Profiling Dataflow Systems on Multiple Abstraction Levels



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Compiling Dataflow Systems are Everywhere! Dataflow systems in different areas























Query

df_sales.join(df_CPUs, col("df_sales.cpuID" === col("df_CPUs.ID"), "inner")

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loopTuples:

0%	%localTid = phi [%1, %]
0.1%	%3 = getelementptr int8
0.1%	%4 = getelementptr int8
2.2%	%5 = load int32 %4, %lo
2.3%	%7 = crc32 i64 59616971
1.5%	%8 = crc32 i64 22314097
1.2%	%9 = rotr i64 %8, 32
2.3%	%10 = xor i64 %7, %9
2.2%	%11 = mul i64 %10, 2685
1.2%	%12 = shr %11, 16
2.4%	%13 = getelementptr int
32.1%	%14 = load int32 %40, i
0.2%	%15 = isnotnull ptr %12
0.3%	condbr %15 %loopHashCha
	1



loopBlocks S 8 %state, i **8 %3, i64** 20 ocalTid 17643560850 79111444414





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Dataflow System

n	table T	
0.	i64 %13	

loopTuples:

0%	%localTid = phi [%1, %loopBlocks %2, %
0.1%	%3 = getelementptr int8 %state, i64 32
0.1%	%4 = getelementptr int8 %3, i64 262144
2.2%	%5 = load int32 %4, %localTid
2.3%	%7 = crc32 i64 5961697176435608501, %5
1.5%	%8 = crc32 i64 2231409791114444147, %5
1.2%	%9 = rotr i64 %8, 32
2.3%	%10 = xor i64 %7, %9
2.2%	%11 = mul i64 %10, 2685821657736338717
1.2%	%12 = shr %11, 16
2.4%	%13 = getelementptr int8 %5, i64 %12
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0.2%	%15 = isnotnull ptr %12
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32.

























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Profiling Dataflow Systems on Multiple Abstraction Levels





Query **Dataflow System** Dataflow Graph Imperative Prog.

Machine IR

x86 Assembly

df_sales.join(df_CPUs, col("df_sales.cpuID") ...)

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Machine IR Results

Profiling Samples

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Tailored Profiling Closing the gap

- Track connection between components of all abstraction levels down to generated code
- Map profiling samples back to higher abstraction levels
- Ingredients
- Tagging Dictionary & Register Tagging

(1) Connection tracking of abstraction components for each lowering step (top-down)

② Store mapping in the *Tagging Dictionary*

(3) Map profiling results to each abstraction level's components (bottom-up)

(4) Aggregate the data for profiling results

for each tuple t in sales • • • if t.price > 500• • •

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Generated Query Code

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Generated Query Code

for each tuple t in sales

if *t*.price > 500

{for each tuple t in sales s -> Scan sales}

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Generated Query Code

for each tuple t in sales

→if t.price > 500

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Generated Query Code

for each tuple t in sales

call malloc(...)

if t.price > 500

call malloc(...)

Tagging Dictionary

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malloc(...)

Generated Query Code

for each tuple t in sales

call malloc(...)

if *t*.price > 500

call malloc(...)

Scan, Filter ?

Tagging Dictionary

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malloc(...)

Generated Query Code

for each tuple t in sales

call malloc(...)

if *t*.price > 500

call malloc(...)

Scan, Filter ?

Call-Stack Sample

Generated Query Code

for each tuple t in sales call malloc(...) if t.price > 500call malloc(...)

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Machine Register

for	ea	ach
	se	tTa
	ca	11
	un.	set
	if	t .
		se
		Са
		un

Generated Query Code

```
n tuple t in sales
ag(Scan)
malloc(...)
tTag()
price > 500
etTag(Filter)
all malloc(...)
nsetTag()
```

Machine Register

Generated Query Code

for each tuple t in sales setTag(Scan) **if** *t*.price > 500 setTag(Filter) call malloc(...) unsetTag()

Machine Register

Scan

Profiling Sample		
	Source Line	Register V
	malloc()	Scan

Generated Query Code

for each tuple t in sales setTag(Scan) call malloc(...) **if** *t*.price > 500 setTag(Filter) → call malloc(...) unsetTag()

Machine Register

Filter

Profiling Sample		
	Source Line	Register \
	malloc()	Filte

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Time per operator

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Time per operator

Context-aware profiling over time

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Time per operator

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Profiling Dataflow Systems on Multiple Abstraction Levels

Context-aware profiling over time

Memory access patterns

Little implementation effort

Component	Lines Added	
Code Gen.	56	
Sample Processing	1176	
Visualizations	510	

- Little implementation effort
- High accuracy

Component	Lines Added
Code Gen.	56
Sample Processing	1176
Visualizations	510
Attributed Samples	98 %

- Little implementation effort
- High accuracy
- Lightweight
- Small query execution overhead
- Tagging Dictionary: populated at compile time
- Register Tagging: 3%
- Cheaper than call-stack sampling

Component	Lines Added
Code Gen.	56
Sample Processing	1176
Visualizations	510

98 %

Impact of Tailored Profiling Where can you apply it?

- Preserve connection information to close gap
- Profiling results on high abstraction levels

Profiling Dataflow Systems on Multiple Abstraction Levels

Impact of Tailored Profiling Where can you apply it?

- Preserve connection information to close gap
- Profiling results on high abstraction levels
- Lightweight, high accuracy
- Easy to integrate
- Applicable to many systems

Impact of Tailored Profiling Where can you apply it?

- Preserve connection information to close gap
- Profiling results on high abstraction levels
- Lightweight, high accuracy
- Easy to integrate
- Applicable to many systems
- Already supported: profiling code on CPUs (multisocket and multicore)
- Future work: heterogenous compute resources, distributed systems

What else can be found in the Paper

- Technical and implementation details
- Tagging Dictionary for multiple abstraction levels
- Register Tagging and call-stack sampling
- Supported optimizations

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- Integration details for our compiling DBMS Umbra

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- Tagging Dictionary for multiple abstraction levels
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- Supported optimizations
- Integration details for our compiling DBMS Umbra
- More detailed evaluation, runtime overhead and use cases

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Thank you for watching!

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