# Class 10: Halloween Candy Mini-Project

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### 2022-10-28

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### Background

In this mini project we will examine 538 halloween can dy data.

```
url <- "https://raw.githubusercontent.com/fivethirtyeight/data/master/candy-power-ranking/
candy <- read.csv(url, row.names=1)
head(candy)</pre>
```

	choco	olate	fruity	caramel	peanutyalm	ondy i	nougat	crispedri	cewafer
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 j	pluribus	sugarpe	ercent pric	eperc	ent wir	percent	
100 Grand	0	1	C	)	0.732	0.8	860 6	6.97173	
3 Musketeers	0	1	C	)	0.604	0.	511 6	7.60294	
One dime	0	0	C	)	0.011	0.3	116 3	32.26109	
One quarter	0	0	C	)	0.011	0.	511 4	6.11650	
Air Heads	0	0	C	)	0.906	0.	511 5	2.34146	
Almond Joy	0	1	C	)	0.465	0.	767 5	0.34755	

### Q1.

How many different candy types are in this dataset?

```
nrow(candy)
```

[1] 85

There are 85 candies in this dataset

### Q2.

How many fruity candy types are in the dataset?

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

[1] 38

There are 38 fruity candies in this dataset

#### What is your favorite Candy?

### Q3

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

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

[1] 76.7686

My favorite candy is Kit Kat and the win percentage is 76.7686.

### Q4

What is the winpercent value for "Kit Kat"?

The win percent value for Kit Kat 76.7686.

#### What is the winpercent value for Tootsie Roll Snack Bars?

The win percent value for Tootsie Roll Snack Bars is 49.653503.

```
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_	_missingcomp	olete_ra	ntmenean	$\operatorname{sd}$	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?

The win percentage average is much higher than most of the other columns as it is in percentage not fraction of the data that is chosen.

### Q7

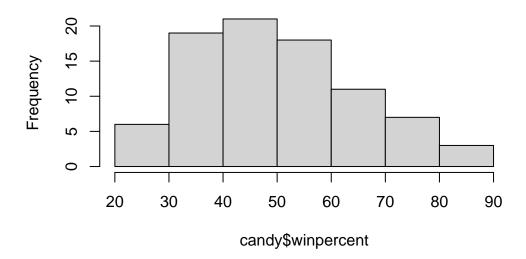
What do you think a zero and one represent for the candy\$chocolate column? The zero and one would represent either yes or no for if the candy is chocolate or not,

### Q8

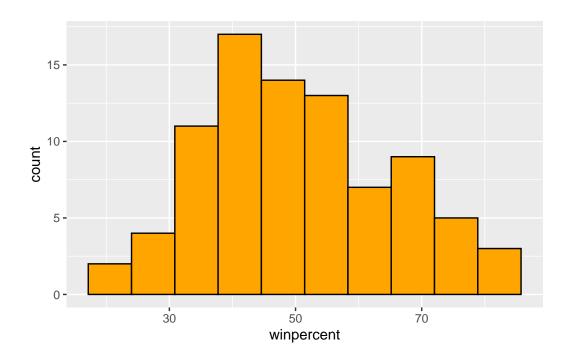
#### Plot a histogram of winpercent values

hist(candy\$winpercent)

# Histogram of candy\$winpercent



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



#### Is the distribution of winpercent values symmetrical?

No, the distribution is not symmetrical

### Q10

Is the center of the distribution above or below 50%? The center of the distribution is below 50%, as the center/median is 47.829754

#### #Q11 On average is the cholocate candy higher or lower ranked than fruit candy?

```
choc_mean_winper <- mean(candy$winpercent[as.logical(candy$chocolate)])
fruit_mean_winper <- mean(candy$winpercent[as.logical(candy$fruity)])
choc_mean_winper > fruit_mean_winper
```

#### [1] TRUE

Is this difference statistically significant?

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

Welch Two Sample t-test

data: choc_stats and fruit_stats
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</pre>
```

Yes this is statistically significant becasue the p-value is less than .05

#### **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
```

		chocolate	fruity	caran	nel j	peanutyaln	nondy	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	
		crispedrio	ewafer	${\tt hard}$	bar	pluribus	sugar	percent	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	Ŀ						
Boston Baked	Beans	23.41782	2						
Chiclets		24.52499	)						
Super Bubble		27.30386	3						
Jawbusters		28.12744	Ŀ						

The least liked candies are Nik L Nip, Boston Baked Beans, Chiclets, Super Bubble, and Jawbusters

### Q14

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

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

	chocolate	fruity	caramel	peanutyalmondy	nougat
Snickers	1	0	1	1	1
Kit Kat	1	0	0	0	0
Twix	1	0	1	0	0
ReeseÕs Miniatures	1	0	0	1	0
ReeseÕs Peanut Butter cup	1	0	0	1	0

crispedricewafer hard bar pluribus sugarpercent

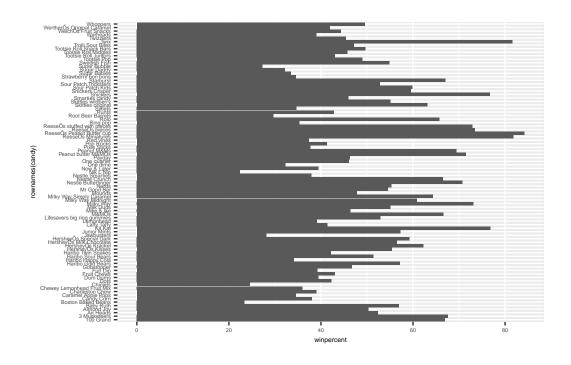
Snickers		0	0	1	0	0.546
Kit Kat		1	0	1	0	0.313
Twix		1	0	1	0	0.546
ReeseÕs Miniatures		0	0	0	0	0.034
ReeseÕs Peanut Butter cup		0	0	0	0	0.720
p	ricepercent	winpe	ercent			
Snickers	0.651	76.	67378			
Kit Kat	0.511	76.	76860			
Twix	0.906	81.	64291			
ReeseÕs Miniatures	0.279	81.	86626			
ReeseÕs Peanut Butter cup	0.651	84.	18029			

The top 5 candies are Snickers, Kit Kat, Twix, Reese's Miniatures, and Reese's Peanut Butter Cup

# Q15.

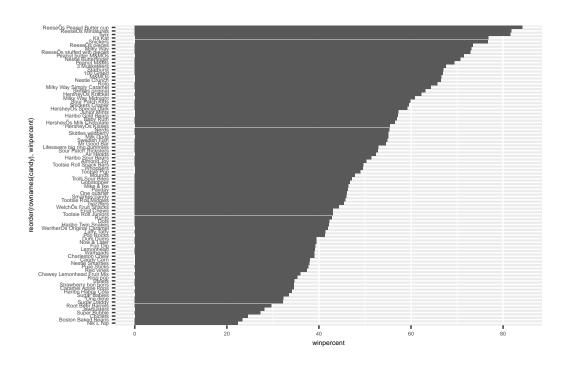
#### Make a first barplot of candy ranking based on winpercent values

ggplot(candy) + aes(winpercent, rownames(candy)) + geom\_col() + theme(text=element\_text(si



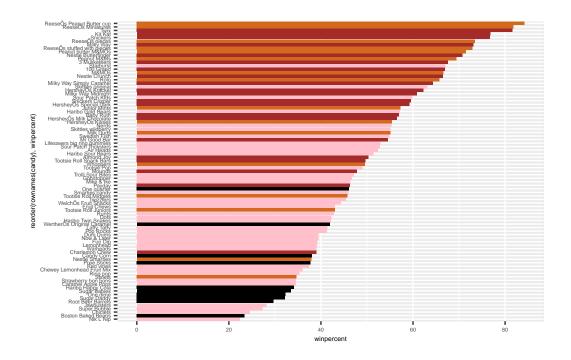
This is quite ugly, use the reorder() function to get the bars sorted by winpercent?

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



```
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_
```



# 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"

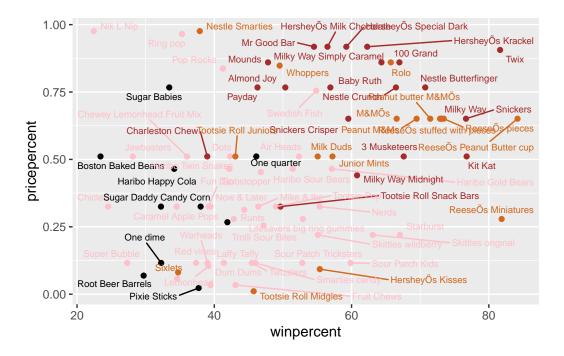
### Taking a look at Pricepercent

```
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=2.5, max.overlaps = 15)
```

Warning: ggrepel: 3 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?

The candy that is ranked highest with the lowest price is the Reese's Miniatures as it has a high win percentage with an overall low price

### **Q20**

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

```
library (dplyr)
most_exp_candy <-candy %>% arrange(pricepercent) %>% tail(5)
```

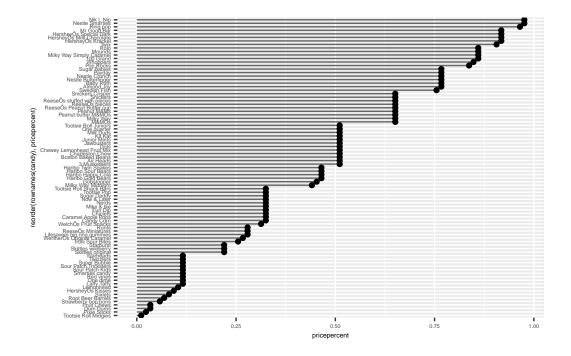
```
most_exp_candy_order <- most_exp_candy %>% arrange(winpercent) %>% head(1)
most_exp_candy_order
```

```
chocolate fruity caramel peanutyalmondy nougat crispedricewafer hard Nik L Nip 0 1 0 0 0 0 0 0 0 bar pluribus sugarpercent pricepercent winpercent Nik L Nip 0 1 0.197 0.976 22.44534
```

Nik L Nip has the lowest ranking (by winpercent) of the 5 most expensive candies.

### **Q21**

#### Make a lollipop chart of the data by pricepercent vs candy

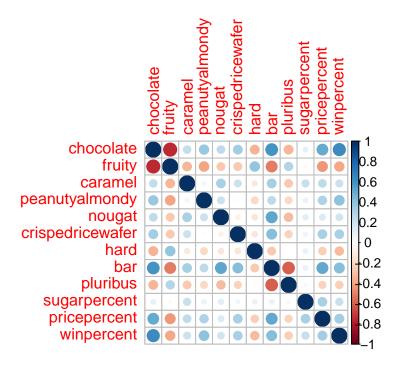


### **Exploring the Correlation Structure**

```
library(corrplot)
```

corrplot 0.92 loaded

cij <- cor(candy)
corrplot(cij)</pre>



# **Q22**

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

Chocolate and Fruity are the most negatively correlated of these categories

#### Similarly, what two variables are most positively correlated?

Chocolate and winpercentage or bar are very positively correlated

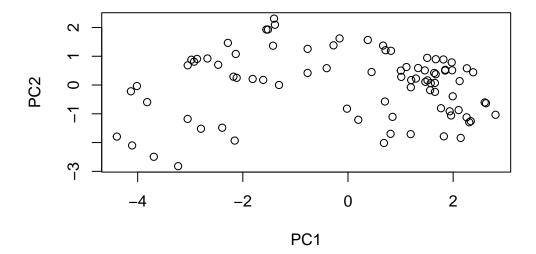
#### **Principle Component Analysis**

```
pca <- prcomp(candy, scale=T)
summary(pca)</pre>
```

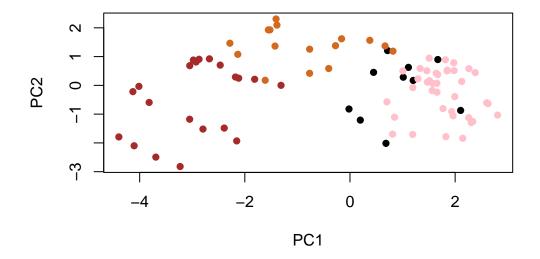
#### Importance of components:

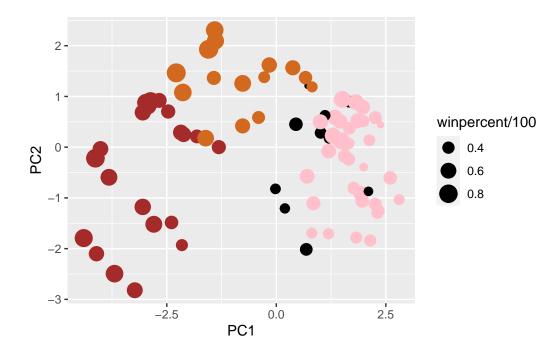
```
PC2
                                        PC3
                                                 PC4
                                                        PC5
                                                                PC6
                                                                        PC7
                          PC1
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)

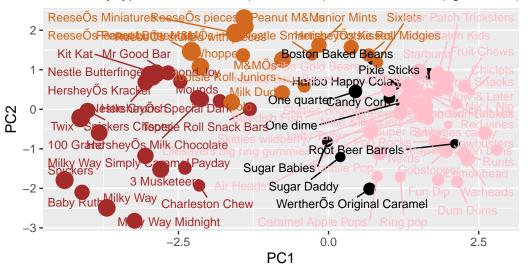




```
library(ggrepel)
```

#### Halloween Candy PCA Space

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



Data from 538

```
library(plotly)

Attaching package: 'plotly'

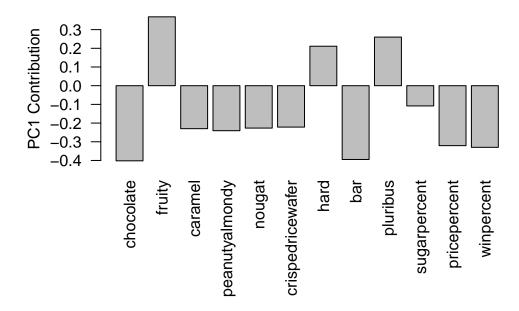
The following object is masked from 'package:ggplot2':
    last_plot

The following object is masked from 'package:stats':
    filter

The following object is masked from 'package:graphics':
    layout

#ggplotly(p)
```

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



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

The original variables that are picked up are fruity, hard, and pluribus Yes I would say this generally makes sense as fruity candies tend to have many in a package and they tend to be harder than many other candies. They are shown as also positively correlated in our correlation plot.