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ON DATA COLLECTION, GRAPH CONSTRUCTION, AND SAMPLING IN TWITTER

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International Symposium on Foundations and Applications of Big Data Analytics (FAB) 2016



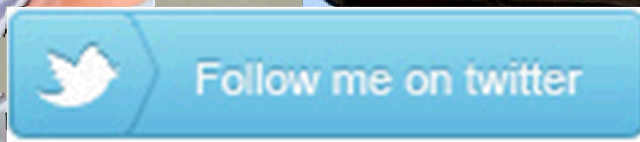
This work was supported by the Laboratory Directed Research and Development program at Sandia National Laboratories, a multi-program laboratory managed and operated by Sandia Corporation, a wholly owned subsidiary of Lockheed Martin Corporation, for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000. SAND NO. 2016-XXXX C.

OVERVIEW

- We present several problems
- We propose some solutions, metrics, and models
 - Fewer than the number of problems

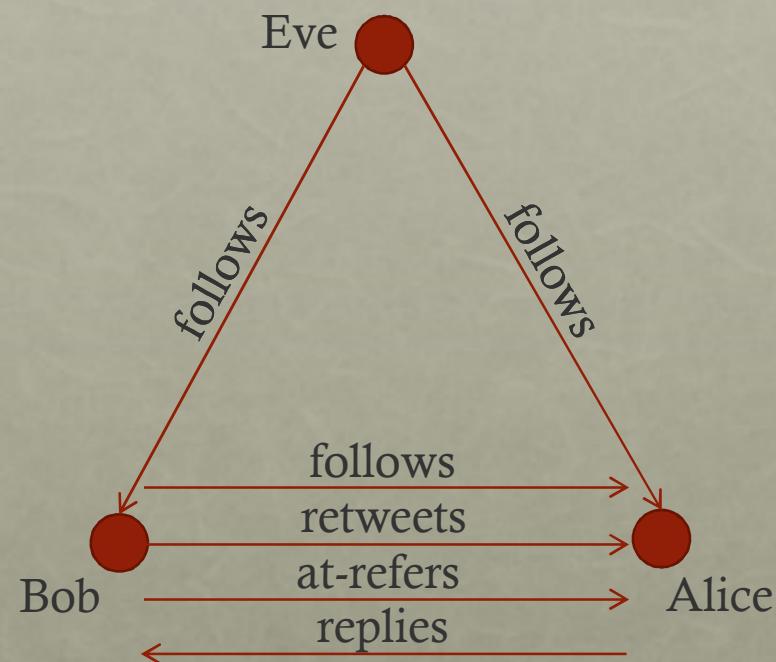


WHY TWITTER GRAPHS?



We need to better understand how information flows through Twitter's network

TWITTER'S GRAPH



- Starting state:
 - Bob follows Alice
 - Eve follows Bob and Alice
- Series of tweets:
 - *Alice*: This is a great article: <http://some.url>
 - *Bob*: (retweeting Alice) This is a great article: <http://some.url>
 - *Bob*: @Alice, that article was great
 - *Alice*: (reply to Bob) Then you'll love this one: <http://other.url>

CAN'T GET ALL OF IT

- Twitter allows anyone free access to their data
 - Severely rate limited
 - Different rates for different query types
- 305M active users*
 - >580 years to get all of those
- Therefore, we must sample
 - How does this affect biasing?

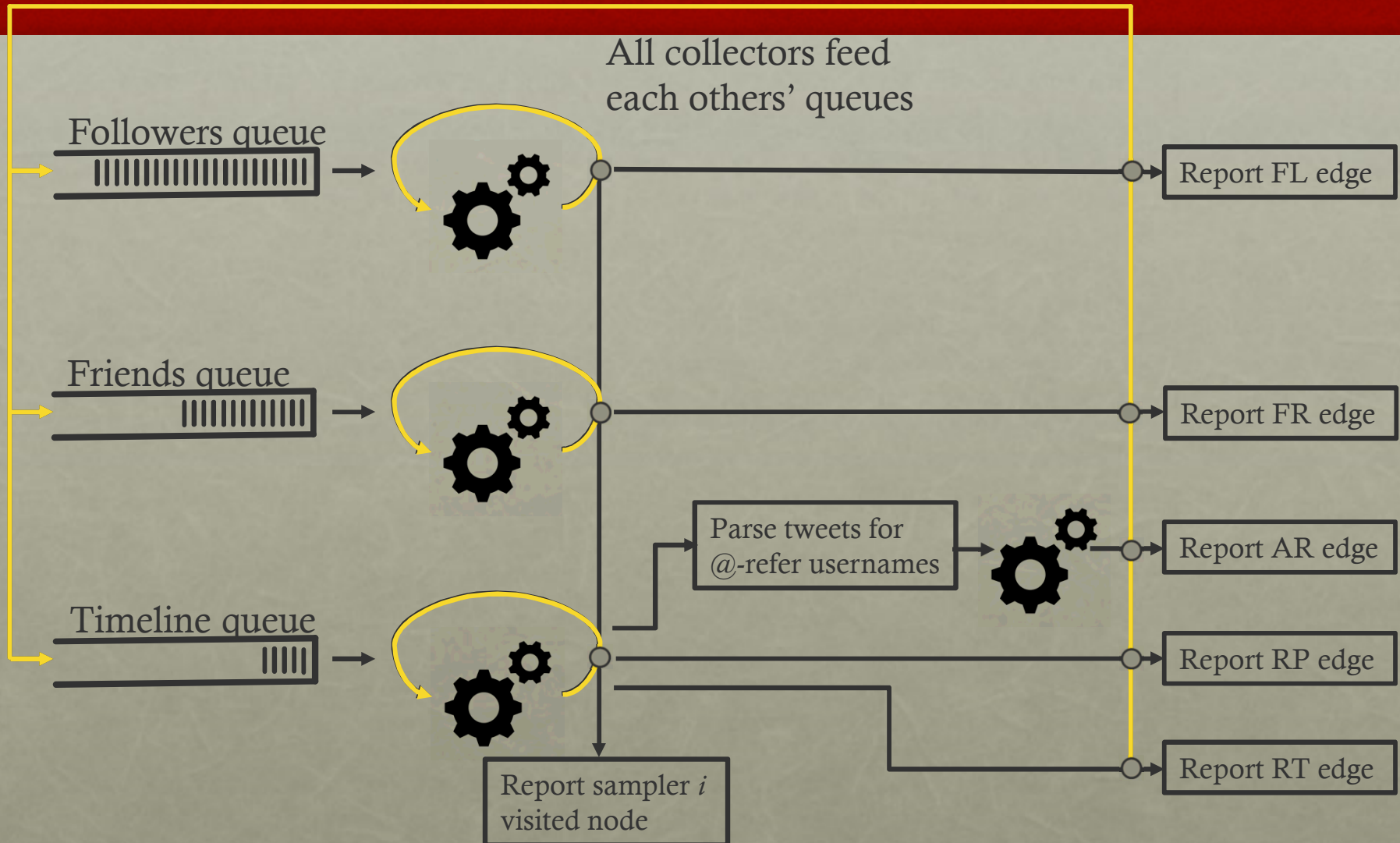
SOME PREVIOUS WORK

- Considerable work on sampling techniques in single-edge-type graphs
 - e.g., (Leskovec, Faloutsos 2006), (Maiya, Berger-Wolf, 2010)
- Sampling introduces bias
 - Random walk finds more high-degree nodes (Lovász, 1993)
 - Bias can be exploited if understood (Maiya, Berger-Wolf, 2011)
- Sampling Twitter
 - Focus on specific edge type (Avrachenkov, et al., 2014)
- Ours appears to be first work focused on sampling in multiple-edge-type networks

GETTING TWITTER

- Twitter provides free access to their data via the Twitter API
 - Different queries for different edge types
 - Rate limits vary for different queries
 - Multiple queries to get full information for a single user
- **Problem 1:** How do you sample different requests at different rates?
 - We propose separate queues for each
- **Problem 2:** How do you keep the queues from sampling different parts of the graph?
 - We propose shared-fed queues

Our Twitter Collector



COLLECTION RESULTS 1

Number of Requests

ID	Duration (days)	Friend	Follower	Timeline
1	7	7,773	7,259	139,540
2	9	8,690	9,002	168,822
3	7	6,511	6,670	118,682

COLLECTION RESULTS 2

Number of Users

ID	Duration (days)	Friend	Follower	Timeline
1	7	4,435	118	13,573
2	9	4,797	878	11,319
3	7	3,780	166	10,050
Ave	Req. per	1.8	37.3	12.3

COLLECTION RESULTS 3

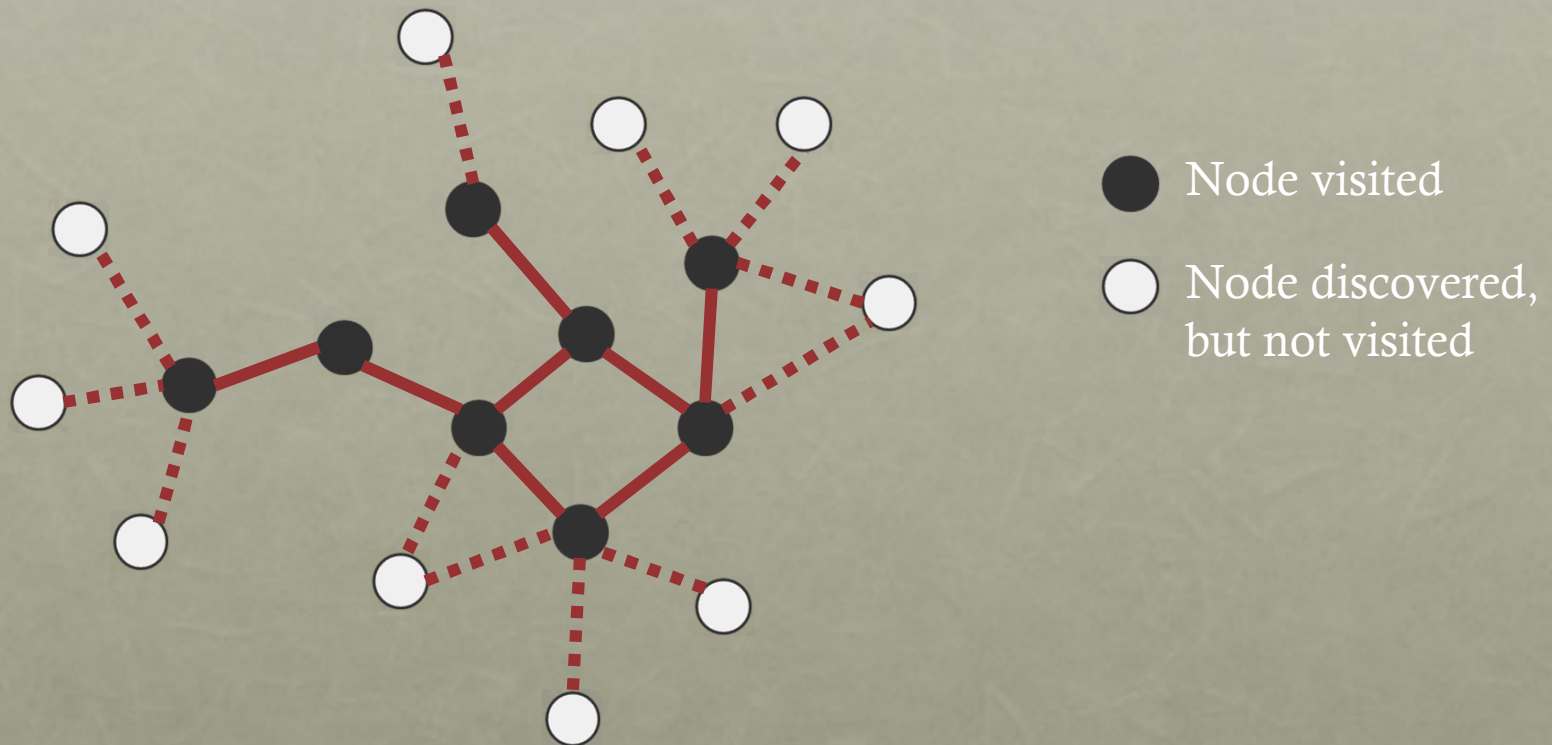
Users with Zero Results

ID	Duration (days)	Friend	Follower	Timeline
1	7	37%	11%	6%
2	9	24%	33%	5%
3	7	44%	22%	20%
Ave	0-queries (hours)	35.6	2.7	1.5

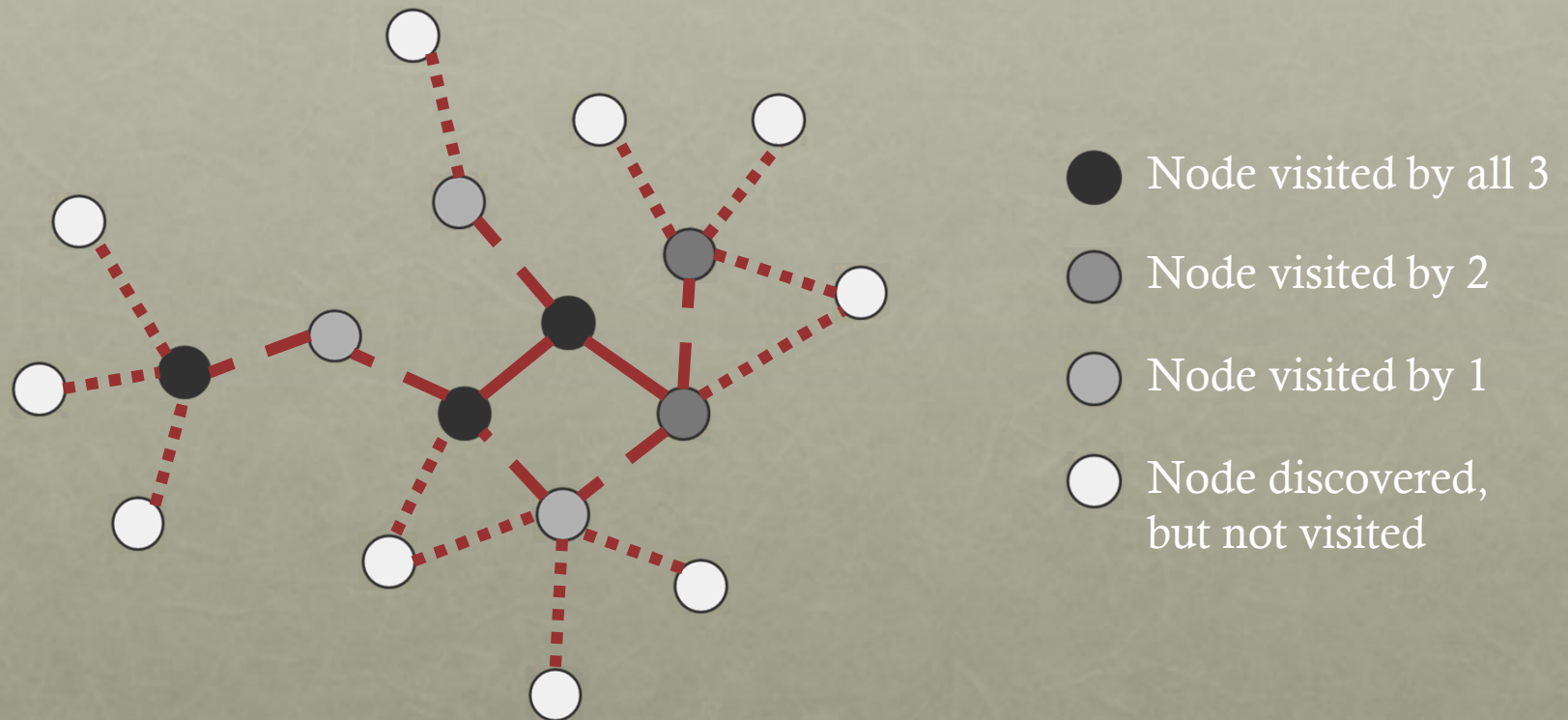
COLLECTION PROBLEMS

- **Problem 3:** Can we avoid more of those zero-hits queries?
 - There may be indications between collectors' results that push away from zero-hits
- **Problem 4:** How would avoiding the highest degree follower nodes affect biasing?

TRADITIONAL SAMPLING



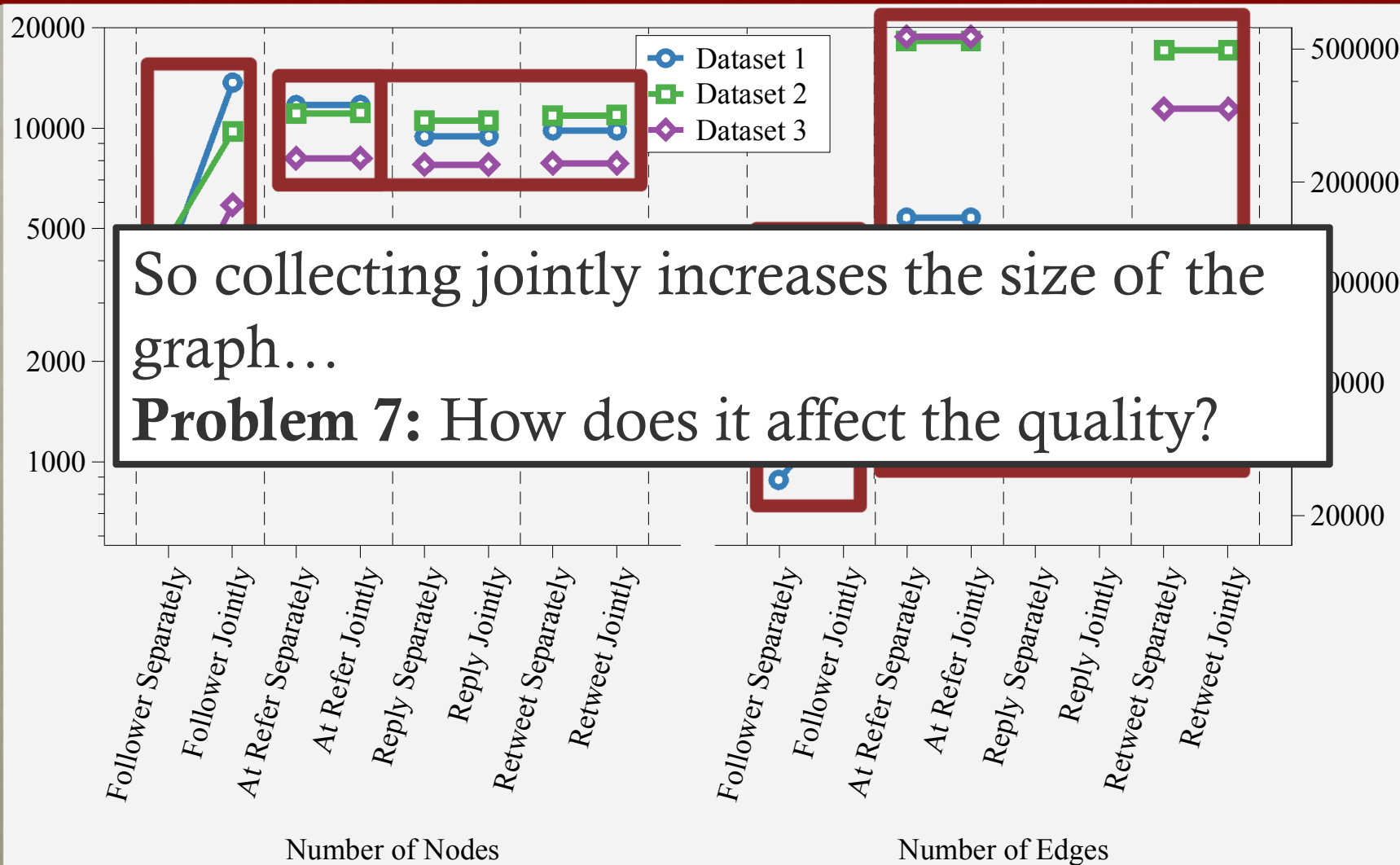
SEMANTIC SAMPLING



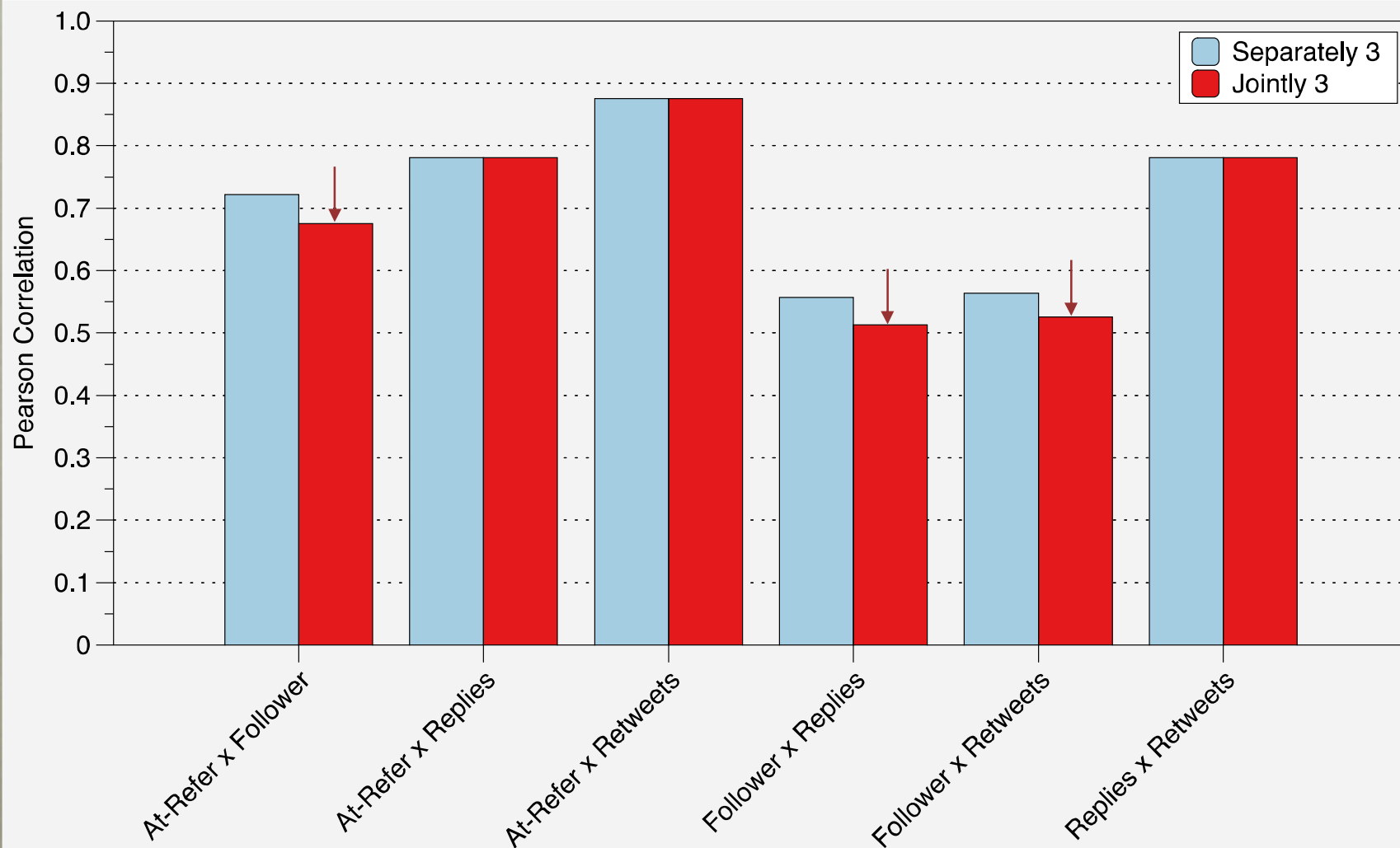
FORMING A GRAPH

- **Problem 5:** How do we define “visited” with multiple collectors?
 - We propose visited by any
- **Problem 6:** Which edges are allowed in the graph?
 - We propose two options
- *Collecting separately* requires the edge-type sampler to visit both ends for an edge of that type to be included
- *Collecting jointly* request some edge-type sampler to visit both ends for an edge of any type to be included

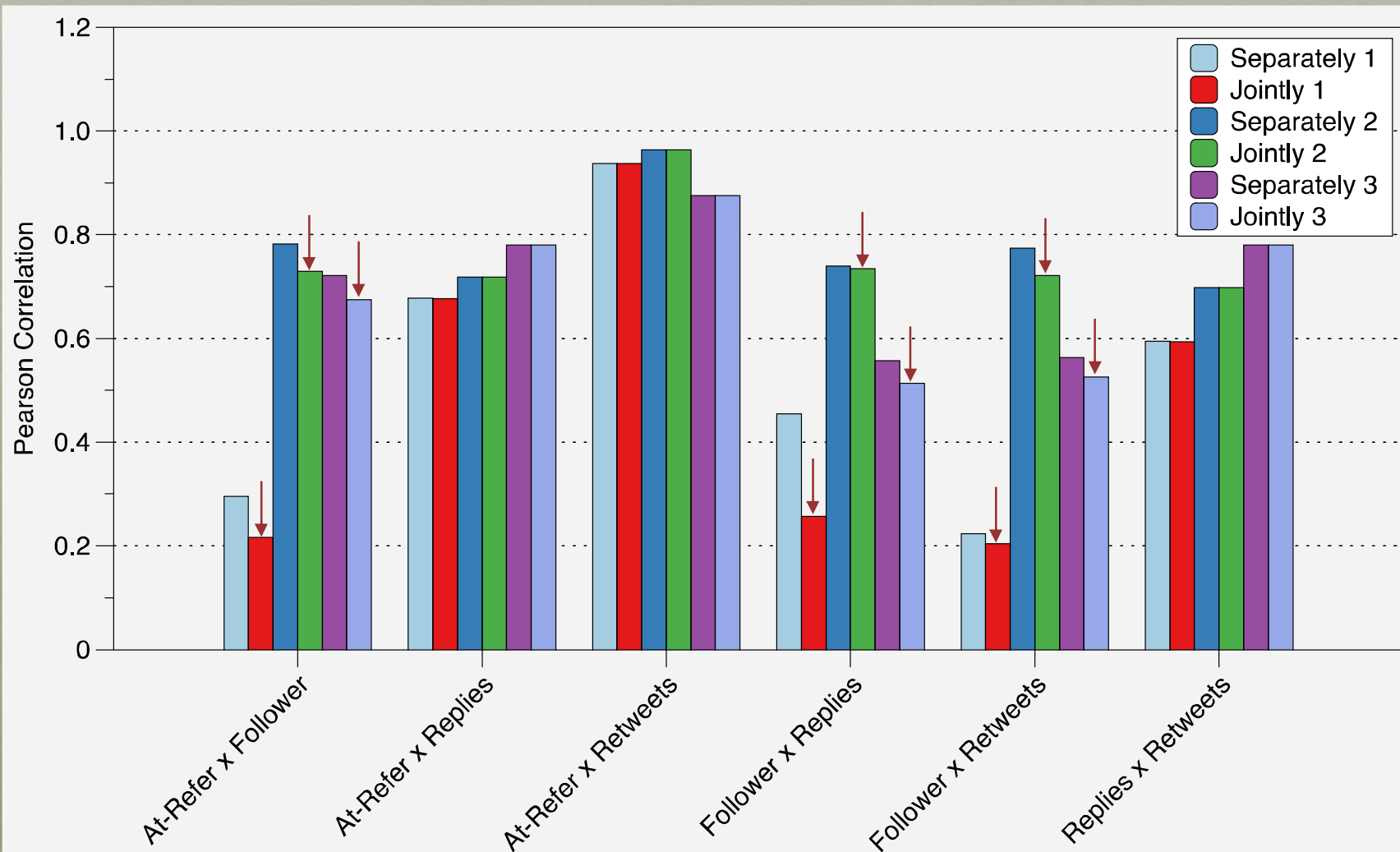
SEPARATELY VS. JOINTLY



GRAPH METRICS



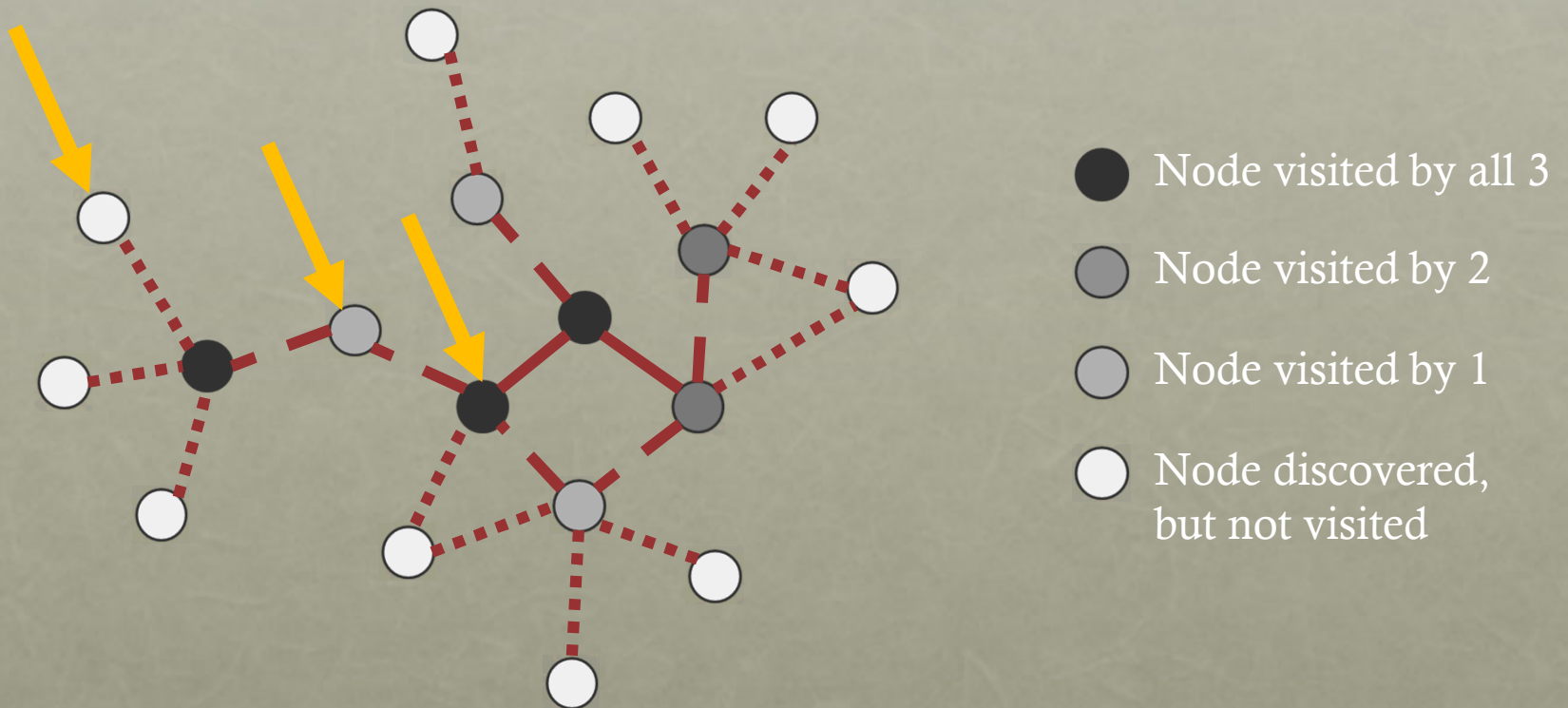
GRAPH METRICS



SUBSAMPLING

- Traditional graph sampling samples a graph using different techniques and analyzes how different techniques affect metrics
- We attempted this with our semantic graph...

SUBSAMPLING



SAMPLING ANALYSIS

Friends

ID	Successes (1 hour)	Failures (1 hour)
1	60	816
2	60	548
3	60	723

Followers

ID	Successes (1 hour)	Failures (1 hour)
1	60	4,589
2	60	28,189
3	60	644

Timeline

ID	Successes (1 hour)	Failures (1 hour)
1	1,200	1,052
2	1,200	1,141
3	1,200	625

- **Problem 8:** We hit the edge of our collect more than getting good results. Why? How can we avoid it?

WHY FAILURES?

- Asynchronous queries mean you get responses from one collector before another
 - The to-visit queue is thus in a different order
- Our graphs are relatively small ... even though collected over many days
- We propose a model on why we are so close to the collection “edge” even at the seed – curse of dimensionality
 - Different collectors add more dimensions to the data
 - Increased dimensionality → decreased density
 - Sparsest collector (followers) makes a narrow dimension

SEMANTIC GRAPHS SPECIFIC?

- Traditional graph sampling has not reported this “edge failure” phenomenon
 - However, they subsampled against the graph itself
 - Hitting a leaf can mean degree 1 in original data or a collection edge
 - We sampled against original collected data
 - Thus, we could differentiate between true leaves and collection-caused leaves
- **Problem 9:** How much does this occur in traditional graphs?

OPEN PROBLEMS

- How do you sample different requests at different rates?
- How do you keep the queues from sampling different parts of the graph?
- Can we avoid more of those zero-hits queries?
- Can we avoid the highest degree follower nodes with minimal biasing?
- How do we define “visited” with multiple collectors?
- Which edges are allowed in the graph?
- How does collecting jointly vs. collecting separately affect graph quality?
- We hit the edge of our collect more than getting good results. Why? How can we avoid it?
- How much does edge-hitting exist in traditional graphs?

WE NEED ANSWERS!



THANKS

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