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For example, we might be able to overcome message latency by judicious use of non-blocking sends and receives

Rather than waiting for a receive to complete, we carry on working on some other part of the computation: later, when the receive has completed, we can go back to that part of the computation

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**Exercise** You wish to make a cup of tea and a sandwich. Do you

(a) make the sandwich then start boiling the kettle; or

(b) start boiling the kettle then make the sandwich?





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- it is hard to program effectively: simple programs are easy to write, but efficient programs usually need experienced programmers
- there are a huge number of variations of messaging: quite often you can replace several calls to MPI functions with one, more complex, MPI function that is more efficient overall

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- it is not naturally dynamic: the number of processors is effectively fixed and cannot vary during the execution of the program. This excludes efficient execution of some kinds of program (later versions of MPI do include MPI\_Comm\_spawn but it's not easy to use)

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The MPI standard is still being developed and updated



# **Exercise** Read about UPC, a (not popular) alternative to MPI, that presents a virtual shared NUMA architecture

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First we need to recall the SIMD architecture and go through the issues it brings



All processors are controlled by just one Control unit, so are all executing the same instruction



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This is data parallelism

There is a shared chunk of *global memory* and each processor has its own chunk of *private memory* 

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Processors can be strung linearly in a *vector* or in a square mesh as an *array* 



Array processor



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Of course, you can use an array as a vector or a vector as an array, with a modest loss of efficiency

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GPUs owe a lot to array processor design: more on this later

The basic idea of SIMD is that we can parallelise loops like

```
for (i = 0; i < 1024; i++) {
    c[i] = a[i] + b[i];
}</pre>
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as

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Exercise Go back and look at OpenMP
```

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So no races, thus no serialisation of the operations is needed

What if there are conflicts? For example

```
for (i = 1; i < 1024; i++) {
    a[i] = a[i] + a[i-1];
}</pre>
```

Here, the new value of a[i] depends on the value of a[i-1]; which will have been updated in the previous iteration of the loop

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takes the original value of a[i-1]

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1	1 1	1 1	1 1	+
1	2	2	2	

This is due to the nature of the original loop: it is actually a *prefix scan* operation

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Prefix scans can be done SIMD, but when parallelising code you have to be aware that is what is happening!

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If your problem is data parallel, it can get huge speedups by running SIMD

If you can get your data to the individual processors fast enough

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The major way to lose efficiency is through data movement

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This kind of asynchronous programming improves efficiency but is much harder to do and to get right