

Class09: Candy Mini Project

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Background

In today's mini-project we will analyze candy data with the exploratory graphics, basic statistics, correlation analysis, and principal component analysis methods we have been learning thus far.

Data Import

The data comes as a CSV file from 538.

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

	chocolate	fruity	caramel	peanuty	almondy	nougat	crisped	rice	wafer
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

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

```
nrow(candy)
```

[1] 85

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

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

[1] 38

Q3. What is your favorite candy (other than Twix) in the dataset and what is its windpercent value?

```
candy["Milky Way", "winpercent"]
```

[1] 73.09956

Q4. What is the winpercent value for “Kit kat”?

```
candy["Kit Kat", "winpercent"]
```

[1] 76.7686

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

```
candy["Tootsie Roll Snack Bars", "winpercent"]
```

[1] 49.6535

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

Table 1: Data summary

Name	candy
Number of rows	85
Number of columns	12
Column type frequency:	
numeric	12
Group variables	None

Variable type: numeric

skim_variable	n_missing	complete_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	
peanutyalmond	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	
crispedricewafer	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?

Yes, winpercent seems to be on a different scale to the majority of the other columns in the dataset. Winpercent values are way larger.

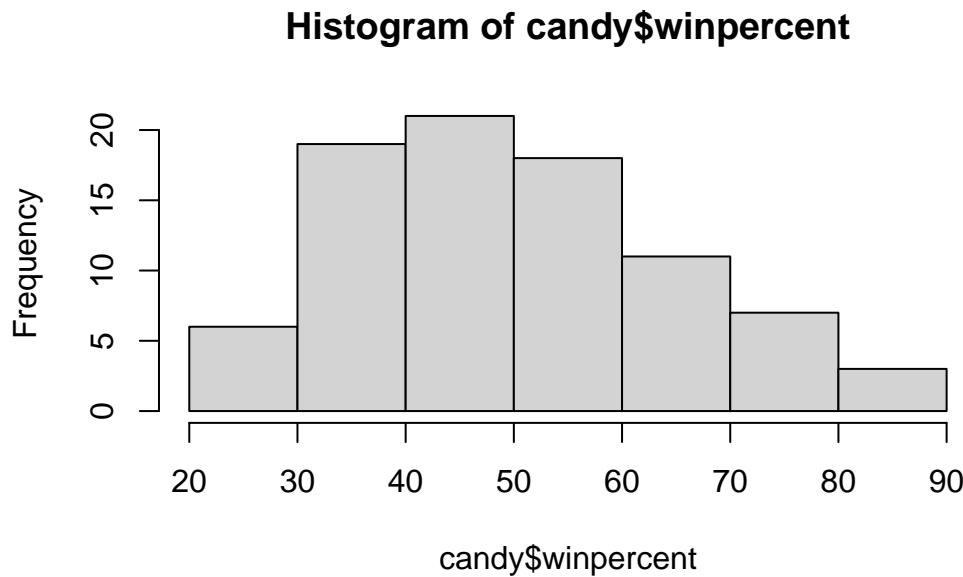
Q7. What do you think a zero and one represent for the candy\$chocolate column?

For the candy\$chocolate column, I believe zero represents no chocolate and one represents chocolate.

Exploratory Analysis

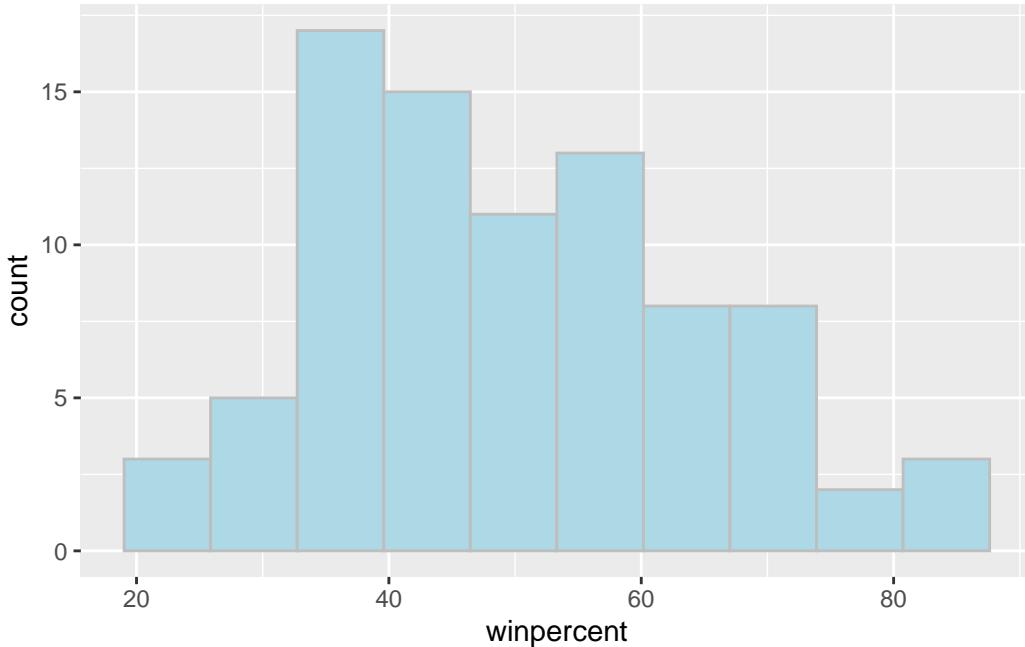
Q8. Plot a histogram of winpercent values using both base R an ggplot2.

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



```
library(ggplot2)

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



Q9. Is the distribution of winpercent values symmetrical?

No, the distribution of winpercent values is not symmetrical.

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

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

[1] 50.31676

The center of the distribution is above 50%.

```
summary(candy$winpercent)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
22.45	39.14	47.83	50.32	59.86	84.18

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

On average, chocolate candy is higher ranked than fruit candy as chocolate (x) = 60.92153, and the mean for fruity (y) = 44.11974.

Q12. Is this difference statistically significant?

No, the difference was not statistically significant as the p-value was very low = 2.781e-08.

```
t.test(candy$winpercent[as.logical(candy$chocolate)],
       candy$winpercent[as.logical(candy$fruity)])
```

Welch Two Sample t-test

```
data: candy$winpercent[as.logical(candy$chocolate)] and candy$winpercent[as.logical(candy$fruity)]
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
```

Overall Candy Rankings

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

The five least liked candy types within the data set are: Nik L Nip, Boston Baked Beans, Chiclets, Supper Bubble, and Jawbusters as shown below:

```
head(candy[order(candy$winpercent), ], 5)
```

	chocolate	fruity	caramel	peanuty	almondy	nougat		
Nik L Nip	0	1	0	0	0	0		
Boston Baked Beans	0	0	0	1	0	0		
Chiclets	0	1	0	0	0	0		
Super Bubble	0	1	0	0	0	0		
Jawbusters	0	1	0	0	0	0		
	crisped	rice	wafer	hard	bar	pluribus	sugarpercent	pricepercent
Nik L Nip	0	0	0	1	0	0.197	0.976	
Boston Baked Beans	0	0	0	1	0	0.313	0.511	
Chiclets	0	0	0	1	0	0.046	0.325	
Super Bubble	0	0	0	0	0	0.162	0.116	
Jawbusters	0	1	0	1	0	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

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

The top 5 all time favorite candy types out of this set are: Reese's Peanut Butter cup, Reese's Miniatures, Twix, Kit Kat, and Snickers.

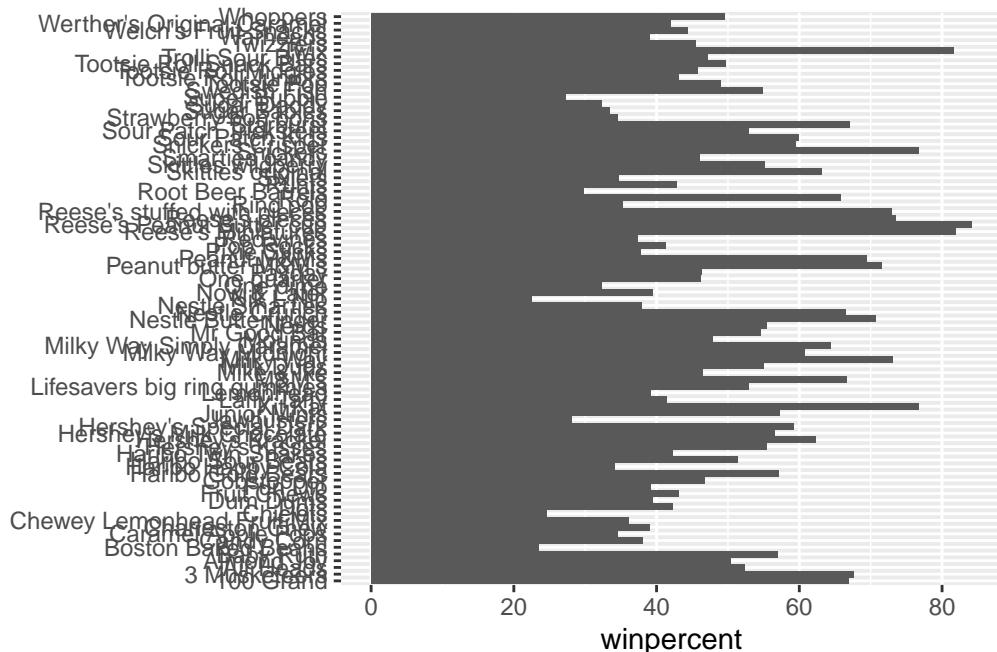
```
head(candy[order(-candy$winpercent), ], 5)
```

	chocolate	fruity	caramel	peanuty	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
	crispedrice	wafers	hard	bar	pluribus	sugarpercent
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				

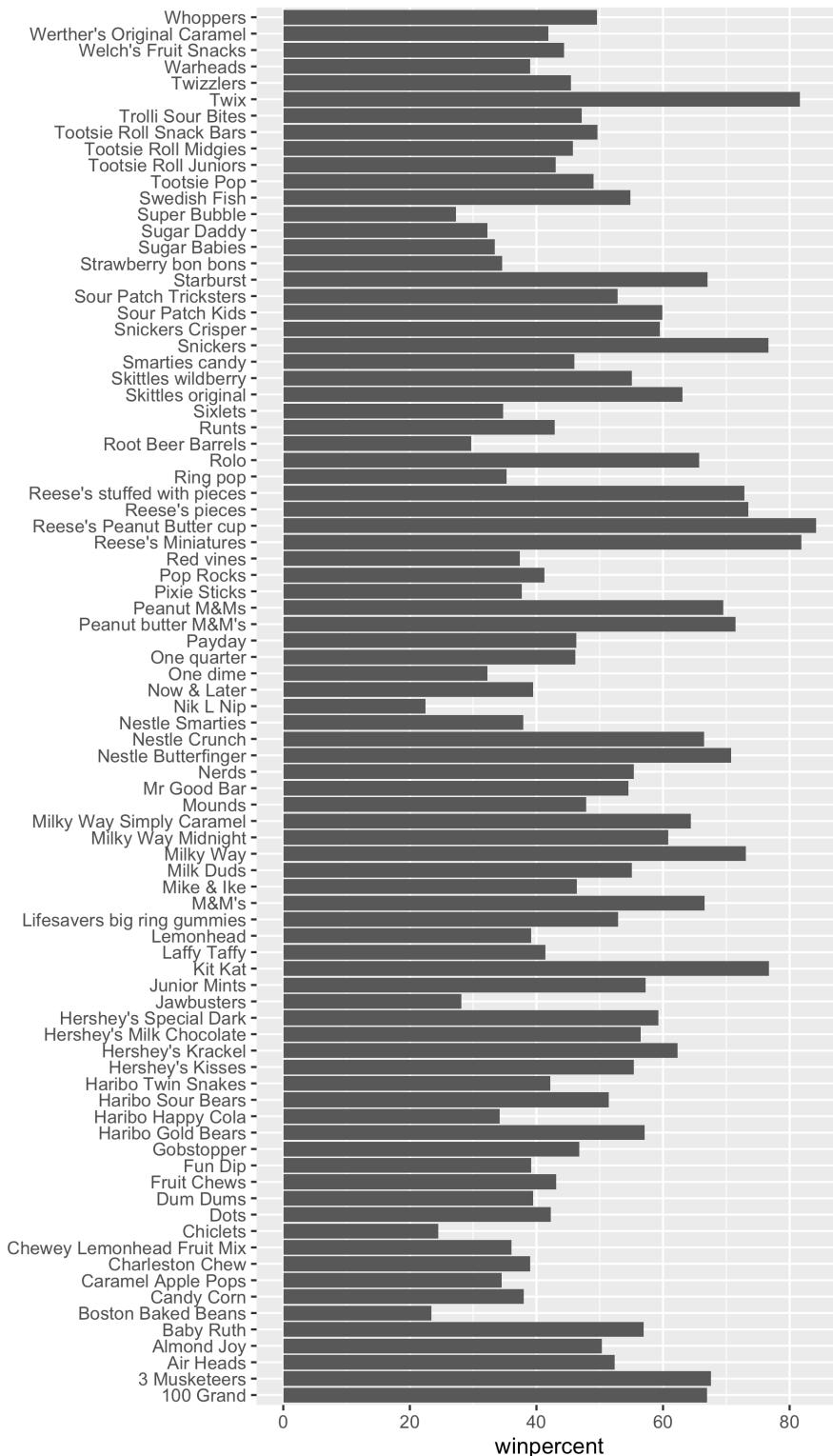
Q15. Make a first barplot of candy ranking based on winpercent values.

```
library(ggplot2)

ggplot(candy) +
  aes(winpercent, rownames(candy)) +
  geom_col()+
  ylab("") #turn off y-label that we don't need
```



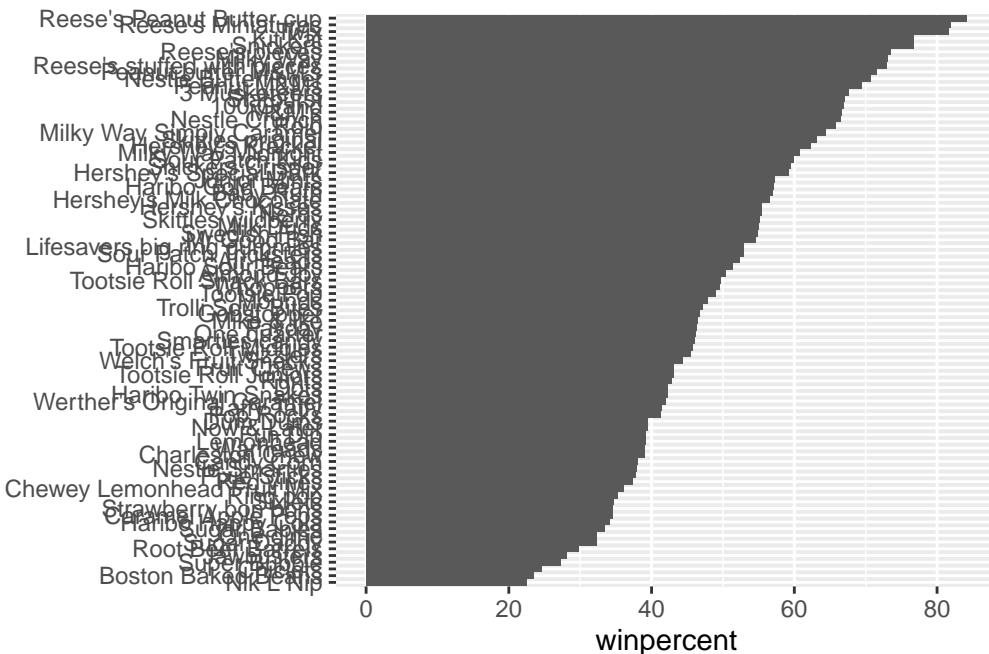
```
ggsave("barplot1.png", height=10, width=6)
```



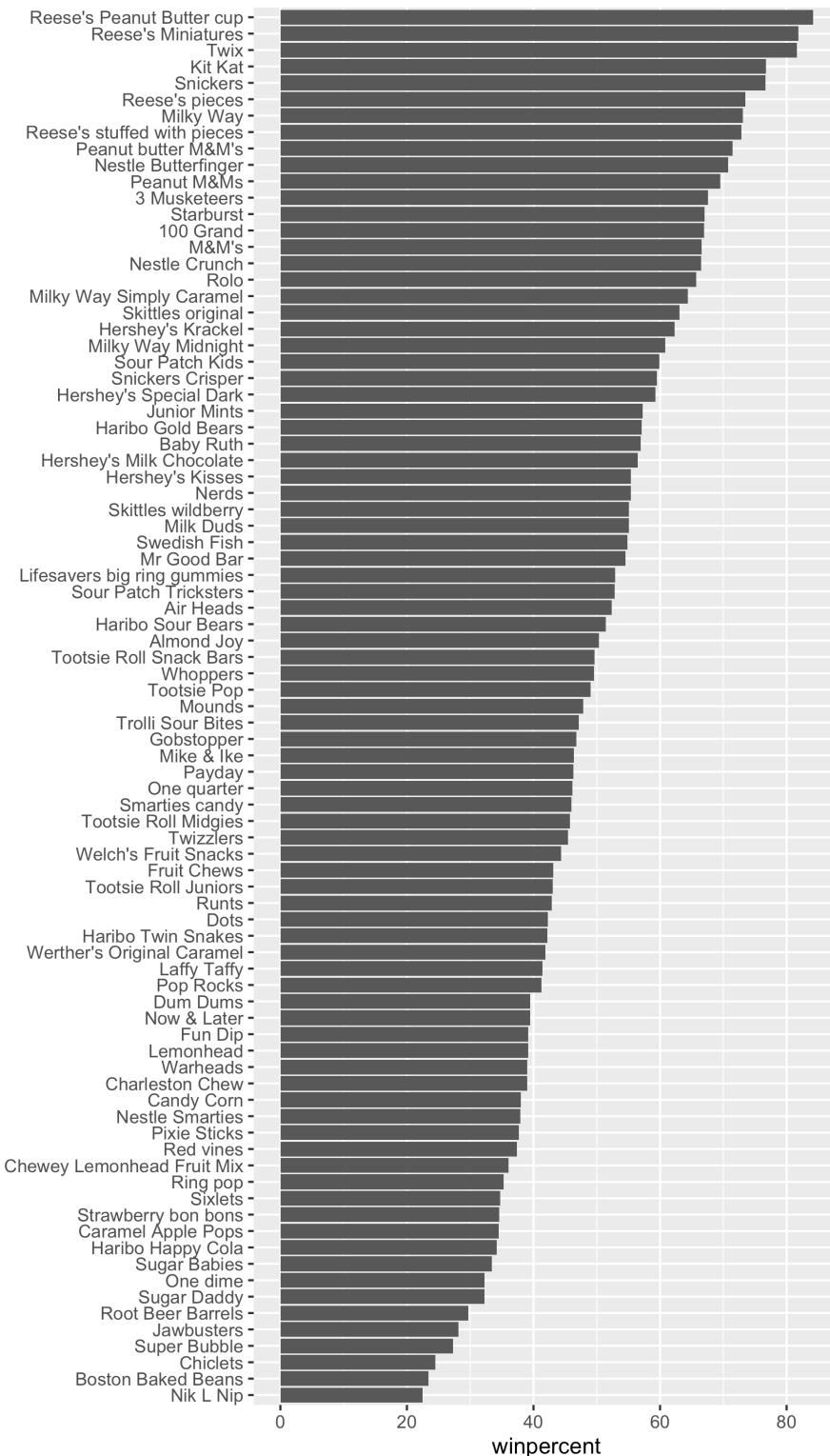
Q16. This is quite ugly, use the reorder() function to get the bars sorted by winpercent?

```
library(ggplot2)

ggplot(candy) +
  aes(winpercent, reorder(rownames(candy), winpercent)) +
  geom_col() +
  ylab("") # turn off Y-label that we don't need
```



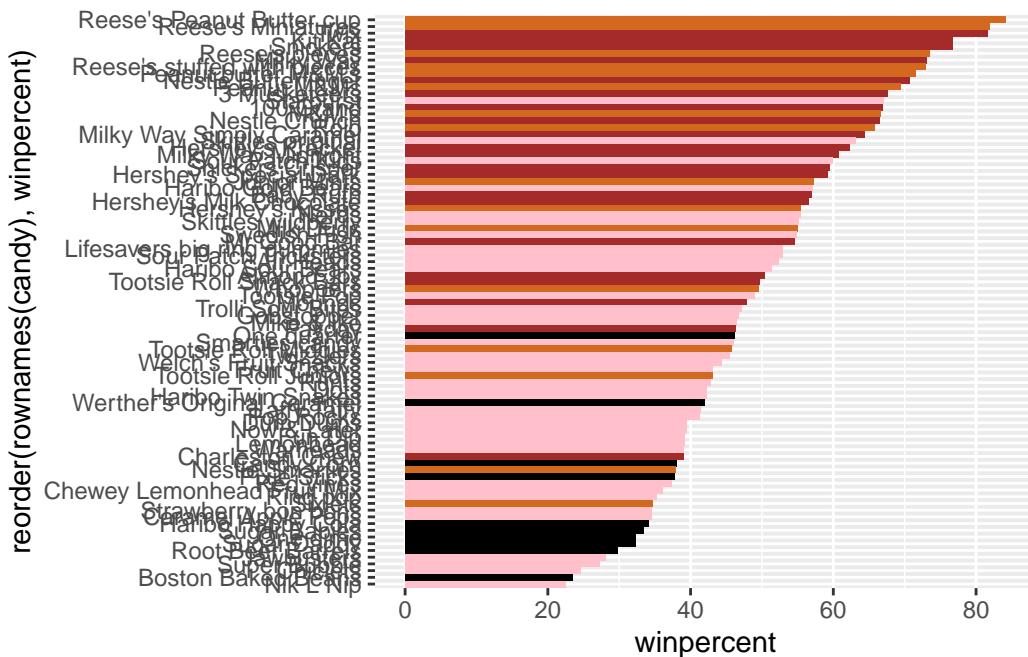
```
ggsave("barplot2.png", height=10, width=6)
```



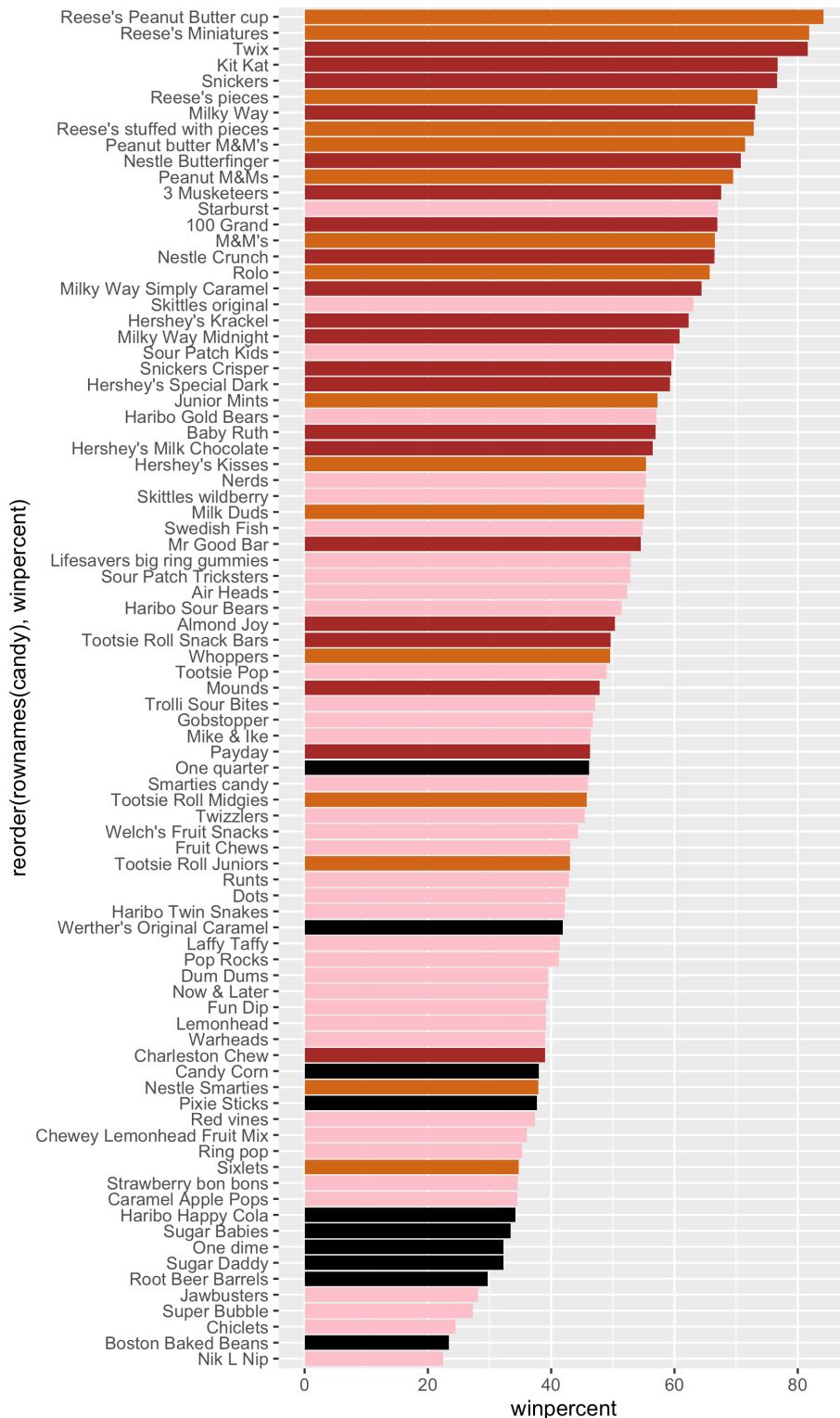
Time to add some useful color

```
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)
```



```
ggsave("barplot3.png", height=10, width=6)
```



Q17. What is the worst ranked chocolate candy?

The worst ranked chocolate candy is Sixlets.

Q18. What is the best ranked fruity candy?

The best ranked fruity candy is Starburst.

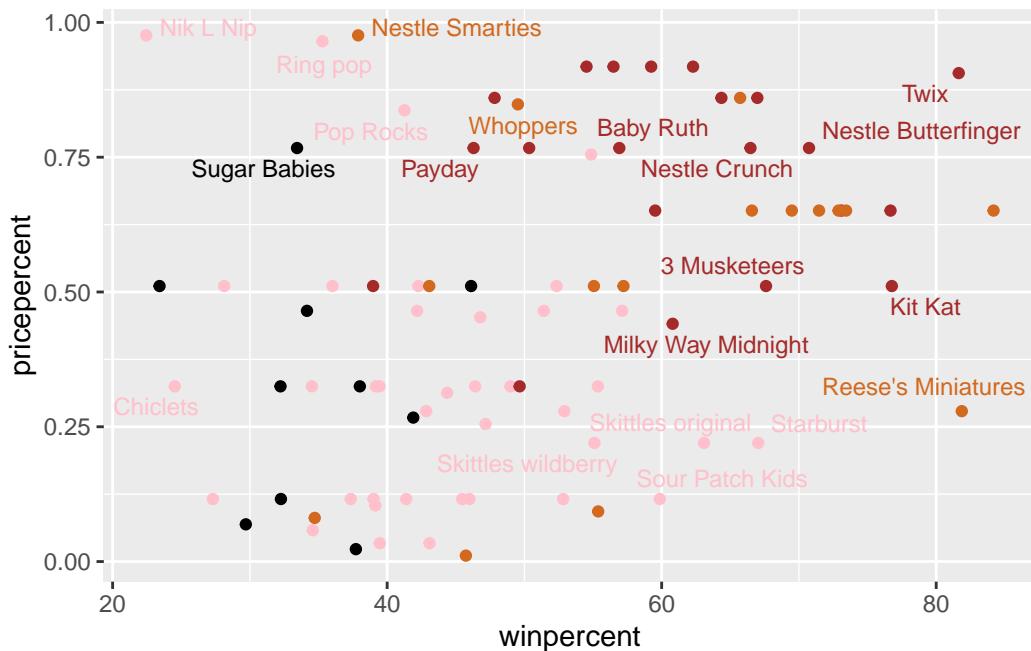
Taking a look at pricepercent

```
candy$my_cols <- my_cols
```

```
library(ggrepel)

ggplot(candy) +
  aes(winpercent, pricepercent, label = rownames(candy)) +
  geom_point(col = my_cols) +
  geom_text_repel(col = my_cols, size = 3.3, max.overlaps = 5)
```

Warning: ggrepel: 65 unlabeled data points (too many overlaps). Consider increasing max.overlaps



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?"

Reese's Miniatures offers the most bang for your buck.

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

The top 5 most expensive candy types in the dataset are: Nik L Nip, Nestle Smarties, Ring pop, Hershey's Krackel, Hershey's Milk Chocolate. Nik L Nip 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

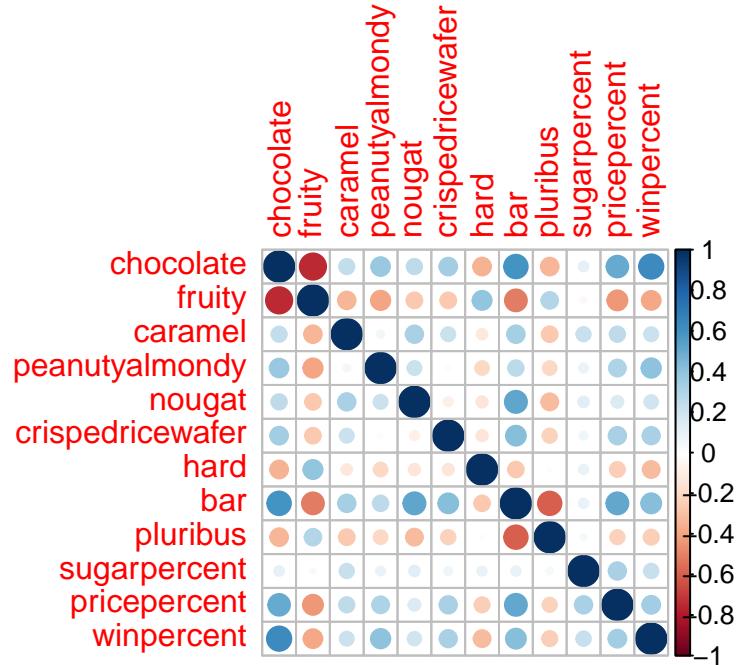
Exploring the correlation structure

Pearson correlation values range from -1 to +1

```
library(corrplot)
```

corrplot 0.95 loaded

```
cij <- cor(candy[sapply(candy, is.numeric)])
corrplot(cij)
```



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

Chocolate and fruity are anti-correlated (red dots).

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

Chocolate and bar are most positively correlated.

Principal Component Analysis

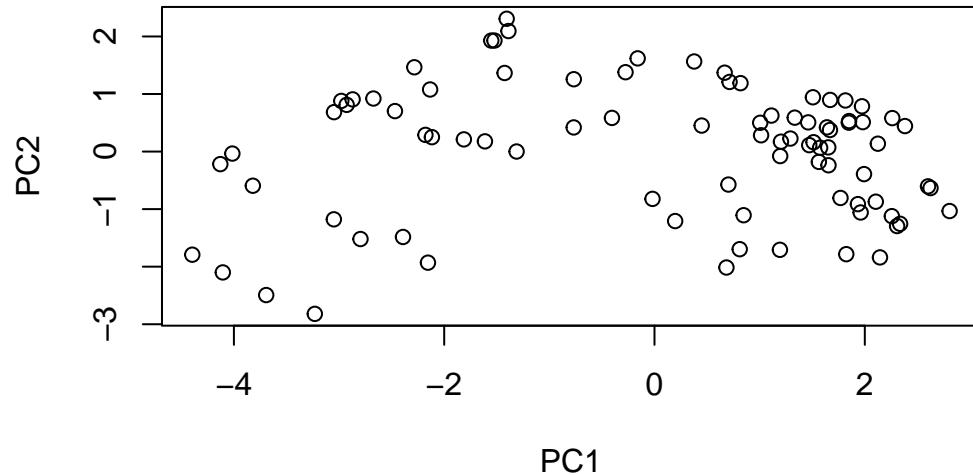
```
pca <- prcomp(candy[sapply(candy, is.numeric)], scale = TRUE)
summary(pca)
```

Importance of components:

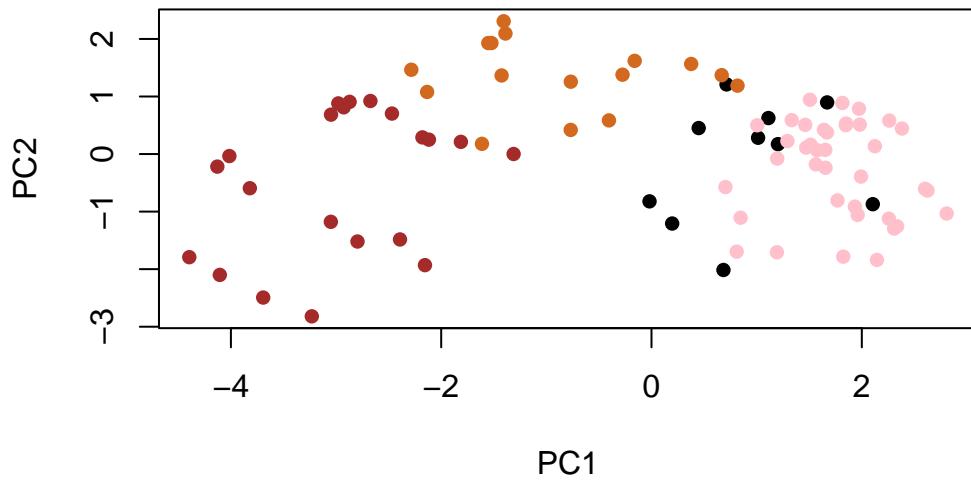
	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	2.0788	1.1378	1.1092	1.07533	0.9518	0.81923	0.81530
Proportion of Variance	0.3601	0.1079	0.1025	0.09636	0.0755	0.05593	0.05539
Cumulative Proportion	0.3601	0.4680	0.5705	0.66688	0.7424	0.79830	0.85369
	PC8	PC9	PC10	PC11	PC12		
Standard deviation	0.74530	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])
```

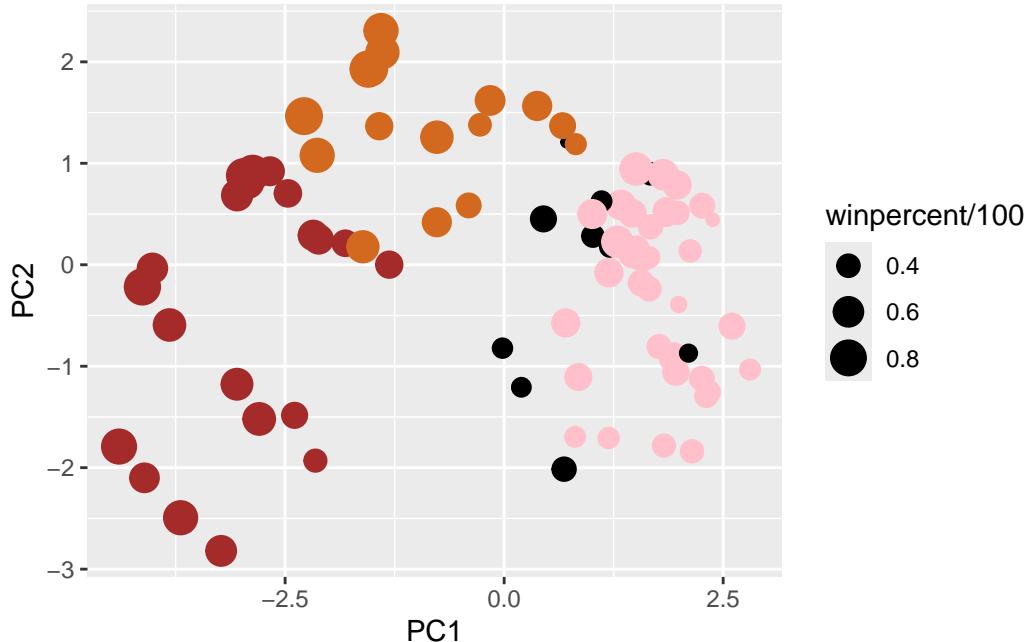


```
plot(pca$x[, 1:2], col=my_cols, pch=16)
```



```
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
```



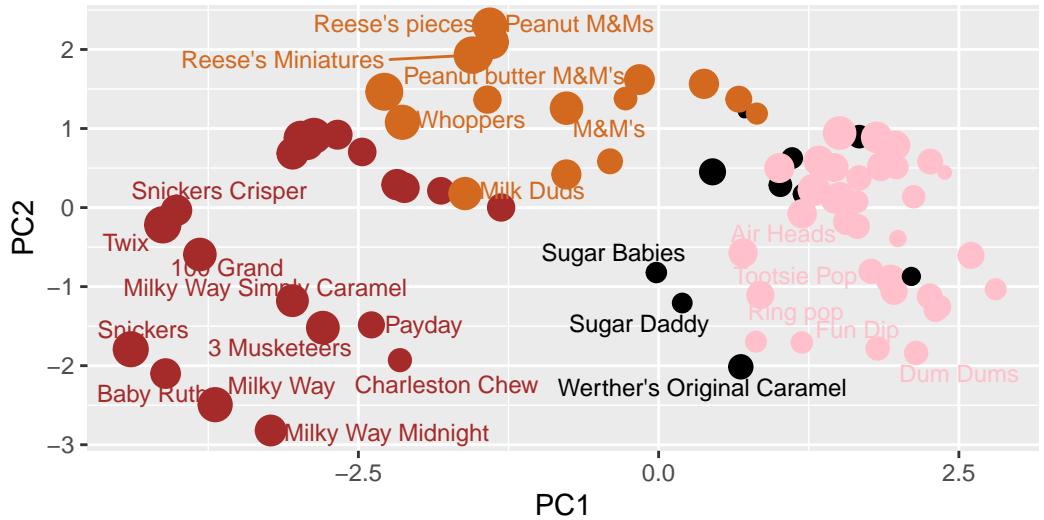
```
library(ggrepel)

p + geom_text_repel(size=3.3, col=my_cols, max.overlaps = 7) +
  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")
```

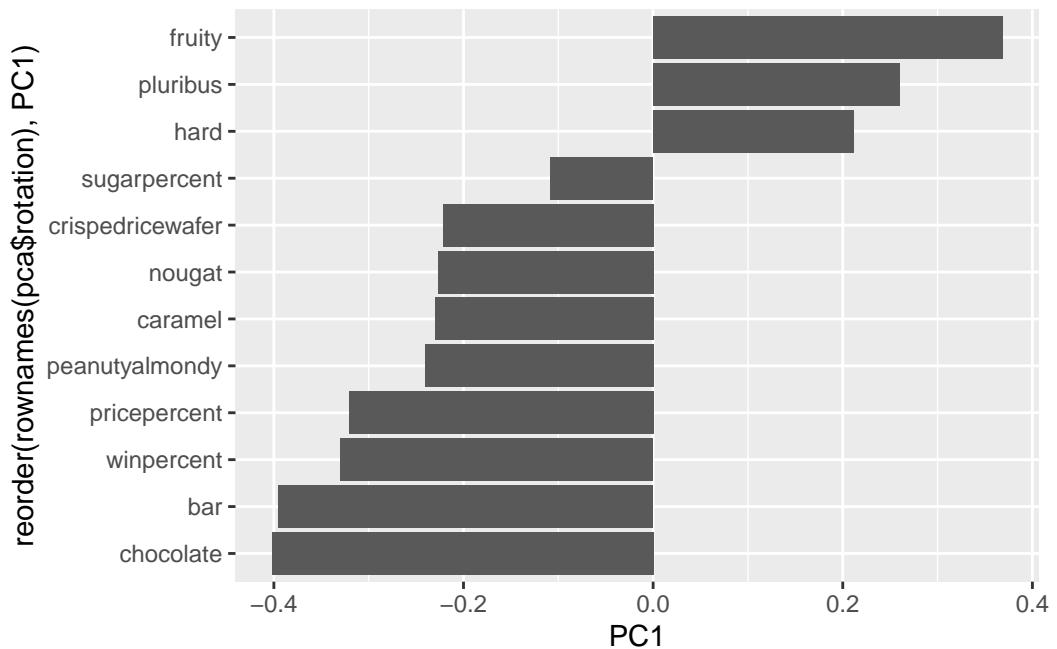
Warning: ggrepel: 59 unlabeled data points (too many overlaps). Consider increasing max.overlaps

Halloween Candy PCA Space

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



```
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 original variables picked up by PC1 in the positive direction are fruity, pluribus, and hard. These make sense to me as they are fruity candies. This relationship was highlighted previously in the correlation plot and the PCA score plot, where chocolate and fruity candy had a negative correlation.

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?

Based on exploration analysis, correlation findings, and PCA results, winning candies tend to be chocolate bars, often with caramel or peanut filling. The correlation analysis showed chocolate + bar as positively correlated with winpercent, while fruity is negatively correlated. The visualization of rankings and scatter plots also showed Reese's, Twix, and Kitkat, which share the characteristics of the “winning candy”. The PCA results further supported this by grouping bar + chocolate together. All in all, these analyses confirm that chocolate-bar candies, tend to be the most successful, especially at reasonable prices.