

Computing Expected Violence Exposure with the Bootstrap

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Overview

Instead of using simple regression to produce a measure of “expected violence”, I implement a bootstrap procedure.

Method

First, I pull in all the data that I will need.

```
movie_ratings <- readRDS("movie_ratings_final.rds")
movie <- readRDS("movie.rds")
dd <- read_csv("fulliblockday.csv")
```

Then, I randomly partition the movies into 5 groups.

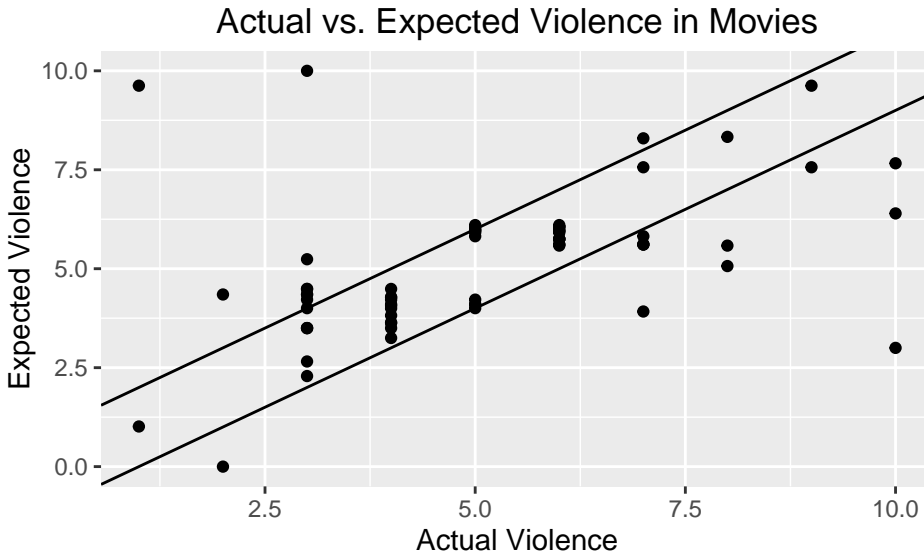
```
set.seed(392)
df <- movie_ratings[sample(nrow(movie_ratings)),] # Shuffle rows

m1 <- lm(Violence ~ Genre+MPAA_Rating, data=df[15:70,])
p1 <- predict(m1, df[1:14,])
m2 <- lm(Violence ~ Genre+MPAA_Rating, data=df[c(1:14, 29:70),])
p2 <- predict(m2, df[15:28,])
m3 <- lm(Violence ~ Genre+MPAA_Rating, data=df[c(1:28, 43:70),])
p3 <- predict(m3, df[c(29:37, 39:42),]) # 38 is the musical
m4 <- lm(Violence ~ Genre+MPAA_Rating, data=df[c(1:42, 57:70),])
p4 <- predict(m4, df[43:56,])
m5 <- lm(Violence ~ Genre+MPAA_Rating, data=df[1:56,])
p5 <- predict(m5, df[57:70,])

Exp_Viol_Boot <- c(p1,p2,p3,p4,p5)
Exp_Viol_Boot <- append(Exp_Viol_Boot, 4, after=37)
Exp_Viol_Boot <- ifelse(Exp_Viol_Boot<0, 0, Exp_Viol_Boot)

df <- df %>%
  bind_cols(data.frame(Exp_Viol_Boot))

ggplot(df, aes(x=Violence, y=Exp_Viol_Boot)) +
  geom_point() +
  labs(title="Actual vs. Expected Violence in Movies",
       y="Expected Violence",
       x="Actual Violence") +
  # geom_abline(slope=1, intercept=0) +
  geom_abline(slope=1, intercept=1) +
  geom_abline(slope=1, intercept=-1)
```



```
cor(Exp_Violence~Exp_Viol_Boot, data=df)
```

```
## [1] 0.8962002
```

We see that there is a decent amount of variation in the measures, which is good for our analyses. As a final note, we confirm that some examples of movies with large residuals are actually unexpectedly violent/nonviolent.

```
foo <- df %>%
  mutate(Resid_Violence = Exp_Viol_Boot-Violence) %>%
  arrange(Resid_Violence)
head(foo)
```

```
##           Title  Genre
## 1      Hannibal Horror
## 2 The Passion of the Christ Drama
## 3    The Perfect Storm Drama
## 4      Scary Movie Comedy
## 5    Jurassic Park 3 Action
## 6      Gladiator Action
##
##                                     Info Year
## 1      R for strong gruesome violence, some nudity and language 2001
## 2                                     R for sequences of graphic violence 2004
## 3                                     PG-13 for language and scenes of peril 2000
## 4 R for strong crude sexual humor, language, drug use and violence 2000
## 5                                     PG-13 for intense sci-fi terror and violence 2001
## 6                                     R for intense, graphic combat 2000
##   MPAA_Rating Violence viol_strong viol_mild viol_non Exp_Violence
## 1          R      10      TRUE      FALSE      FALSE      6.500000
## 2          R      10      TRUE      FALSE      FALSE      7.245825
## 3        PG13       7      FALSE      TRUE      FALSE      5.262526
## 4          R       8      TRUE      FALSE      FALSE      5.915784
## 5        PG13       8      TRUE      FALSE      FALSE      6.005567
## 6          R      10      TRUE      FALSE      FALSE      7.988866
##   Exp_Viol_Boot Resid_Violence
## 1      3.000000      -7.000000
## 2      6.397261      -3.602739
```

```
## 3      3.919039      -3.080961
## 4      5.067278      -2.932722
## 5      5.584124      -2.415876
## 6      7.663662      -2.336338
```

Looking at the top 6 movies that are “more violent than expected”, we see that our measure performs decently well. Many of these films are in fact more violent than one would perhaps anticipate. We note that some movies, such as Hannibal and Gladiator, should be expected to be violent, and thus should not really be on this list. More controls may produce a better measure, but for now, I proceed with the current results.

```
foo <- foo %>%
  arrange(desc(Resid_Violence))
head(foo)
```

```
##           Title           Genre
## 1      Erin Brockovich      Drama
## 2    The Blair Witch Project  Horror
## 3      Meet the Parents      Comedy
## 4 There's Something About Mary Romantic Comedy
## 5           Shark Tale      Adventure
## 6           Ice Age        Adventure
##
##           Info Year MPAA_Rating
## 1           R for language 2000      R
## 2           R 1999              R
## 3 PG-13 for sexual content, drug references and language 2000    PG13
## 4           R for strong comic sexual content and language 1998      R
## 5           PG for mild language and crude humor 2004      PG
## 6           PG for mild peril 2002      PG
##  Violence viol_strong viol_mild viol_non Exp_Violence Exp_Viol_Boot
## 1         1      FALSE      FALSE      TRUE      7.245825      9.624386
## 2         3      FALSE      FALSE      TRUE      6.500000     10.000000
## 3         2      FALSE      FALSE      TRUE      3.932486      4.349131
## 4         3      FALSE      FALSE      TRUE      4.402412      5.239582
## 5         3      FALSE      FALSE      TRUE      4.231507      4.489142
## 6         3      FALSE      FALSE      TRUE      4.231507      4.489142
##  Resid_Violence
## 1      8.624386
## 2      7.000000
## 3      2.349131
## 4      2.239582
## 5      1.489142
## 6      1.489142
```

The top 6 movies that are “less violent than expected” seem to make sense as well, though somewhat less so than the “more violent” movies. R-rated movies with little to no violence seem to cause the model some issues. Again, more controls may help here, but for now I’ll proceed with the current results.

```
movie <- movie %>%
  left_join(select(df, Title, Exp_Viol_Boot), by="Title")

movie <- movie %>%
  mutate(exp_viol_strong = Exp_Viol_Boot>=7, # Note: 7, not 8 because of expected violence model
         exp_viol_mild = Exp_Viol_Boot>4 & Exp_Viol_Boot<7,
         exp_viol_non = Exp_Viol_Boot<=4) %>%
  mutate(more_violent = Exp_Viol_Boot-Violence <= -1,
         as_violent = (Exp_Viol_Boot-Violence >= -1) & (Exp_Viol_Boot-Violence < 1),
```

```

        less_violent = Exp_Viol_Boot-Violence >= 1)

daily_exposure <- movie %>%
  group_by(Date) %>%
  summarise(tickets_tot = sum(Tickets),
            tickets_strong = sum(ifelse(viol_strong, Tickets, 0)),
            tickets_mild = sum(ifelse(viol_mild, Tickets, 0)),
            tickets_non = sum(ifelse(viol_non, Tickets, 0)),
            tickets_exp_strong = sum(ifelse(exp_viol_strong, Tickets, 0)),
            tickets_exp_mild = sum(ifelse(exp_viol_mild, Tickets, 0)),
            tickets_exp_non = sum(ifelse(exp_viol_non, Tickets, 0)),
            tickets_more_violent = sum(ifelse(more_violent, Tickets, 0)),
            tickets_as_violent = sum(ifelse(as_violent, Tickets, 0)),
            tickets_less_violent = sum(ifelse(less_violent, Tickets, 0)),
            tickets_to_violence = tickets_non + 2*tickets_mild + 3*tickets_strong,
            tickets_to_exp_violence = tickets_exp_non + 2*tickets_exp_mild + 3*tickets_exp_strong,
            tickets_to_aggviol = sum(Tickets*Violence),
            tickets_to_aggexpviol = sum(Tickets*Exp_Viol_Boot))

```

Exporting Data

With the full data frame compiled, I export the data for analysis.

```

master_boot <- dd %>%
  mutate(Date = as.Date(mdy, origin="1960-01-01 UTC")) %>%
  left_join(daily_exposure, by="Date") %>%
  filter(year(Date) >= 1998) %>%
  mutate(ln_actstrong = log(tickets_strong),
         ln_actmild = log(tickets_mild),
         ln_actnon = log(tickets_non)) %>%
  mutate(ln_actstrong = ifelse(ln_actstrong=="-Inf", 0, as.numeric(ln_actstrong)),
         ln_actmild = ifelse(ln_actmild=="-Inf", 0, as.numeric(ln_actmild)),
         ln_actnon = ifelse(ln_actnon=="-Inf", 0, as.numeric(ln_actnon))) %>% # Use 0 since nonzero v
  mutate(ln_expstrong = log(tickets_exp_strong),
         ln_expmild = log(tickets_exp_mild),
         ln_expnnon = log(tickets_exp_non)) %>%
  mutate(ln_expstrong = ifelse(ln_expstrong=="-Inf", 0, as.numeric(ln_expstrong)),
         ln_expmild = ifelse(ln_expmild=="-Inf", 0, as.numeric(ln_expmild)),
         ln_expnnon = ifelse(ln_expnnon=="-Inf", 0, as.numeric(ln_expnnon))) %>%
  mutate(ln_more = log(tickets_more_violent),
         ln_as = log(tickets_as_violent),
         ln_less = log(tickets_less_violent)) %>%
  mutate(ln_more = ifelse(ln_more=="-Inf", 0, as.numeric(ln_more)),
         ln_as = ifelse(ln_as=="-Inf", 0, as.numeric(ln_as)),
         ln_less = ifelse(ln_less=="-Inf", 0, as.numeric(ln_less))) %>%
  mutate(ln_ttv = log(tickets_to_violence),
         ln_ttev = log(tickets_to_exp_violence)) %>%
  mutate(ln_ttv = ifelse(ln_ttv=="-Inf", 0, as.numeric(ln_ttv)),
         ln_ttev = ifelse(ln_ttev=="-Inf", 0, as.numeric(ln_ttev))) %>%
  mutate(ln_ttav = log(tickets_to_aggviol),
         ln_ttaev = log(tickets_to_aggexpviol)) %>%
  mutate(ln_ttav = ifelse(ln_ttav=="-Inf", 0, as.numeric(ln_ttav)),
         ln_ttaev = ifelse(ln_ttaev=="-Inf", 0, as.numeric(ln_ttaev)))

```

```
write_csv(master_boot, path="master_boot.csv")  
saveRDS(master_boot, "master_boot.rds")
```