Sentiment Analysis - Ebola

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```
trigger <- read.csv(file="./Trigger_Other.csv", header=TRUE, sep=",", stringsAsFactors = FALSE)</pre>
```

```
concerns=trigger$t_q6
questions=trigger$t_q7
risks=trigger$t_q8
byelaws=trigger$t_q9
discussions=trigger$t_q10
capability=trigger$t_q11
```

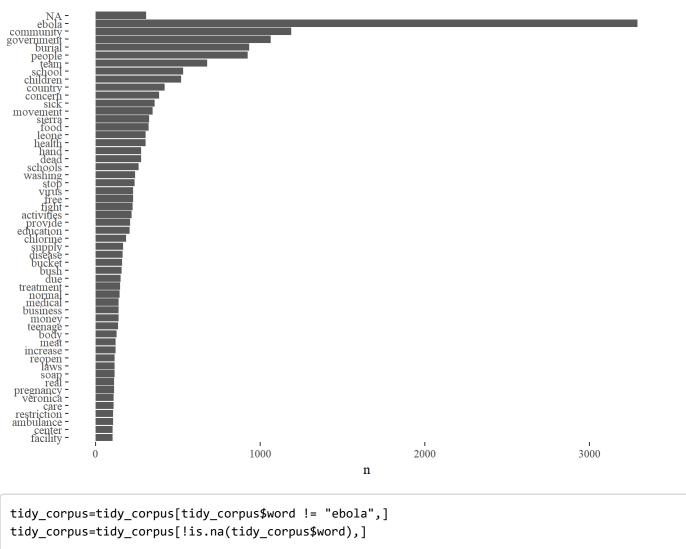
```
concerns_df <- tibble(line = 1:length(concerns),text=concerns )
tidy_corpus <- concerns_df %>%
  unnest_tokens(word, text)
data('stop_words')
tidy_corpus <- tidy_corpus %>%
  anti_join(stop_words)
```

```
## Joining, by = "word"
```

1 Explorations

1.1 Word Counts

```
tidy_corpus %>%
  count(word, sort = TRUE) %>%
  filter(n > 100) %>%
  mutate(word = reorder(word, n)) %>%
  ggplot(aes(word, n)) +
    geom_col() +
    xlab(NULL) +
    coord_flip() +
  theme_tufte()
```



```
tidy_corpus=tidy_corpus[tidy_corpus$word != "ebola",]
tidy_corpus=tidy_corpus[!is.na(tidy_corpus$word),]

corpus_sentiment <- tidy_corpus %>%
  inner_join(get_sentiments("bing")) %>%
  count(word, sentiment) %>%
  spread(sentiment, n, fill = 0) %>%
  mutate(sentiment = positive - negative)
```

```
## Joining, by = "word"
```

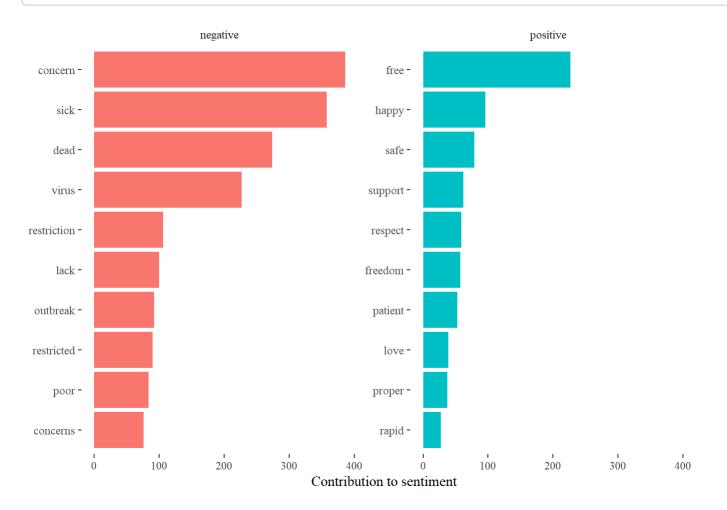
```
word_counts <- tidy_corpus %>%
  inner_join(get_sentiments("bing")) %>%
  count(word, sentiment, sort = TRUE) %>%
  ungroup()
```

```
## Joining, by = "word"
```

1.2 Positive and Negative - Word Frequencies

```
positive_sentiment <- corpus_sentiment %>%
  filter(sentiment>0) %>%
  count(word, sentiment, sort = TRUE) %>%
  ungroup()
word_counts %>%
  group_by(sentiment) %>%
  top_n(10) %>%
  ungroup() %>%
 mutate(word = reorder(word, n)) %>%
  ggplot(aes(word, n, fill = sentiment)) +
  geom_col(show.legend = FALSE) +
  facet_wrap(~sentiment, scales = "free_y") +
  labs(y = "Contribution to sentiment",
       x = NULL) +
  coord_flip() +
  theme_tufte()
```

Selecting by n



1.3 Positive and Negative - Word Clouds

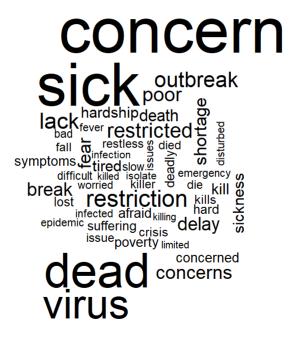
```
layout(matrix(c(1, 2), nrow=2), heights=c(1, 6))
par(mar=rep(0, 4))
plot.new()
text(x=0.5, y=0.5, "Positive Sentiment")

word_counts %>%
  filter(sentiment=='positive') %>%
  with(wordcloud(word, n, max.words = 50))
```

Positive Sentiment



```
layout(matrix(c(1, 2), nrow=2), heights=c(1, 6))
par(mar=rep(0, 4))
plot.new()
text(x=0.5, y=0.5, "Negative Sentiment")
word_counts %>%
  filter(sentiment=='negative') %>%
  with(wordcloud(word, n, max.words = 50))
```



1.4 Sentimentr tests

sentimentr tests

```
element_id sentence_id word_count
##
                                           sentiment
## 1:
                                       4 0.25000000
## 2:
               2
                           1
                                       6 -0.20412415
## 3:
               3
                           1
                                       4 -0.37500000
## 4:
               4
                           1
                                       5 0.04472136
                            1
## 5:
               5
                                       7
                                          0.18898224
```

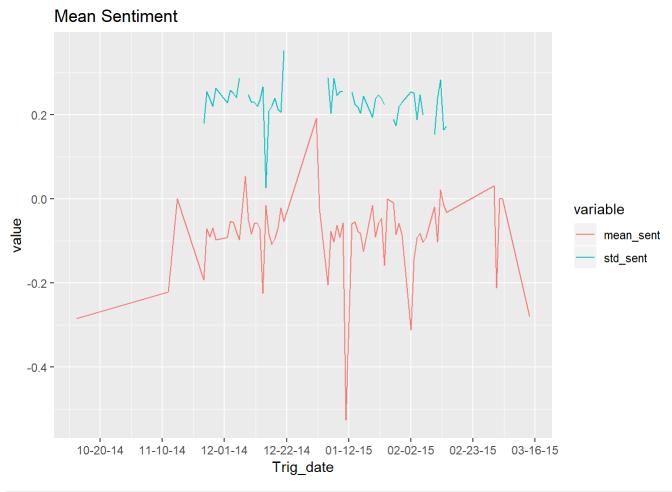
Notes for Sentiment by time: - Two fields: mean sentiment as a function of a day - Approx. Positive slope - Two extremely negative peaks, indicate sentiment about concerns over entire country fell - Terrible news can be a confounding factor

2 Sentiment by time

2.1 Mean Sentiment as a function of day

```
A <- (trigger %>%
  dplyr::select(Trig_date,District,Chiefdom,t_q6))
B <- (trigger$t_q6 %>%
         get_sentences() %>%
  sentiment_by(by=NULL))
sentiment_concerns <- cbind(A,B)</pre>
sentiment_concerns$Trig_date <- as.Date(sentiment_concerns$Trig_date, "%m/%d/%Y")</pre>
concerns_by_day <- sentiment_concerns %>%
  dplyr::select(Trig_date,ave_sentiment)
concerns_by_day <- concerns_by_day %>%
  group_by(Trig_date) %>%
  summarise(mean_sent = mean(ave_sentiment, na.rm = TRUE), std_sent = sd(ave_sentiment, na.rm =
TRUE))
require(ggplot2)
ggplot( data = concerns_by_day, aes( x=Trig_date, y=value, color=variable)) +
  geom_line(aes(y=mean_sent, color = "mean_sent")) +
  geom_line(aes(y=std_sent, color = "std_sent")) +
  ggtitle('Mean Sentiment') +
  scale_x_date(labels = date_format("%m-%d-%y"), breaks = "3 week")
```

```
## Warning: Removed 1 rows containing missing values (geom_path).
```



```
#geom_point()
concerns_by_day
```

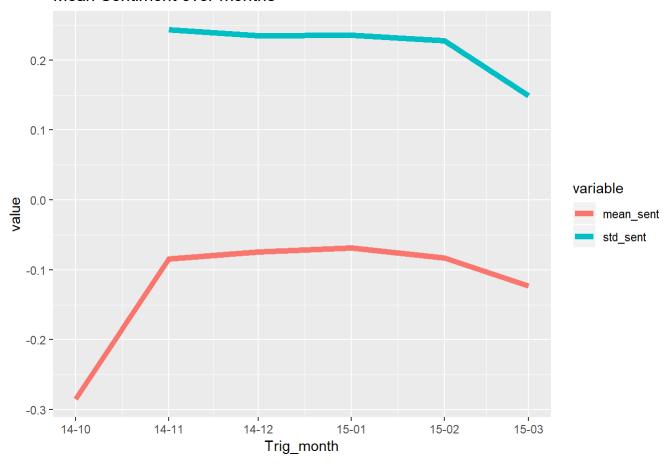
```
## # A tibble: 67 x 3
##
      Trig_date mean_sent std_sent
##
      <date>
                     <dbl>
                               <dbl>
   1 2014-10-12
                   -0.285
##
                             NaN
##
   2 2014-11-12
                   -0.221
                               0.265
   3 2014-11-15
##
                    0
                             NaN
    4 2014-11-24
                   -0.192
                               0.179
##
##
   5 2014-11-25
                   -0.0712
                               0.255
   6 2014-11-26
                   -0.0899
                               0.238
    7 2014-11-27
                   -0.0691
                               0.220
##
##
   8 2014-11-28
                   -0.0978
                               0.263
   9 2014-12-02
                   -0.0912
                               0.228
##
## 10 2014-12-03
                   -0.0530
                               0.258
## # ... with 57 more rows
```

2.2 Mean Sentiment as a function of month

```
A <- (trigger %>%
  dplyr::select(Trig date,District,Chiefdom,t q6))
B <- (trigger$t q6 %>%
         get_sentences() %>%
  sentiment_by(by=NULL))
sentiment concerns cuts <- cbind(A,B)
sentiment_concerns_cuts$Trig_date <- as.Date(sentiment_concerns_cuts$Trig_date, "%m/%d/%Y")</pre>
sentiment_concerns_cuts$Trig_month <- as.Date(cut(sentiment_concerns_cuts$Trig_date,breaks = "mo</pre>
nth"))
sentiment_concerns_cuts$Trig_week <- as.Date(cut(sentiment_concerns_cuts$Trig_date,breaks = "wee</pre>
k",start.on.monday = FALSE))
concerns_by_week <- sentiment_concerns_cuts %>%
  dplyr::select(Trig_week,ave_sentiment)
concerns by week <- concerns by week %>%
  group by(Trig week) %>%
  summarise(mean_sent = mean(ave_sentiment, na.rm = TRUE), std_sent = sd(ave_sentiment, na.rm =
TRUE))
concerns by month <- sentiment concerns cuts %>%
  dplyr::select(Trig_month, ave_sentiment)
concerns_by_month <- concerns_by_month %>%
  group by(Trig month) %>%
  summarise(mean sent = mean(ave sentiment, na.rm = TRUE),
            std sent = sd(ave sentiment, na.rm = TRUE),
            cv_sent = cv(ave_sentiment,na.rm = TRUE))
ggplot( data = concerns_by_month, aes( x=Trig_month, y=value, color=variable)) +
  geom_line(aes(y=mean_sent, color = "mean_sent"), size=2) +
  geom_line(aes(y=std_sent, color = "std_sent"),size =2) +
  ggtitle('Mean Sentiment over months') +
  scale x date(labels = date format("%y-%m"), breaks = "1 month")
```

Warning: Removed 1 rows containing missing values (geom_path).

Mean Sentiment over months

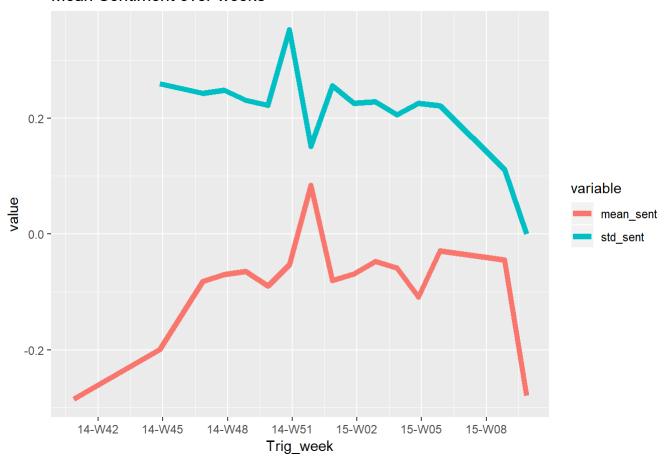


2.3 Mean Sentiment as a function of week

```
ggplot( data = concerns_by_week, aes( x=Trig_week, y=value, color=variable)) +
  geom_line(aes(y=mean_sent, color = "mean_sent"),size=2) +
  geom_line(aes(y=std_sent, color = "std_sent"),size=2) +
  ggtitle('Mean Sentiment over weeks') +
  scale_x_date(labels = date_format("%y-W%W"), breaks = "3 week")
```

Warning: Removed 1 rows containing missing values (geom_path).

Mean Sentiment over weeks



3 Sentiment by space

```
concerns_by_chiefdom <- sentiment_concerns %>%
  dplyr::select(District, Chiefdom, ave_sentiment)
concerns_by_chiefdom <- concerns_by_chiefdom %>%
  group_by(District,Chiefdom) %>%
  summarise(mean_ave = mean(ave_sentiment, na.rm = TRUE))
concerns_by_district <- sentiment_concerns %>%
  dplyr::select(District, ave_sentiment)
concerns_by_district <- concerns_by_district %>%
  group_by(District) %>%
  summarise(mean_sent = mean(ave_sentiment, na.rm = TRUE),
            sd_sent = sd(ave_sentiment,na.rm = TRUE),
            cv_sent = cv(ave_sentiment,na.rm = TRUE))
counts_by_district <- sentiment_concerns %>%
  dplyr::select(District) %>%
  group_by(District) %>%
  summarise(count_sent = n())
head(concerns_by_district)
```

```
## # A tibble: 6 x 4
   District mean_sent sd_sent cv_sent
##
##
    <chr> <dbl> <dbl>
                                 <dbl>
## 1 Bo
               -0.0744 0.218
                                 -293.
## 2 Bombali -0.0556 0.241 -433.
## 3 Bonthe -0.109 0.262 -241.
                                -479.
## 4 Kailahun -0.0451 0.216
## 5 Kambia
               -0.0628
                        0.222
                                 -353.
## 6 Koinadugu -0.0547
                                 -421.
                          0.230
```

3.1 Number of trigger visits across districts

```
shp <- read_sf('./shp1/SLE_adm2.shp')
district_df<-data.frame(shp$NAME_2,shp$geometry)

plt <- ggplot()

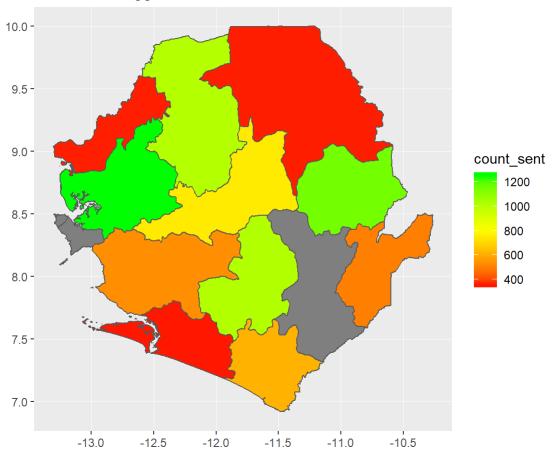
counts_plot <- district_df %>%
  left_join(counts_by_district,by=c("shp.NAME_2"="District"))
```

```
## Warning: Column `shp.NAME_2`/`District` joining factor and character
## vector, coercing into character vector
```

```
#write.csv(data.frame(shp$admin3Name,shp$admin3RefN,shp$admin2Name),file = "ShapeFile3.csv")

plt + geom_sf(data = counts_plot, aes(fill=count_sent)) + scale_fill_gradient2(low='red',high='g reen',mid = 'yellow', midpoint = 800) + ggtitle("Number of trigger visits across Districts")
```

Number of trigger visits across Districts



3.2 Mean sentiment across districts

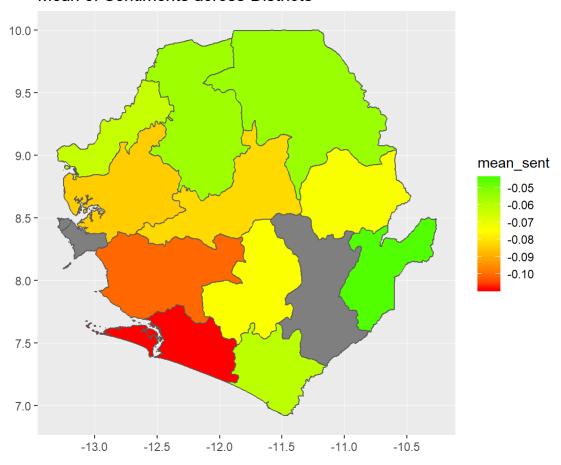
Notes and interpretations about sentiment by space: - Freetown and another district missing from the data - south-western part more negative sentiment - Measures against ebola may be better in other parts of the country

```
concerns_plot<-district_df %>%
  left_join(concerns_by_district,by=c("shp.NAME_2"="District"))
```

Warning: Column `shp.NAME_2`/`District` joining factor and character
vector, coercing into character vector

```
#write.csv(data.frame(shp$admin3Name,shp$admin3RefN,shp$admin2Name),file = "ShapeFile3.csv")
plt <- ggplot()
plt + geom_sf(data = concerns_plot, aes(fill=mean_sent)) + scale_fill_gradient2(low='red',high=
'green', mid='yellow',midpoint=-0.075) + ggtitle("Mean of Sentiments across Districts")</pre>
```

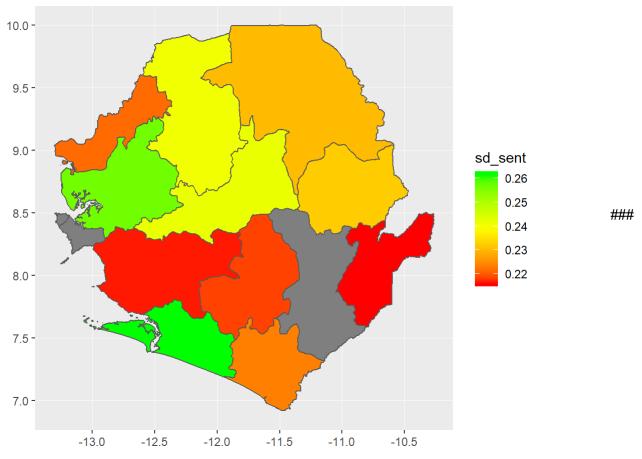
Mean of Sentiments across Districts



3.3 SD of sentiments across districts

```
sd_max <- max(concerns_plot[!is.na(concerns_plot$sd_sent),]$sd_sent)
sd_min <- min(concerns_plot[!is.na(concerns_plot$sd_sent),]$sd_sent)
sd_mid <- (sd_max+sd_min)/2
plt <- ggplot()
plt + geom_sf(data = concerns_plot, aes(fill=sd_sent)) + scale_fill_gradient2(low='red',high='green', mid='yellow', midpoint=sd_mid) + ggtitle("Standard Deviation of Sentiments across Districts")</pre>
```

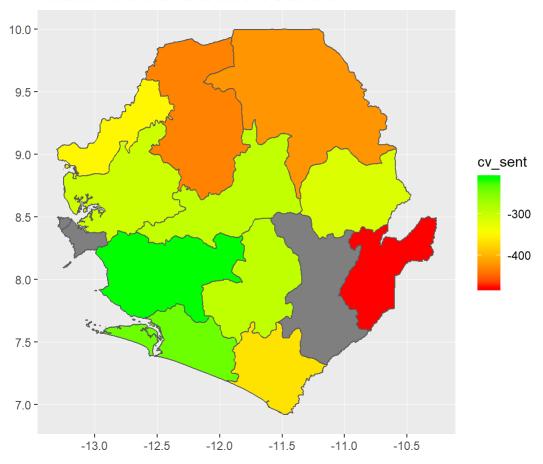
Standard Deviation of Sentiments across Districts



3.4 Co-efficient of variation of sentiments across districts

```
cv_max <- max(concerns_plot[!is.na(concerns_plot$cv_sent),]$cv_sent)
cv_min <- min(concerns_plot[!is.na(concerns_plot$cv_sent),]$cv_sent)
cv_mid <- (cv_max+cv_min)/2
plt <- ggplot()
plt + geom_sf(data = concerns_plot, aes(fill=cv_sent)) + scale_fill_gradient2(low='red',high='gr een', mid='yellow', midpoint=cv_mid) + ggtitle("Variation of Sentiments across Districts")</pre>
```

Variation of Sentiments across Districts



```
head(concerns_plot)
```

```
shp.NAME 2
##
                                       geometry
                                                              sd sent
                                                  mean sent
       Kailahun MULTIPOLYGON (((-10.30196 8... -0.04505028 0.2159075
## 1
         Kenema MULTIPOLYGON (((-11.49417 8...
## 2
## 3
           Kono MULTIPOLYGON (((-11.03017 9... -0.07359921 0.2323381
## 4
        Bombali MULTIPOLYGON (((-11.90307 9... -0.05558179 0.2406574
         Kambia MULTIPOLYGON (((-13.13486 8... -0.06281015 0.2215894
## 5
## 6
      Koinadugu MULTIPOLYGON (((-11.20397 9... -0.05468942 0.2303602
##
       cv_sent
## 1 -479.2589
## 2
## 3 -315.6802
## 4 -432.9789
## 5 -352.7924
## 6 -421.2152
```

4 Sentiment by time and space

4.1 Mean sentiment by months across districts

```
concerns_dis_mon <- sentiment_concerns %>%
   dplyr::select(Trig_date,District, ave_sentiment)

concerns_dis_mon$Trig_date <- as.Date(as.yearmon(concerns_dis_mon$Trig_date))
#concerns_dis_mon$Trig_date <- as.Date(concerns_dis_mon$Trig_date, "%b %Y")

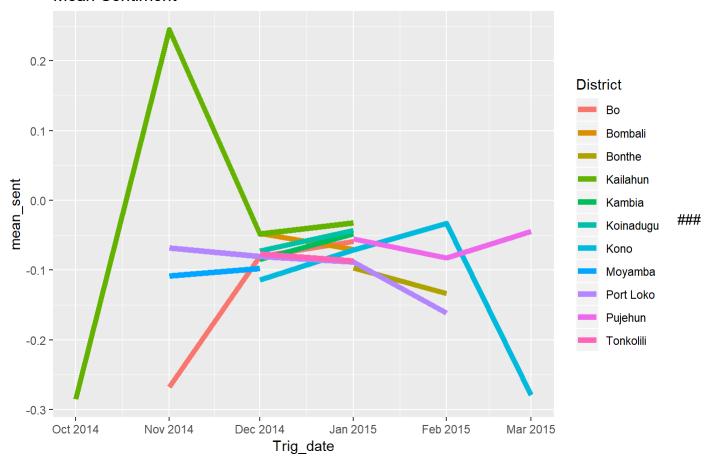
concerns_dis_mon <- concerns_dis_mon %>%
   group_by(District,Trig_date) %>%
   summarise(mean_sent = mean(ave_sentiment, na.rm = TRUE))

concerns_dis_mon
```

```
## # A tibble: 30 x 3
## # Groups:
             District [11]
##
     District Trig_date mean_sent
##
     <chr>
             <date>
                           <dbl>
## 1 Bo
              2014-11-01 -0.268
  2 Bo
             2014-12-01 -0.0802
##
## 3 Bo
             2015-01-01 -0.0589
## 4 Bombali 2014-12-01
                         -0.0473
## 5 Bombali 2015-01-01
                         -0.0705
## 6 Bonthe
             2015-01-01
                         -0.0969
## 7 Bonthe
             2015-02-01
                         -0.133
## 8 Kailahun 2014-10-01
                         -0.285
## 9 Kailahun 2014-11-01
                          0.245
## 10 Kailahun 2014-12-01
                          -0.0483
## # ... with 20 more rows
```

```
ggplot( data = concerns_dis_mon, aes(x=Trig_date, y=mean_sent)) + geom_line(aes(color = Distric
t), size=2) + ggtitle('Mean Sentiment') + scale_x_date(name="Trig_date", date_labels = "%b %Y",
    breaks="month" )
```

Mean Sentiment



4.2 Mean sentiment by weeks across districts

```
concerns_dis_week <- sentiment_concerns %>%
   dplyr::select(Trig_date,District, ave_sentiment)

concerns_dis_week$Trig_week <- as.Date(cut(concerns_dis_week$Trig_date,breaks = "week",start.on.
monday = FALSE))

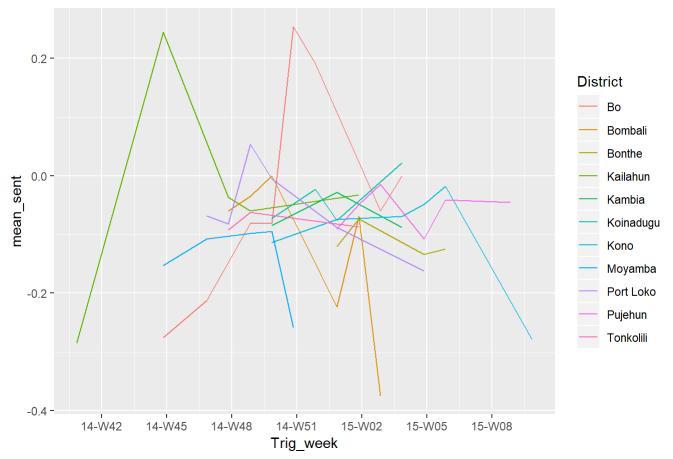
concerns_dis_week <- concerns_dis_week %>%
   group_by(District,Trig_week) %>%
   summarise(mean_sent = mean(ave_sentiment, na.rm = TRUE))

concerns_dis_week
```

```
## # A tibble: 56 x 3
               District [11]
## # Groups:
##
      District Trig_week mean_sent
##
               <date>
                               <dbl>
##
   1 Bo
               2014-11-09
                             -0.276
    2 Bo
               2014-11-23
                             -0.212
##
##
    3 Bo
               2014-12-07
                            -0.0813
##
    4 Bo
               2014-12-14
                            -0.0810
    5 Bo
               2014-12-21
                             0.254
##
                             0.191
##
    6 Bo
               2014-12-28
               2015-01-18
##
    7 Bo
                            -0.0598
    8 Bo
               2015-01-25
##
   9 Bombali
               2014-11-30
                             -0.0594
## 10 Bombali 2014-12-07
                             -0.0349
## # ... with 46 more rows
```

```
ggplot( data = concerns_dis_week, aes(x=Trig_week, y=mean_sent)) + geom_line(aes(color = Distric
t)) + ggtitle('Mean Sentiment over weeks') + scale_x_date(name="Trig_week", date_labels = "%y-W%
W", breaks="3 week" )
```

Mean Sentiment over weeks



4.3 Mean sentiment by months across district (Animation)

```
concerns_month_geo <-district_df %>%
  left_join(concerns_dis_mon,by=c("shp.NAME_2"="District"))
```

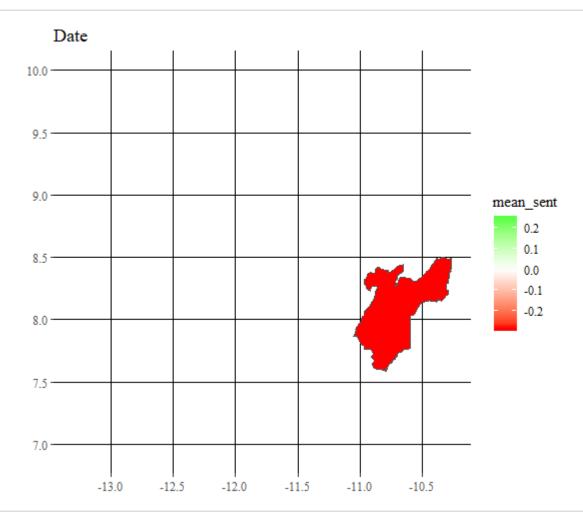
Warning: Column `shp.NAME_2`/`District` joining factor and character
vector, coercing into character vector

concerns_month_geo <- concerns_month_geo[!is.na(concerns_month_geo\$Trig_date),]</pre>

concerns_month_geo_animate <- ggplot() + geom_sf(data = concerns_month_geo, aes(fill=mean_sent,f
rame=Trig_date)) + scale_fill_gradient2(low='red',high='green', mid='white',midpoint=0) + transi
tion_states(Trig_date, wrap=TRUE) + coord_sf() + theme_tufte() + labs(title = "Date")</pre>

Warning: Ignoring unknown aesthetics: frame

animate plot with gganimate
animate(concerns_month_geo_animate, fps=3)



#anim_save("animation.gif", animation=Last_animation())

#concerns_month_geo\$Trig_month

#concerns_month_geo