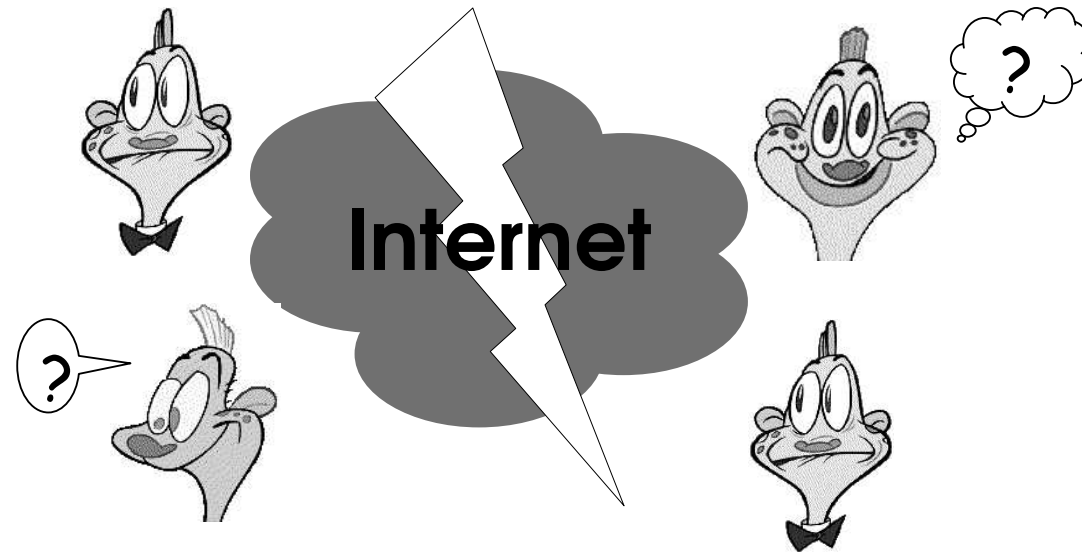


A Game-Theoretic Analysis of TCP

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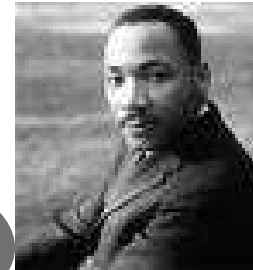
CMU/ICSI/UC Berkeley

1980s: Network Collapse



In the 80s's, naïve behavior caused the network to collapse

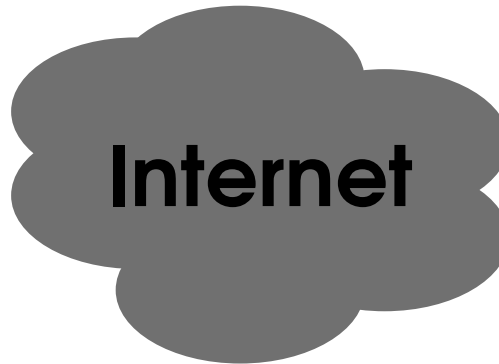
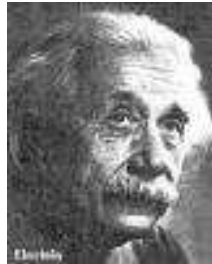
Salvation!



Internet

**Socially responsible congestion control
implemented at end-points was given credit for
saving the Internet**

Salvation?



Can the network survive with greedy (but intelligent) behavior?

Why Bother?

- If greed is bad, today's Internet is stable because –
 - End-points are consciously social and/or
 - It is hard to modify end-hosts to be greedy

We may need mechanisms to monitor, dissuade aggressive behavior

- If not, we need no such mechanism
 - Can rely on end-point behavior for efficient operation

Outline

- The TCP Game
- Results for the TCP Game
- Mechanisms for Nash Equilibrium

The TCP Game

- *TCP Game*
 - Flows attempt to maximize application throughput
 - Flows modify their AIMD parameters (α , β)
 - Must still provide reliability
- What happens at *Nash Equilibrium*?
 - No flow can gain in throughput by unilaterally changing its parameter choice

The TCP Game

- Analyze a simplified version of this Game for...
 - Parameters at the Nash Equilibrium
 - Efficiency at the Nash Equilibrium
 - Link Goodput and per-flow Loss rate
- Study symmetric Nash Equilibria only

Factors Affecting the Nash Equilibrium

(1) End-point's loss recovery mechanism

- Reno vs. SACK (primitive vs. modern)
- Depends on TCP implementation

(2) Loss assignment at routers

- Bursty loss assignment vs. randomized uniform losses

(3) Congestion control parameters of the flows

- How flows are allowed to vary their parameters
- Under complete control of end-point

End-points show greed by adjusting factor (3) alone. Factors (1), (2) are part of the environment.

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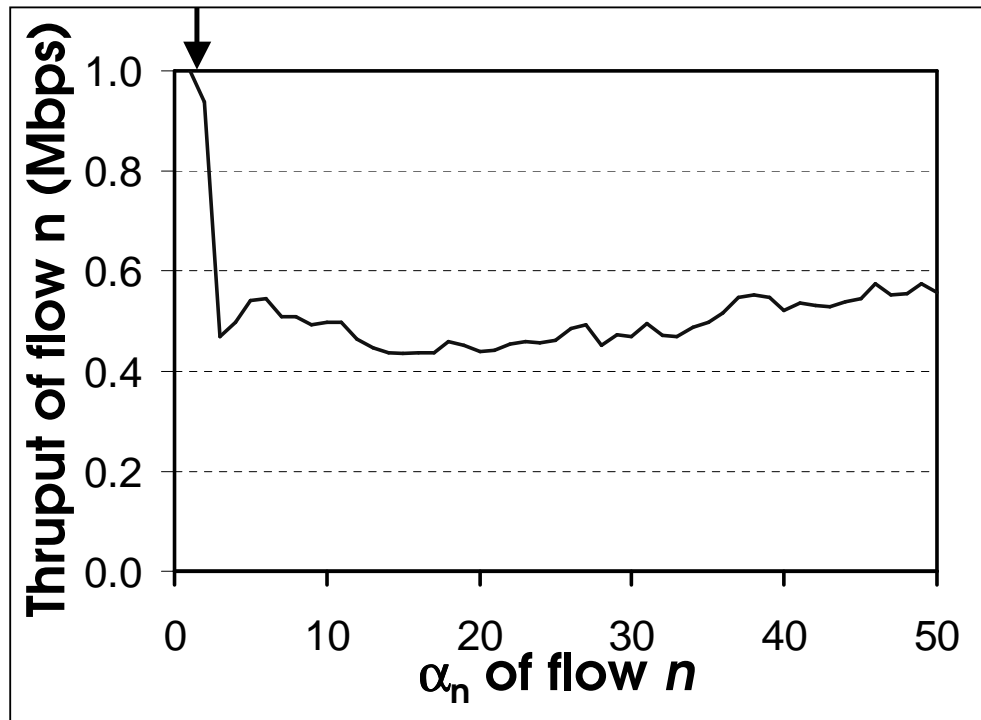
Results – Road-map

- First, consider FIFO droptail buffers
 - Most wide-spread in today's Internet
 - Efficiency at Nash Equilibrium for Tahoe, Reno, SACK-style loss recovery
- Then, discuss RED buffers briefly
 - As above
- Put the results together

FIFO Droptail Buffering

- Droptail buffers punish bursty behavior
 - Unintentionally so
 - Observed by designers of RED AQM
 - Flows with bursty transmission incur losses proportional to their burstiness
 - AIMD flows incur losses in bursts of size $\sim \alpha$ (AI parameter)

Results for FIFO Droptail Buffers – A Sample



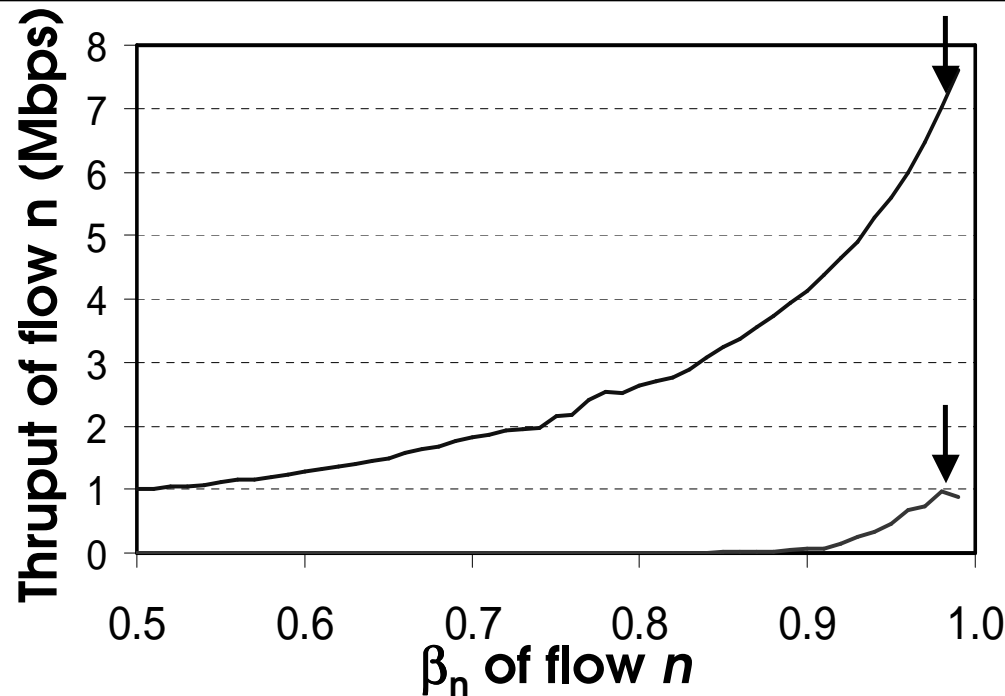
— $\alpha_1 \dots \alpha_{n-1} = 1$

**Reno-style loss recovery
(flows vary α , keeping $\beta=0.5$)**

α_E	Link Goodput	Per-flow Loss rate
1	100%	0.15%

- Greedy flows don't gain by using large α 's
 - Flows observe burst losses
 - Reno's severe reaction (time-outs) kicks in

Results for FIFO Droptail Buffers – A Sample



— $\beta_1 \dots \beta_{n-1} = 0.5$
 - - - $\beta_1 \dots \beta_{n-1} = 0.98$

**Reno-style loss recovery
 (flows vary β , keeping $\alpha=1$)**

β_E	Link Goodput	Per-flow Loss rate
0.98	100%	1.5%

- Greedy flows gain by using $\beta \rightarrow 1$
 - No burst losses since $\alpha=1$

Results for FIFO Droptail Buffers – A Sample

**Reno-style loss recovery
(flows vary α , β together)**

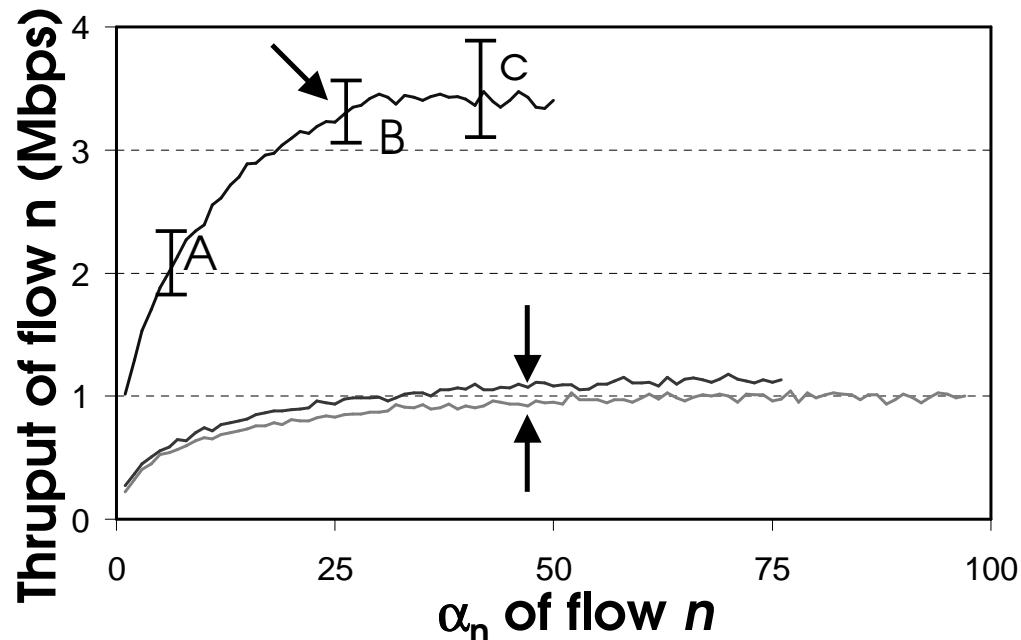
(α_E, β_E)	Link Goodput	Per-flow Loss rate
(1, 0.98)	100%	1.5%

- Nash Equilibrium is efficient!
 - Goodput is high and loss rate is low
 - Greedy behavior might work out
- But unfair
 - Since $\beta \rightarrow 1$ (AIAD)

RED Buffering

- RED buffers “spread” losses uniformly across flows
 - Identical loss %-age across flows irrespective of parameters used
 - Greater greed of a few flows causes a small increase in overall loss rate
 - Bursty flows do not experience burst losses, unlike droptail buffers

Results for RED Buffers – A Sample



— $\alpha_1 \dots \alpha_{n-1} = 1$
 — $\alpha_1 \dots \alpha_{n-1} = 27$
 — $\alpha_1 \dots \alpha_{n-1} = 48$

**SACK-style loss recovery
(flows vary α alone; $\beta=0.5$)**

α_E	Link Goodput	Per-flow Loss rate
48	96%	5%

- Aggression is always good
 - TCP SACK → high loss rate doesn't affect goodput
 - RED → greater aggression will cause minor increase in overall loss rate

Results for RED Buffers – A Sample

SACK-style loss recovery

Flows vary α, β together

(α_E, β_E)	Link Goodput	Per-flow Loss rate
(23, 0.98)	96%	5.70%

- Nash Equilibrium is inefficient
 - Parameter setting is very aggressive
 - Loss rate is high
 - Potential congestion collapse!

Results for the TCP Game – A Summary

	Droptail	RED
Tahoe	$\alpha_E \rightarrow \text{high}, \beta_E \rightarrow 1,$ Inefficient	$\alpha_E \rightarrow \text{high}, \beta_E \rightarrow 1,$ Inefficient
Reno	$\alpha_E = 1, \beta_E \rightarrow 1,$ Efficient, Unfair	$\alpha_E \rightarrow \text{high}, \beta_E \rightarrow 1,$ Inefficient
SACK	$\alpha_E \rightarrow \text{high}, \beta_E \rightarrow 1,$ Inefficient	$\alpha_E \rightarrow \text{high}, \beta_E \rightarrow 1,$ Inefficient

Discussion

Question: Does selfish congestion control endanger network efficiency?

Common Intuition: Yes, since flows would always gain from being more aggressive.

Our Answer: Not necessarily true!

- In the traditional setting (Reno end-points and droptail routers), network operates fine despite selfish behavior
- Selfish behavior very detrimental with modern loss recovery and queue management schemes

Outline

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Mechanisms for Nash Equilibrium

- We need mechanisms to explicitly counter aggressive behavior
- Has been a hot topic in the past
 - Fair Queuing discourages aggressive behavior
 - But needs per-flow state
 - RED-PD, AFD etc. explored lighter mechanisms
 - Aim to ensure *fair* bandwidth allocation

Our requirement is less stringent:

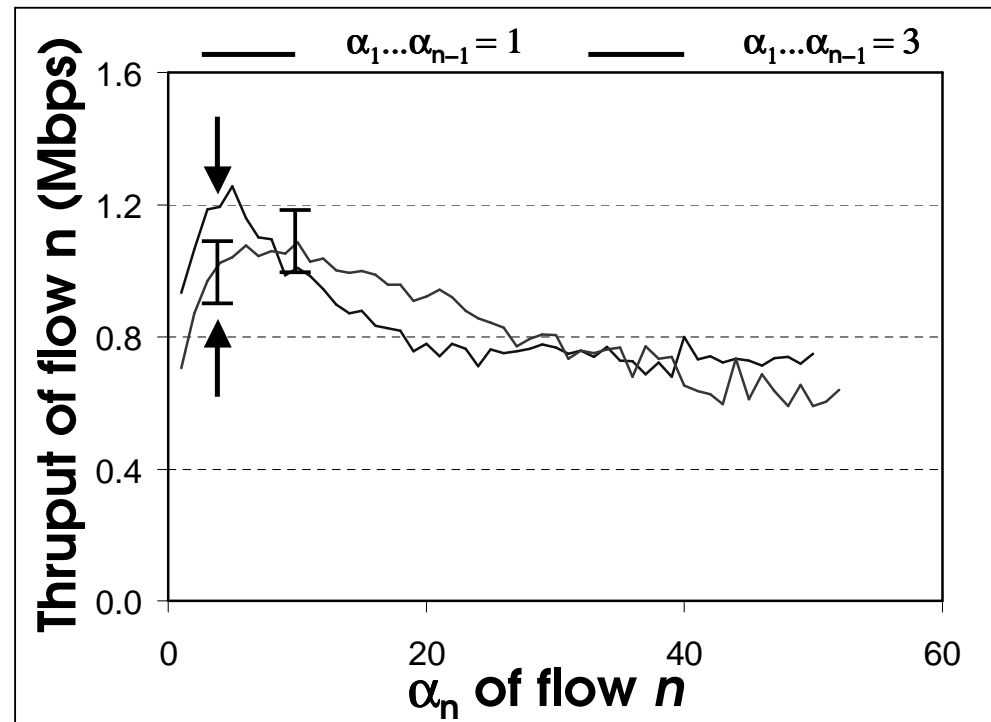
How much preferential dropping is needed to ensure a *reasonable* Nash Equilibrium?

CHOKe+ -- A Simple, Stateless Scheme

- A small modification to RED is enough

- CHOKe+

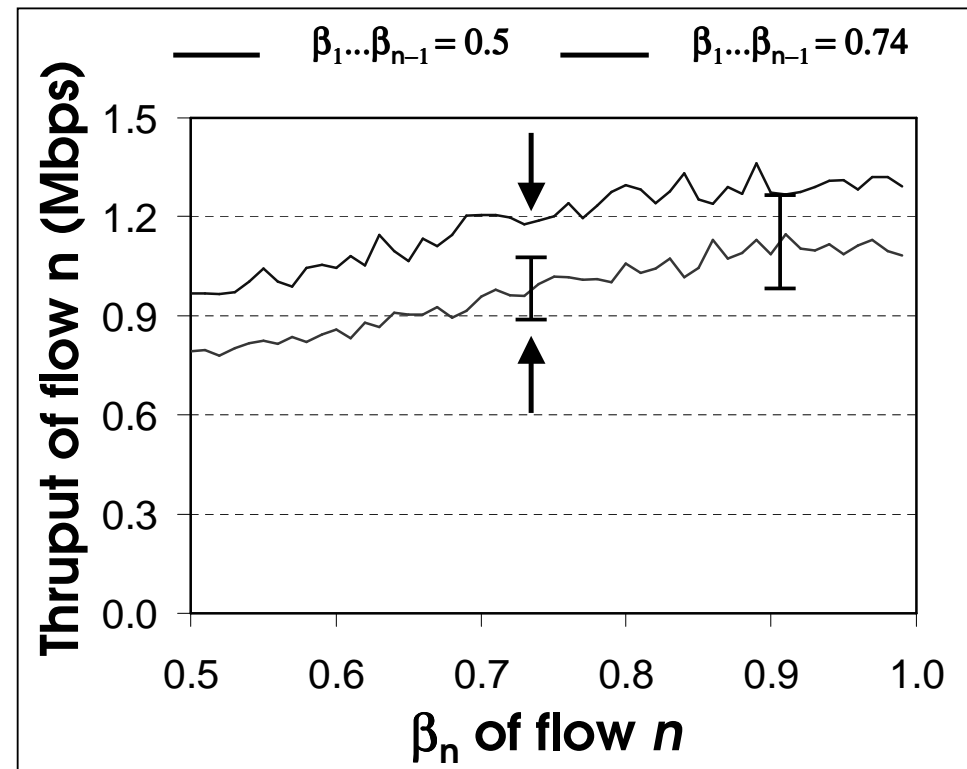
- Simple, stateless
- Provides just the right amount of punishment to aggressive flows
- Makes marginal advantage from greed insignificant
- E.g. SACK flows varying α



α_E	Goodput	Loss rate
3	97%	2.74%

CHOKe+ (Cont.)

- $\beta \rightarrow 1$ at Nash Equilibrium in all cases
 - $\beta < 1$ impossible to ensure without Fair Queuing
- But, CHOKe+ encourages $\beta < 1$
 - Makes aggressive β a risky choice
 - With SACK flows $\beta=0.74$ at Nash Equilibrium



β_E	Goodput	Loss rate
0.74	100%	2.42%

Summary

- Greedy congestion control may not always lead to inefficient operation
 - Traditional Reno host-droptail router setting
- Unfortunately, greedy behavior is bad in most other situations
- Fortunately, it is possible to ensure a desirable Nash Equilibrium via simple, stateless mechanisms

Back-up

- Back-up
 - Back-up
 - Back-up

CHOKe+

- CHOKe would have worked
 - But, enforces too high a drop rate
 - Underutilization at low levels of multiplexing
 - CHOKe+ fixes this problem

The CHOKe+ Algorithm

- For each incoming packet P
 - Pick k packets at random from queue
 - Let m be # packets from the same flow as P
 - Let $0 \leq \gamma_2 < \gamma_1 \leq 1$ be constants
 - If $m > \gamma_1 k$, P and the m packets are dropped
 - Else if $\gamma_2 k \leq m < \gamma_1 k$, drop P and the m packets only if RED were to drop P
 - Else just drop P according to RED

Why AIMD?

- Analysis is more generic than meets the eye
 - Conclusions hold for other congestion control schemes
 - Burstiness is a property of probing
- Widely employed

Why not Change Loss Recovery?

- Historical evaluation
- Very difficult to change
 - Sometimes need bilateral (protocol) support
 - Needs many implementation changes
 - Many design decisions were influenced by system requirements