# **Movie Data Analysis**

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January 2, 2019

#### R Markdown

Movie data analysis

```
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(tidyr)
library(stringr)
library(ggplot2)
setwd('C:/Users/nachi/OneDrive/Desktop/Tredence')
#Read csv file
links = read.csv('links.csv', header = TRUE, fill = TRUE)
movies = read.csv('movies.csv', header = TRUE, fill = TRUE)
ratings = read.csv('ratings.csv', header = TRUE, fill = TRUE)
tags = read.csv('tags.csv', header = TRUE, fill = TRUE)
# Making genre value change that will be required later
movies$genres <- as.character(movies$genres)</pre>
movies$genres[movies$genres == '(no genres listed)'] <- '(no_genres_listed)'</pre>
# Merge all data
d1 <- merge(movies, links, by.x = "movieId", by.y = "movieId", all.x = TRUE)</pre>
d2 <- merge(d1, ratings, by.x = "movieId", by.y = "movieId", all.x = TRUE)</pre>
# d3 \leftarrow merge(d2, tags, by.x = c("movieId", "userId"), by.y = c("movieId", "userId"), all.x
= TRUE)
### Levels of data for the datasets above
# --> links is at Movie Level
# --> movies is at Movie Level
# --> ratings is at Movie, user level
# --> tags is at Movie, user, tag level
# --> d1 is at Movie level
```

```
# --> d2 is at Movie, user level

# Typecasting
d2$title <- as.character(d2$title)
d2$genres <- as.character(d2$genres)
# d2$tag <- as.character(d2$tag)
d2$release_dt <- as.integer(substr(d2$title, nchar(d2$title)-4, nchar(d2$title)-1))</pre>
```

#### NOTE:

Some of the movies do not have the release year mentioned against their name. Since there are very few of them, they have been ignored in the following analysis.

## **Data wrangling**

1. Average rating for each movie released in or after 1996

2. Top 5 most reviewed movies every year after 1994

r1994 <- d2[(d2\$release dt > 1994 & !is.na(d2\$rating)), ]

# Clearing memory space allocated to temporary variables

# r1994\_review <- NULL

r1994 <- NULL

```
m1996 <- d2[d2$release_dt >= 1996,]
# Keeping necessary columns
m1996 <- m1996[c("title", "rating")]  # Filtering by name in this situation. Mostly
done by index
m1996_ratings <- aggregate(m1996, by = list(m1996$title), FUN = mean, na.rm = TRUE)

# Clearing memory space allocated to temporary variables
m1996_ratings$title <- NULL
m1996 <- NULL</pre>
```

```
# Removing NAs that are intrduced because some movies do not have missing dates
r1994 <- r1994[!is.na(r1994$release dt),]
# Keeping necessary columns (not really necessary to do this)
r1994 <- r1994[c("title", "movieId", "release_dt")]
r1994_review <- aggregate(movieId ~ title + release_dt, data = r1994, FUN = length)
names(r1994_review)[3] <- "num_ratings"</pre>
# Getting top 5
ans <- r1994 review \% group by (release dt) \% top n(n = 5, wt = num ratings)
# Imp NOTE: The no. of rows in the answer is not a multiple of 5 as there are movies with
the same no. of ratings
ans <- ans[order(ans$release_dt, -ans$num_ratings),]</pre>
ans <- ans %>%
                                                                   # Creating a rank
required for plotting later
          group by(release dt) %>%
          mutate(rank top5 = order(order(num ratings, decreasing=TRUE)))
```

3. Average rating for "Drama", "Romance" and "Drama and Romance" movies. This has been done using tags not genres - Can be done just as easily by genres

```
# Function to get the list of movies that match the tag(s)
mlist <- function(x)</pre>
  tags_temp <- tags[tags$tag %in% x ,]</pre>
  tags_temp$userId <- NULL</pre>
  temp <- spread(tags_temp, key = tag, value = timestamp)</pre>
  movie list <- unlist(temp[complete.cases(temp),]$movieId)</pre>
  if(length(movie list)==0){
    print("There are no movies that match all the following genres")
    print(x)
    return(NULL)
  } else {
    return(movie_list)
}
result <- function(x)
  if(is.null(x)){
    print("There are no such movies")
    print(x)
    return()
  } else {
    genre data <- d2[d2$movieId %in% tag movie list,]</pre>
    genre_data <- genre_data[c("title", "rating")]</pre>
    genre_rating <- aggregate(rating ~ title, genre_data, FUN = mean, na.rm = TRUE) #</pre>
Movie level ratings
    return(mean(genre_data$rating)) # Genre level rating
  }
}
# Drama
tag_list <- c("drama")</pre>
tag_movie_list <- mlist(tag_list)</pre>
result(tag movie list)
## [1] 4.150209
# Romance
tag_list <- c("romance")</pre>
tag_movie_list <- mlist(tag_list)</pre>
result(tag_movie_list)
## [1] 3.542184
# Drama and Romance
tag list <- c("drama", "romance")</pre>
tag_movie_list <- mlist(tag_list)</pre>
## [1] "There are no movies that match all the following genres"
## [1] "drama" "romance"
```

```
result(tag_movie_list)
## [1] "There are no such movies"
## NULL
## NULL
```

#### **VERY IMP**

NOTE: Tag matching has been done in terms of equality, not as a 'contains' eg: when the tag input is 'drama', movies with 'courtroom drama' have not been selected. But, the function can be easily modified for the 'contains' scenario as well.

NOTE: Both the functions can be integrated into one but the idea remains the same

### 4. Number of customers who rated a movie tagged as "horror" by year

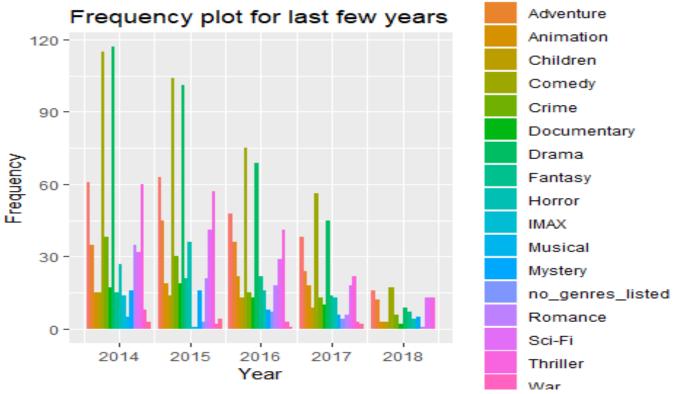
```
# tags_in_question <- c("horror", "drama")</pre>
tags in question <- c("horror")</pre>
movie_list <- unlist(unique(tags[tags$tag %in% tags_in_question,]$movieId))</pre>
temp_df <- d2[d2$movieId %in% movie_list,]</pre>
yearly users <- aggregate(userId ~ release dt, temp df, FUN = length)
print(paste("The yearly number of users that rated a movie with
the",tags_in_question,"tag(s) is/are"))
## [1] "The yearly number of users that rated a movie with the horror tag(s) is/are"
print(yearly_users)
##
     release dt userId
## 1
           1986
                    126
           1990
                     44
## 2
                     25
           1991
## 3
## 4
           2010
                     42
temp_df <- NULL
```

# **Data plotting**

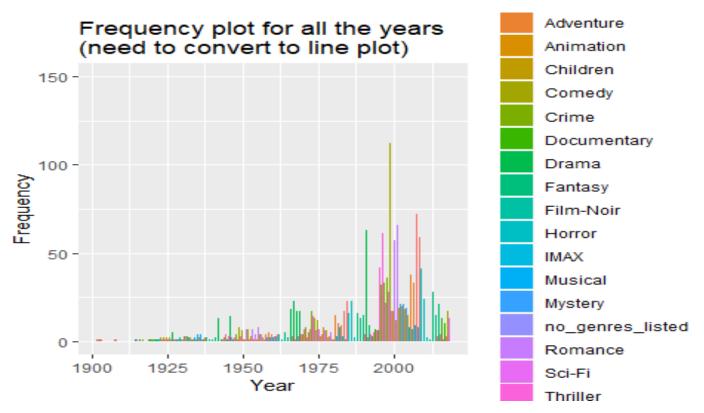
1. Trend of movie genres by the release years i.e. frequency of different genres of movies released each year. If a movie is across multiple genres then count them in all

```
# Using the splitstackshape package
# library(splitstackshape)
# transpose_fn <- function(x)</pre>
# {
#
#
    temp <- x[c("movieId", "genres")]</pre>
    temp_wide <- cSplit(temp, "genres", sep="|")</pre>
#
    temp_long <- melt(temp_wide, id.vars = c("movieId"))</pre>
#
#
   temp_long <- temp_long[!is.na(temp_long$value),]
#
   temp_long$variable <- NULL
#
#
   # Add back movie title info
#
   x \leftarrow merge(x, temp long, by="movieId", all.x = TRUE)
#
   x$genres <- as.character(x$genres)</pre>
#
   x$title <- as.character(x$title)
# x$genres <- NULL
# names(x)[3] \leftarrow "genres"
#
   return(x)
#
# }
# Using the stringr package
library(reshape2)
##
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
       smiths
##
transpose_fn <- function(x)</pre>
  temp <- movies[c("movieId", "genres")]</pre>
  temp$genres <- as.character(temp$genres)</pre>
  abc <- strsplit(temp$genres," | ", fixed = TRUE)</pre>
  abc <- as.data.frame(str_split_fixed(abc,' ', 20))</pre>
  abc <- abc[, colSums(abc != "") != 0]
  abc <- as.data.frame(lapply(abc, gsub, pattern="c\\(", replacement=''))</pre>
  abc <- as.data.frame(lapply(abc, gsub, pattern='[() ",]', replacement=''))</pre>
  abc <- as.data.frame(apply(abc, 2, function(x) gsub("\^$|\^\$", NA, x)))
  abc[] <- lapply(abc, function(x) if(is.factor(x)) as.character(x) else x)</pre>
  temp_wide <- cbind(temp,abc)</pre>
  temp wide$genres <- NULL
  temp_long <- melt(temp_wide, id.vars = c("movieId"))</pre>
  temp_long <- temp_long[!is.na(temp_long$value),]</pre>
  temp_long$variable <- NULL</pre>
  # Add back movie title info
  x <- merge(x, temp_long, by="movieId", all.x = TRUE)</pre>
```

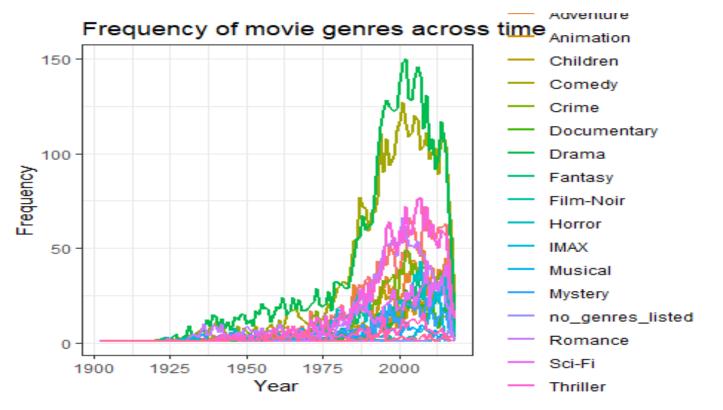
```
x$genres <- as.character(x$genres)</pre>
  x$title <- as.character(x$title)</pre>
  x$genres <- NULL
  names(x)[3] <- "genres"</pre>
  return(x)
}
# Create data for plotting
df <- transpose_fn(movies)</pre>
df$year <- as.integer(substr(df$title, nchar(df$title)-4, nchar(df$title)-1))</pre>
df_plot <- aggregate(movieId ~ genres + year, data = df, FUN = length)</pre>
df_plot <- df_plot[order(-df_plot$year, df_plot$genres, -df_plot$movieId),]</pre>
## Plotting
library(ggplot2)
# Filtering the data for a few years to just visualize the plot
zzz <- df_plot[df_plot$year>2013,]
ggplot(zzz,aes(x = year,y = movieId, fill=genres)) +
  geom_bar(stat="identity", position = "dodge") +
  xlab("Year")+ylab("Frequency") + ggtitle("Frequency plot for last few years")
```



```
# Plot for the entire data
ggplot(df_plot,aes(x = year,y = movieId, fill=genres)) +
   geom_bar(stat="identity", position = "dodge") +
   xlab("Year")+ylab("Frequency") + ggtitle("Frequency plot for all the years \n(need to convert to line plot)")
```

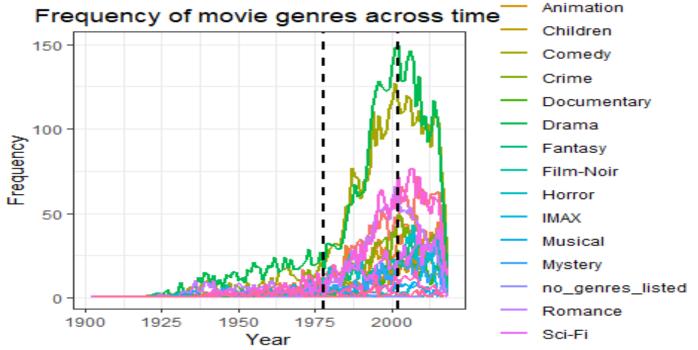


## --> A bar graph is not the best way to visualize the trend. Using a line graph below
ggplot(df\_plot, aes(year, movieId, group = genres, colour = genres)) +
 geom\_line(size = 1) + theme\_bw() +
 xlab("Year")+ylab("Frequency") + ggtitle("Frequency of movie genres across time")



#### 2. Is there a shift in trend?

```
## --> Yes there is
ggplot(df_plot, aes(year, movieId, group = genres, colour = genres)) +
    geom_line(size = 1) + theme_bw() +
```



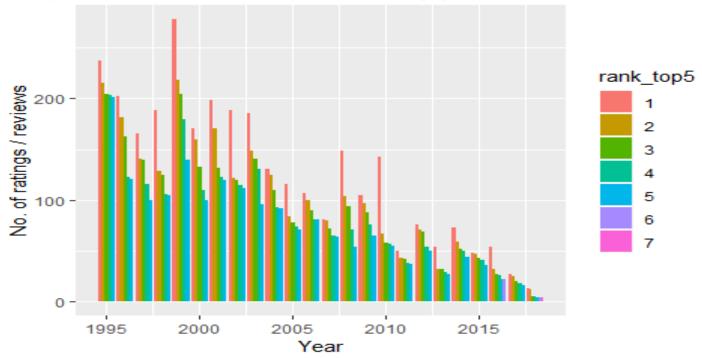
NOTE: There is a stark increase in the number of movies producted after 1978, The upward trend peaks at 2002 and then starts to fall down again (but

# 3. Top 5 most reviewed movies every year after 1994 - would like to see all the years plotted at one go head(ans)

```
## # A tibble: 6 x 4
## # Groups:
               release_dt [2]
     title
##
                                            release dt num ratings rank top5
     <chr>>
                                                 <int>
                                                             <int>
                                                                        <int>
##
## 1 Braveheart (1995)
                                                  1995
                                                               237
                                                                            1
## 2 Toy Story (1995)
                                                  1995
                                                               215
                                                                            2
## 3 Usual Suspects, The (1995)
                                                  1995
                                                               204
                                                                            3
## 4 Seven (a.k.a. Se7en) (1995)
                                                               203
                                                                            4
                                                  1995
                                                                            5
## 5 Apollo 13 (1995)
                                                  1995
                                                               201
## 6 Independence Day (a.k.a. ID4) (1996)
                                                                            1
                                                  1996
                                                               202
ans_temp <- ans
ans temp$rank top5 <- as.character(ans temp$rank top5)</pre>
ggplot(ans_temp,aes(x = release_dt,y = num_ratings, fill=rank_top5)) +
  geom bar(stat="identity", position = "dodge") +
 theme(plot.title = element_text(hjust = 0.5)) +
 # geom_text(aes(label=title), position=position_dodge(width=1), vjust=-0.25, angle=-90)
```

+
xlab("Year")+ylab("No. of ratings / reviews") + ggtitle("Top 5 most reviewed movies
every year after 1994")

# Top 5 most reviewed movies every year after 1994



##### VERY IMP

NOTE: 1) The rankings go from 1 to 7 (not 1 to 5) because year 2016 and 2018 have more than 1 movie with the same number of reviews

2) This chart does not show the movie names. They can be shown by uncommenting the geom\_text() line in the above code. This is not recommended as the plot becomes too cluttered and unreadable. This issue can be resolved by displaying the movie name when you hover over a bar using the mouse. This cannot be done in ggplot, but can be done using the plotly library

## What else can you do with the data?

4. What was the golden age of cinema? - i.e. Comparing movie trends (number of movies) across years and Average rating of movies across time

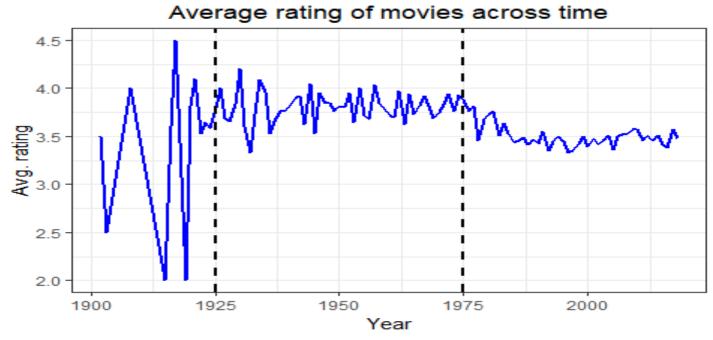
```
library(gtable)
library(grid)
temp_movies <- movies</pre>
temp_movies$title <- as.character(temp_movies$title)</pre>
temp_movies$year <- as.integer(substr(temp_movies$title, nchar(temp_movies$title)-4,</pre>
nchar(temp movies$title)-1))
temp movies agg <- aggregate(title ~ year, data = temp movies, FUN = length)
df temp <- d2[(!is.na(d2$rating) & !is.na(d2$release dt)), ]</pre>
df_temp <- aggregate(rating ~ release_dt, data = df_temp, FUN = mean, na.rm = TRUE)</pre>
# Individual plots
#1 No. of movies
ggplot(temp_movies_agg, aes(year, title)) +
  geom_line(size = 1, colour = "red") + theme_bw() +
 xlab("Year")+ylab("# movies") + ggtitle("Number of movies produced across time") +
  geom vline(xintercept=2016, linetype="dashed", size = 0.75) +
  theme(plot.title = element text(hjust = 0.5),
        plot.caption = element text(hjust = 0)) +
  labs(caption = "NOTE: Decline after 2016 probably due to incomplete data"
       , face="bold", size=5)
```

# Number of movies produced across time



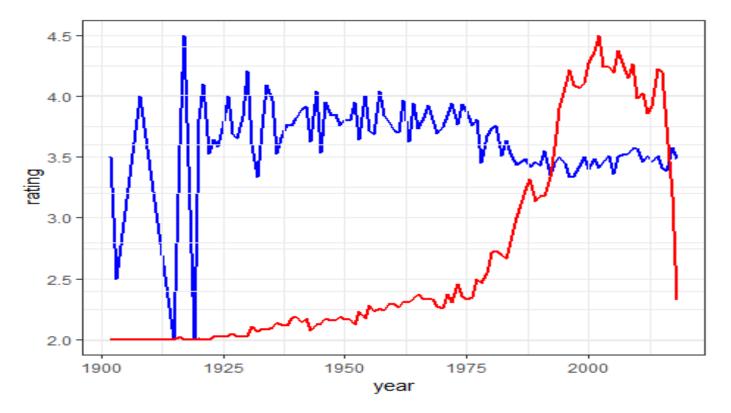
NOTE: Decline after 2016 probably due to incomplete data

```
#2 Average rating
ggplot(df_temp, aes(release_dt, rating)) +
    geom_line(size = 1, colour = "blue") + theme_bw() +
    xlab("Year")+ylab("Avg. rating") + ggtitle("Average rating of movies across time") +
    geom_vline(xintercept=1925, linetype="dashed", size = 0.75) +
```



- 1) The plot shows extreme fluctuations before 1925 because there are very few
- 2) There seems to be a small drop in average rating after 1975

```
# Combining into one plot
df_temp <- cbind(df_temp, temp_movies_agg)</pre>
grid.newpage()
# Two plots
p1 <- ggplot(df temp, aes(year, rating)) + geom line(size = 1, colour = "blue") +
theme bw()
p2 <- ggplot(df_temp, aes(year, title)) + geom_line(size = 1, colour = "red") +
theme_bw() %+replace%
 theme(panel.background = element_rect(fill = NA))
# Extract gtable
g1 <- ggplot_gtable(ggplot_build(p1))</pre>
g2 <- ggplot gtable(ggplot build(p2))</pre>
# Overlapping the panel of 2nd plot on that of the 1st plot
pp <- c(subset(g1$layout, name == "panel", se = t:r))</pre>
g <- gtable_add_grob(g1, g2\$grobs[[which(g2\$layout\$name == "panel")]], pp\$t, pp\$b,
pp$1)
# Drawing
grid.draw(g)
```



## # <Just an extra line>

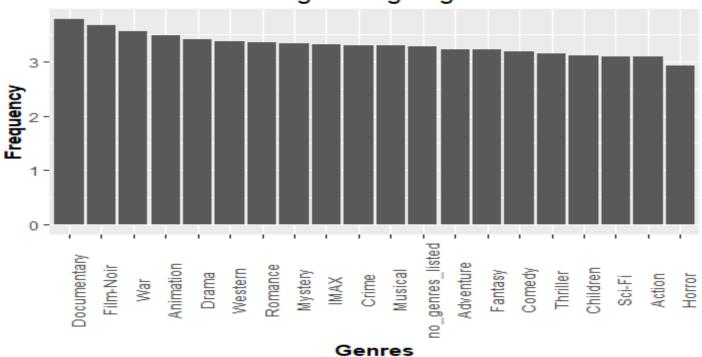
NOTE: The convoluted method of combining graphs has to be done because the scales of both the metrics are not comparable. If they are comparable, then you can just use two geom\_line() commands in the same plot

#### 5. Which genres are the best (in terms of ratings)?

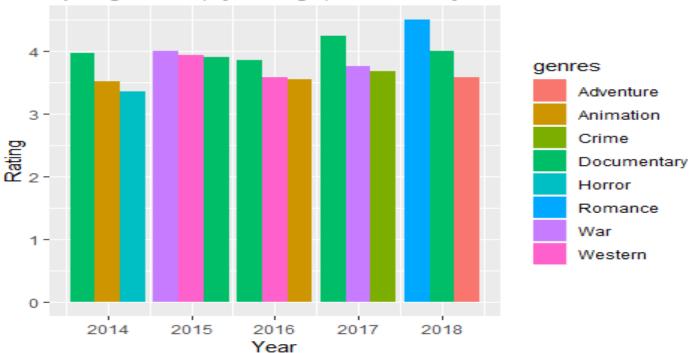
```
mov_ratings <- aggregate(rating ~ title, data = d2, FUN = mean, na.rm = TRUE)
df_temp <- merge(df, mov_ratings, by.x = "title", by.y = "title", all.x = TRUE)
df_temp <- df_temp[(!is.na(df_temp$rating) & !is.na(df_temp$year)), ]

# Overall
df_temp1 <- aggregate(rating ~ genres, data = df_temp, FUN = mean, na.rm = TRUE)
df_temp1 <- df_temp1[order(-df_temp1$rating),]
ggplot(df_temp1,aes(x = reorder(genres, -rating),y = rating), fill = genres) +
    geom_bar(stat="identity", position = "dodge") +
    xlab("Genres")+ylab("Frequency") + ggtitle("Average rating of genres") +
    theme(plot.title = element_text(hjust = 0.5),
        axis.title.x = element_text(face="bold"),
        axis.title.y = element_text(angle = 90),
        axis.title.y = element_text(face="bold"))</pre>
```

# Average rating of genres



# Top 3 genres (by ratings) in recent years



## 6. Most Frequently occuring tags

```
tag_freq <- aggregate(userId ~ tag, data = tags, FUN = length)
names(tag_freq)[2] <- "Tag_Freq"
temp <- tag_freq %>% top_n(n = 10, wt = Tag_Freq)

ggplot(temp,aes(x = reorder(tag, -Tag_Freq),y = Tag_Freq)) +
    geom_bar(stat="identity", position = "dodge") +
    xlab("Tag")+ylab("Frequency") + ggtitle("Frequently occuring tags") +
    theme(plot.title = element_text(hjust = 0.5),
        axis.title.x = element_text(face="bold"),
        axis.title.y = element_text(face="bold"))
```

# Frequently occuring tags

