

Lecture 20

Almost Thread Level Parallelism



Da'Miki

Intel Xeon Phi



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1997: THE FIRST INTEL® TERAFLOP COMPUTER
consisted of: and occupied:
9,298 INTEL PROCESSORS | **72** SERVER CABINETS

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will provide: and occupy:
1 TERAFLOP OF PERFORMANCE | **1** PCIe SLOT



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Review

- Flynn Taxonomy of Parallel Architectures
 - SIMD: *Single Instruction Multiple Data*
 - MIMD: *Multiple Instruction Multiple Data*
 - SISD: Single Instruction Single Data
 - MISD: Multiple Instruction Single Data (unused)
- Intel SSE SIMD Instructions
 - One instruction fetch that operates on multiple operands simultaneously
 - 64/128 bit XMM registers
 - (SSE = Streaming SIMD Extensions)
- Threads and Thread-level parallelism



Intel SSE Intrinsics

- Intrinsics are C functions and procedures for putting in assembly language, including SSE instructions
 - With intrinsics, can program using these instructions indirectly
 - One-to-one correspondence between SSE instructions and intrinsics



Example SSE Intrinsics

Intrinsics:

- Vector data type:
`_m128d`
- Load and store operations:
`_mm_load_pd`
`_mm_store_pd`
`_mm_loadu_pd`
`_mm_storeu_pd`
- Load and broadcast across vector
`_mm_load1_pd`
- Arithmetic:
`_mm_add_pd`
`_mm_mul_pd`

Corresponding SSE instructions:

- MOVAPD/aligned, packed double
- MOVAPD/aligned, packed double
- MOVUPD/unaligned, packed double
- MOVUPD/unaligned, packed double
- MOVSD + shuffling/duplicating
- ADDPD/add, packed double
- MULPD/multiple, packed double



Example: 2 x 2 Matrix Multiply

Definition of Matrix Multiply:

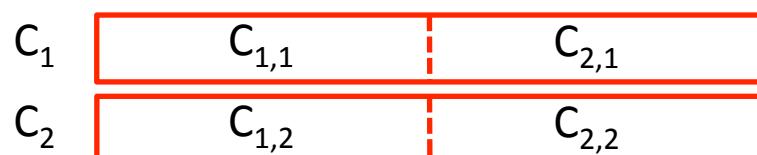
$$C_{i,j} = (A \times B)_{i,j} = \sum_{k=1}^2 A_{i,k} \times B_{k,j}$$

$$\begin{bmatrix} A_{1,1} \\ A_{2,1} \end{bmatrix} \begin{bmatrix} A_{1,2} \\ A_{2,2} \end{bmatrix} \times \begin{bmatrix} B_{1,1} & B_{1,2} \\ B_{2,1} & B_{2,2} \end{bmatrix} = \begin{bmatrix} C_{1,1} = A_{1,1}B_{1,1} + A_{1,2}B_{2,1} \\ C_{2,1} = A_{2,1}B_{1,1} + A_{2,2}B_{2,1} \end{bmatrix} \quad \begin{bmatrix} C_{1,2} = A_{1,1}B_{1,2} + A_{1,2}B_{2,2} \\ C_{2,2} = A_{2,1}B_{1,2} + A_{2,2}B_{2,2} \end{bmatrix}$$
$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \times \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix} = \begin{bmatrix} C_{1,1} = 1*1 + 0*2 = 1 \\ C_{2,1} = 0*1 + 1*2 = 2 \end{bmatrix} \quad \begin{bmatrix} C_{1,2} = 1*3 + 0*4 = 3 \\ C_{2,2} = 0*3 + 1*4 = 4 \end{bmatrix}$$

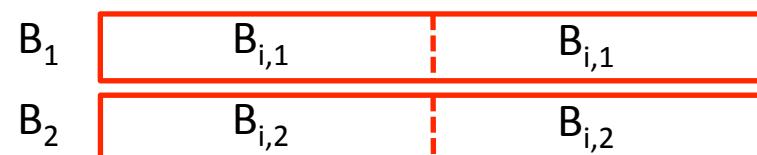


Example: 2 x 2 Matrix Multiply

- Using the XMM registers
 - 64-bit/double precision/two doubles per XMM reg



Stored in memory in Column order



Example: 2 x 2 Matrix Multiply

- Initialization

C_1	0	0
C_2	0	0



Example: 2 x 2 Matrix Multiply

- Initialization

C ₁	0	0
C ₂	0	0

- I = 1

A	A _{1,1}	A _{2,1}
---	------------------	------------------

[`_mm_load_pd`](#): Load 2 doubles into XMM reg, Stored in memory in Column order

B ₁	B _{1,1}	B _{1,1}
B ₂	B _{1,2}	B _{1,2}

[`_mm_load1_pd`](#): SSE instruction that loads a double word and stores it in the high and low double words of the XMM register (duplicates value in both halves of XMM)



Example: 2 x 2 Matrix Multiply

- First iteration intermediate result

C ₁	0+A _{1,1} B _{1,1}	0+A _{2,1} B _{1,1}
C ₂	0+A _{1,1} B _{1,2}	0+A _{2,1} B _{1,2}

c1 = `_mm_add_pd(c1, _mm_mul_pd(a, b1));`
c2 = `_mm_add_pd(c2, _mm_mul_pd(a, b2));`
SSE instructions first do parallel multiplies
and then parallel adds in XMM registers

- | = 1

A	A _{1,1}	A _{2,1}
---	------------------	------------------

`_mm_load_pd`: Stored in memory in Column order

B ₁	B _{1,1}	B _{1,1}
B ₂	B _{1,2}	B _{1,2}

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SSE instructions first do parallel multiplies
and then parallel adds in XMM registers

- | = 2

A	A _{1,2}	A _{2,2}
---	------------------	------------------

`_mm_load_pd`: Stored in memory in Column order

B ₁	B _{2,1}	B _{2,1}
B ₂	B _{2,2}	B _{2,2}

`_mm_load1_pd`: SSE instruction that loads a double word and stores it in the high and low double words of the XMM register (duplicates value in both halves of XMM)



Example: 2 x 2 Matrix Multiply

- Second iteration intermediate result

	$C_{1,1}$	$C_{2,1}$
C_1	$A_{1,1}B_{1,1} + A_{1,2}B_{2,1}$	$A_{2,1}B_{1,1} + A_{2,2}B_{2,1}$
C_2	$A_{1,1}B_{1,2} + A_{1,2}B_{2,2}$	$A_{2,1}B_{1,2} + A_{2,2}B_{2,2}$
	$C_{1,2}$	$C_{2,2}$

- $| = 2$

A	$A_{1,2}$	$A_{2,2}$
---	-----------	-----------

`c1 = _mm_add_pd(c1, _mm_mul_pd(a, b1));`
`c2 = _mm_add_pd(c2, _mm_mul_pd(a, b2));`
SSE instructions first do parallel multiplies
and then parallel adds in XMM registers

`_mm_load_pd`: Stored in memory in
Column order

B ₁	$B_{2,1}$	$B_{2,1}$
B ₂	$B_{2,2}$	$B_{2,2}$

`_mm_load1_pd`: SSE instruction that loads
a double word and stores it in the high and
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Example: 2 x 2 Matrix Multiply (Part 1 of 2)

```
#include <stdio.h>
// header file for SSE compiler intrinsics
#include <emmintrin.h>

// NOTE: vector registers will be represented in
// comments as v1 = [ a | b ]
// where v1 is a variable of type __m128d and
// a, b are doubles

int main(void) {
    // allocate A,B,C aligned on 16-byte boundaries
    double A[4] __attribute__ ((aligned (16)));
    double B[4] __attribute__ ((aligned (16)));
    double C[4] __attribute__ ((aligned (16)));
    int lda = 2;
    int i = 0;
    // declare several 128-bit vector variables
    __m128d c1,c2,a,b1,b2;
```

```
// Initialize A, B, C for example
/* A =
   1 0
   0 1
*/
A[0] = 1.0; A[1] = 0.0; A[2] = 0.0; A[3] = 1.0;

/* B =
   1 3
   2 4
*/
B[0] = 1.0; B[1] = 2.0; B[2] = 3.0; B[3] = 4.0;

/* C =
   0 0
   0 0
*/
C[0] = 0.0; C[1] = 0.0; C[2] = 0.0; C[3] = 0.0;
```



Example: 2 x 2 Matrix Multiply (Part 2 of 2)

```
// used aligned loads to set
// c1 = [c_11 | c_21]
c1 = _mm_load_pd(C+0*lda);
// c2 = [c_12 | c_22]
c2 = _mm_load_pd(C+1*lda);

for (i = 0; i < 2; i++) {
    /* a =
     * i = 0: [a_11 | a_21]
     * i = 1: [a_12 | a_22]
     */
    a = _mm_load_pd(A+i*lda);
    /* b1 =
     * i = 0: [b_11 | b_11]
     * i = 1: [b_21 | b_21]
     */
    b1 = _mm_load1_pd(B+i+0*lda);
    /* b2 =
     * i = 0: [b_12 | b_12]
     * i = 1: [b_22 | b_22]
     */
    b2 = _mm_load1_pd(B+i+1*lda);

    /* c1 =
     * i = 0: [c_11 + a_11*b_11 | c_21 + a_21*b_11]
     * i = 1: [c_11 + a_21*b_21 | c_21 + a_22*b_21]
     */
    c1 = _mm_add_pd(c1,_mm_mul_pd(a,b1));
    /* c2 =
     * i = 0: [c_12 + a_11*b_12 | c_22 + a_21*b_12]
     * i = 1: [c_12 + a_21*b_22 | c_22 + a_22*b_22]
     */
    c2 = _mm_add_pd(c2,_mm_mul_pd(a,b2));
}

// store c1,c2 back into C for completion
_mm_store_pd(C+0*lda,c1);
_mm_store_pd(C+1*lda,c2);

// print C
printf("%g,%g\n%g,%g\n",C[0],C[2],C[1],C[3]);
return 0;
```



Inner loop from gcc -O -S

```
L2: movapd  (%rax,%rsi), %xmm1 //Load aligned A[i,i+1]->m1
    movddup  (%rdx), %xmm0      //Load B[j], duplicate->m0
    mulpd    %xmm1, %xmm0       //Multiply m0*m1->m0
    addpd    %xmm0, %xmm3       //Add m0+m3->m3
    movddup  16(%rdx), %xmm0   //Load B[j+1], duplicate->m0
    mulpd    %xmm0, %xmm1       //Multiply m0*m1->m1
    addpd    %xmm1, %xmm2       //Add m1+m2->m2
    addq     $16, %rax          // rax+16 -> rax (i+=2)
    addq     $8, %rdx           // rdx+8 -> rdx (j+=1)
    cmpq     $32, %rax          // rax == 32?
    jne      L2                 // jump to L2 if not equal
    movapd   %xmm3, (%rcx)     //store aligned m3 into C[k,k+1]
    movapd   %xmm2, (%rdi)     //store aligned m2 into C[l,l+1]
```



You Are Here!

Software

- Parallel Requests
Assigned to computer
e.g., Search “Katz”

- Parallel Threads
Assigned to core
e.g., Lookup, Ads

- Parallel Instructions
>1 instruction @ one time
e.g., 5 pipelined instructions

- Parallel Data
>1 data item @ one time
e.g., Add of 4 pairs of words

- Hardware descriptions
All gates functioning in parallel at same time

Hardware

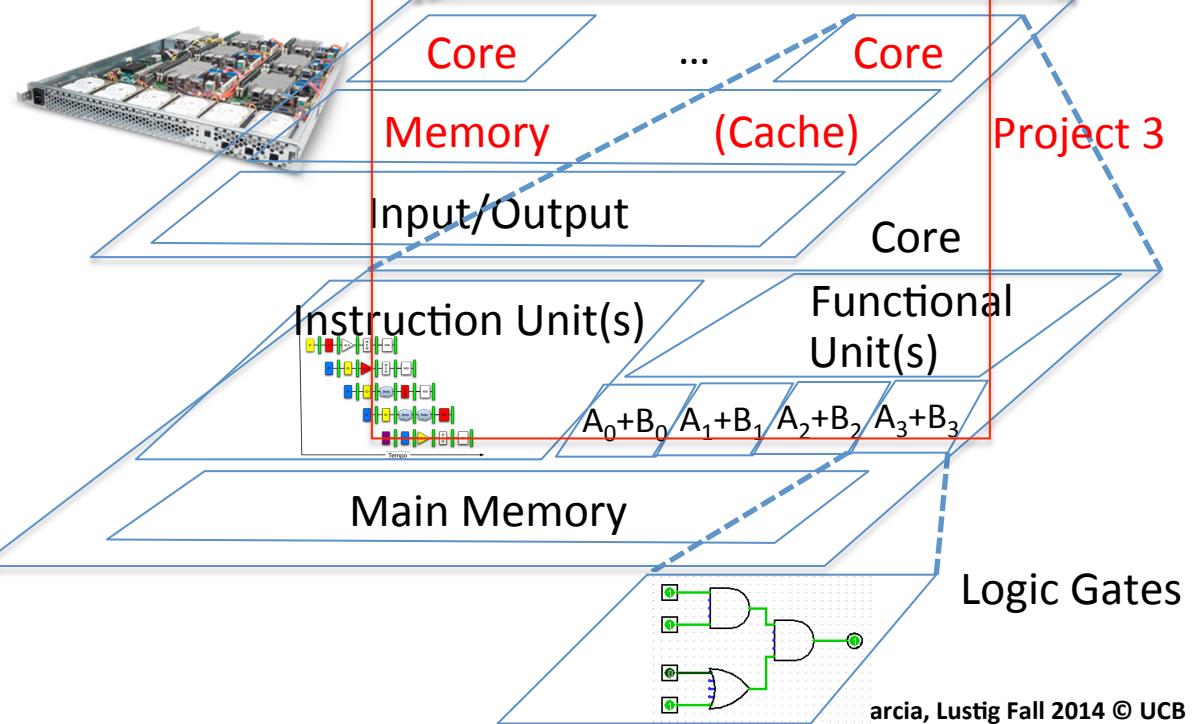
Warehouse Scale Computer



Smart Phone



Harness Parallelism & Achieve High Performance



Thoughts about Threads

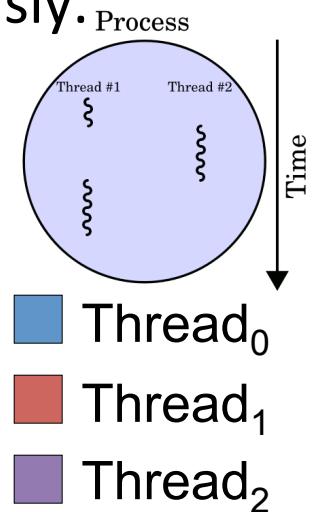
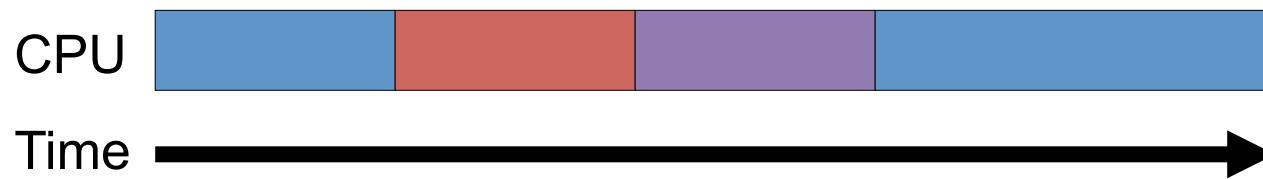


“Although threads seem to be a small step from sequential computation, in fact, they represent a huge step. They discard the most essential and appealing properties of sequential computation: understandability, predictability, and determinism. Threads, as a model of computation, are wildly non-deterministic, and the job of the programmer becomes one of pruning that nondeterminism.”
— *The Problem with Threads*, Edward A. Lee, UC Berkeley, 2006



Background: Threads

- A *Thread* stands for “thread of execution”, is a single stream of instructions
 - A program / process can **split**, or **fork** itself into separate threads, which can (in theory) execute simultaneously.
 - An easy way to describe/think about parallelism
- A single CPU can execute many threads by *Time Division Multiplexing*

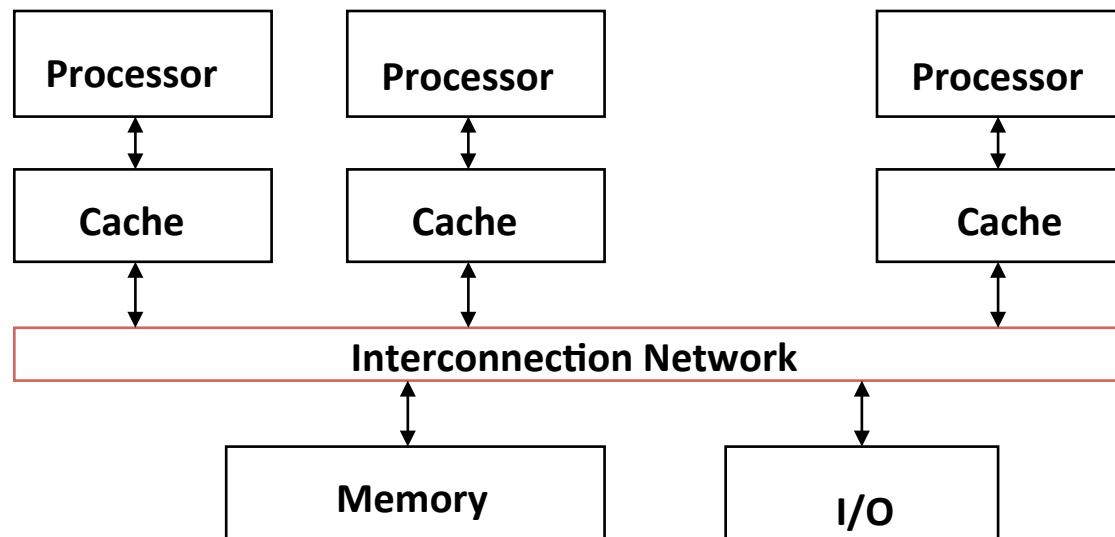


- *Multithreading* is running multiple threads through the same hardware



Parallel Processing: Multiprocessor Systems (MIMD)

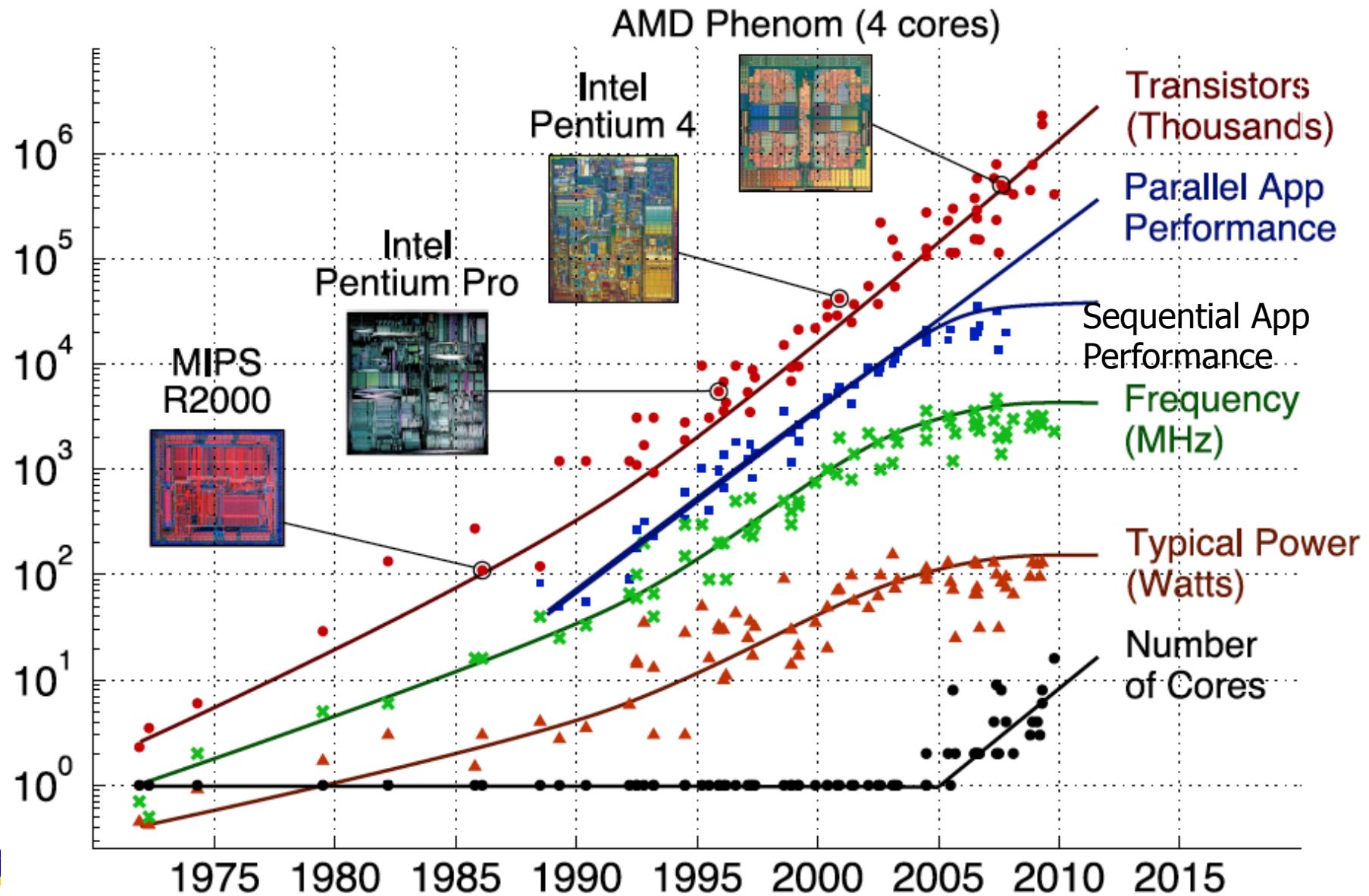
- Multiprocessor (MIMD): a computer system with at least 2 processors



1. Deliver high throughput for independent jobs via job-level parallelism
 2. Improve the run time of a single program that has been specially crafted to run on a multiprocessor - a parallel processing program
- Now Use term **core** for processor ("Multicore") because "Multiprocessor Microprocessor" too redundant



Transition to Multicore



Data partially collected by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond

UCB

Multiprocessors and You

- Only path to performance is parallelism
 - Clock rates flat or declining
 - SIMD: 2X width every 3-4 years
 - 128b wide now, 256b 2011, 512b in 2014?, 1024b in 2018?
 - **Advanced Vector Extensions** are 256-bits wide!
 - MIMD: Add 2 cores every 2 years: 2, 4, 6, 8, 10, ...
- A key challenge is to craft parallel programs that have high performance on multiprocessors as the number of processors increase – i.e., that scale
 - Scheduling, load balancing, time for synchronization, overhead for communication
- Will explore this further in labs and projects



Parallel Performance Over Time

Year	Cores	SIMD bits /Core	Core * SIMD bits	Peak DP FLOPs
2003	2	128	256	4
2005	4	128	512	8
2007	6	128	768	12
2009	8	128	1024	16
2011	10	256	2560	40
2013	12	256	3072	48
2015	14	512	7168	112
2017	16	512	8192	128
2019	18	1024	18432	288
2021	20	1024	20480	320



So, In Conclusion...

- Sequential software is slow software
 - SIMD and MIMD only path to higher performance
- SSE Intrinsics allow SIMD instructions to be invoked from C programs
- MIMD uses multithreading to achieve high parallelism

