Class 10

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Class 10: Halloween Mini-Project

1. Importing candy data

```
candy_file <- "candy-data.csv"

candy = read.csv(candy_file, row.names = 1)
head(candy)</pre>
```

	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 har nluribus		cuarna	ercent priceper	ent wir	nercent

	naru	Dai	b car ibas	sugarpercent	bi rcebei cent	winbercent
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 candy types are in this dataset?

```
candy_types <- nrow(candy)
candy_types</pre>
```

[1] 85

There are 85 different candy types.

Q2. How many fruit candy types are in this dataset?

```
fruity_candy <- sum(candy$fruity)
fruity_candy</pre>
```

There are 38 fruity candy types.

2. What is your favorite candy?

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

candy["Nerds",]\$winpercent

[1] 55.35405

My favorite candy is Nerds and its winpercent value is 55.35.

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

candy["Kit Kat",]\$winpercent

[1] 76.7686

The winpercent value for "Kit Kat" is 76.77.

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

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

[1] 49.6535

The winpercent value for "Tootsie Roll Snack Bars" is 49.65.

library("skimr")
skim(candy)

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 compl	ete_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?

The winpercent column stands out the most from the other columns in the dataset. The winpercent column has a minimum value of 0 and a maximum value of 84.18 while the other columns range from 0 to 1. This suggests that the winpercent column has a different scale.

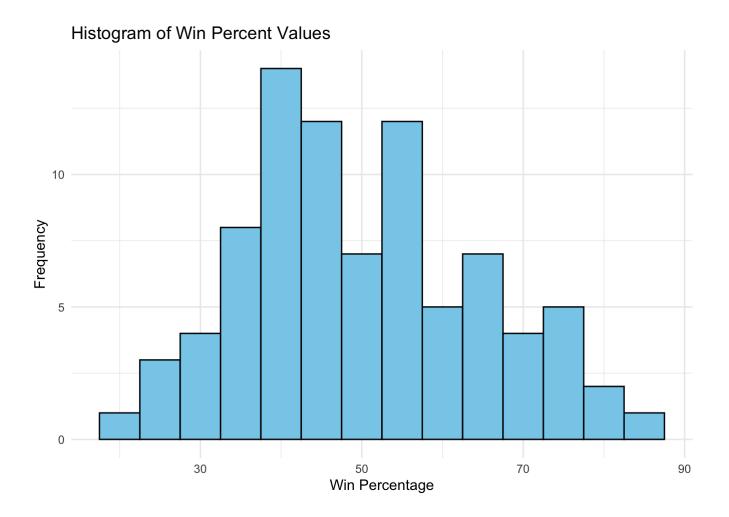
Q7. What do you think a zero and one represent for the candy\$chocolate column?

The 0 most likely stands for candy without chocolate and the 1 stands for candy that contains chocolate.

Q8. Plot a histogram of winpercent values

```
library(ggplot2)

ggplot(candy, aes(x = winpercent)) +
  geom_histogram(binwidth = 5, fill = "skyblue", color = "black") +
  labs(x = "Win Percentage", y = "Frequency", title = "Histogram of Win Percent Values")
  theme_minimal()
```



Q9. Is the distribution of winpercent values symmetrical?

No, the distribution of winpercent values skew right.

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

```
mean_winpercent <- mean(candy$winpercent)
if (mean_winpercent > 50) {
  center <- "above"
} else if (mean_winpercent < 50) {
  center <- "below"
} else {
  center <- "at"
}
center</pre>
```

[1] "above"

The center of the distribution is above 50%.

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

```
mean_chocolate <- mean(candy$winpercent[as.logical(candy$chocolate)])
mean_fruit <- mean(candy$winpercent[as.logical(candy$fruity)])

if (mean_chocolate > mean_fruit) {
   rank <- "higher"
} else if (mean_chocolate < mean_fruit) {
   rank <- "lower"
} else {
   rank <- "equal"
}
rank</pre>
```

[1] "higher"

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

Q12. Is this difference statistically significant?

```
t.test <- t.test(candy$winpercent[as.logical(candy$chocolate)], candy$winpercent[as.logic
t.test</pre>
```

```
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
```

Because the p-value is < 0.05 and the confidence interval (11.44563 to 22.15795) indicates the range of differences in means, the difference between chocolate candy and fruity candy is statistically significant.

3. Overall Candy Rankings

Nik L Nip

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

```
least_liked <- head(candy[order(candy$winpercent), ], n = 5)
least_liked</pre>
```

```
chocolate fruity caramel peanutyalmondy nougat
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	
		crispedricev	vafer	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							
Boston Baked	Beans	23.41782							
Chiclets		24.52499							
Super Bubble		27.30386							
Jawbusters		28.12744							

The five least like candy types 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?

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

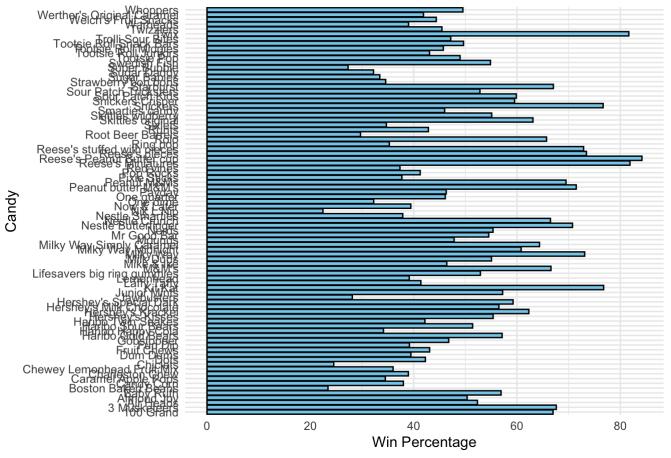
				_	_		
	chocolate	fruity	caran	nel p	peanutyalr	nondy	nougat
Reese's Peanut Butter cup	1	0		0		1	0
Reese's Miniatures	1	0		0		1	0
Twix	1	0		1		0	0
Kit Kat	1	0		0		0	0
Snickers	1	0		1		1	1
	crispedrio	cewafer	hard	bar	pluribus	sugai	rpercent
Reese's Peanut Butter cup		0	0	0	0		0.720
Reese's Miniatures		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
	priceperce	ent win	percer	nt			
Reese's Peanut Butter cup	0.6	551 84	4.1802	29			
Reese's Miniatures	0.2	279 83	1.8662	26			
Twix	0.9	906 83	1.6429	91			
Kit Kat	0.5	511 70	5.7686	50			
Snickers	0.6	551 70	6.6737	78			

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

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

```
ggplot(candy) +
  aes(winpercent, rownames(candy)) +
  geom_col(fill = "skyblue", color = "black") +
  labs(x = "Win Percentage", y = "Candy", title = "Candy Ranking Based on Win Percentage"
  theme_minimal()
```

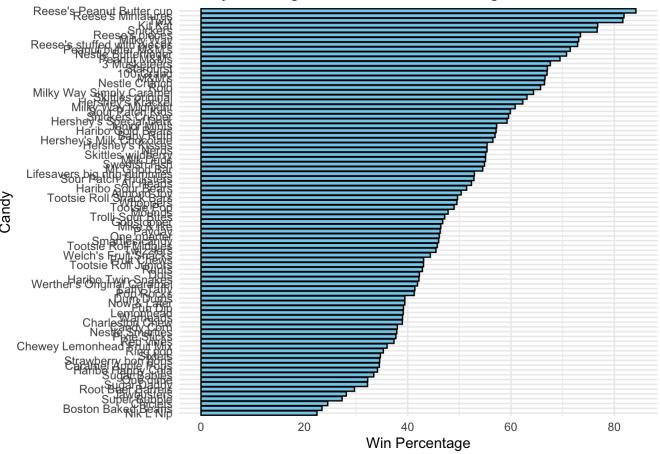
Candy Ranking Based on Win Percentage



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(fill = "skyblue", color = "black") +
  labs(x = "Win Percentage", y = "Candy", title = "Candy Ranking Based on Win Percentage"
  theme_minimal()
```

Candy Ranking Based on Win Percentage

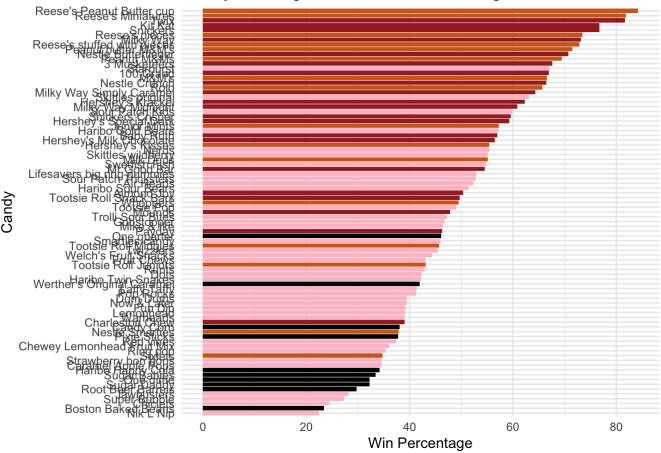


Time to add some useful color

```
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"

# Barplot with colors based on candy type
ggplot(candy) +
    aes(winpercent, reorder(rownames(candy), winpercent)) +
    geom_col(fill = my_cols) +
    labs(x = "Win Percentage", y = "Candy", title = "Candy Ranking Based on Win Percentage"
    theme_minimal()</pre>
```

Candy Ranking Based on Win Percentage



Q17. What is the worst ranked chocolate candy?

Sixlets

Q18. What is the best ranked fruit candy?

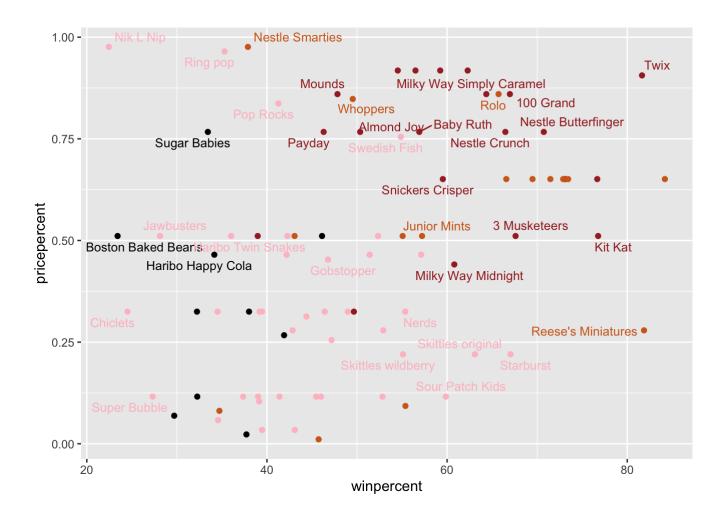
Starburst

4. 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=3.3, max.overlaps = 5)
```

Warning: ggrepel: 50 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?

```
min_price_index <- which.min(candy$pricepercent)
least_money <- rownames(candy)[which.max(candy$winpercent[c(min_price_index)])]
least_money</pre>
```

[1] "100 Grand"

100 Grand offers the most bang for your buck!

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)</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

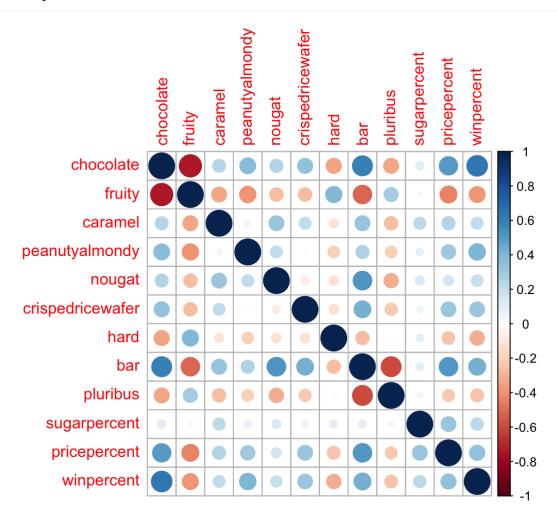
The top 5 most expensive candy types are Nik L Nip, Nestle Smarties, Ring pop, Hershey's Krackel, and Hershey's Milk Chocolate. The least popular is Nik L Nip.

5. 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)?

pricepercent and winpercent

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

sugarpercent and pricepercent

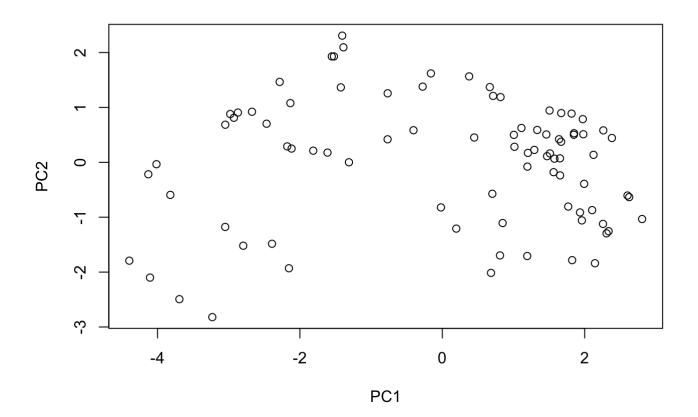
6. Principal Component Analysis

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

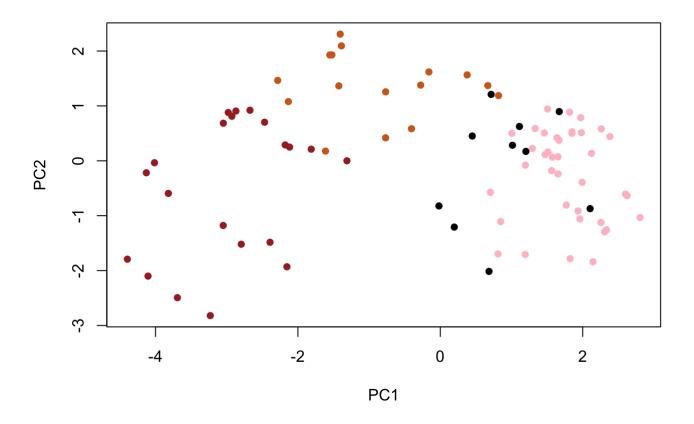
Importance of components:

```
PC4
                          PC1
                                 PC2
                                        PC3
                                                        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
```

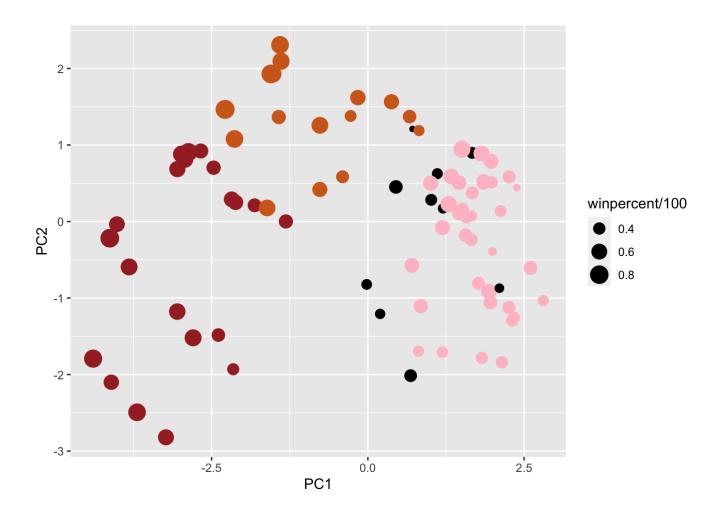
```
plot(pca$x[, 1], pca$x[, 2], xlab = "PC1", ylab = "PC2")
```



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



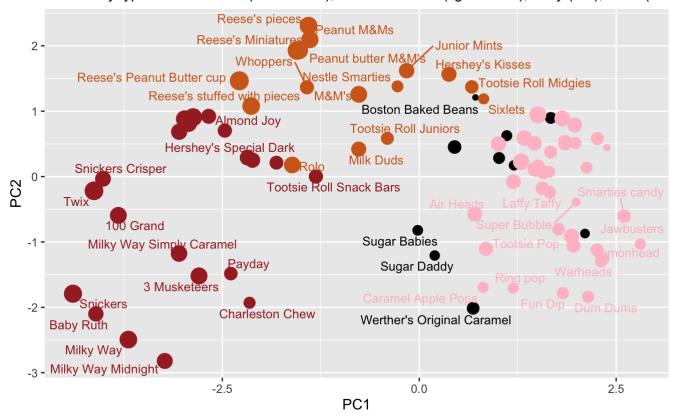
```
# Make a new data-frame with our PCA results and candy data
my_data <- cbind(candy, pca$x[,1:3])</pre>
```



Warning: ggrepel: 39 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), fruity (red), other (blac



Data from 538

library(plotly)

```
Attaching package: 'plotly'

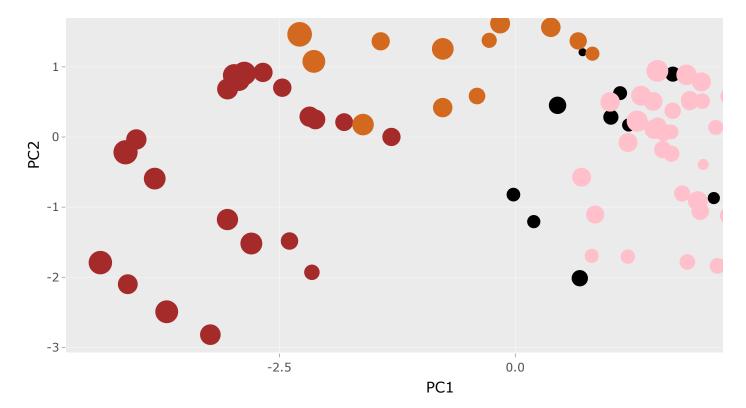
The following object is masked from 'package:ggplot2':
    last_plot

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

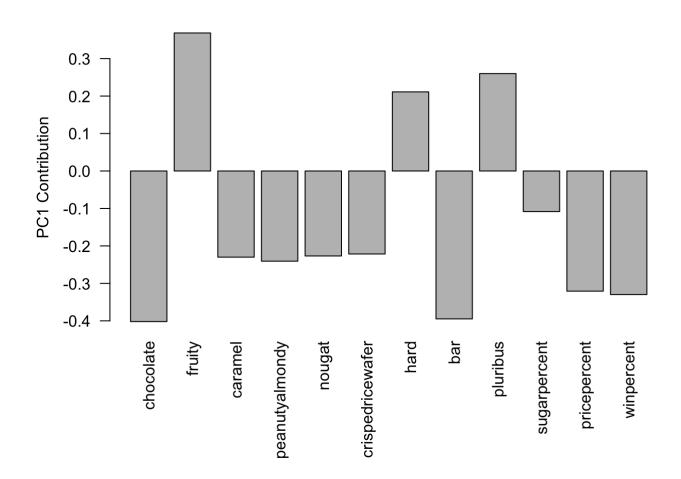
The following object is masked from 'package:graphics':
    layout
```

ggplotly(p)





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



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 strongly by PC1 in the positive direction. Yes, it makes sense to me that the fruity or hard candies coming in a bag or multiple fruity or hard candies in a box contribute positively to PC1. Since PC1 represents variation among the variables, it makes sense since some people like fruity candy and some don't, and some people like hard candy while others don't, and so on and so forth.