

Automatic Detection of MPI Application Structure with Event Flow Graphs

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Trace

- Full temporal order of events is preserved
- A lot of data to store, process, analyze
- Profile (summary)

100x A 42x B 33x C

- 17x D
- Temporal order is not preserved
- Far less data

Implementation in IPM¹

- Keep data in a hash table
- Keys: event (-signatures)
- Values: **statistics** (#calls, duration, ...)

key		#calls	duration			
В		42	23.1			
	A	100	12.0			
ntegrated Performance Monitor ttp://ipm-hpc.sourceforge.net/						

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Something in Between Profiling and Tracing...

Event Flow Graphs (EFGs)

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- Keep a history of the previous event that happened
- Keep track of pairs of events (prev., curr.) instead of single events



- Similar to a control flow graph, but
 - records tansitions that have actually happened in an execution
 - records how many times these transitions have happend









```
int main() {
    MPI_Init(...);
    for(i=1;i<=10;i++) {
        MPI_Bcast(...);
        if(i%2) /* odd */
            MPI_Recv(...);
        else /* even */
            MPI_Send(...);
    }
    MPI_Finalize();
}</pre>
```

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> In this case, the trace cannot be uniquely reconstructed from the EFG.





Temporal EFG (t-EFG):

– A modified version of an EFG that **guarantees** trace recovery

Ideas

- At each node, keep track of which outgoing edge to take next
- Represent this information in a compact way
- t-EFG for the previous example:
 - Edge label describes a **partition** of the iteration space



Using t-EFGs for Trace Compression MAXIMILIANS-UNIVERSITÄT

Runtime data collection is still efficient

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- Around 2% overhead in terms of execution time
- See: [EuroPar '14]: Xavier Aguilar, et al. MPI Trace Compression using Event **Flow Graphs**
- Compression results for some benchmarks [EuroPar '14] (sequence of events only)

Benchmark	# Ranks	Comp. Factor	
AMG	96	1.76x	
GTC	64	46.60x	
MILC	96	39.03x	
SNAP	96	119.23x	Compression!
MiniDFT	40	4.33x	
MiniFE	144	19.93x	
MiniGhost	96	4.85x	



- Compression ratio depends on the structure of the graphs
 - Simple graphs with few nodes and edges correspond to high compression ratios

Benchmark	Avg. Compr. Ratio	Avg. Num of Nodes	Avg. Num of Edges	Avg. Node Cardinality
AMG	1.76	9,384.94	10,586.47	4.59
MiniDFT	4.33	690.30	1,980.38	27.29
SNAP	119.23	28	1,120.26	14,149.22
GTC	46.60	114.5	121.20	109.10







- MiniGhost example application
 - 3160 events in the trace
 - 87 nodes, 90 edges in the EFG

Compressing sequences (chains)

- 13 nodes, 16 edges
- Nested loops (cycles) visible



Application Structure

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- Structure:= loops and their nesting
- Folklore: "big outer loop hypothesis": most scientific applications are dominated by a big outer time-stepping loop
- Detecting Structure
 - If a loop contains MPI calls, the loop will show up as a cycle in the Event Flow Graph



- Detecting cycles in flow graphs is a common requirement for (de-)compilers
 - Many algorithms exist

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 We used an efficient DFS-based algorithm by T. Wei et al., "A New Algorithm for Identifying Loops in Decompilation", 2007





Loop Detection Results

			Outermost Loop(s)			
Benchmark	# Ranks	Total Runtime (sec)	Count	Time in all	Time in dominant	
MiniGhost	96	282.17	1	98.8%	98.8%	
MiniFE	144	133.50	13	78.1%	77.7%	
BT	144	370.59	7	99.4%	99.0%	
LZ	128	347.53	3	99.2%	98.9%	

 "Big outer loop hypothesis" largely holds for these (and other) example benchmarks







So far: post-mortem operation



Now: Online operation



- Steady state?
 - No \rightarrow do nothing
 - Yes \rightarrow perform loop detection
- At main loop header?
 - No \rightarrow do nothing
 - Yes → collect trace for N iterations ("smart data collection")

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- Application structure can be detected online, while the application runs
 - Reduce redundant data, change data granularity, etc
- The event flow graph becomes stable once the application enters its iterative phase
- Our mechanism checks the number of nodes in the graph to detect application stability to trigger the loop detection mechanism







- Six applications representing typical scientific codes
 - MiniGhost
 - MiniFE
 - MiniMD
 - GTC
 - LU
 - BT
- Cray XE6 with 2 twelve-core AMD MagnyCours at 2.1 GHz
 - 32 GB DDR3 memory per node
 - Nodes interconnected with Cray Gemini network



Metric	Mini- Ghost	MiniFE	GTC	MiniMD	BT	LU
Trace size	26 MB	77 MB	48 MB	555 MB	717 MB	7.7 GB
10 iterations trace	4.4 MB	4.1 MB	1.3 MB	788 KB	29 MB	267 MB
% reduced	83%	94.7%	97.3%	99.8%	96%	96.53%

- Detect the application structure on-line to keep tracing information of only 10 iterations of the main loop
- If the application is regular, a few iterations will represent the overall performance behaviour
- Performance results (statistics) still representative







Example: MiniGhost





No source instrumentation needed

Conclusions

- Graphs captured through the PMPI interface
- Some use cases:

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- Map performance data to program structure
- Reduce amount of data collected while application runs
- Converting t-EFGs to trees onging work
 - Exciting possibilities: analysis, modeling, code generation, ...