Shared Memory

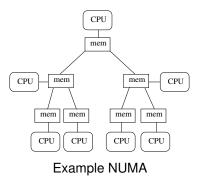
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Shared Memory

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It is well suited to MIMD, but note that SIMD also uses symmetric shared memory, but with a different access pattern

So if symmetric, i.e., uniform access, shared memory does not scale, we can try managing memory in other ways



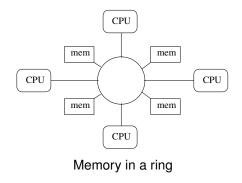
Each processor has a chunk of memory, but can also access memory of other processors, perhaps arranged in a tree

A processor will have fast access to its closest chunk of memory, but slower access to more remote memory

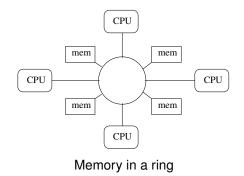
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And different chunks of remote memory will have different access speeds

Of course many other topologies have been tried: star, ring, hypercube, full interconnect, and so on



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This architecture evens out the access time to different chunks of memory a little

These are non-uniform memory access

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It can make a huge difference to the speed of a program if the data is not where it should be

If data is close to the processor that is using it, it will go faster than if the data has to be fetched from further away

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Of course, if data needs to be used by several processors, this becomes a very difficult scheduling problem

NUMA implementations stratify the memory in terms of "distance"

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For example:

- direct connection on the local memory bus
- on the same node
- one hop away
- two hops away
- and so on

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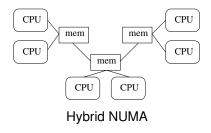
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This is still a matter of great research and development!

And, of course, there are hybrids where CPUs share some memory symmetrically and some memory NUMA





Distributed Memory

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Classifications Distributed Memory

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If the problem is the memory bus bottleneck which means you have to keep cached copies of a value, and then you have the problem of keeping coherence amongst the copies, why not simply *not* have shared memory?

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Distributed memory says each processor's memory it its own and is entirely separate from every other processor's memory

Distributed Memory

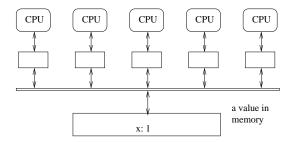
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The variable x on this processor is the same as the x on that processor (assuming SPMD)

Distributed Memory



Shared address space

Distributed Memory

Processors in a distributed memory architecture each have their own, separate, address space

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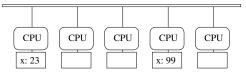
Memory location 42 on one processor is entirely separate from memory location 42 on every other processor

Each processor has their own version of variable x, nothing to do with any other x on other processors

Distributed Memory

Each processor has its own memory

Network

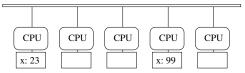


Distributed memory architecture

Distributed Memory

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Distributed memory architecture

Typically connected by a network, rather than an expensive memory bus

Distributed Memory

To get at data on another node a processor sends a *message* to that node, which will reply with the data

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Your code has to change, too

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See MPI (Message Passing Interface) later, but conceptually we have to write

```
x = FetchDouble(remotecpuname, "y");
```

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Both in needing a lot more text, and in needing thought on where to put your data

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Some parallel programming systems (see later) *only* provide messaging across threads (often via mechanisms called *channels*), thus masking the underlying architecture and improving program portability across architectures

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Of course, if you replicate data that gets updated, you immediately have a coherence problem again, but now your own code has to deal with it

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But you do need to put a lot of thought into replicating read-write (mutable) data

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Thus allowing more computation; but at the cost of more complicated programming



Computation vs. Communication

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But this idea of overlapping computation and communication is important and will reappear many times

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Programs have to be written with this in mind: global shared mutable values are simply not a good idea, even in uniprocessor programs!

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The computations do have to be huge!

Distributed Memory

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Scales *very* well as an architecture. Clusters of over a million cores exist: see the TOP500 list

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This is roughly linear (per CPU) price scaling

Distributed Memory

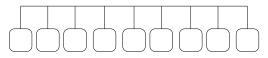
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In a network like

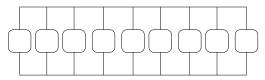


Simple shared network

the single shared network is clearly a bottleneck

Classifications Distributed Memory

So we need to scale the network. There are many choices:

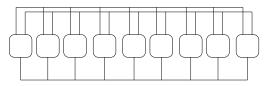


Network with two interfaces

Each processor would use one interface to communicate with processors 0, 2, 4, etc.., and the other interface to processors 1, 3, 5, etc., thus spreading the load

Distributed Memory

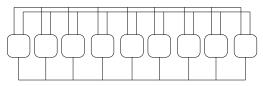
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Network with three interfaces

Distributed Memory

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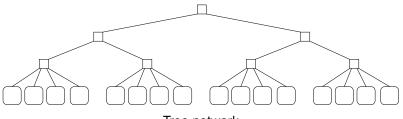


Network with three interfaces

But this gets expensive very quickly

Distributed Memory

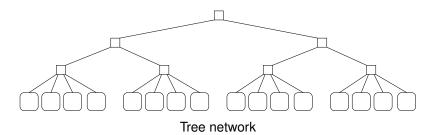
Trees are a good way of connecting things:



Tree network

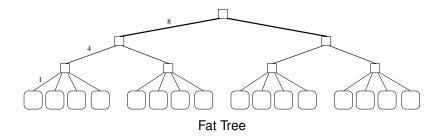
Distributed Memory

Trees are a good way of connecting things:



Though the upper links now are a bottleneck, and we have introduced another non-uniformity

Distributed Memory



In a *fat tree* links up the tree have larger bandwidths, thus allowing full simultaneous bandwidth between each pair of nodes

Distributed Memory

Though the latency between nodes will vary

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In practice, a full fat tree is quite expensive, so real fat trees tend to skimp on the upper links a bit, e.g, 1, 2, 2 in the above diagram would be much cheaper to build (and a "2" would probably be a pair of "1"s)

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Exercise Azure uses a *Clos network* within its datacentres. Read about this

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Adding bandwidth in a network is relatively cheap

But decreasing latency is very expensive whatever the system



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Thus this is also called *distributed virtual memory* and *distributed shared memory*



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But it will be very NUMA to data



Unfortunately, programmers *do* have to care as the speed of a program will be very hard to predict or control, depending on how data is distributed across memory and the particular NUMA architecture it is running on



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A poor programmer will think their life is easier



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Compare with "how fast is x = y?" in VSM

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- in the Operating System, such as *Mosix*. This means all standard system libraries and user code can be used unchanged and a cluster looks like a single big machine: a *single system image* (SSI)
- by the programming language and libraries, such as Cluster OpenMP or Unified Parallel C (see later), so the language may need a bit of learning by the programmer

Virtual Shared/Distributed Virtual Memory

VSM is currently fairly rare in practice, though as NUMA techniques improve, people are starting to talk about *shared memory clusters* as being a viable and useful way to proceed

Latency numbers every programmer should know

| L1 Cache hit | 0.5 ns | 0.5 sec one heart beat |
|--|----------------|---|
| Mutex lock/unlock | 25 ns | 25 sec making coffee |
| Main memory access | 100 ns | 100 sec |
| Read 1MB from memory | 250,000 ns | brushing your teeth 2.9 days a long weekend |
| Round trip within datacentre | 500,000 ns | 5.8 days a short holiday |
| Read 1MB from disk | 30,000,000 ns | 1 year |
| Send a packet California $ ightarrow$ | 150,000,000 ns | 4.8 years |
| $\text{Netherlands} \rightarrow \text{California}$ | | two round trips |
| | | to Mars |

https://gist.github.com/hellerbarde/2843375

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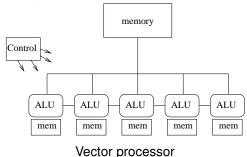
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It has elements of both shared and distributed memory

It is used for data parallel computation

Vectors

A *vector processor* is a SIMD collection of CPUs (actually ALUs), often with a chunk of global shared memory (and a single control unit)



Each processor also has its own chunk of local memory that it operates on

Classifications Vectors

The local memory allows each ALU to work on a different set of values



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Note: this is not cache, but simply per-ALU memory

Cache vs Local

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Cache memory: a fast local copy of a slower memory location. If a value of a variable is cached on different cores, we want all the caches to contain the same value for that variable

Local memory: per core memory (not always fast, by the way!) where we expect to have different values for a given variable in each