ECE590-03 Enterprise Storage Architecture

Fall 2016

Workload profiling and sizing

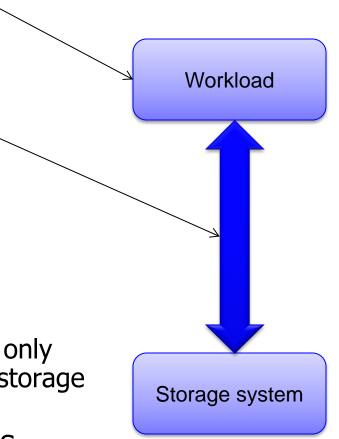
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The problem

- Workload characterization: Determining the IO pattern of an application (or suite of applications)
 - We do so by measuring it, known as workload profiling
- Storage sizing: Determining how much hardware you need to serve a given application (or suite of applications)
- The challenge of characterization and sizing
 - Storage is a complex system!
 - Danger: high penalty for underestimating needs...

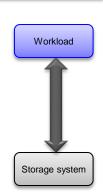
Two kinds of metrics

- Inherent access pattern metrics:
 - Based on the code
- Resulting **performance metrics**:
 - The performance observed when those access patterns hit the storage system
- Sometimes difficult to separate:
 - Common one that's hard to tell: IOPS
 - Did we see 50 IOPS because the workload only made that many requests, or because the storage system could only respond that fast?
 - Was storage system mostly idle? Then IOPS was limited by workload.



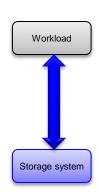
Access pattern metrics

- Random vs. sequential IO
 - Often expressed as random%
 - Alternatives: average distance, seek distance histogram, etc.
- IO size
- IOPS
 - If controller/disk utilization was low, then IOPS represent storage demand (the rate the app asked for)
 - Alternative metric: inter-arrival time (average, histogram, etc.)
- Reads vs. writes
 - Often expressed as read%
 - May also split all of the above by read vs. write (read access pattern often different from write pattern)
- Breaking down application: can we identify separate threads?
 - Is it 50% random, or is there one 100% random thread and one 100% sequential thread?



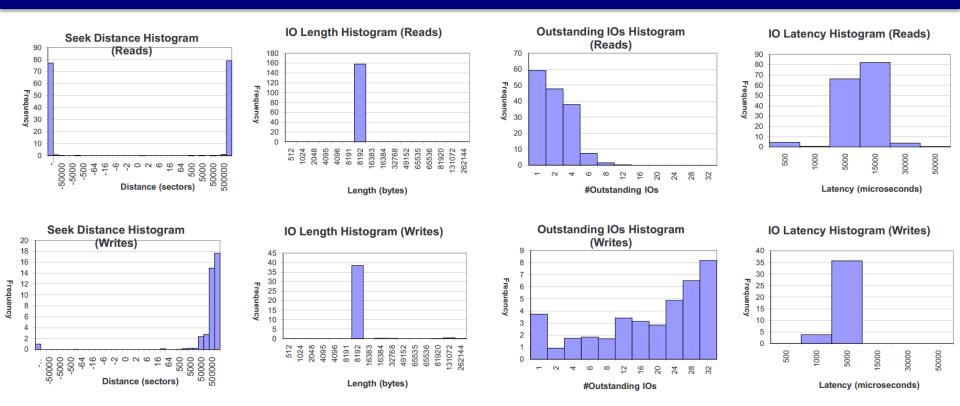
Performance metrics

- IOPS (if storage system was bottleneck)
 - Alternative metric: IO latency (average, histogram, etc.)
 - Alternative metric: throughput (for sequential workloads)



- Queue length: number of IO operations outstanding at a time
 - A measure of IO parallelism

Example of metrics



- Metrics for "DVDStore", a web store benchmark.
 - Random workload (seek distance ≠ 0)
 - IO size = 8k
 - Short read queue, long write queue
 - Reasonable latency (within usual seek time)
 - Seek distance for writes is biased positive (likely due to asynchronous write flushing doing writes in positive order to minimize write seek distance)

How to get these metrics?

- Profiling: Run the workload and measure
- Two problems:
 - How to "run"?
 - Most workloads interact with users
 - Need user behavior to get realistic access pattern!
 - Where to get users?
 - App already in production? Use actual users
 - If not, fake it: **synthetic load generation** (extra program pretends to be users)
 - What about so-called benchmarks?
 - How to "measure"? We'll see in a bit...

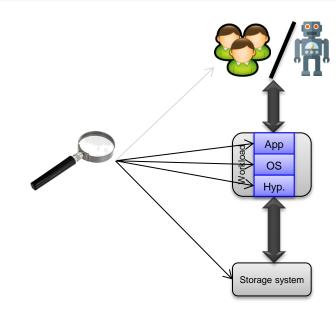


Benchmarks

- **Benchmark**: a program used to generate load in order to measure resulting performance. Various types:
 - The application itself: You literally run the real app with a synthetic load generator.
 - Example: Microsoft Exchange plus LoadGen
 - Application-equivalent: Implements a realistic task from scratch, often with synthetic load generation built in.
 - Example: DVDStore, an Oracle benchmark that literally implements a web-based DVD store.
 - **Task simulator**: Generate an access pattern commonly associated with a certain *type* of workload
 - Example: Swingbench DSS, which generates database requests consistent with computing long-running reports
 - Synthetic benchmark: Generate a mix of load with a specific pattern
 - Example: IOZone, which runs a block device at a given random%, read%, IO size, etc.

Methods of profiling

- App instrumentation
 - Requires code changes
- Kernel instrumentation
 - Can hook at system call level (e.g. strace) or block IO level (e.g. blktrace).
 - Can also do arbitrary kernel instrumentation, hook anything (e.g., systemtap)
- Hypervisor instrumentation
 - Hypervisor sees all I/O by definition
 - Example: vscsiStats in VMware ESX
- Storage controller instrumentation
 - Use built-in performance counters
 - Basically this is kernel instrumentation on the storage controller kernel
- User-level metrics (e.g. latency to load an email)
 - These don't directly help understand storage performance, but they *are* the metrics that users actually care about



Sizing

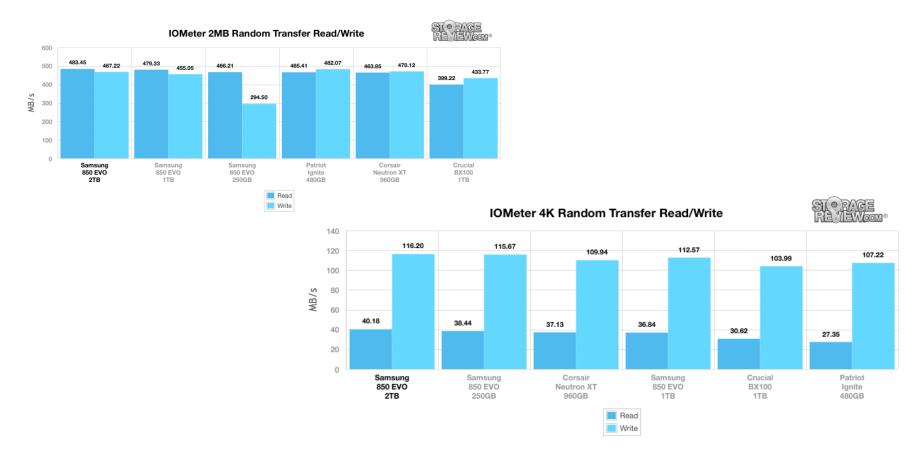
- Now we know how workload acts;
 need to decide how much storage gear we need to buy
- Will present basic rules, but there are complicating factors:
 - Effects of storage efficiency features?
 - Effects of various caches?
 - CPU needs of the storage controller?
 - Result when multiple workloads are combined on one system?
- Real-world sizing of enterprise workloads:
 - For commercial apps, ask the vendor companies with big, expensive, scalable apps have sizing teams that write sizing guides, tools, etc.
 - On the storage system side, ask the system vendor companies with big, expensive, scalable storage systems have sizing teams too.

Disk array sizing

- Recall: In a RAID array, performance is proportional to number of disks; this includes IOPS
- Each disk "provides" some IOPS: IOPS_{disk}
- Our workload profile tells us: IOPSworkload
- Compute $\frac{IOPS_{workload}}{IOPS_{disk}}$: get number of disks needed
- Add overhead: RAID parity disks, hot spares, etc.
- Add safety margin: 20% minimum, >50% if active/active
- Note: this works for SSDs too, IOPS_{disk} is just way bigger

Characterizing disks

- Use synthetic benchmark to find performance in the extremes (100% read, 100% write, 100% seq, 100% random, etc.)
- You did this on HW1...results for Samsung 850 Evo 2TB SSD:



Interpolation-based sizing

- For large/complex storage deployments with mixed workloads, simple IOPS math may break down
- Alternative: measurement with interpolation
 - Beforehand:

```
foreach (synthetic benchmark configuration with access pattern a) foreach (storage system configuration s) set up storage s, generate IO pattern a, record metrics as M[a,s]
```

- Later, given real workload with access pattern $\mathbf{a}_{\mathbf{given}}$ and performance requirements $\mathbf{M}_{\mathbf{required}}$
 - Find points a,s in table where a is near a_{given} and performance M[a,s] > M_{required}
 - Deploy a storage system based on the constellation of corresponding s
 values.
 - Can interpolate storage configurations **s** (with risk)
 - Pessimistic model: Can just pick from systems where a was clearly "tougher" and performance M was still "sufficient"
- Why do all this? Because **s** can include ALL storage config parameters (storage efficiency, cache, config choices, etc.)

Combining workloads

- Rare to have one storage system handle just ONE workload; shared storage on the rise
- Can we simply add workload demands together?
 - Sometimes...it's complicated.
 - Example that works: two random workloads run on separate 3-disk RAIDs will get similar performance running together one 6-disk RAID
 - Example that doesn't: a random workload plus a sequential workload wrecks performance of the sequential workload
 - Random IOs will "interrupt" big sequential reads that would otherwise be combined by OS/controller.

Workload combining

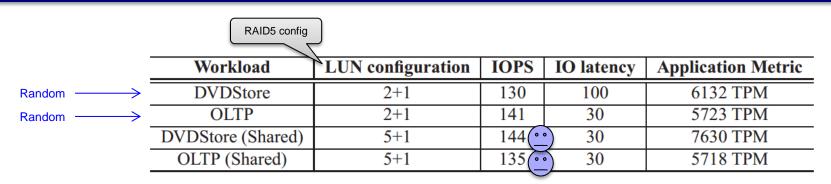


Table 1. Comparison of DVDStore and OLTP when run in isolation and shared mode

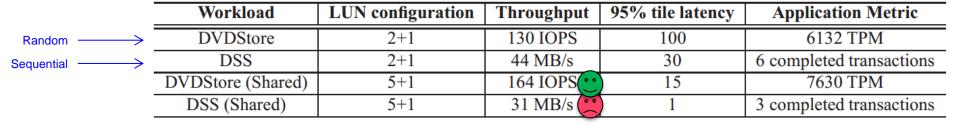


Table 2. Comparison of DVDStore and DSS when run in isolation and shared mode

- "OLTP" = "Online Transaction Processing" (normal user-activity-driven database)
- "DSS" = "Decision Support System" (long-running report on a database)
- Table 2: DVDStore benefits a little from twice as many disks to help with latency, but DSS's sequential IO gets wrecked by the random interruptions to its stream

Conclusion

- To **characterize** a workload, we must **profile** it
 - Run it (generating user input if needed)
 - Measure IO metrics in app/kernel/hypervisor/controller
- Can use workload profile for sizing: to identify storage gear needed
 - Basic rule: provision enough disks for the IOPS you need
 - Past that, look for published guidance from software/hardware vendor
 - Failing that, use successive experiments with differing gear to identify performance trends