

Languages and Abstractions for High-Performance  
Scientific Computing  
CS598 APK

Andreas Kloeckner

Spring 2025

# Outline

## Introduction

Notes

Notes (unfilled, with empty boxes)

Notes (source code on Github)

About This Class

Why Bother with Parallel Computers?

Lowest Accessible Abstraction: Assembly

Architecture of an Execution Pipeline

Architecture of a Memory System

Shared-Memory Multiprocessors

## Machine Abstractions

Performance: Expectation, Experiment, Observation

Performance Oriented Languages and Abstractions

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## Why this class?

- ▶ Setting: Performance-Constrained Code  
When is a code performance-constrained?

A desirable quality (fidelity/capability) is limited by computational cost on a given computer.

- ▶ If your code is performance-constrained, what is the *best* approach?

Use a more efficient method/algorithm.

- ▶ If your code is performance-constrained, what is the *second-best* approach?

Ensure the current algorithm uses your computer efficiently. Observe that this is a *desperate measure*.

## Examples of Performance-Constrained Codes

[4cm]

- ▶ Simulation codes
  - ▶ Weather/climate models
  - ▶ Oil/gas exploration
  - ▶ Electronic structure
  - ▶ Electromagnetic design
  - ▶ Aerodynamic design
  - ▶ Molecular dynamics / biological systems
  - ▶ Cryptanalysis
- ▶ Machine Learning
- ▶ Data Mining

Discussion:

- ▶ In what way are these codes constrained?
- ▶ How do these scale in terms of the problem size?

## What Problem are we Trying To Solve?

$$(C_{ij})_{i,j=1}^{m,n} = \sum_{k=1}^{\ell} A_{ik} B_{kj}$$

- ▶ [Reference BLAS DGEMM code](#)
- ▶ [OpenBLAS DGEMM code](#)

[Demo: intro/DGEMM Performance](#)

Demo Instructions: Compare OpenBLAS against Fortran BLAS on large square matrix

## Goals: What are we Allowed to Ask For?

- ▶ Goal: “make efficient use of the machine”
- ▶ In general: not an easy question to answer
- ▶ In theory: limited by *some* peak machine throughput
  - ▶ Memory Access
  - ▶ Compute
- ▶ In practice: many other limits (Instruction cache, TLB, memory hierarchy, NUMA, registers)

# Class web page

<https://bit.ly/hpcabstr-s25>

contains:

- ▶ Class outline
- ▶ Slides/demos/materials
- ▶ Assignments
- ▶ Discussion Forum
- ▶ Grading Policies
- ▶ Video
- ▶ HW1 (soon)

# Welcome Survey

Please go to:

<https://bit.ly/hpcabstr-s25>

and click on 'Start Activity'.

If you are seeing this later, you can find the activity at [Activity:  
welcome-survey](#).



# Grading / Workload

Four components:

- ▶ Homework: 25%
- ▶ Paper Presentation: 25%
  - ▶ 30 minutes (two per class)
  - ▶ Presentation sessions scheduled throughout the semester
  - ▶ Paper list on web page
  - ▶ Sign-up survey: soon
- ▶ Paper Reactions: 10%
- ▶ Computational Project: 40%

## Open Source <3

These notes (and the accompanying demos) are open-source!

Bug reports and pull requests welcome:

<https://github.com/inducer/hpc-lang-abstr-notes>

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## Approaches to High Performance

- ▶ Libraries (seen)
- ▶ Black-box Optimizing Compilers
- ▶ Compilers with Directives
- ▶ Code Transform Systems
- ▶ “Active Libraries”

Q: Give examples of the latter two.

- ▶ Code Transform System: [CHiLL](#)
- ▶ Active Library: [PyTorch](#)

## Libraries: A Case Study

$$(C_{ij})_{i,j=1}^{m,n} = \sum_{k=1}^{\ell} A_{ik} B_{kj}$$

### Demo: intro/DGEMM Performance

- ▶ Demo Instructions: Compare OpenBLAS on large square and small odd-shape matrices
- ▶ Demo Instructions: Also consider DGEMV version

## Do Libraries Stand a Chance? (in general)

- ▶ Tremendously successful approach — Name some examples

(e.g.) LAPACK, Eigen, UMFPACK, FFTW, Numpy, Deal.ii

- ▶ Saw: Three simple integer parameters suffice to lose 'good' performance

- ▶ Recent efforts: e.g. [Batch BLAS](#)

- ▶ Separation of Concerns

Example: Finite differences – e.g. implement  $\partial_x$ ,  $\partial_y$ ,  $\partial_z$  as separate (library) subroutines — What is the problem?

Data locality: data should be traversed once,  $\partial_x$ ,  $\partial_y$ ,  $\partial_z$  computed together

Separation of concerns: each op. traverses data separately

- ▶ Flexibility and composition

## (Black-Box) Optimizing Compiler: Challenges

Why is black-box optimizing compilation so difficult?

- ▶ Application developer knowledge lost
    - ▶ Simple example: “Rough” matrix sizes
    - ▶ Data-dependent control flow
    - ▶ Data-dependent access patterns
    - ▶ Activities of other, possibly concurrent parts of the program
    - ▶ Profile-guided optimization can recover some knowledge
  - ▶ Obtain proofs of required properties
  - ▶ Size of the search space
- Consider <http://polaris.cs.uiuc.edu/publications/padua.pdf>

# Directive-Based Compiler: Challenges

What is a directive-based compiler?

Demo Instructions: [Show l2dformta\\_qbx from pyfmmlib/vec\\_wrappers.f90.](#)

- ▶ Generally same as optimizing compiler
- ▶ Make use of extra promises made by the user
- ▶ What should the user promise?
- ▶ Ideally: feedback cycle between compiler and user
  - ▶ Often broken in both directions
  - ▶ User may not know what the compiler did
  - ▶ Compiler may not be able to express what it needs
- ▶ Directives: generally not mandatory

## Lies, Lies Everywhere

- ▶ Semantics form a contract between programmer and language/environment
- ▶ Within those bounds, implementation has full freedom
- ▶ True at every level:
  - ▶ Assembly
  - ▶ “High-level” language (C)

Give examples of lies at these levels:

- ▶ Assembly: Concurrent execution
- ▶ “High-level” language (C): (e.g.) strength reduction, eliminated ops

One approach: *Lie to yourself*

- ▶ “Domain-specific languages” ← A fresh language, I can do what I want!
- ▶ Consistent semantics are notoriously hard to develop
  - ▶ Especially as soon as you start allowing subsets of even (e.g.) C’s



# Class Outline

## High-level Sections:

- ▶ Intro, Armchair-level Computer Architecture
- ▶ Machine Abstractions
- ▶ Performance: Expectation, Experiment, Observation
- ▶ Programming Languages for Performance
- ▶ Program Representation and Optimization Strategies
- ▶ Code Generation/JIT

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# Moore's Law

## Moore's Law: The number of transistors on microchips has doubled every two years

Moore's law describes the empirical regularity that the number of transistors on integrated circuits doubles approximately every two years. This advancement is important for other aspects of technological progress in computing – such as processing speed or the price of computers.

Our World  
in Data

### Transistor count

50,000,000,000

10,000,000,000

5,000,000,000

1,000,000,000

500,000,000

100,000,000

50,000,000

10,000,000

5,000,000

1,000,000

500,000

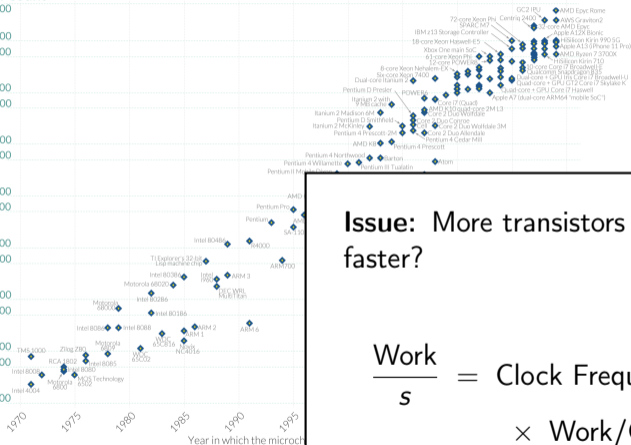
100,000

50,000

10,000

5,000

1,000



Issue: More transistors =  
faster?

$$\frac{\text{Work}}{s} = \text{Clock Frequency} \times \text{Work/Clock}$$

Data source: Wikipedia ([wikipedia.org/wiki/Transistor\\_count](https://wikipedia.org/wiki/Transistor_count))  
OurWorldInData.org – Research and data to make progress against the world's largest problem

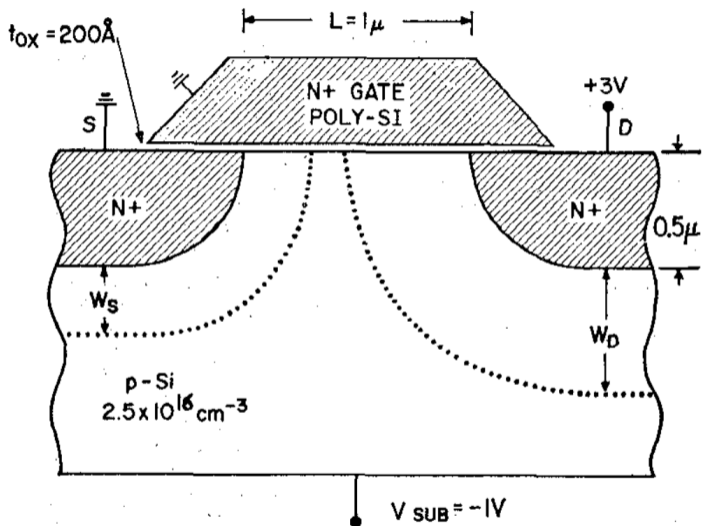
## Dennard Scaling of MOSFETs

Parameter	Factor
Dimension	$1/\kappa$
Voltage	$1/\kappa$
Current	$1/\kappa$
Capacitance	$1/\kappa$
Delay Time	$1/\kappa$
Power dissipation/circuit	$1/\kappa^2$
Power density	1

[Dennard et al. '74, via Bohr '07]

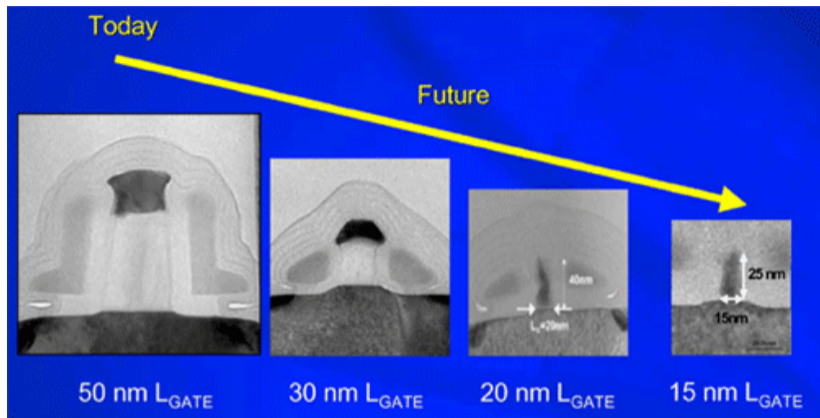
- ▶ Frequency = Delay time<sup>-1</sup>

# MOSFETs ("CMOS" – "complementary" MOS): Schematic



[Dennard et al. '74]

# MOSFETs: Scaling



[Intel Corp.]

- ▶ 'New' problem at small scale:  
Sub-threshold leakage (due to low voltage, small structure)  
Dennard scaling is over – and has been for a while.

## Peak Architectural Instructions per Clock: Intel

CPU	IPC	Year
Pentium 1	1.1	1993
Pentium 3	1.9	1999
Pentium 4 (Gallatin)	1.9	20
Pentium D	2	2005
Pentium M	2.5	2003
Core 2	3	2006
Sandy Bridge...	<4	2011
Skylake	<4	2015
Golden Cove	<6	2021
Lion Cove	<8	2024

[Charlie Brey <http://brej.org/blog/?p=15>, Wikipedia, Intel]

Context: [Lemire: simdjson achieved IPC, '19](#)

Discuss: How do we get out of this dilemma?

## The Performance Dilemma

- ▶ IPC: Brick-ish Wall
- ▶ Clock Frequency: Brick Wall

Ideas:

- ▶ Make one instruction do more copies of the same thing (“SIMD”)
- ▶ Use copies of the same processor (“SPMD”/“MPMD”)

Question: What is the *conceptual* difference between those ideas?

- ▶ SIMD executes multiple program instances in lockstep.
- ▶ SPMD has no synchronization assumptions.



## The Performance Dilemma: Another Look

- ▶ **Really:** A crisis of the 'starts-at-the-top-ends-at-the-bottom' programming model
- ▶ **Tough luck:** Most of our codes are written that way
- ▶ **Even tougher luck:** Everybody on the planet is *trained* to write codes this way

So:

- ▶ **Need:** Different tools/abstractions to write those codes

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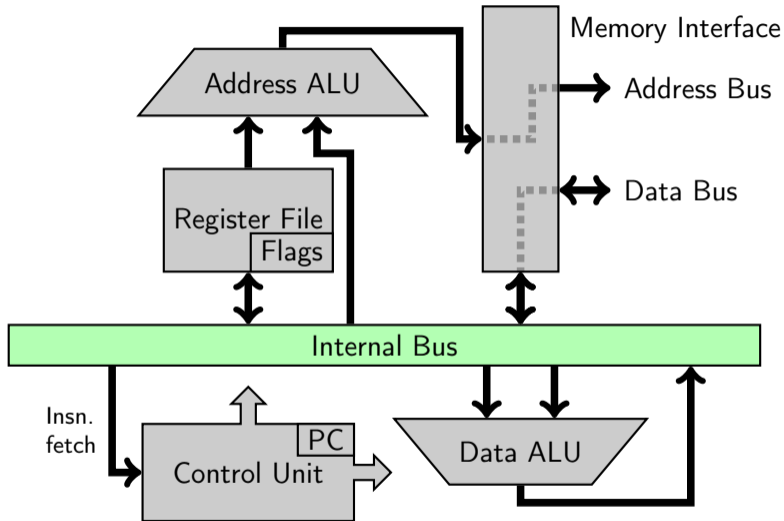
Shared-Memory Multiprocessors

## Machine Abstractions

Performance: Expectation, Experiment, Observation

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## A Basic Processor: Closer to the Truth



- ▶ loosely based on Intel 8086
- ▶ What's a [bus](#)?

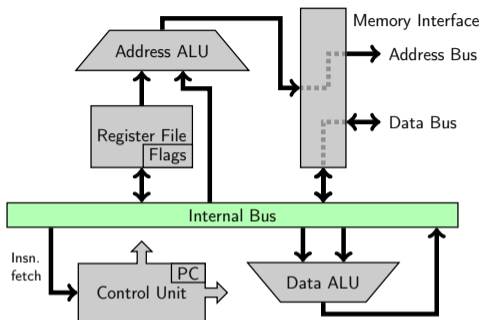
## A Very Simple Program

```
int a = 5;          4:   c7 45 f4 05 00 00 00 movl   $0x5,-0xc(%rbp)
int b = 17;         b:   c7 45 f8 11 00 00 00 movl   $0x11,-0x8(%rbp)
int z = a * b;     12:  8b 45 f4                mov    -0xc(%rbp),%eax
                  15:  0f af 45 f8            imul  -0x8(%rbp),%eax
                  19:  89 45 fc                mov    %eax,-0x4(%rbp)
                  1c:  8b 45 fc                mov    -0x4(%rbp),%eax
```

Things to know:

- ▶ Question: Which is it?
  - ▶ `<opcode> <src>, <dest>`
  - ▶ `<opcode> <dest>, <src>`
- ▶ [Addressing modes](#) (Immediate, Register, Base plus Offset)
- ▶ [0xHexadecimal](#)

## A Very Simple Program: Another Look



```
4:  c7 45 f4 05 00 00 00 movl  $0x5,-0xc(%rbp)
b:  c7 45 f8 11 00 00 00 movl  $0x11,-0x8(%rbp)
12: 8b 45 f4                mov   -0xc(%rbp),%eax
15: 0f af 45 f8            imul -0x8(%rbp),%eax
19: 89 45 fc                mov   %eax,-0x4(%rbp)
1c: 8b 45 fc                mov   -0x4(%rbp),%eax
```

## A Very Simple Program: Intel Form

```
4:      c7 45 f4 05 00 00 00      mov     DWORD PTR [rbp-0xc],0x5
b:      c7 45 f8 11 00 00 00      mov     DWORD PTR [rbp-0x8],0x11
12:     8b 45 f4                    mov     eax,DWORD PTR [rbp-0xc]
15:     0f af 45 f8                imul   eax,DWORD PTR [rbp-0x8]
19:     89 45 fc                    mov     DWORD PTR [rbp-0x4],eax
1c:     8b 45 fc                    mov     eax,DWORD PTR [rbp-0x4]
```

- ▶ “Intel Form”: (you might see this on the net)  
<opcode> <sized dest>, <sized source>
- ▶ Previous: “AT&T Form”: (we’ll use this)
- ▶ Goal: Reading comprehension.
- ▶ Don’t understand an opcode?

[https://en.wikipedia.org/wiki/X86\\_instruction\\_listings](https://en.wikipedia.org/wiki/X86_instruction_listings)

## Assembly Loops

```
int main()                                0:    55                push   %rbp
{                                           1:    48 89 e5          mov    %rsp,%rbp
    int y = 0, i;                          4:    c7 45 f8 00 00 00 00  movl  $0x0,-0x8(%rbp)
    for (i = 0;                             b:    c7 45 fc 00 00 00 00  movl  $0x0,-0x4(%rbp)
        y < 10; ++i)                       12:   eb 0a            jmp   1e <main+0x1e>
        y += i;                             14:   8b 45 fc          mov   -0x4(%rbp),%eax
    return y;                               17:   01 45 f8          add   %eax,-0x8(%rbp)
}                                           1a:   83 45 fc 01      addl  $0x1,-0x4(%rbp)
                                           1e:   83 7d f8 09      cmpl  $0x9,-0x8(%rbp)
                                           22:   7e f0            jle  14 <main+0x14>
                                           24:   8b 45 f8          mov   -0x8(%rbp),%eax
                                           27:   c9              leaveq
                                           28:   c3              retq
```

Things to know:

- ▶ [Condition Codes \(Flags\)](#): Zero, Sign, Carry, etc.
- ▶ [Call Stack](#): Stack frame, stack pointer, base pointer
- ▶ [ABI](#): Calling conventions

# Demos

## [Demo: intro/Assembly Reading Comprehension](#)

Demo: Source-to-assembly mapping

Code to try:

```
int main()  
{  
    int y = 0, i;  
    for (i = 0; y < 100; ++i)  
        y += i*i;  
    return y;  
}
```

Also try <https://godbolt.org> for direct source-to-assembly mapping



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## Modern Processors?

All of this can be built in about 4000 transistors.

(e.g. MOS 6502 in Apple II, Commodore 64, Atari 2600)

So what exactly are Intel/ARM/AMD/Nvidia doing with the other **billions** of transistors?

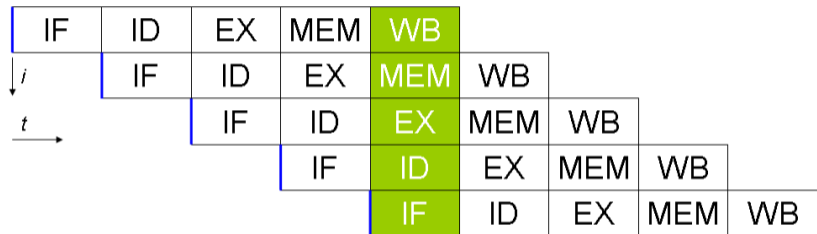
# Execution in a Simple Processor



- ▶ [IF] Instruction fetch
- ▶ [ID] Instruction Decode
- ▶ [EX] Execution
- ▶ [MEM] Memory Read/Write
- ▶ [WB] Result Writeback

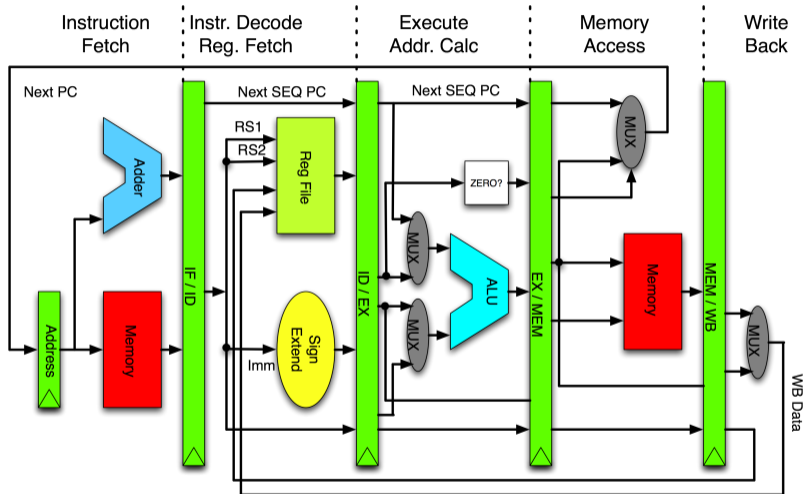
[Wikipedia ©]

## Solution: Pipelining



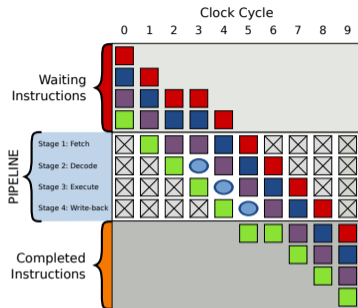
[Wikipedia ©]

# MIPS Pipeline: 110,000 transistors



[Wikipedia ©]

# Hazards and Bubbles



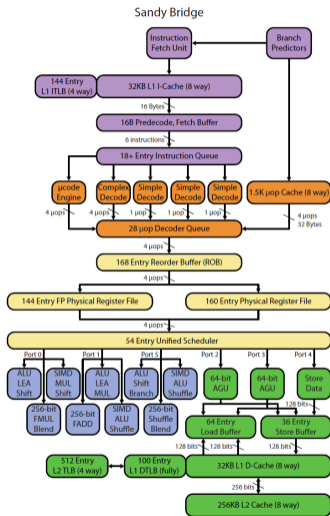
Q: Types of Pipeline Hazards? (aka: what can go wrong?)

- ▶ Data
- ▶ Structural
- ▶ Control

## Demo: intro/Pipeline Performance Mysteries

- ▶ a, a: elapsed time 3.83603 s
- ▶ a, b: elapsed time 2.58667 s
- ▶ a, a unrolled: elapsed time 3.83673 s
- ▶ aa, bb unrolled: elapsed time 1.92509 s
- ▶ a, b unrolled: elapsed time 1.92084 s

# A Glimpse of a More Modern Processor

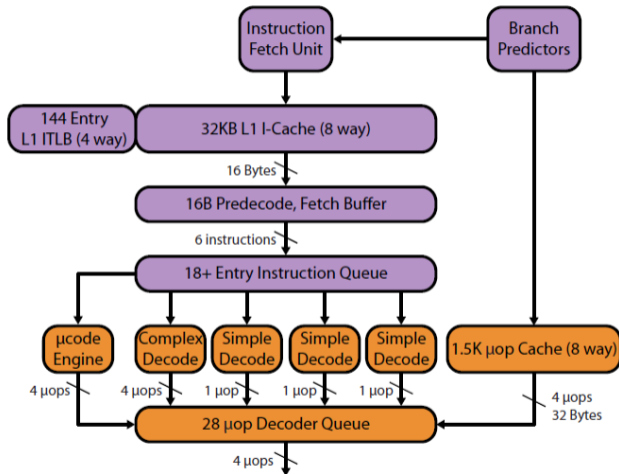


[David Kanter / Realworldtech.com]



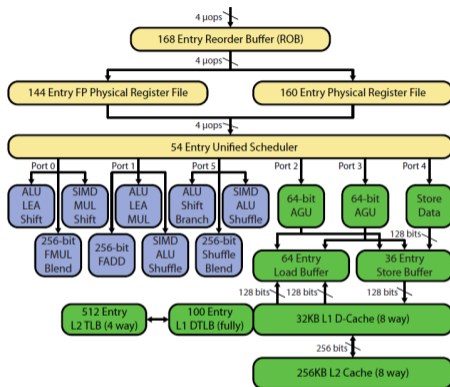
# A Glimpse of a More Modern Processor: Frontend

## Sandy Bridge



[David Kanter / Realworldtech.com]

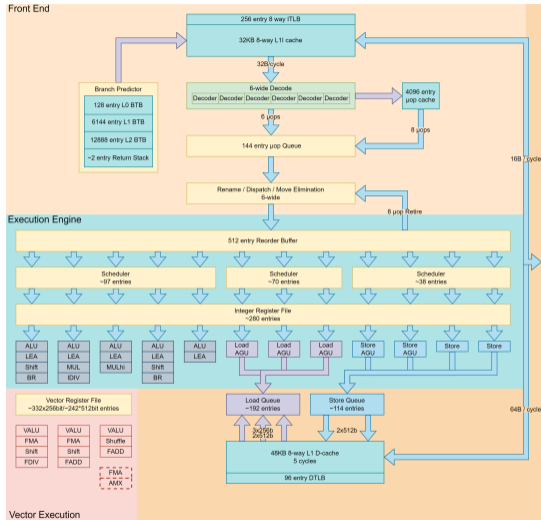
# A Glimpse of a More Modern Processor: Backend



- ▶ **New concept:** Instruction-level parallelism (“ILP”, “superscalar”)
- ▶ Where does the IPC number from earlier come from?

[David Kanter / Realworldtech.com]

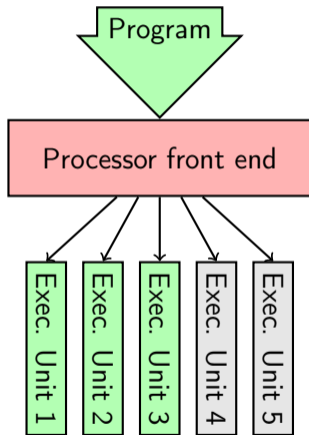
# A Glimpse of a More Modern Processor: Golden Cove



# Demo

[Demo: intro/More Pipeline Mysteries](#)

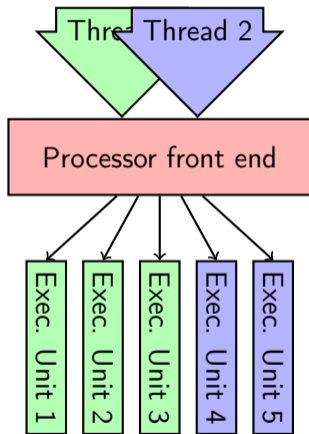
# SMT/"Hyperthreading"



Q: Potential issues?

- ▶  $n \times$  the cache demand!
- ▶ Power?
- ▶ Some people just turn it off and manage their own ILP.

## SMT/"Hyperthreading"



Q: Potential issues?

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## More Bad News from Dennard

Parameter	Factor
Dimension	$1/\kappa$
Line Resistance	$\kappa$
Voltage drop	$\kappa$
Response time	1
Current density	$\kappa$

[Dennard et al. '74, via Bohr '07]

- ▶ The above scaling law is for on-chip interconnects.
- ▶ Current  $\sim$  Power vs. response time

Getting information from

- ▶ processor to memory
- ▶ one computer to the next

is

- ▶ slow (in *latency*)
- ▶ power-hungry



## Somewhere Behind the Interconnect: Memory

Performance characteristics of memory:

- ▶ Bandwidth
- ▶ Latency

*Flops are cheap*

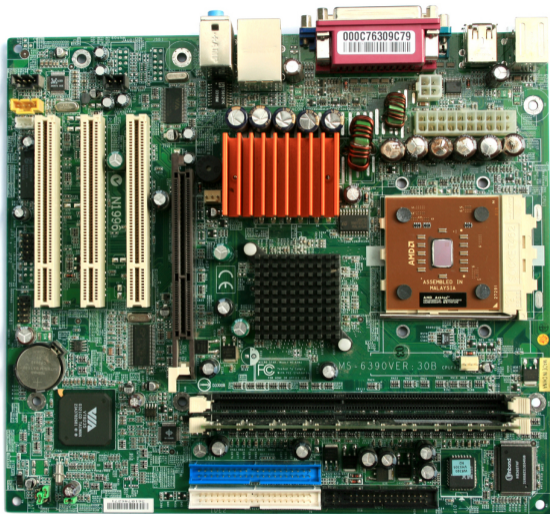
*Bandwidth is money*

*Latency is physics — M. Hoemmen*

Minor addition (but important for us)?

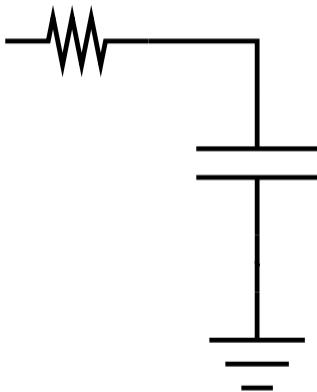
- ▶ Bandwidth is money **and code structure**

# Latency is Physics: Distance

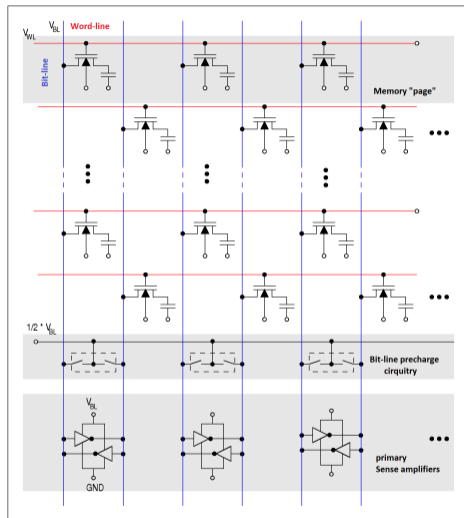


[Wikipedia ©]

## Latency is Physics: Electrical Model



# Latency is Physics: DRAM



[Wikipedia ©]

## Latency is Physics: Performance Impact?

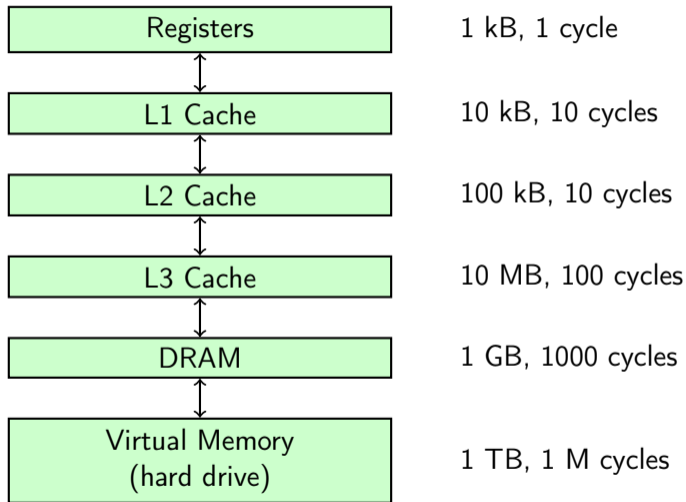
What is the performance impact of high memory latency?

Processor stalled, waiting for data.

Idea:

- ▶ Put a look-up table of recently-used data onto the chip.
- ▶ Cache

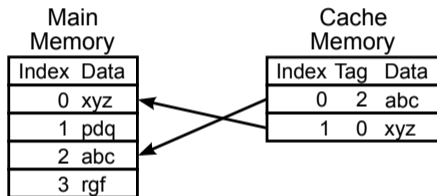
## Memory Hierarchy



## A Basic Cache

Demands on cache implementation:

- ▶ Fast, small, cheap, low power
- ▶ Fine-grained
- ▶ High “hit”-rate (few “misses”)



Design Goals: at odds with each other. Why?

Address matching logic expensive

[Wikipedia ©]

# Caches: Engineering Trade-Offs

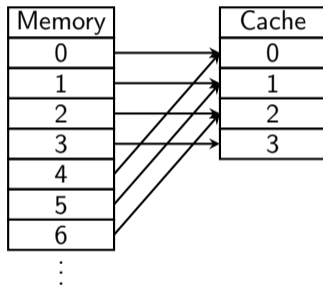
## Engineering Decisions:

- ▶ More data per unit of access matching logic  
→ Larger “Cache Lines”
- ▶ Simpler/less access matching logic  
→ Less than full “Associativity”
- ▶ Eviction strategy
- ▶ Size

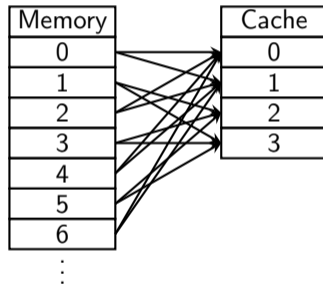


# Associativity

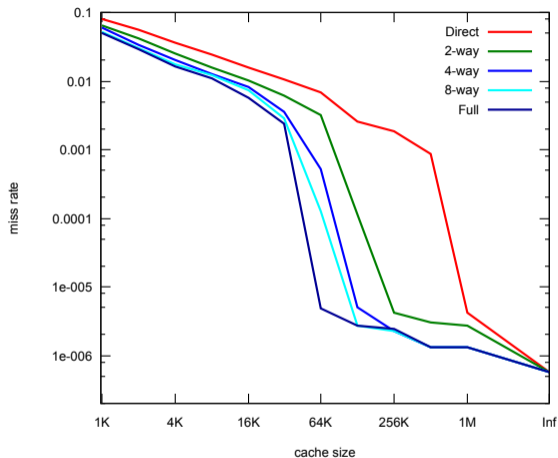
Direct Mapped:



2-way set associative:



## Size/Associativity vs Hit Rate



Miss rate versus cache size on the Integer portion of SPEC CPU2000  
[Cantin, Hill 2003]

## Demo: Learning about Caches

[Demo: intro/Cache Organization on Your Machine](#)

## Experiments: 1. Strides: Setup

```
int go(unsigned count, unsigned stride)
{
    const unsigned array_size = 64 * 1024 * 1024;
    int *ary = (int *) malloc(sizeof(int) * array_size);

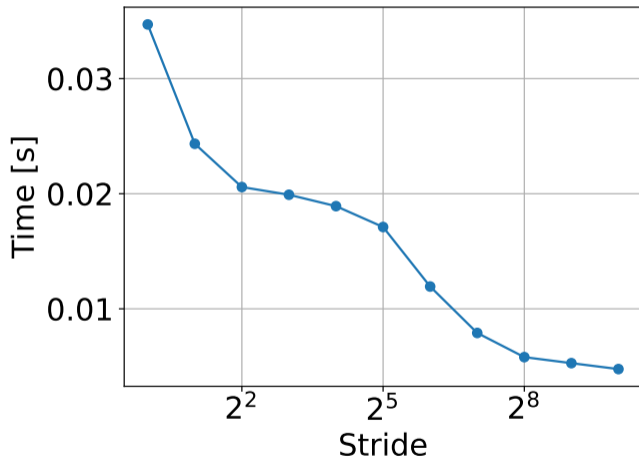
    for (unsigned it = 0; it < count; ++it)
    {
        for (unsigned i = 0; i < array_size; i += stride)
            ary[i] *= 17;
    }

    int result = 0;
    for (unsigned i = 0; i < array_size; ++i)
        result += ary[i];

    free(ary);
    return result;
}
```

What do you expect? [[Ostrovsky '10](#)]

## Experiments: 1. Strides: Results



## Experiments: 2. Bandwidth: Setup

```
int go(unsigned array_size, unsigned steps)
{
    int *ary = (int *) malloc(sizeof(int) * array_size);
    unsigned asm1 = array_size - 1;

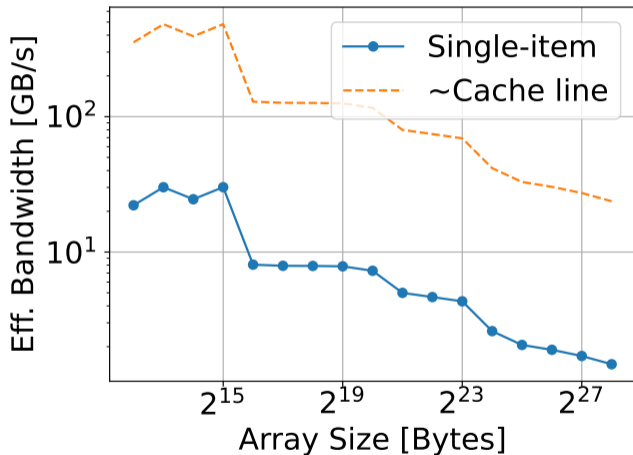
    for (unsigned i = 0; i < 100*steps;)
    {
        #define ONE ary[(i++*16) & asm1] ++;
        #define FIVE ONE ONE ONE ONE ONE ONE
        #define TEN FIVE FIVE
        #define FIFTY TEN TEN TEN TEN TEN
        #define HUNDRED FIFTY FIFTY
        HUNDRED
    }

    int result = 0;
    for (unsigned i = 0; i < array_size; ++i)
        result += ary[i];

    free(ary);
    return result;
}
```

What do you expect? [Ostrovsky '10]

## Experiments: 2. Bandwidth: Results



## Experiments: 3. A Mystery: Setup

```
int go(unsigned array_size, unsigned stride, unsigned steps)
{
    char *ary = (char *) malloc(sizeof(int) * array_size);

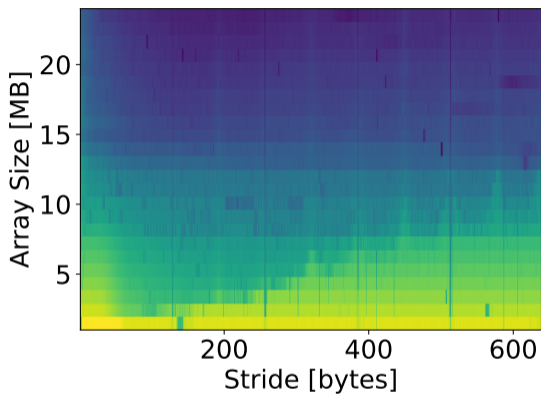
    unsigned p = 0;
    for (unsigned i = 0; i < steps; ++i)
    {
        ary[p] ++;
        p += stride;
        if (p >= array_size)
            p = 0;
    }

    int result = 0;
    for (unsigned i = 0; i < array_size; ++i)
        result += ary[i];

    free(ary);
    return result;
}
```



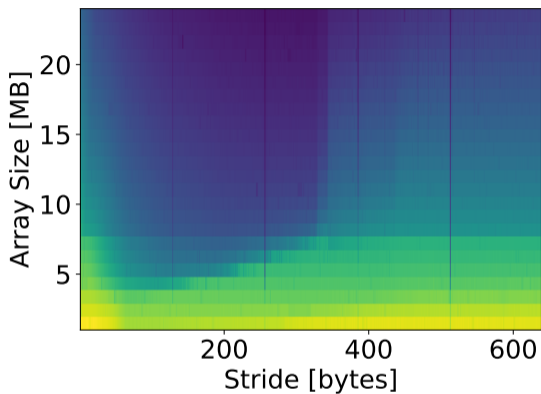
## Experiments: 3. A Mystery: Results



Color represents achieved bandwidth:

- ▶ Yellow: high
- ▶ Blue: low

## Experiments: 3. A Mystery: Results on Sandy Bridge



Color represents achieved bandwidth:

- ▶ Yellow: high
- ▶ Blue: low

## Thinking about the Memory Hierarchy

- ▶ What is a **working set**?
- ▶ What is **data locality** of an algorithm?
- ▶ What does this have to do with caches?

## Case Study: Streaming Workloads

Q: Estimate expected throughput for saxpy on an architecture with caches. What are the right units?

$$z_i = \alpha x_i + y_i \quad (i = 1, \dots, n)$$

- ▶ Units: GBytes/s
- ▶ Net memory accessed:  $n \times 4 \times 3$  bytes
- ▶ Actual memory accessed:  $n \times 4 \times 4$  bytes  
(To read  $z$  read into the cache before modification)

Demo: [https://github.com/lcw/stream\\_ispc](https://github.com/lcw/stream_ispc)

## Special Store Instructions

At least two aspects to keep apart:

- ▶ Temporal Locality: Are we likely to refer to this data again soon? (*non-temporal* store)
- ▶ Spatial Locality: Will (e.g.) the entire cache line be overwritten? (*streaming* store)

What hardware behavior might result from these aspects?

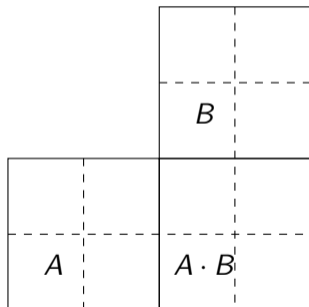
- ▶ Non-temporal: Write past cache entirely (/invalidate), or evict soon
- ▶ Spatial: Do not fetch cache line before overwriting

- ▶ Comment on what a compiler can promise on these aspects.
- ▶ Might these 'flags' apply to loads/prefetches?

(see also: [McCalpin '18])

## Case study: Matrix-Matrix Mult. ('MMM'): Code Structure

- ▶ How would you structure a high-performance MMM?
- ▶ What are sources of concurrency?
- ▶ What should you consider your working set?



- ▶ Sources of concurrency: row, column loop, summation loop (?)
- ▶ Working set: artificially created blocks
- ▶ Provide enough concurrency: SIMD, ILP, Core

## Case study: Matrix-Matrix Mult. ('MMM'): Code Structure

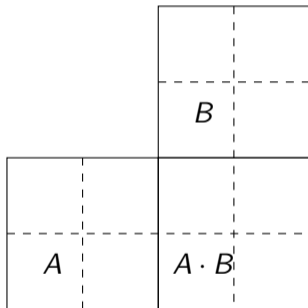
```
for (i)
  for (j)
    for (k)
      c[i,j] += a[i,k]*b[k,j]
```

After:

```
for (itile)
  for (jtile)
    for (ktile)
      for (iinner)
        for (jinner)
          for (kinner)
            c[itile*NB+iinner, ...]
              += ...
```

How many tiles?

$$(N/N_B)^3$$



## Case study: Matrix-Matrix Mult. ('MMM') via Latency

Cost model for MMM in a two-level hierarchy based on latency?

Avg latency per access =

$$(1 - \text{Miss ratio}) \cdot \text{Cache Latency}$$

$$+ \text{Miss ratio} \cdot \text{Mem Latency}$$

Assume: Working set fits in cache, No conflict misses

Calculation (for a single tile):

▶ Total accesses:  $4N_B^3$  ( $N_B$ : block size)

▶ Misses:  $3N_B^2$

▶ Miss rate:

$$\frac{3}{4N_B \cdot \text{cache line size}}$$

[Yotov et al. '07]



## Case study: Matrix-Matrix Mult. ('MMM') via Bandwidth

Cost model for MMM in a two-level hierarchy based on bandwidth?

- ▶ FMA throughput:  $16 \times 2$  SP FMAs per clock (e.g.)
- ▶ Cycle count:  $2N^3/(2 \cdot 32) = N^3/32$
- ▶ Required cache bandwidth:  
(words accessed)/(cycles) =  $4N^3/(N^3/32) = 128$  floats/cycle  
(GB/s?)
- ▶ Total mem. data motion:  
# blocks  $\cdot 4 \cdot$  (block size) =  $(N/N_B)^3 \cdot 4N_B^2 = 4N^3/N_B$
- ▶ Required mem. bandwidth:  
(Mem.motion)/(cycles) =  $4N^3/N_B/(N^3/32) = 128/N_B$   
floats/cycle (GB/s?)
- ▶ What size cache do we need to get to feasible memory bandwidth?

## Case study: Matrix-Matrix Mult. ('MMM'): Discussion

**Discussion:** What are the main simplifications in each model?

Latency:

- ▶ Concurrency/parallelism of memory accesses
- ▶ Miss assumptions
- ▶ (HW) prefetching

Bandwidth:

- ▶ Miss assumptions
- ▶ Multiple cache levels
- ▶ Latency effects

[\[Yotov et al. '07\]](#)

**Q:** How can we analyze cache cost of algorithms in general?

## Hong/Kung: Red/Blue Pebble Game

Simple means of I/O cost analysis: “Red/blue pebble game”

- ▶ A way to quantify I/O cost on a DAG (why a DAG?)
- ▶ “Red Hot” pebbles: data that can be computed on
- ▶ “Blue Cool” pebbles: data that is stored, but not available for computation without I/O

Game Setup:

- ▶ Place blue pebbles on inputs

One ‘turn’:

- ▶ Swap red  $\rightarrow$  blue or blue  $\rightarrow$  red
- ▶ Place red pebble on successor if all predecessors have red
- ▶ Remove any pebble

**Note:** Can allow “Red/Purple/Blue/Black”: more levels

[[Hong/Kung '81](#)]

## Hong/Kung: Red/Blue Pebble Game

Q: What are the cost metrics in this model?

- ▶ I/O Cost: Turn a red into a blue pebble and vice versa
- ▶ Number of red pebbles (corresponding to size of 'near' storage)

What does a Hong/Kung lower bound for Matrix-Matrix look like?

$$\# \text{ turns} \cdot \sqrt{\# \text{red pebbles}} = \Omega(N^3)$$

## Cache-Oblivious Algorithms

**Annoying chore:** Have to pick multiple machine-adapted block sizes in cache-adapted algorithms, one for each level in the memory hierarchy, starting with registers.

**Idea:**

- ▶ Step 1: Express algorithm recursively in divide & conquer manner
  - ▶ Step 2: Pick a strategy to decrease block size
- Give examples of block size strategies, e.g. for MMM:

- ▶ All dimensions
- ▶ Largest dimension

**Result:**

- ▶ Asymptotically optimal on Hong/Kung metric

## Cache-Oblivious Algorithms: Issues

Observed results?

- ▶ 1% of peak if used all the way to scalar level
- ▶ With good base case: 30-60% of peak

What are potential issues on actual hardware?

- ▶ Function call overhead
- ▶ Not conducive to prefetching
- ▶ Too many levels: 1-ish level may suffice

[[Yotov et al. '07](#)]

## Recall: Big-O Notation

Classical Analysis of Algorithms (e.g.):

$$\text{Cost}(n) = O(n^3).$$

Precise meaning? Anything missing from that statement?

**Missing:** 'as  $n \rightarrow \infty$ '

There exists a  $C$  and an  $N_0$  independent of  $n$  so that for all  $n \geq N_0$ ,

$$\text{Cost}(n) \leq C \cdot n^3.$$

## Comment: “Asymptotically Optimal”

Comments on asymptotic statements about cost in relation to high performance?

- ▶ No statement about finite  $n$
- ▶ No statement about the constant

**Net effect:** Having an understanding of asymptotic cost is *necessary*, but *not sufficient* for high performance.

HPC is in the business of minimizing  $C$  in:

$$\text{Cost}(n) \leq C \cdot n^3 \quad (\text{for all } n)$$



# Alignment

*Alignment* describes the process of matching the base address of:

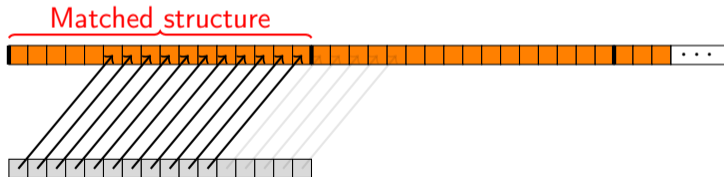
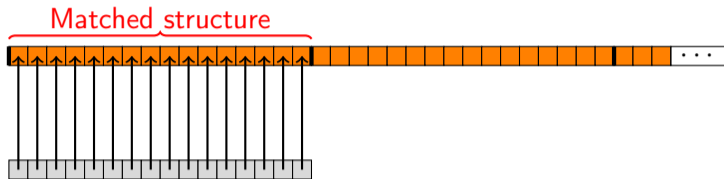
- ▶ Single word: double, float
- ▶ SIMD vector
- ▶ Larger structure

To machine granularities:

- ▶ Natural word size
- ▶ Vector size
- ▶ Cache line
- ▶ Memory page

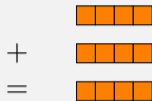
Q: What is the performance impact of misalignment?

# Performance Impact of Misalignment



## SIMD: Basic Idea

What's the basic idea behind SIMD?



What architectural need does it satisfy?

- ▶ Insufficient instruction decode/dispatch bandwidth
- ▶ Tack more operations onto one decoded instruction

Typically characterized by width of data path:

- ▶ SSE: 128 bit (4 floats, 2 doubles)
- ▶ AVX-2: 256 bit (8 floats, 4 doubles)
- ▶ AVX-512: 512 bit (16 floats, 8 doubles)

## SIMD: Architectural Issues

Realization of inter-lane comm. in SIMD? Find instructions. ([Intel Intrinsic Guide](#))

- ▶ Misaligned stores/loads? (no)
- ▶ Broadcast, Unpack+Interleave, Shuffle, Permute
- ▶ Reductions ("horizontal")

Name tricky/inefficient aspects in terms of expressing SIMD:

- ▶ Divergent control flow, Masking, Reconvergence
- ▶ Indirect addressing: gather/scatter
- ▶ Thought experiment: SIMD recursion possible?

## SIMD: Transposes

Why are transposes important? Where do they occur?

- ▶ Whenever SIMD encounters a mismatched data layout
- ▶ For example: MMM of two row-major matrices

Example implementation aspects:

- ▶ HPTT: [[Springer et al. '17](#)]
- ▶ github: [springer13/hptt](#) [8x8 transpose microkernel](#)
- ▶ Q: Why 8x8?

# Outline

## Introduction

Notes

Notes (unfilled, with empty boxes)

Notes (source code on Github)

About This Class

Why Bother with Parallel Computers?

Lowest Accessible Abstraction: Assembly

Architecture of an Execution Pipeline

Architecture of a Memory System

**Shared-Memory Multiprocessors**

## Machine Abstractions

Performance: Expectation, Experiment, Observation

Performance Oriented Languages and Abstractions

## Multiple Cores vs Bandwidth

Assume:

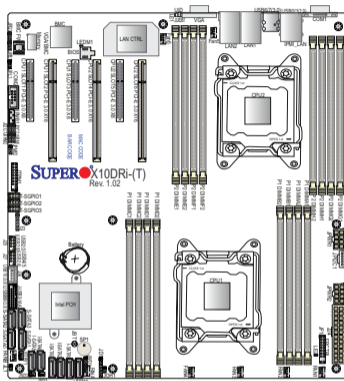
- ▶ memory latency of 100 ns
- ▶ peak DRAM bandwidth of 100 GB/s (per socket)

How many cache lines should be/are in flight at one time?

- ▶  $100\text{ns} \cdot 100\text{GB}/s = 10,000\text{bytes}$
- ▶ About 156 cache lines
- ▶ Historic oops: Older Intel hardware could only handle about 10 pending requests per core at one time
- ▶  $10 \cdot 64/100\text{ns} \approx 6.4\text{GB}/s$

[\[McCalpin '18\]](#)

# Topology and NUMA



[SuperMicro Inc. '15]

Demo:

- ▶ Show `lstopo` on porter, from [hwloc](#).
- ▶ [lstopo on MI300](#)



## Placement and Pinning

Who decides on what core my code runs? How?

- ▶ The OS scheduler: “Oh, hey, look! A free core!”
- ▶ You, explicitly, by **pinning**:
  - ▶ `OMP_PLACES=cores, pthread_setaffinity_np()`

Who decides on what NUMA node memory is allocated?

- ▶ `malloc` uses 'first touch'
- ▶ *You* can decide explicitly (through `libnuma`)

[Demo: intro/NUMA and Bandwidths](#)

What is the main expense in NUMA?

Latency (but it impacts bandwidth by way of queuing)

# Cache Coherence

What is *cache coherence*?

- ▶ As soon as you make a copy of (cache) something, you risk inconsistency with the original
- ▶ A set of guarantees on how (and in what order) changes to memory become visible to other cores

How is cache coherence implemented?

- ▶ Snooping
- ▶ Protocols, operating on cache line states (e.g. "[MESI](#)")

## [Cache coherence simulation](#)

What are the performance impacts?

- ▶ [Demo: intro/Threads vs Cache](#)
- ▶ [Demo: intro/Lock Contention](#)

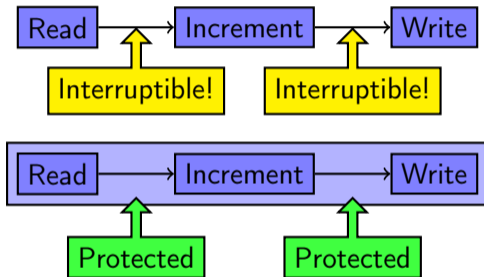
## MESI Sequence to Try

- ▶ CPU0: read a0
- ▶ CPU0: read a0
- ▶ CPU0: write a0
- ▶ CPU0: write a0
- ▶ CPU1: read a0
- ▶ CPU1: write a0
- ▶ CPU1: write a0
- ▶ CPU0: write a0
- ▶ CPU0: write a2
- ▶ CPU0: write a0

Simplifications in the model?

- ▶ Access size = cache line size

## 'Conventional' vs Atomic Memory Update



# Outline

Introduction

## Machine Abstractions

C

OpenCL/CUDA

Convergence, Differences in Machine Mapping

Lower-Level Abstractions: SPIR-V, PTX

Performance: Expectation, Experiment, Observation

Performance-Oriented Languages and Abstractions

Polyhedral Representation and Transformation

# Outline

Introduction

## Machine Abstractions

C

OpenCL/CUDA

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Polyhedral Representation and Transformation

## Atomic Operations: Compare-and-Swap

```
#include <stdatomic.h>
_Bool atomic_compare_exchange_strong(
    volatile A* obj, C* expected, C desired );
```

What does `volatile` mean?

Memory may change at any time, do not keep in register.

What does this do?

- ▶ `(*obj == *expected) ? desired : *obj → *obj.`
- ▶ Return true iff memory contents was as expected.

How might you use this to implement atomic FP multiplication?

Read previous, perform operation, try CAS, maybe retry

# Memory Ordering

Why is Memory Ordering a Problem?

- ▶ Out-of-order CPUs reorder memory operations
- ▶ Compilers reorder memory operations

What's the purpose of different memory orders?

- ▶ Atomicity itself is unaffected
- ▶ Makes sure that 'and then' is meaningful



## Memory ordering semantics

- ▶ `mo_relaxed`
- ▶ `mo_consume`: no r/w reorder *before* in this thread, *data-dependent* releasing writes elsewhere visible (often: compiler only)
- ▶ `mo_acquire`: no r/w reorder *before* in this thread, *all* releasing writes elsewhere visible
- ▶ `mo_release`: no r/w reorder *after* in this thread, *all* writes here visible elsewhere if they acquire the written var
- ▶ `mo_acq_rel`
- ▶ `mo_seq_cst`: `mo_acq_rel` plus a single total order exists (~store queue flush on x86)

## Example: A Semaphore With Atomics

```
#include <stdatomic.h> // mo_→memory_order, a_→atomic
typedef struct { atomic_int v;} naive_sem_t;
void sem_down(naive_sem_t *s)
{
    while (1) {
        while (a_load_explicit(&(s->v), mo_acquire) < 1)
            spinloop_body();
        int tmp=a_fetch_add_explicit(&(s->v), -1, mo_acq_rel);
        if (tmp >= 1)
            break; // we got the lock
        else // undo our attempt
            a_fetch_add_explicit(&(s->v), 1, mo_relaxed);
    }
}
void sem_up(naive_s_t *s) {
    a_fetch_add_explicit(&(s->v), 1, mo_release);
}
```

## C: What is 'order'?

### C11 Committee Draft, December '10, Sec. 5.1.2.3, "Program execution":

- ▶ (3) *Sequenced before* is an asymmetric, transitive, pair-wise relation between evaluations executed by a single thread, which induces a partial order among those evaluations. Given any two evaluations A and B, if A is sequenced before B, then the execution of A shall precede the execution of B. (Conversely, if A is sequenced before B, then B is sequenced after A.) If A is not sequenced before or after B, then A and B are unsequenced. Evaluations A and B are *indeterminately sequenced* when A is sequenced either before or after B, but it is unspecified which. The presence of a sequence point between the evaluation of expressions A and B implies that every value computation and side effect associated with A is sequenced before every value computation and side effect associated with B. (A summary of the *sequence points* is given in annex C.)

Q: Where is this definition used (in the standard document)?

In defining the semantics of atomic operations.

## C: What is 'order'? (Encore)

### C11 Draft, 5.1.2.4 "Multi-threaded executions and data races":

- ▶ All modifications to a particular atomic object M occur in some particular total order, called the *modification order* of M.
- ▶ An evaluation A *carries a dependency* to an evaluation B if ...
- ▶ An evaluation A is *dependency-ordered* before an evaluation B if ...
- ▶ An evaluation A *inter-thread happens before* an evaluation B if ...
- ▶ An evaluation A *happens before* an evaluation B if ...

Why is this so subtle?

- ▶ Many common optimizations depend on the ability to reorder operations.
- ▶ Two options:
  1. Lose the ability to do those optimizations
  2. Specify precisely how much of the order should be externally observable

## C: How Much Lying is OK?

### C11 Committee Draft, December '10, Sec. 5.1.2.3, "Program execution":

- ▶ (1) The semantic descriptions in this International Standard describe the behavior of an abstract machine in which issues of optimization are irrelevant.
- ▶ (2) Accessing a volatile object, modifying an object, modifying a file, or calling a function that does any of those operations are all *side effects*, which are changes in the state of the execution environment.  
[...]

## C: How Much Lying is OK?

- ▶ (4) In the abstract machine, all expressions are evaluated as specified by the semantics. An actual implementation need not evaluate part of an expression if it can deduce that its value is not used and that no needed side effects are produced (including any caused by calling a function or accessing a volatile object).
- ▶ (6) The least requirements on a conforming implementation are:
  - ▶ Accesses to volatile objects are evaluated strictly according to the rules of the abstract machine.
  - ▶ At program termination, all data written into files shall be identical to the result that execution of the program according to the abstract semantics would have produced.
  - ▶ The input and output dynamics of interactive devices shall take place as specified in 7.21.3. The intent of these requirements is that unbuffered or line-buffered output appear as soon as possible, to ensure that prompting messages actually appear prior to a program waiting for input.

This is the observable behavior of the program.

# Arrays

Why are **arrays** the dominant data structure in high-performance code?

- ▶ Performance is mostly achieved with *regular, structured* code (e.g. SIMD, rectangular loops)
- ▶ Arrays are a natural fit for that type of code
- ▶ Abstractions of linear algebra map directly onto arrays

Any comments on C's arrays?

- ▶ 1D arrays: fine, no surprises
- ▶  $n$ D arrays: basically useless: sizes baked into types
  - ▶ Interestingly: Fortran is (incrementally) smarter

## Arrays vs Abstraction

Arrays-of-Structures or Structures-of-Arrays? What's the difference? Give an example.

- ▶ Example: Array of XYZ coordinates:
  - ▶ XYZXYZXYZ...
  - ▶ XXX...YYY...ZZZ...
- ▶ Which of these will be suitable for SIMD? (e.g. computing a norm?)
- ▶ Structures-of-Arrays if at all possible – to expose regularity

Language aspects of the distinction? Salient example?

- ▶ C struct forces you into arrays-of-structures
  - ▶ AoS: more “conceptually sound”
  - ▶ SoA: better for performance
- ▶ Complex numbers



## C and Multi-Dimensional Arrays: A Saving Grace

*// YES:*

```
void f(int m, int n, double (*)[m][n]);
```

*// NO:*

```
struct ary {  
    int m;  
    int n;  
    double (*array)[m][n];  
};
```

*// YES:*

```
struct ary {  
    int m;  
    int n;  
    double a[];  
};
```

# SIMD

Name language mechanisms for SIMD:

- ▶ Inline Assembly
- ▶ Intrinsics
- ▶ Vector Types `typedef int v4si __attribute__((vector_size (16)))`;
- ▶ `#pragma simd`
- ▶ Merging of scalar program instances (in hw/sw)

[Demo: machabstr/Ways to SIMD](#)

## Outer-Loop/inner-Loop Vectorization

Contrast *outer-loop* vs *inner-loop* vectorization.

- ▶ Inner-loop: Inner-most loop vectorized
- ▶ Outer loop: Vectorize a whole kernel. Requires:
  - ▶ Changed memory layout
  - ▶ Must be able to express *all* control flow

**Side q:** Would you consider GPUs outer- or inner-loop-vectorizing?

## Alignment: How?

The old way:

```
int __attribute__((aligned (8))) a_int;
```

Difference between these two?

```
int __attribute__((aligned (8))) * ptr_t_1;  
int * __attribute__((aligned (8))) ptr_t_2;
```

The 'new' way (C/C++11):

```
struct alignas(64) somestruct_t { /* ... */ };  
struct alignas(alignof(other_t))  
    somestruct_t { /* ... */ };  
struct  
    alignas(  
        std::hardware_destructive_interference_size)  
        somestruct_t { /* ... */ };
```

What is *constructive interference*?

## Alignment: Why?

What is the concrete impact of the constructs on the previous slide?

- ▶ Compiler needs to *know* whether data is aligned
  - ▶ Generate the correct instructions (which encode alignment promises)
  - ▶ Stack-allocate memory of the correct alignment
- ▶ Heap-allocated memory needs to actually satisfy the alignment promise!
  - ▶ `posix_memalign`
  - ▶ Hack it by overallocating
  - ▶ In `numpy`: overallocate in bytes, get base address, offset, obtain view

# Pointers and Aliasing

[Demo: machabstr/Pointer Aliasing](#)

# Register Pressure

What if the register working set gets larger than the registers can hold?  
What is the performance impact?

- ▶ “Register Spill”: save/reload code being generated
- ▶ CPU: L1 is relatively fast
- ▶ Other architectures: can be quite dramatic

[Demo: machabstr/Register Pressure](#)

# Object-Oriented Programming

Object-oriented programming: The weapon of choice for encapsulation and separation of concerns!

Performance perspective on OOP?

- ▶ Fine-grain OOP leads to an AoS disaster
- ▶ Long expressions create many temporaries
  - ▶ Memory traffic
- ▶ Return values
- ▶ Run-time polymorphism (virtual methods) lead to fine-grain flow control

**Summary:** No good, very bad. *Must* have sufficient granularity to offset cost.

[Demo: machabstr/Object Orientation vs Performance](#)



# Being Nice to Your Compiler

Some rules of thumb:

- ▶ Use indices rather than pointers
- ▶ Extract common subexpressions
- ▶ Make functions static
- ▶ Use `const`
- ▶ Avoid store-to-load dependencies

What are the concrete impacts of doing these things?

# Outline

Introduction

## Machine Abstractions

C

OpenCL/CUDA

Convergence, Differences in Machine Mapping

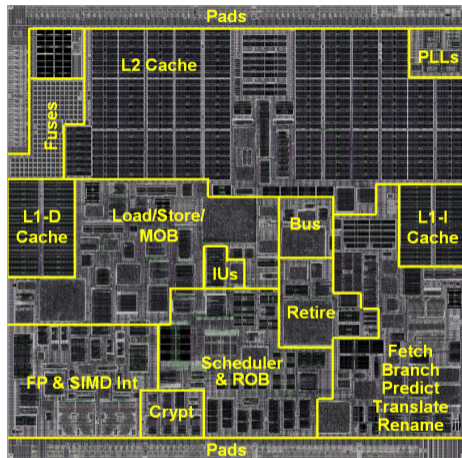
Lower-Level Abstractions: SPIR-V, PTX

Performance: Expectation, Experiment, Observation

Performance-Oriented Languages and Abstractions

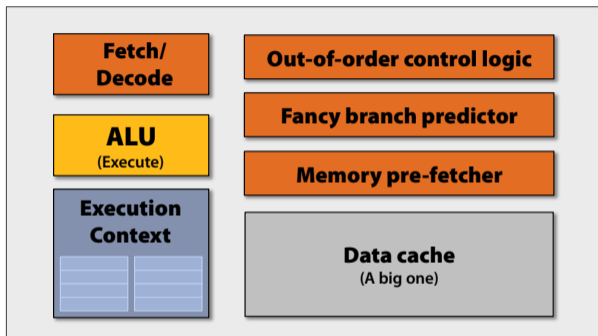
Polyhedral Representation and Transformation

# Chip Real Estate



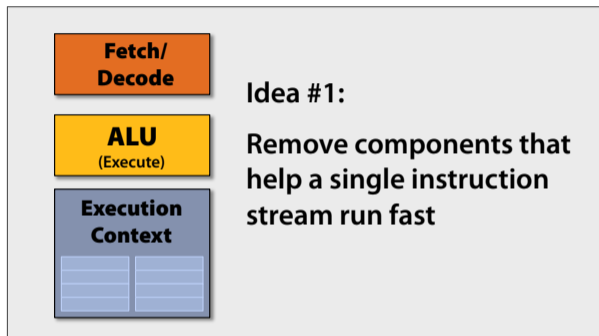
*Die floorplan:* VIA Isaiah (2008).  
65 nm, 4 SP ops at a time, 1 MiB L2.

## “CPU-style” Cores



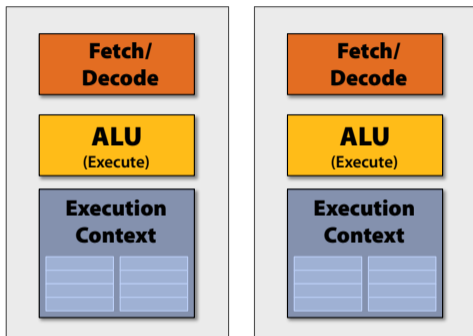
[Fatahalian '08]

## Slimming down



[Fatahalian '08]

## More Space: Double the Number of Cores



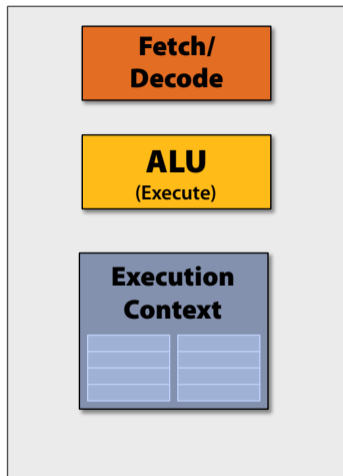
[Fatahalian '08]

## Even more



[Fatahalian '08]

# SIMD



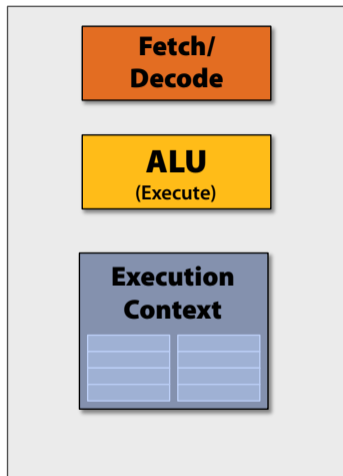
## Idea #2: SIMD

Amortize cost/complexity of managing an instruction stream across many ALUs

[Fatahalian '08]



# SIMD

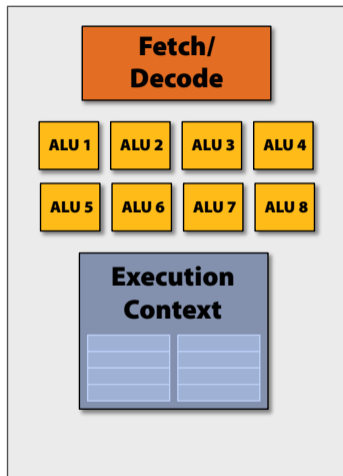


## Idea #2: SIMD

Amortize cost/complexity of managing an instruction stream across many ALUs

[Fatahalian '08]

# SIMD

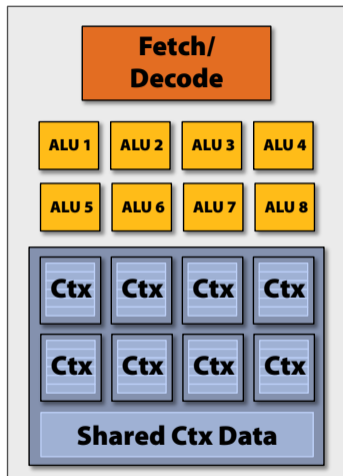


## Idea #2: SIMD

Amortize cost/complexity of managing an instruction stream across many ALUs

[Fatahalian '08]

# SIMD



## Idea #2: SIMD

Amortize cost/complexity of managing an instruction stream across many ALUs

[Fatahalian '08]

# Latency Hiding

- ▶ Latency (mem, pipe) hurts non-OOO cores
- ▶ Do *something* while waiting

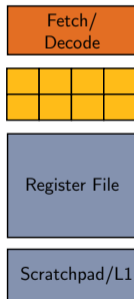
What is the unit in which work gets scheduled on a GPU?

A SIMD vector  
(‘warp’ (Nvidia), ‘Wavefront’ (AMD))

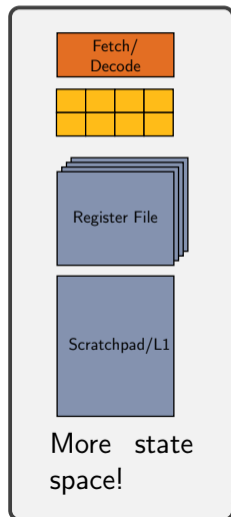
How can we keep busy?

- ▶ More vectors (bigger group)
- ▶ ILP

Before:



After:

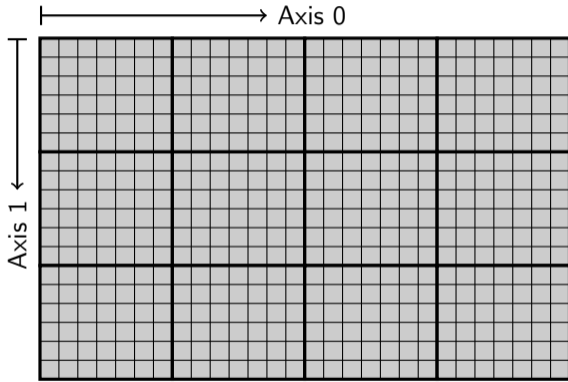


## GPUs: Core Architecture Ideas

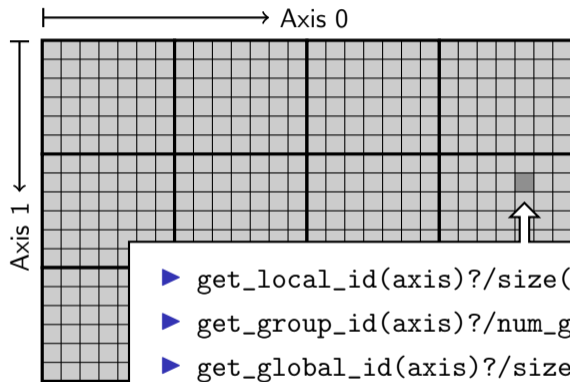
Three core ideas:

- ▶ Remove things that help with latency in single-thread
- ▶ Massive core and SIMD parallelism
- ▶ Cover latency with concurrency
  - ▶ SMT
  - ▶ ILP

# 'SIMT'



## Wrangling the Grid



- ▶ `get_local_id(axis)?/size(axis)?`
- ▶ `get_group_id(axis)?/num_groups(axis)?`
- ▶ `get_global_id(axis)?/size(axis)?`

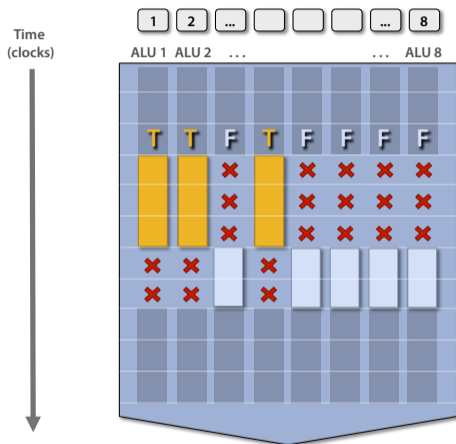
`axis=0,1,2,...`

## Demo CL code

[Demo: machabstr/Hello GPU](#)



# 'SIMT' and Branches



```
<unconditional  
shader code>  
  
if (x > 0) {  
    y = pow(x, exp);  
    y *= Ks;  
    refl = y + Ka;  
} else {  
    x = 0;  
    refl = Ka;  
}  
  
<resume unconditional  
shader code>
```

[Fatahalian '08]

## GPU Abstraction: Core Model Ideas

How do these aspects show up in the model?

- ▶ View concrete counts as an implementation detail
  - ▶ SIMD lane
  - ▶ Core
  - ▶ Scheduling slot
- ▶ Program as if there are infinitely many of them
- ▶ Hardware division is expensive  
Make  $nD$  grids part of the model to avoid it
- ▶ Design the model to expose *extremely* fine-grain concurrency (e.g. between loop iterations!)
- ▶ Draw from the same pool of concurrency to hide latency

## GPU Program 'Scopes'

Hardware	CL adjective	OpenCL	CUDA
SIMD lane	private	Work Item	Thread
SIMD Vector	—	Subgroup	Warp
Core	local	Workgroup	Thread Block
Processor	global	NDRange	Grid

## GPU: Communication

What forms of communication exist at each scope?

- ▶ Subgroup: Shuffles (!)
- ▶ Workgroup: Scratchpad + barrier, local atomics + mem fence
- ▶ Grid: Global atomics

Can we just do locking like we might do on a CPU?

- ▶ Independent forward progress of all threads is not guaranteed:  
no.  
(true until recently)
- ▶ But: Device partitioning can help!

# GPU Programming Model: Commentary

## Advantage:

- ▶ Clear path to scaling in terms of core count
- ▶ Clear path to scaling in terms of SIMD lane count

## Disadvantages:

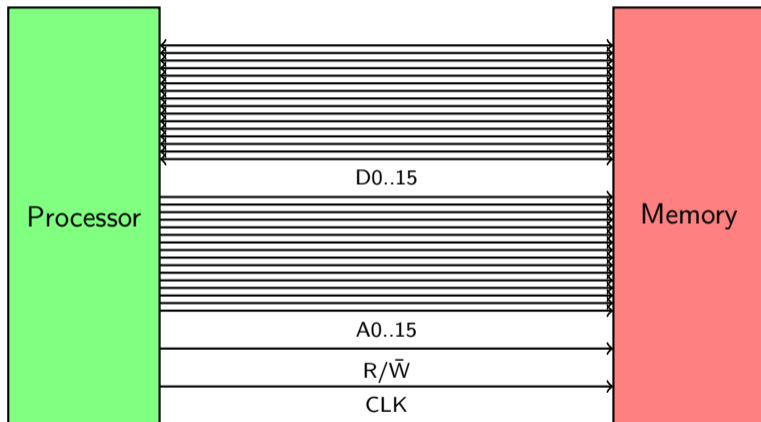
- ▶ “Vector” / “Warp” / “Wavefront”
  - ▶ Important hardware granularity
  - ▶ Poorly/very implicitly represented
- ▶ What is the impact of reconvergence?

## Performance: Limits to Concurrency

What limits the amount of concurrency exposed to GPU hardware?

- ▶ Amount of register space  
**Important:** Size of (per-lane) register file is *variable*
- ▶ Amount of scratchpad space  
Size of (per-group) scratchpad space is *variable*
- ▶ Block size
- ▶ Available ILP
- ▶ Number of scheduler (warp/group) slots (not really)
- ▶ Synchronization

## Memory Systems: Recap



## Parallel Memories

**Problem:** Memory chips have only one data bus.

So how can multiple threads read multiple data items from memory simultaneously?

Broadly:

- ▶ Split a really wide data bus, but have only one address bus
- ▶ Have many 'small memories' ('banks') with separate data and address busses, select by address LSB.

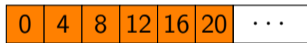
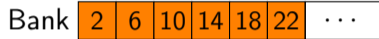
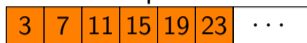
Where does banking show up?

- ▶ Scratchpad
- ▶ GPU register file
- ▶ Global memory



# Memory Banking

Fill in the access pattern:



→ Address



Thread



# Memory Banking

Fill in the access pattern:



```
local_variable[lid(0)]
```

# Memory Banking

Fill in the access pattern:



```
local_variable[BANK_COUNT*lid(0)]
```

# Memory Banking

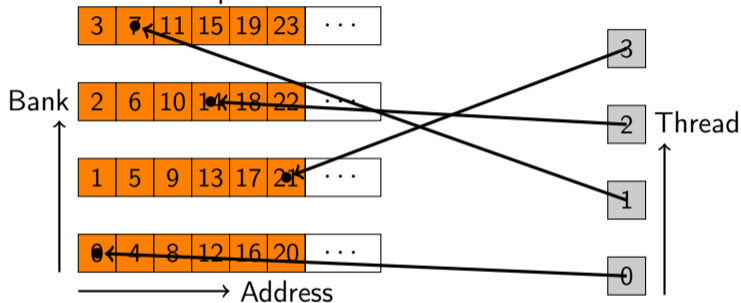
Fill in the access pattern:



```
local_variable[(BANK_COUNT+1)*lid(0)]
```

# Memory Banking

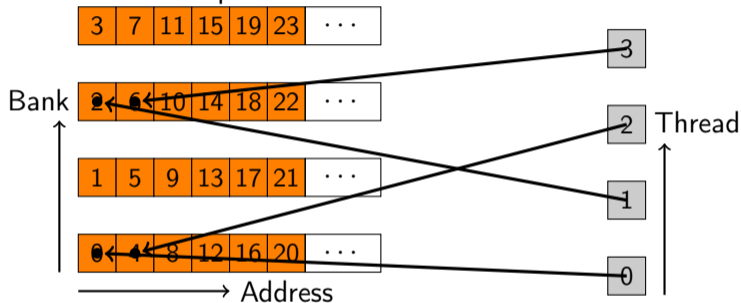
Fill in the access pattern:



```
local_variable[ODD_NUMBER*lid(0)]
```

# Memory Banking

Fill in the access pattern:



```
local_variable[2*lid(0)]
```

# Memory Banking

Fill in the access pattern:



```
local_variable[f(gid(0))]
```

## Memory Banking: Observations

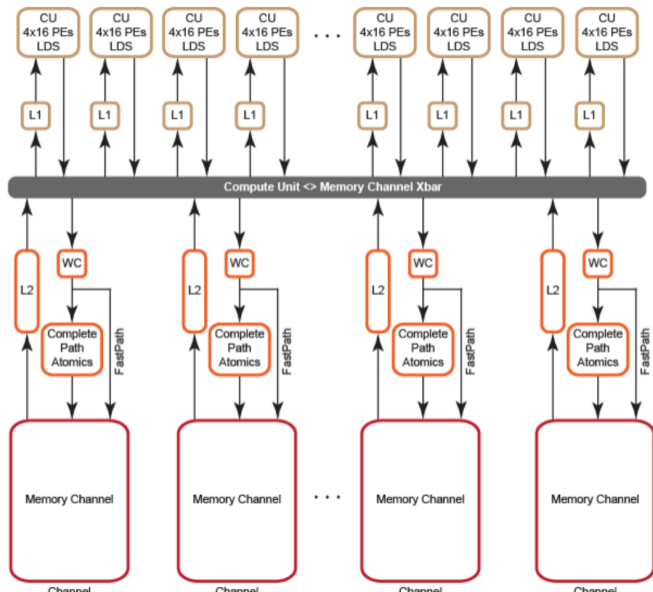
- ▶ Factors of two in the stride: generally bad
- ▶ In a conflict-heavy access pattern, padding can help
  - ▶ Usually not a problem since scratchpad is transient by definition
- ▶ Word size (bank offset) may be adjustable (Nvidia)

Given that unit strides are beneficial on global memory access, how do you realize a transpose?

```
Workgroup size (e.g.): 16x16
__local float tmp[16 * 17];
tmp[lid(0)*17 + lid(1)] = a[lid(1) * 16 + lid(0)];
barrier(CLK_LOCAL_MEM_FENCE);
```

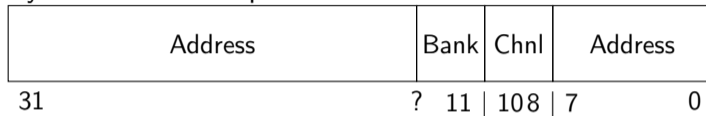


# GPU Global Memory System



# GPU Global Memory Channel Map: Example

Byte address decomposition:



Implications:

- ▶ Transfers between compute unit and channel have granularity
  - ▶ Reasonable guess: warp/wavefront size  $\times$  32bits
  - ▶ Should strive for good utilization ('*Coalescing*')
- ▶ Channel count often *not* a power of two  $\rightarrow$  complex mapping
  - ▶ *Channel conflicts* possible
- ▶ Also *banked*
  - ▶ *Bank conflicts* also possible

## GPU Global Memory: Performance Observations

Key quantities to observe for GPU global memory access:

- ▶ Stride
- ▶ Utilization

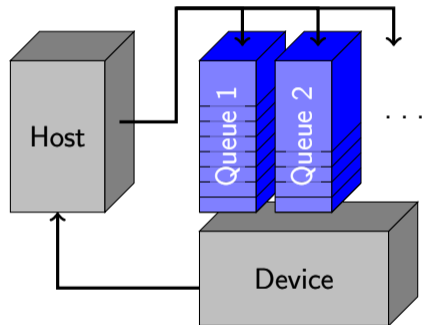
Are there any guaranteed-good memory access patterns?

Unit stride, just like on the CPU

- ▶ Need to consider access pattern *across entire device*
- ▶ *GPU caches*: Use for *spatial*, not for temporal locality
- ▶ Switch available: L1/Scratchpad partitioning
  - ▶ Settable on a per-kernel basis
- ▶ Since GPUs have meaningful caches at this point:  
Be aware of cache annotations (see later)

# Host-Device Concurrency

- ▶ Host and Device run asynchronously
- ▶ Host submits to queue:
  - ▶ Computations
  - ▶ Memory Transfers
  - ▶ Sync primitives
  - ▶ ...
- ▶ Host can wait for:
  - ▶ *drained* queue
  - ▶ Individual “events”
- ▶ Profiling



## Timing GPU Work

How do you find the execution time of a GPU kernel?

- ▶ Do a few 'warm-up' calls to the kernel
- ▶ Drain the queue
- ▶ Start the timer
- ▶ Run the kernel enough times to get to a few milliseconds run time
- ▶ Drain the queue
- ▶ Stop the timer, divide by the number of runs

How do you do this asynchronously?

- ▶ Enqueue 'markers' instead of draining the queue.
- ▶ Find timing of 'markers' after work is complete

## Host-Device Data Exchange

Sad fact: Must get data onto device to compute

- ▶ Transfers can be a bottleneck
- ▶ If possible, overlap with computation
- ▶ Pageable memory incurs difficulty in GPU-host transfers, often entails (another!) CPU side copy
- ▶ “Pinned memory”: unpageable, avoids copy
  - ▶ Various *system-defined* ways of allocating pinned memory

“Unified memory” (CUDA)/“Shared Virtual Memory” (OpenCL):

- ▶ GPU directly accesses host memory
- ▶ Main distinction: Coherence
  - ▶ “Coarse grain”: Per-buffer fences
  - ▶ “Fine grain buffer”: Byte-for-byte coherent (device mem)
  - ▶ “Fine grain system”: Byte-for-byte coherent (anywhere)

## Performance: Ballpark Numbers?

Bandwidth host/device:

PCIe v2: 8 GB/s — PCIe v3: 16 GB/s — NVLink: 200 GB/s

Bandwidth on device:

Registers:  $\sim 10$  TB/s — Scratch:  $\sim 10$  TB/s — Global: 500 GB/s

Flop throughput?

10 TFLOPS single precision – 3 TFLOPS double precision

Kernel launch overhead?

10 microseconds

# Outline

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OpenCL/CUDA

Convergence, Differences in Machine Mapping

Lower-Level Abstractions: SPIR-V, PTX

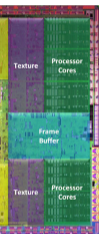
Performance: Expectation, Experiment, Observation

Performance-Oriented Languages and Abstractions

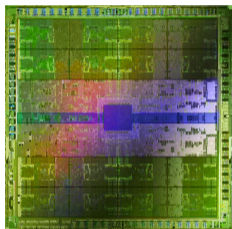
Polyhedral Representation and Transformation



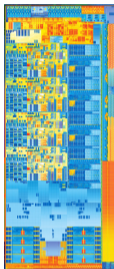
# Die Shot Gallery



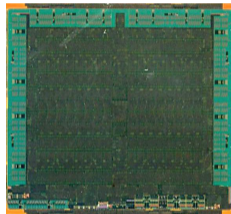
200  
(8)



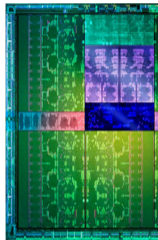
Nv Fermi  
(2010)



Intel IVB  
(2012)



AMD Tahiti  
(2012)



Nv GK110  
(2012?)

# Trends in Processor Architecture

What can we expect from future processor architectures?

- ▶ Commodity chips
- ▶ “Infinitely” many cores
- ▶ “Infinite” vector width
- ▶ Must hide memory latency ( $\rightarrow$  ILP, SMT)
- ▶ Compute bandwidth  $\gg$  Memory bandwidth
- ▶ Bandwidth only achievable by *homogeneity*
- ▶ Can't keep the whole thing powered all the time anyway.  
Consequence?  
Lots of weird stuff springs up. Examples: “Raytracing Cores”,  
“Tensor Cores”

## Common Challenges

What are the common challenges encountered by both CPUs and GPUs?

- ▶ Dealing with Latency (ILP/SMT/Caches)
- ▶ Exposing concurrency
- ▶ Expose a coherent model for talking to SIMD
- ▶ Making memory system complexity manageable

**Goal:** Try to see CPUs and GPUs as points in a design space 'continuum' rather than entirely different things.

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# PTX: Demo

[Demo: machabstr/PTX and SASS](#)

[Nvidia PTX manual](#)

## PTX: Cache Annotations

### Loads:

- .ca Cache at all levels—likely to be accessed again
- .cg Cache at global level (cache in L2 and below and not L1)
- .cs Cache streaming—likely to be accessed once
- .lu Last use
- .cv Consider cached system memory lines stale—fetch again

### Stores:

- .wb Cache write-back all coherent levels
- .cg Cache at global level (cache in L2 and below and not L1)
- .cs Cache streaming—likely to be accessed once
- .wt Cache write-through (to system memory)

Lost/hidden at the C level!

## SPIR-V

*Currently:* C (OpenCL C, GLSL, HLSL) used as intermediate representations to feed GPUs.

Downsides:

- ▶ Compiler heuristics may be focused on human-written code
- ▶ Parsing overhead (preprocessor!)
- ▶ C semantics may not match (too high-level)

SPIR-V:

- ▶ Goal: Common intermediate representation (“IR”) for all GPU-facing code (Vulkan, OpenCL)
- ▶ “Extended Instruction Sets”:
  - ▶ General compute (OpenCL/CUDA) needs: pointers, special functions
- ▶ Different from “SPIR” (tweaked LLVM IR)

## SPIR-V Example

```
%2 = OpTypeVoid
%3 = OpTypeFunction %2 ; void ()
%6 = OpTypeFloat 32 ; 32-bit float
%7 = OpTypeVector %6 4 ; vec4
%8 = OpTypePointer Function %7 ; function-local vec4*
%10 = OpConstant %6 1
%11 = OpConstant %6 2
%12 = OpConstantComposite %7 %10 %10 %11 %10 ; vec4(1.0, 1.0, 2.0, 1.0)
%13 = OpTypeInt 32 0 ; 32-bit int, sign-less
%14 = OpConstant %13 5
%15 = OpTypeArray %7 %14

[...]

%34 = OpLoad %7 %33
%38 = OpAccessChain %37 %20 %35 %21 %36 ; s.v[2]
%39 = OpLoad %7 %38
%40 = OpFAdd %7 %34 %39
      OpStore %31 %40
      OpBranch %29
%41 = OpLabel ; else
%43 = OpLoad %7 %42
%44 = OpExtInst %7 %1 Sqrt %43 ; extended instruction sqrt
%45 = OpLoad %7 %9
%46 = OpFMul %7 %44 %45
      OpStore %31 %46
```



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Performance: Expectation, Experiment, Observation

- Forming Expectations of Performance

- Timing Experiments and Potential Issues

- Profiling and Observable Quantities

- Practical Tools: perf, toplev, likwid

Performance-Oriented Languages and Abstractions

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## Qualifying Performance

- ▶ What is *good* performance?
- ▶ Is speed-up (e.g. GPU vs CPU? C vs Matlab?) a meaningful way to assess performance?
- ▶ How else could one *form an understanding* of performance?

Modeling: how understanding works in science

[Hager et al. '17](#)

[Hockney et al. '89](#)

## A Story of Bottlenecks

Imagine:

- ▶ A bank with a few service desks
- ▶ A revolving door at the entrance

What situations can arise at *steady-state*?

- ▶ Line inside the bank (good)
- ▶ Line at the door (bad)

What numbers do we need to characterize performance of this system?

- ▶  $P_{\text{peak}}$ : [task/sec] Peak throughput of the service desks
- ▶  $I$ : [tasks/customer] Intensity
- ▶  $b$ : [customers/sec] Throughput of the revolving door

## A Story of Bottlenecks (cont'd)

- ▶  $P_{\text{peak}}$ : [task/sec] Peak throughput of the service desks
- ▶  $I$ : [tasks/customer] Intensity
- ▶  $b$ : [customers/sec] Throughput of the revolving door

What is the aggregate throughput?

Bottleneck is either

- ▶ the service desks (good) or
- ▶ the revolving door (bad).

$$P \leq \min(P_{\text{peak}}, I \cdot b)$$

[Hager et al. '17](#)

## Application in Computation

Translate the bank analogy to computers:

- ▶ Revolving door: typically: Memory interface
- ▶ Revolving door throughput: Memory bandwidth [bytes/s]
- ▶ Service desks: Functional units (e.g. floating point)
- ▶  $P_{\text{peak}}$ : Peak FU throughput (e.g.: [flops/s])
- ▶ Intensity: e.g. [flops/byte]

Which parts of this are task-dependent?

- ▶ All of them! This is not a model, it's a *guideline* for making models.
- ▶ Specifically  $P_{\text{peak}}$  varies substantially by task

## A Graphical Representation: 'Roofline'

Plot (often log-log, but not necessarily):

- ▶ X-Axis: Intensity
- ▶ Y-Axis: Performance

What does our inequality correspond to graphically?

$$P \leq \min(P_{\text{peak}}, I \cdot b)$$



What does the shaded area mean?

Achievable performance

## Example: Vector Addition

```
double r, s, a[N];  
for (i=0; i<N; ++i)  
    a[i] = r + s * a[i];}
```

Find the parameters and make a prediction.

Machine model:

- ▶ Memory Bandwidth: e.g.  $b = 10$  GB/s
- ▶  $P_{\text{peak}}$ : e.g. 4 GF/s

Application model:

- ▶  $l = 2$  flops / 16 bytes = 0.125 flops/byte

↑  
performance



## Refining the Model

- ▶  $P_{\max}$ : Applicable peak performance of a loop, assuming that data comes from the fastest data path (this is not necessarily  $P_{\text{peak}}$ )
- ▶ Computational intensity (“work” per byte transferred) over the slowest data path utilized
- ▶  $b$ : Applicable peak bandwidth of the slowest data path utilized

[Hager et al. '17](#)

## Calibrating the Model: Bandwidth

Typically done with the STREAM benchmark.

Four parts: Copy, Scale, Add, Triad  $a[i] = b[i] + s \cdot c[i]$

Do the four measurements matter?

- ▶ No—they're a crude attempt at characterizing intensity.
- ▶ On a modern machine, all four measurements should be identical.

Any pitfalls?

Streaming stores, remember?

[McCalpin: STREAM](#)

## Calibrating the Model: Peak Throughput

Name aspects that should/could be factored in when determining peak performance:

- ▶ Types of operation (FMA? Or only adds?)
- ▶ SIMD
- ▶ Pipeline utilization / operation latency
- ▶ Throughput of faster datapaths

## Practical Tool: IACA

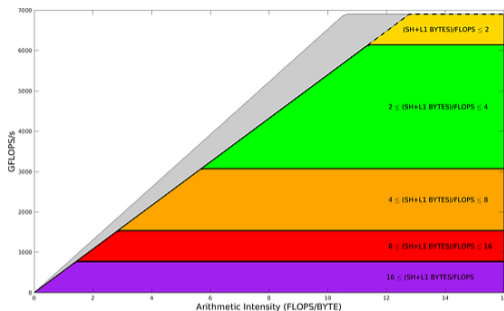
**Question:** Where to obtain an estimate of  $P_{\max}$ ?

**Demo:** [perf/Forming Architectural Performance Expectations](#)

Questions:

- ▶ What does IACA do about memory access? / the memory hierarchy?

## An Example: Exploring Titan V Limits



- ▶ Memory bandwidth: 652 GB/s theoretical, 540 GB/s achievable
- ▶ Scratchpad / L1 throughput:  
 $80 \text{ (cores)} \times 32 \text{ (simd width)} \times 4 \text{ (word bytes)} \times 1.2 \text{ (base clock)} \approx 12.288 \text{ TB/s}$
- ▶ Theoretical peak flops of 6.9 TFLOPS/s [Wikipedia]

[Warburton '18](#)

## Rooflines: Assumptions

What assumptions are built into the roofline model?

- ▶ Perfect overlap  
(What would imperfect overlap be in the bank analogy?)
- ▶ Only considers the dominant bottleneck
- ▶ Throughput-only (!)  
No latency effects, no start-up effects, only steady-state

Important to remember:

- ▶ It is what you make of it—the better your calibration, the more info you get
- ▶ But: Calibrating on experimental data loses predictive power (e.g. SPMV)

## Modeling Parallel Speedup: A 'Universal' Scaling Law

Develop a model of *throughput*  $X(N)$  for a given load  $N$ , assuming execution resources scale with  $N$ .

$$X(N) = \frac{\gamma N}{1 + \alpha \cdot (N - 1) + \beta N \cdot (N - 1)}$$

What do the individual terms model?

- ▶  $\gamma$ : Throughput increase per load increase
- ▶  $\alpha$ : **Contention** due to waiting/queueing for shared resources/sequential sections ("Amdahl's law")
- ▶  $\beta$ : **Incoherence penalty** due to waiting for data to become coherent through *point-to-point exchange*

[[Gunther '93](#)]

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Machine Abstractions

**Performance: Expectation, Experiment, Observation**

Forming Expectations of Performance

**Timing Experiments and Potential Issues**

Profiling and Observable Quantities

Practical Tools: perf, toplev, likwid

Performance-Oriented Languages and Abstractions

Polyhedral Representation and Transformation



## Combining Multiple Measurements

How can one combine multiple performance measurements? (e.g. “average speedup”?)

*Example:* Which computer should you buy?

Execution time [s]	Computer A	Computer B	Computer C
Program 1	1	10	20
Program 2	1000	100	20

	Computer A	Computer B	Computer C
Arithmetic mean	500.5	55	20
Geometric mean	31.622	31.622	20

## Combining Multiple Measurements: Observations

	Computer A	Computer B	Computer C
Arithmetic mean	500.5	55	20
Geometric mean	31.622	31.622	20

- ▶ Depending on normalization, the arithmetic mean will produce an arbitrary ranking
- ▶ Geometric mean  $\sqrt[n]{a_1 \cdots a_n}$ : consistent ranking
- ▶ Is geomean **good**? (What is the meaning of multiplying times?)

Take-home message:

- ▶ Be mindful of units when combining measurements (e.g. sums of times make sense, but products of times may not)
- ▶ Avoid combined measurements if you can
- ▶ Ideally: purposefully choose a weighting

## Timing Experiments: Pitfalls

What are potential issues in timing experiments? (What can you do about them?)

- ▶ Warm-up effects (do a few runs before timing to only time steady state)
- ▶ Timing noise
  - ▶ Know your timer granularity
  - ▶ Know your clock kinds (wall, monotone, process)
  - ▶ Know your clock sources (RTC, PIT, APIC, TSC (nominal), TSC (actual))
  - ▶ Know your overheads (function call/kernel launch)
  - ▶ Make sure your timing granularity is appropriate (On-node: one second is a reasonable number)

## Timing Experiments: Pitfalls (part 2)

What are potential issues in timing experiments? (What can you do about them?)

- ▶ NUMA placement (use `numactl`, `libnuma`, or respect first-touch)
- ▶ Thread migration between cores (and resulting cache effects)
  - ▶ Pin your threads to cores
- ▶ Uninitialized pages are never fetched
  - ▶ Is `calloc` good enough?
- ▶ Frequency Scaling
  - ▶ Turn it off or run long enough for thermal steady-state
  - ▶ Understand how RAPL (“Running average power limit”) and “power leases” work
  - ▶ Realize there’s a dependency on what instructions you execute
- ▶ Noise from other users

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# Profiling: Basic Approaches

Measurement of “quantities” relating to performance

- ▶ Exact: Through binary instrumentation (valgrind/Intel Pin/...)
- ▶ Sampling: At *some* interval, examine the program state

We will focus on profiling by *sampling*.

Big questions:

- ▶ What to measure?
- ▶ At what intervals?

## Defining Intervals: Performance Counters

A *performance counter* is a counter that increments every time a given **event** occurs.

What events?

- ▶ [Demo: perf/Using Performance Counters](#)
- ▶ see also [Intel SDM, Volume 3](#)

Interaction with performance counters:

- ▶ Read repeatedly from user code
- ▶ Interrupt program execution when a threshold is reached
- ▶ Limited resource!
  - ▶ Only a few available: 4-8 per core
  - ▶ Each can be configured to count one type of event
  - ▶ **Idea:** Alternate counter programming at some rate (requires steady-state execution!)

## Profiling: What to Measure

- ▶ Raw counts are hard to interpret
- ▶ Often much more helpful to look at *ratios* of counts per core/subroutine/loop/...

What ratios should one look at?

[Demo: perf/Using Performance Counters](#)



## Profiling: Useful Ratios

Basic examples:

- ▶  $(\text{Events in Routine 1}) / (\text{Events in Routine 2})$
- ▶  $(\text{Events in Line 1}) / (\text{Events in Line 2})$
- ▶  $(\text{Count of Event 1 in X}) / (\text{Count of Event 2 in X})$

Architectural examples:

- ▶ instructions / cycles
- ▶ L1-dcache-load-misses / instructions
- ▶ LLC-load-misses / instructions
- ▶ stalled-cycles-frontend / cycles
- ▶ stalled-cycles-backend / cycles

Issue with 'instructions' as a metric?

May or may not correlate with 'amount of useful work'

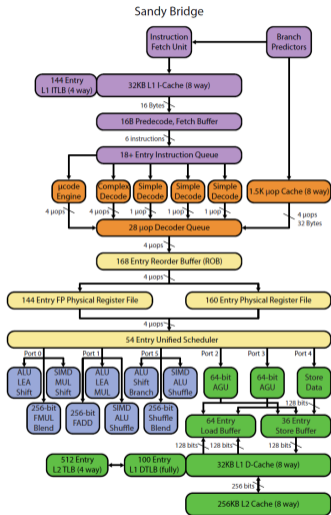
## “Top-Down” Performance Analysis

**Idea:** Account for useful work per available issue slot  
What is an issue slot?

A clock cycle that passed at an interface to an execution pipeline

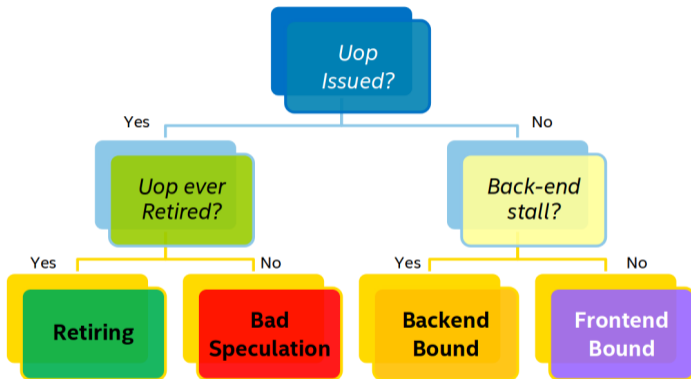
[[Yasin '14](#)]

# Issue Slots: Recap



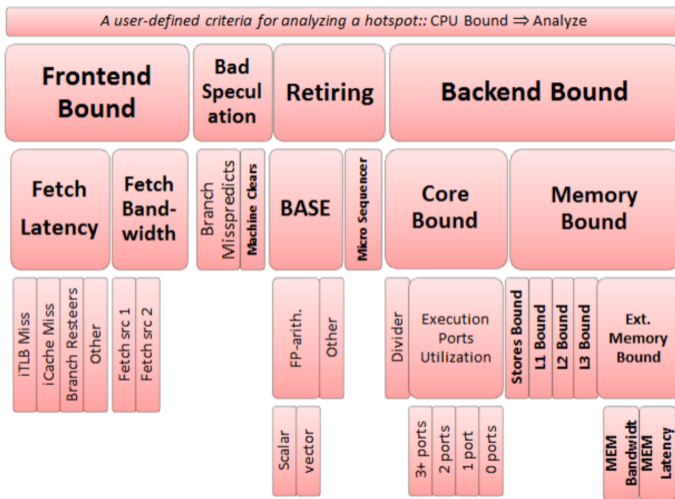
[David Kanter / Realworldtech.com]

## What can happen to an issue slot: at a high level?



[Yasin '14]

# What can happen to an issue slot: in detail?



[Yasin '14]

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## Demo: Performance Counters

Show the rest of:

[Demo: perf/Using Performance Counters](#)

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Performance-Oriented Languages and Abstractions

- Expression Trees

- Parallel Patterns and Array Languages

Polyhedral Representation and Transformation



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Expression Trees

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# Expression Trees and Term Rewriting

Demos:

- ▶ [Demo: lang/01 Expression Trees](#)
- ▶ [Demo: lang/02 Traversing Trees](#)
- ▶ [Demo: lang/03 Defining Custom Node Types](#)
- ▶ [Demo: lang/04 Common Operations](#)

How do expression trees come to be? (not our problem here)

- ▶ Partitioning, classification of input stream into tokens (lexing)
- ▶ Extraction of higher-level constructs from token stream (parsing)
  - ▶ Recursive descent
  - ▶ Table/automata-based (e.g. Yacc, ANTLR, PLY, boost::spirit)

## Embedded languages

Main challenge: Obtaining a syntax tree. Approaches?

- ▶ Symbolic execution (seen above, runtime)
- ▶ Type system abuse
  - ▶ [Demo: lang/Expression Templates](#)
- ▶ boost::metaparse (string → tree at compile time)
- ▶ “Reflection”
  - ▶ [Demo: lang/05 Reflection in Python](#)

## Macros: Goals and Approaches

What is a macro?

- ▶ In C: Simple textual replacement with parameters
- ▶ Generally: any type of compile-time computation (that operates on the code)
  - ▶ Question: How would you express loops in the C preprocessor?

What data do macro systems operate on?

- ▶ Character-streams
- ▶ Syntax/expression trees

## Macros: Textual and Syntactic, Hygiene

Macros: What can go wrong if you're not careful?

```
#define INCI(i) do { int a=0; ++i; } while(0)
int main(void)
{
    int a = 4, b = 8;
    INCI(a);
    INCI(b);
    printf("a is now %d, b is now %d\n", a, b);
    return 0;
}
```

How can the problem above be avoided?

Ensure macro-internal identifiers (e.g. a above)

# Towards Execution

[Demo: lang/06 Towards Execution](#)

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## Reduction

$$y = f(\dots f(f(x_1, x_2), x_3), \dots, x_N)$$

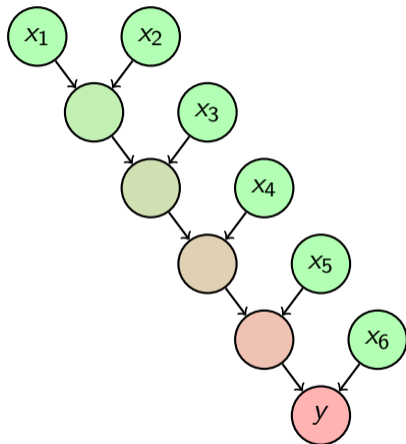
where  $N$  is the input size.

Also known as

- ▶ Lisp/Python function `reduce` (Scheme: `fold`)
- ▶ C++ STL `std::accumulate`



## Reduction: Graph



## Approach to Reduction



Can we do better?

“Tree” very imbalanced. What property of  $f$  would allow ‘rebalancing’?

$$f(f(x, y), z) = f(x, f(y, z))$$

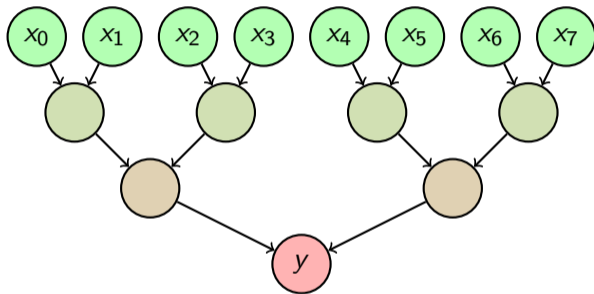
Looks less improbable if we let

$$x \circ y = f(x, y):$$

$$x \circ (y \circ z) = (x \circ y) \circ z$$

Has a very familiar name: *Associativity*

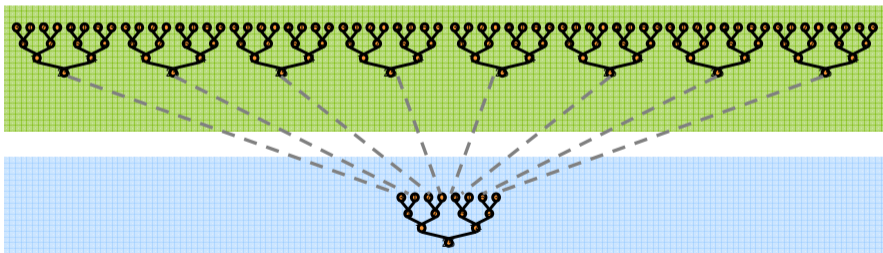
## Reduction: A Better Graph



Processor allocation?

## Mapping Reduction to SIMD/GPU

- ▶ Obvious: Want to use tree-based approach.
- ▶ Problem: Two scales, Work group and Grid
  - ▶ to occupy both to make good use of the machine.
- ▶ In particular, need synchronization after each tree stage.
- ▶ Solution: Use a two-scale algorithm.



*In particular:* Use multiple grid invocations to achieve inter-workgroup synchronization.

But no. Not even close.

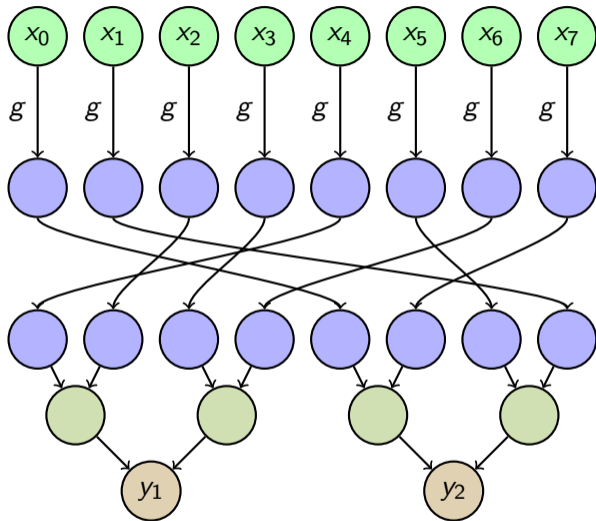
Sounds like this:

$$y = f(\dots f(f(g(x_1), g(x_2)), \\ g(x_3)), \dots, g(x_N))$$

where  $N$  is the input size.

- ▶ Lisp naming, again
- ▶ Mild generalization of reduction

## Map-Reduce: Graph



## Scan

$$y_1 = x_1$$

$$y_2 = f(y_1, x_2)$$

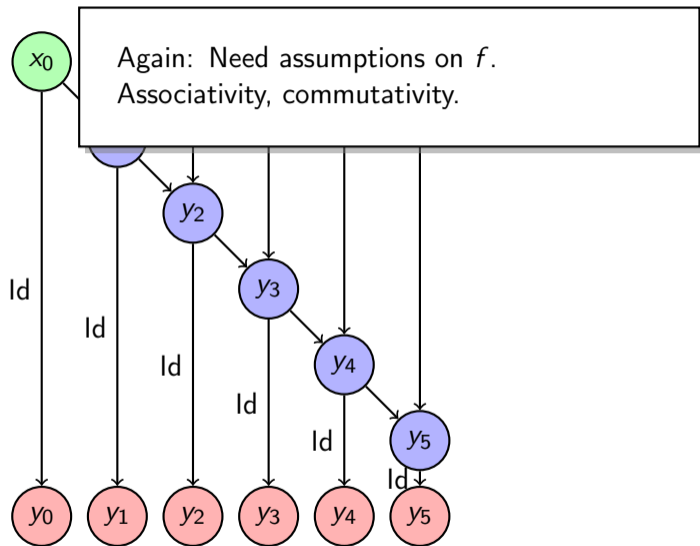
$$\vdots = \vdots$$

$$y_N = f(y_{N-1}, x_N)$$

where  $N$  is the input size. (Think:  $N$  large,  $f(x, y) = x + y$ )

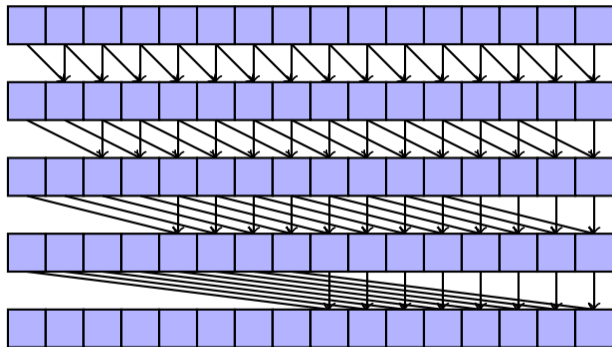
- ▶ Prefix Sum/Cumulative Sum
- ▶ Abstract view of: loop-carried dependence
- ▶ Also possible: Segmented Scan

## Scan: Graph





## Scan: Implementation



Work-efficient?

## Scan: Implementation II

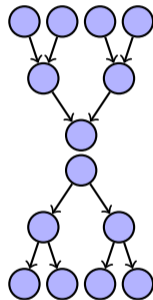
Two sweeps: Upward, downward, both tree-shape

On upward sweep:

- ▶ Get values  $L$  and  $R$  from left and right child
- ▶ Save  $L$  in local variable  $Mine$
- ▶ Compute  $Tmp = L + R$  and pass to parent

On downward sweep:

- ▶ Get value  $Tmp$  from parent
- ▶ Send  $Tmp$  to left child
- ▶ Sent  $Tmp+Mine$  to right child



## Scan: Examples

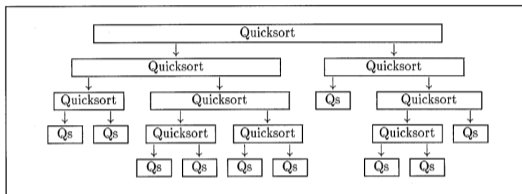
Name examples of Prefix Sums/Scans:

- ▶ Anything with a loop-carried dependence
- ▶ One row of Gauss-Seidel
- ▶ One row of triangular solve
- ▶ Segment numbering if boundaries are known
- ▶ Low-level building block for many higher-level algorithms algorithms, e.g. predicate filter, sort
- ▶ FIR/IIR Filtering
- ▶ [Blelloch '93](#)

## Data-parallel language: Goals

**Goal:** Design a full data-parallel programming language

**Example:** What should the (asymptotic) execution time for Quicksort be?



$O(\log(N))$

**Question:** What parallel primitive could be used to realize this?

- ▶ Segmented Scan, i.e. a scan with data-dep. boundaries
- ▶ Any basic scan operation can be segmented while retaining

## NESL Example: String Search

```
teststr = "string strap asop string" : [char]
>>> candidates = [0:#teststr-5];
candidates = [0, 1, 2, 3, .... : [int]
>>> {a == 's: a in teststr -> candidates};
it = [T, F, F, F, F, F, F, T, F, F....] : [bool]
>>> candidates = {c in candidates;
...     a in teststr -> candidates | a == 's};
candidates = [0, 7, 13, 20, 24] : [int]
>>> candidates = {c in candidates;
...     a in teststr -> {candidates+1:candidates}
...     | a == 't};
```

- ▶ Work and depth of this example?
- ▶ NESL specifies work and depth for its constructs
- ▶ How can scans be used to realize this?

[Blelloch '95](#)

# Array Languages

## Idea:

- ▶ Operate on entire array at once
- ▶ Inherently data-parallel

## Examples:

- ▶ APL, numpy
- ▶ Tensorflow (talk on Friday), Pytorch

## Important axes of distinction:

- ▶ Lazy or eager
- ▶ Imperative (with in-place modification) or pure/functional

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Polyhedral Model: What?

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Polyhedral Model: What?



## Basic Object: Presburger Set

Think of the **problem statement** here as representing an arbitrary-size (e.g.: dependency) graph.

*Presburger sets* correspond to a subset of predicate logic acting on tuples of integers.

**Important:** Think of this as a mathematical tool that can be used in many settings.

# Basic Object: Presburger Set

Terms:

- ▶ Variables, Integer Constants
- ▶  $+$ ,  $-$
- ▶  $\lfloor \cdot / d \rfloor$

Predicates:

- ▶  $(\text{Term}) \leq (\text{Term})$
- ▶  $(\text{Pred}) \wedge (\text{Pred}), (\text{Pred}) \vee (\text{Pred}), \neg(\text{Pred})$
- ▶  $\exists v : (\text{Pred})(v)$

Sets: integer tuples for which a predicate is true

[Verdoolaege '13](#)

## Presburger Sets: Reasoning

What's "missing"? Why?

- ▶ Multiplication, Division
- ▶ Most questions become undecidable in its presence

Why is this called 'quasi-affine'?

- ▶ Affine:  $\vec{a} \cdot \vec{x} + b$ .
- ▶ Quasi: inclusion of modulo/existential quantifier

## Presburger Sets: Reasoning

What do the resulting sets have to do with polyhedra? When are they convex?

- ▶ Each constraint specifies a half-space
- ▶ Intersection of half-spaces is a convex polyhedron
- ▶ Unions can be used to make non-convex polyhedra

Why polyhedra? Why not just rectangles?

- ▶ Rectangular domains are not closed under many transformations
- ▶ E.g. strip-mining

## Demo: Constructing and Operating on Presburger Sets

[Demo: lang/Operating on Presburger Sets](#)

## Making Use of Presburger Sets

- ▶ Loop Domains
- ▶ Array Access Relations (e.g. write, read: per statement)
- ▶ Schedules, with “lexicographic time”
- ▶ Dependency graphs
- ▶ (E.g. cache) interference graphs

Q: Specify domain and range for the relations above.

## Example: Dependency Graph

Given:

- ▶ Write access relation  $W$ : Loop domain  $\rightarrow$  array indices
- ▶ Read access relation  $R$
- ▶ Schedule  $S$  for statement  $S_i$ : Loop domain  $D \rightarrow$  lex. time of statement instance
- ▶ Relation  $\prec$ : Lexicographic 'before'

Find the dependency graph:

$$D = ((W^{-1} \circ R) \cup (R^{-1} \circ W) \cup (W^{-1} \circ W)) \cap (S \prec S)$$

[Verdoolaege '13](#)

## Example: Last Instance

Given:

- ▶ Write access relation  $W$ : Loop domain  $\rightarrow$  array indices
- ▶ Read access relation  $R$
- ▶ Schedule  $S$  for statement  $S_i$ : Loop domain  $D \rightarrow$  lex. time of statement instance
- ▶ Relation  $\prec$ : Lexicographic 'before'

Find the statement instances accessing array element:

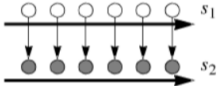

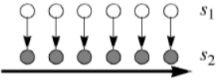
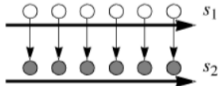
$$(R \cup W)^{-1}$$

Find the *last* statement instance accessing array element:

$$\text{lexmax}((R \cup W)^{-1})$$

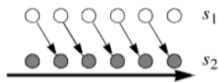


# Primitive Transformations (I)

SOURCE CODE	PARTITION	TRANSFORMED CODE
<pre>for (i=1; i&lt;=N; i++)   Y[i] = Z[i]; /*s1*/ for (j=1; j&lt;=N; j++)   X[j] = Y[j]; /*s2*/</pre> 	<p>Fusion</p> $s_1 : p = i$ $s_2 : p = j$	<pre>for (p=1; p&lt;=N; p++){   Y[p] = Z[p];   X[p] = Y[p]; }</pre> 
<pre>for (p=1; p&lt;=N; p++){   Y[p] = Z[p];   X[p] = Y[p]; }</pre> 	<p>Fission</p> $s_1 : i = p$ $s_2 : j = p$	<pre>for (i=1; i&lt;=N; i++)   Y[i] = Z[i]; /*s1*/ for (j=1; j&lt;=N; j++)   X[j] = Y[j]; /*s2*/</pre> 

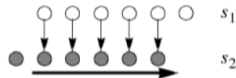
# Primitive Transformations (II)

```
for (i=1; i<=N; i++) {
  Y[i] = Z[i]; /*s1*/
  X[i] = Y[i-1]; /*s2*/
}
```

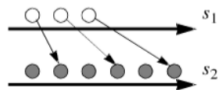


Re-indexing  
 $s_1 : p = i$   
 $s_2 : p = i - 1$

```
if (N>=1) X[1]=Y[0];
for (p=1; p<=N-1; p++){
  Y[p]=Z[p];
  X[p+1]=Y[p];
}
if (N>=1) Y[N]=Z[N];
```

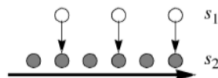


```
for (i=1; i<=N; i++)
  Y[2*i] = Z[2*i]; /*s1*/
for (j=1; j<=2N; j++)
  X[j]=Y[j]; /*s2*/
```

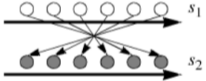
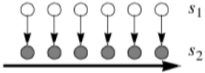


Scaling  
 $s_1 : p = 2 * i$   
 $(s_2 : p = j)$

```
for (p=1; p<=2*N; p++){
  if (p mod 2 == 0)
    Y[p] = Z[p];
  X[p] = Y[p];
}
```



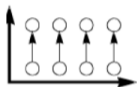
# Primitive Transformations (III)

SOURCE CODE	PARTITION	TRANSFORMED CODE
<pre> for (i=0; i&lt;=N; i++)   Y[N-i] = Z[i]; /*s1*/ for (j=0; j&lt;=N; j++)   X[j] = Y[j]; /*s2*/           </pre> 	<p>Reversal</p> $s_1 : p = N - i$ $(s_2 : p = j)$	<pre> for (p=0; p&lt;=N; p++){   Y[p] = Z[N-p];   X[p] = Y[p]; }           </pre> 

[Aho/Ullman/Sethi '07]

# Primitive Transformations (IV)

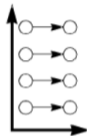
```
for (i=1; i<=N; i++)
  for (j=0; j<=M; j++)
    Z[i,j] =
      Z[i-1,j];
```



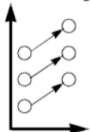
Permutation

$$\begin{bmatrix} p \\ q \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix}$$

```
for (p=0; p<=M; p++)
  for (q=1; q<=N; q++)
    Z[q,p] = Z[q-1,p]
```



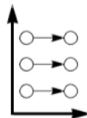
```
for (i=1; i<=N+M-1; i++)
  for (j=max(1,i+N);
       j<=min(i,M); j++)
    Z[i,j] =
      Z[i-1,j-1];
```



Skewing

$$\begin{bmatrix} p \\ q \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} i \\ j \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

```
for (p=1; p<=N; p++)
  for (q=1; q<=M; q++)
    Z[p,q-p] =
      Z[p-1,q-p-1]
```



## Example: Last Instance

Given:

- ▶ Dependency relation  $D$

Check whether a transformed schedule  $S'$  is valid:

$$\vec{i} \rightarrow \vec{j} \in D \Rightarrow S'(\vec{i}) \prec S'(\vec{j})$$

## A peek under the hood: Fourier-Motzkin Elimination

INPUT: A polyhedron  $S$  with variables  $x_1, \dots, x_n$

OUTPUT:  $x_1, \dots, x_{n-1}$

- ▶ Let  $C$  be all the constraints in  $S$  involving  $x_n$ .
- ▶ For every pair of a lower and an upper bound on  $x_m$  in  $C$ :

$$\begin{aligned}L &\leq c_1 x_n, \\ &\leq c_2 x_n \leq U,\end{aligned}$$

create the new constraint  $c_2 L \leq c_1 U$ .

- ▶ Divide by the GCD of  $c_1, c_2$  if applicable.
- ▶ If the new constraint is not satisfiable,  $S$  was empty.
- ▶ Let  $S' = S \setminus C \cup \bigcup_{L,U} \{c_2 L \leq c_1 U\}$ .

[Aho/Ullman/Sethi '07]

Q: How can this help implement an emptiness check?