

A Simulation Study of the Stochastic Compensation Effect for Packet Reordering in Multipath Data Streaming

Dmitry Korzun, Dmitriy Kuptsov, Andrei Gurtov

Department of Computer Science, Petrozavodsk State University
Helsinki Institute for Information Technology, Aalto University

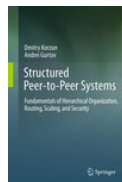
The study is partially supported by the Russian Fund for Basic Research
and the Ministry of Education and Science of the Russian Federation

UKSim-AMSS 9th IEEE European Modelling Symposium
on Mathematical Modelling and Computer Simulation
6–8 October, 2015, Madrid, Spain

Multipath Data Streaming

- A source schedules packets $1, 2, \dots, N$ to split the stream among the different paths $1, 2, \dots, M$, proportionally to their rates
- The rates are estimated based on such observable path characteristics as delay or bottleneck bandwidth
- Application domains:
 - ▶ Data transfer in the Internet: multi-homed hosts and TCP sessions
 - ▶ Wireless communication: multipath traffic in overlay networks
 - ▶ Load balancing: splitting a single flow across multiple network paths
 - ▶ Peer-to-Peer systems: many neighbors to forward a packet

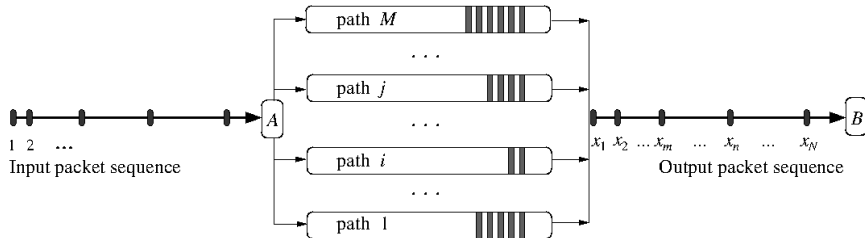
- D. Korzun and A. Gurtov.
Structured Peer-to-Peer Systems:
Fundamentals of Hierarchical Organization,
Routing, Scaling, and Security.
Springer, 2013



Packet Reordering: Motivation

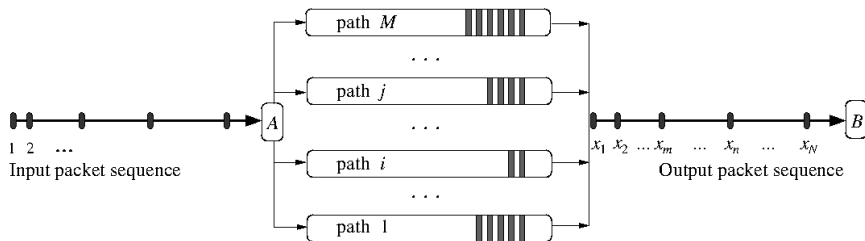
- Multipath data streaming $A \rightarrow B$
- Source A applies rate-proportional scheduling, either
 - ▶ deterministic (e.g., round-robin),
 - ▶ or randomized (e.g., Bernoulli scheme)
- Since network path characteristics are random there may be a large number of out-of-order packets at the destination
- Degradation of the application performance, especially when distant packets are reordered
- Destination B has to keep a large resequencing buffer for sorting incoming packets
- Randomized strategy at A may lead to improvements:
 - ▶ Variability of system state parameters are partially compensated by randomizing input parameters
 - ▶ The stochastic compensation effect

Packet Reordering: Problem Statement (1/2)



- Discrete uniform time $n = 1, 2, \dots, N$ to serve N -packet stream:
 - ▶ At time instance n the next packet is forwarded (asynchronous departure process)
 - ▶ Each packet n is instantaneously dispatched to a path i using a scheduler at A
- Source A always has data to send and the forwarding cost is negligible for all paths $i = 1, 2, \dots, M$

Packet Reordering: Problem Statement (2/2)



- Let $S_n^{(i)} > 0$ be the end-to-end delay of packet n in path i
- Destination B reassembles the sequence of packets
 - ▶ The n th position in the output sequence is occupied by the packet of input sequence number x_n

Assumptions

- No network loss, $S_n^{(i)} < \infty$
↪ the output is a permutation of the input sequence
- No bandwidth bottlenecks
↪ an arbitrary number of packets can be sent sequentially to path i without affecting the delay

- $\{S_n^{(i)}\}$ is i.i.d with generic element $S^{(i)}$
and the mean delay $\tau_i = E[S^{(i)}]$, $0 < \tau_i < \infty$,

$$p_i = \frac{\mu_i}{\mu_i + \dots + \mu_M}, \quad i = 1, \dots, M. \quad (1)$$

- No complete knowledge of path states (e.g., on-the-fly packets)
 - ▶ A can estimate μ_i
 - ▶ The case of parallel non-observable queues in queuing systems
- All M paths are order-preserving:
for any two packets following the same path if $x_n < x_m$ then $n < m$
(i.e., FCFS discipline)

Metrics of Packet Reordering (1/2)

- Let $1 \leq n < m \leq N$ be packet positions in the output sequence
- Set $r_{nm} = m - n$ if $x_n > x_m$ and $r_{nm} = 0$ otherwise
- The reorder distance probability: an arbitrary output packet n is reordered on distance k with probability

$$d_k = \mathbb{P}[r_{n,n+k} > 0]. \quad (2)$$

- Let $r_n = n - x_n$ be displacement of packet n from its original position. The displacement probability distribution:

$$f_k = \mathbb{P}[r_n = k], \quad -N < k < N, \quad (3)$$

i.e., f_k shows the frequency of packets of displacement k

Metrics of Packet Reordering (2/2)

- For each given n , consider the last reordered packet m . Then the maximum reorder distance is on average

$$\rho_{\max}(N) = \frac{1}{N} \sum_{n=1}^N \max_{n < m} r_{nm}.$$

Estimation of resequencing buffer size needed at the destination

- Reorder entropy: the total disorder of a packet sequence,

$$\rho_{\text{ent}}(N) = - \sum_{f_k > 0} f_k \ln f_k.$$

- Concentrated (low ρ_{\max}) or dispersed (high ρ_{\max}) displacement
 - ▶ If no reordering ($f_0 = 1$), then the minimum $\rho_{\text{ent}} = 0$ is achieved
 - ▶ The maximum $\rho_{\text{ent}} = \ln(2N + 1)$ is for uniformly displaced packets

Scheduling Strategies: Deterministic vs. Randomized

- WRR scheduler operates in loop $i = 1, 2, \dots, M$
 - ▶ Each iteration assigns a batch of subsequent packets to path i
 - ▶ The number of packets in a batch for path i is fixed to Cp_i , where C is a constant common for all paths
 - ▶ Lengthy packet batches are constructed for fast paths when there are slow paths
- BN scheduler forwards any packet n to path i with probability p_i
 - ▶ Packet batches of variable length appear
 - ▶ If $p_i = p = 1/M$ for any path i then batches are assigned in accordance with the geometric distribution
 - ▶ Batch of l successive packets $n, n + 1, \dots, n + l - 1$ has the probability $(1 - p)p^l$

Simulation Model: Path Delay Distributions

- *Exponential*: PDF $f_i(x) = \lambda_i e^{-\lambda_i x}$ has mean $\tau_i = 1/\lambda_i$ and variance $\sigma_i^2 = 1/\lambda_i^2$
- *Power-law*: PDF $f_i(x) = \frac{\alpha_i - 1}{s_i^{\min}} \left(\frac{x}{s_i^{\min}} \right)^{-\alpha_i}$ has the minimal value s_i^{\min}
 - ▶ If $\alpha_i \leq 2$ the mean is infinite, otherwise $\tau_i = \frac{\alpha_i - 1}{\alpha_i - 2} s_i^{\min}$
 - ▶ If $\alpha_i \leq 3$ the variance is infinite, otherwise $\sigma_i^2 = \frac{\alpha_i - 1}{\alpha_i - 3} (s_i^{\min})^2$
 - ▶ equivalent to Pareto distribution (by substitution $\alpha_{\text{prt}} = \alpha - 1$)
 - ▶ continuous counterpart of the Zipf-like distribution

Simulation Model: Experiments of Group I

The number of paths is varied as $M = 2, 3, \dots, 15$

- *Identical*: Every path has the same delay probability distribution parameters: $\forall i S^{(i)} = S$
- *Similar*: The delay probability distribution parameters are varied such that $\forall i S^{(i)} \approx S$
- *Slow & fast*: Given M paths are classified as slow and fast. If i is slow and j is fast then $\tau_i \gg \tau_j$.
The share of slow paths is $0 < q_{\text{slow}} < 1$

| Pattern | Exponential | | Power-law | | | |
|-----------|------------------------|---|------------------------------------|------------------|---------------------|-----------------------|
| | $\lambda \sim U[a, b]$ | $\tau_{\text{avg}} = \sigma_{\text{avg}}$ | $\alpha_{\text{prt}} \sim U[a, b]$ | s_{min} | τ_{avg} | σ_{avg} |
| Identical | 0.3 | 3.3 | 2.3 | 10 | 17.7 | 21.3 |
| Similar | 0.3, 0.4 | 2.9 | 2, 3 | 10, 20 | 25 | 22.4 |
| Slow | 0.1, 0.12 | 9 | 2, 2.3 | 10, 30 | 37.4 | 65.8 |
| Fast | 0.3, 0.4 | 2.9 | 3.3, 3.4 | 1, 5 | 4.3 | 2 |

Simulation Model: Experiments of Group II

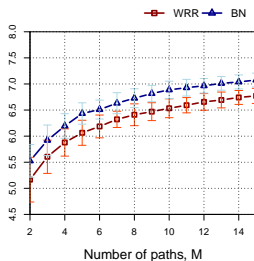
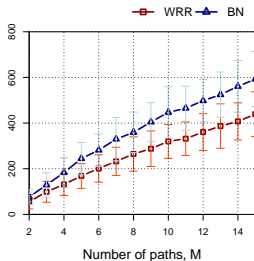
The number of paths is fixed to $M = 2$

- Fixed path: delay distribution parameters are fixed
- Varied path: parameters are varied reducing the mean delay

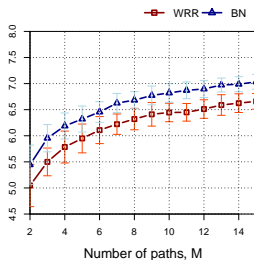
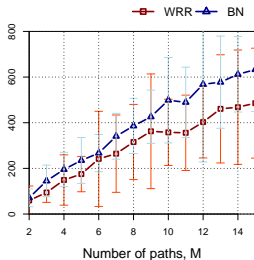
| | Exponential | | Power-law, $s^{\min} \sim U[2, 3]$ | | |
|-------------|------------------------|---|------------------------------------|---------------------|-----------------------|
| | $\lambda \sim U[a, b]$ | $\tau_{\text{avg}} = \sigma_{\text{avg}}$ | $\alpha_{\text{prt}} \sim U[a, b]$ | τ_{avg} | σ_{avg} |
| Fixed path | 1.00, 2.00 | 0.67 | 2.5, 2.7 | 4.06 | 3.25 |
| Varied path | 0.01, 0.02 | 66.67 | 1.5, 1.8 | 6.35 | ∞ |
| | 0.03, 0.05 | 25.00 | 2.0, 2.3 | 4.67 | 8.23 |
| | 0.10, 0.20 | 6.67 | 2.5, 2.8 | 4.02 | 3.06 |
| | 0.30, 0.50 | 2.50 | 3.0, 3.3 | 3.66 | 1.92 |
| | 1.00, 2.00 | 0.67 | 3.5, 3.8 | 3.44 | 1.41 |
| | 2.50, 3.00 | 0.36 | 4.0, 4.3 | 3.29 | 1.10 |
| | 3.00, 3.50 | 0.31 | 4.5, 4.8 | 3.18 | 0.91 |
| | 3.50, 4.00 | 0.27 | 5.0, 5.3 | 3.10 | 0.77 |

Identical paths in streaming $N = 10^4$ packets

■ Exponential delay

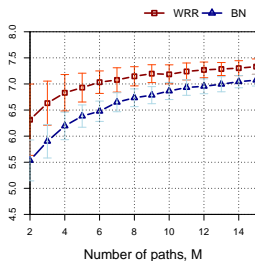
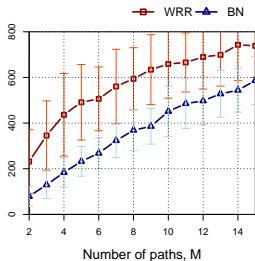


■ Power-law delay

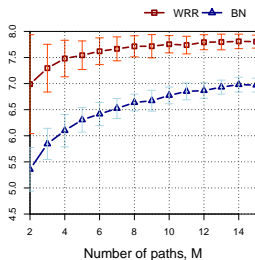
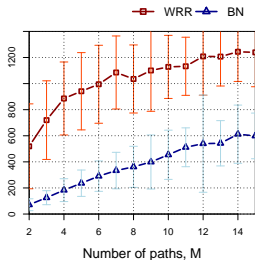


Similar paths for $N = 10^4$

■ Exponential delay

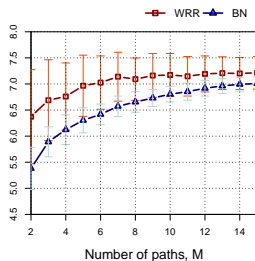
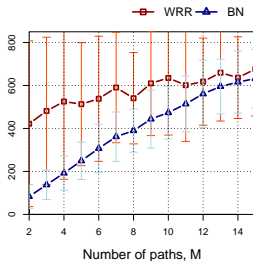


■ Power-law delay

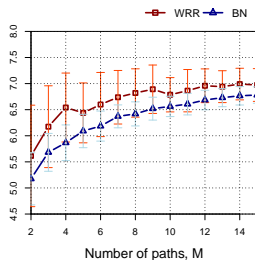
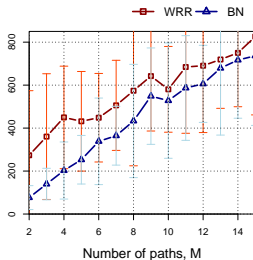


Slow & fast paths for $N = 10^4$ and $q_{\text{slow}} = 75\%$

■ Exponential delay

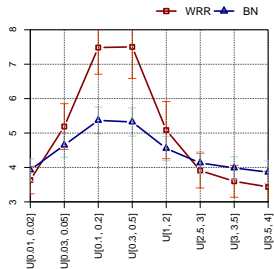
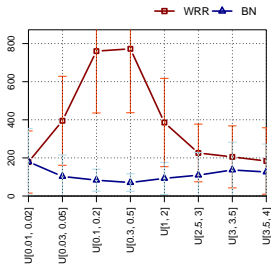


■ Power-law delay

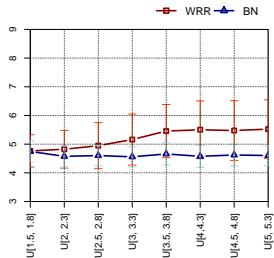
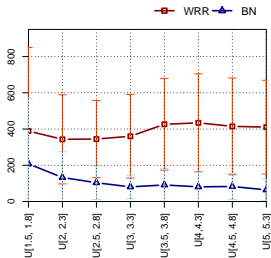


One path is fixed and the other is varied

■ Varied exponential delay



■ Varied power-law delay



Conclusion

- We compared weighted round-robin scheduler (deterministic) and Bernoulli scheduler (randomized) using simulation experiments
- Our hypothesis: randomness of transmission delay can be compensated with randomizing the packet scheduling
- Our results show that the randomization reduces packet reordering, and the stochastic compensation effect has a place for certain multipath configurations
- We expect that utilization of this effect in networking applications can essentially improve the application performance when the path diversity of underlying networks is high

Thank you

E-mail: dkorzun@cs.karelia.ru