# Performance Modeling

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Slides as of 14/01/22 11:57:44

# Today, I am out of my depth

- · Giuliano knows the theory of this much better than me
- But, I know how a CPU works :-)
- So, I get to tell you the practical side of things

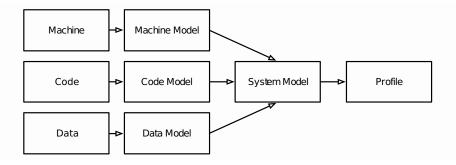
# Where we stand

- We can (empirically) determine performance metrics of hardware & software systems if we have access to
  - hardware to run it on
  - the code
  - the input
- What happens if we lack one of these?
  - We need to model it!
- Why would that happen, you ask?

# Why would we need analytical performance modeling

- When we want to know performance "on the cheap" (i.e. without running)
  - For charging before execution
  - For provisioning systems
  - Other reasons?

# System (Model) Aspects



Alright, let's model something!

Before we start...

# Operating assumptions

- We make simplifying assumptions about the input
  - We assume a known distribution (usually uniform without correlation)
- We do not model system noise
  - Could be caused by scheduling, other processes, external factors,  $\ldots$
- In this lecture, we assume single-threaded, deterministic code
  - Modeling contention in parallel systems is an open research topic

# Performance modeling approaches

### • Two approaches:

### Numerical/Experimental Model

- A series of datapoints describing the observed behavior of the system
- Useful to describe system behaviour for humans
- Predictive power depends on interpretation (example is coming up)

### Analytical Model

- A formal characterization of the relationship between parameters and performance metrics
- Often difficult to interpret for humans (moderately useful to describe system)
- Prediction is performed by evaluating the model

# Numerical models step 1: gathering data

What we want			
	Parameter	Metric	
	0	1	
	1	0	
	2	3	
	3	2	
	4	2	
	5	4	
	0.5	0	
	1.5	1	
	2.5	3.2	
	3.5	1.9	
	4.5	3	
	5.5	6	

### But how do we get pristine results like this?

# Numerical models step 1: gathering data

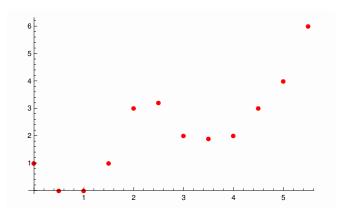
- Through Microbenchmarking
- "*Microbenchmarks* are small, specially designed programs use to test some specific portion of a system"

# Numerical models step 1: gathering data

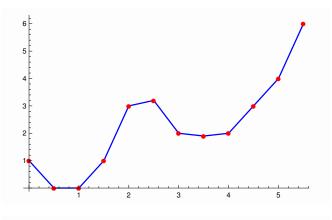
### A Memory Subsystem Microbenchmark

```
extern int* input;
extern size_t N; // some large constant
extern size_t stride; // the parameter of our experiment
int sum = 0;
for(size_t i = 0; i < N; i += stride) {
   sum += input[stride];
}
```

# Numerical models step 2: interpret

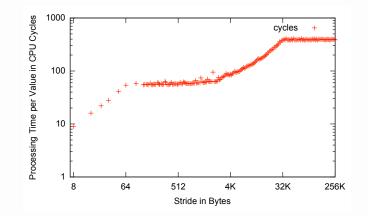


# Numerical models step 2: interpret



• Prediction through, for example, interpolation

# Numerical models step 2: interpret



# Numerical models pro/cons

- Advantages
  - Easy to get (if the system is available)
  - Based on ground truth
  - Relatively easy to interpret
- Disadvantages
  - · Generalize poorly (i.e., cannot easily be applied to new environments)
  - Massive amounts of experimental data needed for high-dimensional system parameter spaces
  - Limited accuracy for/confidence in prediction (data may be missing, inaccurate, ...)
  - · Limited interpretability: contributing factors are (at best) implicit
  - · Limited insight: how does the system actually work?

The alternative:

### The alternative: analytical models

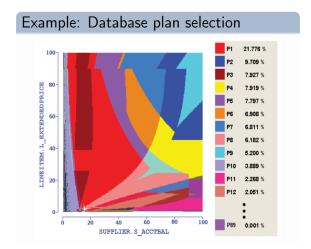
$$DA'\_total(R_1, R_2) = \sum_{j=\text{abs}|h_{R_1} - h_{R_2}|+1}^{\max(h_{R_1}, h_{R_2})-1} \begin{cases} DA(R_1, j) + DA(R_2, j'), \text{ if } h_{R_1} > h_{R_2} \\ DA(R_1, j') + DA(R_2, j), \text{ if } h_{R_1} < h_{R_2} \end{cases} + \sum_{j=1}^{\text{abs}|h_{R_1} - h_{R_2}|} \begin{cases} DA(R_1, j), \text{ if } h_{R_1} > h_{R_2} \\ 2 \cdot DA(R_2, j), \text{ if } h_{R_1} < h_{R_2} \end{cases}$$
(12)

A Model for an R-Tree Theodoridis et al.: Cost Models for Join Queries in Spatial Databases

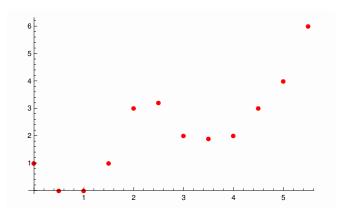
# Analytical models

- Analytical model development is more an art than a craft
- · Requires detailed understanding of the system
  - The parameters
  - The effects
- Requires extensive validation
  - Results always questionable
- Often end up very complicated to deal with edge cases

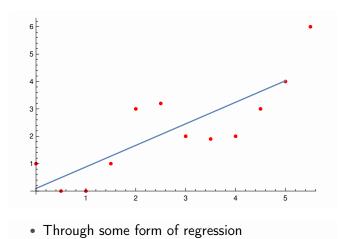
# Analytical models are often complex



# Turning empirical models into analytical ones...



# Turning empirical models into analytical ones...



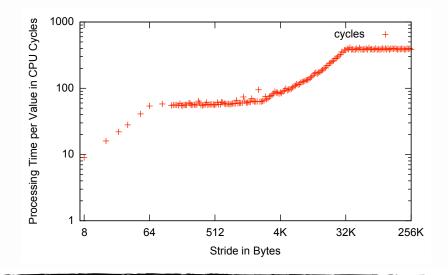
### How is this different from numerical modeling?

### Admittedly, the line is blurry!

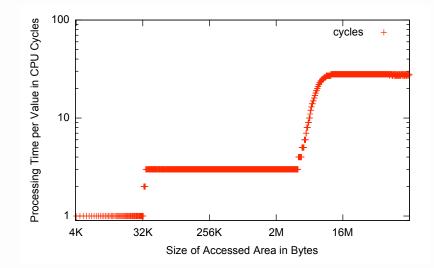
# I have decided that interpolation is numerical while regression is analytical

But some things really cannot be done using numerical modeling?

### How do you model that...



### ... or that?



# ... or that?

#### For completeness, here is the code

# We need to apply $\ensuremath{\mathsf{AI}}$

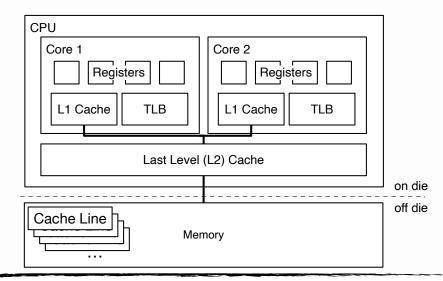
Actual Intelligence

(and some simplifying assumptions)

# Analytical model ingredients

- A *Characteristic Equation* (potentially with parameters) An equation that describes the behavior of the target metric of your experiment or system in dependence of a varied parameter
  - In our examples: stride and data size
- Values for system parameters
  - In our examples: *access latency, access granularity (block size)* and *capacity* of the caches

As seen in [Manegold et al., Generic database cost models for hierarchical memory systems] What do we know about the system we are trying to model



# System parameters

Variable	Description	
$B_0$ :	Size of a General Purpose Register of the CPU	
$l_0$ :	Access Latency of the Level 1 Cache	
$C_0$ :	Capacity of a General Purpose Register of the CPU	
$B_1$ :	Size of a cache line of the Level 1 cache	
$l_1$	Access Latency of the Level 2 Cache	
$C_1$ :	Capacity of the Level 1 Cache	
$B_2$ :	Size of a cache line of the Level 2 cache	
$l_2$	Access Latency of the main memory	
$C_2$ :	Capacity of the Level 2 Cache	
$B_3$ :	Size of a Memory Page	
$l_3$ :	Lookup time in the Page Table	
$C_3$ :	Number of Memory Pages in the TLB tims Page size	

### A characteristic, non-linear equation

### $T_{\ensuremath{\mathit{Mem}}}$ average time for a memory access

$$s = stride$$

$$T_{Mem} = l_3 \cdot min\left(1, \frac{s}{B_3}\right) + l_2 \cdot min\left(1, \frac{s}{B_2}\right)$$

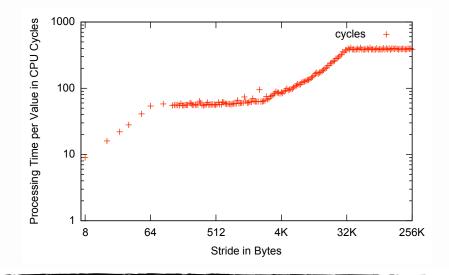
$$+ l_1 \cdot min\left(1, \frac{s}{B_1}\right) + l_0 \cdot min\left(1, \frac{s}{B_0}\right)$$

### A characteristic, non-linear equation

### $T_{\ensuremath{\mathit{Mem}}}$ average time for a memory access

$$T_{Mem} = \begin{cases} 10 & \text{size} < \text{C1} \\ 10 + \text{l1} & \text{size} < \text{C2} \\ 10 + \text{l1} + \text{l3} & \text{size} < \text{C3} \\ 10 + \text{l1} + \text{l2} + \text{l3} & \text{Otherwise} \end{cases}$$

#### Fitting the characteristic equation



#### Demo Time!

## Demo Time!

https://www.wolframcloud.com/obj/hlgr/Published/CPUModel.nb

## System parameters determined through fitting characteristic equation

Variable	Value
$B_0$ :	1 word (64 bit)
$l_0$ :	1 cycle
$C_0$ :	1 word
$B_1$ :	8 words
$l_1$	3 cycles
$C_1$ :	4096 words
$B_2$ :	8 words
$l_2$	55 cycles
$C_2$ :	786432 words
$B_3$ :	512 words
$l_3$ :	1 cycle
$C_3$ :	131072 words

## A note

- Some of these can be read from documentation
- However, self tuning systems
  - require less work/expertise
  - are more resilient
  - scale forward (i.e., work on future architectures)
  - and are sometimes more accurate...

## Modeling Memory Access

#### Let's model this

```
extern int* input1; // uniform random data
extern int* input2; // random data
int sum = 0;
for(size_t i = 0; i < inputSize; i++) {
    sum += input2[input1[i]];
}</pre>
```

#### Parameters

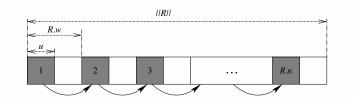
#### Memory Regions

- it's length (R.n), i.e., the number of stored tuples and
- it's width (*R.w*), the size of a tuple in processor words (we will assume a processor with 64bit words).
- The size of the region ( $\|R\|$ ) is defined as the product of length and width.

#### Access Patterns

• u the number of words read in each access

## Modeling sequential access

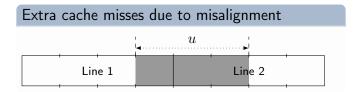


#### Estimating the number of cache misses - not examinable

If 
$$R.w - u < B$$
  
 $M_i^s (s\_trav) = \frac{R.w \cdot R.n}{B_i}$ 
If  $R.w - u > B$   
 $M_i^s (s\_trav) = R.n \cdot \left[\frac{u}{B_i}\right]$ 

[Pirk, Holger, et al. "Cache conscious data layouting for in-memory databases."]

#### Estimating the number of cache misses - not examinable

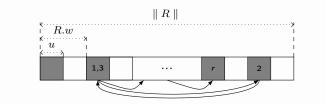


## Estimating the number of cache misses - not examinable

If 
$$R.w - u > B$$

$$M_i^s(s\_trav) = R.n \cdot \left( \left\lceil \frac{u}{B_i} \right\rceil + \frac{(u-1) \mod B_i}{B_i} \right)$$

# Modeling random access (with repetitive access to elements)



• Additional parameter r, number of accesses

## Modeling complex patterns

 $\mathcal{P}_1 \oplus \mathcal{P}_2$  the sequential execution of the access patterns  $\mathcal{P}_1$  and  $\mathcal{P}_2$  $\mathcal{P}_1 \odot \mathcal{P}_2$  the concurrent execution of access patterns.

## Example

#### Code

```
extern int* input1; // uniform random data, 1024 value
extern int* input2; // random data, 64 values
int sum = 0;
for(size_t i = 0; i < inputSize; i++) {
    sum += input2[input1[i]];
}
```

#### Access pattern description

$$s\_trav(R.w = 1, u = 1, R.n = 1024)$$
  
 $\odot rr\_acc(R.w = 1, u = 1, R.n = 64, r = 1024)$ 

## Example

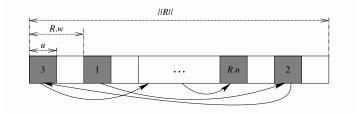
#### Let's model this

```
extern struct{int a; int b; int c;}* input1; // uniform random data, 1024 value
extern int* input2; // random data, 64 values
int sum = 0;
for(size_t i = 0; i < inputSize; i++) {
        sum += input2[input1[i].a];
}</pre>
```

#### Access pattern description

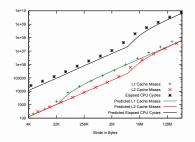
$$s\_trav(R.w = 3, u = 1, R.n = 1024)$$
  
 $\odot rr\_acc(R.w = 1, u = 1, R.n = 64, r = 1024)$ 

## Modeling random access without repetitive access

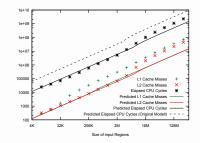


#### Results

#### Hash Join Build



#### Hash Join Probe



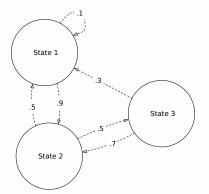
$$s\_trav() \odot r\_trav()$$

$$\oplus s\_trav() \odot rr\_acc() \odot s\_trav()$$

## Modeling dynamic effects using stochastic methods

- Some effects/components have dynamic state
- State can influence behavior and performance
- Analytical models are, by definition, stateless
- Stochastical models/processes can form the bridge between the two
  - Many exist: random walks, gaussian processes, levy-processes...
  - and most importantly: Markov Processes/Chains
  - This is one of Giuliano's core research interests
- (I am using them when I have to)

## (Discrete) Markov chains



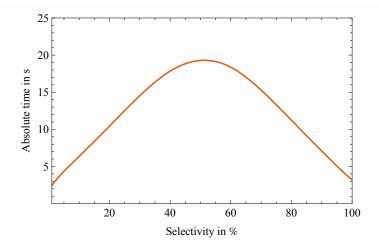
- · Basically a finite-state machine with transition probabilities
- They have "the Markov property": the next state is only dependent on the previous state and a random variable

## Modeling code

#### A simple loop

```
extern int* input; // uniform random ints between 0 and 100
int sum = 0;
for(size_t i = 0; i < inputSize; i++) {
        if(input[i] > s) {
            sum += input[i];
        }
}
```

## Modeling code



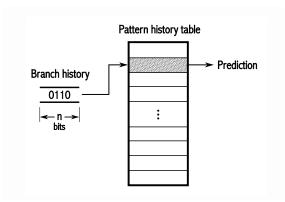
## Modeling code

#### A simple loop

```
extern int* input; // uniform random data
int sum = 0;
for(size_t i = 0; i < inputSize; i++) {
        if(input[i] > 20) {
            sum += input[i];
        }
}
```

What is the branch misprediction rate?

## Branch predictors

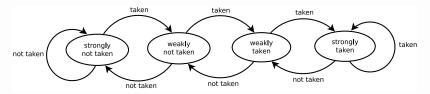


## Let's think about this in Markov terms



• Implemented in a *saturated counter* 

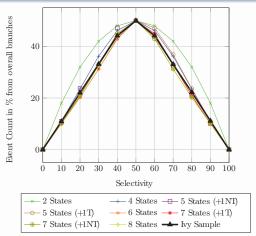
## The solution



- Probability of the branch predictor predicting taken:
  - The probability of it being in one of the states on the right
  - We can calculate the probability of it being in any state as the **stationary distribution**
- Branch misprediction rate:
  - $(P(pred\_taken) * P(act\_not\_taken)) + (P(pred\_not\_taken) * P(act\_taken))$

#### Validation

## Comparing stationary distribution with performance counter



## Modeling the entire system

- Is usually infeasible due to scale and noise
- We need to apply modeling with care
- Step 1: identify parts of the code that matter for performance using a profiler
  - Hot code sections (vtune calls this "bottlenecks")
- Step 2: re-create their relevant behavior in a controlled environment
- Step 3: Model
- Step 4: Validate
- We will practise this in the next interactive session

## Provide feedback, please!



https://co339.pages.doc.ic.ac.uk/feedback/modeling

#### Get the slides online



https://co339.pages.doc.ic.ac.uk/decks/PerformanceModeling.pdf