# Locality

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### Efficiency matters



#### Figure 1: The Line, Saudi Arabia

# Efficiency matters



Figure 2: Devices are complicated

# Outline

- Principle of locality
- Types of locality
- Approaches that exploit locality
  - Caching
  - Prefetching
  - Partitioning
- Locality examples
  - Data structure layout
  - Locality in locking primitives
  - Locality in NUMA machines

# Locality refers to the idea that interactions or effects are limited to immediate, adjacent areas.

- In computing: Locality refers to the efficiency of data access and processing
- Modern computers are designed using the principle of locality
  - Caches, predictive loading, faster storage transfer

### Efficient data movement is all that matters

- Time/energy cost: moving data
  - One compute unit to a storage unit (CPU  $\leftarrow \rightarrow$  memory)
  - One storage unit to another (disk  $\leftarrow \rightarrow$  memory)
- Communication links also need spaces
  - Buses, networks are bottlenecks

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At the end, we want to minimize data movement or have data ready when we want to work with it

Fundamental limitations exists:

- Packing computation and memory in a limited space
- Shrinking distance among units:
  - Failure of Dennard scaling
  - Cooling is becoming an issue even with 3D chips

### An example of complexity: the memory hierarchy

- Time scale for CPU to access data (or data movement latency):
  - L1 access:  $\sim 1 \text{ns}$
  - L2 access: ~4ns
  - L3 access (local): ~12-20ns
  - L3 access (remote): ~30-90ns
  - Local DRAM: ~80ns
  - Remote DRAM:  $\sim 130\text{-}200\text{ns}$
  - Byte addressable non-volatile memory: ~300ns
  - SSD: ~2-40us
  - Remote machine:  $\sim 2us +$
  - HDD:  $\sim 10 \text{ms}$

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How do we ensure that we can keep up with this complexity?

- Two-level memory hierarchies in the ATLAS computer
  - Main memory + auxiliary storage
- Demand paging
- Backbone of multi-programming



"When it was first observed in the 1960s, thrashing was an unexpected, sudden drop in throughput of a multiprogrammed system ... I explained the phenomenon in 1968 and showed that a **working-set memory controller** would stabilize the system ..."

- Peter D. Denning

Describes the set of information that a process needs to access in a given period of time to carry out its information.

- Model program's memory behavior over time
- Working set of a program:
  - *Programmer's view*: Smallest collection of information present in main memory to assure efficient execution of a program
  - System's view: The set of most recently referenced pages

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- The working set is a reflection of the current active locality of reference for a process

- Locality allows the concept of working set to be effective
- Locality determines which resources are required with some degree of accuracy
- Without locality, unable to predict future resource requirements
  - Leads to inefficient systems

- Temporal locality
- Ø Spatial locality
- Ø Network locality

#### Repeatedly access same memory locations over time period

- Frequent access to sum's memory location illustrates *temporal locality*
- Other examples:
  - Function call and recursion
  - Caching data



Figure 3: sum access

#### Access nearby memory locations within a small time frame

- Consecutive memory access of array
- Other examples:
  - Sequential vs random access of storage media
  - Accessing memory in a row-by-row fashion

nt array[1000]; // assume this array is already populated with data
nt sum = 0;

for (int i = 0; i < 1000; i++) {
 sum += array[i]; // consecutive memory locations are accessed</pre>

Figure 4: array access

### Access to a memory location nearby is faster than access to a memory location that is farther

- Examples:
  - Caches in CPUs: L1, L2, LLC
  - Multi-socket machines
  - Content delivery networks (CDNs)
- Minimize the latency and bandwidth requirements by minimizing the distance for data access

### Locality becomes important for today's machines



Figure 5: Simplified view of a 4-socket machine

- Accessing data from the local socket is faster than accessing from the remote socket
  - Described as non-uniform memory access (NUMA)

### • Caching

- Prefer sequential access over random access
- Partitioning of data or computation

Use cases: working set, lock algorithms, out-of-core graph algorithms, distributed  $k\nu$  stores

Keep a working set of data close to the CPU that is used frequently

- Ubiquitous in systems
  - CPU caches
  - MMUs: TLB
  - Networks (edge caches)
  - OS/DB buffers; storage device controller, DRAMs in storage

Sequential access is faster than random access

- Comes from the physical properties of devices
  - Hard drives
    - $\bullet\,$  Mechanically moving parts: seek time >> transfer time
    - Reading a byte is not cheaper than reading a page
  - Flash/solid state devices: only large blocks can be written
  - DRAM
    - Block addressing and transfer via the bus
    - TLBs (again)
- Examples: write-ahead logging, block nested loop joins

Splitting up the parts of resources and using divide and conquer

- Decomposing an embarrassingly parallel tasks
  - Embarrassing parallel jobs: Do not require any synchronization
    - Can work independently
  - Decompose a large piece of the job, and process them in parallel
  - Ex. Map/reduce

Splitting up the parts of resources and using divide and conquer

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  - Ex. Map/reduce
- But they are not applicable everywhere
  - Non-uniform distribution of access in a key-value store
  - Synchronizing tasks

- Locality starts impacting when the cost to access/modify/move data changes by a huge factor.
- Several scenarios to keep in mind with respect to locality:
  - Minimizing data movement
    - Caching, partitioning for parallel computation and movement
    - Involves either moving computation to data or moving data to the computation unit
  - Data layout for efficient fetching of data
    - Sequential vs random
  - Overlapping computation and data movement
    - Prefetching

- Oata structure layout
- Ø Locking primitives minimizing data movement
- ONUMA: Data structure replication and partitioning

- When accessing memory, CPU accesses data in a way that impacts application's performance
- Two data structures as an example:
  - Arrays
  - Tree data structure

### Arrays

- Matching storage layout with the looping order of algorithms
  - Sequential vs random access
  - Example: Matrix
- Stored as A11, A12, ..., A1n, A21, A22, A2n, ..., Amn
- Loop: for i in 1  $\dots$  n { for j in 1  $\dots$  m { Aij  $\dots$  }} efficient
- Loop: for j in 1  $\dots$  m { for i in 1  $\dots$  n { Aij  $\dots$  }} inefficient

A11	A21	 Am1
A12	A22	 Am2
A1n	A2n	 Amn

- Align storage layout with use cases if possible
  - Loop reordering in compilers
  - Sorting

### Locality with respect to locks

- Locks are the basic building blocks for concurrent systems
- Locks:
  - Provide mutually exclusive access to shared data
  - $\bullet\,$  Order waiters accessing the critical section >
- Lock algorithms try to minimize the movement of shared data



Figure 6: Threads going to access a file protected by a lock

### Spin locks basic behavior

- Waiters wait for their turn
- Locks serialize the access: Introduce sequential bottleneck



Figure 7: Basic spinlock (taken from Art of Multiprocessor Programming)

# Locks first try to minimize contention

- Contention: Threads writing to the same cache line (shared data)
- Hardware maintains a consistent state of the shared data using the coherence protocol



Figure 8: Lock latency

- TAS broadcasts to everyone of the lock situation
  - Saturates memory bandwidth (different from locality)
- Queue lock: Maintains a queue of waiters and notify next in line without bothering others
  - Minimizes shared data contention (cache line)

#### • Let's consider a NUMA machine

• Accessing the local socket is faster than remote socket



#### Figure 9: Accessing in non-NUMA fashion

- Let's consider a NUMA machine
  - Accessing the local socket is faster than remote socket



Figure 10: Accessing in NUMA fashion

• Group lock waiters from one socket, process them, and then pass to another socket

- Comprises of multiple locks (n+1)
  - A global lock
  - NUMA node lock on each node



Figure 11: Cohort lock

- Acquire: First acquire the local node lock, then acquire the global lock
- Release: First release the global lock, then release the local lock
- Maintain locality of data: minimize cache-line bouncing
  - Passes the lock within the same socket multiple times before releasing the global lock

### Need to localize shared data



- Critical section data is transferred for each lock acquire
- The wait for lock increases with increasing thread count

### Need to localize shared data



- Critical section data is transferred for each lock acquire
- The wait for lock increases with increasing thread count
- Q. How can we localize shared data?

- Locality: Keep all shared data on one core
- Use a server client model
  - Clients send request to server (encode their critical section function)
  - Server processes request on client's behalf
- Shared data is ALWAYS accessed by one core!

# Data placement in NUMA machines

- Goal: Keep application's data close to the computation
  - Latency is problematic for memory sensitive applications
  - Bandwidth is an issue for memory intensive applications



- Allocate memory using first touch or interleaved policy
  - First touch: allocating from the local node first
  - Interleaved: Allocate memory using round robin
- Use page migration during application execution (AutoNUMA)

### Realizing locality at various levels

- From caches to CPU
  - Ex: data structure layout: arrays vs linked list
- From one CPU to another
  - HPC algorithms, synchronization primitives
- From memory to LLC
  - Ex: graph algorithms, packet processing
- From one NUMA domain to another NUMA domain
  - Ex: data structures, synchronization primitives (locks)
- From SSD to memory
  - Ex: Paging, out-of-core graph processing applications
- From NIC to memory:
  - Ex: Remote memory, paging

- Locality is one of the most important principles
  - Started from virtual memory; now applicable everywhere
- Three types of locality: temporal, spatial, network
- Locality is applicable across the whole stack