## ENHANCING DISTRIBUTED TRACING TO ORDER EVENTS

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# DISTRIBUTED SYSTEMS TODAY Google

Distributed systems surround us.

Some examples:

- Google
- Facebook
- Cellular networks\*

Distributed systems are

Networks of devices/machines ("nodes"), such as computers or servers that communicate with one another to complete tasks

Q snacks for scooby-doo			×	Ļ
	Google Search	I'm Feeling Lucky		



## WHAT IS **DISTRIBUTED TRACING?**

*Distributed tracing* is when a software tracks the flow of service requests in a distributed system

This is useful for:

- Debugging
- Regulation
- Analyzing performance



https://www.scaleyourapp.com/difference-between-centr alized-decentralized-distributed-systems-explained/

https://static.googleusercontent.com/media/research.goo gle.com/en//pubs/archive/36356.pdf

https://www.uplers.com/wp-content/uploads/2020/04/Top -5-Debugging-Tools-Your-Front-end-Developers-Should-Have-Hands-On.jpg

## THE IDEAL TRACE: CAUSALITY

 $A_{S}$ 

A<sub>F</sub>

 $\mathsf{B}_{\mathsf{S}}$ 

 $\mathsf{B}_\mathsf{E}$ 

Happens-before relationship

 A<sub>s</sub> must occur for B<sub>s</sub> to
 occur

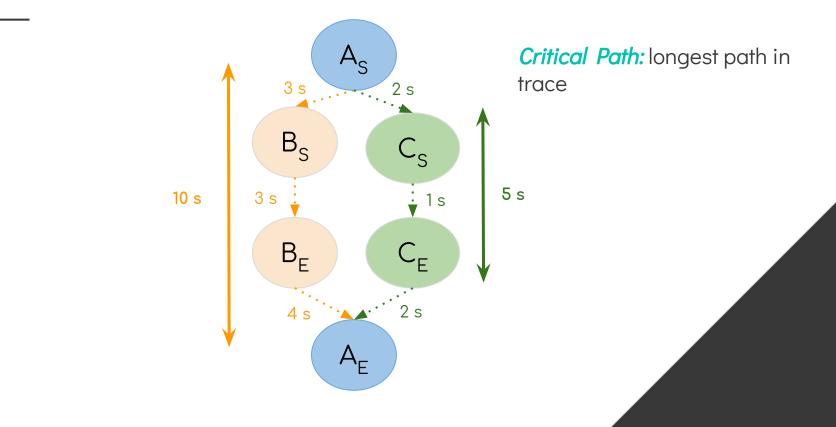
- Concurrent relationship

   Events could happen in any order or simultaneously
  - Synchronization

 $A_{S}$ 

 $\mathsf{B}_{\mathsf{S}}$ 

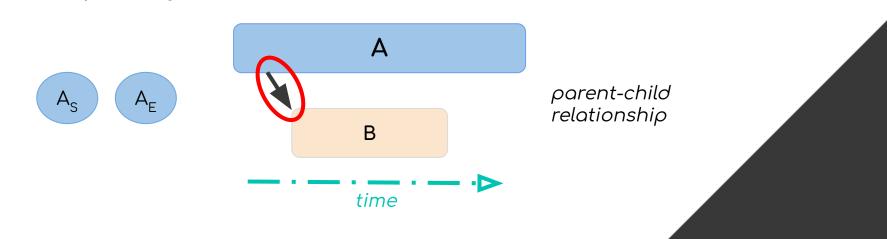
## THE IDEAL TRACE: APPLICATIONS



## TRACING: **SPANS**

Span (definition):

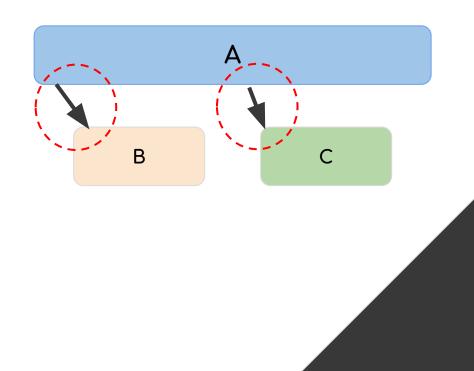
"The 'span' is the primary building block of a distributed trace, representing an individual unit of work done in a distributed system." – OpenTracing.io



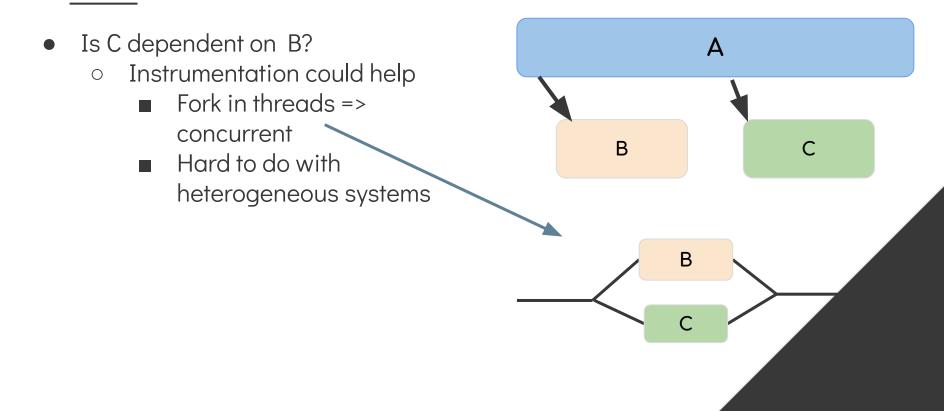


## TRACING: THE CHALLENGES

- Caller-callee relationships
  - B dependent on A
  - C dependent on A
- Is C dependent on B?



## TRACING: THE CHALLENGES





# CAN WE USE **BIG DATA** TO GET CLOSER TO THE IDEAL TRACE?





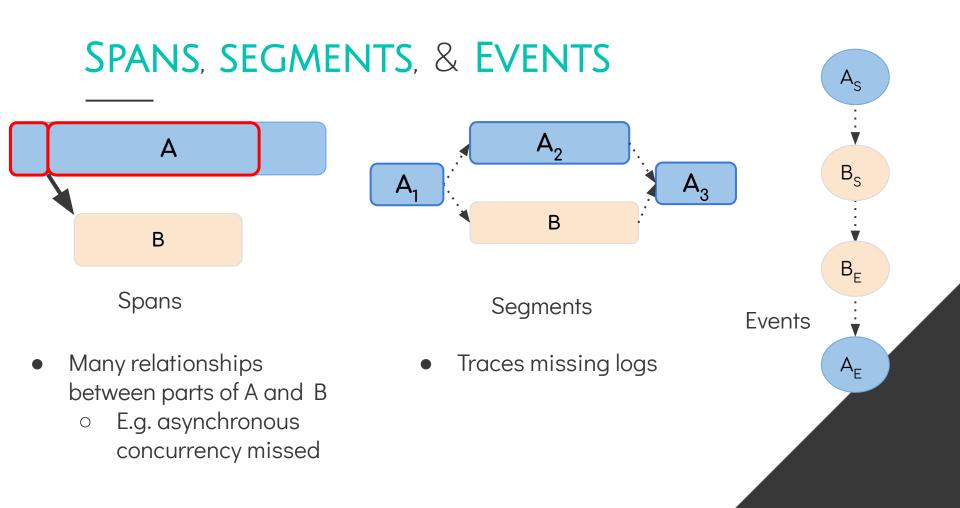
## OVERVIEW

#### 01 DISTRIBUTED TRACING

#### 02 THE MYSTERY MACHINE

03 SCOOBY SYSTEMS

04 FUTURE WORK



## THE MYSTERY MACHINE: GCM

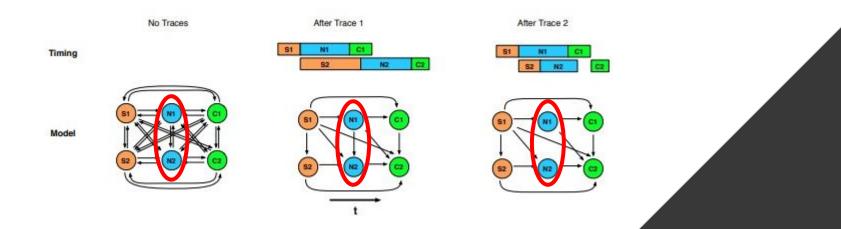


The Mystery Machine produces a Global Causal Model (GCM)

- Uses a segment-based model.
- shows happens-before dependencies between every segment across the traces such that the dependencies hold for *every trace*.

## THE MYSTERY MACHINE: HOW IT WORKS

- First creates all hypothetical happens-before edges between all segments to create GCM
- Then takes happens-before relationships across the traces and removes violated edges from GCM
- Reaches final GCM once it has iterated through all traces



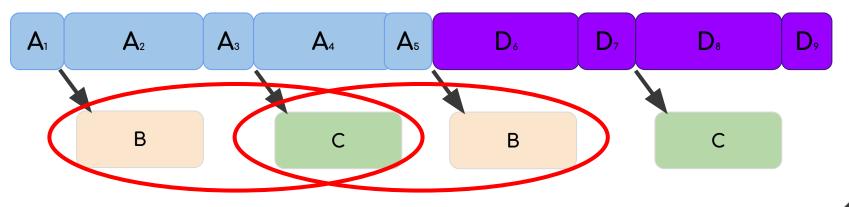
## THE MYSTERY MACHINE: LIMITATIONS

- **Problem:** Assumes enough natural variation
- The Mystery Machine assumes Facebook-specific things
- **Problem:** Assumes that all traces are correct, leaving no room for error such as:
  - Clock skew
  - Anomalies in structure (caused by bugs)





## THE MYSTERY MACHINE: LIMITATIONS



- **Problem:** Mystery Machine assumes that there are no repeats of the same segment or event
- GCM starts with interconnected B and C
  - Sees в -> с; removes с -> в from model
  - Sees c -> b; removes b -> c from model

But what if **B** -> **c** is the true structure?



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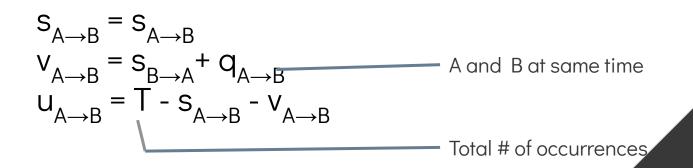
#### 04 FUTURE WORK

## User-Defined Threshold

Address **rigidity** of *The Mystery Machine* Don't eliminate edge after <u>just one</u> violation

 $s_{A \rightarrow B} = \# \text{ of successes of } A \rightarrow B$  $v_{A \rightarrow B} = \# \text{ of violations of } A \rightarrow B$  $u_{A \rightarrow B} = A$ , B do not both occur

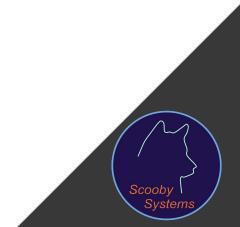
- Chose <u>success-based model</u> contrary to *The Mystery Machine* 
  - Counts successes of each relationship
  - $\circ$  User has freedom to choose threshold in terms of  $s_{A \rightarrow B}$  and  $v_{A \rightarrow B}$

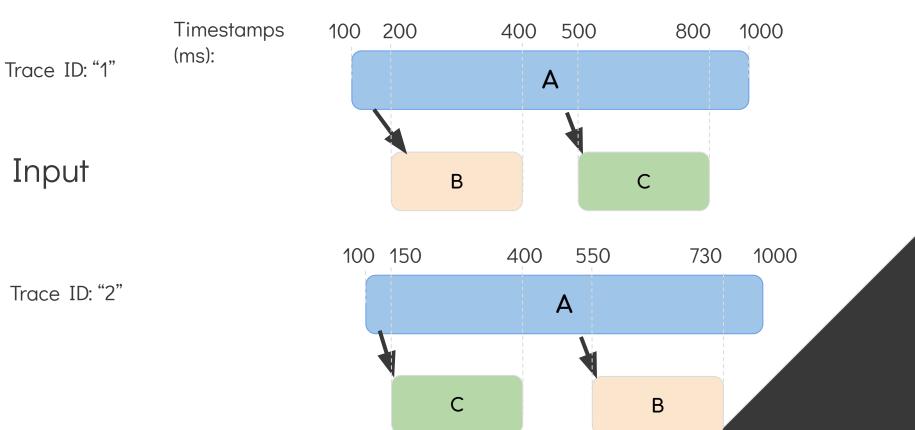


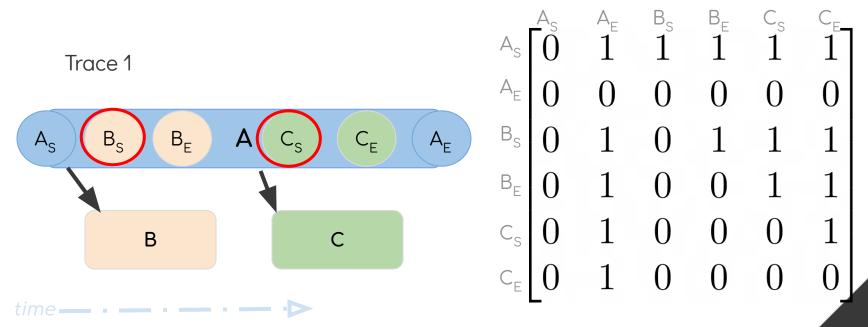
## SCOOBY SYSTEMS

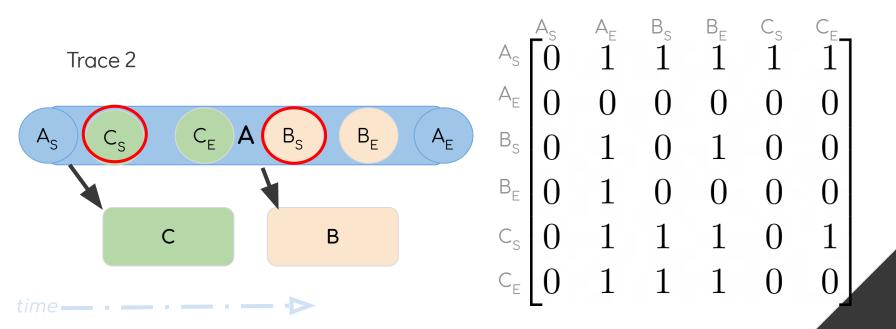
- Preprocessing
  - Create Causal Model
- Main Algorithm
  - for each trace
    - for each pair of spans
      - update causal model
- Apply Threshold
  - $\circ$  For each relationship in causal model
    - if (passes threshold) add edge to final model











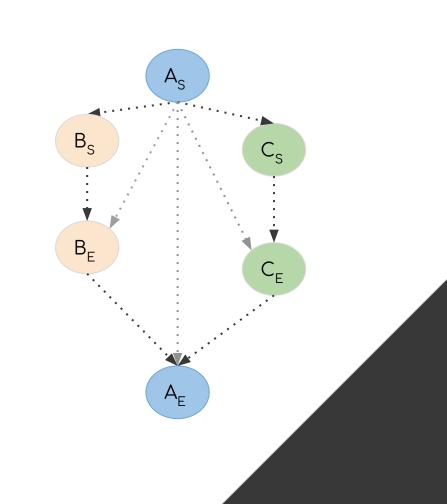
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#### **THRESHOLD** $\left[ 0 \right]$ ()()s/T > 90% () () ( ) $\mathbf{2}$ $\mathbf{2}$ () () () () () $\mathbf{2}$ $\mathbf{2}$ ()()U ()

## GCM VISUALIZED

0 1 1 0 ( ) ( )



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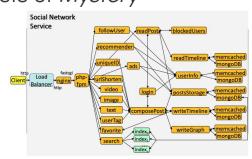
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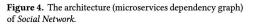
#### 04 FUTURE WORK

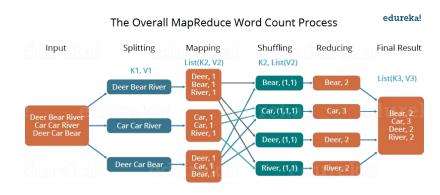
## NEXT STEPS

- Implement Scooby Systems in Hadoop

   Scalable
- Threshold-based determination of edges
- Proposing solutions / further analysis of Mystery Machine limitations
- Evaluation
  - DeathStarBench







# ACKNOWLEDGEMENTS

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Thanks to our **families** 

🗃 slidesgo

Thank **YOU** for listening!

# Any Questions?

