Master Informatics Eng.

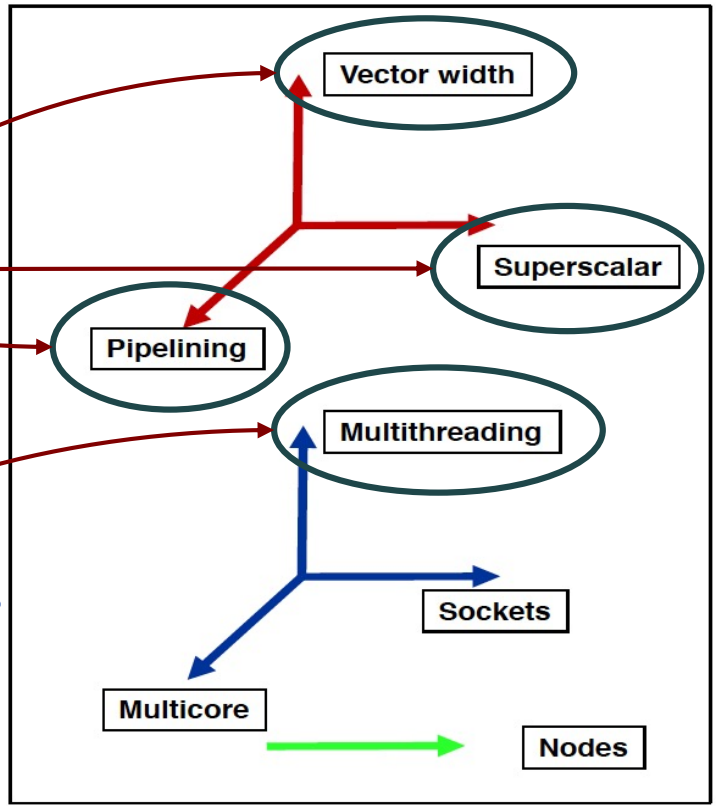
2021/22

A.J.Proença

At core level:
Multithreading and Data Parallelism

Key issues for parallelism in a single-core

- **Currently under discussion:**
  - pipelining:
    reviewed in the `combine` example
  - superscalar:
    idem, but some more now
  - data parallelism:
    vector computers &
    vector extensions to scalar processors
  - multithreading:
    alternative approaches



Multithreading

Performing multiple threads of execution in parallel

- Share all resources but replicate registers, PC/IP, etc.
- Fast switching between threads

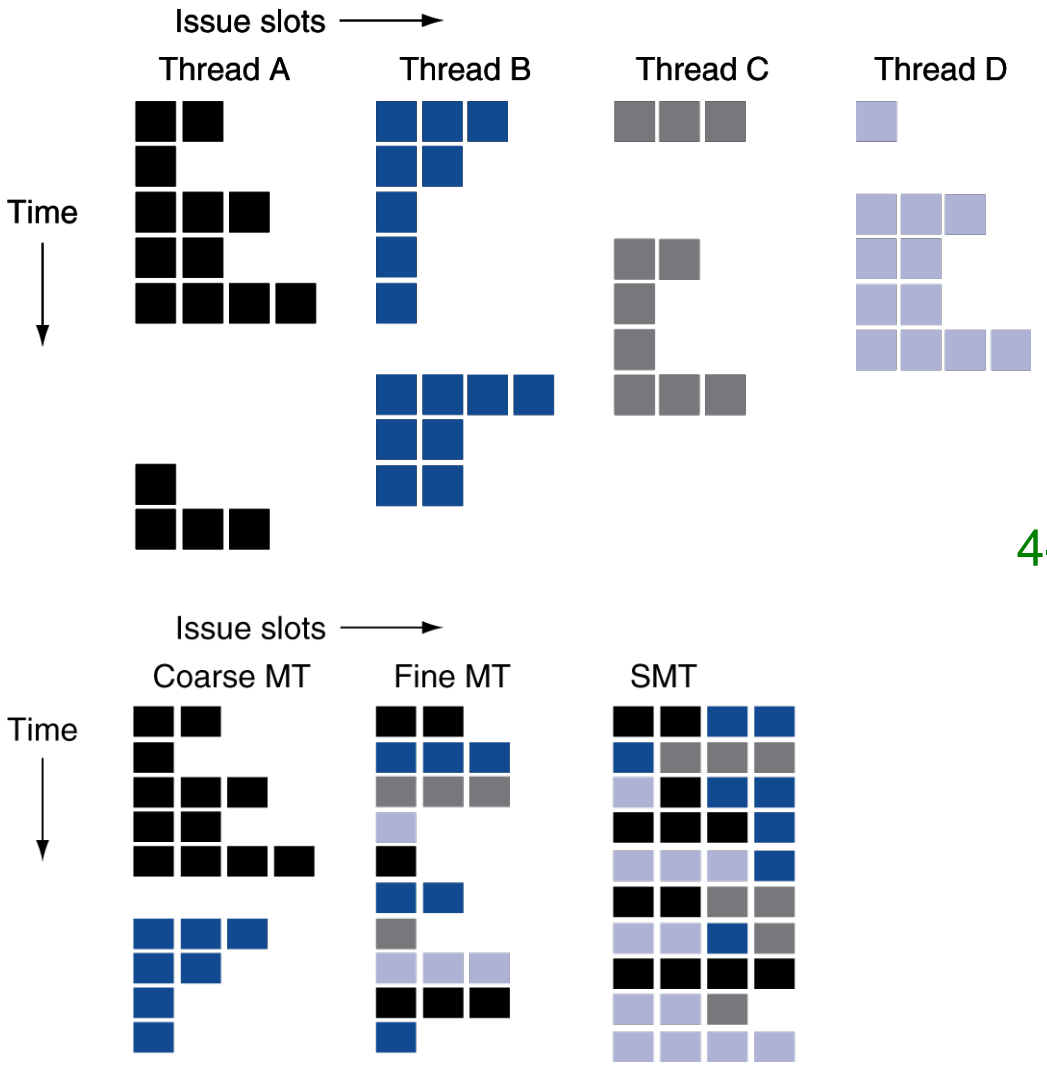
1. Fine-grain multithreading / time-multiplexed MT
   - Switch threads after each cycle
   - Interleave instruction execution
   - If one thread stalls, others are executed
2. Coarse-grain multithreading
   - Only switch on long stall (e.g., L2-cache miss)
   - Simplifies hardware, but doesn't hide short stalls (eg, data hazards)
3. Simultaneous multithreading *(next slide…)*

3. Simultaneous Multithreading

- In multiple-issue dynamically scheduled processor
  - Schedule instructions from multiple threads
  - Instructions from independent threads execute when function units are available
  - Within threads, dependencies handled by scheduling and register renaming
- Example: Intel from Pentium-4 HT
  - Two threads: duplicated registers, shared function units and caches

***HT**: Hyper-Threading, Intel trade mark for their SMT implementation*
*MT in Xeon Phi KNC: 4-way SMT with time-mux MT, **not HT**!*

Multithreading Example



4-way superscalar

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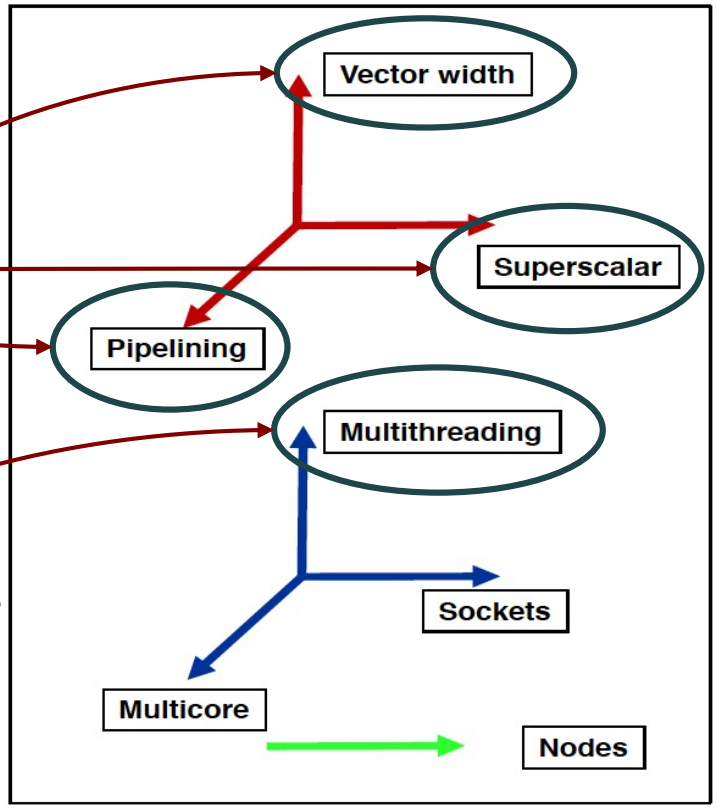


Instruction and Data Streams

Flynn's Taxonomy of Computers *

		Data Streams	
		Single	Multiple
Instruction Streams	Single	**SISD**: Intel Pentium 4	**SIMD**: SSE instructions of x86
	Multiple	**MISD**: No examples today	**MIMD**: Intel Xeon e5345

- SPMD: Single Program Multiple Data
  - A parallel program on a MIMD computer
  - Conditional code for different processors

** Mike Flynn, "Very High-Speed Computing Systems", Proc. of IEEE, 1966*

Introduction

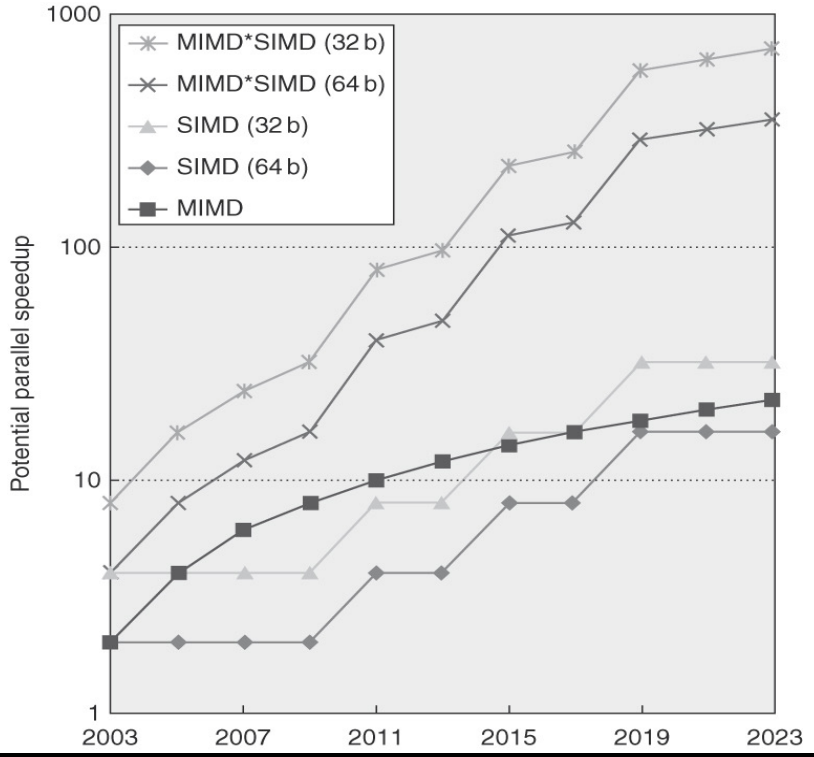
- SIMD architectures can exploit significant data-level parallelism for:
  - matrix-oriented scientific computing
  - media-oriented image and sound processing
- SIMD is more energy efficient than MIMD
  - only needs to fetch one instruction per data operation
  - makes SIMD attractive for personal mobile devices
- SIMD allows programmers to continue to think sequentially

SIMD Parallelism

- Vector architectures
- SIMD & extensions
- Graphics Processor Units (GPU) *(another set of slides)*

- For x86 processors:
  - Expected grow:
    2 more cores/chip/year
  - SIMD width:
    2x every 4 years
  - Potential speedup:
    SIMD 2x that from MIMD!

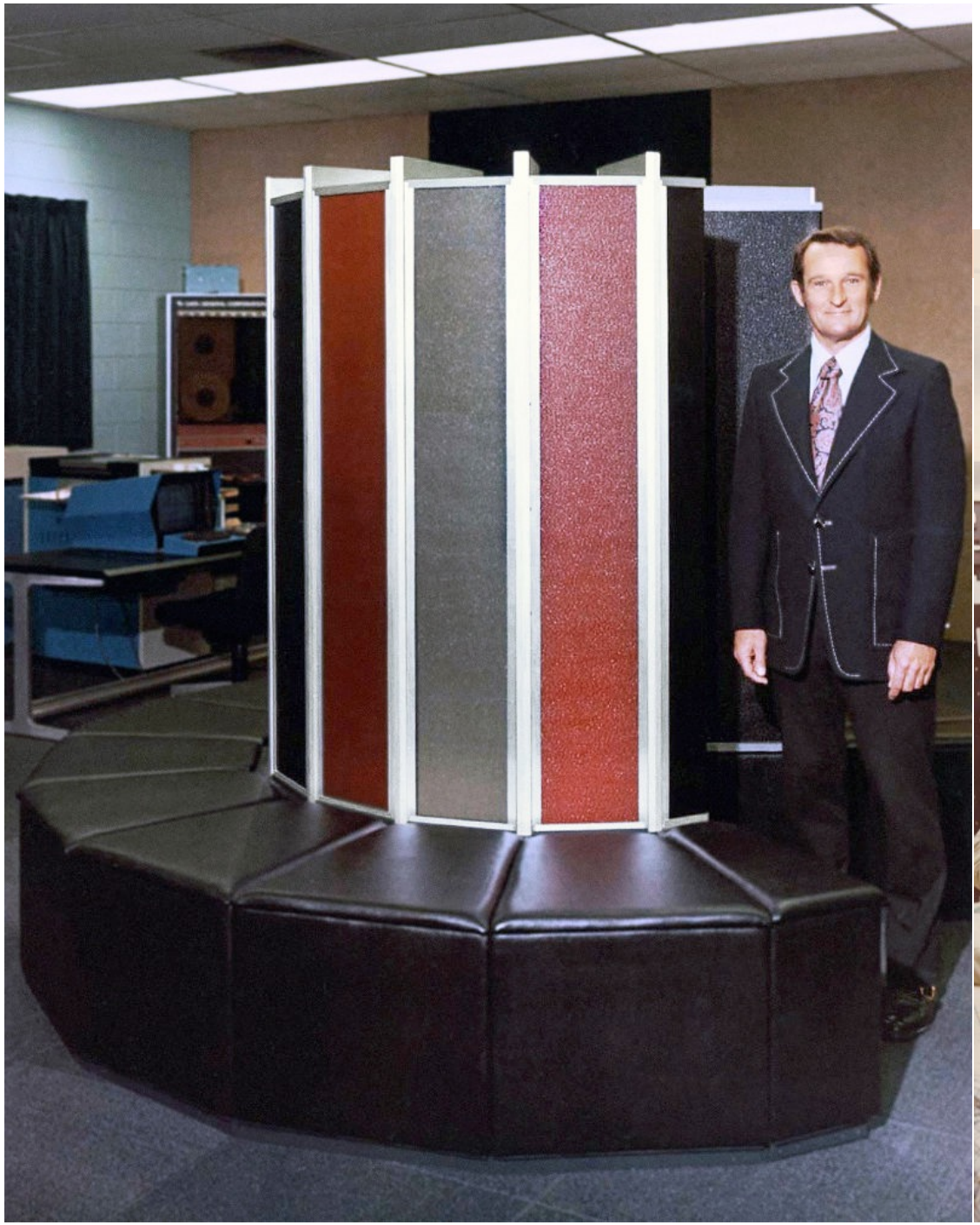
2012 !

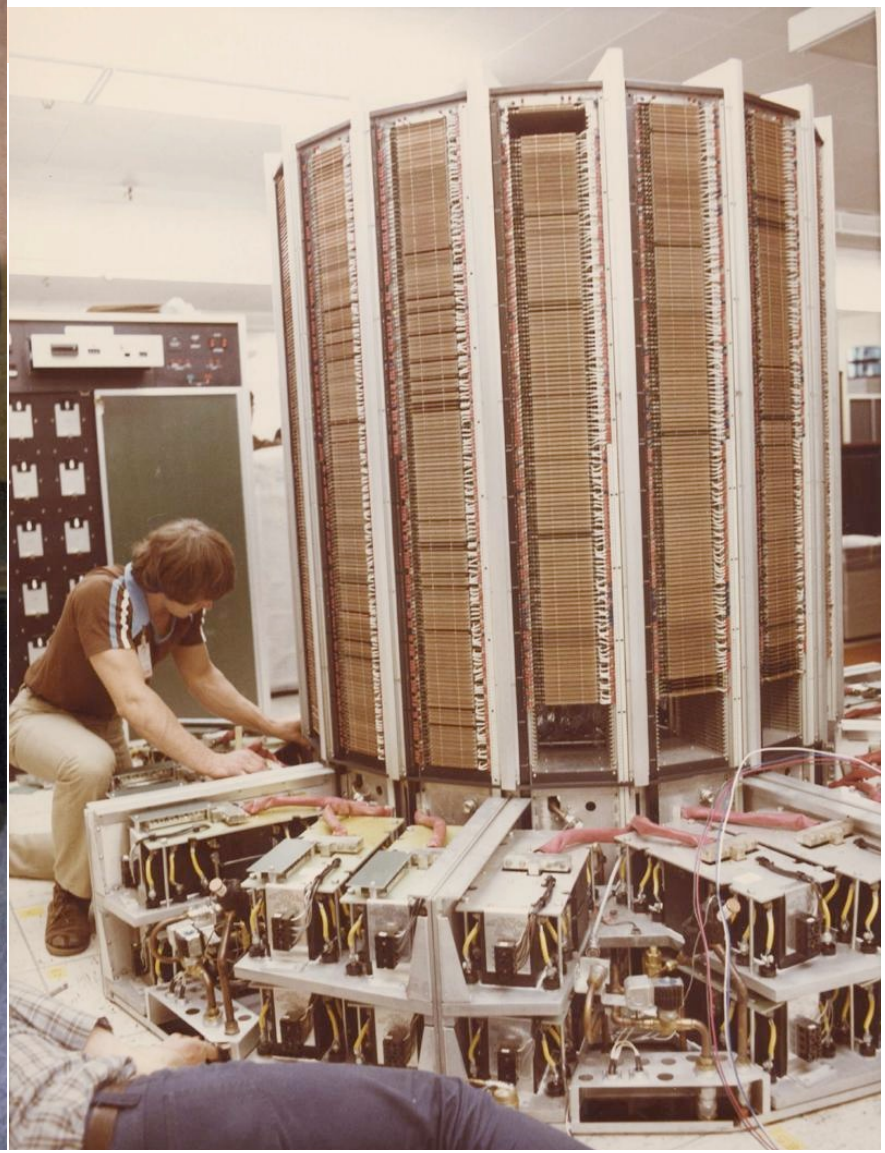


Vector Architectures

- Basic idea:
  - Read sets of data elements *(gather from memory)* into "vector registers"
  - Operate on those registers
  - Store/scatter the results back into memory
- Registers are controlled by the compiler
  - Used to hide memory latency
  - Leverage memory bandwidth

Cray-1 Supercomputer (1976)





Challenges

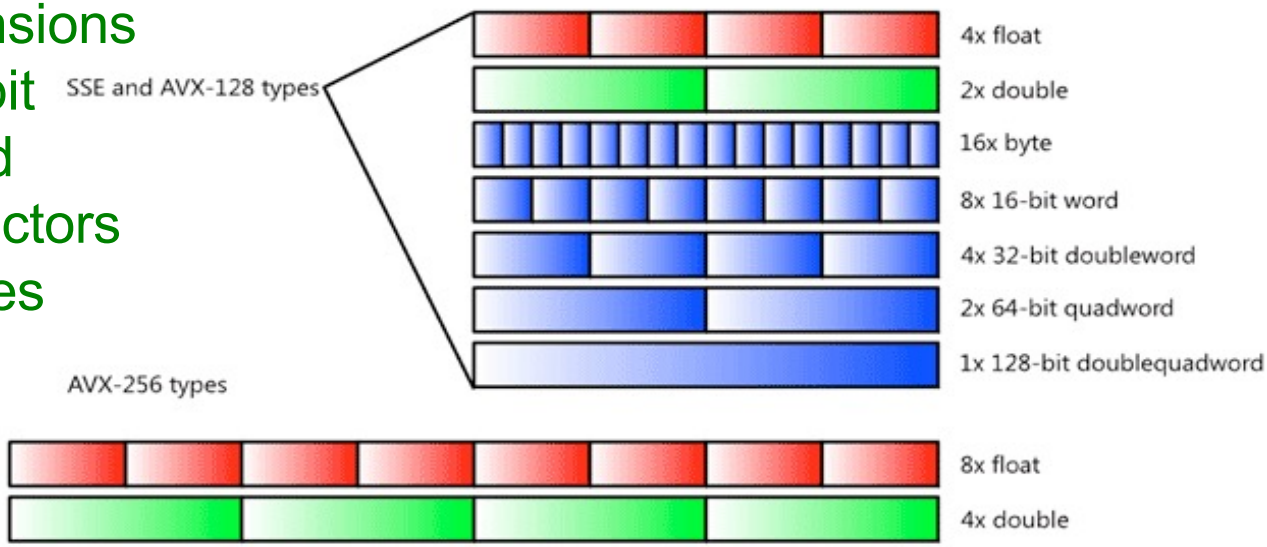
- Start up time
  - Latency of vector functional unit
  - Assume the same as Cray-1
    - Floating-point add => 6 clock cycles
    - Floating-point multiply => 7 clock cycles
    - Floating-point divide => 20 clock cycles
    - Vector load => 12 clock cycles
- Improvements:
  - > 1 element per clock cycle
  - Non-64 wide vectors
  - IF statements in vector code
  - Memory system optimizations to support vector processors
  - Multiple dimensional matrices
  - Sparse matrices
  - Programming a vector computer

Vector Programming

- Compilers are a key element to give hints on whether a code section will vectorize or not
- Check if loop iterations have data dependencies, otherwise vectorization is compromised
- Vector Architectures have a too high cost, but simpler variants are currently available on off-the-shelf devices, as SIMD extensions to the scalar processor; however:
  - most do not support non-unit stride => care must be taken in the design of data structures
  - same applies for gather-scatter...

SIMD Extensions

- Media applications operate on data types narrower than the native word size
  - Intel SIMD extensions started with 64-bit wide vectors and grew to wider vectors and more facilities
  - Current AVX generation is 512-bit wide

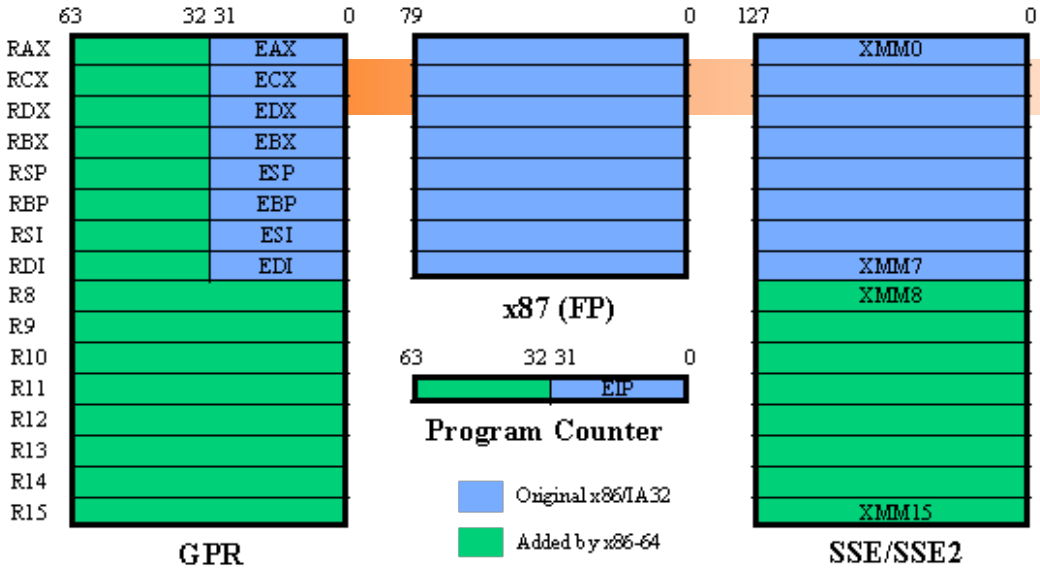


- Limitations, compared to vector architectures:
  - Number of data operands encoded into op code
  - No sophisticated addressing modes (strided, scatter-gather, but…)
  - No mask registers

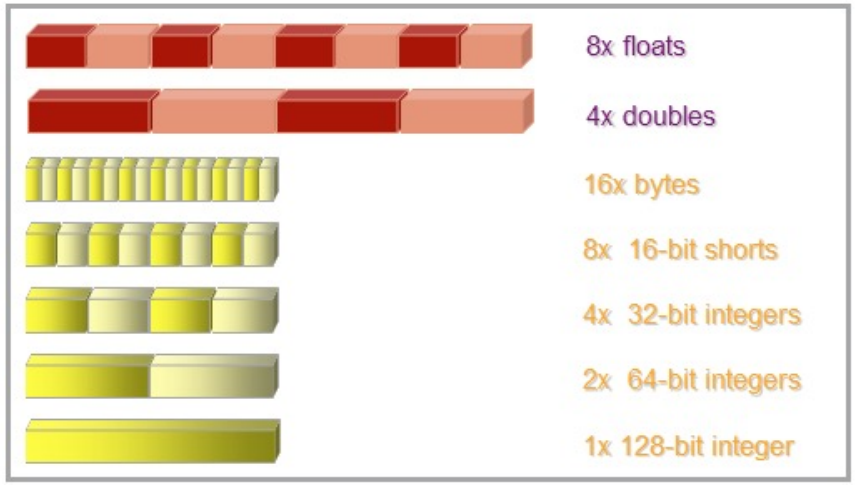
SIMD Implementations

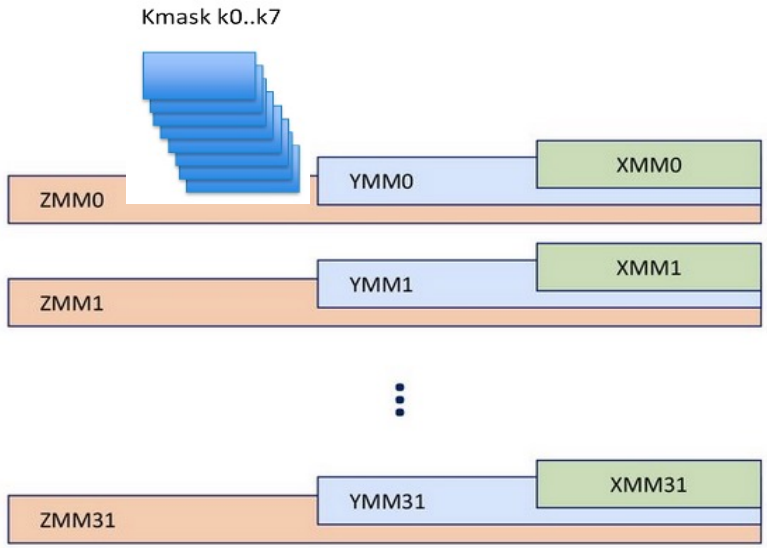
- Intel implementations:
  - MMX (1996)
    - Eight 8-bit integer ops or four 16-bit integer ops
  - Streaming SIMD Extensions (SSE) (1999)
    - Eight 16-bit integer ops
    - Four 32-bit integer/fp ops or two 64-bit integer/fp ops
  - Advanced Vector eXtensions (AVX) (2010...)
    - Eight 32-bit fp or four 64-bit fp ops (integers only in AVX-2)
    - 512-bits wide in AVX-512 (and also in Larrabee & Phi-KNC)
  - Operands **must / should be in consecutive and aligned** memory locations
- AMD Zen/Epyc (Opteron follow-up): up to AVX-2
- Armv8 (64-bit) architecture: NEON & SVE

Registers for vector processing in Intel 64

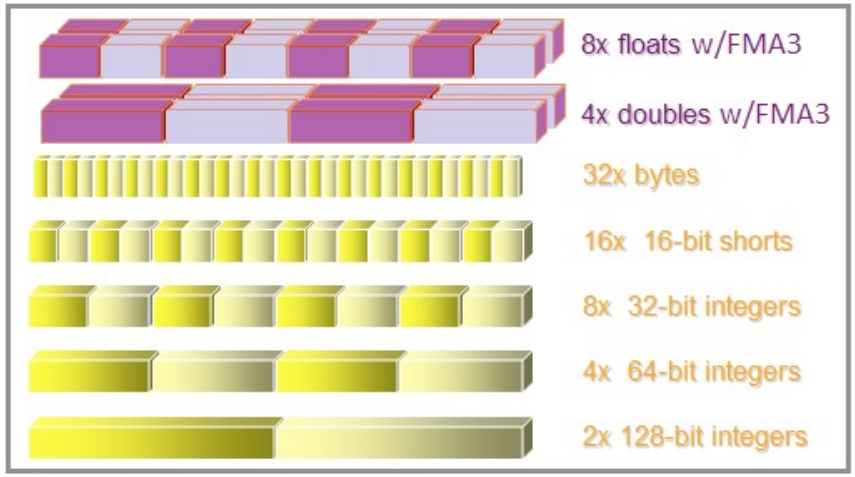


Advanced Vector eXtensions, AVX 1.0





- XMM0 – XMM15
  - 128-bit registers
  - SSE
- YMM0 – YMM15
  - 256-bit registers
  - AVX, AVX2
- ZMM0 – ZMM31
  - 512-bit registers
  - AVX-512



AVX 2.0

AMD only supports AVX 2

Intel evolution to the AVX-512

Intel® AVX Technology

256b AVX1
16 SP / 8 DP Flops/Cycle

256b AVX2
32 SP / 16 DP Flops/Cycle (FMA)

512b AVX-512
64SP / 32 DP Flops/Cycle (FMA)

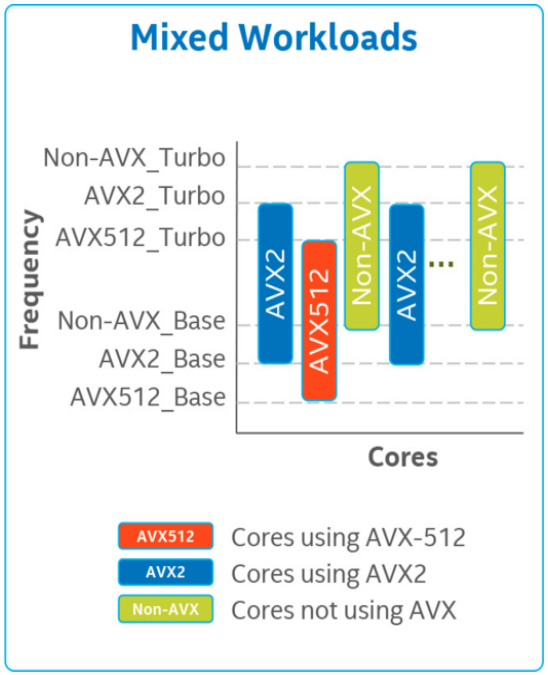
AVX	AVX2
256-bit basic FP	Float16 (IVB 2012)
16 registers	256-bit FP FMA
NDS (and AVX128)	256-bit integer
Improved blend	PERMD
MASKMOV	Gather
Implicit unaligned	

AVX-512
- 512-bit FP/Integer
- 32 registers
- 8 mask registers
- Embedded rounding
- Embedded broadcast
- Scalar/SSE/AVX "promotions"
- HPC additions
- Transcendental support
- Gather/Scatter

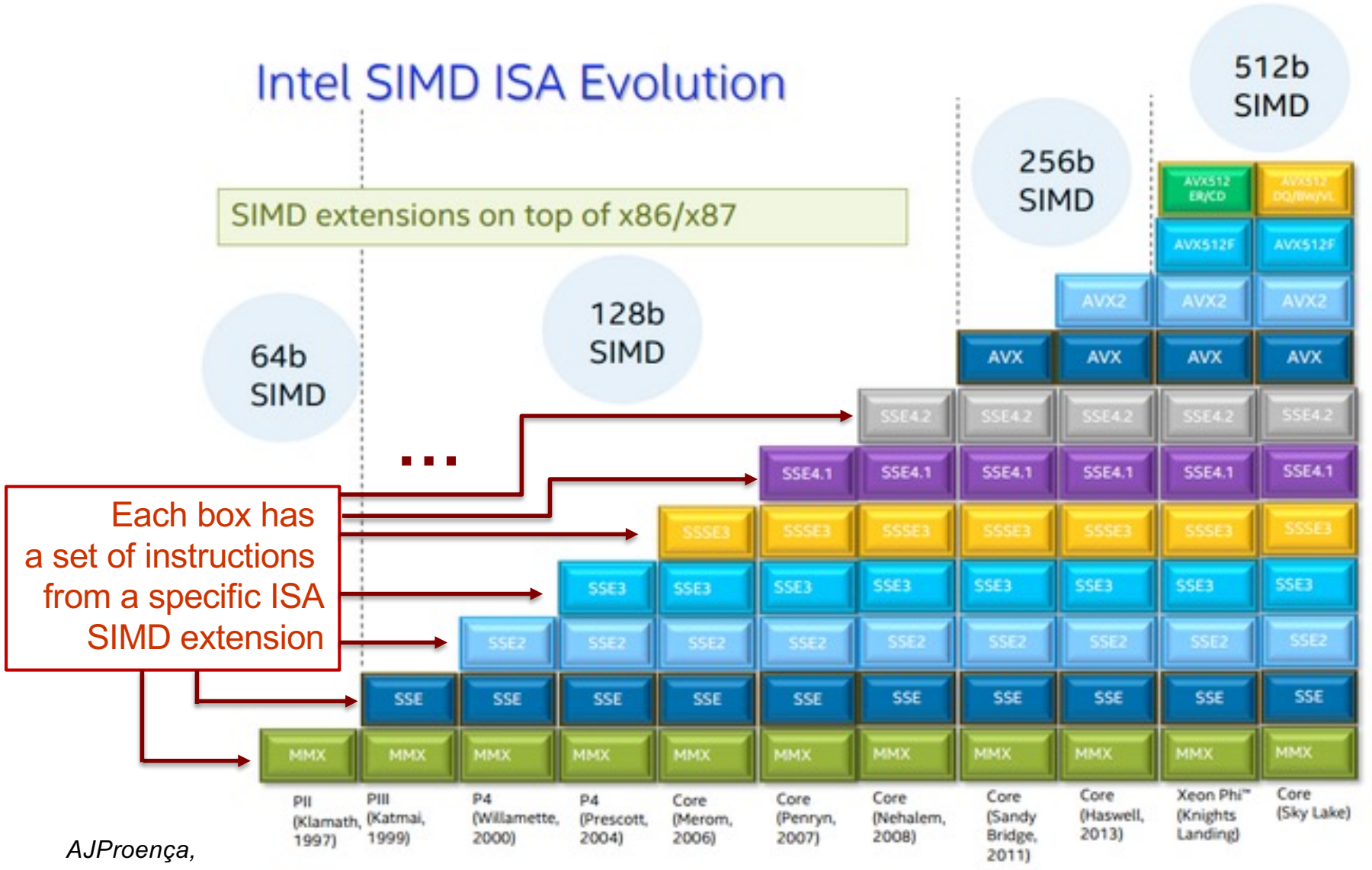
SNB 2011 — **Sandy Bridge**

HSW 2013 — **Haswell**

~~Future~~ Processors (KNL & SKX)
~~in planning, subject to change~~



Intel SIMD ISA evolution



The AVX-512 across Intel devices

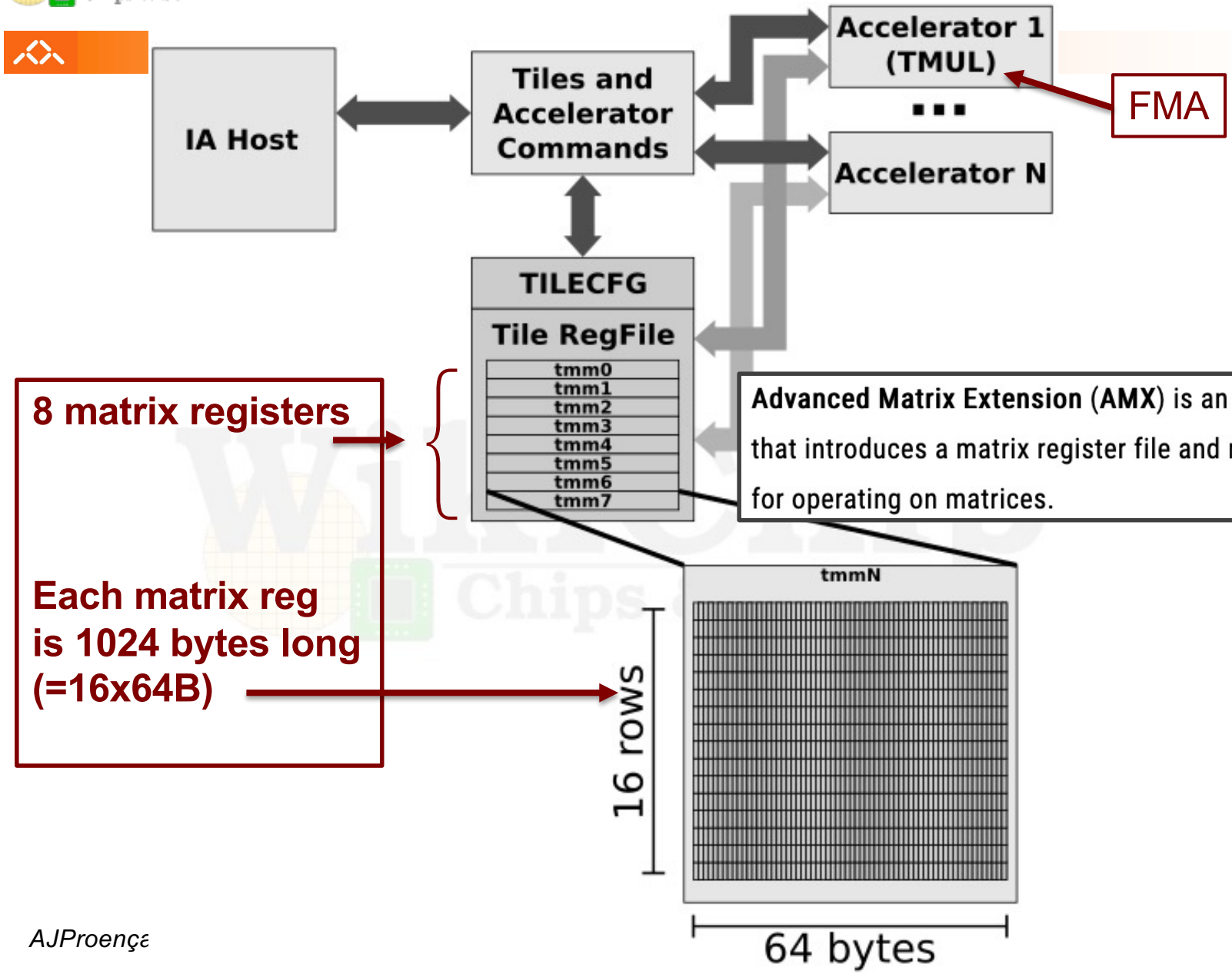
Microarchitecture	Support Level													
	F	CD	ER	PF	BW	DQ	VL	IFMA	VBMI	4FMAPS	VNNI	4VNNIW	VPOPCNTDQ	BF16
Knights Landing	✔	✔	✔	✔	✘	✘	✘	✘	✘	✘	✘	✘	✘	✘
Knights Mill	✔	✔	✔	✔	✘	✘	✘	✘	✘	✔	✔	✔	✔	✘
Skylake (server)	✔	✔	✘	✘	✔	✔	✔	✘	✘	✘	✘	✘	✘	✘
Cascade Lake	✔	✔	✘	✘	✔	✔	✔	✘	✘	✘	✔	✘	✘	✘
Cannon Lake	✔	✔	✘	✘	✔	✔	✔	✔	✔	✘		✘	✘	✘
Ice Lake (server)	✔	✔	✘	✘	✔	✔	✔	✔	✔	✔	✔	✘	✘	✘
Cooper Lake	✔	✔	✘	✘	✔	✔	✔	✘	✘	✘	✔	✘	✘	✔

AVX512F - AVX-512 Foundation
AVX512CD - AVX-512 Conflict Detection
AVX512BW - AVX-512 Byte and Word
AVX512DQ - AVX-512 Doubleword and Quadword
AVX512VL - AVX-512 Vector Length
AVX512IFMA - AVX-512 Integer Fused Multiply-Add
AVX512_VNNI - AVX-512 Vector Neural Network Instructions
AVX512_BF16 - AVX-512 BFloat16 Instructions

"I hope AVX512 dies a painful death, and that Intel starts fixing real problems…"
Linus Torvalds

Intel Advanced Matrix Extension (AMX)

(expected in 2021)



https://fuse.wikichip.org/news/3600/the-x86-advanced-matrix-extension-amx-brings-matrix-operations-to-debut-with-sapphire-rapids/

ARM architecture

ARM (stylised in lowercase as **arm**, previously an acronym for **Advanced RISC Machines** and originally **Acorn RISC Machine**) is a family of reduced instruction set computing (RISC) architectures for computer processors, configured for various environments. Arm Ltd. develops the architecture and licenses it to other companies, who design their own products that implement one of those architectures—including systems-on-chips (SoC) and systems-on-modules (SoM) that

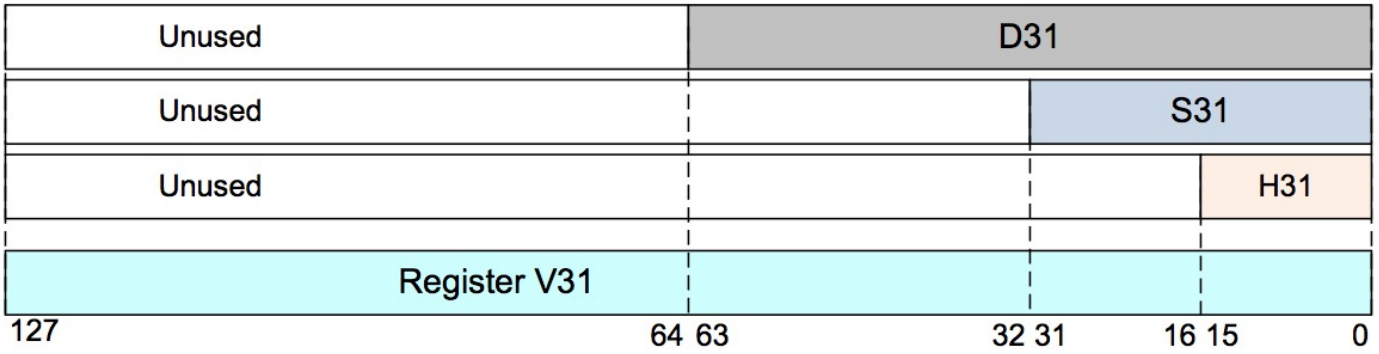
History [edit]

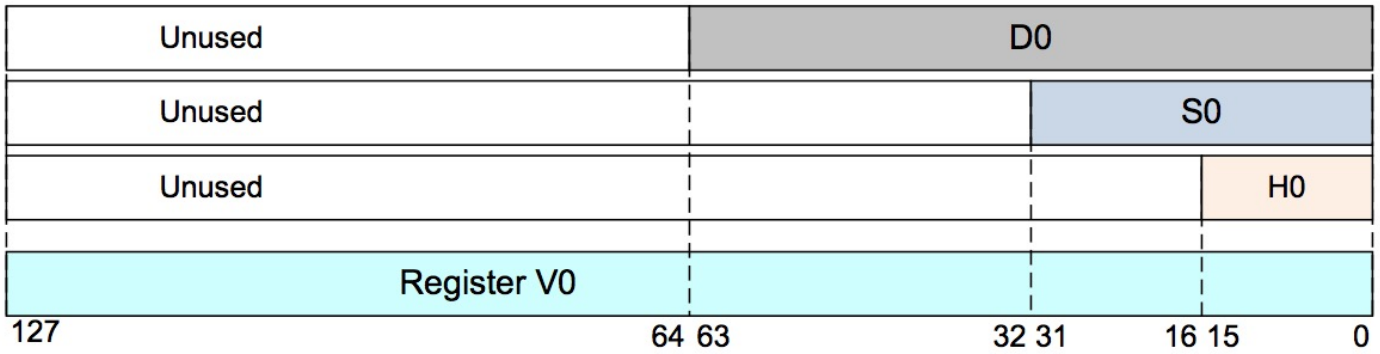
BBC Micro [edit]

Main article: BBC Micro

Acorn Computers' first widely successful design was the BBC Micro, introduced in December 1981. This was a relatively conventional machine based on the MOS 6502 CPU but ran at roughly double the performance of competing designs like the Apple II due to its use of faster DRAM. Typical DRAM of the era ran at about 2 MHz; Acorn arranged a deal with Hitachi for a supply of faster 4 MHz parts.[16]

NEON vector & FP registers in Armv8 (64-bit)





DP register

SP register

HP register

Vector register

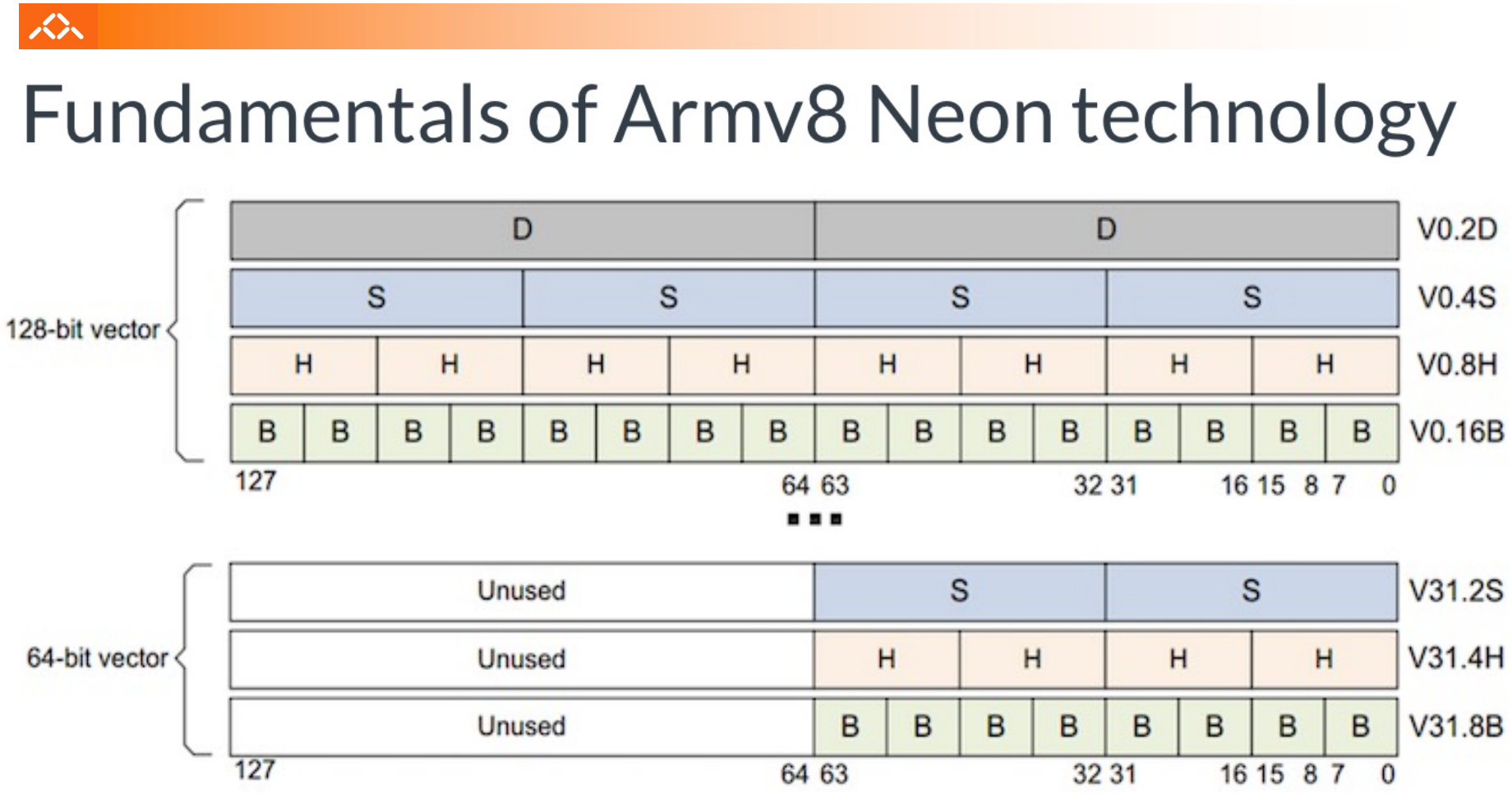
32x 128-bit registers

Figure 4-10 Arrangement of floating-point values

Note

16-bit floating-point is supported, but only as a format to be converted from or to. It is not supported for data processing operations.

NEON vector & FP registers in Armv8 (64-bit)



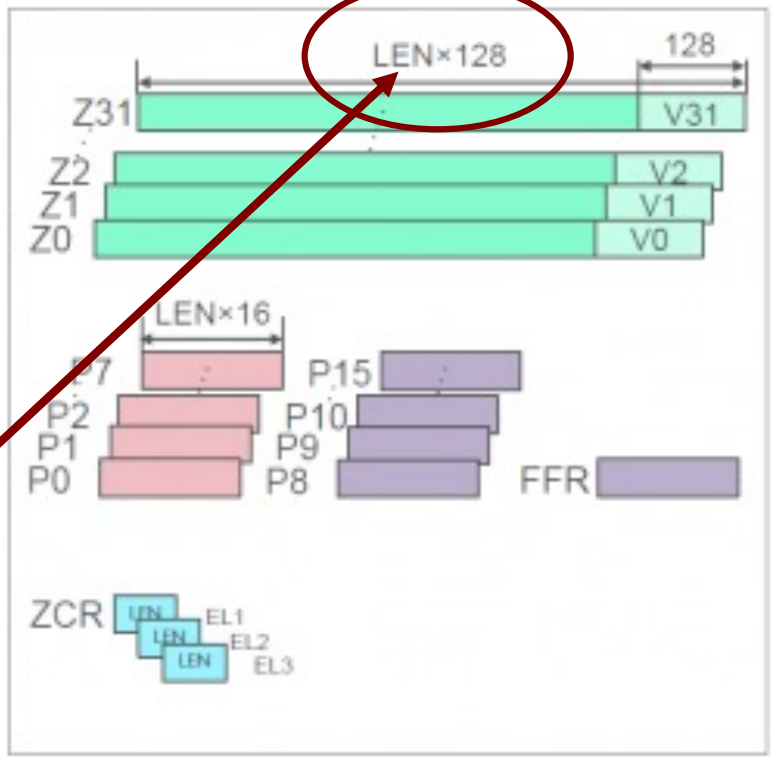
https://developer.arm.com/documentation/102474/0100/Fundamentals-of-Armv8-Neon-technology

Armv8-A Scalable Vector Extension (SVE)

SVE architectural state

Vector size: up to **2048**

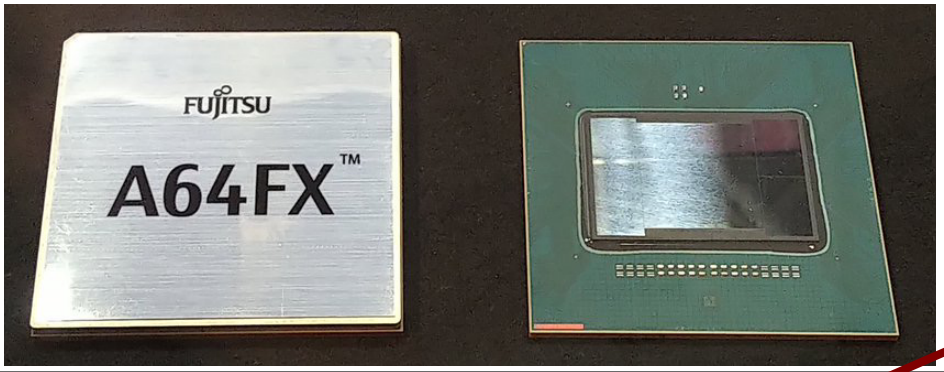
- Scalable vector registers
  - Z0-Z31 extending NEON's V0-V31
    - DP & SP floating-point
    - 64, 32, 16 & 8-bit integer
- Scalable predicate registers
  - P0-P7 lane masks for ld/st/arith
  - P8-P15 for predicate manipulation
  - FFR *first fault register*
- Scalable vector control registers
  - ZCR_ELx vector length (LEN=1..16)
  - Exception / privilege level EL1 to EL3



5 ©ARM 2016

ARM

The 1st implementation of SVE: Fujitsu A64FX



ARM64

Fujitsu's A64FX Arm Chip:
SVE 4x128

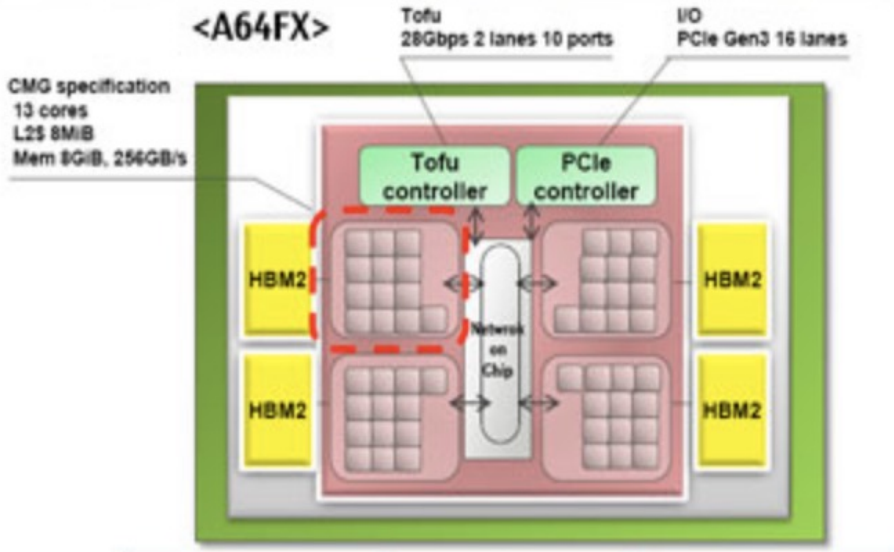
A64FX Chip Overview

FUJITSU

- **Architecture Features**
  - Armv8.2-A (AArch64 only)
  - SVE 512-bit wide SIMD
  - 48 computing cores + 4 assistant cores*

*All the cores are identical

  - HBM2 32GiB
  - Tofu 6D Mesh/Torus
    28Gbps x 2 lanes x 10 ports
  - PCIe Gen3 16 lanes



Beyond vector extensions

- Vector/SIMD-extended architectures are hybrid approaches
  - mix (super)**scalar + vector** op capabilities on a single device
  - **highly pipelined** approach to reduce memory access penalty
  - **tightly-closed access to shared memory**: lower latency
- Evolution of vector/SIMD-extended architectures
  - **computing accelerators optimized for number crunching (GPU)**
  - **add support for matrix multiply + accumulate operations**; **why?**
    - most scientific, engineering, AI & finance applications use matrix computations, namely the dot product: multiply and accumulate the elements in a row of a matrix by the elements in a column from another matrix
    - manufacturers typically call these extension **Tensor Processing Unit** (TPU)
  - **support for half-precision FP & 8-bit integer; why?**
    - machine learning using neural nets is becoming very popular; to compute the model parameter during training phase, intensive matrix products are used and with very low precision (is adequate!)

Reading suggestions

From CAQA 5th Ed

- Concepts and challenges in ILP: section 3.1
- Pipelining: basic and intermediate concepts: App. C
- Exploiting ILP w/ multiple issue & static scheduling: 3.7
- Exploiting ILP w/ dyn sched, multiple issue & specul: 3.8
- Multithread: exploiting TLP on uniprocessors: 3.12
- Vector architecture: 4.2
- SIMD instruction set extensions for multimedia: 4.3
- Graphic processing units: *(later…)* 4.4
- Detecting and enhancing loop-level parallelism: 4.5

Slides on Vector Computing from previous year

- in http://gec.di.uminho.pt/mei/cp/T13-ArqVect.pdf