Topic_model_TMN

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Load Libraries

```
library(tidyverse)
library(topicmodels)
library(tidytext)
library(lexicon)
library(factoextra)
```

Read in dataset

```
movies <- read.csv("movie_plots.csv", header = TRUE)</pre>
```

Unnest tokens using tidytext:

Joining with `by = join_by(word)`

Remove common names using lexicon:

```
data("freq_first_names")
data("freq_last_names")

first_names <- tolower(freq_first_names$Name) #convert to lower case
last_names <- tolower(freq_last_names)

plot_word_counts <- plot_word_counts |> filter(!(word %in% first_names)))
plot_word_counts <- plot_word_counts |> filter(!(word %in% last_names)))
```

Cast word counts to document term matrix

```
plots_dtm <- plot_word_counts |> cast_dtm(Movie.Name, word, n)
```

Look at dimensions of matrix

```
#Distinct words
dim(plot_word_counts |> distinct(word))[1]
```

[1] 13394

```
dim(movies)
```

[1] 1077 2

LDA with 30 topics:

```
plots_lda <- LDA(plots_dtm, k = 30, control = list(seed = 1066))</pre>
```

Retrieving Gammas:

```
plots_gamma <- tidy(plots_lda, matrix = "gamma")
#gamma is per-document-per-topic probabilities
#each values is an estimated proportion of words from that document that are generated
#from that topic</pre>
```

Retrieving Betas:

```
plots_beta <- tidy(plots_lda, matrix = "beta")
#beta is the probability of that term being generated from that topic</pre>
```

Pivoting the plots_gamma table wider so we can cluster by gammas for each topic

Create 8 clusters for 8 genres:

```
set.seed(8)
plots_gamma_wider_no_na <- plots_gamma_wider |> drop_na()
cluster <- kmeans(plots_gamma_wider_no_na |> select(-document), 10)
plot_clusters <- fviz_cluster(cluster, data = plots_gamma_wider_no_na |> select(-document))
```

Look into genres in each cluster. Read in data with genres:

```
english_movies_with_genres <- read.csv("movie_plots_with_genres.csv")
clusters <- cluster[["cluster"]]
plots_gamma_wider$cluster <- clusters</pre>
```

Let's get the colors used in each cluster for further EDA:

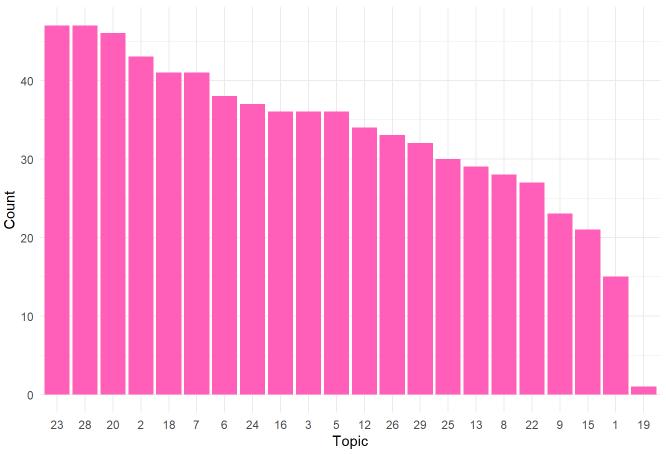
```
# Extract the colors used in the plot
colors_used <- ggplot_build(plot_clusters)$data[[2]]$colour

# Display the unique colors used
unique(colors_used)</pre>
```

```
[1] "#F8766D" "#D89000" "#A3A500" "#39B600" "#00BF7D" "#00BFC4" "#00B0F6" [8] "#9590FF" "#E76BF3" "#FF62BC"
```

Let's look at individual cluster to understand a bit more of context, start with cluster10 because that cluster seems to cover the most range on the cluster map.

```
cluster_10 <- plots_gamma_wider |> filter(cluster == 10)
#let's look at the max values of gammas for each movie and see which topic they fall into
cluster_10$max_value <- apply(cluster_10[, 2:31], 1, max)</pre>
#let's get the topic number of the max gammas for each row
cluster_10$max_colname <- apply(cluster_10[, 2:31], 1,</pre>
                                 function(x) colnames(cluster_10[, 2:31])[which.max(x)])
#lets plot the count of each topic that has the max gamma value for a movie
# Create a data frame with the counts
count_cluster10 <- as.data.frame(table(cluster_10$max_colname))</pre>
count_cluster10$Var1 <- factor(count_cluster10$Var1,</pre>
                                levels = count cluster10$Var1[order(-count cluster10$Freq)])
ggplot(count_cluster10, aes(x = Var1, y = Freq)) +
  geom_bar(stat = "identity", fill = "#FF62BC") +
  labs(title = "Count of Topics that Contain Max Gammas in Cluster 10",
       x = "Topic", y = "Count") +
 theme_minimal()
```

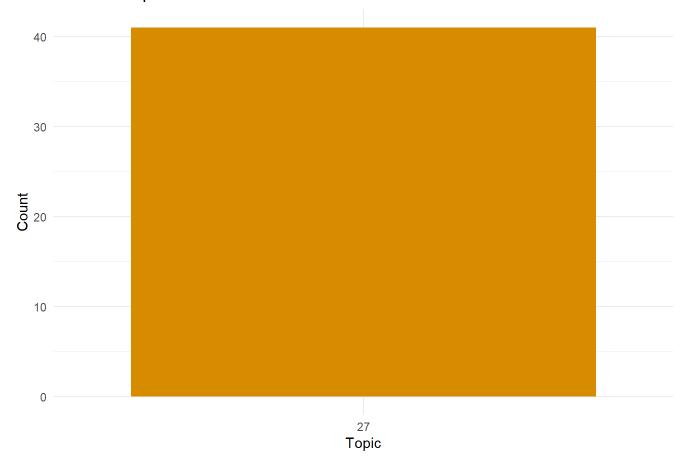


Let's do the same thing for cluster1 and compare with the results from cluster10:



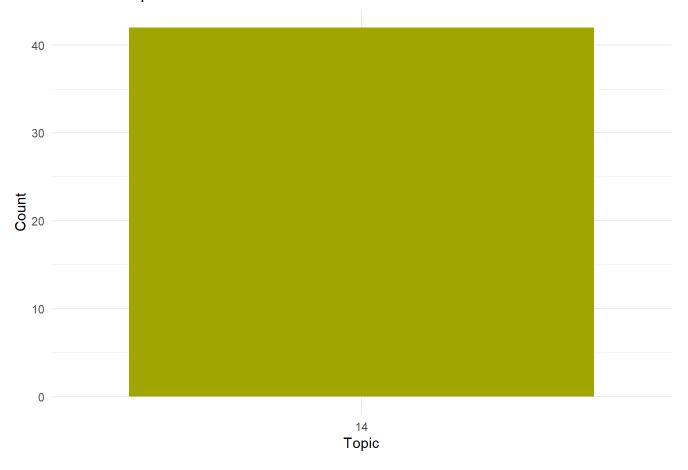
Observation: Interesting that all of cluster 1 count for max gammas is topic 4. Topic 4 was not counted as max at all in cluster 10 so perhaps cluster 10 and 4 genres might not be so closely related... let's look at more individual clusters.

Let's look at cluster2:



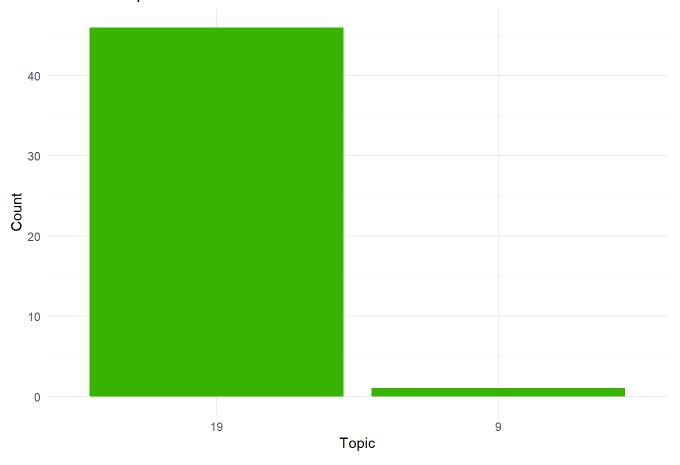
Observation: Cluster2 contains all max gammas from Topic 27, which does not show up as max gammas counts in Cluster10.

Let's look at Cluster3.



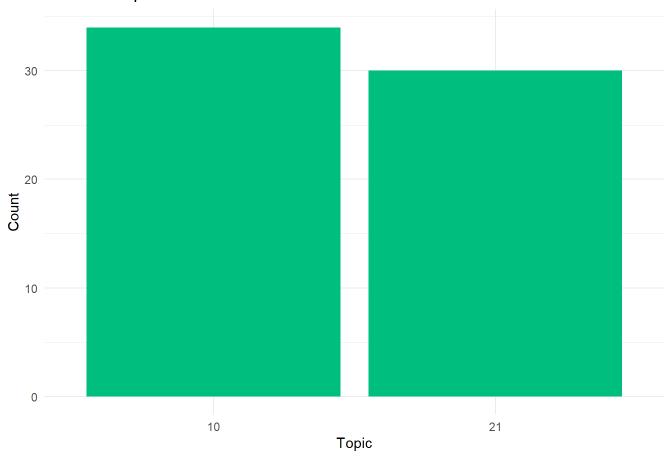
Observation: Cluster3 contains all max gammas from Topic 14, which does not show up as max gammas counts in Cluster10.

Let's look at Cluster4.



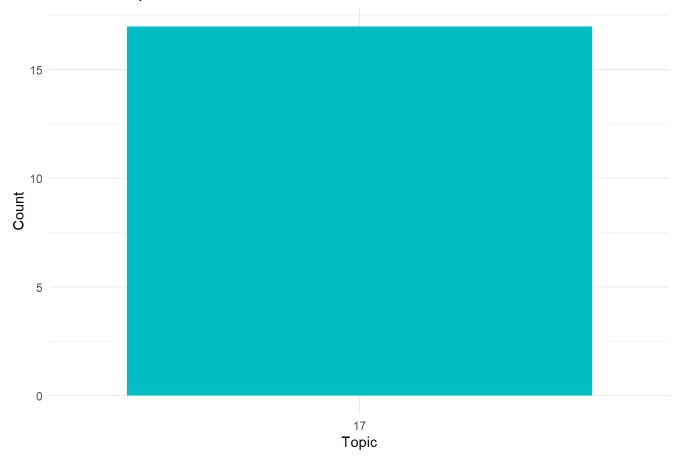
Observation: majority of max gammas counts fall in topic 19 with a few falling in topic 9. Both of these tops show up in the plot of max gamma counts for cluster 10. However, with cluster 10, topic 9 has a much higher count than topic 19, so I would say there's likely not too much overlap between cluster 4 and cluster 10.

Let's look at Cluster5:



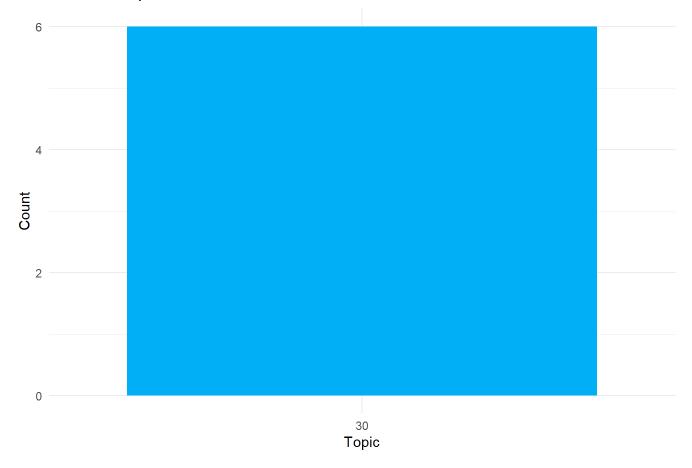
Observation: Cluster 5 seems to have a mix between topic 10 and topic 21. No overlap with cluster 10.

Let's look at Cluster 6:



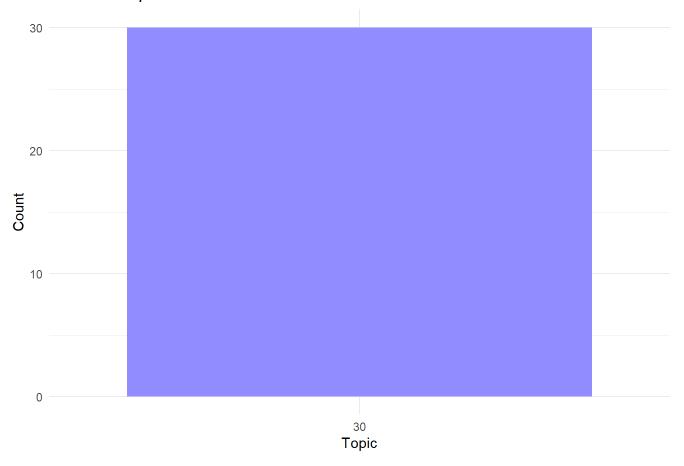
Observation: All max gammas in cluster 6 falls in Topic 17. No overlap with topic 10.

Let's look at Cluster7.



Observation: All max counts in cluster 7 falls in topic 30. Not much overlap with cluster 10.

Let's look at Cluster8:



Observation: All max gammas in cluster 8 falls in Topic 30. No overlap with topic 10. But perhaps overlap with topic 7.

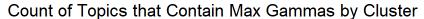
Let's look at Cluster9:

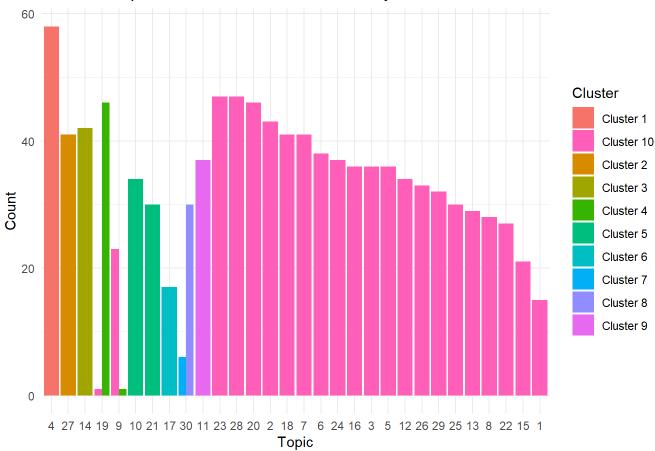


Observation: All max gammas in cluster 9 falls in Topic 11. No overlap with topic 10.

Let's plot the count of all the max topics in each cluster:

```
count_cluster1$Cluster <- "Cluster 1"</pre>
count_cluster2$Cluster <- "Cluster 2"</pre>
count_cluster3$Cluster <- "Cluster 3"</pre>
count_cluster4$Cluster <- "Cluster 4"</pre>
count_cluster5$Cluster <- "Cluster 5"</pre>
count_cluster6$Cluster <- "Cluster 6"</pre>
count_cluster7$Cluster <- "Cluster 7"</pre>
count_cluster8$Cluster <- "Cluster 8"</pre>
count_cluster9$Cluster <- "Cluster 9"</pre>
count_cluster10$Cluster <- "Cluster 10"</pre>
# Combine the data frames using rbind()
combined_counts <- rbind(count_cluster1, count_cluster2, count_cluster3, count_cluster4,</pre>
                           count_cluster5, count_cluster6, count_cluster7, count_cluster8,
                           count_cluster9, count_cluster10)
ggplot(combined_counts, aes(x = Var1, y = Freq, fill = Cluster)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(title = "Count of Topics that Contain Max Gammas by Cluster",
       x = "Topic", y = "Count") +
```





Observation: Generally, it seems like there's not much overlap between the clusters except for cluster 7 and 8 which might be able to be combine into one. But each clusters seems to have a unique topic, which might've been the point of the clusters...let's look closer at cluster 7 and 8 and topic 30.

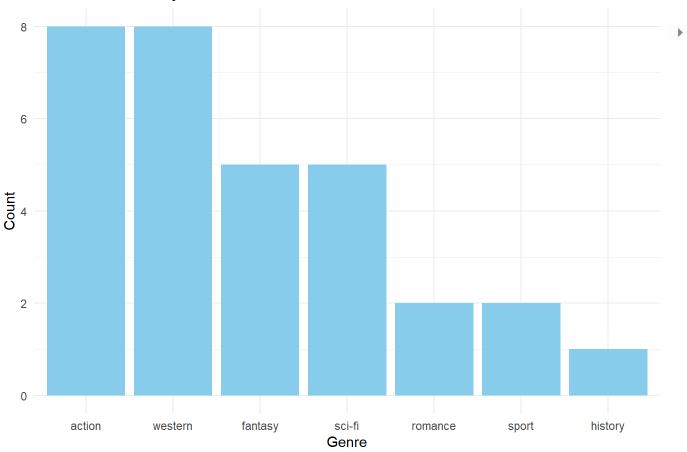
```
# Get the top 10 terms for topic 30
terms_topic_30 <- terms(plots_lda, 10)[, 30]
print(terms_topic_30)</pre>
```

```
[1] "father"    "kidnapped" "land"    "fight"    "women"
[6] "morrell"    "monster"    "daltons"    "government" "return"
```

Observation: looking at these terms, perhaps cluster 7 and 8 are suspense-action kidnapping movies?

```
# Merge cluster8 with movie_with_genres df to check our estimation
movie_plots_with_genres <- read.csv("movie_plots_with_genres.csv", header = T)
movie_plots_with_genres <- as.data.frame(movie_plots_with_genres)</pre>
```

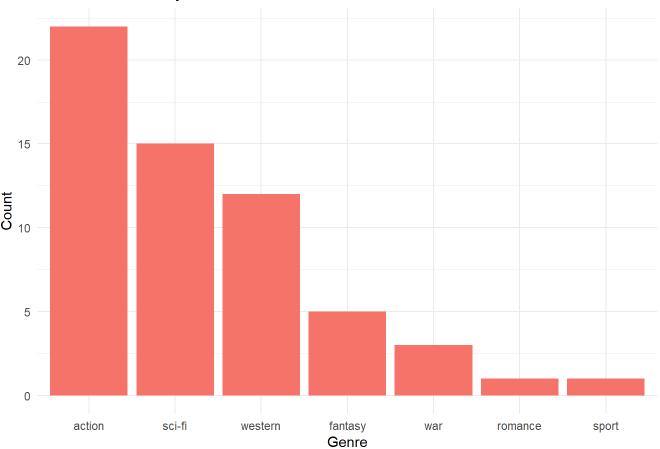
Count of Movies by Genre from Cluster 7 & 8



Let's do the same thing and check the genres of Cluster 1.

```
ggplot(merged_data1, aes(x = Genre)) +
    geom_bar(fill = "#F8766D") +
    labs(title = "Count of Movies by Genre from Cluster 1", x = "Genre", y = "Count") +
    theme_minimal()
```

Count of Movies by Genre from Cluster 1



```
terms_topic_4 <- terms(plots_lda, 10)[, 4]
print(terms_topic_4)</pre>
```

```
[1] "life" "war" "woman" "death" "brothers" "cowboy"
[7] "line" "drug" "world" "police"
```

Observation: Topic 4 also seems to be action. Most of these genres show up under action and western not sure what else to do to make it more accurate...this is as far as I got...