SciDAC







center for information

technology integration

UNIVERSITY OF MICHIGAN



LIEL DATA LABO

FRECIE MELLON UNIN

**Carnegie Mellon** 

PETASCALE DATA STORAGE INSTITUTE

- 3 universities, 5 labs, G. Gibson, CMU, PI
- SciDAC @ Petascale storage issues
  - www.pdsi-scidac.org
  - Community building: ie. PDSW-SC07 (Sun 11th)

Scientific Discovery through Advanced Computing

- APIs & standards: ie., Parallel NFS, POSIX
- Failure data collection, analysis: ie., CFDR
- Performance trace collection & benchmark publication
- IT automation applied to HEC systems & problems
- Novel mechanisms for core (esp. metadata, wide area)

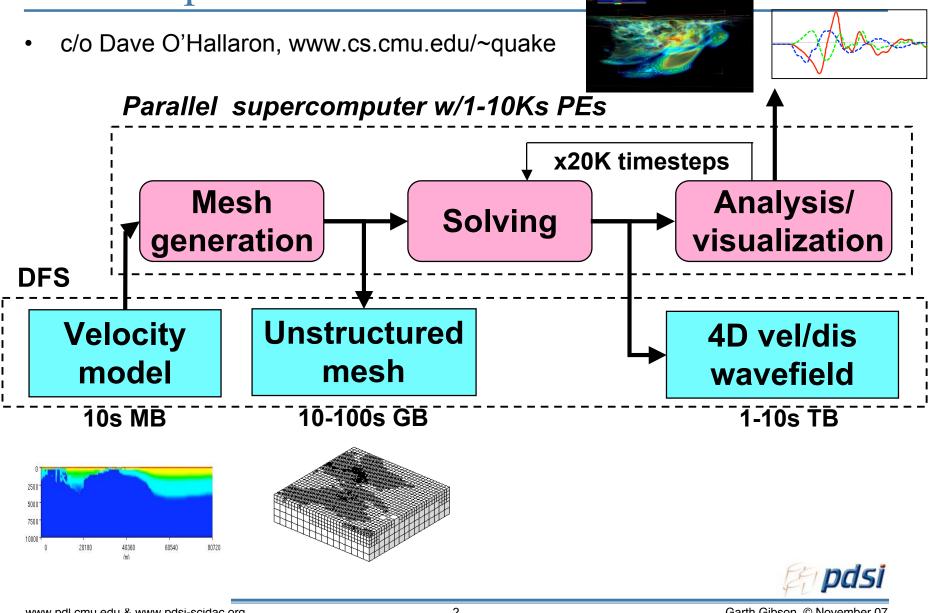




Parallel Data Laboratory

NATIONAL LABORATORY - EST.1943

#### Earthquake simulation



www.pdl.cmu.edu & www.pdsi-scidac.org

# Motivation: Earthquake analysis

- Serialization of model, mesh and wavefield into one database file
  - Total is generally much too large to be entirely loaded into memory
  - Multiple regions of the file contain data of different types
  - Different types of data see different access patterns
    - B-tree on mesh gets random tiny accesses
    - Mesh descriptors get random medium fetches
    - Wavefields get larger accesses, but concurrency makes it look random
- Basic problem: file system sees no internal structure information
  - Prefetching, allocation, caching strategies assume all data is of same type
  - Mixed access patterns strongly confuse file system policies
  - Fall back on safe, probably slower policies: no prefetching, serial locking
- What if file system knew more about "compound, out-of-core" files?
  - Hints? Tradition is too few codes issue them, too few systems use them, too hard to debug performance implications of hints
  - What alternatives?
    - Partial schema definitions: regions that have different types of access
    - Access Methods: embedded some structures in FS (B-trees, arrays ...)



### Motivation: file count scales w/ FLOPS?

- Understanding File Systems at Rest (www.pdsi-scidac.org/fsstats/) •
  - Just getting started with data collection, but already see lots of tiny files
  - Manipulation of small files, and attributes of small files is a growing problem
  - Small things cost mechanical positioning -- orders of magnitude slower
  - Large files may in fact be collections of small things (HDF, netCDF, etc)
    - If sequentially loaded into memory and written from memory, cool
    - But if accessed "out-of-core", really much the same problem
    - Gets worse as numbers of things gets larger, if not sequential access file size: count=12338026 average=18058 580506

	count=12338926 average=18958.589506										
min=0 max=757630040											
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www.pdl.cmu.edu & www.pdsi-scidac.org

Garth Gibson © November 07

## Motivation: HugeDirs -> AttributeDB

- CMU adding directories of 1B to 1T small files to PVFS
  - Partitioning entries into buckets distributed over all storage servers
  - Building index structure for insert(), delete(), lookup()
    - no range query, but unordered complete scan must be fast (readdir)
  - Making highly parallel, minimal bottlenecks, minimal coherency
  - Believe this is extensible to range queries (concurrent B-trees)
    - More challenging to fully load balance if key access patterns arbitrary
    - Probable next steps to do this however
- Simple extension to not create actual files, just manipulate records
  - Special directory is "key, opaque data" tuples
  - Use for extended attributes of small files to optimize to access patterns
    - le., type: mp3, code, pdf, checkpoint, .... Find all by type
    - le., timestamp of last modification, find N oldest
    - Ie,. Size, find N largest
  - Apply powerful aggregation structures as well as indices: ie., fastbits
- Could this be used for variable stores?



# Motivation: ChangeLog Representation

- Fastest checkpoint might be a sequential series of "variable=value"
  - Instead of seeking to serialized location, just append operation to log
  - Each thread writes strictly sequential log of operations
  - "Meaning" of set of logs is applying log to (possibly null) initial database
- Decouple writing logs from applying logs to serialized database
  - Optimize each separately; pipeline from compute to IO nodes
  - Defer serializing by just storing changelogs for later application
  - Some checkpoints never read, so never serialized
  - If read before serialized, trigger serialization (or something smarter)
  - Represent logs as attributes of database; that is, hide in FS (directory?)
  - Important tricks: "block logs" so each can be applied in parallel later; type logs so "no overwrite" can be known and any order allowed
- Not always best representation
  - If data sequential anyway, operation encoding probably larger
  - If read intermingled with write, might force inefficient serialization



## Issues: Library vs File System

- Library is user code bound into application
  - File has structure if the right library is bound into app touching it
  - Rapid development of new "ease of programming" APIs
  - Optimization for actual file system is best guess of library writers
    - Especially a problem if ease of programming was big goal
- File system is system code independent of application
  - File structure known to file system, facilitating optimization
  - Slower development and proliferation as file system is a broad service
  - FS is always-there service; easy delayed processing, rebalancing, recovery
  - FS is inherently distributed, aware of actual servers, changes in servers
- Deployment of new file system APIs?
  - Changes in parallel file systems in progress: pNFS, HEC Posix extensions
- Mitigation for deployment issues: canoncial representations
  - Any file system offering a specialized API also offers:
    - Canonicalization (tar, untar) routines and library implementation written to canonical representation, which does not have to be (as) fast
    - [special case of the backup problem & NDMP solution]





• Select and work through a few examples

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