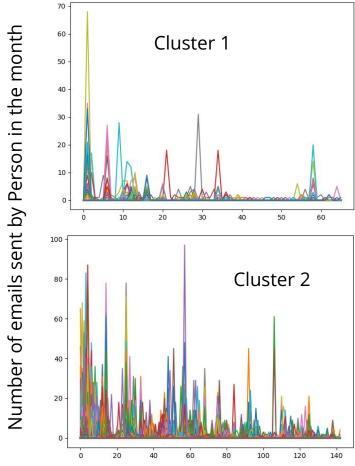
Some Research and Methodologies with IETF Data

Priyanka Sinha, RASPRG Meeting IETF116

Outline

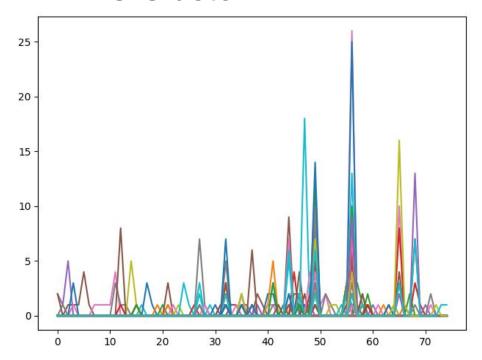
- Summary from AID 2021 workshop participation and further discussions
- Relevant Methodologies for IRTF RASPRG Research
- How is IETF data valuable for Open Research?
- Summary of Hackathon



Time in months since v6ops WG started

- First level grouping of homophilic participants by their temporal activity within v6ops working group in 10 clusters
 - People in Cluster 1 are interested in different aspects of v6ops than those in Cluster 2.
 - Jari Arkko, Mirja Kuhlewind in Cluster 1
 - Timothy Baldwin, Michael
 Richardson, Paul A. Vixie in Cluster
 2
 - Fred Baker in another cluster
 - Activity peaks consistently overlap indicating they are interested in similar issues - help identify advocates? Leaders?
- Entity Disambiguation Same name, different email address found in same cluster - eg Eric Klein, Linjian Song, Christopher Liljenstolpe, m7m7, etc

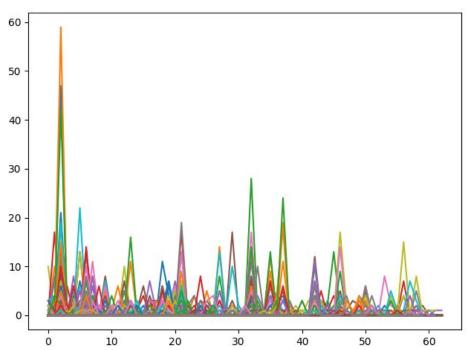
MPLS Cluster



- Adrian Farrel afarrel@juniper.net
- Gaurav agrawal <u>gaurav.agrawal@huawei.com</u>
- Himanshu Shah hshah@force10networks.com
- 정태식 <u>cts@etri.re.kr</u>
- 吴吉朋 wujipeng@gmail.com
- SP sp543@nyu.edu
- Raghu <u>raghav.rao@wipro.com</u>
- Arashmid Akhavain <u>arashmid@nortelnetworks.com</u>
- AtrJoh@netscape.net
- Kullberg Alan-G19424 <u>alan.kullberg@motorola.com</u>
- DECRAENE Bruno RD-CORE-ISS <u>bruno.decraene@francetelecom.com</u>
- "Naidu, Venkata" Venkata.Naidu@Marconi.com

With activity peaks aligned, these people possibly have the similar viewpoint or interest within MPLS WG. Further text mining on their content, would provide say topics and opinions for new participants at IETF to find shepherds and advocates.

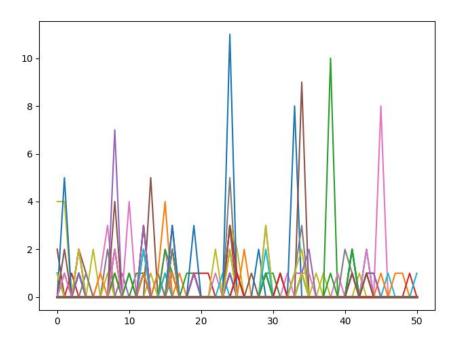
IDR Cluster



- Tony Li tli@juniper.net
- Jakob Heitz <u>iheitz@redback.com</u>
- Gargi Nalawade <u>nalawade@redback.com</u>
- Saikat Ray <u>ray.saikat@ericsson.com</u>
- Adrian Farrel adrian@olddog.co.uk
- Dhruv Dhody <u>dhruv.ietf@gmail.com</u>
- Yimin Shen yshen@juniper.net

Another example of entity disambiguation of emails and affiliations from temporal activity. I know some of these people have worked together in person earlier, even though their affiliation has changed, even last name! Likely have similar technical opinions and help IETF to speed process of consensus and maybe even adoption.

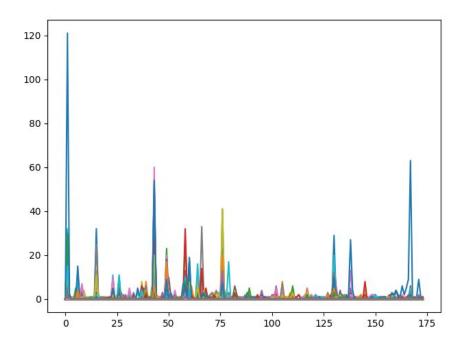
CCAMP Cluster



- ogino <u>ogino@kddilabs.jp</u>
- Wataru İmajuku imajuku.wataru@lab.ntt.co.jp
- Acee Lindem <u>acee@cisco.com</u>
- Dhruv Dhody <u>dhruv.ietf@gmail.com</u>
- "Linwei (Wei)" wei.linwei@huawei.com
- Don Fedyk <u>dwfedyk@nortel.com</u>
- FEDYK Don
 <u>Donald.Fedyk@alcatel-lucent.com</u>
- David Ward <u>dward@juniper.net</u>
- VICTOR LOPEZ ALVAREZ
 victor.lopezalvarez@telefonica.com
- Francesco Lazzeri <u>francesco.lazzeri@ericsson.com</u>

Seemingly different demographics and affiliations may have similar opinions or interests. Motivation for IETF to look beyond demographics and affiliations.

All Combined Cluster



- Adrian Farrel <u>adrian@olddog.co.uk</u>, <u>olddog@clara.co.uk</u>, <u>drian-nomcom@olddog.co.uk</u>
- "Dutta, Pranjal (Pranjal)"
 <u>pdutta@alcatel-lucent.com</u>,
 pranjal.dutta@alcatel-lucent.com
- 정연쾌 <u>vkjeong@etri.re.kr</u>
- <u>E.T.Metz@telecom.tno.nl</u>
- Jakob Heitz iheitz+041207@redback.com
- Balaji Pitta venkatachalapathy
 <u>balaji pv@hotmail.com</u>

Entity disambiguation from temporal analytics helps identify the person, downstream text mining for richer analytics for IETF

Causal Learning Dialogue Outcomes

Attribute	Predicate	Value Domain
sentiment	SENTIMENT_VERY_NEG SENTIMENT_NEG SENTIMENT_LOW_POS SENTIMENT_MEDIUM_POS SENTIMENT_HIGH_POS	<-0.2 [-0.2,0) [0,0.5) [0.5,0.8) >0.8
EMPATH	CATEGORY_ABSENT CATEGORY_LOW CATEGORY_MEDIUM CATEGORY_HIGH	=0 (0,1) [1,2) >=2

Formal method text mining to study IETF mailing list

in speeding consensus.

conversations, their outcomes and causes of the same. May help



DOMESTIC WORK ABSENT **AL EMERGENCY ABSE**

FAMILY ABSENT && !SENTIMENT POS ##[0:3000] FAMILY HIGH && SENTIMENT_VERY_POS##[0:10000] DOMESTIC WORK ABSENT

> +1 19:37:27 Yes I agree with you but I really don like when it comes and lick you...! +1 21:36:11 ha ha ha! thats due to their loving nature, I guess :) +6 12:12:47 Cleaning the desktop????? +8 08:49:34 yaaa frm inside.... P +64 13:12:40 what a story....nd people still fall for it +64 14:36:22 yaa..thts the trick...... :) :P +64 15:01:06 yeah jus like u fell for it +64 20:27:50 yaa like evryone i also fell for it...] aftr tht i posted it here... +65 11:27:47 u seem so happy that u fell for it :) •93 19:58:05 So Hutch/Vodafone dog has got a new job.... Cleaning or systems screen from inside.... :P

Multigraph Edge Features for Graph Mining Peer Groups

Enron data set

	#words	#vocabulary	EMPATH	#emails sent	#emails recvd	sentiment	#entities	#capital words
avg	1047	949	0.004	2	3	0.3	109	222
max	101549	92491	4.31	226	343	22.8	9972	21504
min	1	1	0	1	1	-4.3	0	0

		#urls	#verbs	#auxiliaries	#symbols	#numbers	#nouns	#adjectives	#adverbs	#pronouns
	avg	10	141	198	18	38	337	57	32	36
	max	1002	13870	19680	2189	5910	32349	5403	3420	3848
	min	0	0	0	0	0	0	0	0	0

• StackExchange data set

	#questions	#comments	comment sentiment	popularity	EMPATH
avg	1.17	1.17	0.14	3.12	0.002
max	23	23	5.28	276	0.63
min	1	1	-1	0	0

Findings from Computational Psycholinguistics

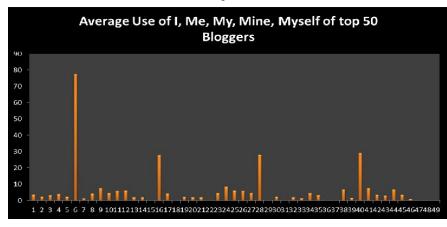
Coverage Evaluation

Proposed Method	#Words in dictionary	#Words found in dataset		%usage in dataset	%coverage of users
HEXACO	245	152	0.08	0.22	13
HEXACO Extension	2,108	1,999	1.07	3.95	50.18
LIWC	4,487	3,993	2.16	43.77	90.51

Sentiment versus Traits

Corelation Between	Honesty	Emotionality	Extraversion	Agreeableness	Conscientious ness	Openness
posts versus feedback score	0.0882	0.1770	0.3362	0.3287	0.3471	0.4005
posts score versus positive emotions	-0.0914	0.2040	0.0873	0.0419	0.3476	-0.0937
Posts score versus negative emotions	-0.1316	0.1963	-0.0520	-0.0810	0.2556	0.0064
Positive emotions in posts versus feedback	0.3171	0.2700	0.3844	0.4334	0.3368	0.4689
Negative emotions in posts versus feedback	0.3147	0.5070	0.1544	0.4677	0.3845	0.3438

LIWC Stability Observation



Observations

- Openness is independent of use of emotive words
- Open, agreeable, extraverted, conscientious people evoke similar traits in the comments they receive from people.
- Use of 1st person personal pronouns is an indication of depression. Here the person has posted tragic short love stories.

Why study inclusion in the IETF

- IETF is a voluntary global organization with its communication, activities recorded and available for analysis and study
- Diversity, Inclusion and Representation make the consensus process robust.
- Do consensus mechanisms depend on in-person meeting, side talks and in person advocacy?
 - This puts participants who engage in remote-only mode (perhaps due to geographical, personal constraints) at a disadvantage
- People are multi-dimensional and their sense of community may not be appropriated by their observable demographic indicators
 - Data driven identification of influence that some members may have on consensus building

IRTF RASPRG Relevant Research Directions

- Analyzing the development of the make up of standard setting communities, their diversity, and the impact it has on standard-setting.
 - Study Contextual Integrity in terms of Group Behavior. Research to find out computational models of people's behavior from their perspective of privacy.

- Understanding the decision making processes that lead to the production of publications.
 - Identify key people who may help provide efficient and valuable consensus on drafts, proposals, rfcs and standards.

Open Research Problems for which IETF Data is valuable

- Good data source to build a Universal Behavior Model quantitative studies for RASP RG
 - Large engagement from various demographics.
 - Semi Formal text. Less Noisy. Not short text.
 - Many turns of dialogues. Several messages in a Thread.
 - Active open source tools BigBang
- Provide data driven insights into personality and group behavior for a population useful in say evaluating autonomous AI for vehicles - help IETF WG with human factors insight
- Provide data driven insights into diversity, norms
- A big data resource for psychologists and cognitive scientists to test their hypotheses computationally - confirm qualitative studies with data driven insights for RASP RG

Summary of Hackathon

Thank You

- https://scholar.google.co.in/citations?hl=en&user= ctlSowAAAAJ&view_op=list_works&sortby=pub date
- https://www.youtube.com/@PriyankaSinhaMahap atra
- https://www.linkedin.com/in/priyanka1982
- https://twitter.com/priyanka_iitg