

# Data 115 Final Project: Nintendo Game Analysis

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## The Big Question

What would have been the ideal game for Nintendo to release during the 2010s? How do North America's and Japan's tastes differ in this hypothetical game?

This analysis attempts to find characteristics (genre, platform, franchise) of a Nintendo game that would have sold the most number of copies during the 2010s. This curiosity extends to determine if Japan, the home country of Nintendo, would have differing preferences in games than North America.

## The Data

The data used for this analysis was taken from [Kaggle](#). The original poster, Gregory Smith, claims to have scraped the data from [vgdata](#) using a script based on BeautifulSoup using Python. The record consists of more than 16,500 games with sales (defined by copies sold) greater than 100,000. However, two games were removed due to incompleteness, and little else is mentioned about this dataset's processing or its original purpose.

The variables in this dataset are the game's name, platform, year released, genre, and publisher. It also includes the number of copies sold in millions from North America, Europe, Japan, other (rest of the world), and total globally. Lastly, a rank is included based on global sales. Unfortunately, the original author does not specify the period of time the sales were based on, but judging on the most recent games included, it can assume it is not current as of December 2022.

Nonetheless, I selected this data for this project because I have a personal interest in video games and a soft spot for Nintendo. This data will also be sufficient to answer my big question. Sadly, the most recent Nintendo game in this dataset is from 2016 and does not include the newest console release, the Nintendo Switch, which saw massive success and brought an onslaught of new games. Therefore, modern predictions are not appropriate, which is why I am limiting my research question to the 2010s.

## Initial Cleaning and Processing

```
vgsales.df = read.csv("~/Documents/School/Data 115/FinalProject/vgsales.csv")
```

### Managing Missing Years:

I first noticed some missing values in the Year column. To handle this, I converted Year into a numeric value, then replaced each NA with the mean of all the years (2006).

```
#replacing n/a with mean
vgsales.df$Year = as.numeric(vgsales.df$Year)
summary(vgsales.df$Year)
#mean of years = 2006
vgsales.df$Year[is.na(vgsales.df$Year)] = 2006
```

### Splitting Up Nintendo:

It was at this point I decided to narrow my research question to Nintendo publishing, so a new data frame was created to only hold Nintendo games.

```
nintendo.df = vgsales.df[vgsales.df$Publisher == "Nintendo",]
summary(nintendo.df)
```

Rank	Name	Year	Genre	NA_Sales
Min. : 1	Length:703	Min. :1983	Length:703	Min. : 0.000
1st Qu.: 719	Class :character	1st Qu.:2000	Class :character	1st Qu.: 0.005
Median : 2335	Mode :character	Median :2005	Mode :character	Median : 0.370
Mean : 3861	Platform	Mean :2004	Publisher	Mean : 1.162
3rd Qu.: 6010	Length:703	3rd Qu.:2009	Length:703	3rd Qu.: 0.960
Max. :16545	Class :character	Max. :2016	Class :character	Max. :41.490
	Mode :character		Mode :character	

EU_Sales	JP_Sales	Other_Sales	Global_Sales
Min. : 0.0000	1st Qu.: 0.0900	Min. :0.0000	Min. : 0.010
1st Qu.: 0.0000	Min. : 0.0000	1st Qu.:0.0000	1st Qu.: 0.290
Median : 0.1200	Median : 0.2800	Median :0.0300	Median : 0.890
Mean : 0.5956	Mean : 0.6478	Mean :0.1356	Mean : 2.541
3rd Qu.: 0.4800	3rd Qu.: 0.7300	3rd Qu.:0.0900	3rd Qu.: 2.250
Max. :29.0200	Max. :10.2200	Max. :8.4600	Max. :82.740

The resulting Nintendo data frame holds 703 games, which is quite the decrease from the original data set. But this size decreased further by only including games released in the 2010 decade.

```
nintendoRecent.df = nintendo.df[nintendo.df$Year >= 2010,]
```

This left a sample size of 170. Not ideal compared to the original data set, but not terrible.

### Creating “Franchise” Column:

Franchises play a major role in sales especially for a brand like Nintendo. Because Nintendo houses a large number of dominating gaming franchises, a new column was created to reflect the associated franchise of the game. To accomplish this, specific words were searched in the game's name to determine if a game was marketed within a certain franchise.

```
library(stringr)

#formalize the name strings to lowercase
nintendo.df$name = str_to_lower(nintendo.df$name)

nintendo.df$Franchise = "Other"
nintendo.df$Franchise[str_count(nintendo.df$name, "pokemon") != 0] = "Pokemon"
nintendo.df$Franchise[str_count(nintendo.df$name, "pokémon") != 0] = "Pokemon"
nintendo.df$Franchise[str_count(nintendo.df$name, "mario") != 0] = "Mario"
nintendo.df$Franchise[str_count(nintendo.df$name, "yoshi") != 0] = "Yoshi"
nintendo.df$Franchise[str_count(nintendo.df$name, "zelda") != 0] = "Zelda"
nintendo.df$Franchise[str_count(nintendo.df$name, "animal crossing") != 0] = "Animal Crossing"
nintendo.df$Franchise[str_count(nintendo.df$name, "donkey kong") != 0] = "Donkey Kong"
nintendo.df$Franchise[str_count(nintendo.df$name, "kirby") != 0] = "Kirby"
nintendo.df$Franchise[str_count(nintendo.df$name, "wii") != 0] = "Wii Exclusive"
nintendo.df$Franchise[str_count(nintendo.df$name, "smash bro") != 0] = "Smash Bros"
```

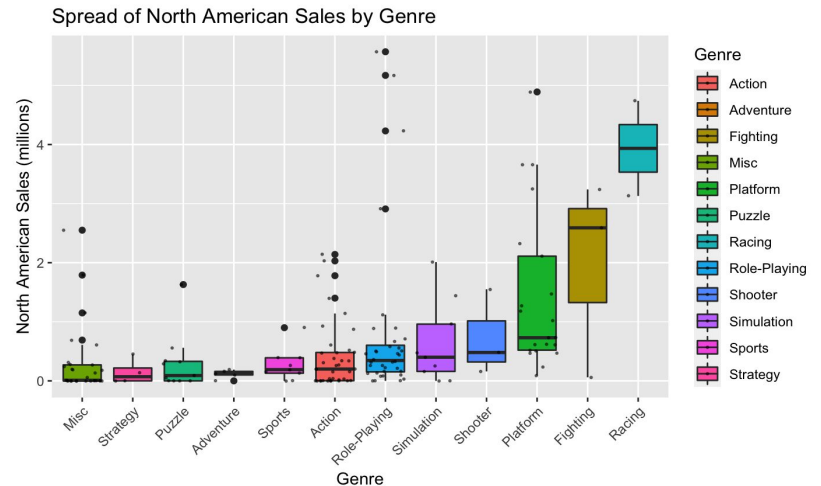
It is important to note that duplicates in names were ignored as these duplicates may have different platforms, and therefore technically a different entry.

## Exploratory Analysis

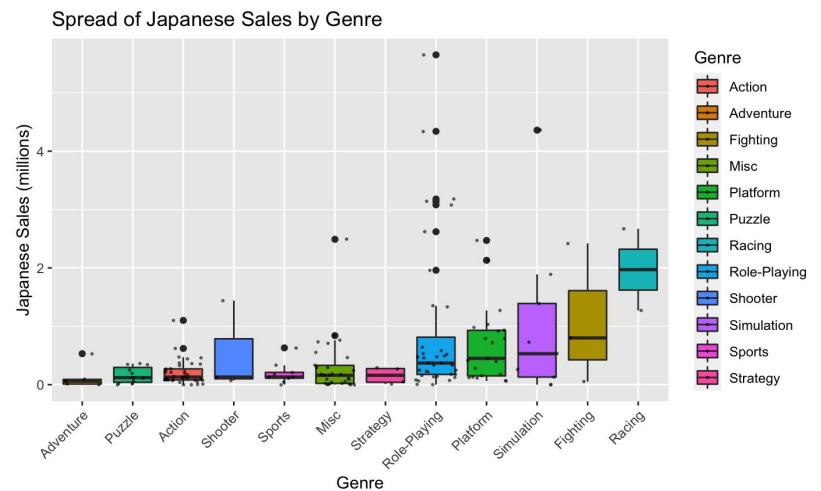
This section examines how each variable (genre, platform, franchise) influenced the sales from North America, Japan, and globally.

### Genre:

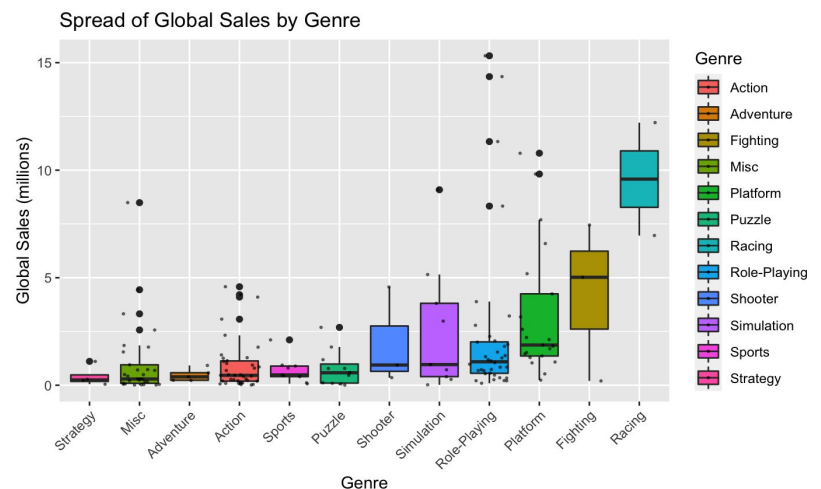
```
ggplot(nintendoRecent.df, aes(x =
reorder(Genre, NA_Sales, FUN =
median), y = NA_Sales, fill= Genre)) +
  geom_boxplot() +
  scale_x_discrete(guide =
guide_axis(angle = 45)) +
  labs(x = "Genre", y = "North
American Sales (millions)", title =
"Spread of North American Sales by
Genre") +
  geom_jitter(color="black", size=0.4,
alpha=0.5)
```



```
ggplot(nintendoRecent.df, aes(x =
reorder(Genre, JP_Sales, FUN =
median), y = JP_Sales, fill= Genre)) +
  geom_boxplot() +
  scale_x_discrete(guide =
guide_axis(angle = 45)) +
  labs(x = "Genre", y = "Japanese
Sales (millions)", title = "Spread of
Japanese Sales by Genre")+
  geom_jitter(color="black", size=0.4,
alpha=0.5)
```



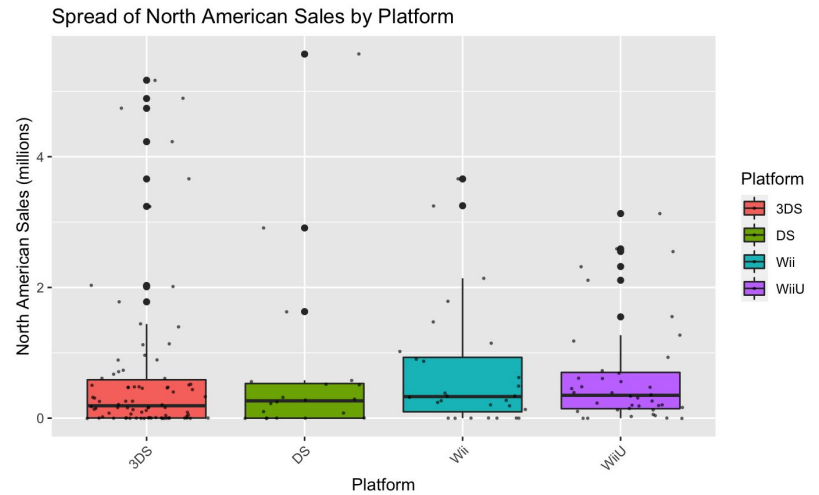
```
ggplot(nintendoRecent.df, aes(x =
reorder(Genre, Global_Sales, FUN =
median), y = Global_Sales, fill=
Genre)) +
  geom_boxplot() +
  scale_x_discrete(guide =
guide_axis(angle = 45)) +
  labs(x = "Genre", y = "Global Sales
(millions)", title = "Spread of Global
Sales by Genre")+
  geom_jitter(color="black", size=0.4,
alpha=0.5)
```



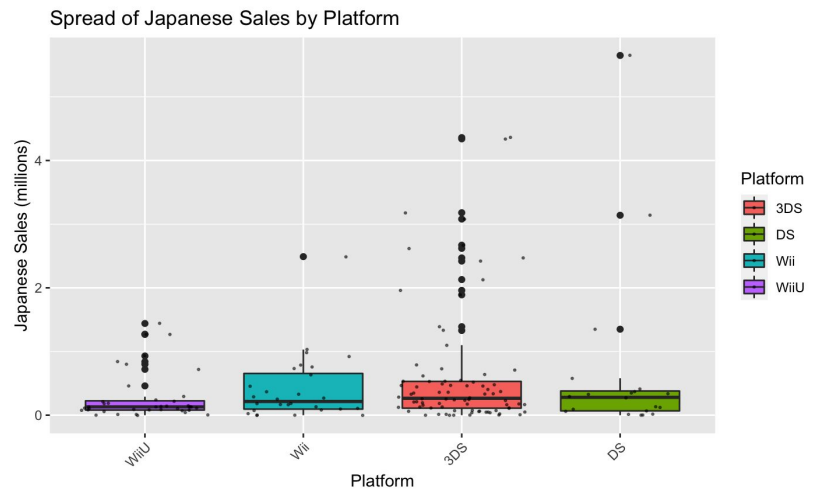
From these plots, it becomes evident that racing is the genre with the highest median sales across all three regions with fighting in second.

## Platform:

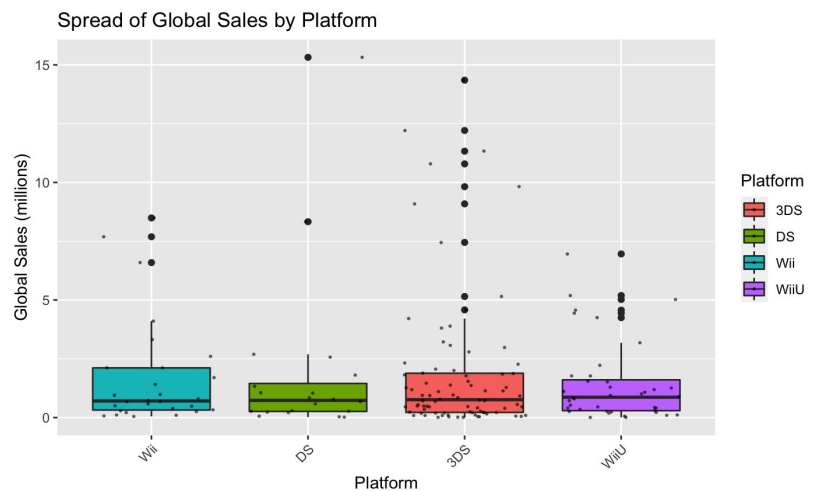
```
ggplot(nintendoRecent.df, aes(x =
reorder(Platform, NA_Sales, FUN =
median), y = NA_Sales, fill =
Platform)) +
  geom_boxplot() +
  scale_x_discrete(guide =
guide_axis(angle = 45)) +
  labs(x = "Platform", y = "North
American Sales (millions)", title =
"Spread of North American Sales by
Platform") +
  geom_jitter(color="black", size=0.4,
alpha=0.5)
```



```
ggplot(nintendoRecent.df, aes(x =
reorder(Platform, JP_Sales, FUN =
median), y = JP_Sales, fill =
Platform)) +
  geom_boxplot() +
  scale_x_discrete(guide =
guide_axis(angle = 45)) +
  labs(x = "Platform", y = "Japanese
Sales (millions)", title = "Spread of
Japanese Sales by Platform") +
  geom_jitter(color="black", size=0.4,
alpha=0.5)
```



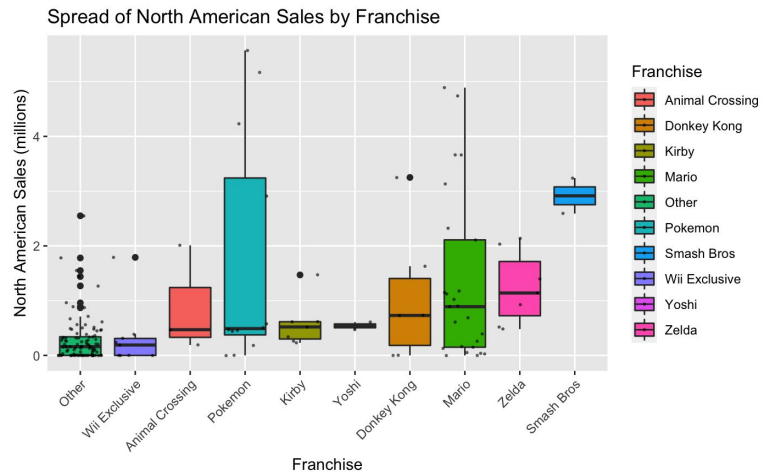
```
ggplot(nintendoRecent.df, aes(x =
reorder(Platform, Global_Sales, FUN =
median), y = Global_Sales, fill =
Platform)) +
  geom_boxplot() +
  scale_x_discrete(guide =
guide_axis(angle = 45)) +
  labs(x = "Platform", y = "NGlobal
Sales (millions)", title = "Spread of
Global Sales by Platform") +
  geom_jitter(color="black", size=0.4,
alpha=0.5)
```



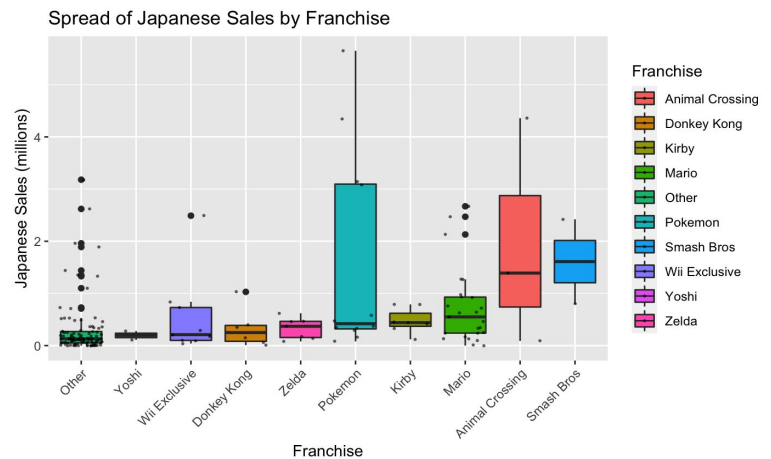
Based on these plots, we can conclude that the platform has very little to do with sales success as all medians have very little difference between each other, and the three region's leading platforms are inconsistent.

## Franchise:

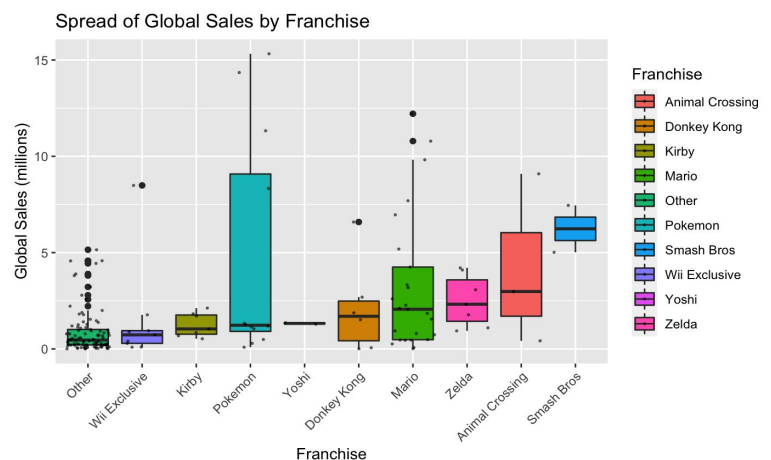
```
ggplot(nintendoRecent.df, aes(x =
  reorder(Franchise, NA_Sales, FUN = median),
  y = NA_Sales, fill = Franchise)) +
  geom_boxplot() +
  scale_x_discrete(guide = guide_axis(angle = 45)) +
  labs(x = "Franchise", y = "North American
  Sales (millions)", title = "Spread of North
  American Sales by Franchise") +
  geom_jitter(color="black", size=0.4,
  alpha=0.5)
```



```
ggplot(nintendoRecent.df, aes(x =
  reorder(Franchise, JP_Sales, FUN =
  median), y = JP_Sales, fill = Franchise)) +
  geom_boxplot() +
  scale_x_discrete(guide =
  guide_axis(angle = 45)) +
  labs(x = "Franchise", y = "Japanese
  Sales (millions)", title = "Spread of
  Japanese Sales by Franchise") +
  geom_jitter(color="black", size=0.4,
  alpha=0.5)
```



```
ggplot(nintendoRecent.df, aes(x =
  reorder(Franchise, Global_Sales, FUN =
  median), y = Global_Sales, fill =
  Franchise)) +
  geom_boxplot() +
  scale_x_discrete(guide =
  guide_axis(angle = 45)) +
  labs(x = "Franchise", y = "Global Sales
  (millions)", title = "Spread of Global
  Sales by Franchise") +
  geom_jitter(color="black", size=0.4,
  alpha=0.5)
```



As seen in these plots, Smash Bros appears to lead all three regions in sales based on franchise. Animal Crossing comes in second for Japan and global, whereas North America's second is held by Zelda. Interestingly, Pokemon in all three regions has an extremely wide interquartile range.

From these nine plots, we can estimate that platform will have little effect on sales. On the other hand, genre and franchise may have a much higher effect. While all three regions (North America, Japan, global) appear to appreciate the Smash Bros. franchise and racing genre the most, North America seems to appreciate the Animal Crossing franchise far less than Japan, but is more receptive to Zelda. Universally, Pokemon games appear to receive varying degrees of success as each has a large spread. Finally, franchises made exclusively for the Wii and those unaffiliated with larger franchises (other) consistently rank last across all regions.

## Modeling

To model this data, various multiple regressions were used where the independent variables were platform, genre, and franchise, and the dependent variables were North American sales, Japanese sales, and global sales.

This method of analysis was chosen because of its ability to predict a numerical output (sales) based on multiple categorical input (platform, genre, franchise) for an ideal hypothetical game.

The pro to this method are the predictions the model will allow given various input variables. However, the con is that the input variables have a high chance of collinearity. This results in a model that is not optimally fitted.

Alternative methods were scarce given the current statistical knowledge I possess.

### Checking for Collinearity:

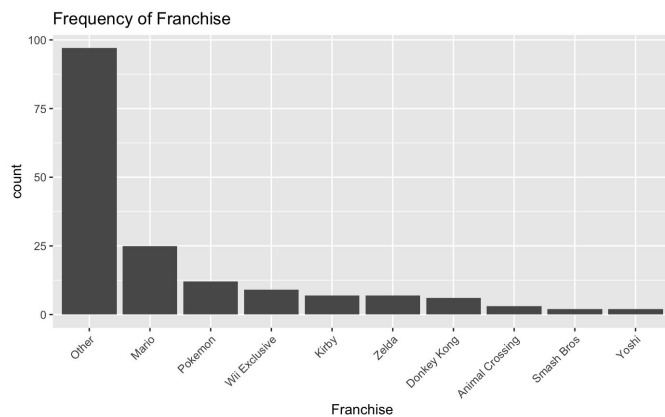
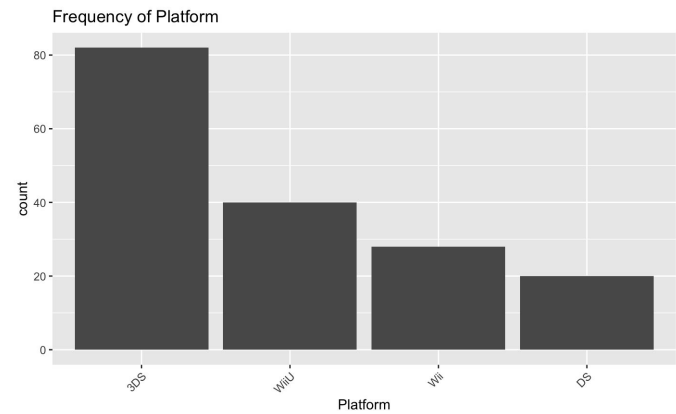
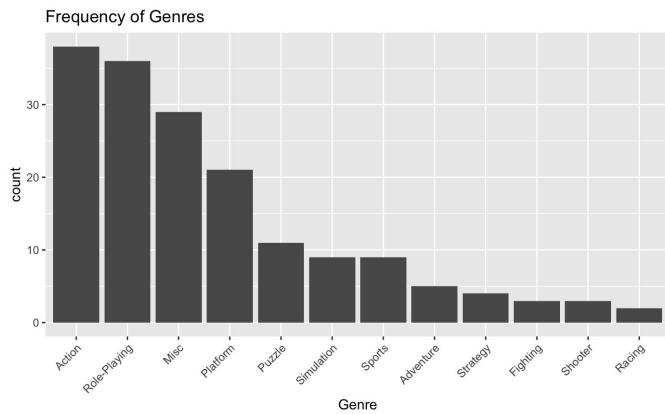
There was great suspicion of collinearity between genre and franchise because franchises typically stick to one or two genres of gameplay. To determine this suspicion's accuracy, a frequency table between genre and franchise was created.

```
table(nintendoRecent.df$Genre, nintendoRecent.df$Franchise)
```

	Animal Crossing	Donkey Kong	Kirby	Mario	Other	Pokemon	Smash Bros	Wii Exclusive	Yoshi	Zelda
Action	0	0	2	2	24	2	0	1	0	7
Adventure	0	0	0	0	5	0	0	0	0	0
Fighting	0	0	0	0	1	0	2	0	0	0
Misc	1	0	0	4	16	2	0	6	0	0
Platform	0	3	5	7	4	0	0	0	2	0
Puzzle	0	3	0	1	7	0	0	0	0	0
Racing	0	0	0	2	0	0	0	0	0	0
Role-Playing	0	0	0	3	25	8	0	0	0	0
Shooter	0	0	0	0	3	0	0	0	0	0
Simulation	2	0	0	0	7	0	0	0	0	0
Sports	0	0	0	6	1	0	0	2	0	0
Strategy	0	0	0	0	4	0	0	0	0	0

As expected, this table is extremely sparse and uneven. In fact, when creating a bar chart of each input variable, it is evident that each group's sizes are wildly uneven.

```
library(forcats)
ggplot(nintendoRecent.df, aes(x = fct_infreq(Genre))) + geom_bar()
+
  labs(x = "Genre", title = "Frequency of Genres") +
  scale_x_discrete(guide = guide_axis(angle = 45))
gsgplot(nintendoRecent.df, aes(x = fct_infreq(Platform))) +
geom_bar() +
  labs(x = "Platform", title = "Frequency of Platform") +
  scale_x_discrete(guide = guide_axis(angle = 45))
ggplot(nintendoRecent.df, aes(x = fct_infreq(Franchise))) +
geom_bar() +
  labs(x = "Franchise", title = "Frequency of Franchise") +
  scale_x_discrete(guide = guide_axis(angle = 45))
```



While this uneven disbursement does not bode well for the model's fit, a more formal check for collinearity was performed using Chi-squared tests.

```
chisq.test(nintendoRecent.df$Genre, nintendoRecent.df$Franchise)
chisq.test(nintendoRecent.df$Genre, nintendoRecent.df$Platform)
chisq.test(nintendoRecent.df$Platform, nintendoRecent.df$Franchise)
```

Pearson's Chi-squared test

```
data: nintendoRecent.df$Genre and nintendoRecent.df$Franchise
X-squared = 322.15, df = 99, p-value < 2.2e-16
```

Pearson's Chi-squared test

```
data: nintendoRecent.df$Genre and nintendoRecent.df$Platform
X-squared = 46.066, df = 33, p-value = 0.06496
```

Pearson's Chi-squared test

```
data: nintendoRecent.df$Platform and nintendoRecent.df$Franchise
X-squared = 43.986, df = 27, p-value = 0.02075
```

With p-values of less than 0.5 for the first and third Chi-squared tests, it is concluded that genre and franchise, and platform and franchise are not independent of each other and thus, have collinearity.

Given the wild unevenness of cell sizes and the results of the Chi-squared tests, the model made will not accurately reflect the data as well as one would hope. Despite this, interesting insights can still be made however good the fit.

## Making the Models:

For each region, a simple model was made for genre, platform, and franchise before combining all three into a multiple regression model.

## North America:

```
summary(lm(NA_Sales ~ Genre, nintendoRecent.df))
summary(lm(NA_Sales ~ Platform, nintendoRecent.df))
summary(lm(NA_Sales ~ Franchise, nintendoRecent.df))
nintendoRecentNA.lm = lm(NA_Sales ~ Genre + Franchise + Platform, data = nintendoRecent.df)
summary(nintendoRecentNA.lm)
```

Call:

```
lm(formula = NA_Sales ~ Genre, data = nintendoRecent.df)
```

Residuals:

Min	1Q	Median	3Q	Max
-1.9033	-0.4158	-0.2491	0.0617	4.7558

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.4158	0.1529	2.720	0.00726 **
GenreAdventure	-0.3018	0.4483	-0.673	0.50181
GenreFighting	1.5475	0.5651	2.738	0.00688 **
GenreMisc	-0.1172	0.2324	-0.504	0.61478
GenrePlatform	1.0328	0.2562	4.031	8.63e-05 ***
GenrePuzzle	-0.1176	0.3226	-0.365	0.71596
GenreRacing	3.5192	0.6836	5.148	7.74e-07 ***
GenreRole-Playing	0.3984	0.2192	1.818	0.07101 .
GenreShooter	0.3142	0.5651	0.556	0.57900
GenreSimulation	0.2164	0.3493	0.620	0.53644
GenreSports	-0.1480	0.3493	-0.424	0.67237
GenreStrategy	-0.2683	0.4953	-0.542	0.58885

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.9423 on 158 degrees of freedom

Multiple R-squared: 0.27, Adjusted R-squared: 0.2192  
F-statistic: 5.313 on 11 and 158 DF, p-value: 3.966e-07

Call:

```
lm(formula = NA_Sales ~ Franchise, data = nintendoRecent.df)
```

Residuals:

Min	1Q	Median	3Q	Max
-1.710	-0.282	-0.152	0.093	3.860

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.8900	0.5385	1.653	0.1003
FranchiseDonkey Kong	0.1667	0.6595	0.253	0.8008
FranchiseKirby	-0.3114	0.6436	-0.484	0.6291
FranchiseMario	0.4412	0.5699	0.774	0.4400
FranchiseOther	-0.6080	0.5467	-1.112	0.2678
FranchisePokemon	0.8200	0.6020	1.362	0.1751
FranchiseSmash Bros	2.0250	0.8514	2.378	0.0186 *
FranchiseWii Exclusive	-0.5656	0.6218	-0.910	0.3644
FranchiseYoshi	-0.3500	0.8514	-0.411	0.6816
FranchiseZelda	0.3443	0.6436	0.535	0.5934

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.9327 on 160 degrees of freedom

Multiple R-squared: 0.2759, Adjusted R-squared: 0.2351  
F-statistic: 6.773 on 9 and 160 DF, p-value: 3.276e-08

Call:

```
lm(formula = NA_Sales ~ Platform, data = nintendoRecent.df)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.7157	-0.6128	-0.3984	-0.0330	4.8790

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.643049	0.118789	5.413	2.13e-07 ***
PlatformDS	0.047951	0.268263	0.179	0.858
PlatformWiiU	0.072666	0.235447	0.309	0.758
PlatformWiiU	0.009201	0.207456	0.044	0.965

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.076 on 166 degrees of freedom

Multiple R-squared: 0.0006802, Adjusted R-squared: -0.01738  
F-statistic: 0.03766 on 3 and 166 DF, p-value: 0.9902

Call:

```
lm(formula = NA_Sales ~ Genre + Franchise + Platform, data = nintendoRecent.df)
```

Residuals:

Min	1Q	Median	3Q	Max
-1.4810	-0.3437	-0.0717	0.1831	3.6801

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.67709	0.55078	1.229	0.2209
GenreAdventure	-0.05727	0.41035	-0.140	0.8892
GenreFighting	-0.02334	0.84046	-0.028	0.9779
GenreMisc	-0.12710	0.22820	-0.557	0.5784
GenrePlatform	1.24738	0.27970	4.460	1.63e-05 ***
GenrePuzzle	0.04470	0.31180	0.143	0.8862
GenreRacing	3.20247	0.63727	5.025	1.45e-06 ***
GenreRole-Playing	0.31929	0.21228	1.504	0.1347
GenreShooter	0.67875	0.50354	1.348	0.1798
GenreSimulation	0.40699	0.34309	1.186	0.2375
GenreSports	-0.36387	0.35076	-1.037	0.3013
GenreStrategy	0.05380	0.44365	0.121	0.9036
FranchiseDonkey Kong	-0.30876	0.65216	-0.473	0.6366
FranchiseKirby	-1.10706	0.64235	-1.723	0.0869 .
FranchiseMario	0.07952	0.56197	0.141	0.8877
FranchiseOther	-0.59375	0.52285	-1.136	0.2580
FranchisePokemon	0.80391	0.58959	1.364	0.1748
FranchiseSmash Bros	2.28532	1.13648	2.011	0.0462 *
FranchiseWii Exclusive	-0.31037	0.59838	-0.519	0.6048
FranchiseYoshi	-1.36040	0.81947	-1.660	0.0990 .
FranchiseZelda	0.53374	0.62615	0.852	0.3954
PlatformDS	0.08957	0.22176	0.404	0.6869
PlatformWiiU	0.26049	0.19706	1.322	0.1883
PlatformWiiU	-0.04814	0.17340	-0.278	0.7817

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.8236 on 146 degrees of freedom

Multiple R-squared: 0.4847, Adjusted R-squared: 0.4036  
F-statistic: 5.972 on 23 and 146 DF, p-value: 3.774e-12

From the North American simple models and their p-values, it is observed that genres of fighting, platforming, and racing are statistically significant, meaning that their influence on North American sales is not due to chance. Platform has no significance, but the Smash Bros. franchise does.

From the multiple regression model, a similar result is achieved. However, fighting as a genre is not statistically significant.



Japan:

```
summary(lm(JP_Sales ~ Genre, nintendoRecent.df))
summary(lm(JP_Sales ~ Platform, nintendoRecent.df))
summary(lm(JP_Sales ~ Franchise, nintendoRecent.df))
nintendoRecentJP.lm = lm(JP_Sales ~ Genre + Franchise + Platform, data = nintendoRecent.df)
summary(nintendoRecentJP.lm)
```

Call:

```
lm(formula = JP_Sales ~ Genre, data = nintendoRecent.df)
```

Residuals:

```
      Min       1Q   Median       3Q      Max
-1.0400 -0.3578 -0.1266  0.1354  4.6972
```

Coefficients:

```
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    0.20658    0.13137   1.572 0.117846
GenreAdventure -0.06858    0.38526  -0.178 0.858947
GenreFighting   0.88342    0.48567   1.819 0.070808 .
GenreMisc       0.10342    0.19969   0.518 0.605239
GenrePlatform   0.47866    0.22020   2.174 0.031216 *
GenrePuzzle     -0.04203    0.27727  -0.152 0.879700
GenreRacing     1.76342    0.58752   3.001 0.003124 **
GenreRole-Playing 0.74620    0.18835   3.962 0.000112 ***
GenreShooter    0.34009    0.48567   0.700 0.484803
GenreSimulation  0.82564    0.30022   2.750 0.006652 **
GenreSports    -0.01213    0.30022  -0.040 0.967810
GenreStrategy  -0.05158    0.42570  -0.121 0.903716
---
```

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
Residual standard error: 0.8098 on 158 degrees of freedom
Multiple R-squared:  0.1847,    Adjusted R-squared:  0.128
F-statistic: 3.254 on 11 and 158 DF,  p-value: 0.0004951
```

Call:

```
lm(formula = JP_Sales ~ Franchise, data = nintendoRecent.df)
```

Residuals:

```
      Min       1Q   Median       3Q      Max
-1.8567 -0.2921 -0.1721  0.0645  4.0800
```

Coefficients:

```
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    1.9467    0.4540   4.288 3.11e-05 ***
FranchiseDonkey Kong -1.6133    0.5561  -2.901 0.004239 **
FranchiseKirby   -1.4710    0.5427  -2.711 0.007448 **
FranchiseMario  -1.2131    0.4805  -2.525 0.012555 *
FranchiseOther   -1.6446    0.4610  -3.568 0.000476 ***
FranchisePokemon -0.3767    0.5076  -0.742 0.459151
FranchiseSmash Bros -0.3367    0.7179  -0.469 0.639722
FranchiseWii Exclusive -1.3944    0.5243  -2.660 0.008613 **
FranchiseYoshi   -1.7517    0.7179  -2.440 0.015775 *
FranchiseZelda   -1.6167    0.5427  -2.979 0.003341 **
---
```

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
Residual standard error: 0.7864 on 160 degrees of freedom
Multiple R-squared:  0.2215,    Adjusted R-squared:  0.1777
F-statistic: 5.059 on 9 and 160 DF,  p-value: 5.171e-06
```

Call:

```
lm(formula = JP_Sales ~ Platform, data = nintendoRecent.df)
```

Residuals:

```
      Min       1Q   Median       3Q      Max
-0.6790 -0.4270 -0.2337 -0.0635  4.9710
```

Coefficients:

```
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    0.63695    0.09500   6.705 3.02e-10 ***
PlatformDS      0.04205    0.21454   0.196  0.8449
PlatformWii    -0.22659    0.18830  -1.203  0.2305
PlatformWiiU   -0.35595    0.16591  -2.145  0.0334 *
```

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
Residual standard error: 0.8603 on 166 degrees of freedom
Multiple R-squared:  0.03343,    Adjusted R-squared:  0.01596
F-statistic: 1.914 on 3 and 166 DF,  p-value: 0.1293
```

Call:

```
lm(formula = JP_Sales ~ Genre + Franchise + Platform, data = nintendoRecentJP.lm)
```

Residuals:

```
      Min       1Q   Median       3Q      Max
-1.4461 -0.3227 -0.0613  0.1784  3.8941
```

Coefficients:

```
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    1.716666    0.494499   3.472 0.000681 ***
GenreAdventure    0.003611    0.368420   0.010 0.992194
GenreFighting   -0.113365    0.754579  -0.150 0.880785
GenreMisc       -0.055841    0.204879  -0.273 0.785578
GenrePlatform    0.571319    0.251120   2.275 0.024357 *
GenrePuzzle     0.036963    0.279944   0.132 0.895137
GenreRacing     1.562225    0.572151   2.730 0.007104 **
GenreRole-Playing 0.565091    0.190587   2.965 0.003537 **
GenreShooter    0.587294    0.452086   1.299 0.195966
GenreSimulation  0.525916    0.308036   1.707 0.089888 .
GenreSports    -0.223472    0.314917  -0.710 0.479070
GenreStrategy    0.073166    0.398322   0.184 0.854516
FranchiseDonkey Kong -1.564687    0.585525  -2.672 0.008390 **
FranchiseKirby   -1.557615    0.576720  -2.701 0.007735 **
FranchiseMario  -1.155896    0.504552  -2.291 0.023398 *
FranchiseOther   -1.553300    0.469424  -3.309 0.001180 **
FranchisePokemon -0.505697    0.529343  -0.955 0.340990
FranchiseSmash Bros 0.159694    1.020360   0.157 0.875849
FranchiseWii Exclusive -0.883444    0.537243  -1.644 0.102244
FranchiseYoshi   -1.939991    0.735734  -2.637 0.009275 **
FranchiseZelda   -1.284296    0.562172  -2.285 0.023782 *
PlatformDS      -0.020135    0.199098  -0.101 0.919586
PlatformWii     -0.104609    0.176923  -0.591 0.555253
PlatformWiiU    -0.305989    0.155679  -1.966 0.051253 .
---
```

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
Residual standard error: 0.7394 on 146 degrees of freedom
Multiple R-squared:  0.3719,    Adjusted R-squared:  0.273
F-statistic: 3.759 on 23 and 146 DF,  p-value: 4.917e-07
```

From the Japanese simple models and their p-values, it is observed that genres of platform, racing, role-playing, and simulation are statistically significant. Interestingly, fighting did not reach significance. The only significant platform is the WiiU. All franchises were significant except for Pokemon and Smash Bros.

From the multiple regression model, platform, racing, and role-playing are statistically significant in genres. All franchises are significant except Pokemon, Smash Bros., and Wii exclusives, and no platforms are significant.

Global:

```
summary(lm(Global_Sales ~ Genre, nintendoRecent.df))
summary(lm(Global_Sales ~ Platform, nintendoRecent.df))
summary(lm(Global_Sales ~ Franchise, nintendoRecent.df))
nintendoRecent.lm = lm(Global_Sales ~ Genre + Franchise + Platform, data = nintendoRecent.df)
summary(nintendoRecent.lm)
```

```
Call:
lm(formula = Global_Sales ~ Genre, data = nintendoRecent.df)

Residuals:
    Min       1Q   Median       3Q      Max
-4.0233 -1.0397 -0.5938  0.3053 12.9069

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)    0.9347    0.4005   2.334 0.020856 *
GenreAdventure -0.4667    1.1745  -0.397 0.691608
GenreFighting   3.2886    1.4806   2.221 0.027760 *
GenreMisc       0.1149    0.6087   0.189 0.850507
GenrePlatform   2.3205    0.6713   3.457 0.000702 ***
GenrePuzzle    -0.1593    0.8453  -0.188 0.850774
GenreRacing     8.6503    1.7911   4.830 3.21e-06 ***
GenreRole-Playing 1.4783    0.5742   2.575 0.010953 *
GenreShooter    1.0186    1.4806   0.688 0.492473
GenreSimulation  1.6653    0.9152   1.820 0.070722 .
GenreSports    -0.2403    0.9152  -0.263 0.793236
GenreStrategy  -0.5222    1.2977  -0.402 0.687919
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.469 on 158 degrees of freedom
Multiple R-squared:  0.2292,    Adjusted R-squared:  0.1756
F-statistic: 4.272 on 11 and 158 DF,  p-value: 1.45e-05
```

```
Call:
lm(formula = Global_Sales ~ Franchise, data = nintendoRecent.df)

Residuals:
    Min       1Q   Median       3Q      Max
-4.5917 -0.8165 -0.5165  0.3085 10.6383

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)    4.1633    1.4079   2.957 0.00358 **
FranchiseDonkey Kong -2.0400    1.7244  -1.183 0.23854
FranchiseKirby   -2.9148    1.6828  -1.732 0.08518 .
FranchiseMario  -0.9425    1.4900  -0.633 0.52792
FranchiseOther   -3.2968    1.4295  -2.306 0.02238 *
FranchisePokemon  0.5183    1.5741   0.329 0.74237
FranchiseSmash Bros  2.0717    2.2261   0.931 0.35346
FranchiseWii Exclusive -2.6400    1.6257  -1.624 0.10637
FranchiseYoshi   -2.8383    2.2261  -1.275 0.20415
FranchiseZelda   -1.6619    1.6828  -0.988 0.32485
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.439 on 160 degrees of freedom
Multiple R-squared:  0.2385,    Adjusted R-squared:  0.1956
F-statistic: 5.567 on 9 and 160 DF,  p-value: 1.136e-06
```

```
Call:
lm(formula = Global_Sales ~ Platform, data = nintendoRecent.df)

Residuals:
    Min       1Q   Median       3Q      Max
-1.9000 -1.4479 -0.9964  0.0872 13.4000

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)    1.90793    0.30221   6.313 2.4e-09 ***
PlatformDS      0.01207    0.68249   0.018 0.986
PlatformWii    -0.18257    0.59900  -0.305 0.761
PlatformWiiU   -0.46043    0.52779  -0.872 0.384
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 2.737 on 166 degrees of freedom
Multiple R-squared:  0.004983,    Adjusted R-squared:  -0.013
F-statistic: 0.2771 on 3 and 166 DF,  p-value: 0.8419
```

```
Call:
lm(formula = Global_Sales ~ Genre + Franchise + Platform, data = nintendoRecent.df)
```

```
Residuals:
    Min       1Q   Median       3Q      Max
-4.0472 -1.0803 -0.2008  0.5009 10.2228

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)    3.34088    1.47223   2.269 0.024718 *
GenreAdventure  0.04394    1.09687   0.040 0.968101
GenreFighting  -0.16387    2.24655  -0.073 0.941952
GenreMisc      -0.13637    0.60997  -0.224 0.823400
GenrePlatform  2.88917    0.74764   3.864 0.000167 ***
GenrePuzzle    0.24368    0.83346   0.292 0.770417
GenreRacing     7.84825    1.70342   4.607 8.82e-06 ***
GenreRole-Playing 1.12913    0.56742   1.990 0.048467 *
GenreShooter    1.94890    1.34596   1.448 0.149772
GenreSimulation 1.57145    0.91709   1.714 0.088741 .
GenreSports    -0.92577    0.93758  -0.987 0.325080
GenreStrategy   0.17639    1.18589   0.149 0.881965
FranchiseDonkey Kong -2.64972    1.74324  -1.520 0.130674
FranchiseKirby  -4.18787    1.71703  -2.439 0.015927 *
FranchiseMario  -1.33455    1.50216  -0.888 0.375776
FranchiseOther  -2.97701    1.39758  -2.130 0.034839 *
FranchisePokemon  0.59905    1.57597   0.380 0.704414
FranchiseSmash Bros  3.32757    3.03784   1.095 0.275157
FranchiseWii Exclusive -1.41722    1.59949  -0.886 0.377050
FranchiseYoshi  -4.63547    2.19045  -2.116 0.036023 *
FranchiseZelda  -0.72036    1.67371  -0.430 0.667541
PlatformDS      0.02813    0.59276   0.047 0.962220
PlatformWii     0.24470    0.52674   0.465 0.642939
PlatformWiiU    -0.53916    0.46349  -1.163 0.246626
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.202 on 146 degrees of freedom
Multiple R-squared:  0.4336,    Adjusted R-squared:  0.3444
F-statistic:  4.86 on 23 and 146 DF,  p-value: 1.213e-09
```

From the global simple models and their p-values, it is observed that genres of fighting, platforming, racing, and role-playing are statistically significant. Platform has no significance, but the “other” franchise does.

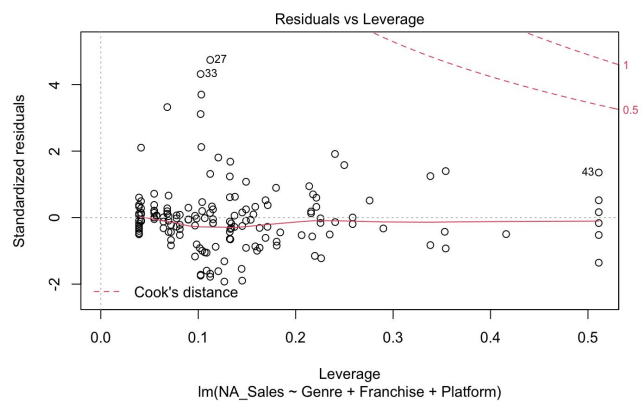
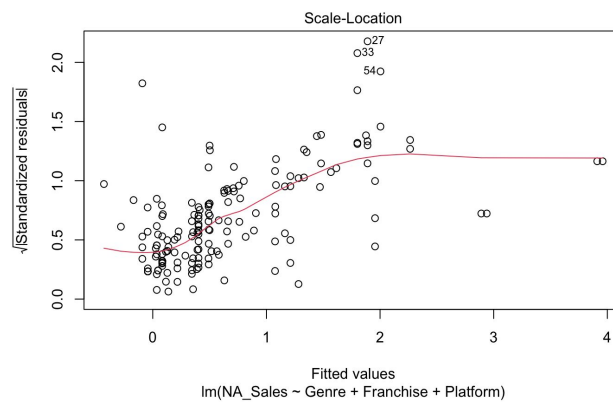
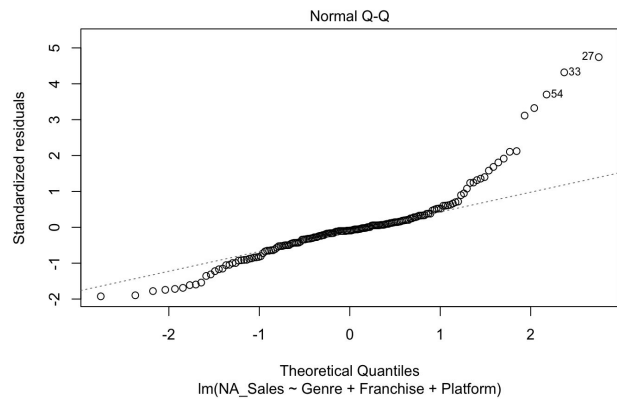
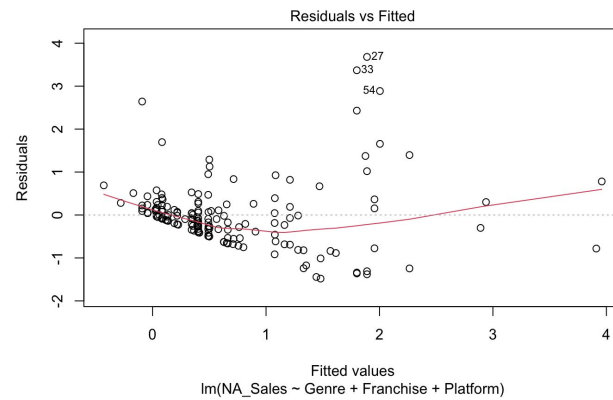
From the multiple regression model, platform, racing, and role-playing are statistically significant. Franchises Kirby, “other,” and Wii exclusives are also significant, and no platforms are significant.

### Evaluating the Models:

To evaluate the fit of the multiple regressions model for each region, diagnostic plots were made.

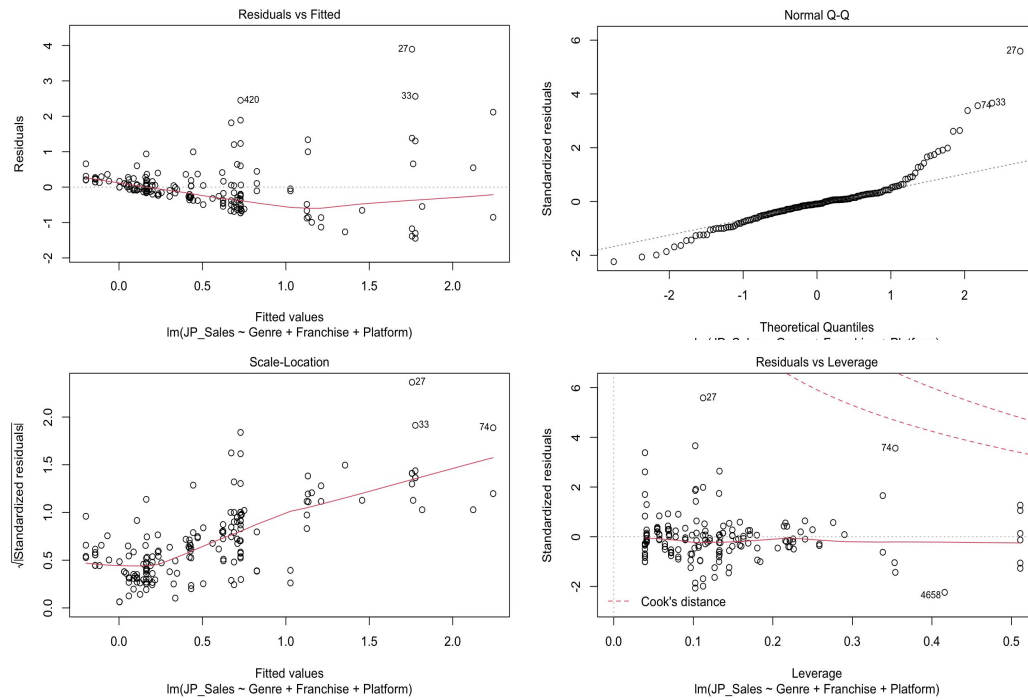
#### North America:

```
par(mfrow = c(2,2))
plot(nintendoRecentNA.lm)
```

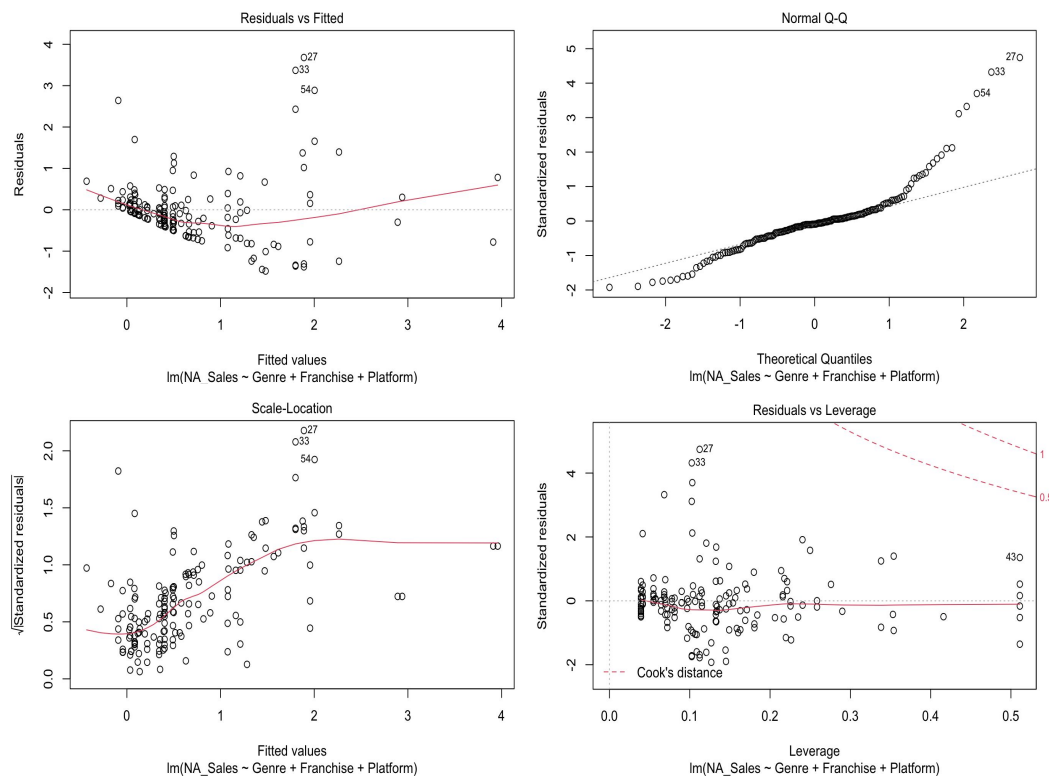


Japan:

```
par(mcfrow = c(2,2))
plot(nintendoRecentJP.lm)
```

Global:

```
par(mcfrow = c(2,2))
plot(nintendoRecent.lm)
```



According to these diagnostic plots, all three regression models are not of the highest fit. However, the underlying cause is known and therefore, can be taken into account when further accessing predictions.

## Predicting Sales from Models

To predict what combination of attributes an ideal Nintendo game during the 2010s would be, multiple data frames containing different combinations of the attributes were fed into the predictions. The chosen attributes were based on the highest and interesting spread of sales from each region as seen in the box plots in the exploratory analysis.

### North America:

Genre: racing, fighting, platform, shooter, simulation, role-playing

Platform: WiiU, Wii, DS, 3 DS

Franchise: Smash Bros., Zelda, Mario, Donkey Kong, Pokemon

```
NAplatform = c(rep(as.factor(c("WiiU")), 30), rep(as.factor(c("Wii")), 30),
rep(as.factor(c("DS")), 30), rep(as.factor(c("3DS")), 30))

NAgenre = as.factor(rep(c(rep("Racing", 5), rep("Fighting", 5), rep("Platform", 5),
rep("Shooter", 5), rep("Simulation", 5), rep("Role-Playing", 5)), 4))

NAfranchise = as.factor(rep(c("Smash Bros", "Zelda", "Mario", "Donkey Kong", "Pokemon"), 24))

newNAadf = data.frame(Platform = NAplatform, Franchise = NAfranchise, Genre = NAgenre)
newNAadf = cbind(newNAadf, NA_Sales.pred = predict(nintendoRecentNA.lm, newNAadf))
arrange(newNAadf, -NA_Sales.pred)
```

	<b>Platform</b> <fctr>	<b>Franchise</b> <fctr>	<b>Genre</b> <fctr>	<b>NA_Sales.pred</b> <dbl>
31	Wii	Smash Bros	Racing	6.4253661
61	DS	Smash Bros	Racing	6.2544402
91	3DS	Smash Bros	Racing	6.1648738
1	WiiU	Smash Bros	Racing	6.1167368
35	Wii	Pokemon	Racing	4.9439588
65	DS	Pokemon	Racing	4.7730329
95	3DS	Pokemon	Racing	4.6834665
32	Wii	Zelda	Racing	4.6737834
5	WiiU	Pokemon	Racing	4.6353295
62	DS	Zelda	Racing	4.5028575

Based on this model, the ideal Nintendo game during the 2010s for North America is...

Wii racing game in Smash Bros. franchise.

### Japan:

Genre: racing, fighting, simulation, platform, role-playing

Platform: DS, 3 DS, Wii

Franchise: Smash Bros, Animal Crossing, Mario, Kirby, Pokemon

```
JPplatform = c(rep(as.factor(c("DS")), 25), rep(as.factor(c("3DS")), 25),
rep(as.factor(c("Wii")), 25))

JPgenre = as.factor(rep(c(rep("Racing", 5), rep("Fighting", 5), rep("Simulation", 5),
rep("Platform", 5), rep("Role-Playing", 5)), 3))

JPfranchise = as.factor(rep(c("Smash Bros", "Animal Crossing", "Mario", "Kirby", "Pokemon"),
15))

newJPdf = data.frame(Platform = JPplatform, Franchise = JPfranchise, Genre = JPgenre)
newJPdf = cbind(newJPdf, JP_Sales.pred = predict(nintendoRecentJP.lm, newJPdf))
arrange(newJPdf, -JP_Sales.pred)
```

	<b>Platform</b> <fctr>	<b>Franchise</b> <fctr>	<b>Genre</b> <fctr>	<b>JP_Sales.pred</b> <dbl>
26	3DS	Smash Bros	Racing	3.43858442
1	DS	Smash Bros	Racing	3.41844966
51	Wii	Smash Bros	Racing	3.33397501
27	3DS	Animal Crossing	Racing	3.27889065
2	DS	Animal Crossing	Racing	3.25875589
52	Wii	Animal Crossing	Racing	3.17428123
30	3DS	Pokemon	Racing	2.77319364
5	DS	Pokemon	Racing	2.75305889
55	Wii	Pokemon	Racing	2.66858423
41	3DS	Smash Bros	Platform	2.44767862

Based on this model, the ideal Nintendo game during the 2010s for Japan is...

3 DS racing game in Smash Bros. franchise.

### Global:

Genre: racing, fighting, platform, role-playing, simulation

Platform: WiiU, 3DS, DS, Wii

Franchise: Smash Bros., Animal Crossing, Zelda, Mario, Pokemon

```
platform = c(rep(as.factor(c("WiiU")), 25), rep(as.factor(c("3DS")), 25), rep(as.factor(c("DS")),
25), rep(as.factor(c("Wii")), 25))

genre = as.factor(rep(c(rep("Racing", 5), rep("Fighting", 5), rep("Platform", 5), rep("Role-Playing",
5), rep("Simulation", 5)), 4))

franchise = as.factor(rep(c("Smash Bros", "Animal Crossing", "Zelda", "Mario", "Pokemon"), 20))

newdf = data.frame(Platform = platform, Franchise = franchise, Genre = genre)
newdf = cbind(newdf, Global_Sales.pred = predict(nintendoRecent.lm, newdf))
arrange(newdf, -Global_Sales.pred)
```

	<b>Platform</b> <fctr>	<b>Franchise</b> <fctr>	<b>Genre</b> <fctr>	<b>Global_Sales.pred</b> <dbl>
76	Wii	Smash Bros	Racing	14.761404
51	DS	Smash Bros	Racing	14.544828
26	3DS	Smash Bros	Racing	14.516703
1	WiiU	Smash Bros	Racing	13.977545
80	Wii	Pokemon	Racing	12.032882
55	DS	Pokemon	Racing	11.816306
30	3DS	Pokemon	Racing	11.788180
77	Wii	Animal Crossing	Racing	11.433835
5	WiiU	Pokemon	Racing	11.249023
52	DS	Animal Crossing	Racing	11.217259

Based on this model, the ideal Nintendo game during the 2010s for the globe is...

Wii racing game in Smash Bros. franchise.

## Results

Across all three regions (North America, Japan, Global), racing games in the Smash Bros. franchise appear to contribute to the highest sales in their respective regions. Platform had a minimal impact on sales as one can infer that games are typically only released on the latest console. Differences in gaming taste between North America and Japan are small, but Japan appears to purchase games in the Animal Crossing franchise more than North America.

However, the models used to determine these results are flawed due to uneven cells and collinearity. Nonetheless, the results are still supported by the box plots of each input variable as seen in the exploratory analysis section.

## Future Analysis

The original question in mind for this analysis asked what the ideal Nintendo game would be. However, with the outdated data set given, the question was restricted to ask what the ideal Nintendo game would be in the 2010s. Therefore, I would like to continue to answer this question with more accuracy given a fully updated data set that includes games released for the Nintendo Switch.

Additionally, I believe more aspects of a game would accurately reflect what factors are more attractive to sales as genre and franchise can only convey so much. If columns included whether a game has multiplayer, solo, or both, whether it is linear gameplay or is open world, and what its intended age range is, then I believe that I could gather more useful results.

## Conclusion

This analysis found that across all three regions (North America, Japan, Global), racing games in the Smash Bros. franchise seem to contribute to the highest sales while the platform had little impact. However, the models used to find these results did not have as sufficient of a fit as one would hope due to uneven cell sizes and collinearity.

Although the results of this analysis are not as strong as initially hoped, the success of recent games released by Nintendo offer alternative support. According to Nintendo's corporate [website](#), the top selling Nintendo game as of September 2022 is Mario Kart 8 Deluxe with 48.41 million copies sold. The second most selling game is Animal Crossing: New Horizons with 40.17 million copies sold. Finally, the third most selling game is Super Smash Bros. Ultimate with 29.53 million copies sold.

Mario Kart is a racing game with a large array of Nintendo characters across franchises. This character variety is a large commonality with the Super Smash Bros. franchise and may be the factor that entices players. Couple these aspects with a multiplayer and party atmosphere, and this may be another variable that contributes to the success of Nintendo games. While Animal Crossing: New Horizons has little similarities between these two games, I predict it is the release date that drove the sales as this game was released March 20, 2020 at the early onset of the Pandemic. With isolation just beginning, a relaxing solo-player simulator game seemed to appeal the most to the public.

Upon reviewing the success of the newer games, future analysis can continue to answer the original research question of what the ideal Nintendo game would be. Additions to data could include updated games and consoles, and more attributes such as the number of players, the linearity of gameplay, and intended audience. These new factors and new games could help to reduce some of the problems this analysis faced by adding more to the sample size and reducing collinearity.