

QoS CONTROL AND INTERNET TRAFFIC

QoS (quality of service) control

- feedback traffic control
 - closed-loop/adaptive control
 - small time scale: msec
 - end-to-end and/or with router support
- resource provisioning
 - open-loop control: allocate/reserve resources in advance
 - large time scale: seconds, minutes, and longer
 - router-based control with admission control at edge

Packet scheduling:

→ at routers/switches

→ default: FIFO with drop-tail packet drop

Granularity

- per-flow scheduling

→ e.g., single TCP session, UDP streaming session

- aggregate-flow scheduling

→ classes of traffic: e.g., platinum, gold, silver, bronze

Scheduling algorithms:

- priority scheduling
 - each flow or class assigned priority
 - simple but can lead to starvation
- weighted fair queueing
 - assign each flow/class a weight
 - generalizes round-robin
 - prevents starvation
 - requires traffic rate control to maintain service differentiation
- normalized weighted fair queueing
 - imposes strict ordering of service differentiation
- real-time scheduling
 - e.g., EDF (earliest deadline first)

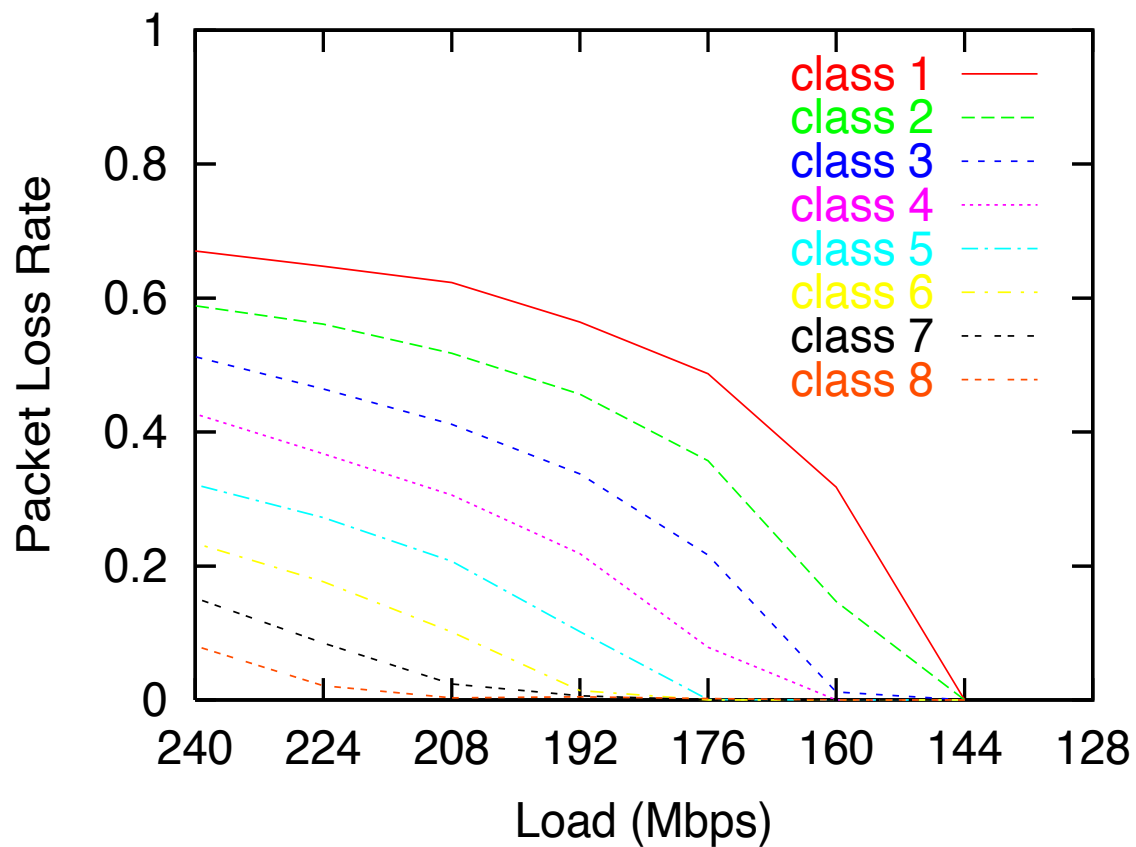
QoS control:

- per-user (or flow) reservation for super-quality service
→ guaranteed service
- shared service classes (platinum, gold, silver, bronze)
for good service
→ differentiated service

Internet standards:

- IETF IntServ
→ RSVP protocol
→ analogous to leasing a line
- IETF DiffServ
→ different types of router behavior
→ AF, EF, Cisco's LLQ for VoIP

Cisco 7206VXR router: packet loss rate

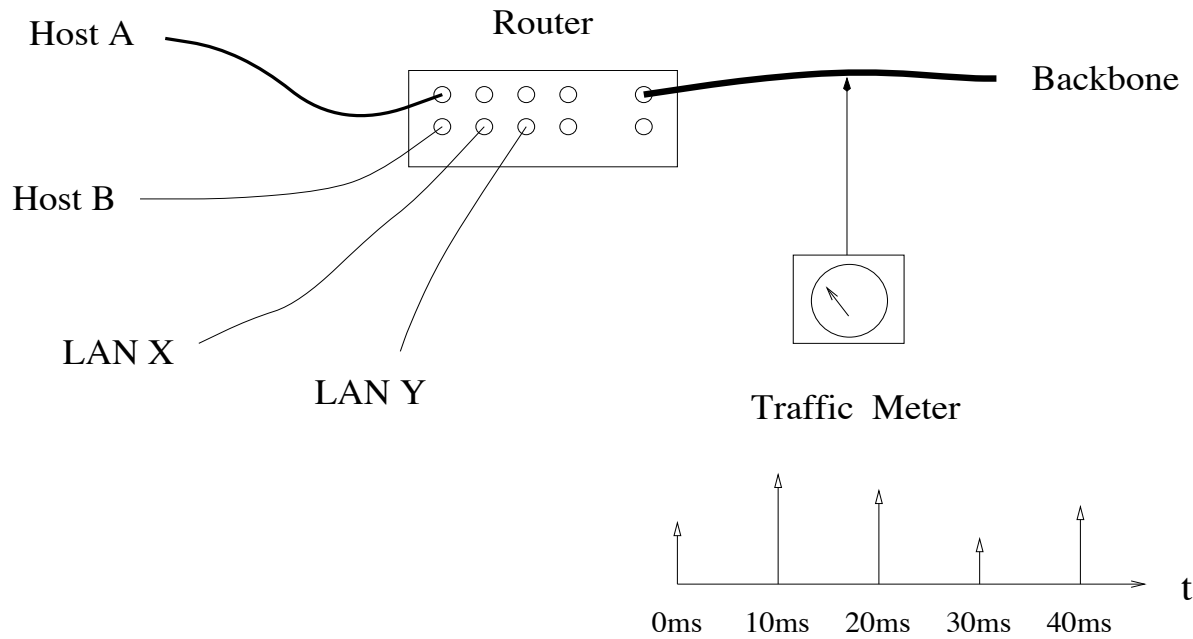


→ 8 classes

→ OC-3 link

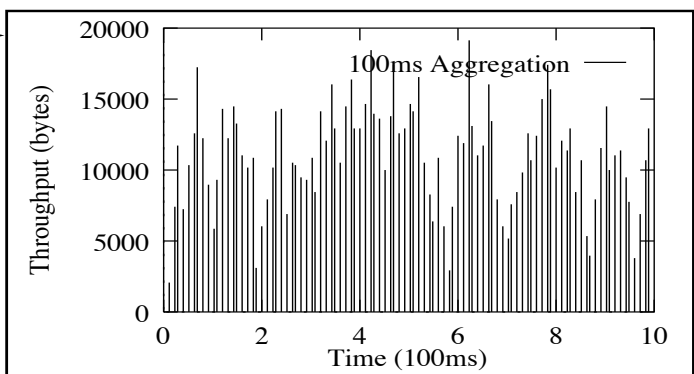
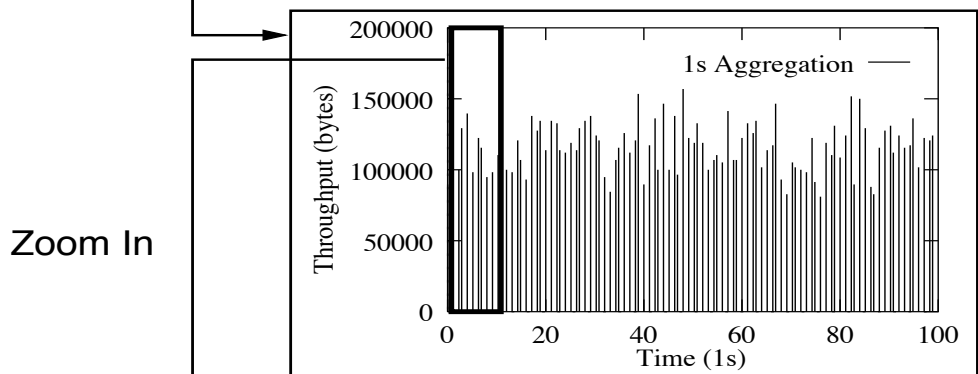
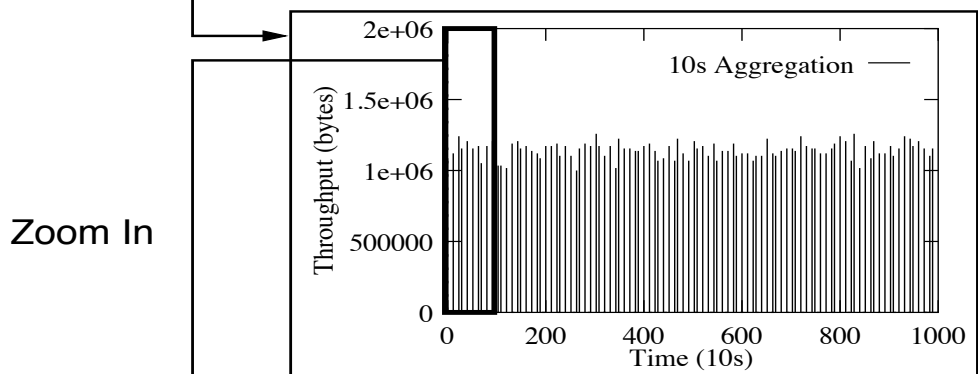
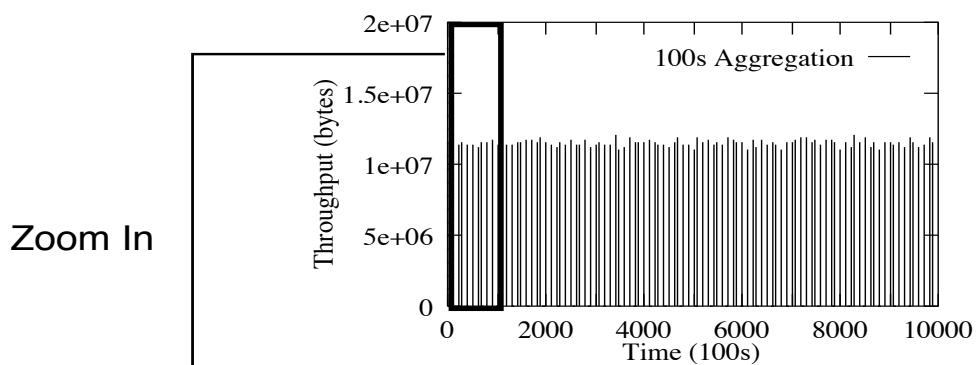
→ varying offered load

What does Internet traffic look like?



→ logging

→ traffic time series (at 10ms granularity)



Deaggregation

Aggregation

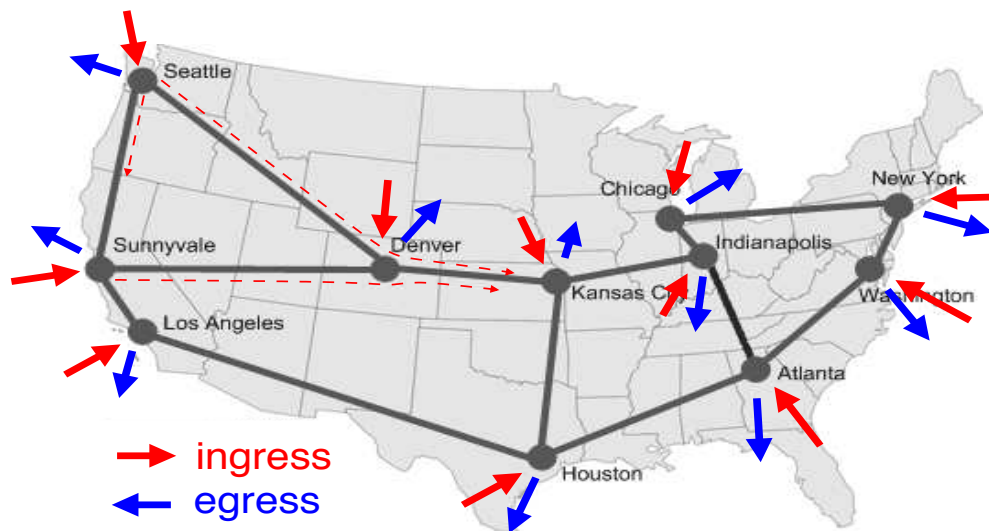
Aggregated traffic becomes flat

- “flat is good” rule for QoS provisioning
- bandwidth dimensioning
- technically: law of large numbers in action
- not correlated in time
- efficient and happy customers

Also aggregation over multiple users

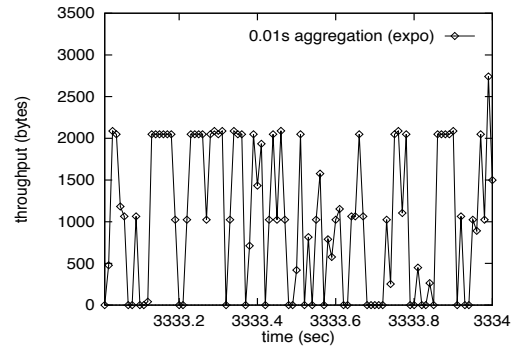
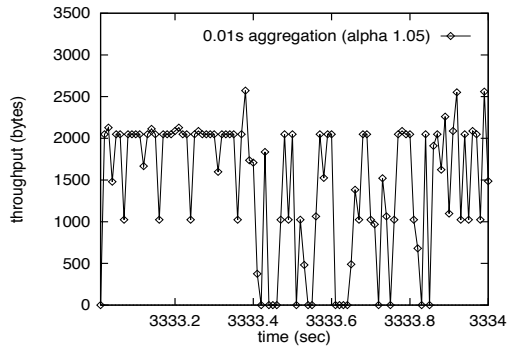
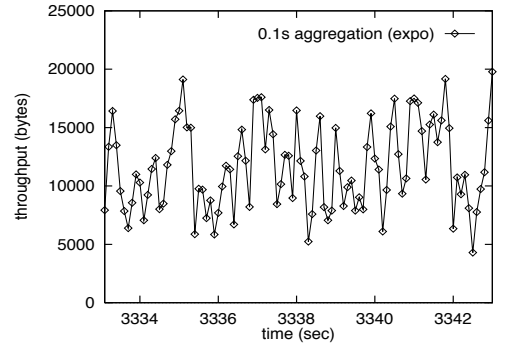
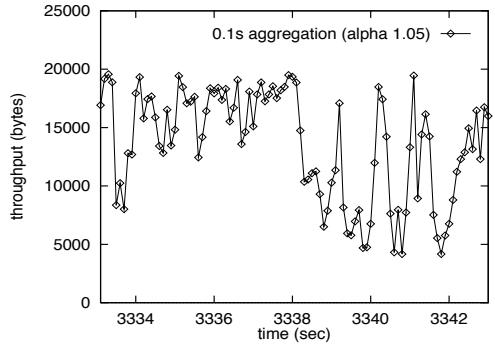
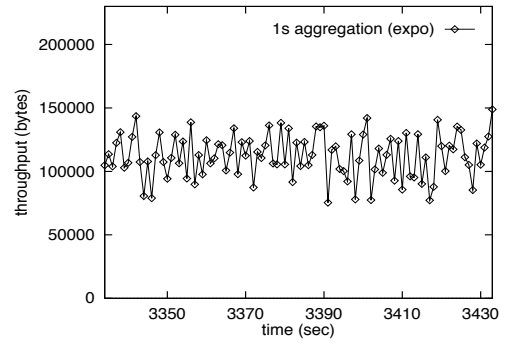
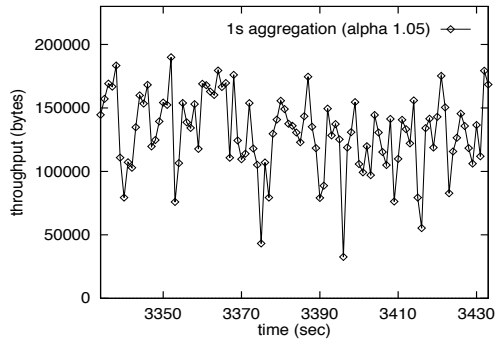
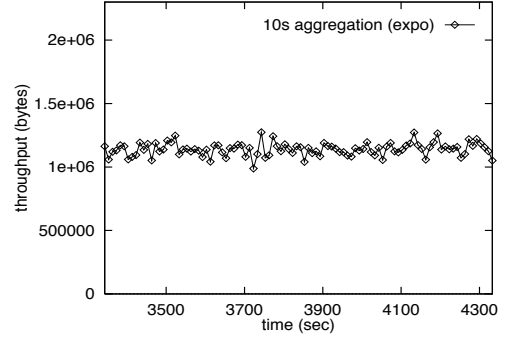
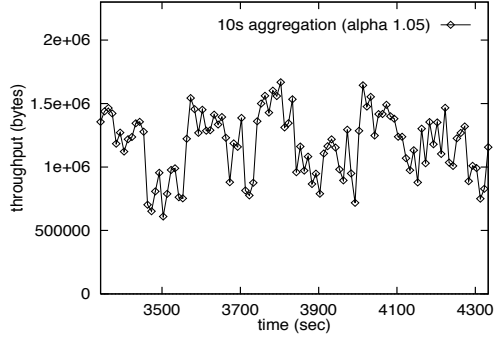
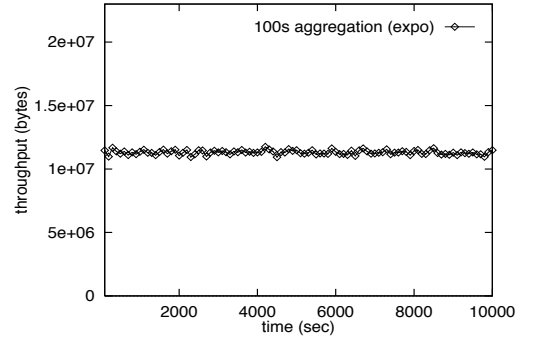
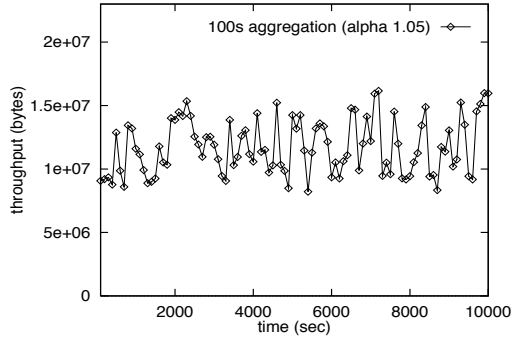
- called statistical multiplexing
- assuming independence between different users
- nice normal distribution shape
- allows non-peak reservation
- consider peak-to-mean ratio: less costly

- LLN: principal engineering tool used by large transit providers and large access providers
 - “largeness” is key
 - even though components are random, system is well-behaved and predictable
 - apply at ingress/egress and backbone links
 - measurement-based tool: traffic matrix

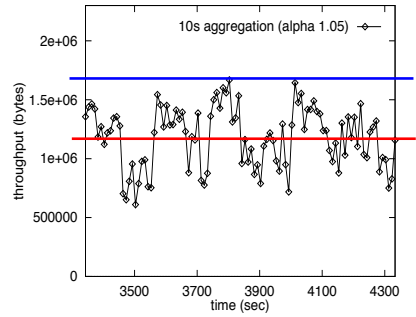
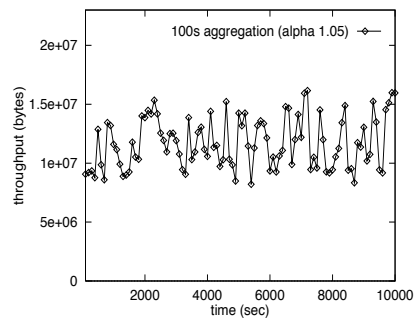


Back to Internet traffic:

- what does it look like?
- doesn't become flat with time aggregation
- stays bursty!
- in a peculiar fashion: self-similar or fractal

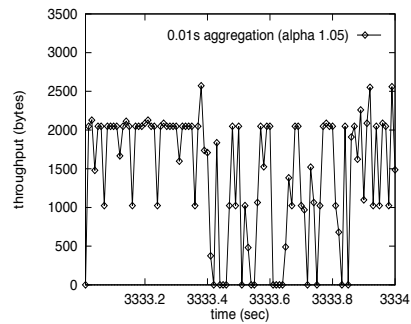
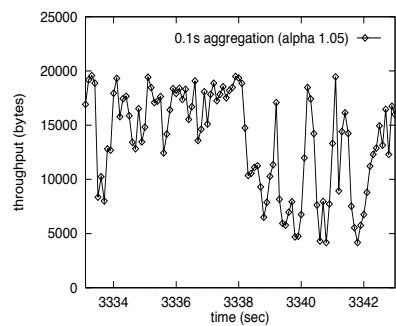
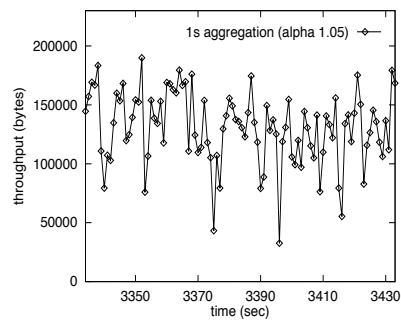


Internet: self-similar

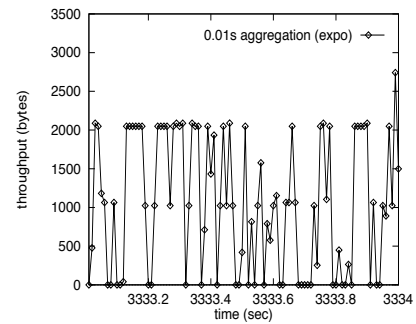
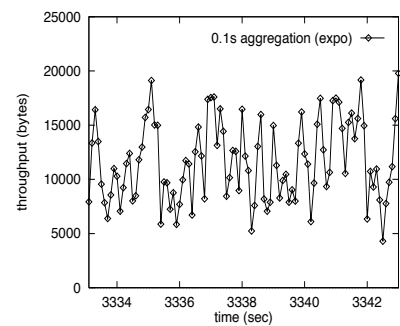
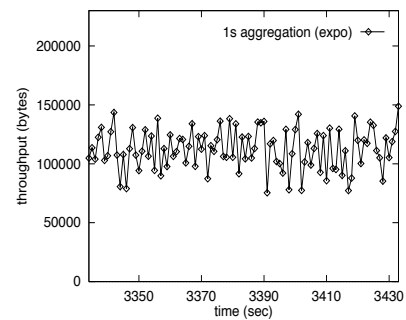
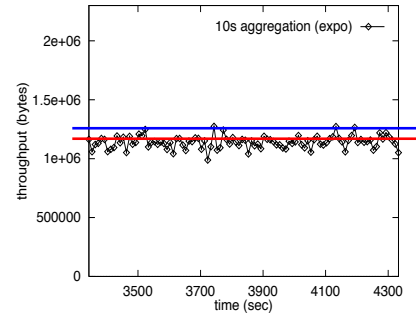
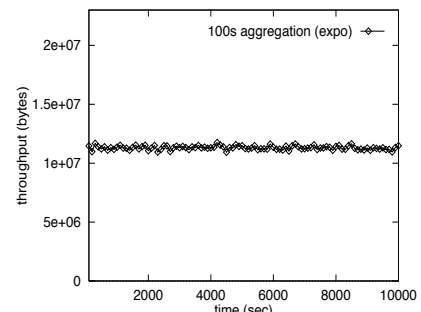


peak

avrg



Telephony: Poisson-like



Consequences:

- cannot use “flat is good” method anymore
- intrinsic trade-off between QoS and efficiency
- bad news for QoS provisioning
- traffic must be correlated in time (why?)

We know session arrivals are (approximately) Poisson; what about session lifetimes?

Important fact: TCP session lifetimes are heavy-tailed

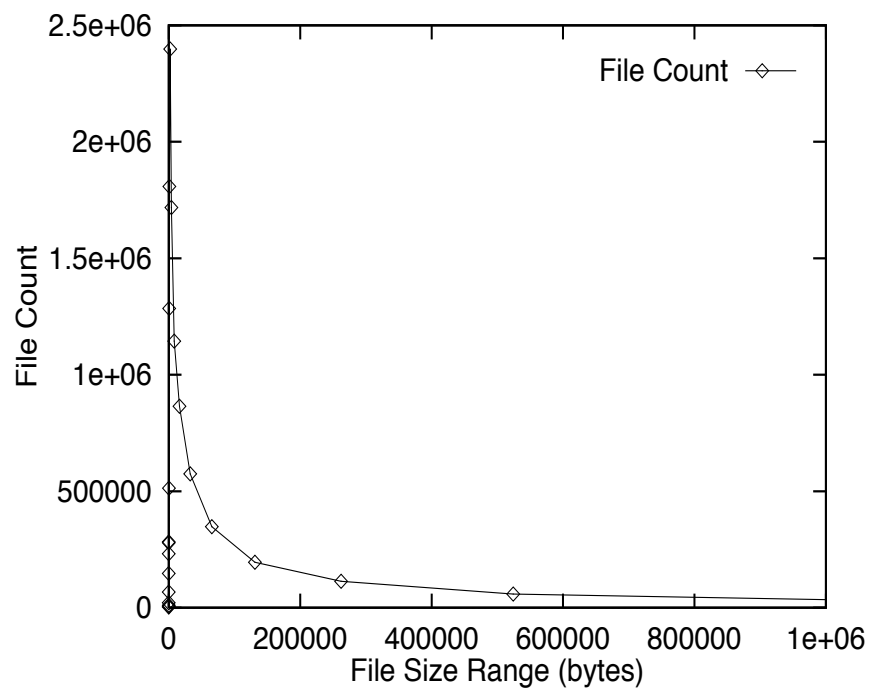
- $\Pr\{Z > x\} \approx x^{-\alpha}$
- as opposed to: $\Pr\{Z > x\} \approx e^{-bx}$
- exponent: $1 < \alpha < 2$ (closer to 1)
- note: different from Internet connectivity power-law
- much more likely session will last a long time
- has finite mean but infinite variance

Why would TCP session lifetimes be heavy-tailed?

- TCP traffic makes up bulk of Internet traffic

Tale of elephants and mice:

→ UNIX and WWW file systems



→ many small ones (mice)

→ a few very large ones (elephants)

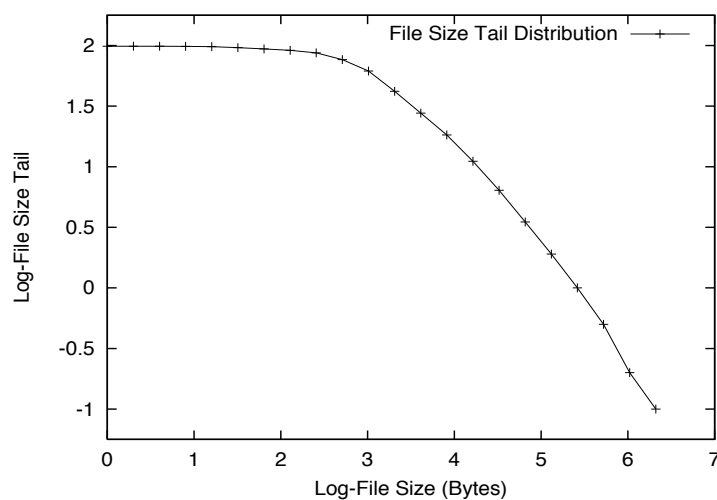
→ 10% consumes 90% of bandwidth

How to check if files sizes are heavy-tailed?

Since $\Pr\{Z > x\} \approx x^{-\alpha}$, take logarithm on both sides:

$$\longrightarrow \log \Pr\{Z > x\} \approx -\alpha \log x$$

$\longrightarrow$ linear function with negative slope $-\alpha$



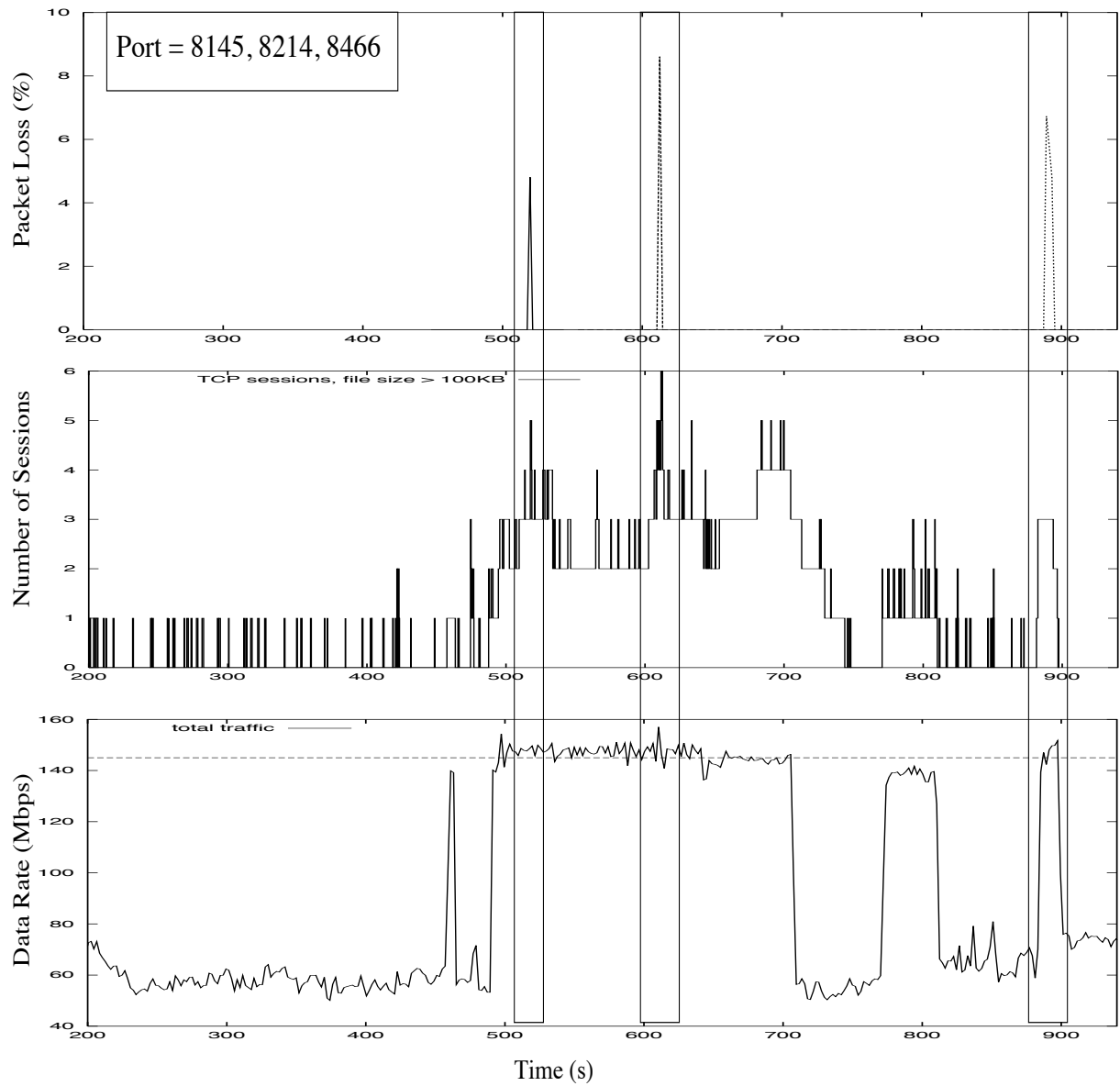
$\longrightarrow$ holds true for large x

$\longrightarrow$ what's the slope α ?

$\longrightarrow$ we don't care about details of small sizes (why?)

Elephants in action:

→ at backbone router



Can the problem be solved?

→ no: as long as elephants and mice holds

Turns out to be a wide-spread phenomenon is sociology, networks, and elsewhere

→ size (population) of cities

→ popularity

→ frequency of words in books

→ etc.

In the real world:

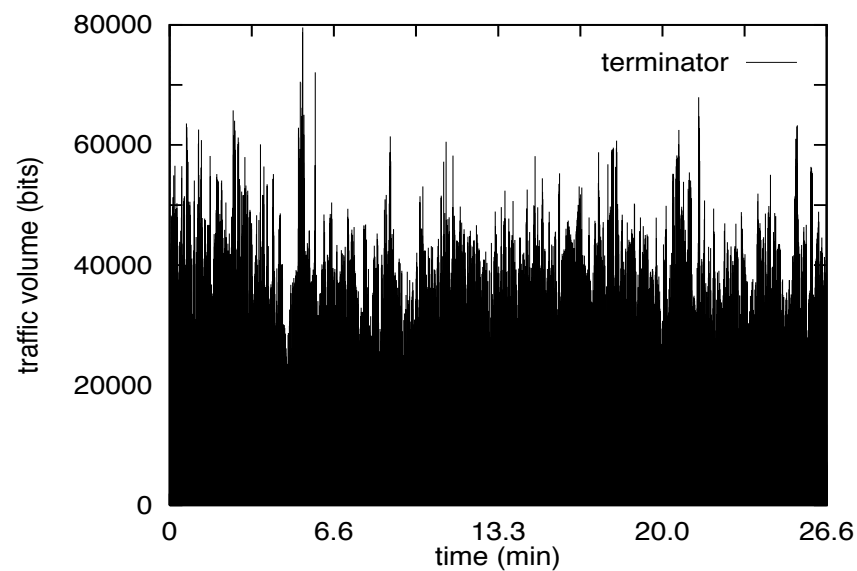
→ norm: skewed distribution of sizes

→ power-law

**Bandwidth allocation rule of thumb: 50%
overprovision**

Real-time VBR source profile

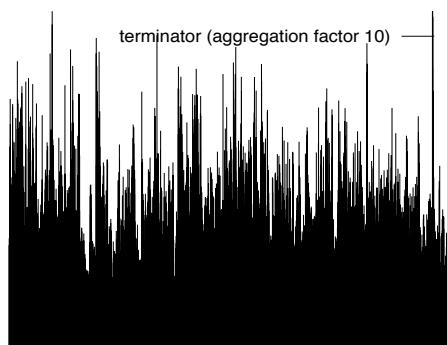
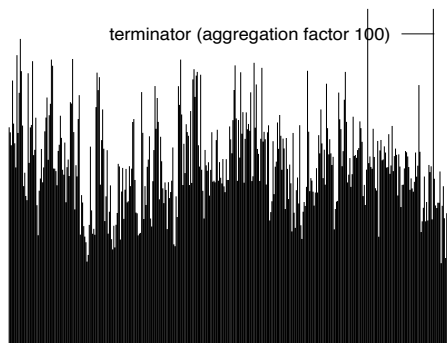
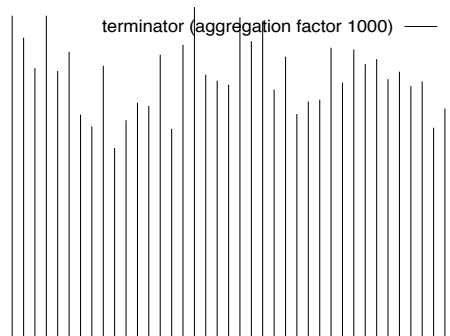
Consider traffic profile of compressed video (e.g., MPEG)



→ periodic real-time application

→ period $T = 1/f$ (f : frame rate)

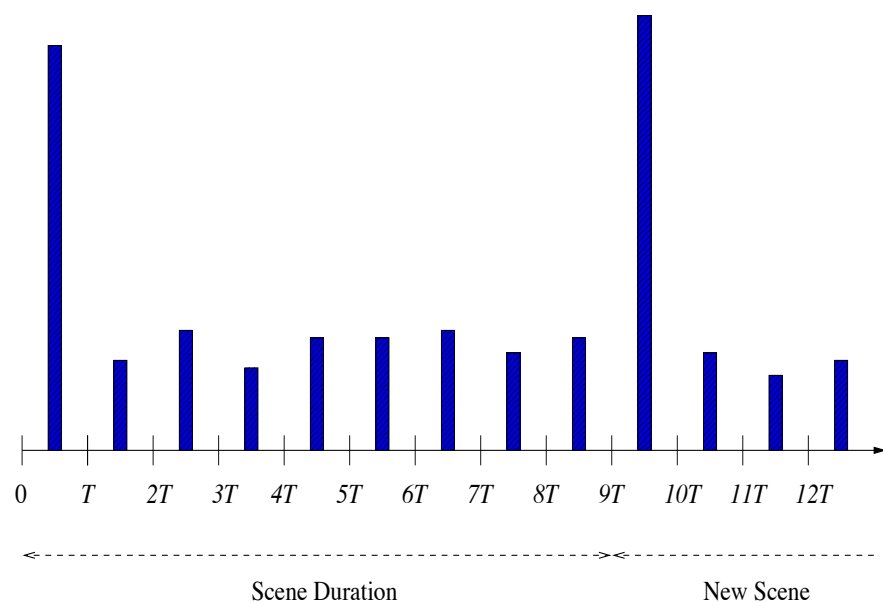
Burstiness structure:



→ burstiness persists across multiple time scales

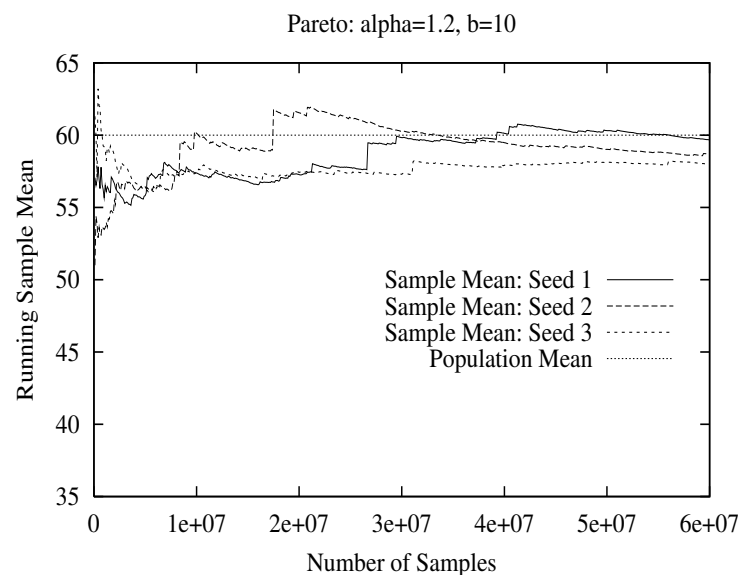
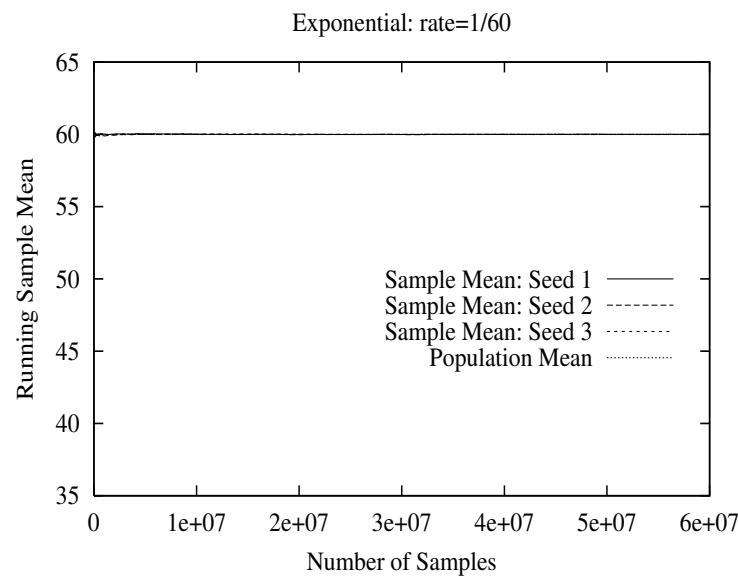
Possible causes:

- Heavy-tailed scene durations
 - facilitates inter-frame compression
- Repeating GOP pattern
 - e.g., I B B P B B



Note on sampling and convergence.

Sample mean convergence rate: exponential vs. Pareto



extremely slow convergence: concern when
generating artificial workload