

# Understanding and Improving Computational Science Storage Access through Continuous Characterization

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#### Motivation

- Leadership computing systems are used for a variety of scientific applications
- Understanding production I/O behavior on these systems is important for several reasons
- For scientists:
  - Is my application performing well?
  - Can I get more bang for my buck?
- For administrators:
  - How are the storage resources being used?
  - What applications and resources need to be tuned?
- For researchers and system planners:
  - What are the trends as applications scale up?
  - How do we design the next system?
  - What research avenues are the most promising?



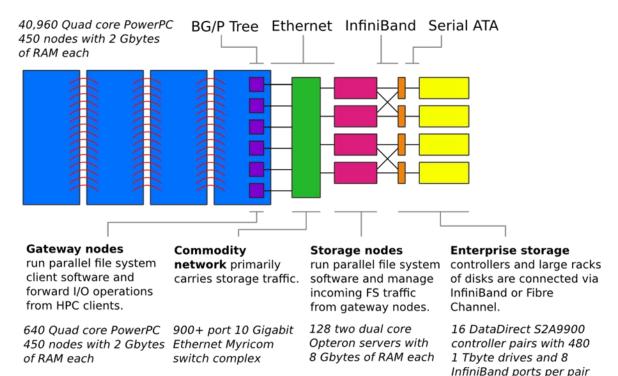
**Our goal**: to observe I/O patterns of the majority of applications running on our leadership platform, without perturbing their execution, with enough detail to gain insight and aid in performance debugging.

# The challenge of collecting data at scale: How do we observe a leadership storage system in its natural habitat?

Target system:

Intrepid IBM BlueGene/P

Argonne National Laboratory



- Open science, capability workload
- Sensitive to performance overhead in production
- Too many applications to instrument manually

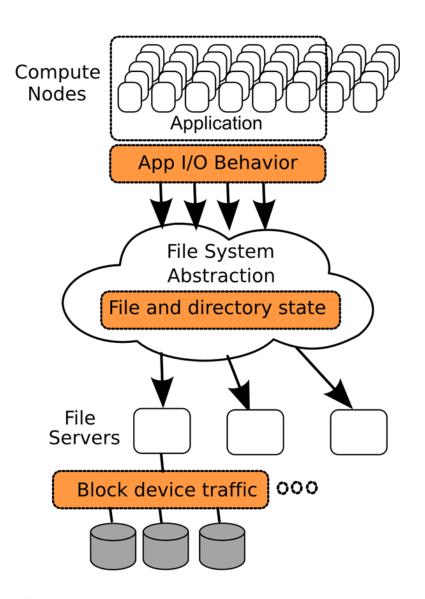


#### **Technical hurdles**

- How do we observe applications?
  - Application or protocol level tracing
    - Performance overhead and volume of data: untenable for system-wide use
  - Benchmarks
    - Isolated examples don't reflect system diversity
  - Statistical sampling
    - May miss critical features
- What APIs or protocols should we instrument?
  - MPI-IO? POSIX? HLL? File System? I/O Forwarding?
- Deluge of information available from multiple system components
  - Inconsistent format
  - What is the overhead?
- How do we correlate application behavior with system activity?



#### Three views of I/O



#### Tools used for instrumentation:

- Darshan: instruments and characterizes I/O function calls at the application level (ANL)
- Fsstats: collects static information about aggregate file attributes such as file size and access time (Shobhit Dayal, CMU)
- Iostat: part of the Sysstat tool suite, can be used to report block device utilization statistics (Sebastien Godard)



#### Darshan

- Darshan records counters, histograms, and strategically chosen timestamps related to I/O activity (not a complete trace of each operation)
- POSIX, POSIX stream, MPI-IO, and limited HDF5 and PNetCDF functions
- Access patterns, access sizes, I/O time, alignment, datatypes, etc.
- Link-time wrappers inserted via modifications to the default MPI compiler scripts
- Minimal overhead during execution
- Reduction, compression, and storage is performed at MPI\_Finalize() time
- "Application level" is important: we observe the application's intentions, rather than the system software's interpretation of those intentions
- Inspired by the 1990s Charisma project, Kotz and Nieuwejaar



#### **Fsstats and iostat**

#### Fsstats:

- http://www.pdsi-scidac.org/fsstats/
- Walks a specified directory tree
- Creates aggregate histograms and usage summaries
- We developed a small wrapper to run fsstats in parallel across a collection of user directories and merge the results

#### lostat:

- http://sebastien.godard.pagesperso-orange.fr/
- Reports data from /proc about utilization on each block device
- Can be run continuously to report information over regular intervals
- We developed wrappers to run iostat on each file server and filter results only include GPFS and PVFS block devices. Logs were post-processed to create aggregate summaries.

```
[pcarns@pcarns-laptop ~]$ iostat -x -d -m
Linux 2.6.38-8-generic (pcarns-laptop) 05/24/2011 __i686_ (2 CPU)
Device: rrqm/s wrqm/s r/s w/s rMB/s wMB/s avgrq-sz avgqu-sz await r_await w_await svctm %util
sda 0.17 5.16 4.58 2.89 0.11 0.03 38.71 0.65 87.37 33.23 173.05 3.98 2.97
```

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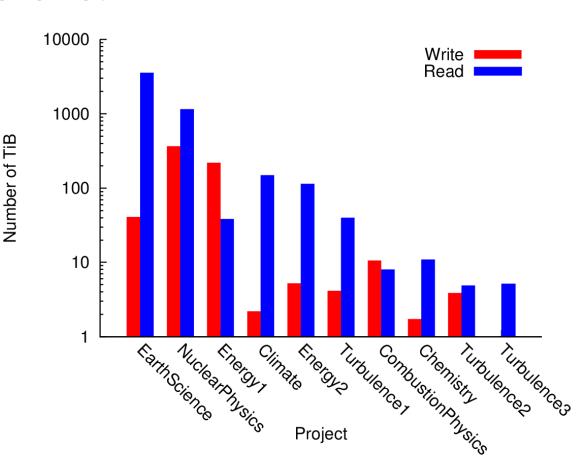
# Studying a production storage system

- In the paper we analyzed a two month window of data collected from January to March of 2010
- This presentation will cover a subset of the findings:
  - What can we learn about trends in I/O patterns and overall system usage?
  - What level of detail can we obtain for specific applications, and how are the most I/O intensive applications performing?
  - How can this data assist in identifying applications with tuning needs?
  - How can this data be used to influence future storage research?
- Application instrumentation:
  - 6,500 jobs (25% of all core hours) were instrumented
  - Examples from 38 distinct science and engineering project allocations
- File system contents (GPFS and PVFS):
  - Roughly 191 million files, sampled at beginning and end of study
- Block device traffic (16 DDN 9900s):
  - Continuous sampling at 60 second intervals, 8 petabytes of total traffic



# Which instrumented applications were the most data-intensive?

Amount of data accessed by projects instrumented with Darshan



- Contradicts assumptions about write-dominated workloads in HPC
- Data usage varies wildly across scientific domains

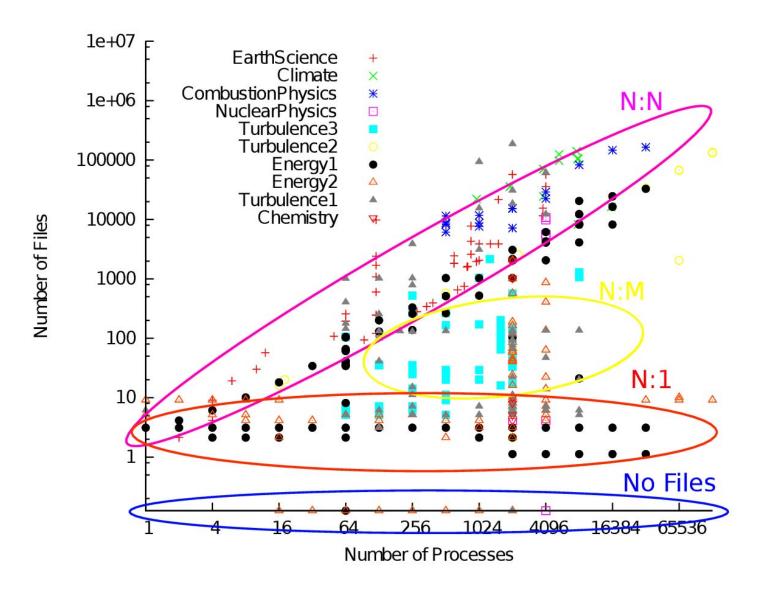


# What I/O strategies do production applications really use?

- Almost as many different I/O strategies as there are applications....
- APIs:
  - Mostly POSIX or MPI-IO, some use of HDF5 and PNetCDF
- Number of files:
  - N:N, N:1, N:0, N:(N/x) and N:(N\*x) mapping of processes to files
  - N:(N/x) includes several examples of manual aggregation: subsets of processes performing I/O on behalf of others without using MPI-IO functionality
- I/O operations:
  - Anywhere from 0.01% to 95% of I/O time is spent performing metadata operations rather than actually moving data
  - Access sizes ranging from 1 byte to 1 GByte



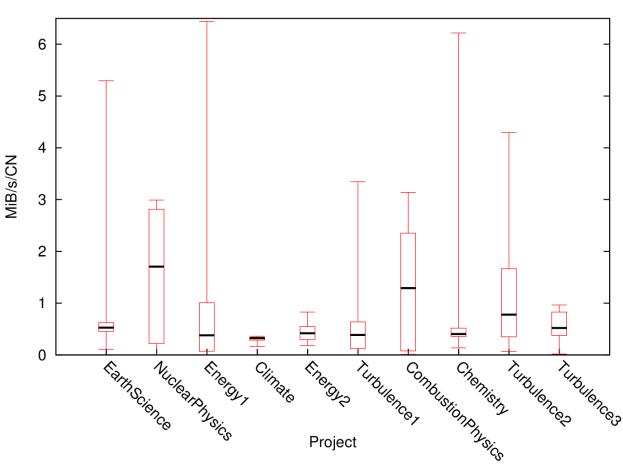
# I/O strategy example: Files per process





### How successful are the various I/O strategies?

- We developed a normalized metric to compare I/O performance across jobs without explicit instrumentation
- The accuracy is limited, but it helps to provide some initial indication of which applications are most interesting
- Data is filtered to only include jobs with 1024 processes or more and at least 512 MiB of data

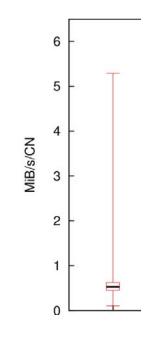


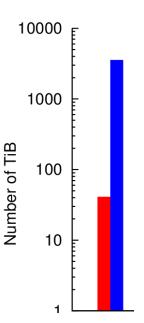
$$MiB/s/CN = \left(\frac{\sum_{rank=0}^{n-1} (bytes_r + bytes_w)}{max_{rank=0}^{n-1} (t_{md} + t_r + t_w)}\right)/N_{cn}$$



# Project case study: EarthScience

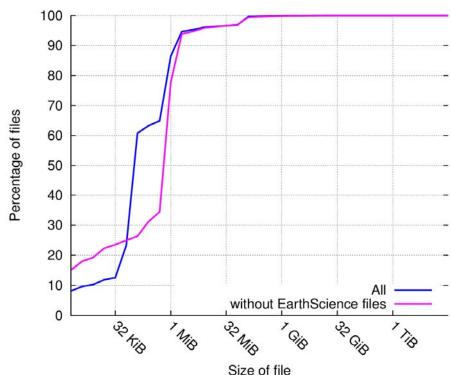
- Most jobs used 4096 cores
- Performance is relatively low, despite some positive characteristics
  - relatively large access size, ranging from 100 KiB to 4 MiB
- File usage:
  - Over 1 million files accessed by some script jobs
  - One read or write per file
  - Contributed 88% of new files during study, but only 15% of storage capacity
- 95% of I/O time was spent performing metadata activity
- This is essentially a data-intensive analysis code, should work well but the file organization leads to staggering metadata overhead on Intrepid

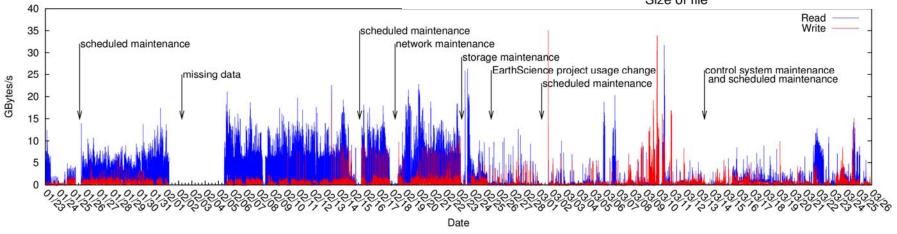




# EarthScience impact

- 96 million files (out of 191 million total), mostly under 128 KiB in size, skewed analysis of file sizes
- Reads went from 78% to 50% of all traffic when this project reduced its activity in February



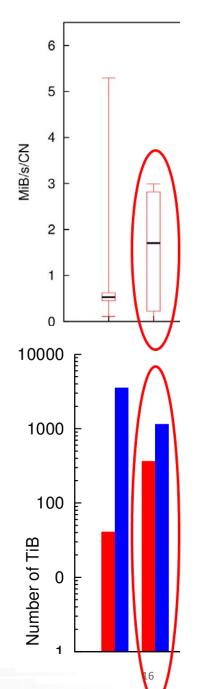




# Project case study: NuclearPhysics

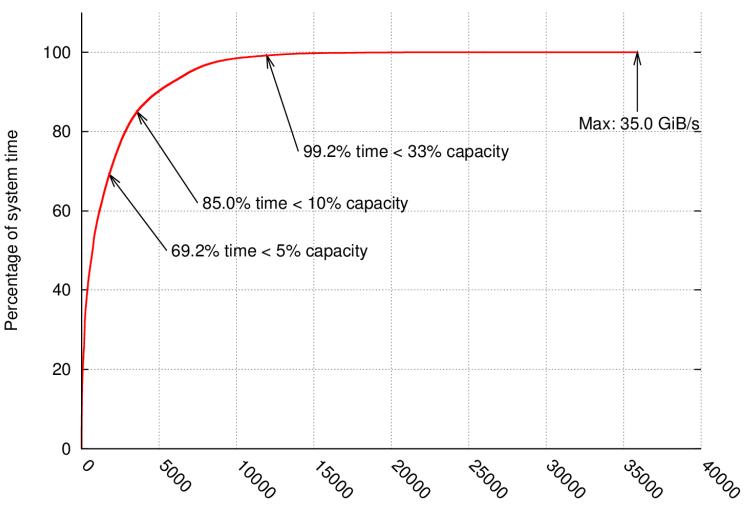
#### Broad range of performance:

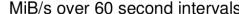
- but very little variability among jobs with consistent command line and job size
- The gap comes from use of two distinctly different executables:
- 1<sup>st</sup> executable:
  - example of manual aggregation: 512 of 4096 processes perform I/O
  - Read 1 TiB and write 500 GiB per job
  - Large access size, balanced I/O, pretty good performance
- 2<sup>nd</sup> executuable:
  - Rank-0 I/O for 2048 or 4096 processes
  - Poor relative performance
  - Is this a problem?
  - Not yet: only 1% of run time is spent performing I/O



### Trends: burstiness

Cumulative distribution of aggregate throughput on 1 minute intervals

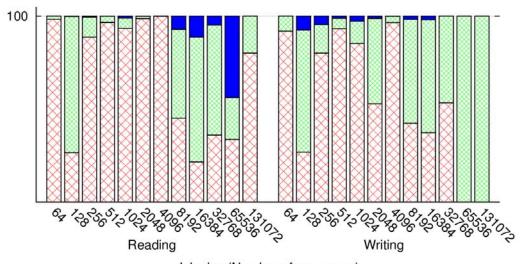




MiB/s over 60 second intervals
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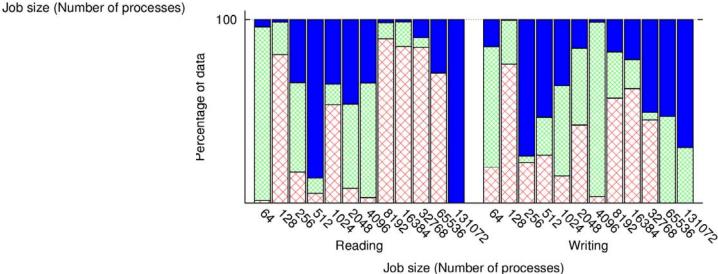
# Trends in application behavior vs. scale



Percentage of data

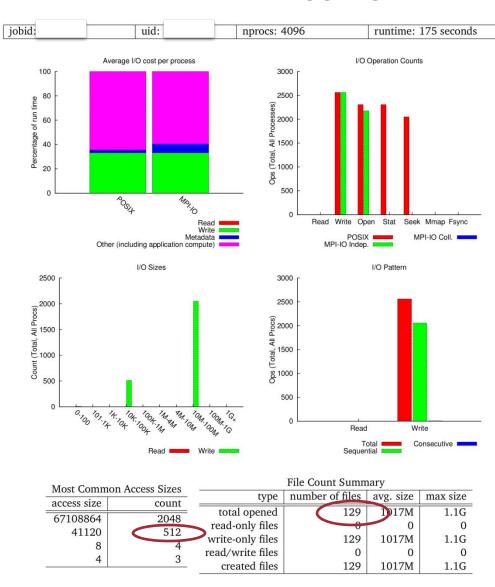
POSIX MPI-IO Libraries

- Increased use of MPI-IO at larger scales
- Increased sharing of files among subsets of processes at larger scales



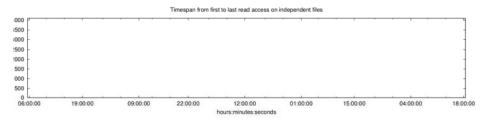
Shared Unique Partial Shared

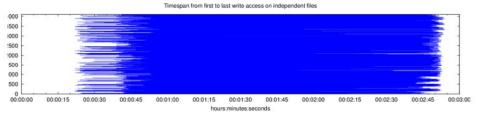
# Examples of debugging and tuning with Darshan

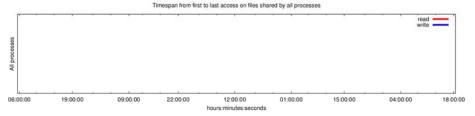


- Example output from job summary tool, available to all users
- Behavioral bug example:
- Mismatch between number of files vs. number of header writes
- The same header is being overwritten 4 times in each data file

# Examples of debugging and tuning with Darshan







|                      | Average I/O per process  |                    |
|----------------------|--------------------------|--------------------|
|                      | Cumulative time spent in | Amount of I/O (MB) |
|                      | I/O functions (seconds)  |                    |
| Independent reads    | 0.000000                 | 0.000000           |
| Independent writes   | 57.985348                | 32.004902          |
| Independent metadata | 4.362344                 | N/A                |
| Shared reads         | 0.000000                 | 0.000000           |
| Shared writes        | 0.000000                 | 0.000000           |
| Shared metadata      | 0.000000                 | N/A                |

| Data Transfer Per Filesystem |              |         |         |         |
|------------------------------|--------------|---------|---------|---------|
| File System                  | Write        |         | Read    |         |
|                              | MiB          | Ratio   | MiB     | Ratio   |
| /pvfs-surveyor               | 131092.07819 | 0.00000 | 0.00000 | 0.00000 |

- Possible performance bug:
- Why do some processes begin writing 25 seconds later than others?





#### **Broad observations**

- It is possible to instrument entire leadership systems during production activity
- System studies paint a different picture of I/O behavior than what we might expect from case studies and anecdotes
- Applications do crazy things (depending on your point of view!)
- There are several reasons:
  - Artifacts from porting:
    - Tuning strategies vary wildly across systems
    - This is a burden for application developers
  - Fumbling around in the dark:
    - How do users find out about I/O behavior and how to improve it?
    - Manual instrumentation is time consuming and prone to error
  - No one universal I/O strategy is consistently "the best"
- I/O behavior varies depending on the scale of the job



# Opportunities for system software

- Useful observations for system software developers:
  - Frequent periods of "mostly" idle time, due to capability workload, maintenance, etc.
  - Files are either deleted quickly, or left unmodified for extended periods of time
- How can we take advantage of those characteristics?
  - Replication, compression, erasure coding, hierarchical storage, reorganization, etc.
- Where are our APIs (or advocacy of those APIs) falling short?
  - Many apps are rolling their own layout and aggregation strategy, despite efforts to provide tools in MPI-IO and HLLs



#### Future work

- Darshan everywhere, all the time, resistance is futile
  - Improve coverage
  - Deploy on more systems that have different job characteristics for comparison
  - TACC example:
    - LD\_PRELOAD instrumentation, coverage of mpich1, mpich2, openmpi and several compilers
    - Tuning Darshan itself for different I/O systems
- Continue file system and device instrumentation, possibly expand into vendorspecific instrumentation
- Automate how we leverage the data:
  - identification of applications that need help
  - tuning suggestions (for application or system parameters)
- Correlation and prediction of characteristics vs. expected performance
- Upcoming Darshan releases



#### Is the data and software available?

# Yes! Announcing the Darshan Data Repository: http://www.mcs.anl.gov/darshan/data

- Anonymized version of data from the time period studied in the paper
- Just Darshan data for now, not iostat or fsstats
- We hope to post more data in the future
- Darshan 2.1.1 (available in a few days) will be able to parse anonymized logs, in the mean time please use svn trunk
- Darshan itself has been available for over a year
- Open source and portable across a variety of systems



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- Thank you to the ALCF management, administrators, and power users at Argonne
  - Positive, helpful responses from everyone involved
- Thank you to the authors of fsstats, iostats, and the Charisma project



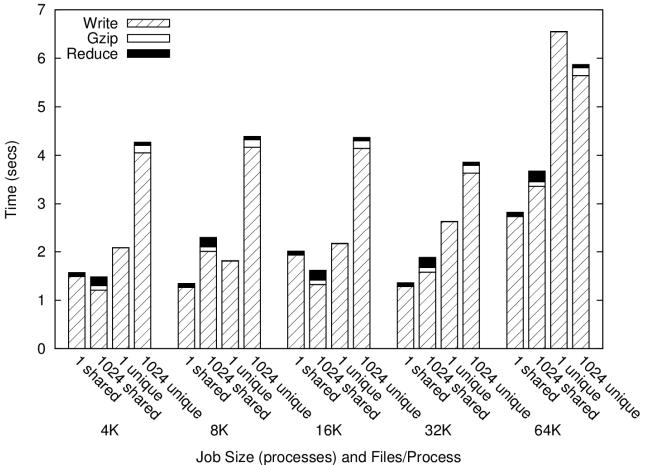
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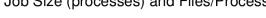
http://www.mcs.anl.gov/darshan



# Darshan overhead at job shutdown time

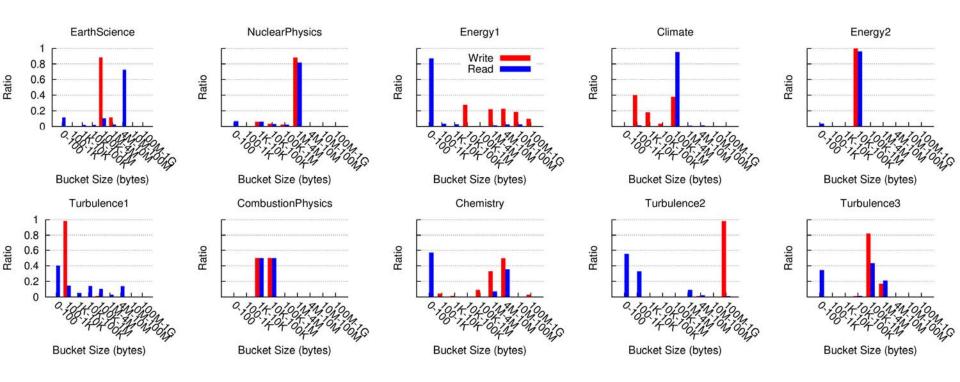


Largest example: 67 million files characterized





# Access size histograms per project





# I/O Characteristics by project

| Project         |          | Percent Time | Cumulative | Files    | Creates  |      |         | MiB      |
|-----------------|----------|--------------|------------|----------|----------|------|---------|----------|
| 1               | MiB/s/CN | in I/O       | md cost    | per proc | per proc | seq. | aligned | per proc |
| EarthScience    | 0.69     | 36.18%       | 95%        | 140.67   | 98.87    | 64%  | 97%     | 1779.48  |
| NuclearPhysics  | 1.53     | 0.55%        | 55%        | 1.72     | 0.63     | 100% | 0%      | 234.57   |
| Energy1         | 0.77     | 39.22%       | 31%        | 0.26     | 0.16     | 87%  | 36%     | 66.35    |
| Climate         | 0.31     | 3.97%        | 82%        | 3.17     | 2.44     | 97%  | 5%      | 1034.92  |
| Energy2         | 0.44     | 0.001%       | 3%         | 0.02     | 0.01     | 86%  | 11%     | 24.49    |
| Turbulence 1    | 0.54     | 0.13%        | 64%        | 0.26     | 0.13     | 77%  | 25%     | 117.92   |
| CombustionPhysi | cs 1.34  | 11.73%       | 67%        | 6.74     | 2.73     | 100% | 0%      | 657.37   |
| Chemistry       | 0.86     | 1.68%        | 21%        | 0.20     | 0.18     | 42%  | 47%     | 321.36   |
| Turbulence2     | 1.16     | 5.85%        | 81%        | 0.53     | 0.03     | 67%  | 50%     | 37.36    |
| Turbulence3     | 0.58     | 0.01%        | 1%         | 0.03     | 0.01     | 100% | 1%      | 40.40    |



#### Trends: burstiness

Breakdown of time periods in which throughput was below 5% of peak capacity

| Duration  |       | Cumulative | Percentage    |
|-----------|-------|------------|---------------|
| (minutes) | Count | Minutes    | of Total Time |
| 1         | 1420  | 1420       | 1.7%          |
| 2-5       | 2259  | 7053       | 8.4%          |
| 6-10      | 775   | 5882       | 7.0%          |
| 11-20     | 383   | 5530       | 6.6%          |
| 21-30     | 104   | 2581       | 3.1%          |
| 31-40     | 50    | 1756       | 2.1%          |
| 41-50     | 30    | 1369       | 1.6%          |
| 51-60     | 19    | 1052       | 1.3%          |
| > 60      | 169   | 30935      | 37.1%         |

- Data is only 1 minute granularity, so some extremely short bursts are averaged out
- More precision needed to observe time periods with zero activity

