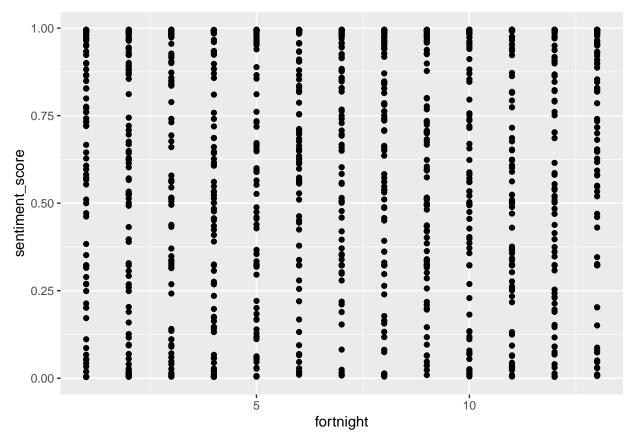
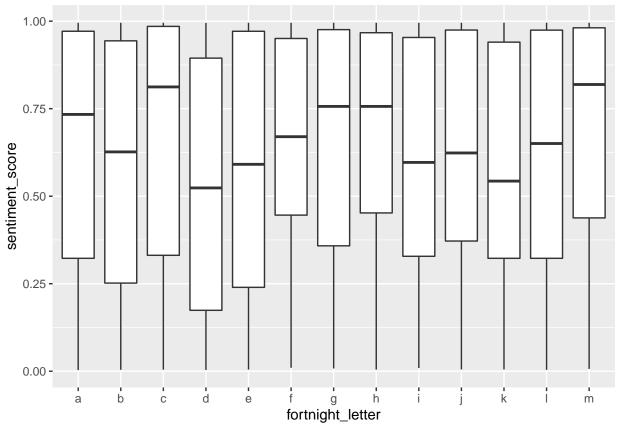
Datafest Data Analysis

```
#US analysis import
US_analysis = read_excel("C:\\Users\\gtham\\OneDrive - Pomona College\\A - DATAFEST\\Analysis Datasets\
US_analysis_music <- US_analysis %>%
  filter(video_category == "Music")
US_analysis_travel <- US_analysis %>%
  filter(video_category == "Travel and Events")
US_analysis_people <- US_analysis %>%
  filter(video_category == "People and Blogs")
US_analysis_entertainment <- US_analysis %>%
  filter(video_category == "Entertainment")
US analysis news <- US analysis %>%
  filter(video_category == "News and Politics")
US_analysis_how_to <- US_analysis %>%
  filter(video_category == "How-to and Style")
US_analysis_education <- US_analysis %>%
  filter(video_category == "Education")
US_analysis_science <- US_analysis %>%
  filter(video_category == "Science and Technology")
#full US data data summaries
ggplot(US_analysis) +
  geom_point(aes(x = fortnight, y = sentiment_score))
```



```
ggplot(US_analysis) +
geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

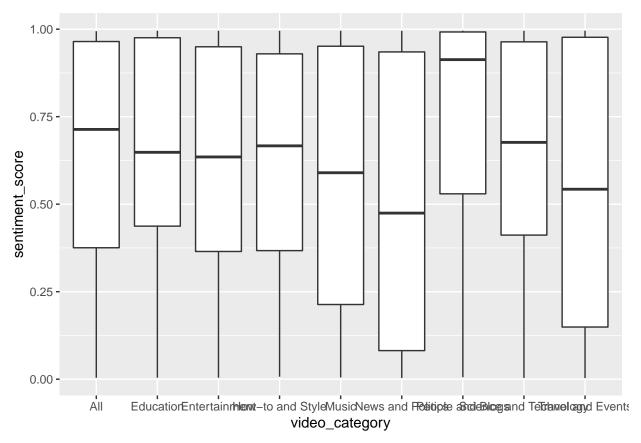


```
US_analysis %>%
  group_by(fortnight) %>%
  summarize(mean(sentiment_score))

## # A tibble: 13 x 2
## fortnight `mean(sentiment_score)`
## <dbl> <dbl>
```

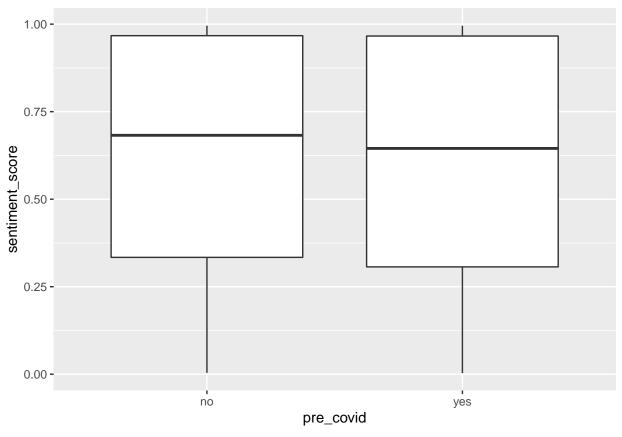
```
##
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.625
##
    2
               2
                                     0.579
               3
    3
                                     0.642
##
               4
##
    4
                                     0.518
               5
##
    5
                                     0.569
               6
##
    6
                                     0.642
##
    7
               7
                                     0.667
               8
    8
                                     0.669
##
##
    9
               9
                                     0.596
              10
## 10
                                     0.617
## 11
                                     0.562
              11
## 12
              12
                                     0.606
## 13
              13
                                     0.675
```

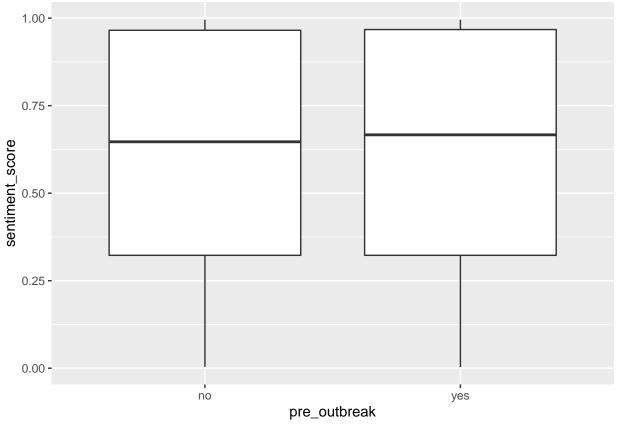
```
ggplot(US_analysis) +
  geom_boxplot(aes(x = video_category, y = sentiment_score))
```



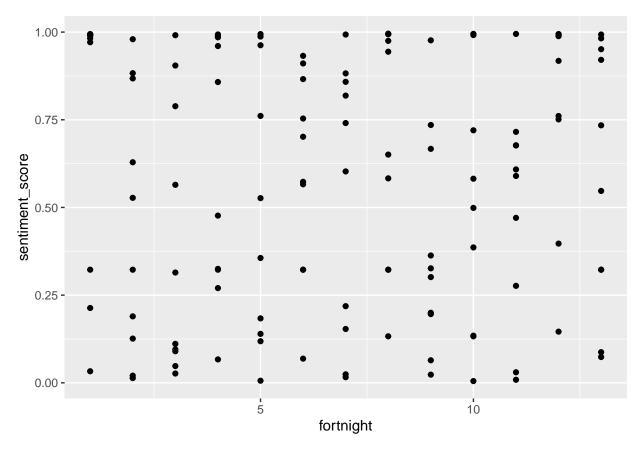
```
##
##
     <chr>
                                                <dbl>
## 1 All
                                                0.645
## 2 Education
                                                0.649
## 3 Entertainment
                                                0.612
## 4 How-to and Style
                                                0.623
## 5 Music
                                                0.564
## 6 News and Politics
                                                0.503
## 7 People and Blogs
                                                0.750
## 8 Science and Technology
                                                0.635
## 9 Travel and Events
                                                0.534
```

```
ggplot(US_analysis) +
geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```

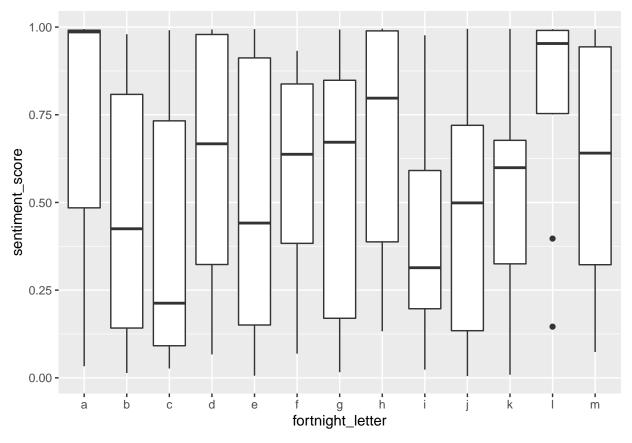




```
US_analysis %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
##
     pre_outbreak `mean(sentiment_score)`
##
     <chr>>
                                      <dbl>
## 1 no
                                      0.611
## 2 yes
                                      0.614
\# data \ summary \ and \ analysis \ for \ music \ dataset
ggplot(US_analysis_music) +
 geom_point(aes(x = fortnight, y = sentiment_score))
```



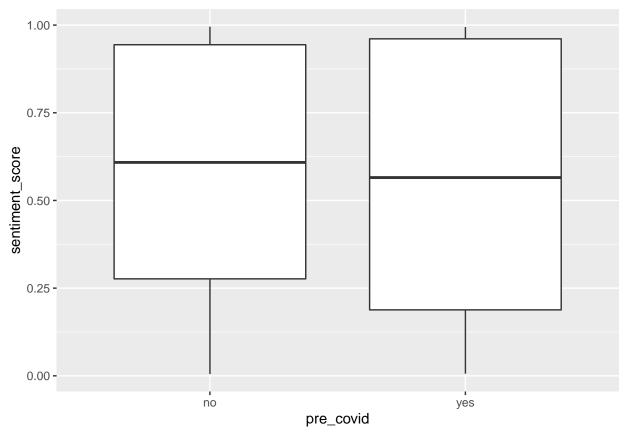
```
ggplot(US_analysis_music) +
geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

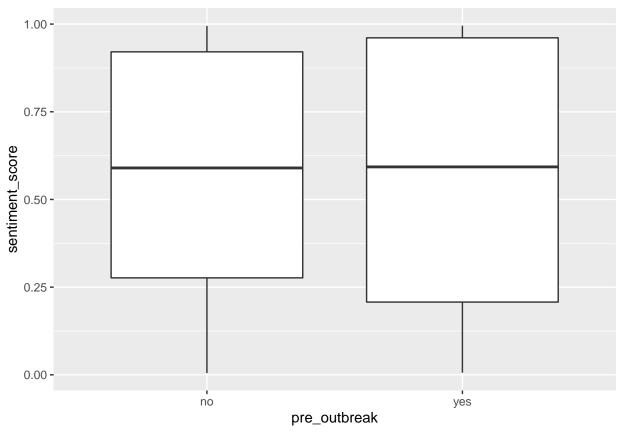


```
US_analysis_music %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
      fortnight `mean(sentiment_score)`
##
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.748
##
    2
               2
                                     0.456
               3
    3
                                     0.393
##
               4
##
                                     0.625
               5
##
    5
                                     0.504
               6
##
    6
                                     0.602
##
    7
               7
                                     0.531
               8
                                     0.691
##
    8
##
    9
               9
                                     0.385
              10
## 10
                                     0.494
                                     0.505
## 11
              11
## 12
              12
                                     0.793
## 13
              13
                                     0.593
```

```
ggplot(US_analysis_music) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```

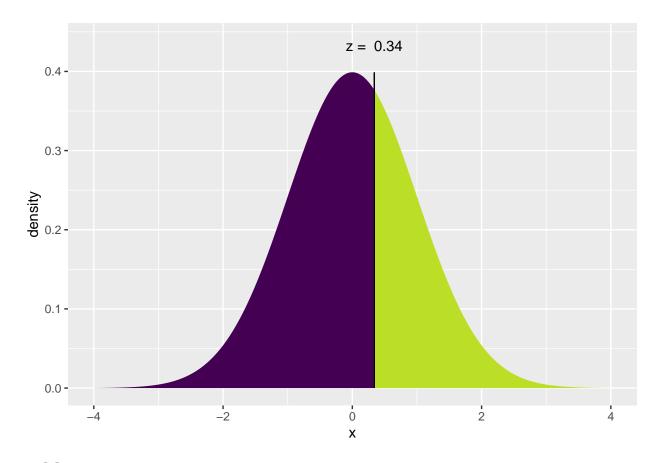




```
US_analysis_music %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>
                                     <dbl>
## 1 no
                                     0.555
                                     0.569
## 2 yes
\#two\ proportion\ z	ext{-}test\ for\ music\ dataset
#null hypothesis: the true proportion of positive sentiment music videos published precovid and postcov
count(US_analysis_music, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
     <1g1>
##
                           <int>
## 1 FALSE
                              69
## 2 TRUE
                              60
m_num_precovid = 60
m_num_postcovid = 69
m_num = 129
US_analysis_music %>%
 filter(pre_covid == "yes") %>%
```

count(sentiment_score > 0.5)

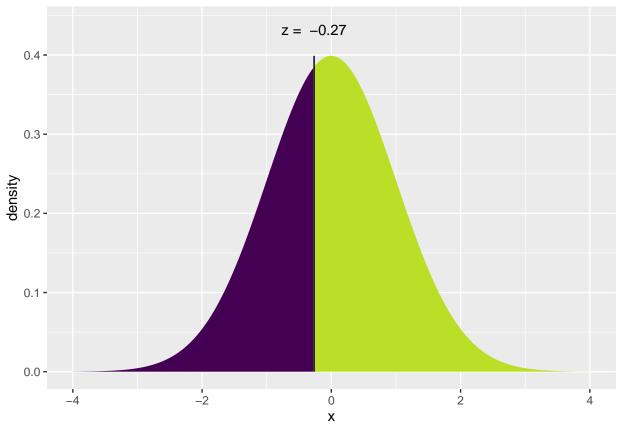
```
## # A tibble: 2 x 2
##
    `sentiment_score > 0.5`
##
     <1g1>
                             <int>
## 1 FALSE
                                27
## 2 TRUE
                                 33
p_hat_1_m_pos = 33/60
US_analysis_music %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
                                n
##
     <lg1>
                             <int>
## 1 FALSE
                                29
## 2 TRUE
                                40
p_hat_2_m_pos = 40/69
p_hat_m_pos = (33+40)/(60+69)
sd \leftarrow sqrt((((p_hat_m_pos)*(1-p_hat_m_pos))/60)+(((p_hat_m_pos)*(1-p_hat_m_pos))/69))
z_score <- ((p_hat_2_m_pos-p_hat_1_m_pos)-0)/sd</pre>
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 0.3396) = P(Z \le 0.3396) = 0.6329
## P(X > 0.3396) = P(Z > 0.3396) = 0.3671
##
```



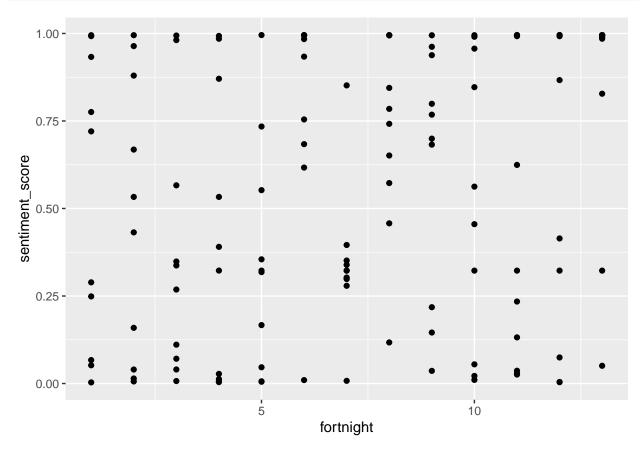
```
## [1] 0.7341715
#outbreak music
count(US_analysis_music, pre_outbreak == "yes")
## # A tibble: 2 x 2
     `pre_outbreak == "yes"`
##
     <1g1>
                              <int>
## 1 FALSE
                                49
## 2 TRUE
                                80
m_num_preoutbreak = 80
m_num_postoutbreak = 49
m_num = 129
US_analysis_music %>%
  filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                34
## 2 TRUE
                                46
p_hat_1_m_pos = 46/80
```

US_analysis_music %>%

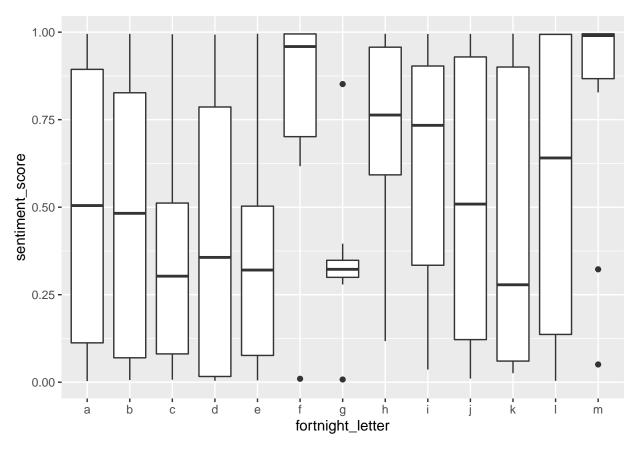
```
filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
     `sentiment_score > 0.5`
##
     <1g1>
                              <int>
## 1 FALSE
                                 22
## 2 TRUE
                                 27
p_hat_2_m_pos = 27/49
p_hat_m_pos = (46+27)/(80+49)
sd \leftarrow sqrt((((p_hat_m_pos)*(1-p_hat_m_pos))/80)+(((p_hat_m_pos)*(1-p_hat_m_pos))/49))
z_{score} \leftarrow ((p_{at_2_m_pos_p_hat_1_m_pos)-0})/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -0.2667) = P(Z \le -0.2667) = 0.3948
  P(X > -0.2667) = P(Z > -0.2667) = 0.6052
##
```



```
#data summary travel
ggplot(US_analysis_travel) +
geom_point(aes(x = fortnight, y = sentiment_score))
```



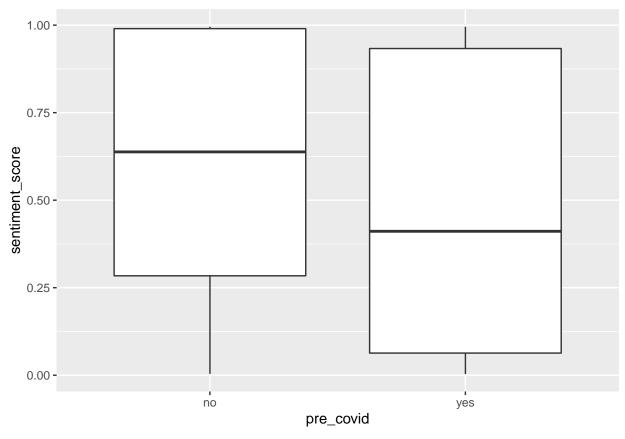
```
ggplot(US_analysis_travel) +
  geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

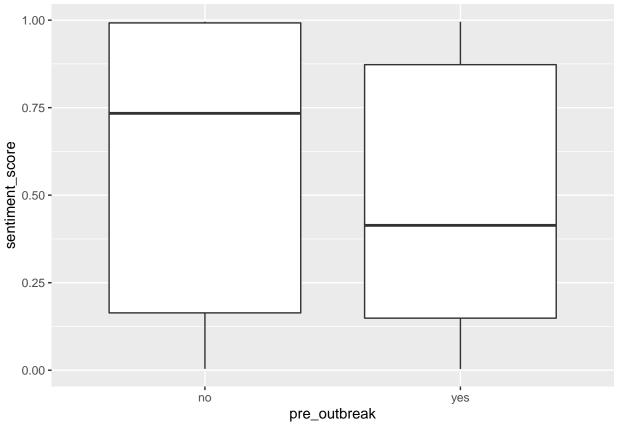


```
US_analysis_travel %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
##
      fortnight `mean(sentiment_score)`
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.508
##
    2
               2
                                     0.469
               3
                                     0.372
##
               4
##
                                     0.414
               5
##
    5
                                     0.350
               6
    6
                                     0.796
##
##
    7
               7
                                     0.347
               8
                                     0.715
##
    8
##
    9
               9
                                     0.624
              10
## 10
                                     0.522
                                     0.439
## 11
              11
              12
## 12
                                     0.566
## 13
              13
                                     0.815
```

```
ggplot(US_analysis_travel) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```



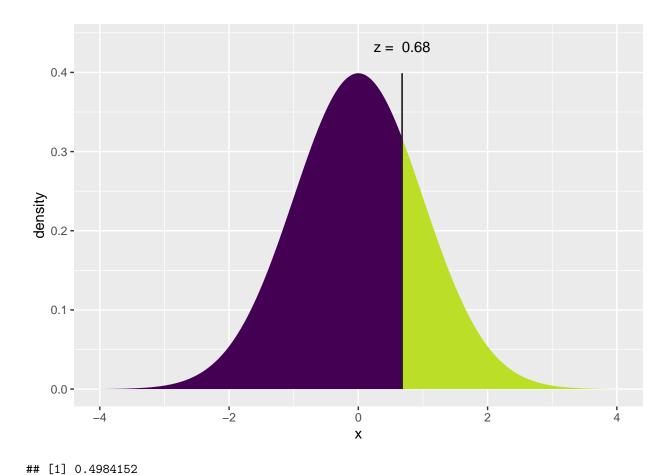


```
US_analysis_travel %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>>
                                     <dbl>
## 1 no
                                     0.593
## 2 yes
                                    0.497
#precovid travel
#null hypothesis: the true proportion of positive sentiment travel videos published precovid and postco
count(US_analysis_travel, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
                              n
     <1g1>
##
                          <int>
## 1 FALSE
                             70
## 2 TRUE
                             60
t_num_precovid = 60
t_num_postcovid = 70
t_num = 130
```

US_analysis_travel %>%

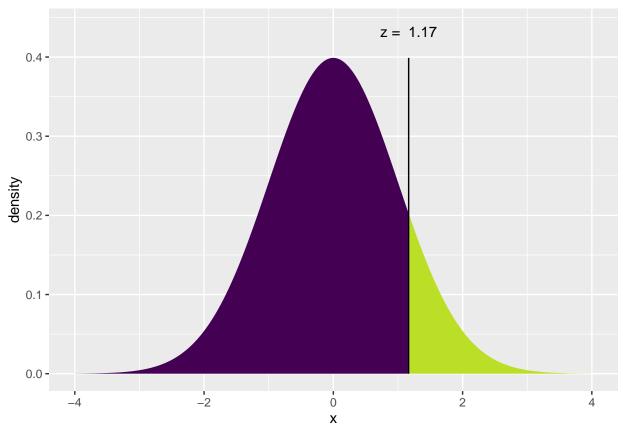
filter(pre_covid == "yes") %>%
count(sentiment_score > 0.5)

```
## # A tibble: 2 x 2
##
    `sentiment_score > 0.5`
     <1g1>
                             <int>
## 1 FALSE
                                31
## 2 TRUE
                                 29
p_hat_1_t_pos = 29/60
US_analysis_travel %>%
 filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
                                n
## <lgl>
                              <int>
## 1 FALSE
                                32
## 2 TRUE
                                38
p_hat_2_t_pos = 38/70
p_hat_t_pos = (29+38)/(60+70)
sd \leftarrow sqrt((((p_hat_t_pos)*(1-p_hat_t_pos))/60)+(((p_hat_t_pos)*(1-p_hat_t_pos))/70))
z_score <- ((p_hat_2_t_pos-p_hat_1_t_pos)-0)/sd</pre>
\#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 0.677) = P(Z \le 0.677) = 0.7508
## P(X > 0.677) = P(Z > 0.677) = 0.2492
##
```

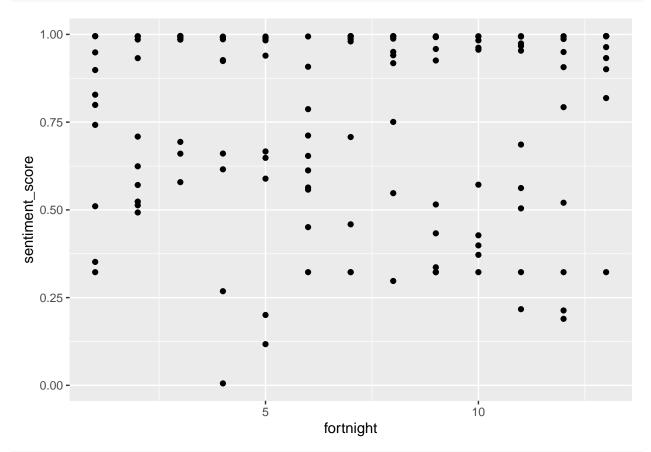


```
#outbreak travel
count(US_analysis_travel, pre_outbreak == "yes")
## # A tibble: 2 x 2
    `pre_outbreak == "yes"`
##
     <1g1>
                             <int>
## 1 FALSE
                                50
                                80
## 2 TRUE
t_num_preoutbreak = 80
t_num_postoutbreak = 50
t_num = 130
US_analysis_travel %>%
 filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                             <int>
## 1 FALSE
                                42
## 2 TRUE
                                38
p_hat_1_t_pos = 38/80
US_analysis_travel %>%
```

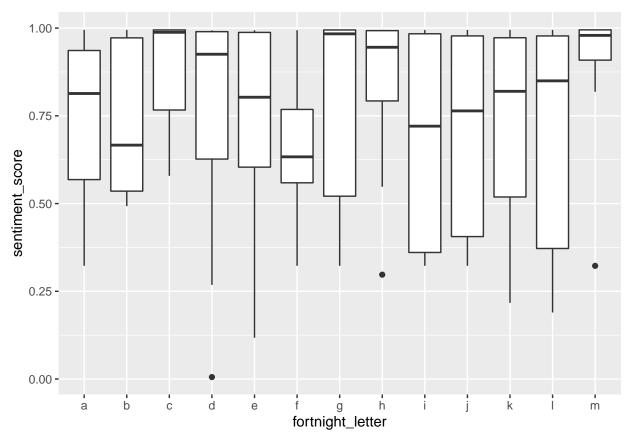
```
filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
    `sentiment_score > 0.5`
##
     <1g1>
                              <int>
## 1 FALSE
                                 21
## 2 TRUE
                                 29
p_hat_2_t_pos = 29/50
p_hat_t_pos = (38+29)/(80+50)
sd \leftarrow sqrt((((p_hat_t_pos)*(1-p_hat_t_pos))/80)+(((p_hat_t_pos)*(1-p_hat_t_pos))/50))
z_{score} \leftarrow ((p_{at_2_t_pos_p_hat_1_t_pos_0})/sd
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 1.165) = P(Z \le 1.165) = 0.8781
## P(X > 1.165) = P(Z > 1.165) = 0.1219
##
```



```
#data summary people and blogs
ggplot(US_analysis_people) +
  geom_point(aes(x = fortnight, y = sentiment_score))
```



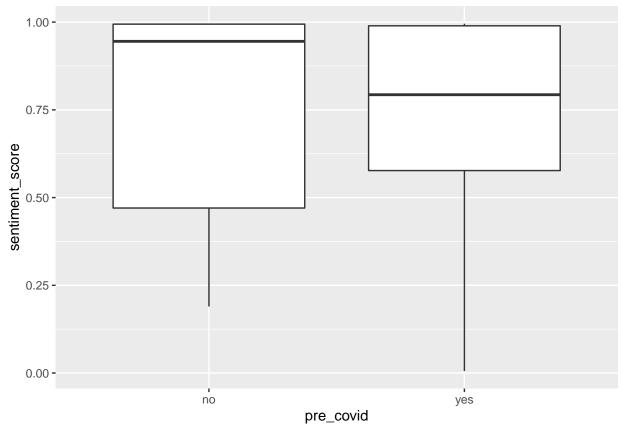
```
ggplot(US_analysis_people) +
  geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

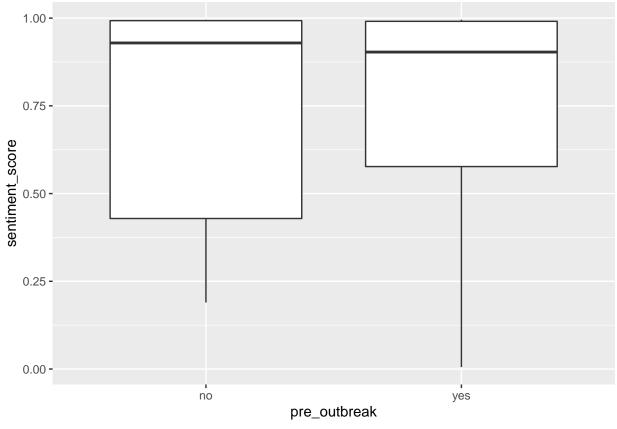


```
US_analysis_people %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
##
      fortnight `mean(sentiment_score)`
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.739
##
    2
               2
                                     0.734
               3
    3
                                     0.887
##
               4
##
                                     0.736
               5
##
    5
                                     0.712
               6
    6
                                     0.656
##
##
    7
               7
                                     0.776
               8
                                     0.838
##
    8
##
    9
               9
                                     0.679
              10
## 10
                                     0.698
## 11
                                     0.718
              11
## 12
              12
                                     0.687
## 13
              13
                                     0.891
```

```
ggplot(US_analysis_people) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```





```
US_analysis_people %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>>
                                     <dbl>
## 1 no
                                     0.735
                                     0.760
## 2 yes
#precovid people
count(US_analysis_people, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
##
                              n
##
     <1g1>
                          <int>
## 1 FALSE
                             70
## 2 TRUE
                             60
p_num_precovid = 60
p_num_postcovid = 70
p_num = 130
US_analysis_people %>%
  filter(pre_covid == "yes") %>%
  count(sentiment_score > 0.5)
```

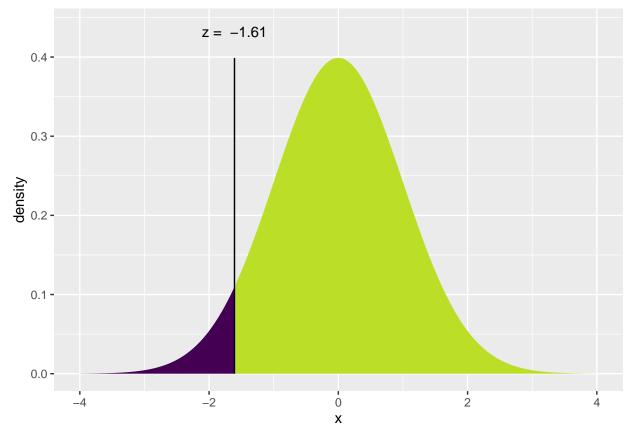
A tibble: 2 x 2

```
`sentiment_score > 0.5`
##
                                 n
##
     <1g1>
                              <int>
## 1 FALSE
                                 9
## 2 TRUE
                                 51
p_hat_1_p_s = 51/60
US_analysis_people %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
   `sentiment_score > 0.5`
##
   <lg1>
                              <int>
## 1 FALSE
                                 18
## 2 TRUE
                                 52
p_hat_2_p_s = 52/70
p_hat_p_s = (51+52)/(60+70)
sd \leftarrow sqrt((((p_hat_p_pos)*(1-p_hat_p_pos))/60)+(((p_hat_p_pos)*(1-p_hat_p_pos))/70))
z\_score \leftarrow ((p\_hat\_2\_p\_pos-p\_hat\_1\_p\_pos)-0)/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -1.501) = P(Z \le -1.501) = 0.06664
## P(X > -1.501) = P(Z > -1.501) = 0.9334
##
```

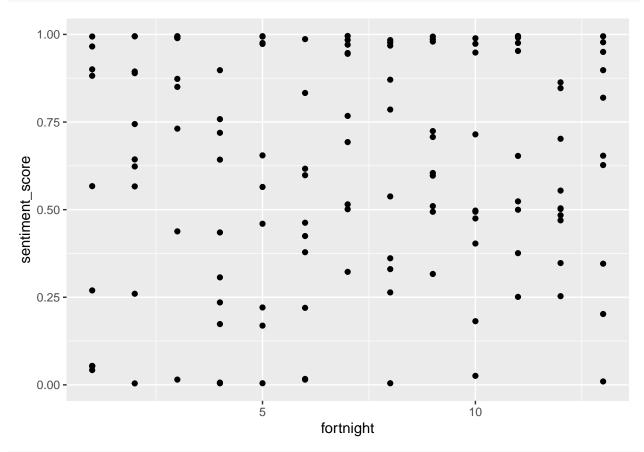


```
## [1] 0.1332855
#outbreak people
count(US_analysis_people, pre_outbreak == "yes")
## # A tibble: 2 x 2
     `pre_outbreak == "yes"`
     <1g1>
##
                             <int>
## 1 FALSE
                                50
## 2 TRUE
                                80
p_num_preoutbreak = 80
p_num_postoutbreak = 50
p_num = 130
US_analysis_people %>%
 filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                             <int>
## 1 FALSE
                                13
## 2 TRUE
                                67
p_hat_1_p_s = 67/80
US_analysis_people %>%
```

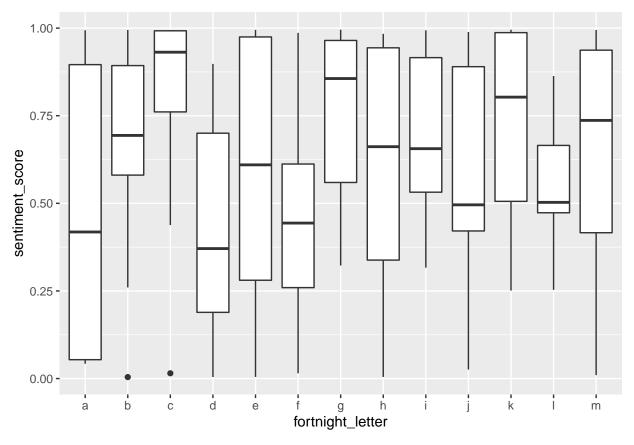
```
filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
     `sentiment_score > 0.5`
##
     <1g1>
                              <int>
## 1 FALSE
                                 14
## 2 TRUE
                                 36
p_hat_2_p_s = 36/50
p_hat_p_os = (67+36)/(80+50)
sd \leftarrow sqrt((((p_hat_p_pos)*(1-p_hat_p_pos))/80)+(((p_hat_p_pos)*(1-p_hat_p_pos))/50))
z_{score} \leftarrow ((p_{at_2p_pos_p_hat_1p_pos_0})/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
  P(X \le -1.607) = P(Z \le -1.607) = 0.05406
  P(X > -1.607) = P(Z > -1.607) = 0.9459
##
```



```
#data summary entertainment
ggplot(US_analysis_entertainment) +
  geom_point(aes(x = fortnight, y = sentiment_score))
```



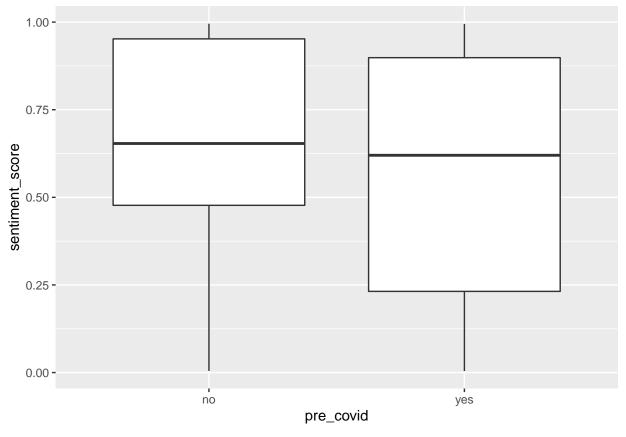
```
ggplot(US_analysis_entertainment) +
  geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

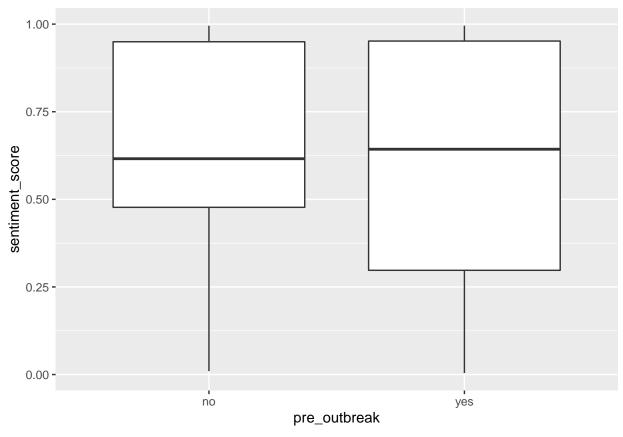


```
US_analysis_entertainment %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
      fortnight `mean(sentiment_score)`
##
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.478
##
    2
               2
                                     0.661
               3
                                     0.787
##
               4
##
                                     0.418
               5
##
    5
                                     0.601
               6
    6
                                     0.455
##
##
    7
               7
                                     0.764
               8
    8
                                     0.608
##
##
    9
               9
                                     0.691
              10
## 10
                                     0.570
## 11
              11
                                     0.721
              12
## 12
                                     0.553
## 13
              13
                                     0.648
```

```
ggplot(US_analysis_entertainment) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```

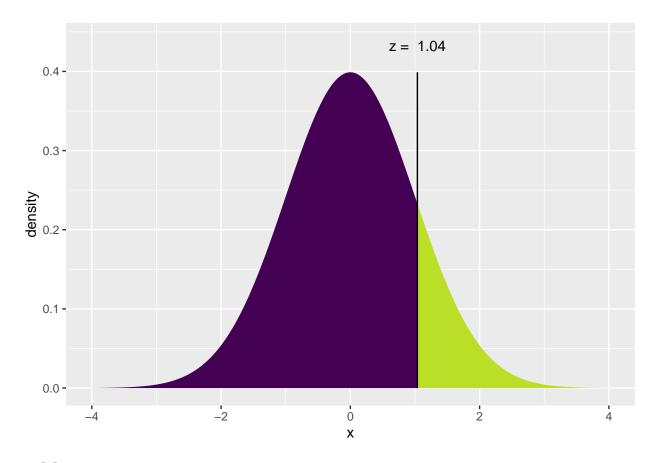




```
US_analysis_entertainment %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>
                                     <dbl>
## 1 no
                                     0.637
                                     0.597
## 2 yes
#precovid entertainment
count(US_analysis_entertainment, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
##
                              n
##
     <1g1>
                           <int>
## 1 FALSE
                             70
## 2 TRUE
                             60
num_precovid = 60
num_postcovid = 70
num = 130
US_analysis_entertainment %>%
  filter(pre_covid == "yes") %>%
  count(sentiment_score > 0.5)
```

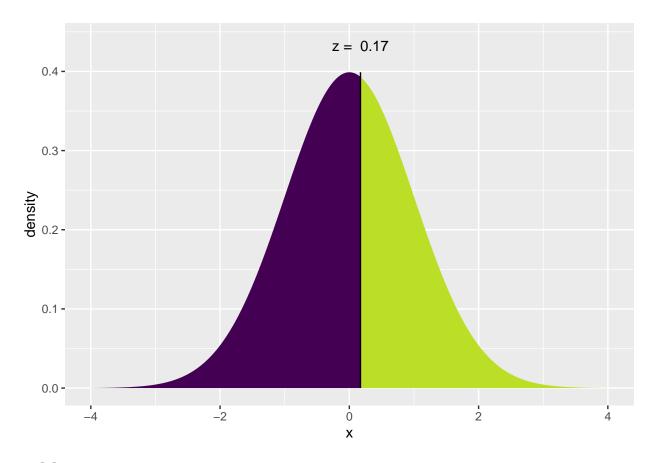
A tibble: 2 x 2

```
`sentiment_score > 0.5`
##
                                 n
##
     <1g1>
                              <int>
## 1 FALSE
                                 25
## 2 TRUE
                                 35
*proportion of positive sentiment videos precovid from sample
p_hat1 = 35/60
US_analysis_entertainment %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 23
## 2 TRUE
                                 47
{\it \#proportion of positive sentiment videos postcovid from sample}
p_hat2 = 47/70
p_hat = (35+47)/(60+70)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/60)+(((p_hat)*(1-p_hat))/70))
z_{score} \leftarrow ((p_{hat2-p_hat1})-0)/sd
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 1.038) = P(Z \le 1.038) = 0.8503
   P(X > 1.038) = P(Z > 1.038) = 0.1497
##
```

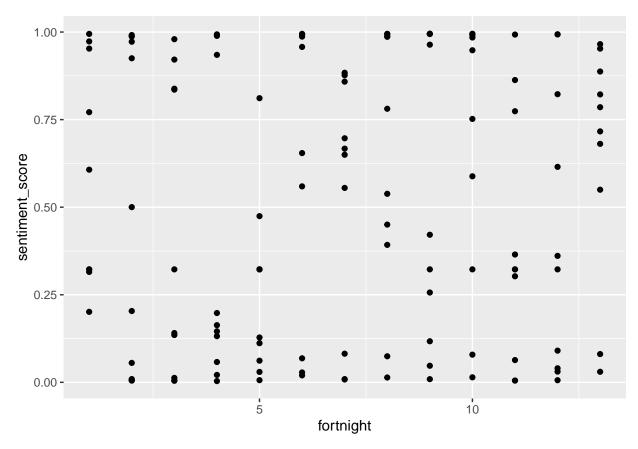


```
#outbreak entertainment
count(US_analysis_entertainment, pre_outbreak == "yes")
## # A tibble: 2 x 2
     `pre_outbreak == "yes"`
     <1g1>
##
                              <int>
## 1 FALSE
                                 50
## 2 TRUE
                                 80
num_preoutbreak = 80
num_postoutbreak = 50
num = 130
US_analysis_entertainment %>%
  filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 30
## 2 TRUE
                                 50
{\it \#proportion of positive sentiment videos\ preoutbreak\ from\ sample}
p_hat1 = 50/80
```

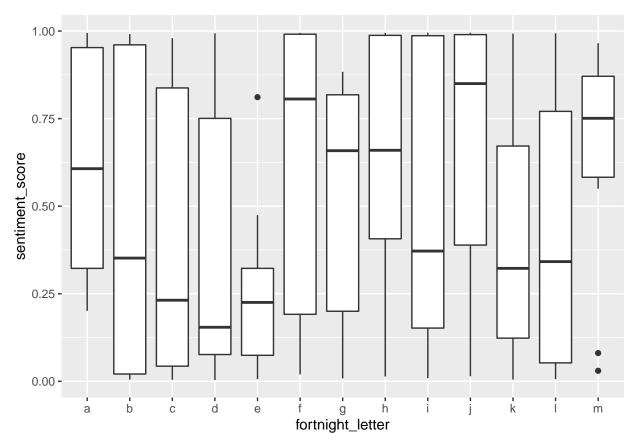
```
US_analysis_entertainment %>%
 filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
   `sentiment_score > 0.5`
##
     <lg1>
                              <int>
## 1 FALSE
                                 18
## 2 TRUE
                                 32
*proportion of positive sentiment videos postoutbreak from sample
p_hat2 = 32/50
p_hat = (50+32)/(80+50)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/80)+(((p_hat)*(1-p_hat))/50))
z\_score \leftarrow ((p\_hat2-p\_hat1)-0)/sd
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 0.1724) = P(Z \le 0.1724) = 0.5684
## P(X > 0.1724) = P(Z > 0.1724) = 0.4316
##
```



```
#data summary news and politics
ggplot(US_analysis_news) +
geom_point(aes(x = fortnight, y = sentiment_score))
```



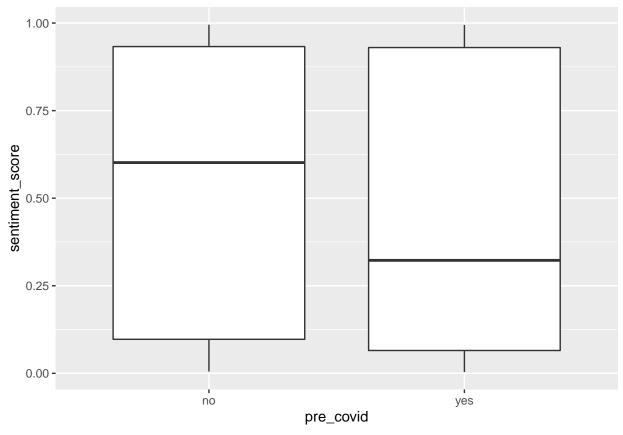
```
ggplot(US_analysis_news) +
geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

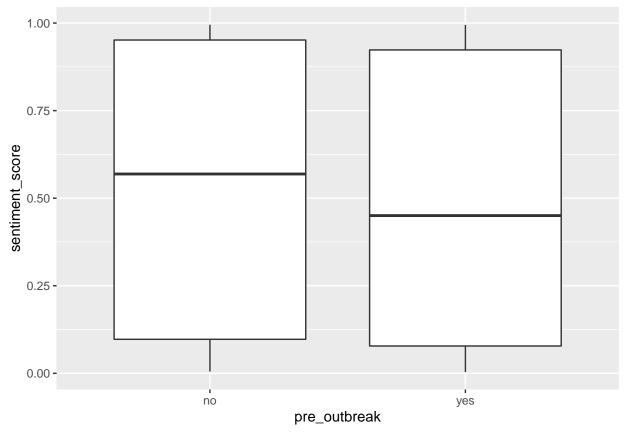


```
US_analysis_news %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
      fortnight `mean(sentiment_score)`
##
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.607
##
    2
               2
                                     0.466
               3
                                     0.420
##
               4
##
                                     0.364
               5
                                     0.259
##
    5
               6
    6
                                     0.626
##
##
    7
               7
                                     0.529
               8
                                     0.621
##
    8
##
    9
               9
                                     0.512
              10
## 10
                                     0.667
                                     0.402
## 11
              11
              12
## 12
                                     0.428
## 13
              13
                                     0.647
```

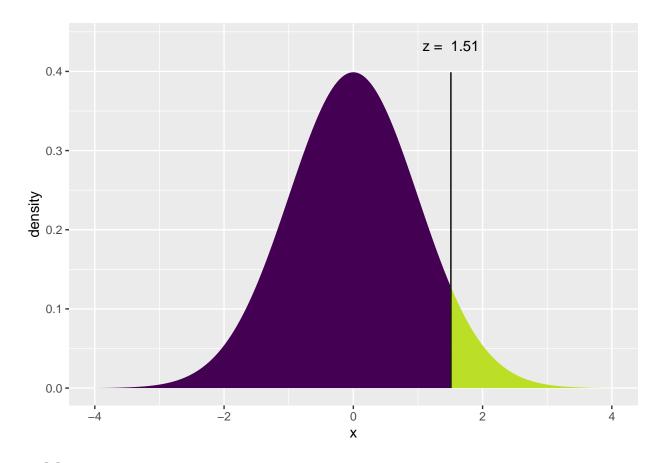
```
ggplot(US_analysis_news) +
geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```





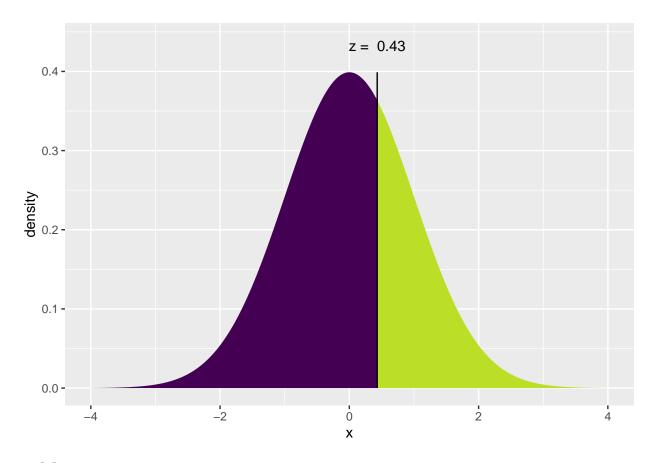
```
US_analysis_news %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
                                     <dbl>
##
     <chr>>
## 1 no
                                     0.531
## 2 yes
                                     0.485
#precovid news
count(US_analysis_news, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
##
                              n
##
     <1g1>
                          <int>
## 1 FALSE
                             70
## 2 TRUE
                             59
num_precovid = 59
num_postcovid = 70
num = 129
US_analysis_news %>%
  filter(pre_covid == "yes") %>%
  count(sentiment_score > 0.5)
```

```
`sentiment_score > 0.5`
##
                                 n
##
     <1g1>
                              <int>
## 1 FALSE
                                 34
## 2 TRUE
                                 25
*proportion of positive sentiment videos precovid from sample
p_hat1 = 25/59
US_analysis_news %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 31
## 2 TRUE
                                 39
{\it \#proportion of positive sentiment videos postcovid from sample}
p_hat2 = 39/70
p_hat = (25+39)/(59+70)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/59)+(((p_hat)*(1-p_hat))/70))
z_{score} \leftarrow ((p_{hat2-p_hat1})-0)/sd
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 1.51) = P(Z \le 1.51) = 0.9345
   P(X > 1.51) = P(Z > 1.51) = 0.06554
##
```

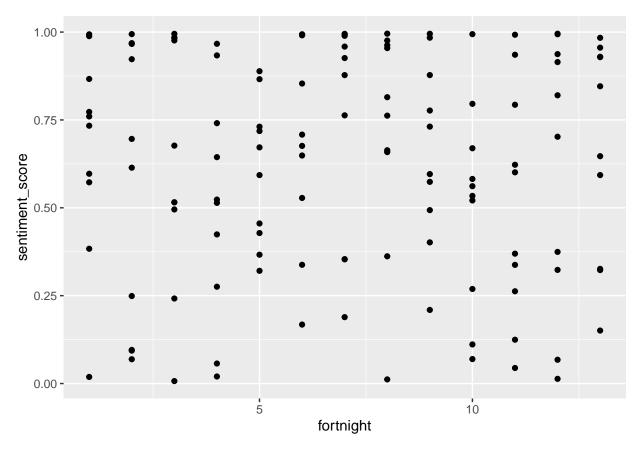


```
## [1] 0.1310897
#outbreak news
count(US_analysis_news, pre_outbreak == "yes")
## # A tibble: 2 x 2
    `pre_outbreak == "yes"`
##
     <1g1>
                              <int>
## 1 FALSE
                                 50
## 2 TRUE
                                 79
num_preoutbreak = 79
num_postoutbreak = 50
num = 129
US_analysis_news %>%
 filter(pre_outbreak == "yes") %>%
 count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
    `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 41
## 2 TRUE
                                 38
{\it \#proportion of positive sentiment videos\ preoutbreak\ from\ sample}
p_hat1 = 38/79
```

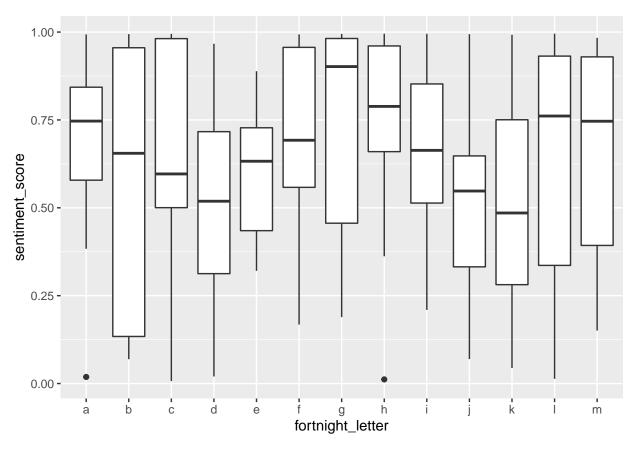
```
US_analysis_news %>%
 filter(pre_outbreak == "no") %>%
 count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
   `sentiment_score > 0.5`
##
     <lg1>
                              <int>
## 1 FALSE
                                 24
## 2 TRUE
                                 26
*proportion of positive sentiment videos postoutbreak from sample
p_hat2 = 26/50
p_hat = (38+26)/(79+50)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/79)+(((p_hat)*(1-p_hat))/50))
z\_score \leftarrow ((p\_hat2-p\_hat1)-0)/sd
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 0.4315) = P(Z \le 0.4315) = 0.6669
## P(X > 0.4315) = P(Z > 0.4315) = 0.3331
##
```



```
#data summary how-to and style
ggplot(US_analysis_how_to) +
  geom_point(aes(x = fortnight, y = sentiment_score))
```



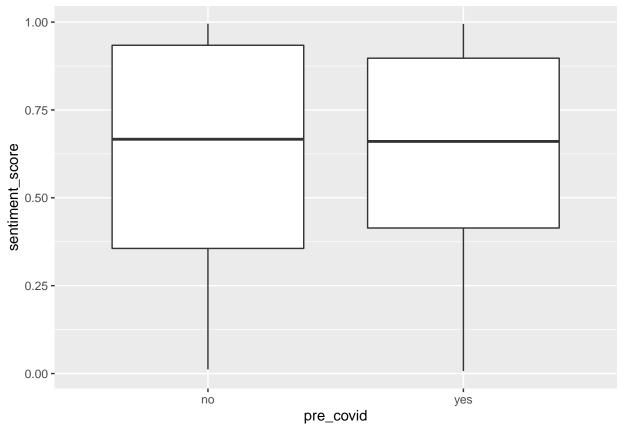
```
ggplot(US_analysis_how_to) +
geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

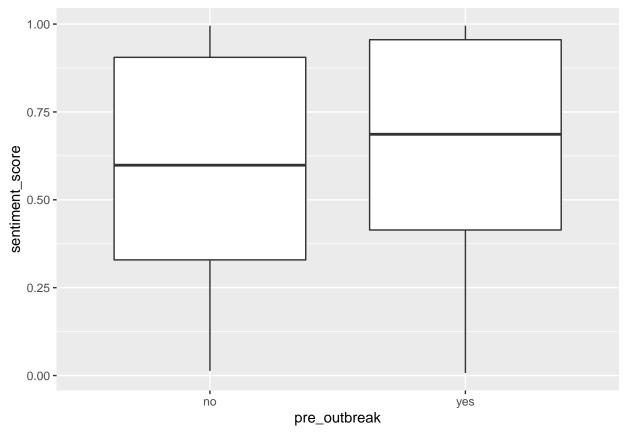


```
US_analysis_how_to %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
##
      fortnight `mean(sentiment_score)`
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.669
##
    2
               2
                                     0.567
               3
    3
                                     0.639
##
               4
##
    4
                                     0.510
               5
##
    5
                                     0.604
               6
##
    6
                                     0.690
##
    7
               7
                                     0.740
               8
                                     0.716
    8
##
##
    9
               9
                                     0.664
              10
## 10
                                     0.511
## 11
              11
                                     0.508
              12
## 12
                                     0.614
## 13
              13
                                     0.668
```

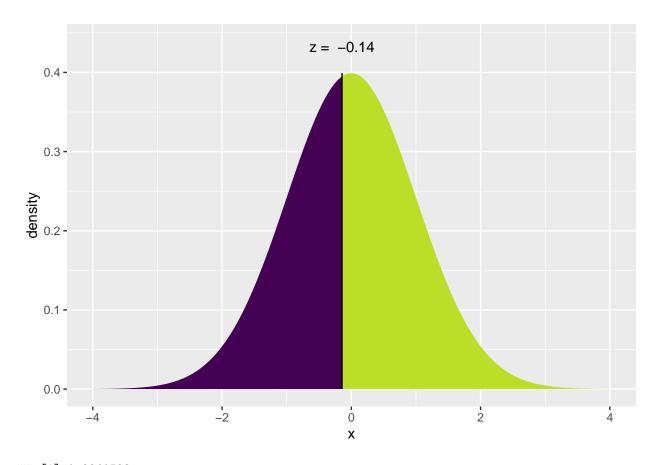
```
ggplot(US_analysis_how_to) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```





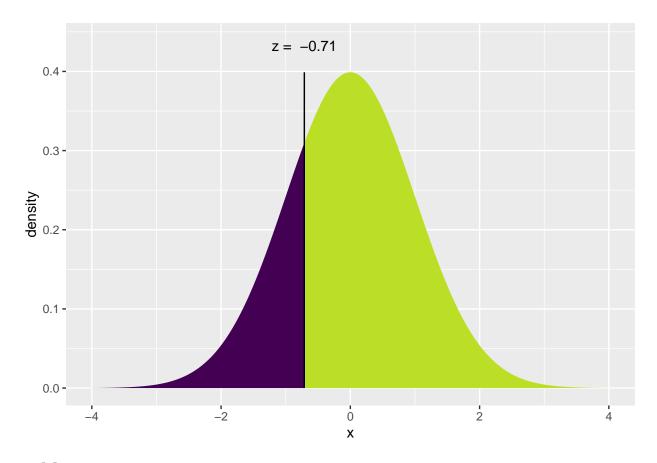
```
US_analysis_how_to %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>>
                                     <dbl>
## 1 no
                                     0.593
## 2 yes
                                     0.642
#precovid how-to
count(US_analysis_how_to, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
##
                              n
##
     <1g1>
                           <int>
## 1 FALSE
                             70
## 2 TRUE
                             60
num_precovid = 60
num_postcovid = 70
num = 130
US_analysis_how_to %>%
  filter(pre_covid == "yes") %>%
  count(sentiment_score > 0.5)
```

```
`sentiment_score > 0.5`
##
                                n
##
     <1g1>
                              <int>
## 1 FALSE
                                 19
## 2 TRUE
                                 41
*proportion of positive sentiment videos precovid from sample
p_hat1 = 41/60
US_analysis_how_to %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 23
## 2 TRUE
                                 47
{\it \#proportion of positive sentiment videos postcovid from sample}
p_hat2 = 47/70
p_hat = (41+47)/(60+70)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/60)+(((p_hat)*(1-p_hat))/70))
z_score \leftarrow ((p_hat2-p_hat1)-0)/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -0.1447) = P(Z \le -0.1447) = 0.4425
   P(X > -0.1447) = P(Z > -0.1447) = 0.5575
##
```

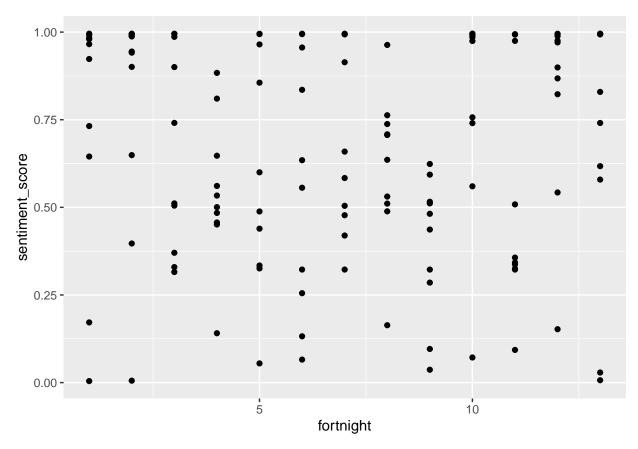


```
## [1] 0.8849523
#outbreak how-to
count(US_analysis_how_to, pre_outbreak == "yes")
## # A tibble: 2 x 2
     `pre_outbreak == "yes"`
     <1g1>
##
                              <int>
## 1 FALSE
                                 50
## 2 TRUE
                                 80
num_preoutbreak = 80
num_postoutbreak = 50
num = 130
US_analysis_how_to %>%
  filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 24
## 2 TRUE
                                 56
{\it \#proportion of positive sentiment videos\ preoutbreak\ from\ sample}
p_hat1 = 56/80
```

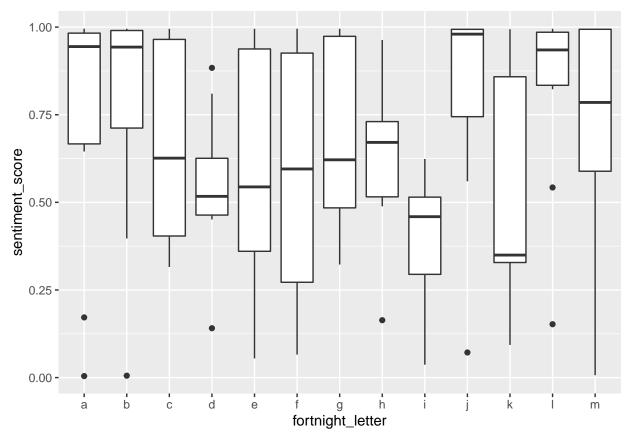
```
US_analysis_how_to %>%
  filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
   `sentiment_score > 0.5`
##
     <1g1>
                              <int>
## 1 FALSE
                                 18
## 2 TRUE
                                 32
*proportion of positive sentiment videos postoutbreak from sample
p_hat2 = 32/50
p_hat = (56+32)/(80+50)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/80)+(((p_hat)*(1-p_hat))/50))
z\_score \leftarrow ((p\_hat2-p\_hat1)-0)/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -0.7117) = P(Z \le -0.7117) = 0.2383
## P(X > -0.7117) = P(Z > -0.7117) = 0.7617
##
```



```
#data summary education
ggplot(US_analysis_education) +
  geom_point(aes(x = fortnight, y = sentiment_score))
```



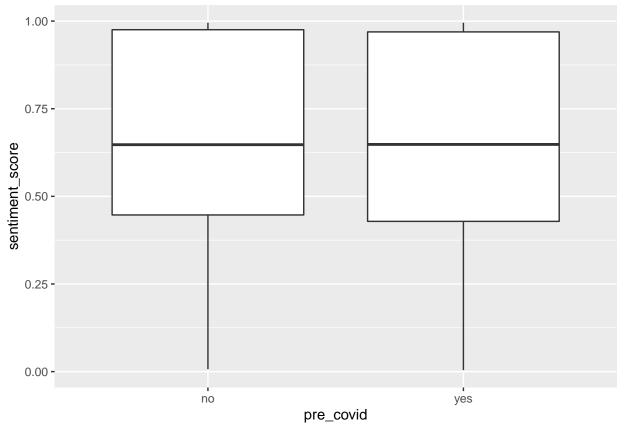
```
ggplot(US_analysis_education) +
geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

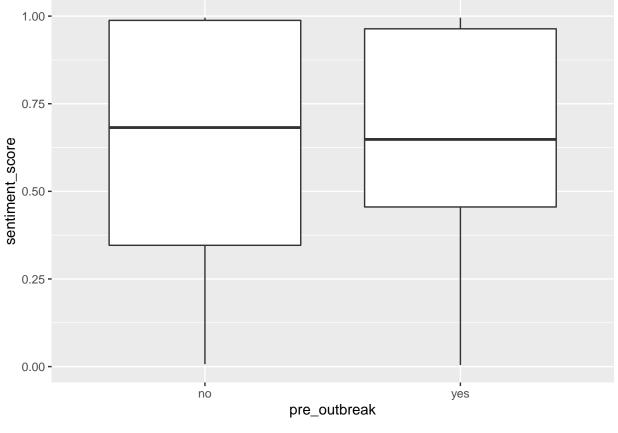


```
US_analysis_education %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
##
      fortnight `mean(sentiment_score)`
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.739
##
    2
               2
                                     0.781
               3
                                     0.665
##
               4
##
                                     0.547
               5
##
    5
                                     0.605
               6
    6
                                     0.575
##
##
    7
               7
                                     0.686
               8
                                     0.621
##
    8
##
    9
               9
                                     0.390
              10
## 10
                                     0.806
                                     0.525
## 11
              11
## 12
              12
                                     0.821
## 13
              13
                                     0.678
```

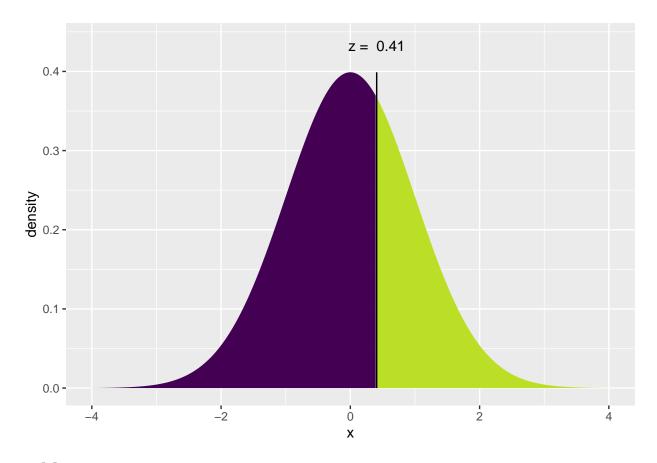
```
ggplot(US_analysis_education) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```





```
US_analysis_education %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>
                                     <dbl>
## 1 no
                                     0.644
## 2 yes
                                    0.652
#precovid education
count(US_analysis_education, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
##
                              n
##
     <1g1>
                          <int>
## 1 FALSE
                             70
## 2 TRUE
                             60
num_precovid = 60
num_postcovid = 70
num = 130
US_analysis_education %>%
  filter(pre_covid == "yes") %>%
  count(sentiment_score > 0.5)
```

```
`sentiment_score > 0.5`
##
                                 n
##
     <1g1>
                              <int>
## 1 FALSE
                                 20
## 2 TRUE
                                 40
*proportion of positive sentiment videos precovid from sample
p_hat1 = 40/60
US_analysis_education %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 21
## 2 TRUE
                                 49
{\it \#proportion of positive sentiment videos postcovid from sample}
p_hat2 = 49/70
p_hat = (40+49)/(60+70)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/60)+(((p_hat)*(1-p_hat))/70))
z_{score} \leftarrow ((p_{hat2-p_hat1})-0)/sd
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 0.4077) = P(Z \le 0.4077) = 0.6583
   P(X > 0.4077) = P(Z > 0.4077) = 0.3417
##
```

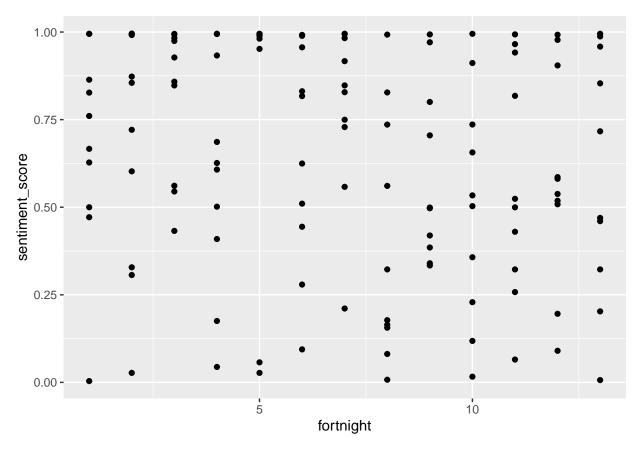


```
#outbreak education
count(US_analysis_education, pre_outbreak == "yes")
## # A tibble: 2 x 2
     `pre_outbreak == "yes"`
##
     <1g1>
                              <int>
## 1 FALSE
                                 50
## 2 TRUE
                                 80
num_preoutbreak = 80
num_postoutbreak = 50
num = 130
US_analysis_education %>%
  filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 25
## 2 TRUE
                                 55
{\it \#proportion of positive sentiment videos\ preoutbreak\ from\ sample}
p_hat1 = 55/80
```

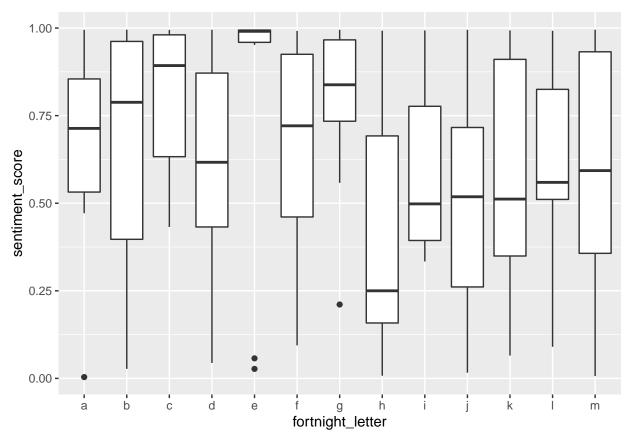
```
US_analysis_education %>%
 filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
   `sentiment_score > 0.5`
##
     <lg1>
                              <int>
## 1 FALSE
                                 16
## 2 TRUE
                                 34
*proportion of positive sentiment videos postoutbreak from sample
p_hat2 = 34/50
p_hat = (55+34)/(80+50)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/80)+(((p_hat)*(1-p_hat))/50))
z\_score \leftarrow ((p\_hat2-p\_hat1)-0)/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -0.08953) = P(Z \le -0.08953) = 0.4643
## P(X > -0.08953) = P(Z > -0.08953) = 0.5357
##
```



```
#data summary science and technology
ggplot(US_analysis_science) +
  geom_point(aes(x = fortnight, y = sentiment_score))
```



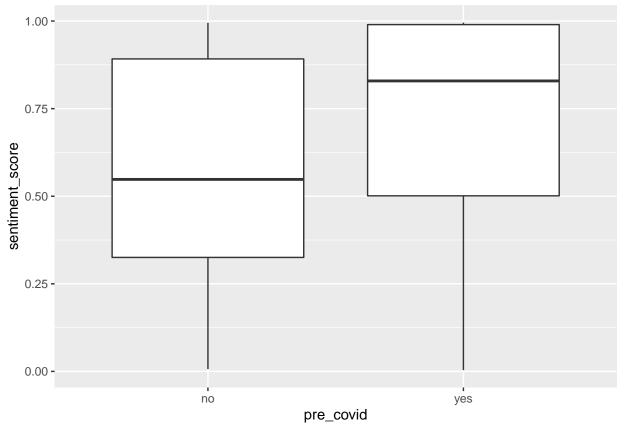
```
ggplot(US_analysis_science) +
geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

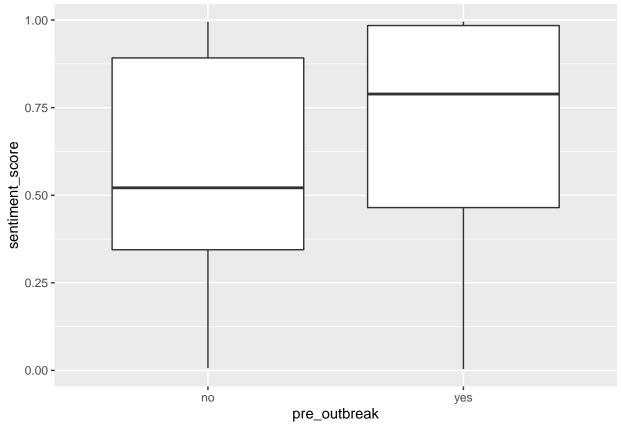


```
US_analysis_science %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
##
      fortnight `mean(sentiment_score)`
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.671
##
    2
               2
                                     0.670
               3
    3
                                     0.812
##
               4
##
                                     0.597
               5
##
    5
                                     0.798
               6
    6
                                     0.654
##
##
    7
               7
                                     0.781
               8
                                     0.403
##
    8
##
    9
               9
                                     0.594
              10
## 10
                                     0.506
                                     0.582
## 11
              11
## 12
              12
                                     0.589
## 13
              13
                                     0.597
```

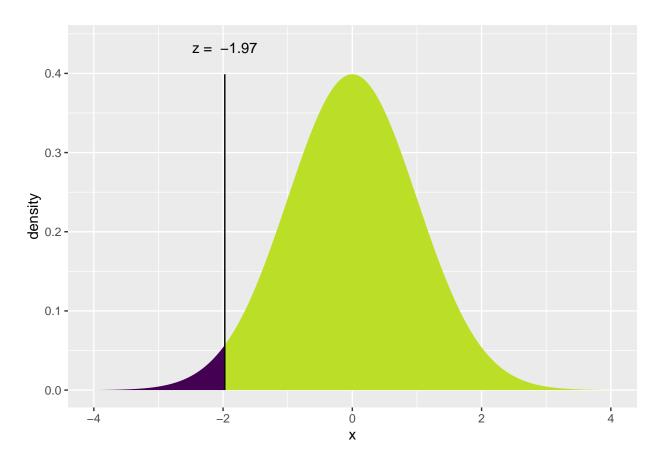
```
ggplot(US_analysis_science) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```





```
US_analysis_science %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>>
                                     <dbl>
## 1 no
                                     0.574
## 2 yes
                                    0.673
#precovid scitech
count(US_analysis_science, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
##
                              n
##
     <1g1>
                          <int>
## 1 FALSE
                             70
## 2 TRUE
                             60
num_precovid = 60
num_postcovid = 70
num = 130
US_analysis_science %>%
  filter(pre_covid == "yes") %>%
  count(sentiment_score > 0.5)
```

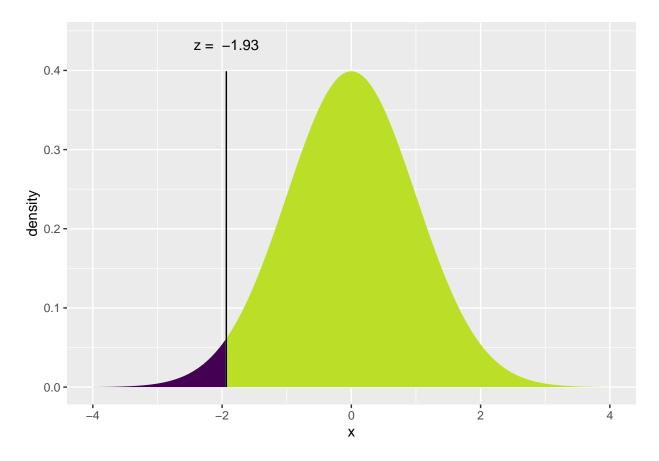
```
`sentiment_score > 0.5`
##
                                 n
##
     <1g1>
                              <int>
## 1 FALSE
                                 15
## 2 TRUE
                                 45
*proportion of positive sentiment videos precovid from sample
p_hat1 = 45/60
US_analysis_science %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 29
## 2 TRUE
                                 41
{\it \#proportion of positive sentiment videos postcovid from sample}
p_hat2 = 41/70
p_hat = (45+41)/(60+70)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/60)+(((p_hat)*(1-p_hat))/70))
z_score \leftarrow ((p_hat2-p_hat1)-0)/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -1.973) = P(Z \le -1.973) = 0.02422
   P(X > -1.973) = P(Z > -1.973) = 0.9758
##
```



```
## [1] 0.0484472
```

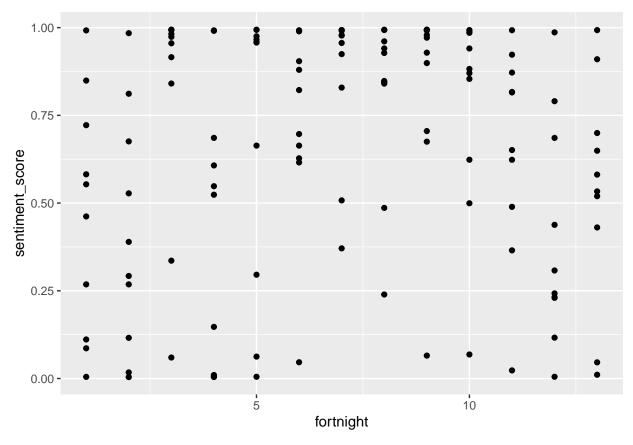
```
#outbreak scitech
count(US_analysis_science, pre_outbreak == "yes")
## # A tibble: 2 x 2
     `pre_outbreak == "yes"`
     <1g1>
##
                              <int>
## 1 FALSE
                                 50
## 2 TRUE
                                 80
num_preoutbreak = 80
num_postoutbreak = 50
num = 130
US_analysis_science %>%
 filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 22
## 2 TRUE
                                 58
{\it \#proportion of positive sentiment videos\ preoutbreak\ from\ sample}
p_hat1 = 58/80
```

```
US_analysis_science %>%
  filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
   `sentiment_score > 0.5`
##
##
     <1g1>
                              <int>
## 1 FALSE
                                 22
## 2 TRUE
                                 28
*proportion of positive sentiment videos postoutbreak from sample
p_hat2 = 28/50
p_hat = (58+28)/(80+50)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/80)+(((p_hat)*(1-p_hat))/50))
z\_score \leftarrow ((p\_hat2-p\_hat1)-0)/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -1.934) = P(Z \le -1.934) = 0.02654
## P(X > -1.934) = P(Z > -1.934) = 0.9735
##
```

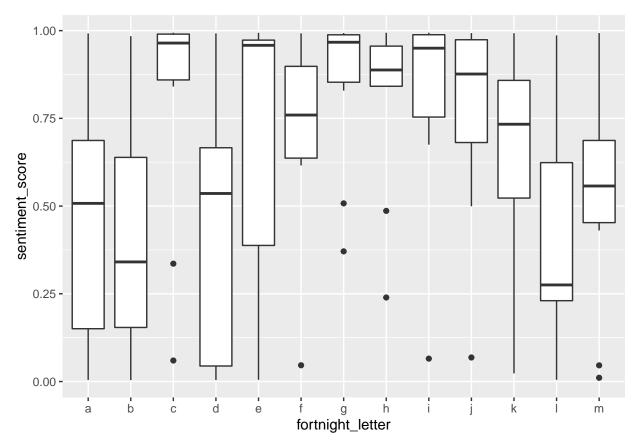


```
#Youtube API All Categories
US_analysis_all <- US_analysis %>%
  filter(video_category == "All")

#data summary all categories
ggplot(US_analysis_all) +
  geom_point(aes(x = fortnight, y = sentiment_score))
```



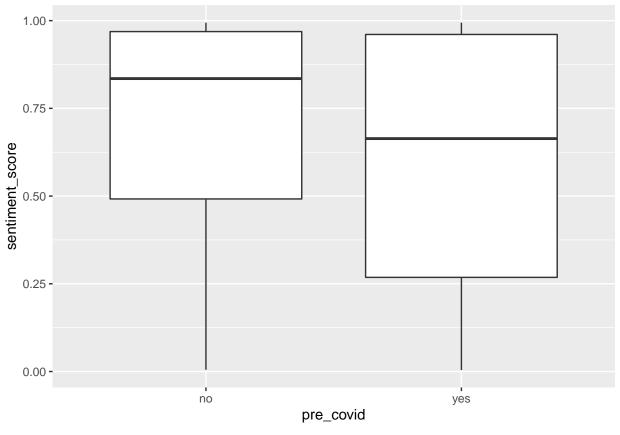
```
ggplot(US_analysis_all) +
geom_boxplot(aes(x = fortnight_letter, y = sentiment_score))
```

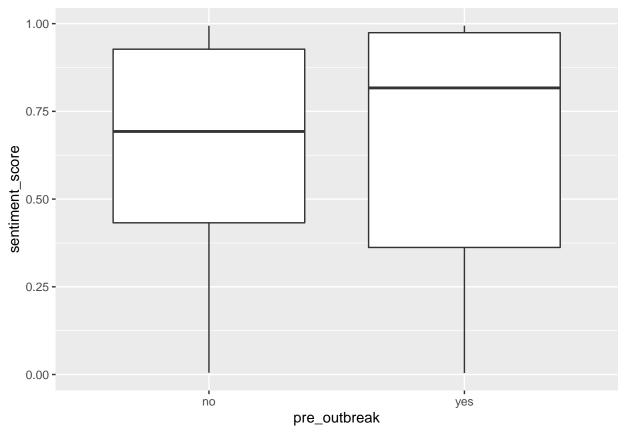


```
US_analysis_all %>%
group_by(fortnight) %>%
summarize(mean(sentiment_score))
```

```
## # A tibble: 13 x 2
##
      fortnight `mean(sentiment_score)`
           <dbl>
##
                                     <dbl>
##
    1
               1
                                     0.463
##
    2
               2
                                     0.409
               3
    3
                                     0.804
##
               4
##
                                     0.451
               5
##
    5
                                     0.687
               6
    6
                                     0.724
##
##
    7
               7
                                     0.852
               8
                                     0.808
##
    8
##
    9
               9
                                     0.820
## 10
              10
                                     0.771
                                     0.657
## 11
              11
## 12
              12
                                     0.403
## 13
              13
                                     0.537
```

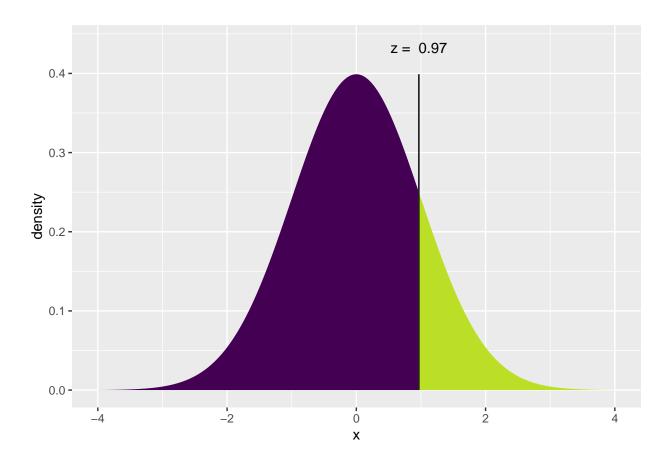
```
ggplot(US_analysis_all) +
  geom_boxplot(aes(x = pre_covid, y = sentiment_score))
```





```
US_analysis_all %>%
  group_by(pre_outbreak) %>%
  summarize(mean(sentiment_score))
## # A tibble: 2 x 2
     pre_outbreak `mean(sentiment_score)`
##
     <chr>>
                                     <dbl>
## 1 no
                                     0.638
## 2 yes
                                     0.650
#precovid all categories
count(US_analysis_all, pre_covid == "yes")
## # A tibble: 2 x 2
     `pre_covid == "yes"`
##
                              n
##
     <1g1>
                          <int>
## 1 FALSE
                             70
## 2 TRUE
                             60
num_precovid = 60
num_postcovid = 70
num = 130
US_analysis_all %>%
 filter(pre_covid == "yes") %>%
  count(sentiment_score > 0.5)
```

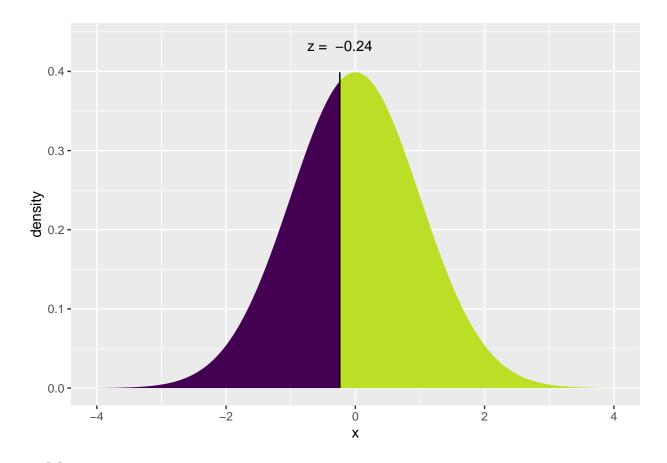
```
`sentiment_score > 0.5`
##
                                 n
##
     <1g1>
                              <int>
## 1 FALSE
                                 21
## 2 TRUE
                                 39
*proportion of positive sentiment videos precovid from sample
p_hat1 = 39/60
US_analysis_all %>%
filter(pre_covid == "no") %>%
count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 19
## 2 TRUE
                                 51
{\it \#proportion of positive sentiment videos postcovid from sample}
p_hat2 = 51/70
p_hat = (39+51)/(60+70)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/60)+(((p_hat)*(1-p_hat))/70))
z_{score} \leftarrow ((p_{hat2-p_hat1})-0)/sd
#p-value
2* (1-xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le 0.9676) = P(Z \le 0.9676) = 0.8334
   P(X > 0.9676) = P(Z > 0.9676) = 0.1666
##
```



```
## [1] 0.3332287
```

```
#outbreak all categories
count(US_analysis_all, pre_outbreak == "yes")
## # A tibble: 2 x 2
     `pre_outbreak == "yes"`
     <1g1>
##
                              <int>
## 1 FALSE
                                 50
## 2 TRUE
                                 80
num_preoutbreak = 80
num_postoutbreak = 50
num = 130
US_analysis_all %>%
  filter(pre_outbreak == "yes") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
##
     `sentiment_score > 0.5`
     <1g1>
                              <int>
## 1 FALSE
                                 24
## 2 TRUE
                                 56
{\it \#proportion of positive sentiment videos\ preoutbreak\ from\ sample}
p_hat1 = 56/80
```

```
US_analysis_all %>%
  filter(pre_outbreak == "no") %>%
  count(sentiment_score > 0.5)
## # A tibble: 2 x 2
## `sentiment_score > 0.5`
##
     <lg1>
                              <int>
## 1 FALSE
                                 16
## 2 TRUE
                                 34
*proportion of positive sentiment videos postoutbreak from sample
p_hat2 = 34/50
p_hat = (56+34)/(80+50)
sd \leftarrow sqrt((((p_hat)*(1-p_hat))/80)+(((p_hat)*(1-p_hat))/50))
z\_score \leftarrow ((p\_hat2-p\_hat1)-0)/sd
#p-value
2* (xpnorm(z_score, 0, 1))
##
## If X \sim N(0, 1), then
## P(X \le -0.2404) = P(Z \le -0.2404) = 0.405
## P(X > -0.2404) = P(Z > -0.2404) = 0.595
##
```



```
## [1] 0.8100434
#Two independent samples t-tests; Comparing two independent means
#pre_covid music
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_music)
##
   Welch Two Sample t-test
##
##
## data: sentiment_score by pre_covid
## t = 0.26554, df = 122.07, p-value = 0.791
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1093453 0.1432249
## sample estimates:
##
   mean in group no mean in group yes
           0.5714842
                             0.5545444
#pre_outbreak music
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_music)
##
##
   Welch Two Sample t-test
## data: sentiment_score by pre_outbreak
## t = -0.20763, df = 106.16, p-value = 0.8359
\#\# alternative hypothesis: true difference in means is not equal to 0
```

```
## 95 percent confidence interval:
## -0.1410260 0.1142872
## sample estimates:
## mean in group no mean in group yes
          0.5553141
                             0.5686835
#pre covid travel and events
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_travel)
##
##
   Welch Two Sample t-test
## data: sentiment_score by pre_covid
## t = 1.3442, df = 123.54, p-value = 0.1814
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.04274451 0.22364743
## sample estimates:
## mean in group no mean in group yes
          0.5754868
                             0.4850353
#pre outbreak travel and events
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_travel)
##
## Welch Two Sample t-test
## data: sentiment_score by pre_outbreak
## t = 1.3771, df = 97.353, p-value = 0.1716
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.04260254 0.23574101
## sample estimates:
## mean in group no mean in group yes
          0.5931672
                             0.4965979
##
#pre covid people and blogs
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_people)
##
## Welch Two Sample t-test
##
## data: sentiment score by pre covid
## t = 0.23163, df = 127.47, p-value = 0.8172
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.08407956 0.10637403
## sample estimates:
## mean in group no mean in group yes
##
          0.7553102
                             0.7441629
#pre_outbreak people and blogs
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_people)
##
## Welch Two Sample t-test
##
```

```
## data: sentiment_score by pre_outbreak
## t = -0.49143, df = 94.86, p-value = 0.6243
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1263419 0.0762046
## sample estimates:
   mean in group no mean in group yes
          0.7347385
                             0.7598071
#pre covid entertainment
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_entertainment)
##
##
  Welch Two Sample t-test
##
## data: sentiment_score by pre_covid
## t = 1.4485, df = 112.72, p-value = 0.1503
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.03089568 0.19889013
## sample estimates:
## mean in group no mean in group yes
          0.6508300
                             0.5668328
#pre_outbreak entertainment
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_entertainment)
##
## Welch Two Sample t-test
##
## data: sentiment_score by pre_outbreak
## t = 0.71369, df = 119.51, p-value = 0.4768
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.07097912 0.15098488
## sample estimates:
## mean in group no mean in group yes
          0.6366792
                             0.5966764
#pre_covid news and politics
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_news)
## Welch Two Sample t-test
## data: sentiment_score by pre_covid
## t = 1.3033, df = 120.9, p-value = 0.195
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.04643392 0.22534759
## sample estimates:
## mean in group no mean in group yes
          0.5436469
                             0.4541901
#pre outbreak news and politics
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_news)
```

```
##
## Welch Two Sample t-test
## data: sentiment_score by pre_outbreak
## t = 0.66178, df = 105.59, p-value = 0.5096
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.09242914 0.18504418
## sample estimates:
## mean in group no mean in group yes
          0.5310914
                             0.4847839
#pre_covid how-to and style
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_how_to)
## Welch Two Sample t-test
## data: sentiment_score by pre_covid
## t = 0.33917, df = 125.74, p-value = 0.735
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.08987314 0.12705021
## sample estimates:
## mean in group no mean in group yes
          0.6316795
                             0.6130910
##
#pre_outbreak how-to and style
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_how_to)
##
## Welch Two Sample t-test
##
## data: sentiment_score by pre_outbreak
## t = -0.86642, df = 103.57, p-value = 0.3883
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.16017409 0.06277054
## sample estimates:
   mean in group no mean in group yes
          0.5931298
                             0.6418316
#pre_covid education
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_education)
##
## Welch Two Sample t-test
##
## data: sentiment_score by pre_covid
## t = -0.093264, df = 123.83, p-value = 0.9258
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1144924 0.1041882
## sample estimates:
## mean in group no mean in group yes
##
          0.6468118
                            0.6519638
```

```
#pre_outbreak education
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_education)
##
## Welch Two Sample t-test
##
## data: sentiment_score by pre_outbreak
## t = -0.14281, df = 94.887, p-value = 0.8867
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1235517 0.1069695
## sample estimates:
## mean in group no mean in group yes
          0.6440874
##
                             0.6523785
#pre_covid science and technology
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_science)
##
## Welch Two Sample t-test
##
## data: sentiment_score by pre_covid
## t = -2.1768, df = 124.91, p-value = 0.03138
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.23154610 -0.01101184
## sample estimates:
## mean in group no mean in group yes
          0.5789204
##
                             0.7001993
#pre_outbreak science and technology
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_science)
##
## Welch Two Sample t-test
##
## data: sentiment_score by pre_outbreak
## t = -1.7595, df = 109.71, p-value = 0.08128
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.21146936 0.01256699
## sample estimates:
## mean in group no mean in group yes
##
          0.5736945
                             0.6731457
#pre_covid all categories
t.test(sentiment_score ~ pre_covid, alternative = "two.sided", data = US_analysis_all)
## Welch Two Sample t-test
## data: sentiment_score by pre_covid
## t = 1.6947, df = 117.06, p-value = 0.09279
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.01734725 0.22313145
```

```
## sample estimates:
## mean in group no mean in group yes
          0.6926683
                            0.5897762
#pre_outbreak categories
t.test(sentiment_score ~ pre_outbreak, alternative = "two.sided", data = US_analysis_all)
##
## Welch Two Sample t-test
##
## data: sentiment_score by pre_outbreak
## t = -0.19637, df = 109.63, p-value = 0.8447
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1332495 0.1092246
## sample estimates:
## mean in group no mean in group yes
##
           0.6377874
                            0.6497999
```