

Recent Progress on a Statistical Network Calculus

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Dagstuhl, October 2002

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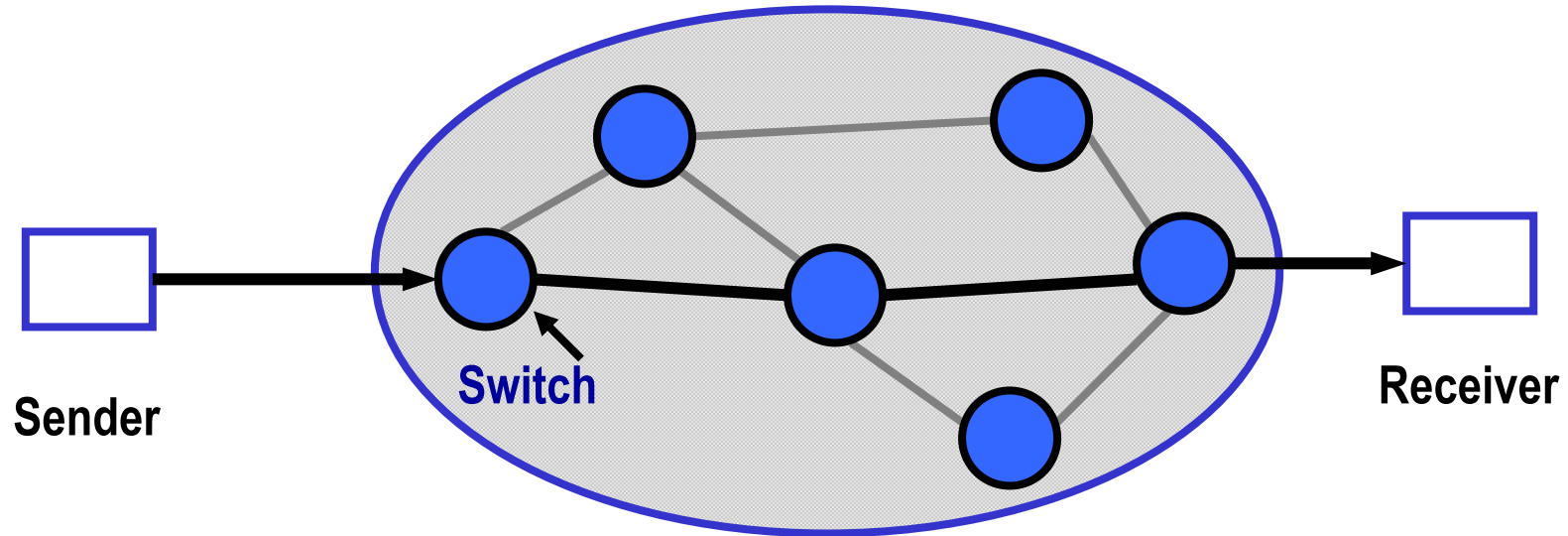
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Contents

- R. Boorstyn, A. Burchard, J. Liebeherr, C. Oottamakorn. "Statistical Service Assurances for Packet Scheduling Algorithms", IEEE Journal on Selected Areas in Communications. Special Issue on Internet QoS, Vol. 18, No. 12, pp. 2651-2664, December 2000.
- A. Burchard, J. Liebeherr, and S. D. Patek. "A Calculus for End-to-end Statistical Service Guarantees." (2nd revised version), Technical Report, May 2002.
- J. Liebeherr, A. Burchard, and S. D. Patek , "Statistical Per-Flow Service Bounds in a Network with Aggregate Provisioning", Infocom 2003.
- C. Li, A. Burchard, J. Liebeherr, "Calculus with Effective Bandwidth", July 2002.

Service Guarantees



- A **deterministic service** gives worst-case guarantees

$$\text{Delay} \leq d$$

- A **statistical service** provides probabilistic guarantees

$$\Pr[\text{Delay} \geq d] \leq \varepsilon \quad \text{or} \quad \Pr[\text{Loss} \geq l] \leq \varepsilon$$

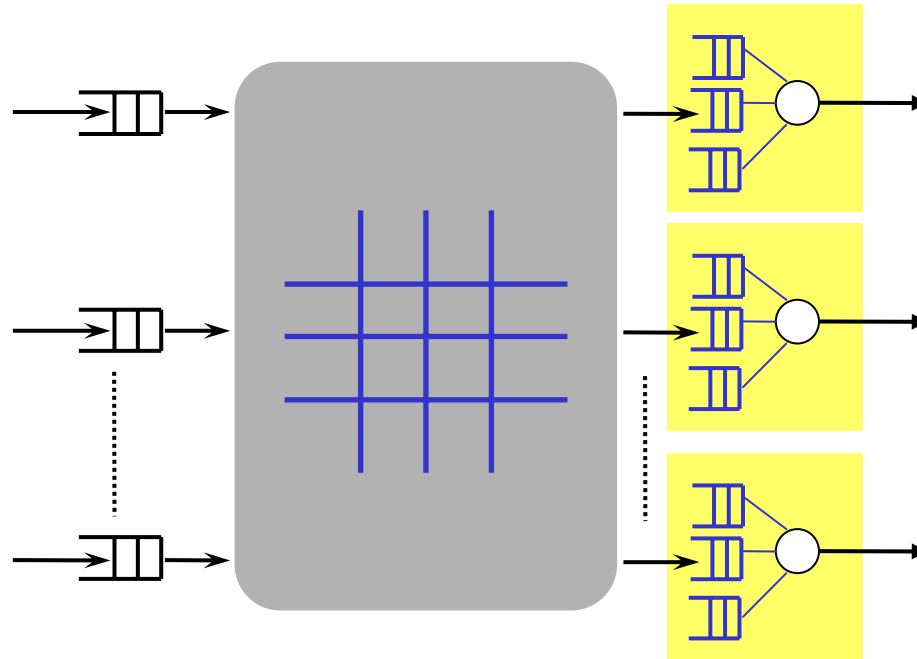
Multiplexing Gain

$$\left(\begin{array}{l} \text{Resource needed} \\ \text{to support QoS} \\ \text{for } N \text{ flows} \end{array} \right) \ll N \cdot \left(\begin{array}{l} \text{Resource needed} \\ \text{to support QoS} \\ \text{for 1 flow} \end{array} \right)$$

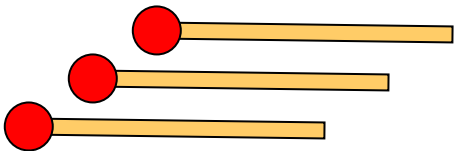
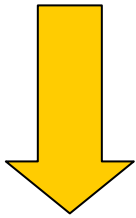
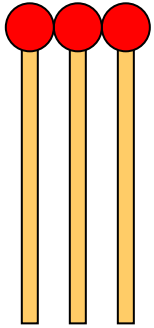
Sources of multiplexing gain:

- Traffic Conditioning (Policing, Shaping)
- Scheduling
- Statistical Multiplexing of Traffic

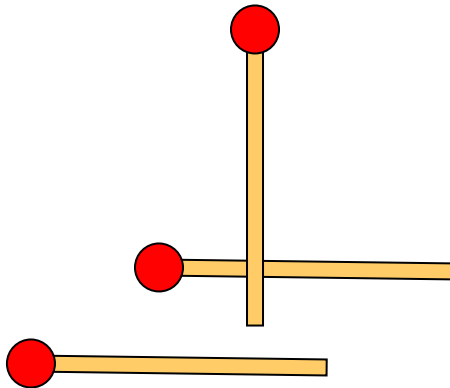
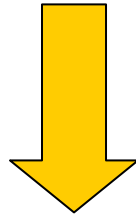
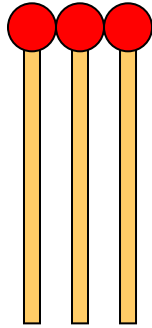
Scheduling



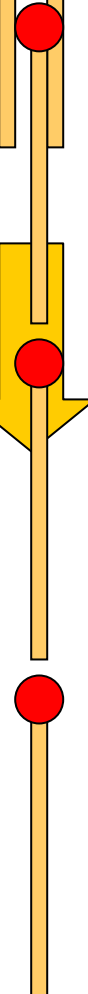
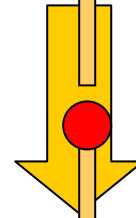
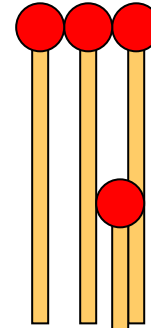
- Scheduling algorithm determines the order in which traffic is transmitted



Expected case

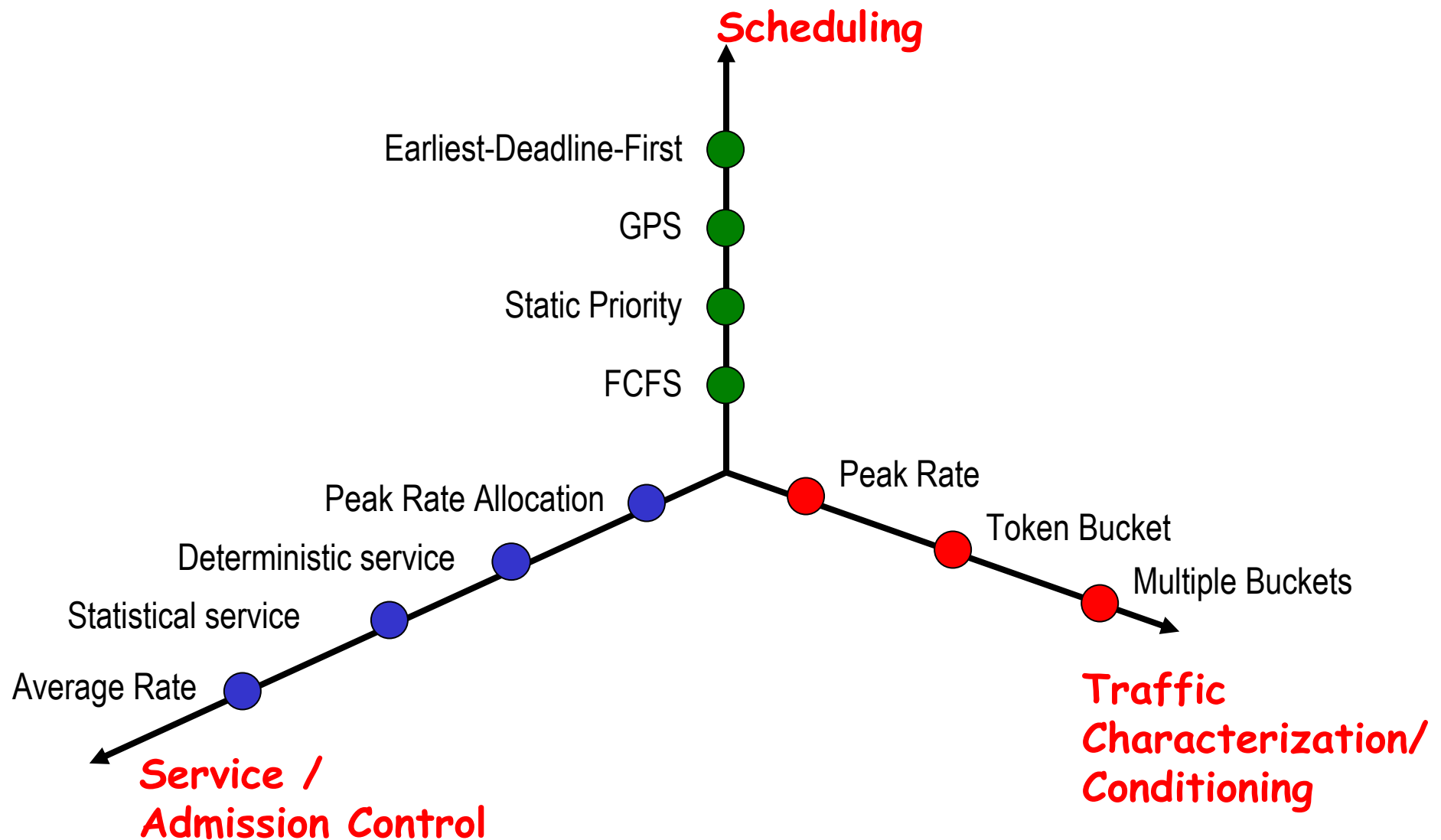


Probable worst-case

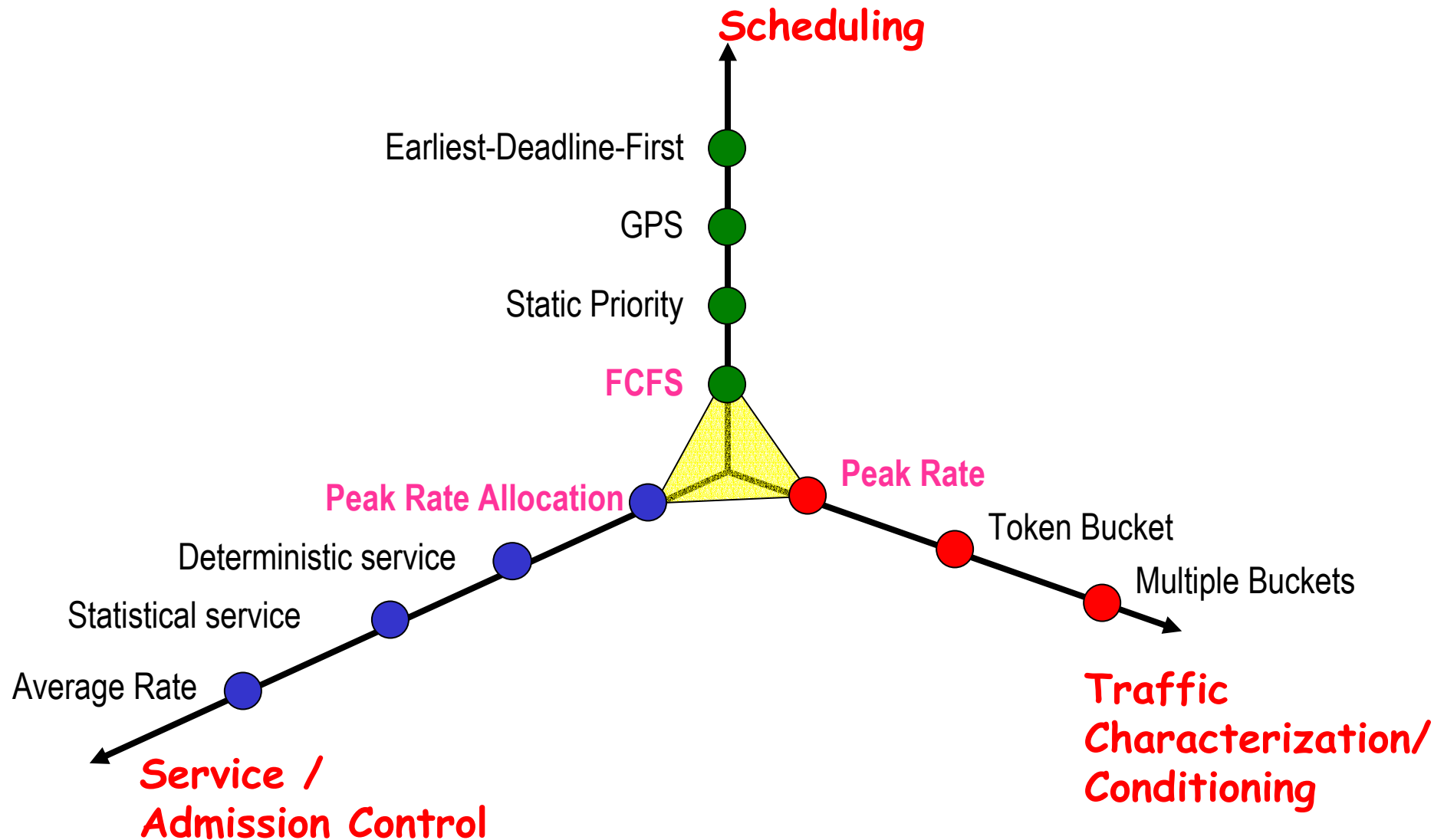


Deterministic worst-case

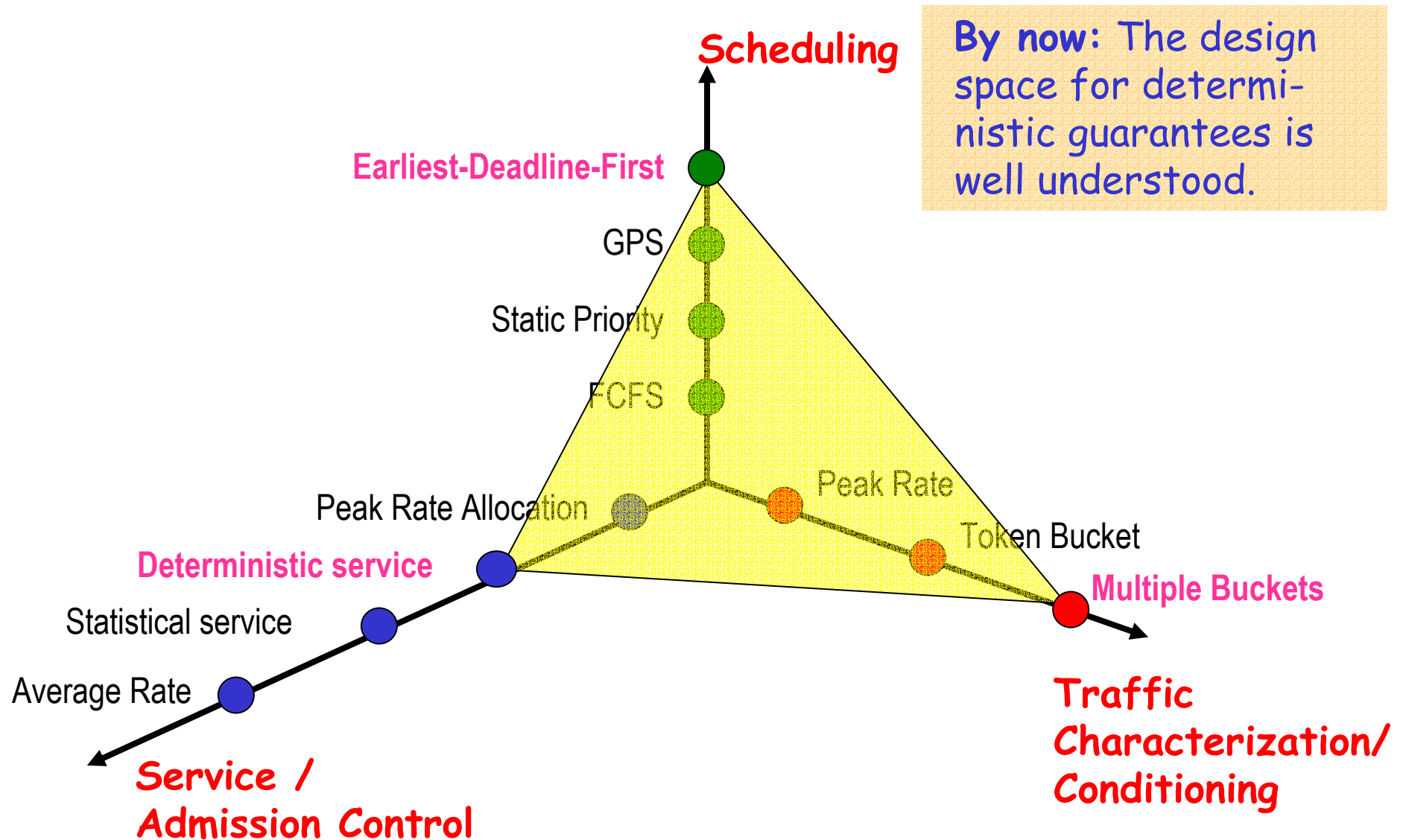
Designing Networks for Multiplexing Gain



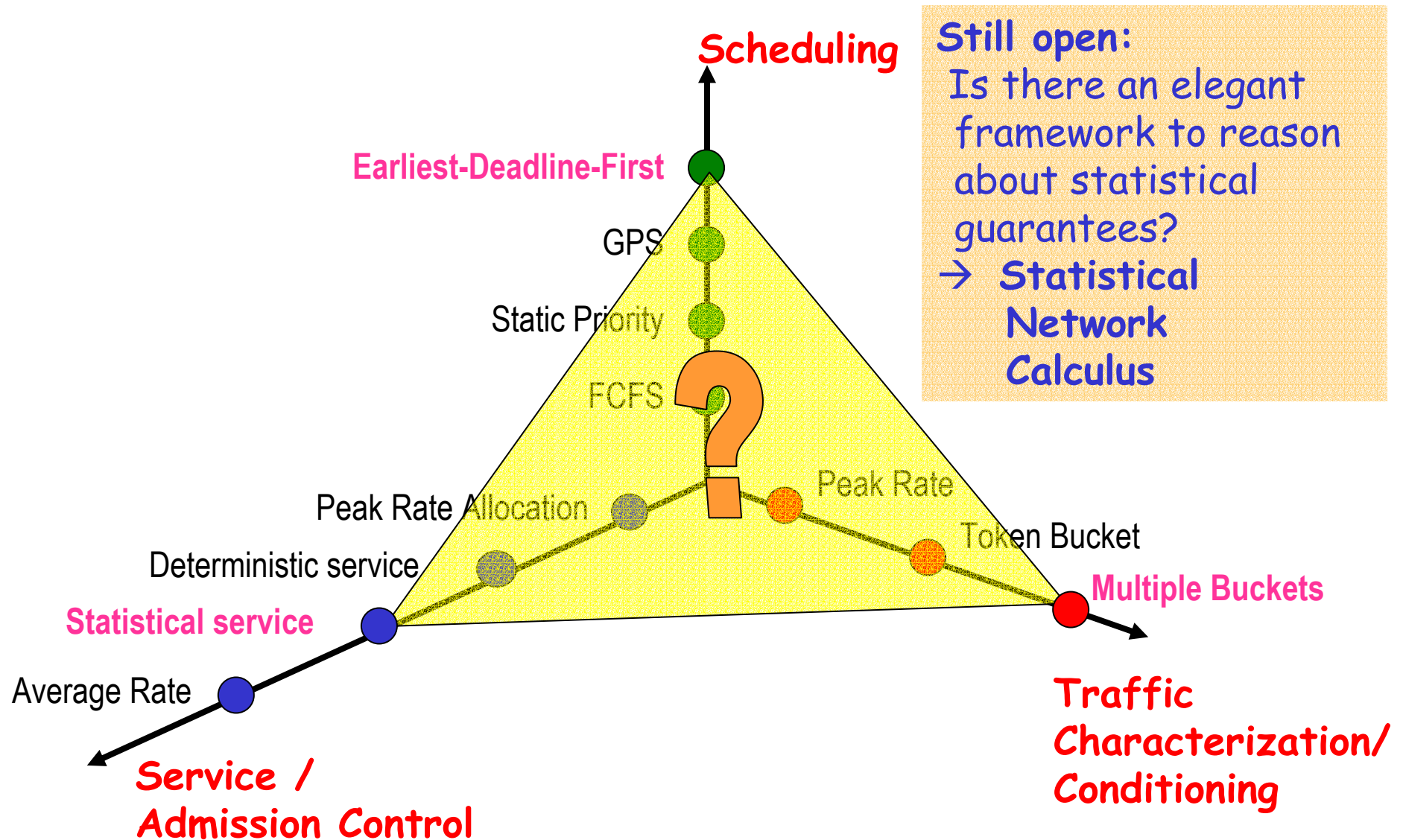
Designing Networks for Multiplexing Gain



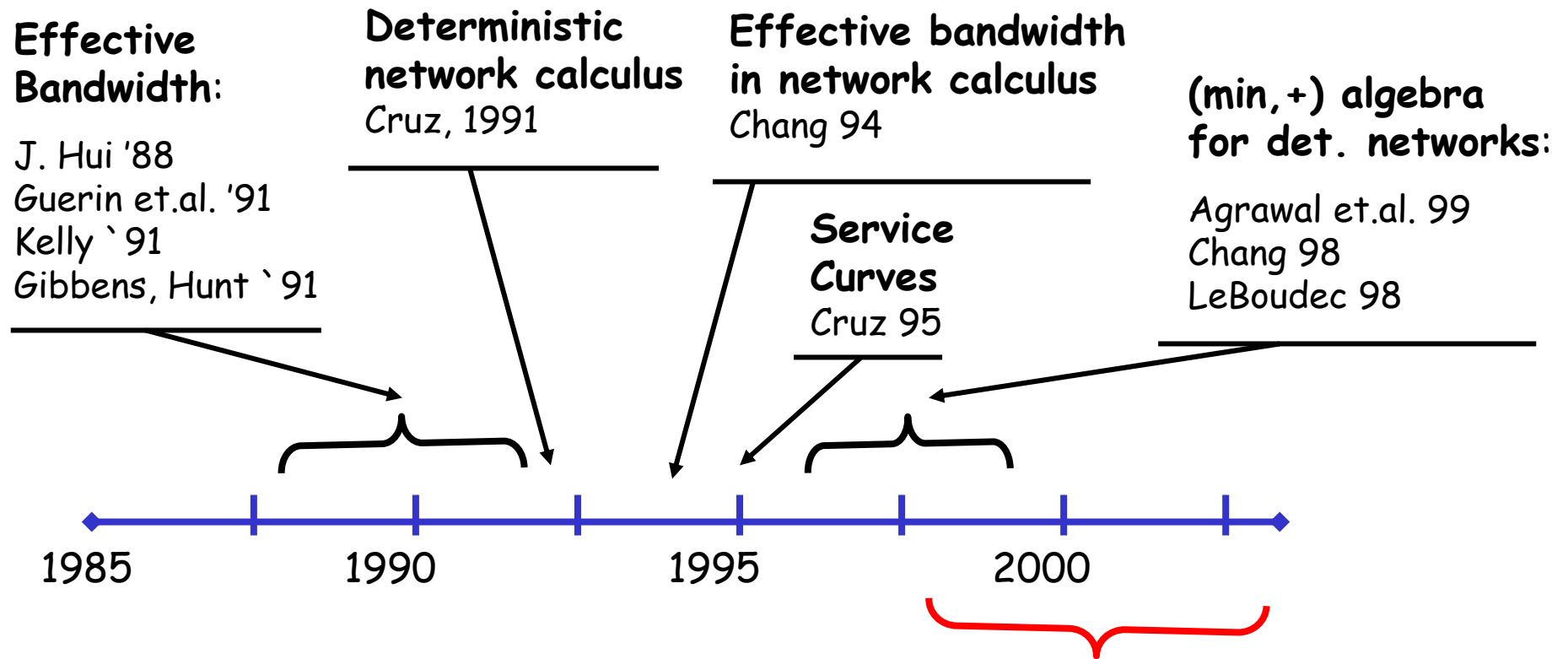
Designing Networks for Multiplexing Gain



Designing Networks for Multiplexing Gain



Related Work (small subset)



Motivation for our work on statistical network calculus:

- (1) Maintain elegance of deterministic calculus
- (2) Exploit know-how of statistical multiplexing

Source Assumptions

Arrivals $A_j(t, t+\tau)$ are random processes

Deterministic Calculus:

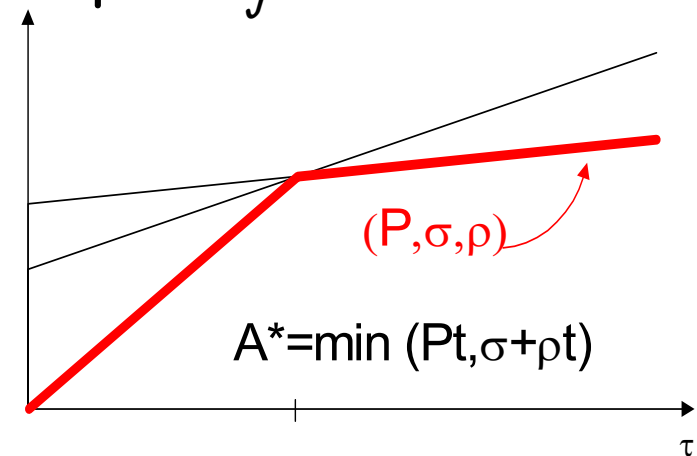
(A1) Additivity: For any $t_1 < t_2 < t_3$, we have:

$$A_j(t_1, t_2) + A_j(t_2, t_3) = A_j(t_1, t_3)$$

(A2) Subadditive Bounds: Traffic A_j is constrained by a subadditive deterministic envelope A_j^* as follows

$$A_j(t, t + \tau) \leq A_j^*(\tau) \quad , \forall t, \forall \tau$$

with $\rho = \lim_{\tau \rightarrow \infty} A_j^*(\tau) / \tau$



Source Assumptions

Statistical Calculus:

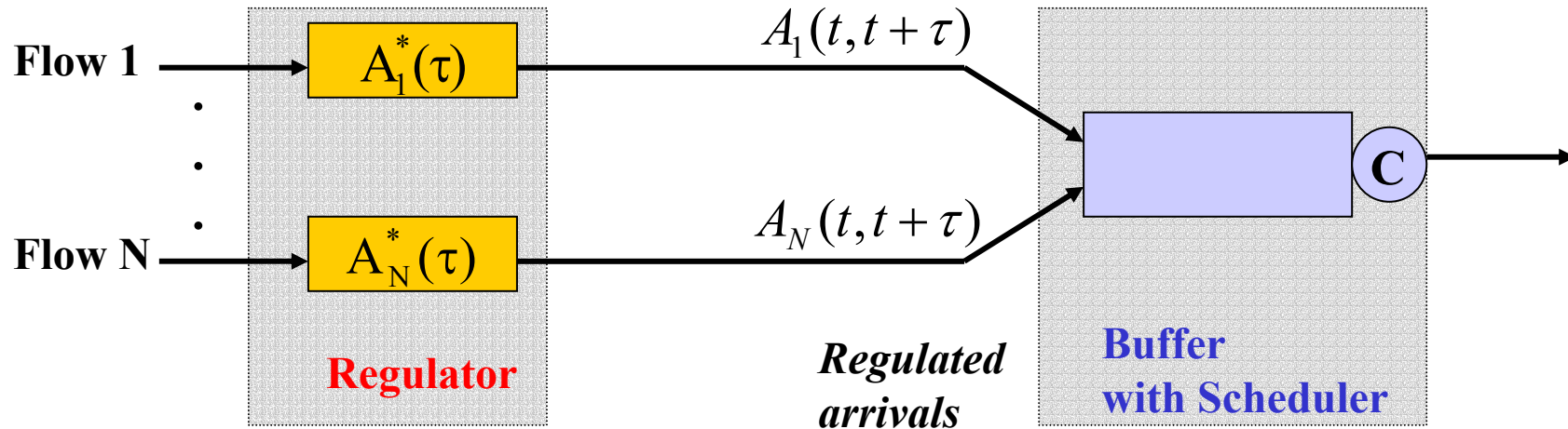
(A1) +(A2)

(A3) **Stationarity:** The A_j are stationary random variables

(A4) **Independence:** The A_i and A_j ($i \neq j$) are stochastically independent

(No assumptions on arrival distribution!)

Aggregating Arrivals



Arrivals from multiple flows:
$$A_C = \sum_j A_j$$

Deterministic Calculus:

Worst-case of multiple flows is sum of the worst-case of each flow

$$A_C(t, t + \tau) \leq \sum_j A_j^*(\tau)$$

Aggregating Arrivals

Statistical Calculus:

To bound aggregate arrivals we define a function that is a bound on the sum of multiple flows with high probability → "Effective Envelope"

- Effective envelopes are non-random functions

- effective envelope $\mathcal{G}_C^\varepsilon$:

$$Pr\{A_C(t, t + \tau) \leq \mathcal{G}_C^\varepsilon(\tau)\} \geq 1 - \varepsilon \quad \forall t, \tau$$

- strong effective envelope $\mathcal{H}_C^{l, \varepsilon}$:

$$Pr\{\forall [t, t + \tau] \subseteq I_\ell : A_C(\tau) \leq \mathcal{H}_C^{l, \varepsilon}(\tau)\} \geq 1 - \varepsilon \quad \forall I_\ell$$

Obtaining Effective Envelopes

$$\mathcal{G}_C^\varepsilon(t) = \inf_{s>0} \frac{1}{s} \left(\sum_{j \in C} \log \bar{M}_j(s, t) - \log \varepsilon \right)$$

$$\text{with } \bar{M}_j(s, t) = 1 + \frac{\rho_j t}{A_j^*(t)} (e^{s A_j^*(t)} - 1)$$

$$\mathcal{H}_C^{\ell, \varepsilon'}(t) \leq \mathcal{G}_C^\varepsilon(\gamma t + a), \quad 0 \leq t \leq \ell$$

$$\begin{aligned} \text{with } \quad \varepsilon' &\leq \varepsilon \cdot \frac{\ell \sqrt{\gamma} + 1}{a \sqrt{\gamma} - 1} \\ a &\in (0, \ell) \\ \gamma &> 1 \end{aligned}$$

Effective vs. Deterministic Envelopes

$$A^* = \min(Pt, \sigma + \rho t)$$

Type 1 flows:

$$P = 1.5 \text{ Mbps}$$

$$\rho = .15 \text{ Mbps}$$

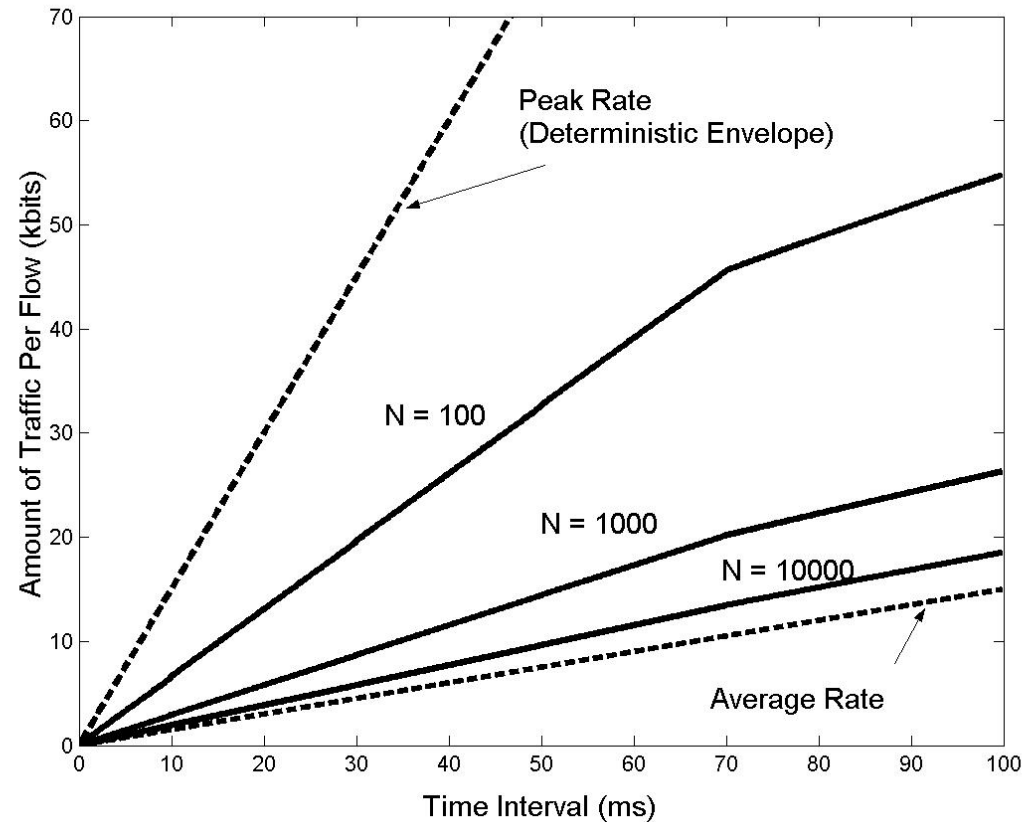
$$\sigma = 95400 \text{ bits}$$

Type 2 flows:

$$P = 6 \text{ Mbps}$$

$$\rho = .15 \text{ Mbps}$$

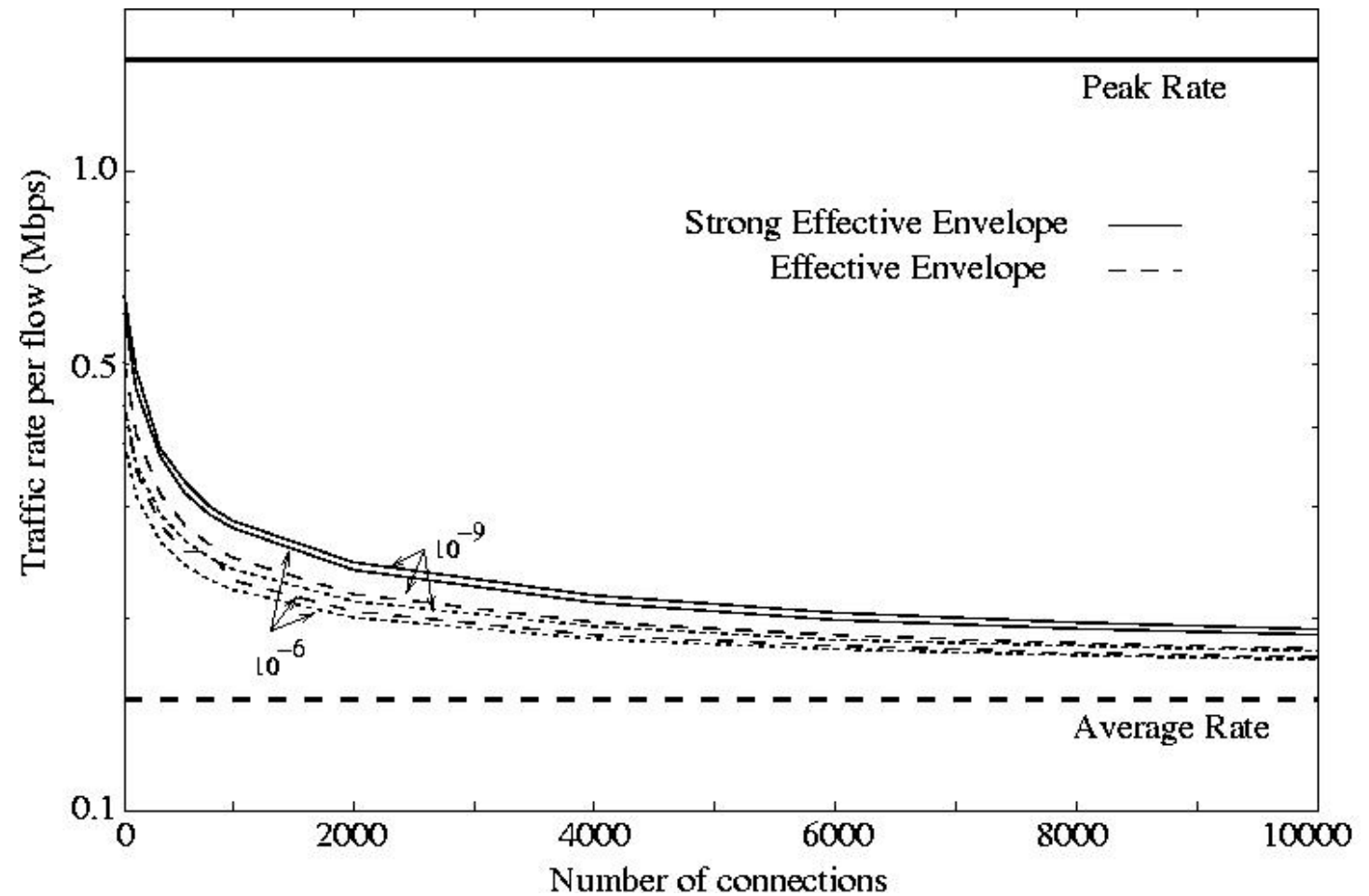
$$\sigma = 10345 \text{ bits}$$



Type 1 flows

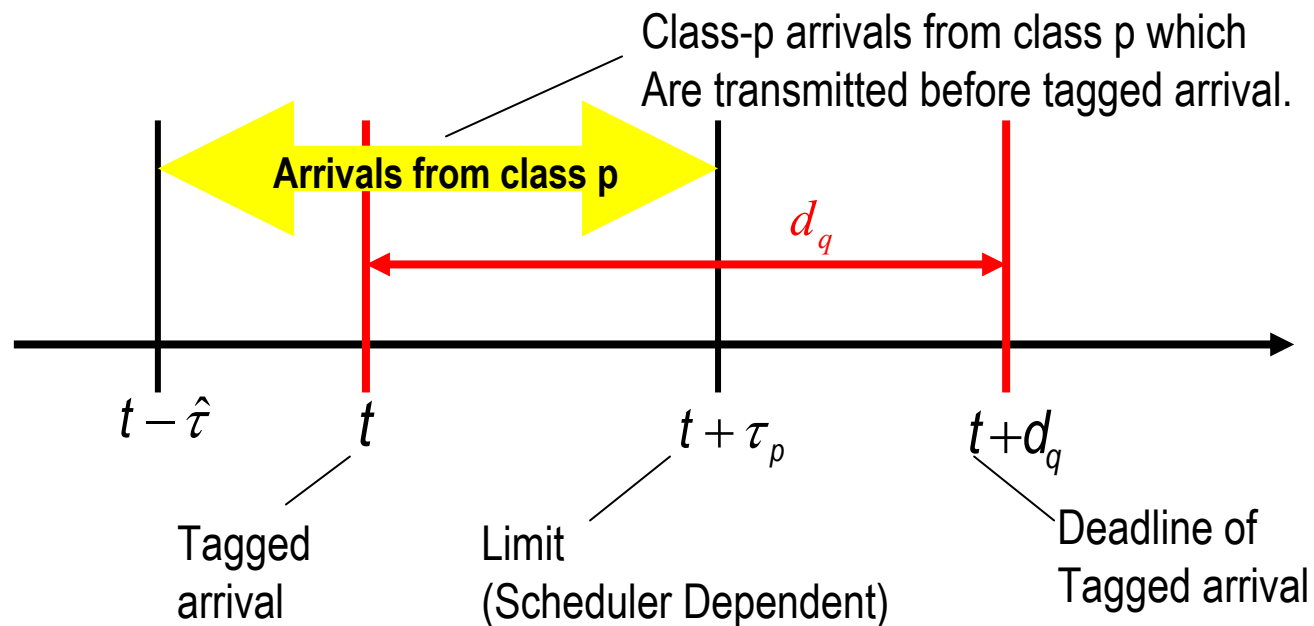
Effective vs. Deterministic Envelopes

Traffic rate at $t = 50$ ms
Type 1 flows



Scheduling Algorithms

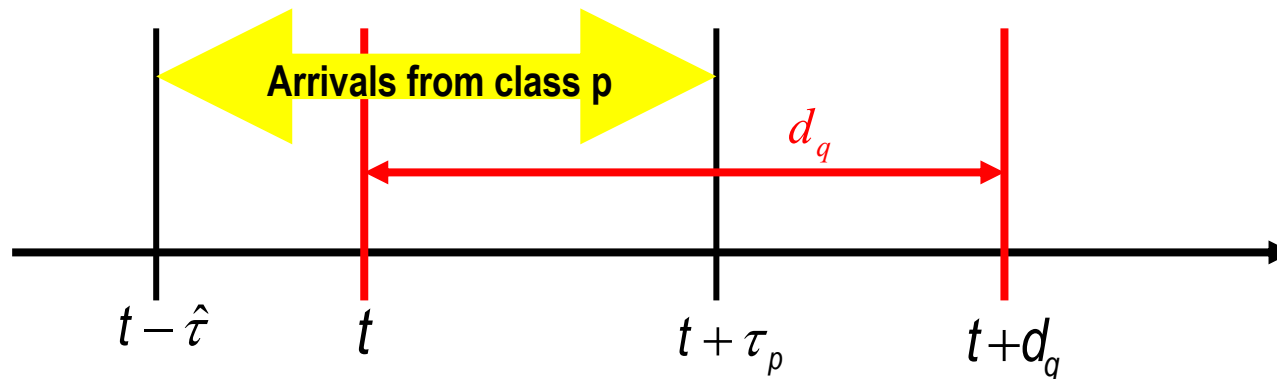
- Consider a work-conserving scheduler with rate R
- Consider class- q arrival at t with $t+d_q$:



- Tagged arrival has no delay bound violation if

$$\sup_{\hat{\tau}} \left\{ \sum_p A_{C_p}(t - \hat{\tau}, t + \tau_p) - R(\hat{\tau} + d_q) \right\} \leq 0$$

Scheduling Algorithms



$$\sup_{\hat{\tau}} \left\{ \sum_p A_{C_p}(t - \hat{\tau}, t + \tau_p) - (\hat{\tau} + d_q) \right\} \leq 0$$

with

FIFO: $\tau_p = 0.$

SP: $\tau_p = -\hat{\tau} \ (p > q)$, $0 \ (p = q)$, $d_q \ (p < q).$

EDF: $\tau_p = \max\{-\hat{\tau}, d_q - d_p\}.$

Admission Control for Scheduling Algorithms

with Deterministic Envelopes:

$$\sup_{\hat{\tau}} \left\{ \sum_p A_j^* (\tau_p + \hat{\tau}) - \hat{\tau} \right\} \leq d_q$$

with Effective Envelopes:

$$\sup_{\hat{\tau}} \left\{ \sum_p \mathbf{G}_{C_p}^{\varepsilon/Q} (\tau_p + \hat{\tau}) - \hat{\tau} \right\} \leq d_q$$

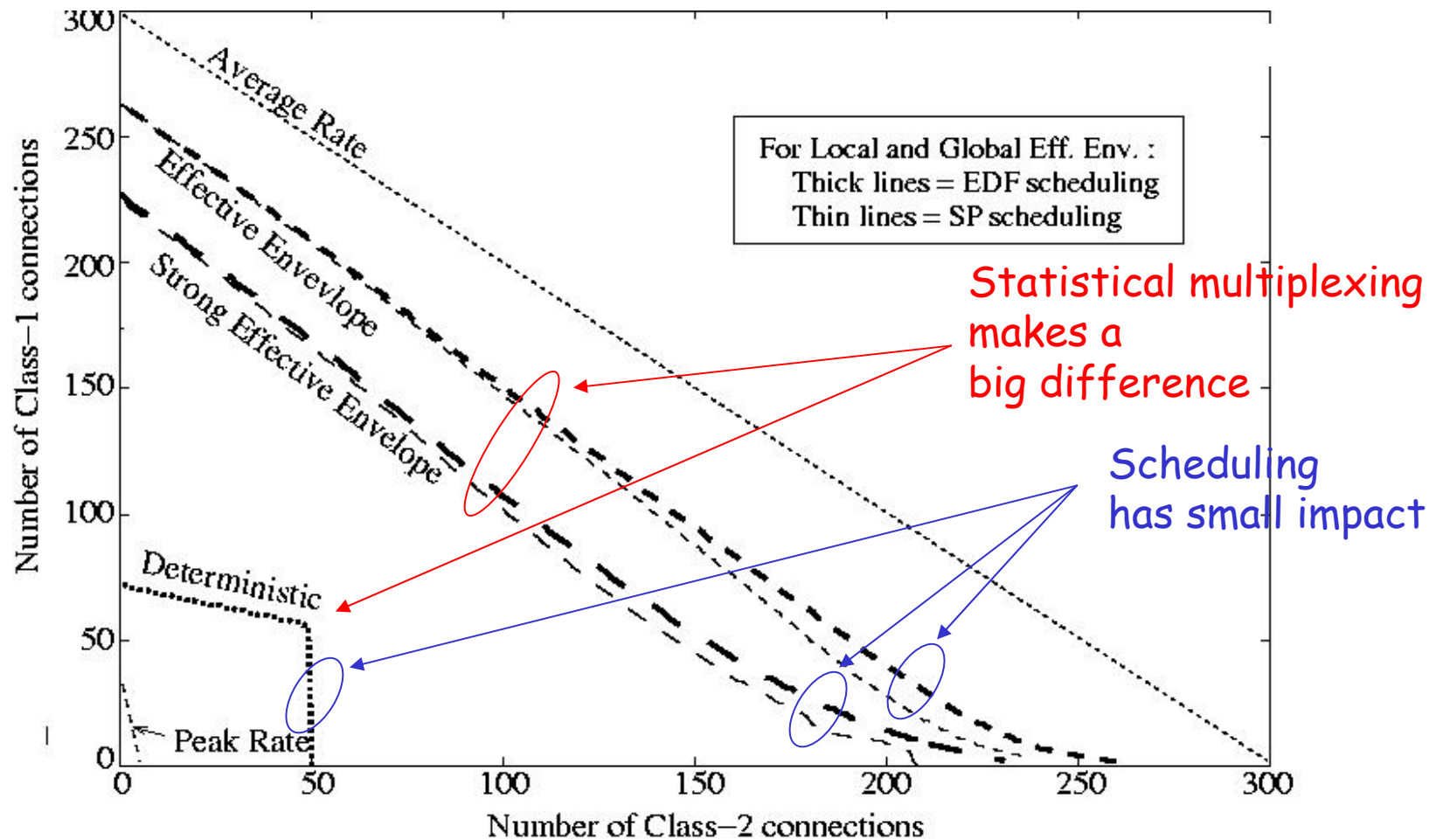
with Strong Effective Envelopes:

$$\sup_{\hat{\tau}} \left\{ \sum_p \mathbf{H}_{C_p}^{l,\varepsilon/Q} (\tau_p + \hat{\tau}) - \hat{\tau} \right\} \leq d_q$$

Effective vs. Deterministic Envelope

$C = 45$ Mbps, $\varepsilon = 10^{-6}$

Delay bounds: Type 1: $d_1 = 100$ ms, Type 2: $d_2 = 10$ ms,



Effective Envelopes and Effective Bandwidth

Effective Bandwidth (Kelly, Chang)

$$\alpha(s, \tau) = \sup_{t \geq 0} \left\{ \frac{1}{s\tau} \log E[e^{s(A[t+\tau] - A[t])}] \right\}$$

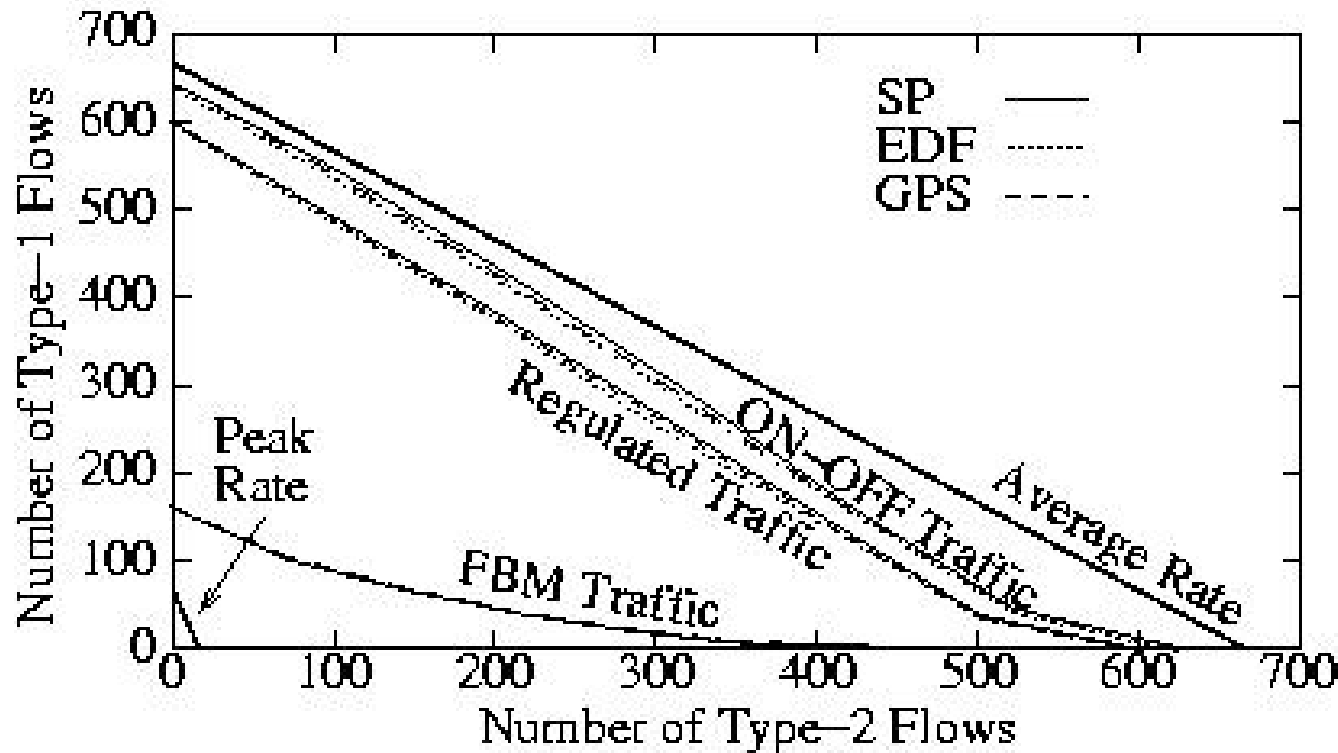
$$s, \tau \in (0, \infty)$$

Given $\alpha(s, \tau)$, an effective envelope is given by

$$\mathcal{G}^\varepsilon(\tau) = \inf_{s > 0} \left\{ \tau \alpha(s, \tau) - \frac{\log \varepsilon}{s} \right\}$$

Effective Envelopes and Effective Bandwidth

Now, we can calculate statistical service guarantees for schedulers and traffic types



Schedulers:

SP - Static Priority

EDF - Earliest
Deadline First

GPS - Generalized
Processor Sharing

Traffic:

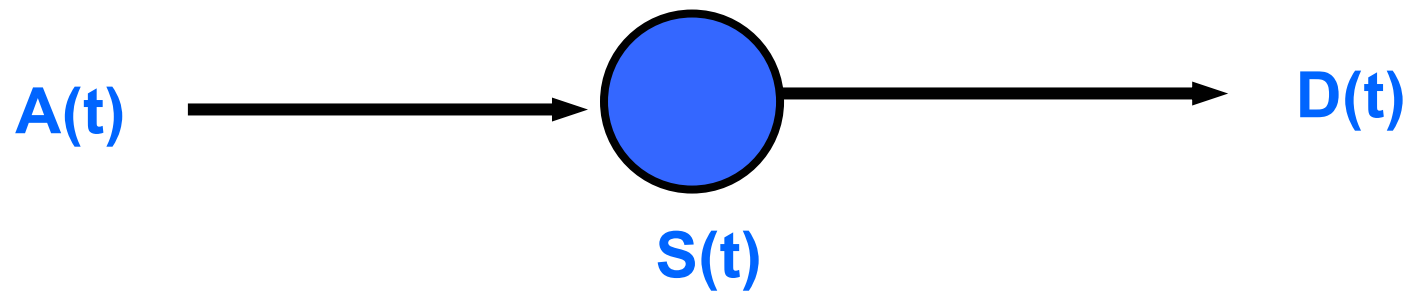
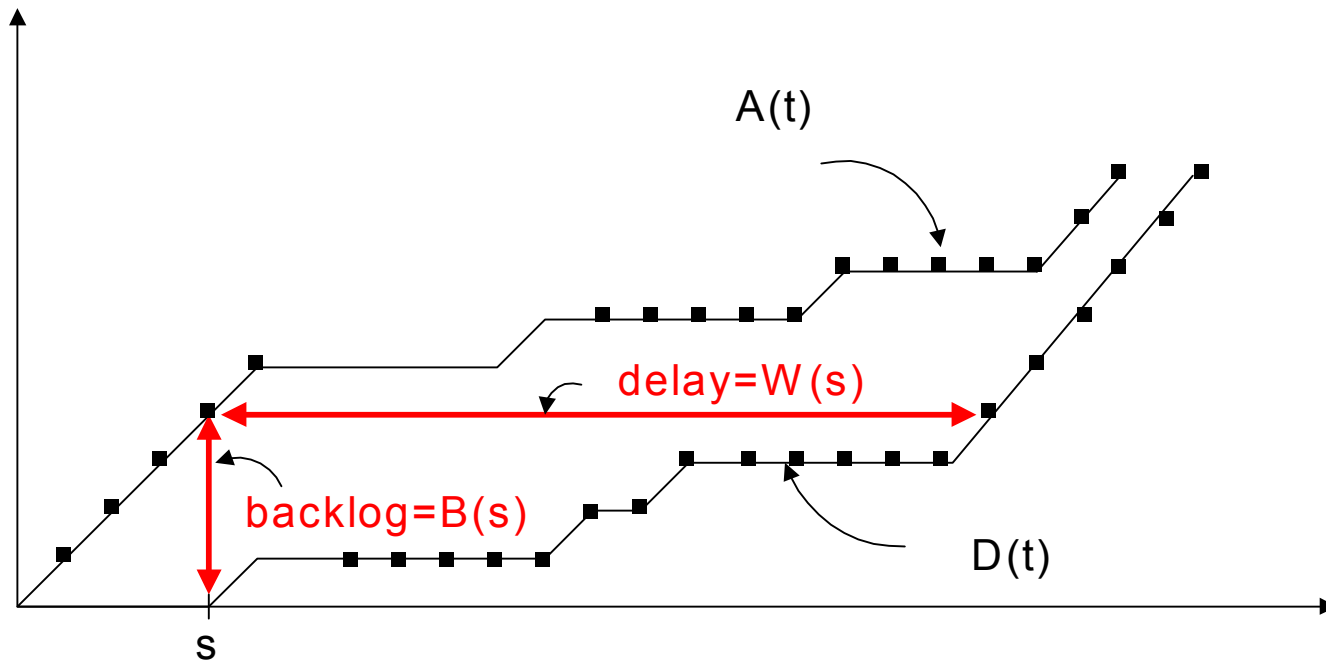
Regulated - leaky
bucket

On-Off - On-off
source

FBM - Fractional
Brownian Motion

$C = 100$ Mbps, $\epsilon = 10^{-6}$

Statistical Network Calculus with Min-Plus Algebra



Convolution and Deconvolution operators

- **Convolution operation:**

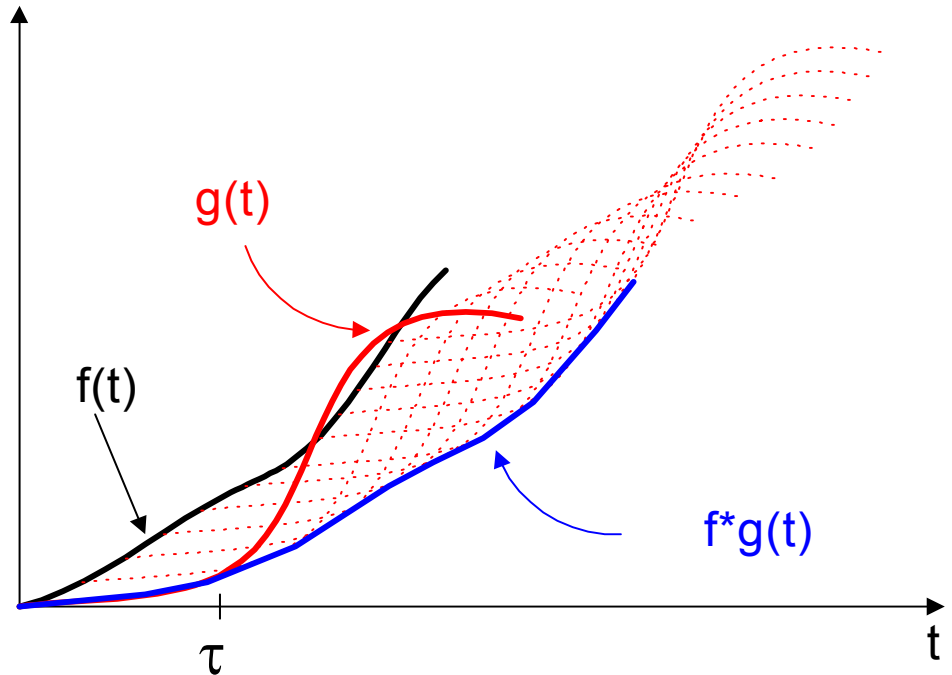
$$f * g(t) = \inf_{\tau \in [0, t]} f(t - \tau) + g(\tau)$$

- **Deconvolution operation**

$$f \otimes g(t) = \sup_{\tau \in [0, t]} f(t + \tau) - g(\tau)$$

- **Impulse function:**

$$\delta_{\tau}(t) = \begin{cases} \infty & , t > \tau \\ 0 & , t \leq \tau \end{cases}$$



Service Curves (Cruz 1995)

A (minimum) service curve for a flow is a function S such that:

$$D(t) \geq A * S(t) \quad , \forall t \geq 0$$

Examples:

- Constant rate service curve: $S(t) = c \cdot t$
- Service curve with delay guarantees: $S(t) = \delta_d(t)$

Network Calculus Main Results (Cruz, Chang, LeBoudec)

1. Output Envelope: $A^* \otimes S$ is an envelope for the departures:

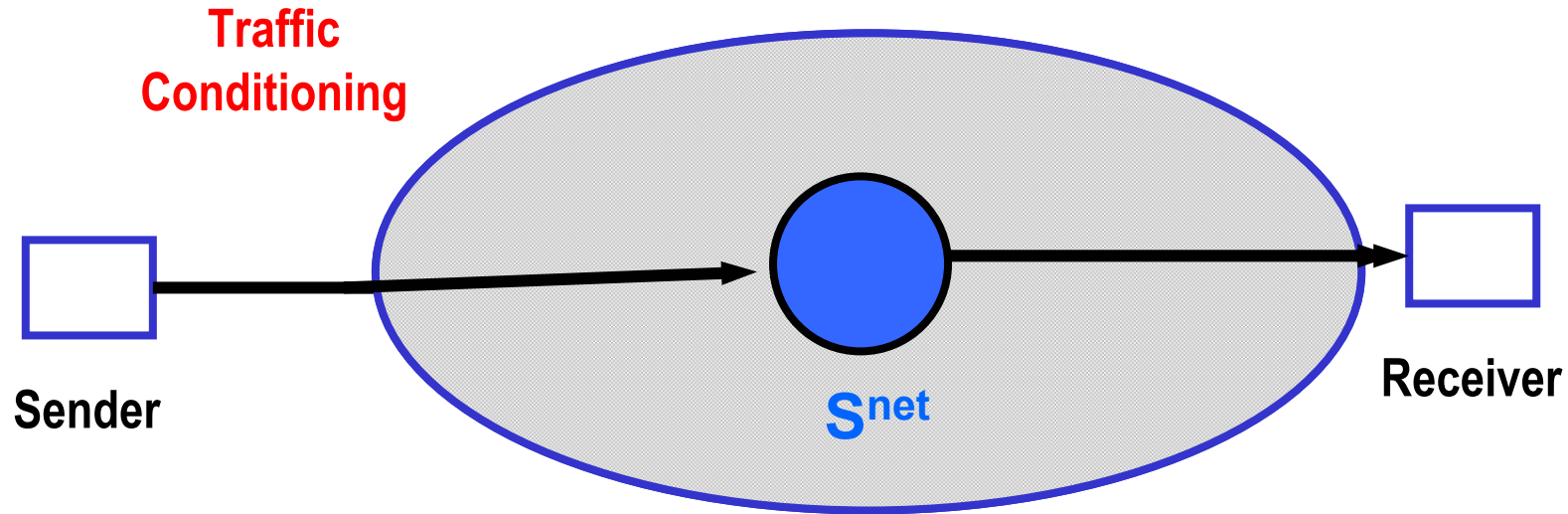
$$A^* \otimes S(t) \geq D(t + \tau) - D(\tau)$$

2. Backlog bound: $A^* \otimes S(0)$ is an upper bound for the backlog B

3. Delay bound: An upper bound for the delay is

$$d_{\max} \geq \inf_{\tau \in [0, t]} \left\{ d \geq 0 \mid \forall t \geq 0 : A^*(t - d) \leq S(t) \right\}$$

Network Service Curve (Cruz, Chang, LeBoudec)



Network Service Curve:

If S^1 , S^2 and S^3 are service curves for a flow at nodes, then

$$S_{net} = S^1 * S^2 * S^3$$

is a service curve for the entire network.

Statistical Network Calculus

A (minimum) service curve for a flow is a function S such that:

$$D(t) \geq A * S(t) \quad , \forall t \geq 0$$

A (minimum) effective service curve for a flow is a function S^ε such that:

$$\Pr[D(t) \geq A * S^\varepsilon(t)] \geq 1 - \varepsilon \quad , \forall t \geq 0$$

Statistical Network Calculus Theorems

1. **Output Envelope:** $A^* \otimes S^\varepsilon$ is an envelope for the departures:

$$\Pr[A^* \otimes S^\varepsilon(t) \geq D(t+\tau) - D(\tau)] \geq 1 - \varepsilon, \forall t, \tau \geq 0$$

2. **Backlog bound:** $A^* \otimes S^\varepsilon(0)$ is an upper bound for the backlog

$$\Pr[B(t) \leq A^* \otimes S^\varepsilon(0)] \geq 1 - \varepsilon, \forall t \geq 0$$

3. **Delay bound:** A probabilistic upper bound for the delay

$$d_{\max} \geq \inf_{\tau \in [0, t]} \left\{ d \geq 0 \mid \forall t \geq 0 : A^*(t-d) \leq S^\varepsilon(t) \right\}$$

, i.e.,

$$\Pr[W(t) \leq d_{\max}] \geq 1 - \varepsilon, \forall t \geq 0$$

Effective Network Service Curve

Network Service Curve:

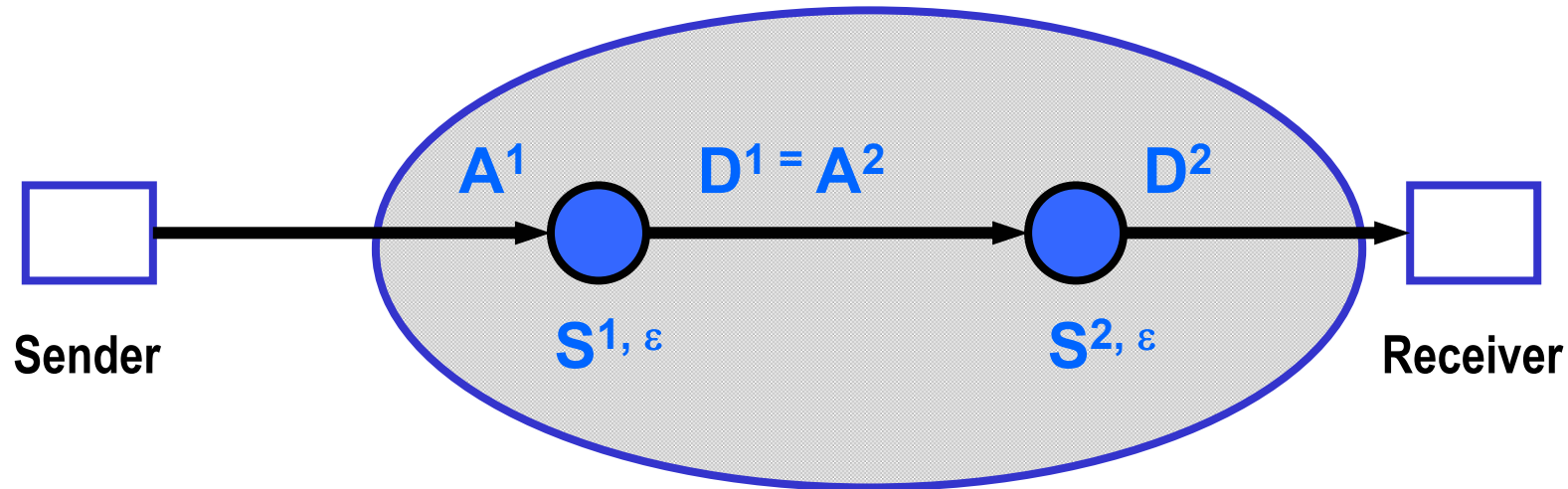
If $S^{1,\varepsilon}, S^{2,\varepsilon} \dots S^{H,\varepsilon}$ are effective service curves for a flow at nodes, then

$$\Pr [D(t) \geq A * (S^{1,\varepsilon} * S^{2,\varepsilon} * \dots * S^{H,\varepsilon} * \delta_{H a})(t)] \geq 1 - \frac{Ht\varepsilon}{a}$$

Unfortunately, this network service is not very useful!

A "good" network service curve can be obtained by working with a modified service curve definition

What is the cause of the problem with the network effective service curve?

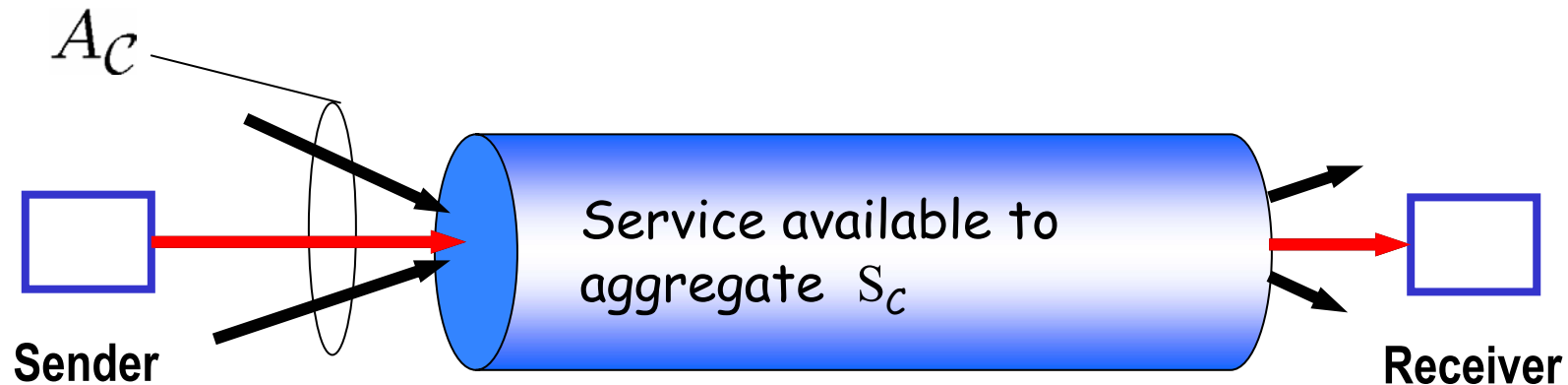


In the convolution

$$D^2(t) \geq A^2 * S^{2, \varepsilon}(t) = \inf_{\tau \in [0, t]} A^2(t - \tau) + S^{2, \varepsilon}(\tau)$$

the range $[0, t]$ where the infimum is taken is a random variable that does not have an a priori bound.

Statistical Per-Flow Service Bounds

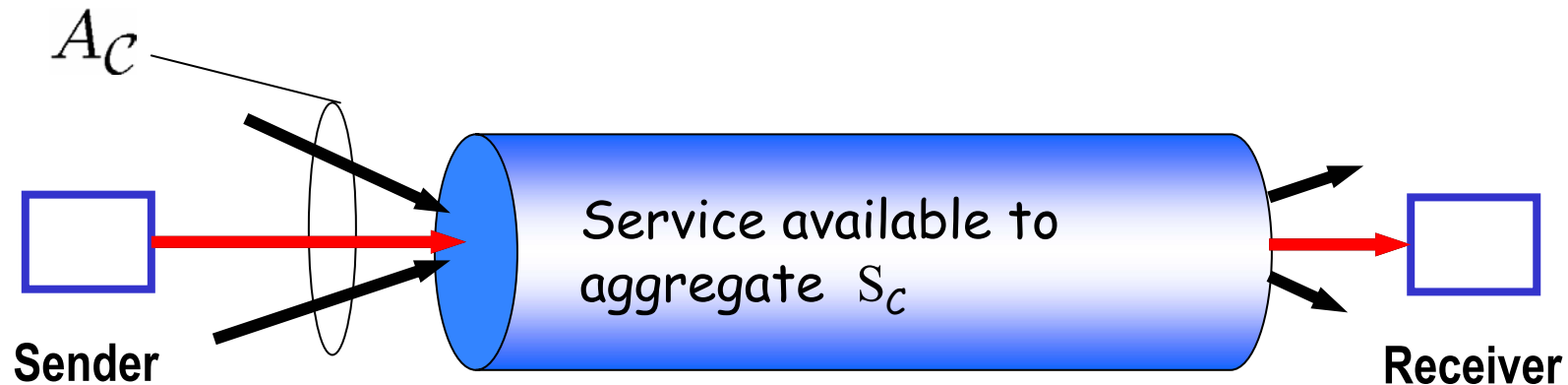


Given:

- Service guarantee to aggregate (S_c) is known
- Total Traffic $A_c = \sum_j A_j$ is known

What is a lower bound on the service seen by a single flow?

Statistical Per-Flow Service Bounds

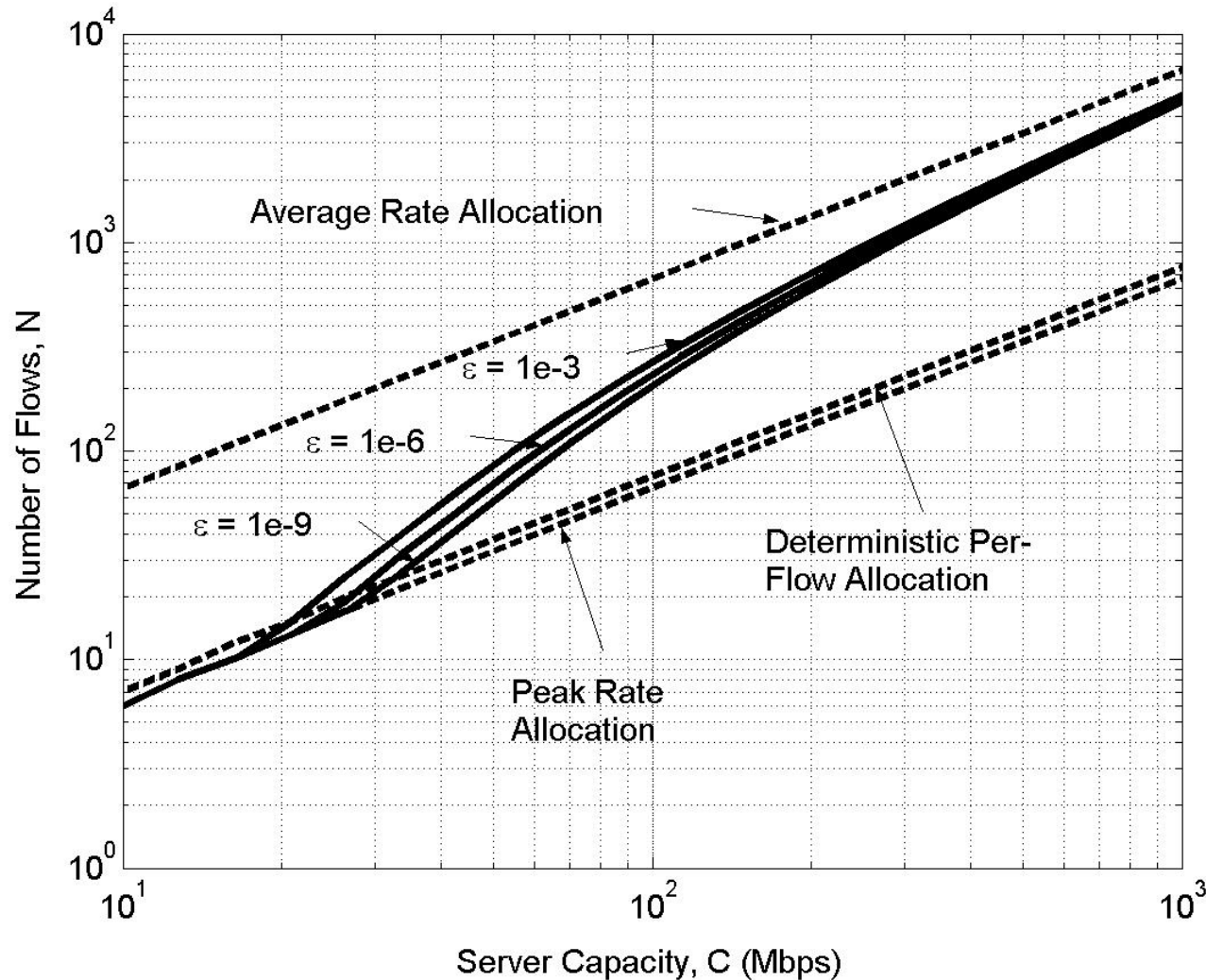


Can show:

$$S_j^{\varepsilon_1 + \varepsilon_2} = [S_c - \mathcal{H}_c^{T^{\varepsilon_1}, \varepsilon_2}]_+$$

is an effective service curve for a flow where
 $\mathcal{H}^{T^{\varepsilon_1}, \varepsilon_2}$ is a strong effective envelope and
 T^{ε_1} is a probabilistic bound on the busy period

Number of flows that can be admitted



Type 1 flows:

Goal: probabilistic
delay bound
 $d=10\text{ms}$

$$S_C(t) = C t$$

Conclusions

- Convergence of deterministic and statistical analysis with new constructs:
 - **Effective envelopes**
 - **Effective service curves**
- Preserves much (but not all) of the deterministic calculus
- **Open issues:**
 - So far: Often need bound on busy period or other bound on "relevant time scale".
 - Many problems still open for multi-node calculus