

Class 09: Halloween Mini Project

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##Halloween Mini Project

Today is Halloween and we will apply lots of the analysis methods and R graphics approaches to find out all about typical Halloween candy.

Importing candy data

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

	chocolate	fruity	caramel	peanut	almond	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	sugar	percent	price	percent	win	percent
100 Grand	0	1	0		0.732		0.860	66.97	173
3 Musketeers	0	1	0		0.604		0.511	67.60	294
One dime	0	0	0		0.011		0.116	32.26	109
One quarter	0	0	0		0.011		0.511	46.11	650
Air Heads	0	0	0		0.906		0.511	52.34	146
Almond Joy	0	1	0		0.465		0.767	50.34	755

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 in the dataset and what is it's winpercent value?

```
#My favorite candy in the dataset is Junior Mints  
candy["Junior Mints", ]$winpercent
```

[1] 57.21925

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

`skimr::skim()` function gives you a quick overview of a given dataset

```
#install skimr package install.packages("skimr")  
library("skimr")  
skim(candy)
```

Table 1: Data summary

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

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 dataset?

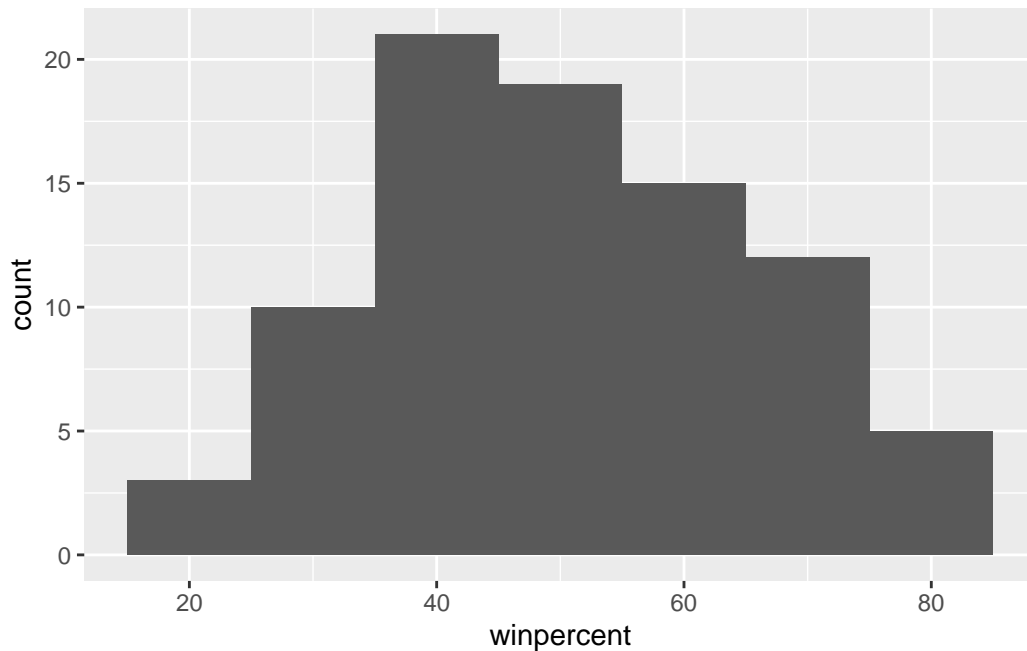
Yes, the winpercent appears to be on a different scale.

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

Zero represents that the candy is not considered a chocolate and one represents it is.

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?

The distribution of winpercent values are not symmetrical past on the histogram.

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

The center of distribution is below 50%.

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

```
candy_chocolate <- candy$winpercent[as.logical(candy$chocolate)]
candy_fruity <- candy$winpercent[as.logical(candy$fruity)]

mean(candy_chocolate) > mean(candy_fruity)
```

[1] TRUE

On average, chocolate candy is higher ranked than fruit candy.

Q12. Is this difference statistically significant?

```
t.test(candy_chocolate, candy_fruity)
```

Welch Two Sample t-test

```
data: candy_chocolate and 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
```

The difference is statistically significant because the p-value is >0.05 .

Overall Candy Rankings

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

```
#using tidyverse
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	peanutyalmondy	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	ricewafer	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					

```
#by ascending order
```

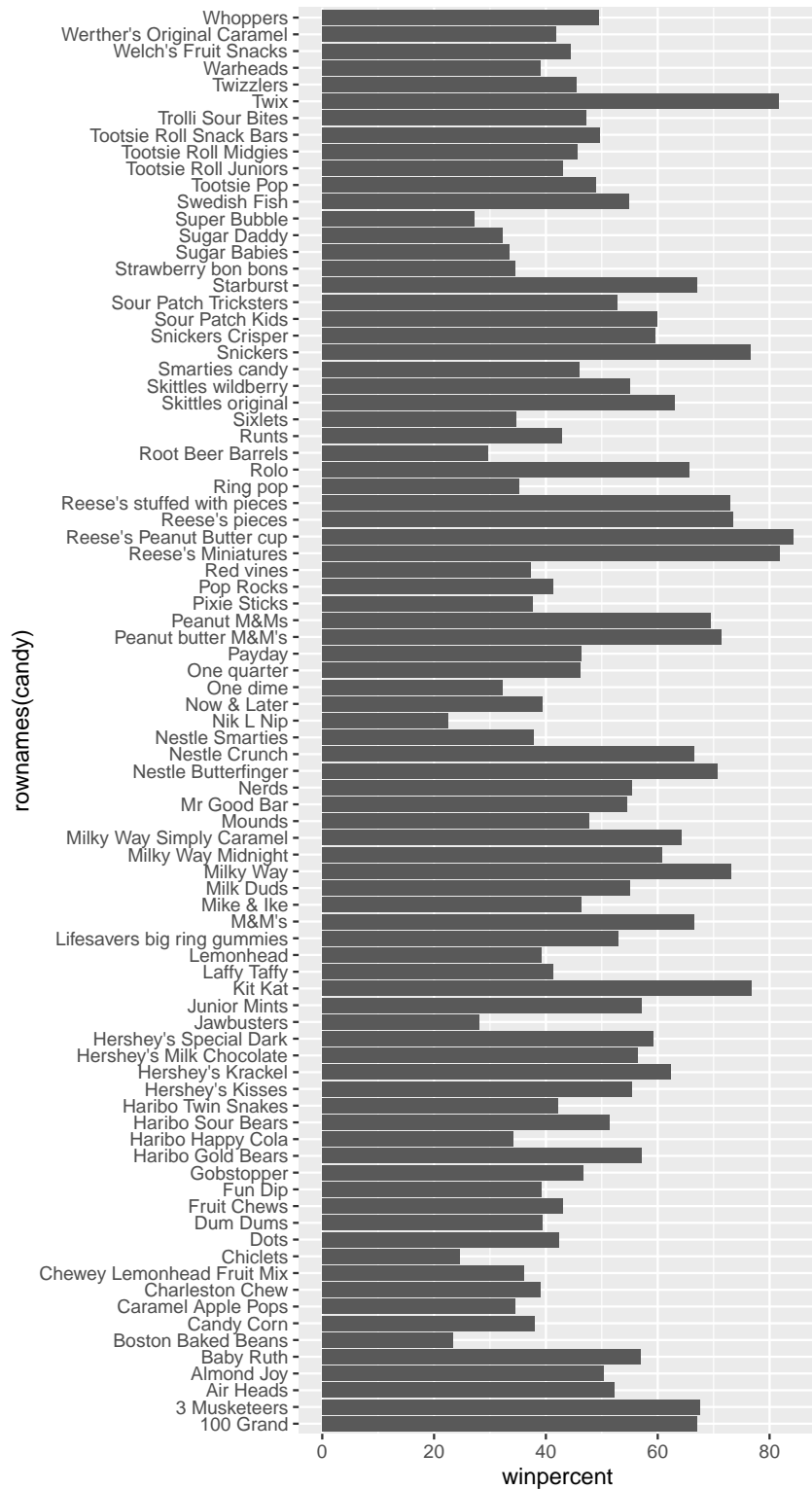
Q14. What are the top 5 all time favorite candy types out of this set?

```
candy %>%
  arrange(winpercent) %>%
  tail(5)
```

	chocolate	fruity	caramel	peanut	almondy	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	ricewafer	hard bar	pluribus	sugarpercent	
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
	pricepercent	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				

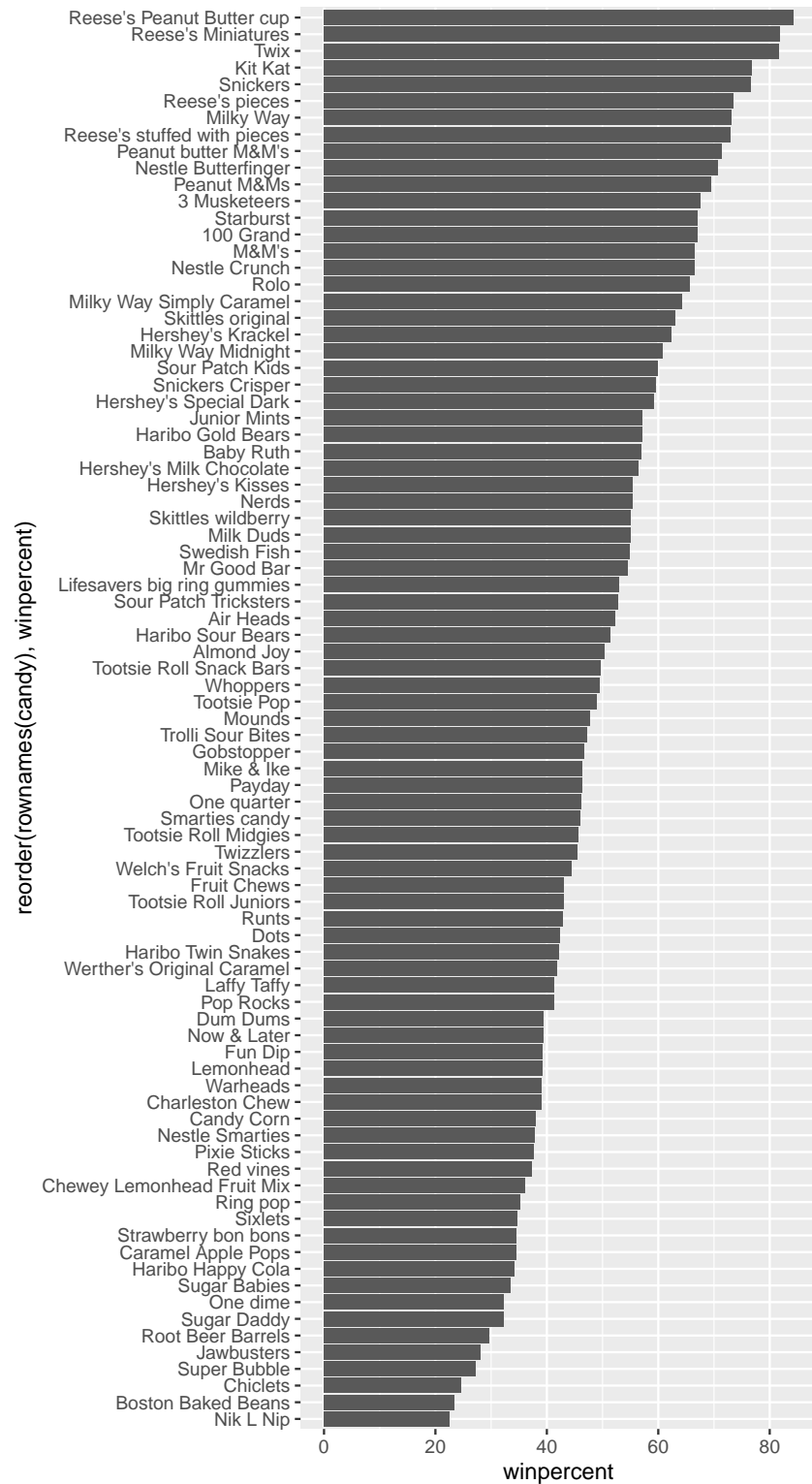
Q15. Make a first barplot of candy ranking based on winpercent values.

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



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

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

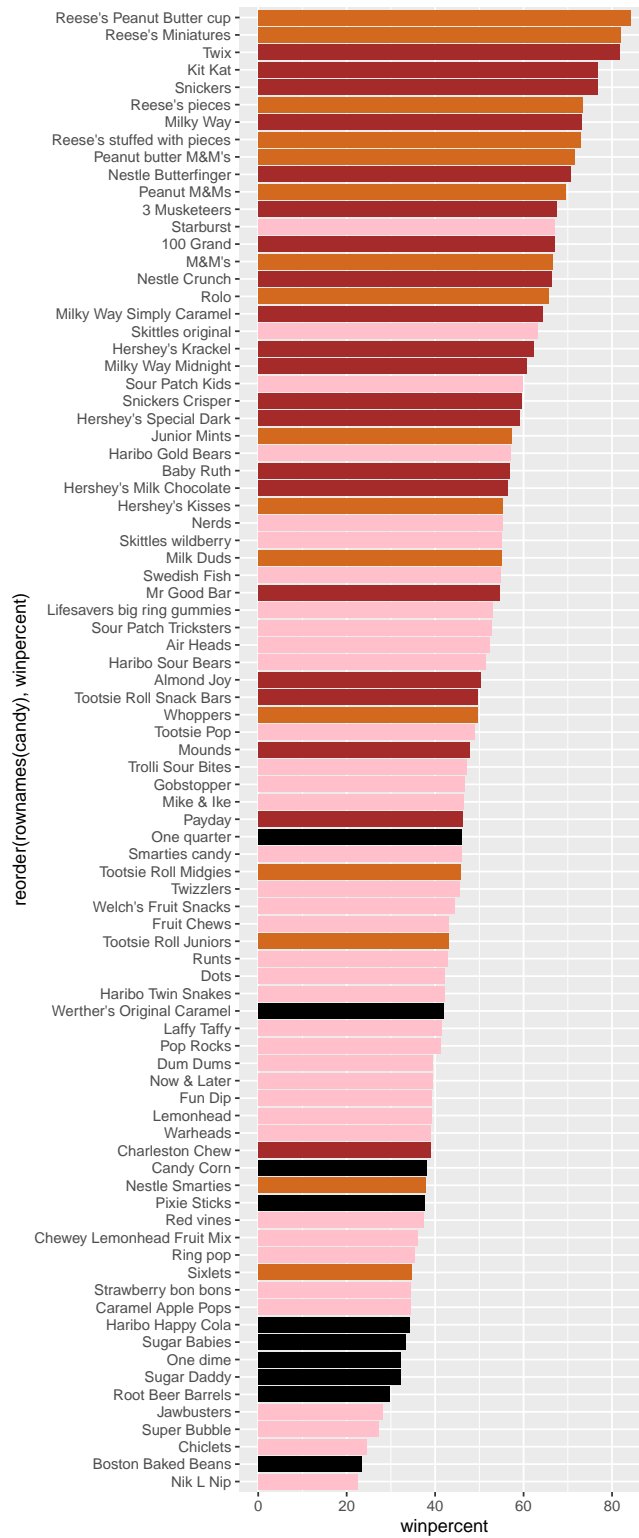


Time to add some useful color :)

Create color vectors for each candy type. `fill=my_cols` for `geom_col()`

```
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?

Sixlets is the worst ranked chocolate candy.

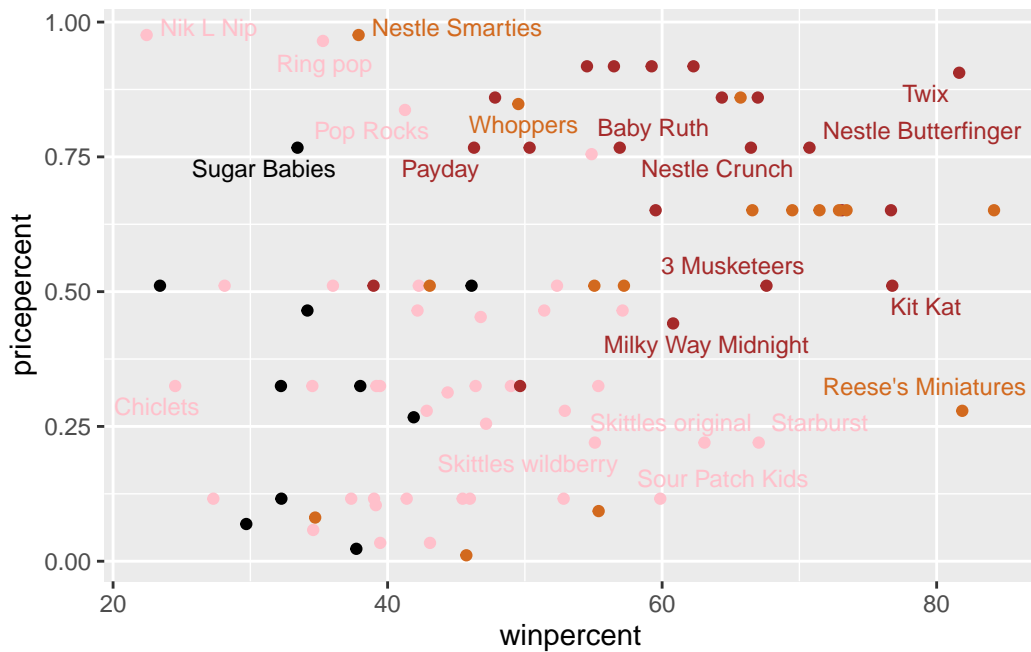
Q18. What is the best ranked fruity candy?

Starburst is the best ranked fruity candy.

Taking a look at pricepercent

```
#install ggrepel package `install.packages("ggrepel")`  
library(ggrepel)  
  
# How about a plot of price vs win  
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 Miniatures

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

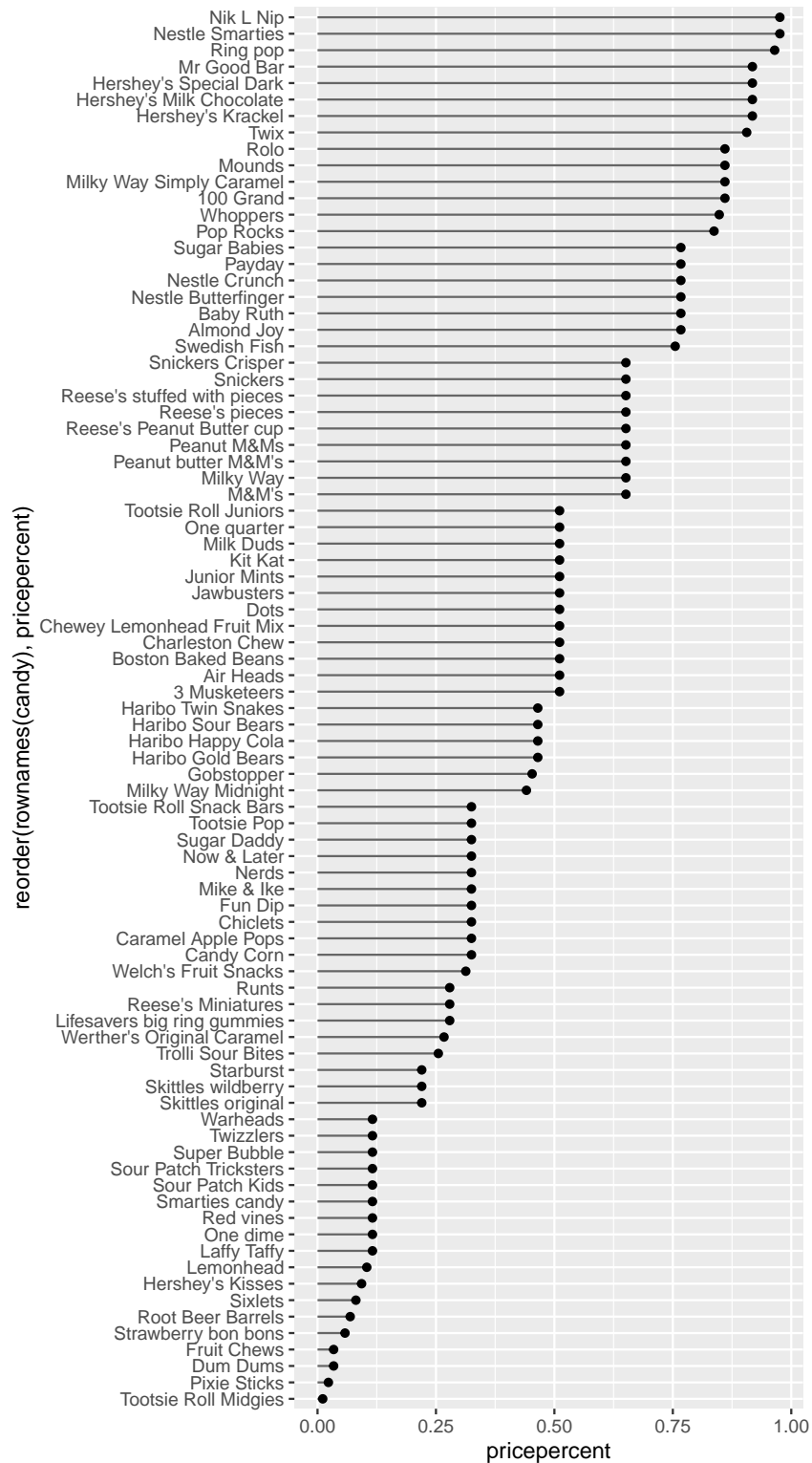
```
ord <- order(candy$pricepercent, decreasing = TRUE)
head( candy[ord,c(11,12)], n=5 )
```

	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

```
#11,12 are the position for pricepercent and winpercent
```

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()`.

```
# Make a lollipop chart of pricepercent
ggplot(candy) +
  aes(pricepercent, reorder(rownames(candy), pricepercent)) +
  geom_segment(aes(yend = reorder(rownames(candy), pricepercent),
                  xend = 0), col="gray40") +
  geom_point()
```

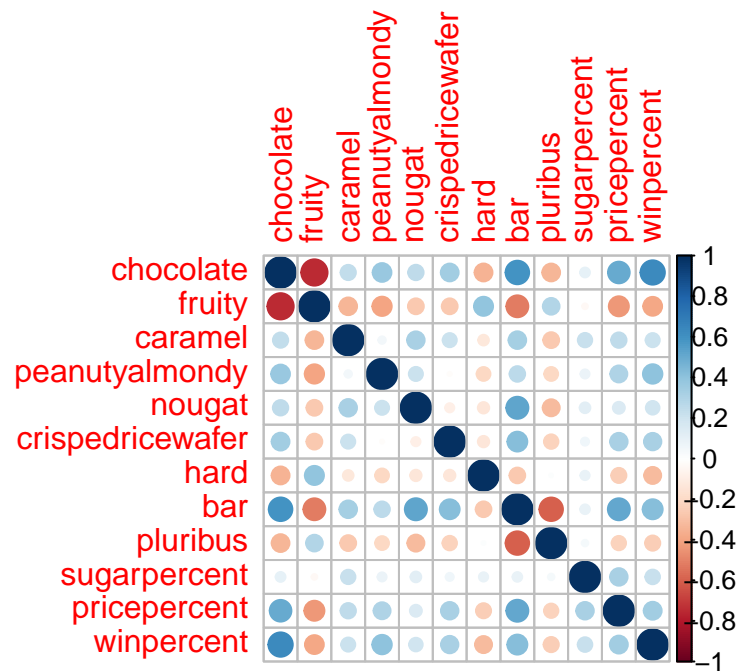


##Exploring the correlation structure

```
#install corrplot package install.packages("corrplot")  
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)?

Fruit and Chocolate variables are anti-correlated.

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

Chocolate and winpercent are most positively correlated.

##Principal Component Analysis

Let's apply PCA using the prcomp() function to our candy dataset remembering to set the scale=TRUE argument.

The default is scale= FALSE

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

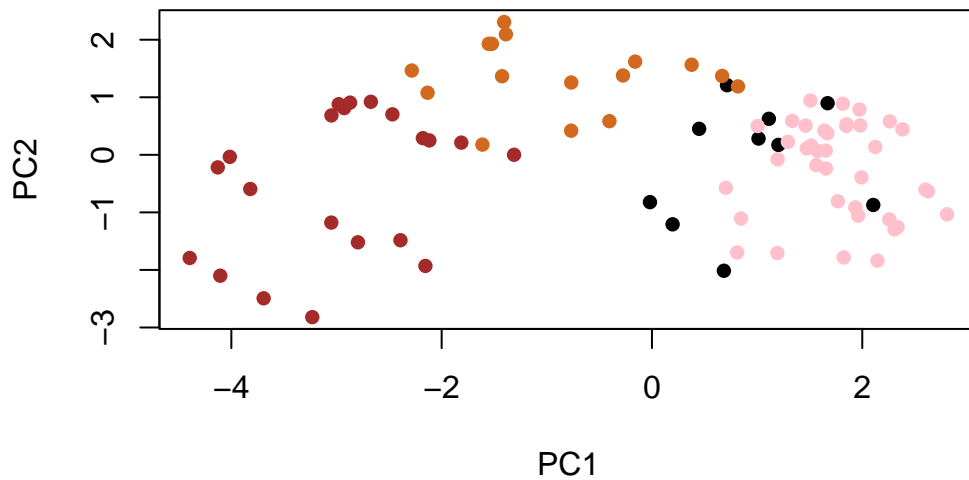
Importance of components:

	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 main PCA score plot of PC1 vs PC2.

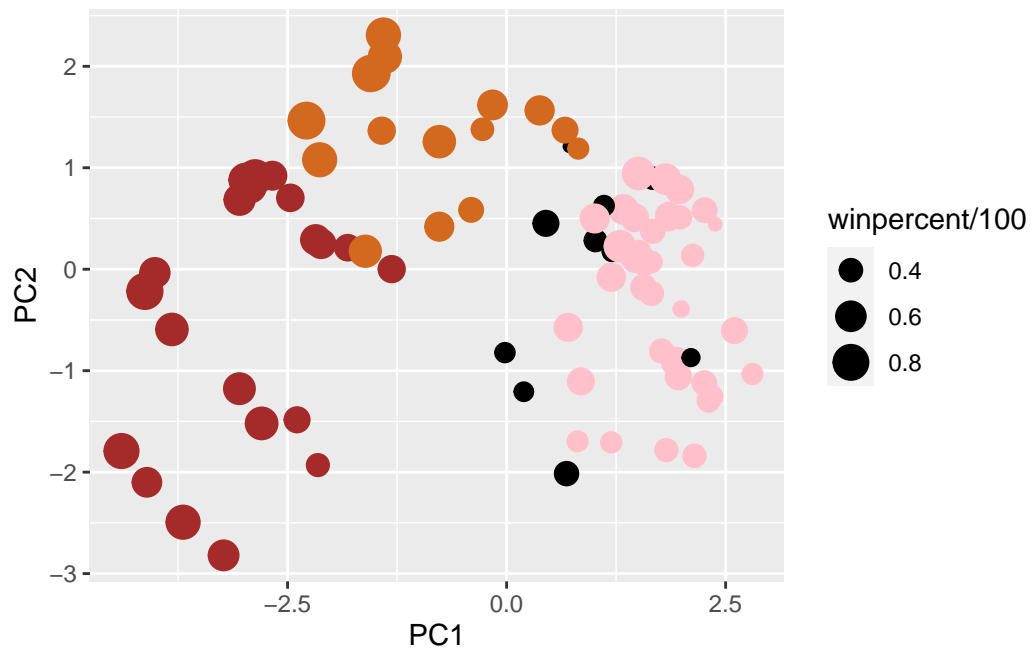
```
plot(pca$x[,1:2], col=my_cols, pch=16)
```



```
# Make a new data-frame with our PCA results and candy data
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



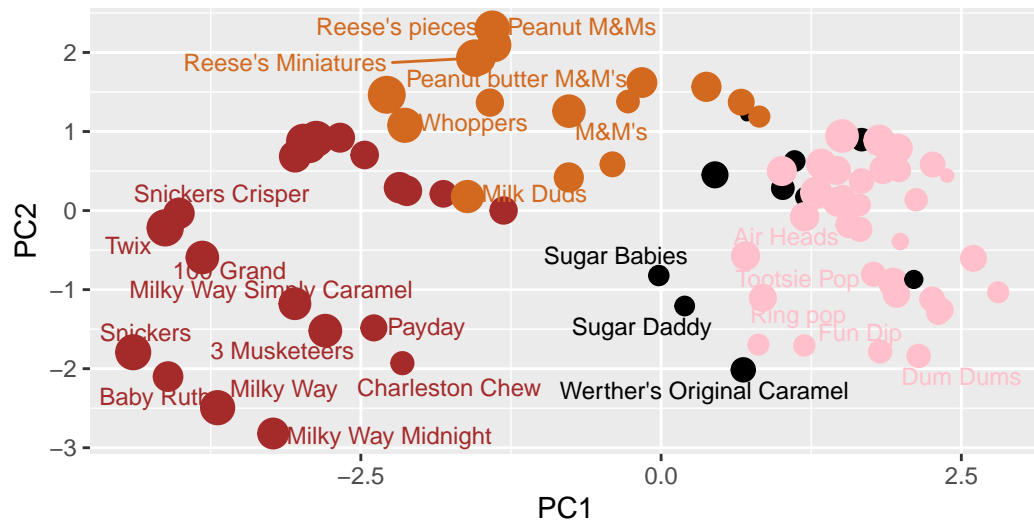
```
library(ggrepel)
```

```
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

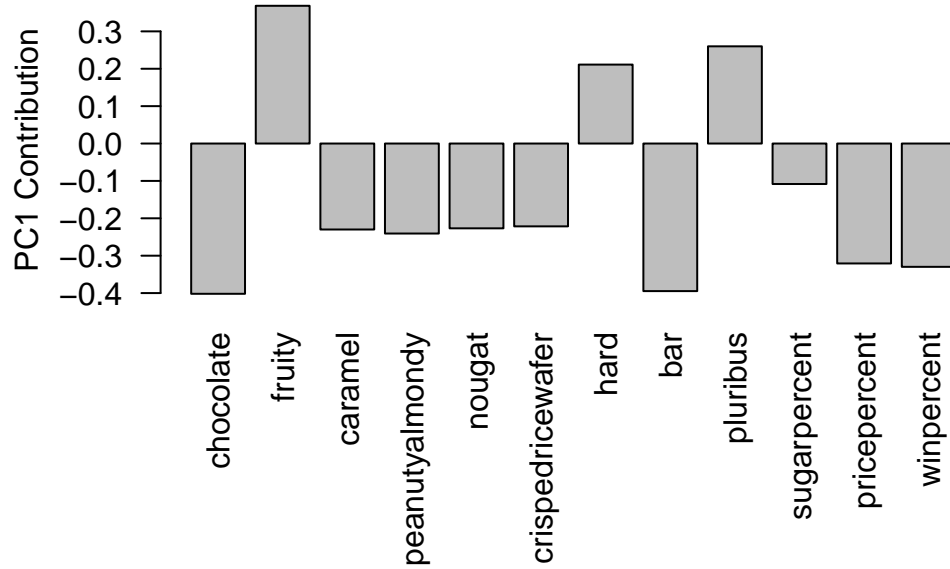
Colored by type: chocolate bar (dark brown), chocolate other (light brown),



Data from 538

```
#install plotly package `install.package("plotly")`
#library(plotly)
#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 most strongly by PC1 in the positive direction. This makes sense as these variables are all positively correlated with one another.