Honors Fellows Social Media Analysis

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Introduction

Goal: Come Up with Strategies to Increase Social Media Presence for Honors Fellows

Summary: In order to come up with strategies to increase social media presence, we will be analyzing Instagram posts made by the Honors Fellows Instagram account, @honorsfellowsucla, and the engagement they recieved. We will start by summarizing the data collected from these posts, and then we will move on to analyze how the content affected the engagement of each post. We will be using the statistical programing language R to do this. Through this, we hope to find the ideal content types that will increase engagement for @honorsfellowsucla, and therefore increase the social media presence for Honors Fellows as a whole.

Summary of Data

```
honors_data <- read.csv("Honors_Data.csv", stringsAsFactors = TRUE)
honors_data$Date.Posted <- as.character(honors_data$Date.Posted)</pre>
total_likes <- sum(honors_data$Number.of.Likes)</pre>
total_comments <- sum(honors_data$Number.of.Comments)</pre>
ratio <- round(total_likes / (total_comments * 10))</pre>
honors_data
      Post.. Date.Posted Number.of.Comments Number.of.Likes
                                                                          Category
                                                                            People
                                                                            Events
```

```
People
                                                                      Events
                                                                      People
                                                                      People
                                                                      Events
                                                                      Events
                                                                      Events
                                                                      People
                                                                      Events
                                                            28 Announcements
                                                                      People
                                                            15 Announcements
                                                             22 Announcements
                                                             12 Announcements
                                                             12 Announcements
                                                             11 Announcements
                                                             10 Announcements
Data was collected from the 19 posts on @uclahousing's Instagram page. The data includes the post number, the date posted, the number of
comments, the number of likes, and the category of the post. Each post was put into one of three categories, listed below.
```

[1] "Announcements" "Events" "People"

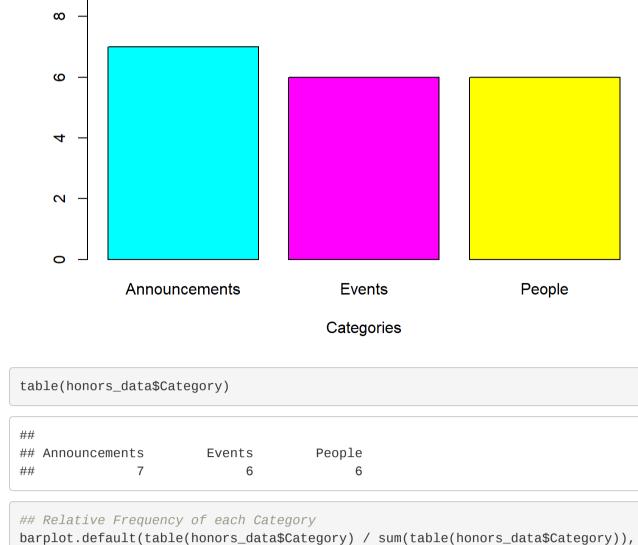
```
Post engagement can be seen in two ways- through the number of likes each post gets and the number of comments each post gets. In order to
combine these into one meaningful statistic, we will be taking the total number of total number of likes @uclahousing has so far, 502, and dividing it
by the total number of comments 9, and then by 10, giving us the proportion 6, and using this number as how much each comment is "worth". Each
post will now be given an engagement score based on their total number of likes and comments, with a like being worth 1 point and a comment
being worth 6 points.
```

honors_data <- cbind(honors_data, "Engagement.Score" = honors_data\$Number.of.Likes + (honors_data\$Number.of.Comme Now, we can summarize our data. ## Graph of Total Frequency of Each Category

barplot.default(table(honors_data\$Category), main = "Total Frequency of Each Category", col = 5:10, ylim = c(0, 10), xlab = "Categories")

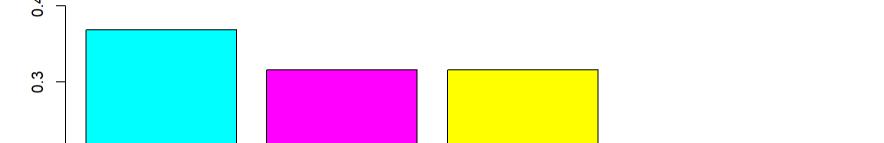
levels(honors_data\$Category)

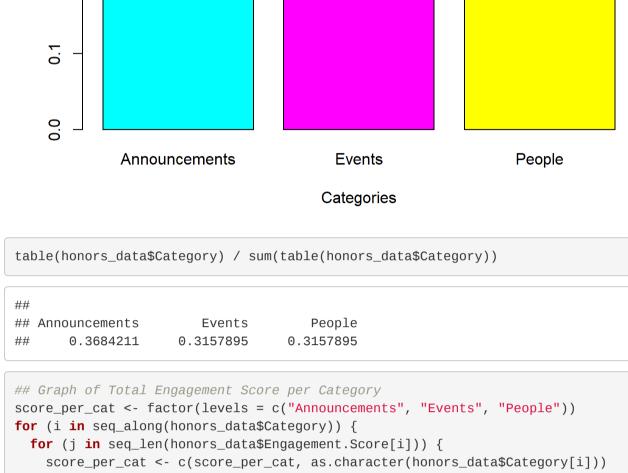
```
Total Frequency of Each Category
```



main = "Relative Frequency of each Category", col = 5:10, ylim = c(0, .4), xlab = "Categories")

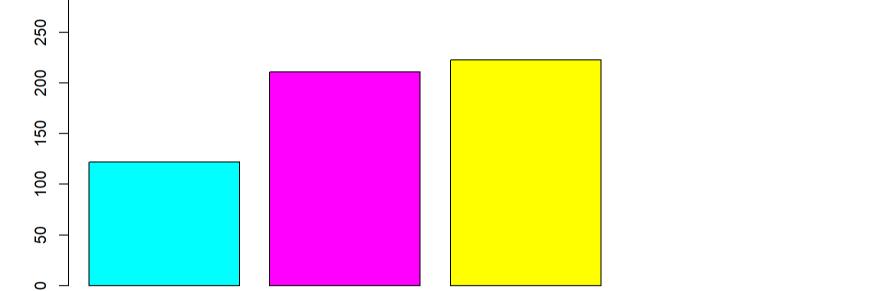
Relative Frequency of each Category 0.4





barplot(table(score_per_cat), main = "Total Engagement Score per Category",

```
col = 5:10, ylim = c(0, 300), xlab = "Categories")
                Total Engagement Score per Category
300
```



People

table(score_per_cat) ## score_per_cat ## Announcements Events People ## 122 211 223 ## Graph of Posts per Year years <- factor(levels = c("2018", "2019", "2020"))for (i in seq_along(honors_data\$Date.Posted)) { years[i] <- substr(honors_data\$Date.Posted[i], nchar(honors_data\$Date.Posted[i]) - 3, nchar(honors_data\$Date.Po</pre> sted[i])) barplot(table(years), main = "Posts per Year", col = 10:15, ylim = c(0, 20), xlab = "Years") Posts per Year 20

Events

Categories

Announcements

15

table(years)

300

40

30

20

10

1.0

cat_freq_year

Announcements

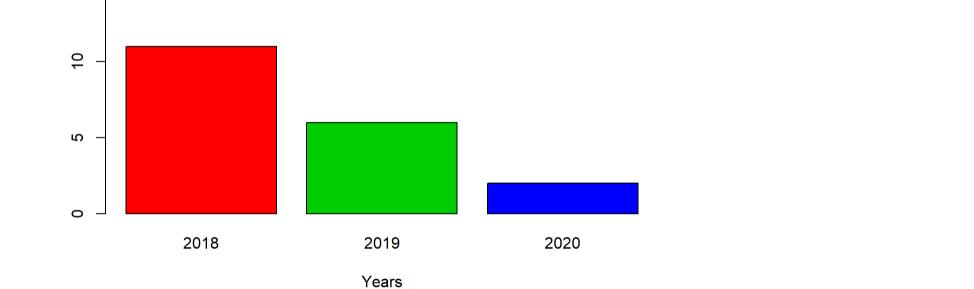
2018

increase in posts made under the Announcements category.

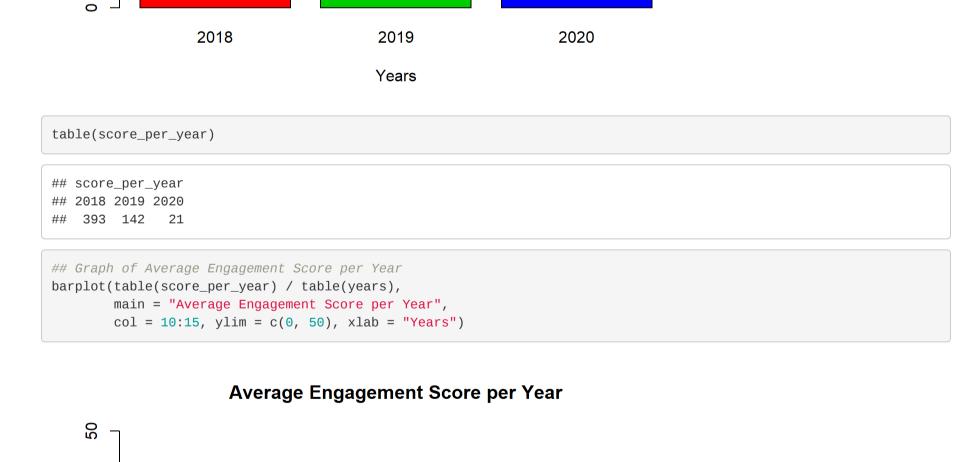
for (i in seq_along(honors_data\$Category)) {

Events People

2018



years ## 2018 2019 2020 11 6 ## Graph of Total Engagement Score per Year $score_per_year <- factor(levels = c("2018", "2019", "2020"))$ for (i in seq_along(years)) { for (j in seq_len(honors_data\$Engagement.Score[i])) { score_per_year <- c(score_per_year, as.character(years[i]))</pre> barplot(table(score_per_year), main = "Total Engagement Score per Year", col = 10:15, ylim = c(0, 500), xlab = "Years")**Total Engagement Score per Year** 500 400





2020

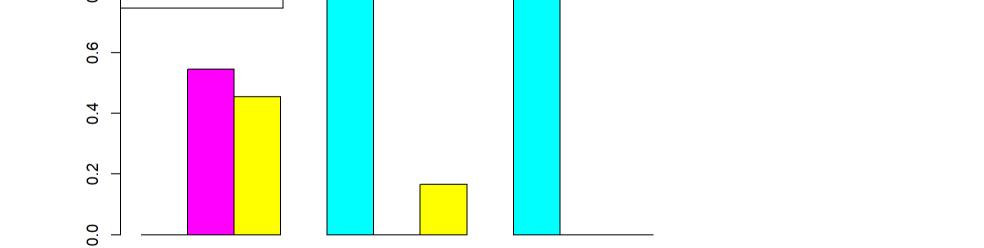
2019

barplot(cat_freq_year, main = "Frequency of Each Category per Year", col = 5:7, ylim = c(0,1), xlab = "Years", beside = TRUE) legend("topleft", c("Announcements", "Events", "People"), fill = 5:10)

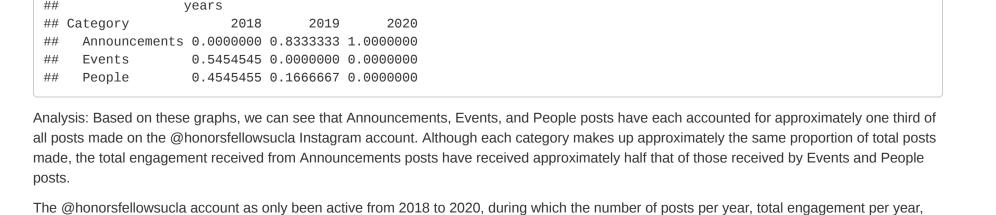
Frequency of Each Category per Year

2019

Years



2020



Content vs Post Engagement The content of each post was categorized into one of three categories: Announcements, Events, People. The median post engagement by category is as follows events <- numeric(0)

and average engagement per year fell over time. This coincided with a decrease in posts made under the Events and People categories and an

if (honors_data\$Category[i] == "Events") { events <- c(events, honors_data\$Engagement.Score[i])</pre> } announcements <- numeric(0)</pre>

for (i in seq_along(honors_data\$Category)) { if (honors_data\$Category[i] == "Announcements") { announcements <- c(announcements, honors_data\$Engagement.Score[i])</pre>

```
people <- numeric(0)</pre>
for (i in seq_along(honors_data$Category)) {
 if (honors_data$Category[i] == "People") {
    people <- c(people, honors_data$Engagement.Score[i])</pre>
}
                                      Announcements
                                                                         Events
                                                                                                   People
                                                   12
                                                                           36.5
                                                                                                     39.5
```

Summary of Findings From this data, we can see that posts of the Events and People categories have a significantly higher median post engagement than posts made in the announcements category. Because of this, it can be recommended that @honorsfellowsucla should increase the number of Events and People posts, which include posts such as photographs of people at events, interviews with honors fellows, etc, and decrease the number of Announcements posts, which include posts such as flyers, invitations for groups to join, etc. in order to increase their engagement and overall social media presence