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Future High-Performance Networks



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PRESENTED BY

Managing the data tsunami on future generation
communication networks



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Outline

- General background – Why care?
- Research challenges – What's the problem?
- High speed data center networks – How to fix it?
- Self-learning networks – Why do we need them?
- Long-term Research Path – Where are we going?
- Conclusion

15 Second Research Philosophy

My Two Rules of Data Center Network Design:

#1 Avoid moving data whenever
possible

#2 If you must move data do it as
efficiently as possible

Background

- High performance networks (HPNs) and Cloud
 - Cost – HPC is a small market (\$s)
 - Scale – HPC is a small market (volume)
 - Viability – HPC is a small market (risk)
- HPNs can only keep up if they also help Clouds
 - Clouds starting to need lower latency
 - Higher bandwidth always helps
 - Clouds catching up here already
 - Network tuning problem for HPC and Cloud

Background

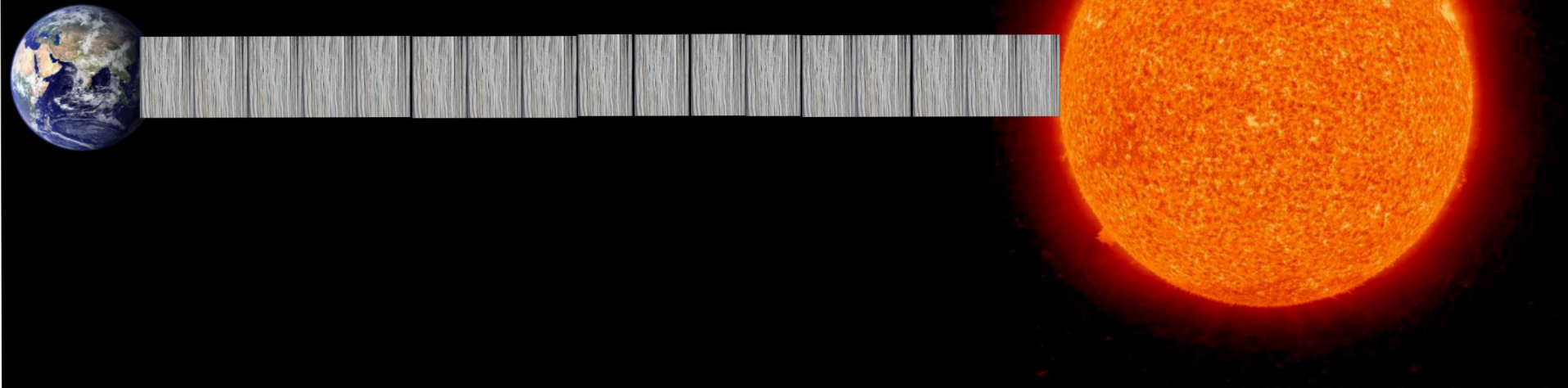
- From Cisco¹:
 - Annual global IP traffic will reach 3.3 ZB by 2021
 - Global IP traffic 3X by 2021 (127X 2005)
 - Smartphone traffic will exceed PC traffic by 2021
- Where does all this data go?
 - Data centers
- Data centers are the hotspots of the internet
 - HPC centers have same problem (CERN)

¹Cisco whitepaper: Cisco Visual Networking Index: Forecast and Methodology, 2016–2021 June 6, 2017

Just how much data?

- If you printed text files:

To the sun 15 times

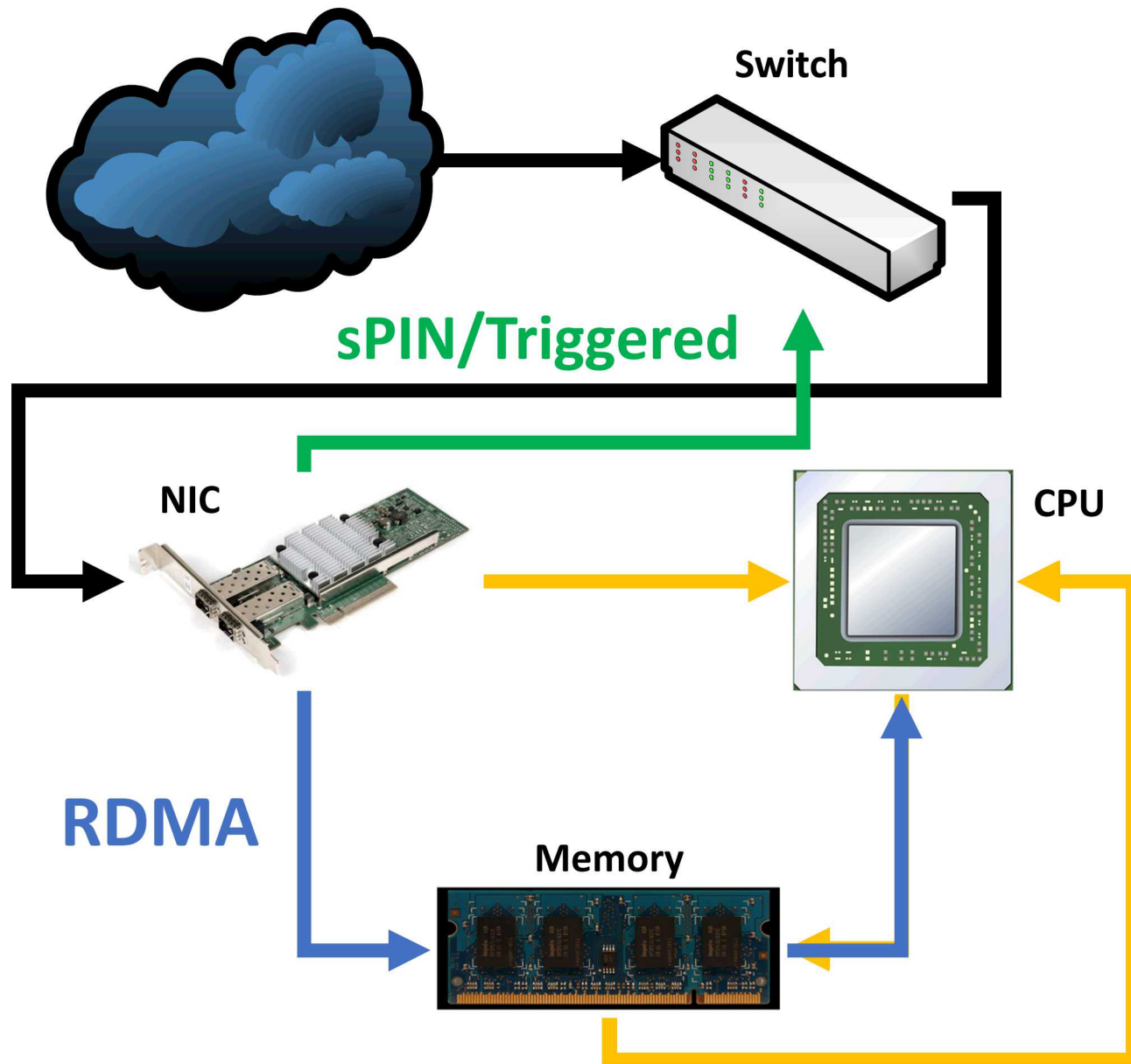


How do we handle all this data?

- Data center processing adds even more data to local network
- Key concerns: bandwidth and latency
- Leading edge – high performance networks
 - Mostly for scientific computing
 - Expensive
 - Only a few systems ever use them
 - Lucky if you can sell more than 10 big systems

What makes HPNs fast?

- Not like a normal network interface (NIC)
- Can write data directly to memory
 - No CPU or OS involvement
 - No copying (zero copy)
 - Called “OS Bypass” or user-level NICs
- Handles some message processing
 - Checksums (correctness)
 - Matching (data steering)



Traditional
Processing

What makes HPNs fast?

- Switches provide minimal features
 - Fast – switching times in nanoseconds
- Efficient send-side
 - Command queues are fast
 - Addresses are known in advance
 - User-level drivers
 - No kernel level delays

Why not HPNs everywhere?

- Expensive
- Compute is cheap
- Sockets code can be slow
 - Negates the benefit of an HPN
- No need yet (but it's coming)
 - Not compelling business case for all uses
 - Latency is still acceptable
 - Consumer devices don't need too much bandwidth

So why do we care?

- Cloud will start dictating HPN design!
- Wireless 5G and 6G driving up demand
 - Latency way down, bandwidth up
- Machine learning everything
 - Unacceptable latency?
 - Alexa – wait a couple seconds - annoying
 - Self-driving car – deadly
 - Humans can wait a long time
 - Other computers cannot

What's so difficult? Make it faster

- Sockets
 - Legacy – super easy to code
 - Everything coded for it
- But...An onload model design
 - Onload – CPU does it
- Latency needs means we need offload
 - Offload – NIC does it
- Fundamentally different designs

Making it faster

- Design an offload model
 - Easy to onload something designed for offload
 - Not easy to offload and onload design
- TCP offload engines are complicated
 - Also expensive
- Can we do better?
 - Don't re-write code!

Research Challenges

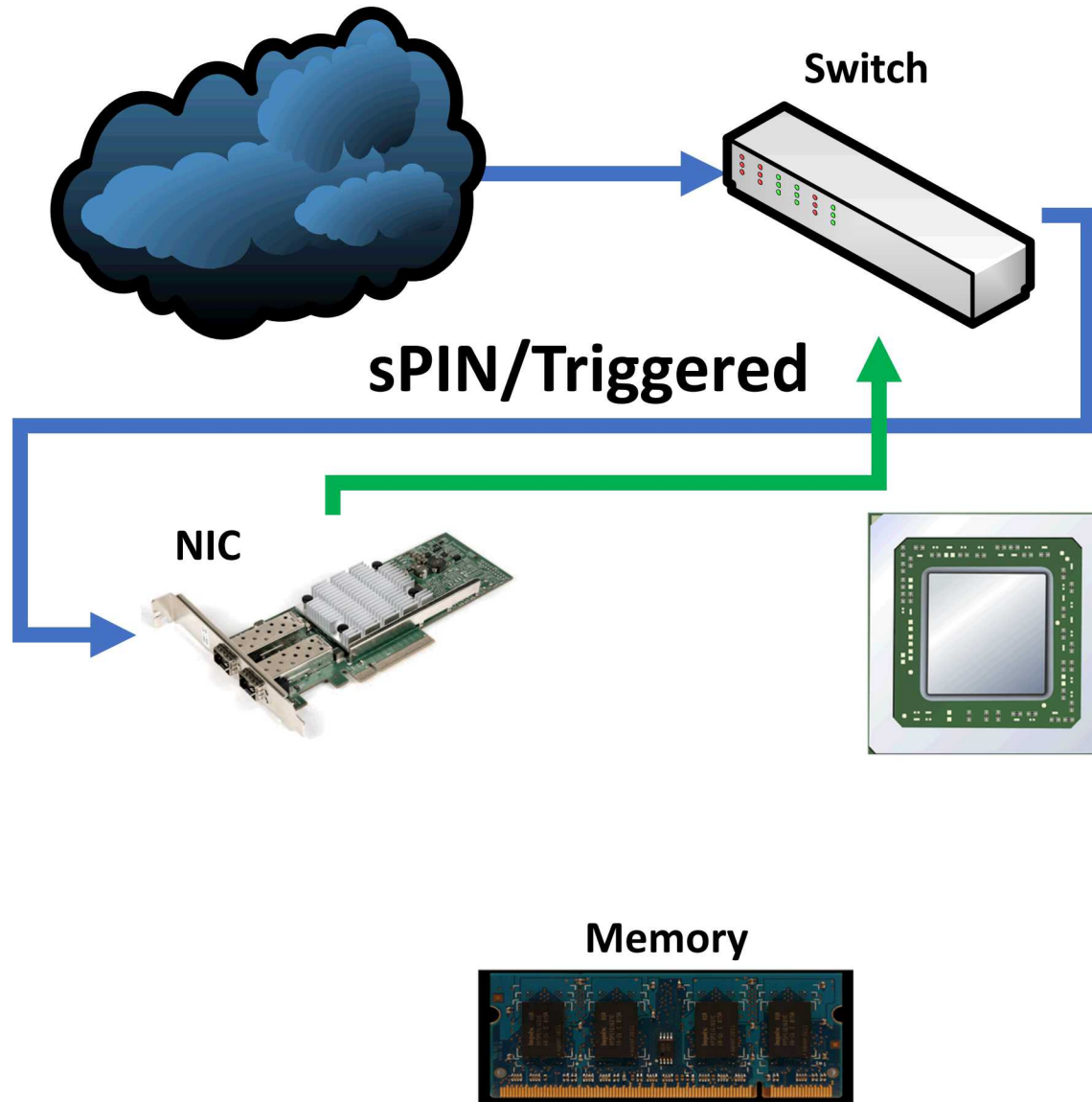
- My research is to make HPNs:
 - Useful – compatible with sockets
 - Fast – best exotic networks
 - Reliable – best off the shelf hardware
 - Flexible – software defined networking flexible
 - Adaptable – adaptive to conditions
 - Deployable – not just in data centers
- We can learn from what Cloud does better
 - Clouds are reliable...



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Avoid moving data

- Challenges in HPC and data centers – Power/Energy
- Moving data is expensive
 - In time and energy
- Orders of magnitude more energy to move from component to component
- Avoid copies
- Work on data where it is



My Two Rules of Data Center Network Design:

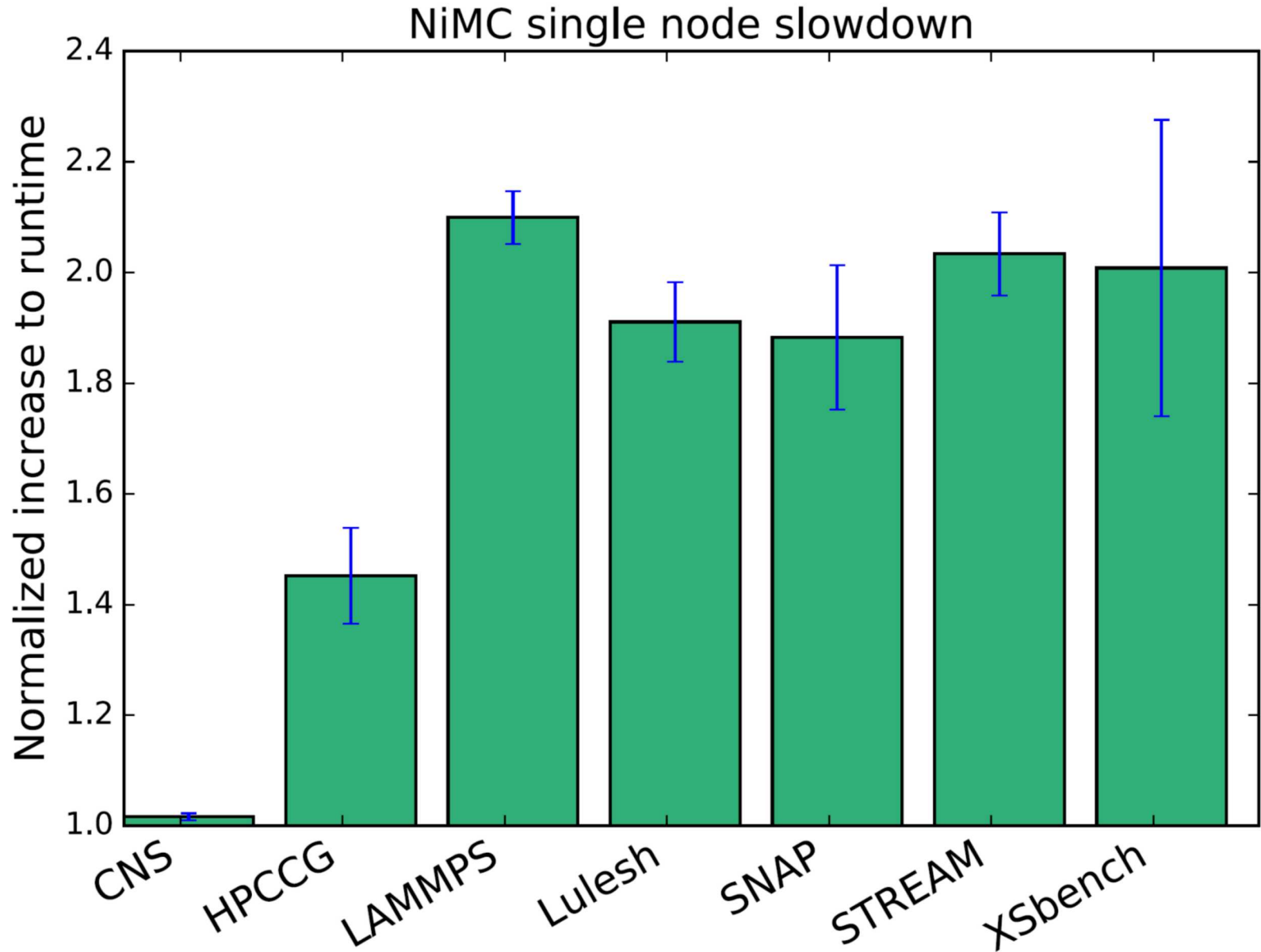
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Research Challenges

- Fast and Useful:
 - RDMA – Direct Memory Access
 - Competes with applications
- Hypothesis:
- Contention for memory resources should be observable and significant
- Corollary:
- If contention exists, we can avoid it

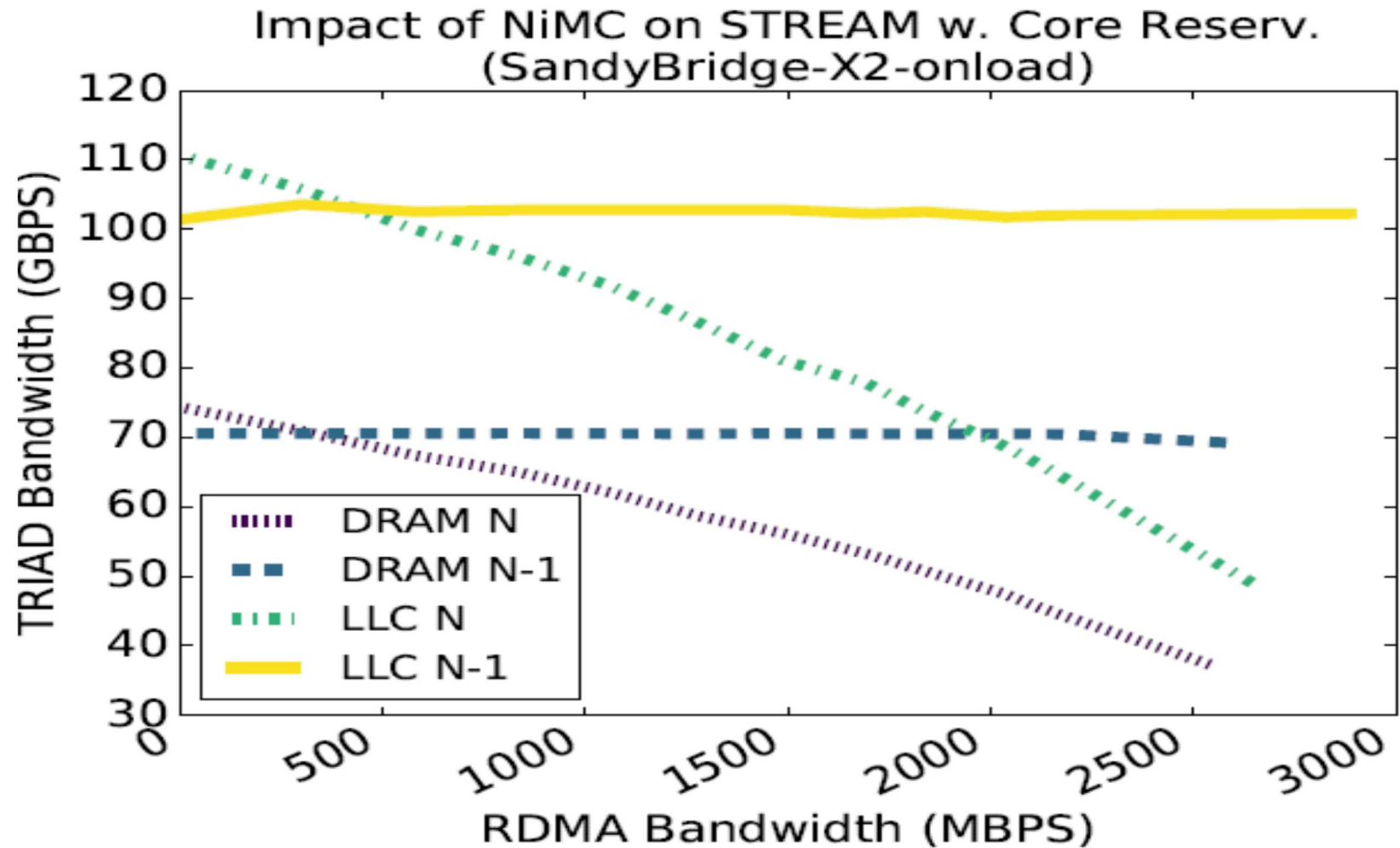
Network-induced Memory Contention



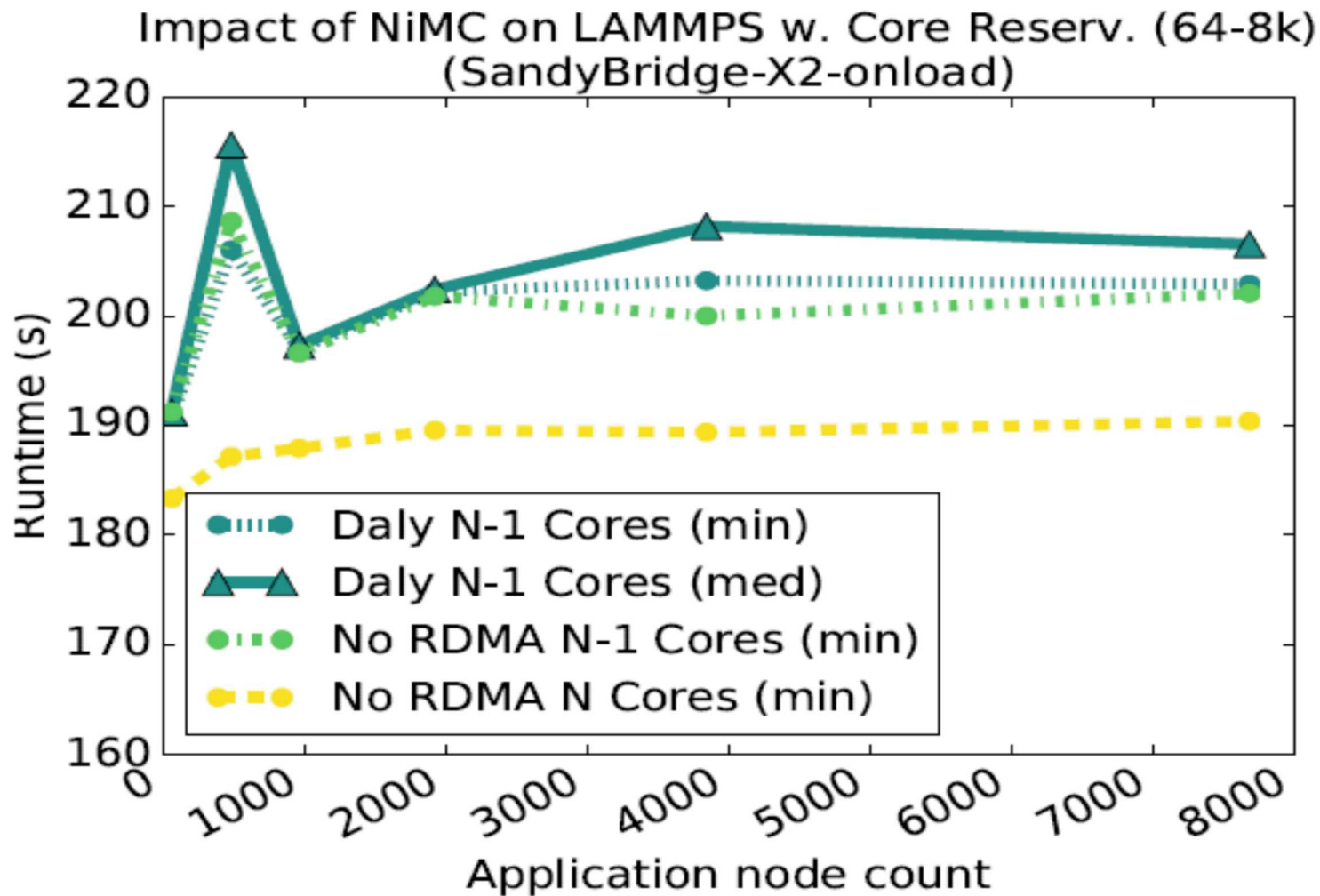
NiMC Problem

- NiMC is a problem
- Can we detect it?
 - Groves, Grant, Arnold, “NiMC: Characterizing and Eliminating Network-Induced Memory Contention”, IPDPS 2016
- What can we do if we can detect it?
 - Groves, Grant, Gonzales, Arnold, “Unraveling Network-induced Memory Contention: Deeper Insights with Machine Learning” TPDS Vol 29 Iss 8

More NiMC Data



Core reservation solution



Machine Learning Results

- Online data collection is limited to performance monitoring counters
 - Can only read 3-4 at a time on most CPUs
- Choosing the right 3 characteristics
 - Can identify NiMC 99.5 times out of 100
- Worst case scenario
 - False positives
 - Unnecessary slowdowns
 - Solution – slow stream first, then allocate new resources

Vision of the Future

- sPIN: Need non-deadline based methods too
 - Leverage/extend existing hardware
 - Break deadline – allow rescheduling
- NiMC: The RDMA model is part of the problem
 - RDMA is the local memory subsystem model
 - Issues with knowing when operations complete
 - Reserved resources per peer

Research Vision

- Self-learning NICs
 - Use triggered operations
 - Current work: Turing complete
 - Speed grows with line rate
 - Eviction of programs
 - Reschedule – feed back into pipeline
 - No longer dependent on remote packets

Research Vision

- Breaking dependency on streams
 - Use spare cycles to “learn”
 - Adjust network parameters
 - Yield whenever work to do
 - NICs are idle 90%+ of the time
 - When doing compute intensive work
 - Min. speed: 1 operation per packet
 - SERDES rates mean 3+GHz speeds

Research Vision

- Accelerate Machine Learning
 - Prep data
 - Work on data
- Use simple ML Techniques
 - Good area for collaboration
- Move techniques out to the compute
 - Rule #1

RDMA-next

- Re-design RDMA
 - Memory model -> Operations model
 - Know how much data to expect
 - Build in knowledge
 - Don't know how much data?
 - Build in buffer data thresholds
 - Abstract away specific resource allocation
 - No more reservations required

Summary

- #1 Avoid moving data whenever possible
 - sPIN today
 - Self-learning NICs tomorrow
- #2 If you must move data do it as efficiently as possible
 - NiMC and machine learning today
 - RDMA-next tomorrow



Thank you



Questions?