

The Relevance of Long-Range Dependence in Disk Traffic and Implications for Trace Synthesis

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I/O Workload Modeling and Synthesis

- ◆ Block-level disk I/O trace
 - Timestamp, r/w, offset, size
- ◆ Important because ...
 - Real traces are difficult to obtain
 - Performance analysis and architecture of storage systems depend upon traces and simulations
- ◆ Difficult because ...
 - Disk traffic is bursty
 - No consensus on what a “good” model should capture
 - We focus on performance-related characteristics
- ◆ Still an unsolved problem [Ganger]

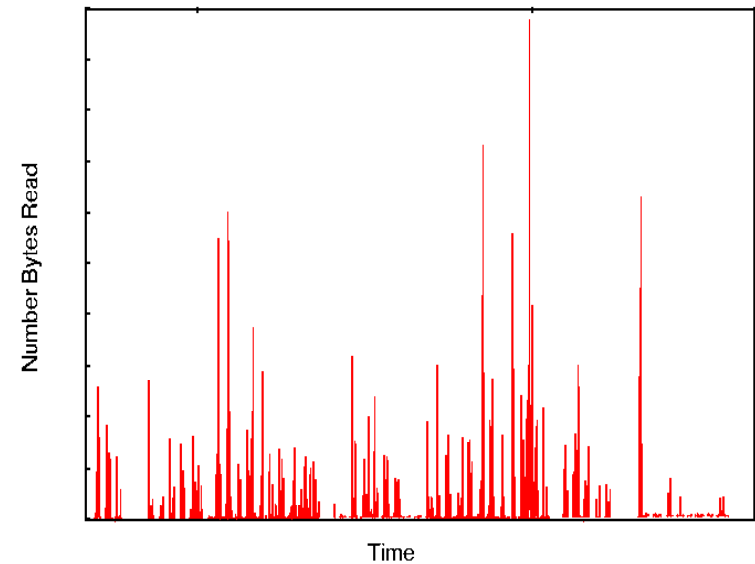
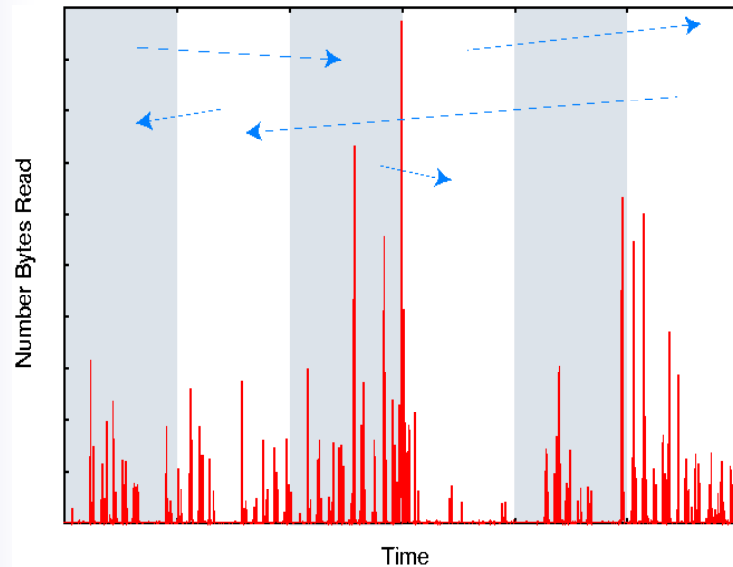


Long-Range Dependence in Disk Traffic

- ◆ Long-range dependence
 - Non-negligible event correlations across large time spans
 - Measured by the Hurst coefficient (0.5 ~ 1.0)
- ◆ Observed at ...
 - Network traffic [Leland et al.]
 - Web traffic [Crovella and Bestavros]
 - File system traffic [Gribble et al.]
 - Disk I/O traffic [Gomez and Santonja]
- ◆ Is long-range behavior really important?



Removing Long-Range Dependence



- ◆ Shuffling trace intervals removes long range dependence
- ◆ The technique was originally proposed in [Grossglauser and Bolot]

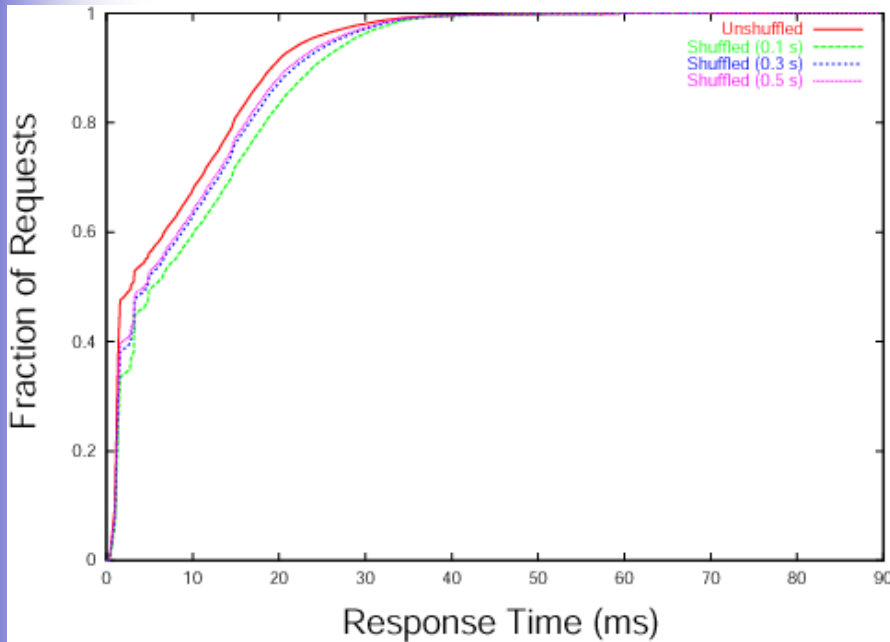


Experimental Methodology

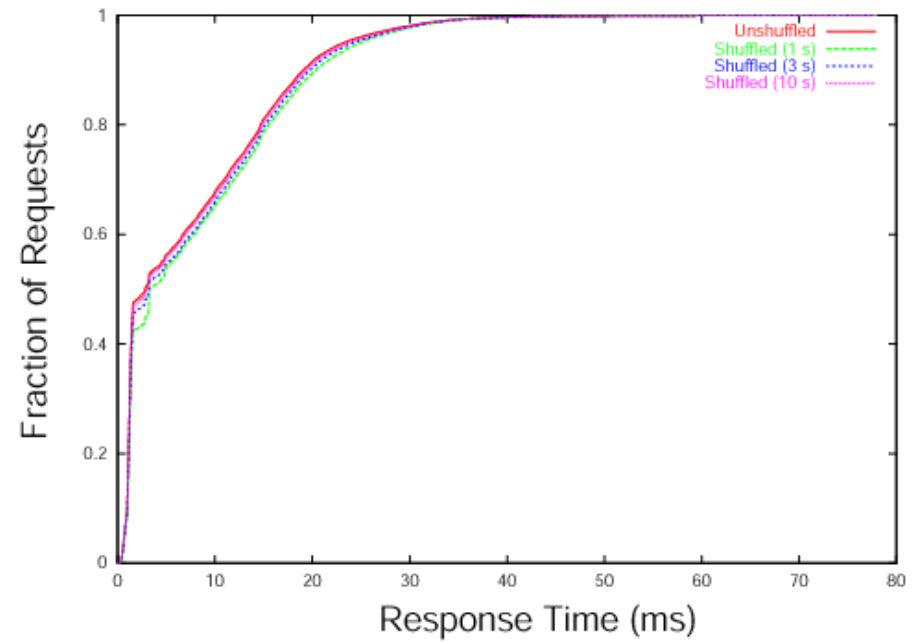
- ◆ HPL Pantheon I/O subsystem simulator [Wilkes]
- ◆ Metric: root mean squared (RMS) horizontal distance between the CDF of I/O response times [Ruemmler and Wilkes]
- ◆ Workloads [Ruemmler and Wilkes]
 - Cello news server traces
 - Random accesses
 - Snake file server traces
 - Sequential accesses



Results For Snake File Server



Small shuffling intervals

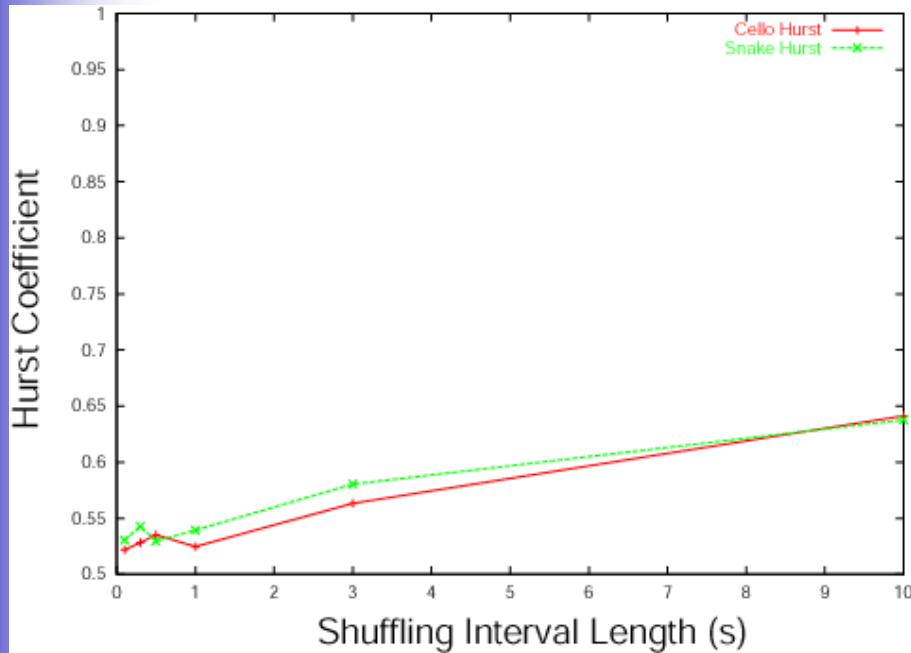


Large shuffling intervals

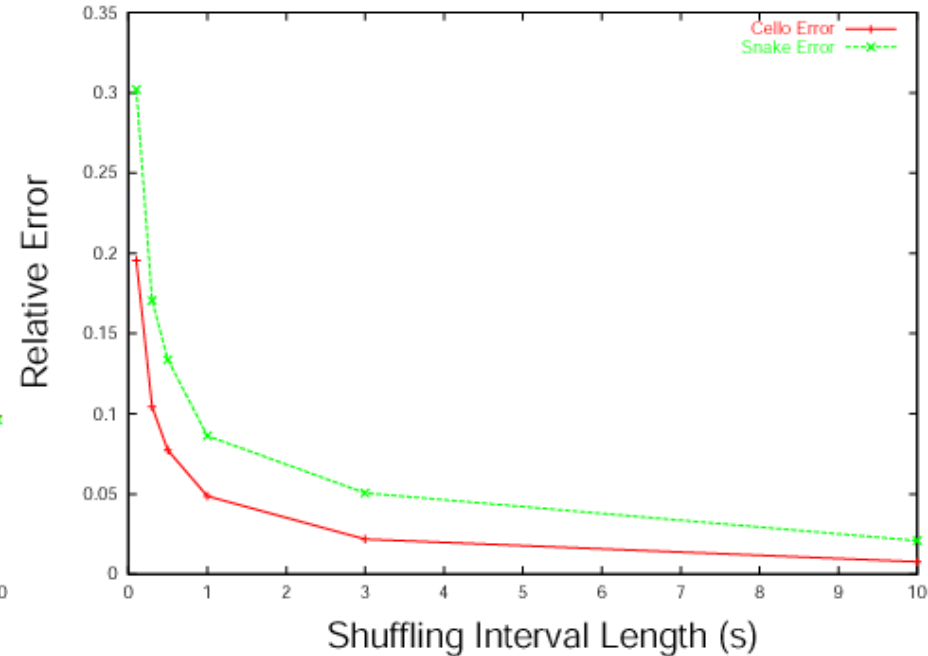
- ◆ Shuffling at very small time scales perturbs disk behavior
- ◆ Correlations longer than one second do not affect disk response times appreciably



Experimental Results



Hurst coefficient



Relative error (RMS)

- ◆ Long-range dependence is removed by shuffling
- ◆ Relative error is still small
- ◆ Dependence at large time scales in I/O traffic does not significantly affect disk behaviors with respect to response times



Implications of the Irrelevance of LRD

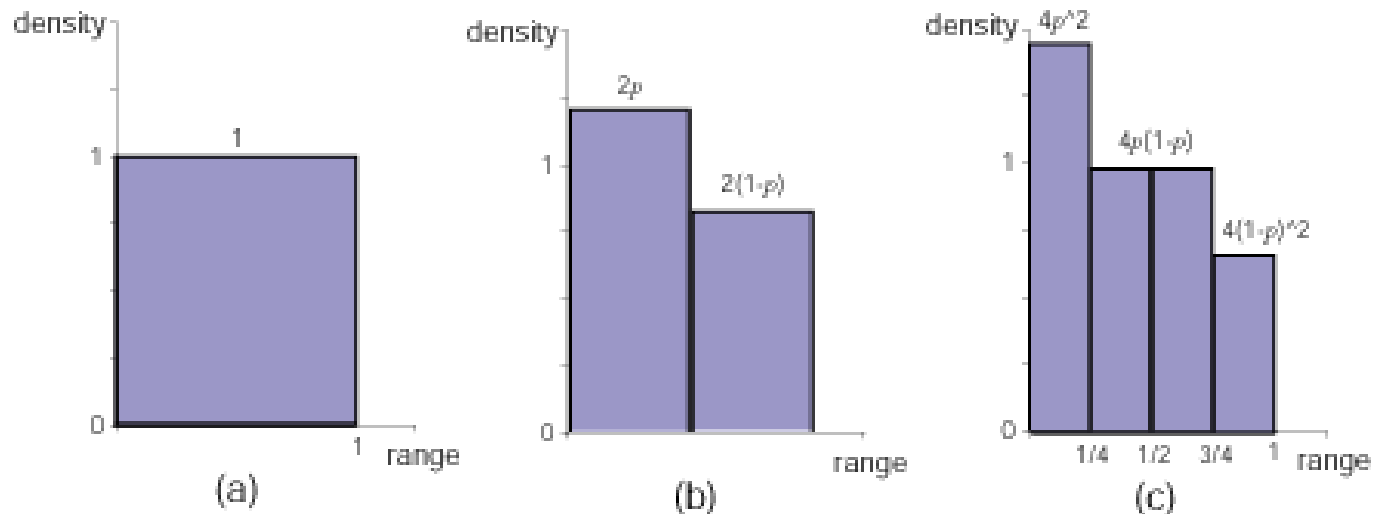
- ◆ Implications to workload modeling and synthesis
 - Short trace intervals can be considered as independent
- ◆ Binomial multifractals for local burstiness in I/O access patterns
- ◆ Reconstruct quality synthetic workloads from a small set of representative trace intervals, which are selected using cluster analysis based on metrics related to disk behaviors (in another paper)



Multifractal-Based I/O Trace Synthesis

◆ Binomial multifractals

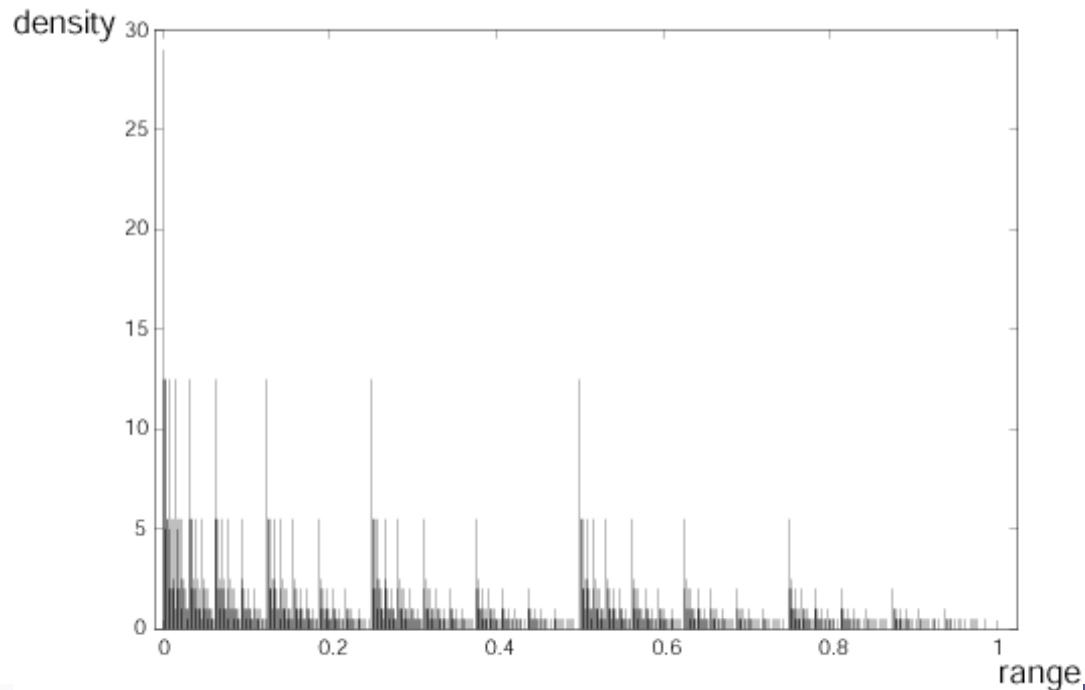
- 80/20 “laws”
- Bias p (> 0.5)
- Previously used in I/O traffic modeling by [Wang and Madhyastha et al.]



Multifractal-Based I/O Trace Synthesis

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Estimating Bias p

- ◆ Entropy values at different aggregation levels

$$Y_t^{(n)}(k) = \int_{k2^{-n}}^{(k+1)2^{-n}} Y_t dt, \text{ where } k = 0, 1, \dots, 2^n - 1$$

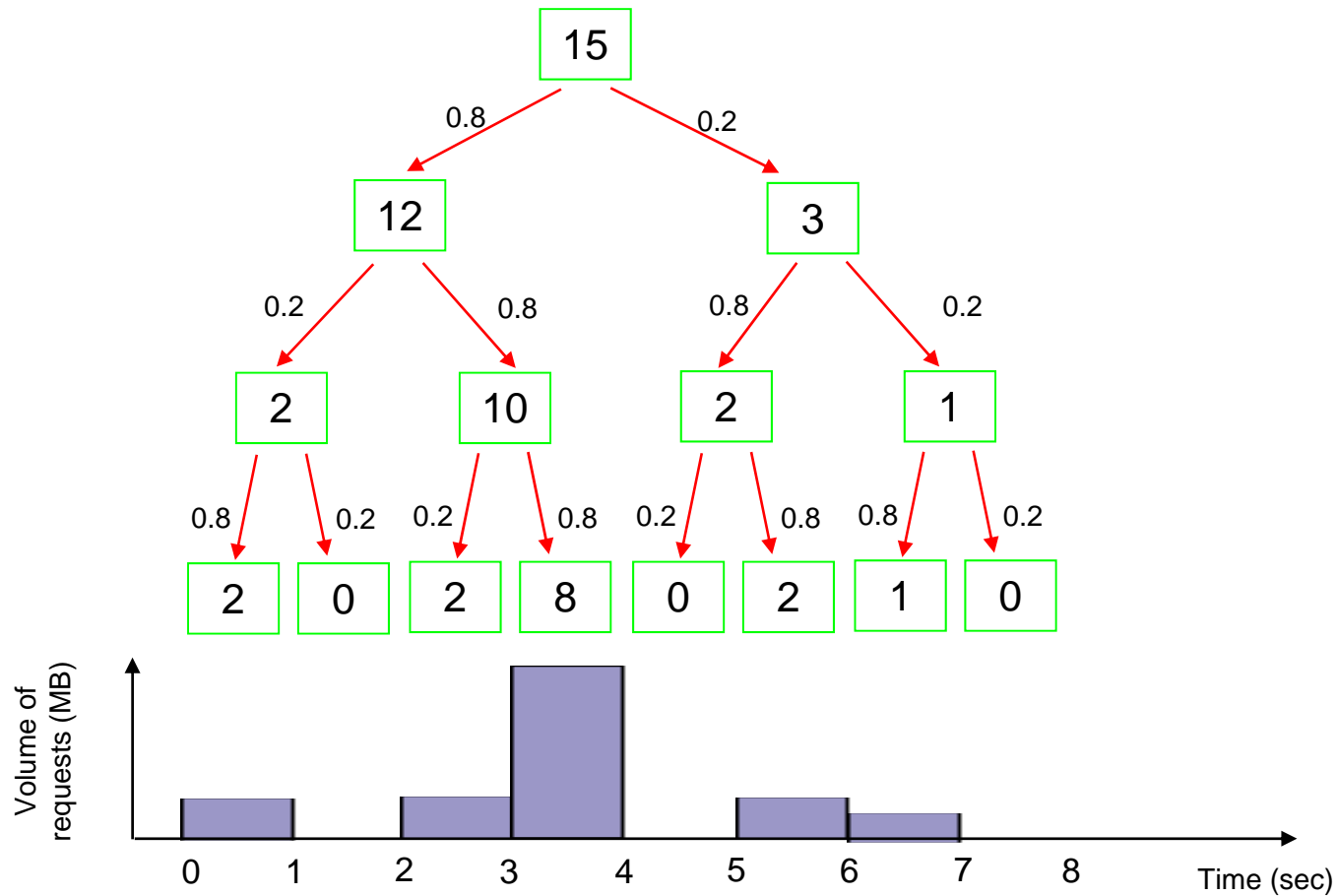
$$E_p^{(n)} = - \sum_{k=0}^{2^n-1} \frac{Y_t^{(n)}(k)}{\int_0^1 Y_t dt} \log_2 \frac{Y_t^{(n)}(k)}{\int_0^1 Y_t dt}$$

$$E_p^{(1)} = -p \log_2 p - (1-p) \log_2 (1-p)$$



Binomial Multifractal Generation

Bias p : 0.8, time interval length l : 8 sec, volume of requests m : 15 MB



Synthetic Trace Generation

Input: original trace file f , interval length s

Output: a time-stamped request sequence

$$((t_1, m_1), (t_2, m_2), \dots, (t_n, m_n))$$

For each non-empty interval w in f

Calculate bias p

$mass$ = aggregated request size in interval w

Binomial-Multifractal-Generation ($p, s, mass$)

map local time-stamps to real time-stamps

End For



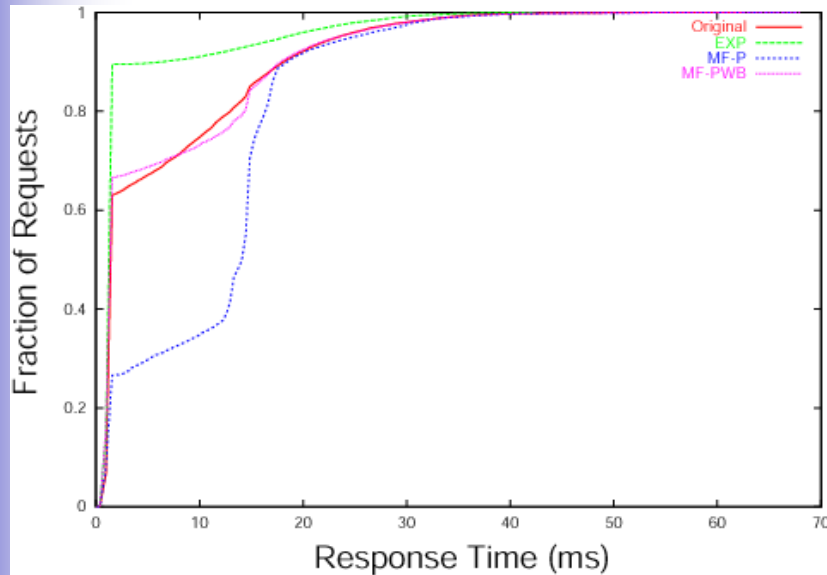
Synthesis Method Description

	EXP	MF-P (CMU)	MF-PWB
Interarrival pattern	Exponential	Binomial Multifractals	Binomial Multifractals
Timescale	N/A	Large (1 day)	Small (seconds)
Synthesizes	Interarrival time	Interarrival time and request size	Interarrival time and request size
Does not synthesize*	Request size, sector ID, r / w	Sector ID, r / w	Sector ID, r / w

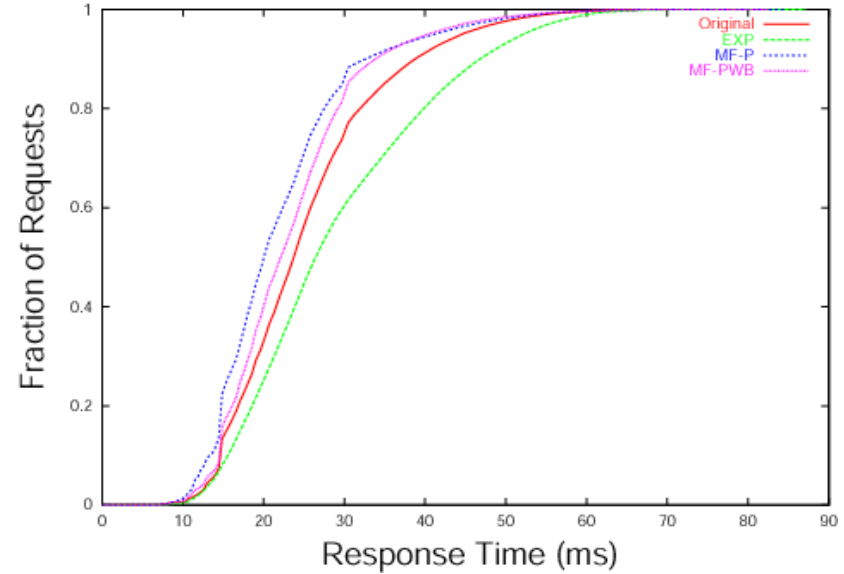
* The information that is not synthesized is retained from real traces



Synthetic Traces



Snake file server

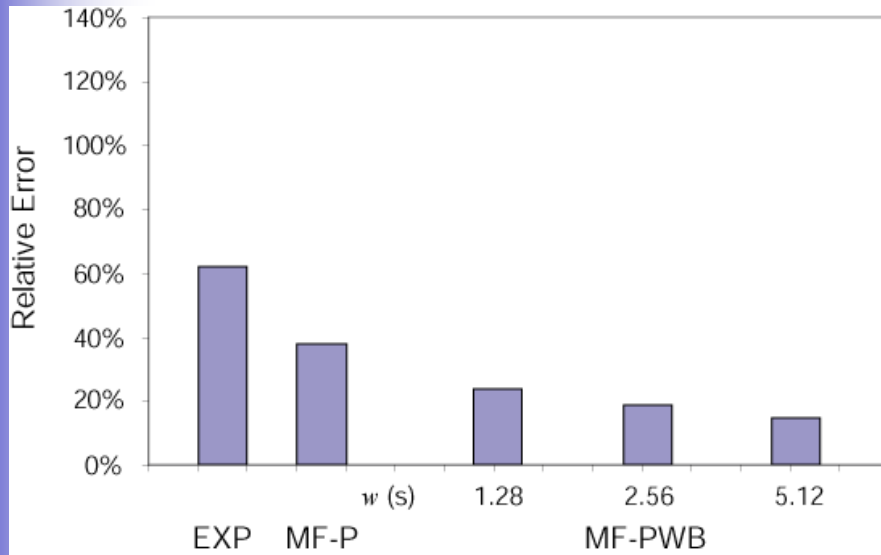


Cello news server

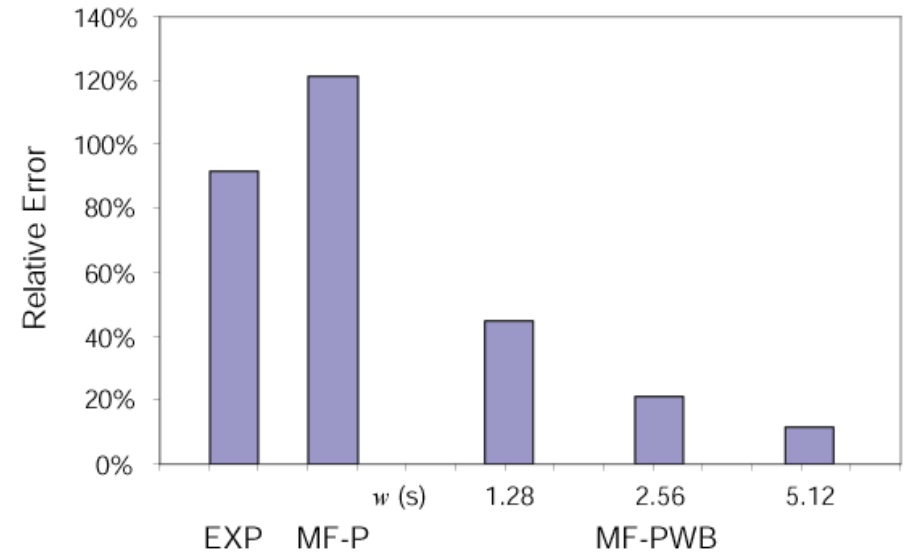
- ◆ Binomial multifractals models the interarrival pattern well
- ◆ Exponential does not
- ◆ Generating at small timescales improves synthesis quality



Experimental Results



Snake file server (Day 1)

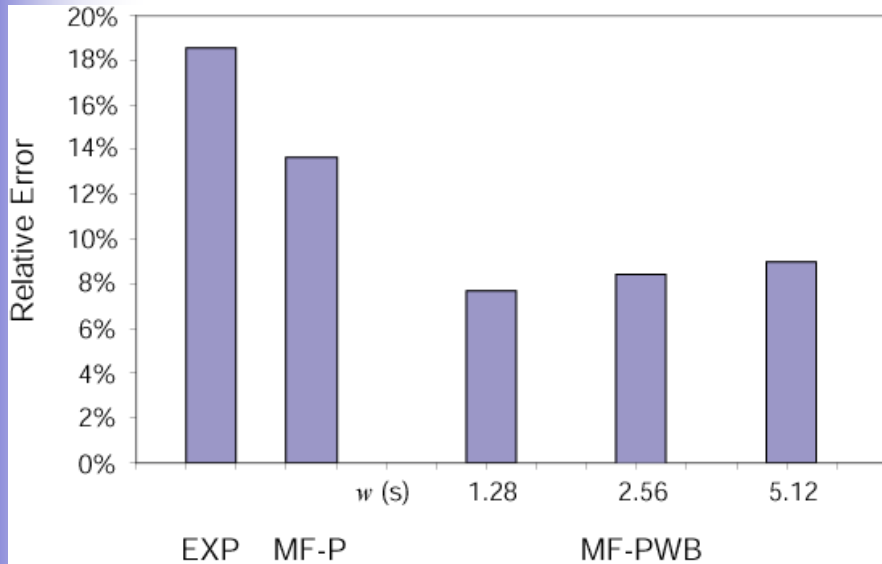


Snake file server (Day 2)

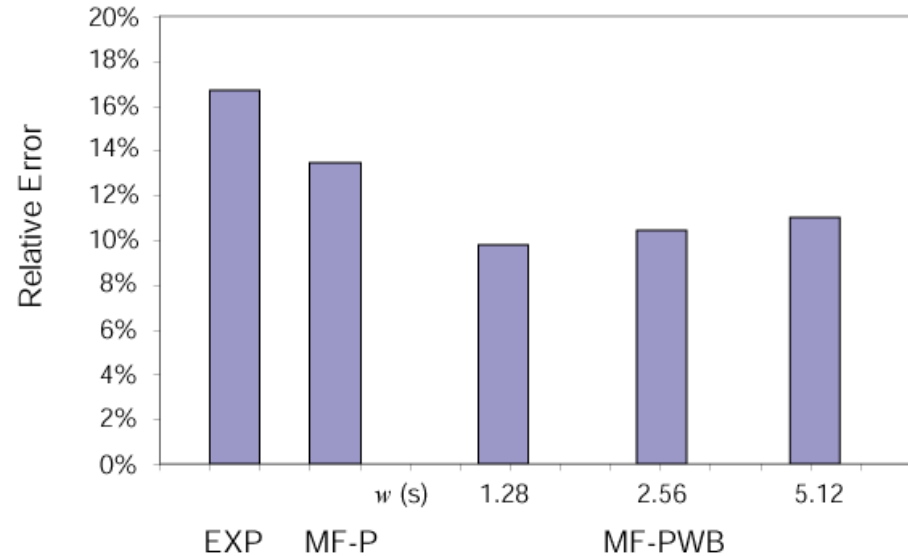
- ◆ MF-PWB method can generate the most accurate synthetic traces



Experimental Results



Cello news server (Day 1)



Cello news server (Day 2)

- ◆ The generation interval length matters
- ◆ The majority of synthesis errors by MF-P and MF-PWB comes from synthetic arrival time (results shown in paper)



Conclusions and Future Work

- ◆ Long range dependence is not important for reproducing certain performance metrics
- ◆ Binomial multifractals well capture local burstiness in I/O interarrival patterns
 - Achieve a 8 –12% demerit factor for I/O response times on random and sequential workloads
- ◆ Automatically determine appropriate interval length
- ◆ Need to quantify spatial locality
- ◆ PQRS model [Wang et al.]
 - Use a two-dimensional version of binomial multifractals
 - Model both spatial and temporal localities as well as their correlations



Thank You!

◆ Acknowledgements

- Our shepherds Julian Satran and Robert Chadduck
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- UCSC Storage Systems Research Center

◆ More information:

- <http://www.soe.ucsc.edu/~tara/stargroup>
- <http://www.soe.ucsc.edu/~hongbo/publications.html>

◆ Questions?

