

By NEWS event, proportion missing values per attribute

```
> apply(d[,1:7],2,function(x) mean(is.na(x)))
  n_aff tot_de  n_inj country  n_hl tot_aff  date
0.2794 0.1180 0.7343 0.0000 0.9671 0.1996 0.0000
```

per-event aggregation

```
> ee=d %>% group_by(eid) %>%
summarise(num_articles=n(),tot_de=tot_de[1],tot_aff=tot_aff[1],c
ountry=country[1],e_type=e_type[1]) %>% filter(!is.na(tot_de)
& !is.na(tot_aff))
```

Filter to articles mentioning events that have both total deaths and total affected data.

per type:

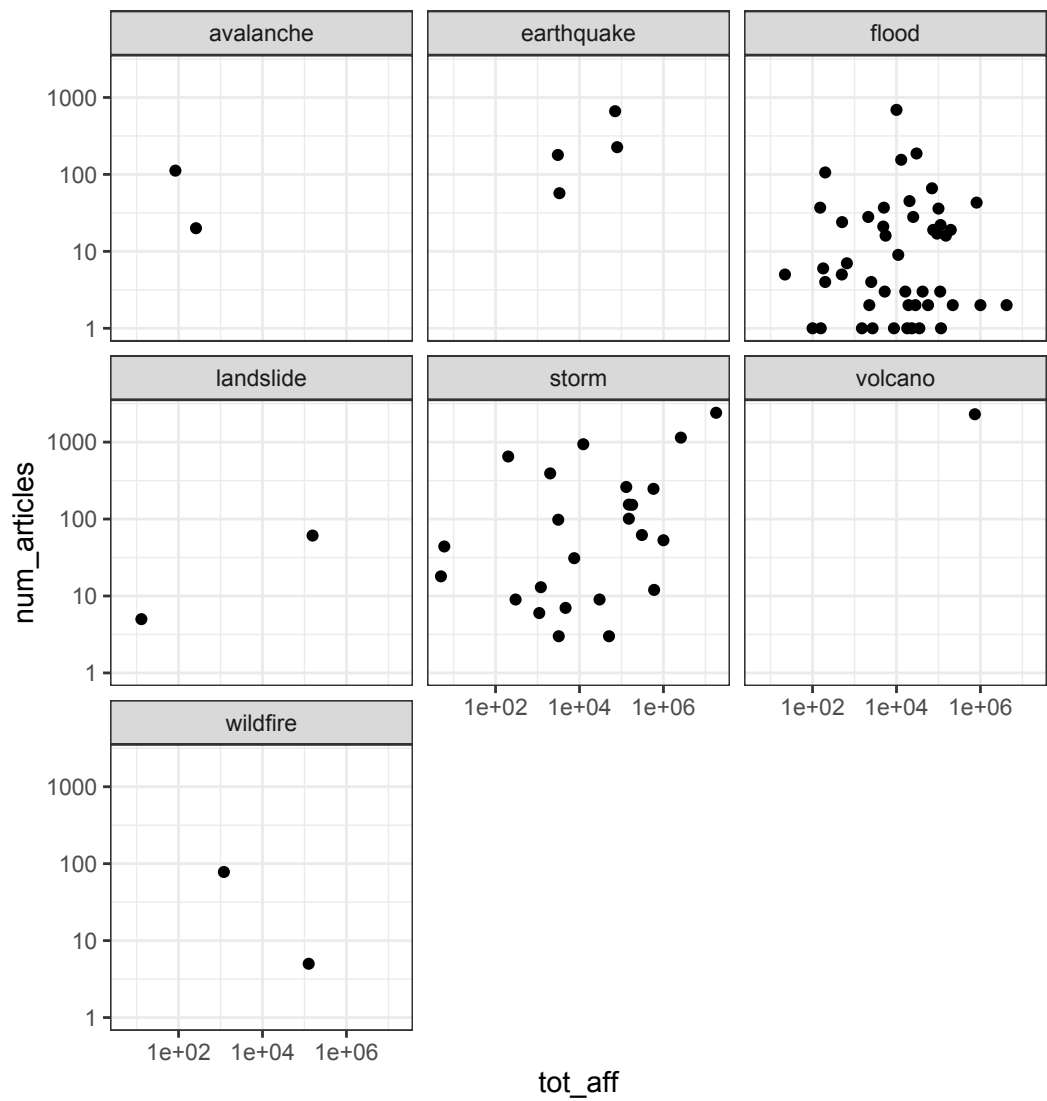
```
> ee %>% filter(!is.na(tot_de) & !is.na(tot_aff)) %>%
group_by(e_type) %>% summarise(n_events=n(),
n_articles=sum(num_articles), a_per_e=n_articles/n_events,
mean(tot_de),sd(tot_de),mean(tot_aff),sd(tot_aff))
`summarise()` ungrouping output (override with `.groups`
argument)
```

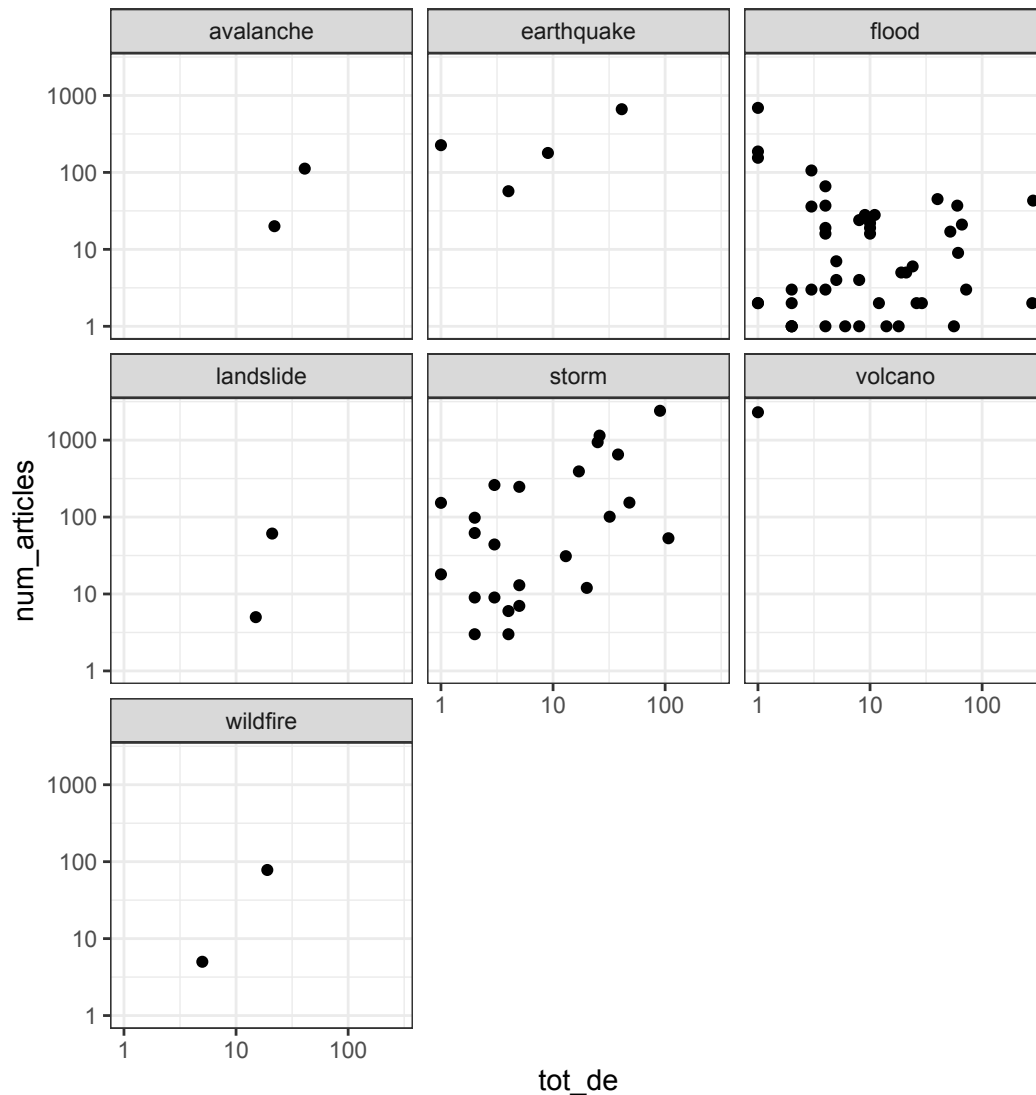
A tibble: 7 x 8

e_type	n_events	n_articles	a_per_e	`mean(tot_de)` `sd(tot_de)`	`mean(tot_aff)` `sd(tot_aff)`
<chr>	<int>	<int>	<dbl>	<dbl>	<dbl>
1 avalanche	2	132	66	31.5	
	172	124.			
2 earthquake	4	1125	281.	13.8	
	38981.	41471.			
3 flood	47	1687	35.9	27.1	
	162579.	629731.			
4 landslide	2	66	33	18	
	77932.	110193.			
5 storm	24	6814	284.	19.1	
	992157.	3665481.			
6 volcano	1	2302	2302	1	
NA	736802	NA			
7 wildfire	2	83	41.5	12	
	63600	88247.			

Does impact of the disaster affect coverage?

A little bit for storms. rho=.471 for deaths, rho=.371 for affecteds.





```
> ee%>%group_by(e_type) %>%
  summarise(n_events=n(),cor(tot_de,num_articles,method='spearman'
),cor(tot_aff,num_articles,method='spearman')) %>%
  filter(n_events>2)
`summarise()` ungrouping output (override with `.groups`
argument)
# A tibble: 3 x 4
  e_type      n_events `cor(tot_de, num_articles, method = "s...
`cor(tot_aff, num_articles, method = "...
  <chr>         <int>         <dbl>
<dbl>
1 earthqua...         4           0.4
0.6
2 flood         47          -0.0152
0.0194
3 storm        24           0.471
```

0.371

For all floods and storms, (geometric) mean number of articles per country where the event took place.

```
> ee %>% filter(e_type %in% c('flood','storm','earthquake')) %>%
group_by(e_type,country) %>%
summarise(n_events=n(),gm_arts=exp(mean(log(num_articles)))) %>%
arrange(-gm_arts) %>% print(n=99)
`summarise()` regrouping output by 'e_type' (override with
`.groups` argument)
# A tibble: 50 x 4
# Groups:   e_type [3]
  e_type      country n_events gm_arts
  <chr>      <chr>      <int>   <dbl>
1 storm      NPL                2 1659.
2 flood      GBR                1  690.
3 earthquake GBR                1  663
4 storm      GBR                1  650.
5 storm      PHL                2  482.
6 storm      ESP                1  392.
7 earthquake VAT                1  226.
8 earthquake SAU                1  179
9 storm      SLB                1  154.
10 storm     CAN                1  153.
11 flood     ZAF                1  106.
12 storm     SGP                1  101.
13 storm     FJI                1   98.
14 flood     CAN                2   83.5
15 flood     IND                2   72.4
16 flood     KGZ                1   66.0
17 earthquake NGA                1   57.
18 storm     PAK                1   53.
19 storm     NZL                1  44.0
20 flood     UGA                1   36
21 storm     AFG                1   31
22 storm     USA                5   27.9
23 flood     AUS                1   19.0
24 storm     CZE                1   18.0
25 flood     TUR                1   17
26 flood     KEN                3   16.0
27 flood     TJK                1   16.0
28 storm     CHN                1   12
29 flood     PAK                2   11.2
30 flood     USA                7   10.9
31 flood     SGP                1    9.
```

32	storm	VNM	1	7.00
33	storm	IND	1	6
34	flood	IRN	2	4.47
35	flood	IDN	3	3.85
36	flood	NGA	3	3.53
37	flood	RWA	2	3.
38	storm	NGA	1	3.
39	storm	THA	1	3.
40	flood	ISR	2	2.65
41	flood	CHN	1	2
42	flood	GHA	1	2
43	flood	IRL	2	2
44	flood	MWI	1	2
45	flood	AFG	1	1
46	flood	ARM	1	1
47	flood	HKG	1	1
48	flood	TWN	1	1
49	flood	YEM	1	1
50	flood	ZWE	1	1