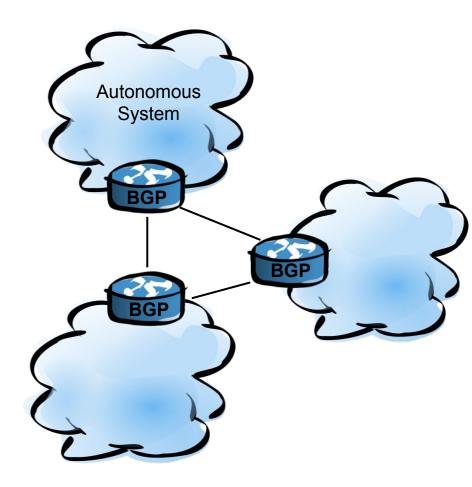
Automatic Inference of BGP Location Communities

Brivaldo A. Silva JrPaulo MolOsvaldo FonsecaÍtalo CunhaRonaldo A. FerreiraEthan Katz-Bassett

SIGMETRICS 2022 Mumbai, India

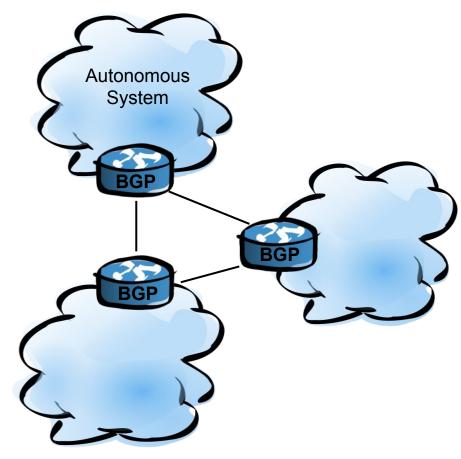


BGP: the routing protocol used to exchange Internet routes and reachability information between autonomous systems



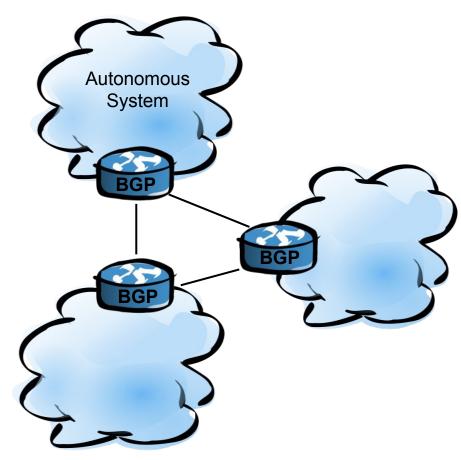


- BGP: the routing protocol used to exchange Internet routes and reachability information between autonomous systems
- BGP is an old protocol used to sustain the current need of stability, reliability, and complex policies.



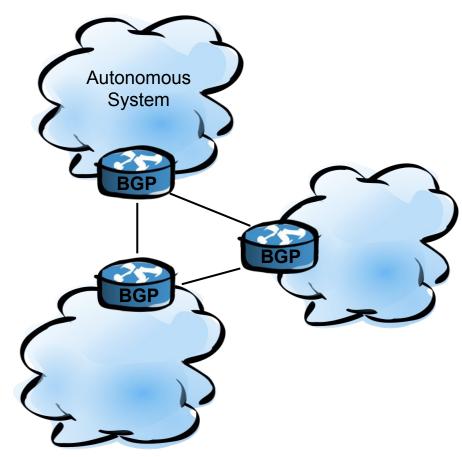


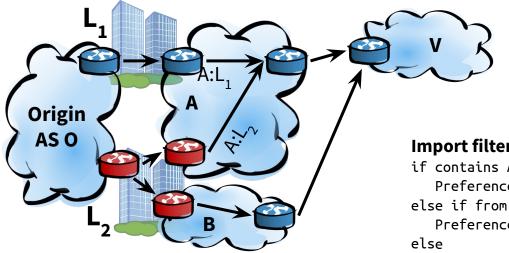
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- BGP: the routing protocol used to exchange Internet routes and reachability information between autonomous systems
- BGP is an old protocol used to sustain the current need of stability, reliability, and complex policies.
- Operators explore all available options to improve the "user experience".
- Network operators have increasingly relied on the communities attribute to instrument BGP.

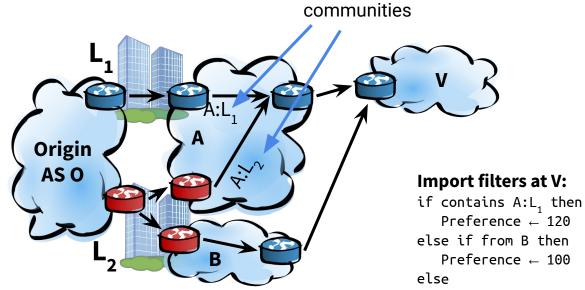




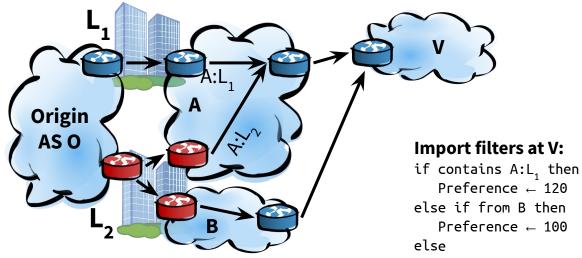
Import filters at V:

if contains $A:L_1$ then Preference $\leftarrow 120$ else if from B then Preference \leftarrow 100 Preference \leftarrow 80



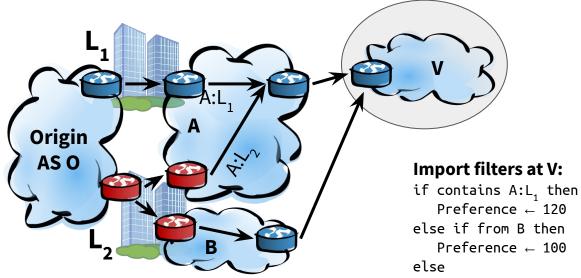






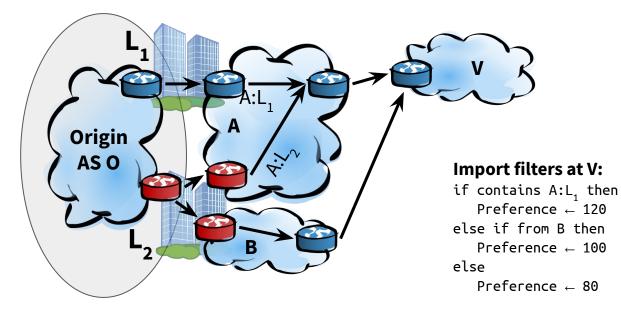
Preference \leftarrow 80



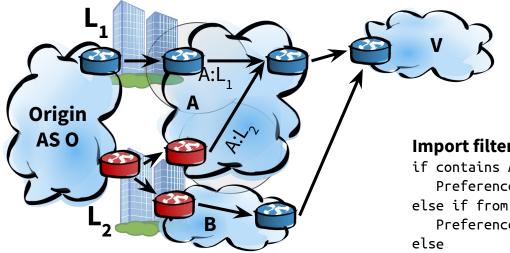


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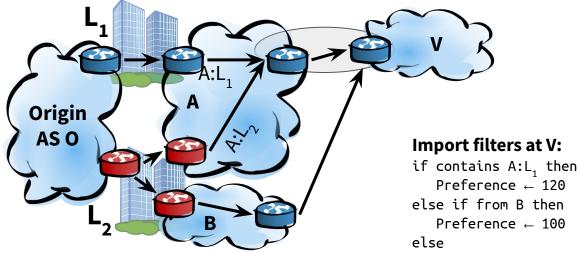




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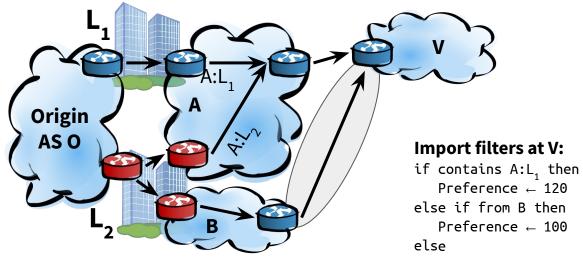
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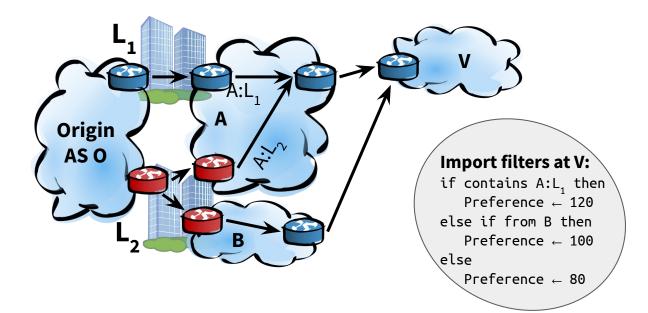
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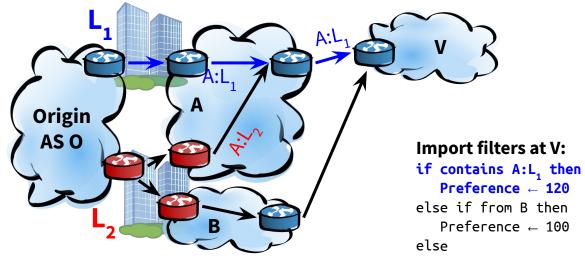


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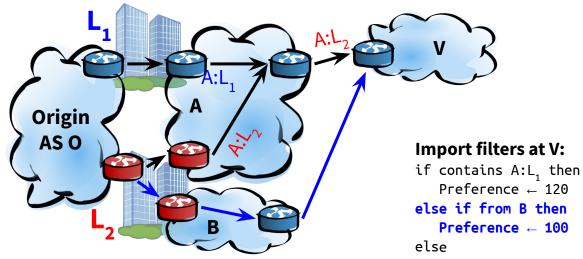




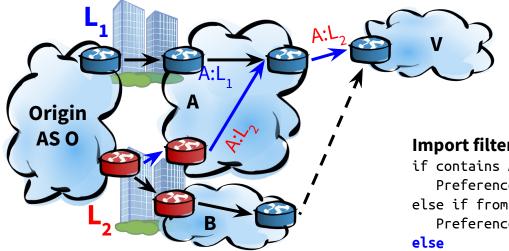








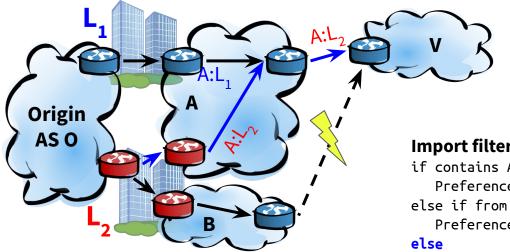




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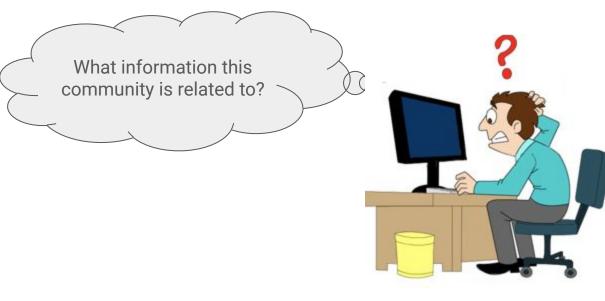


• Community semantics are not standardized



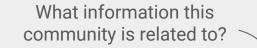


- Community semantics are not standardized
- □ Hard to know the semantics of a specific community





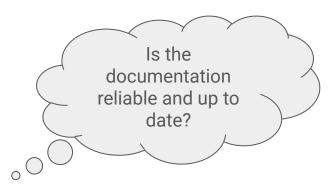
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- Hard to know the semantics of a specific community
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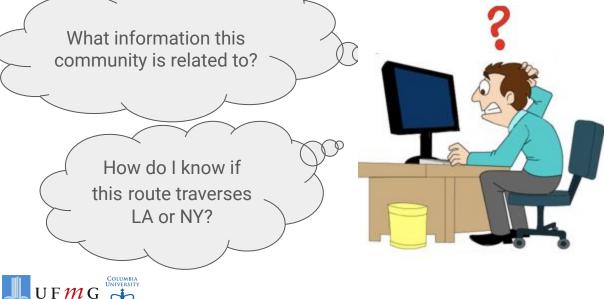






- Community semantics are not standardized
- Hard to know the semantics of a specific community
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What information this community is related to?

How do I know if

this route traverses

LA or NY?

COLUMBIA UNIVERSITY

F*M*(

- Community semantics are not standardized
- Hard to know the semantics of a specific community
- Hard to know the community for some specific goal



What we have so far?

- □ Current practical applications of BGP communities:
 - Help identify outages (Giotsas, et al, 2017)





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What we have so far?

• Current practical applications of BGP communities:

- Help identify outages (Giotsas, et al, 2017)
- Determine the AS behaviour (remove or not communities) (Krenc, et al, 2021)
- Correct values for blackhole communities (Giotsas, et al, 2017)





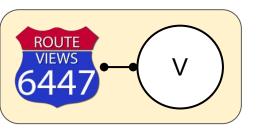
Algorithm

Inference of Location Communities

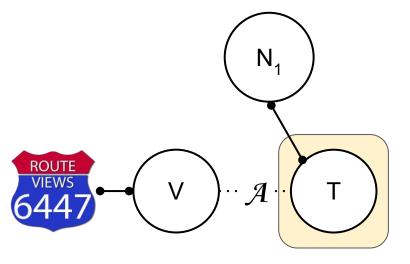


Collector



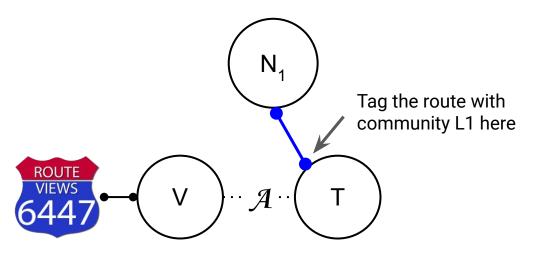




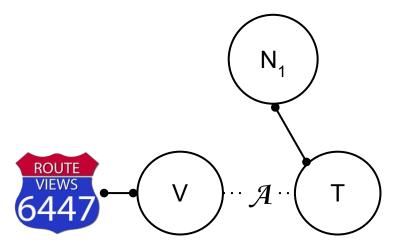


target AS T

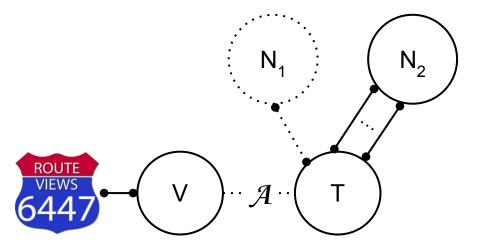




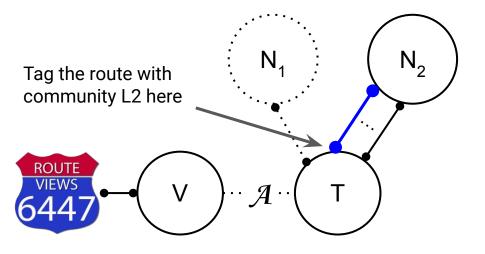




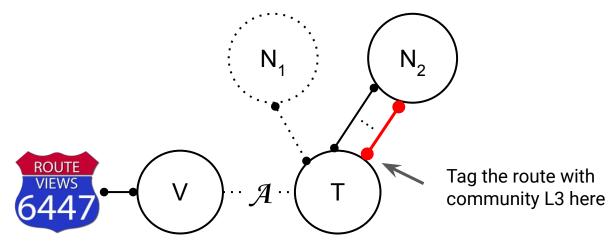




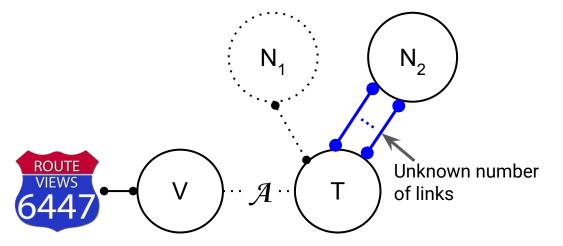




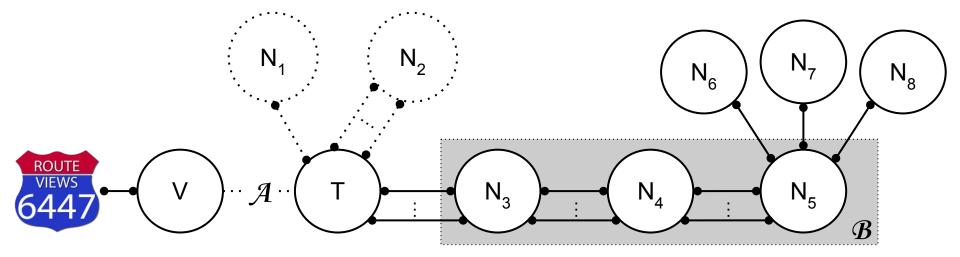




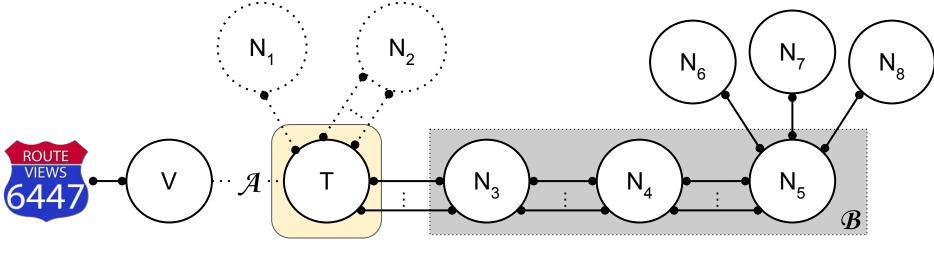






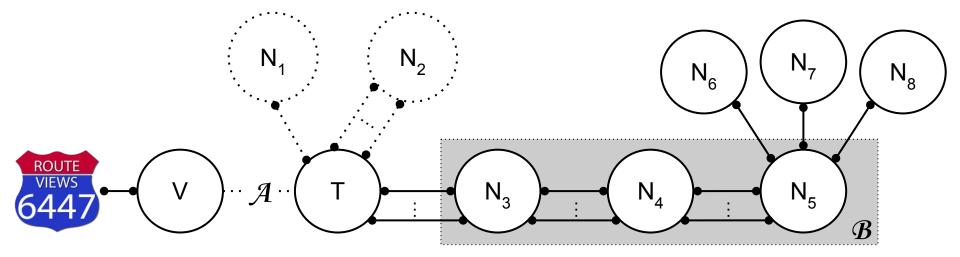




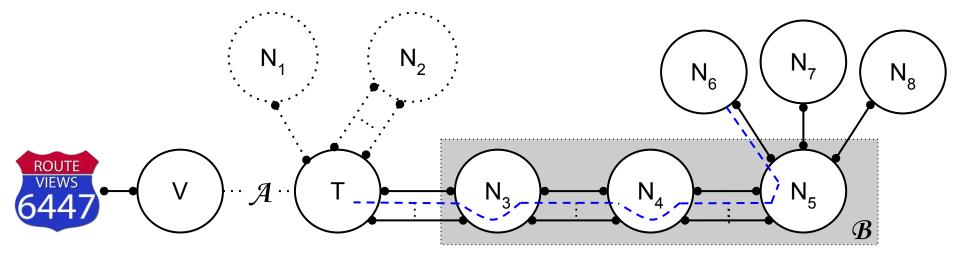


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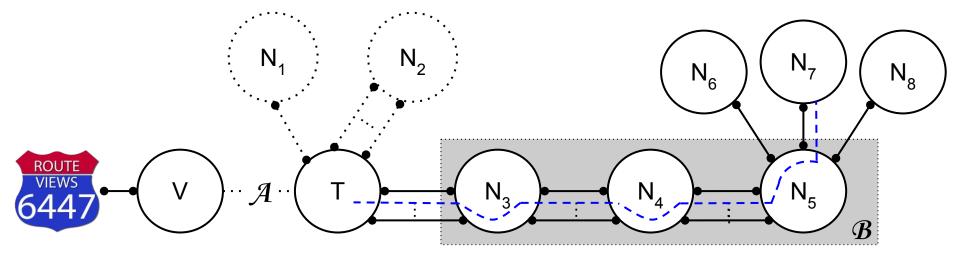




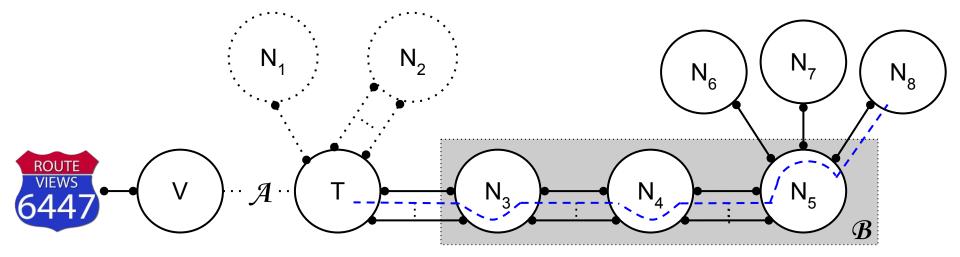




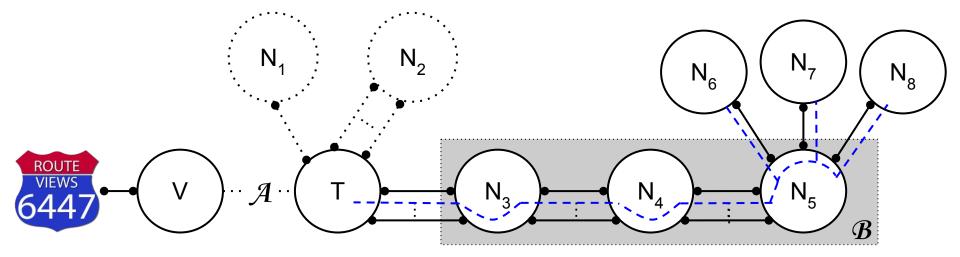




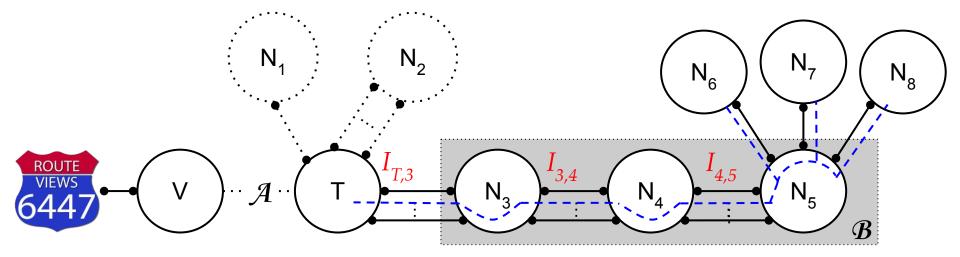




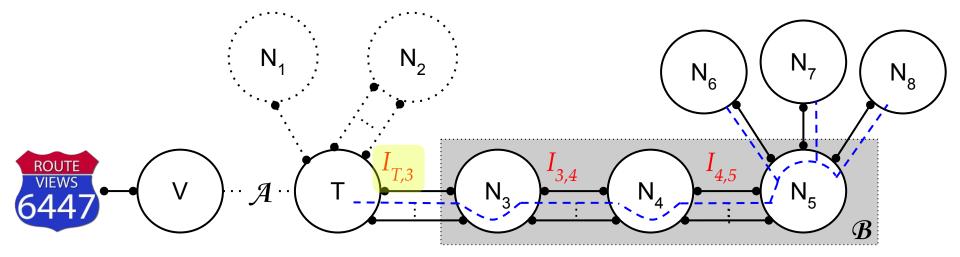




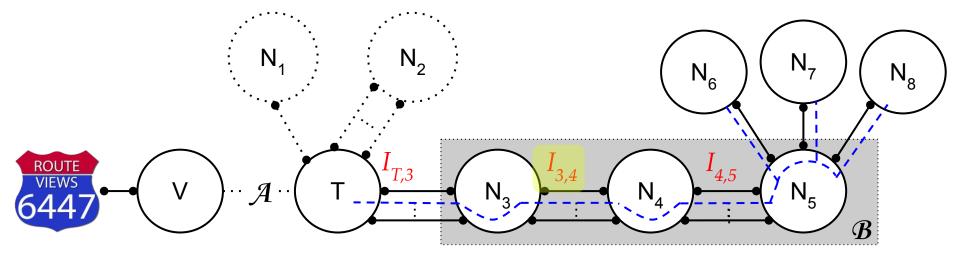




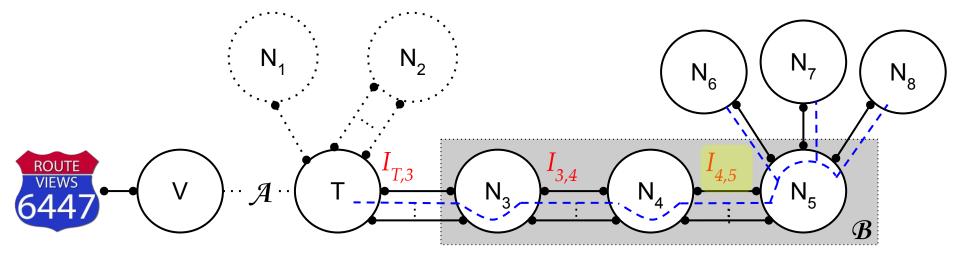






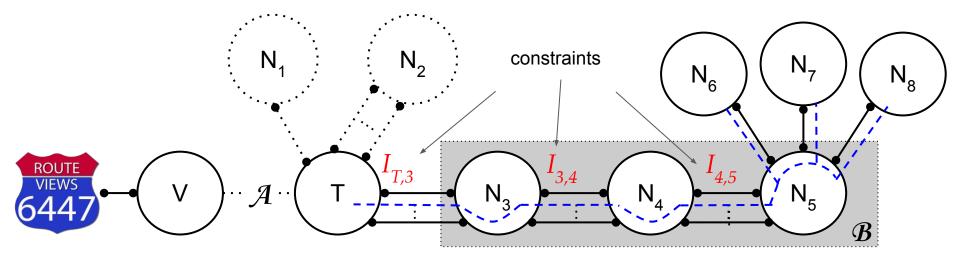






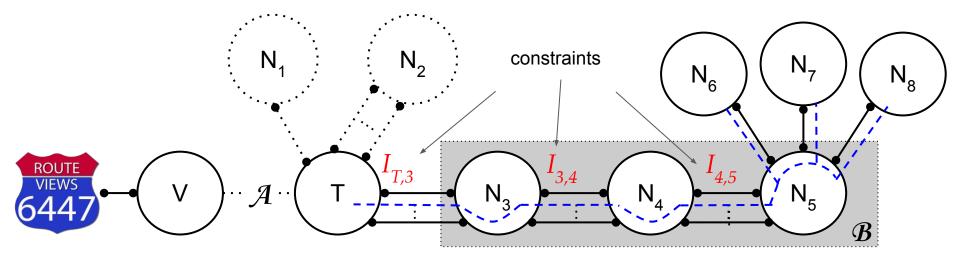


Evaluate Location Communities (limitation)





Evaluate Location Communities (summary)





Results and Dataset

		Commun	ity Type		CAIDA
Network (AS)	Geo	Dev/link	RELATION	ACTION	
Verizon (701)	0	0	0	11	0
NTT (2914)	93	0	2	44	39
GTT (3257)	10,000*	$11,\!000*$	1,783*	13,023*	68
Deutsche Telekom (3320)	24	0	3	0	17
Level $3 (3356)$	178	0	2	5	82
PCCW Global (3491)	44	0	0	21	24
Lumen (3549)	239	239	239	87	28
Orange (5511)	46	0	0	55	11
Zayo (6461)	804*	0	6	152	0
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Project	Project Collectors		V	VPs		Total ASes (thousands)		$\begin{array}{c} \text{Routes} \\ \text{(millions)} \end{array}$	
Year	2017	2020	2017	2020	2017	2020	2017	2020	
RV	17	20	192	232	61	72	96	184	
RIPE	20	20	330	510	61	72	115	311	
Isolario	4	5	83	145	60	72	66	209	
Total (uniq)	41	45	529	738	61	73	277	704	



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			Inf.
CONFIGURATION	Precision	Recall	Recall
Prioritize precision	0.93	0.72	0.89
Default configuration	0.91	0.80	0.87
Prioritize recall	0.87	0.81	0.89



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Automatic Inference vs CAIDA

Community				Comm	IUNITIES
Type	DATABASE	RECALL	PRECISION	Total	Correct
Geolocation	CAIDA	0.21	0.86	303	261
	Inferences	0.77			
Location	Inferences	0.80	0.91	1081	983



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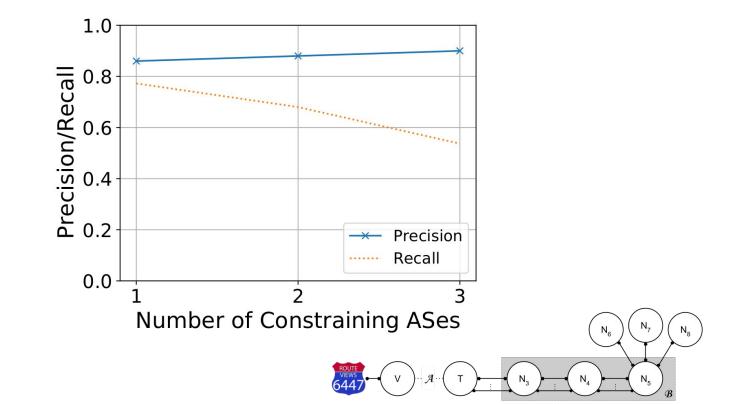


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	Inferences	0.77			
Location	Inferences	0.80	0.91	1081	983



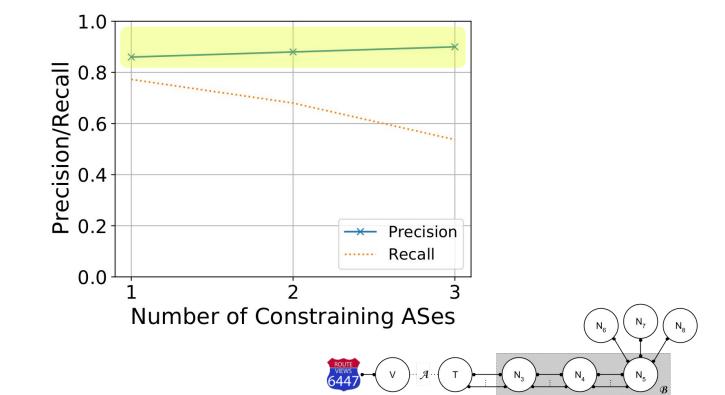
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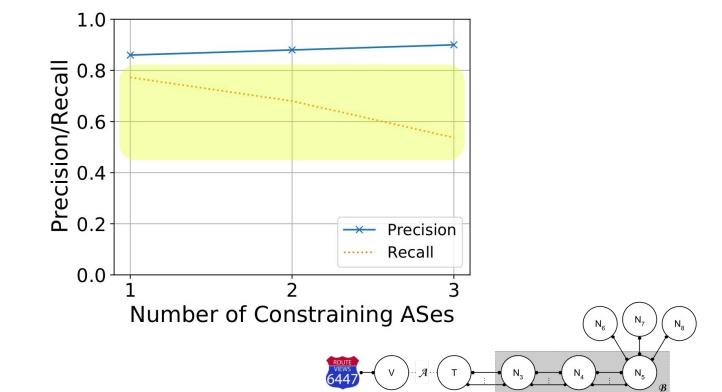
COLUMBIA UNIVERSITY

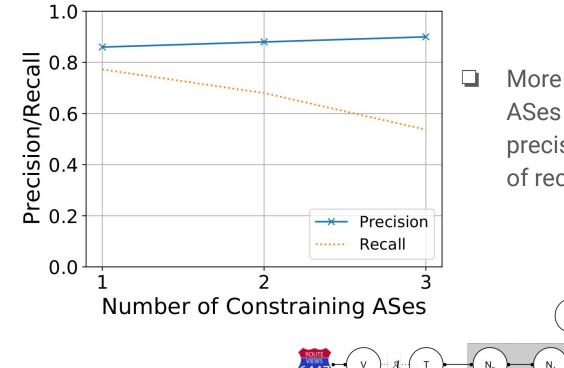
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More constraining ASes lead to higher precision at the cost of recall.

N_

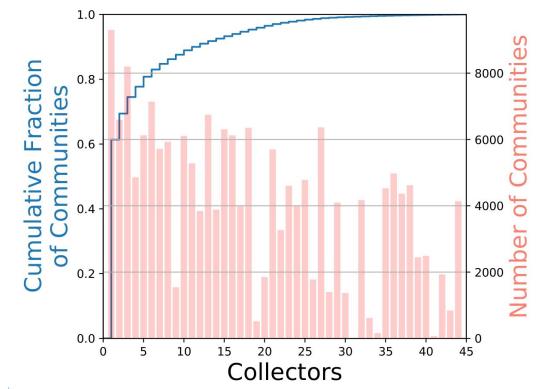
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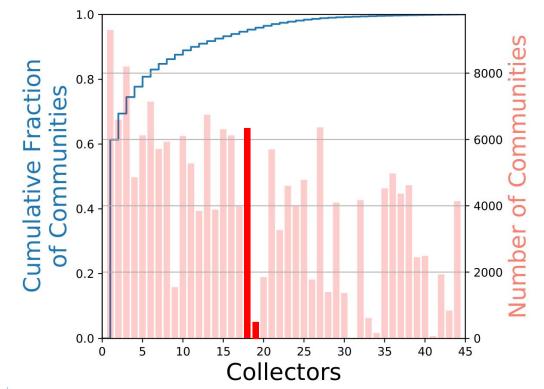
N₈



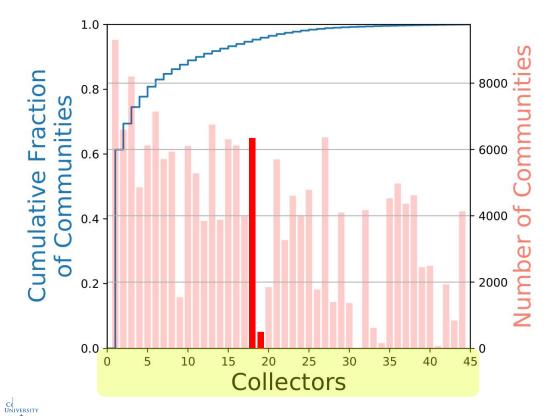
Contribution on Inference per Collector



Contribution on Inference per Collector



Contribution on Inference per Collector



UF*M*

Additional collectors would support more inferences.

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- Our work is the first we are aware of to use routing announcements to infer the semantics of BGP communities.
- Our algorithm automatically infers location communities and achieves high precision (93%) and recall (81%).
- Our manually-built ground truth DB, as well as the code to generate the location DB are publicly available.



Thank you!