

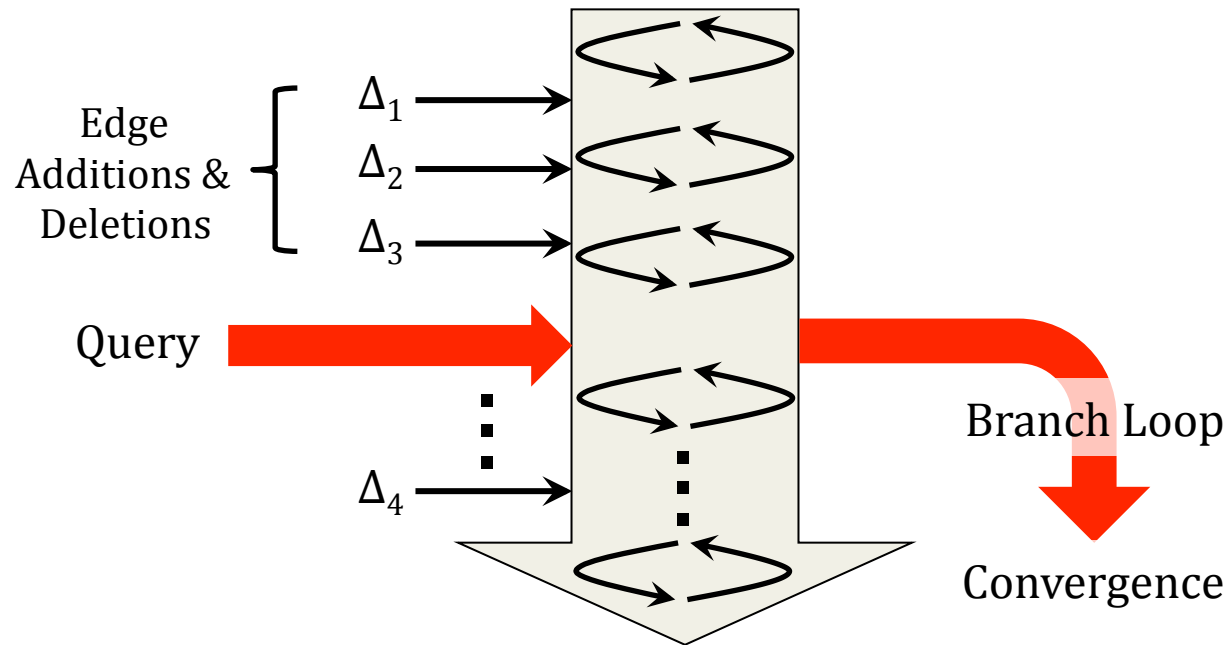
KickStarter: Fast and Accurate Computations on Streaming Graphs via Trimmed Approximations

Keval Vora, Rajiv Gupta and Guoqing Xu



Streaming Graph Processing

- Graph changes rapidly as computation proceeds
- Incremental processing
 - Maintain “profitable” approximation



The Good, the Bad and the Ugly

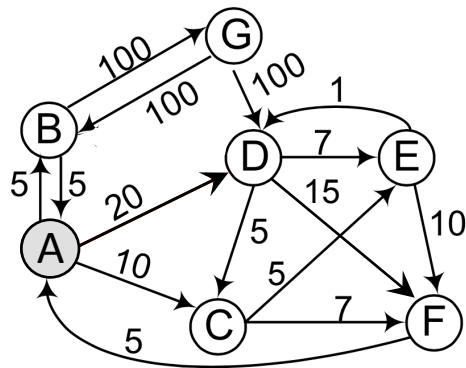
- Correctness & performance



The Good Scenario



SSWP $v.path \leftarrow \max_{e \in \text{inEdges}(v)} (\min(e.source.path, e.weight))$

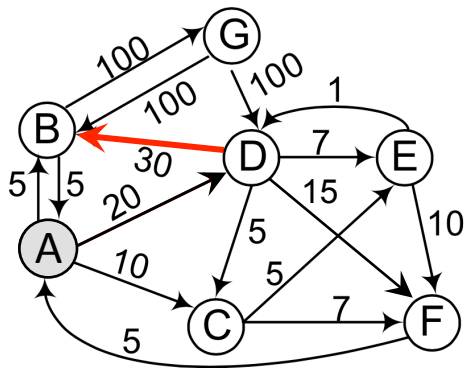


A	B	C	D	E	F	G
∞	5	10	20	7	15	5

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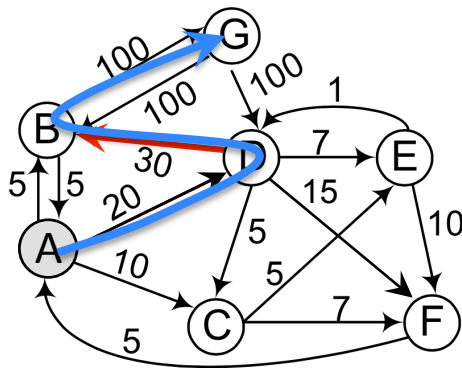


A	B	C	D	E	F	G
∞	5	10	20	7	15	5
Add D \rightarrow B						

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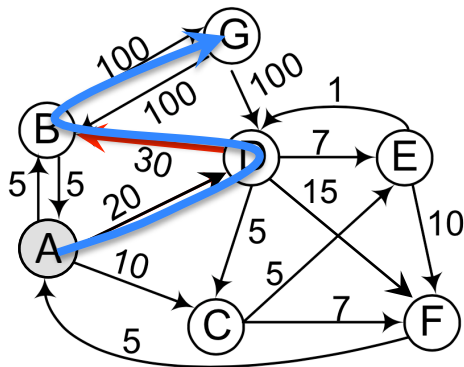


A	B	C	D	E	F	G
∞	5	10	20	7	15	5
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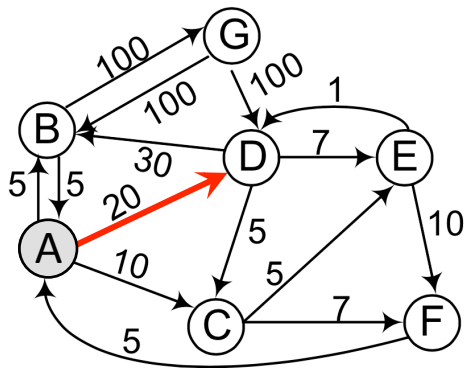


A	B	C	D	E	F	G
∞	5	10	20	7	15	5
Add D \rightarrow B						
∞	20	10	20	7	15	5
∞	20	10	20	7	15	20

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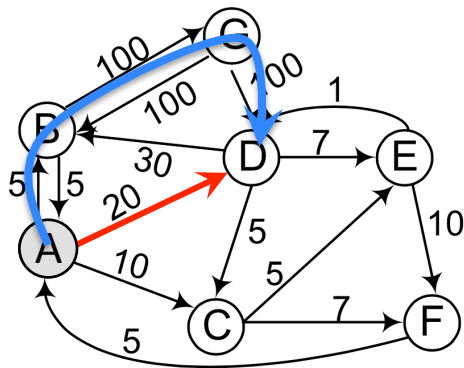


A	B	C	D	E	F	G
∞	20	10	20	7	7	20
Delete A \rightarrow D						

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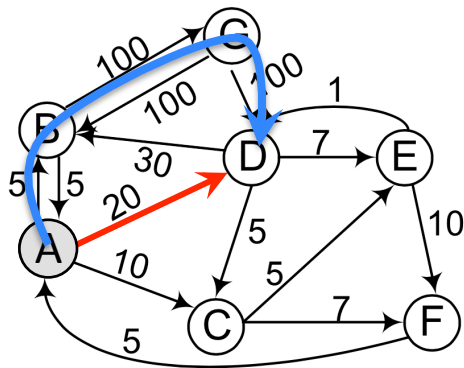


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∞	20	10	20	7	7	20
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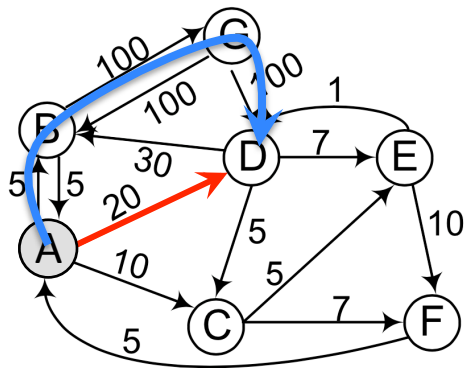


A	B	C	D	E	F	G
∞	20	10	20	7	7	20
Delete A \rightarrow D						
...						
∞			20			

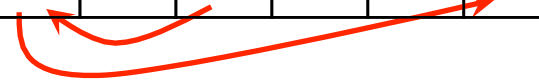
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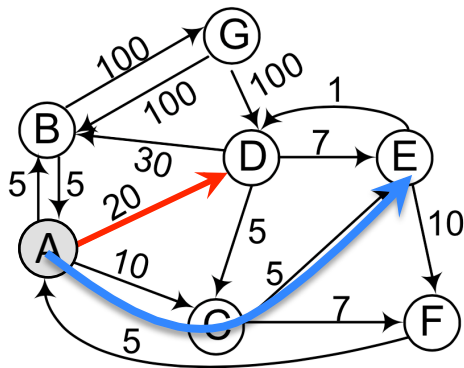
A	B	C	D	E	F	G
∞	20	10	20	7	7	20
Delete A \rightarrow D						
...						
∞	20		20			20



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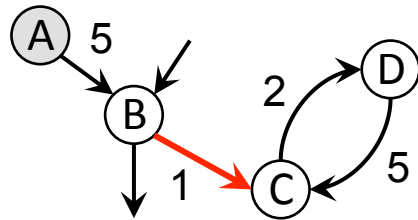
A	B	C	D	E	F	G
∞	20	10	20	7	7	20
Delete A \rightarrow D						
...						
∞	20	10	20	7	7	20



The Ugly Scenario



SSSP $v.path \leftarrow \min_{e \in \text{inEdges}(v)} (e.source.path + e.weight)$

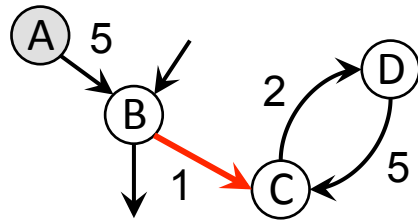


A	B	C	D
0	5	6	8
Delete B → C			

The Ugly Scenario



SSSP $v.path \leftarrow \min_{e \in \text{inEdges}(v)} (e.source.path + e.weight)$



A	B	C	D
0	5	6	8
Delete B → C			
0	5	6	8
0	5	13	8
0	5	13	15
0	5	20	15
0
0	5	MAX	MAX

The Good, the Bad and the Ugly

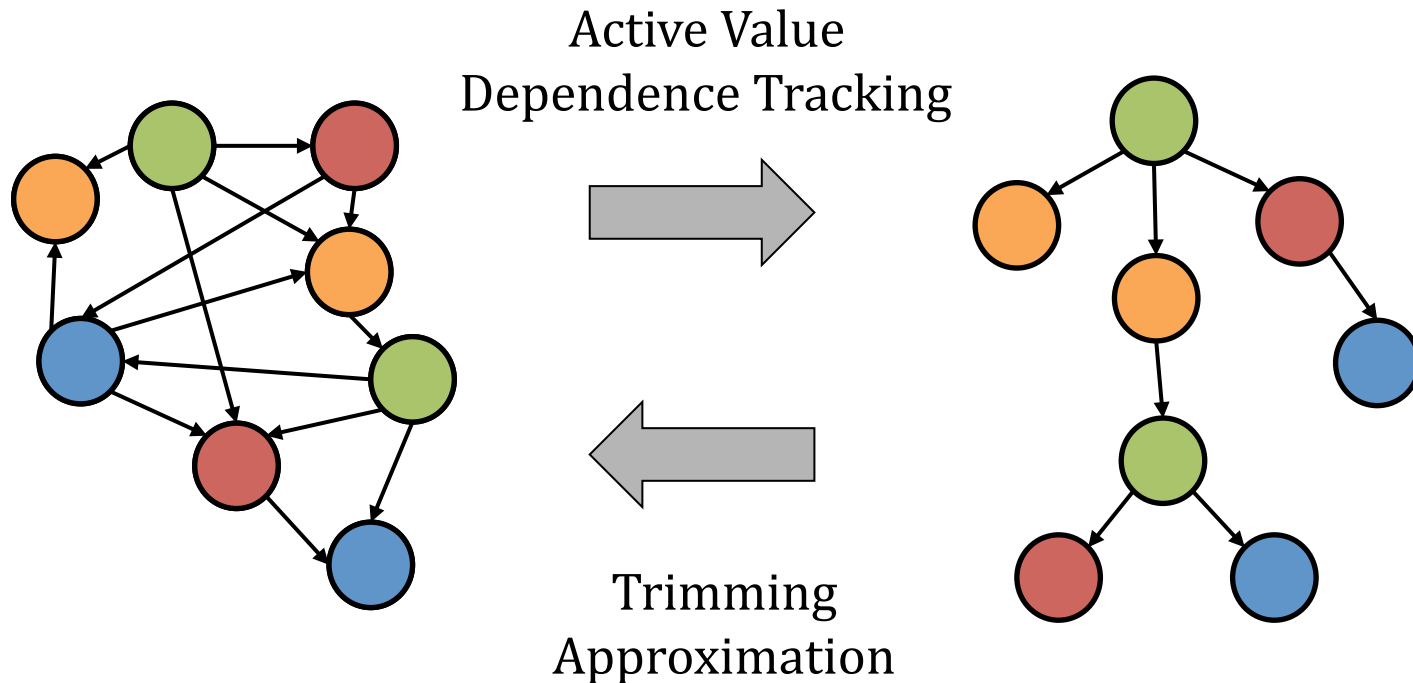
- Correctness & performance

Edge Deletions



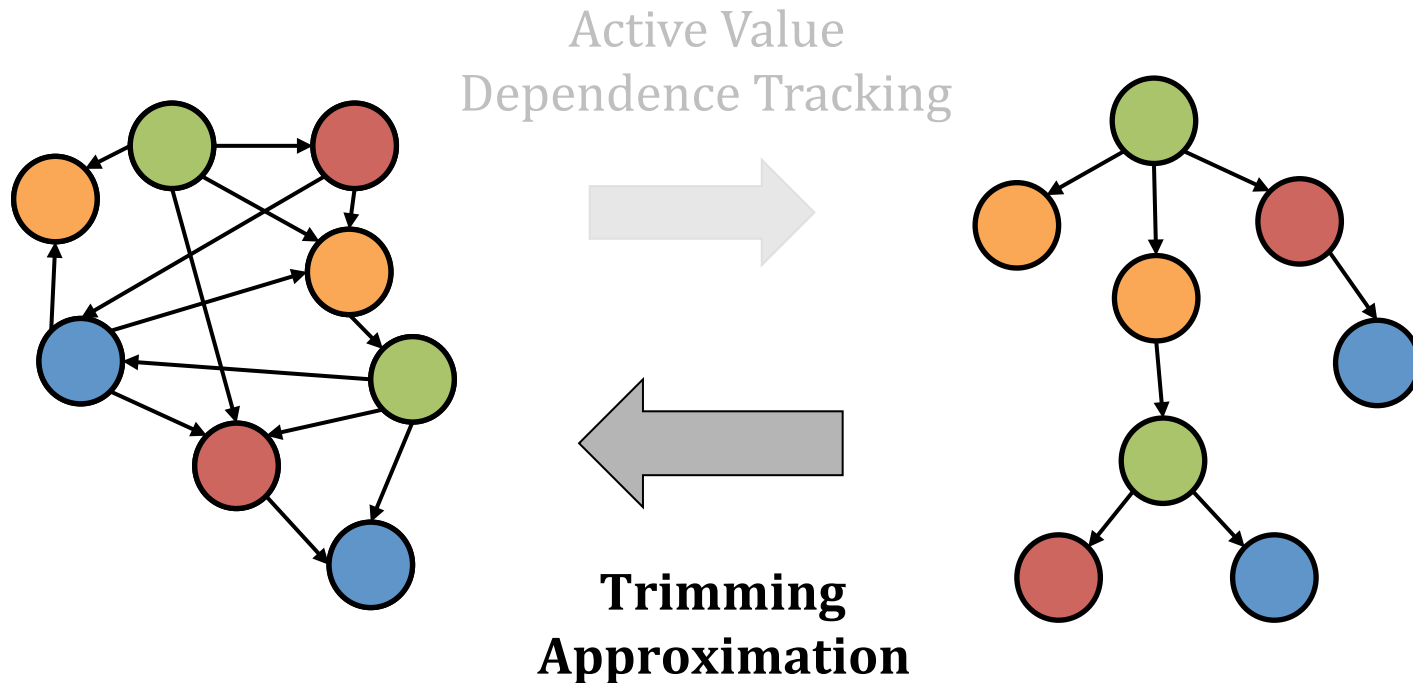
KickStarter

- Maintain value dependences during computation
 - $a \xrightarrow{LT} b$ iff b 's value resulted from a



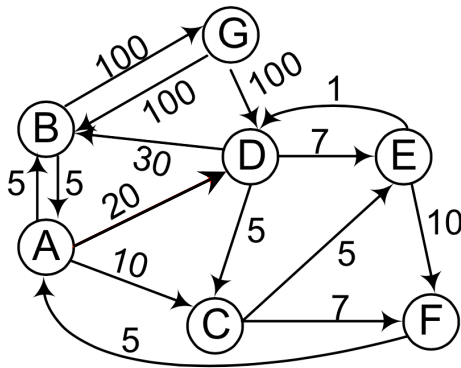
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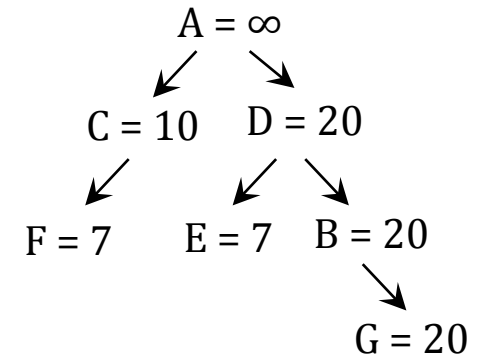


Trimming via Value Dependence

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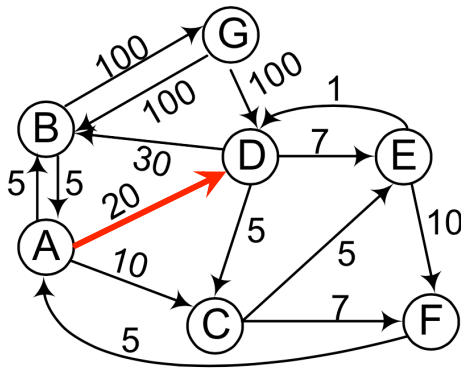
A	B	C	D	E	F	G
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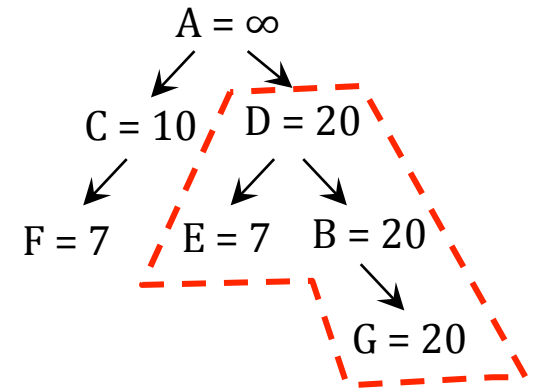
SSWP $v.path \leftarrow \max_{e \in \text{inEdges}(v)} (\min(e.source.path, e.weight))$

Trimming via Value Dependence

- Compute safe approximations
 - Can be done using the same vertex function



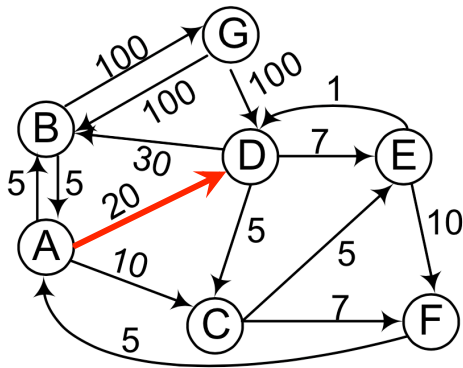
A	B	C	D	E	F	G
∞	20	10	20	7	7	20
Delete A \rightarrow D						



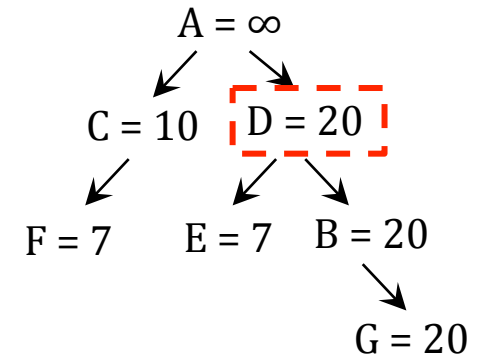
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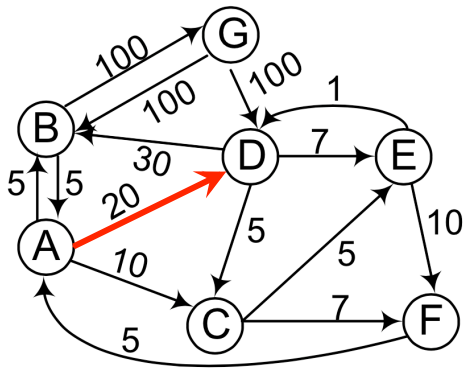
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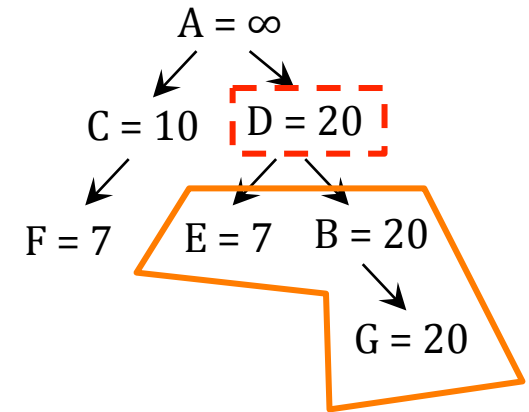
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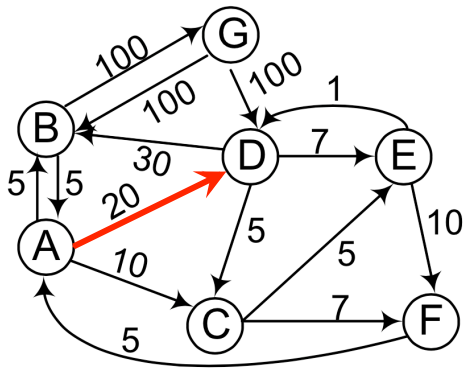
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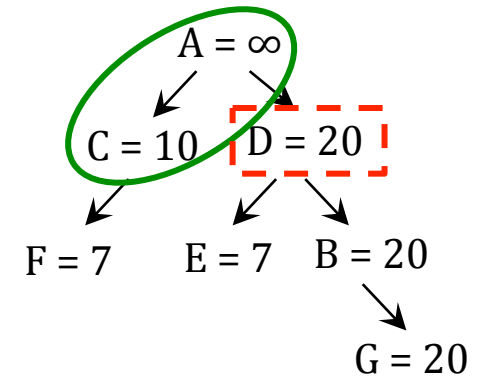
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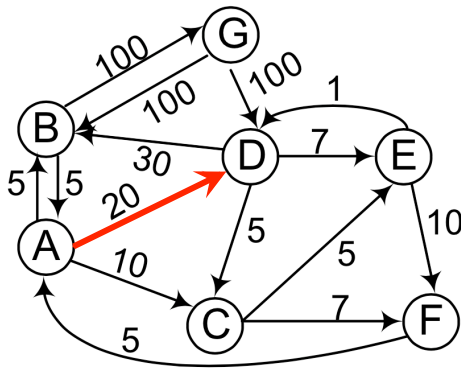
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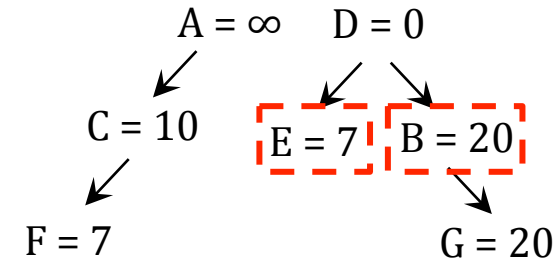
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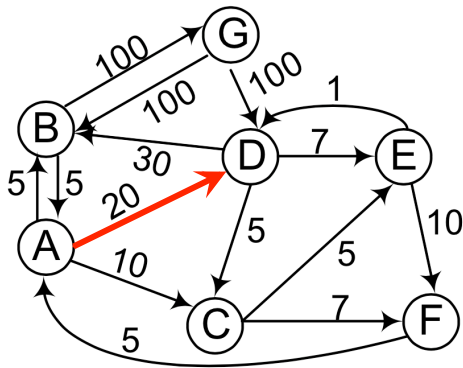
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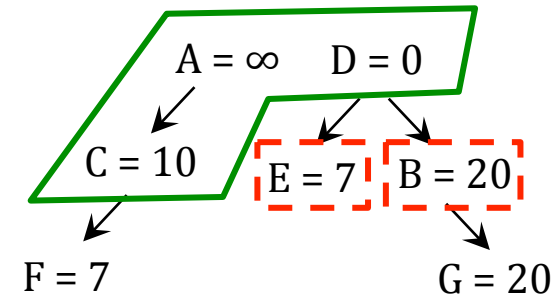
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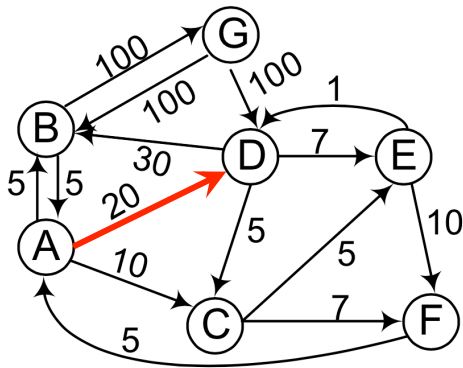
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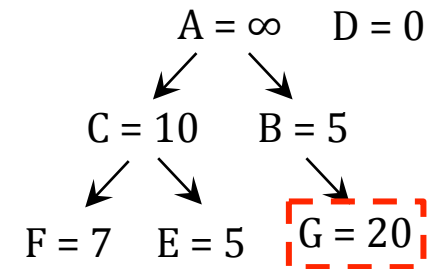
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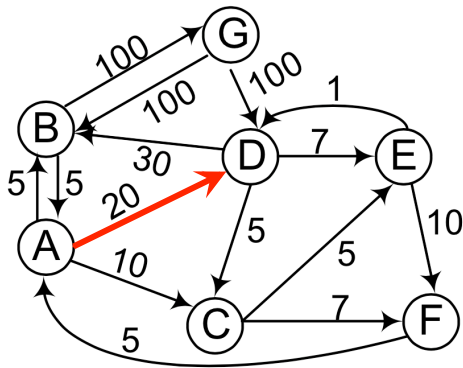
A	B	C	D	E	F	G
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Delete A \rightarrow D						
∞	20	10	0	7	7	20
∞	5	10	0	5	7	20



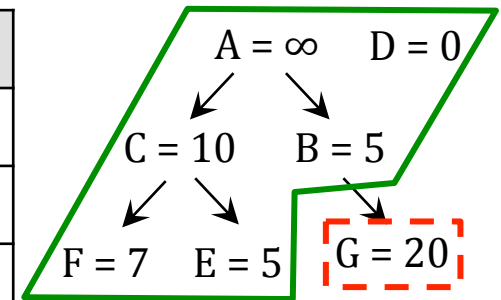
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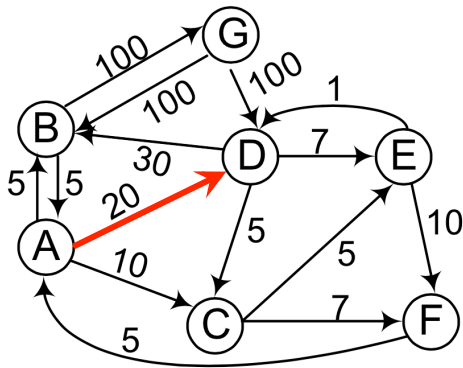
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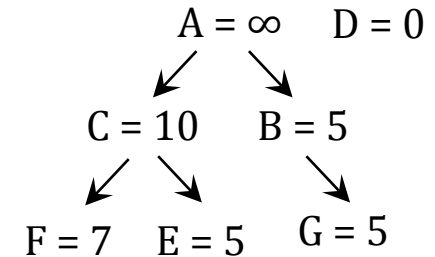
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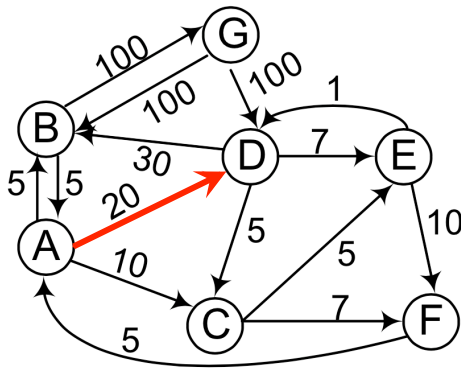


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Delete A \rightarrow D						
∞	20	10	0	7	7	20
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Trimming Complete						

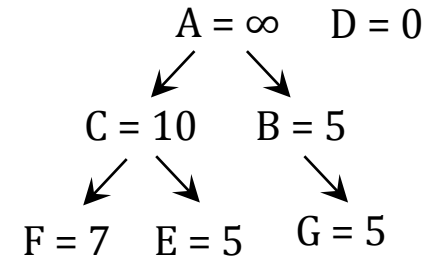


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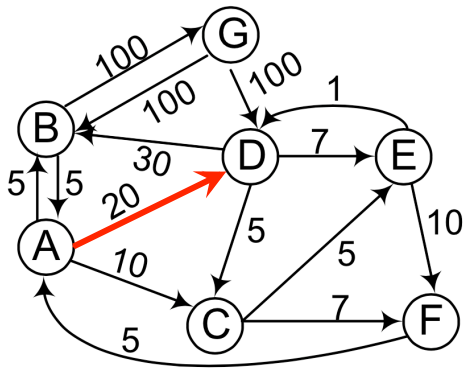


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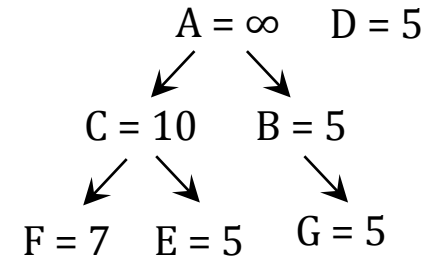


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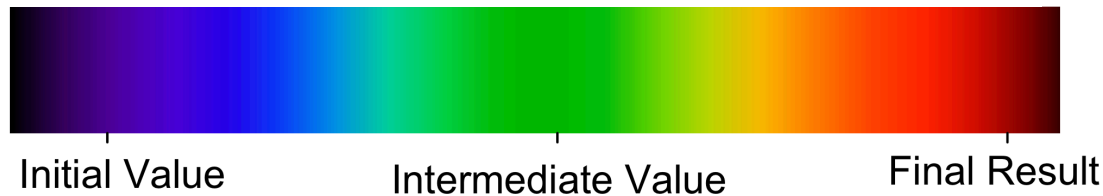


A	B	C	D	E	F	G
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∞	20	10	0	7	7	20
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Trimming Complete						
∞	5	10	5	5	7	5



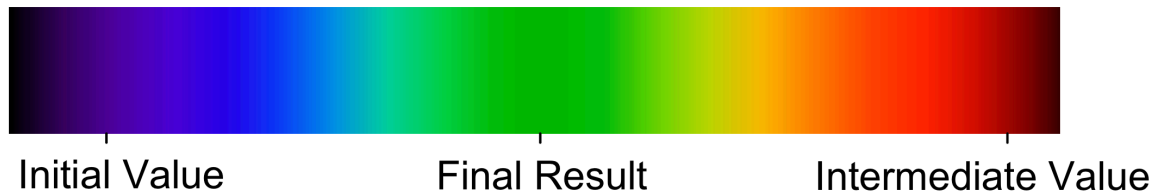
Safety

$$\mathbf{SSWP} \quad v.path \leftarrow \max_{e \in \text{inEdges}(v)} (\min(e.source.path, e.weight))$$



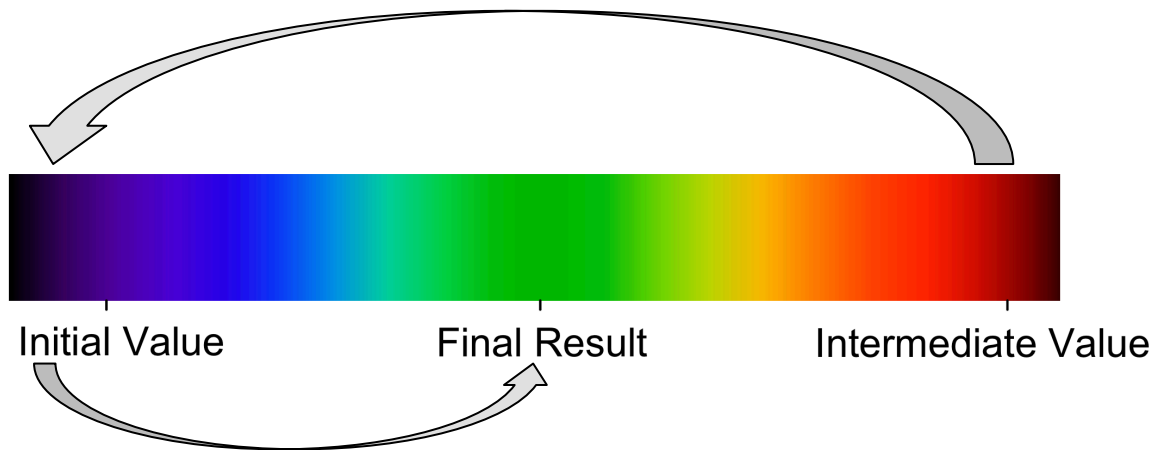
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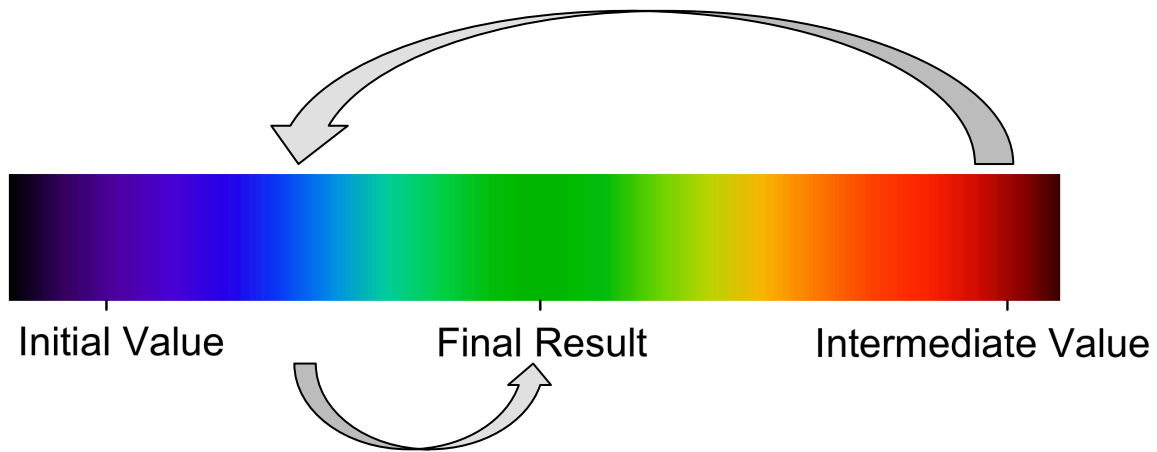
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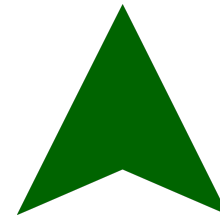
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Monotonic Graph Algorithms

- Vertex values exhibit increasing/decreasing values



- WidestPaths (SSWP)

Monotonic Graph Algorithms

- Vertex values exhibit increasing/decreasing values



- ShortestPaths (SSSP)



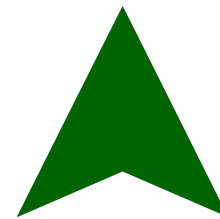
- WidestPaths (SSWP)

Monotonic Graph Algorithms

- Vertex values exhibit increasing/decreasing values



- ShortestPaths (SSSP)
- ConnectedComponents
- MinimumSpanningTree
 - BreadFirstSearch
 - FacilityLocation



- WidestPaths (SSWP)
- Reachability

Experimental Setup

- 16-node EC2 cluster: 8-core/16GB nodes
- Monotonic algorithms

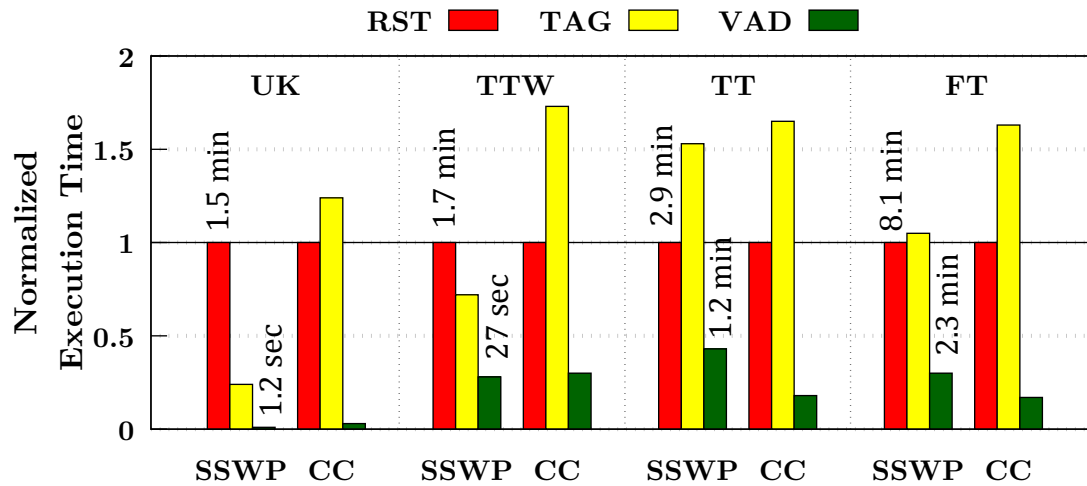
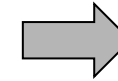
Bad Scenario	SingleSourceWidestPaths (SSWP) ConnectedComponents (CC)
Ugly Scenario	SingleSourceShortestPaths (SSSP) BreadthFirstSearch (BFS)

Experimental Setup

- Streaming graph datasets constructed using [SIGMOD'16]
 - Fixed point achieved at 50% edges
 - Remaining edges treated as edge additions
 - Edge deletions sampled from loaded graph
- Rate of update stream (100K-1M updates per query)
- Edge deletion ratio (10-50%)

Graphs	#Edges	#Vertices
Friendster (FT)	2.5B	68.3M
Twitter (TT)	2.0B	52.6M
Twitter (TTW)	1.5B	41.7M
UKDomain (UK)	1.0B	39.5M

Trimming for Safety

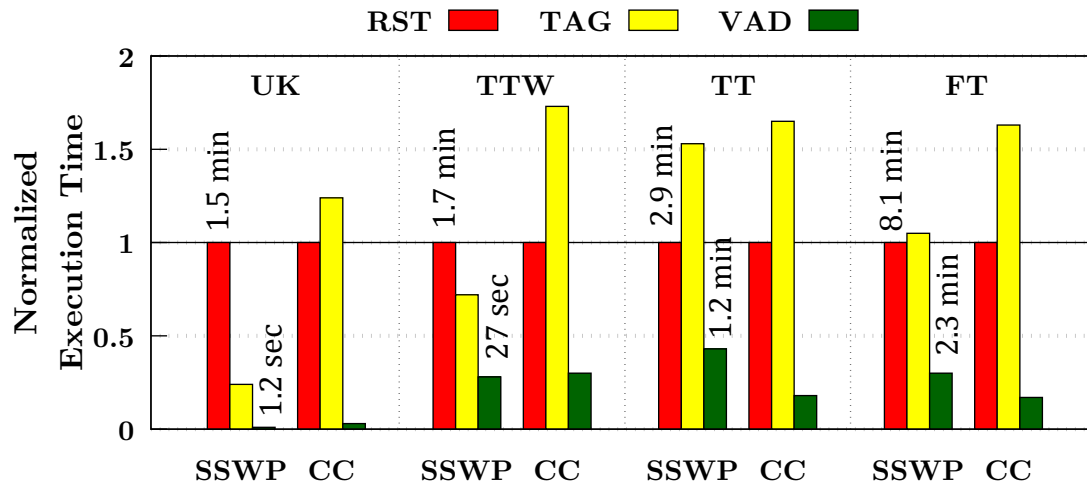
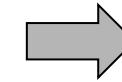


	SSWP	CC
RST/ VAD	17.7x	10x
TAG/ VAD	6.2x	13.7x

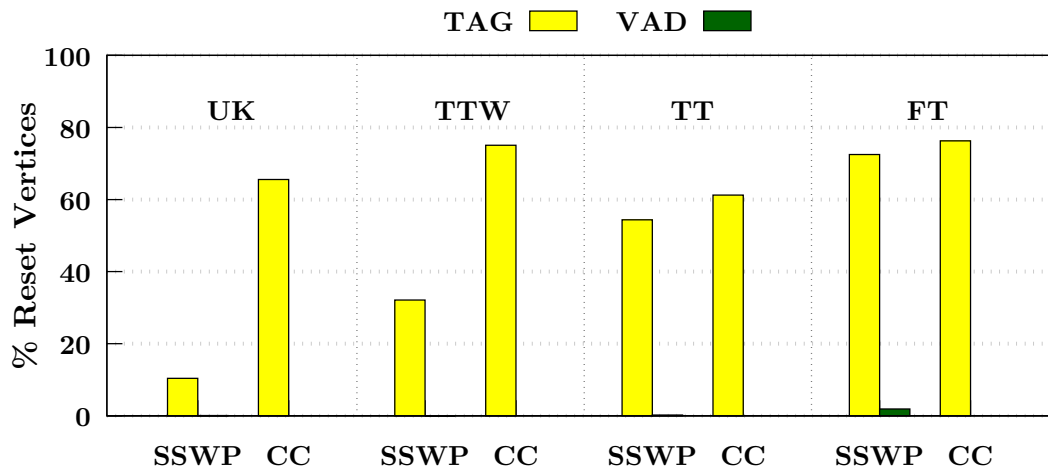
RST: Reset all vertex values

- 100K updates per query
- 30% deletion rate

Trimming for Safety



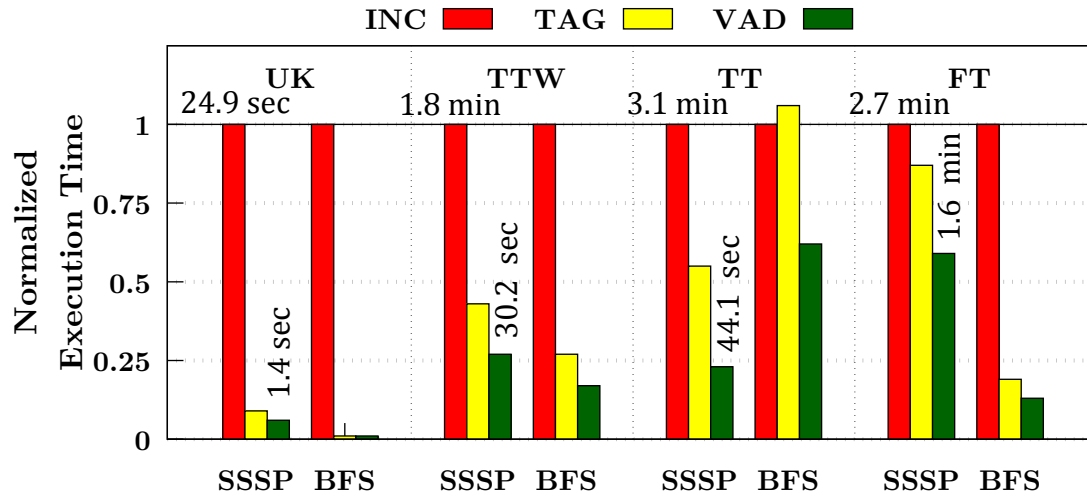
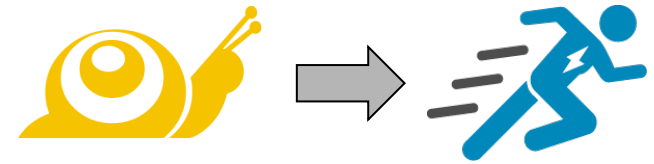
	SSWP	CC
RST/ VAD	17.7x	10x
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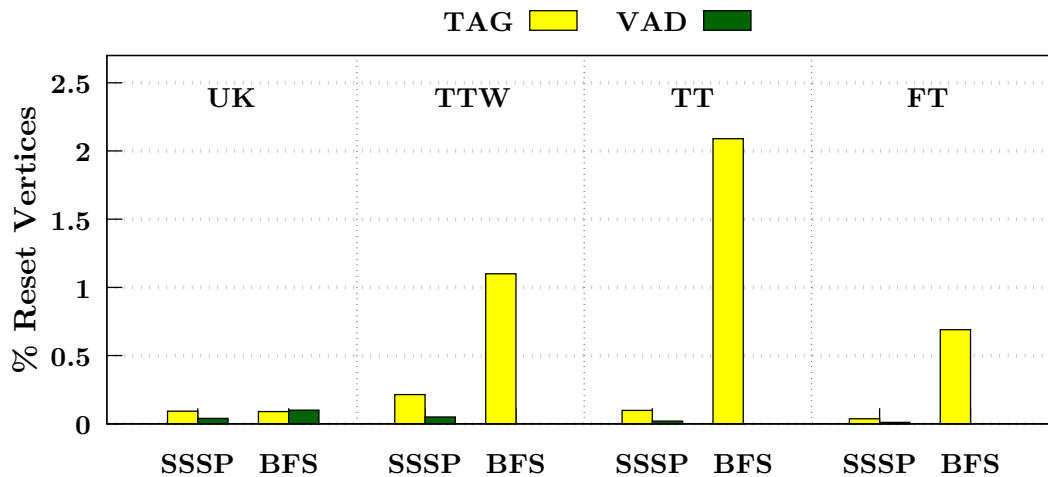
RST: Reset all vertex values

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Trimming for Performance



	SSSP	BFS
INC/ VAD	23.7x	8.5x
TAG/ VAD	1.5x	1.7x



INC: Incremental processing
(no resets)

- 100K updates per query
- 30% deletion rate

More Results

- Individual query performance
 - Trimming v/s computation time
- Effectiveness of trimming over resetting
- Varying update rate
- Varying edge deletion ratio
- Dependence tracking overhead

Summary

- Incremental processing in presence of edge deletions
- Trimming approximation
 - Reuse safe and profitable values
- Active dependence tracking based trimming
 - Up to 8.5-23.7x speedups



Thanks

