G²: A Graph Processing System for Diagnosing Distributed Systems

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Cost of Debugging

• The huge printing presses for a major Chicago newspaper began malfunctioning ...



\$10,000 = \$1 + \$9,999

Most bugs can be fixed quickly, however identifying the root causes is hard.

Motivation

- Diagnosing distributed systems is frustrating
 - Execution is too complex to comprehend
 - Tons of logs, but correlations are missing
 - Lost in the information sea







- Finds correlated information.
- Facilitates better summarization and reasoning
- Is fast and easy to use

Contribution

- Graph based diagnosis for distributed systems
 - Execution graph to capture correlations
 - Graph based diagnosis operators
 - Slicing for finding & filtering
 - HierarchicalAggregation for summarization
- Declarative diagnosis queries
 - Integrated with Microsoft LINQ
- Distributed engine
 - Integrated relational computation and graph traversal
 - Optimizations based on the characteristics of the execution graph and diagnosis operators

Outline

- Model
- Engine
- Programming
- Evaluation

Capture Correlations



Execution is Graph



Slicing: Find the correlated subgraph and filter others by traversing the execution graph



Slicing: Find the correlated subgraph and filter others by traversing the execution graph





Understand Execution Graph

- Execution graph is rather huge
 - A 2-hour SCOPE/Dryad graph has over 1.2 billion vertices,
 0.54 billion edges, and lots of user payload(logs)
- Connected subgraph is also huge
 - However, intra-machine interactions are much more than inter-machine ones(91% vs 9% in SCOPE/Dryad graph)
- Graph structure data is relatively small
 - User payload is over 64% in storage
- Iterative access to graph structure data
 - Concurrent traversals
 - Aggregation follows slicing

Optimize Graph Access

- Diagnosing tool as a distributed system
- Optimal partition on graph data
 - At machine boundary initially. Dynamic partitioning.
 - Local data is stored in database
- Caching
 - Graph structure data in memory
 - Retrieve payload only when necessary
- Prefetching
 - Get vertex properties during slicing, instead of during aggregation

Understand Slicing & HierarchicalAggregation

- Latency is an issue
 - More than 200 hops sometimes, due to deep paths
- Rigorous synchronization is not efficient
 - Different from Page Rank/Belief Propagation
- Aggregation repeatedly colors local vertices with the same aggregation identity
 - Lots of local messages

Optimize Fast Execution Graph Traversal

- Batched Asynchronous Graph Traversal
 - Explore local vertices until reaching cross-partition edges without synchronization

• Partition-level interface

- One traversal worker on each partition
- Direct access to the whole local graph data
- Local vertices could be condensed into super nodes in advance





Play with G²

- Capture the graph
 - Manual annotation, Binary rewriter and dynamic instrumentation

• Write simple C# queries

- Reuse existing relational operators in LINQ
- Slicing(Chopping) / HierarchicalAggregation
- Local Extensions: Diff, CriticalPath, ...
- Provide diagnosis wizards in Visual Studio



Evaluation

Systems	LOC(K)	Func#	Edge#	Event#	Raw(MB)	DB(MB)	Time(min)	Node#
Berkeley DB	172	46164	92502	186597	14	29	2	3
G ²	27	267,728	634,704	1,212,778	85	231	17	60
SCOPE/Dryad	1,577	3,128,105	8,964,168	20,106,457	1,226	3,269	120	60

Table 1: Per node graph statistics

Systems	Annotated Edge#	Annotated CS#	Instrumented Func#	Rules
Berkeley DB	2	2	1,542	23
G ²	9	11	197	10
SCOPE/Dryad	17	13	730	5

Table 2: Instrumentation statistics

- 60 machines
- 2 GHZ dual core
- 8 GB memory
- Two 1 TB disk
- 1 Gb Ethernet

End to End Query Performance



5770 random queries on the SCOPE/Dryad Graph

Events.Where (e => ...) .Slicing(Slice.Forward) .HierarchicalAggregate(e => e.Val.Process.ID);

Related Work

- Execution Model
 - Path based analysis
 - Pure log analysis
 - Static analysis
- Distributed Execution Engine and Storage
 - Graph systems
 - Map-reduce alike systems
- Diagnosis Platform
 - Cloud9: Testing as a service
 - Dapper: path analysis atop of BigTable

Conclusion

- Graph based diagnosis for distributed systems
 - Slicing for filtering the logs
 - HierarchicalAggregation for summarization
- Graph engine with specific requirements
 - Integrated relational computation and graph traversal support
 - Batched asynchronous graph traversal and partition-level interface for better performance



Thanks!

Generations for log structure and related tools: Text

- Unstructured text
 - Format: [AUTO: time, component, log level, pid, tid, location], `printf' message
 - Aggregation: by the meta information or keywords in the unstructured message
- Pros
 - Free style format
 - Easy to process: grep
- Cons
 - May miss many implicit dependencies among log entries without shared tag (e.g., request id)

Generations for log structure and related tools: Paths

- Path-based aggregation
 - Format:
 - [ANNOTATION: path id] + unstructured text
 - Optional: [ANNOTATION] dependencies among log entries belonging to the same path are captured
 - Aggregation: by the user request id (path id)
- Pros
 - Effective for request-centric analysis and modeling
 - The logs are partitioned by request id, and each partition can usually be handled by single machine
 - A nice balance between usability and the overhead
- Cons
 - Cut off interactions between requests, which is common in distributed systems, such as batching
 - Path is statically defined by the pre-defined `requests' only



Scaling Performance



Graph Traversal Interface

```
IQueryable<T> GraphTraversal<TWorker>(
        this Graph<TV, TE> g,
        IQueryable<Vertex<TV, TE>> startVertices
        ) where TWorker : GPartitionWorker<TV, TE, _, T>;
class GPartitionWorker<TV, TE, TMsg, T>
{
    public Vertex<TV, TE> GetLocalVertex(ID VertexID);

— public void SendMessage(ID VertexID, TMsg msg);

    public void WriteOutput(T val);
    public virtual void Initialize(VertexIterator<TV, TE>);

public virtual void OnMessage(Vertex<TV, TE>, TMsg msg);

    public virtual void Finalize();
}
```

```
class GPartitionSlicingWorker<TV, TE> : GPartitionWorker<TV, TE, bool, Vertex<TV, TE>>
{
    private HashSet<ID> VisitedVertices;
    public override void Initialize(VertexIterator<TV,TE> inits)
         foreach (var v in inits)
             SendMessage(v.ID, true);
    }
    public override void OnMessage(Vertex<TV,TE> v, bool msg)
    {
        if (VisitedVertices.Contains(v.ID)) return;
        VisitedVertices.Add(v.ID);
        WriteOutput(v);
        foreach (var e in v.OutEdges)
            if (e.IsCausal())
                SendMessage(e.DstVertexID, true);
}
```

Experience using G²



Deployment Issues

- Capture the correlations
 - Instrument the network and thread pool libraries to capture the asynchronous transitions among threads and machines
- Store and process the logs
 - Option 1: dedicated graph engine (G²)
 - Pros: complete support of G² diagnosis queries
 - Cons: interference to host systems
 - Option 2: in-app graph engine with latest logs
 - Pros: lightweight, easy to deploy
 - Cons: limited memory cache capacity (latest logs only)