

class10miniproect

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

	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

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

```
nrow(candy)
```

```
[1] 85
```

We can use the `ncol()` function to find that there are 85 different types of candy being compared.

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

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

```
[1] 38
```

We can use the sum function with a row argument to find that there are 38 fruity types.

```
candy[as.logical(candy$chocolate),]
```

	chocolate	fruity	caramel	peanutyalmondy	nougat
100 Grand	1	0	1	0	0
3 Musketeers	1	0	0	0	1
Almond Joy	1	0	0	1	0
Baby Ruth	1	0	1	1	1
Charleston Chew	1	0	0	0	1
Hershey's Kisses	1	0	0	0	0
Hershey's Krackel	1	0	0	0	0
Hershey's Milk Chocolate	1	0	0	0	0
Hershey's Special Dark	1	0	0	0	0
Junior Mints	1	0	0	0	0
Kit Kat	1	0	0	0	0
Peanut butter M&M's	1	0	0	1	0
M&M's	1	0	0	0	0
Milk Duds	1	0	1	0	0
Milky Way	1	0	1	0	1
Milky Way Midnight	1	0	1	0	1
Milky Way Simply Caramel	1	0	1	0	0
Mounds	1	0	0	0	0
Mr Good Bar	1	0	0	1	0
Nestle Butterfinger	1	0	0	1	0
Nestle Crunch	1	0	0	0	0
Peanut M&Ms	1	0	0	1	0
Reese's Miniatures	1	0	0	1	0
Reese's Peanut Butter cup	1	0	0	1	0
Reese's pieces	1	0	0	1	0
Reese's stuffed with pieces	1	0	0	1	0
Rolo	1	0	1	0	0
Sixlets	1	0	0	0	0
Nestle Smarties	1	0	0	0	0
Snickers	1	0	1	1	1
Snickers Crisper	1	0	1	1	0

Tootsie Pop	1	1	0	0	0
Tootsie Roll Juniors	1	0	0	0	0
Tootsie Roll Midgies	1	0	0	0	0
Tootsie Roll Snack Bars	1	0	0	0	0
Twix	1	0	1	0	0
Whoppers	1	0	0	0	0
	crisped	rice	wafer	hard bar	pluribus sugarpercent
100 Grand		1	0	1	0.732
3 Musketeers		0	0	1	0.604
Almond Joy		0	0	1	0.465
Baby Ruth		0	0	1	0.604
Charleston Chew		0	0	1	0.604
Hershey's Kisses		0	0	0	1.127
Hershey's Krackel		1	0	1	0.430
Hershey's Milk Chocolate		0	0	1	0.430
Hershey's Special Dark		0	0	1	0.430
Junior Mints		0	0	0	1.197
Kit Kat		1	0	1	0.313
Peanut butter M&M's		0	0	0	1.825
M&M's		0	0	0	1.825
Milk Duds		0	0	0	1.302
Milky Way		0	0	1	0.604
Milky Way Midnight		0	0	1	0.732
Milky Way Simply Caramel		0	0	1	0.965
Mounds		0	0	1	0.313
Mr Good Bar		0	0	1	0.313
Nestle Butterfinger		0	0	1	0.604
Nestle Crunch		1	0	1	0.313
Peanut M&Ms		0	0	0	1.593
Reese's Miniatures		0	0	0	0.034
Reese's Peanut Butter cup		0	0	0	0.720
Reese's pieces		0	0	0	1.406
Reese's stuffed with pieces		0	0	0	0.988
Rolo		0	0	0	1.860
Sixlets		0	0	0	1.220
Nestle Smarties		0	0	0	1.267
Snickers		0	0	1	0.546
Snickers Crisper		1	0	1	0.604
Tootsie Pop		0	1	0	0.604
Tootsie Roll Juniors		0	0	0	0.313
Tootsie Roll Midgies		0	0	0	1.174
Tootsie Roll Snack Bars		0	0	1	0.465
Twix		1	0	1	0.546

Whoppers	1	0	0	1	0.872
	pricepercent	winpercent			
100 Grand	0.860	66.97173			
3 Musketeers	0.511	67.60294			
Almond Joy	0.767	50.34755			
Baby Ruth	0.767	56.91455			
Charleston Chew	0.511	38.97504			
Hershey's Kisses	0.093	55.37545			
Hershey's Krackel	0.918	62.28448			
Hershey's Milk Chocolate	0.918	56.49050			
Hershey's Special Dark	0.918	59.23612			
Junior Mints	0.511	57.21925			
Kit Kat	0.511	76.76860			
Peanut butter M&M's	0.651	71.46505			
M&M's	0.651	66.57458			
Milk Duds	0.511	55.06407			
Milky Way	0.651	73.09956			
Milky Way Midnight	0.441	60.80070			
Milky Way Simply Caramel	0.860	64.35334			
Mounds	0.860	47.82975			
Mr Good Bar	0.918	54.52645			
Nestle Butterfinger	0.767	70.73564			
Nestle Crunch	0.767	66.47068			
Peanut M&Ms	0.651	69.48379			
Reese's Miniatures	0.279	81.86626			
Reese's Peanut Butter cup	0.651	84.18029			
Reese's pieces	0.651	73.43499			
Reese's stuffed with pieces	0.651	72.88790			
Rolo	0.860	65.71629			
Sixlets	0.081	34.72200			
Nestle Smarties	0.976	37.88719			
Snickers	0.651	76.67378			
Snickers Crisper	0.651	59.52925			
Tootsie Pop	0.325	48.98265			
Tootsie Roll Juniors	0.511	43.06890			
Tootsie Roll Midgies	0.011	45.73675			
Tootsie Roll Snack Bars	0.325	49.65350			
Twix	0.906	81.64291			
Whoppers	0.848	49.52411			

Q3. What is your favorite candy in the dataset and what is it's winpercent value?

```
candy["Nestle Butterfinger", ]$winpercent
```

```
[1] 70.73564
```

My favourite candy is butterfingers, which is very popular with a `winpercent` of 70.74%. >Q4. What is the `winpercent` value for “Kit Kat”?

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

```
[1] 76.7686
```

Kit Kats have a `winpercent` of 76.77. Not sure why given that they’re vile bars of cardboard. >Q5. What is the `winpercent` value for “Tootsie Roll Snack Bars”?

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

```
[1] 49.6535
```

This candy has a 49.65% `winpercent`, and isn’t very popular.

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

```
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	
peanutyalmondy	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	

Yes. The `winpercent` column is on a very different scale.

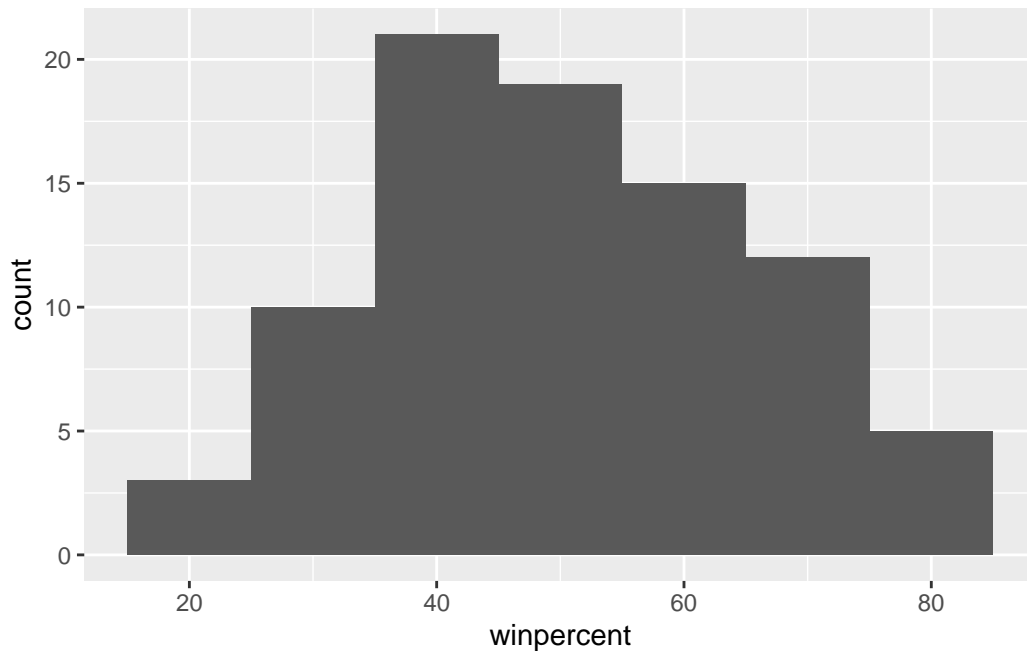
Q7. What do you think a zero and one represent for the `candy$chocolate` column?

The 1s and 0s show whether or not the candy include chocolate.

Q8. Plot a histogram of `winpercent` values.

```
library(ggplot2)

ggplot(candy) +
  aes(winpercent) +
  geom_histogram(binwidth=10)
```



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%?

The center of the distribution is above 50%. >Q11. On average is chocolate candy higher or lower ranked than fruit candy?

```
choc.inds <- as.logical(candy$chocolate)
choc.win <- candy[choc.inds,]$winpercent
mean(choc.win)
```

```
[1] 60.92153
```

We can see that chocolate wins about 61% of the time.

```
fruit.inds <- as.logical(candy$fruit)
fruit.win <- candy[fruit.inds,]$winpercent
mean(fruit.win)
```

```
[1] 44.11974
```

and that fruit wins about 44% of the time.

Q12. Is this difference statistically significant?

```
t.test(choc.win, fruit.win)
```

Welch Two Sample t-test

```
data: choc.win and fruit.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
```

We can use the `ttest()` function to show that this is a statistically significant difference.

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

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

	crisped	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

We can list the set by `winpercent` and then list the bottom five to see the least liked candies are Nik L Nips, Boston Baked Beans, Chiclets, Super Bubbles, and Jawbusters. >Q14. What are the top 5 all time favorite candy types out of this set?

```
tail(candy[order(candy$winpercent),], n=5)
```

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

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

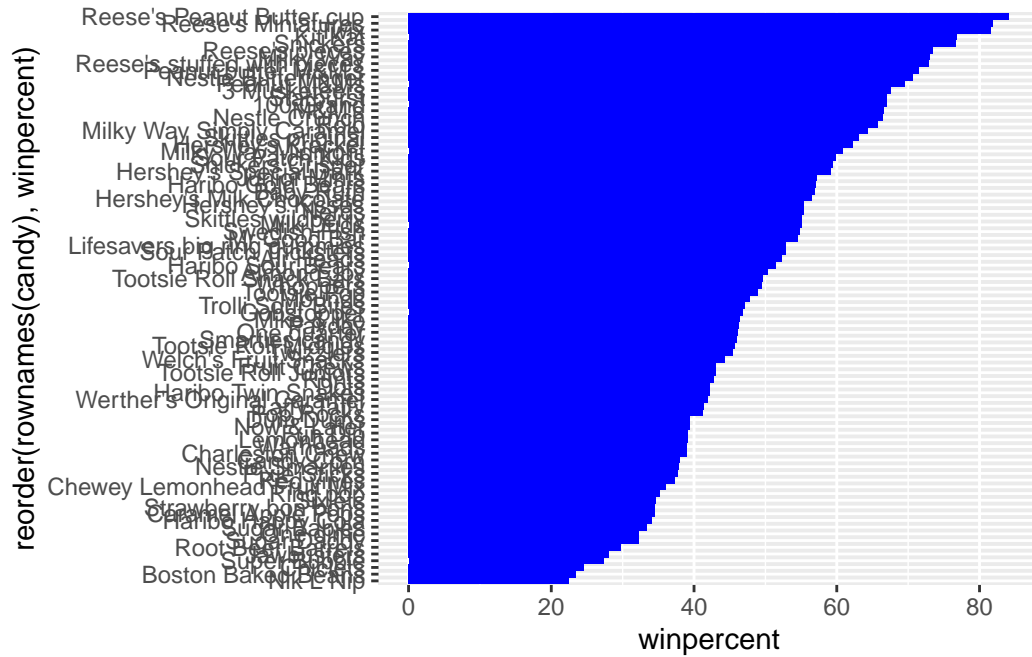
	price	percent	winpercent
Snickers	0.651	76.67378	
Kit Kat	0.511	76.76860	
Twix	0.906	81.64291	
Reese's Miniatures	0.279	81.86626	
Reese's Peanut Butter cup	0.651	84.18029	

You can use the `tail` function to see that Snickers, Kit Kats (idk why), Twix bars, Reese's minis, and Reese's PB cups are the most popular.

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

```
library(ggplot2)

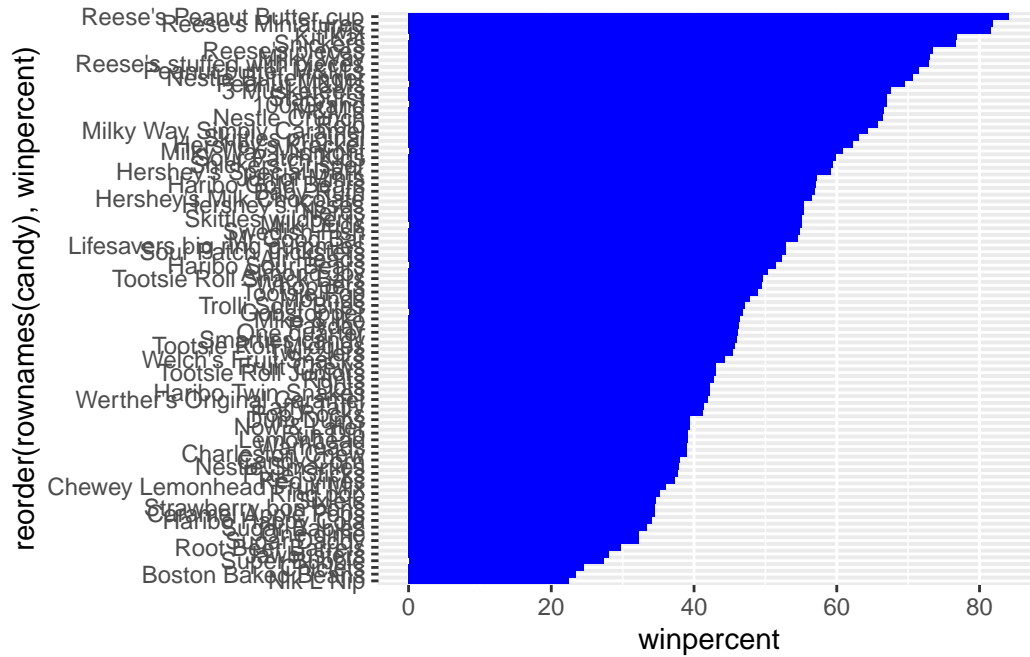
ggplot(candy) +
  aes(winpercent, reorder(rownames(candy), winpercent)) +
  geom_col(fill="blue")
```



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

```
library(ggplot2)

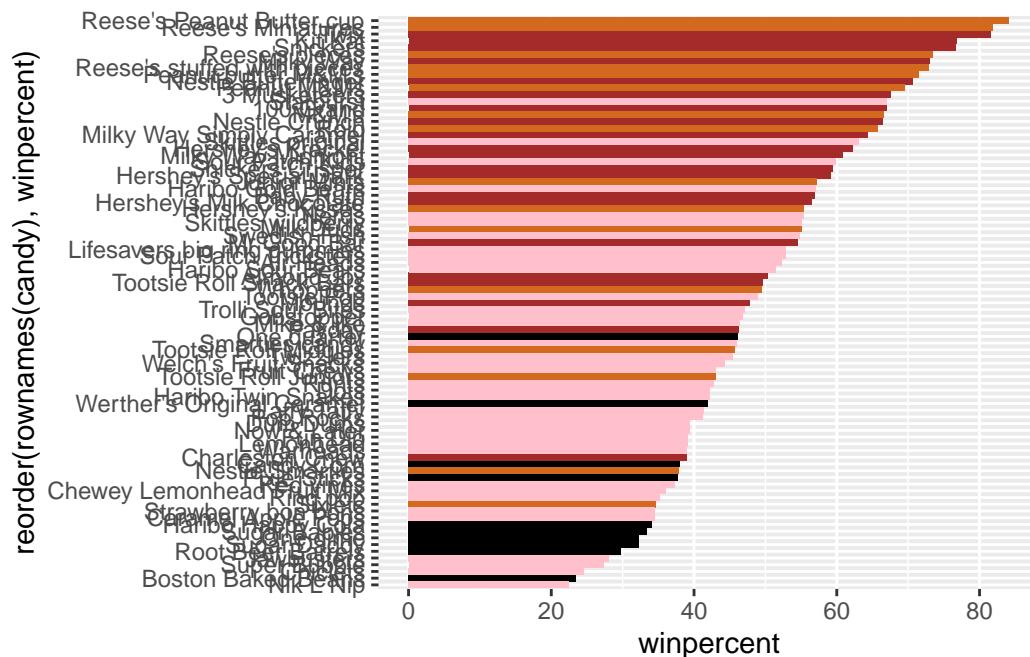
ggplot(candy) +
  aes(winpercent, reorder(rownames(candy), winpercent)) +
  geom_col(fill="blue")
```



```
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"

library(ggplot2)

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



Q17. What is the worst ranked chocolate candy?

We can see from the helpfully colour-coded graph that Sixlets are the lowest ranking chocolate candy.

Q18. What is the best ranked fruity candy?

We can see from the helpfully colour-coded graph that Starburst are the highest ranking fruity candy.

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?

```
library(ggrepel)

ggplot(candy) +
  aes(winpercent, pricepercent, label=rownames(candy)) +
  geom_point(col=my_cols) +
  geom_text_repel(col=my_cols, size=1.8, max.overlaps = 50)
```



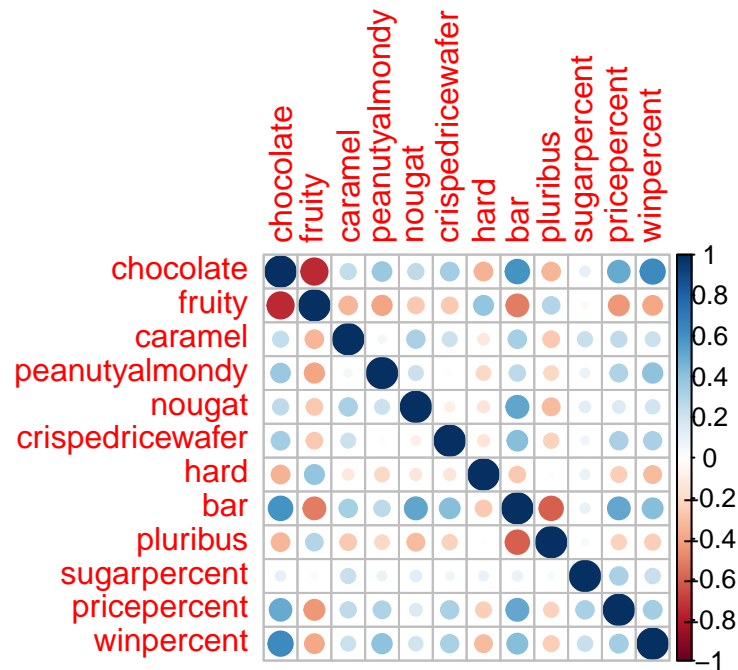
We can generate a plot that compares `pricepercent` and `winpercent` and see that Reese's miniatures and Starburst are the most popular for the price point. >Q20. What are the top 5 most expensive candy types in the dataset and of these which is the least popular?

The most expensive candies are Nik L Dips, Ring Pops, Nestle Smarties, Hershey's Krackel, and Hershey's Milk Chocolates. The least popular are Nik L Dips.

```
library(corrplot)
```

```
corrplot 0.92 loaded
```

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



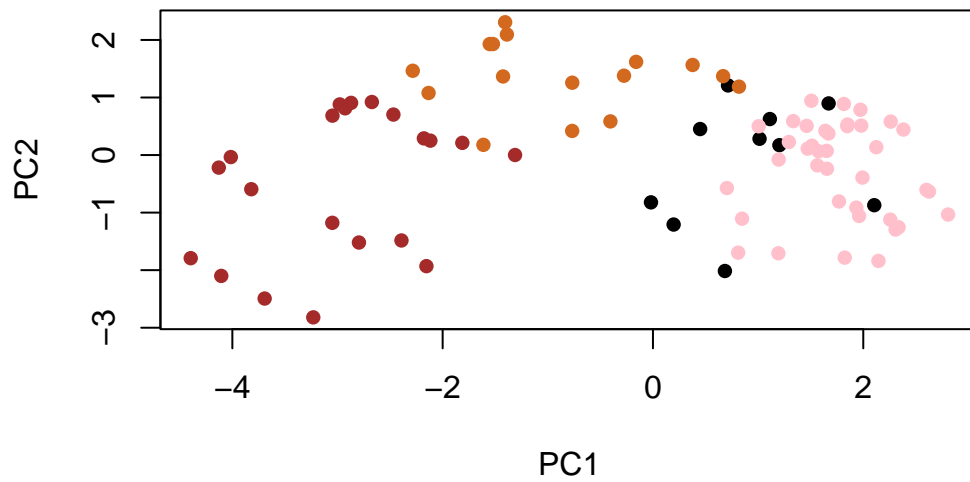
Q22. Examining this plot what two variables are anti-correlated (i.e. have minus values)?

Fruity chocolates are highly anti-correlated.

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

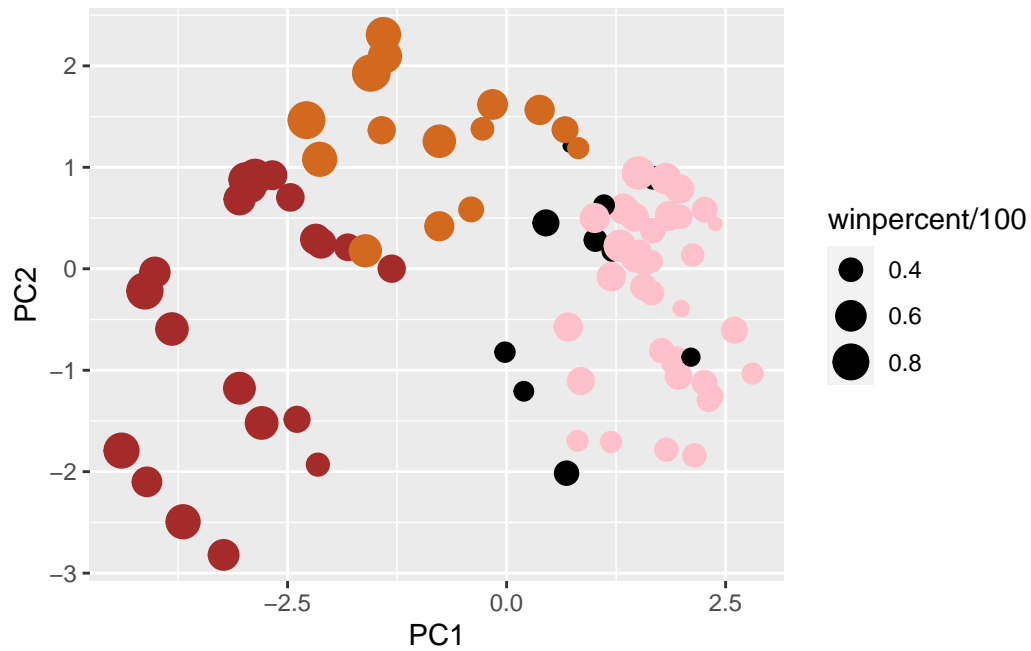
The most positively correlated are the same variables, but also chocolate and winpercent.

```
pca <- prcomp(candy, scale=TRUE)#First we run a PCA of the data
plot(pca$x[,1:2], col=my_cols, pch=16) #then we give it some colors and make the dot size
```



```
my_data <- cbind(candy, pca$x[,1:3])
p <- ggplot(my_data) + #plot the data nicely in ggplot
  aes(x=PC1, y=PC2,
      size=winpercent/100,
      text=rownames(my_data),
      label=rownames(my_data)) +
  geom_point(col=my_cols)
```

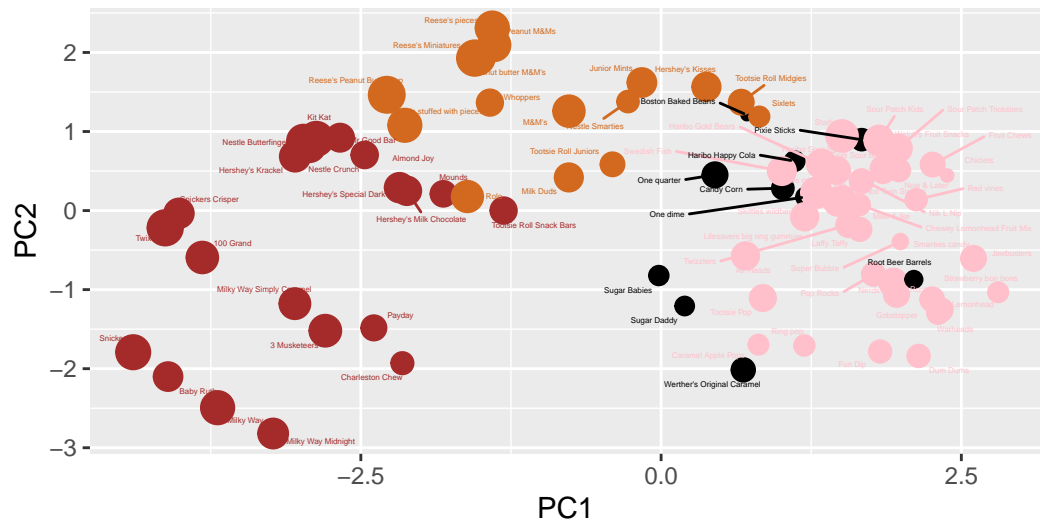
p



```
library(ggrepel)

p + geom_text_repel(size=1, col=my_cols, max.overlaps = 50) + #remember to change the size
  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")
```

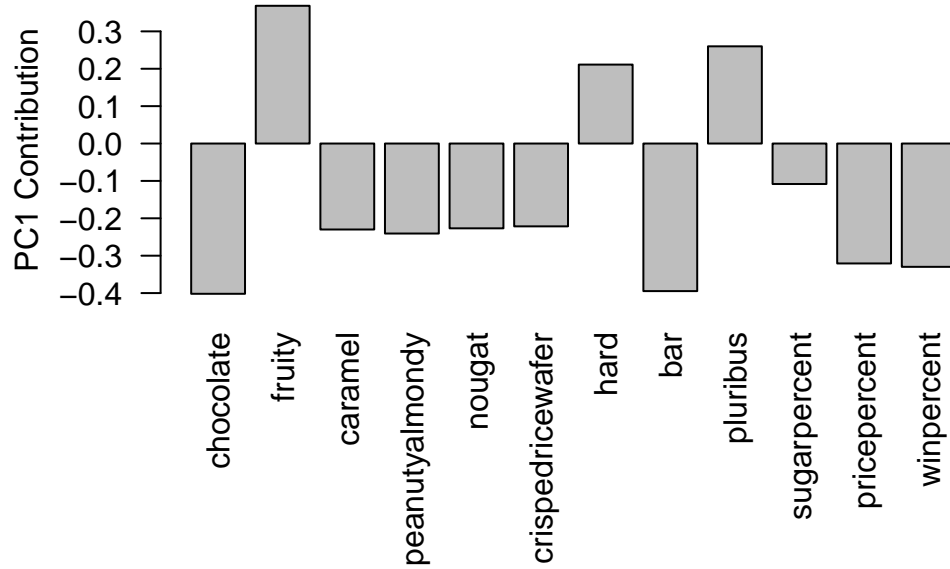

Colored by type: chocolate bar (dark brown), chocolate other (light brown),



Data from 538

```
#library(plotly) this would load up the ability to mouse over the data to see what it is,
#ggplotly(p)
```

```
par(mar=c(8,4,2,2))
barplot(pca$rotation[,1], las=2, ylab="PC1 Contribution")
```



Q24. What original variables are picked up strongly by PC1 in the positive direction? Do these make sense to you?

Fruity, Hard, and Pluribus are picked up in the positive direction because they're highly associated, which changes the way that it's plotted in PCA. This makes sense because many fruity candies come in an assorted package.