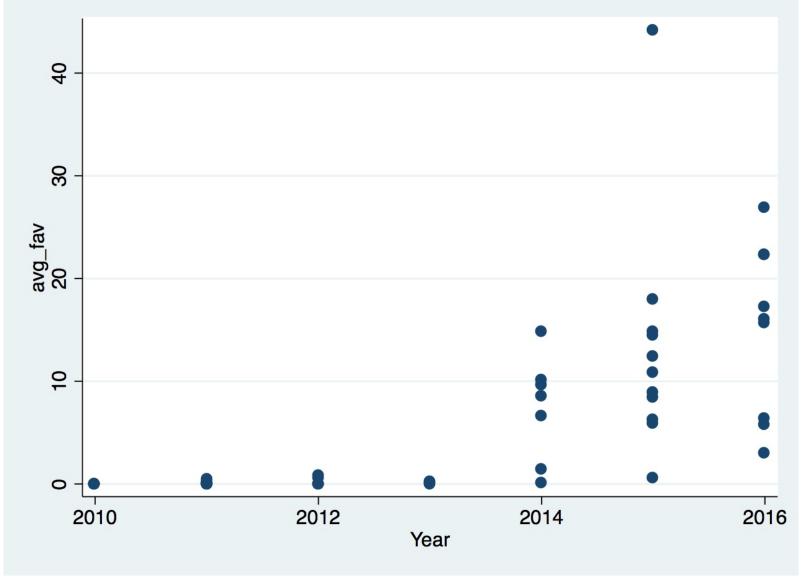
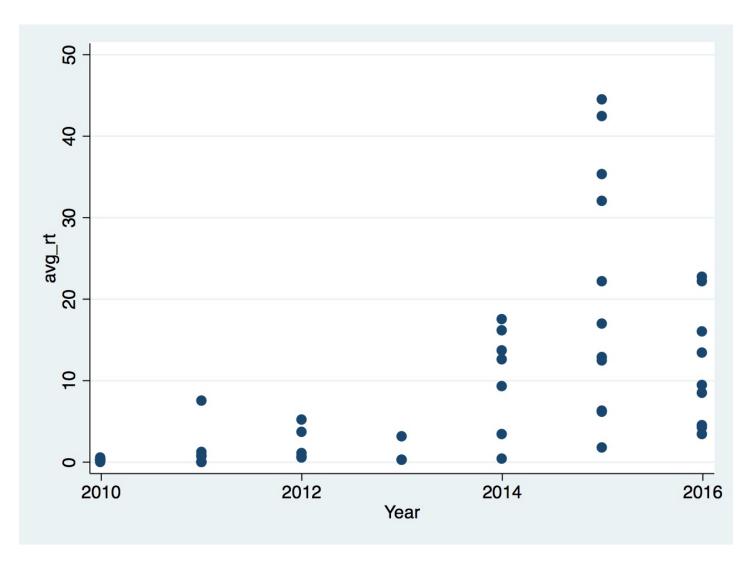
# Preliminary Results Analysis of twitter activity as a measure of interest

Date: 03.13.2017

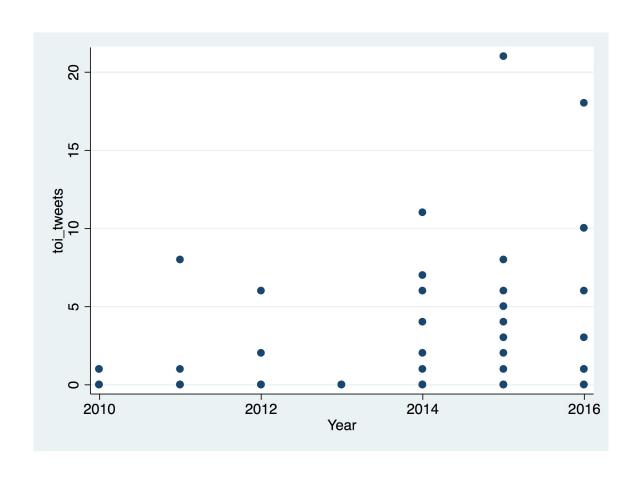
#### Fav activity increasing with time



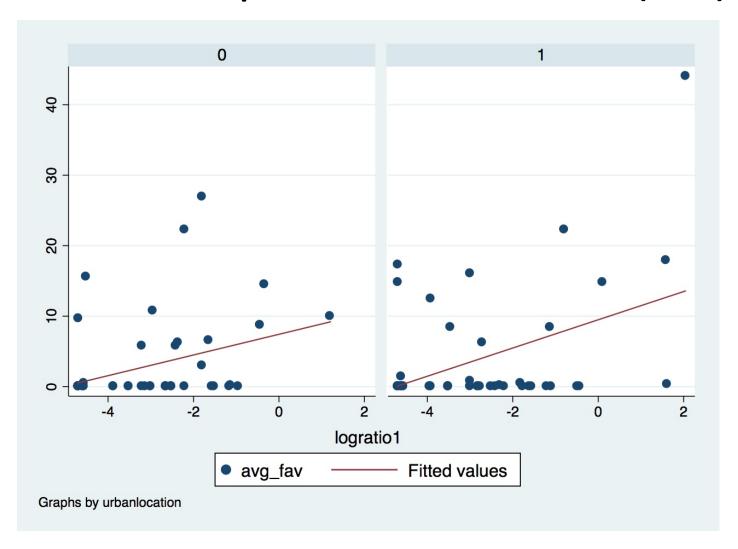
#### RT activity increasing with time



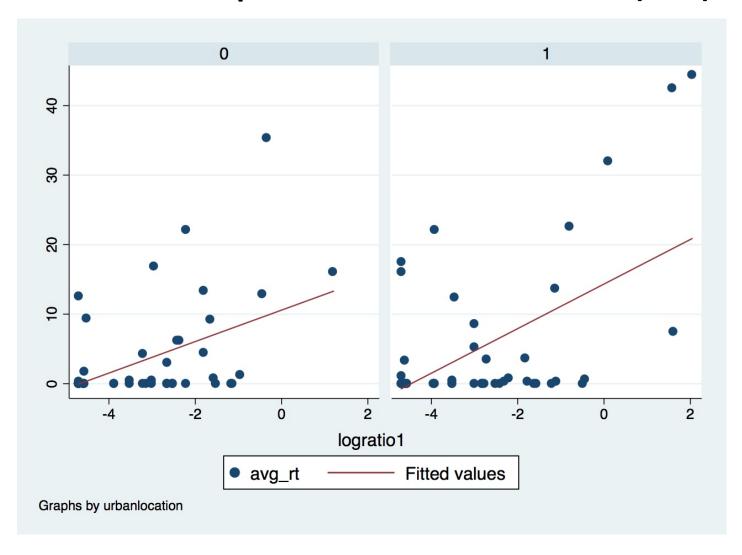
#### Tweets increasing with time



## Interest measures show predicted relationship with each other (fav)



## Interest measures show predicted relationship with each other (RT)



## For tweeted disasters, higher RT activity seen for urban disasters

. bys urbanlocation: sum avg\_rt if tweeted==1

avg rt	24	10.76744	13.3626	0	44.41071
Variable	0bs	Mean	Std. Dev.	Min	Max
· urbanlocat	ion = 1				
avg_rt	23	7.678985	8.842567	0	35.25
Variable	0bs	Mean	Std. Dev.	Min	Max
> urbanlocat	ion = 0				

## For tweeted disasters, higher fav activity seen for urban disasters

```
. bys urbanlocation: sum avg_fav if tweeted==1
-> urbanlocation = 0
    Variable
                       0bs
                                   Mean
                                           Std. Dev.
                                                            Min
                                                                        Max
     avg_fav
                        23
                              6.385839
                                           7.647204
                                                                   26.90625
-> urbanlocation = 1
    Variable
                                           Std. Dev.
                       0bs
                                   Mean
                                                            Min
                                                                        Max
     avg_fav
                                           10.72251
                        24
                               7.755237
                                                                   44.09018
```

#### Urbanlocation no longer significant in explaining avg\_fav count as measure of interest

. reg avg\_fav logdeaths i.distype i.year urbanlocation if tweeted==1

	Source	SS	df	MS	Number of obs	=	47
-					F(12, 34)	=	3.28
	Model	2120.25564	12	176.68797	Prob > F	=	0.0032
	Residual	1832.68223	34	53.9024187	R-squared	=	0.5364
_					Adj R-squared	=	0.3727
	Total	3952.93787	46	85.933432	Root MSE	=	7.3418

avg_fav	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
logdeaths	.5046856	.8027513	0.63	0.534	-1.126701	2.136073
distype						
Extreme Temperature	4580165	4.778219	-0.10	0.924	-10.16853	9.252494
Flood	1.585045	4.087165	0.39	0.701	-6.721073	9.891163
Mass movement (dry)	3.227875	9.311604	0.35	0.731	-15.69558	22.15133
Storm	-1.814857	4.811726	-0.38	0.708	-11.59346	7.963746
year						
2011	796516	4.799644	-0.17	0.869	-10.55057	8.957534
2012	-1.666104	5.422057	-0.31	0.761	-12.68505	9.352842
2013	-2.081565	5.653427	-0.37	0.715	-13.57071	9.40758
2014	6.720936	4.612729	1.46	0.154	-2.653258	16.09513
2015	12.01773	4.491183	2.68	0.011	2.890545	21.14491
2016	13.57317	4.578107	2.96	0.006	4.269338	22.87701
urbanlocation	3.139481	2.577723	1.22	0.232	-2.099083	8.378045
_cons	-2.961057	5.620564	-0.53	0.602	-14.38342	8.461304

## As compared to interest measured through google trends

. reg logratio1 logdeaths i.distype i.year urbanlocation

	Source	SS	df	MS	Number of obs	=	103
-					F(13, 89)	=	5.43
	Model	122.584196	13	9.42955357	Prob > F	=	0.0000
	Residual	154.638107	89	1.73750683	R-squared	=	0.4422
_					Adj R-squared	=	0.3607
	Total	277.222304	102	2.71786572	Root MSE	=	1.3181

logratio1	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
logdeaths	.2814007	.0999045	2.82	0.006	.0828926	.4799088
distype						
Extreme Temperature	-4.36207	.723282	-6.03	0.000	-5.799216	-2.924924
Flood	-2.896685	.6008257	-4.82	0.000	-4.090513	-1.702857
Landslide	-2.483376	.8253432	-3.01	0.003	-4.123315	8434365
Mass movement (dry)	-2.872433	1.51975	-1.89	0.062	-5.892144	.1472788
Storm	-4.04006	.6127435	-6.59	0.000	-5.257568	-2.822551
year						
2011	.1837595	.5018559	0.37	0.715	8134174	1.180936
2012	4221598	.5553821	-0.76	0.449	-1.525692	.6813725
2013	8127649	.5082197	-1.60	0.113	-1.822587	.1970568
2014	.3564589	.4693818	0.76	0.450	5761927	1.28911
2015	1799507	.4412534	-0.41	0.684	-1.056712	.6968104
2016	4118767	.4865303	-0.85	0.400	-1.378602	.5548487
urbanlocation	. 6379847	.2934805	2.17	0.032	.0548451	1.221124
_cons	992174	.7018526	-1.41	0.161	-2.38674	.4023921

### Source: toi analysis

. bys urbanlocation: sum toi\_tweets

urbanlocat	ion = 0				
Variable	0bs	Mean	Std. Dev.	Min	Max
toi_tweets	57	.8596491	2.271238	0	11
urbanlocat:	ion = 1				
Variable	0bs	Mean	Std. Dev.	Min	Max
toi_tweets	46	2.326087	4.590107	0	21
bys urbanlo	cation: sum to			v	21
<b>bys urbanlo</b> -> urbanlocat	cation: sum to			•	21
	cation: sum to			Min	Max
-> urbanlocat	cation: sum to	oi_tweets if	toi_tweets>0		
-> urbanlocat Variable	cation: sum to	oi_tweets if  Mean	toi_tweets>0  Std. Dev.	Min	Max
<pre>variable toi_tweets</pre>	cation: sum to	oi_tweets if  Mean	toi_tweets>0  Std. Dev.	Min	Max

#### Source: toi analysis

> urbanlocati	lon = 0					
Variable	0bs	Mean	Std. Dev.	Min	Max	
toi_max_rt	12	29.68611	24.82903	2	100	
toi_max_fav	12	25.28889	18.33872	0	57.5	
> urbanlocati	on = 1					
Variable	0bs	Mean	Std. Dev.	Min	Max	
toi_max_rt	18	28.15483	30.0126	0	101.625	
oi_max_fav	18	24.17652	27.16208	0	108.625	
bys urbanl	ocation: sum	toi_max_rt	toi_max_fav			
> urbanloca Variable		s Me	an Std. De		Min 0	
→ urbanloca Variable toi_max_rt	tion = 0 Ob:	s Me 7 6.2497	an Std. De <b>08 16.4371</b>	7		100
> urbanloca Variable	tion = 0    Ob:	s Me 7 6.2497	an Std. De <b>08 16.4371</b>	7	0	Max 100 57.5
<pre>vurbanloca Variable toi_max_rt toi_max_fav</pre>	tion = 0    Ob:	6.2497 5.3239	an Std. De <b>08 16.4371</b>	7 3	0	100

9.460378

20.51902

108.625

toi\_max\_fav