Halloween Mini Project

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Table of contents

1.Importing candy data	1
2. What is your favorite candy?	2
3. Overall Candy Rankings	8
Time to add some useful color	13
4. Taking a look at pricepercent	15
5 Exploring the correlation structure	19
6. Principal Component Analysis	20

Today we will examine data from 538 on common Halloween candy. In particular we will use ggplot, dplyr, and PCA to make sense of this multivariate dataset

1.Importing candy data

```
candy <- read.csv("candy-data.txt", row.names=1)
head(candy)</pre>
```

	choco	olate	fruity	caramel	peanut	tyalmondy	nougat	crispedr	icewafer
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	priceper	cent wi	npercent	
100 Grand	0	1	()	0.732	0	.860	66.97173	
3 Musketeers	0	1	()	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?

Ans. There are 85 different candy types in this dataset.

nrow(candy)

[1] 85

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

Ans. There are 38 fruity candy types in this dataset.

sum(candy\$fruity)

[1] 38

How many chocolate candy are there in this dataset?

Ans. There are 37 chocolate candy in this dataset.

sum(candy\$chocolate)

[1] 37

2. What is your favorite candy?

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

Ans. My favorite candy in the dataset is Peanut M&Ms. The winpercent value is 69.48379.

```
candy["Peanut M&Ms", "winpercent"]
```

[1] 69.48379

candy["Peanut M&Ms",]\$winpercent

[1] 69.48379

Q4. What is the winpercent value for "Kit Kat"?

Ans. The winpercent value for Kit Kat is 76.7686.

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

[1] 76.7686

Q5. What is the winpercent value for "Tootsie Roll Snack Bars"?

Ans. The winpercent value for Tootsie Roll Snack Bars is 49.6535.

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

[1] 49.6535

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

Ans. The winpercent variable is on a different scale compared to the majority of the other columns in the dataset. The other variables range from 0 to 1, but the winpercent variable range is much higher than 1.

N.B It looks like the winpercent row in the skim_variable column is on a different scale than the others (0-100% rather than 0-1). I will need to scale this dataset before analysis like PCA.

```
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_	_missingcom	plete_ra	ntmenean	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	

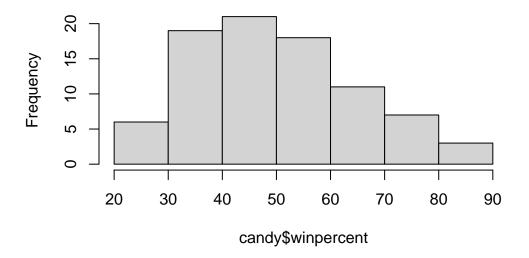
Q7. What do you think a zero and one represent for the candy\$\text{chocolate column}?

Ans. The zero for the candy\$chocolate column represents candies that aren't chocolate. The one represents the candies that contain chocolate.

Q8. Plot a histogram of winpercent values.

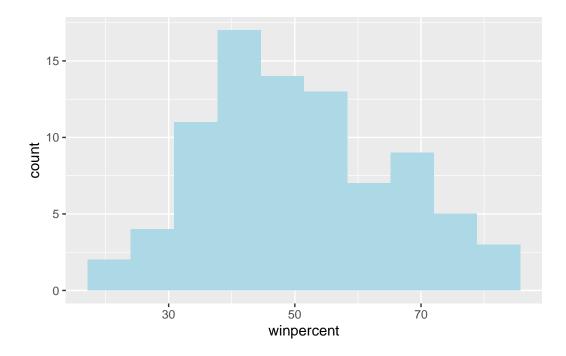
hist(candy\$winpercent)

Histogram of candy\$winpercent



```
library(ggplot2)

ggplot(candy) +
  aes(x=winpercent) +
  geom_histogram(bins=10, fill="lightblue")
```



Q9. Is the distribution of winpercent values symmetrical?

Ans. No, the distribution of winpercent value is skewed to the right.

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

Ans. The center of the distribution is below 50% based on the histogram, with the center being around 45%. We can also see this by using the summary() function to figure out the median, 47.83, which is lower than 50.

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?

Ans. The chocolate candy is ranked higher than the fruit candy.

- Step 1: Find all "chocolate" candy
- Step 2: Find their "winpercent" values
- Step 3: Summarize these values
- Step 4: Find all "fruit" candy

- Step 5: Find their "winpercent" values
- Step 6: Summarize these values
- Step 7: Compare the two summary values
- 1. Find all chocolate candy

```
choc.inds <- candy$chocolate == 1</pre>
```

2. Find their winpercent values

```
choc.win <- candy[choc.inds,]$winpercent</pre>
```

3. Summarize these values

```
choc.mean <- mean(choc.win)
choc.mean</pre>
```

[1] 60.92153

4. Find all fruit candy

```
fruity.inds <- candy$fruity == 1</pre>
```

5. Find their winpercent values

```
fruity.win <- candy[fruity.inds,]$winpercent</pre>
```

6. Summarize these values

```
fruity.mean <- mean(fruity.win)
fruity.mean</pre>
```

[1] 44.11974

7. Compare the two summary values The chocolate winpercent is higher than the fruity candy.

```
choc.mean
```

[1] 60.92153

fruity.mean

[1] 44.11974

Q12. Is this difference statistically significant?

Ans. This difference is statistically signficant, due to the p-value being very low, 2.871e-08.

t.test(choc.win, fruity.win)

```
Welch Two Sample t-test
```

```
data: choc.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
```

3. Overall Candy Rankings

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

Ans. The 5 least liked candy types in this dataset is Nik L Nip, Boston Baked Beans, Chiclets, Super Bubble, and Jawbusters.

```
# Not that useful - it just sorts the values
sort(candy$winpercent)
```

```
[1] 22.44534 23.41782 24.52499 27.30386 28.12744 29.70369 32.23100 32.26109 [9] 33.43755 34.15896 34.51768 34.57899 34.72200 35.29076 36.01763 37.34852 [17] 37.72234 37.88719 38.01096 38.97504 39.01190 39.14106 39.18550 39.44680 [25] 39.46056 41.26551 41.38956 41.90431 42.17877 42.27208 42.84914 43.06890 [33] 43.08892 44.37552 45.46628 45.73675 45.99583 46.11650 46.29660 46.41172 [41] 46.78335 47.17323 47.82975 48.98265 49.52411 49.65350 50.34755 51.41243 [49] 52.34146 52.82595 52.91139 54.52645 54.86111 55.06407 55.10370 55.35405
```

```
[57] 55.37545 56.49050 56.91455 57.11974 57.21925 59.23612 59.52925 59.86400 [65] 60.80070 62.28448 63.08514 64.35334 65.71629 66.47068 66.57458 66.97173 [73] 67.03763 67.60294 69.48379 70.73564 71.46505 72.88790 73.09956 73.43499 [81] 76.67378 76.76860 81.64291 81.86626 84.18029
```

```
x <- c(10, 1, 100)
order(x)
```

[1] 2 1 3

x[order(x)]

[1] 1 10 100

The order() function tells us how it arrange the elements of the input to make them sorted - i.e. how to order them

We can determine the order of winpercent to make them sorted and use that order to arrange the whole dataset.

```
order.inds <- order(candy$winpercent)
head(candy[order.inds, ])</pre>
```

	chocolate	fruity	cara	ו ום	neanutvalm	nondv	ກດນອາ	
Nik L Nip	0	1	cara	0	peanabyan	nonay O	0	
-	•			_		1	•	
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	
Root Beer Barrels	0	0		0		0	0	
	crispedrio	ewafer	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
Root Beer Barrels		0	1	0	1		0.732	0.069
	winpercent	;						
Nik L Nip	22.44534	<u> </u>						
Boston Baked Beans	23.41782	2						
Chiclets	24.52499)						

Super Bubble 27.30386 Jawbusters 28.12744 Root Beer Barrels 29.70369

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

Ans. The top 5 all time favorite candy types are Reese's Peanut Butter Cup, Snickers, Kit Kat, Twix, and Reese's Miniatures.

tail(candy[order.inds,])

	chocolate	fruity	caran	nel j	peanutyalm	ondy	nougat
Reese's pieces	1	0		0		1	0
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
	crispedrio	cewafer	${\tt hard}$	bar	pluribus	sugar	percent
Reese's pieces		0	0	0	1		0.406
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
	priceperce	ent winp	percer	nt			
Reese's pieces	0.6	351 73	3.4349	99			
Snickers	0.6	351 76	6.6737	78			
Kit Kat	0.5	511 76	5.7686	30			
Twix	0.9	906 83	1.6429	91			
Reese's Miniatures	0.2	279 83	1.8662	26			
Reese's Peanut Butter cup	0.6	351 84	4.1802	29			

order.winpercent.decrease <- order(candy\$winpercent, decreasing=TRUE)
head(candy[order.winpercent.decrease,])</pre>

	${\tt chocolate}$	${\tt fruity}$	caramel	${\tt peanutyalmondy}$	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
Reese's pieces	1	0		0		1	0
	crispedricewa	afer	${\tt hard}$	bar	pluribus	sugarp	ercent
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
Reese's pieces		0	0	0	1		0.406
	pricepercent	winp	percer	nt			
Reese's Peanut Butter cup	0.651	84	1.1802	29			
Reese's Miniatures	0.279	83	1.8662	26			
Twix	0.906	83	1.6429	91			
Kit Kat	0.511	76	5.7686	30			
Snickers	0.651	76	6.6737	78			
Reese's pieces	0.651	73	3.4349	99			

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) %>%
  tail(5)
```

	${\tt chocolate}$	fruity	caramel	${\tt peanutyalmondy}$	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

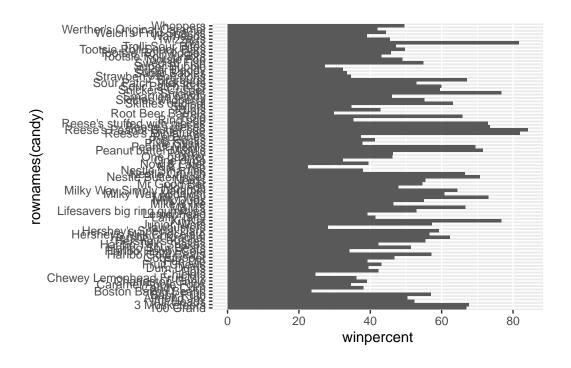
	crispedricewaf	er	hard	bar	pluribus	sugarpercent
Snickers		0	0	1	0	0.546
Kit Kat		1	0	1	0	0.313
Twix		1	0	1	0	0.546
Reese's Miniatures		0	0	0	0	0.034
Reese's Peanut Butter cup		0	0	0	0	0.720
	pricepercent w	inp	ercer	nt		
Snickers	0.651	76	6.6737	78		
Kit Kat	0.511	76	5.7686	60		
Twix	0.906	81	.6429	91		
Reese's Miniatures	0.279	81	.8662	26		
Reese's Peanut Butter cup	0.651	84	1.1802	29		

Q Which approach do you prefer and why?

I prefer the dyplyr because it requires less parenthesis and commas, which makes it harder for me to make a mistake with a comma.

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

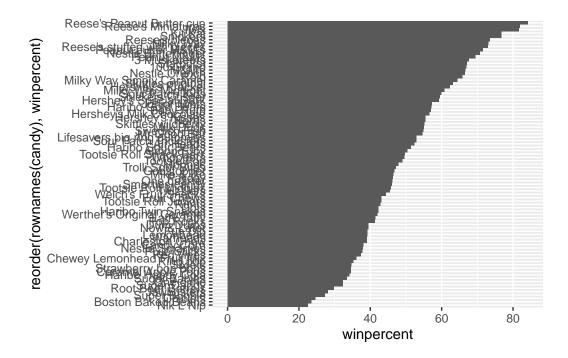
```
ggplot(candy) +
  aes(x=winpercent, rownames(candy)) +
  geom_col()
```



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

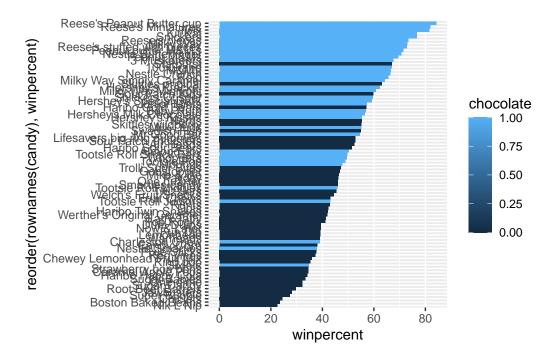
Let's reorder it:

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



Time to add some useful color

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



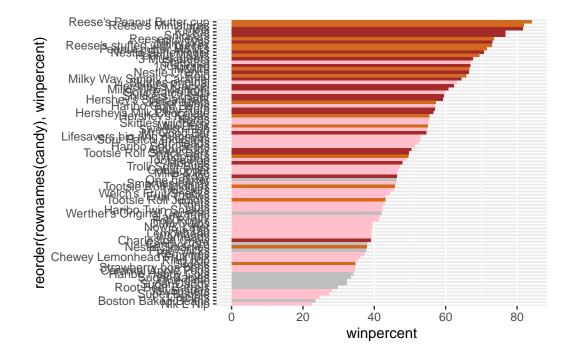
We need to make our own separate color vector where we can spell out what candy is colored a particular color.

```
mycols <- rep("gray", nrow(candy))
mycols[candy$chocolate == 1] <- "chocolate"
mycols[candy$bar == 1] <- "brown"
mycols[candy$fruity == 1] <- "pink"
mycols</pre>
```

```
[1] "brown"
                  "brown"
                               "gray"
                                            "gray"
                                                         "pink"
                                                                      "brown"
 [7] "brown"
                  "gray"
                               "gray"
                                            "pink"
                                                         "brown"
                                                                      "pink"
                                                         "pink"
[13] "pink"
                  "pink"
                               "pink"
                                            "pink"
                                                                      "pink"
[19] "pink"
                  "gray"
                               "pink"
                                            "pink"
                                                         "chocolate"
                                                                      "brown"
                               "pink"
[25] "brown"
                  "brown"
                                            "chocolate" "brown"
                                                                      "pink"
                  "pink"
                               "chocolate"
                                           "chocolate" "pink"
                                                                      "chocolate"
[31] "pink"
[37] "brown"
                  "brown"
                               "brown"
                                            "brown"
                                                         "brown"
                                                                      "pink"
                  "brown"
                               "pink"
                                            "pink"
                                                         "brown"
                                                                      "chocolate"
[43] "brown"
[49] "gray"
                  "pink"
                               "pink"
                                            "chocolate" "chocolate" "chocolate"
[55] "chocolate"
                  "pink"
                               "chocolate"
                                            "gray"
                                                         "pink"
                                                                      "chocolate"
                                                                      "brown"
[61] "pink"
                  "pink"
                               "chocolate" "pink"
                                                         "brown"
[67] "pink"
                  "pink"
                               "pink"
                                            "pink"
                                                         "gray"
                                                                      "gray"
                                            "chocolate" "chocolate" "brown"
[73] "pink"
                  "pink"
                               "pink"
[79] "pink"
                  "brown"
                               "pink"
                                            "pink"
                                                         "pink"
                                                                      "gray"
```

[85] "chocolate"

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



as.logical(c(1,0,1))

[1] TRUE FALSE TRUE

Q17. What is the worst ranked chocolate candy?

Ans. The worst ranked chocolate candy is Sixlets.

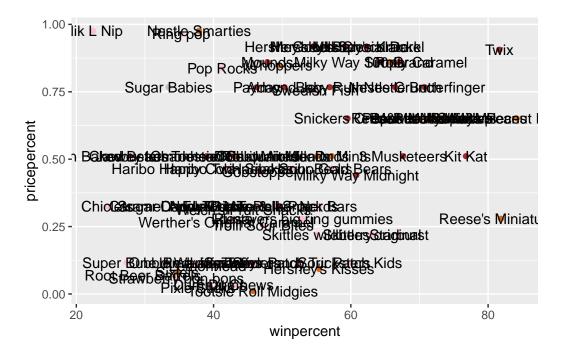
Q18. What is the best ranked fruity candy?

Ans. The best ranked fruity candy is Starburst.

4. Taking a look at pricepercent

Make a plot of winpercent (x-axis) vs pricepercent (y-axis)

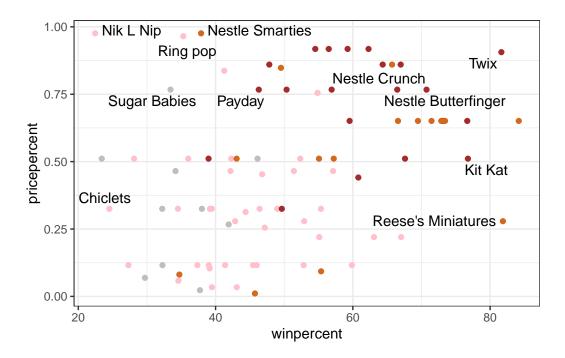
```
ggplot(candy) +
  aes(winpercent, pricepercent, label=rownames(candy)) +
  geom_point(col=mycols) +
  geom_text()
```



To avoid the overplotting of the text labels, we can use the add on package ggrepel

```
ggplot(candy) +
  aes(winpercent, pricepercent, label=rownames(candy)) +
  geom_point(col=mycols) +
  geom_text_repel(max.overlaps = 5) +
  theme_bw()
```

Warning: ggrepel: 74 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?

Ans. Reeses Miniatrues has one of the highest winpercent and one of the lowest pricepercent.

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

```
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
```

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

```
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

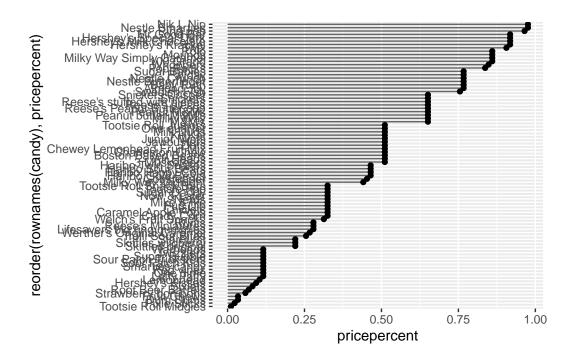
Q20. What are the top 5 most expensive candy types in the dataset and of these which is the least popular?

Ans. The top 5 most expensive candy types in the dataset are Hershey's Special Dark, Mr Good Bar, Ring pop, Nik L Nip, and Nestle Smarties. The least popular of these 5 is the Nik L Nip.

```
order.pricepercent <- order(candy$pricepercent)
tail(candy[order.pricepercent, ],n=5)</pre>
```

		chocolate	fruity	carar	nel j	peanutyalr	nondy	nougat	
Hershey's Special	Dark	1	0		0		0	0	
Mr Good Bar		1	0		0		1	0	
Ring pop		0	1		0		0	0	
Nik L Nip		0	1		0		0	0	
Nestle Smarties		1	0		0		0	0	
		crispedrio	cewafer	${\tt hard}$	bar	pluribus	sugar	percent	
Hershey's Special	Dark		0	0	1	0		0.430	
Mr Good Bar			0	0	1	0		0.313	
Ring pop			0	1	0	0		0.732	
Nik L Nip			0	0	0	1		0.197	
Nestle Smarties			0	0	0	1		0.267	
pricepercent winpercent									
Hershey's Special	\mathtt{Dark}	0.9	918 59	9.2361	12				
Mr Good Bar		0.9	918 54	4.5264	1 5				
Ring pop		0.9	965 35	5.2907	76				
Nik L Nip		0.9	976 22	2.4453	34				
Nestle Smarties		0.9	976 37	7.887	19				

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().



5 Exploring the correlation structure

Now that we have explored the data set a little, we will see how the variables interact with one another.

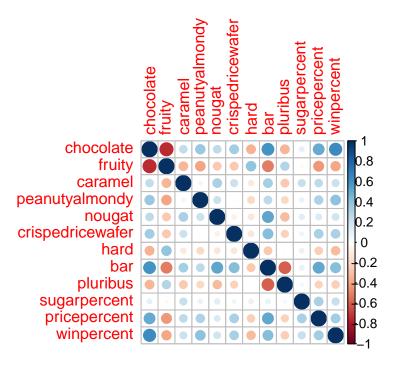
First we will use correlation and view the results with the **corrplot** package to plot a correlation matrix.

cij <- cor(candy)</pre>

library(corrplot)

corrplot 0.95 loaded

corrplot(cij)



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

Ans. Two variables that are anti-correlated are chocolate and fruity candy.

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

Ans. Two variables that are most positively correlated are chocolate and winpercent.

6. Principal Component Analysis

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

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

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
```

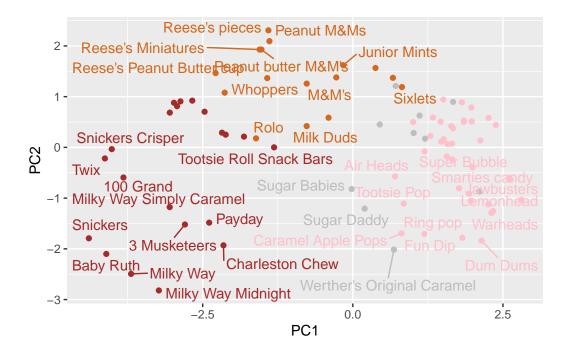
```
attributes(pca)
```

```
$names
[1] "sdev" "rotation" "center" "scale" "x"
$class
[1] "prcomp"
```

Let's plot our main results as our PCA "score plot"

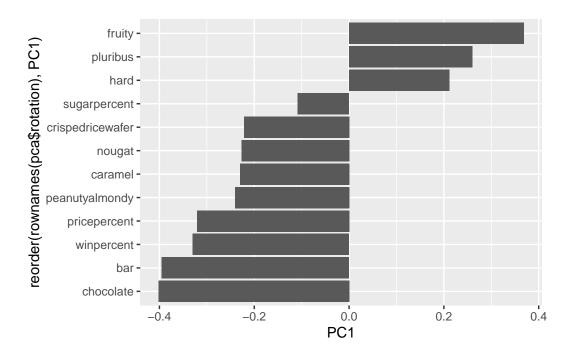
```
ggplot(pca$x) +
aes(PC1, PC2, label=rownames(pca$x)) +
geom_point(col=mycols) +
geom_text_repel(col=mycols)
```

Warning: ggrepel: 48 unlabeled data points (too many overlaps). Consider increasing max.overlaps



Finally, let's look at the how the original variables contribute to the PCs, start with PC1

```
ggplot(pca$rotation) +
  aes(PC1, reorder(rownames(pca$rotation), PC1)) +
  geom_col()
```



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

Ans. The original variables that are picked up strongly by PC1 in the positive direction are fruity, pluribus, and hard. This makes sense to me because the graphs shown above have all the fruity candy on the left side of the graph with PC1 on the x-axis.