

Fairness, Flow Control & Multi-User Games



Peter Key

peterkey@microsoft.com

Partners



- Frank Kelly, Richard Gibbens (Stats Lab)
- Derek McAuley, Dave Stewart (Microsoft Res Ltd)

Current Solutions



eg *ATM*

- 'Police users'
- Tight QoS guarantees
- Too complex for users
- C.O. (inelastic bias)

■ eg *Internet*

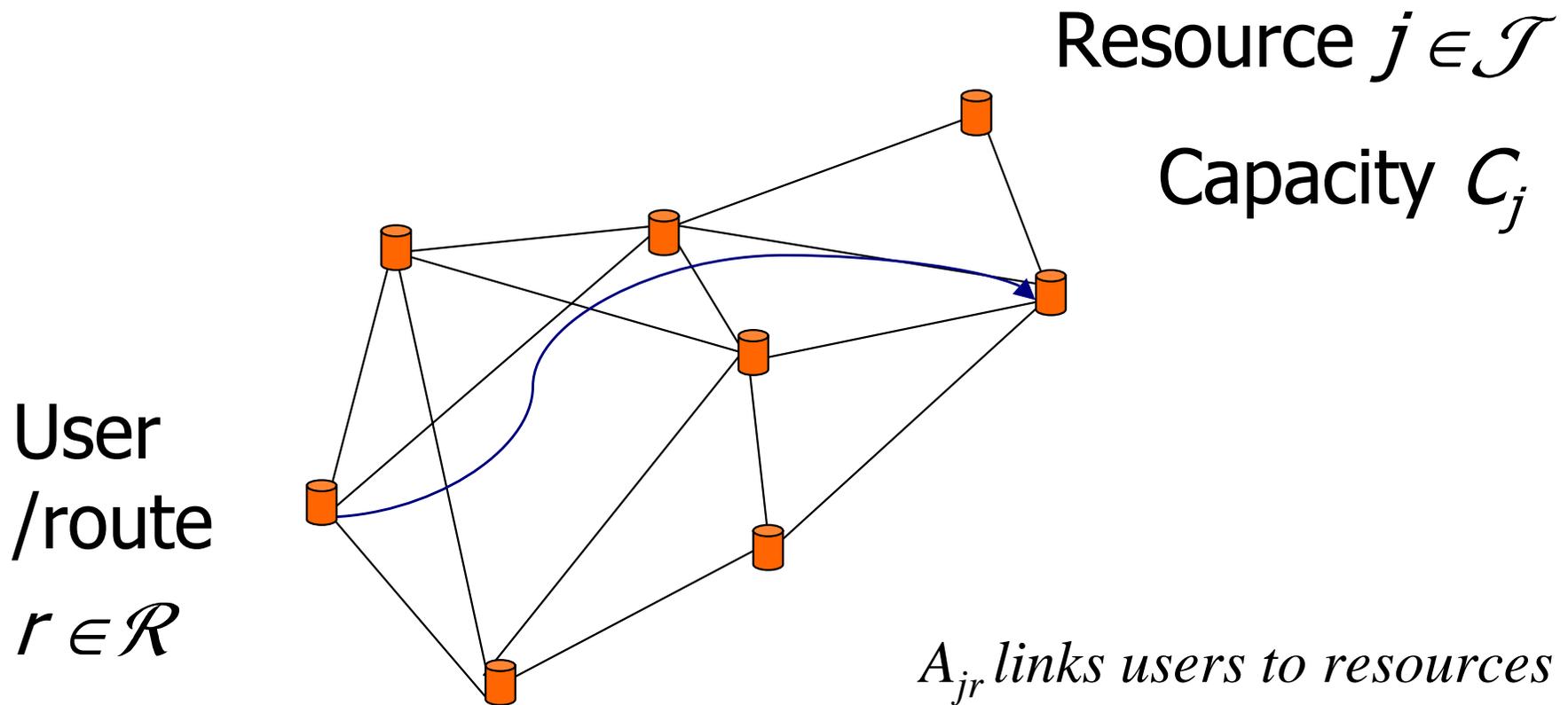
- 'Best effort' (!)=1 QoS
- Low usage users subsidise heavy users
- Relies on 'nice' users (eg compliant TCP)
- Fails to account for congestion externality

What is QoS?

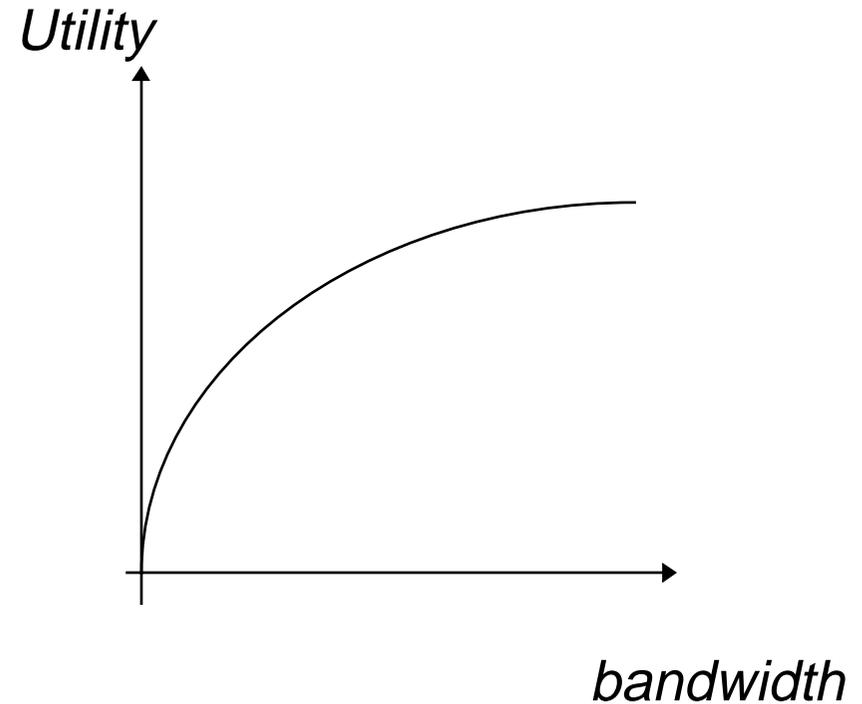
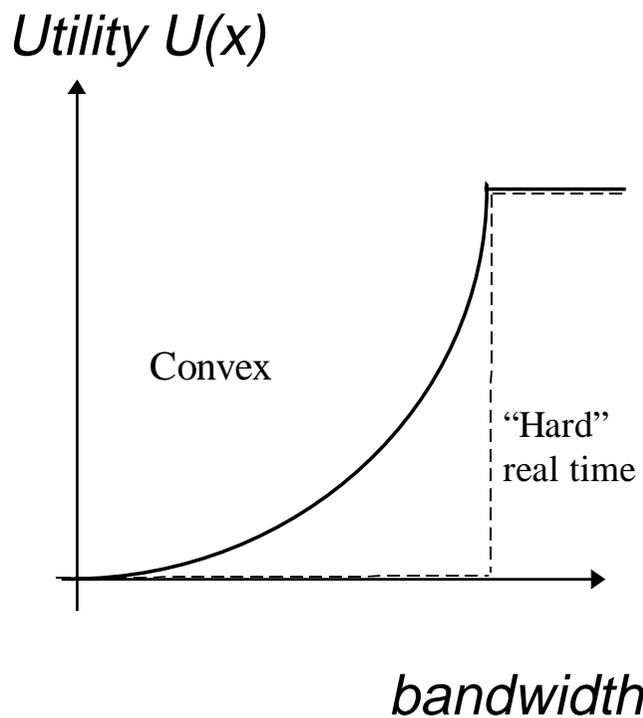


- Current thinking tends toward definitions involving mappings on a multidimensional space (eg CCITT series of recommendations have a whole layered structure)
- Most SLAs for telecommunications define quantities that are not measurable (especially by provider).
- Is this complexity necessary - other services are differentiated by a few criteria?
- Only need to be have non-identical measures to apply Cake-slicing theorem!

Resource system ('network')



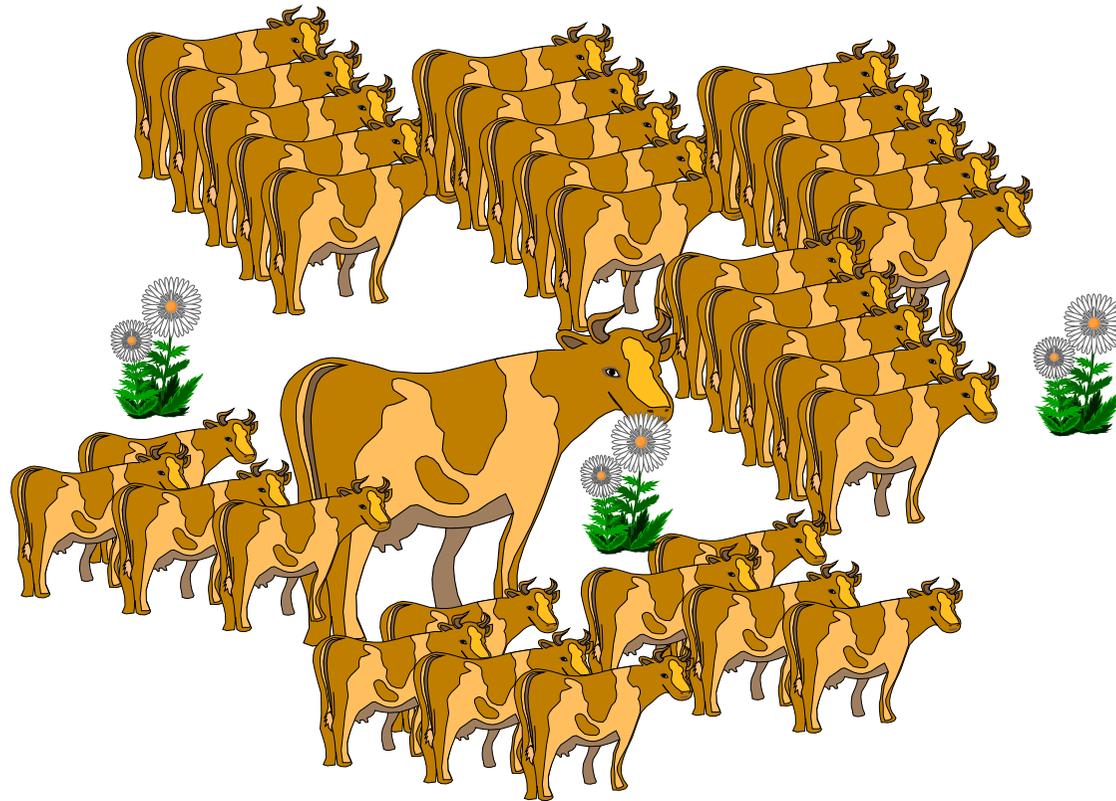
Service Requirements (Shenker)



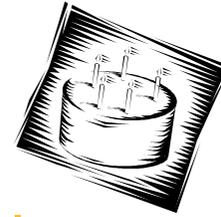
Real Time
Limited capacity → Rejection
Telcos

Data
No calls rejected
Share bandwidth
The Internet!

The problem of the commons



Fairness



- There are lots of ways to be fair!
- In a network context, Max-Min and Max-Utilisation are like “boundary” values
- Proportional fairness has links with Nash arbitration scheme



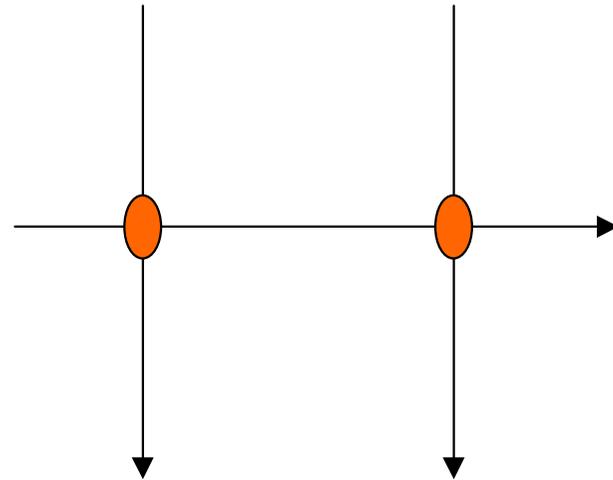
What is fair??



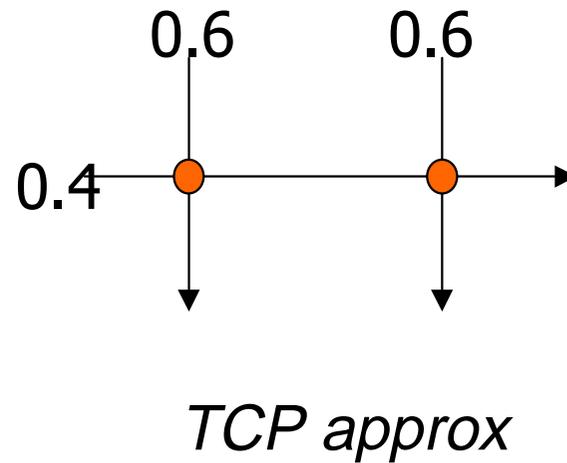
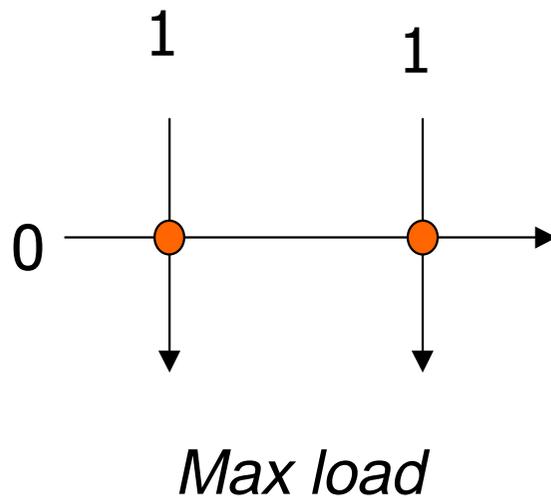
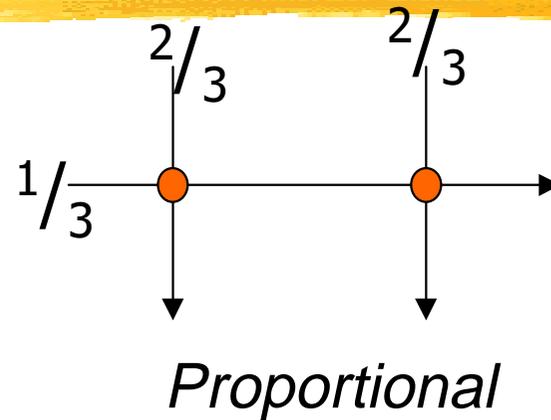
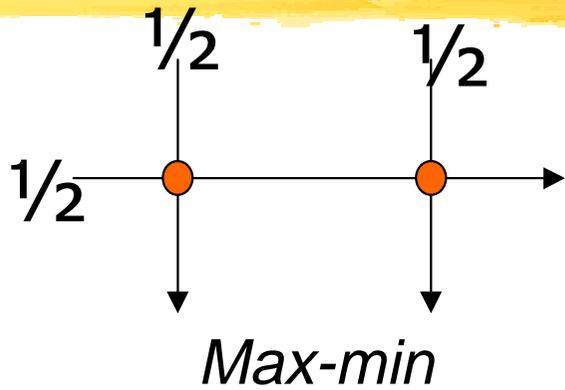
- As many definitions as utility functions (preference functions on x s.t. $Ax \leq C$)
- Max-min fair popular
- Proportional fairness advocated by eg Kelly
- Revenue maximisation another option

Fairness Examples

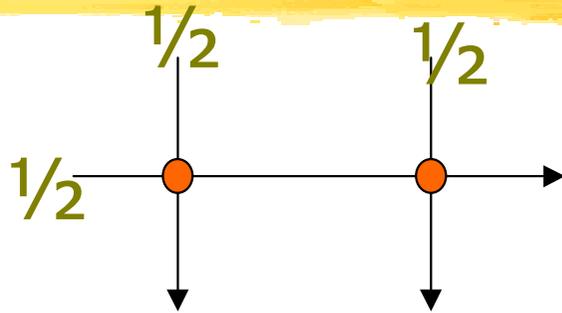
- Two resources
- Each with capacity normalised to 1
- Vertical streams use one resource
- Horizontal streams use two resources



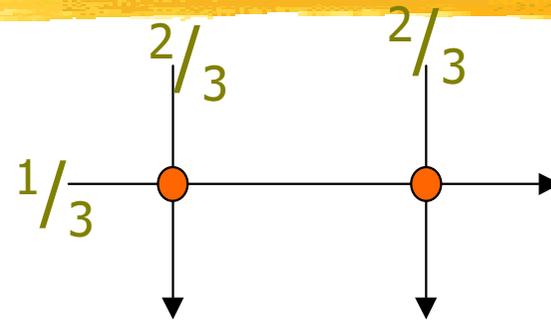
Fairness Examples, eg



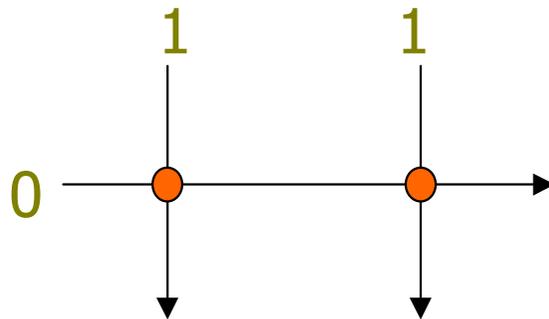
Fairness Examples, eg



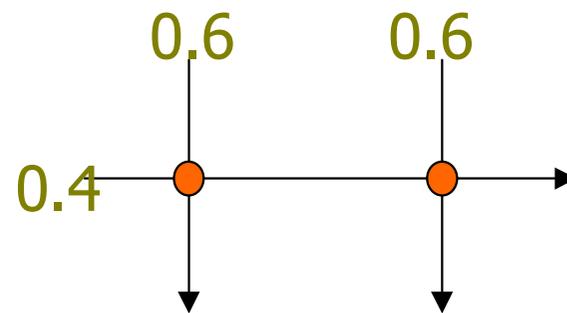
Max-min
 $U(x) = -(-\text{Log}(x))^\infty$



Proportional
 $U(x) = \text{Log}(x)$

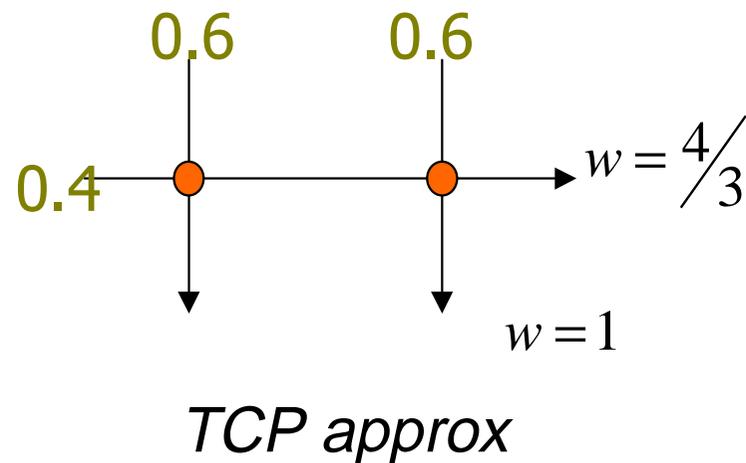
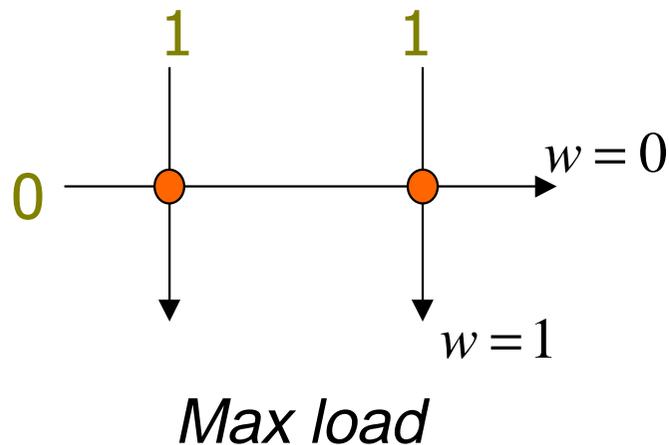
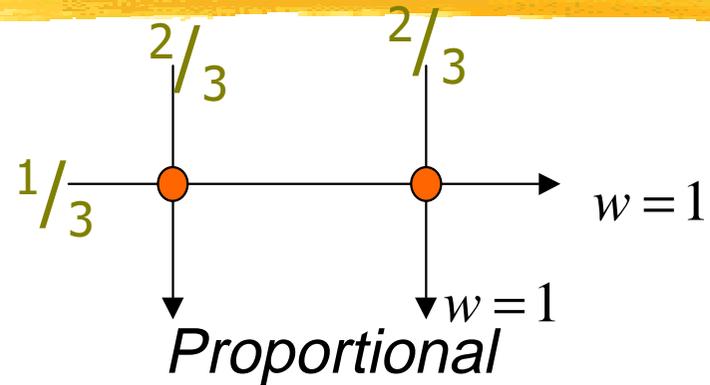
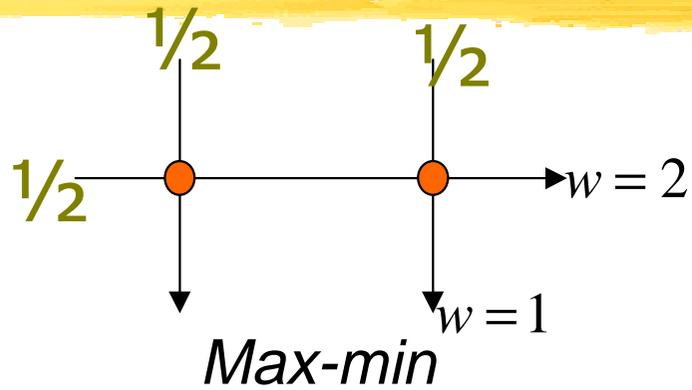


Max load
 $U(x) = \text{Log}(1+x)$



TCP approx
 $U(x) = -(-\text{Log}(x))^{3/2}$

Fairness Examples, prop. fair prices



Some problems with TCP (CA mode)

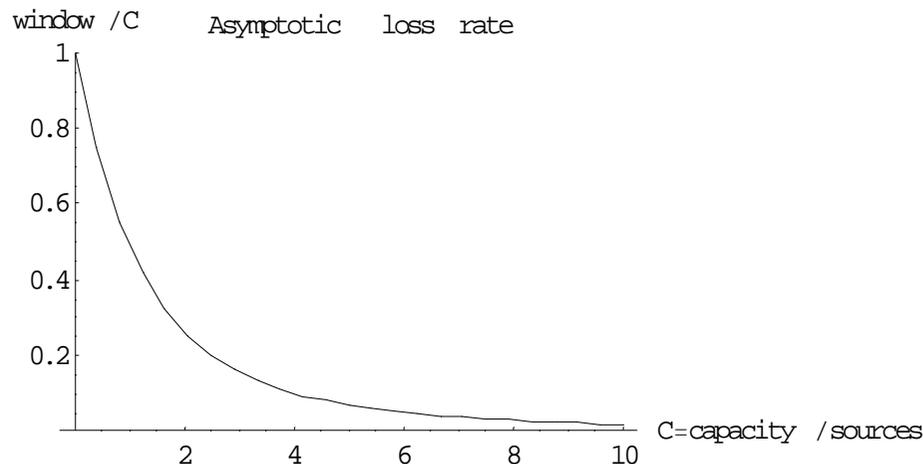
- Lose cells
 - | hence keeps network at high load (but ...)
 - | slow to react (queue size rather than rate)
- If p is probability lose cell, throughput $\propto 1/\sqrt{p}$

(should be $1/p$)
- Does not scale, eg if double capacity
 - | wrong shadow prices
 - | only works if "capacity" $>$ number of sources, where capacity = (bottleneck \times rtt/mss)

Simple TCP model with feedback

Resource loses excess load, n number of sources,
 nC capacity (= n x resource capacity * RTT / MSS),

RTT = 1ms, cap = 1MB/s, C = 244



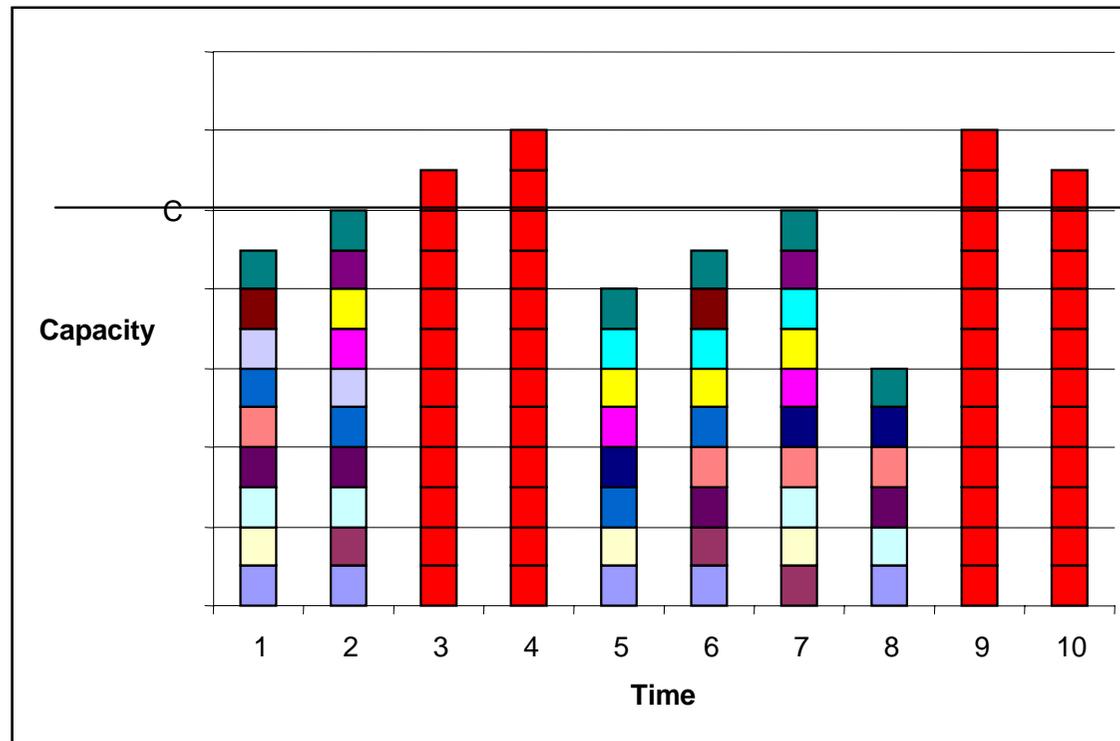
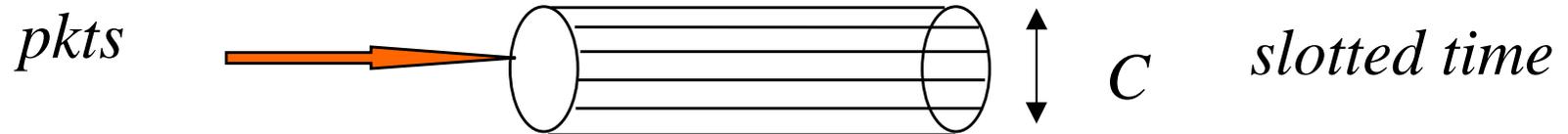
$$\text{window} = \frac{1}{2} (C + \sqrt{8 + C^2})$$

Framework (Gibbens & Kelly)

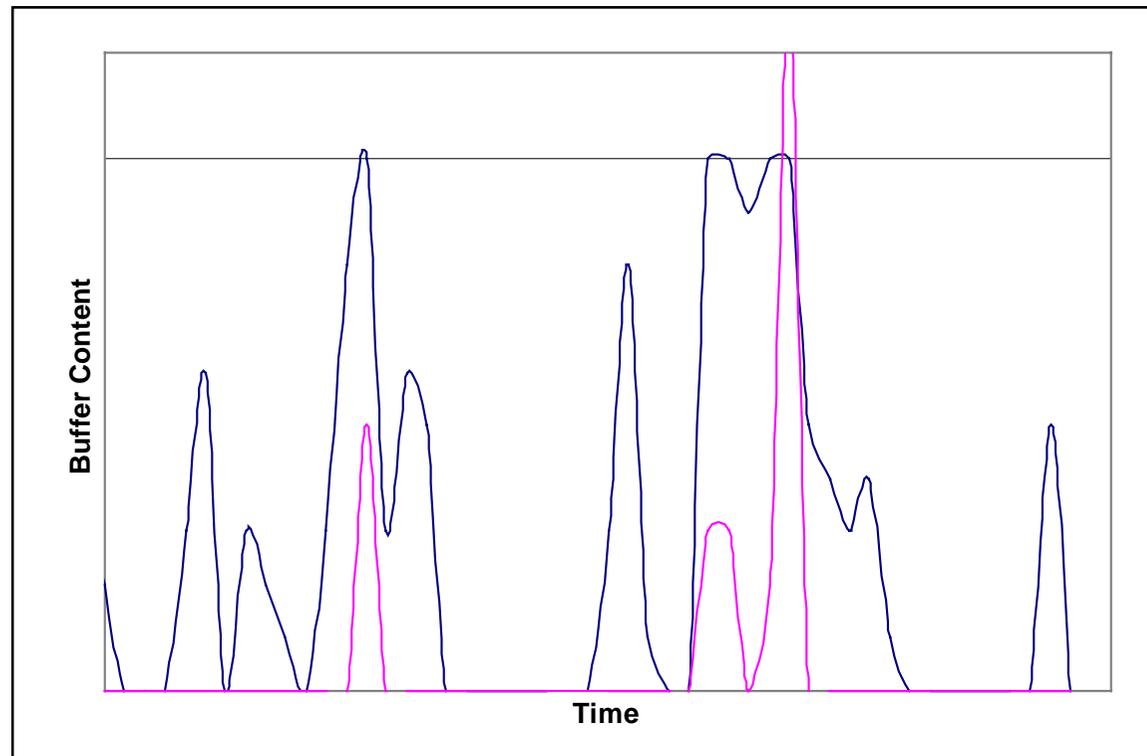
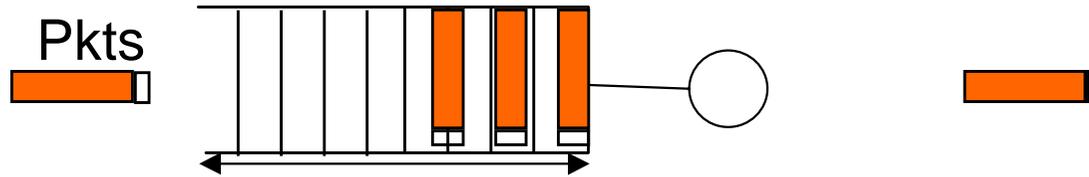


- Users generate load
- Feedback signal is “sample shadow price”
 - eg in simple slotted model, mark every packet if load exceeds capacity: *shadow price* as describes effect of adding extra packet
 - if a buffered system, mark packets from start of busy period to last packet lost in this period

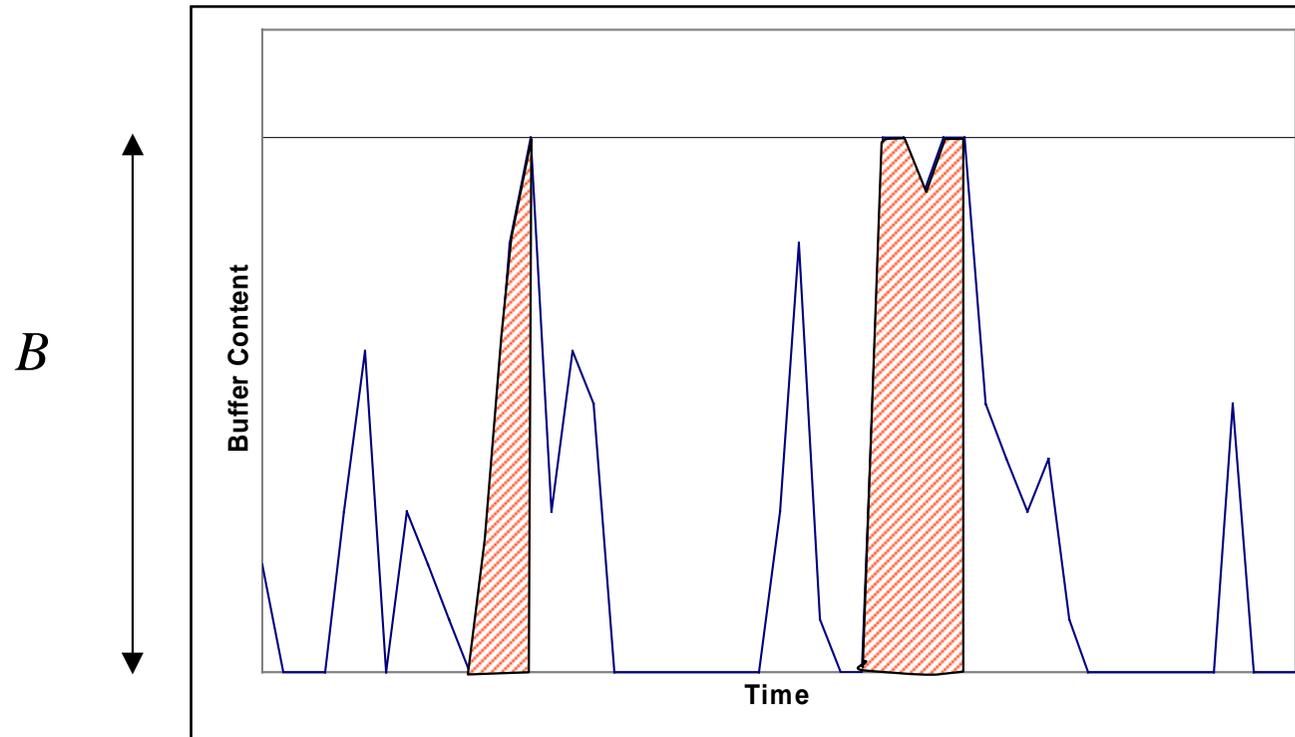
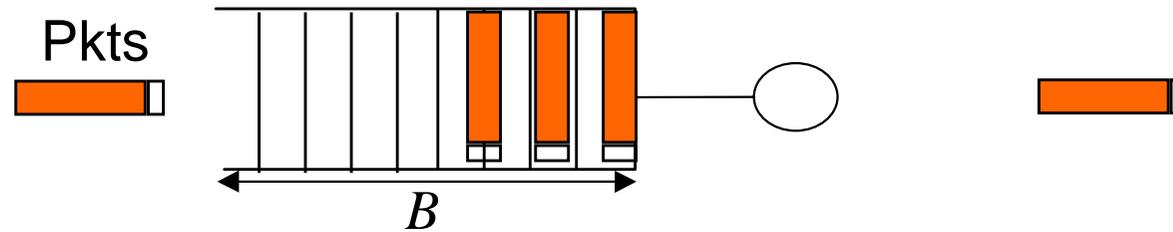
Sample Path Shadow Prices



Sample path Shadow prices, buffered model



Shadow path shadow prices -buffered model



Features of Efficient Pricing Scheme



- Those with the highest cost of delay get served first
- Prices send the right signals for capacity expansion
 - | If all of the congestion revenues are reinvested in new capacity, then capacity will be expanded to the point where its marginal value is equal to its marginal cost
- eg “Smart market” approach of Mackie-Mason/Varian, though some issues with their model
- Prices related to congestion

Optimisation Framework (for fairness)

$$\text{Max } U = \sum_r U(x_r)$$

$$\text{s.t. } Ax \leq C$$

System optimum

U is utility

$$\text{Max } U = \sum_r U(x_r) - \sum_j C_j(y)$$

$$y = \sum_r A_{jr} x_r$$

C is cost function, eg

$$C_j = (y - C_j)^+$$

$$\text{Max } U(x_r) - tx_r$$

User optimum

Solution

Consistent set of taxes (prices) and load exist s.t.

$$U'(x_r) = \sum_{j \in r} \mu_j \quad \text{if } x_r > 0$$

$$\mu_j = \frac{dU}{dC_j} \quad \text{constrained case}$$

$$\mu_j = \frac{dC_j}{dy} \quad \text{unconstrained}$$

$$U'(x_r) = t_r$$

Eg Network chooses taxes, user chooses load, solution is network, user and System optimal. But dependent on Utility function, so

Example

$$C_j(y) = E(Y - c_j)^+ \Rightarrow \mu_j = \frac{dC_j}{dy} = \Pr\{Y \geq c_j\}$$

if Y Poisson and $E(Y) = y$.

Also,

$$E(x_r I\{Y \geq c\}) = x_r \Pr\{Y \geq c_j\}$$

Hence mark packet if exceed capacity

If $U(x_r) = w_r \text{Log}(x_r)$, then

$$w_r = x_r \Pr\{Y \geq c_j\}$$

Adaptive (prop fair) scheme

Suggests the adaptive approach

$$\frac{d}{dt} x_r(t) = \kappa_r \left(w_r(t) - x_r(t) \sum_{j \in r} \mu_j(t) \right)$$

where $\mu_j(t) = p_j \left(\sum_{j \in r} x_r(t) \right)$ $\frac{d}{dy} C_j(y) = p_j(y)$

$C_j(y)$ is cost function on resource j when load is y

maximises $\sum_r U_r(x_r) - \sum_j C_j \left(\sum_{j \in r} x_r \right)$

Example - elastic control

$$x_{t+1} = x_t + \kappa (w_t - f(t))$$

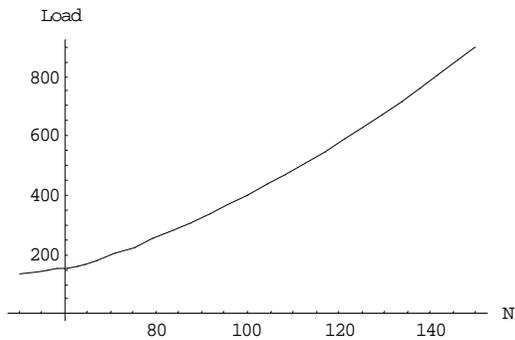
w_t reflects willingness to pay,

$f(t)$ is feedback received from the network

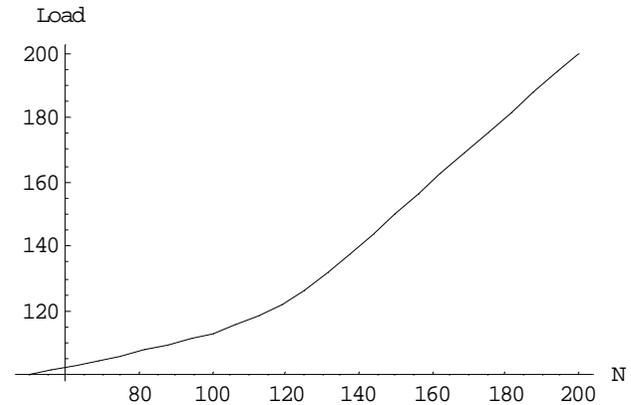
eg

$f(t) = x_t$ if (resource/ bottleneck overloaded) else = 0

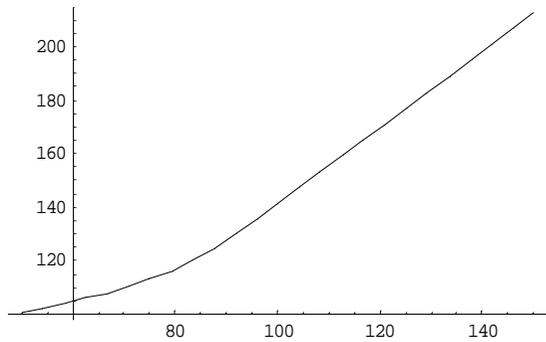
TCP & TCP like schemes



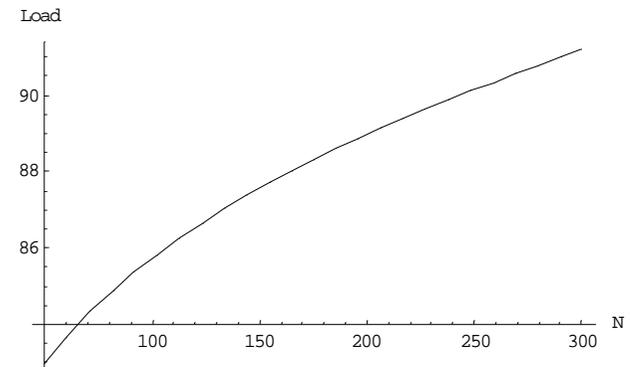
TCP



$w=1$

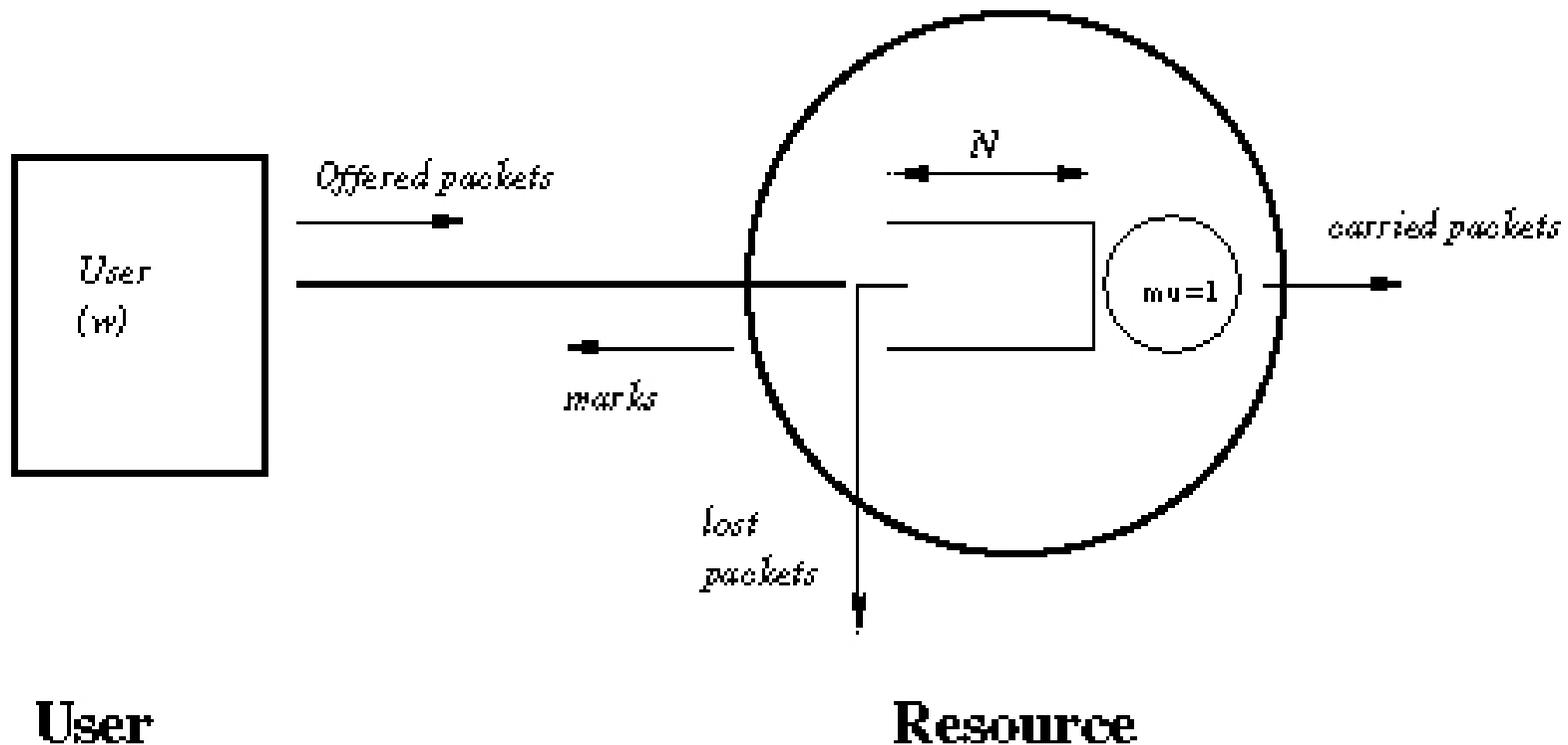


TCP with shadow price

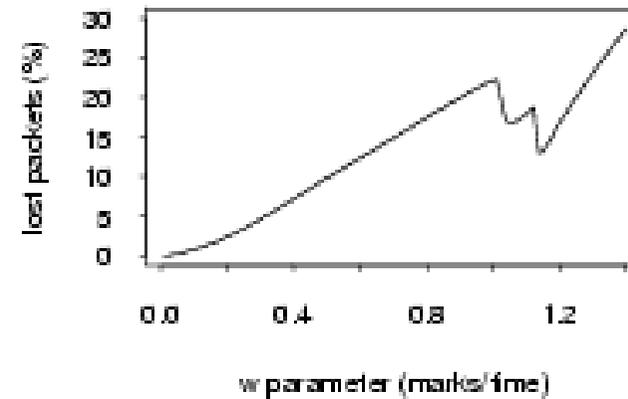
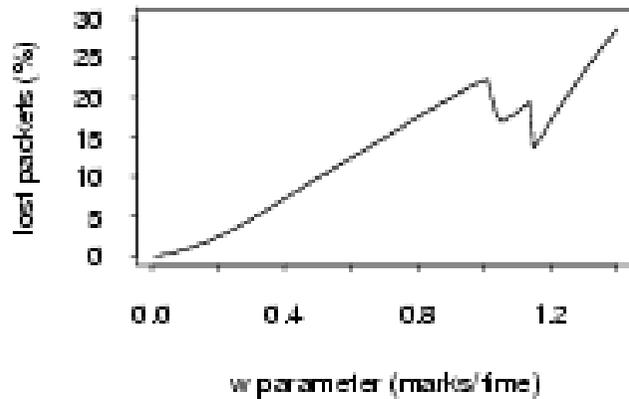
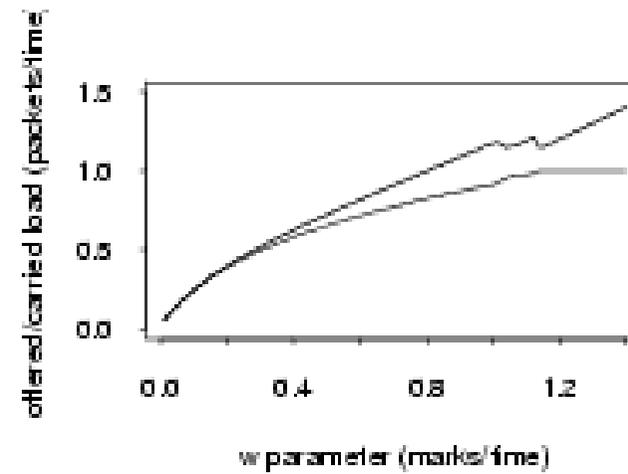
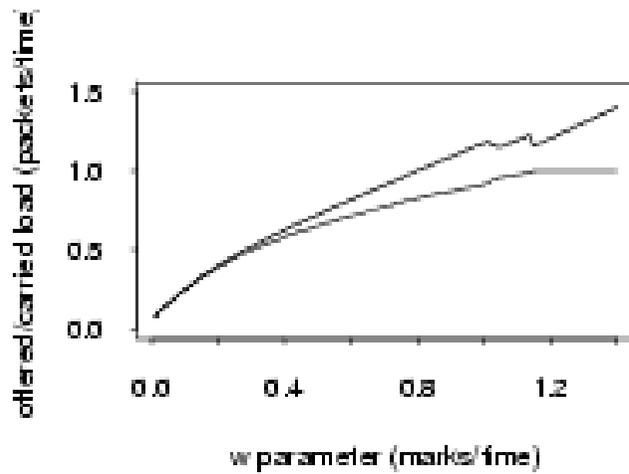


$w=0.05$

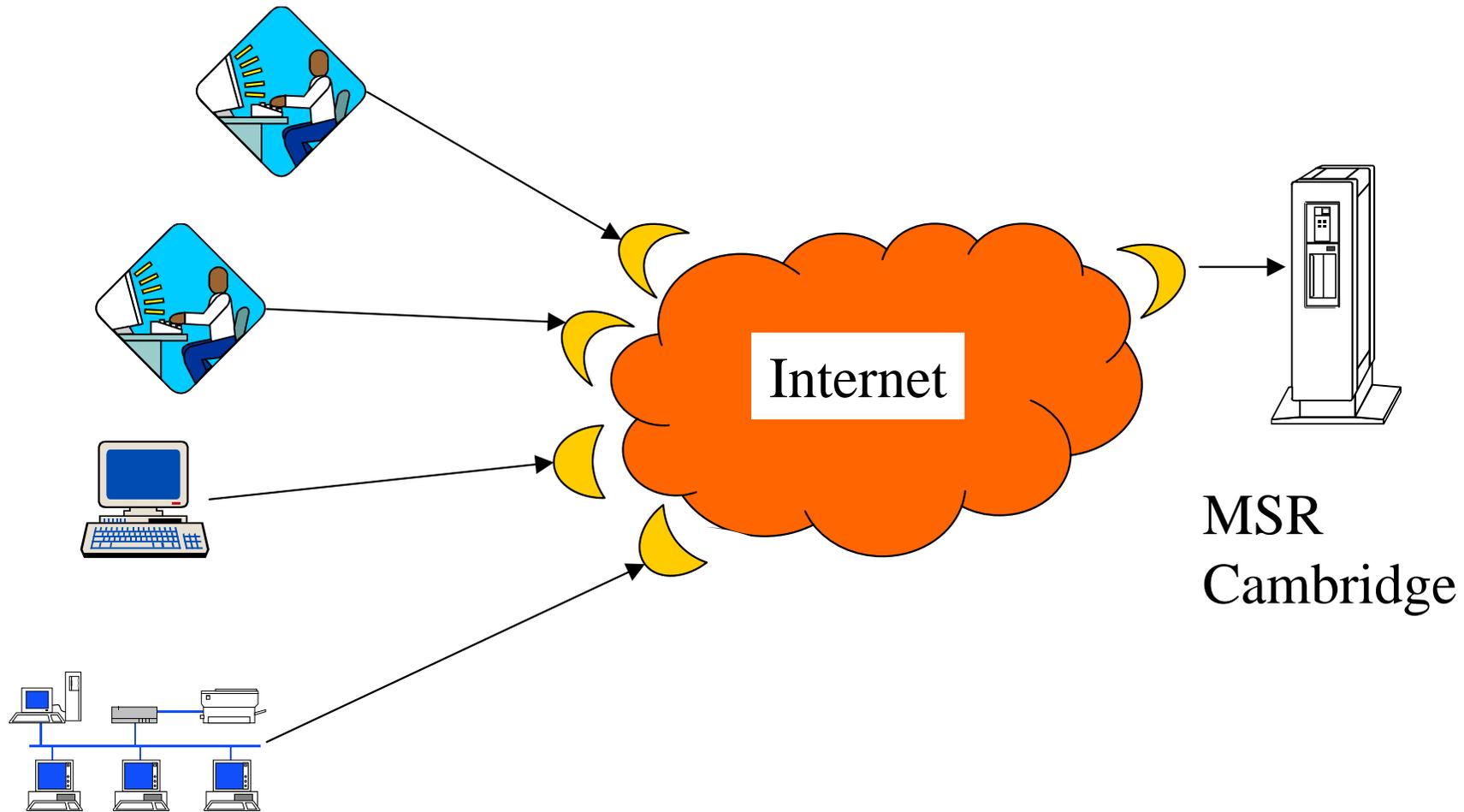
Single node network



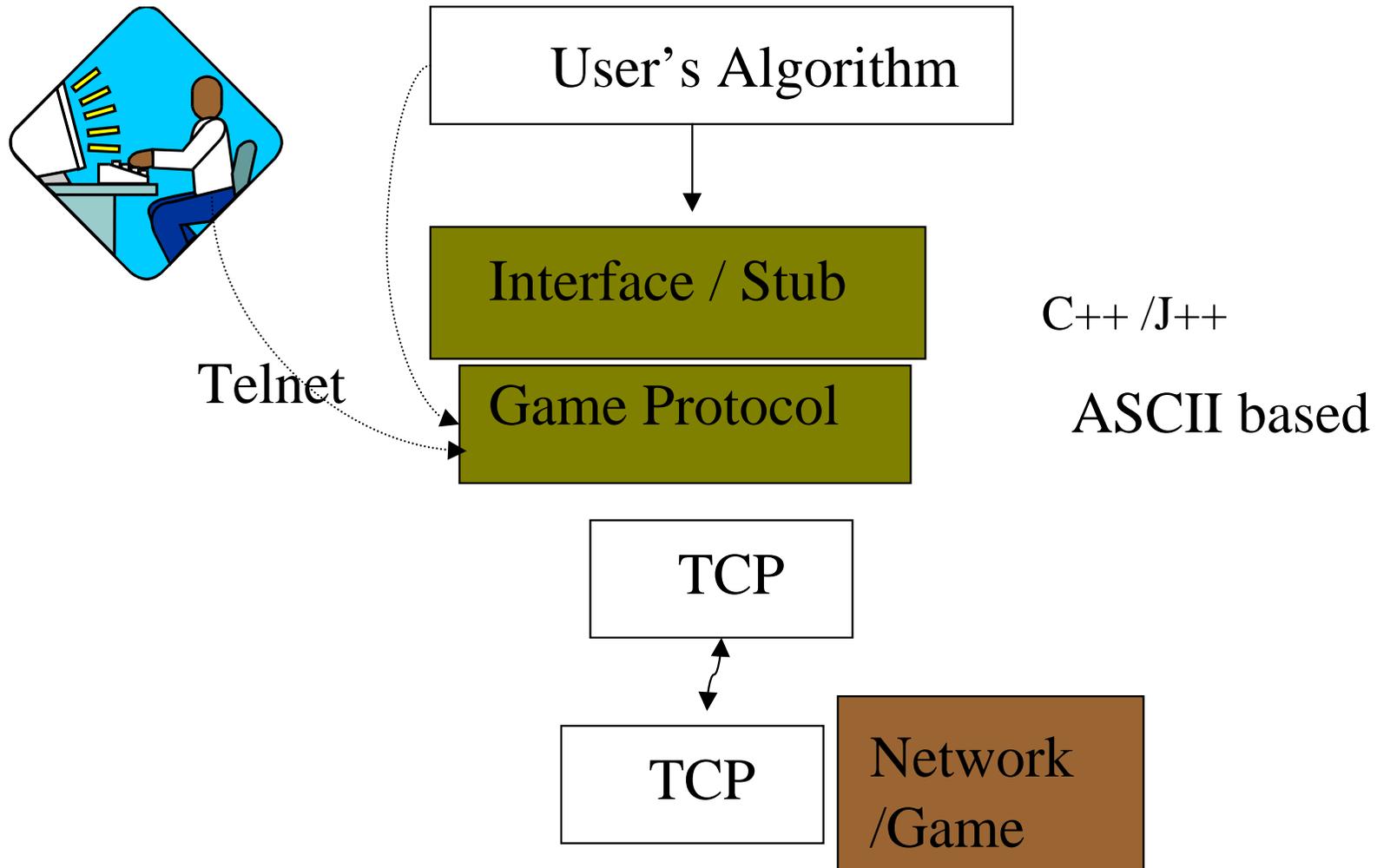
Buffered Model



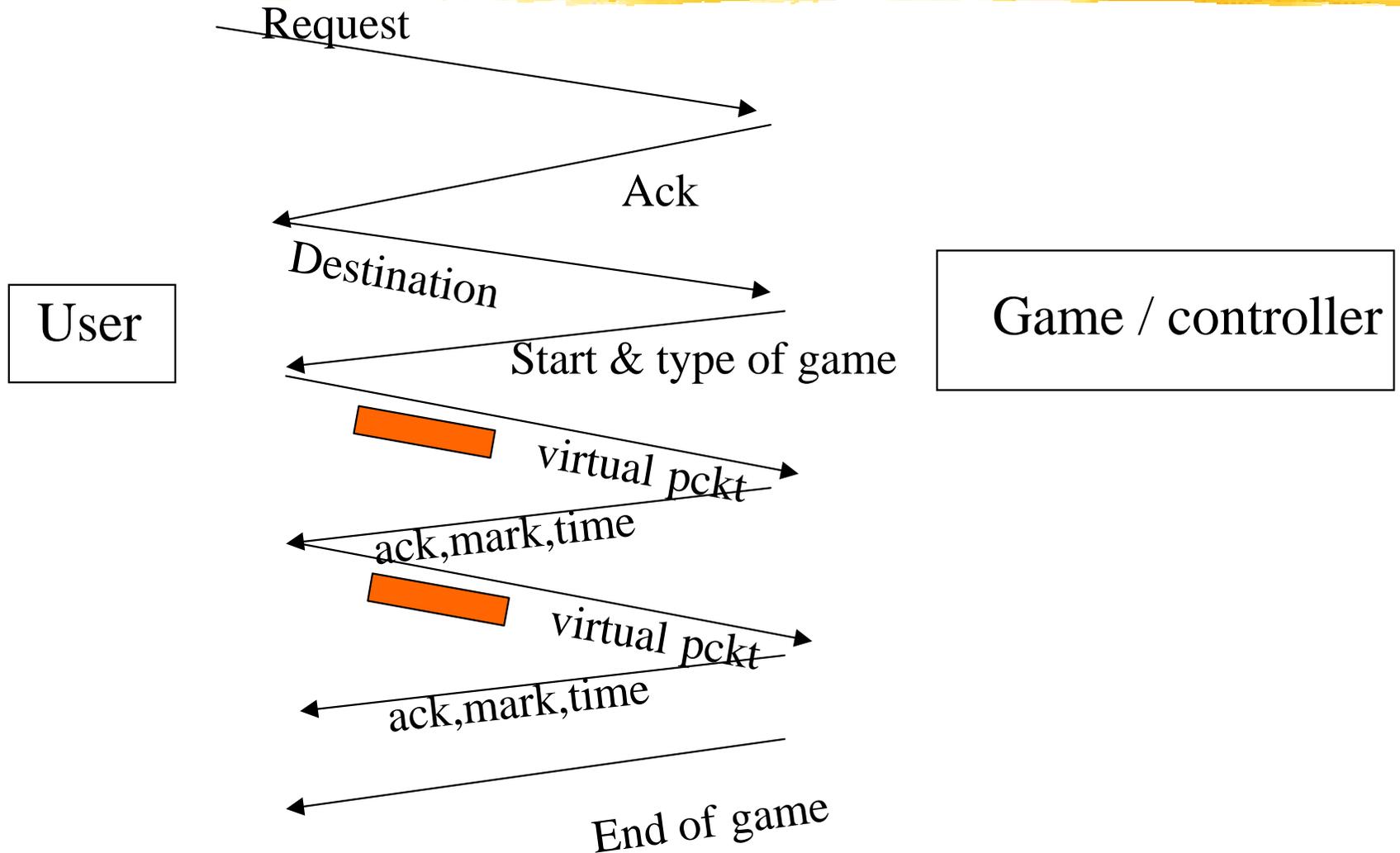
Distributed multi-player game



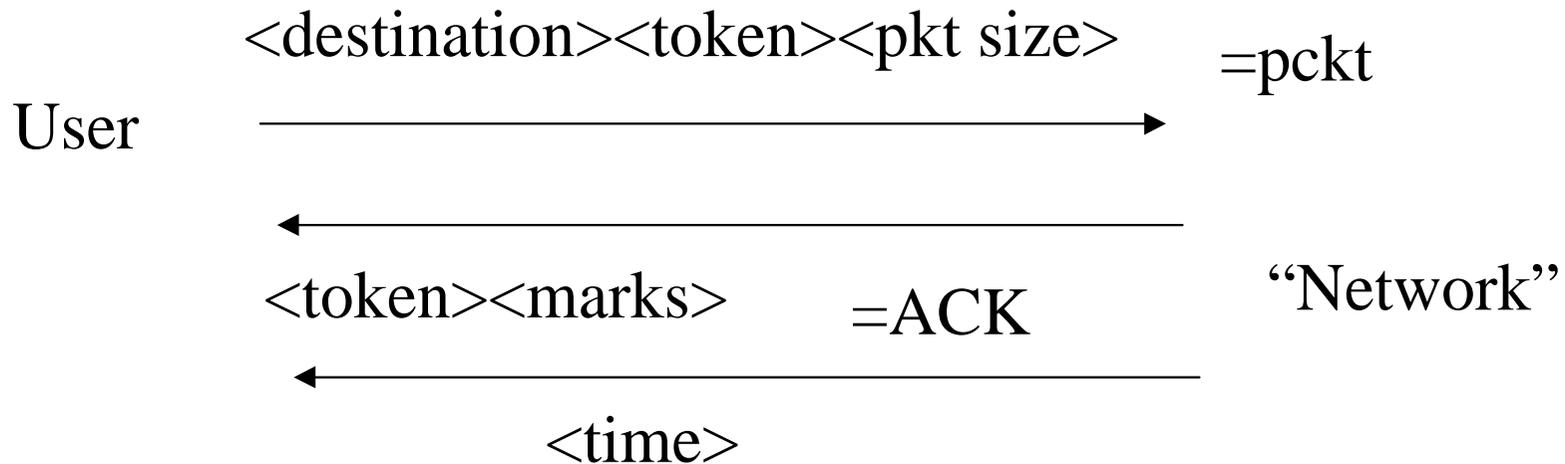
Protocol structures



Information flow



Protocol



Notes:

- All single word (32bit) unsigned,
- pkt size integer
- marks integer
- Time=sec.μsec
- Token generated by user
- Corrupted/lost token=packet drop

Example Objectives

Assumes notified cost per mark

- Maximise (ave. thruput - ave. cost)
- Max Discounted $\sum(\text{thruput} - \text{cost})$
- For given utility function, max $\sum(\text{utility} - \text{cost})$
- Transfer an amount of data $F(\text{file})$ at min cost
- Transfer F in set time T at min cost
- Transfer F as quickly as possible at min cost
- Given fixed budget, maximise transfer

Iterative Approach



- New User plays on test harness
- Plays against controlled load
 - (eg against copies of single game or against sample from random population)
- Plays against other users each with same objective
- Plays against others with multiple objectives

Disciplines



- Computer science
- Control Theory
- Game Theory / econometrics
- Stochastic Decision Theory
- Optimisation / Dynamic Programming



Let the games begin!