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They have a language *Heterogeneous-Compute Interface for Portability* (HIP) that is very similar to CUDA and runs on AMD and NVIDIA hardware

In fact, they have a CUDA to HIP translator to aid porting code

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But it is clear each of CUDA, HIP and OneAPI are “best used” with the hardware of their respective developers

GPUs

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Exercise Read about these

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This is more like an OpenMP for GPUs

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Exercise Is that a good or a bad thing?

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A freely available version of OpenACC for NVIDIA GPUs is available and GCC also supports it (but only on Nvidia and AMD): this may help OpenACC to become more popular

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See the current Top 500

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So the accidents of history brought us to where we are today

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One processor that has had a lot of attention recently is the Apple M1

Coprocessors

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- 4 fast CPU cores
- 4 energy efficient CPU cores
- 8 GPU cores (24,576 threads)
- up to 16GB memory in a *unified memory* architecture
- 16 core neural processing unit (NPU) (11 trillion ops/sec)
- a digital signal processor (DSP)
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This takes coprocessing to new levels!

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Advanced Exercise Read about the memory consistency features used by the M1 to support compatability with x86 code

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Initially for support of speech recognition, it could probably be used for more general deep learning

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As well as doing some graphics. . .

The End

End of Lectures

Future sessions will be problems classes, and going through past papers