# Introduction to Commodity SIMD

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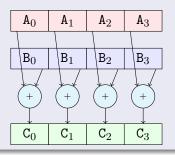
## Single Instruction Multiple Data (SIMD)

- Vector instructions implement the SIMD concept.
- Multiple vector elements are combined in a single operation.

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### Example: <4 x i16> Vector Addition



- Four 16-bit additions at once.
- Replaces the loop:
  - for (int i=0; i<4; i++) {
     C[i] = A[i] + B[i];
    }</pre>
- Intel paddw MMX instruction.

## MMX, SSE, AVX, AVX-512

- 1997: Intel Pentium processors introduce 64-bit SIMD (MMX).
- 1999: 128-bit floating point SIMD vectors (SSE).
- 2000: SSE2 adds 128-bit integer vector operations.
  - Widespread Intel/AMD standard: all x86-64 processors.
- SSE3 (2004), SSSE3 (2005), SSE4 (2008).
- 256-bit SIMD: Intel AVX (2011), AVX2 (2013).
- 512-bit SIMD: Intel AVX-512 (2017)

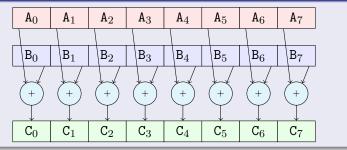
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#### Not Just Intel/AMD

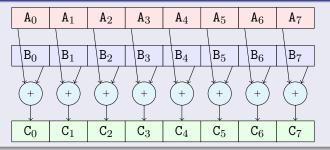
- Altivec/VMX (2004): 128-bit SIMD
- ARM Neon (2008): 128-bit SIMD
- ARM SVE: scalable SIMD up to 2048 bits (in progress)

### 128-bit SSE2: <8 x i16> Vector Addition



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#### AVX2, AVX-512

• 256-bit AVX2: <16 x i16> vector addition in a single operation.

• 512-bit AVX-512: <32 x i16> vector addition in a single operation.

# Why SIMD?

## SIMD Costs

#### • Intel has added hundreds of SIMD instructions over two decades.

- More SIMD instructions added than any other kind.
- Substantial transistor count, chip area devoted to SIMD.
- Larger cores, so fewer cores per package possible.

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## SIMD Benefits

- SIMD naturally supports data parallel applications.
  - Graphics, signal and image processing.
  - Physical simulation.
  - Database queries, financial analytics.
- SIMD has natural advantages over multicore.
  - Cost of instruction fetch/decode divided by SIMD vector length.
  - SIMD ALUs share common control and data path logic.
  - Synchronization of parallel execution is automatic.

### Several New Instruction Families

- AVX-512F: Foundation core 32/64 bit operations (extending AVX).
- AVX-512DQ: New doubleword/quadword (32/64-bit) operations.
- AVX-512BW: AVX-2 byte/word operations extended to 512 bits.
- AVX-512VBMI: Full byte-level permutation selecting from 128 bytes.
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### Systematic New Features

- Systematic masking and blending using bitmask registers.
- Constant parameter broadcasting, rounding and exception control.
- Register count increased from 16 to 32.
- Ternary logic all possible 3-bit Boolean functions.

# Opportunity and Challenge

### Opportunity

- Extensive SIMD parallelism offers the potential to dramatically speed-up applications.
- The expected speed-up is potentally very large.
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## Challenge

- Existing sequential programs generally cannot be autovectorized.
  - Too many sequential dependencies between data elements.
  - Programmer code optimizations often obscure parallelizable logic.
- Language technology may limit access to SIMD capabilities.
- Text processing may involve variable-length items not well-matched to fixed SIMD field and register widths.
- Data parallel algorithmic approaches may be hard to find.