

# Class 10: Halloween Candy Mini Project

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As it is nearly Halloween and the half way point in the quarter , let's do a mini project to figure out the best candy!

## Importing candy data

```
candy <- read.csv("candy-data.txt", 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
```

```
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr     1.1.4     v readr     2.1.5
v forcats   1.0.1     v stringr   1.5.2
v ggplot2   4.0.0     v tibble    3.3.0
v lubridate 1.9.4     v tidyverse 1.3.1
v purrr     1.1.0
-- Conflicts -----
x dplyr::filter() masks stats::filter()
x dplyr::lag()   masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become non-conflicting
```

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
```

## What is your favorite candy?

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

```
library(dplyr)
candy["Twix", ]$winpercent
```

```
[1] 81.64291
```

```
candy["Sour Patch Kids", ]$winpercent
```

```
[1] 59.864
```

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
```

There is a useful `skim()` function in the `skimr` package that can help give you a quick overview of a given dataset. Let’s install this package and try it on our candy data.

```
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	

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
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?

The win percent is on a 0-100 scale, the rest are 0-1

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

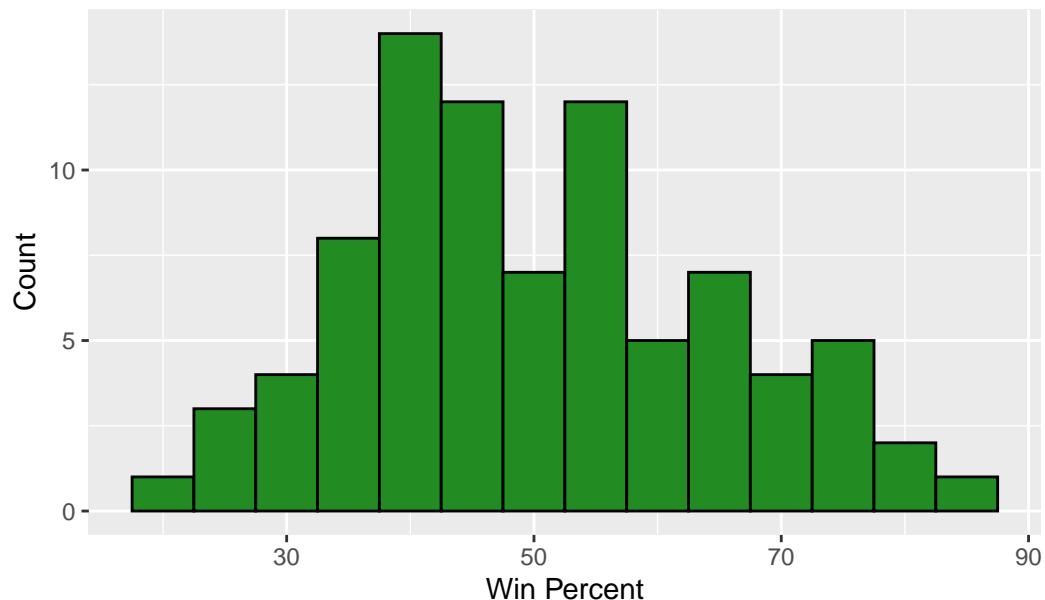
0 = the candy **does not** contain chocolate 1 = the candy **does** contain chocolate

Q8. Plot a histogram of winpercent values

```
library(ggplot2)

ggplot(candy, aes(x = winpercent)) +
  geom_histogram(binwidth = 5, fill = "forestgreen", color = "black") +
  labs(title = "Histogram of Candy Win Percent", x = "Win Percent", y = "Count")
```

### Histogram of Candy Win Percent

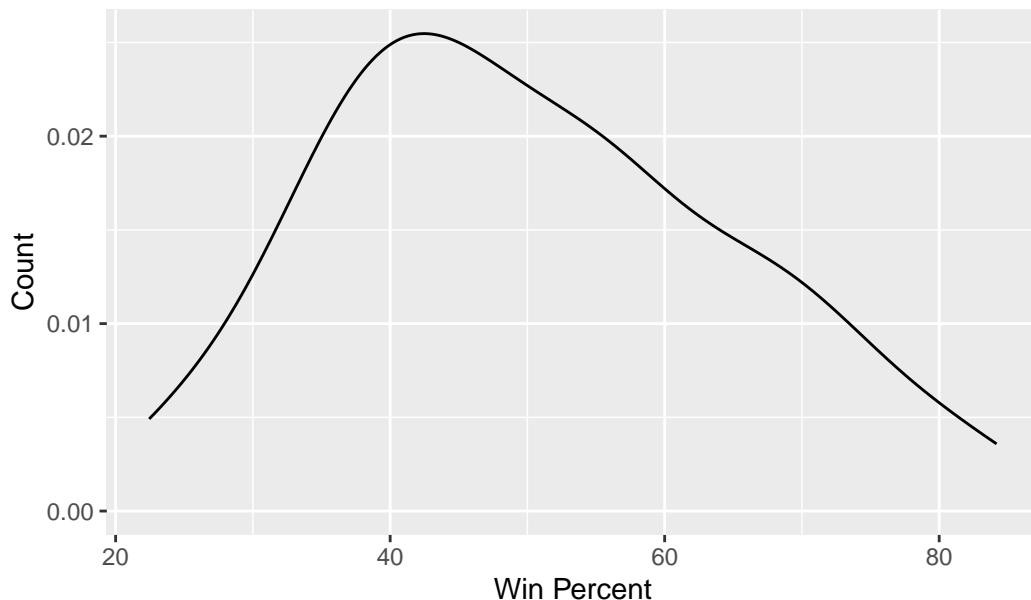


Q9. Is the distribution of winpercent values symmetrical?

```
ggplot(candy, aes(x = winpercent)) +  
  geom_density(binwidth = 5) +  
  labs(title = "Histogram of Candy Win Percent", x = "Win Percent", y = "Count")
```

Warning in geom\_density(binwidth = 5): Ignoring unknown parameters: `binwidth`

## Histogram of Candy Win Percent



No it is not

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

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

[1] 50.31676

```
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?

```
# Chocolate candies  
mean(candy$winpercent[as.logical(candy$chocolate)])
```

[1] 60.92153

```
# Fruity candies  
mean(candy$winpercent[as.logical(candy$fruity)])
```

```
[1] 44.11974
```

Q12. Is this difference statistically significant?

```
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
```

Difference is statistically significant ( $p < 0.05$ )

## Overall Candy Rankings

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

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

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

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					

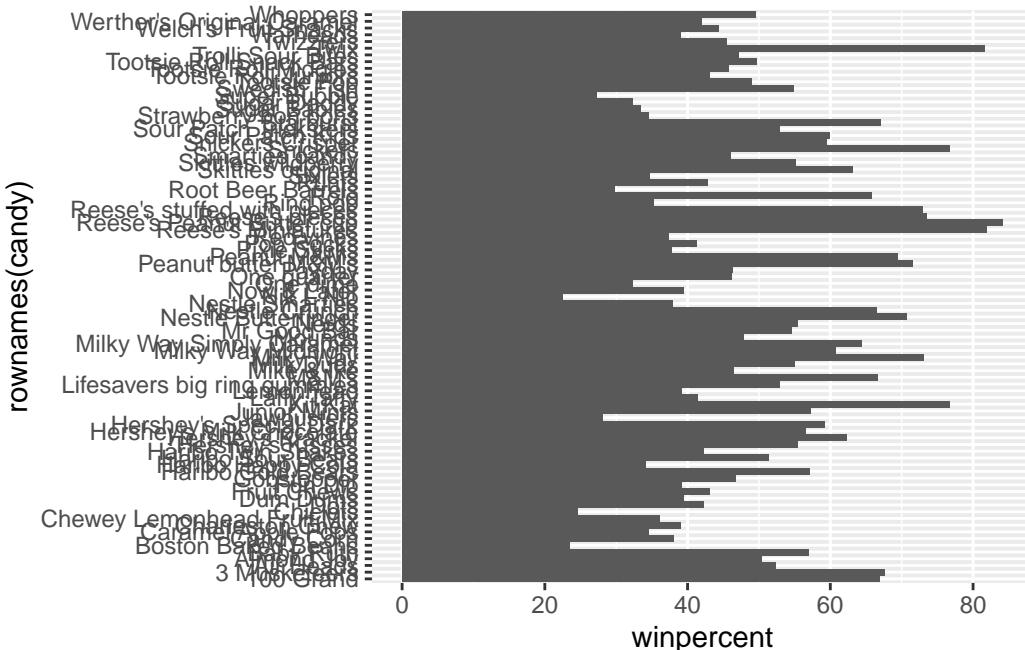
Q14. What are the top 5 all time favorite candy types out of this set?

```
candy %>% arrange(desc(winpercent)) %>% head(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				

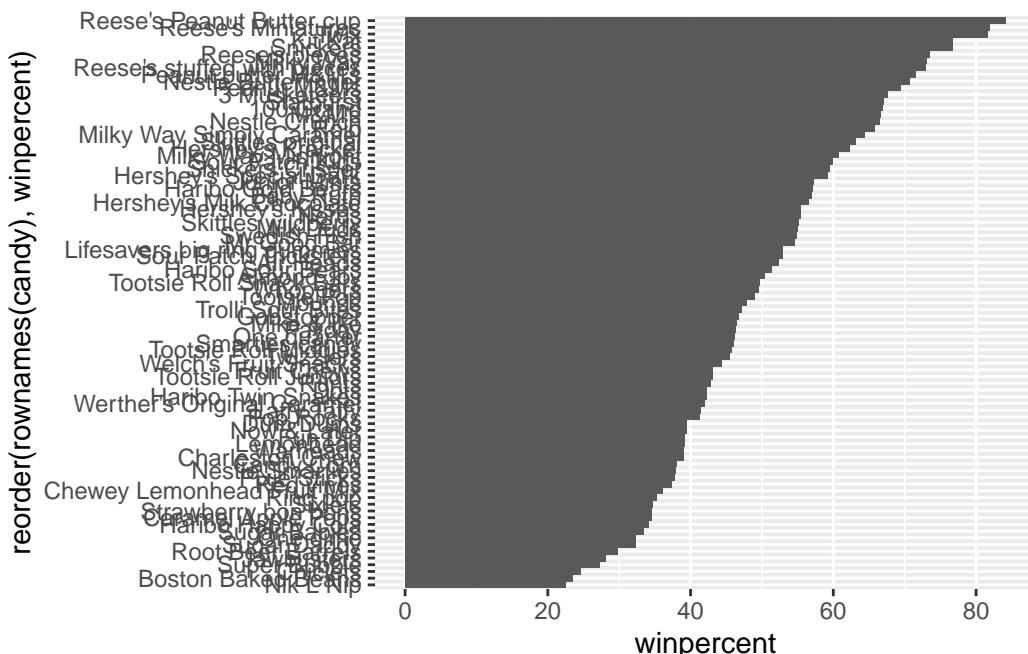
Q15. Make a first barplot of candy ranking based on winpercent values.

```
ggplot(candy) +
  aes(winpercent, rownames(candy)) +
  geom_bar(stat = "identity")
```



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

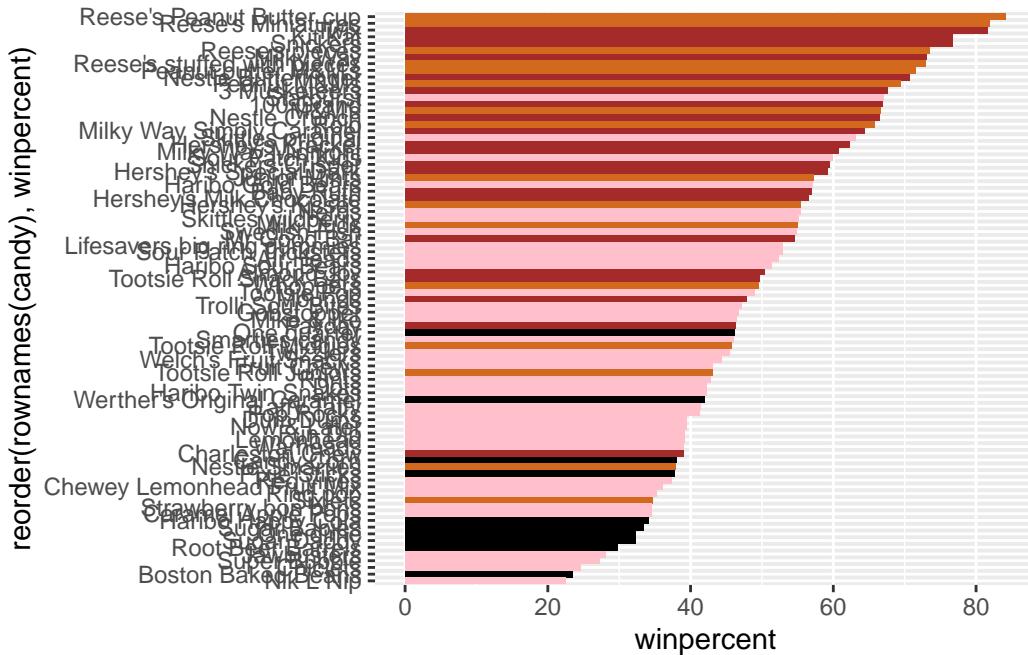
```
ggplot(candy) +  
  aes(winpercent, reorder(rownames(candy), winpercent)) +  
  geom_bar(stat = "identity")
```



```

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)
  
```



Now, for the first time, using this plot we can answer questions like: > Q17. What is the worst ranked chocolate candy?

```
candy_choc <- candy[as.logical(candy$chocolate), ]
rownames(candy_choc)[which.min(candy_choc$winpercent)]
```

[1] "Sixlets"

Q18. What is the best ranked fruity candy?

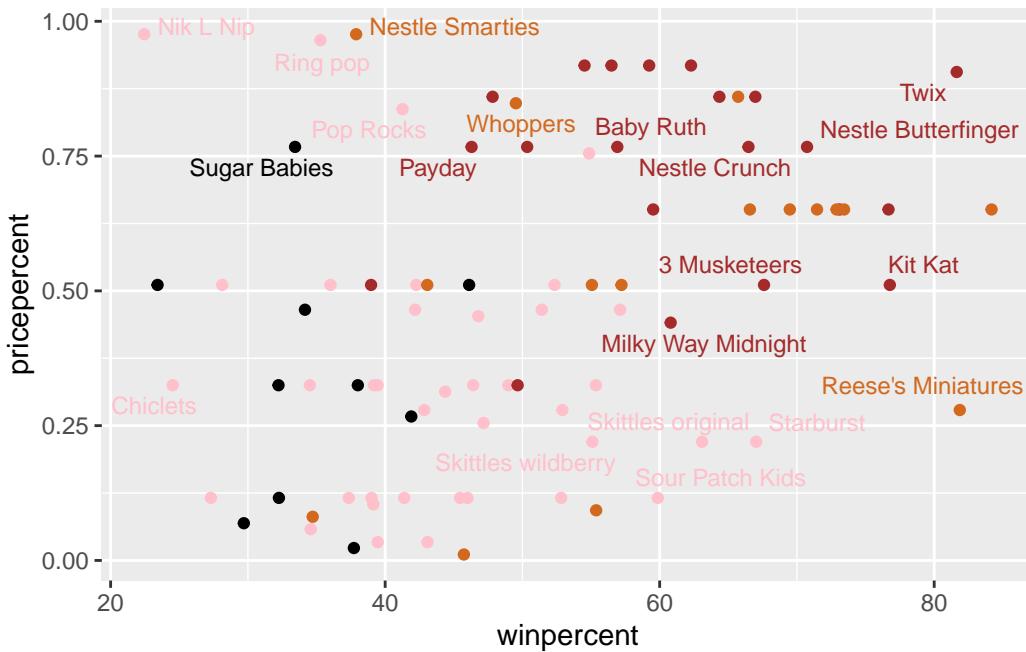
```
candy_fruity <- candy[as.logical(candy$fruity), ]
rownames(candy_fruity)[which.max(candy_fruity$winpercent)]
```

[1] "Starburst"

## Taking a look 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 type is the highest ranked in terms of winpercent for the least money - i.e. offers the most bang for your buck?

```
# Calculate win/price ratio
bang_for_buck <- candy$winpercent / candy$pricepercent

# Find candy with highest ratio
rownames(candy)[which.max(bang_for_buck)]
```

[1] "Tootsie Roll Midgies"

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

Order by pricepercent descending

```
expensive_candies <- candy[order(-candy$pricepercent), ]
```

Top 5 most expensive

```
top5_expensive <- head(expensive_candies, 5)
top5_expensive
```

	chocolate	fruity	caramel	peanut	almond	nougat
Nik L Nip	0	1	0	0	0	0
Nestle Smarties	1	0	0	0	0	0
Ring pop	0	1	0	0	0	0
Hershey's Krackel	1	0	0	0	0	0
Hershey's Milk Chocolate	1	0	0	0	0	0
	crispedrice	wafers	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				

Least popular among these 5

```
rownames(top5_expensive)[which.min(top5_expensive$winpercent)]
```

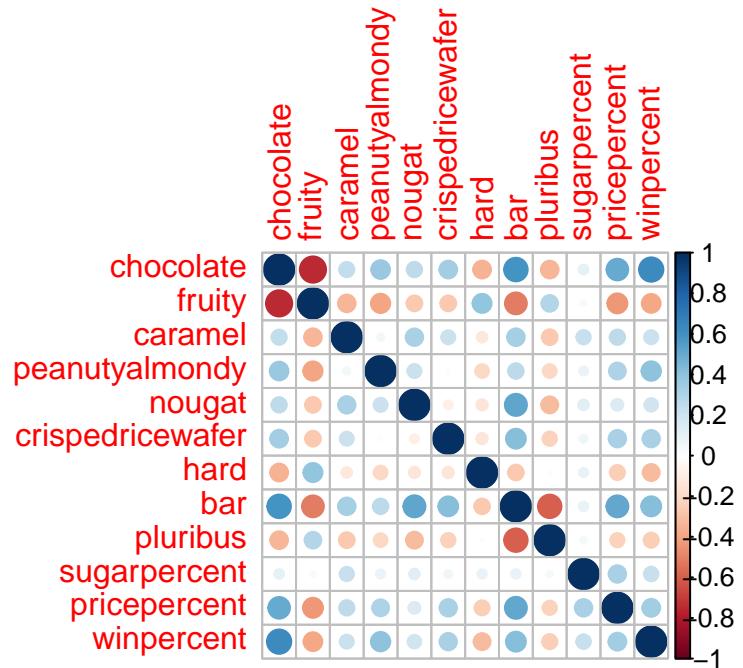
```
[1] "Nik L Nip"
```

## Exploring the correlation structure

```
library(corrplot)
```

```
corrplot 0.95 loaded
```

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



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

chocolate and fruity

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

chocolate and bars ; most candy bars contain chocolate

## Principal Component Analysis

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

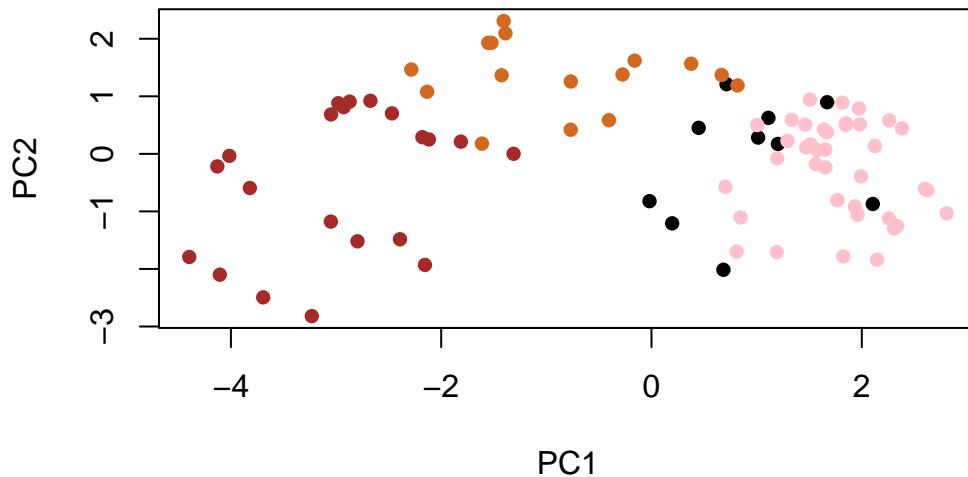
Importance of components:

	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 PC1 vs PC2

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



Combine PCA scores with original candy data

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

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)

library(ggrepel)

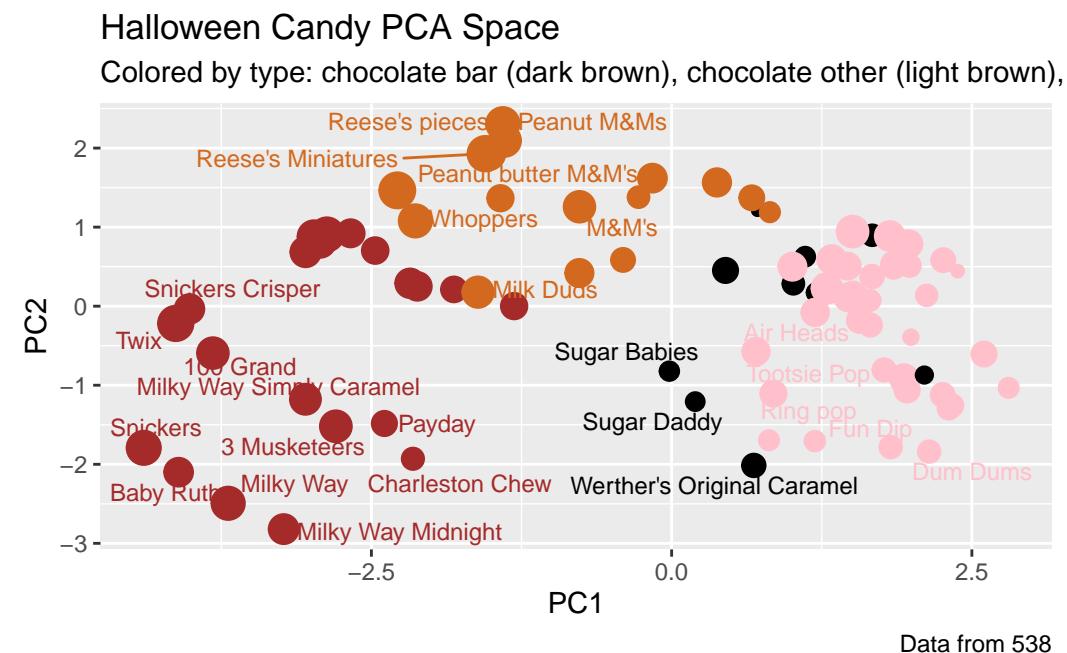
p + geom_text_repel(size=3.3, col=my_cols, max.overlaps = 7) +
  theme(legend.position = "none") +
```

```

  labs(title="Halloween Candy PCA Space",
       subtitle="Colored by type: chocolate bar (dark brown), chocolate other (light brown),
       caption="Data from 538")

```

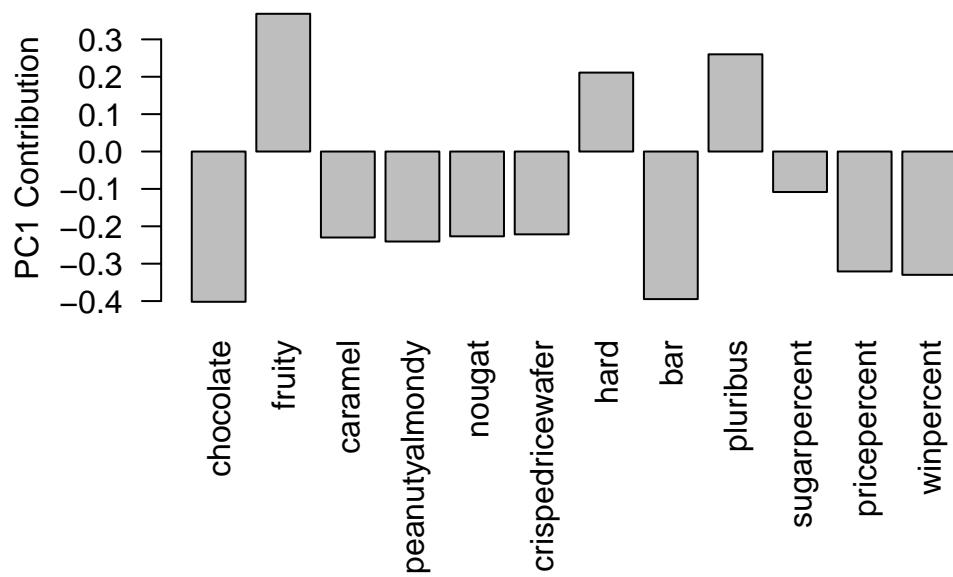
Warning: ggrepel: 59 unlabeled data points (too many overlaps). Consider increasing max.overlaps



```

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?

PC1 picks up variables like chocolate, bar, peanutyalmondy, and pluribus strongly in the positive direction. This makes sense because these features often occur together in candies. Conversely, fruity contributes negatively, separating fruity candies from chocolatey bars along PC1.