

Halloween Mini Project

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Importing Candy data

```
candy_url <- url("https://raw.githubusercontent.com/fivethirtyeight/data/master/candy-power")
candy <- read.csv(candy_url, header = TRUE, 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

Q1: how many different candies are in this dataset?

85

Q2: how many fruity candy types are in this data set?

```
sum(candy[,2])
```

```
[1] 38
```

What is my favorite candy?

```
candy["Root Beer Barrels", ]$winpercent
```

```
[1] 29.70369
```

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

```
[1] 76.7686
```

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

```
[1] 49.6535
```

using library `skimr`

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

Table 1: Data summary

Name	candy
Number of rows	85

Table 1: Data summary

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	

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

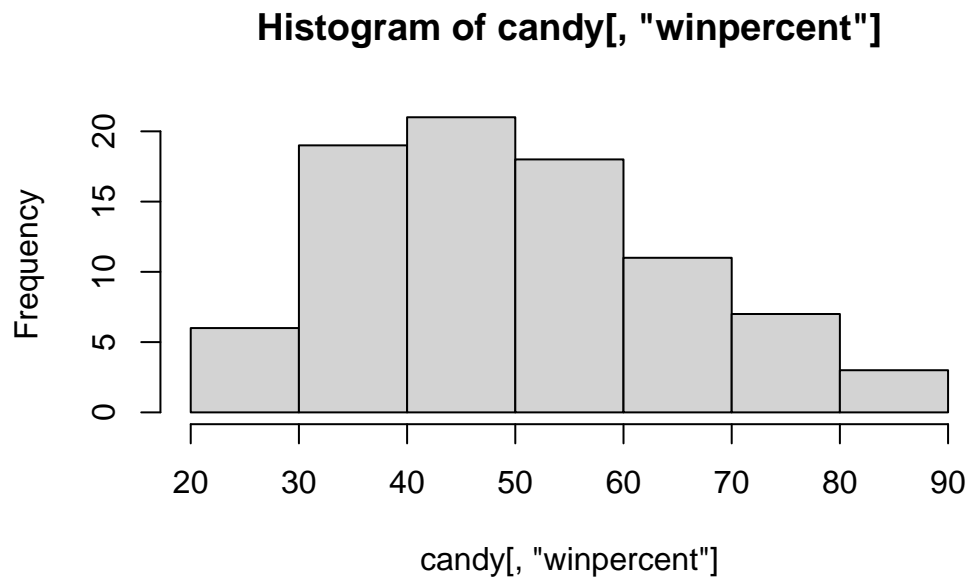
the winpercent is a different scale because the mean is 50 vs ~0-1

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

The zero vs one represents the identity of a candy being or containing chocolate

Q8: Plot a histogram of winpercent values

```
hist(candy[, "winpercent"])
```



Q9: is the distribution of winpercent values symetrical?

No

Q10 is the center of the distributino above or below 50%?

Below 50%

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

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

```
[1] 60.92153
```

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

```
[1] 44.11974
```

Q12: is the difference statically 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 data is significantly different, that chocolate, on average is ranked higher than fruity

Q13: What are the five least likely candy types in this set?

```
ordered_by_winpercent <- candy[order(candy[, "winpercent"]),]  
head(ordered_by_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

```
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

	crispedrice	wafer	hard bar	pluribus	sugarpercent	pricepercent	
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

```
# <- candy %>% group_by(winpercent)
#by_winpercent %>% arrange(desc(winpercent))
```

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

```
ordered_by_winpercent <- candy[order(candy[, "winpercent"], decreasing = TRUE),]
head(ordered_by_winpercent, n=5)
```

	chocolate	fruity	caramel	peanut	almond	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	wafer	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

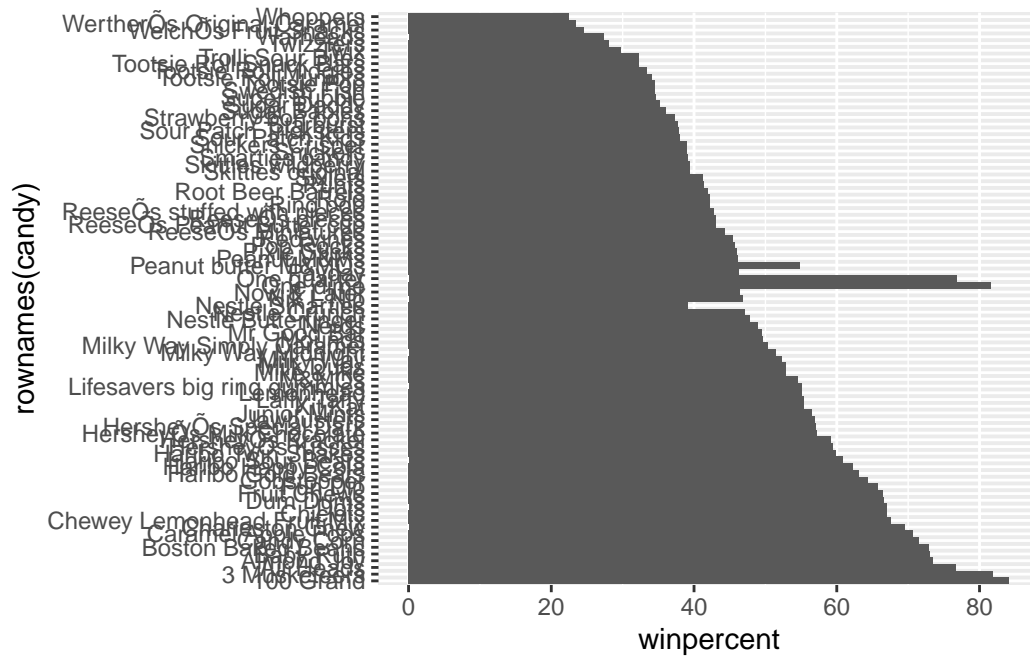
Snickers	0	0	1	0	0.546
	pricepercent	winpercent			
Reese's Peanut Butter cup	0.651	84.18029			
Reese's Miniatures	0.279	81.86626			
Twix	0.906	81.64291			
Kit Kat	0.511	76.76860			
Snickers	0.651	76.67378			

```
library("dplyr")
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		
	crisped	rice	wafer	hard	bar	pluribus	sugar	pricepercent
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							

Q15/16: Make a barplot of candy rankings based on winpercent values

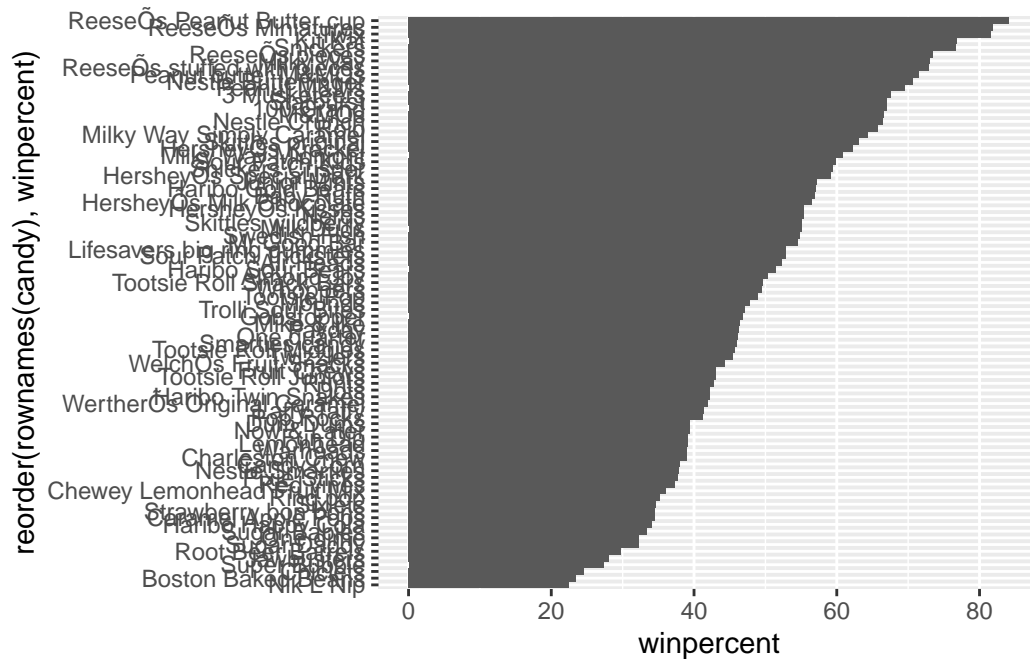
```
library("ggplot2")
ggplot(ordered_by_winpercent)+
  aes(winpercent,rownames(candy))+
  geom_col()
```



#You can do saving certain

```
#ggsave("mybarplot.png",height=1200,width=800,units=c("px"),dpi=300)
```

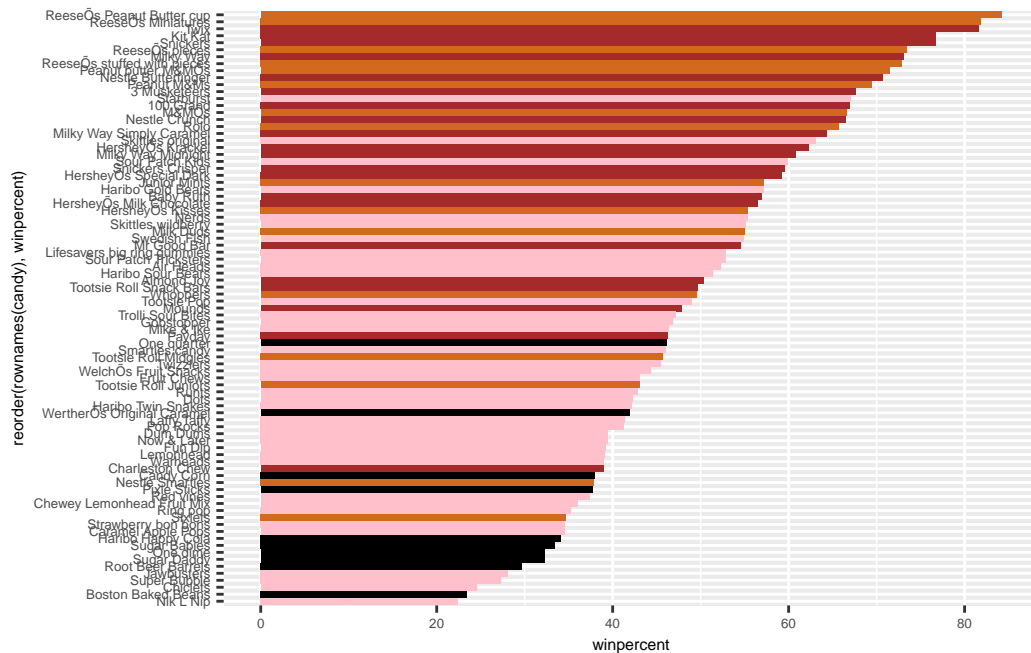
```
library("ggplot2")
ggplot(candy)+
  aes(winpercent,reorder(rownames(candy),winpercent))+
  geom_col()
```



Q17: Coloring the bar plot via type

```
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)+theme(text=element_text(size=6))
```



```
ggsave("mybarplot_withcol.png")
```

Saving 5.5 x 3.5 in image

Worst ranking chocolate is sixlets

Q18: What is the best ranked fruity candy

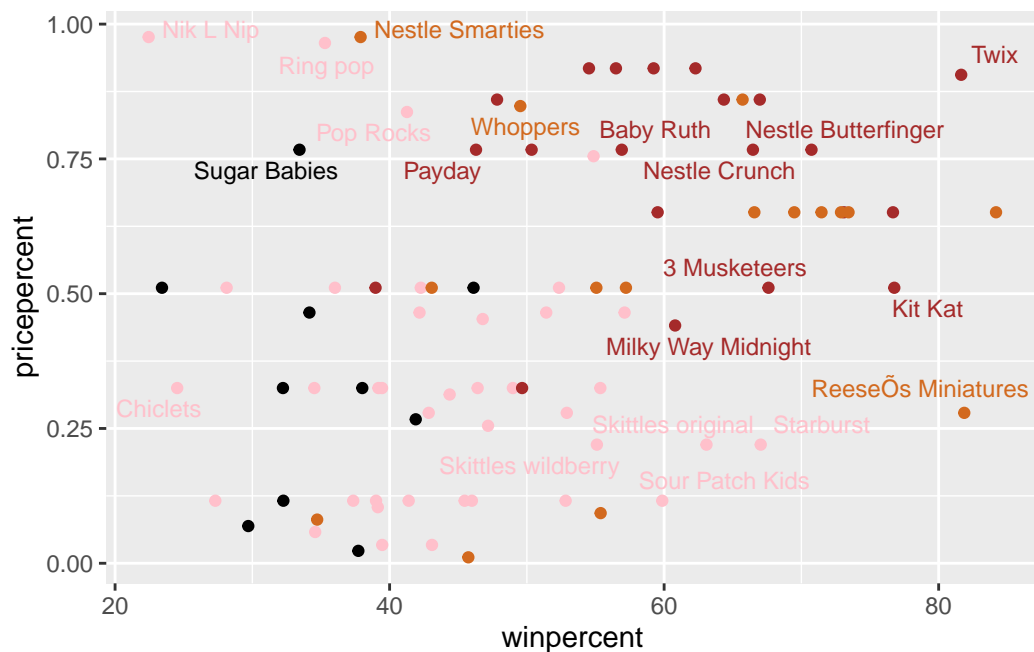
Star bursts

Looking at pricepercent

```
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 is the highest rank and the lowest price

Reese's miniatures

Q20: What are the top 5 most expensive candy types

```
ordered_by_pricepercent <- candy[order(candy[,"pricepercent"],decreasing = TRUE),]
head(ordered_by_pricepercent, n=5)
```

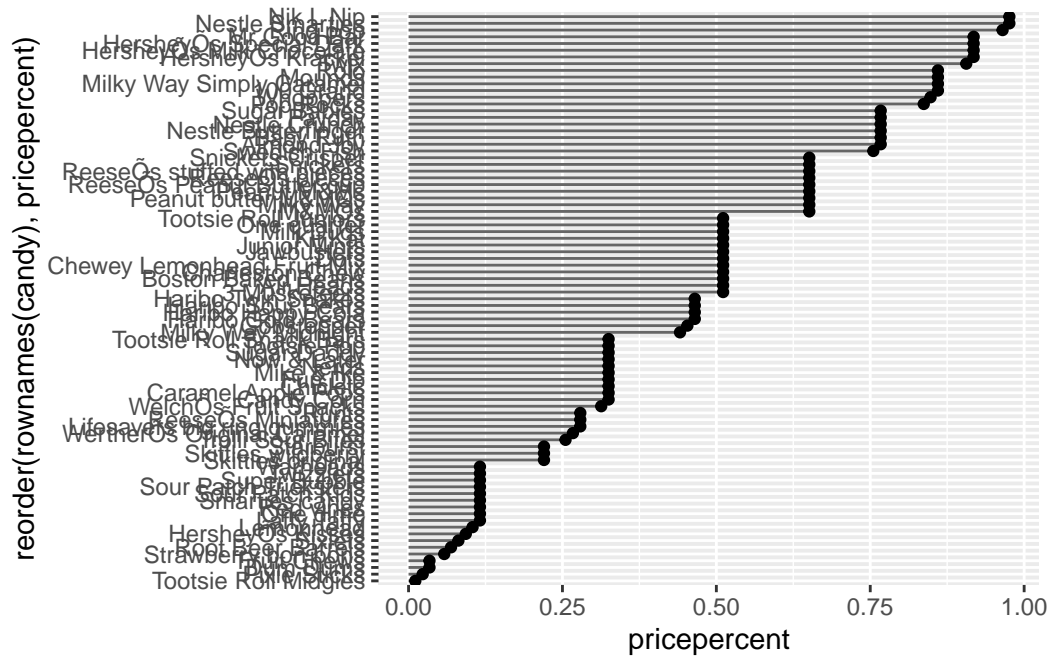
	chocolate	fruity	caramel	peanuty	almondy	nougat
Nik L Nip	0	1	0		0	0
Nestle Smarties	1	0	0		0	0
Ring pop	0	1	0		0	0
Hershey's Krackel	1	0	0		0	0
Hershey's Milk Chocolate	1	0	0		0	0

	crisped	ricewafer	hard	bar	pluribus	sugarpercent
Nik L Nip	0	0	0		1	0.197
Nestle Smarties	0	0	0		1	0.267
Ring pop	0	1	0		0	0.732
Hershey's Krackel	1	0	1		0	0.430
Hershey's Milk Chocolate	0	0	1		0	0.430

	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

Q21: Make a barplot again with `geom_col()` this time using `pricepercent`

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

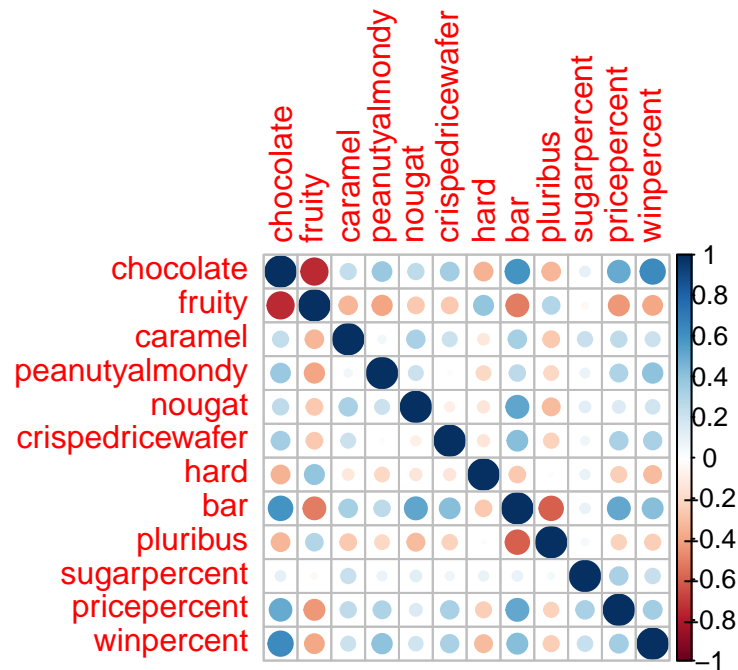


Exploring correlation structure

```
library(corrplot)
```

corrplot 0.92 loaded

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



Q22: Examining this plot what two variables are anti-correlated?

Fruity with chocolate and pluribus with bar

which makes sense, very few candies are fruity and chocolate-y and very few candies are in a bar form with multiple bars

Q23:SWhat two variables are most positively correlated?

Chocolate with winpercent, and chocolate with Bar

Principal component analysis

Note: we need `scale=TRUE` to be entered because we want to scale the winpercent which is abnormally scaled wthih the rest of the data.

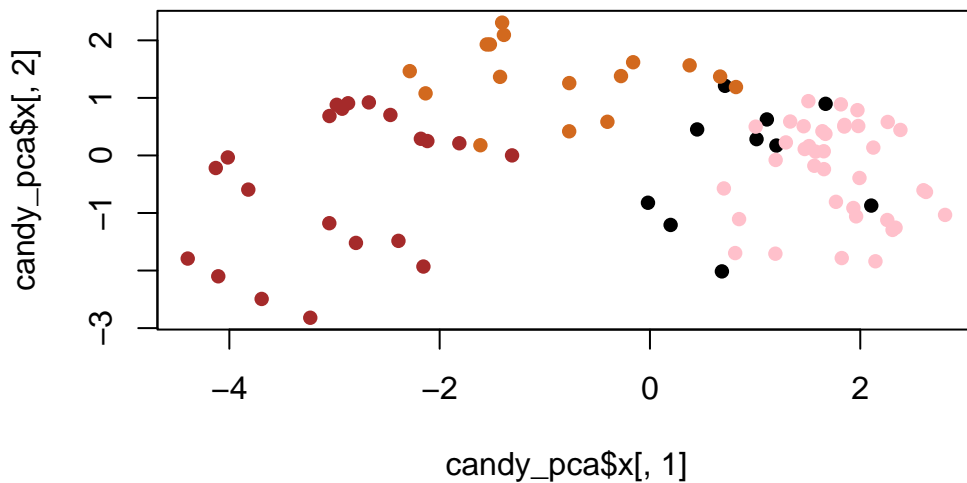
```
candy_pca <- prcomp(candy,scale=TRUE)
summary(candy_pca)
```


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(candy_pca$x[,1],candy_pca$x[,2],col=my_cols,pch=16)
```



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

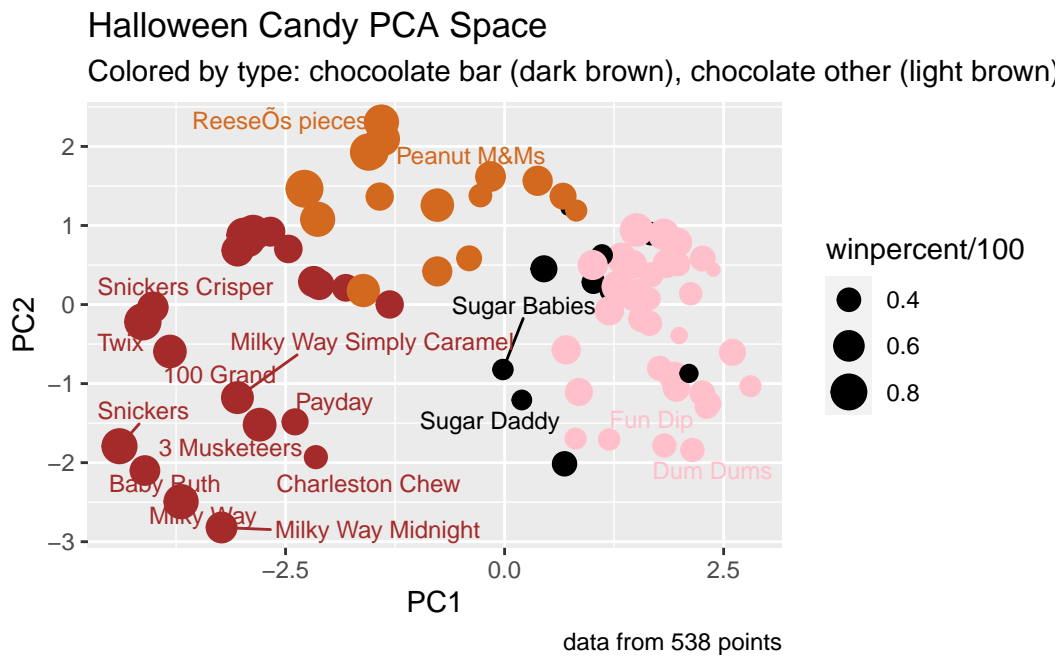
plot_1 <- ggplot(my_data)+
  aes(x=PC1, y=PC2,
      size=winpercent/100,
      text=rownames(my_data),
      label=rownames(my_data))+
  geom_point(col=my_cols)+
```

```

    geom_text_repel(size=3.3,col=my_cols,max.overlaps = 7)+
  labs(title="Halloween Candy PCA Space",
       subtitle="Colored by type: chocoolate bar (dark brown), chocolate other (light brown)",
       caption="data from 538 points")
plot_1

```

Warning: ggrepel: 68 unlabeled data points (too many overlaps). Consider increasing max.overlaps



```

#library(plotly)
#ggplotly(plot_1)

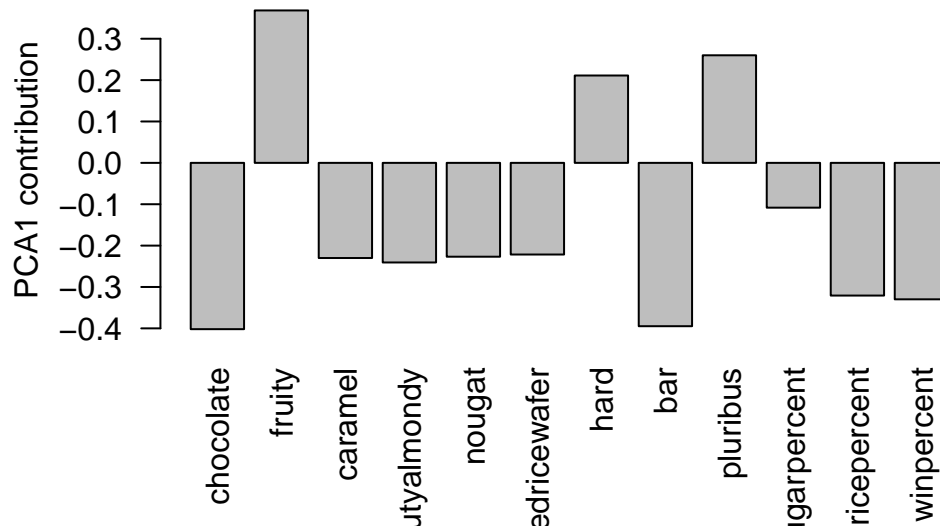
```

Q24: What original variables are picked up strongly by PC1 in the positive direction?

```

barplot(candy_pca$rotation[,1], las=2,ylab="PCA1 contribution")

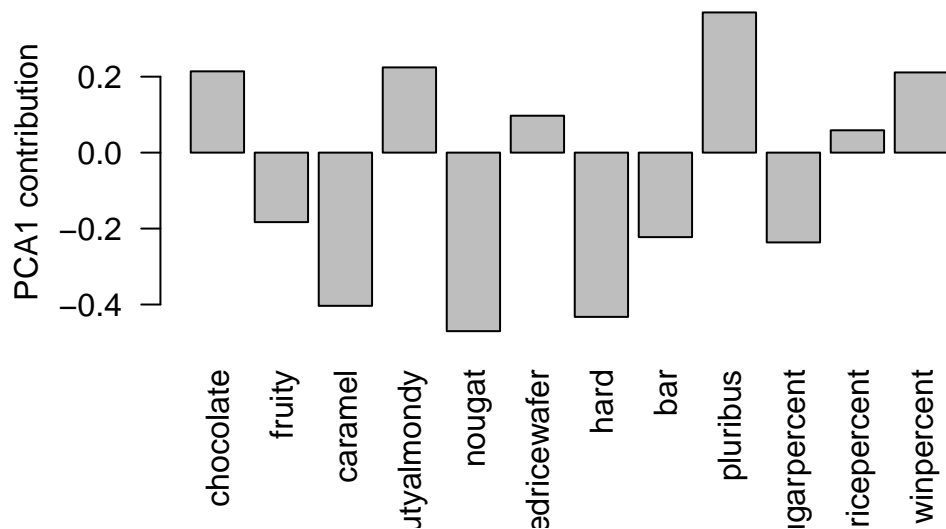
```



This means that fruity, hard, and multiple pieces of candy coming in a bag or a box are all tightly correlated with each other. Similarly chocolate, caramel, peanuts & almonds, nougat, crispy, bar, high sugar, high price, and high win percent are all correlated with each other.

What about PC2?

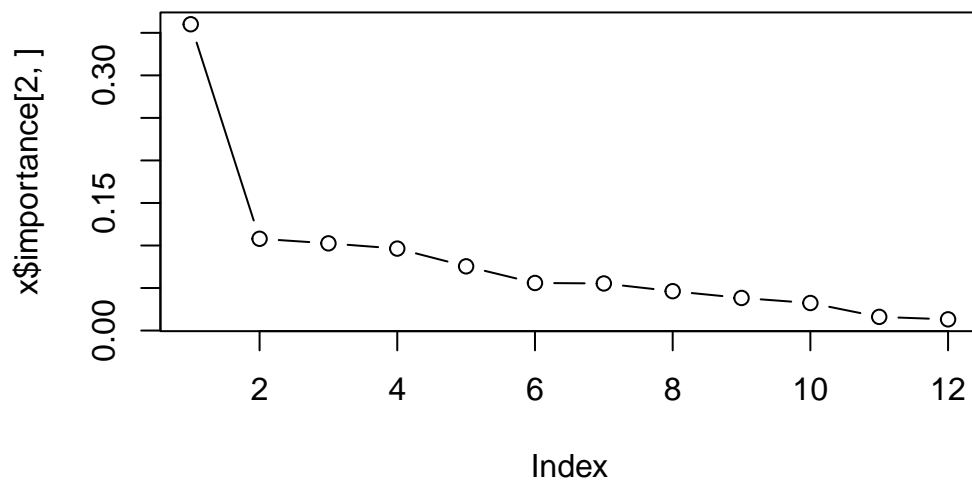
```
barplot(candy_pca$rotation[,2], las=2, ylab="PCA1 contribution")
```



PC2 shows us that caramel/nougat and hard are the opposite of nutty and multiple things in a single bag.

To find out what the intrinsic dimensionality of a data set is you can plot the `candy_pca` summary of variance to determine the inflection point

```
x <- summary(candy_pca)
plot(x$importance[2,], type="b")
```



This shows us that two dimensions is perfectly sufficient for data separation.