

Class 9: Candy Mini-Project

Austin Teel (A17293709)

Table of contents

Background	1
Learning objectives	1
Importing Candy Data	2
What is in the data set	2
What is your favorite candy	3
Exploratory Analysis	4
Overall candy Rankings	7
Time to add some useful color	11
Taking a look at pricepercent	12
Optional making the graph look better.	14
Exploring the correlation structure	15
Principle Component Analysis	16
Summary	20

Background

We will be looking at a candy data set of the top rated candy and answer questions that we find after exploring the data.

Learning objectives

The learning objectives are to make scatterplots that have `ggrepel()` labels on scatter plots as well as `plotly()`. We also must interpret correlations through relationships of the variables as well as comparing through PCA.

Importing Candy Data

We must import the csv file first.

```
candy <- read.csv("candy-data.csv", row.names = 1)
head(candy)
```

	chocolate	fruity	caramel	peanutyalmondy	nougat	crispedricewafer
100 Grand	1	0	1	0	0	1
3 Musketeers	1	0	0	0	1	0
One dime	0	0	0	0	0	0
One quarter	0	0	0	0	0	0
Air Heads	0	1	0	0	0	0
Almond Joy	1	0	0	1	0	0

	hard	bar	pluribus	sugarpercent	pricepercent	winpercent
100 Grand	0	1	0	0.732	0.860	66.97173
3 Musketeers	0	1	0	0.604	0.511	67.60294
One dime	0	0	0	0.011	0.116	32.26109
One quarter	0	0	0	0.011	0.511	46.11650
Air Heads	0	0	0	0.906	0.511	52.34146
Almond Joy	0	1	0	0.465	0.767	50.34755

What is in the data set

Q1. How many different candy types are in this dataset?

```
nrow(candy)
```

```
[1] 85
```

There are 85 candy types in the data this is found through our `nrow` function.

Q2. How many fruity candy types are in the dataset?

```
sum(candy$fruity)
```

```
[1] 38
```

There are 38 fruity candies in the data set.

What is your favorite candy

```
candy[candy$competitorname == "3 Musketeers", ]$winpercent
```

```
numeric(0)
```

Q3. What is your favorite candy (other than Twix) in the dataset and what is it's winpercent value?

My favorite candy is 3 Musketeers and as seen above the win percent of this candy is 67.60294.

Q4. What is the winpercent value for "Kit Kat"?

```
candy[candy$competitorname == "Kit Kat", ]$winpercent
```

```
numeric(0)
```

The win percent of kit cat is 76.7686 as seen above.

Q5. What is the winpercent value for "Tootsie Roll Snack Bars"?

```
candy[candy$competitorname == "Tootsie Roll Snack Bars", ]$winpercent
```

```
numeric(0)
```

The win percent of Tootsie Roll Snack Bars is 49.6535.

```
library("skimr")  
skim(candy)
```

Table 1: Data summary

Name	candy
Number of rows	85
Number of columns	12
<hr/>	
Column type frequency:	
numeric	12

Group variables	None
-----------------	------

Variable type: numeric

skim_vari- able	n_miss- ing	com- plete_rate	mean	sd	p0	p25	p50	p75	p100	hist
chocolate	0	1	0.44	0.50	0.00	0.00	0.00	1.00	1.00	
fruity	0	1	0.45	0.50	0.00	0.00	0.00	1.00	1.00	
caramel	0	1	0.16	0.37	0.00	0.00	0.00	0.00	1.00	
peanutyal- mondy	0	1	0.16	0.37	0.00	0.00	0.00	0.00	1.00	
nougat	0	1	0.08	0.28	0.00	0.00	0.00	0.00	1.00	
crispedrice- wafer	0	1	0.08	0.28	0.00	0.00	0.00	0.00	1.00	
hard	0	1	0.18	0.38	0.00	0.00	0.00	0.00	1.00	
bar	0	1	0.25	0.43	0.00	0.00	0.00	0.00	1.00	
pluribus	0	1	0.52	0.50	0.00	0.00	1.00	1.00	1.00	
sugarpercent	0	1	0.48	0.28	0.01	0.22	0.47	0.73	0.99	
pricepercent	0	1	0.47	0.29	0.01	0.26	0.47	0.65	0.98	
winpercent	0	1	50.32	14.71	22.45	39.14	47.83	59.86	84.18	

Q6. Is there any variable/column that looks to be on a different scale to the majority of the other columns in the dataset?

The variable that is different is the winpercent column due to the fact that instead of 0-1 it is out of 100%.

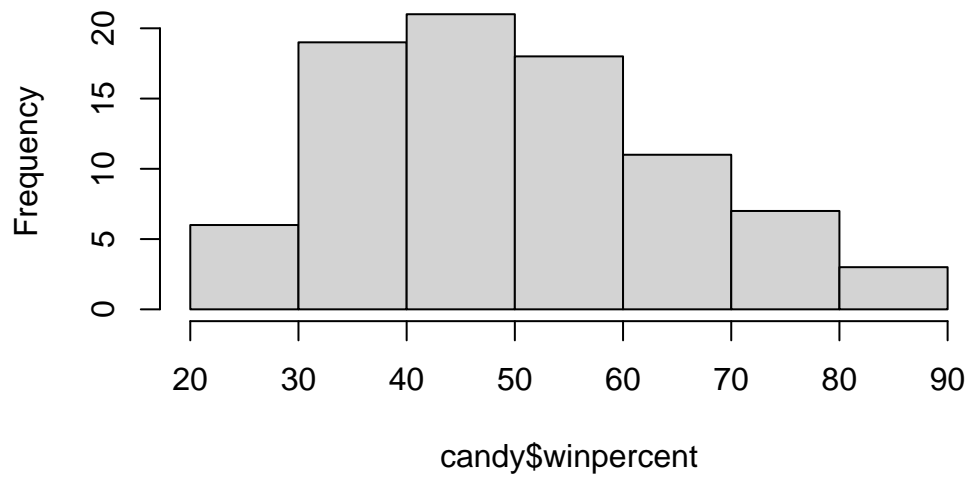
Q7. What do you think a zero and one represent for the `candy$chocolate` column?

The zero and the one represent if that candy has chocolate. zero

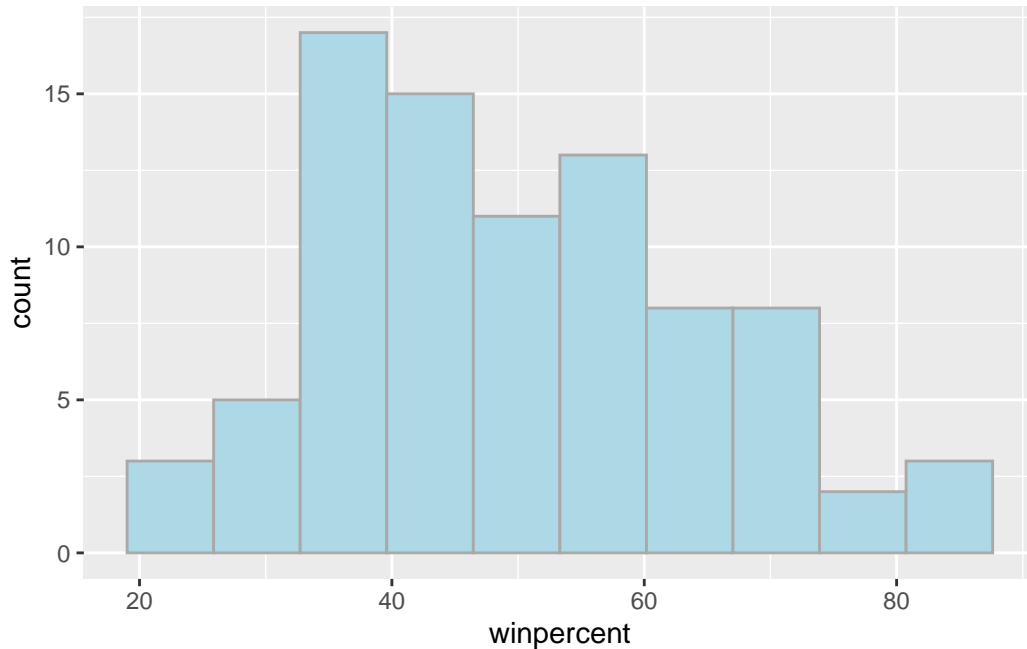
Exploratory Analysis

```
hist(candy$winpercent, breaks=8)
```

Histogram of candy\$winpercent



```
library(ggplot2)
ggplot(candy)+
  aes(winpercent)+
  geom_histogram(bins=10, col="darkgray", fill="lightblue")
```



Q8. Plot a histogram of winpercent values using both base R and ggplot2.

See graphs above.

Q9. Is the distribution of winpercent values symmetrical?

The distribution of winpercent values is not symmetrical and it is skewed to the right.

Q10. Is the center of the distribution above or below 50%?

```
mean(candy$winpercent)
```

```
[1] 50.31676
```

```
median(candy$winpercent)
```

```
[1] 47.82975
```

The center of the distribution mean wise is 50.3% while the median win percent is 47.8%

Q11. On average is chocolate candy higher or lower ranked than fruit candy?

```
chocolate_win <- candy$winpercent[as.logical(candy$chocolate)]
```

```
fruity_win <- candy$winpercent[as.logical(candy$fruity)]
```

```
mean(chocolate_win)
```

```
[1] 60.92153
```

```
mean(fruity_win)
```

```
[1] 44.11974
```

On average the chocolate candy is ranked higher than fruity candy in the average win percent.

Q12. Is this difference statistically significant?

```
t.test(chocolate_win,fruity_win)
```

Welch Two Sample t-test

data: chocolate_win and fruity_win

t = 6.2582, df = 68.882, p-value = 2.871e-08

alternative hypothesis: true difference in means is not equal to 0

95 percent confidence interval:

11.44563 22.15795

sample estimates:

mean of x mean of y

60.92153 44.11974

This difference is statistically different

Overall candy Rankings

Q13. What are the five least liked candy types in this set?

```
library(dplyr)
```

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':

filter, lag

The following objects are masked from 'package:base':

intersect, setdiff, setequal, union

```
candy |> arrange(winpercent) |> head(5)
```

	chocolate	fruity	caramel	peanut	almond	nougat		
Nik L Nip	0	1	0		0	0		
Boston Baked Beans	0	0	0		1	0		
Chiclets	0	1	0		0	0		
Super Bubble	0	1	0		0	0		
Jawbusters	0	1	0		0	0		

	crisp	rice	wafer	hard	bar	pluribus	sugar	percent	price	percent
Nik L Nip				0	0	0	1	0.197		0.976
Boston Baked Beans				0	0	0	1	0.313		0.511
Chiclets				0	0	0	1	0.046		0.325
Super Bubble				0	0	0	0	0.162		0.116
Jawbusters				0	1	0	1	0.093		0.511

	winpercent
Nik L Nip	22.44534
Boston Baked Beans	23.41782
Chiclets	24.52499
Super Bubble	27.30386
Jawbusters	28.12744

The arranged data set says that the candies with the lowest 5 win percents are

- 1 Nik L Nip
- 2 Boston Baked Beans
- 3 Chiclets
- 4 Super Bubble
- 5 Jawbusters

Q14. What are the top 5 all time favorite candy types out of this set?

```
head(candy[order(candy$winpercent, decreasing= TRUE ),], n=5)
```

	chocolate	fruity	caramel	peanut	almondy	nougat
Reese's Peanut Butter cup	1	0	0		1	0
Reese's Miniatures	1	0	0		1	0
Twix	1	0	1		0	0
Kit Kat	1	0	0		0	0
Snickers	1	0	1		1	1

	crisped	rice	wafer	hard	bar	pluribus	sugar	percent
Reese's Peanut Butter cup		0	0	0		0		0.720
Reese's Miniatures		0	0	0		0		0.034
Twix		1	0	1		0		0.546
Kit Kat		1	0	1		0		0.313
Snickers		0	0	1		0		0.546

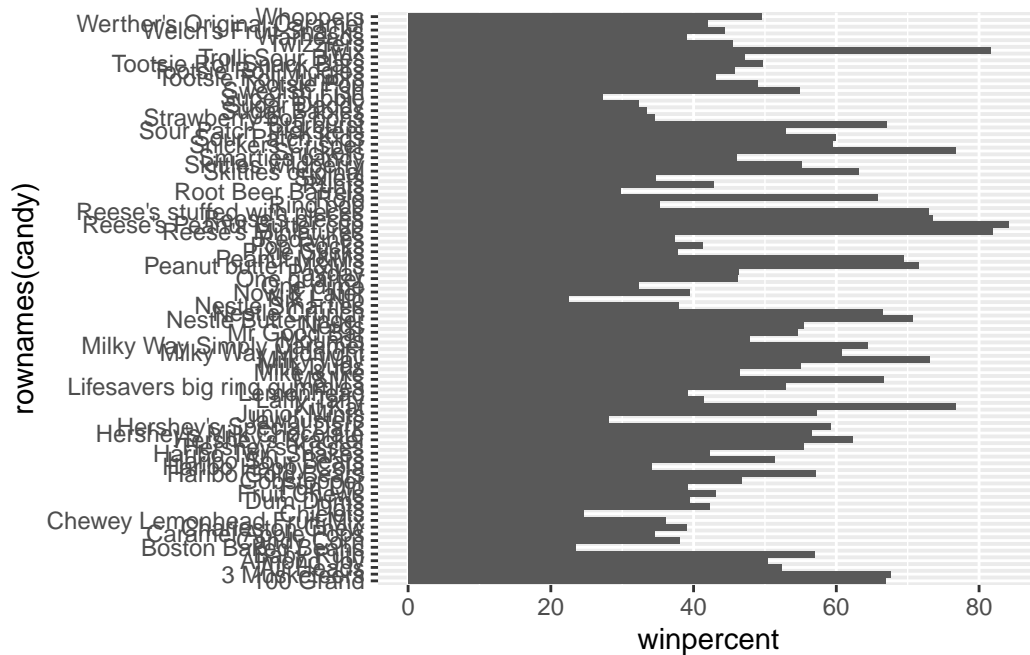
	price	percent	winpercent
Reese's Peanut Butter cup	0.651		84.18029
Reese's Miniatures	0.279		81.86626
Twix	0.906		81.64291
Kit Kat	0.511		76.76860
Snickers	0.651		76.67378

The arranged data set says that the candies with the highest 5 win percents are 1 Reese's Peanut Butter cup 2 Reese's Miniatures 3 Twix 4 Kit Kat 5 Snickers

Q15. Make a first barplot of candy ranking based on winpercent values.

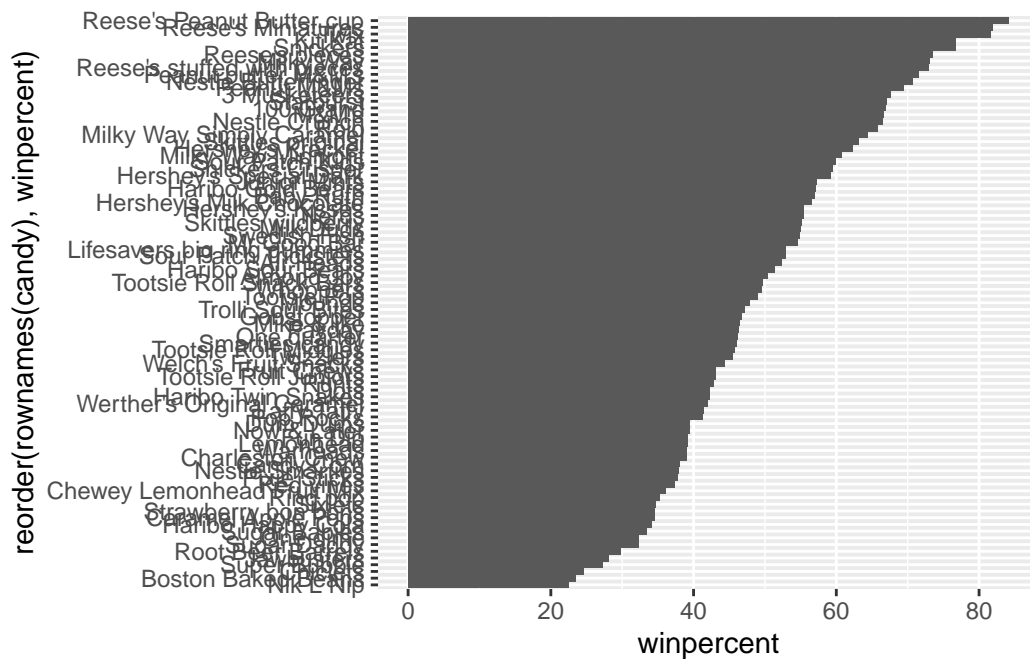
```
candy$name <- rownames(candy)

ggplot(candy) +
  aes(y = rownames(candy), x = winpercent) +
  geom_col()
```



Q16. This is quite ugly, use the `reorder()` function to get the bars sorted by winpercent?

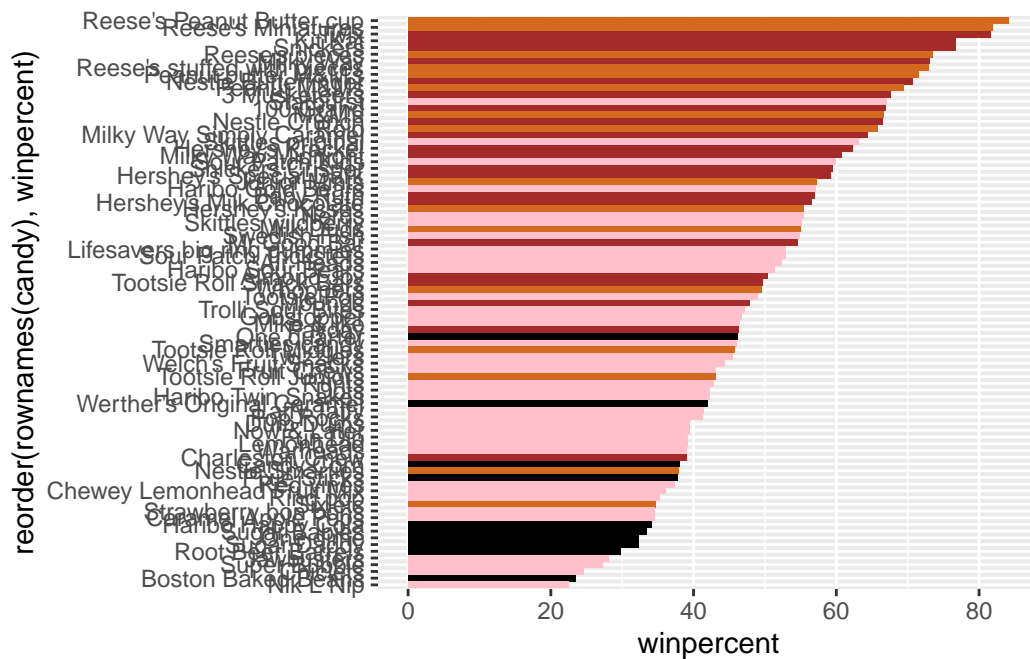
```
ggplot(candy) +
  aes(y = rownames(candy), x = winpercent) +
  aes( winpercent, reorder(rownames(candy),winpercent)) +
  geom_col()
```



Time to add some useful color

```
my_cols=rep("black", nrow(candy))
my_cols[as.logical(candy$chocolate)] = "chocolate"
my_cols[as.logical(candy$bar)] = "brown"
my_cols[as.logical(candy$fruity)] = "pink"

ggplot(candy) +
  aes(winpercent, reorder(rownames(candy),winpercent)) +
  geom_col(fill=my_cols)
```



Q17. What is the worst ranked chocolate candy?

The worst ranked chocolate candy based off of the colored barplot is the candy “sixlets”

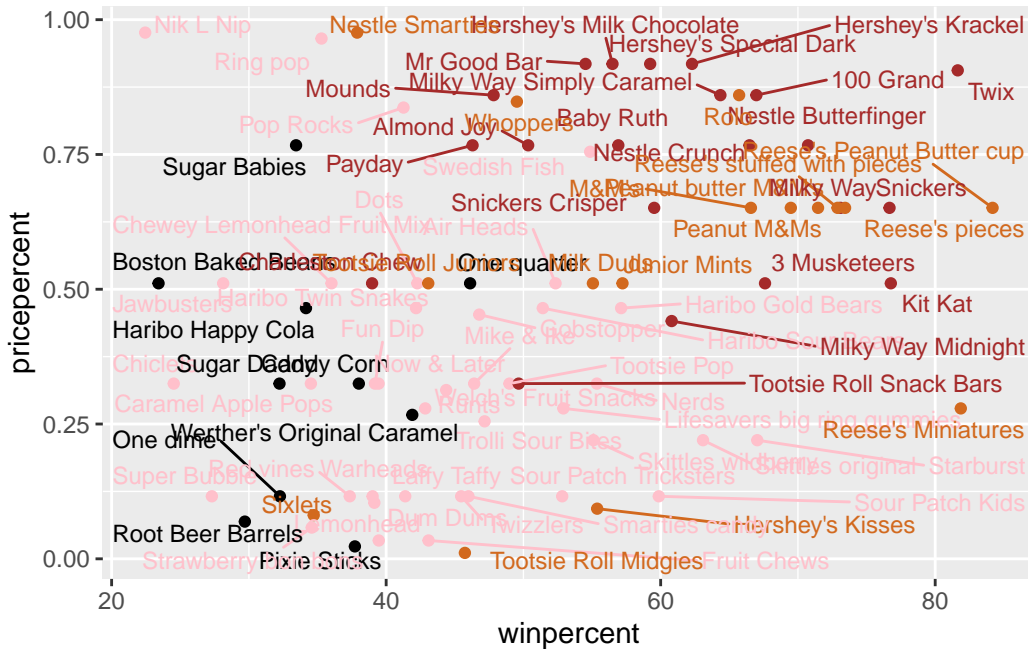
Q18. What is the best ranked fruity candy?

The best ranked fruity candy based off of the colored bar plot is the candy “starburst”

Taking a loof at pricepercent

```
library(ggrepel)

ggplot(candy) +
  aes(x=winpercent, y=pricepercent, label=rownames(candy)) +
  geom_point(col=my_cols) +
  geom_text_repel(col=my_cols, size=3.3, max.overlaps = 38)
```



Q19. Which candy type is the highest ranked in terms of winpercent for the least money - i.e. offers the most bang for your buck?

```
ord <- order(candy$winpercent, decreasing = TRUE)
head( candy[ord,c(11,12)], n=5 )
```

	pricepercent	winpercent
Reese's Peanut Butter cup	0.651	84.18029
Reese's Miniatures	0.279	81.86626
Twix	0.906	81.64291
Kit Kat	0.511	76.76860
Snickers	0.651	76.67378

The most bang for your buck candy based on the order above as well as observations from the graph above is the candy “Reese’s Miniatures”. We can see that this candy has the highest win percent in the top 5 with the least amount of price percent.

Q20. What are the top 5 most expensive candy types in the dataset and of these which is the least popular?

```
ord <- order(candy$pricepercent, decreasing = TRUE)
head( candy[ord,c(11,12)], n=5 )
```

	pricepercent	winpercent
Nik L Nip	0.976	22.44534
Nestle Smarties	0.976	37.88719
Ring pop	0.965	35.29076
Hershey's Krackel	0.918	62.28448
Hershey's Milk Chocolate	0.918	56.49050

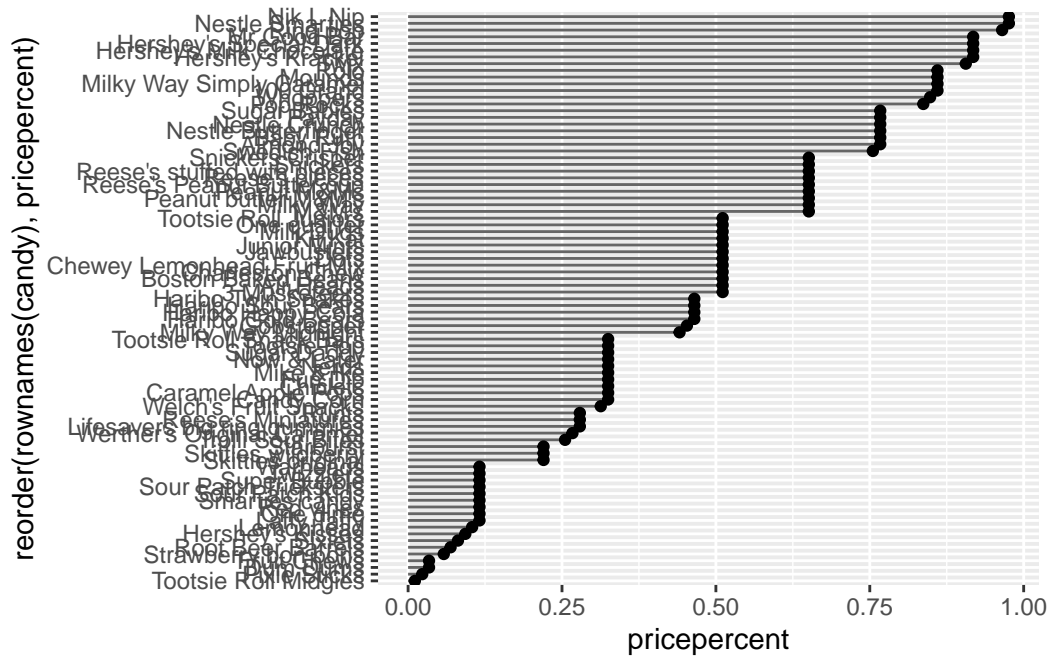
The top 5 most expensive candies are 1. Nik L Nip
 2. Nestle Smarties
 3. Ring pop 4. Hershey's Krackel
 5. Hershey's Milk Chocolate

The least popular of these candies is the most expensive one "Nik L Nip" and this can be seen on the order above as well as the graph above.

Optional making the graph look better.

Q21. Make a barplot again with `geom_col()` this time using `pricepercent` and then improve this step by step, first ordering the x-axis by value and finally making a so called "dot chat" or "lollipop" chart by swapping `geom_col()` for `geom_point()` + `geom_segment()`.

```
ggplot(candy) +
  aes(pricepercent, reorder(rownames(candy), pricepercent)) +
  geom_segment(aes(yend = reorder(rownames(candy), pricepercent),
                    xend = 0), col="gray40") +
  geom_point()
```

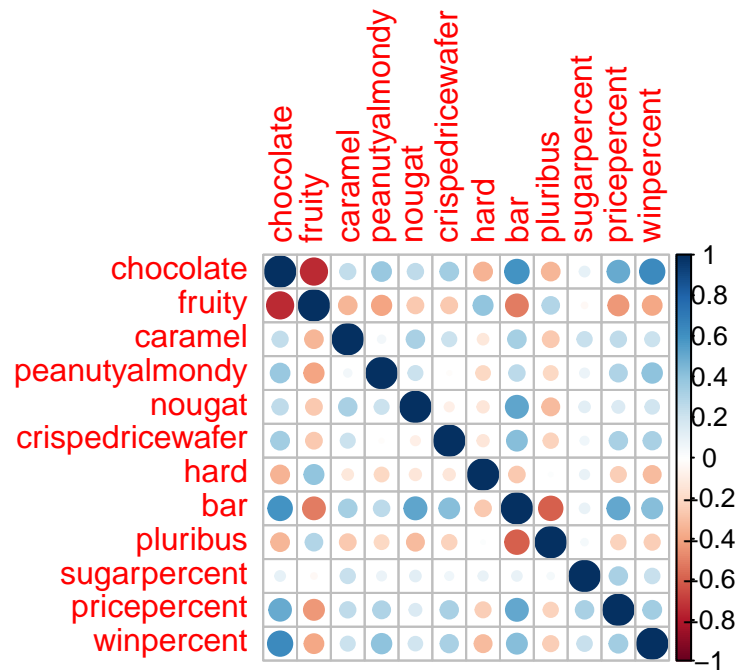


Exploring the correlation structure

```
numeric_candy <- candy[sapply(candy, is.numeric)]
library(corrplot)
```

corrplot 0.95 loaded

```
cij <- cor(numeric_candy)
corrplot(cij)
```



Q22. Examining this plot what two variables are anti-correlated (i.e. have minus values)?

Based on the plot the two variables that are anti-correlated are chocolate and fruity.

Q23. Similarly, what two variables are most positively correlated?

The two variables that are mosy positively correlated are win percent and Chocolate.

Principle Component Analysis

```
pca <- prcomp(numeric_candy, scale=TRUE)
summary(pca)
```

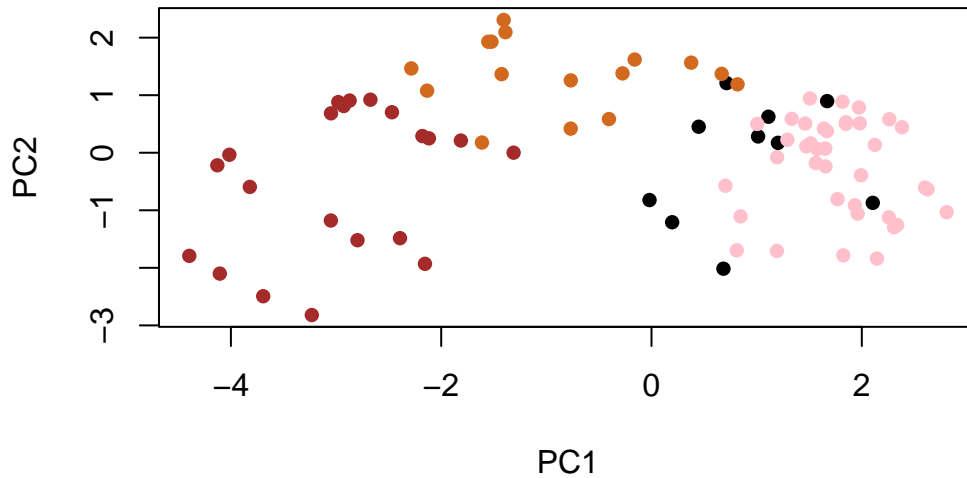
Importance of components:

	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	2.0788	1.1378	1.1092	1.07533	0.9518	0.81923	0.81530
Proportion of Variance	0.3601	0.1079	0.1025	0.09636	0.0755	0.05593	0.05539
Cumulative Proportion	0.3601	0.4680	0.5705	0.66688	0.7424	0.79830	0.85369

	PC8	PC9	PC10	PC11	PC12
Standard deviation	0.74530	0.67824	0.62349	0.43974	0.39760
Proportion of Variance	0.04629	0.03833	0.03239	0.01611	0.01317

Cumulative Proportion 0.89998 0.93832 0.97071 0.98683 1.00000

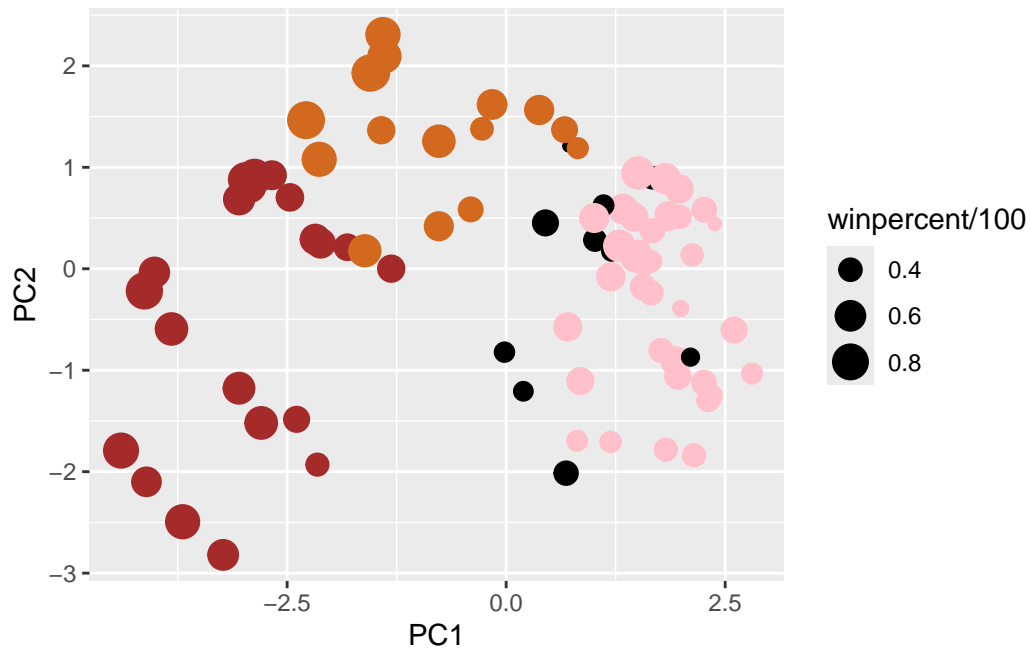
```
plot(pca$x[,1:2], col=my_cols, pch=16)
```



```
my_data <- cbind(candy, pca$x[,1:3])

p <- ggplot(my_data) +
  aes(x=PC1, y=PC2,
      size=winpercent/100,
      text=rownames(my_data),
      label=rownames(my_data)) +
  geom_point(col=my_cols)
```

p

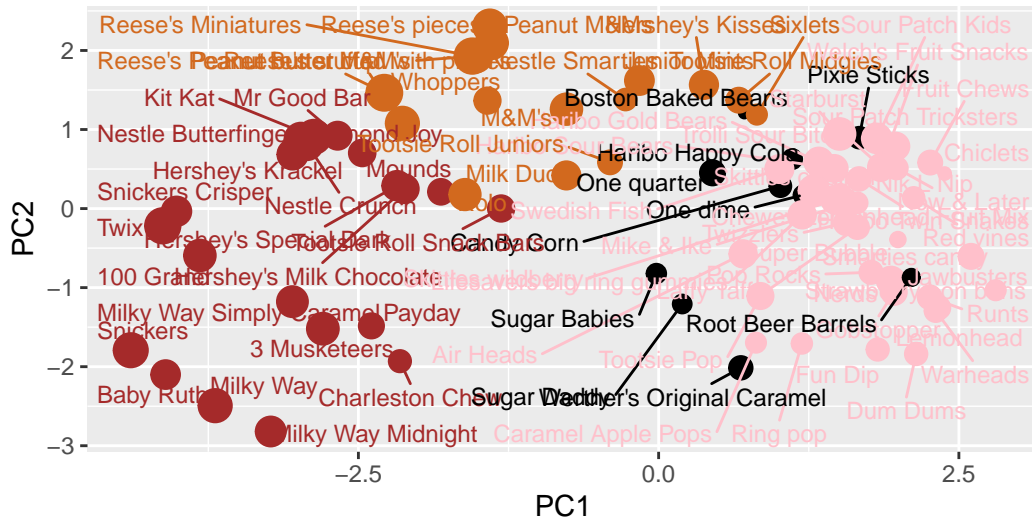


```
library(ggrepel)

p + geom_text_repel(size=3.3, col=my_cols, max.overlaps = 38) +
  theme(legend.position = "none") +
  labs(title="Halloween Candy PCA Space",
        subtitle="Colored by type: chocolate bar (dark brown), chocolate other (light brown),",
        caption="Data from 538")
```

Halloween Candy PCA Space

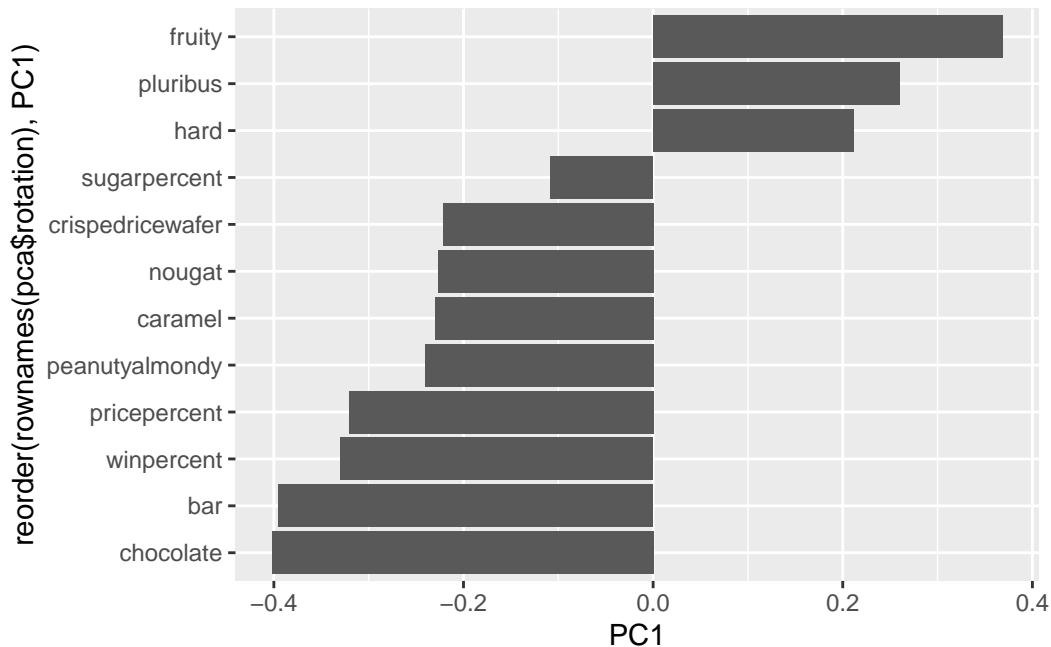
Colored by type: chocolate bar (dark brown), chocolate other (light brown),



Data from 538

```
#library(plotly)
#ggplotly(p)
```

```
ggplot(pca$rotation) +  
  aes(PC1, reorder(rownames(pca$rotation), PC1)) +  
  geom_col()
```



Q24. Complete the code to generate the loadings plot above. What original variables are picked up strongly by PC1 in the positive direction? Do these make sense to you? Where did you see this relationship highlighted previously?

The three variables that are picked up strongly positive in PC1 are fruity, pluribus, and hard. This makes sense because PC1 is representative of these three candies. These candies do not win a lot the relationship was previously highlighted in the popularity plot.

Summary

Q25. Based on your exploratory analysis, correlation findings, and PCA results, what combination of characteristics appears to make a “winning” candy? How do these different analyses (visualization, correlation, PCA) support or complement each other in reaching this conclusion?

The combination of characteristics that appears to make a “winning” candy is of course being a chocolate bar and this can be seen with the graphs that map the popularity amongst the candies. We can also see that in the PCA graph that plots PCA amounts all of this qualitative factors that make popular candies are in the opposite direction of the non popular candy qualities that make up PCA1.