Visualization of Data Movements and Accesses

Til Mohr





Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
    for row in 0..2 {
      sum += matrix.get(row, column);
8 sum
```

Matrix:



Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix:

Matrix in Memory:

1 2 3 4



Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix: Matrix in Memory: 0 1 1 2 3 4

Current Item: Cache:

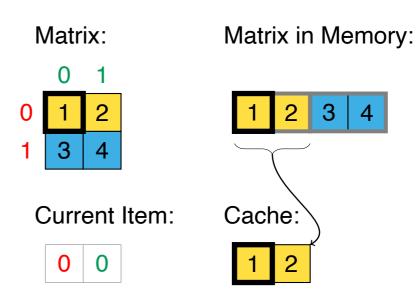
0 0





Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```







Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix: Matrix in Memory: 0 1 1 2 3 4

Current Item: Cache:

1 0

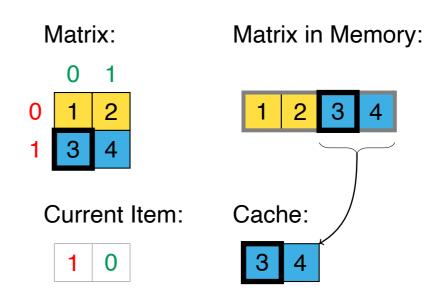
1 2





Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```



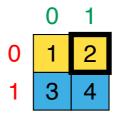




Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix: Matrix in Memory:





Current Item: Cache:

0 1

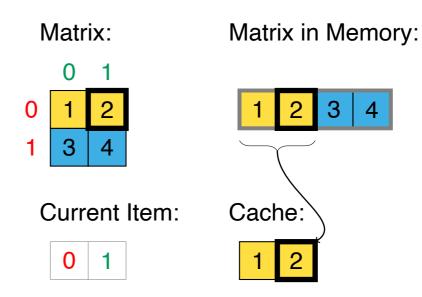
3 4





Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```



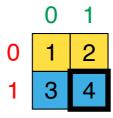




Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix: Matrix in Memory:





Current Item: Cache:

1 1

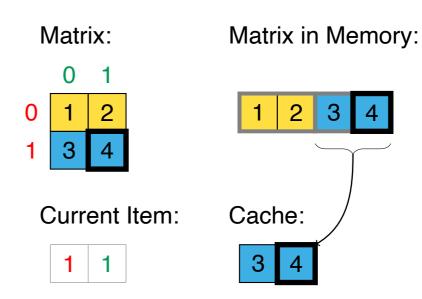
1 2





Listing 1: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for column in 0..2 {
4   for row in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```







Listing 2: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for row in 0..2 {
4   for column in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```



Current Item: Cache:

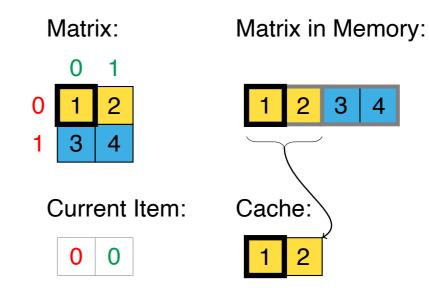
0 0





Listing 2: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for row in 0..2 {
4   for column in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

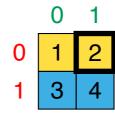




Listing 2: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for row in 0..2 {
4   for column in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix:



Matrix in Memory:



Current Item:



Cache:

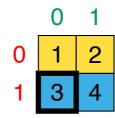




Listing 2: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for row in 0..2 {
4   for column in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix:



Matrix in Memory:



Current Item:



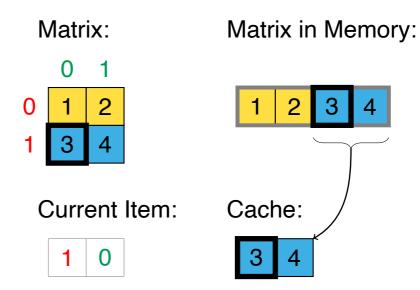






Listing 2: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for row in 0..2 {
4   for column in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```



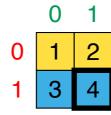




Listing 2: Matrix Summation

```
1 let matrix = Matrix::random(2, 2);
2 let mut sum = 0;
3 for row in 0..2 {
4   for column in 0..2 {
5     sum += matrix.get(row, column);
6   }
7 }
8 sum
```

Matrix: Matri



Matrix in Memory:



Current Item:







Outline

- Memory-Related Performance Problems
 - Data Locality
 - Processor-Memory Performance Gap
- Overview of the Optimization Workflow
- Data Gathering Approaches
- Visualization Techniques
- Specific Optimization Tool
- Conclusion



Memory-Related Performance Problems I Data Locality

$$t_{avg} = p \cdot t_c + (1 - p) \cdot t_m \tag{1}$$

t_{avg}: average access time

p : cache hit percentage

 t_c : cache access time

 t_m : memory access time

$$t_c \ll t_m$$
 (2)

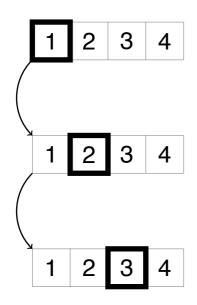




Memory-Related Performance Problems I Data Locality

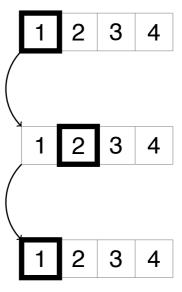
Spacial Locality

 Data that is referenced spatially close together is likely to be accessed in the near future



Temporal Locality

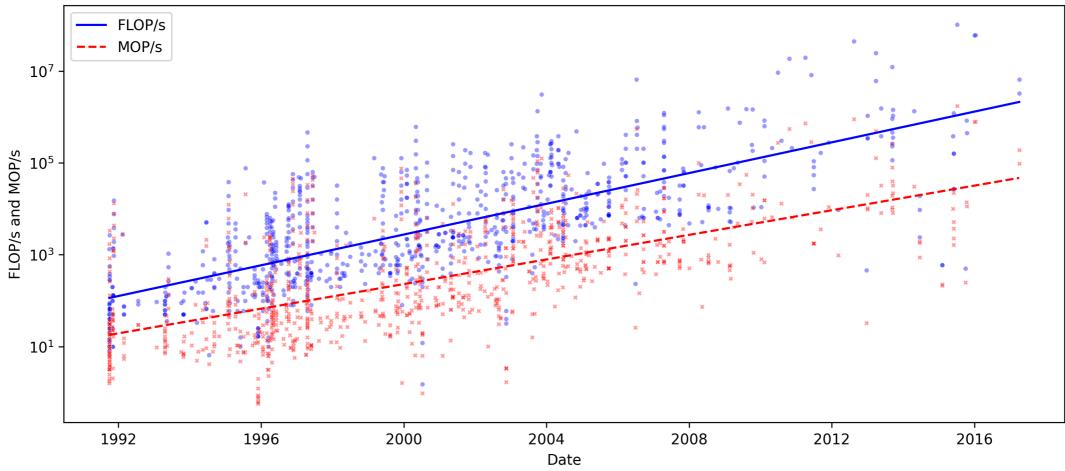
 Data that is referenced in the near past is likely to be accessed in the near future





Memory-Related Performance Problems I Processor-Memory Performance Gap

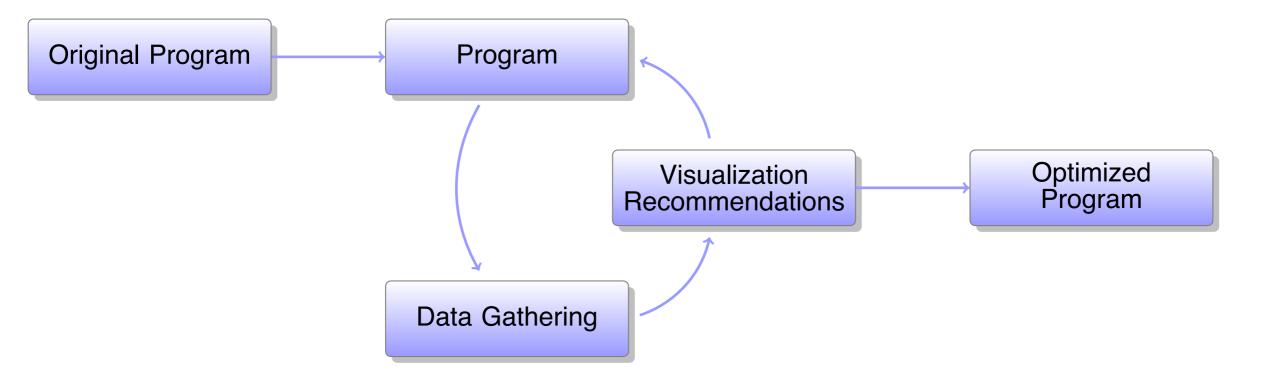
FLOP/s and MOP/s vs Date





Overview of the Optimization Workflow

Visualization-Guided Optimization







Data Gathering Approaches

Goal: Acquire Memory-Related Data for Visualization

- Data Accesses
 - Memory locations / variables
 - Frequencies
- Data access patterns
 - Nested loops
- Cache performance
 - Hit/miss rates
 - Utilization
 - Amount of data transfer in between different cache levels and main memory



Data Gathering Approaches I Dynamic Analyis

Run Program and Capture Memory-Related Information

- Hardware counters
 - Counts cache hits/misses
- Tracing / profiling
- Store source code references alongside memory-related information
- Very accurate
 - Real program data
 - Actual physical hardware

- (3) Can be very slow
- (3) Possible large overhead for very granular data
- Cannot easily analyze just parts of the program



Data Gathering Approaches I Static Analyis

Analyze the Programs Source Code for Data Accesses

- Extract any data access information purely from the source code
- Compile the program into a Data-Flow Oriented IR
- Statistics gathered by analyzing the IR
 - Algorithmic intensity
 - Volume of data circulating in the program

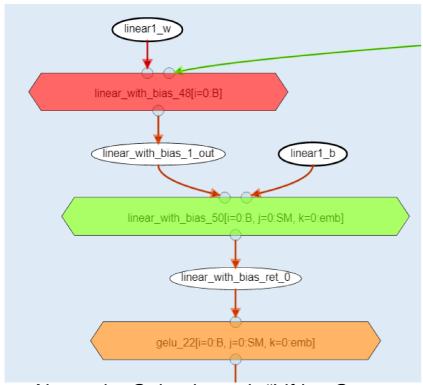
Source: Alexandru Calotoiu et al. "Lifting C semantics for dataflow optimization". In: Proceedings of the 36th ACM International Conference on Supercomputing. 2022, pp. 1-13.





Data Gathering Approaches I Static Analyis

Analyze the Programs Source Code for Data Accesses



Source: Alexandru Calotoiu et al. "Lifting C semantics for dataflow optimization". In: Proceedings of the 36th ACM International Conference on Supercomputing. 2022, pp. 1-13.

- Very fast
- Provides holistic view of the program and its performance
- Very abstract analysis
 - Memory layout of data is not considered
 - Hardware architecture unknown
 - No information about real-world cache performance





Data Gathering Approaches I Cache Simulation

Imitate the Programs Memory Accesses on a Simulated Cache Hierarchy

- Replicate actual hardware through software
 - Cache hierarchy (size, associativity, etc.)
 - Cache replacement policies
 - Cache coherence protocols
- Simulate the programs memory-wise on the simulated hardware
 - Memory (de-)allocations
 - Data accesses
- Very detailed
 - Insights about the memory-layout of data
 - Enables step-by-step analysis of the caches
- ② Allows to analyze only parts of the program

Requires precise parameterization





Visualization Techniques

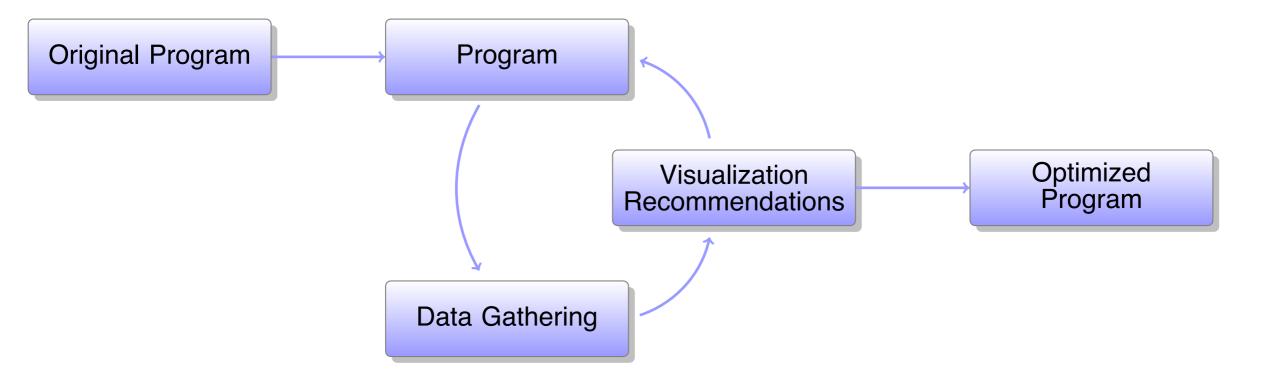
Goal: Display Bottlenecks (and explain them)





Overview of the Optimization Workflow

Visualization-Guided Optimization





References I

Alexandru Calotoiu et al. "Lifting C semantics for dataflow optimization". In: *Proceedings of the 36th ACM International Conference on Supercomputing*. 2022, pp. 1-13.

