



StackMine – Performance Debugging in the Large via Mining Millions of Stack Traces

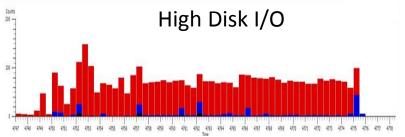
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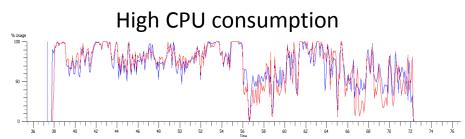




Performance Issues in the Real World

- One of top user complaints
- Impacting large number of users every day
- High impact on usability and productivity





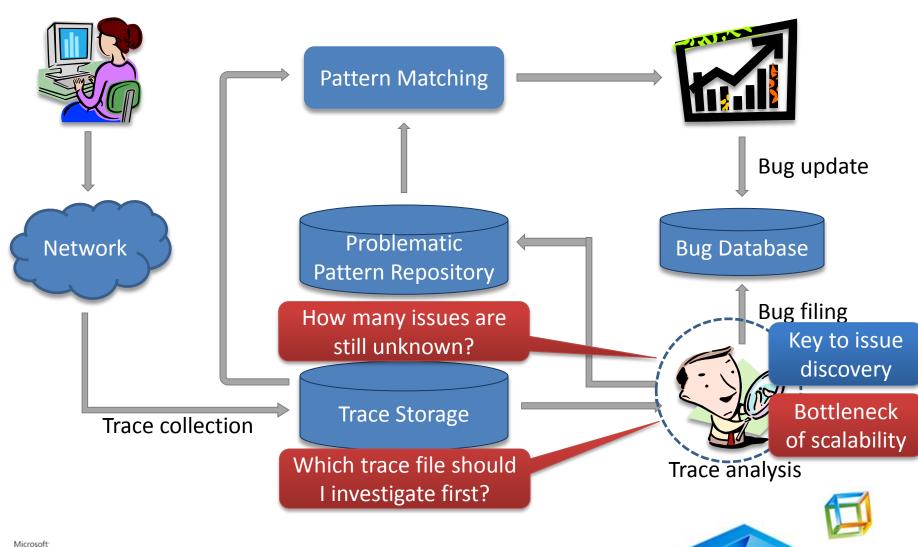
Given **limited** time and resource **before** software release, **development-site** testing and debugging become **insufficient** to ensure satisfactory software performance.







Performance Debugging in the Large







Problem Definition

Input: Runtime traces

collected from millions of users

Output: Program execution patterns causing the most impactful performance problems

inspected by performance analysts







Goal

Conduct systematic discovery & analysis of program execution patterns

- Efficient handling of large-scale trace sets
- Automatic discovery of new patterns
- Effective prioritization of investigation





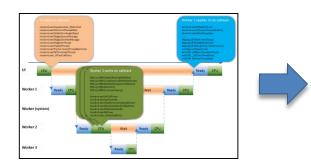


Challenges



Large-scale trace data

- TBs of trace files and increasing
- Millions of events in single trace stream



Highly complex analysis

- Numerous program runtime combinations triggering performance problems
- Multi-layer runtime components from application to kernel being intertwined

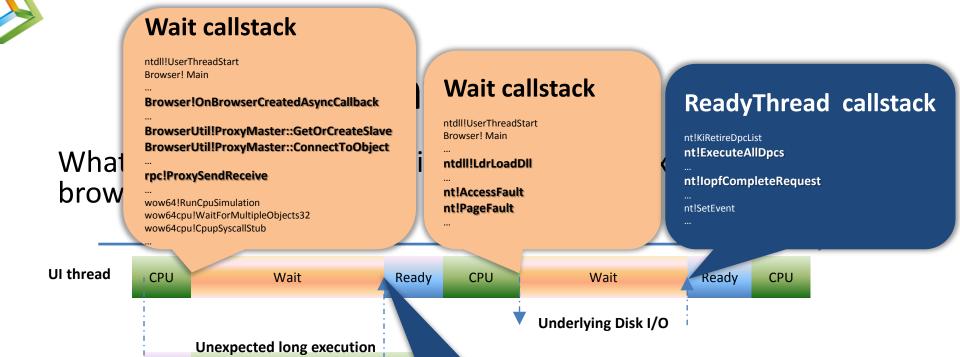


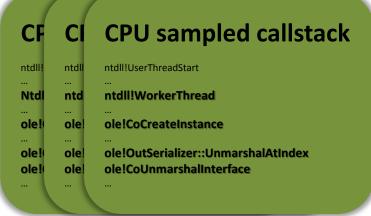
Combination of expertise

 Generic machine learning tools without domain knowledge guidance do not work well



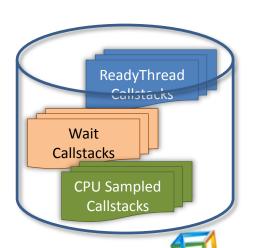






CPU

ReadyThread callstack ntdll!UserThreadStart ... rpc!LrpcloComplete ... user32!PostMessage ... win32k!SetWakeBit nt!SetEvent ...



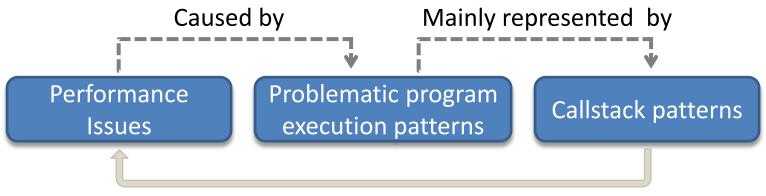


Worker thread Ready



Approach

Formulated as a callstack mining and clustering problem



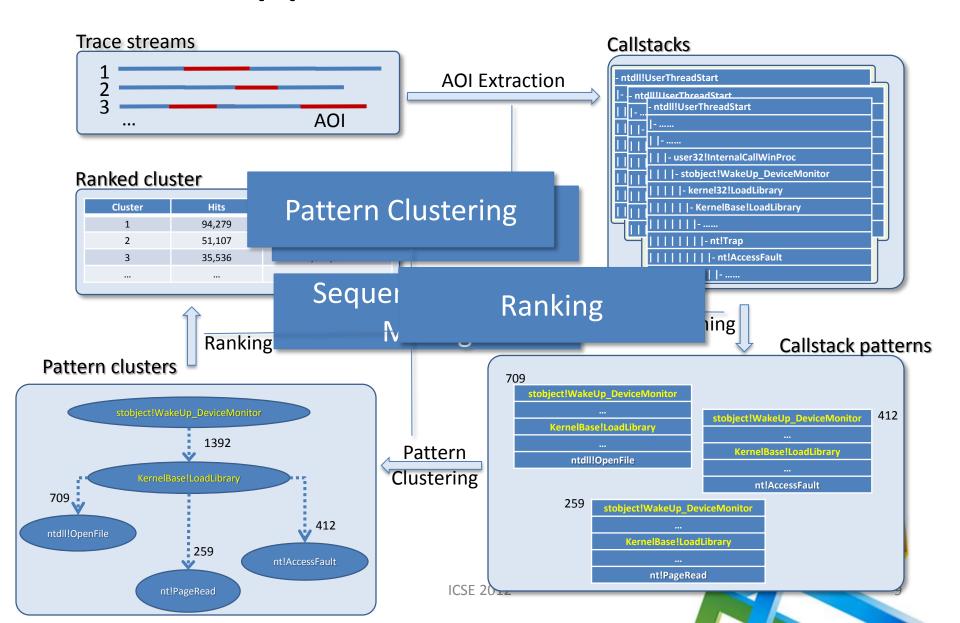
Discovered by mining & clustering costly patterns





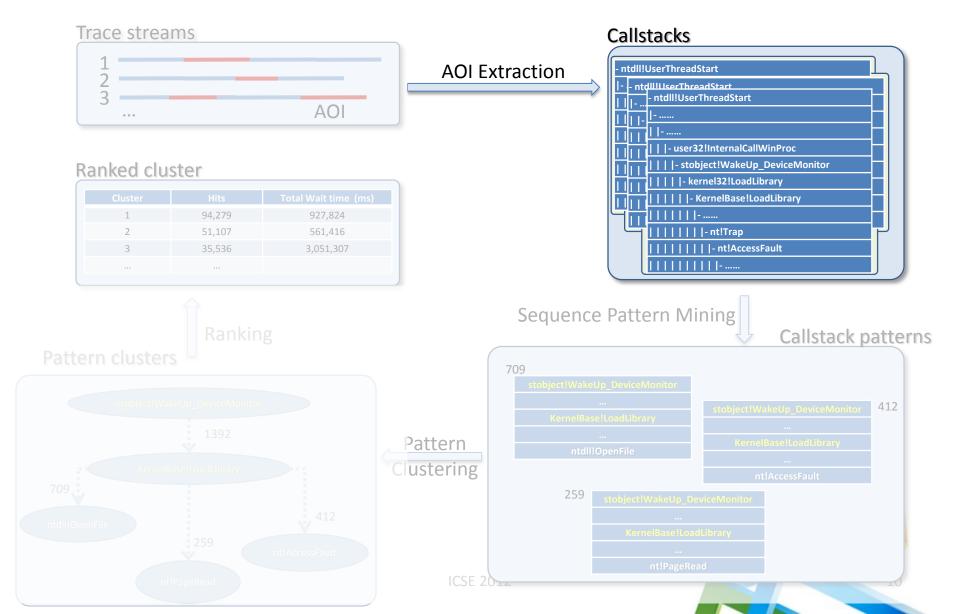


Approach – Workflow





Approach – AOI Extraction





Approach – AOI Extraction Motivation

Runtime traces capture both

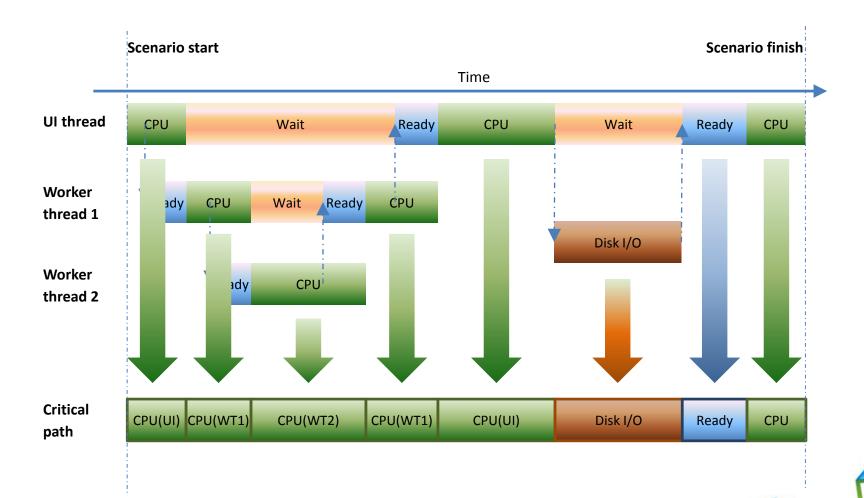
- Relevant executions for performance issue
 - E.g., executions relevant to browser-tab creation
- Irrelevant executions for performance issues
 - E.g., executions of concurrently executed IM
 - Noisy data for mining
 - Huge investigation scope induced





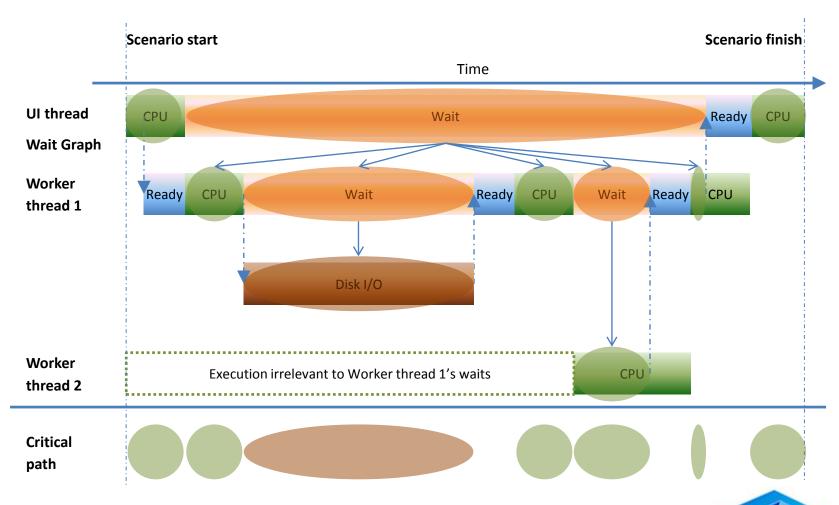


Approach – AOI Extraction Critical Path





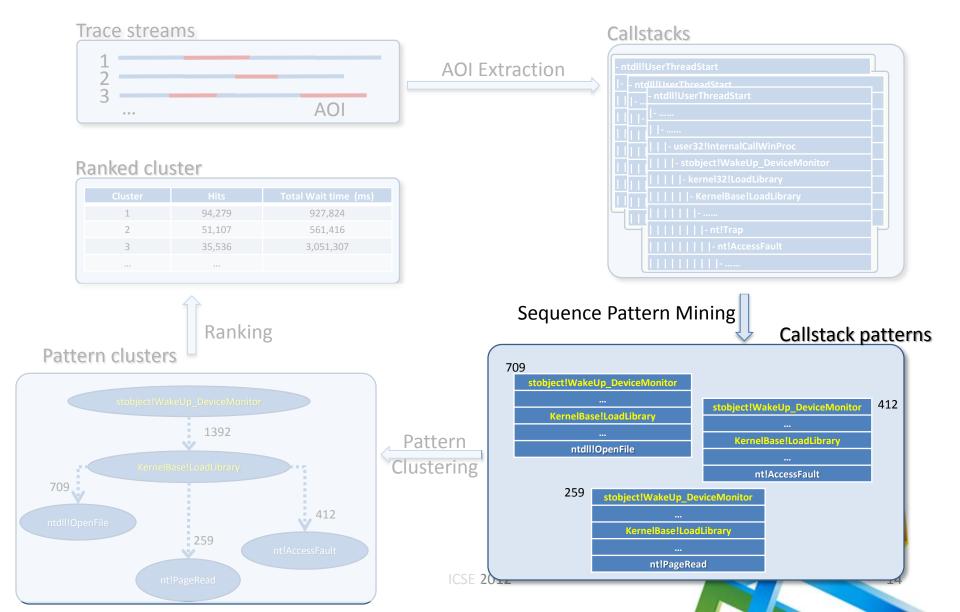
Approach – AOI Extraction Wait Graph







Approach – Callstack Pattern Mining



Approach – Callstack Pattern Mining Costly Sequence Pattern

Example

Sequence	Wait time
АВ	4 ms
ABCDF	1 ms
ABCEF	3 ms
ABGF	2 ms

Cost threshold *T*5 ms (or 50% of total)



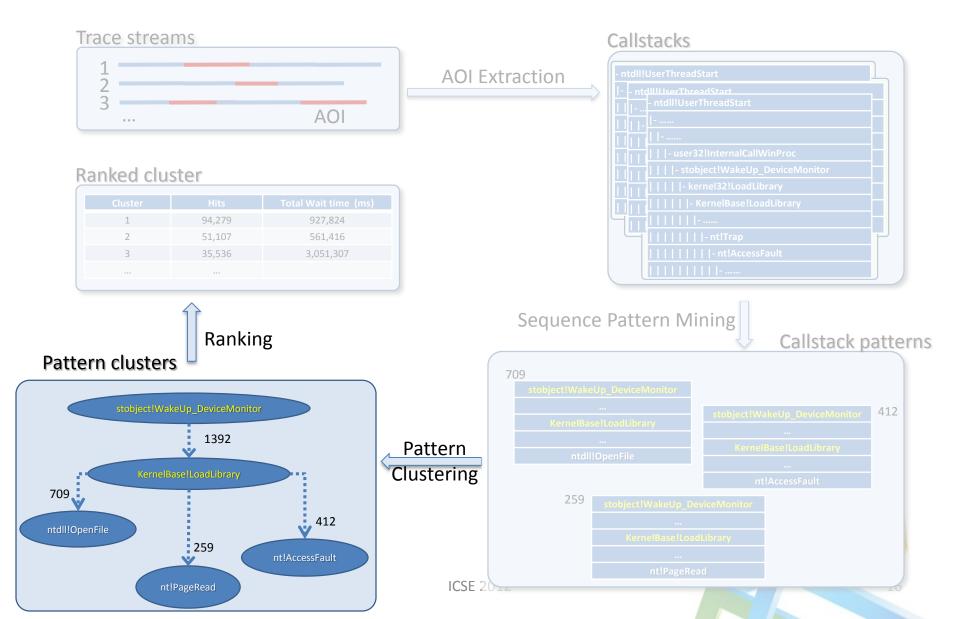
All Patterns	Support	Closed Patterns	Support	Maximal Patterns	Support
A, B, AB	10	A, B, AB	10	A, B, AB	10
F, AF, BF, ABF	6	F, AF, BF, ABF	6	F, AF, BF, ABF	6

- Non-consecutive <u>costly</u> sub-sequence as callstack pattern
- Costly maximal patterns are compact





Approach – Callstack Pattern Clustering





Approach – Callstack Pattern Clustering

Motivation

- Same issue often reflected by variant patterns
- Defect often hidden in invariant parts of variant patterns

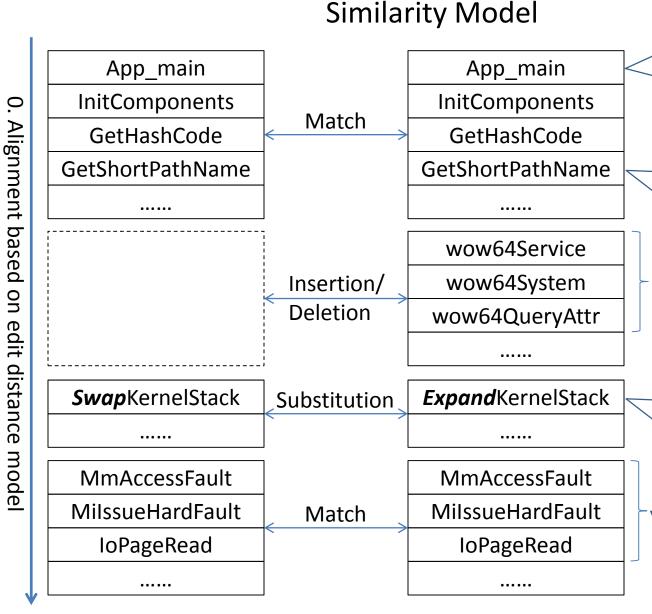
Goal

- Precise measurement of issue impact for better prioritization
- Comprehensive issue representation with pattern variations for quick and precise fixing





Approach – Callstack Pattern Clustering



- Common-purpose function: weight ↓
 Uni() is small
 - Special-purpose function: weight↑
 Uni() is large
 - Variant part representing non-essential factors
- Similar names implying relevant functionalities
 - 5. Constant call path: weight↓ FBi() + BBi() is small





Technical Highlights

- Machine learning for system domain
 - Formulate the discovery of problematic execution patterns as callstack mining & clustering
 - Systematic mechanism to incorporate domain knowledge
- Interactive performance analysis system
 - Parallel mining infrastructure based on HPC + MPI
 - Visualization aided interactive exploration







Evaluation – Windows 7 Study

- Task: since Dec 2010, a continued effort to improve Windows performance
 - Analysts from one performance analysis team for Microsoft Windows
 - Hunt for hidden performance bugs that caused common impact on Windows Explorer UI response
 - Based on over 6,000 trace streams
- Data
 - 921 qualified out of 1,000 randomly sampled trace streams
 - 181 million callstacks in total
 - 140 million wait callstacks
 - 41 million CPU sampled callstacks







Evaluation – Windows 7 Study Research Questions

 RQ1. How much does StackMine improve practices of performance debugging in the large?

• *RQ2*. How **well** do the derived performance signatures **capture** performance bottlenecks?

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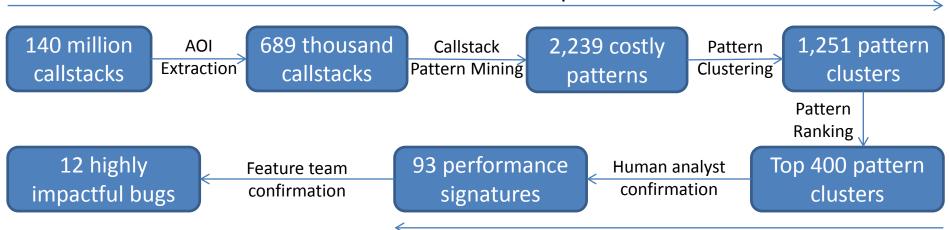
 RQ3. How much does StackMine outperform alternative techniques?



Evaluation – Windows 7 Study RQ1. Overall Improvement of Practices

- Traditional approach would take 20~60 days
- Using StackMine 18 hours

10 hours of automatic computation



8 hours of one human analyst's review





Evaluation – Windows 7 Study RQ2. Performance Bottleneck Coverage

 Performance Bottleneck Coverage (PBC) of a set of performance signatures

$$PBC = \frac{Total\ time\ of\ performance\ signatures}{Total\ time\ of\ collected\ trace\ streams}$$

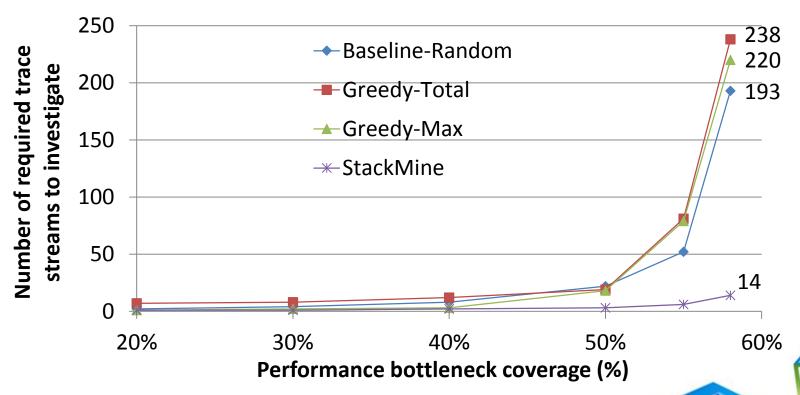
- The higher PBC achieved, the lower possibility that high-impact performance bugs remain not captured
- 58.26% PBC achieved by the 93 signatures





Evaluation – Windows 7 Study RQ3. Comparison with Alternative Techniques

StackMine requires only **7.2%**, **5.8%**, and **6.3%** of trace streams required by the other three techniques





Impact



"We believe that the MSRA tool is highly valuable and much more efficient for mass trace (100+ traces) analysis. For **1000 traces**, we believe the tool **saves us 4-6 weeks** of time to create new signatures, which is quite a significant productivity boost."

- from Development Manager in Windows

Highly effective new issue discovery on Windows mini-hang





Continuous impact on future Windows versions







Conclusion

- The first formulation and real-world deployment of performance debugging in the large
 - as a data mining problem on callstacks

- A mining-clustering mechanism for reducing costlypattern mining results
 - based on domain-specific characteristics of callstacks

 Industrial impact on using StackMine in performance debugging in the large for Microsoft Windows







Acknowledgment

Our partners in Microsoft product teams

The researchers from Microsoft Research









Q&A

Thank you!

