
Network Transport Layer: Network Resource Allocation Framework

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<https://qiaoxiang.me/courses/cnns-xmuf21/index.shtml>

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Admin.

- ❑ Lab assignment 3 & 4 due on Nov. 30
- ❑ Lab assignment 5 overview this afternoon

Outline

- ❑ Admin and recap
- ❑ Transport congestion control
 - what is congestion (cost of congestion)
 - basic congestion control alg.
 - TCP/Reno congestion control
 - TCP Cubic
 - TCP/Vegas
 - network wide resource allocation
 - general framework
 - objective function: axiom derivation of network-wide objective function
 - algorithm: general distributed algorithm framework
 - application: TCP/Reno TCP/Vegas revisited

Recap: TCP/Reno Throughput Modeling

$$\Delta W = \begin{cases} \frac{1}{W} & \text{if the packet is not lost} \\ -\frac{W}{2} & \text{if packet is lost} \end{cases}$$

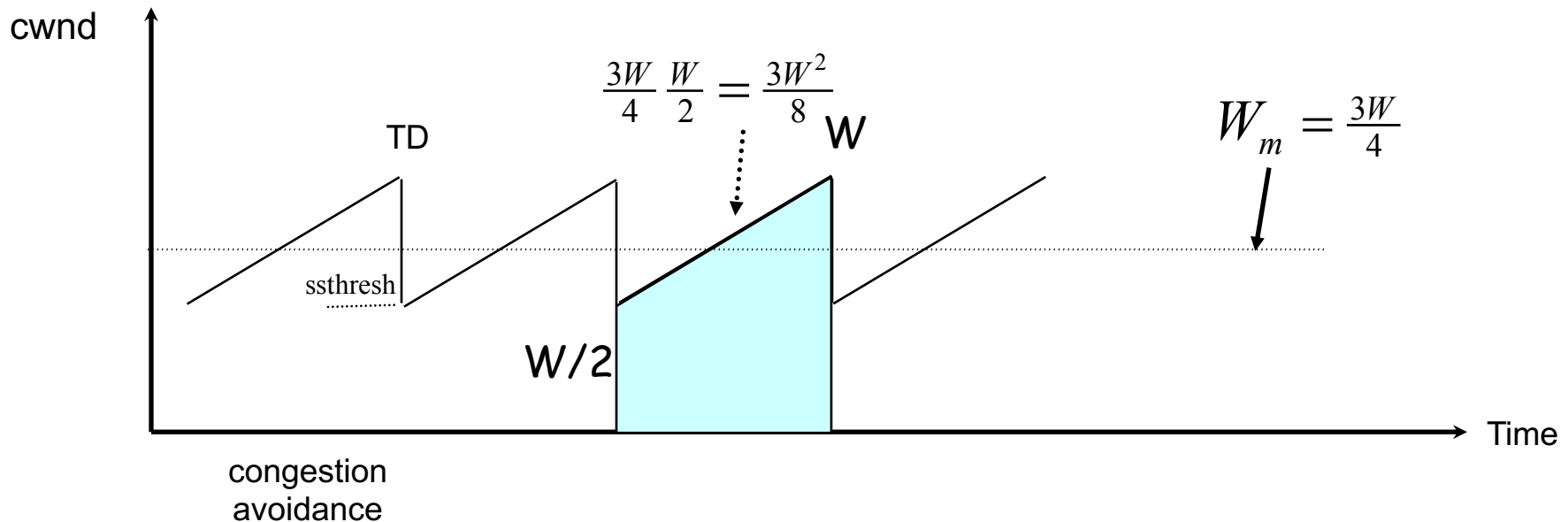
$$\text{mean of } \Delta W = (1-p)\frac{1}{W} + p(-\frac{W}{2}) = 0$$

$$\Rightarrow \text{mean of } W = \sqrt{\frac{2(1-p)}{p}} \approx \frac{1.4}{\sqrt{p}}, \text{ when } p \text{ is small}$$

$$\Rightarrow \text{throughput} \approx \frac{1.4S}{RTT\sqrt{p}}, \text{ when } p \text{ is small}$$

This is called the TCP throughput sqrt of loss rate law.

Recap: TCP/Reno Throughput Modeling



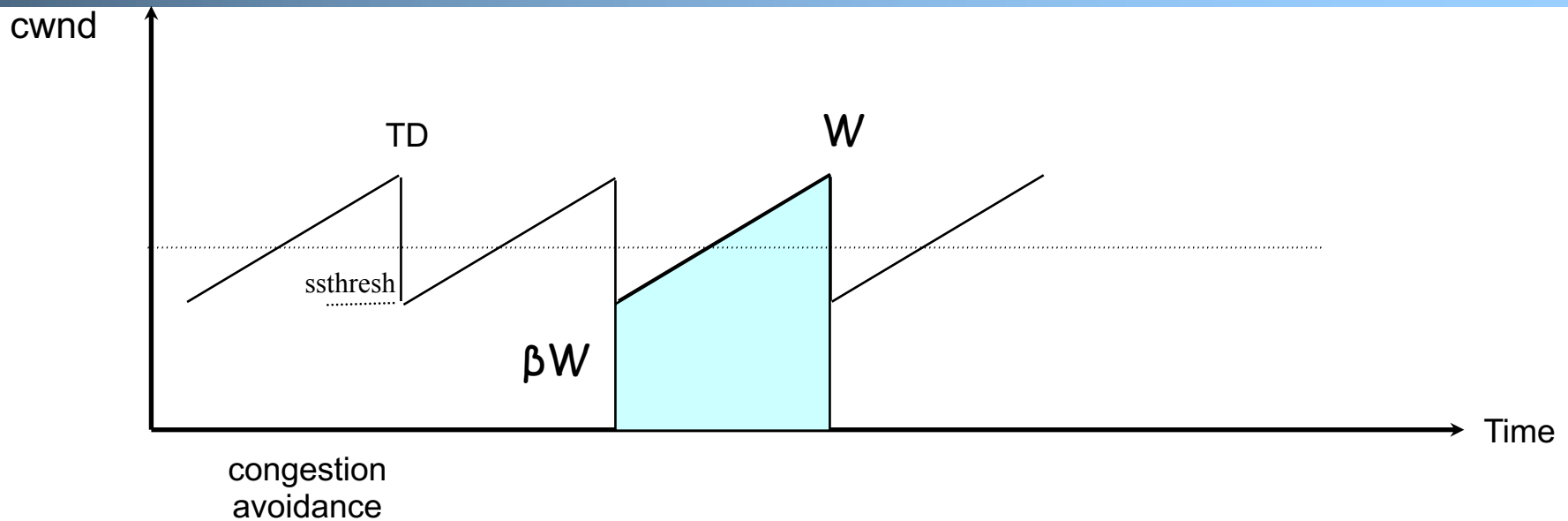
Total packets sent per cycle = $(W/2 + W)/2 * W/2 = 3W^2/8$

Assume one loss per cycle $\Rightarrow p = 1/(3W^2/8) = 8/(3W^2)$

$$\Rightarrow W = \frac{\sqrt{8/3}}{\sqrt{p}} = \frac{1.6}{\sqrt{p}}$$

$$\Rightarrow \text{throughput} = \frac{S}{RTT} \frac{3}{4} \frac{1.6}{\sqrt{p}} = \boxed{\frac{1.2S}{RTT \sqrt{p}}}$$

Recap: Generic AIMD and TCP Friendliness



$$\text{Total packets sent per cycle} = \frac{\beta W + W}{2} \frac{(1-\beta)W}{\alpha} = \frac{(1-\beta)(1+\beta)}{2\alpha} W^2$$

$$\text{Assume one loss per cycle } p = \frac{2\alpha}{(1-\beta)(1+\beta)W^2} \quad W = \sqrt{\frac{2\alpha}{(1-\beta)(1+\beta)p}}$$

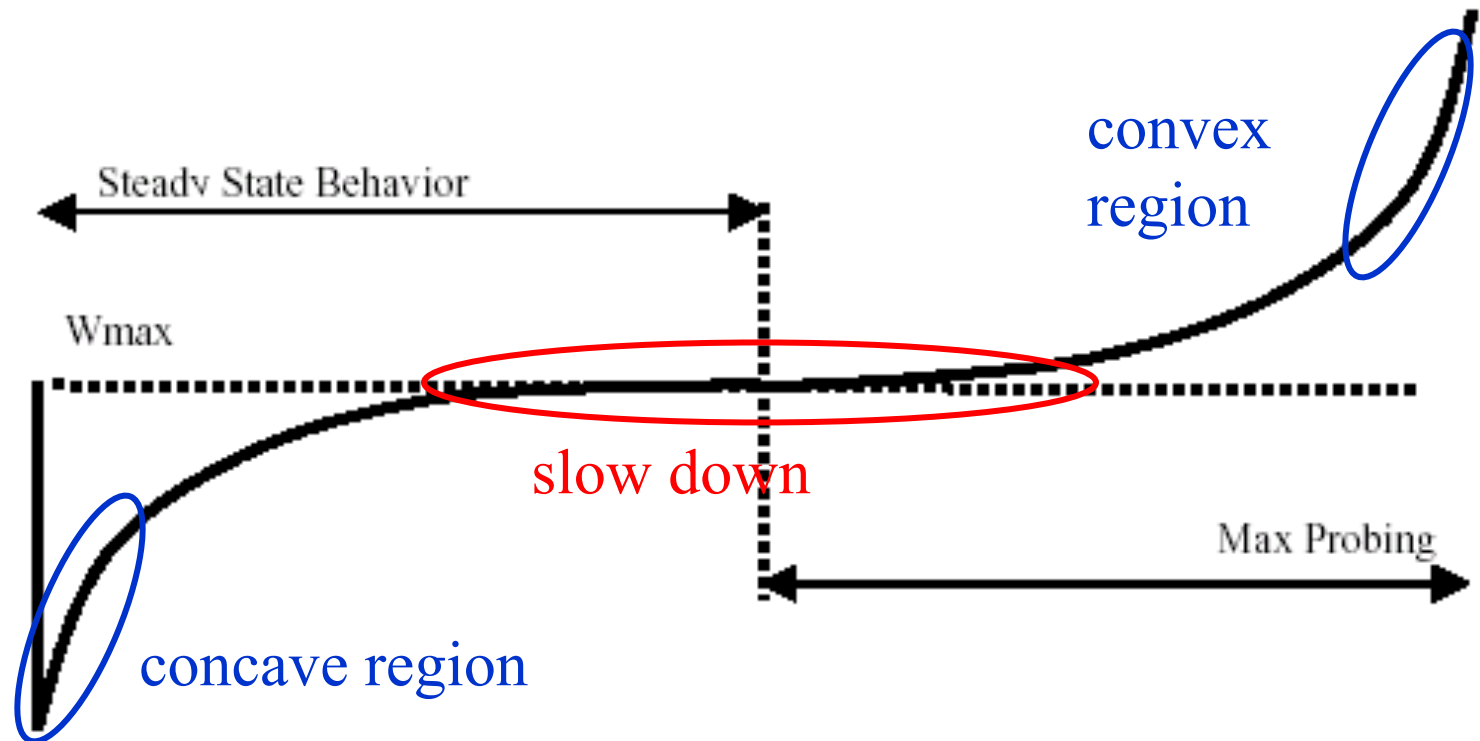
$$t_{\text{put}} = \frac{W_m S}{RTT} = \frac{S}{RTT} \frac{(1+\beta)W}{2} = \frac{S}{RTT} \sqrt{\frac{\alpha(1+\beta)}{2(1-\beta)p}}$$

$$\text{TCP friendly} \Rightarrow \alpha = 3 \frac{1-\beta}{1+\beta}$$

$$\beta' = 1 - \beta$$

Recap: TCP Cubic

$$W_{\text{tcp}(t)} = W_{\text{max}} \beta' + 3 \frac{1-\beta'}{1+\beta'} \frac{t}{RTT}$$



$$W_{\text{cubic}} = C(t - K)^3 + W_{\text{max}} \quad K = \sqrt[3]{W_{\text{max}} \beta / C}$$

where C is a scaling factor, t is the elapsed time from the last window reduction, and β is a constant multiplication decrease factor

Outline

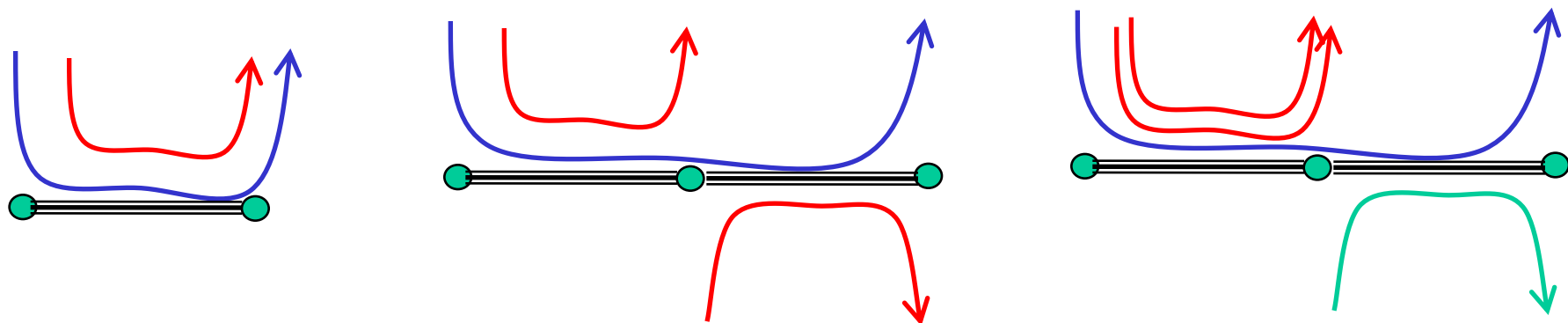
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Motivation

- ❑ So far our discussion is implicitly on a network with a single bottleneck link; this simplifies design and analysis:
 - efficiency/optimality (high utilization)
 - fully utilize the bandwidth of the link
 - fairness (resource sharing)
 - each flow receives an *equal* share of the link's bandwidth

Network Resource Allocation

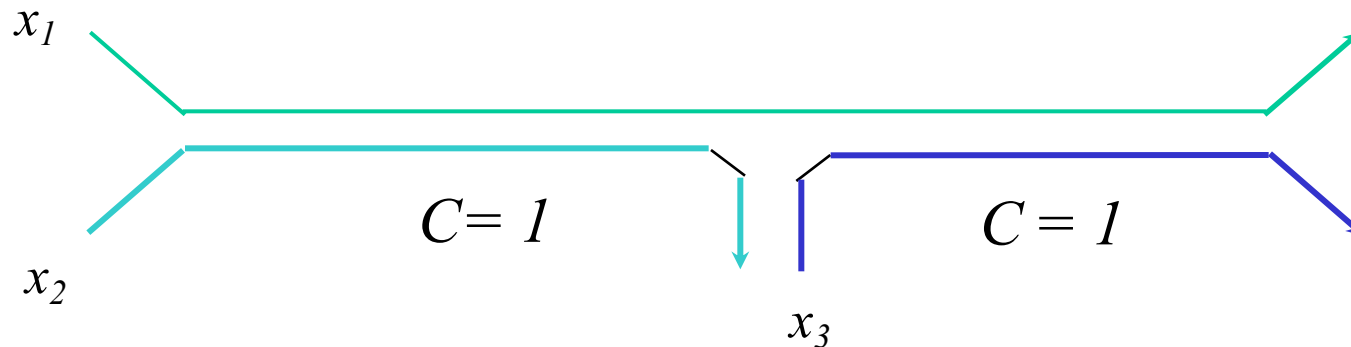
- It is important to understand and design protocols for a general network topology
 - how **will** TCP allocate resource in a **general topology**?
 - how **should** resource be allocated in a **general topology**?



Example: TCP/Reno Rates

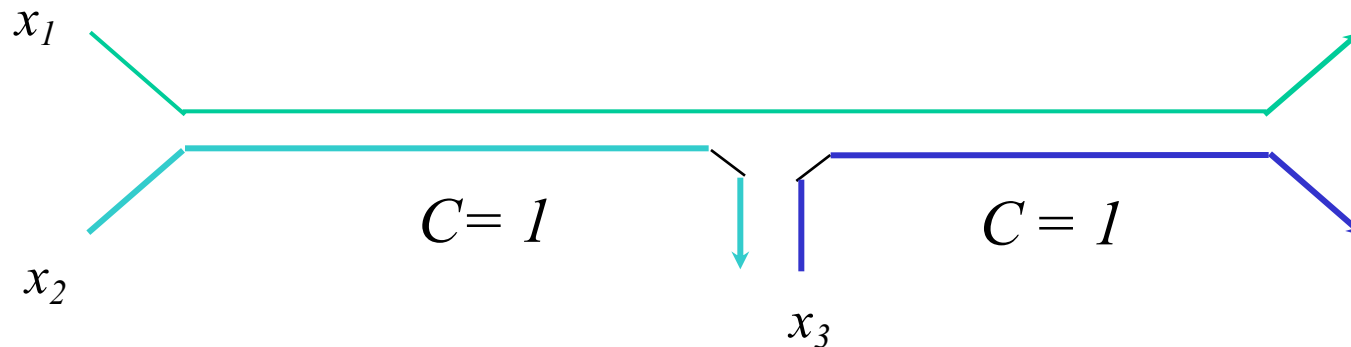
■ Rates:

$$x_1 = \frac{1}{1+2\sqrt{2}} = 0.26$$
$$x_2 = x_3 = 0.74$$



Example: TCP/Vegas Rates

■ Rates : $x_1 = 1/3$
 $x_2 = x_3 = 2/3$



Example: Max-min Fairness

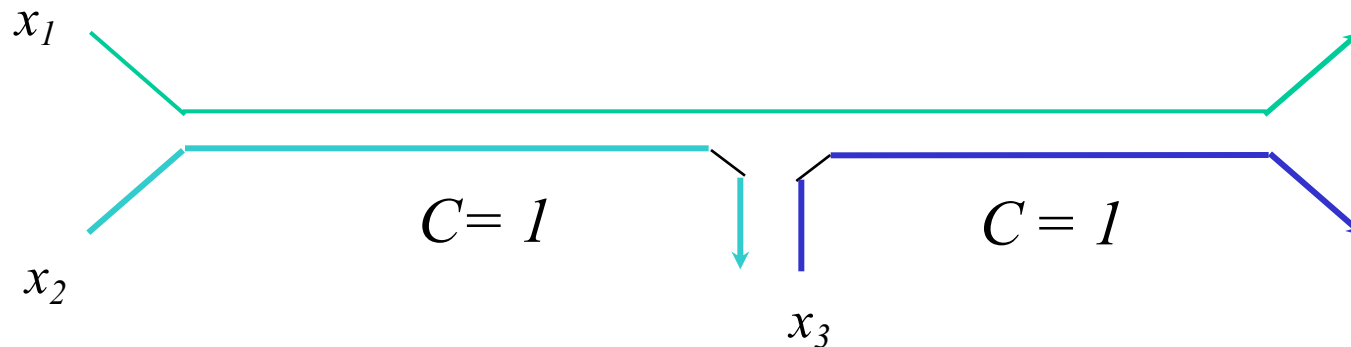


- Max-min fairness: maximizes the throughput of the flow receiving the minimum (of resources)
 - Justification: John Rawls, *A Theory of Justice* (1971)
 - http://en.wikipedia.org/wiki/John_Rawls
 - This is a resource allocation scheme used in ATM and some other network resource allocation proposals

Example: Max-Min

$$\begin{array}{ll} \max_{x_f \geq 0} & \min\{x_f\} \\ \text{subject to} & x_1 + x_2 \leq 1 \\ & x_1 + x_3 \leq 1 \end{array}$$

■ Rates: $x_1 = x_2 = x_3 = 1/2$



Framework: Network Resource Allocation Using Utility Functions

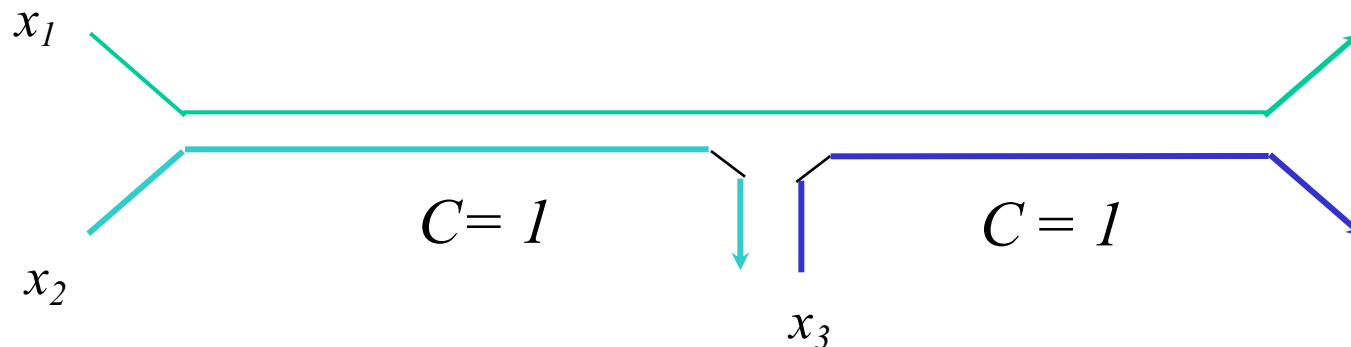
- A set of flows F
- Let x_f be the rate of flow f , and the utility to flow f is $U_f(x_f)$.
- Maximize aggregate utility, subject to capacity constraints

$$\begin{array}{ll} \max & \sum_{f \in F} U_f(x_f) \\ \text{subject to} & \sum_{f: f \text{ uses link } l} x_f \leq c_l \text{ for any link } l \\ \text{over} & x \geq 0 \end{array}$$

Example: Maximize Throughput

$$\begin{array}{ll} \max_{x_f \geq 0} & \sum_f x_f \\ \text{subject to} & x_1 + x_2 \leq 1 \\ & x_1 + x_3 \leq 1 \end{array} \quad U_f(x_f) = x_f$$

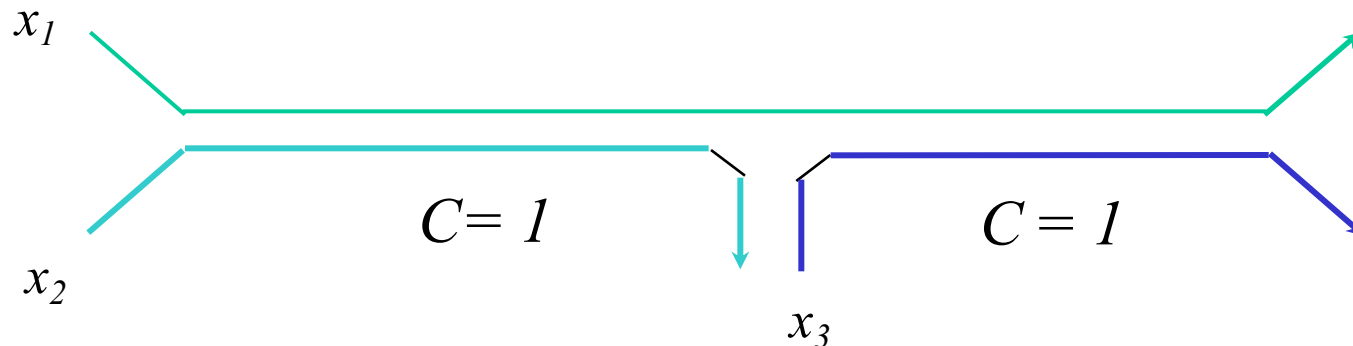
■ Optimal: $x_1 = 0$
 $x_2 = x_3 = 1$



Example: Proportional Fairness

$$\begin{array}{ll} \max_{x_f \geq 0} & \sum_f \log x_f \\ \text{subject to} & x_1 + x_2 \leq 1 \\ & x_1 + x_3 \leq 1 \end{array} \quad U_f(x_f) = \log(x_f)$$

■ Optimal: $x_1 = 1/3$
 $x_2 = x_3 = 2/3$



Example 3: a "Funny" Utility Function

$$\max_{x_f \geq 0} \quad -\frac{1}{4x_1} - \frac{1}{x_2} - \frac{1}{x_3}$$

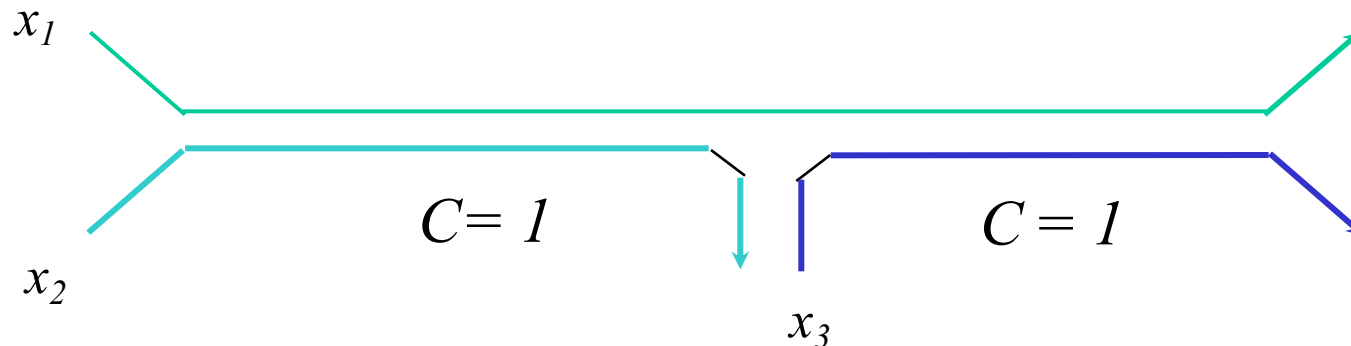
subject to

$$x_1 + x_2 \leq 1$$

$$x_1 + x_3 \leq 1$$

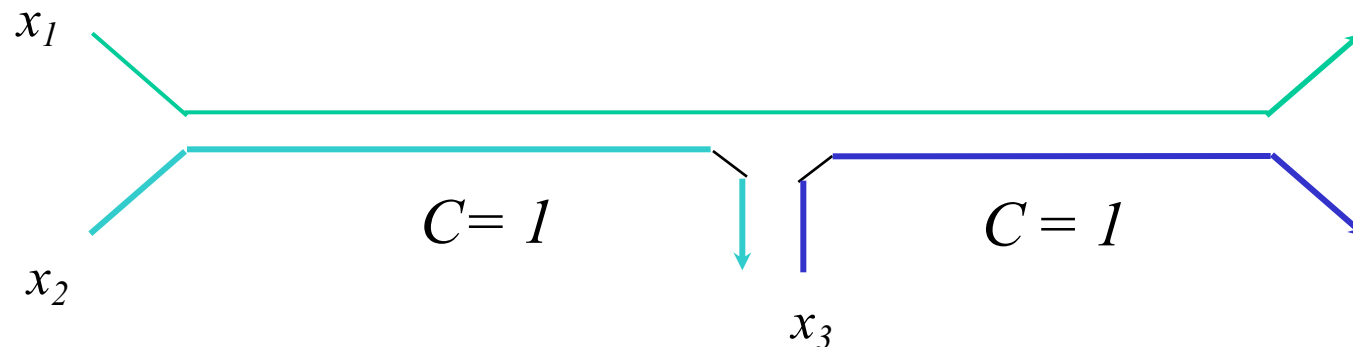
$$U_f(x_f) = -\frac{1}{RTT^2 x_f}$$

■ Optimal: $x_1 = \frac{1}{1+2\sqrt{2}} = 0.26$
 $x_2 = x_3 = 0.74$



Summary: Allocations

Objective	Allocation (x_1 , x_2 , x_3)		
TCP/Reno	0.26	0.74	0.74
TCP/Vegas	$1/3$	$2/3$	$2/3$
Max Throughput	0	1	1
Max-min	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$
Max sum $\log(x)$	$1/3$	$2/3$	$2/3$
Max sum of $-1/(RTT^2 x)$	0.26	0.74	0.74



Questions

$$\begin{array}{ll}
 \max & \sum_{f \in F} U_f(x_f) \\
 \text{subject to} & \sum_{f: f \text{ uses link } l} x_f \leq c_l \text{ for any link } l \\
 \text{over} & x \geq 0
 \end{array}$$

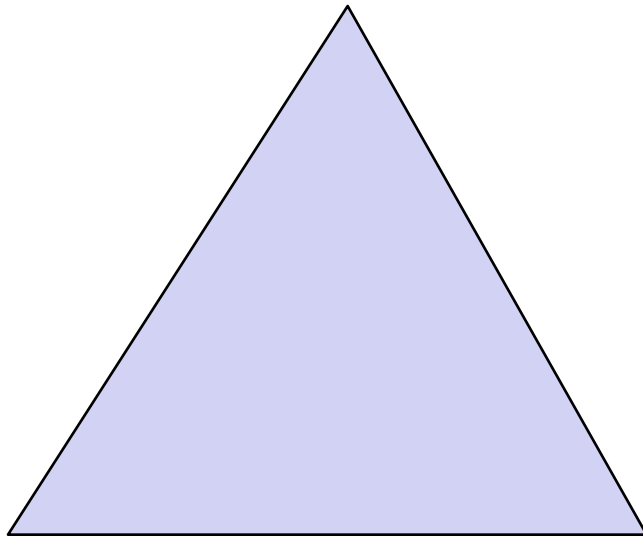
- Forward engineering: systematically
 - design objective function
 - design distributed alg to achieve objective
- Science/reverse engineering: what do TCP/Reno, TCP/Vegas achieve?

Objective	Allocation (x1, x2, x3)		
TCP/Reno	0.26	0.74	0.74
TCP/Vegas	1/3	2/3	2/3
Max throughput	0	1	1
Max-min	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$
Max sum $\log(x)$	1/3	2/3	2/3
Max sum of $-1/(RTT^2 x)$	0.26	0.74	0.74

Outline

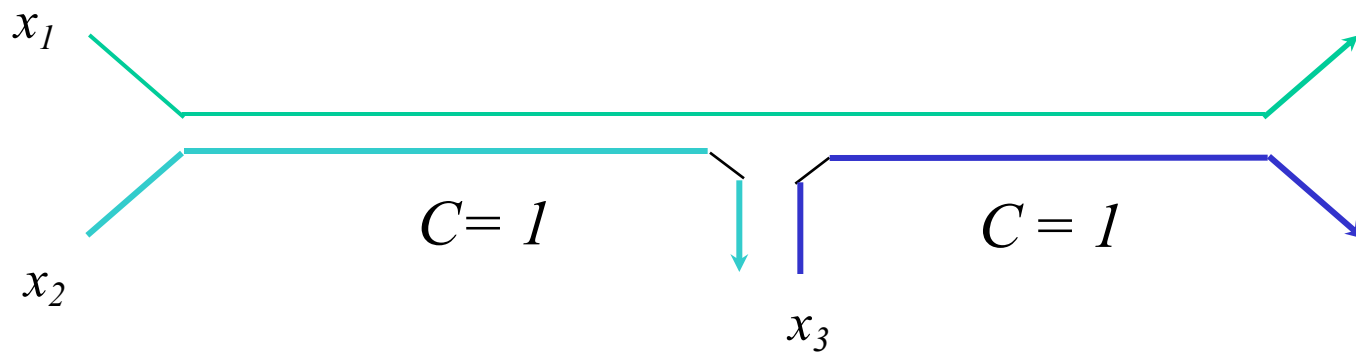
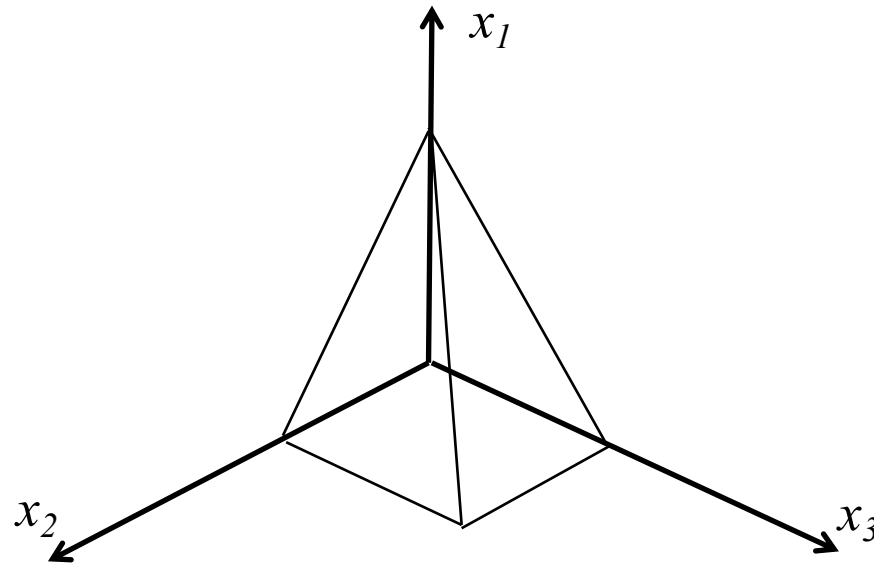
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 - objective function: an example of an axiom derivation of network-wide objective function

Network Bandwidth Allocation Using Nash Bargain Solution (NBS)



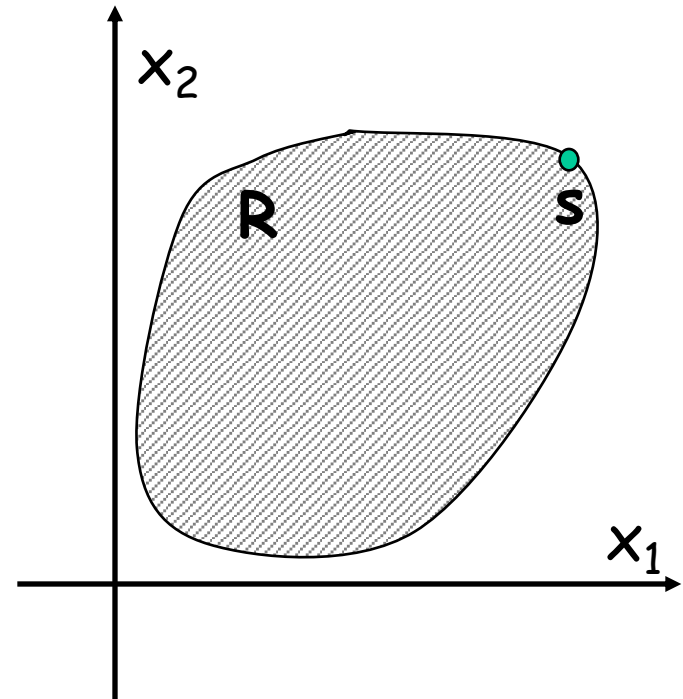
- High level picture
 - given the feasible set of bandwidth allocation, we want to pick an allocation point that is efficient and fair
- The determination of the allocation point should be based on "first principles" (axioms)

Network Bandwidth Allocation: Feasible Region



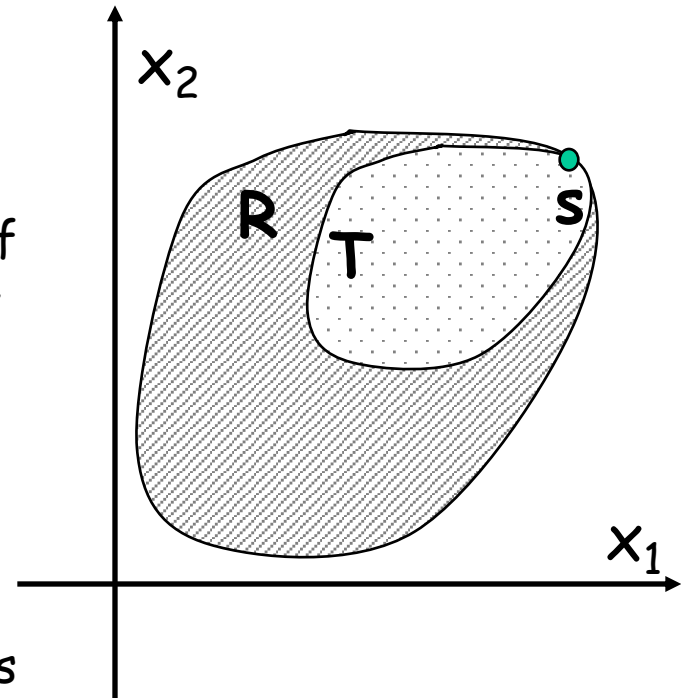
Nash Bargain Solution (NBS)

- Assume a finite, convex feasible set in the first quadrant
- Axioms



Nash Bargain Solution (NBS)

- Assume a finite, convex feasible set in the first quadrant
- Axioms
 - Pareto optimality
 - impossibility of increasing the rate of one user without decreasing the rate of another
 - symmetry
 - a symmetric feasible set yields a symmetric outcome
 - invariance of linear transformation
 - the allocation must be invariant to linear transformations of users' rates
 - independence of irrelevant alternatives
 - assume s is an allocation when feasible set is R , $s \in T \subset R$, then s is also an allocation when the feasible set is T



Nash Bargain Solution (NBS)

- Surprising result by John Nash (1951)

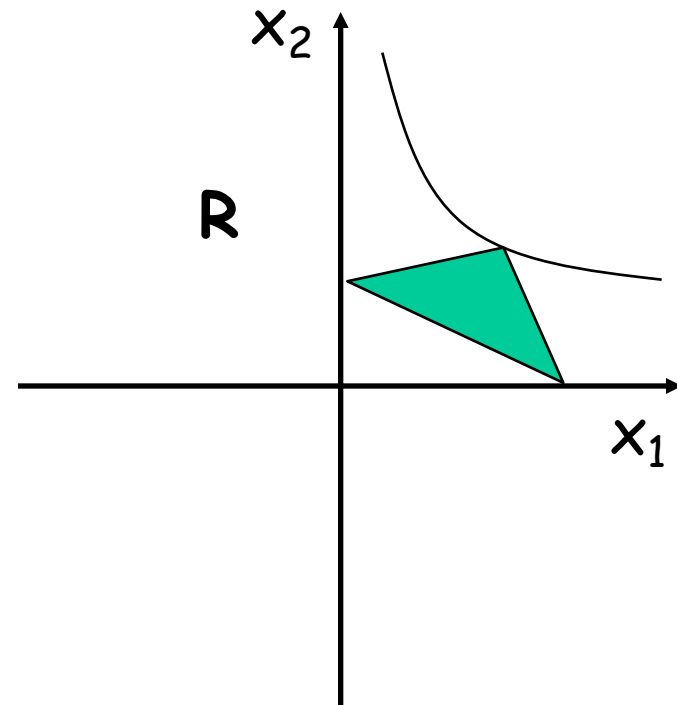
- the rate allocation point is the feasible point which maximizes

$$x_1 x_2 \cdots x_F$$

- This is equivalent to maximize

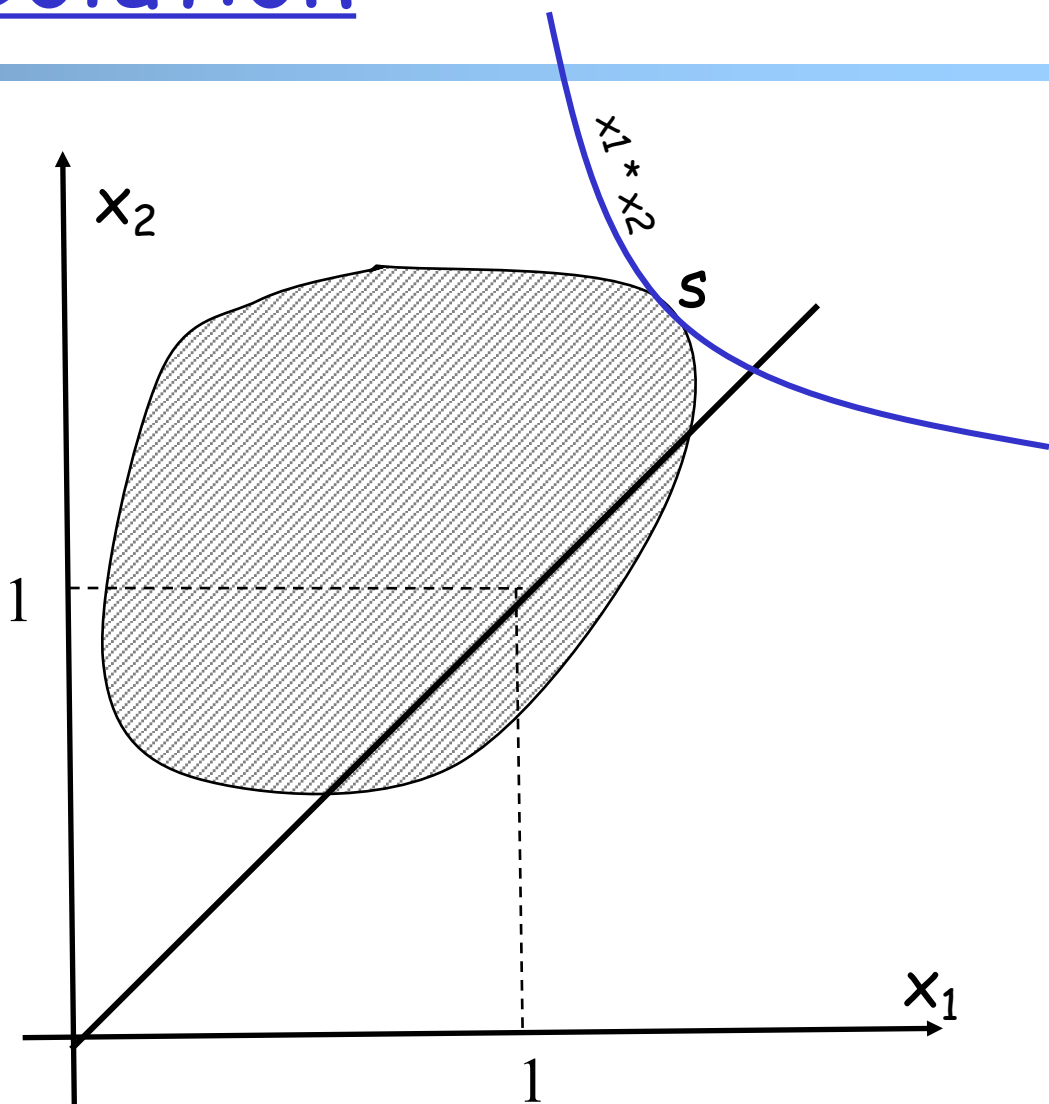
$$\sum_f \log(x_f)$$

- In other words, assume each flow f has utility function $\log(x_f)$
- I will give a proof for $F = 2$
 - think about $F > 2$



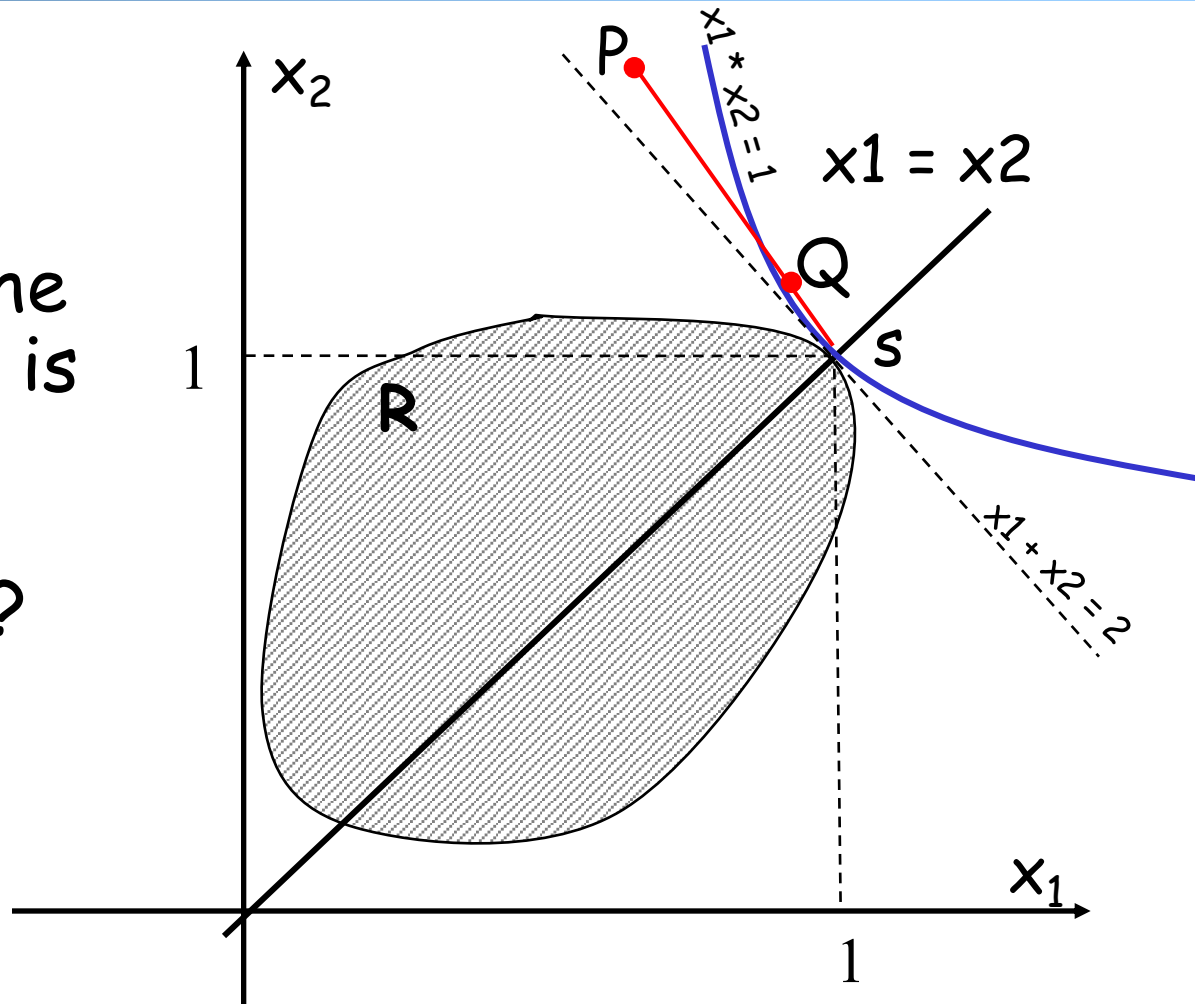
Nash Bargain Solution

- Assume s is the feasible point which maximizes $x_1 * x_2$
- Scale the feasible set so that s is at $(1, 1)$
 - how?



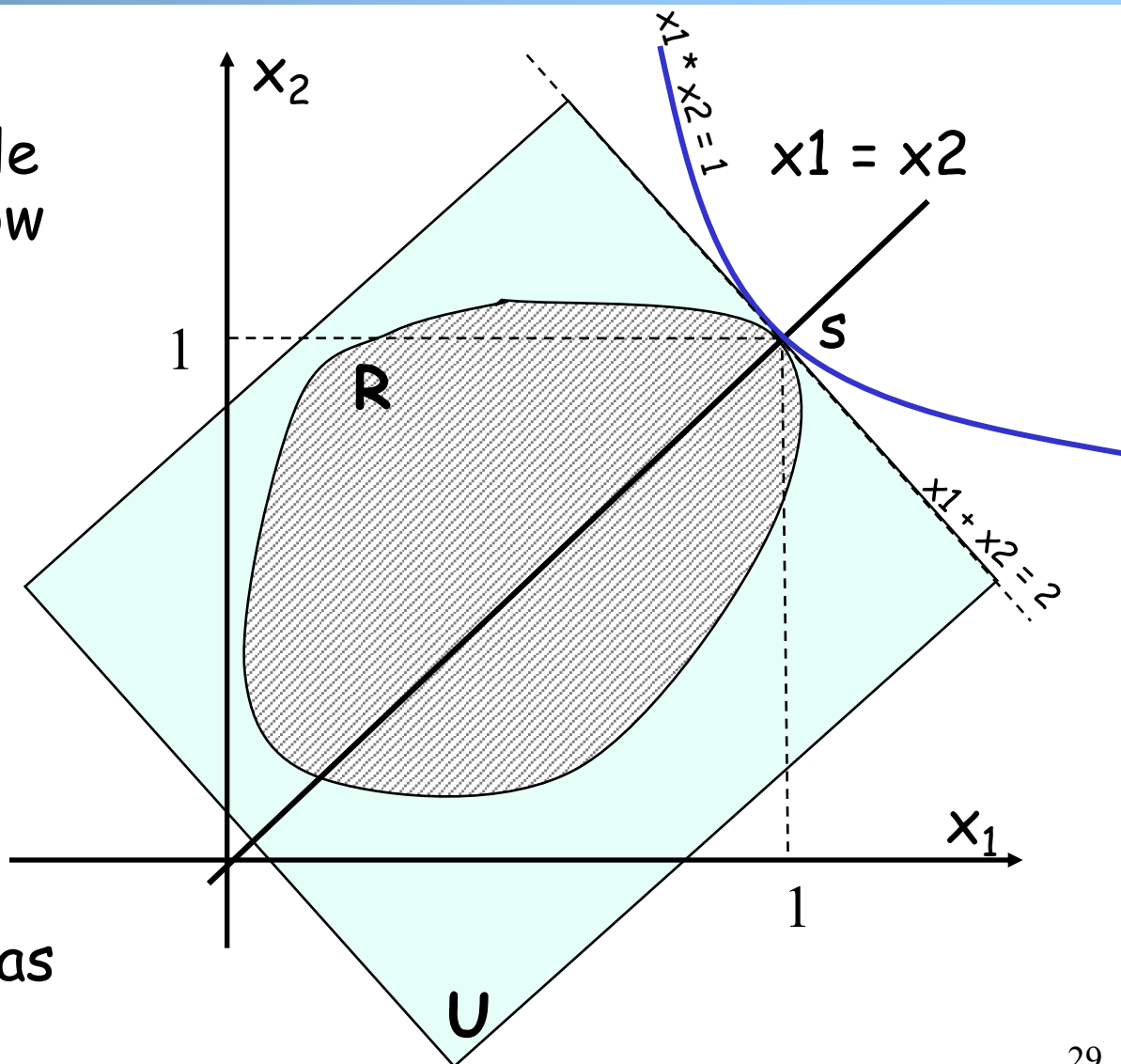
Nash Bargain Solution

Question: after the transformation, is there any feasible point with $x_1 + x_2 > 2$?



Nash Bargain Solution

- Consider the symmetric rectangle U containing the now feasible set
→ According to symmetry and Pareto, s is the allocation when feasible set is U
- According to independence of irrelevant alternatives, the allocation of R is s as well.



NBS \Leftrightarrow Proportional Fairness

- Allocation is proportionally fair if for any other allocation, aggregate of proportional changes is non-positive, e.g. if x_f is a proportional-fair allocation, and y_f is any other feasible allocation, then require

$$\sum_f \frac{y_f - x_f}{x_f} \leq 0$$

Questions to Think

- Vary the axioms and see if you can derive any objective functions

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Recall: Resource Allocation Framework

□ The Resource-Allocation Problem:

$$\begin{array}{ll} \max & \sum_{f \in F} U_f(x_f) \\ \text{subject to} & \sum_{f: f \text{ uses link } l} x_f \leq c_l \text{ for any link } l \\ \text{over} & x \geq 0 \end{array}$$

□ Goal: Design a distributed alg to solve the problem.

□ Discussion:

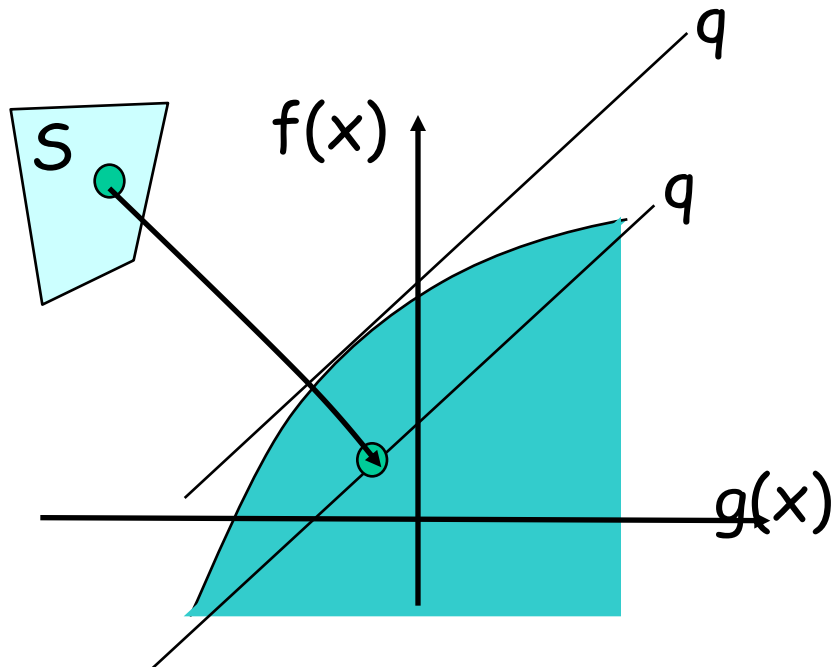
- What are typical approaches to solve optimization, e.g.,?
 $\max U(x)$
- Why is the Resource-Allocation problem hard to solve by a distributed algorithm?

A Two-Slide Summary of Constrained Convex Optimization Theory

$$\begin{array}{ll} \max & \sum_{f \in F} U_f(x_f) \\ \text{subject to} & Ax \leq C \\ \text{over} & x \geq 0 \end{array}$$

$$\begin{array}{ll} \max & f(x) \\ \text{subject to} & g(x) \leq 0 \\ \text{over} & x \in S \end{array}$$

$f(x)$ concave
 $g(x)$ linear
 S is a convex set



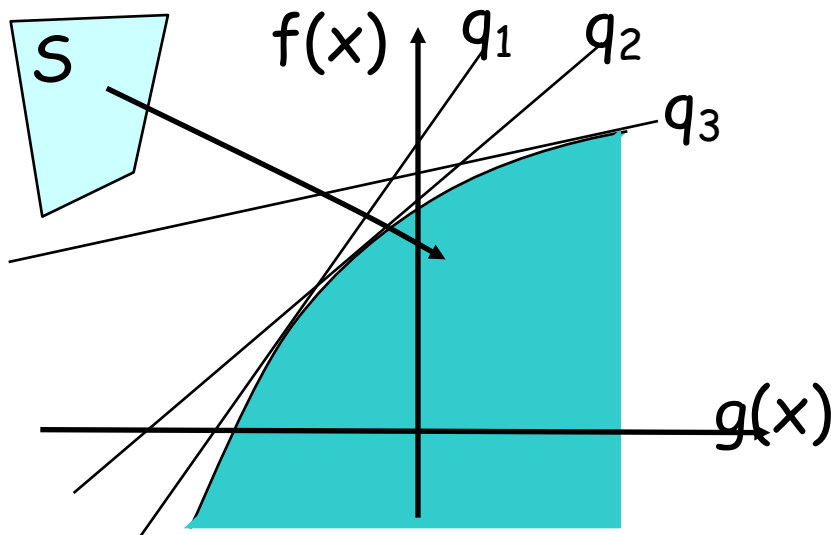
- Map each x in S , to $[g(x), f(x)]$
- Top contour of map is concave
- Easy to read solution from contour
- For each slope q (≥ 0), computes $f(x) - q g(x)$ of all mapped $[f(x), g(x)]$

$$D(q) = \max_{x \in S} (f(x) - qg(x))$$

A Two-Slide Summary of Constrained Convex Optimization Theory

$$\begin{array}{ll} \max & f(x) \\ \text{subject to} & g(x) \leq 0 \\ \text{over} & x \in S \end{array}$$

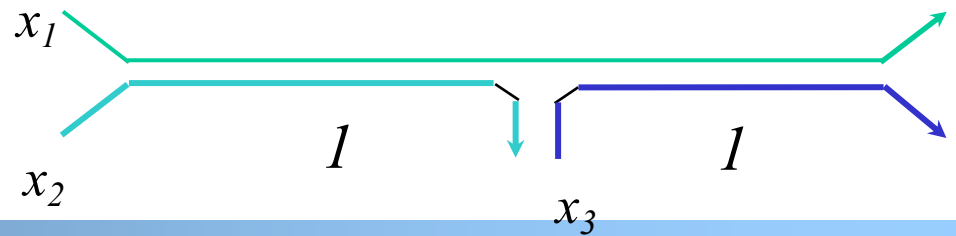
$f(x)$ concave
 $g(x)$ linear
 S is a convex set



$$D(q) = \max_{x \in S} (f(x) - qg(x))$$

- $D(q)$ is called the dual;
- q (≥ 0) are called prices in economics
- $D(q)$ provides an upper bound on obj.
- According to optimization theory: when $D(q)$ achieves minimum over all q (≥ 0), then the optimization objective is achieved.

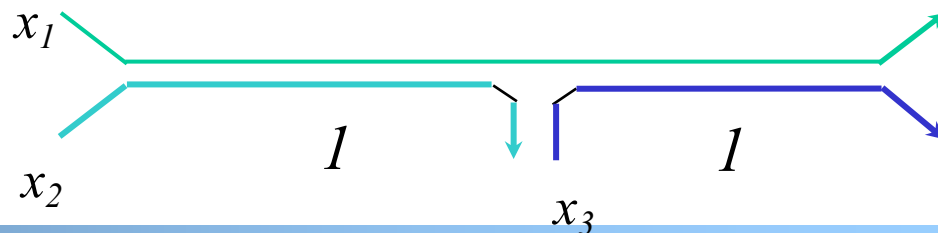
Dual of the Primal



$$\begin{array}{ll} \max & \sum_{f \in F} U_f(x_f) \\ \text{subject to} & \sum_{f: f \text{ uses link } l} x_f \leq c_l \text{ for any link } l \\ \text{over} & x \geq 0 \end{array}$$

$$D(q) = \max_{x_f \geq 0} \left(\sum_f U_f(x_f) - \sum_l q_l \left(\sum_{f: \text{uses } l} x_f - c_l \right) \right)$$

Dual of the Primal



$$\begin{aligned} D(q) &= \max_{x_f \geq 0} \left(\sum_f U_f(x_f) - \sum_l q_l \left(\sum_{f: \text{uses } l} x_f - c_l \right) \right) \\ &= \max_{x_f \geq 0} \sum_f \left(U_f(x_f) - x_f \sum_{l: f \text{ uses } l} q_l \right) + \sum_l q_l c_l \\ &= \sum_f \max_{x_f \geq 0} \left(U_f(x_f) - x_f \sum_{l: f \text{ uses } l} q_l \right) + \sum_l q_l c_l \end{aligned}$$

Distributed Optimization: User Problem

- Given p_f (=sum of dual var q_l along the path) flow f chooses rate x_f to maximize:

$$\begin{array}{ll} \max_{x_f} & U_f(x_f) - x_f p_f \\ \text{over} & x_f \geq 0 \end{array}$$

- Using the price signals, the optimization problem of each user is independent of each other!

Distributed Optimization: User Problem

$$\begin{array}{ll} \max_{x_f} & U_f(x_f) - x_f p_f \\ \text{over} & x_f \geq 0 \end{array}$$

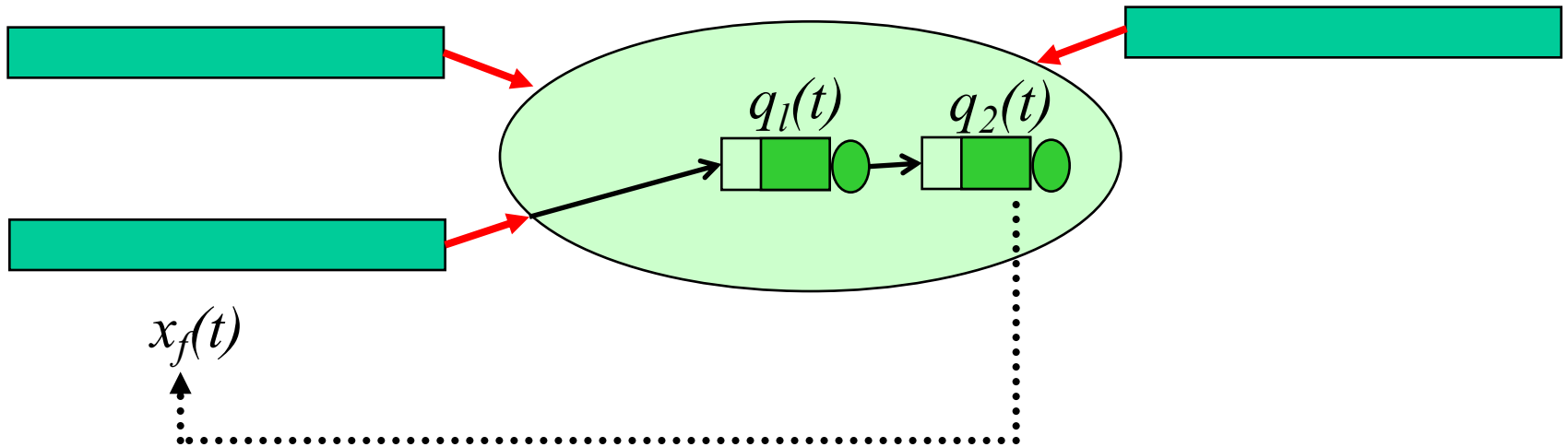
How should flow f adjust x_f locally?

$$\Delta x_f \propto U'_f(x_f) - p_f$$

At equilibrium (i.e., at optimal), x_f satisfies:

$$U'_f(x_f) - p_f = 0$$

Interpreting Congestion Measure



$$p_f = \sum_{f \text{ uses } l} q_l$$

$$\Delta x_f \propto U'_f(x_f) - p_f$$

Distributed Optimization: Network Problem

$$D(q) = \sum_f \max_{x_f \geq 0} \left(U_f(x_f) - x_f \sum_{l: f \text{ uses } l} q_l \right) + \sum_l q_l c_l$$

The network (i.e., link l) adjusts the link signals q_l (assume after all flows have picked their optimal rates given congestion signal)

$$\min_{q \geq 0} \tilde{D}(q) = \sum_l q_l (c_l - \sum_{f: f \text{ uses } l} x_f)$$

Distributed Optimization: Network Problem

$$\min_{q \geq 0} D(q) = \sum_l q_l (c_l - \sum_{f: f \text{ uses } l} x_f)$$

how should link l adjust q_l locally?

$$\Delta q_l \propto -\frac{\partial D(q)}{q_l}$$

$$\frac{\partial}{\partial q_l} D(q) = c_l - \sum_{f: \text{uses } l} x_f$$

$$\Delta q_l \propto \sum_{f: \text{uses } l} x_f - c_l$$

System Architecture

□ SYSTEM(U):

$$\begin{array}{ll}\max & \sum_{f \in F} U_f(x_f) \\ \text{subject to} & \sum_{f: f \text{ uses link } l} x_f \leq c_l \text{ for any link } l \\ \text{over} & x \geq 0\end{array}$$

□ USER_f:

$$\Delta x_f \propto U'_f(x_f) - p_f$$

$$\begin{array}{ll}\max_{x_f} & U_f(x_f) - x_f p_f \\ \text{over} & x_f \geq 0\end{array}$$

□ NETWORK:

$$\Delta q_l \propto -\frac{\partial D(q)}{q_l}$$

$$\min_{q \geq 0} \tilde{D}(q) = \sum_l q_l (c_l - \sum_{f: f \text{ uses } l} x_f)$$

Decomposition Theorem

- There exist vectors \mathbf{p} , \mathbf{w} and \mathbf{x} such that
 1. $w_f = p_f x_f$ for $f \in F$
 2. w_f solves $USER_f(U_f; p_f)$
 3. \mathbf{x} solves $NETWORK(\mathbf{w})$
- The vector \mathbf{x} then also solves $SYSTEM(\mathbf{U})$.

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TCP/Reno Dynamics

$$\Delta x_f \propto U'_f(x_f) - p_f$$

$$\Delta W_{pkt} = (1 - p) \frac{1}{W} - p \frac{W}{2}$$


$$\Delta W_{RTT} = \Delta W_{pkt} W = (1 - p) - p \frac{W^2}{2} \cong 1 - p \frac{W^2}{2}$$

$$\Delta x = \frac{\Delta W_{RTT}}{RTT} = \frac{1}{RTT} - \frac{RTT}{2} p x^2$$

$$= \frac{RTT}{2} x^2 \left(\frac{2}{x^2 RTT^2} - p \right)$$

TCP/Reno Dynamics $\Delta x_f \propto U'_f(x_f) - p_f$

$$\Delta x = \frac{RTT}{2} x^2 \left(\frac{2}{x^2 RTT^2} - p \right)$$

$$U'_f(x_f) - p_f$$


$$\Rightarrow U'_f(x_f) = \left(\frac{\sqrt{2}}{x_f RTT} \right)^2 \quad \Rightarrow U_f(x_f) = -\frac{2}{RTT^2 x_f}$$

TCP/Vegas Dynamics

$$\Delta x_f \propto U'_f(x_f) - p_f$$

$$\Delta W_{\text{RTT}} \approx -(w - xRTT_{\min} - \alpha)$$

$$\Delta x = \frac{\Delta W_{\text{RTT}}}{\text{RTT}} = -\left(\frac{w}{\text{RTT}} - \frac{x}{\text{RTT}}\text{RTT}_{\min} - \frac{\alpha}{\text{RTT}}\right)$$

$$= -\frac{w}{\text{RTT}} + \frac{x}{\text{RTT}}\text{RTT}_{\min} + \frac{\alpha}{\text{RTT}}$$

$$= -x + \frac{x}{\text{RTT}}\text{RTT}_{\min} + \frac{\alpha}{\text{RTT}}$$

$$= \frac{x}{\text{RTT}}(-\text{RTT} + \text{RTT}_{\min} + \frac{\alpha}{x})$$


$$= \frac{x}{\text{RTT}}\left(\frac{\alpha}{x} - (\text{RTT} - \text{RTT}_{\min})\right)$$

$$\begin{aligned}\Delta W &\simeq \alpha - \left(W - \frac{\text{RTT}_{\min}}{\text{RTT}}W\right) \\ &\simeq \alpha - \left(W - \frac{\text{RTT}_{\min}}{\text{RTT}}x\text{RTT}\right) \\ &\simeq -(W - x\text{RTT}_{\min} - \alpha)\end{aligned}$$

TCP/Vegas Dynamics

$$\Delta x_f \propto U'_f(x_f) - p_f$$

$$\Delta x = \frac{x}{RTT} \left(\frac{\alpha}{x} - (RTT - RTT_{\min}) \right)$$

$$U'_f(x_f) - p_f$$


$$\Rightarrow U'_f(x_f) = \frac{\alpha}{x}$$

$$\Rightarrow U_f(x_f) = \alpha \log(x_f)$$

Summary: TCP/Vegas and TCP/Reno

□ Pricing signal is queueing delay T_{queueing}

$$x_f = \frac{\alpha}{T_{\text{queueing}}}$$

$$U'_f(x_f) = T_{\text{queueing}}$$

$$\Rightarrow U'_f(x_f) = \frac{\alpha}{x_f}$$

$$\Rightarrow U_f(x_f) = \alpha \log(x_f)$$

□ Pricing signal is loss rate p

$$x_f = \frac{\alpha}{RTT \sqrt{p}}$$

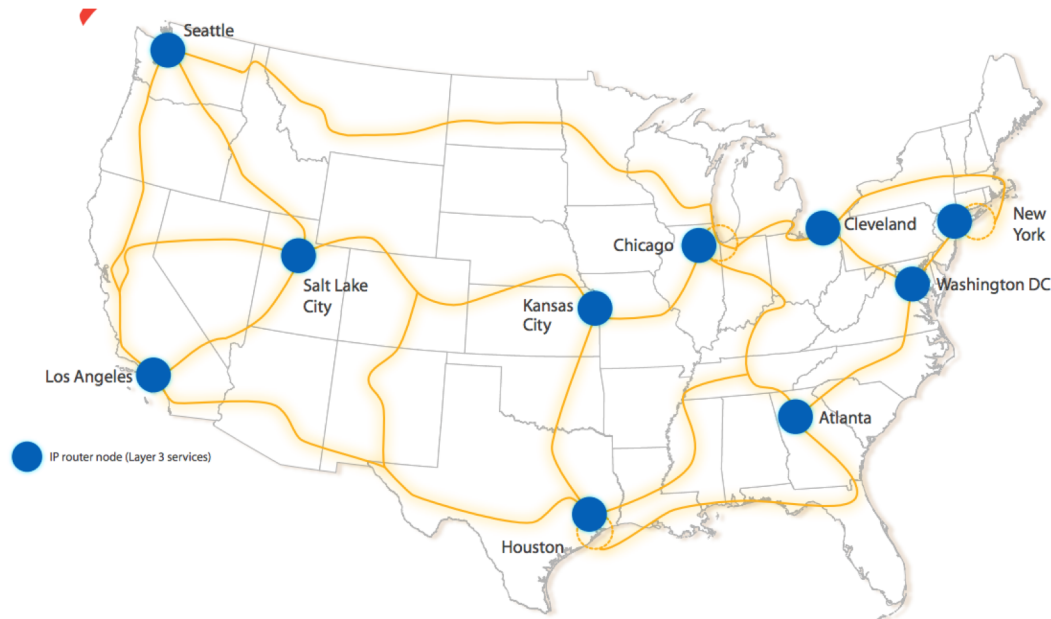
$$U'_f(x_f) = p$$

$$\Rightarrow U'_f(x_f) = \left(\frac{\alpha}{x_f RTT} \right)^2$$

$$\Rightarrow U_f(x_f) = -\frac{\alpha'}{RTT^2 x_f}$$

Discussion

- ❑ Assume that you are given a set of flows deployed at a given network topology.
- ❑ What is a simple way to predict TCP rate allocation?



Summary: Resource Allocation Frameworks

□ Forward (design) engineering:

- how to determine objective functions
- given objective, how to design effective alg

$$\begin{array}{ll} \max & \sum_{f \in F} U_f(x_f) \\ \text{subject to} & Ax \leq C \\ \text{over} & x \geq 0 \end{array}$$

□ Reverse (understand) engineering:

- understand current protocols (what are the objectives of TCP/Reno, TCP/Vegas?)

□ Additional pointers:

- <http://www.statslab.cam.ac.uk/~frank/pf/>