

# On congestion control, multipath routing, and admission control

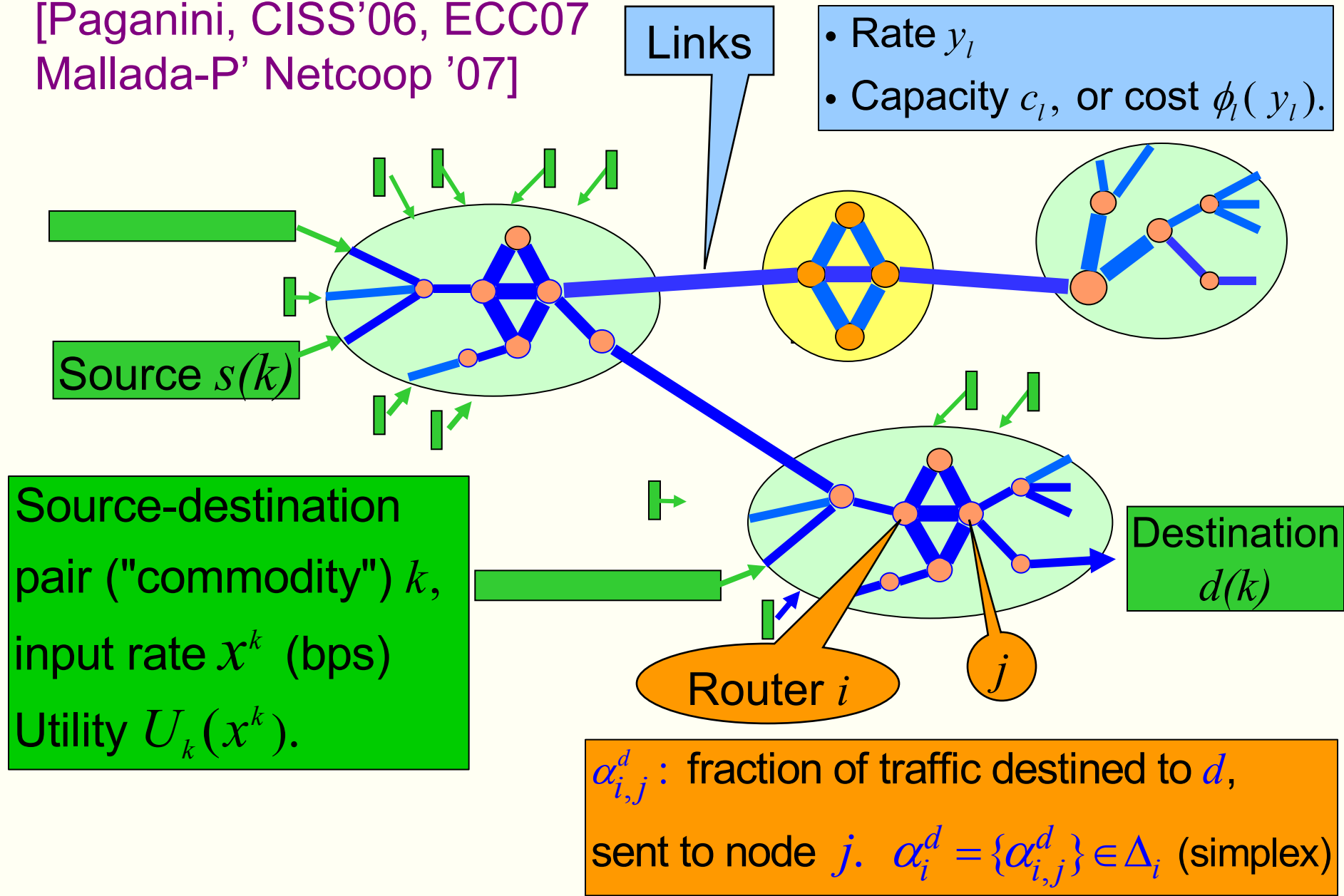
Fernando Paganini

Universidad ORT Uruguay

- Part I (with Enrique Mallada). Combined congestion control and node-based multipath routing: new results on stability since CISS'06.
- Part II (with Andrés Ferragut, in CISS'08 paper). Achieving network stability and user fairness through admission control of TCP connections.

# Congestion control with multipath routing

[Paganini, CISS'06, ECC07  
Mallada-P' Netcoop '07]

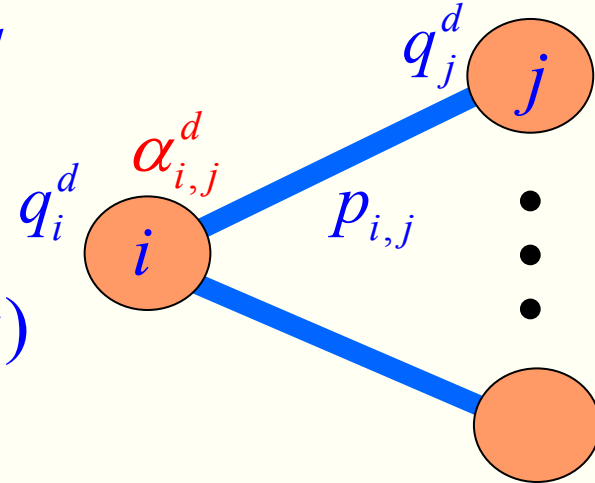


# Congestion prices

- Define: link congestion price  $p_l$ . (e.g.,  $p_l = \phi'(y_l)$  or  $\frac{dp_l}{dt} = \gamma_l [y_l - c_l]$ .)
- $q_i^d$ : average price from node  $i$  to destination  $d$

$$q_d^d = 0, \quad q_i^d = \sum_{j:(i,j) \in L} \alpha_{i,j}^d (p_{i,j} + q_j^d), \quad i \neq d$$

$$q^k := q_{s(k)}^{d(k)}, \quad \text{average price seen by source } s(k)$$

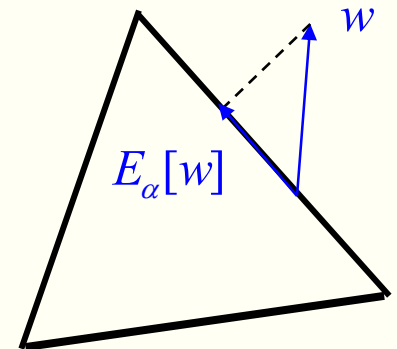


## Multipath routing control.

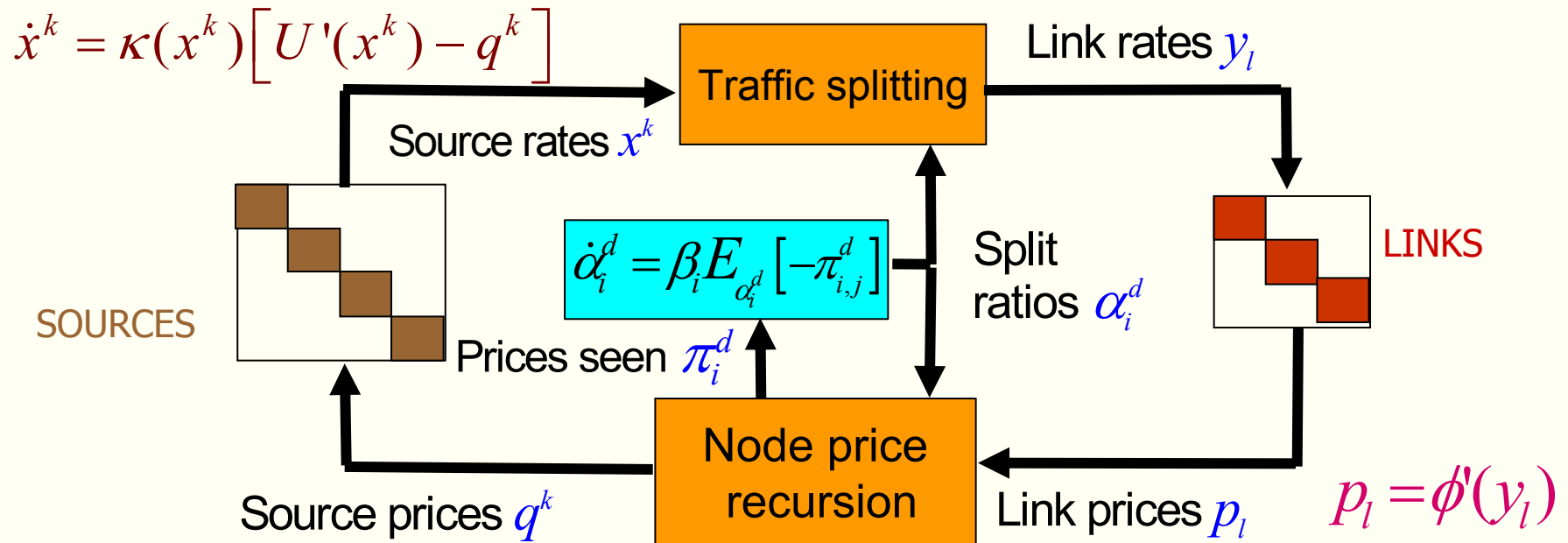
Routers control  $\alpha_i^d := \{\alpha_{i,j}^d\}_{(i,j) \in L}$  based on seen prices  $\pi_i^d := \{p_{i,j} + q_j^d\}_{(i,j) \in L}$

- First choice (essentially from Gallager '77):  
follow negative price gradient.  $\dot{\alpha}_i^d = \beta_i E_{\alpha_i^d} [-\pi_i^d]$ .

The projection  $E_{\alpha_i^d}$  keeps  $\alpha_i^d \in \Delta_i$ .



# Primal congestion control under gradient control of routing fractions



**Theorem:** with these dynamics the system converges globally to the optimum of the SURPLUS problem,

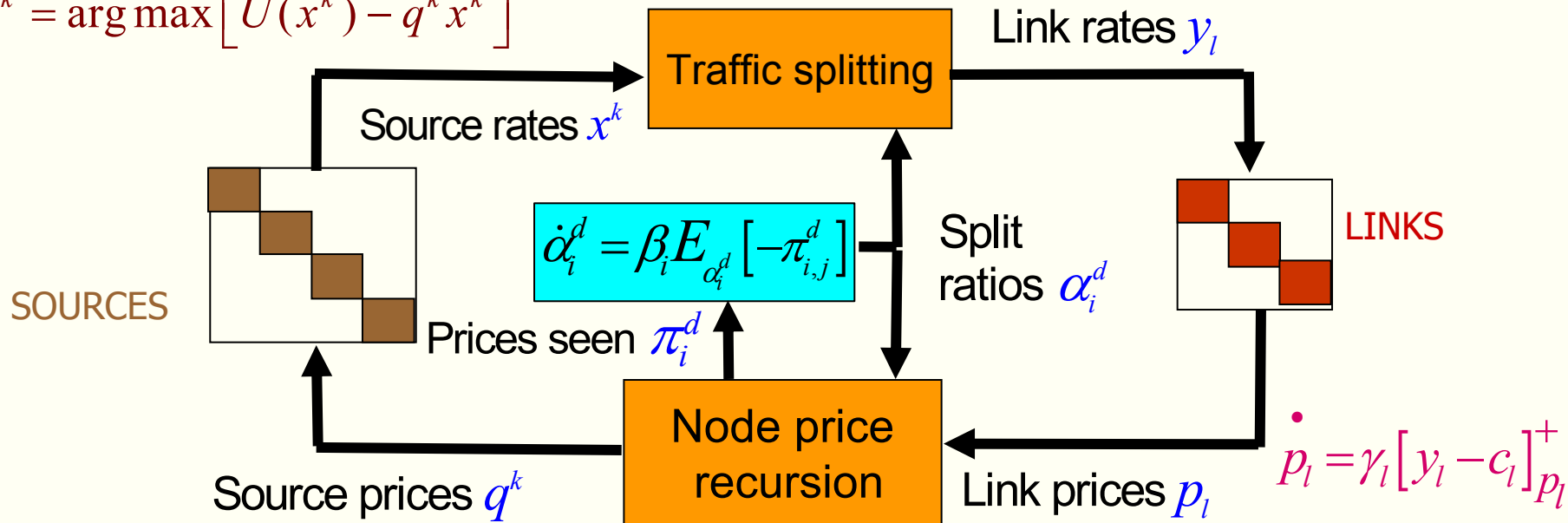
$$\max S := \sum U^k(x^k) - \sum \phi_l(y_l)$$

[See P' CISS '06], extends Gallager '77.

Also in [Xi & Yeh CISS'06], combined with wireless power control.

# Dual congestion control under gradient control of routing fractions

$$x^k = \arg \max [U(x^k) - q^k x^k]$$



**Equilibrium:** solution to optimal WELFARE problem,

$$\max \sum U_k(x^k) \quad \text{subject to } y \leq c$$

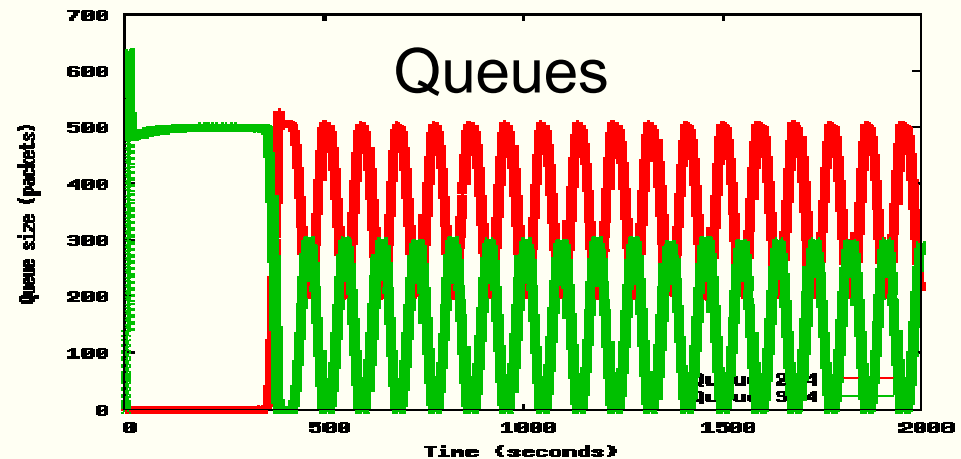
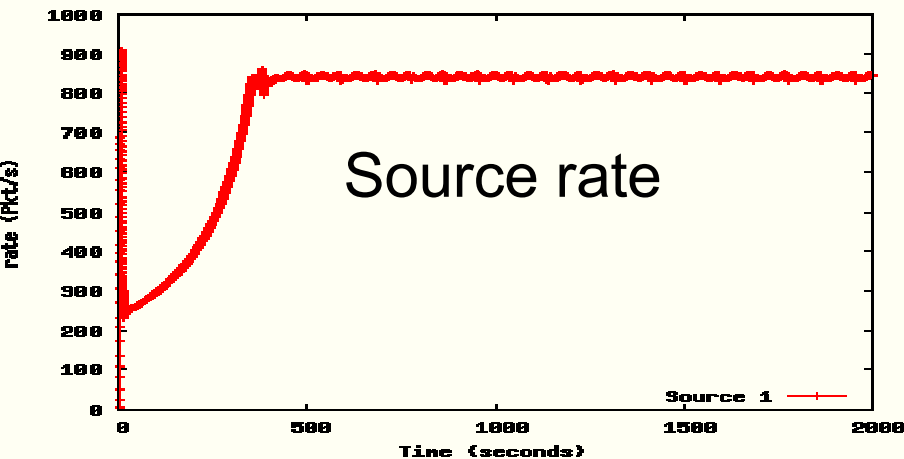
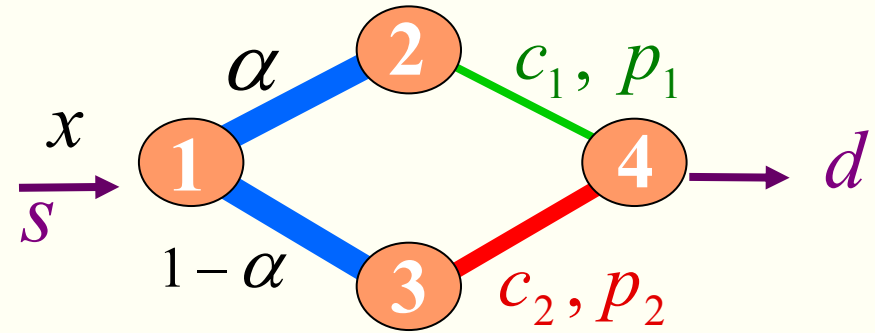
Convergence to equilibrium?

No! Simple examples exhibit harmonic oscillations.

Which model is correct for queuing delay:

Static  $p_l = \varphi(y_l)$  or integrator  $\dot{p}_l = \frac{1}{c_l} [y_l - c_l]_p^+$ ?

Example of instability,  
packet simulation in ns2  
Single source, two bottlenecks.

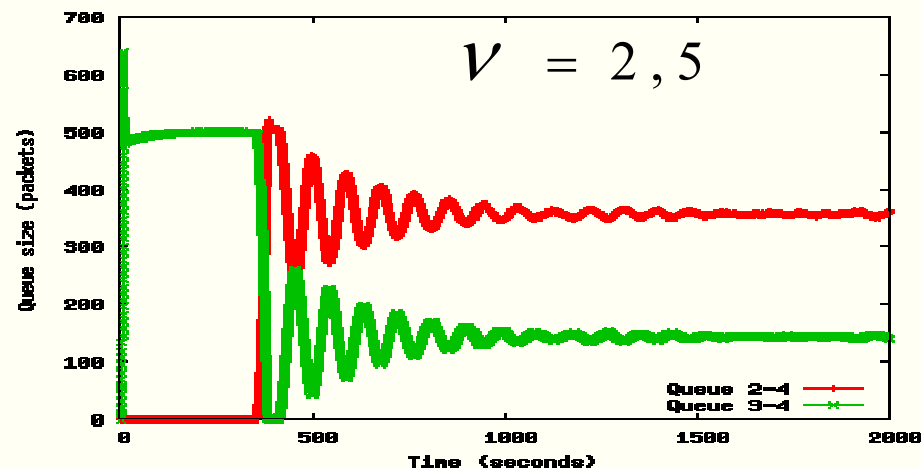
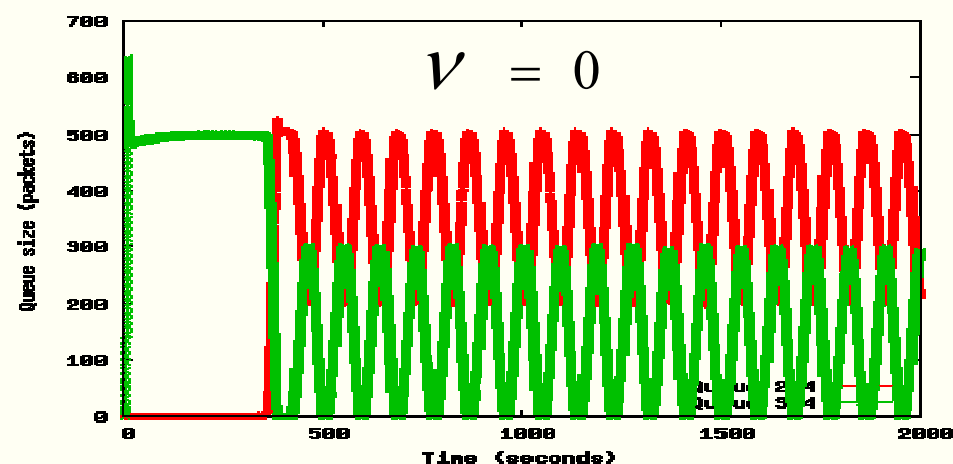


- Dual model of queue is more appropriate.
- Indeed, it predicts correctly oscillation period.

# Solving the problem

Adapt  $\alpha_i^d$  based on anticipated (rather than current) price  $\pi_i^{d*} = \pi_i^d + V_i \dot{\pi}_i^d$

In control terms, add derivative action. Same equilibrium. Simulations:



**Theorems** [P'-Mallada, submitted to ToN, CDC'08]

the equilibrium point (optimum  $\max \sum U^k(x^k)$ ) is:

- locally asymptotically stable in an arbitrary network
- globally asymptotically stable in a network of parallel links.

**Packet implementation:** variants of TCP-FAST and RIP.

# On congestion control, multipath routing, and admission control

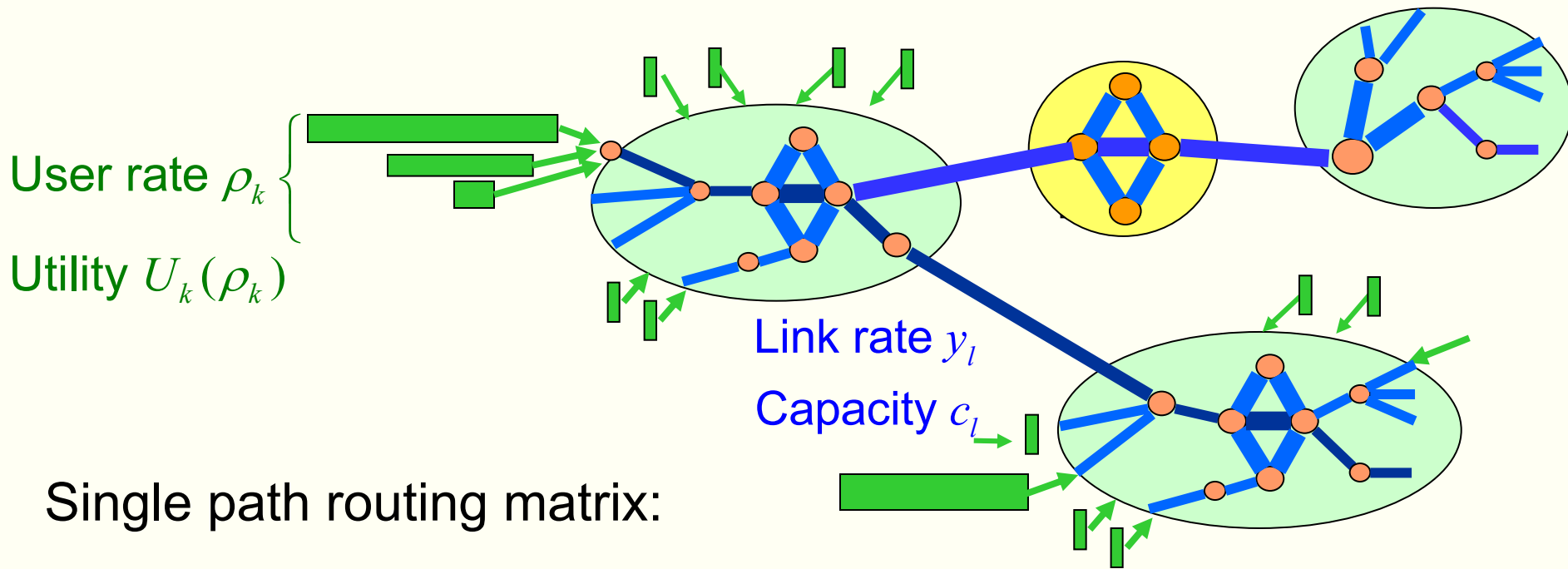
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# Stability and user-level fairness



Single path routing matrix:

$$R_{lk} = \begin{cases} 1 & \text{if user } k \text{ uses link } l \\ 0 & \text{otherwise} \end{cases}$$

$$y = R \rho$$

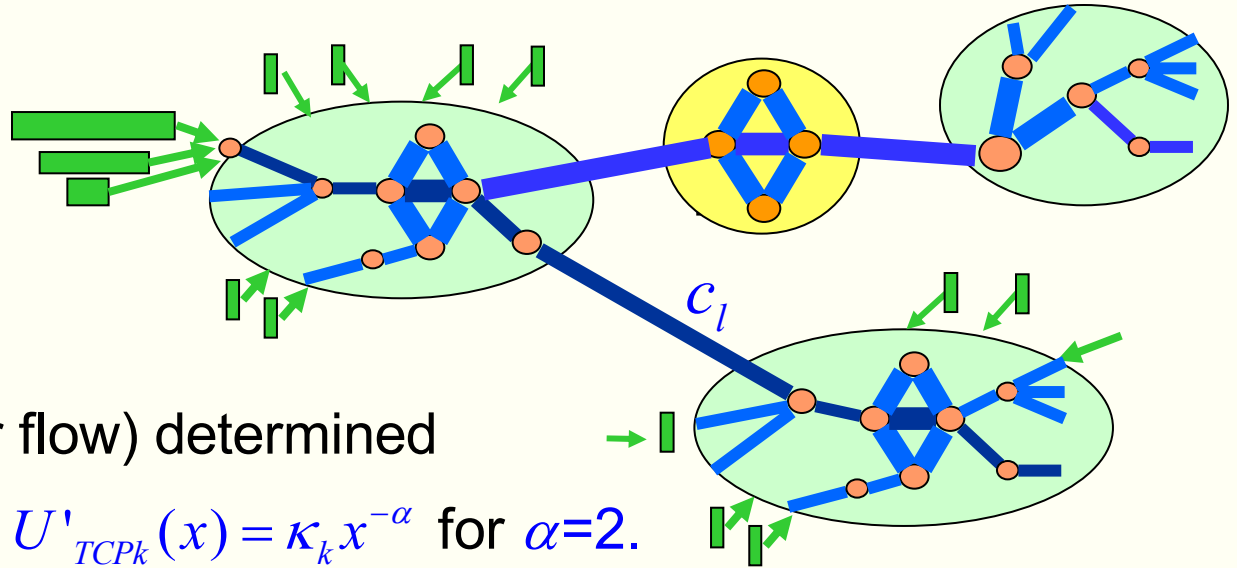
*KELLY'S SYSTEM  
PROBLEM*

$$\max_{\rho} \sum_k \underbrace{U_k(\rho_k)}_{\text{USER UTILITY FUNCTION}}, \quad \text{subject to } \underbrace{R\rho \leq c}_{\text{LINK CAPACITY CONSTRAINTS}}$$

# Contrast with flow-level fairness of TCP

Rate  $x_k$  per flow  $\rightarrow$   
 Utility  $U_{TCPk}(x_k)$   
 $n_k$  flows per user.

TCP utility  $U_{TCPk}$  (per flow) determined  
 by the protocol, e.g.,  $U'_{TCPk}(x) = \kappa_k x^{-\alpha}$  for  $\alpha=2$ .



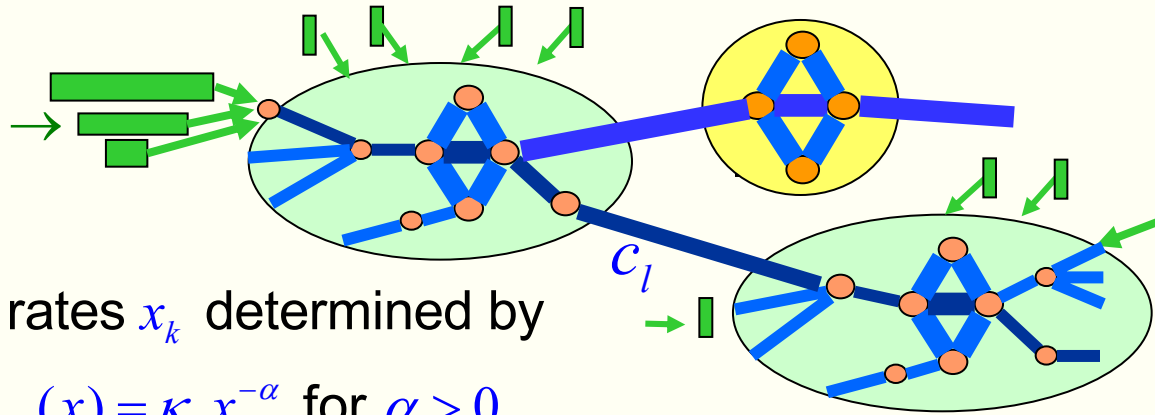
$$\boxed{\text{TCP : NETWORK PROBLEM}} : \max_x \sum_k n_k \underbrace{U_{TCPk}(x_k)}_{\text{TCP UTILITY FUNCTION}} \quad \text{subject to} \quad \sum_k R_{lk} \underbrace{n_k x_k}_{\rho_k} \leq c_l$$

- Without control of number of connections, fairness per flow is moot (Briscoe'07).
- Incentives to employ many TCP flows (e.g., p2p) . Tragedy of the commons?
- If we could control  $n_k$ , can we induce with TCP the SYSTEM problem allocation?  
 (similar to "user problem" setting weight parameter in Kelly-Maulloo-Tan '98)

# On stochastic stability of a network served by TCP

[deVeciana, Lee, Konstantopoulos '99, Bonald-Massoulié '01]

User: Poisson ( $\lambda_k$ ) arrivals,  $\exp(\mu_k)$  workloads.



For each fixed  $\{n_k\}$ , service rates  $x_k$  determined by TCP congestion control  $U'_{TCPk}(x) = \kappa_k x^{-\alpha}$  for  $\alpha > 0$ .

Result:  $n_k$  Markov chain  $\{n_k\}$  stable if and only if  $\sum_k R_{lk} \frac{\lambda_k}{\mu_k} < c_l \quad \forall l$ .

**Remark: congestion control ensures neither stability nor fairness.**

- Both stability, and resource allocation depend solely on users' "open loop" demands  $\frac{\lambda_k}{\mu_k}$ .
- Fairness choice per flow (e.g., value of  $\alpha$ ) has minimal impact. A heavy user will compensate a low TCP rate by increasing  $n_k$ , until  $\rho_k$  serves demand, if feasible. If not  $n_k$ 's grow without bounds.

# Closing the loop on $n_k$ for user-level fairness

Assume that for fixed  $n_k$ , the flow rate  $x_k$  is determined by TCP:

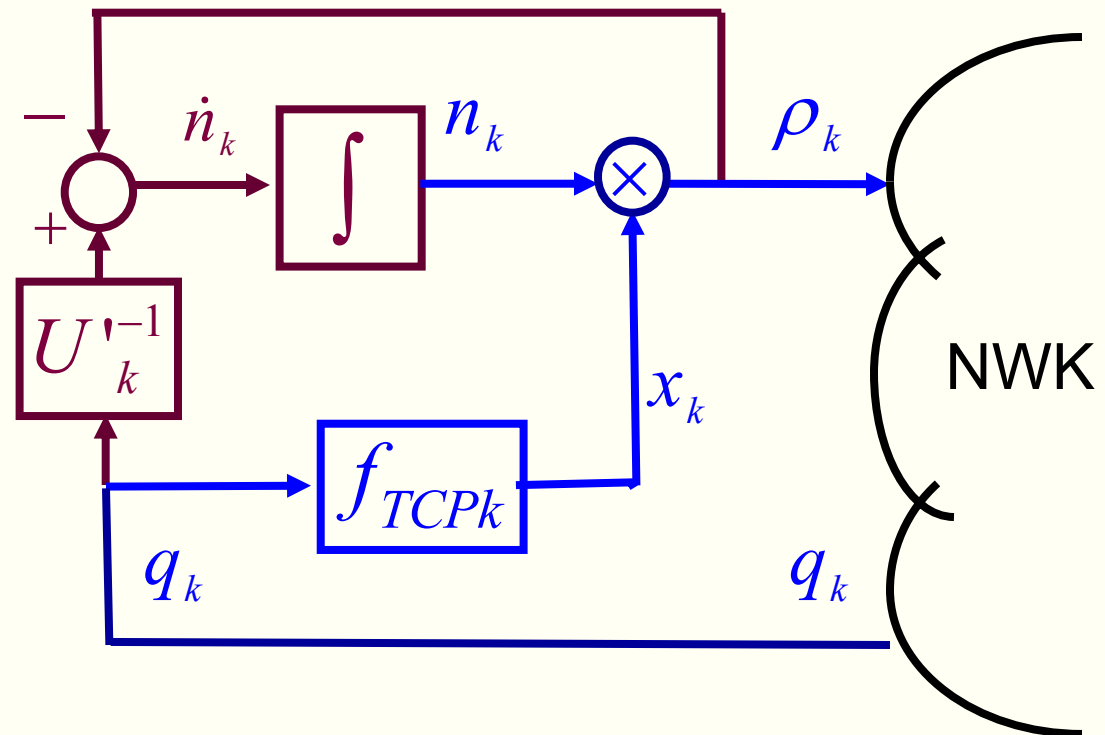
$x_k = f_{TCPk}(q_k)$  where  $q_k$  is the congestion price seen by the source, and  $f_{TCPk} = (U'_{TCPk})^{-1}$ , TCP demand curve. The user rate is  $\rho_k = n_k x_k$ .

Objective: control  $n_k$  so that the system converges to an equilibrium where  $\rho_k = n_k x_k$  solves  $\max_{\rho} \sum_k U_k(\rho_k)$ , s.t.  $R\rho \leq c$ , with utilities defined by users.

Control law for continuous  $n_k$ :

$$\dot{n}_k = \beta \left( U_k'^{-1}(q_k) - \rho_k \right).$$

Other recent work on controlling no. of flows:  
Chen - Zakhor '06  
(for TCP over wireless).

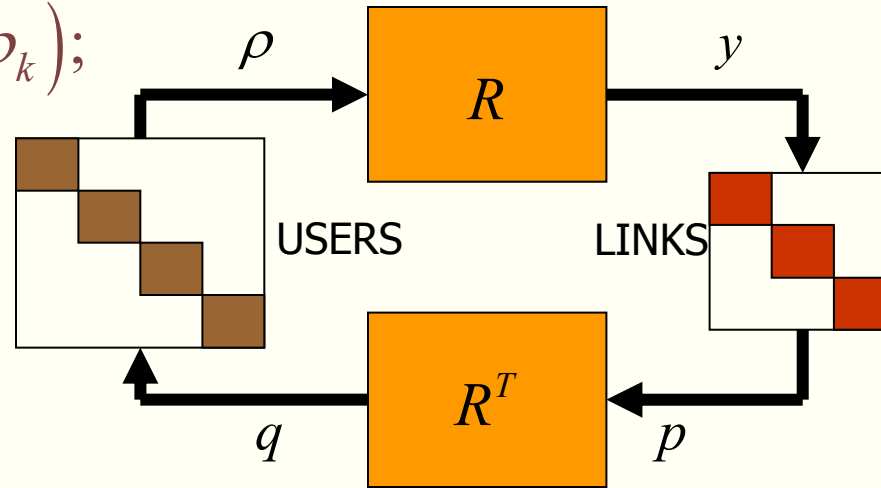


# Analysis using dual TCP congestion control,

$$\dot{n}_k = \beta \left( U_k'^{-1}(q_k) - \rho_k \right);$$

$$\rho_k = n_k x_k;$$

$$x_k = f_{TCPk}(q_k).$$



$$\dot{p}_l = \gamma_l [y_l - c_l]_{p_l}^+$$

## Theorem 1 (arbitrary network).

The equilibrium satisfies  $\max_{\rho} \sum_k U_k(\rho_k)$ , subject to  $R\rho \leq c$ , and is locally asymptotically stable. Proof: passivity argument (as in Wen-Arcak '03).

## Theorem 2 (single bottleneck).

Assume time-scale separation: for fixed  $n = \{n_k\}$ , let  $\hat{q}_k(n)$ ,  $\hat{x}_k(n)$  be the equilibrium values from dual congestion control, and  $\hat{\rho}_k(n) = n_k \hat{x}_k(n)$ . Then the "slow" dynamics  $\dot{n}_k = \beta \left( U_k'^{-1}(\hat{q}_k(n)) - \hat{\rho}_k(n) \right)$  are globally convergent to a point  $n^*$  where the corresponding  $\hat{\rho}_k(n^*)$  are at the optimum welfare point.

# From fluid control to admission control.

In practice,  $n_k$  is discrete (number of TCP connections). Furthermore:

- Real-time control at **sources'** (application layer) is impractical, incentives?
- Killing an ongoing TCP connection to reduce  $n_k$  is undesirable.

More practical alternative:

- Control increase of  $n_k$  (admit new connections), rely on natural termination.
- Admission control carried out by edge router.
- User utility  $U_k(\rho_k)$  describes the SLA: **admit new connection**  $\Leftrightarrow U_k'^{-1}(q_k) > \rho_k$

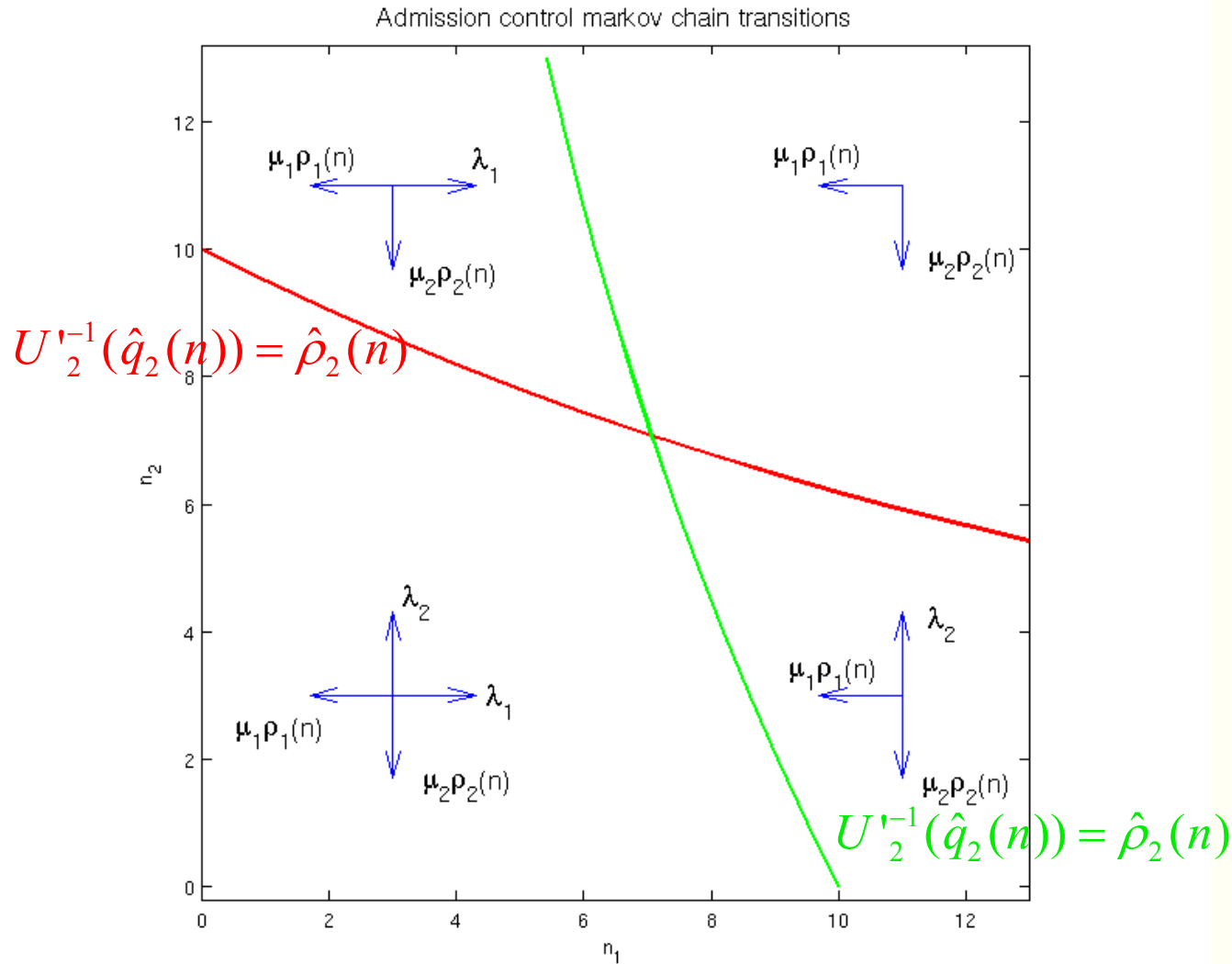
**Stochastic model.** Poisson( $\lambda_k$ ) arrivals, exp( $\mu_k$ ) workloads.

**Active sessions served with rate  $x_k$  obtained from the network.**

Continuous time Markov chain with state  $n = \{n_k\}$ .

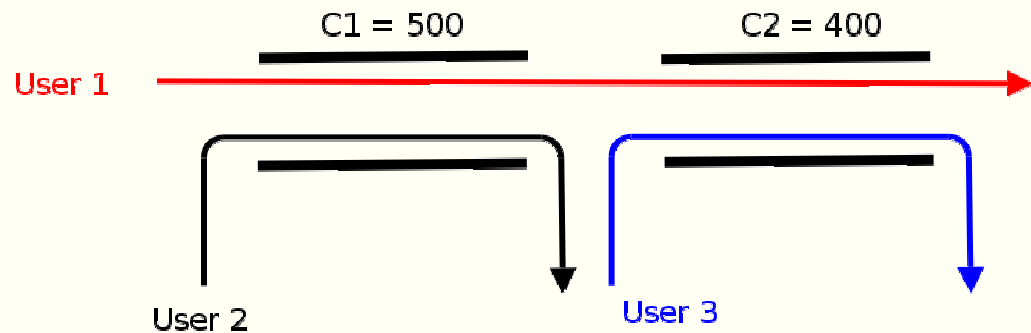
Transition rates:  $q_{n,n+e_k} = \lambda_k \mathbf{1}\{U_k'^{-1}(\hat{q}_k(n)) > \hat{\rho}_k(n)\}$ ;  $q_{n,n-e_k} = \mu_k \hat{\rho}_k(n)$

Single bottleneck, two users, utility  $U_k(\rho_k)=K\log(\rho_k)$ .  
 State space and transition rates:



Irreducible set  
 around  $n=0$   
 is bounded  
 $\Rightarrow$  Stability,  
 Independently  
 of  $\lambda_k, \mu_k$

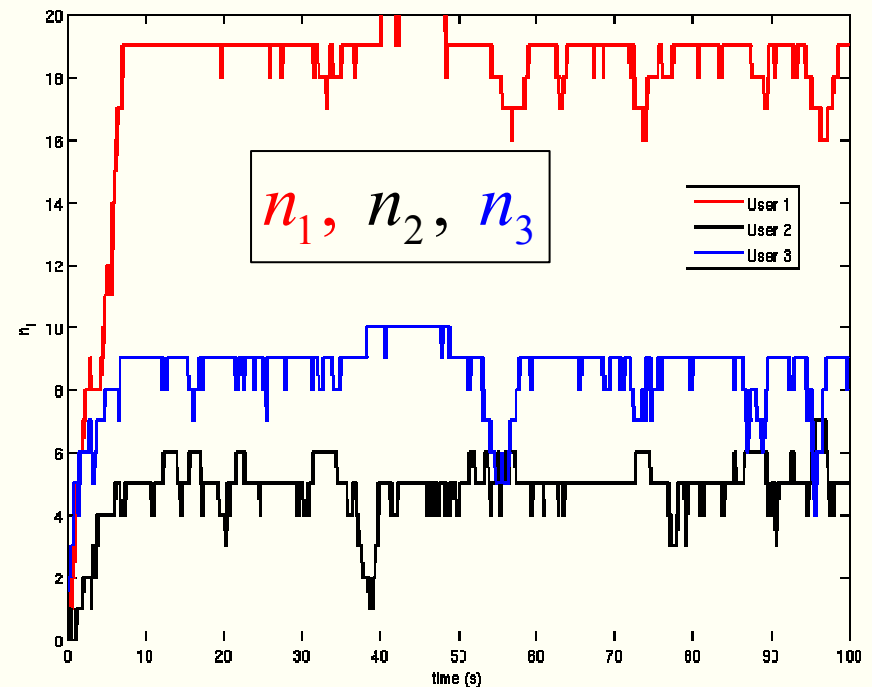
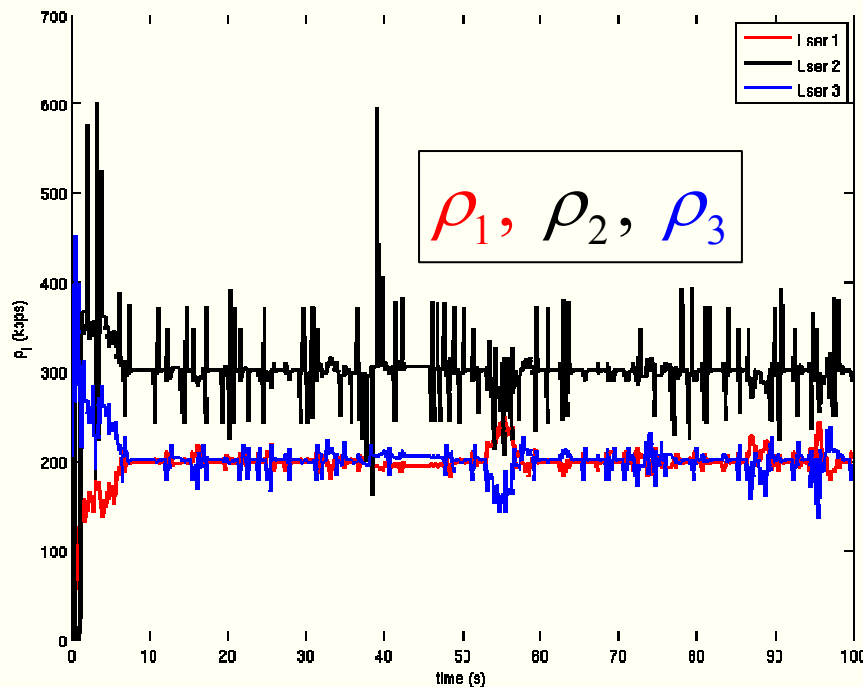
# Fairness? Simulations: JAVA-based tool with random arrivals and workload, simulated dual congestion control.



TCP utility  $U'_{TCPk}(x) = \kappa_k x^{-2}$ .

User utility  $U'_k(x) = \kappa' x^{-5}$ .

emulates max-min fairness.

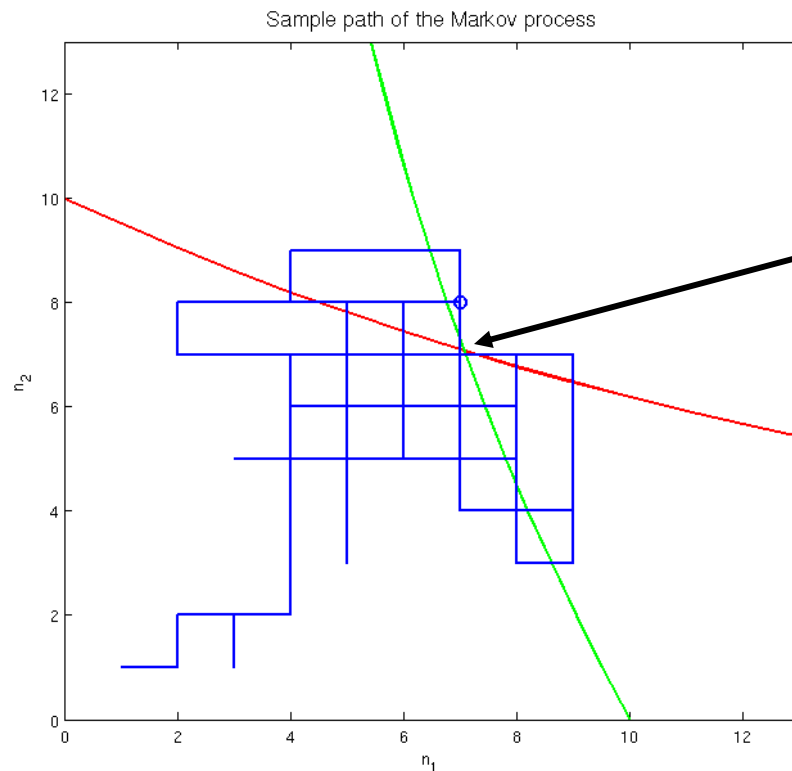




# Fluid modeling of admission control.

Simulation results show admission control achieves optimal allocation, provided the offered loads  $\frac{\lambda_k}{\mu_k}$  are larger than the equilibrium fairshare  $\rho_k^*$ .

We seek analytical proof, and also understanding of the non-greedy case.

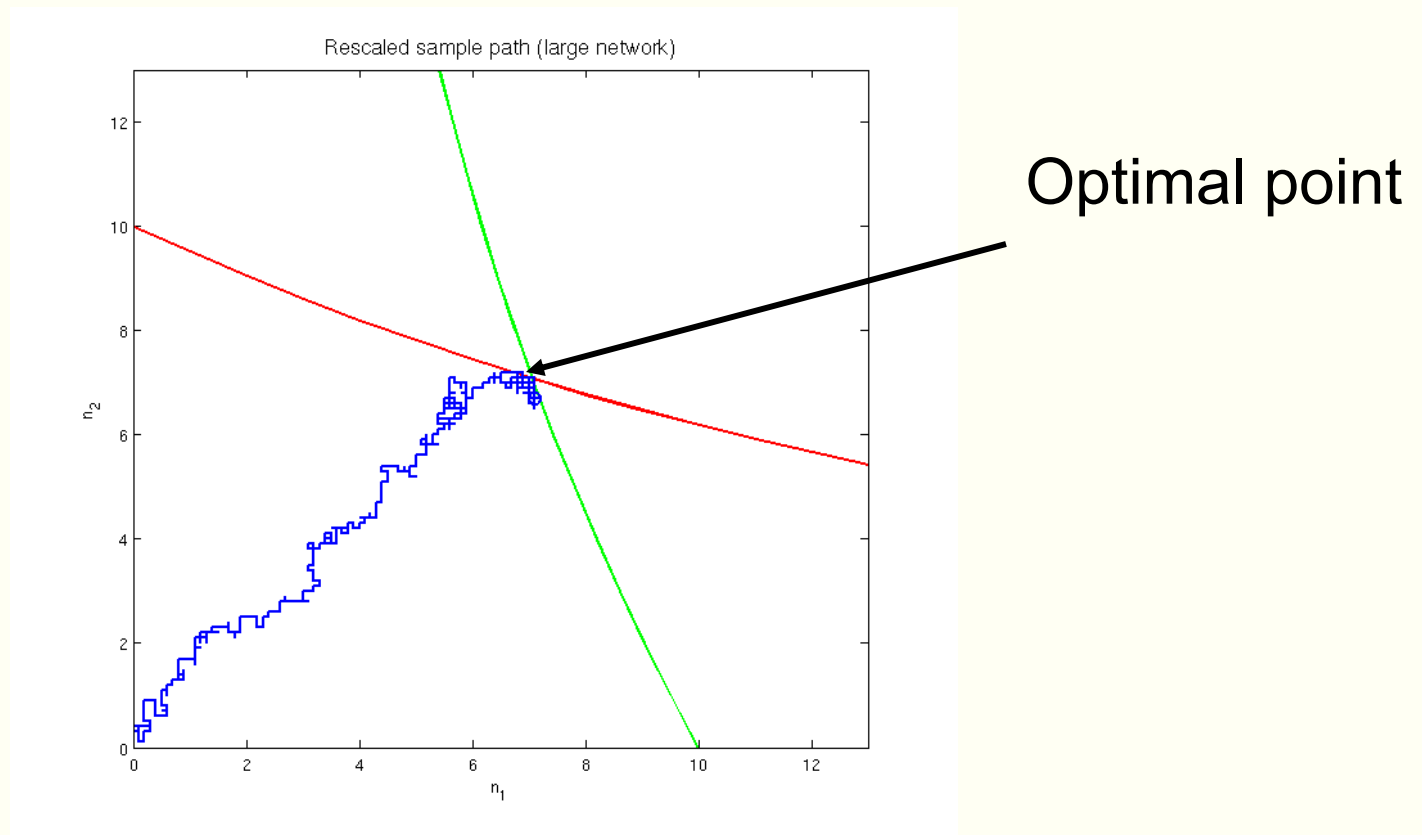


Optimal point

# Fluid modeling of admission control.

Try a large network asymptotic, scaling capacity and user demand curve.

$$c^{(L)} = c_0 L, \quad U_k^{(L)}(\rho) = U_k^{(1)}\left(\frac{\rho}{L\rho_0}\right). \quad \text{Rescaled simulation plot } \frac{n_k^{(L)}}{L}$$

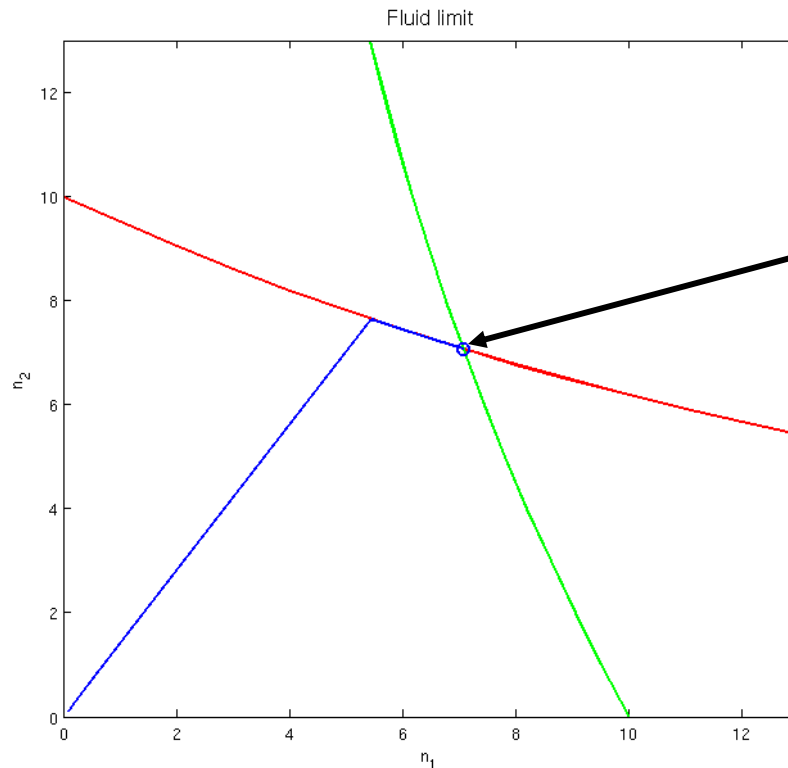


# Fluid limit.

$$\dot{n}_k = \lambda_k \mathbf{1}_{\{U_k^{-1}(\hat{q}_k(n)) > \hat{\rho}_k(n)\}} - \mu_k \hat{\rho}_k(n)$$

For  $\frac{\lambda_k}{\mu_k} > \rho_k^*$  (optimal fairshare) fluid simulations converge to optimal point.

We can prove it in simple cases.

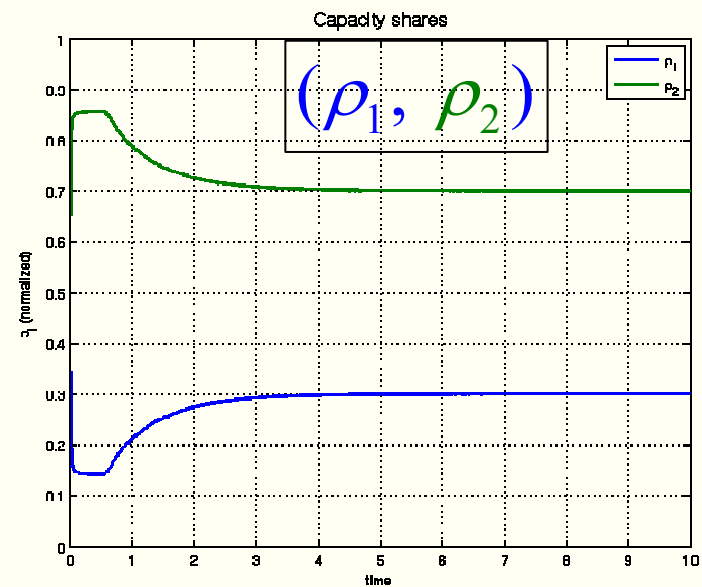
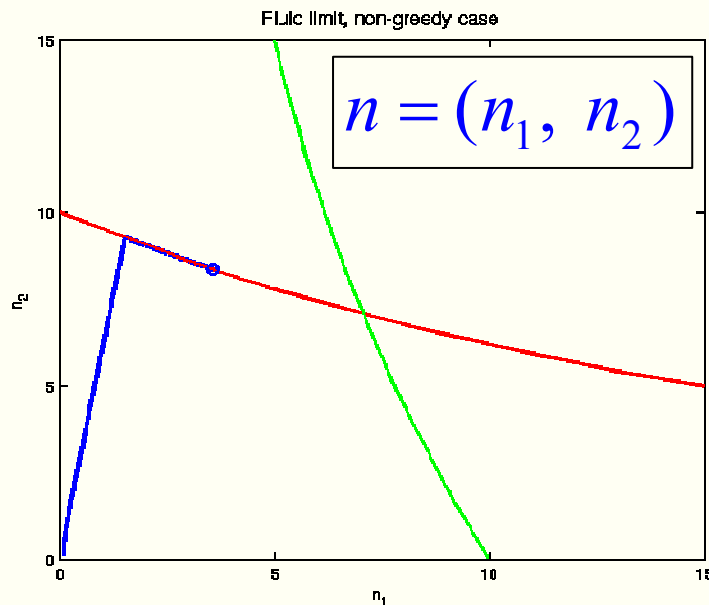


Optimal point

# Fluid limit for the case of non-greedy users:

$$\dot{n}_k = \lambda_k \mathbf{1}_{\left\{U_k'^{-1}(\hat{q}_k(n)) > \hat{\rho}_k(n)\right\}} - \mu_k \hat{\rho}_k(n); \text{ assume a certain user has } \frac{\lambda_k}{\mu_k} < \rho_k^*$$

Example:  $c=1$ , user 1 with  $\frac{\lambda_1}{\mu_1} = 0.3$ , greedy user 2.



Conjecture (verified in simulations so far): converges to solution of  $\max_{\rho} \sum_k \bar{U}_k(\rho_k)$ , s.t.  $R\rho \leq c$ , where  $\bar{U}_k(\cdot)$  corresponds to a demand curve saturated at rate  $\rho_k = \frac{\lambda_k}{\mu_k}$ . Non-greedy user is protected.

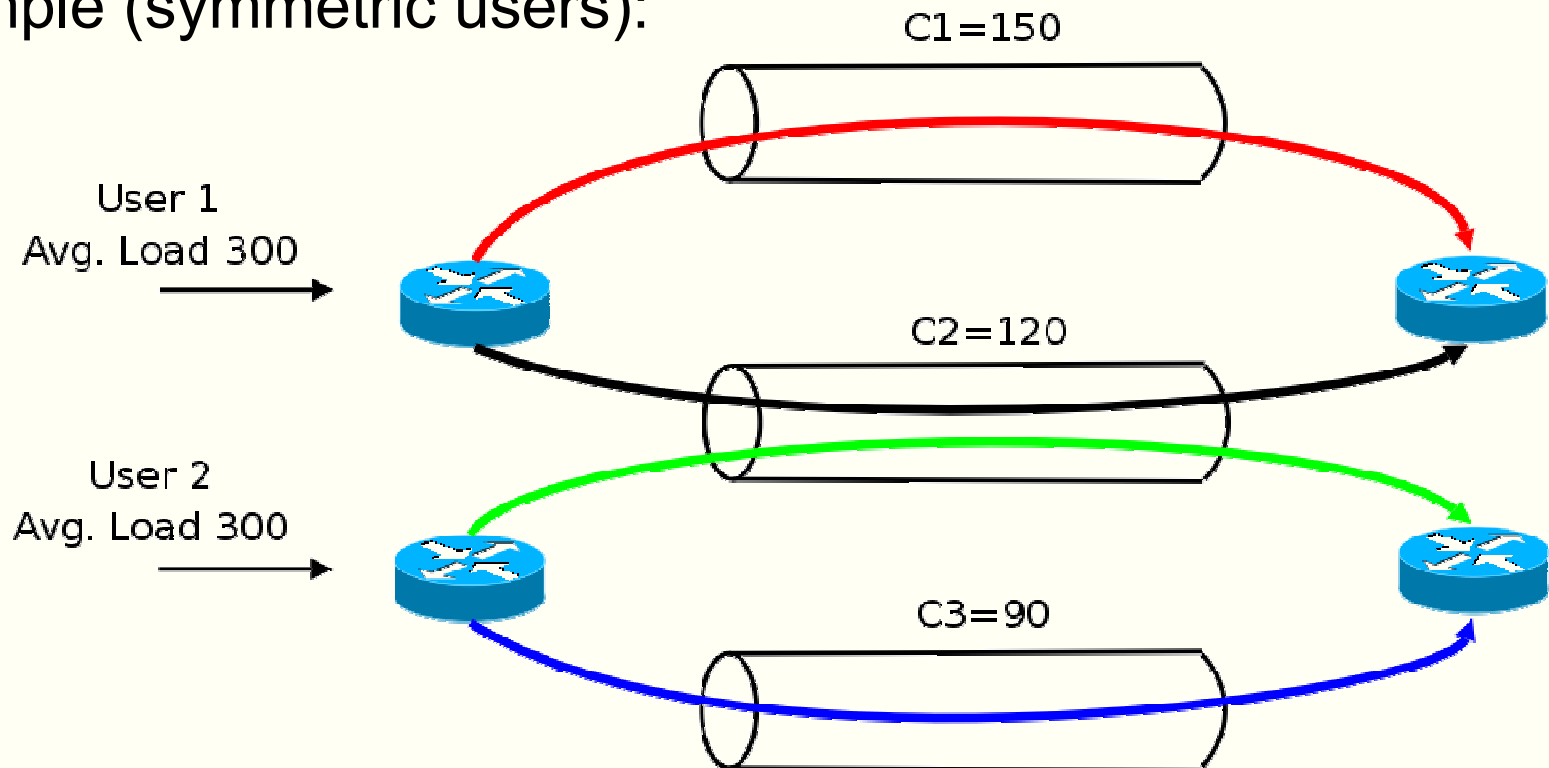
# Back to multipath, work in progress.

Suppose: edge router can choose in which path to route a new flow.

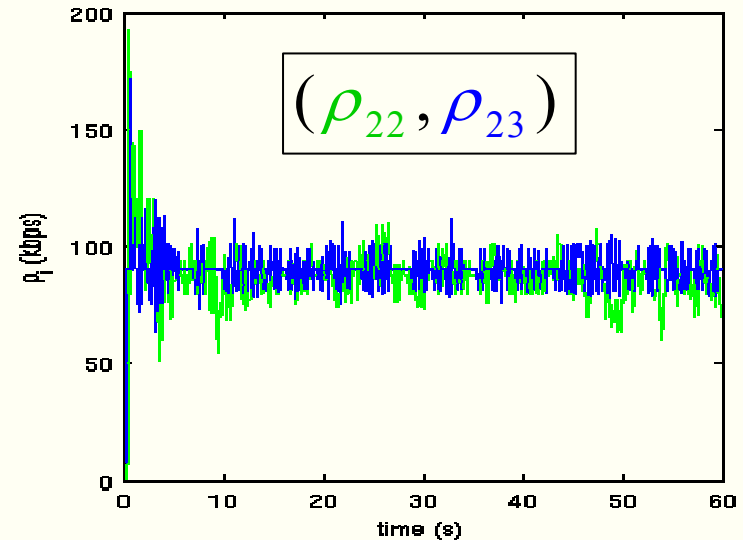
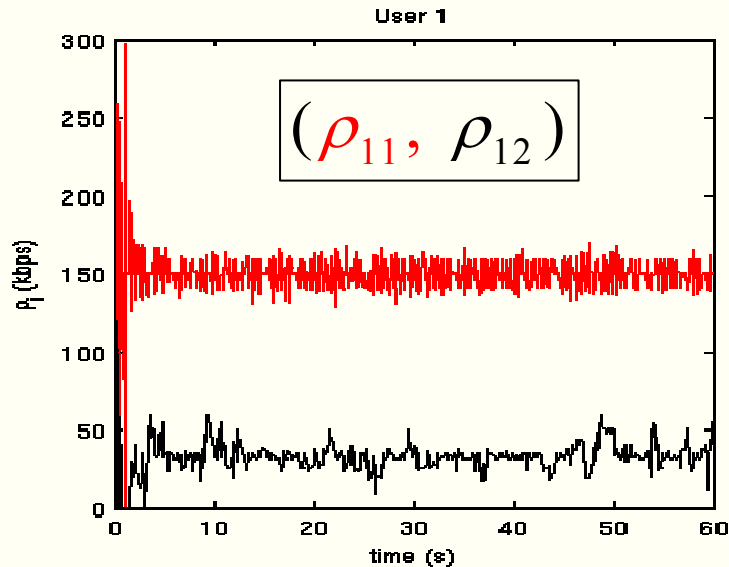
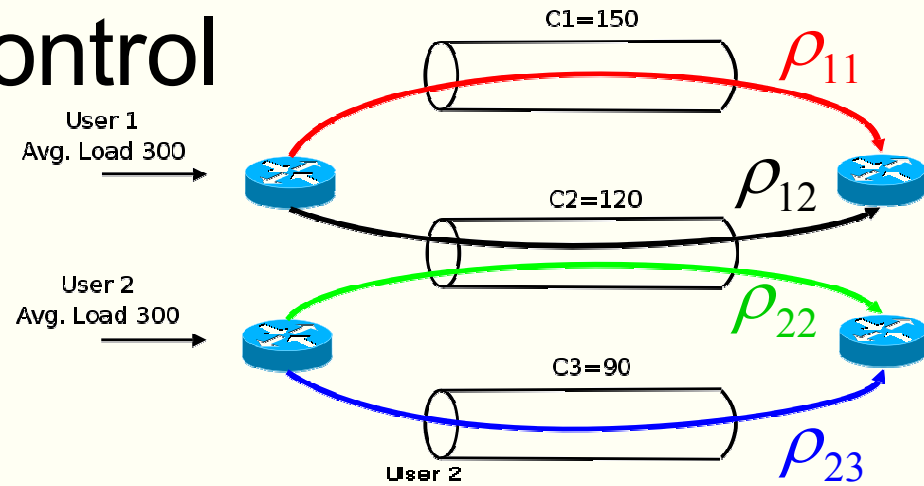
Given: prices  $q_k^r$  of the various candidate routes, a natural policy is:

- Admit new connection  $\Leftrightarrow \min_r q_k^r < U'_k(\rho_k)$ , where  $\rho_k = \sum_r \rho_k^r$
- If admitted, select cheapest path.

Example (symmetric users):



# Multipath admission control Simulations



Conjecture: converges to optimal multipath welfare allocation,

$$\max_{\rho} \sum_k U_k(\rho_k), \text{ subject to } \rho_k = \sum_r \rho_k^r, \quad \sum_k \sum_{r \in l} \rho_k^r \leq c_l.$$

# Conclusions and Future Work

- We studied two cross-layer resource allocation problems:
  - I. Congestion control and multipath routing.
  - II. Congestion control and admission control
- Objective: welfare optimization. We designed decentralized control laws based on prices, achieve these equilibria.
- Dynamic analysis: local stability proofs in arbitrary networks, global results for simpler cases.
- Beware on simplistic models for delay!
- Simulation studies confirm and generalize the above theory.
- Future work:
  - Part I: Global proof in arbitrary networks, loss-based implementation.
  - Part II: Stability proofs for fluid limit model, ns2 implementation.
  - Combination: multipath admission control.