



Twitter rape threats and the **D**iscourse of Online Misogyny (DOOM)

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Supported by the ESRC: [grant number ES/L008874/1]





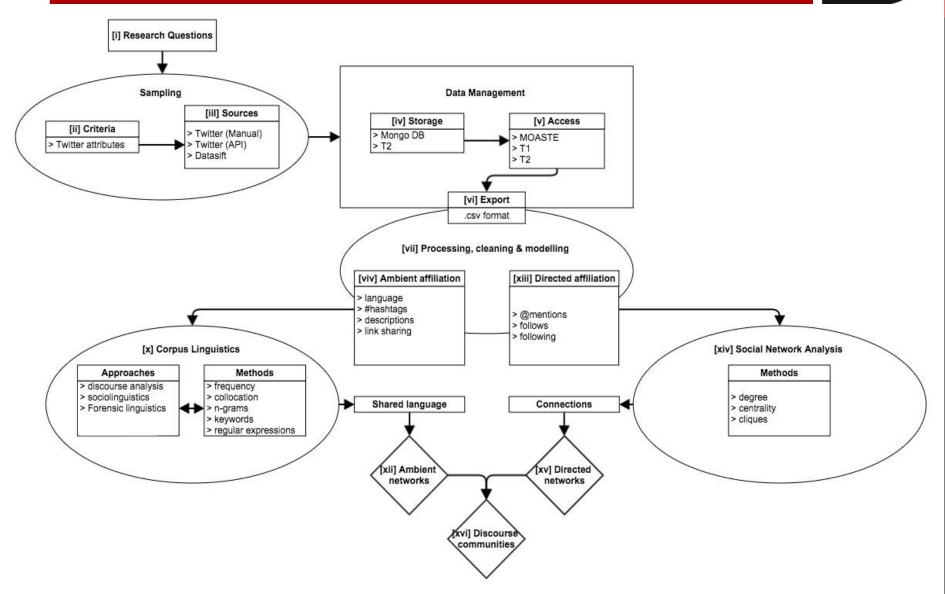
- What are we investigating? Hate speech, sexist abuse, violent forms of misogyny online
- Why? Investigative bodies etc. pressured to act against increasingly violent online misogyny, but remarkable lack of research
- The aims of DOOM understand online misogyny, address public concerns, and inform/advise policy, practice, legislation



- Why this corpus and event?
 - Known case of misogynistic online abuse
 - Known target
 - Legal proceedings
 - John Nimmo, Isabella Sorley, Peter Nunn
- How did we gather it?
 - Seed corpus and DataSift
- How did we analyse it?
 - Corpus linguistics (CL) AntConc & R
 - Social network analysis (SNA) Gephi & R
 - Discourse analysis (DA)

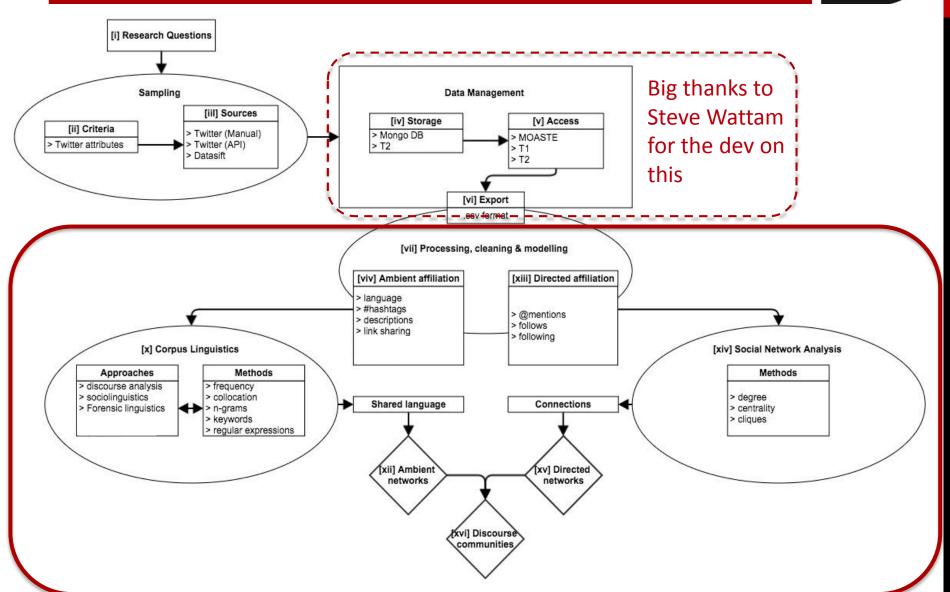


CoCoA Framework - Corpus-assisted **Co**mmunity **A**nalysis





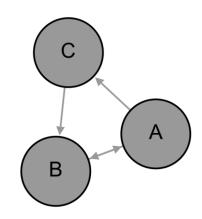
CoCoA Framework - Corpus-assisted **Co**mmunity **A**nalysis





Direct affiliation

- Connections with 'directionality'
- 'directed graphs'^[1] representing relationships ('edges') between actors ('nodes')
 - Asymmetric (e.g. I follow you, you don't follow me)
 - Symmetric (e.g. we mention each other in our tweets)



[3] cf. Scott, J. (2013). Social Network Analysis. 3rd Ed. London: Sage.



Direct affiliation



EDL News @edlnews · Jan 19

Do Nigel Farage's jack booted blue rinse army want to legalise chuffing on a huge joint? #ukip #kipperstoners pic.twitter.com/mQKkczlBht





13.1 ★ ***



The Voice Of Reason @TheVoiceOfRea13 - Jan 19 @edlnews oh fuck off u sad little lefty cunt



13 ★ …





EDL News @edlnews · Jan 19 @TheVoiceOfRea13 First kipper tantrum of the day.





The Voice Of Reason @TheVoiceOfRea13 · Jan 19

@edlnews love to know wat sort of reward you think your goin to receive for opposing anything the majority wants u sad cunt



scott cheadle @schead · Jan 19

@TheVoiceOfRea13 @edInews Ah, the voice of reason. The voice of UKIP. The voice of an illiterate oink unable to string a sentence together.

13.1 ★1 …



The Voice Of Reason @TheVoiceOfRea13 · Jan 19

@schead @edlnews eat shit do gooding prick u haven't got a fuckin brain in your head beleive ul be rewarded for bein a lefty cunt

43.1 ★ ***



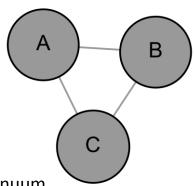
EDL News @edlnews · Jan 19

@TheVoiceOfRea13 @schead superb kipper meltdown, first class! lol



Ambient affiliation^[1,2]

- "to commune with others without necessarily engaging in direct conversational exchanges" [2]
- Shared behaviours/characteristics
- undirected graphs^[3]



[1] Zappavigna, M. (2012) Discourse of Twitter and social media. London: Continuum.

[2]Zappavigna, M. (2013) Enacting identity in microblogging through ambient affiliation. *Discourse & Communication*. 8(2). pp. 209-228.

[3] cf. Scott, J. (2013). Social Network Analysis. 3rd Ed. London: Sage.



Ambient affiliation



BernieBurnette @BB5163 · Jan 9

#UK: Manchester Muslims demanding laws to ban criticism of their paedophile "prophet". #UKIP #CharlieHebdo #EDL #DDL #PVV #Bluehand #TCOT





James Bond @EnglandsTerrier · 3h
How its made.. #Bluehand pic.twitter.com/iDAD6CTNR2



Can anyone tell me why the hell we have Muslim Council of Great Britain... or what the hell they do (other than claim benefits!!)

#bluehand





Andy V @avjestica · Jan 8





Modelling Twitter data for different kinds of analysis

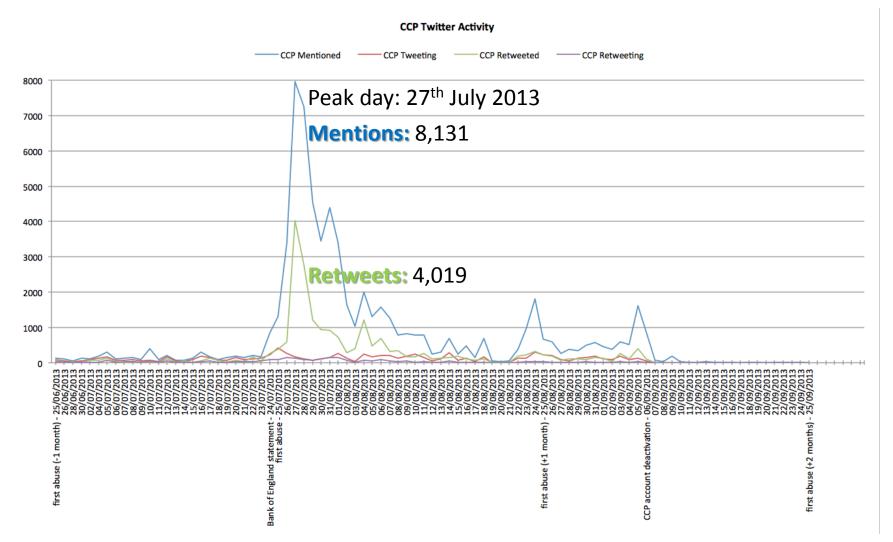
- Twitter data -> ambient affiliation
 - Corpora of tweet text
 - Corpora of descriptions text
 - Frequency lists
 - Word | N-gram | Hashtag | link shares | etc.
- Twitter data -> direct affiliation
 - Interaction 'edgelists'
 - Mentions
 - Retweets
 - Follower/friends relationships
- BONUS: Twitter data -> Timeseries
 - tweets | retweets | mentions | link shares | etc. over time



Any tweets which mention or originate from oCCriadoPerez between 25/06/2013 and 25/09/2014



DOOM dataset





1. How are rape threats constructed?

- What language is used as part of misogynistic abuse online?
- What language is used in relation to it?

2. Who is talking about rape threats?

- Who is talking about (or making) threats online?
- Do they form networks?
- If so, how?



- Extracted top 20 (or fewer) most frequent words (less stopwords) in tweets for each day (92 days)
 - Finds most freq. talk per day
- Aggregated 92 lists of top 20 (or fewer) words into single frequency list
 - List of most freq. words during entire period that accounts for daily variation
- Final list = 456. we looked in more detail at the top twenty...



Analysis – most frequent lexical words overall

Rank	Word	Freq	Rank	Word	Freq
1	Just	4688	11	Can	2694
2	Abuse	4037	12	Rape	2589
3	Twitter	3971	13	Get	2549
4	Don't (dont)	3614	14	Good	2234
5	Women	3388	15	Men	1896
6	People	3208	16	You're (youre)	1763
7	Threats	2996	17	Will	1399
8	Like	2963	18	Support	1277
9	ľm (im)	2830	19	One	981
10	Think	2812	20	Well	937

Overall corpus: top twenty words



Rank	Word	Freq	Rank	Word	Freq
1	Just	4688	11	Can	2694
2	Abuse	4037	12	Rape	2589
3	Twitter	3971	13	Get	2549
appalling ab					2234
5	(one of the	e most com	mon phras	es overall)	1896
	use button,	2200	16	You're	1763
7	Threats	apuse on t	wittei	Will	1399
9		2830	19	One	981

Overall corpus: top twenty words



Defining risky users

- Low risk (147 accounts)
 - General insults and abuse
 - Sexualised, misogynistic remarks
 - Affiliation with high risk users
- High risk (61 accounts)
 - Intent to menace (i.e. to cause fear, threat of harm)
 - (Sexually) aggressive/threatening
 - Incitement to suicide
 - (Repeated) harassment



- Overall proportion of abusive tweets
 - UPPER estimate: 1,957 tweets, from 208 accounts
 - From low and high risk users
 - 2.5:100 tweets abusive
 - LOWER estimate: 705 tweets, from 61 accounts
 - From high risk users
 - 0.9:100 tweets abusive



Rank	Word	Keyness	Rank	Word	Keyness
1	rape	270.5	10	Loool	75.7
2	Cunt	189.2	12	Pussy	73.7
3	Lol	173.4	13	Raping	73.3
4	Bitch	161.4	14	Penis	68.3
5	Raep	158.6	15	Fucking	60.1
6	Jews	119.6	16	Your	59.1
7	Faggot	104.3	17	Ass	56.5
8	Nigger	92.8	18	Gay	51.1
9	Me	83.8	19	Cake	49.8
10	Γ	80.6	20	Cock	49.7

High and low risk: top twenty **key**words



Rank	Word	Keyness	Rank	Word	Keyness
1	rape	270.5	10	Loool	75.7
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Rape



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7			17 'ra	ep train'	56 .5
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Misogyny



Rank	Word	Keyness	Rank	Word	Keyness
1	rape	270.5	10	Loool	75.7
3					73.3
5	Raep	158.6	15	Fucking	60.1
7	Faggot	104.3			56.5
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Homophobia



Rank	Word	Keyness	Rank	Word	Keyness
1	rape	270.5	10	Loool	75.7
3					
5					
6	Jews	119.6			
7					
8	Nigger	92.8			
9	Me	83.8	19	Cake	49.8

Racism/anti-Semitism



Rank	Word	Keyness	Rank	Word	Keyness
1	rape	270.5	10	Loool	75.7
			12	Pussy	73.7
3	Lol	173.4	13	Raping	73.3
			14	Penis	68.3
5	Raep	158.6	15	Fucking	60.1
7			17	Ass	56.5
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			20	Cock	49.7

Genitalia/anatomy

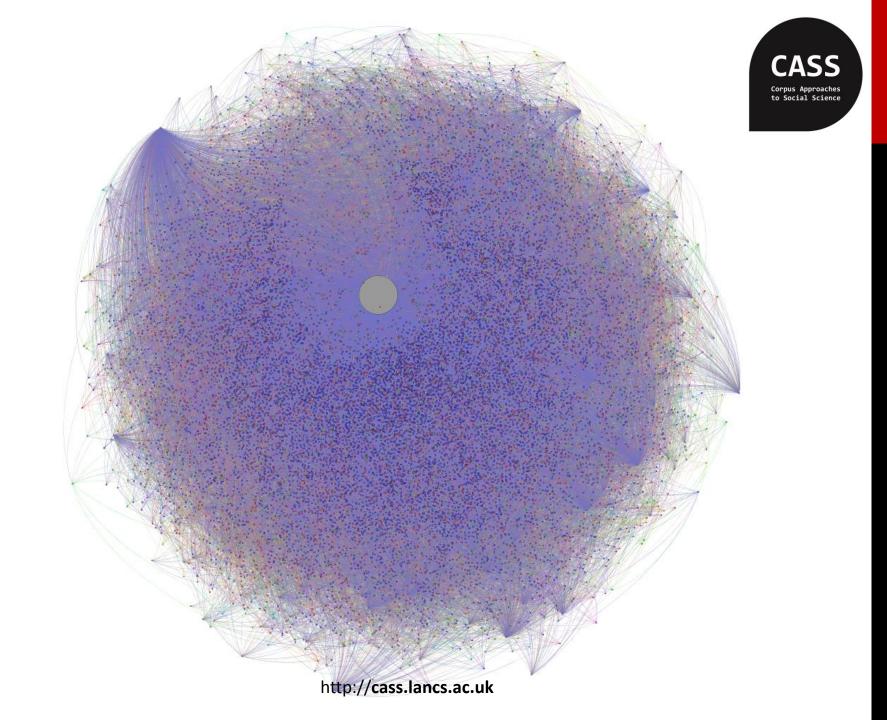


1. How are rape threats constructed?

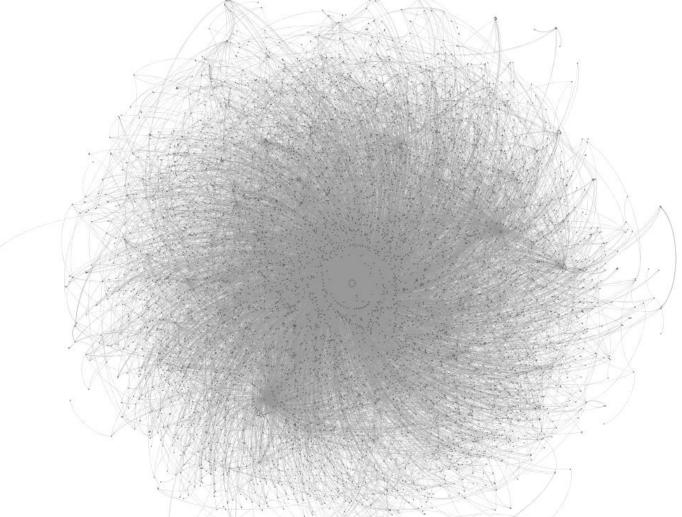
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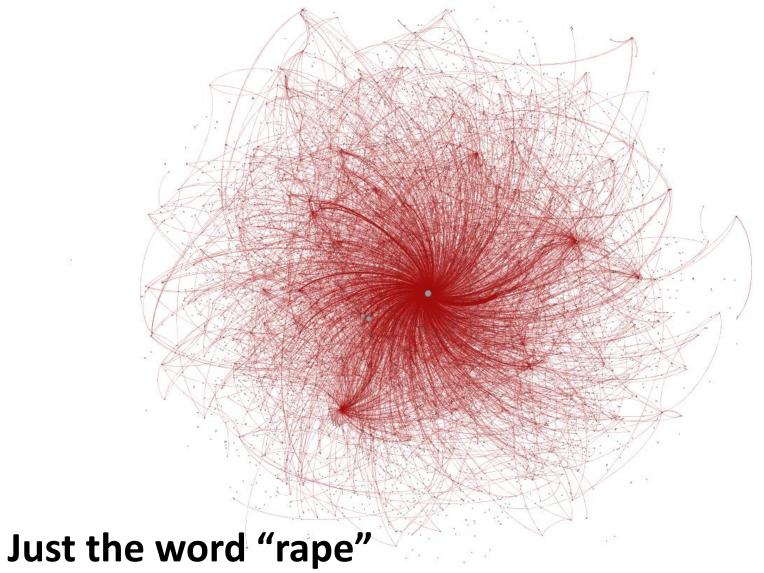






Conversations inc. "rape", "threats", "abuse"



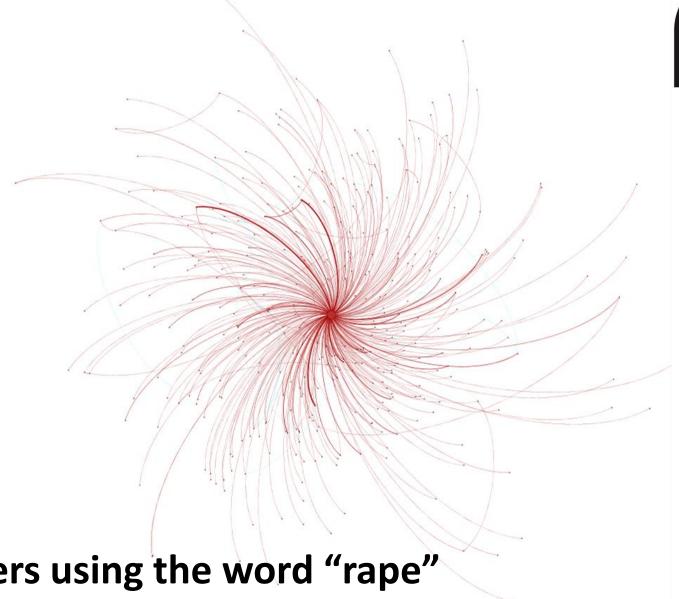




Who is talking about rape threats?

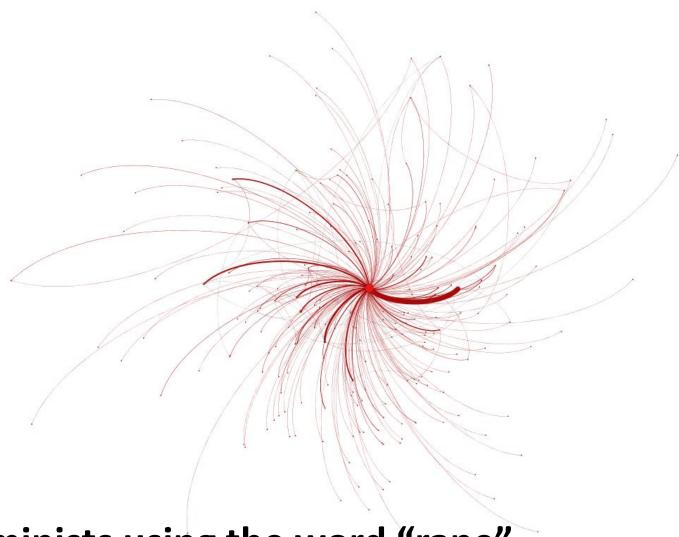
	Word in description	Frequency of occurrence in unique user descriptions (no risk)
Writer/media	Writer	1060
	Editor	431
	Journalist	354
	Books	351
	Blogger	335
	Author	316
	Media	369
Political/activism	Feminist	953
	Politics	598
Education	Student	505
	Teacher	286
	History	284
	Geek	337
Women (relational)	Women	379
	Mum	428
	Mother	339
	Wife	302





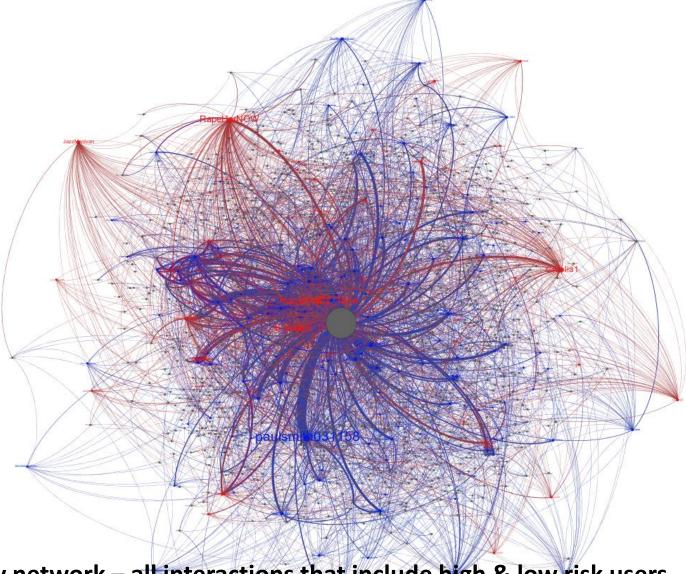
Writers using the word "rape"





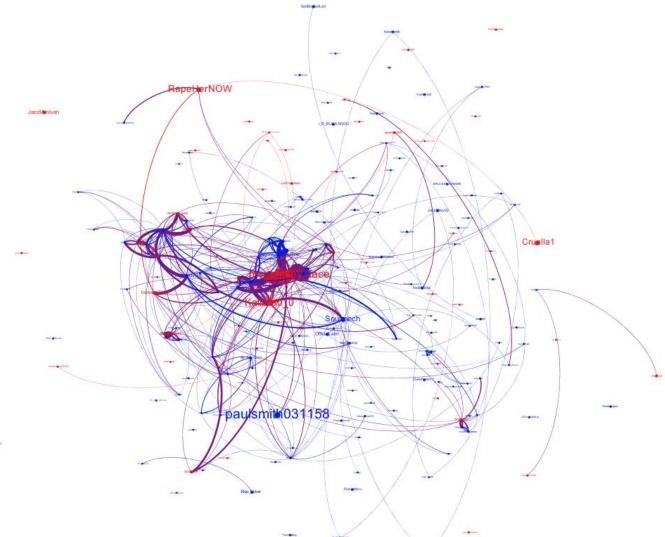
Feminists using the word "rape"





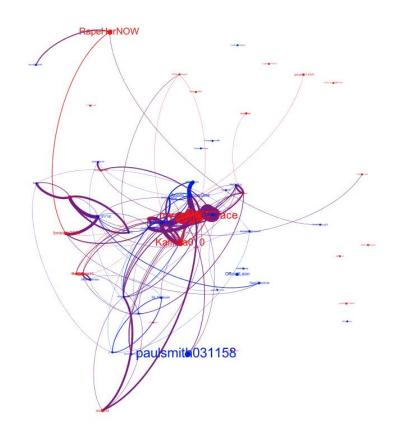
Risky network – all interactions that include high & low risk users





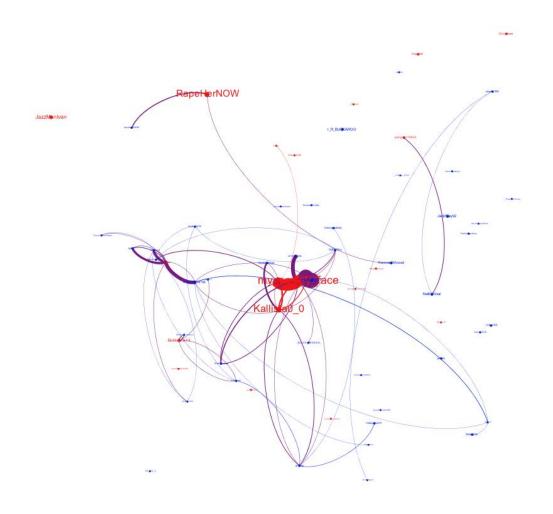
Risky network – high & low risk & interactions between (no-risk removed)





Talk in risky network: rape ("rape", "raep", "raping")





Talk in risky network: misogyny ("cunt", "bitch")

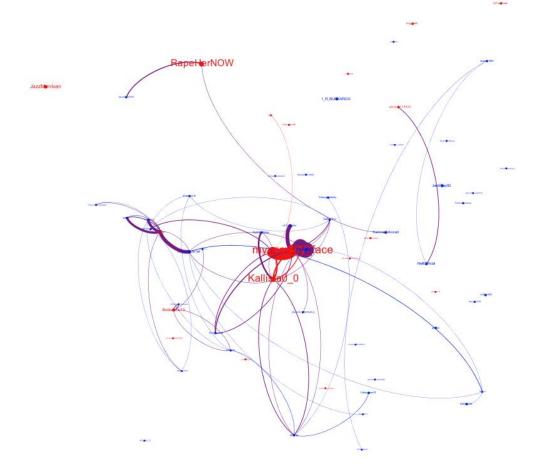
http://cass.lancs.ac.uk



Who is talking about rape threats?

	Word in description	Frequency of occurrence in user descriptions
Group affiliation	.*sec .*sek e.g. #uncletomsec #stfusec #cuntsec #idgafsec	494
	anon #anonymous #cananon #anonmus anonymous	67
	League #rustleleague	56
Racism	Niggas	42
	Nigger	11
Rape	#rapecr3w	101
	Raping	15
	Rape	7

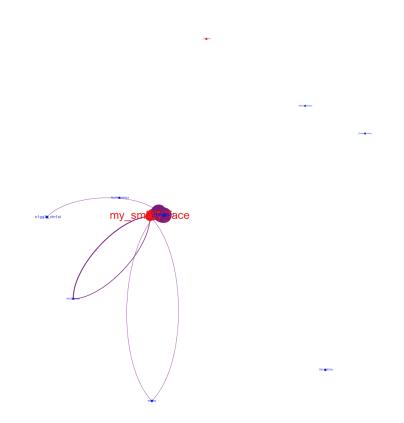




Talk in risky network: misogyny ("cunt", "bitch")

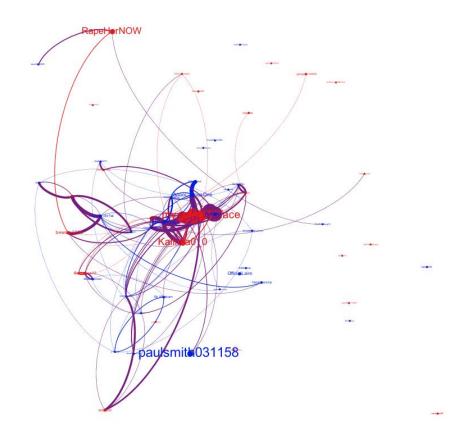
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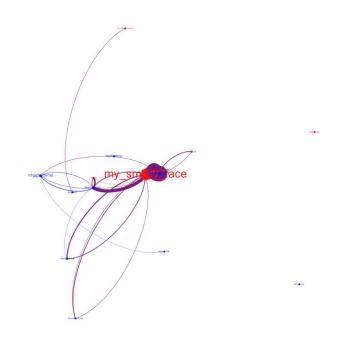
Risky groups (*sec/*sek) & correlation w/risky talk: misogyny ("cunt", "bitch")





Talk in risky network: rape ("rape", "raep", "raping")





Risky groups (*sec/*sek) & correlation w/risky talk: rape ("rape", "raep", "raping")



1. How are rape threats constructed?

- Use of language relating to
 - Violence (e.g. abuse, rape)
 - Homophobia
 - Racism
 - "Faithism" (e.g. Anti-Semitism)
 - Genitalia and anatomy



2. Who is talking about rape threats?

- Non-risky users: writers, activists, etc.
 - (i.e. word analysis alone is not enough too much "noise")
- Risky users: forming coherent networks using ingroup markers
 - in bios, e.g. *sec/*sek
 - in language, e.g. "raep" (also "raep train", "conductor", etc.)
 - in who they affiliate with, mention, retweet, etc.



• Future work:

- Generating linguistic profiles for abusive online behaviour (e.g. hate speech, threats)
- Methods for detecting abusive online behaviour
- Identifying escalation of that behaviour (individual or group) – possibly threat/threat-level indication?
- Investigating how risky communities (networks) develop and evolve