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CS36

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Final Project: Analysis of the MET Collection

Introduction

This project will analyze data from the Metropolitan Museum of Art in New York that describes their collection. This dataset contains a variable called `Is Highlight`, which according to the MET's website, is a boolean variable set to true if an artwork is "a popular and important artwork in the collection". I'm really interested to see what variables predict an artwork's status as a popular and important work and seeing what trends I can find. Here's three specific questions on that theme.

Question 1: Which variables have the highest correlation with `isHighlight`? Question 2: If I'm artist and want the MET to think my art is "a popular and important artwork in the collection" what kind of art should I make? Question 3: What time periods and cultures are the best-represented among the MET's "popular and important artwork"?

I answer these questions in much more robust detail in the Conclusions section, but here's a topline summary of my results.

1. The variables with the largest and most robust effect on whether an artwork is a highlight have to do with the medium. Specifically, Stone Sculpture and Paintings are the most successful.
2. Culture seems to have no impact on the highlight worthiness of your art most of the time, but there are some edge cases.
3. What year you created your work has no impact on whether your art is a highlight, but what year the MET bought your work is extremely robustly associated with highlight status, but at a very low level.

The code for this project is opensourced on Github. You can find it [here](#).

Data Source

The data for this project was sourced from The Metropolitan Museum of Art Open Access CSV. This is a dataset of roughly 470,000 artworks in the MET's collection, on display and in their warehouses. On the MET's repo for this data, they say that it is "generated from our internal database". The dataset was published under a Creative Commons Zero license and the MET all copyright and related rights to the dataset. The MET has an incentive to have an accurate database of all the art in their collection and to release it to the public, so the dataset is likely as high-quality as the MET's data scientists can make it. That said, I don't think the dataset is natively stored in a CSV. There were some issues parsing data, especially gender data, that seem related to an imperfect translation to CSV. Also most big datasets aren't usually stored in CSV, it's just a convenient format to store your data in if you want to opensource it.

Data Ethics

The project has several ethical concerns.

I analyzed the Culture variable in the dataset, but doing this required that I both tidy and flatten the data. As a result, nearly all sub-cultures were removed as independent classifications, and I only kept the top 20 or so largest cultures- all cultures with more than 1000 pieces in the museum. This should not be taken to be an accurate and full depiction of the cultural diversity of objects at the MET museum, since all small cultures and most subcultures were removed. For context, the dataset originally had a total of around 7300 cultures, and my analysis flattened it down to 23.

Additionally, the way I parsed the gender variable right now marks everyone who has a gender that isn't explicitly marked Female as Male. I looked through the data briefly and couldn't find any cases of this messing up but I'm sure it does sometimes. This also posits Male as implicitly the default gender which makes this method of data cleaning problematic, especially going forward, but it seems pretty accurate here because the MET's collection is so Male dominated. And of course there's lots of cases where the art was made by more than one person or by a company.

Finally, I'm concerned that this project could be taken as an attempt to figure out what kind of art is the 'most valuable'. This would both be an inaccurate reading of what I'm trying to do and of what I think about art in general. I'm trying to figure out what the MET thinks is highlight-worthy art and what, if any, variables in their publicly released dataset predict that. Art in itself does not have any value—it only exists relative to the people who made it and the people who engage with it.

Data Import, Cleaning and Tidying, and Exploration

Data Import

First things first, let's

Import our libraries-

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.4
## v forcats    1.0.0      v stringr   1.5.0
## v ggplot2    3.4.4      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.0
## v purrr      1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
library(splines)
library(modelr)
```

And load our data-

```
met_objects <- read_csv("openaccess/MetObjects.csv")
```

```
## Warning: One or more parsing issues, call `problems()` on your data frame for details,
## e.g.:
##   dat <- vroom(...)
##   problems(dat)

## Rows: 484956 Columns: 54
## -- Column specification -----
## Delimiter: ","
## chr (45): Object Number, Gallery Number, Department, Object Name, Title, Cul...
## dbl (5): Object ID, AccessionYear, Constituent ID, Object Begin Date, Objec...
## lgl (4): Is Highlight, Is Timeline Work, Is Public Domain, Metadata Date
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

Cleaning and Tidying

Variables that seem like they could possibly be useful for predicting Is Highlight

1. Department
2. AccessionYear
3. Culture
4. Artist End Date
5. Artist Gender
6. Object Date
7. Classification

I suspect that some of these might be categorical variables with a lot of unique values that we should collapse. Let's test that theory by making some data grids

```
unique_department <- data_grid(met_objects, Department)
unique_culture <- data_grid(met_objects, Culture)
unique_classification <- data_grid(met_objects, Classification)
```

Looking the sizes of the data grids using our sidebar, we're going to need to figure out how to collapse unique values in everything except Department.

Let's start with culture!

Here's some code that collapses this a bunch- down to about 3,000 unique cultures from 7,300 or so originally.

```
unique_culture[1, "Culture"] <- "Chugach, Native American"
unique_culture <- unique_culture %>%
  mutate(Culture_Group = str_extract(Culture, "^[^,;()]+")) %>%
  mutate(Culture_Group = ifelse(
    str_detect(Culture_Group, "^(possibly|probably)"),
    str_extract(Culture_Group, "(?<=(possibly|probably)\\s)\\w+"),
    Culture_Group
  )) %>%
  mutate(Culture_Group = str_to_title(Culture_Group))
```

Now let's merge this with our original dataframe so that we can filter by top cultures.

```
met_objects <- met_objects %>%
  left_join(unique_culture, "Culture") %>%
  mutate(Culture_Group = as.factor(Culture_Group)) %>%
  mutate(Culture_Group = fct_lump_min(Culture_Group, min = 1000))

met_objects %>%
  select(Culture_Group) %>%
  group_by(Culture_Group) %>%
  count() %>%
  arrange(desc(n))
```

```
## # A tibble: 24 x 2
## # Groups:   Culture_Group [24]
##   Culture_Group      n
##   <fct>          <int>
## 1 <NA>          276768
## 2 Other           42313
## 3 American       29348
## 4 French          23836
## 5 Greek           20653
## 6 Japan           16941
## 7 China           13533
## 8 Italian          10679
## 9 British          10614
```

```
## 10 German          7793
## # i 14 more rows
```

Okay now let's do classification

It seems like we could filter a bunch by just taking the first word.

```
unique_classification <- unique_classification %>%
  mutate(Classification_Group = str_extract(Classification, "^[^[-,;(|)]+") %>%
  mutate(Classification_Group = str_to_title(Classification_Group))
```

Making it a factor and taking only the top groups.

```
met_objects <- met_objects %>%
  left_join(unique_classification, "Classification") %>%
  mutate(Classification_Group = as.factor(Classification_Group)) %>%
  mutate(Classification_Group = fct_lump_min(Classification_Group, min = 1000))
```

Cleaning up object date

```
met_objects <- met_objects %>%
  mutate(
    parsed_year = `Object Date`,
    parsed_year = if_else(
      str_detect(parsed_year, "\\d{4}"),
      str_extract(parsed_year, "\\d{4}"),
      NA_character_
    ),
    parsed_year = if_else(
      is.na(parsed_year) & str_detect(`Object Date`, "\\d{3}"),
      str_extract(`Object Date`, "\\d{3}"),
      parsed_year
    ),
    parsed_year = if_else(
      is.na(parsed_year) & str_detect(`Object Date`, "\\d+\\s*(BCE|CE|BC)"),
      if_else(
        str_detect(`Object Date`, "BCE|BC"),
        as.character(-as.numeric(str_extract(`Object Date`, "\\d+"))),
        str_extract(`Object Date`, "\\d+")
      ),
      parsed_year
    ),
    parsed_year = if_else(
      is.na(parsed_year) & str_detect(`Object Date`, "\\d{1,2}(st|nd|rd|th)"),
      as.character(
        as.numeric(str_extract(`Object Date`, "\\d{1,2}")) * 100 - 50
      ),
      parsed_year
    ),
    parsed_year = as.numeric(parsed_year)
  )
```

Cleaning AccessionYear- much less work!

```
met_objects$AccessionYear <- as.double(substr(met_objects$AccessionYear, 1, 4))
```

Cleaning Gender

```
met_objects <- met_objects %>%
  mutate(parsed_gender = ifelse(is.na(`Artist Gender`), NA,
                                ifelse(str_detect(`Artist Gender`, "[Ff]emale"), "Female", "Male")))
```

Data Transformation/Visualization/Modeling

Messing with the data types to make glm happy and selecting just the variables we need

```
met_objects$Culture_Group <- as.character(met_objects$Culture_Group)
met_objects$Classification_Group <- as.character(met_objects$Classification_Group)
selected_variables <- met_objects %>%
  select(`Is Highlight`, Department, AccessionYear, Culture_Group, parsed_gender, parsed_year, Classification_Group)
  mutate(Department = ifelse(is.na(Department), 'Unknown', Department)) %>%
  mutate(parsed_gender = ifelse(is.na(parsed_gender), 'Unknown', parsed_gender)) %>%
  mutate(Culture_Group = ifelse(is.na(Culture_Group), 'Unknown', Culture_Group)) %>%
  mutate(Classification_Group = ifelse(is.na(Classification_Group), 'Unknown', Classification_Group))
```

Okay now that we've got our data cleaned it's time to answer our questions.

Let's set our seed and split up our model.

```
set.seed(123)
big_met_training <- sample_n(selected_variables, nrow(selected_variables) * 0.8)
big_met_testing <- anti_join(selected_variables, big_met_training)
```

```
## Joining with `by = join_by(`Is Highlight`, Department, AccessionYear,
## Culture_Group, parsed_gender, parsed_year, Classification_Group)`
```

Setting up our baseline model...

```
big_met_testing %>%
  mutate(baseline_prediction = FALSE) %>%
  summarize(sum(baseline_prediction == `Is Highlight`)/nrow(big_met_testing))
```

```
## # A tibble: 1 x 1
##   `sum(baseline_prediction == `Is Highlight`)/nrow(big_met_testing)`
##                                                                 <dbl>
## 1                                                                 0.970
```

Some more data type shenanigans and making the model

```
big_met_training$Culture_Group <- as.character(big_met_training$Culture_Group)
big_met_training$Classification_Group <- as.character(big_met_training$Classification_Group)
big_met_training$Department <- as.character(big_met_training$Department)
big_met_training$parsed_gender <- as.character(big_met_training$parsed_gender)

met_model <- glm(`Is Highlight` ~ Department + AccessionYear + Culture_Group +
  parsed_gender + parsed_year + Classification_Group,
  data = big_met_training, family='binomial')
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

We're getting a warning here that some probabilities are numerically zero or 1. That's to be expected—some of these variables might have a near-zero relationship with the thing we're trying to measure.

Here's a summary of our model. Looks like there's quite a few significant variables!

```
summary(met_model)
```

```
##
```

```
## Call:
## glm(formula = `Is Highlight` ~ Department + AccessionYear + Culture_Group +
##      parsed_gender + parsed_year + Classification_Group, family = "binomial",
##      data = big_met_training)
##
## Coefficients:
##
##              Estimate Std. Error
## (Intercept)      -2.035e+01  1.620e+00
## DepartmentArms and Armor      -1.184e+00  2.776e-01
## DepartmentArts of Africa, Oceania, and the Americas -1.248e+00  2.287e-01
## DepartmentAsian Art          -5.796e-01  2.015e-01
## DepartmentCostume Institute    1.459e+01  3.001e+03
## DepartmentDrawings and Prints -2.427e+00  2.908e-01
## DepartmentEgyptian Art        1.545e+01  3.001e+03
## DepartmentEuropean Paintings   2.021e-02  2.536e-01
## DepartmentEuropean Sculpture and Decorative Arts -2.206e+00  2.370e-01
## DepartmentGreek and Roman Art  -8.664e-01  4.800e-01
## DepartmentIslamic Art         -5.814e-01  2.508e-01
## DepartmentMedieval Art        -9.152e-01  2.439e-01
## DepartmentModern and Contemporary Art -9.765e-01  2.410e-01
## DepartmentMusical Instruments   3.033e-01  3.897e-01
## DepartmentPhotographs         -1.071e+00  3.634e-01
## DepartmentRobert Lehman Collection  4.299e-01  2.265e-01
## DepartmentThe American Wing    1.691e+01  3.001e+03
## DepartmentThe Cloisters        -5.423e-01  2.510e-01
## DepartmentThe Libraries        4.067e+01  5.348e+03
## AccessionYear                8.596e-03  7.937e-04
## Culture_GroupAmerican Or European -1.288e+00  1.009e+00
## Culture_GroupBritish           9.387e-02  2.049e-01
## Culture_GroupChina            -1.674e+00  2.960e-01
## Culture_GroupChinese          -1.078e+00  4.546e-01
## Culture_GroupCoptic           -1.501e+01  4.826e+02
## Culture_GroupCypriot          -1.206e+00  6.739e-01
## Culture_GroupDutch            7.188e-01  5.177e-01
## Culture_GroupEtruscan         1.468e-01  6.094e-01
## Culture_GroupEuropean        -1.462e+01  3.390e+02
## Culture_GroupFrench           5.191e-01  1.451e-01
## Culture_GroupGerman           1.003e+00  1.958e-01
## Culture_GroupGreek            9.417e-01  3.733e-01
## Culture_GroupIndian           1.322e-01  4.742e-01
## Culture_GroupIndonesia        -4.219e-01  4.010e-01
## Culture_GroupIran             -6.082e-01  4.539e-01
## Culture_GroupItalian          3.399e-01  1.985e-01
## Culture_GroupJapan            -2.413e+00  3.940e-01
## Culture_GroupJapanese         1.064e+00  2.508e-01
## Culture_GroupOther            3.784e-01  1.347e-01
## Culture_GroupRoman            4.517e-01  3.766e-01
## Culture_GroupRussian          4.951e-01  5.959e-01
## Culture_GroupSpanish          5.709e-01  3.317e-01
## Culture_GroupUnknown          6.538e-02  1.489e-01
## parsed_genderMale            -6.388e-01  1.236e-01
## parsed_genderUnknown         -9.471e-01  9.710e-02
## parsed_year                 -1.401e-04  7.239e-05
## Classification_GroupAlbums    -1.416e+01  5.142e+02
```

## Classification_GroupArmor Parts	-1.398e+00	1.094e+00
## Classification_GroupBooks	-1.383e+01	1.370e+02
## Classification_GroupBronzes	9.197e-01	6.175e-01
## Classification_GroupCeramics	8.819e-02	4.392e-01
## Classification_GroupChordophone	1.073e+00	2.651e-01
## Classification_GroupCodices	1.537e+00	4.817e-01
## Classification_GroupCoins	-1.526e+01	6.073e+02
## Classification_GroupCut Paper	3.743e-01	1.102e+00
## Classification_GroupDrawings	5.883e-01	4.439e-01
## Classification_GroupEnamels	1.236e+00	5.391e-01
## Classification_GroupEphemera	-1.406e+01	4.418e+02
## Classification_GroupGems	-4.527e-01	7.954e-01
## Classification_GroupGlass	-2.445e-01	4.873e-01
## Classification_GroupGold And Silver	4.989e-01	6.572e-01
## Classification_GroupIdiophone	-7.428e-01	4.643e-01
## Classification_GroupJade	-1.369e+01	6.474e+02
## Classification_GroupJewelry	-6.859e-01	5.958e-01
## Classification_GroupMedals And Plaquettes	-1.414e+01	4.369e+02
## Classification_GroupMetal	1.259e-01	5.114e-01
## Classification_GroupMetalwork	3.178e-01	4.422e-01
## Classification_GroupMiscellaneous	3.508e-01	5.494e-01
## Classification_GroupNegatives	-1.513e+01	2.505e+02
## Classification_GroupOther	1.393e-01	4.026e-01
## Classification_GroupPaintings	1.954e+00	4.315e-01
## Classification_GroupPhotographs	-3.186e-01	5.051e-01
## Classification_GroupPrints	-8.771e-01	4.806e-01
## Classification_GroupSculpture	1.899e+00	4.338e-01
## Classification_GroupStone	1.118e-01	4.736e-01
## Classification_GroupStone Sculpture	2.258e+00	5.772e-01
## Classification_GroupStucco	-1.542e+01	4.652e+02
## Classification_GroupSword Furniture	-2.824e+00	1.106e+00
## Classification_GroupSwords	8.169e-01	5.556e-01
## Classification_GroupTerracottas	-7.393e-01	8.017e-01
## Classification_GroupTextiles	-2.210e-01	4.472e-01
## Classification_GroupUnknown	-1.603e+01	3.001e+03
## Classification_GroupVases	-2.702e+00	6.265e-01
## Classification_GroupWood	1.304e+00	4.708e-01
## Classification_GroupWoodwork	1.385e+00	4.800e-01
##	z value Pr(> z)	
## (Intercept)	-12.561	< 2e-16 ***
## DepartmentArms and Armor	-4.264	2.01e-05 ***
## DepartmentArts of Africa, Oceania, and the Americas	-5.457	4.83e-08 ***
## DepartmentAsian Art	-2.876	0.004021 **
## DepartmentCostume Institute	0.005	0.996121
## DepartmentDrawings and Prints	-8.348	< 2e-16 ***
## DepartmentEgyptian Art	0.005	0.995894
## DepartmentEuropean Paintings	0.080	0.936469
## DepartmentEuropean Sculpture and Decorative Arts	-9.308	< 2e-16 ***
## DepartmentGreek and Roman Art	-1.805	0.071110 .
## DepartmentIslamic Art	-2.318	0.020423 *
## DepartmentMedieval Art	-3.752	0.000175 ***
## DepartmentModern and Contemporary Art	-4.052	5.08e-05 ***
## DepartmentMusical Instruments	0.778	0.436327
## DepartmentPhotographs	-2.946	0.003221 **

## DepartmentRobert Lehman Collection	1.898	0.057721	.
## DepartmentThe American Wing	0.006	0.995504	
## DepartmentThe Cloisters	-2.161	0.030731	*
## DepartmentThe Libraries	0.008	0.993933	
## AccessionYear	10.831	< 2e-16	***
## Culture_GroupAmerican Or European	-1.276	0.201792	
## Culture_GroupBritish	0.458	0.646884	
## Culture_GroupChina	-5.655	1.56e-08	***
## Culture_GroupChinese	-2.372	0.017706	*
## Culture_GroupCoptic	-0.031	0.975192	
## Culture_GroupCypriot	-1.790	0.073486	.
## Culture_GroupDutch	1.388	0.164987	
## Culture_GroupEtruscan	0.241	0.809656	
## Culture_GroupEuropean	-0.043	0.965607	
## Culture_GroupFrench	3.577	0.000347	***
## Culture_GroupGerman	5.125	2.97e-07	***
## Culture_GroupGreek	2.523	0.011646	*
## Culture_GroupIndian	0.279	0.780365	
## Culture_GroupIndonesia	-1.052	0.292710	
## Culture_GroupIran	-1.340	0.180265	
## Culture_GroupItalian	1.712	0.086856	.
## Culture_GroupJapan	-6.123	9.16e-10	***
## Culture_GroupJapanese	4.245	2.19e-05	***
## Culture_GroupOther	2.808	0.004979	**
## Culture_GroupRoman	1.199	0.230388	
## Culture_GroupRussian	0.831	0.406071	
## Culture_GroupSpanish	1.721	0.085206	.
## Culture_GroupUnknown	0.439	0.660661	
## parsed_genderMale	-5.169	2.35e-07	***
## parsed_genderUnknown	-9.755	< 2e-16	***
## parsed_year	-1.935	0.052934	.
## Classification_GroupAlbums	-0.028	0.978037	
## Classification_GroupArmor Parts	-1.278	0.201262	
## Classification_GroupBooks	-0.101	0.919592	
## Classification_GroupBronzes	1.489	0.136425	
## Classification_GroupCeramics	0.201	0.840861	
## Classification_GroupChordophone	4.049	5.15e-05	***
## Classification_GroupCodices	3.190	0.001421	**
## Classification_GroupCoins	-0.025	0.979956	
## Classification_GroupCut Paper	0.340	0.734171	
## Classification_GroupDrawings	1.325	0.185037	
## Classification_GroupEnamels	2.293	0.021853	*
## Classification_GroupEphemera	-0.032	0.974615	
## Classification_GroupGems	-0.569	0.569235	
## Classification_GroupGlass	-0.502	0.615782	
## Classification_GroupGold And Silver	0.759	0.447821	
## Classification_GroupIdiophone	-1.600	0.109663	
## Classification_GroupJade	-0.021	0.983131	
## Classification_GroupJewelry	-1.151	0.249687	
## Classification_GroupMedals And Plaquettes	-0.032	0.974188	
## Classification_GroupMetal	0.246	0.805485	
## Classification_GroupMetalwork	0.719	0.472330	
## Classification_GroupMiscellaneous	0.639	0.523131	
## Classification_GroupNegatives	-0.060	0.951816	


```
## Classification_GroupOther          0.346 0.729268
## Classification_GroupPaintings      4.529 5.93e-06 ***
## Classification_GroupPhotographs    -0.631 0.528166
## Classification_GroupPrints         -1.825 0.068015 .
## Classification_GroupSculpture       4.378 1.20e-05 ***
## Classification_GroupStone          0.236 0.813347
## Classification_GroupStone Sculpture 3.913 9.13e-05 ***
## Classification_GroupStucco         -0.033 0.973549
## Classification_GroupSword Furniture -2.553 0.010683 *
## Classification_GroupSwords         1.470 0.141464
## Classification_GroupTerracottas    -0.922 0.356443
## Classification_GroupTextiles       -0.494 0.621257
## Classification_GroupUnknown        -0.005 0.995739
## Classification_GroupVases          -4.313 1.61e-05 ***
## Classification_GroupWood           2.769 0.005625 **
## Classification_GroupWoodwork       2.885 0.003919 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 22858 on 372000 degrees of freedom
## Residual deviance: 18694 on 371916 degrees of freedom
## (15963 observations deleted due to missingness)
## AIC: 18864
##
## Number of Fisher Scoring iterations: 19
```

Our accuracy rate on the test data

```
big_met_testing %>%
  add_predictions(met_model, type='response') %>%
  mutate(predicted_outcome = ifelse(pred>=0.5, TRUE, FALSE)) %>%
  mutate(is_correct = `Is Highlight` == predicted_outcome) %>%
  summarize(mean(is_correct, na.rm=TRUE))

## # A tibble: 1 x 1
##   `mean(is_correct, na.rm = TRUE)`
##   <dbl>
## 1 0.973
```

So our model is slightly better than the baseline, since $0.9560013 > 0.95562$

This might seem small but over the course of our 484956 observations, the delta between those two ($0.9560013 - 0.95562$) will contribute to nearly 185 more correct predictions!

```
(0.9560013-0.95562) * 484956
```

```
## [1] 184.9137
```

Let's put our coefficients in a dataframe

```
model_summary <- summary(met_model)
coefficients_table <- model_summary$coefficients
coef_data <- as.data.frame(coefficients_table)
coef_data$Variable <- rownames(coefficients_table)
coef_data %>%
  arrange(desc(Estimate))
```

##	Estimate	Std. Error
## DepartmentThe Libraries	4.066600e+01	5.348208e+03
## DepartmentThe American Wing	1.690884e+01	3.000991e+03
## DepartmentEgyptian Art	1.544516e+01	3.000991e+03
## DepartmentCostume Institute	1.459073e+01	3.000991e+03
## Classification_GroupStone Sculpture	2.258317e+00	5.772015e-01
## Classification_GroupPaintings	1.954255e+00	4.315272e-01
## Classification_GroupSculpture	1.899399e+00	4.338098e-01
## Classification_GroupCodices	1.536913e+00	4.817439e-01
## Classification_GroupWoodwork	1.384702e+00	4.800268e-01
## Classification_GroupWood	1.303711e+00	4.708479e-01
## Classification_GroupEnamels	1.236124e+00	5.391046e-01
## Classification_GroupChordophone	1.073220e+00	2.650777e-01
## Culture_GroupJapanese	1.064403e+00	2.507510e-01
## Culture_GroupGerman	1.003270e+00	1.957511e-01
## Culture_GroupGreek	9.416556e-01	3.732747e-01
## Classification_GroupBronzes	9.196509e-01	6.175315e-01
## Classification_GroupSwords	8.169456e-01	5.556092e-01
## Culture_GroupDutch	7.187707e-01	5.176619e-01
## Classification_GroupDrawings	5.883166e-01	4.438770e-01
## Culture_GroupSpanish	5.709483e-01	3.317066e-01
## Culture_GroupFrench	5.191278e-01	1.451167e-01
## Classification_GroupGold And Silver	4.988604e-01	6.572145e-01
## Culture_GroupRussian	4.951237e-01	5.959401e-01
## Culture_GroupRoman	4.517349e-01	3.766468e-01
## DepartmentRobert Lehman Collection	4.298678e-01	2.265074e-01
## Culture_GroupOther	3.783939e-01	1.347366e-01
## Classification_GroupCut Paper	3.743428e-01	1.102361e+00
## Classification_GroupMiscellaneous	3.508008e-01	5.493909e-01
## Culture_GroupItalian	3.398936e-01	1.985104e-01
## Classification_GroupMetalwork	3.177982e-01	4.421892e-01
## DepartmentMusical Instruments	3.033293e-01	3.896776e-01
## Culture_GroupEtruscan	1.467970e-01	6.094440e-01
## Classification_GroupOther	1.393404e-01	4.026028e-01
## Culture_GroupIndian	1.322351e-01	4.742260e-01
## Classification_GroupMetal	1.259344e-01	5.114001e-01
## Classification_GroupStone	1.118311e-01	4.736401e-01
## Culture_GroupBritish	9.386546e-02	2.049037e-01
## Classification_GroupCeramics	8.819160e-02	4.392177e-01
## Culture_GroupUnknown	6.537907e-02	1.489271e-01
## DepartmentEuropean Paintings	2.021299e-02	2.535844e-01
## AccessionYear	8.595885e-03	7.936665e-04
## parsed_year	-1.401021e-04	7.238699e-05
## Classification_GroupTextiles	-2.209708e-01	4.472469e-01
## Classification_GroupGlass	-2.445488e-01	4.873068e-01
## Classification_GroupPhotographs	-3.186147e-01	5.050900e-01
## Culture_GroupIndonesia	-4.219066e-01	4.009772e-01
## Classification_GroupGems	-4.527326e-01	7.954132e-01
## DepartmentThe Cloisters	-5.423416e-01	2.510211e-01
## DepartmentAsian Art	-5.795502e-01	2.014789e-01
## DepartmentIslamic Art	-5.814162e-01	2.507743e-01
## Culture_GroupIran	-6.081938e-01	4.538961e-01
## parsed_genderMale	-6.388359e-01	1.235890e-01
## Classification_GroupJewelry	-6.858738e-01	5.958366e-01

## Classification_GroupTerracottas	-7.393051e-01	8.017073e-01
## Classification_GroupIdiophone	-7.427585e-01	4.643088e-01
## DepartmentGreek and Roman Art	-8.663718e-01	4.800441e-01
## Classification_GroupPrints	-8.770601e-01	4.806047e-01
## DepartmentMedieval Art	-9.151804e-01	2.439230e-01
## parsed_genderUnknown	-9.471442e-01	9.709542e-02
## DepartmentModern and Contemporary Art	-9.764528e-01	2.409822e-01
## DepartmentPhotographs	-1.070541e+00	3.634131e-01
## Culture_GroupChinese	-1.078187e+00	4.546018e-01
## DepartmentArms and Armor	-1.183716e+00	2.776142e-01
## Culture_GroupCypriot	-1.206081e+00	6.738625e-01
## DepartmentArts of Africa, Oceania, and the Americas	-1.248127e+00	2.287026e-01
## Culture_GroupAmerican Or European	-1.287765e+00	1.008854e+00
## Classification_GroupArmor Parts	-1.397916e+00	1.093861e+00
## Culture_GroupChina	-1.673788e+00	2.959985e-01
## DepartmentEuropean Sculpture and Decorative Arts	-2.206081e+00	2.369996e-01
## Culture_GroupJapan	-2.412727e+00	3.940162e-01
## DepartmentDrawings and Prints	-2.427079e+00	2.907509e-01
## Classification_GroupVases	-2.702013e+00	6.264619e-01
## Classification_GroupSword Furniture	-2.823982e+00	1.106191e+00
## Classification_GroupJade	-1.368870e+01	6.474290e+02
## Classification_GroupBooks	-1.383347e+01	1.370359e+02
## Classification_GroupEphemera	-1.405919e+01	4.418168e+02
## Classification_GroupMedals And Plaquettes	-1.413798e+01	4.369465e+02
## Classification_GroupAlbums	-1.415507e+01	5.141670e+02
## Culture_GroupEuropean	-1.461904e+01	3.390450e+02
## Culture_GroupCoptic	-1.500754e+01	4.825909e+02
## Classification_GroupNegatives	-1.513445e+01	2.504596e+02
## Classification_GroupCoins	-1.525889e+01	6.073385e+02
## Classification_GroupStucco	-1.542411e+01	4.651727e+02
## Classification_GroupUnknown	-1.602714e+01	3.000991e+03
## (Intercept)	-2.035361e+01	1.620386e+00
##	z value	Pr(> z)
## DepartmentThe Libraries	0.007603669	9.939332e-01
## DepartmentThe American Wing	0.005634420	9.955044e-01
## DepartmentEgyptian Art	0.005146687	9.958936e-01
## DepartmentCostume Institute	0.004861971	9.961207e-01
## Classification_GroupStone Sculpture	3.912527226	9.133524e-05
## Classification_GroupPaintings	4.528696048	5.934880e-06
## Classification_GroupSculpture	4.378415460	1.195453e-05
## Classification_GroupCodices	3.190310677	1.421199e-03
## Classification_GroupWoodwork	2.884633898	3.918691e-03
## Classification_GroupWood	2.768857293	5.625327e-03
## Classification_GroupEnamels	2.292921252	2.185254e-02
## Classification_GroupChordophone	4.048699435	5.150304e-05
## Culture_GroupJapanese	4.244861395	2.187286e-05
## Culture_GroupGerman	5.125231076	2.971731e-07
## Culture_GroupGreek	2.522688288	1.164616e-02
## Classification_GroupBronzes	1.489237070	1.364250e-01
## Classification_GroupSwords	1.470360200	1.414642e-01
## Culture_GroupDutch	1.388494531	1.649865e-01
## Classification_GroupDrawings	1.325404713	1.850370e-01
## Culture_GroupSpanish	1.721245154	8.520635e-02
## Culture_GroupFrench	3.577313777	3.471434e-04

## Classification_GroupGold And Silver	0.759052669	4.478211e-01
## Culture_GroupRussian	0.830827905	4.060709e-01
## Culture_GroupRoman	1.199359231	2.303883e-01
## DepartmentRobert Lehman Collection	1.897808829	5.772127e-02
## Culture_GroupOther	2.808397459	4.978874e-03
## Classification_GroupCut Paper	0.339582923	7.341706e-01
## Classification_GroupMiscellaneous	0.638526678	5.231309e-01
## Culture_GroupItalian	1.712220939	8.685596e-02
## Classification_GroupMetalwork	0.718692811	4.723302e-01
## DepartmentMusical Instruments	0.778410804	4.363269e-01
## Culture_GroupEtruscan	0.240870431	8.096555e-01
## Classification_GroupOther	0.346099023	7.292683e-01
## Culture_GroupIndian	0.278844055	7.803645e-01
## Classification_GroupMetal	0.246254211	8.054855e-01
## Classification_GroupStone	0.236109921	8.133474e-01
## Culture_GroupBritish	0.458095476	6.468838e-01
## Classification_GroupCeramics	0.200792471	8.408609e-01
## Culture_GroupUnknown	0.439000421	6.606612e-01
## DepartmentEuropean Paintings	0.079709135	9.364686e-01
## AccessionYear	10.830600551	2.465331e-27
## parsed_year	-1.935459880	5.293389e-02
## Classification_GroupTextiles	-0.494069007	6.212574e-01
## Classification_GroupGlass	-0.501837413	6.157819e-01
## Classification_GroupPhotographs	-0.630807756	5.281662e-01
## Culture_GroupIndonesia	-1.052195991	2.927096e-01
## Classification_GroupGems	-0.569179174	5.692346e-01
## DepartmentThe Cloisters	-2.160541556	3.073077e-02
## DepartmentAsian Art	-2.876481477	4.021358e-03
## DepartmentIslamic Art	-2.318483529	2.042306e-02
## Culture_GroupIran	-1.339940768	1.802646e-01
## parsed_genderMale	-5.169036988	2.353033e-07
## Classification_GroupJewelry	-1.151110416	2.496868e-01
## Classification_GroupTerracottas	-0.922163342	3.564434e-01
## Classification_GroupIdiophone	-1.599708017	1.096634e-01
## DepartmentGreek and Roman Art	-1.804775473	7.110982e-02
## Classification_GroupPrints	-1.824909432	6.801470e-02
## DepartmentMedieval Art	-3.751922850	1.754835e-04
## parsed_genderUnknown	-9.754777927	1.759854e-22
## DepartmentModern and Contemporary Art	-4.051971230	5.078791e-05
## DepartmentPhotographs	-2.945797826	3.221229e-03
## Culture_GroupChinese	-2.371716286	1.770568e-02
## DepartmentArms and Armor	-4.263887635	2.009005e-05
## Culture_GroupCypriot	-1.789802560	7.348566e-02
## DepartmentArts of Africa, Oceania, and the Americas	-5.457421557	4.830985e-08
## Culture_GroupAmerican Or European	-1.276463483	2.017917e-01
## Classification_GroupArmor Parts	-1.277965306	2.012617e-01
## Culture_GroupChina	-5.654719365	1.561011e-08
## DepartmentEuropean Sculpture and Decorative Arts	-9.308376126	1.298016e-20
## Culture_GroupJapan	-6.123420017	9.158784e-10
## DepartmentDrawings and Prints	-8.347622636	6.965069e-17
## Classification_GroupVases	-4.313132447	1.609576e-05
## Classification_GroupSword Furniture	-2.552887268	1.068341e-02
## Classification_GroupJade	-0.021143170	9.831314e-01
## Classification_GroupBooks	-0.100947776	9.195919e-01

## Classification_GroupEphemera	-0.031821306 9.746146e-01
## Classification_GroupMedals And Plaquettes	-0.032356309 9.741879e-01
## Classification_GroupAlbums	-0.027530096 9.780369e-01
## Culture_GroupEuropean	-0.043118283 9.656072e-01
## Culture_GroupCoptic	-0.031097840 9.751915e-01
## Classification_GroupNegatives	-0.060426704 9.518158e-01
## Classification_GroupCoins	-0.025124189 9.799559e-01
## Classification_GroupStucco	-0.033157810 9.735487e-01
## Classification_GroupUnknown	-0.005340617 9.957388e-01
## (Intercept)	-12.560963289 3.460933e-36

	Varial
## DepartmentThe Libraries	DepartmentThe Librar
## DepartmentThe American Wing	DepartmentThe American W
## DepartmentEgyptian Art	DepartmentEgyptian A
## DepartmentCostume Institute	DepartmentCostume Instit
## Classification_GroupStone Sculpture	Classification_GroupStone Sculpt
## Classification_GroupPaintings	Classification_GroupPaintin
## Classification_GroupSculpture	Classification_GroupSculpt
## Classification_GroupCodices	Classification_GroupCodi
## Classification_GroupWoodwork	Classification_GroupWoodw
## Classification_GroupWood	Classification_GroupW
## Classification_GroupEnamels	Classification_GroupEnam
## Classification_GroupChordophone	Classification_GroupChordoph
## Culture_GroupJapanese	Culture_GroupJapan
## Culture_GroupGerman	Culture_GroupGerm
## Culture_GroupGreek	Culture_GroupGre
## Classification_GroupBronzes	Classification_GroupBronz
## Classification_GroupSwords	Classification_GroupSwo
## Culture_GroupDutch	Culture_GroupDut
## Classification_GroupDrawings	Classification_GroupDrawin
## Culture_GroupSpanish	Culture_GroupSpan
## Culture_GroupFrench	Culture_GroupFren
## Classification_GroupGold And Silver	Classification_GroupGold And Sil
## Culture_GroupRussian	Culture_GroupRuss
## Culture_GroupRoman	Culture_GroupRom
## DepartmentRobert Lehman Collection	DepartmentRobert Lehman Collect
## Culture_GroupOther	Culture_GroupOtl
## Classification_GroupCut Paper	Classification_GroupCut Pap
## Classification_GroupMiscellaneous	Classification_GroupMiscellane
## Culture_GroupItalian	Culture_GroupItal
## Classification_GroupMetalwork	Classification_GroupMetalw
## DepartmentMusical Instruments	DepartmentMusical Instrumen
## Culture_GroupEtruscan	Culture_GroupEtrus
## Classification_GroupOther	Classification_GroupOtl
## Culture_GroupIndian	Culture_GroupInd
## Classification_GroupMetal	Classification_GroupMe
## Classification_GroupStone	Classification_GroupSto
## Culture_GroupBritish	Culture_GroupBrit
## Classification_GroupCeramics	Classification_GroupCeram
## Culture_GroupUnknown	Culture_GroupUnkn
## DepartmentEuropean Paintings	DepartmentEuropean Paintin
## AccessionYear	AccessionY
## parsed_year	parsed_y
## Classification_GroupTextiles	Classification_GroupTexti

```

## Classification_GroupGlass
## Classification_GroupPhotographs
## Culture_GroupIndonesia
## Classification_GroupGems
## DepartmentThe Cloisters
## DepartmentAsian Art
## DepartmentIslamic Art
## Culture_GroupIran
## parsed_genderMale
## Classification_GroupJewelry
## Classification_GroupTerracottas
## Classification_GroupIdiophone
## DepartmentGreek and Roman Art
## Classification_GroupPrints
## DepartmentMedieval Art
## parsed_genderUnknown
## DepartmentModern and Contemporary Art
## DepartmentPhotographs
## Culture_GroupChinese
## DepartmentArms and Armor
## Culture_GroupCypriot
## DepartmentArts of Africa, Oceania, and the Americas
## Culture_GroupAmerican Or European
## Classification_GroupArmor Parts
## Culture_GroupChina
## DepartmentEuropean Sculpture and Decorative Arts
## Culture_GroupJapan
## DepartmentDrawings and Prints
## Classification_GroupVases
## Classification_GroupSword Furniture
## Classification_GroupJade
## Classification_GroupBooks
## Classification_GroupEphemera
## Classification_GroupMedals And Plaquettes
## Classification_GroupAlbums
## Culture_GroupEuropean
## Culture_GroupCoptic
## Classification_GroupNegatives
## Classification_GroupCoins
## Classification_GroupStucco
## Classification_GroupUnknown
## (Intercept)

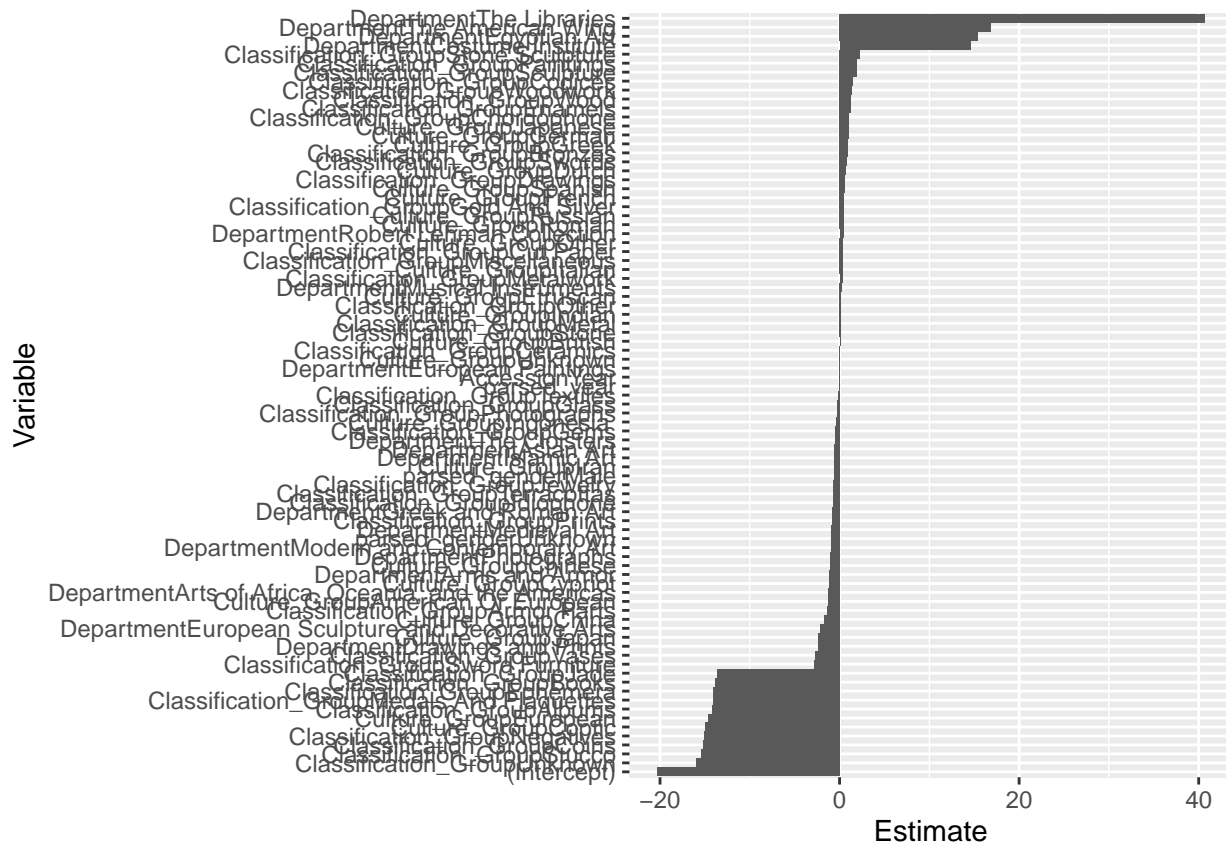
```

Now let's graph them!

```

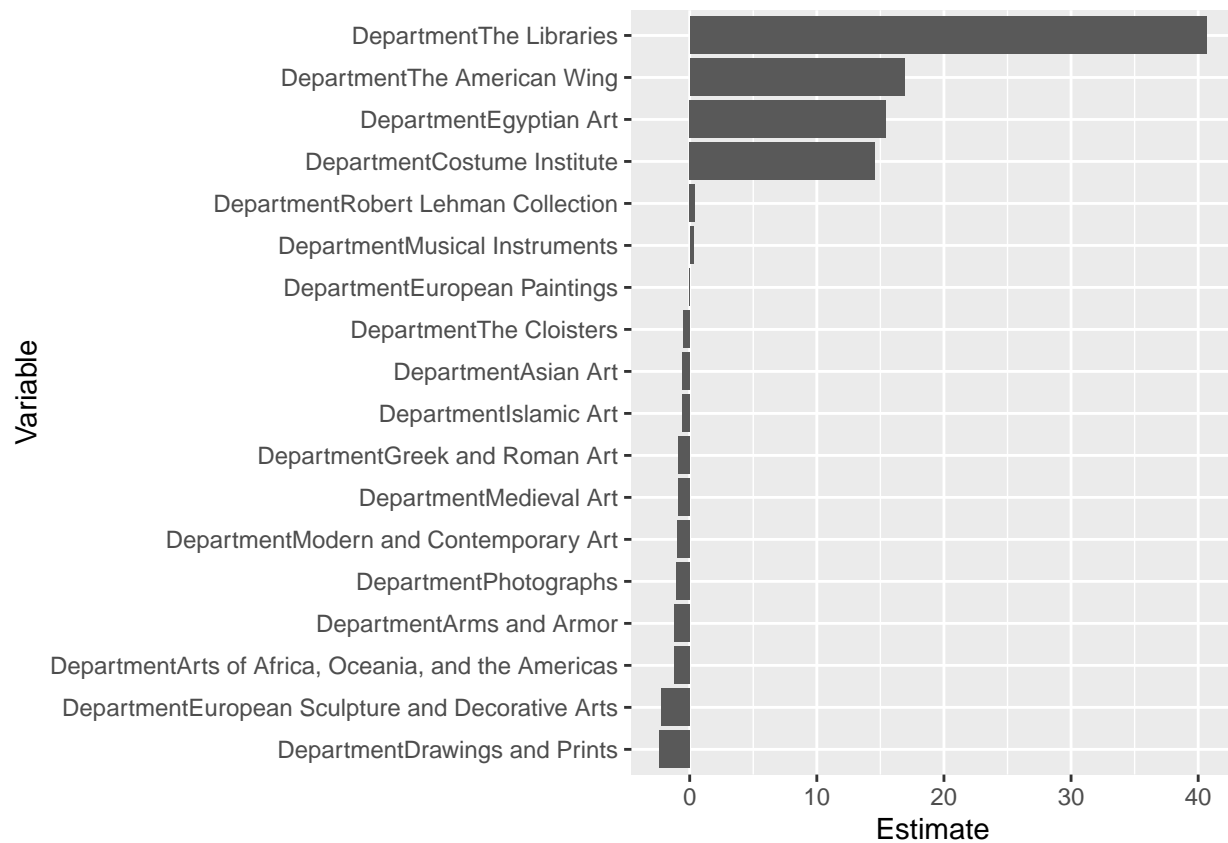
coef_data %>%
  mutate(Variable = reorder(Variable, Estimate)) %>%
  ggplot() +
  geom_bar(aes(y = Variable, x = Estimate), stat = "identity")

```

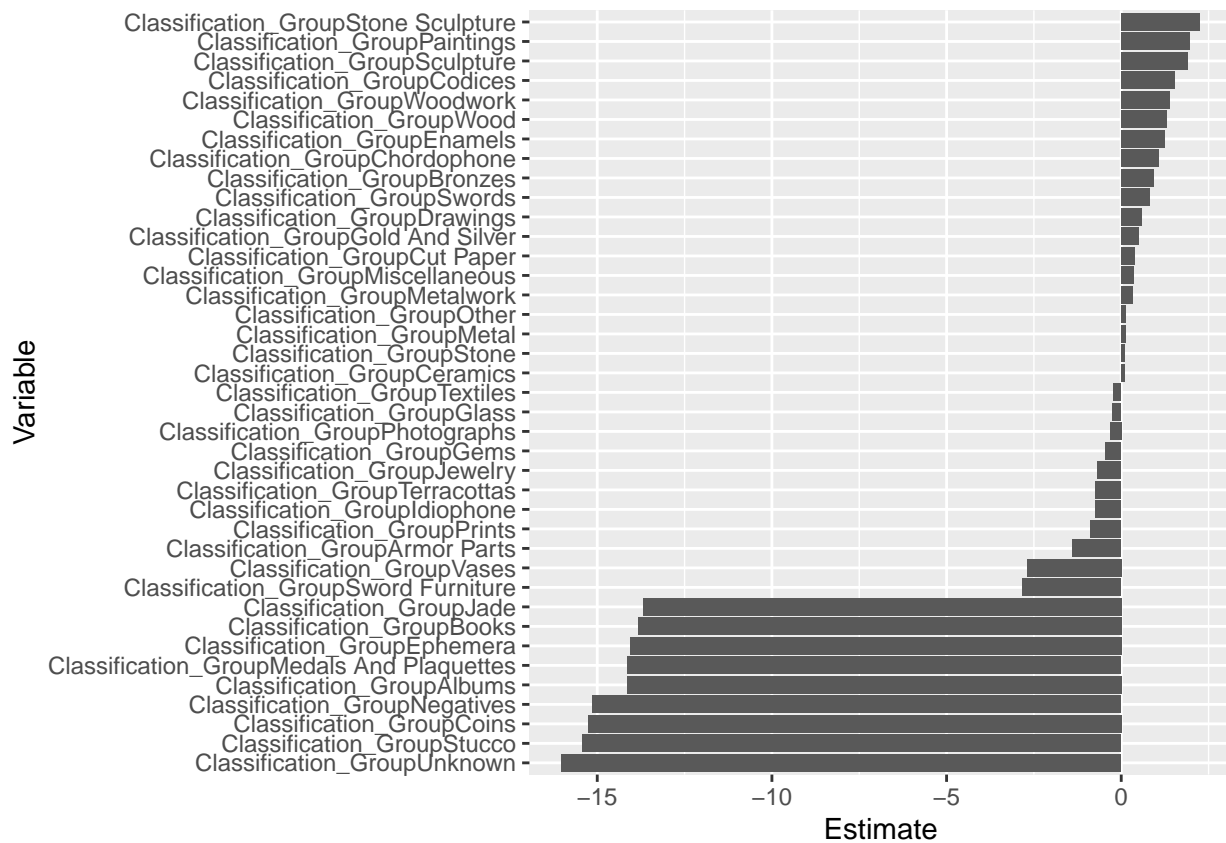


Let's make plots for each of the types of categorical variables so we can see them more clearly.

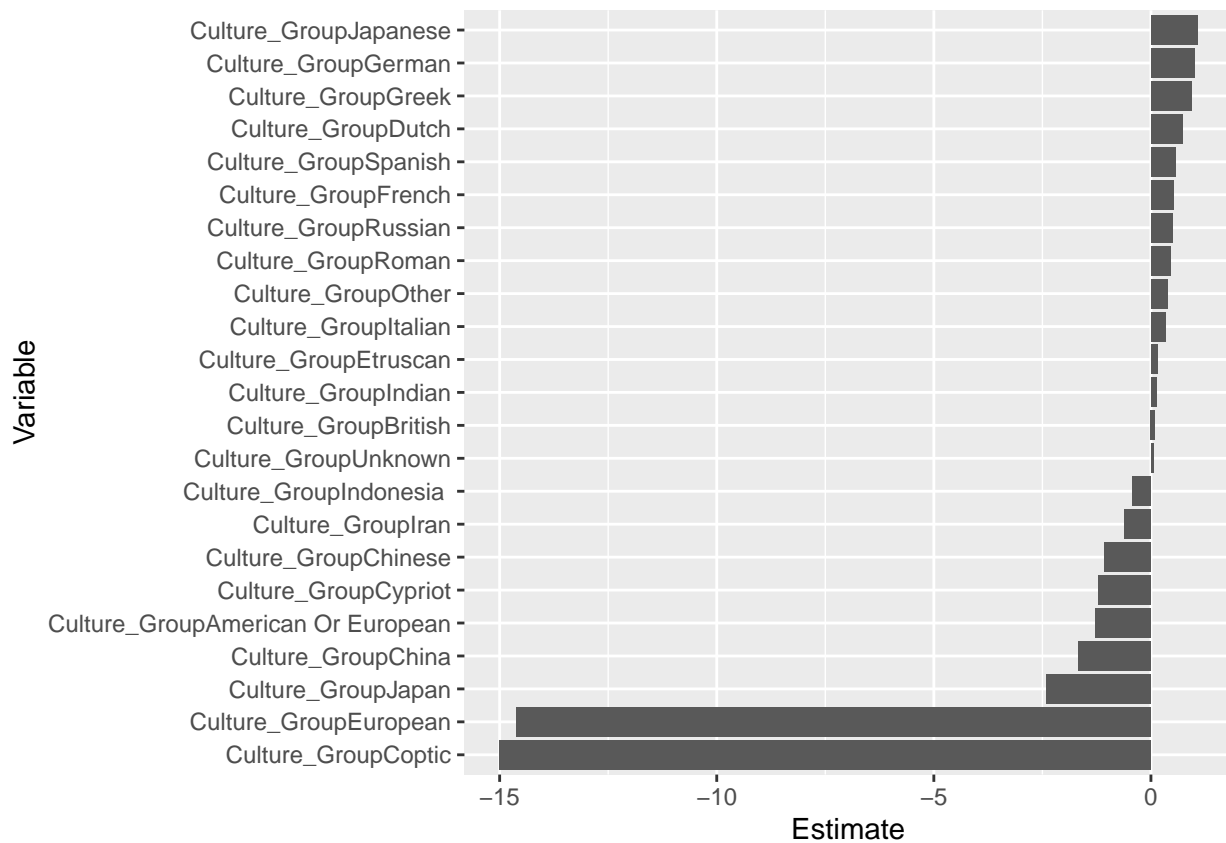
```
coef_data %>%
  filter(str_detect(Variable, "Department")) %>%
  mutate(Variable = reorder(Variable, Estimate)) %>%
  ggplot() +
  geom_bar(aes(y = Variable, x = Estimate), stat = "identity")
```



```
coef_data %>%
  filter(str_detect(Variable, "Classification")) %>%
  mutate(Variable = reorder(Variable, Estimate)) %>%
  ggplot() +
  geom_bar(aes(y = Variable, x = Estimate), stat = "identity")
```

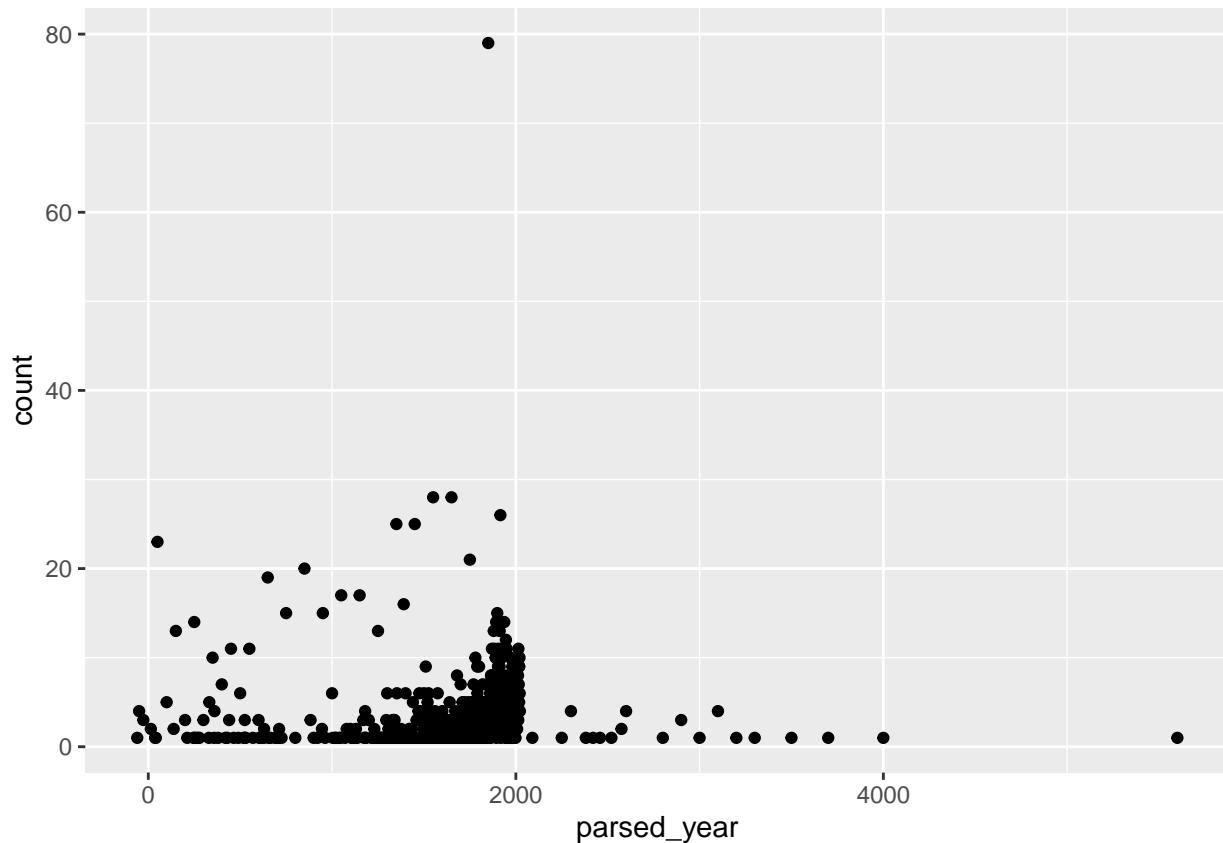
```
coef_data %>%
  filter(str_detect(Variable, "Culture")) %>%
  mutate(Variable = reorder(Variable, Estimate)) %>%
  ggplot() +
  geom_bar(aes(y = Variable, x = Estimate), stat = "identity")
```



Let's do some more analysis on time period though, because even if there's no linear relationship there could still be a relationship.

```
big_met_training %>%
  filter(`Is Highlight` == TRUE) %>%
  select(parsed_year, `Is Highlight`) %>%
  group_by(parsed_year) %>%
  summarize(count = n()) %>%
  ggplot() +
  geom_point(aes(x = parsed_year, y = count))
```

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

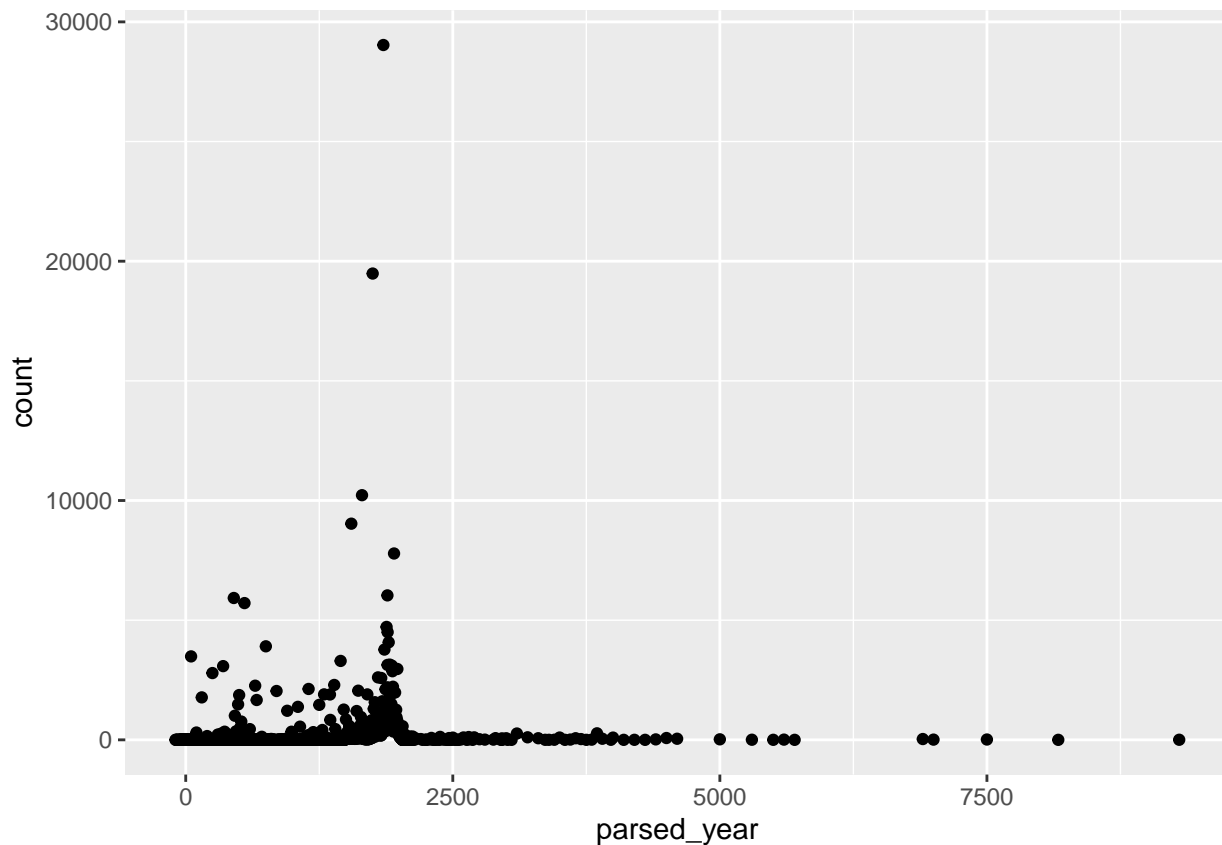


We get a warning on this and the next few graphs that one row was removed. Don't see any need to worry about that since we have hundreds of thousands of rows.

That looks encouraging, but let's make sure we're not just looking at a trend in the overall number of objects.

```
big_met_training %>%
  group_by(parsed_year) %>%
  summarize(count = n()) %>%
  ggplot() +
  geom_point(aes(x=parsed_year, y=count))
```

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



These plots look really weird but they're just artifacts of how we've parsed the year data. This is roughly how we should expect it to look—let's check out the distinct values.

```
big_met_training %>%
  select(parsed_year) %>%
  distinct() %>%
  arrange(desc(parsed_year))
```

```
## # A tibble: 1,496 x 1
##   parsed_year
##   <dbl>
## 1      9300
## 2      8168
## 3      7500
## 4      7000
## 5      6900
## 6      5700
## 7      5600
## 8      5500
## 9      5300
## 10     5000
## # i 1,486 more rows
```

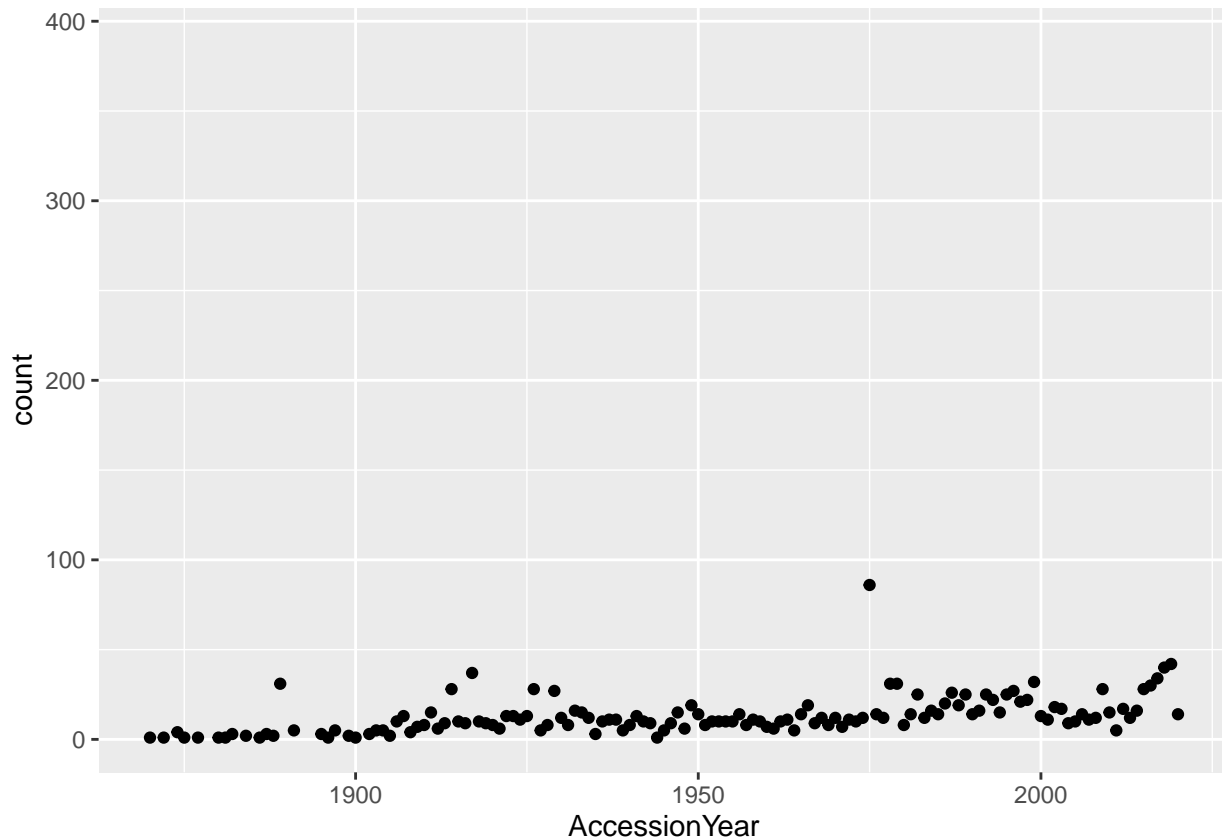
So it seems like there isn't any special relationship with time.

Let's check it out with our AccessionYear too.

```
big_met_training %>%
  filter(`Is Highlight` == TRUE) %>%
```

```
select(AccessionYear, `Is Highlight`) %>%
group_by(AccessionYear) %>%
summarize(count = n()) %>%
ggplot() +
geom_point(aes(x = AccessionYear, y = count))
```

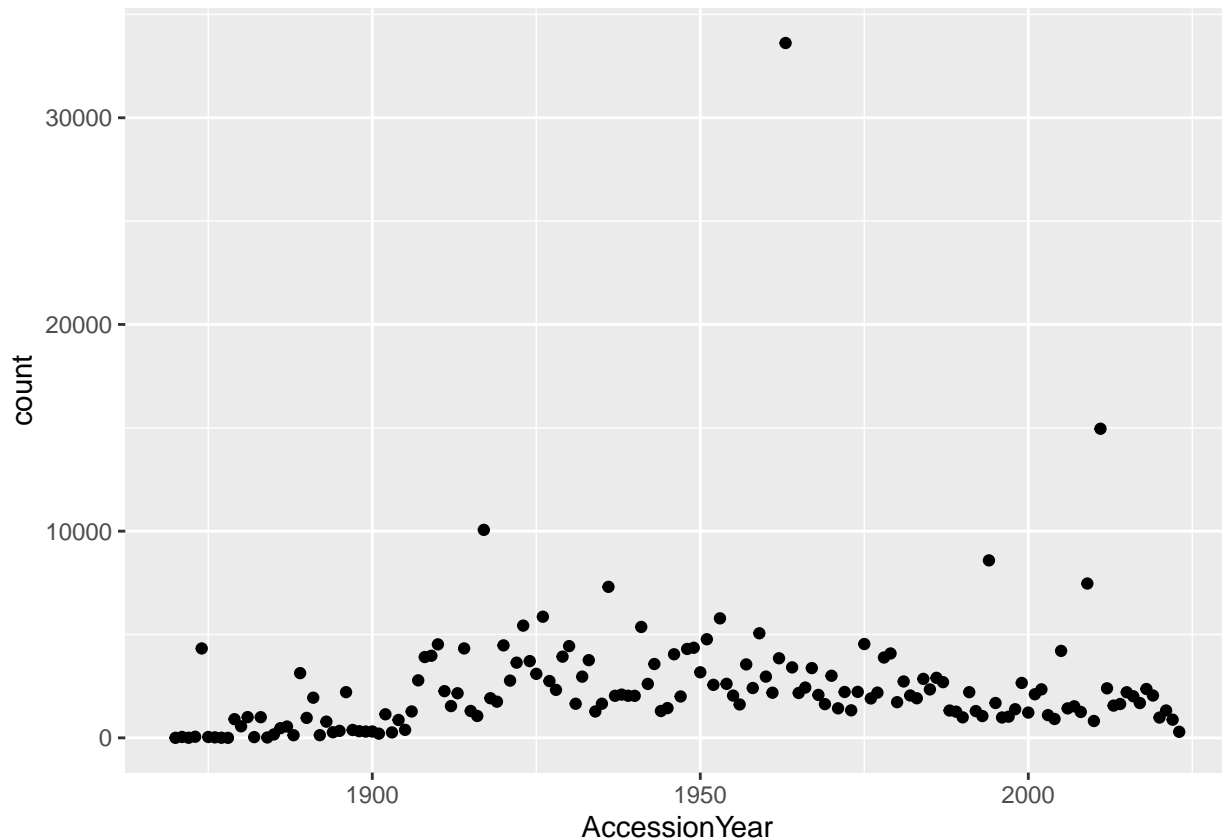
Warning: Removed 1 rows containing missing values (`geom_point()`).



This looks like a linear relationship but let's check the overall data.

```
big_met_training %>%
group_by(AccessionYear) %>%
summarize(count = n()) %>%
ggplot() +
geom_point(aes(x=AccessionYear, y=count))
```

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



No huge differences here but it's somewhat interesting that all the outliers disappeared.

Conclusions

Main Questions

Question 1: Which variables have the highest correlation with isHighlight? Let's see which of our variables are significant.

```
coef_data$Significant <- coef_data$`Pr(>|z|)` < 0.05
significant_vars <- coef_data[coef_data$Significant, ]
significant_vars %>%
  arrange(desc(Estimate))
```

##	Estimate	Std. Error
## Classification_GroupStone Sculpture	2.258316653	0.5772015178
## Classification_GroupPaintings	1.954255435	0.4315271801
## Classification_GroupSculpture	1.899399405	0.4338097704
## Classification_GroupCodices	1.536912709	0.4817439006
## Classification_GroupWoodwork	1.384701586	0.4800268023
## Classification_GroupWood	1.303710548	0.4708478663
## Classification_GroupEnamels	1.236124422	0.5391046121
## Classification_GroupChordophone	1.073219967	0.2650777080
## Culture_GroupJapanese	1.064403208	0.2507509927
## Culture_GroupGerman	1.003269862	0.1957511470
## Culture_GroupGreek	0.941655592	0.3732746515
## Culture_GroupFrench	0.519127829	0.1451166605
## Culture_GroupOther	0.378393914	0.1347365962
## AccessionYear	0.008595885	0.0007936665

```

## DepartmentThe Cloisters -0.542341575 0.2510211263
## DepartmentAsian Art -0.579550189 0.2014788531
## DepartmentIslamic Art -0.581416181 0.2507743419
## parsed_genderMale -0.638835890 0.1235889571
## DepartmentMedieval Art -0.915180366 0.2439230235
## parsed_genderUnknown -0.947144231 0.0970954170
## DepartmentModern and Contemporary Art -0.976452802 0.2409821656
## DepartmentPhotographs -1.070541480 0.3634130864
## Culture_GroupChinese -1.078186555 0.4546018262
## DepartmentArms and Armor -1.183715599 0.2776141635
## DepartmentArts of Africa, Oceania, and the Americas -1.248126701 0.2287026370
## Culture_GroupChina -1.673788227 0.2959984605
## DepartmentEuropean Sculpture and Decorative Arts -2.206081038 0.2369995592
## Culture_GroupJapan -2.412726975 0.3940162472
## DepartmentDrawings and Prints -2.427078710 0.2907508899
## Classification_GroupVases -2.702013331 0.6264619425
## Classification_GroupSword Furniture -2.823981773 1.1061913345
## (Intercept) -20.353605233 1.6203856953
## z value Pr(>|z|)
## Classification_GroupStone Sculpture 3.912527 9.133524e-05
## Classification_GroupPaintings 4.528696 5.934880e-06
## Classification_GroupSculpture 4.378415 1.195453e-05
## Classification_GroupCodices 3.190311 1.421199e-03
## Classification_GroupWoodwork 2.884634 3.918691e-03
## Classification_GroupWood 2.768857 5.625327e-03
## Classification_GroupEnamels 2.292921 2.185254e-02
## Classification_GroupChordophone 4.048699 5.150304e-05
## Culture_GroupJapanese 4.244861 2.187286e-05
## Culture_GroupGerman 5.125231 2.971731e-07
## Culture_GroupGreek 2.522688 1.164616e-02
## Culture_GroupFrench 3.577314 3.471434e-04
## Culture_GroupOther 2.808397 4.978874e-03
## AccessionYear 10.830601 2.465331e-27
## DepartmentThe Cloisters -2.160542 3.073077e-02
## DepartmentAsian Art -2.876481 4.021358e-03
## DepartmentIslamic Art -2.318484 2.042306e-02
## parsed_genderMale -5.169037 2.353033e-07
## DepartmentMedieval Art -3.751923 1.754835e-04
## parsed_genderUnknown -9.754778 1.759854e-22
## DepartmentModern and Contemporary Art -4.051971 5.078791e-05
## DepartmentPhotographs -2.945798 3.221229e-03
## Culture_GroupChinese -2.371716 1.770568e-02
## DepartmentArms and Armor -4.263888 2.009005e-05
## DepartmentArts of Africa, Oceania, and the Americas -5.457422 4.830985e-08
## Culture_GroupChina -5.654719 1.561011e-08
## DepartmentEuropean Sculpture and Decorative Arts -9.308376 1.298016e-20
## Culture_GroupJapan -6.123420 9.158784e-10
## DepartmentDrawings and Prints -8.347623 6.965069e-17
## Classification_GroupVases -4.313132 1.609576e-05
## Classification_GroupSword Furniture -2.552887 1.068341e-02
## (Intercept) -12.560963 3.460933e-36
##

```

```

## Classification_GroupStone Sculpture Classification_GroupStone Sculpture
## Classification_GroupPaintings Classification_GroupPaintings

```

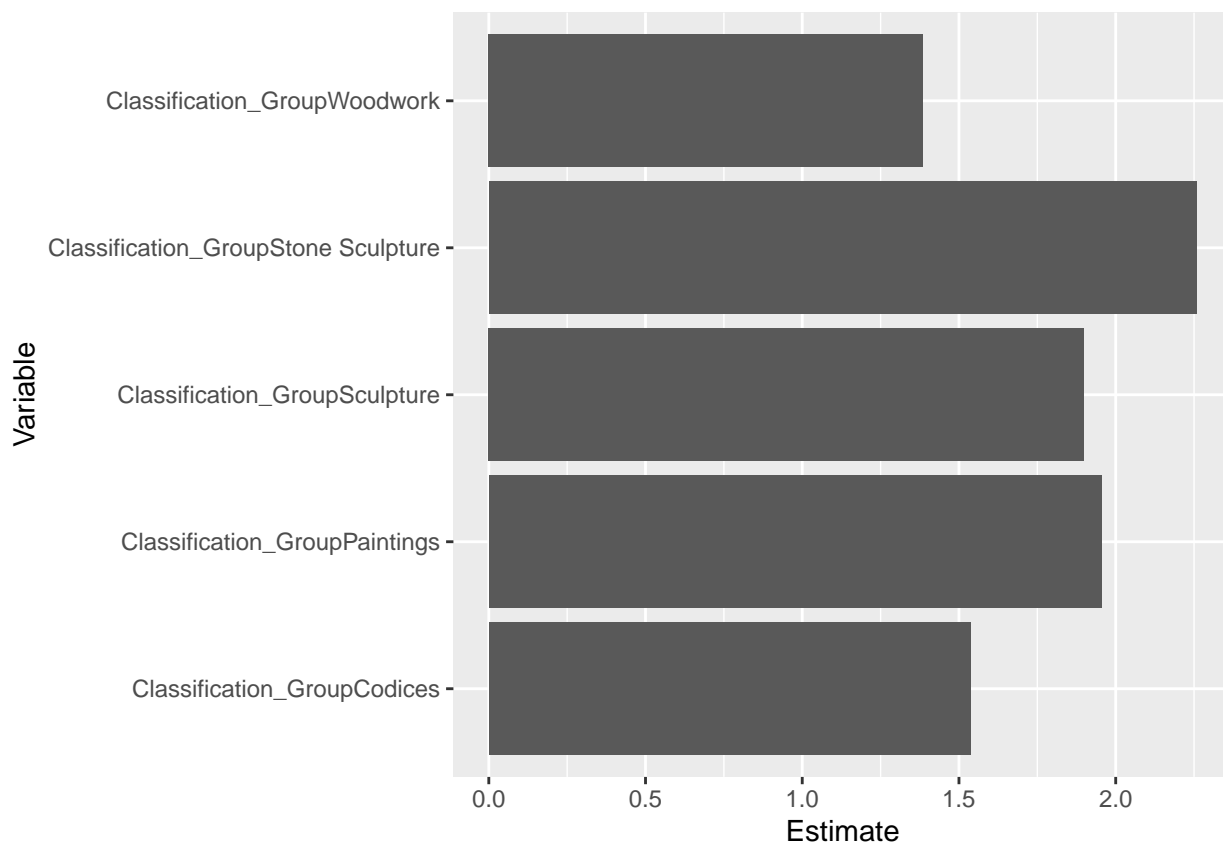
## Classification_GroupSculpture	Classification_GroupSculpture
## Classification_GroupCodices	Classification_GroupCodices
## Classification_GroupWoodwork	Classification_GroupWoodwork
## Classification_GroupWood	Classification_GroupWood
## Classification_GroupEnamels	Classification_GroupEnamels
## Classification_GroupChordophone	Classification_GroupChordophone
## Culture_GroupJapanese	Culture_GroupJapanese
## Culture_GroupGerman	Culture_GroupGerman
## Culture_GroupGreek	Culture_GroupGreek
## Culture_GroupFrench	Culture_GroupFrench
## Culture_GroupOther	Culture_GroupOther
## AccessionYear	AccessionYear
## DepartmentThe Cloisters	DepartmentThe Cloisters
## DepartmentAsian Art	DepartmentAsian Art
## DepartmentIslamic Art	DepartmentIslamic Art
## parsed_genderMale	parsed_genderMale
## DepartmentMedieval Art	DepartmentMedieval Art
## parsed_genderUnknown	parsed_genderUnknown
## DepartmentModern and Contemporary Art	DepartmentModern and Contemporary Art
## DepartmentPhotographs	DepartmentPhotographs
## Culture_GroupChinese	Culture_GroupChinese
## DepartmentArms and Armor	DepartmentArms and Armor
## DepartmentArts of Africa, Oceania, and the Americas	DepartmentArts of Africa, Oceania, and the Americas
## Culture_GroupChina	Culture_GroupChina
## DepartmentEuropean Sculpture and Decorative Arts	DepartmentEuropean Sculpture and Decorative Arts
## Culture_GroupJapan	Culture_GroupJapan
## DepartmentDrawings and Prints	DepartmentDrawings and Prints
## Classification_GroupVases	Classification_GroupVases
## Classification_GroupSword Furniture	Classification_GroupSword Furniture
## (Intercept)	(Intercept)
##	Significant
## Classification_GroupStone Sculpture	TRUE
## Classification_GroupPaintings	TRUE
## Classification_GroupSculpture	TRUE
## Classification_GroupCodices	TRUE
## Classification_GroupWoodwork	TRUE
## Classification_GroupWood	TRUE
## Classification_GroupEnamels	TRUE
## Classification_GroupChordophone	TRUE
## Culture_GroupJapanese	TRUE
## Culture_GroupGerman	TRUE
## Culture_GroupGreek	TRUE
## Culture_GroupFrench	TRUE
## Culture_GroupOther	TRUE
## AccessionYear	TRUE
## DepartmentThe Cloisters	TRUE
## DepartmentAsian Art	TRUE
## DepartmentIslamic Art	TRUE
## parsed_genderMale	TRUE
## DepartmentMedieval Art	TRUE
## parsed_genderUnknown	TRUE
## DepartmentModern and Contemporary Art	TRUE
## DepartmentPhotographs	TRUE
## Culture_GroupChinese	TRUE


```
## DepartmentArms and Armor TRUE
## DepartmentArts of Africa, Oceania, and the Americas TRUE
## Culture_GroupChina TRUE
## DepartmentEuropean Sculpture and Decorative Arts TRUE
## Culture_GroupJapan TRUE
## DepartmentDrawings and Prints TRUE
## Classification_GroupVases TRUE
## Classification_GroupSword Furniture TRUE
## (Intercept) TRUE
```

We have 32 significant variables—all of which have z scores above 2.5. We can see that the top five most positively associated variables are all Classification Groups— Stone Sculpture, Paintings, Sculpture, Codices, and Woodwork.

Let's graph them!

```
significant_vars %>%
  filter(Estimate>1.35) %>%
  ggplot() + geom_bar(aes(y=Variable, x=Estimate), stat='identity')
```



Question 2: If I'm artist and want the MET to think my art is “a popular and important artwork in the collection” what kind of art should I make? Let's remember- the variables we analyzed were Department, AccessionYear, Culture, Artist End Date, Artist Gender, Object Date, and Classification.

Some of these variables are beyond your control—for instance when you're born, what your gender is, and when you make your art. Culture might be under your control somewhat if you're a multicultural artist, but for most people you probably can't manipulate that variable either. So all we're down to is Classification and Department.

Let's look at the significant variables in those dimensions.

```
significant_vars %>%
  filter(str_detect(Variable, "Department|Classification")) %>%
  arrange(desc(Estimate))
```

##	Estimate	Std. Error
## Classification_GroupStone Sculpture	2.2583167	0.5772015
## Classification_GroupPaintings	1.9542554	0.4315272
## Classification_GroupSculpture	1.8993994	0.4338098
## Classification_GroupCodices	1.5369127	0.4817439
## Classification_GroupWoodwork	1.3847016	0.4800268
## Classification_GroupWood	1.3037105	0.4708479
## Classification_GroupEnamels	1.2361244	0.5391046
## Classification_GroupChordophone	1.0732200	0.2650777
## DepartmentThe Cloisters	-0.5423416	0.2510211
## DepartmentAsian Art	-0.5795502	0.2014789
## DepartmentIslamic Art	-0.5814162	0.2507743
## DepartmentMedieval Art	-0.9151804	0.2439230
## DepartmentModern and Contemporary Art	-0.9764528	0.2409822
## DepartmentPhotographs	-1.0705415	0.3634131
## DepartmentArms and Armor	-1.1837156	0.2776142
## DepartmentArts of Africa, Oceania, and the Americas	-1.2481267	0.2287026
## DepartmentEuropean Sculpture and Decorative Arts	-2.2060810	0.2369996
## DepartmentDrawings and Prints	-2.4270787	0.2907509
## Classification_GroupVases	-2.7020133	0.6264619
## Classification_GroupSword Furniture	-2.8239818	1.1061913
##	z value	Pr(> z)
## Classification_GroupStone Sculpture	3.912527	9.133524e-05
## Classification_GroupPaintings	4.528696	5.934880e-06
## Classification_GroupSculpture	4.378415	1.195453e-05
## Classification_GroupCodices	3.190311	1.421199e-03
## Classification_GroupWoodwork	2.884634	3.918691e-03
## Classification_GroupWood	2.768857	5.625327e-03
## Classification_GroupEnamels	2.292921	2.185254e-02
## Classification_GroupChordophone	4.048699	5.150304e-05
## DepartmentThe Cloisters	-2.160542	3.073077e-02
## DepartmentAsian Art	-2.876481	4.021358e-03
## DepartmentIslamic Art	-2.318484	2.042306e-02
## DepartmentMedieval Art	-3.751923	1.754835e-04
## DepartmentModern and Contemporary Art	-4.051971	5.078791e-05
## DepartmentPhotographs	-2.945798	3.221229e-03
## DepartmentArms and Armor	-4.263888	2.009005e-05
## DepartmentArts of Africa, Oceania, and the Americas	-5.457422	4.830985e-08
## DepartmentEuropean Sculpture and Decorative Arts	-9.308376	1.298016e-20
## DepartmentDrawings and Prints	-8.347623	6.965069e-17
## Classification_GroupVases	-4.313132	1.609576e-05
## Classification_GroupSword Furniture	-2.552887	1.068341e-02
##		Variable
## Classification_GroupStone Sculpture		Classification_GroupStone Sculpture
## Classification_GroupPaintings		Classification_GroupPaintings
## Classification_GroupSculpture		Classification_GroupSculpture
## Classification_GroupCodices		Classification_GroupCodices
## Classification_GroupWoodwork		Classification_GroupWoodwork
## Classification_GroupWood		Classification_GroupWood

## Classification_GroupEnamels	Classification_GroupEnamels
## Classification_GroupChordophone	Classification_GroupChordophone
## DepartmentThe Cloisters	DepartmentThe Cloisters
## DepartmentAsian Art	DepartmentAsian Art
## DepartmentIslamic Art	DepartmentIslamic Art
## DepartmentMedieval Art	DepartmentMedieval Art
## DepartmentModern and Contemporary Art	DepartmentModern and Contemporary Art
## DepartmentPhotographs	DepartmentPhotographs
## DepartmentArms and Armor	DepartmentArms and Armor
## DepartmentArts of Africa, Oceania, and the Americas	DepartmentArts of Africa, Oceania, and the Americas
## DepartmentEuropean Sculpture and Decorative Arts	DepartmentEuropean Sculpture and Decorative Arts
## DepartmentDrawings and Prints	DepartmentDrawings and Prints
## Classification_GroupVases	Classification_GroupVases
## Classification_GroupSword Furniture	Classification_GroupSword Furniture
##	Significant
## Classification_GroupStone Sculpture	TRUE
## Classification_GroupPaintings	TRUE
## Classification_GroupSculpture	TRUE
## Classification_GroupCodices	TRUE
## Classification_GroupWoodwork	TRUE
## Classification_GroupWood	TRUE
## Classification_GroupEnamels	TRUE
## Classification_GroupChordophone	TRUE
## DepartmentThe Cloisters	TRUE
## DepartmentAsian Art	TRUE
## DepartmentIslamic Art	TRUE
## DepartmentMedieval Art	TRUE
## DepartmentModern and Contemporary Art	TRUE
## DepartmentPhotographs	TRUE
## DepartmentArms and Armor	TRUE
## DepartmentArts of Africa, Oceania, and the Americas	TRUE
## DepartmentEuropean Sculpture and Decorative Arts	TRUE
## DepartmentDrawings and Prints	TRUE
## Classification_GroupVases	TRUE
## Classification_GroupSword Furniture	TRUE

It seems like most of our variables are actually part of these two categories.

Looking at the most positively and negatively associated variables, it seems like you're best off if you make Stone Sculptures or Paintings, but manage to avoid being put in the "Drawing and Prints" or "European Sculpture and Decorative Arts" departments.

Question 3: What time periods and cultures are the best-represented among the Met's "popular and important artwork"? Let's check out our significant variables again!

```
significant_vars %>%
  filter(str_detect(Variable, "Culture|Year")) %>%
  arrange(desc(Estimate))
```

##		Estimate	Std. Error	z value	Pr(> z)
##	Culture_GroupJapanese	1.064403208	0.2507509927	4.244861	2.187286e-05
##	Culture_GroupGerman	1.003269862	0.1957511470	5.125231	2.971731e-07
##	Culture_GroupGreek	0.941655592	0.3732746515	2.522688	1.164616e-02
##	Culture_GroupFrench	0.519127829	0.1451166605	3.577314	3.471434e-04
##	Culture_GroupOther	0.378393914	0.1347365962	2.808397	4.978874e-03
##	AccessionYear	0.008595885	0.0007936665	10.830601	2.465331e-27

```
## Culture_GroupChinese -1.078186555 0.4546018262 -2.371716 1.770568e-02
## Culture_GroupChina -1.673788227 0.2959984605 -5.654719 1.561011e-08
## Culture_GroupJapan -2.412726975 0.3940162472 -6.123420 9.158784e-10
##
## Variable Significant
## Culture_GroupJapanese Culture_GroupJapanese TRUE
## Culture_GroupGerman Culture_GroupGerman TRUE
## Culture_GroupGreek Culture_GroupGreek TRUE
## Culture_GroupFrench Culture_GroupFrench TRUE
## Culture_GroupOther Culture_GroupOther TRUE
## AccessionYear AccessionYear TRUE
## Culture_GroupChinese Culture_GroupChinese TRUE
## Culture_GroupChina Culture_GroupChina TRUE
## Culture_GroupJapan Culture_GroupJapan TRUE
```

Something weird is going on here—the Japanese culture group is doing great, but the Japan culture group is doing really poorly. You’d expect them to not be significant if this were the case but they both are—in fact they both have Z scores over 4! So there must be some quirk in whether the MET labels art as Japanese or Japan that is really indicative of the propensity of that art to be a highlight.

The other best-off culture groups are the Germans, the Greeks, and the French.

Interestingly, while AccessionYear has a low coefficient, it’s extremely robust, with P functionally equivalent to zero. This is a little surprising because parsed_year is not even a significant variable. The difference between the two is that parsed_year describes when the art was made, whereas AccessionYear describes the year the MET acquired the art. So you’re better off having the MET acquire your art later on, although not much.

Next Steps

An interesting next step would be to try this sort of analysis on other large art museums and see if the model holds. I looked at a few other museums but none I could find had an equivalent to the `Is Highlight` variable.

Another place you could go is start tracking what the museums put in their gallery’s versus what they put in the warehouses and see how that correlates to the sort of variables I looked at—that would get at a similar question but probably in a more universally applicable way.

Sources

Code: <https://github.com/RowanGray472/MET-data-analysis> Data: <https://github.com/metmuseum/openaccess>