AVESTIMAN MERCHANNER

Many-Task Computing on Grids, Clouds, and Supercomputers

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> Computer Science, Illinois Institute of Technology January 22nd, 2010

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 - Argonne Leadership Computing Facility





Alok Choudhary

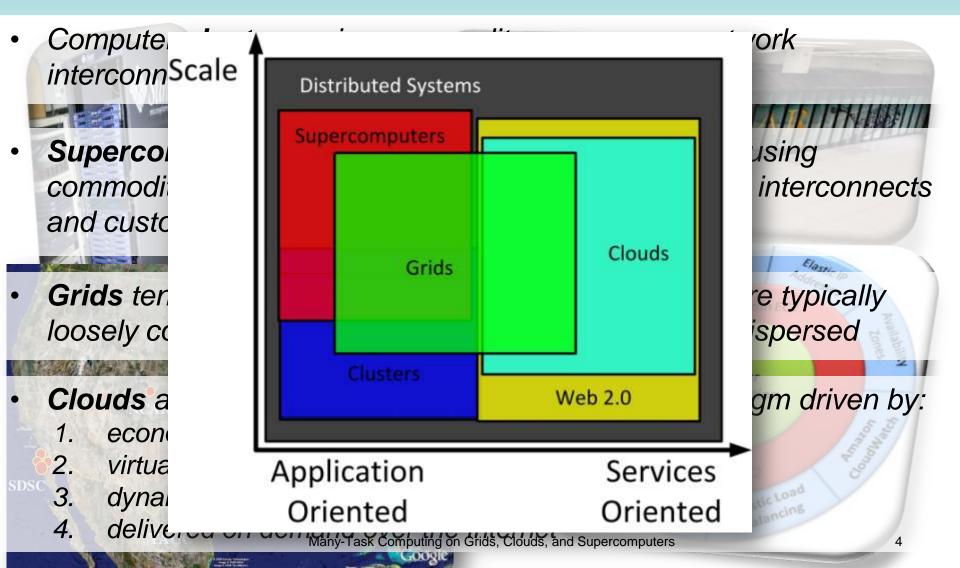
Ian Foster

- NASA: Ames Research Center, GSRP Program
- Over 60 Collaborators
 - Ian Foster (UC/ANL), Rick Stevens (UC/ANL), Alex Szalay (JHU), Jim Gray (MSR), Pete Beckman (ANL), Jerry Yan (NASA ARC), Mike Wilde (UC/ANL), Douglas Thain (ND), Amitabh Chaudhary (ND), Yong Zhao (MS), Zhao Zhang (UC), Catalin Dumitrescu (FNAL), Matei Ripeanu (UBC), Alok Choudhary (NU), and many more... 2



- Defining Many-Task Computing
- Motivation: Scalability Challenges
- Novel Resource Management Techniques
- Performance Evaluation: Micro-benchmarks
- Performance Evaluation: Applications
- Contributions
- Future Work

Clusters, Grids, Clouds, and Supercomputers



HPC, HTC, MTC

HPC: High-Performance Computing

- Synonymous with supercomputing
- Tightly-coupled applications
- Implemented using Message Passing Interface (MPI), needs low latency networks

Input

Hi

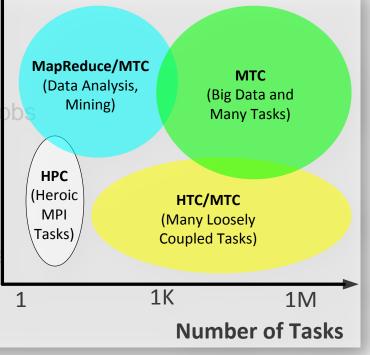
Measured in FLOPS

HTC: High-Throughput Computing Data Size

- Typically applied in clusters and grids
- Loosely-coupled applications with sequential Med
- Measured in operations per month or years

MTC: Many-Task Computing

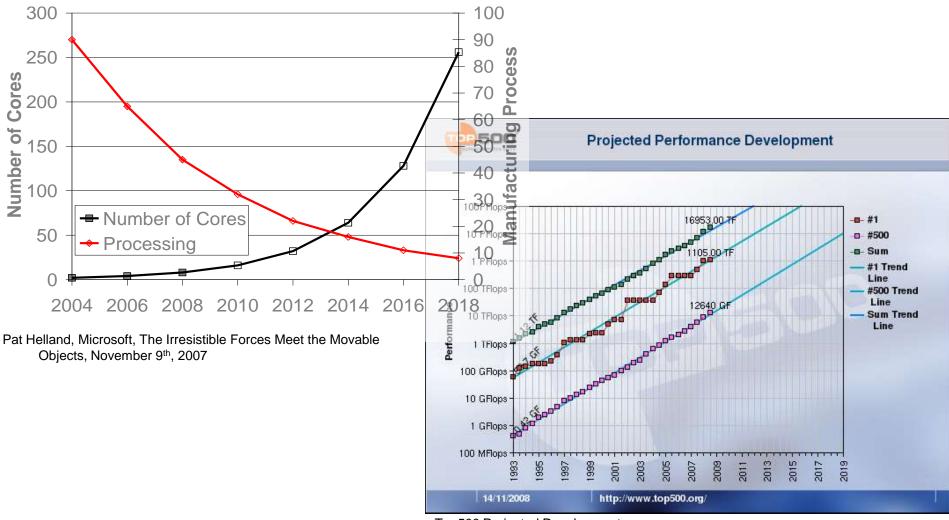
- Bridge the gap between HPC and HTC
- Applied in clusters, grids, and supercomputer
- Loosely coupled apps with HPC orientations
- Many activities coupled by file system ops
- Many resources over Mahousk time in period Sclouds, and Supercomputers [MTAGS08] "Many-Task Computing for Grids and Supercomputers"



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Projected Growth Trends

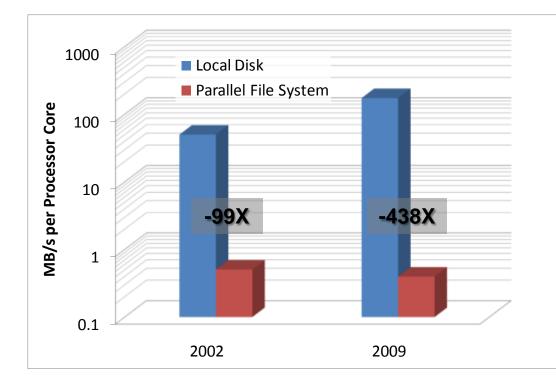


Top500 Projected Development,

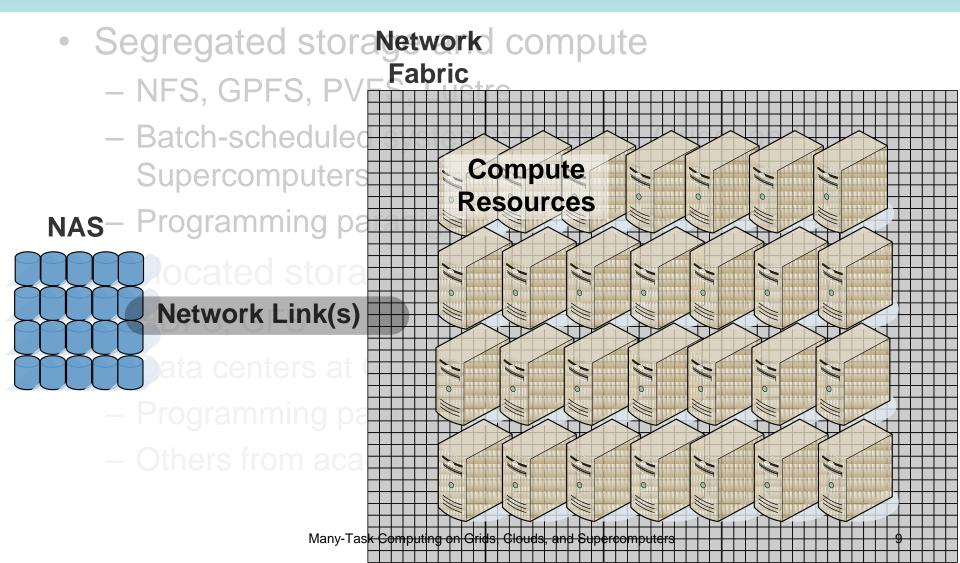
http://www.top500.org/lists/2008/11/performance_development

Growing Storage/Compute Gap

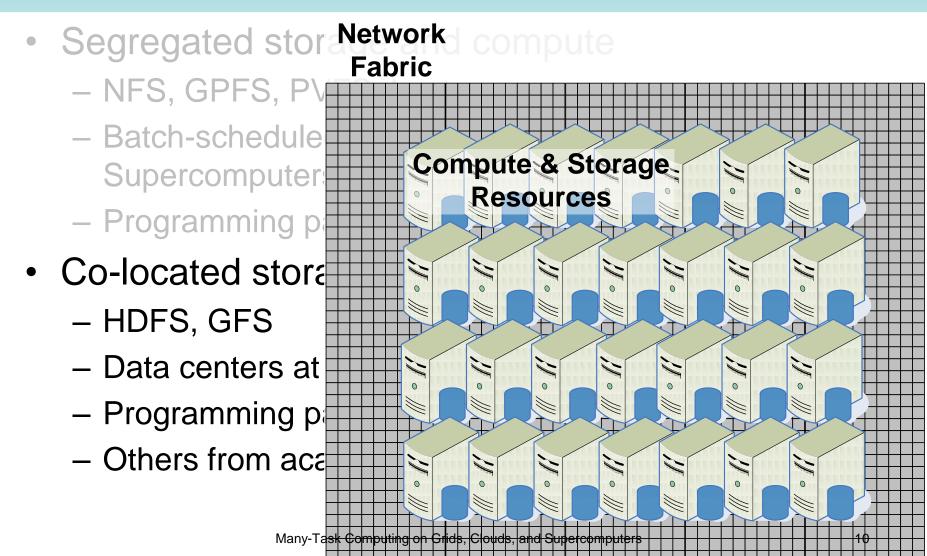
- 2002:
 - Local disk
 - ANL/UC TG Site (70GB SCSI)
 - Parallel File System
 - 2002: IBM Blue Gene/L (GPFS, 1GB/s)
- 2009:
 - Local disk
 - PADS (RAID-0, 6 drives 750GB SATA)
 - Parallel File System
 - IBM Blue Gene/P (GPFS, 65GB/s)



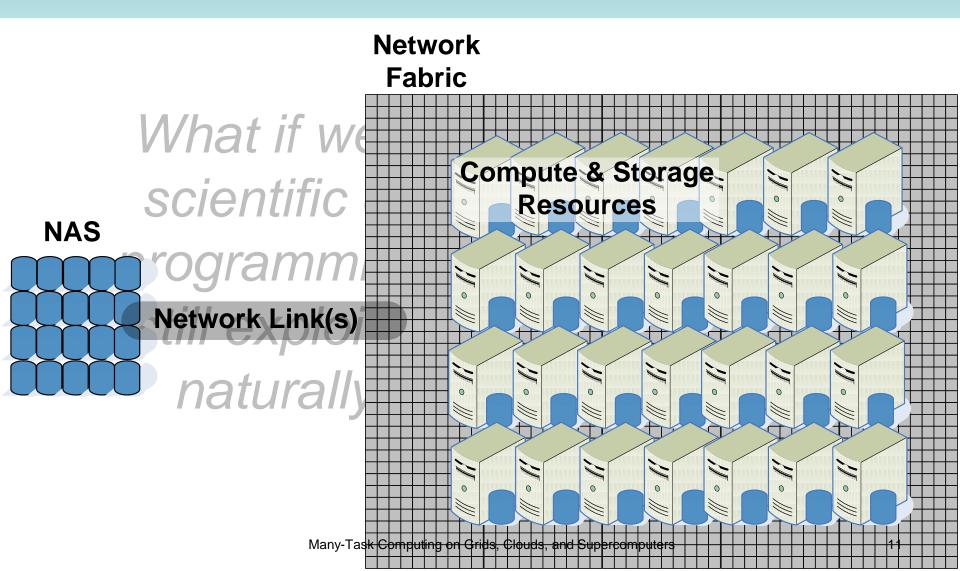
State of the Art: Storage Systems



State of the Art: Storage Systems



Combine State of the Art Systems



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Techniques to Support MTC

- Streamlined task dispatching
- Dynamic resource provisioning
 - Multi-level scheduling
 - Resources are acquired/released in response to demand
- Data diffusion
 - Data diffuses from archival storage to transient resources
 - Resource "caching" allows faster responses to subsequent requests
 - Co-locate data and computations to optimize performance

[[]HPDC09] "The Quest for Scalable Support of Data Intensive Workloads in Distributed Systems"

[[]DIDC09] "Towards Data Intensive Many-Task Computing"

[[]SC08] "Towards Loosely-Coupled Programming on Petascale Systems"

[[]DADC08] "Accelerating Large-scale Data Exploration through Data Diffusion"

[[]UC07] "Harnessing Grid Resources with Data-Centric Task Farms"

[[]SC07] "Falkon: a Fast and Light-weight tasK executiON framework"

[[]TG07] "Dynamic Resource Provisioning in Grid Environments"

Theoretical and Practical Exploration

- Abstract model
 - Models the efficiency and speedup of entire workloads
 - Captures techniques to support MTC
 - Streamlined task dispatching, dynamic resource provisioning, data diffusion
 - Lead to proof of O(NM) competitive caching
- Middleware to support MTC
 - Falkon: a fast a light-weight execution framework
 - Reference Implementation of the abstract model
 - Swift: A Parallel Programming System

Middleware Support: Falkon

- Goal: enable the rapid and efficient execution of many independent jobs on large compute clusters
- Combines three components:
 - a streamlined task dispatcher
 - resource provisioning through multi-level scheduling techniques
 User Task Dispatcher Data-Aware Scheduler

Wait Queue

Available Resources

(GRAM4)

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Execu

Provisioned Resources

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- data diffusion and dat co-located computational a
- Integration into Swift
 Provisioning
 - Applications cover man medicine, chemistry, e

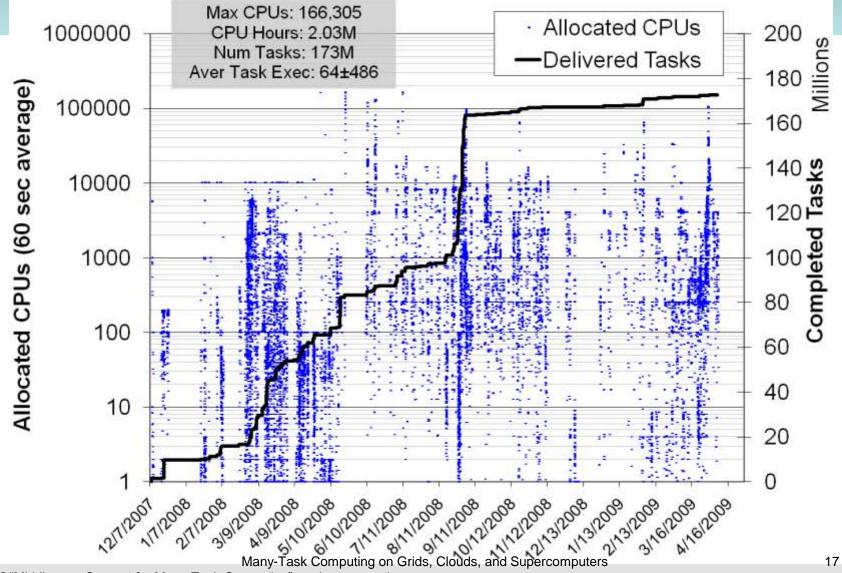
[SciDAC09] "Extreme-scale scripting: Opportunities for large task-parallel applications [SC08] "Towards Loosely-Coupled Programming on Petascale Systems" [Globus07] "Falkon: A Proposal for Project Globus Incubation" [SC07] "Falkon: a Fast and Light-weight tasK executiON framework" [SWF07] "Swift: Fast, Reliable, Loosely Coupled Parallel Computation"

Falkon Project

- Falkon is a real system
 - Late 2005: Initial prototype, AstroPortal
- January - Novem Workload http: 160K CPUs - Febry 1M tasks 60 sec per task Imple • 2 CPU years in 453 sec (~1K Throughput: 2312 tasks/sec some and times the soll of 85% efficiency Sou - Yong Znao, Zi

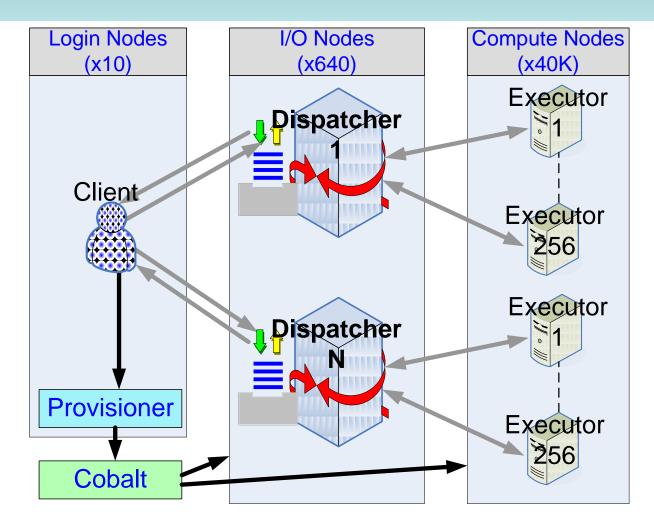
[Globus07] "Falkon: A Proposal for Project GlobusMaoybatistnComputing on Grids, Clouds, and Supercomputers [CLUSTER10] "Middleware Support for Many-Task Computing"

Falkon Activity History (16 months)



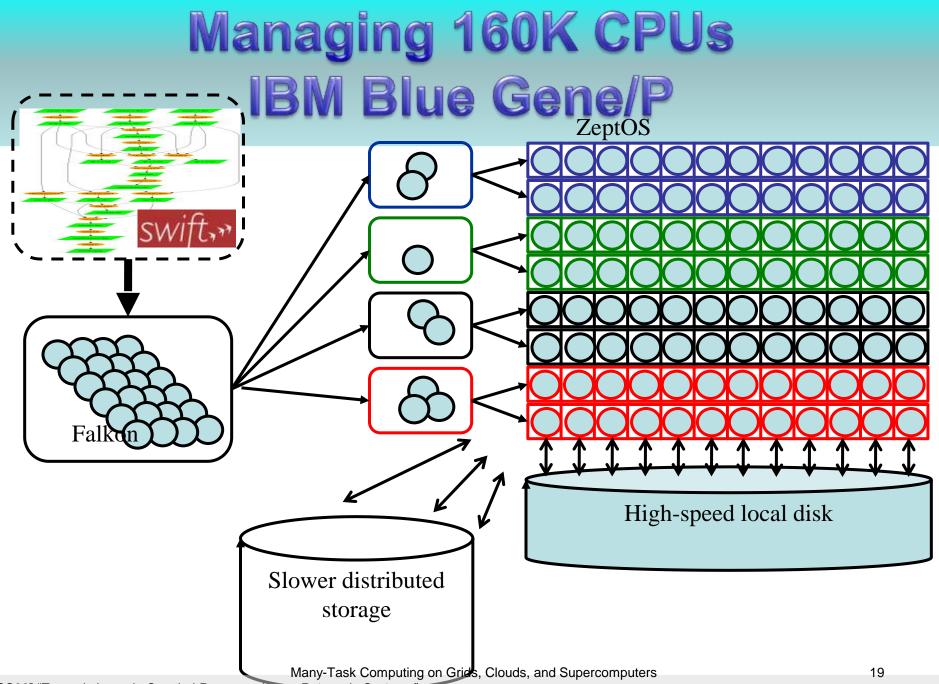
[TPDS09] "Middleware Support for Many-Task Computing", under preparation

Distributed Falkon Architecture



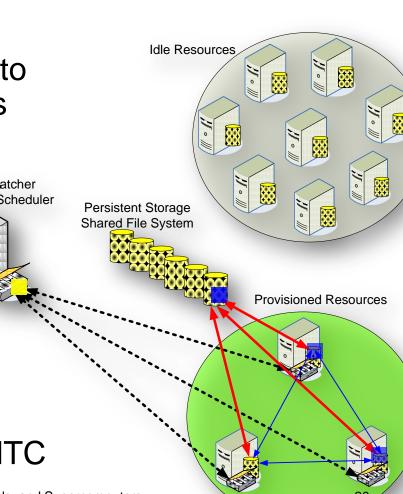
Many-Task Computing on Grids, Clouds, and Supercomputers

[SC08] "Towards Loosely-Coupled Programming on Petascale Systems"



Data Diffusion

- Resource acquired in response to demand
- Data diffuse from archival storage to newly acquired transient resources
- Resource "caching" allows faster responses to subsequent requests atcher Data-Aware Scheduler
- Resources are released when demand drops
- Optimizes performance by coscheduling data and computations
- Decrease dependency of a shared/parallel file systems
- Critical to support data intensive MTC



Many-Task Computing on Grids, Clouds, and Supercomputers

[DADC08] "Accelerating Large-scale Data Exploration through Data Diffusion"

Scheduling Policies

- FA: first-available

 simple load balancing
- MCH: max-cache-hit – maximize cache hits
- MCU: max-compute-util

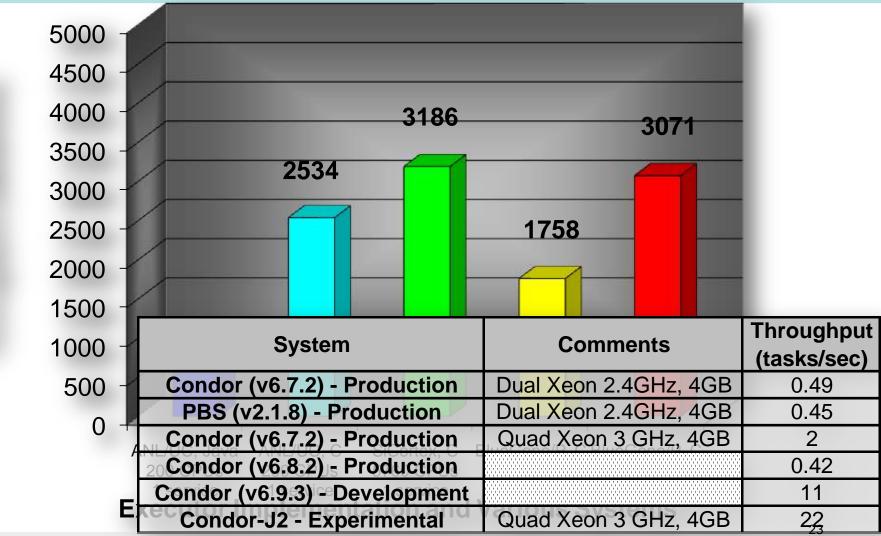
 maximize processor utilization
- GCC: good-cache-compute

 maximize both cache hit and processor utilization at the same time

Outline

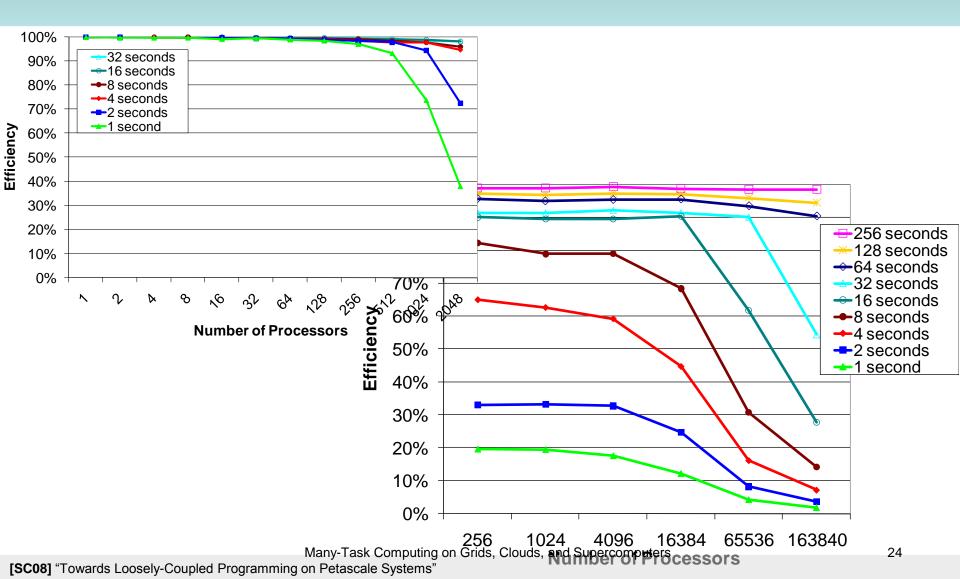
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Dispatch Throughput



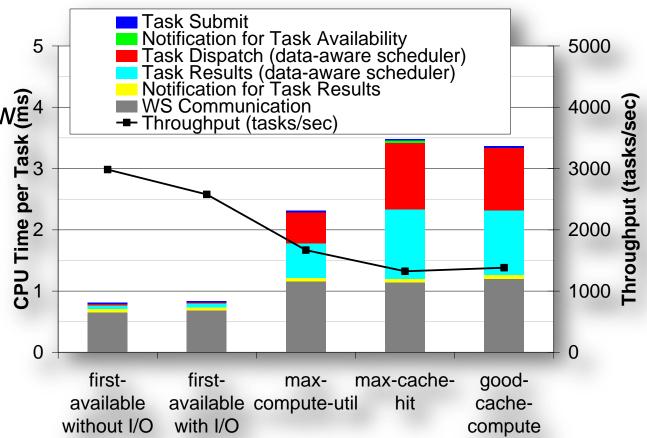
[SC08] "Towards Loosely-Coupled Programming on Petascale Systems"

Execution Efficiency



Data-Aware Scheduler Profiling

- 3GHz dual CPUs
- ANL/UC TG with 128 processors
- Scheduling window 2500 tasks
- Dataset
 - 100K files
 - 1 byte each
- Tasks
 - Read 1 file
 - Write 1 file



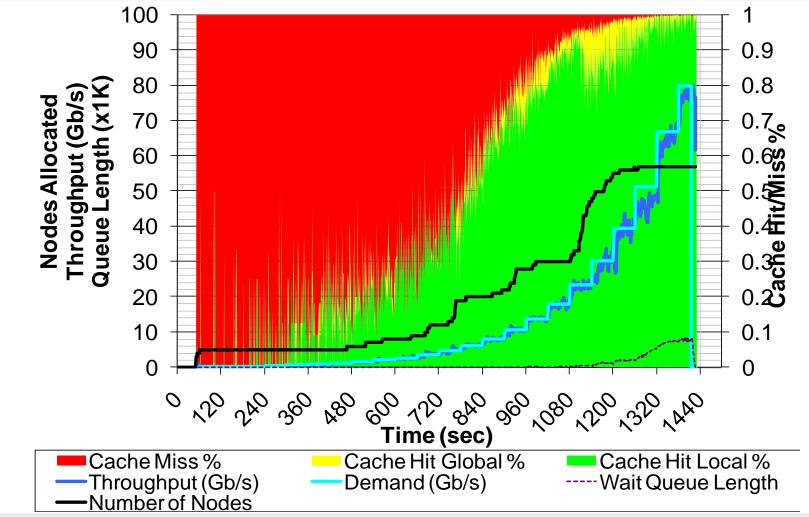
Synthetic Workloads

- Monotonically Increasing Workload
 - Emphasizes increasing loads
- Sine-Wave Workload

 Emphasizes varying loads
- All-Pairs Workload
 - Compare to best case model of active storage
- Image Stacking Workload (Astronomy)
 - Evaluate data diffusion on a real large-scale dataintensive application from astronomy domain

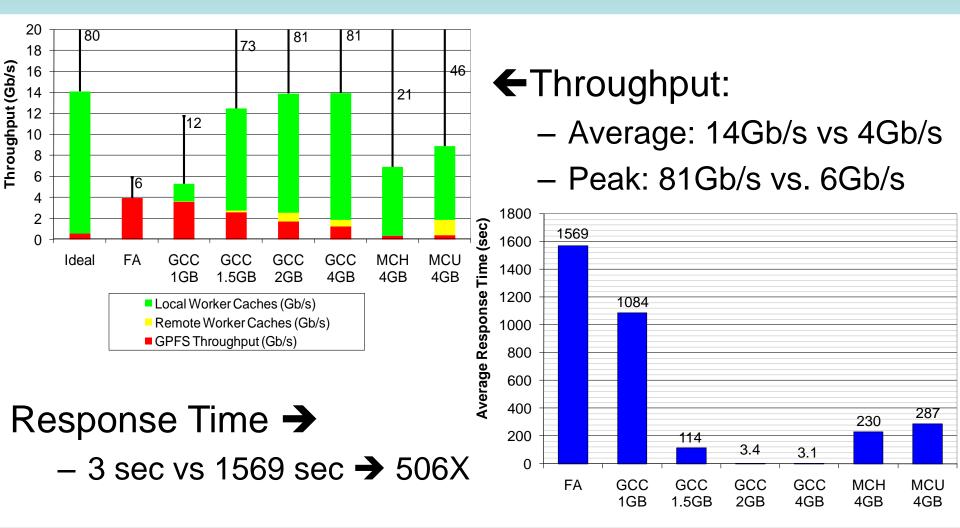
[[]DADC08] "Accelerating Large-scale Data Exploration through Data Diffusion" [HPDC09] "The Quest for Scalable Support of Data Intensive Applications in Distributed Systems" [DIDC09] "Towards Data Intensive Many-Task Computing"

Data Diffusion Monotonically Increasing Workload



[HPDC09] "The Quest for Scalable Support of Data Intensive Applications in Distributed Systems" **[DIDC09]** "Towards Data Intensive Many-Task Computing"

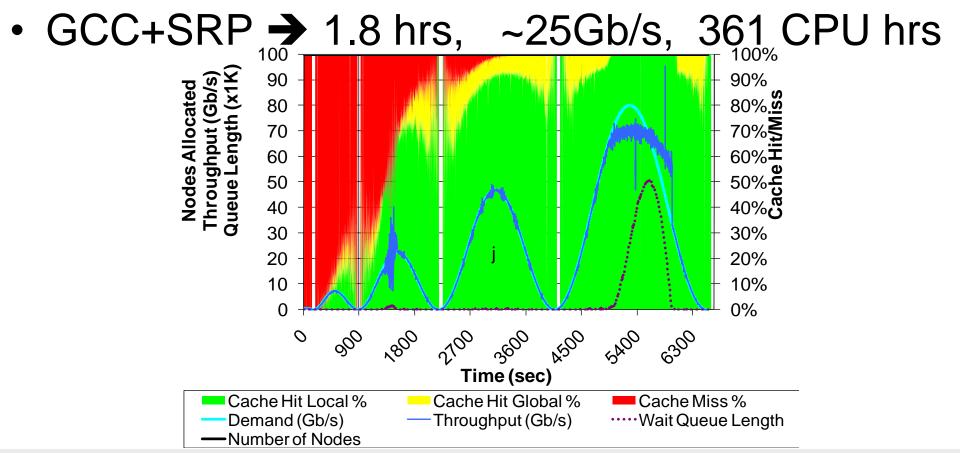
Data Diffusion Monotonically Increasing Workload



[HPDC09] "The Quest for Scalable Support of Data Intensive Applications in Distributed Systems" **[DIDC09]** "Towards Data Intensive Many-Task Computing"

Data Diffusion Sine-Wave Workload

• GPFS → 5.7 hrs, ~8Gb/s, 1138 CPU hrs



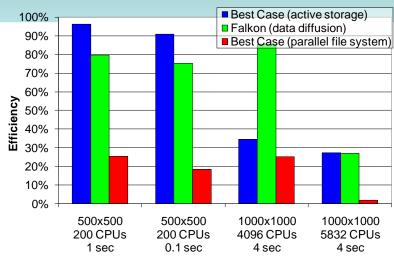
[HPDC09] "The Quest for Scalable Support of Data Intensive Applications in Distributed Systems", under review **[DIDC09]** "Towards Data Intensive Many-Task Computing", under review

Data Diffusion vs. Active Storage All-Pairs Workload

- Pull vs. Push
 - Data Diffusion
 - Pulls *task* working set
 - Incremental spanning forest
 - Active Storage:
 - Pushes workload working set to all nodes
 - Static spanning tree

Christopher Moretti, Douglas Thain, University of Notre Dame

[HPDC09] "The Quest for Scalable Support of Data Intensive Applications in Distributed **[DIDC09]** "Towards Data Intensive Many-Task Computing", under review



Experiment				
Experiment	Approach	Local Disk/Memory (GB)	Network (node-to-node) (GB)	Shared File System (GB)
500x500 200 CPUs 1 sec	Best Case (active storage)	6000	1536	12
	Falkon (data diffusion)	6000	1698	34
500x500 200 CPUs 0.1 sec	Best Case (active storage)	6000	1536	12
	Falkon (data diffusion)	6000	1528	62
1000x1000 4096 CPUs 4 sec	Best Case (active storage)	24000	12288	24
	Falkon (data diffusion)	24000	4676	384
1000x1000 5832 CPUs 4 sec	Best Case (active storage)	24000	12288	24
	^{ler re} \ #alk on (data diffusion)	24000	30 3867	906

Outline

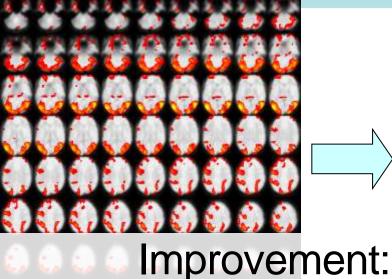
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Applications Medical Imaging: fMRI









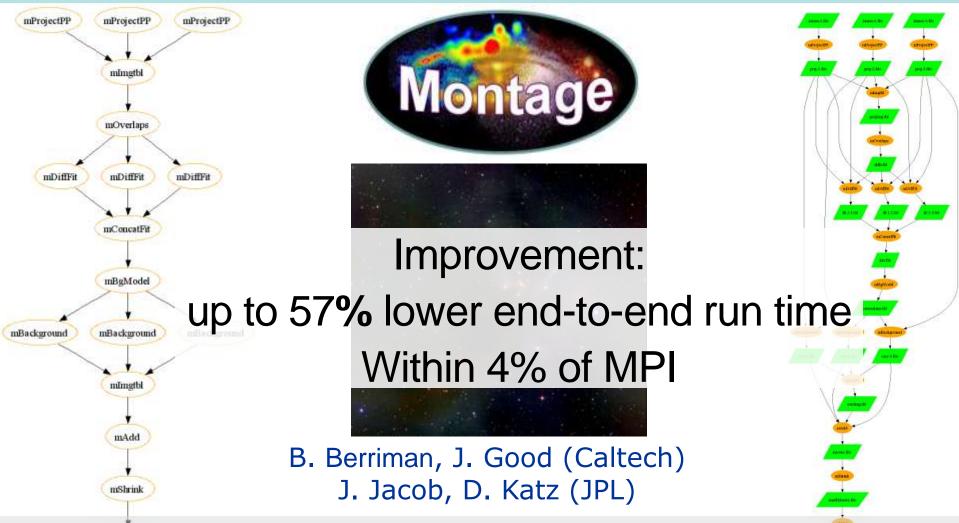
Wide up to 90% lower end-to-end run time

- Testing, interactive analysis, production runs
- Data mining

– Parameter studies

[SC07] "Falkon: a Fast and Light-weight tasK executiON framework" [SWF07] "Swift: Fast, Reliable, Loosely Coupled Parallel Computation"

Applications Astronomy: Montage



[SC07] "Falkon: a Fast and Light-weight tasK executiON framework" [SWF07] "Swift: Fast, Reliable, Loosely Coupled Parallel Computation"

Applications Molecular Dynamics: MolDyn

- Determination of free energies in aqueous solution
 - Antechamber coordinates
 - Charmm solution
 - Charmm free energy

Improvement:

up to 88% lower end-to-end run time

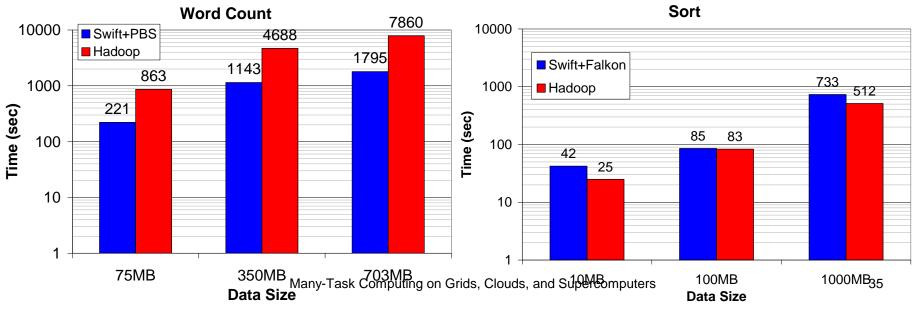
5X more scalable

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[NOVA08] "Realizing Fast, Scalable and Reliable Scientific Computations in Grid Environments"

Applications Word Count and Sort

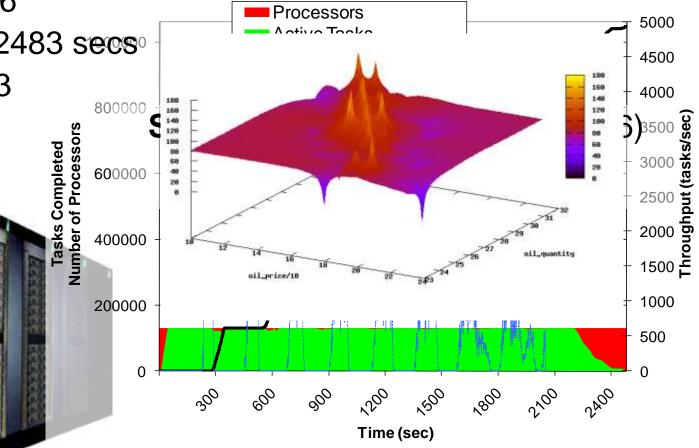
- Classic benchmarks for MapReduce
 - Word Count
 - Sort
- Swift and Falkon performs similar or better than Hadoop (on 32 processors)



Applications Economic Modeling: MARS

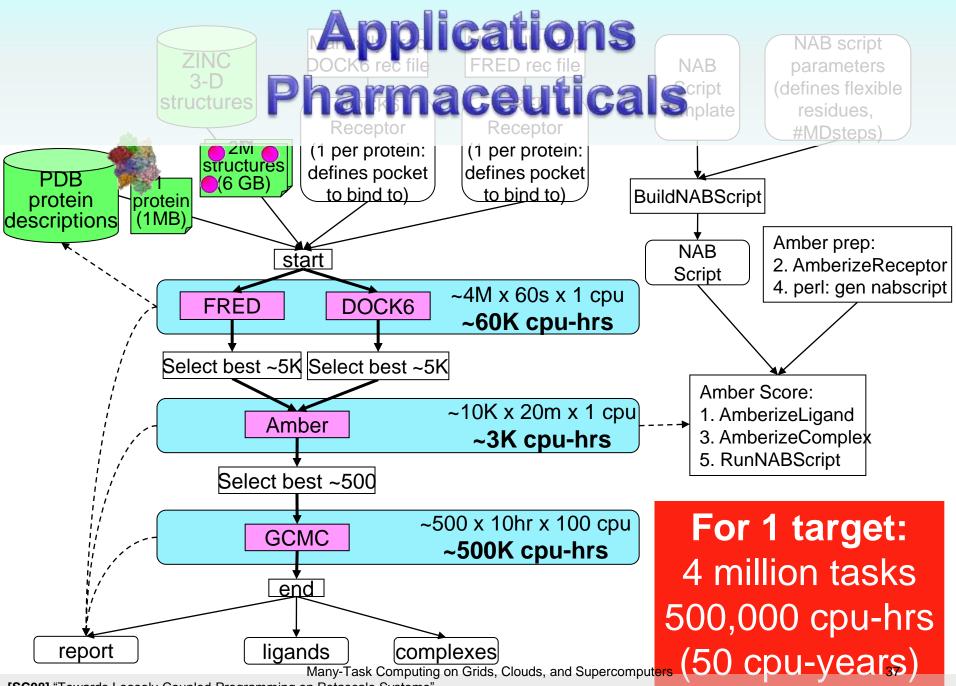
- CPU Cores: 130816
- Tasks: 1048576
- Elapsed time: 2483 secs
- CPU Years: 9.3

nd.



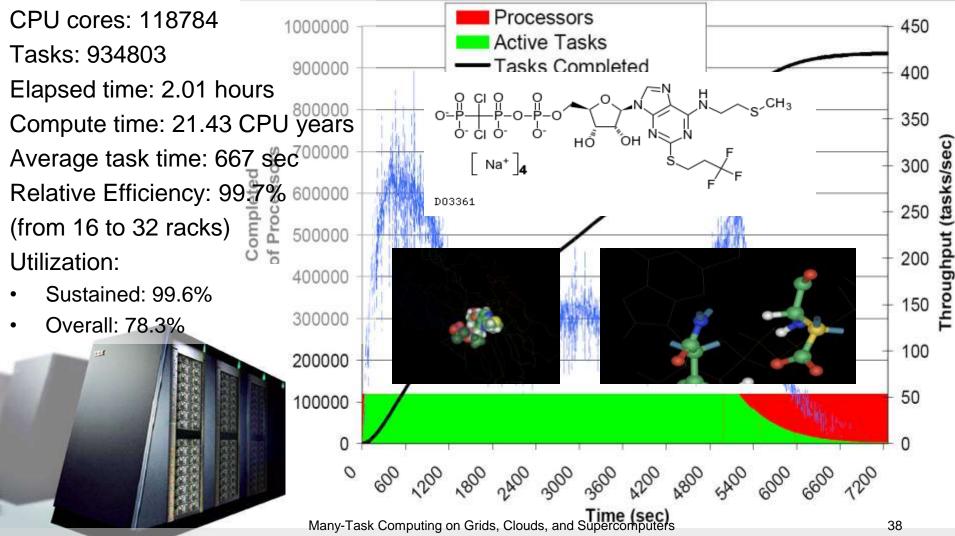
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Applications Pharmaceuticals: DOCK

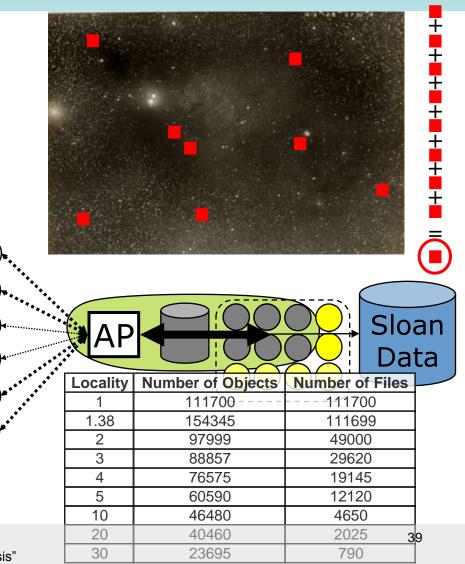


[SC08] "Towards Loosely-Coupled Programming on Petascale Systems"

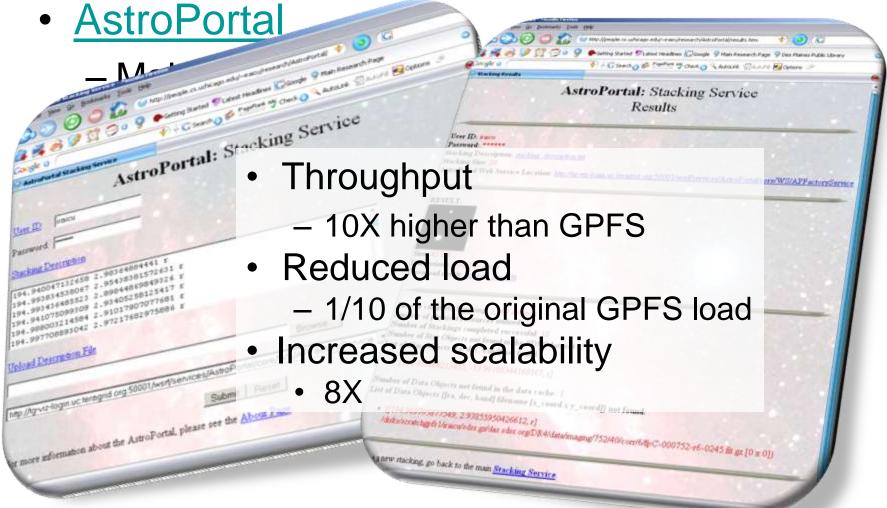
Applications Astronomy: AstroPortal

- Purpose
 - On-demand "stacks" of random locations within ~10TB dataset
- Challenge
 - Processing Costs:
 - O(100ms) per object
 - Data Intensive:
 - 40MB:1sec
 - Rapid access to 10-10K "random" files

— Time-varying load [DADC08] "Accelerating Large-scale Data Exploration through Data Diffusion" [TG06] "AstroPortal: A Science Gateway for Large-scale Astronomy Data Analysis"



Applications Astronomy: AstroPortal



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- There is more to HPC than tightly coupled MPI, and more to HTC than embarrassingly parallel long jobs
 - MTC: Many-Task Computing
 - Addressed real challenges in resource management in large scale distributed systems to enable MTC
 - Covered many domains (via Swift and Falkon): astronomy, medicine, chemistry, molecular dynamics, economic modelling, and data analytics
- Identified that data locality is critical at large-scale → data diffusion
 - Integrated streamlined task dispatching with data aware scheduling
 - Heuristics to maximize real world performance
 - Suitable for varying, data-intensive workloads
 - Proof of O(NM) Competitive Caching

Publications and Service

- Publications
 - 51 articles and proposals
 - 66 formal presentations
 - 854 citations → H-Index 14
- Activities for broader community engagement
 - ScienceCloud: <u>ACM Workshop on Scientific Cloud Computing</u>, 2010
 - TPDS: <u>IEEE Transactions on Parallel and Distributed Systems, Special Issue on</u> <u>Many-Task Computing, 2010</u>
 - MTAGS: <u>ACM Workshop on Many-Task Computing on Grids and</u> <u>Supercomputers, 2009</u>
 - MTAGS: IEEE Workshop on Many-Task Computing on Grids and Supercomputers, 2008
 - BegaJob: <u>Bird of Feather Session "How to Run One Million Jobs", at</u> <u>IEEE/ACM SC08, 2008</u>
 - 53 more activities as a program committee member and/or reviewer Many-Task Computing on Grids, Clouds, and Supercomputers

Teaching

- Introduction to Programming (Spring 2010)
 - Northwestern University, Instructor
- Hot Topics in Distributed Systems: Data-Intensive Computing (Winter 2010)
 - Northwest 9 more courses at University of Chicago and
 - <u>http://www</u> Wayne State University
- Networks an
 Advanced Network Design, Introduction to
 - University Programming for the World Wide Web I, Introduction to
 - <u>http://dsl.c</u> Frogramming for the Wond Wide Web I, Introduction to Computer
- Grid Computer Science 1 & 2, Introduction to Computer
 - University
 Systems , Fundamentals of Computer Programming I
 - http://www
 in Scheme, Problem Solving & Programming in C++,
- Introduction Data Structures & Abstraction in C++
 - Purdue University, Lab TA
- Data Structures and Algorithm Analysis in C++ (2002)
 - University of Michigan, Adjunct Assistant Professor

Mythbusting

- Embarrassingly Happily parallel apps are trivial to run
 - Logistical problems can be tremendous
- Loosely coupled apps do not require "supercomputers"
 - Total computational requirements can be enormous
 - Individual tasks mit be tightly coupled Workloads frequently involve large amounts of I/O
 - Makhaven'tefoundnthersolutionabyeting
 - Costs to run "supercomputers" per FLOP is among the hontymous
- Loosely coupled apps do not require specialized system software
 - Their requirements on the job submission and storage systems can be extremely large
- Shared/parallel file systems are good for all applications
 - They don't scale proportionally with the compute resources
 - Data intensive applications don't perform and scale well
 - Growing compute/storage gap

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Future Work

- Many-Task Computing and Data Intensive Computing
 - Develop theoretical and practical aspects of building efficient and scalable support for MTC
 - Build a new distributed data-aware execution fabric that will support HPC, MTC, and HTC
 - Support interactive HPC applications
- Cloud Computing
 - Enable scientific computing on clouds (e.g. Magellan project)
 - Enable HPC through novel networking
- Many-Core Computing
 - Apply data-aware scheduling of jobs from distributed systems
 - \rightarrow OS scheduling of threads
 - Parallel programming models
 - → automatic parallelization using DAG-based data-flow techniques,

Many-Task Computing on Grids, Clouds, and Supercomputers

Future Work (cont)

- Maintain and expand collaborations
 - Government Labs: ANL, FNAL, NASA, LBNL,
 - Academia: UC, CI, UIC, DePaul, NU, ND, Purdue, UIUC, IU, UWM, WSU, LSU, UNC, USF, UBC, DU, MU
 - Industry: Accenture, IBM, MS, Google, Yahoo, Amazon, Intel
- Interdisciplinary research
 - Bring HEC to the science domain
 - Many applicable domains: astronomy, astrophysics, economic modeling, pharmaceuticals, chemistry, bioinformatics, neuroscience, data analytics, data mining, biometrics, molecular docking, structural equation modeling, posttranslational protein modification, climate modeling
- Combining Education with Research – NSF CDI, NSF REU, NSF CPATH
- Apply for funding from NSF, DOE, NIH, and industry
 DOE INCITE, NSF PetaApps, NSF HECURA, NIH R01

More Information

- More information: <u>http://www.eecs.northwestern.edu/~iraicu/</u>
- Related Projects:
 - Falkon: http://dev.globus.org/wiki/Incubator/Falkon
 - Swift: <u>http://www.ci.uchicago.edu/swift/index.php</u>
- People contributing ideas, slides, source code, applications, results, etc
 - Ian Foster, Alex Szalay, Rick Stevens, Mike Wilde, Jim Gray, Catalin Dumitrescu, Yong Zhao, Zhao Zhang, Gabriela Turcu, Ben Clifford, Mihael Hategan, Allan Espinosa, Kamil Iskra, Pete Beckman, Philip Little, Christopher Moretti, Amitabh Chaudhary, Douglas Thain, Quan Pham, Atilla Balkir, Jing Tie, Veronika Nefedova, Sarah Kenny, Gregor von Laszewski, Tiberiu Stef-Praun, Julian Bunn, Andrew Binkowski, Glen Hocky, Donald Hanson, Matthew Cohoon, Fangfang Xia, Mike Kubal, Alok Choudhary...
- Funding:
 - **NASA**: Ames Research Center, Graduate Student Research Program
 - **DOE**: Office of Advanced Scientific Computing Research
 - NSF: TeragGrid and Computing Research Innovation Fellow Program