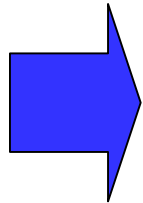
- Otherwise: revert to Reno (loss unavoidable)

Vegas Utility

■ Equilibrium $(x, p) = (F, G)$



$$U_i(x_i) = \alpha_i d_i \log x_i$$



Proportional fairness

Validation (L. Wang, Princeton)

	Source 1	Source 3	Source 5
RTT (ms)	17.1 (17)	21.9 (22)	41.9 (42)
Rate (pkts/s)	1205 (1200)	1228 (1200)	1161 (1200)
Window (pkts)	20.5 (20.4)	27 (26.4)	49.8 (50.4)
Avg backlog (pkts)	9.8 (10)		

measured theory

- NS-2 simulation, single link, capacity = 6 pkts/ms
- 5 sources with different propagation delays, $\alpha_s = 2$ pkts/RTT

Persistent congestion

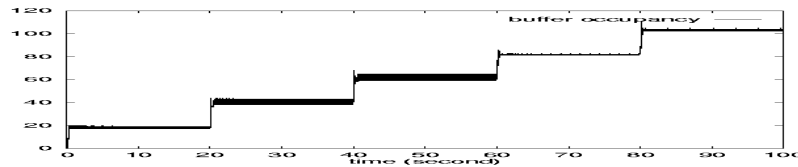
- Vegas exploits buffer process to compute prices (queueing delays)
- Persistent congestion due to
 - Coupling of buffer & price
 - Error in propagation delay estimation
- Consequences
 - Excessive backlog
 - Unfairness to older sources

Theorem (Low, Peterson, Wang '02)

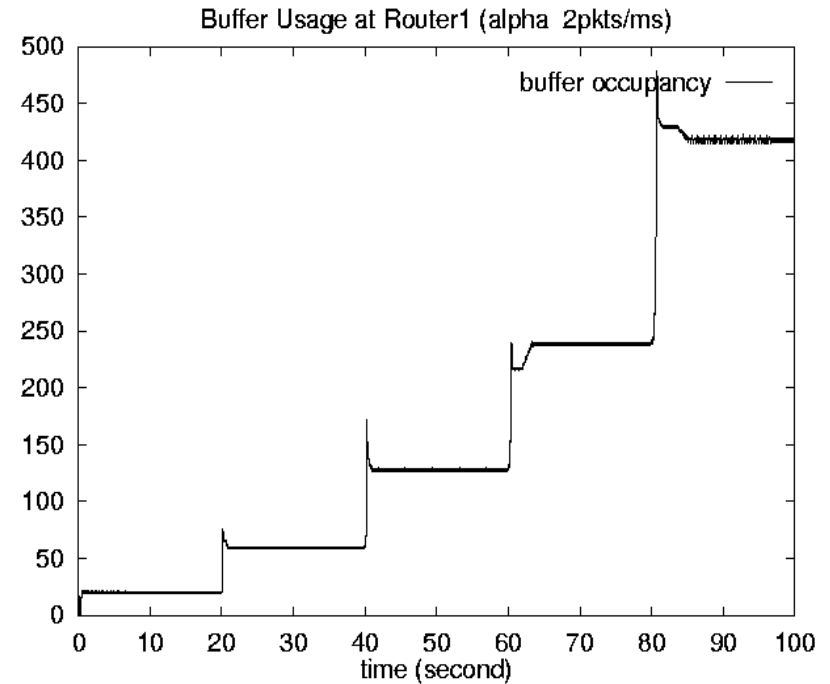
A relative error of ε_s in propagation delay estimation distorts the utility function to

$$\hat{U}_s(x_s) = (1 + \varepsilon_s) \alpha_s d_s \log x_s + \varepsilon_s d_s x_s$$

Validation (L. Wang, Princeton)



Without estimation error



With estimation error

- Single link, capacity = 6 pkt/ms, $\alpha_s = 2$ pkts/ms, $d_s = 10$ ms
- With finite buffer: Vegas reverts to Reno

Validation (L. Wang, Princeton)

Source rates (pkts/ms)

#	src1	src2	src3	src4	src5
1	5.98 (6)				
2	2.05 (2)	3.92 (4)			
3	0.96 (0.94)	1.46 (1.49)	3.54 (3.57)		
4	0.51 (0.50)	0.72 (0.73)	1.34 (1.35)	3.38 (3.39)	
5	0.29 (0.29)	0.40 (0.40)	0.68 (0.67)	1.30 (1.30)	3.28 (3.34)

#	queue (pkts)	baseRTT (ms)
1	19.8 (20)	10.18 (10.18)
2	59.0 (60)	13.36 (13.51)
3	127.3 (127)	20.17 (20.28)
4	237.5 (238)	31.50 (31.50)
5	416.3 (416)	49.86 (49.80)

Outline

Protocol (Reno, Vegas, RED, REM/PI...)

$$\begin{aligned}x(t+1) &= F(p(t), x(t)) \\ p(t+1) &= G(p(t), x(t))\end{aligned}$$

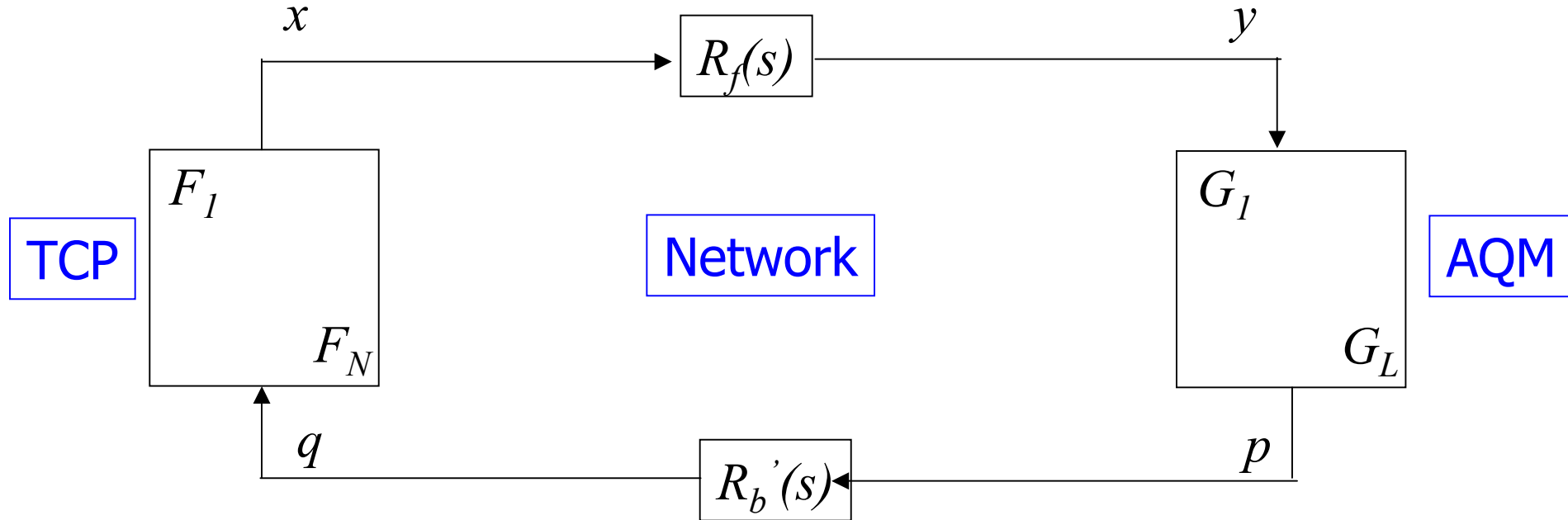
Equilibrium

- Performance
 - Throughput, loss, delay
- Fairness
- Utility

Dynamics

- Local stability
- Cost of stabilization

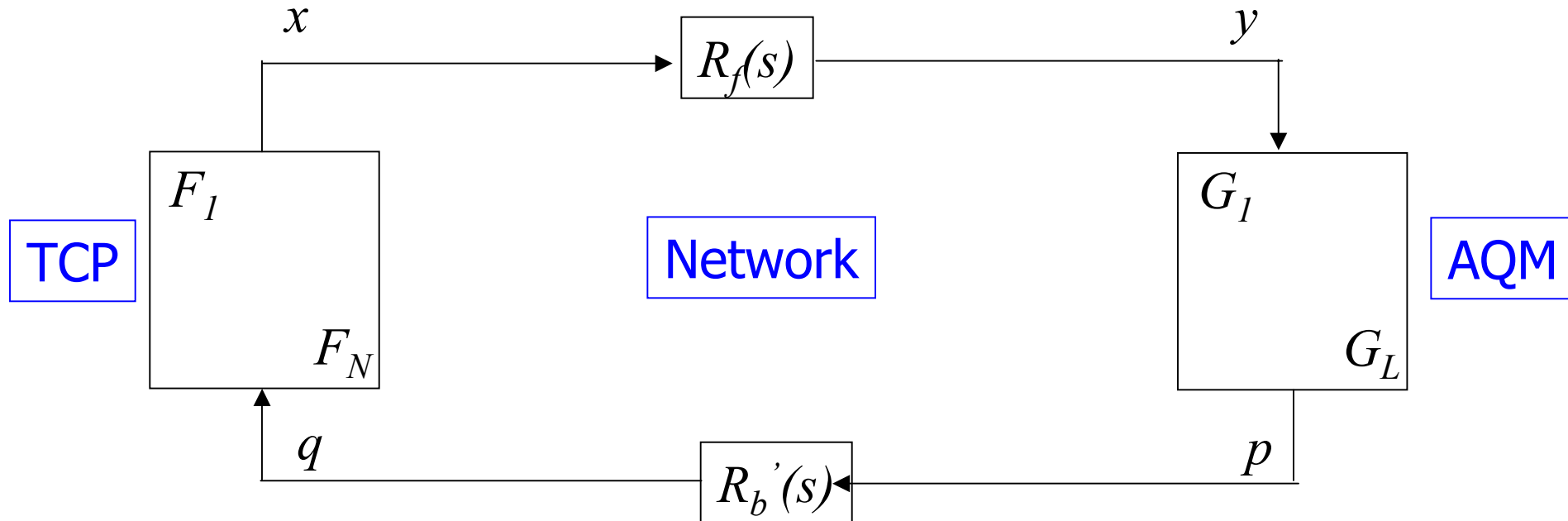
Vegas model



$$F_i = x_i(t) + \frac{1}{(d_i + q_i(t))^2} \operatorname{sgn}(\alpha_i d_i - x_i(t) q_i(t))$$

$$G_l = \left(p_l(t) + \frac{y_l(t)}{c_l} - 1 \right)^+$$

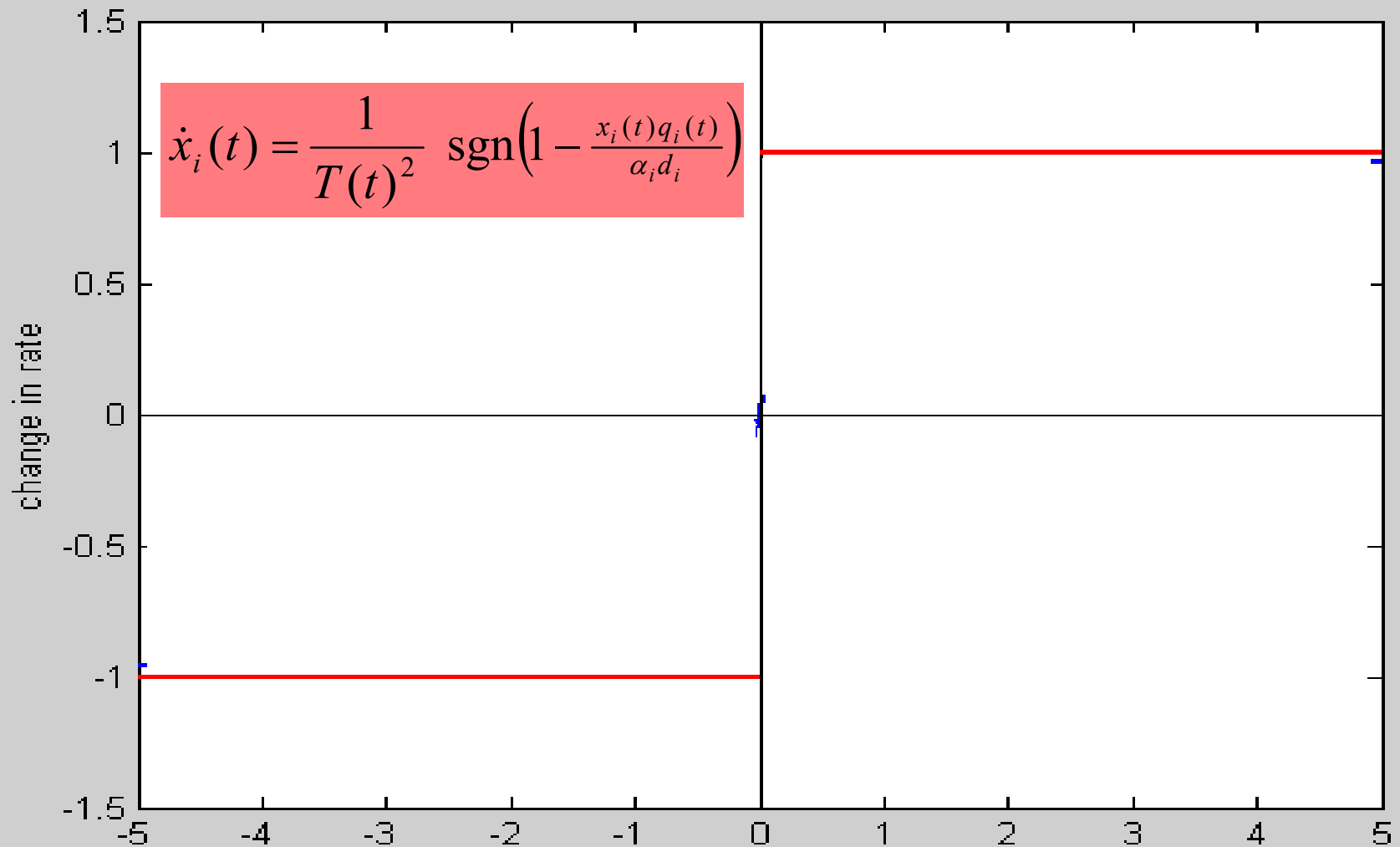
Vegas model



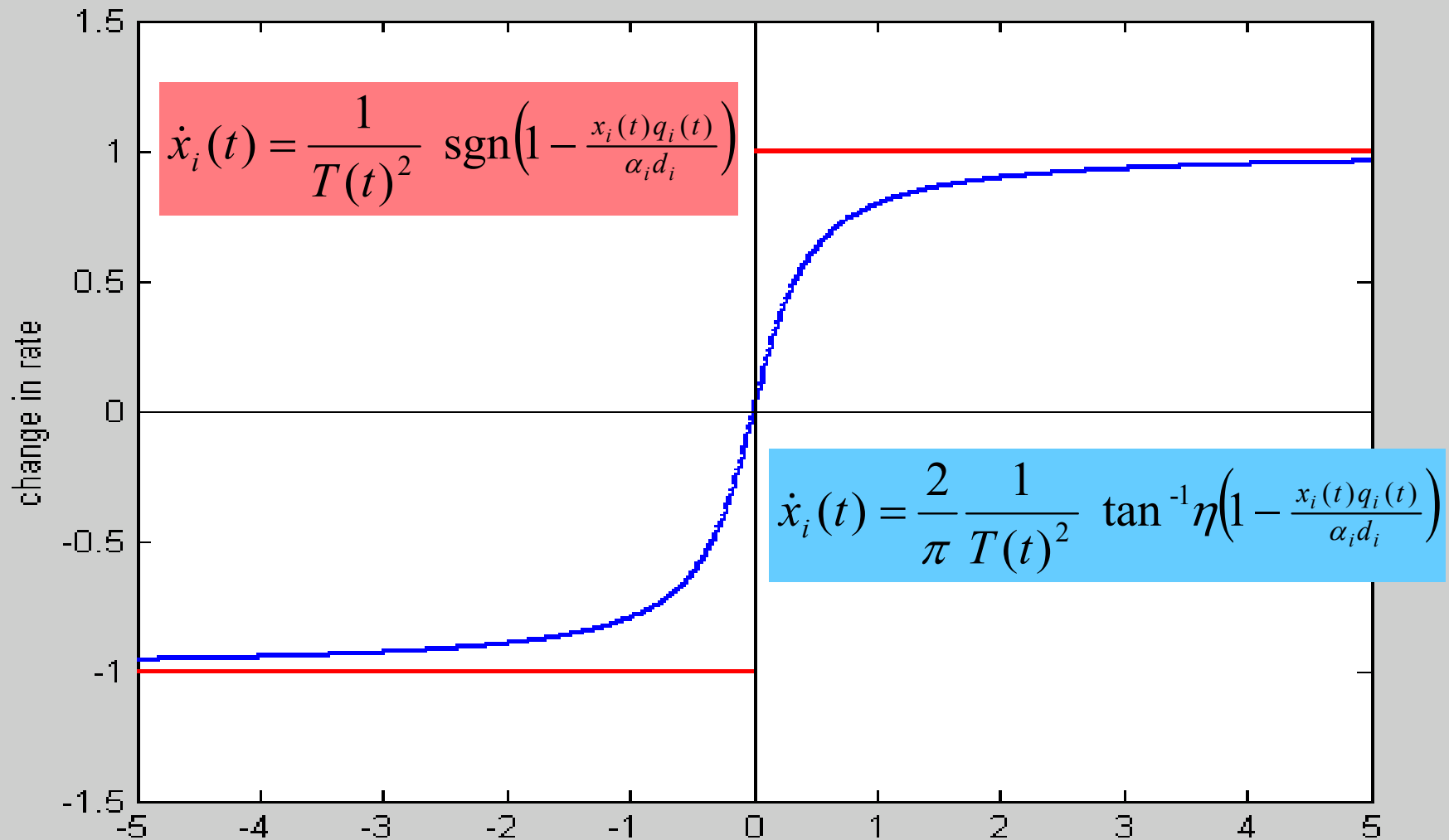
$$[R_f]_{li} = e^{-s\bar{\tau}_{li}} \quad \text{if source } i \text{ uses link } l$$

$$[R_b]_{li} = e^{-s\bar{\tau}_{li}} \quad \text{if source } i \text{ uses link } l$$

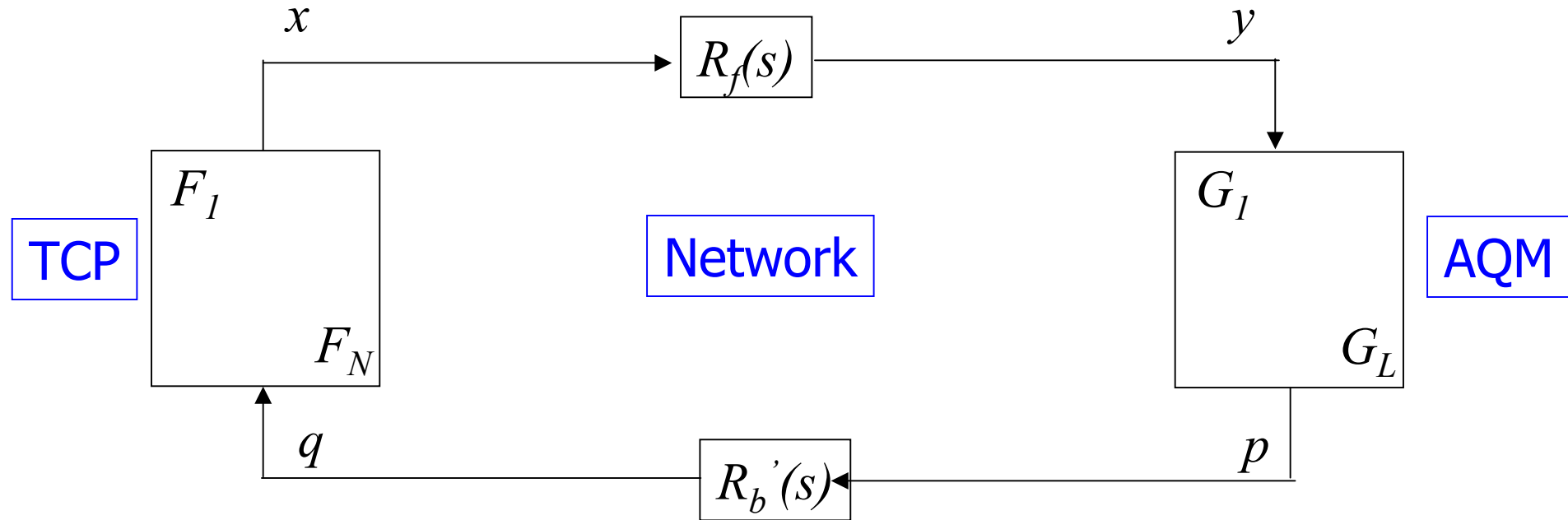
Approximate model



Approximate model



Linearized model



$$\partial \dot{x}_i = -\frac{x_i}{q_i} \frac{a_i}{sT_i + a_i} \partial q_i$$

$$\partial \dot{p}_l = -\frac{\gamma}{c_l} \partial y_l$$

$$a_i = -\frac{2\eta}{\pi} \frac{1}{x_i T_i}$$

γ controls equilibrium delay

Stability

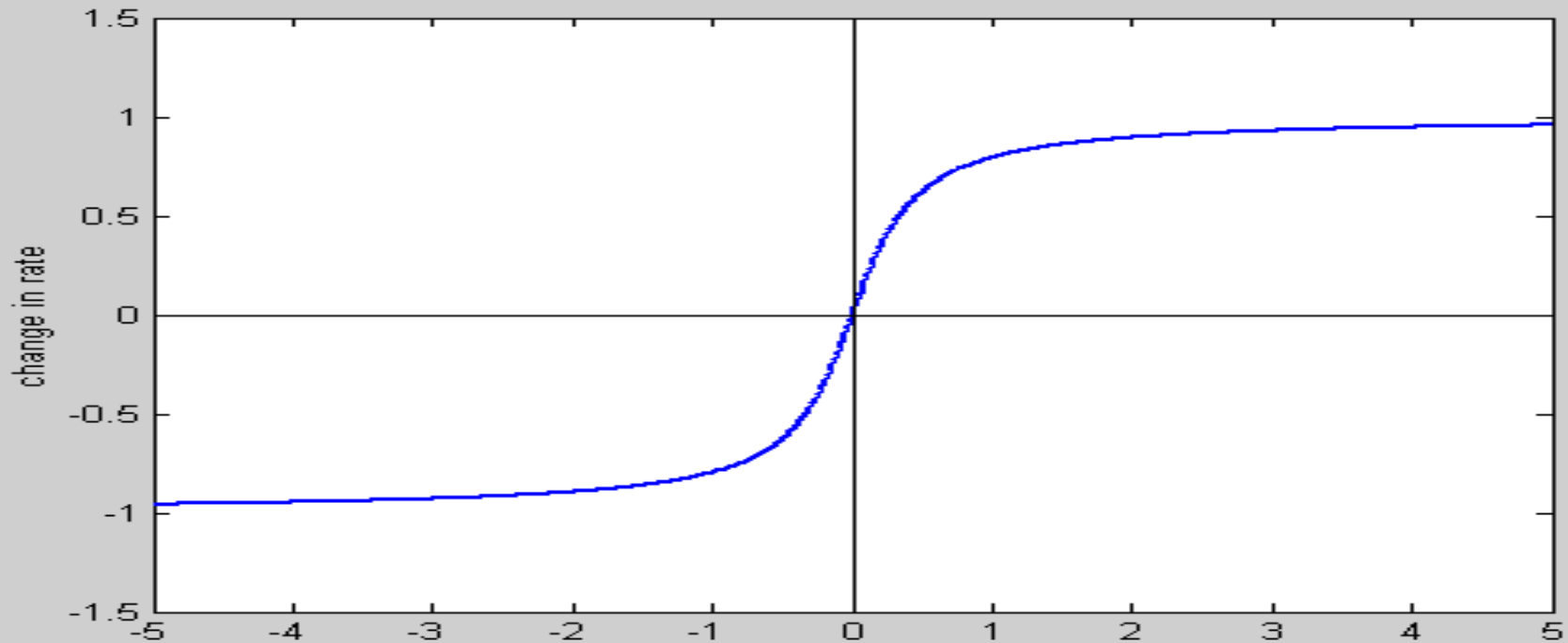
Theorem (Choe & Low, '02)

Locally asymptotically stable if

$$\frac{\text{link queueing delay}}{\text{round trip time}} > \frac{a_i}{\min a_i} \frac{\sin \omega_c}{\omega_c} > 0.63$$

Cannot be satisfied with >1 bottleneck link!

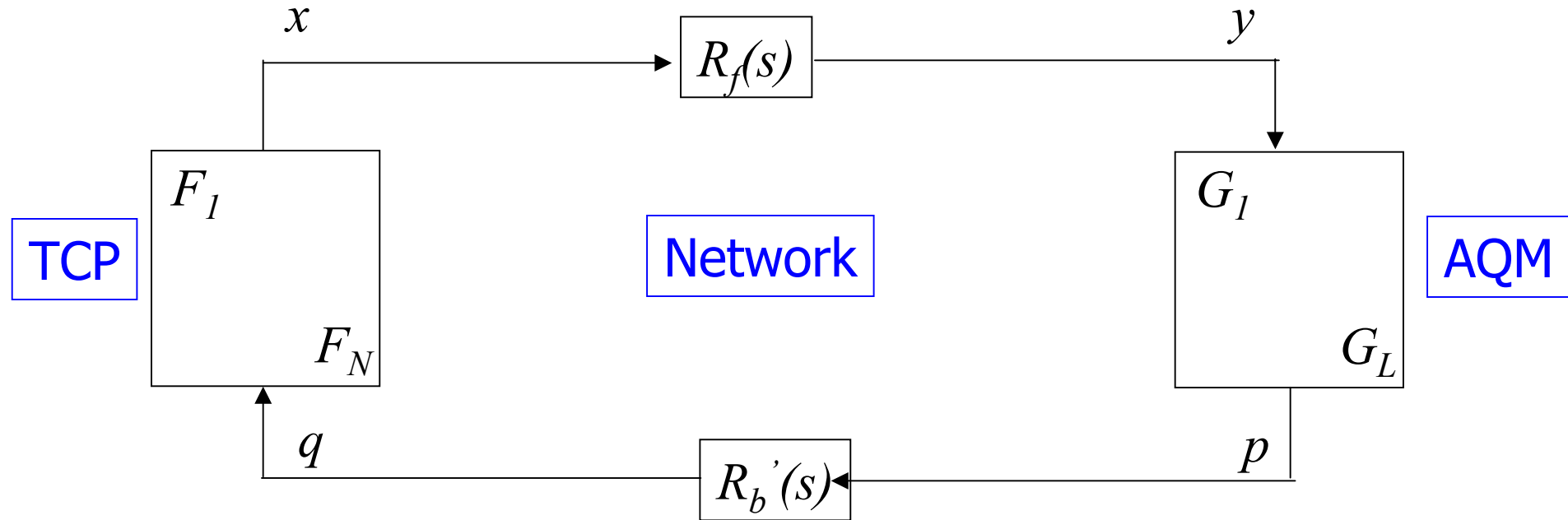
Stabilized Vegas



$$\dot{x}_i(t) = \frac{2}{\pi} \frac{1}{T(t)^2} \tan^{-1} \eta \left(1 - \frac{x_i(t) q_i(t)}{\alpha_i d_i} \right)$$

$$\dot{x}_i(t) = \frac{2}{\pi} \frac{1}{T(t)^2} \tan^{-1} \eta \left(1 - \frac{x_i(t) q_i(t)}{\alpha_i d_i} - \kappa_i \dot{q}_i(t) \right)$$

Linearized model

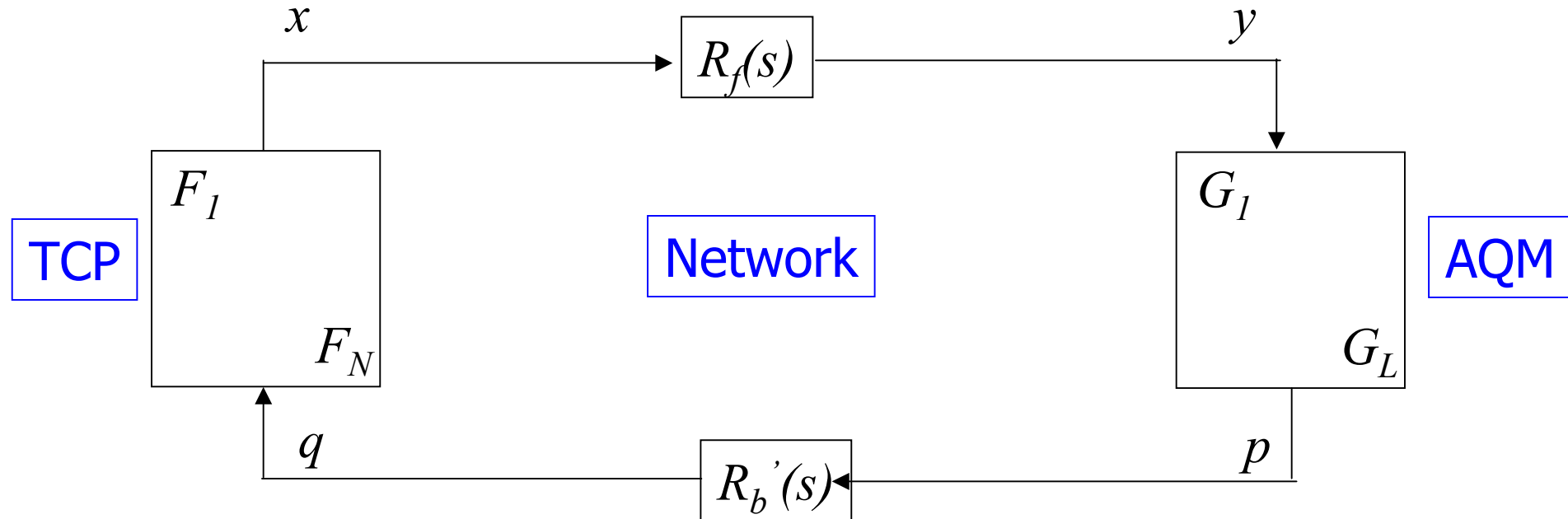


$$\partial \dot{x}_i = -\frac{x_i}{q_i} \frac{a_i}{sT_i + a_i} \partial q_i$$

$$\partial \dot{p}_l = -\frac{\gamma}{c_l} \partial y_l$$

γ controls equilibrium delay

Linearized model



$$\partial \dot{x}_i = -b_i \frac{s + a_i}{s + \mu_i a_i} \partial q_i$$

$$\partial \dot{p}_l = -\frac{\gamma}{c_l} \partial y_l$$

choose $a_i = a$, $\alpha_i = \mu$

γ controls equilibrium delay

Stability

Theorem (Choe & Low, '02)

Locally asymptotically stable if

$$M \frac{\text{round trip time}}{\text{round trip queueing}} < \sigma(a, \mu)$$

example

■ $LHS < 10 * 10 = 100$

■ $a = 0.1, \mu = 0.015 \rightarrow \sigma(a, \mu) = 120$

Stability



Theorem (Choe & Low, '02)

Locally asymptotically stable if

$$M \frac{\text{round trip time}}{\text{round trip queueing}} < \sigma(a, \mu)$$

Application

- Stabilized TCP with current routers
- Queueing delay as congestion measure has the right scaling
- Incremental deployment with ECN

Vertical decomposition

■ Utility maximization

$$\begin{aligned} \text{Primal : } & \max_R \max_{x_i \geq 0} \sum_i U_i(x_i) \\ & \text{subject to } y_l \leq c_l, \quad \forall l \in L \end{aligned}$$

$$\text{Dual : } \min_{p \geq 0} \sum_i \max_{x_i \geq 0} \left(U_i(x_i) - x_i \min_{R_i} \sum_l R_{li} p_l \right) + \sum_l p_l c_l$$

Shortest path routing!

Vertical decomposition

■ Utility maximization

$$\begin{aligned} \text{Primal : } & \max_R \max_{x_i \geq 0} \sum_i U_i(x_i) \\ & \text{subject to } y_l \leq c_l, \quad \forall l \in L \end{aligned}$$

$$\text{Dual : } \min_{p \geq 0} \sum_i \max_{x_i \geq 0} \left(U_i(x_i) - x_i \min_{R_i} \sum_l R_{li} p_l \right) + \sum_l p_l c_l$$

Can shortest-path routing (IP) and TCP/AQM maximize utility?

Vertical decomposition



Theorem (Wang, Li, Low, Doyle '02)

Primal problem is NP-hard

- Cannot be solved by shortest-path routing and TCP/AQM
- Shortest path routing based on prices can be unstable
- Even when stable, there can be duality gap
- How well does TCP/AQM/IP solve it approximately?



Papers



netlab.caltech.edu

- A duality model of TCP flow controls (*ITC, Sept 2000*)
- Optimization flow control, I: basic algorithm & convergence (*ToN, 7(6), Dec 1999*)
- Understanding Vegas: a duality model (*J. ACM, 2002*)
- Scalable laws for stable network congestion control (*CDC, 2001*)
- REM: active queue management (*Network, May/June 2001*)