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Traffic Characteristics and Network Planning



Global Crossing®

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What to expect?

- A methodology to analyze your traffic, and apply the results to the planning process
- Practical approach
- An example from Global Crossing's network
- BUT, your network might be different in:
 - Scale
 - SLA's
 - Applications
 - Etc...

QoS in Backbone Networks

- Requirements are:
 - low delay
 - low jitter
 - low packet loss
- Common practice in backbone networks is overprovisioning:
 - Enough capacity in the network to meet demands
 - In peak times, and under failure conditions
- Prevent significant queue buildup

QoS in Backbone Networks

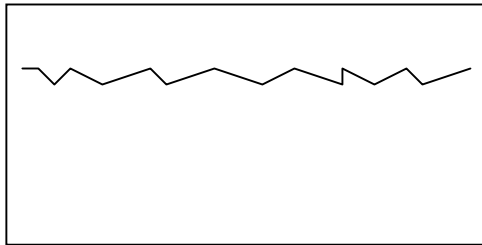
- The overprovisioning approach is effective
 - See Packet Design presentation at NANOG 22 [1]
- But capital is limited today...
- Can we do better than the rules-of-thumb:
 - "upgrade at 40% or 50% utilization"
 - "maximum 75% utilization under failure"
- Is aggregated traffic well-behaved enough to do "tight" capacity planning?

Related work: Opposite views (!)

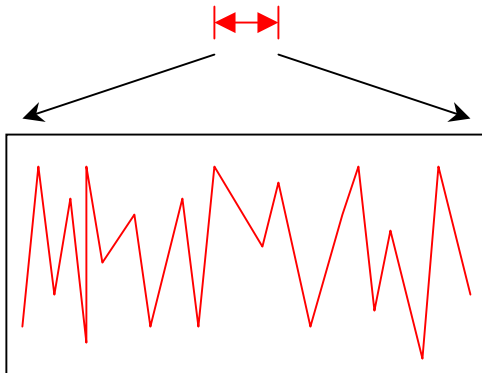
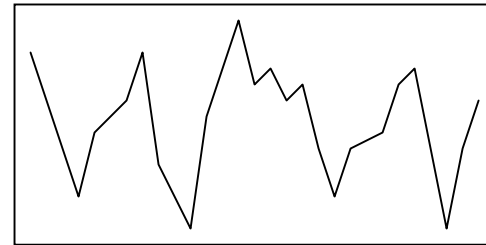
- M/M/1 queuing formula
- Markovian
 - Poisson-process
 - Infinite number of sources
- "Circuits can be operated at over 99% utilization, with delay and jitter well below 1ms" [2] [3]
- Self-Similarity
- Traffic is bursty at many or all timescales
- "Scale-invariant burstiness (i.e. self-similarity) introduces new complexities into optimization of network performance and makes the task of providing QoS together with achieving high utilization difficult" [4]

Opposite views

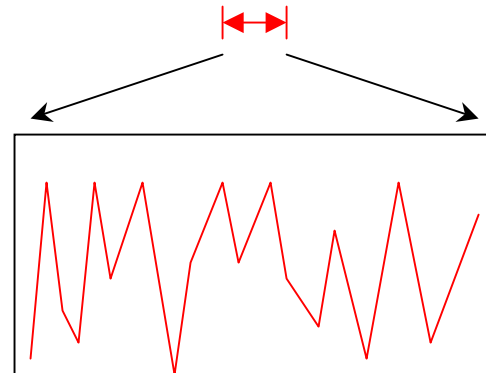
- M/M/1 queuing formula
- Self-Similarity



Long-term



Short-term



Network Planning Framework

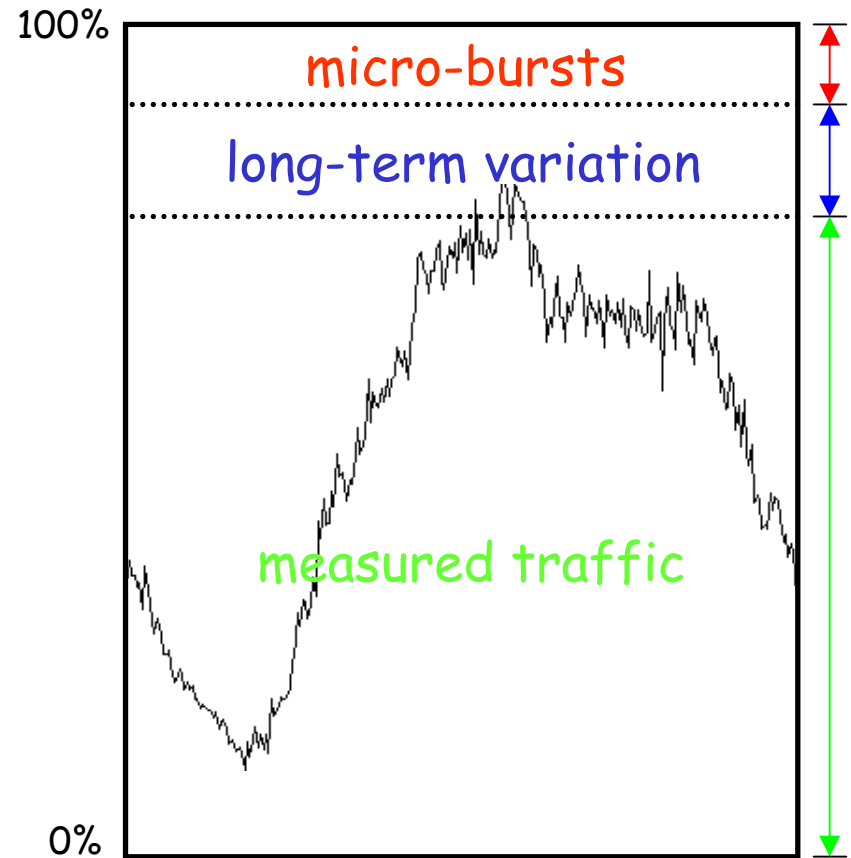
- Demand Characterization
 - Long-term: days/weeks timeframe
 - Short-term: dynamics at sub-5-min timescale
- Failure Analysis
 - Determine failure scenarios and SRLG's
- Simulation and Optimization
 - Determine minimum capacity deployment to meet objectives under normal and failure conditions

Demand Characterization

- Long-term
 - Robust estimation of 5-minute peak values
 - E.g. 95-percentile over day or week
 - Estimate "unforeseen" events
 - Calculate growth rate
- Short-term
 - Critical scale for queuing (1ms)
 - Determine overprovisioning factor that will prevent queue buildup against micro-bursts

Demand Characterization

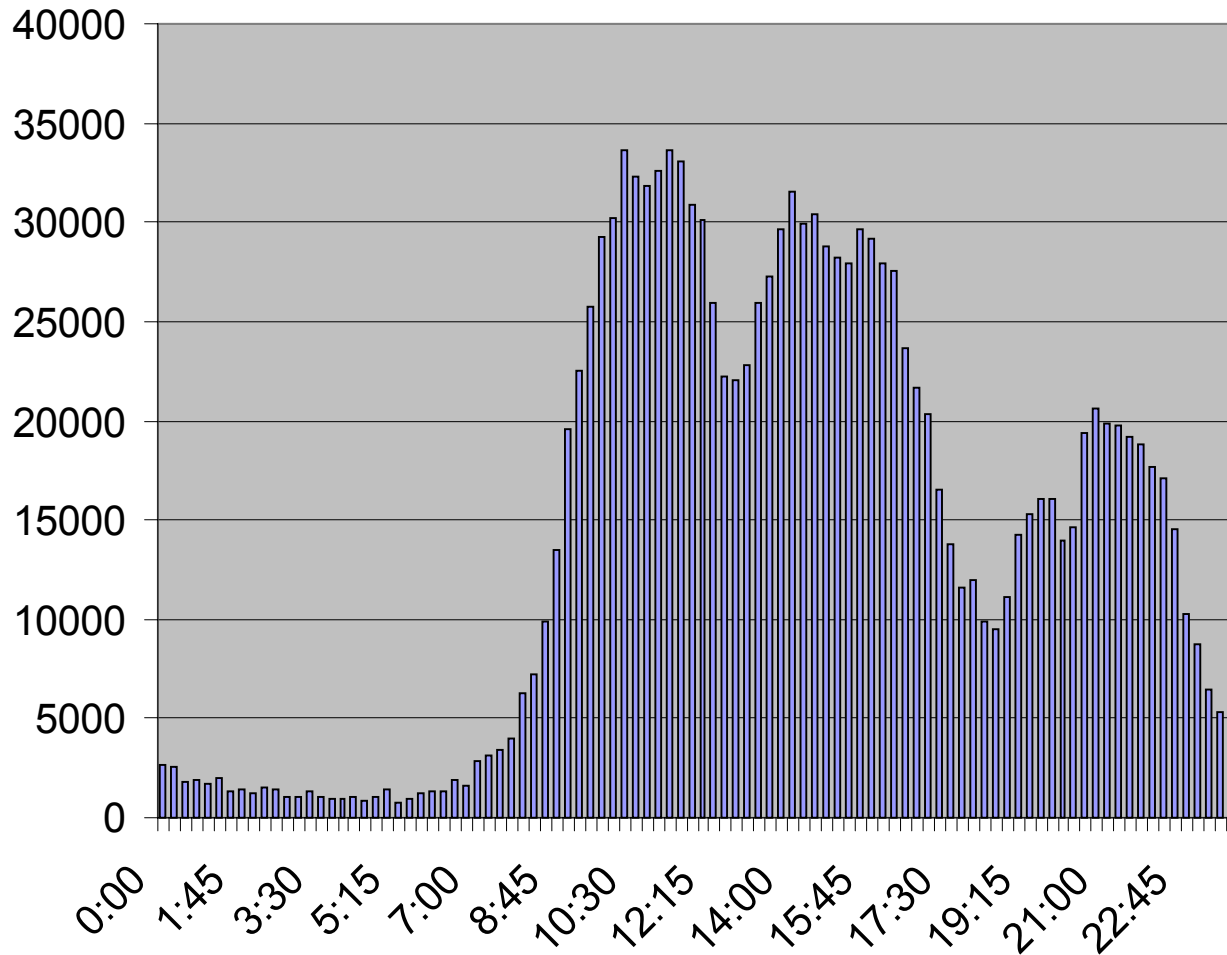
- Measured Traffic
 - P95 (day/week)
- Long-term variation
 - P95 to peak ratio
 - "unforeseen" events
- Micro-bursts
 - Short timescale traffic dynamics
- But let's first take a look at the telephony world...



Telephony Traffic

(inter-city on 6/3/2002)

Centi-Erlang



Voice Capacity Planning (Some) Assumptions

- Erlang B:
 - Call arrivals are random (Poisson)
 - Blocked calls are cleared
- Extended Erlang B:
 - Includes a retry percentage
- Erlang C
 - Blocked calls are queued (*"your call is very important to us, blah, blah..."*)

Voice Capacity Planning Example

- 1 Erlang = 1 hour of calls
 - Average numbers of calls in an hour
- Busy Hour Traffic: about 330 Erlang
- Erlang B formula (for 330 Erlang):
 - Blocking 1% -> 354 lines required
 - Blocking 0.1% -> 376 lines required
- "Overprovisioning" for 1% blocking: 7.3%
- "Overprovisioning" for 0.1% blocking: 13.9%

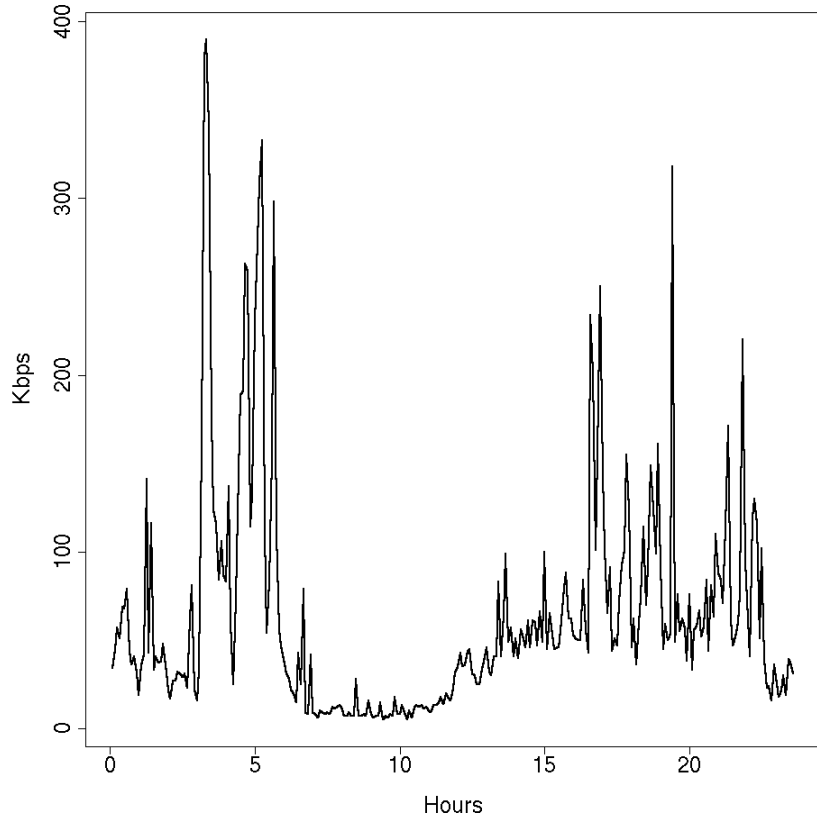
IP Capacity Planning

- Measurement data
 - E.g. 5-min average utilization
- Performance objectives
 - E.g. packet loss $< 0.1\%$, jitter $< 20\text{ms}$
 - End-to-end: convert to per-hop objective
- But we don't have an "Erlang formula"...
- Two paths towards a solution:
 - 1) Model the traffic, and fit parameters
 - 2) Experimentally derive guidelines

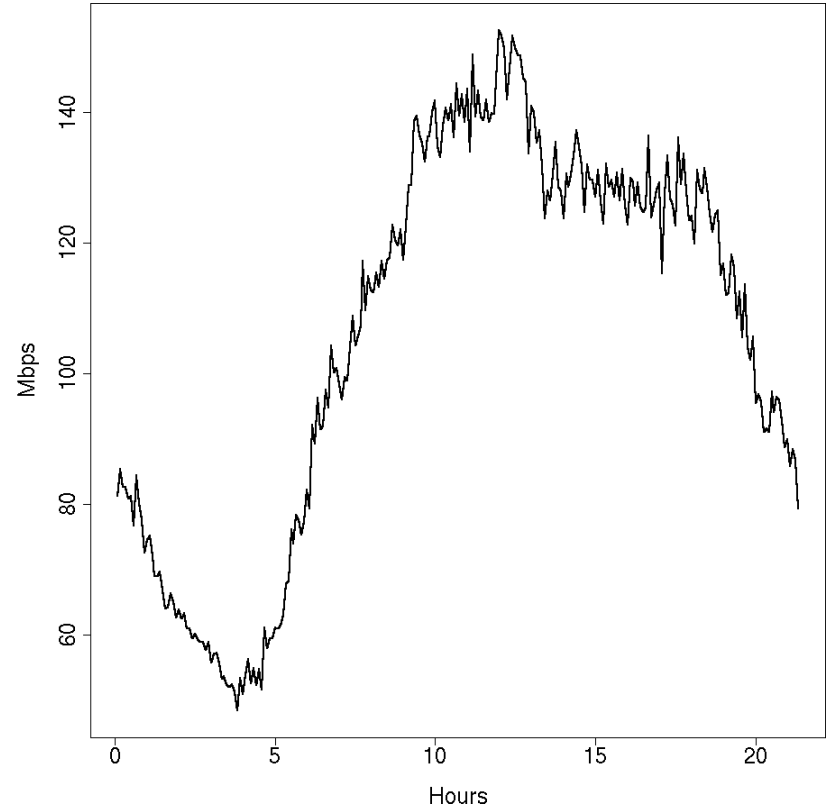
Long-term Traffic Characterization

- Investigate burstiness in 5-min measurements over days/weeks
- Bursty traffic: peaks are very large compared to average
 - I.e. the distribution is Heavy-Tailed
 - Mean and 95-percentile do not represent the traffic very well
 - Planning becomes very difficult
- Collect (SNMP) and analyze network data
 - Traffic Matrix via NetFlow or MPLS mesh

High- vs Low-Bandwidth Demands



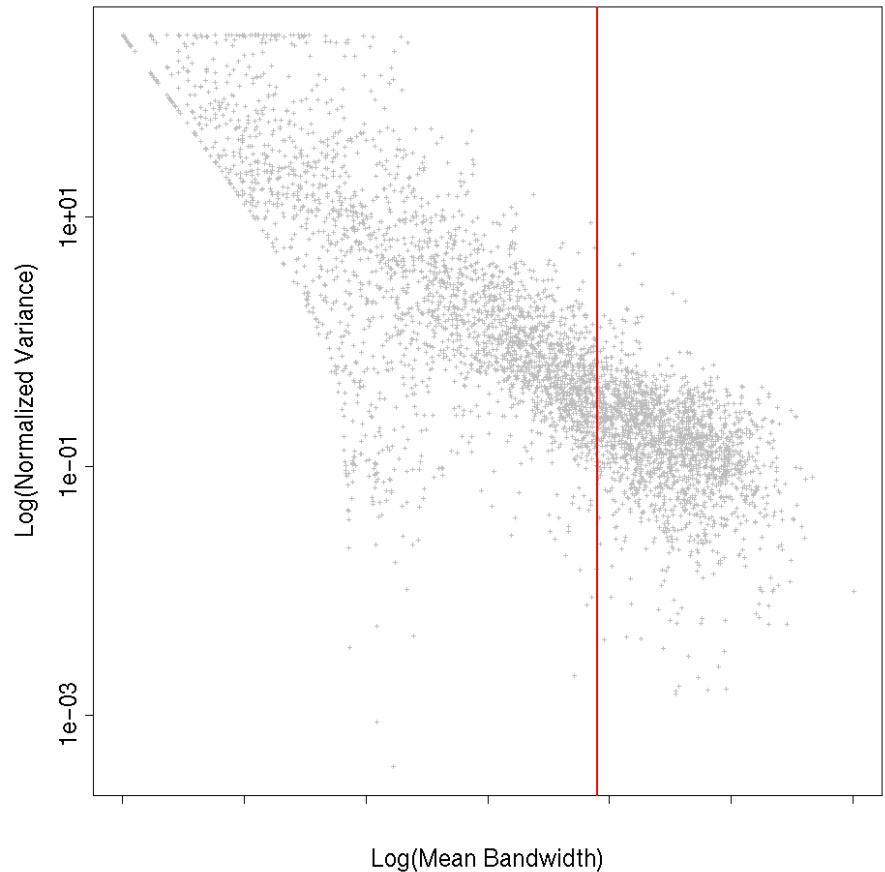
Cleveland -> Denver
Mean=64Kbps, Max=380Kbps
P95=201Kbps, alpha=1.8 (tail index)



Washington D.C. -> Copenhagen
Mean=106Mbps, Max=152Mbps
P95=144Mbps, alpha=21 (tail index)

Variance vs Bandwidth

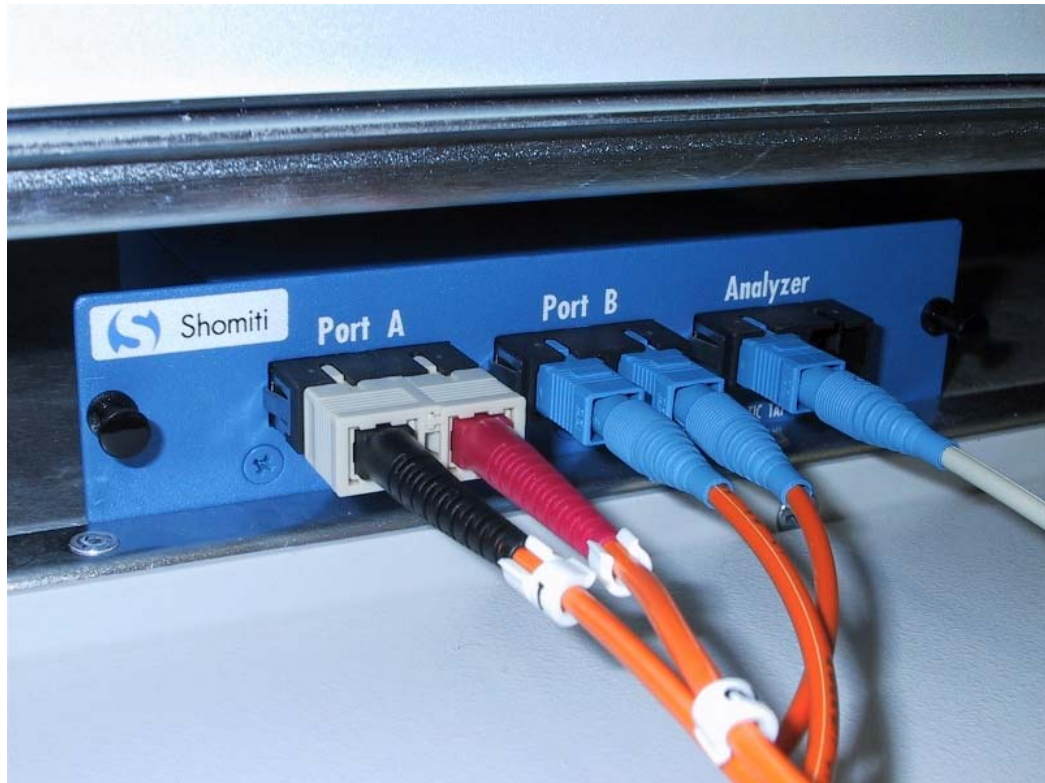
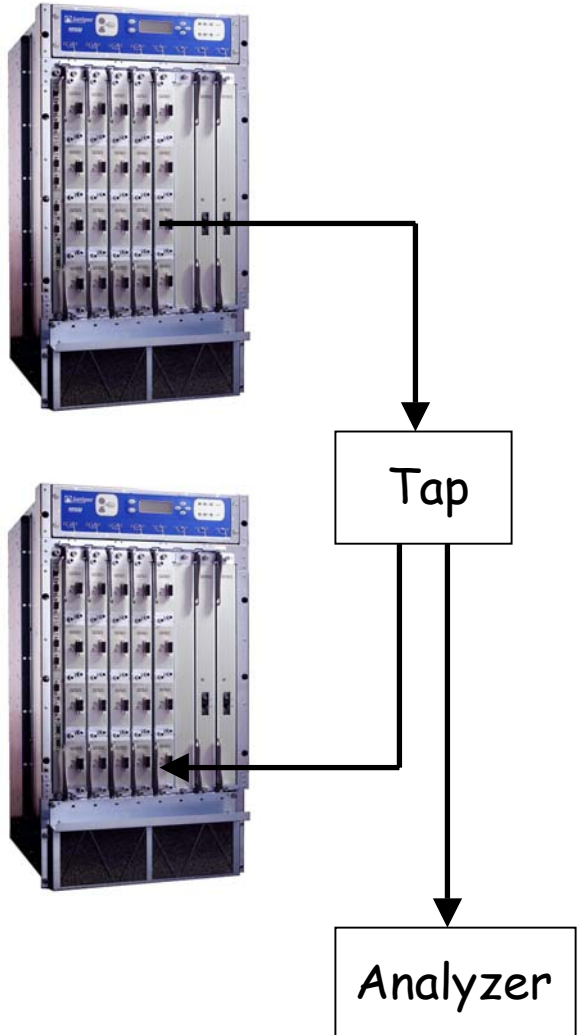
- Around 8200 demands between core routers
- Relative variance decreases with increasing bandwidth [5]
- Vertical red line is 0.5 Mbps
- High-bandwidth demands seem well-behaved
- 98% of traffic is carried by the demands larger than 0.5 Mbps



Short-term Traffic Characterization

- Investigate burstiness within 5-min intervals
- Measurements at critical timescale for queuing, like 1ms or 10ms
- Only at specific locations
 - Complex setup
 - A lot of data
- Analyze statistical properties

Fiber Tap (Gigabit Ethernet)

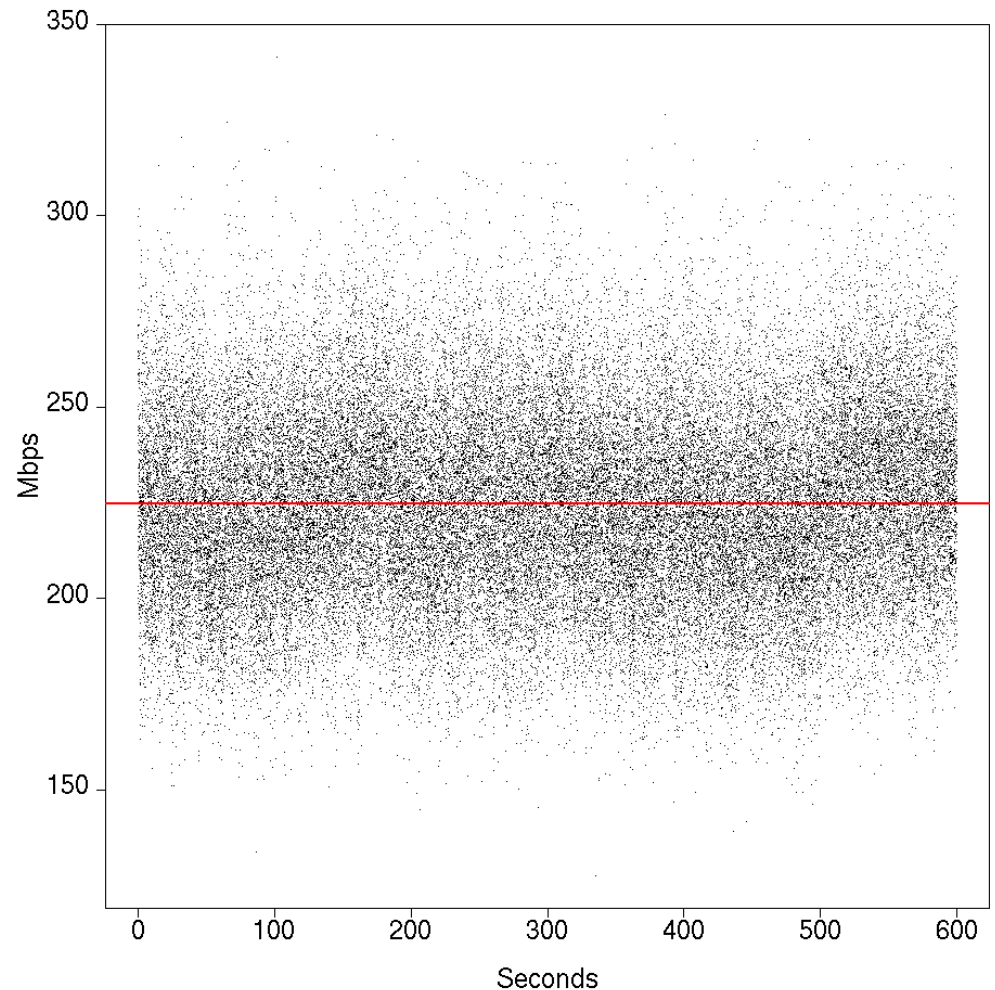


Raw Results

10 min. of data, 10ms scale

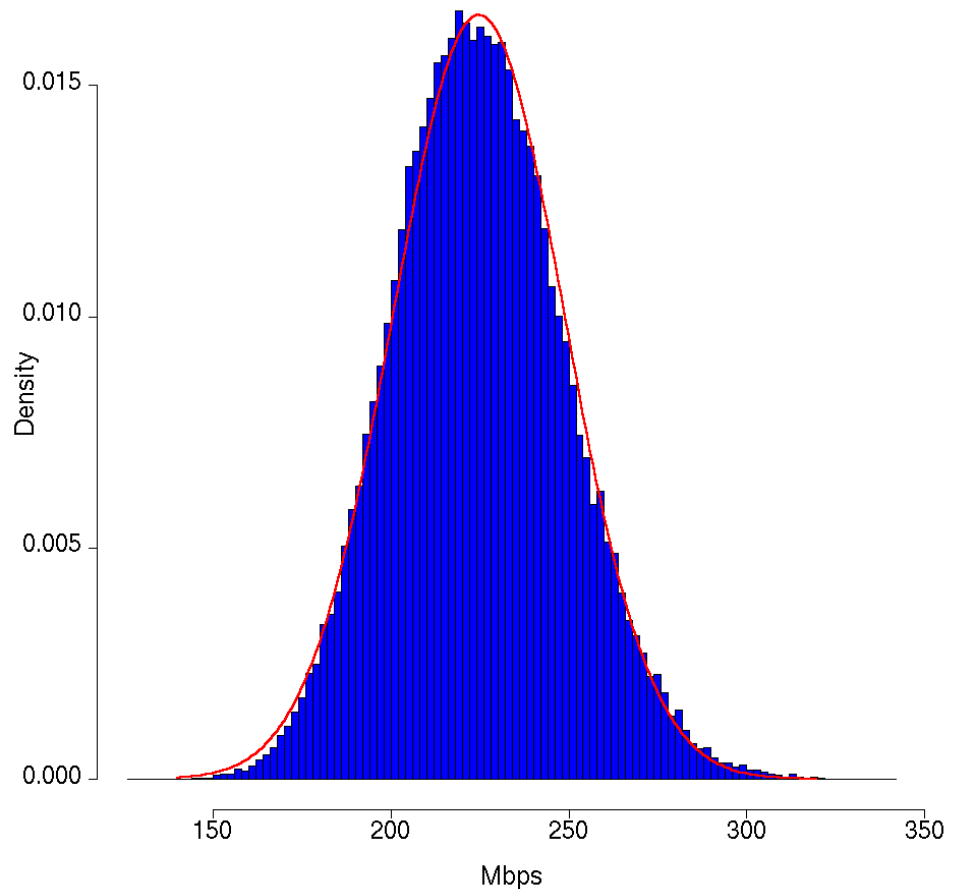
- Mean = 225 Mbps
- Max. = 342 Mbps
- Min. = 128 Mbps

- 95-percentile: 266 Mbps
- 5-percentile: 187 Mbps



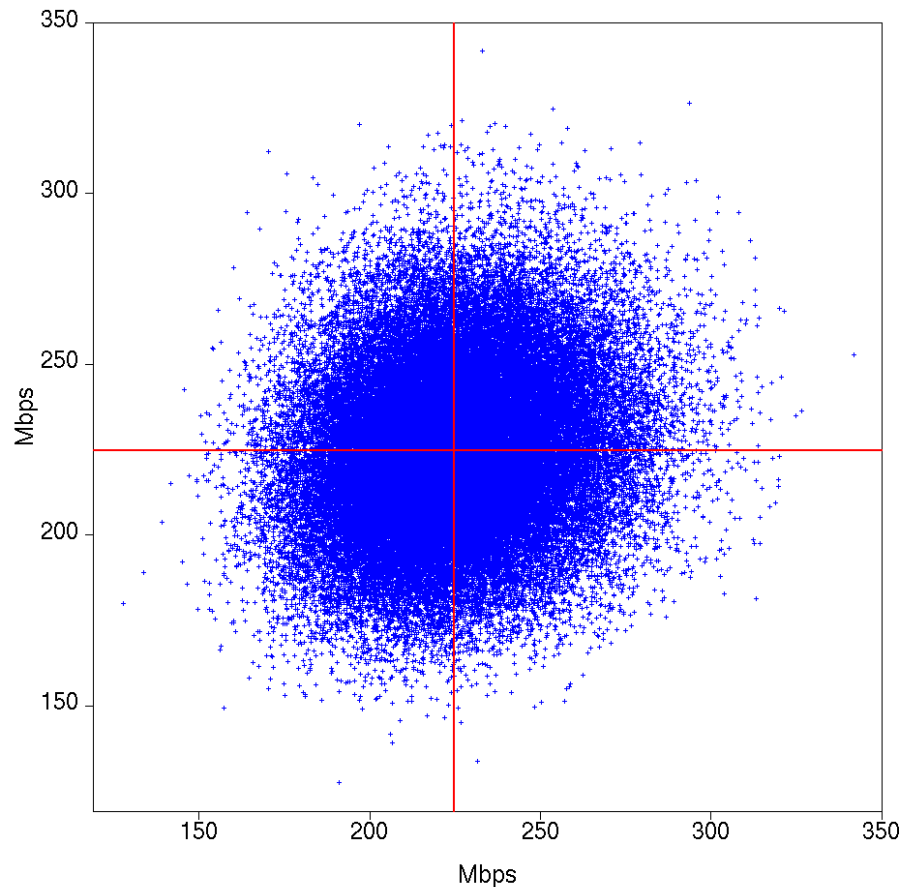
Traffic Distribution Histogram (10ms scale)

- Fits normal probability distribution very well (Std. dev. = 24 Mbps)
- No Heavy-Tails
- Suggests small overprovisioning factor



Autocorrelation Lag Plot (10ms scale)

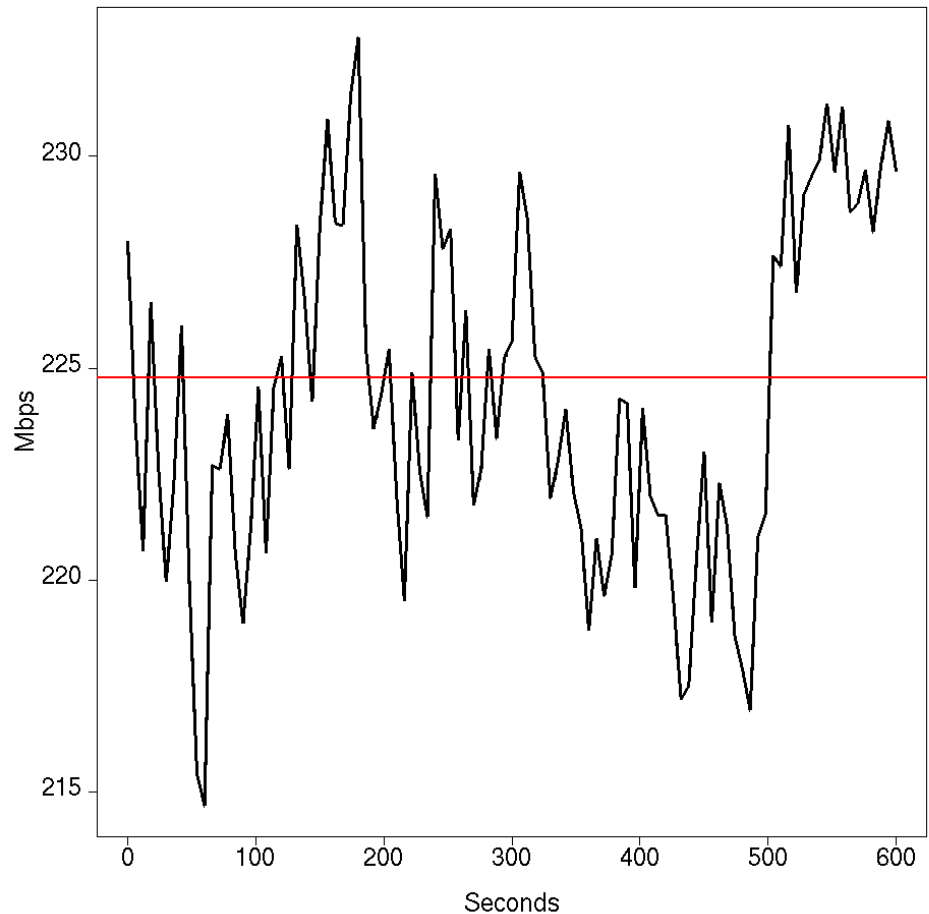
- Scatterplot for consecutive samples
- Are periods of high usage followed by other periods of high usage?
- Autocorrelation at 10ms is 0.16 (=uncorrelated)



Utilization

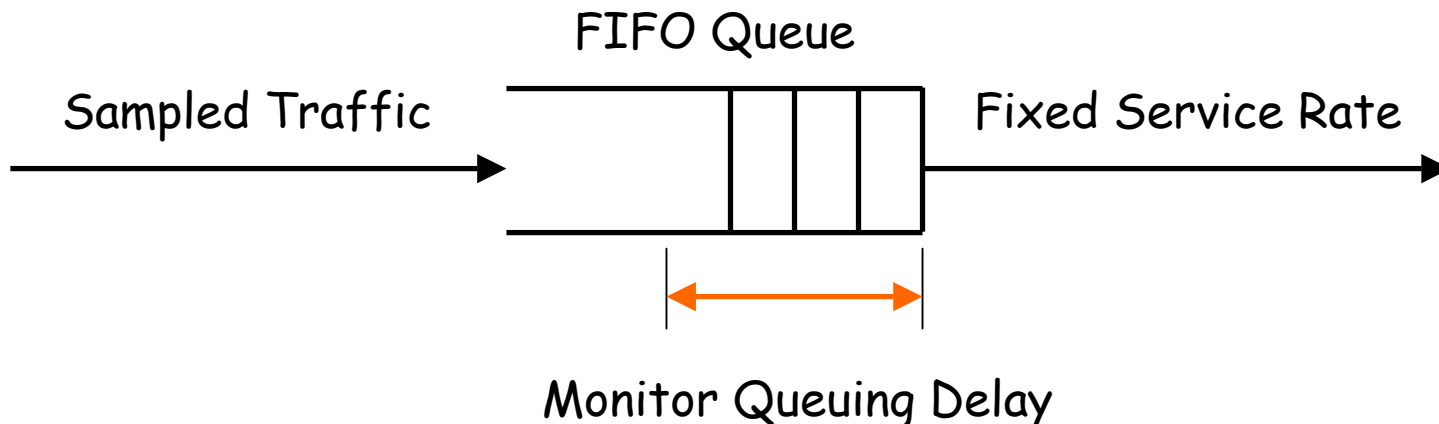
10 min. of data, 10 sec. scale

- Mean = 225 Mbps
- Max. = 233 Mbps
- Min. = 214 Mbps
- Clearly longer derivations from the mean
- High autocorrelation at 10 sec. (0.65)



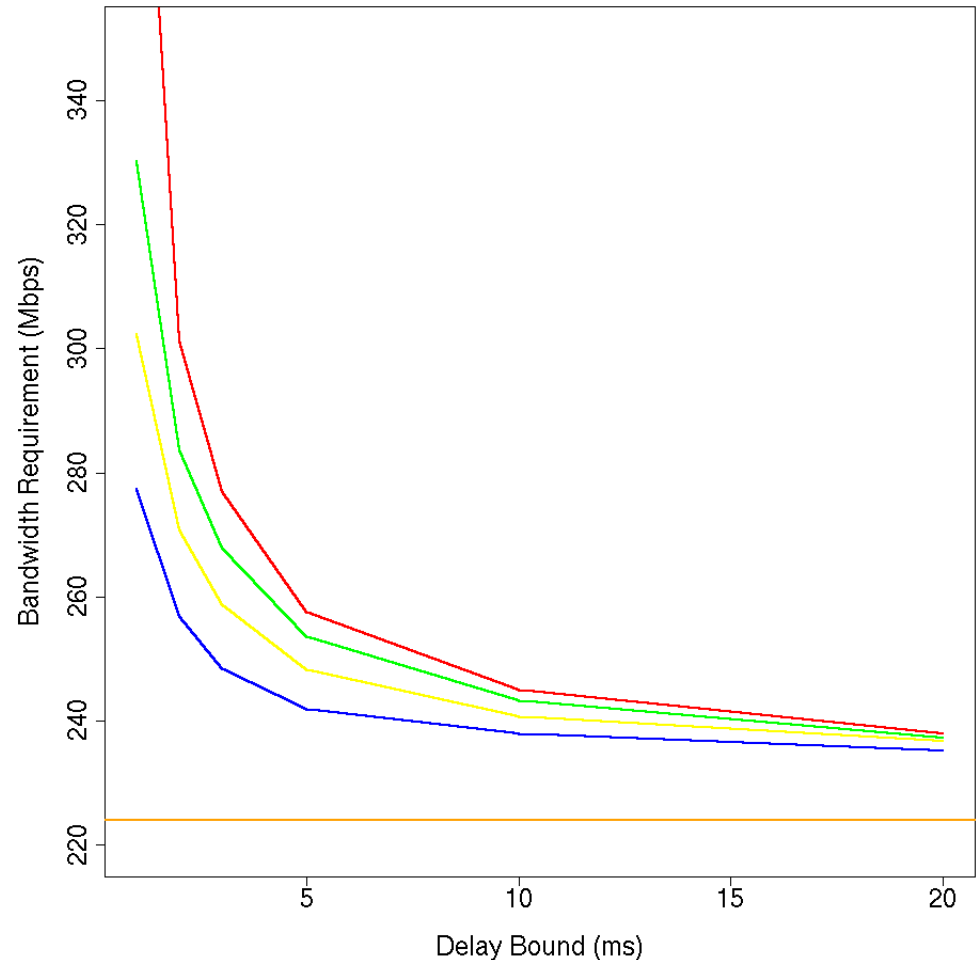
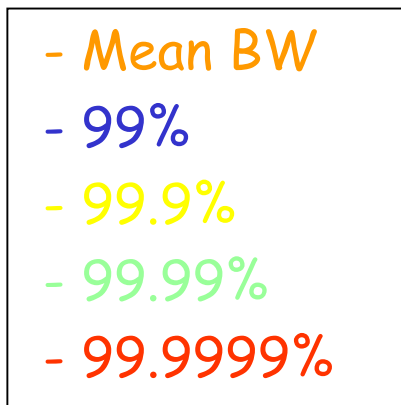
Queuing Simulation

- Feed sampled traffic data into FIFO queue (1ms)
- Fix Service Rate and max. Queuing Delay
- Measure amount of traffic that violates the delay bound
- Repeat for different Service Rates and Queuing Delays



Bandwidth Requirement vs Delay Bound

- How much Bandwidth is needed to meet the Delay Bound for a certain percentage of the traffic?



Bandwidth Requirements

Numeric Results

- Example 1
 - 5ms delay bound
 - 99.9999% of the traffic (10^{-6})
 - BW required: 257 Mbps
 - "Overprovisioning": 14%
- Example 2
 - 10ms delay bound
 - 99.9% of the traffic (10^{-3})
 - BW required: 241 Mbps
 - "Overprovisioning": 7%

Bandwidth Requirements Numeric Results (draft)

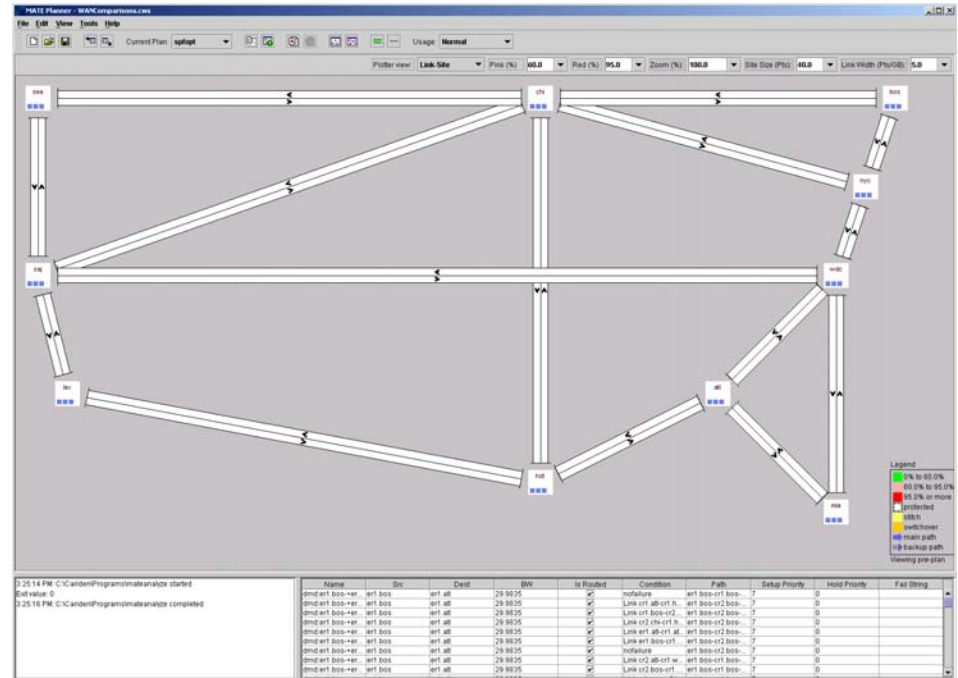
- Synthesized data: 704Mbps
 - 5ms delay bound
 - 99.9999% of the traffic (10^{-6})
 - BW required: 755 Mbps
 - "Overprovisioning": 7.2%
- Synthesized data: 1228Mbps
 - 5ms delay bound
 - 99.9999% of the traffic (10^{-6})
 - BW required: 1271 Mbps
 - "Overprovisioning": 3.5%

Back to the Framework

- Demand Characterization
 - Long-term well-behaved traffic
 - Overprovisioning for short-term bursts can be experimentally derived
- How to use this for planning purposes?
- Failure Analysis
 - Determine failure scenarios
 - E.g. single link failures, routers, SRLG, etc...
- Input for simulation

Simulation

- Feed demands and overprovisioning factors into simulation tool
- Run simulation for normal and failure scenarios
- Optimize Capacity Deployment and Routing (IGP or MPLS based) to meet requirements



- Tools like MATE (Cariden) and NPAT (WANDL)

How does Diff-Serv fit in this picture?

- All traffic in one class (no Diff-Serv) might require large overprovisioning factor for tight objectives (e.g. low delay/jitter for VoIP)
- Prioritizing that traffic (using a SPQ) would make the overprovisioning factor only applicable to that class
- The rest of the available bandwidth can be filled with less sensitive traffic
- But don't deploy too many classes...

Conclusions

- Not “Theory of Everything”, but empirical approach
- Backbone traffic is well-behaved enough to do meaningful network planning, but is not completely “smooth”
- Need several small timescale measurements to cover various types and rates of traffic

What did we learn from this example?

- On a Gigabit Ethernet (backbone) link a 'considerable' overprovisioning percentage is required to bound delay/jitter to a few milliseconds (in the order of 5-10%), on top of your overprovisioning for failures
- There is a good reason to deploy DiffServ to take care of really sensitive/critical traffic

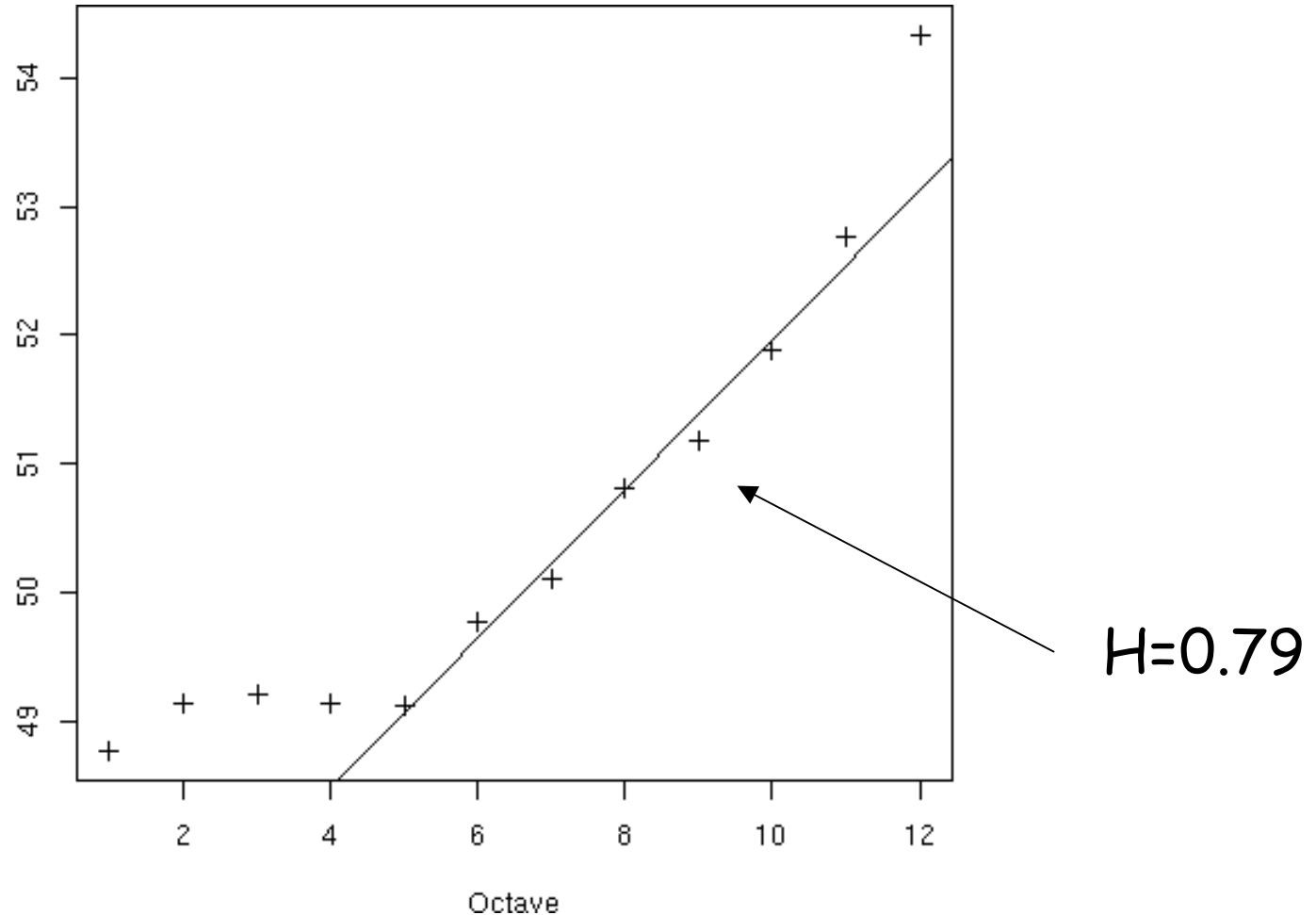
Extra Slides

Abry-Veitch Estimator

- Wavelet decomposition
 - Discrete wavelet transform
 - Time-scale wavelet domain
- Detail variance estimation
 - Coefficients squared and averaged over time ($u[j]$)
- Analysis using *Logscale Diagram*
 - Plot $\log(u[j])$ vs octave j
- LRD parameter estimation
 - *Hurst* parameter H from slope of plot

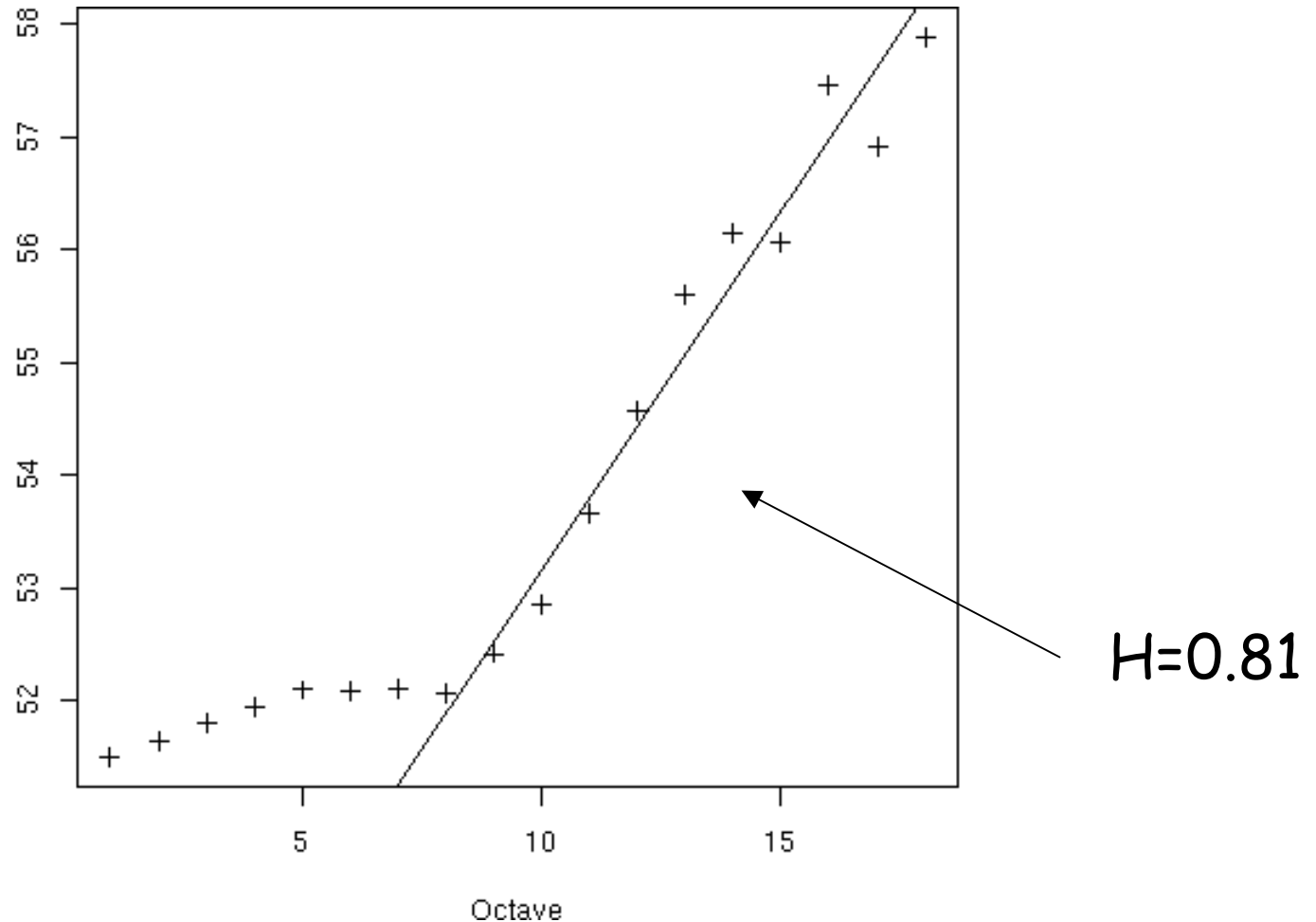
Logscale Diagram

10 min. of data, 10ms samples



Logscale Diagram

60 min. of data, 1ms samples



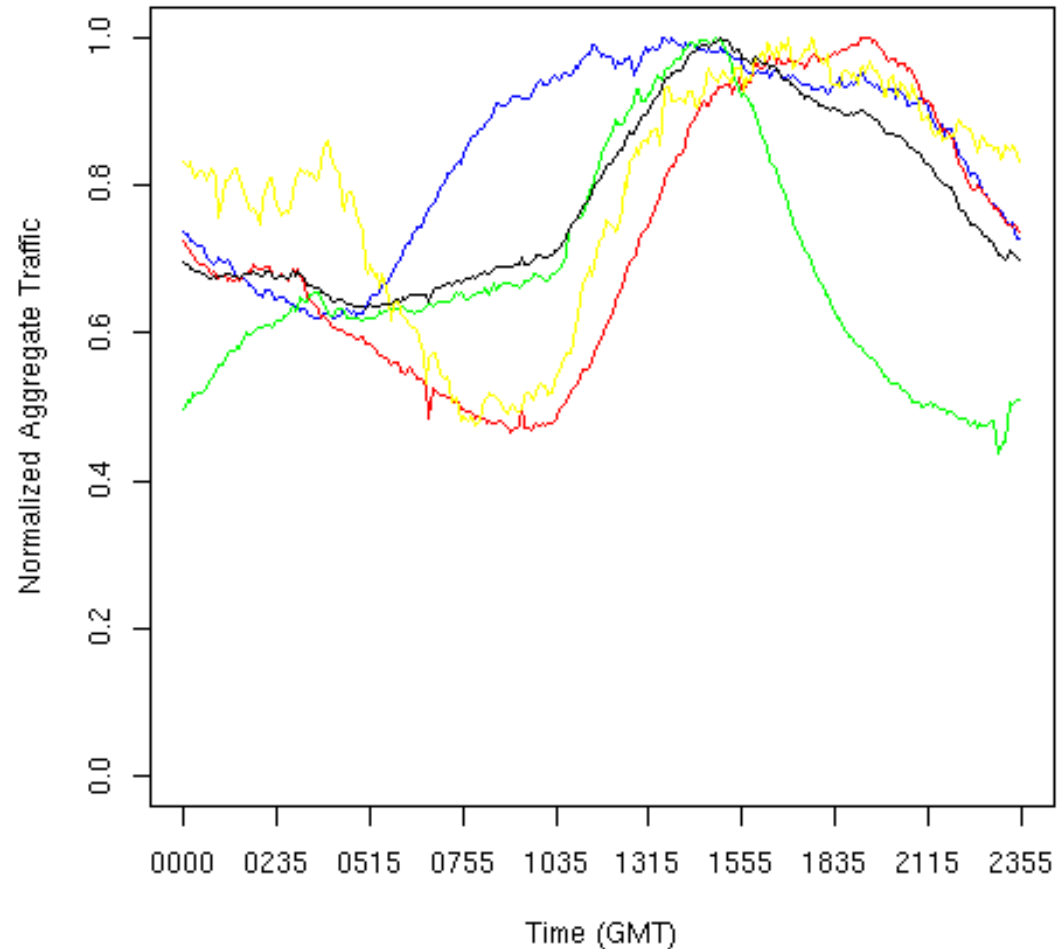
Geographical Traffic Profiles

- Does a world-wide network create utilization efficiencies because customers are distributed over several time zones?
- I.e. do Asian and European customer use the US network during non-peak hours?
- Yes... and No...
- Regional peaks overlap, around 3pm GMT
- Depends also on traffic ratios

Geographical Traffic Profiles

Customer Traffic

- Total
- N. America
- Europe
- Asia
- S. America



Acknowledgements

- Arman Maghbouleh (Cariden)
- Haobo Yu (Packet Design)
- Clarence Filsfils (Cisco)
- Fergal Toomey (Corvil)
- Richard Rensman (KPN)

- Upcoming Paper:
Realizing QoS with Efficient Network Design,
Steven Gordon, Arman Maghbouleh, Vishal
Sharma, Thomas Telkamp

Questions?

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References

- [1] Steve Casner, Cengiz Alaettinoglu and Chia-Chee Kuan, *A Fine-Grained View of High-Performance Networking*, NANOG 22 <http://www.nanog.org/mtg-0105/casner.html>
- [2] Chris Liljenstolpe, *Design Issues in Next Generation Carrier Networks*, MPLS 2001 Conference
- [3] Peter Lothberg, *A View of the Future: The IP-Only Internet*, NANOG 22, <http://www.nanog.org/mtg-0105/lothberg.html>
- [4] Zafer Sahinoglu and Sirin Tekinay, *On Multimedia Networks: Self-Similar Traffic and Network Performance*, IEEE Communications Magazine, January 1999
- [5] Robert Morris and Dong Lin, *Variance of Aggregated Web Traffic*, IEEE INFOCOM 2000, Tel Aviv, March 2000, pages 360-366