

Halloween Candy

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Import Candy Data

Read the csv file after downloading it to project directory.

```
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) #number of candy types
```

```
[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?

```
#skittles wildberry are my favorite candy  
candy["Skittles wildberry", ]$winpercent
```

```
[1] 55.1037
```

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
```

Using skim function

Install package.

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

Table 1: Data summary

Name	candy
Number of rows	85
Number of columns	12
Column type frequency:	

Table 1: Data summary

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	
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?

Winpercent appears to be on a different scale from the majority of other variables in the dataset.

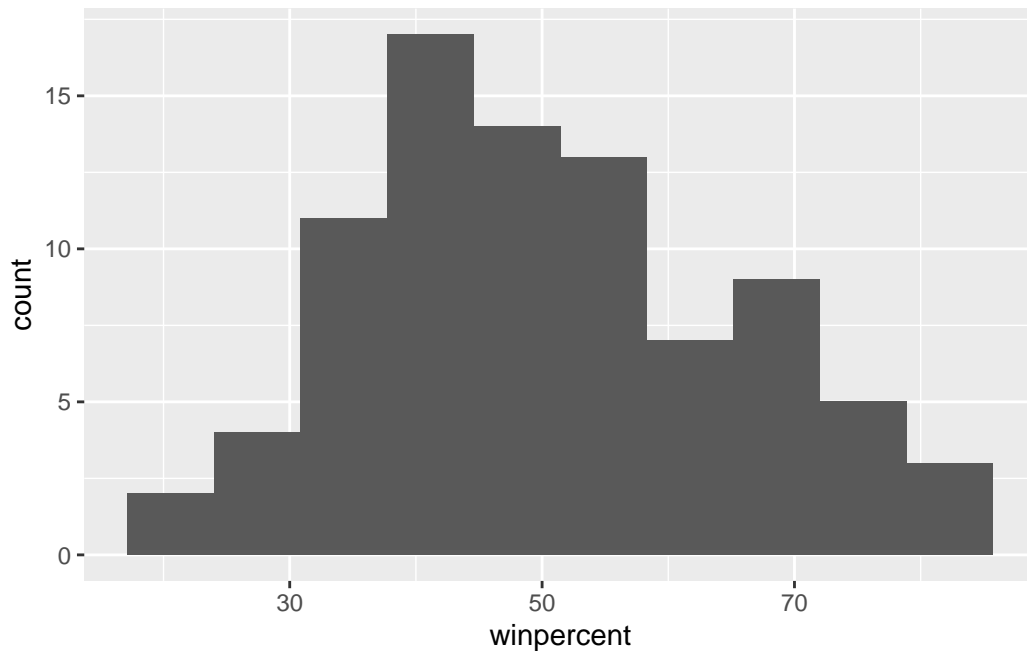
Q7. What do you think a zero and one represent for the candy\$chocolate column?

0 represents a no or not chocolate, and 1 represents yes for it is chocolate.

Histogram Plot

Q8. Plot a histogram of winpercent values

```
library(ggplot2)
ggplot(candy)+
  aes(winpercent)+
  geom_histogram(bins = 10)
```



```
#hist(candy$winpercent)
```

Q9. Is the distribution of winpercent values symmetrical?

The distribution is not symmetrical.

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

```
t.test(wp <- candy$winpercent)
```

One Sample t-test

```
data: wp <- candy$winpercent
t = 31.527, df = 84, p-value < 2.2e-16
alternative hypothesis: true mean is not equal to 0
95 percent confidence interval:
 47.14295 53.49058
sample estimates:
mean of x
 50.31676
```

```
median(wp)
```

```
[1] 47.82975
```

The center of distribution, the mean, is above 50.

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

```
mean(wp[as.logical(candy$chocolate)]) > mean(wp[as.logical(candy$fruity)])
```

```
[1] TRUE
```

Chocolate candy is higher ranked than fruity.

Q12. Is this difference statistically significant?

```
t.test(wp[as.logical(candy$chocolate)], wp[as.logical(candy$fruity)])
```

Welch Two Sample t-test

data: wp[as.logical(candy\$chocolate)] and wp[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

p-value is 2.871e-08 which is < 0.05 so there is a statistical significance between chocolate and fruity ratings.

Overall Candy Rankings

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

```
#install.packages("dplyr")
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_ascending <- candy %>%
  arrange(winpercent)
head(candy_ascending)
```

	chocolate	fruity	caramel	peanut	almond	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
Root Beer Barrels	0	0	0		0	0

	crisped	rice	wafer	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
Root Beer Barrels		0	1	0		1	0.732	0.069

	winpercent
Nik L Nip	22.44534
Boston Baked Beans	23.41782
Chiclets	24.52499
Super Bubble	27.30386
Jawbusters	28.12744
Root Beer Barrels	29.70369

The five least like candies are the first 5 rows of `candy_ascending`.

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

```
candy_descending <- candy %>%
  arrange(desc(winpercent))
head(candy_descending)
```

	chocolate	fruity	caramel	peanut	almond	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

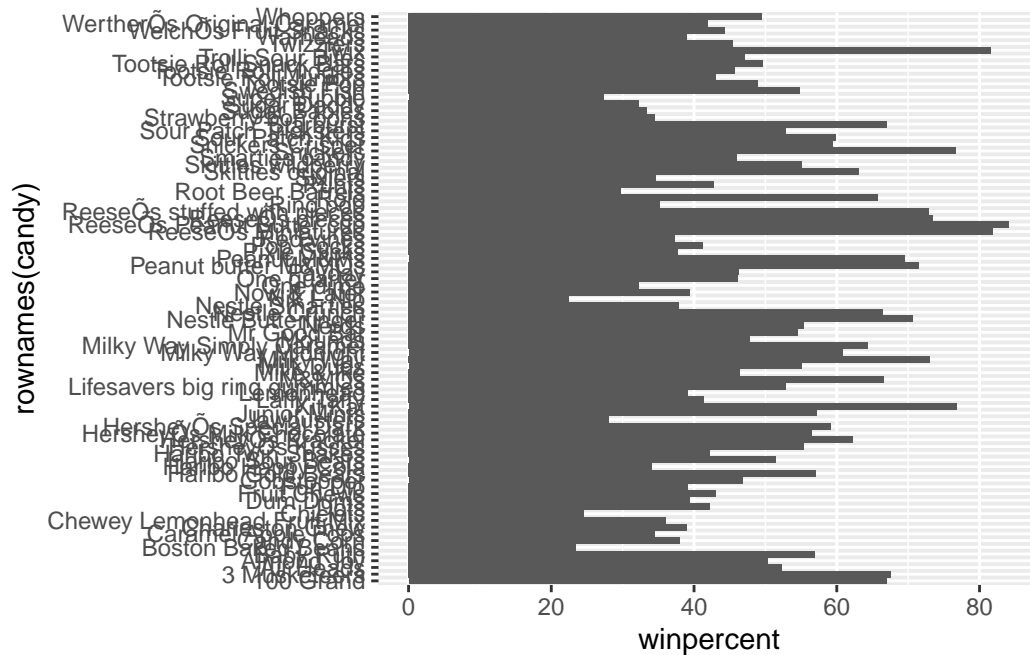
	crisp	rice	wafer	hard	bar	pluribus	sugar
Reese's Peanut Butter cup		0	0	0		0	0.720
Reese's Miniatures		0	0	0		0	0.034
Twix		1	0	1		0	0.546
Kit Kat		1	0	1		0	0.313
Snickers		0	0	1		0	0.546
Reese's pieces		0	0	0		1	0.406

	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	
Reese's pieces	0.651	73.43499	

The top 5 all time favorites are the first 5 rows in `candy_descending`.

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

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



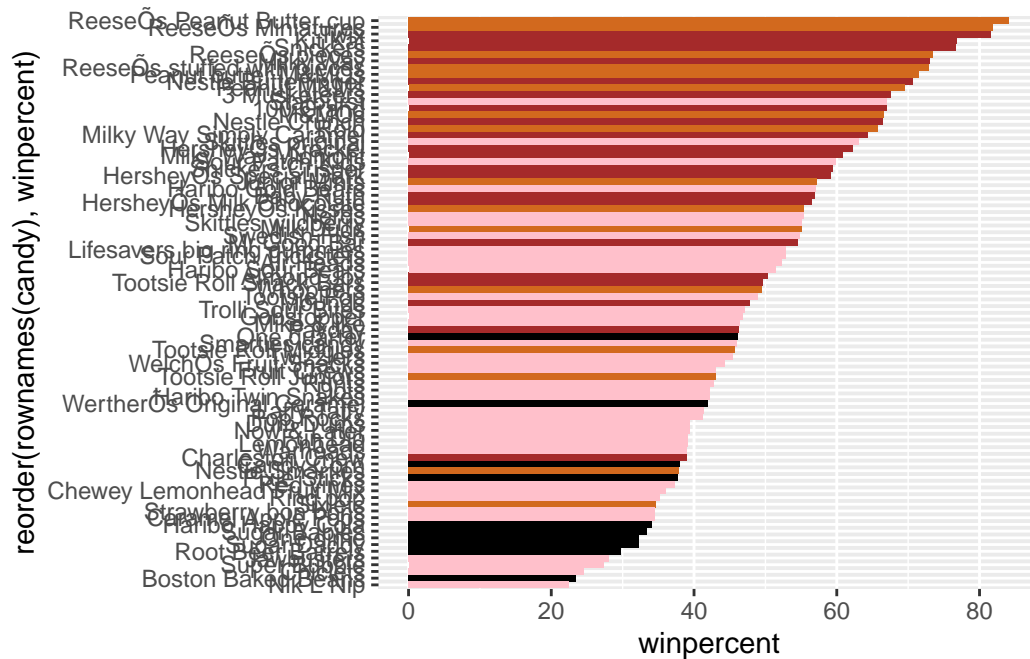
Q16. This is quite ugly, use the `reorder()` function to get the bars sorted by winpercent?

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

Now we shall add colours.

```
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 <- barplot + geom_col(fill = my_cols)
barplot
```

Q17. What is the worst ranked chocolate candy?

Sixlets is the worst ranked chocolate candy.

Q18. What is the best ranked fruity candy?

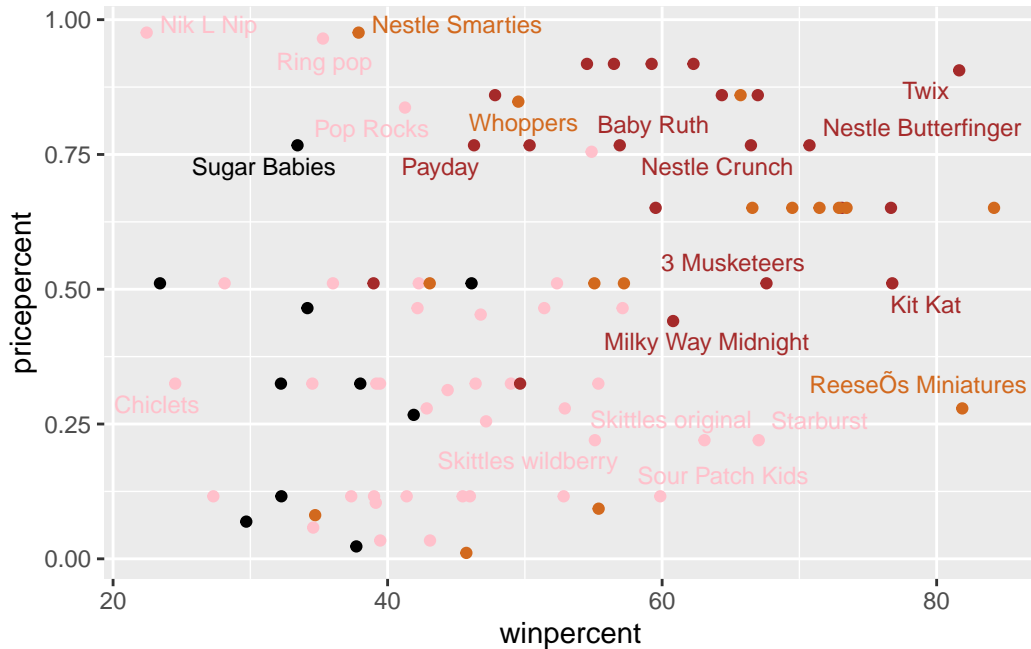
Starbursts is the best ranked fruity candy.

Looking 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: 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's Miniatures are the highest winpercent with the lowest pricepercent

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

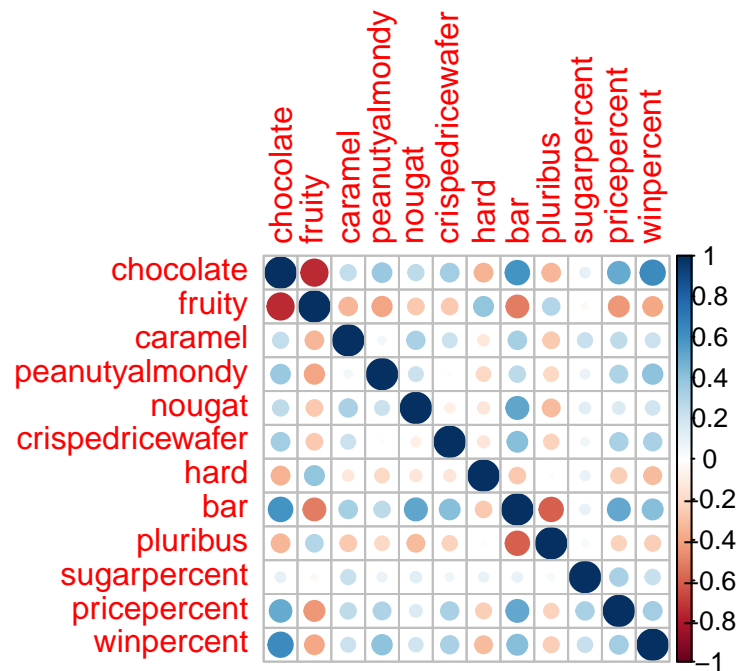
Nik L Nip is the most expensive candy with the least popularity.

Correlation Structure

```
#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)?

fruity and chocolate, pluribus and bar, and all the rest that have red colored dots.

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

winpercent and chocolate.

PCA

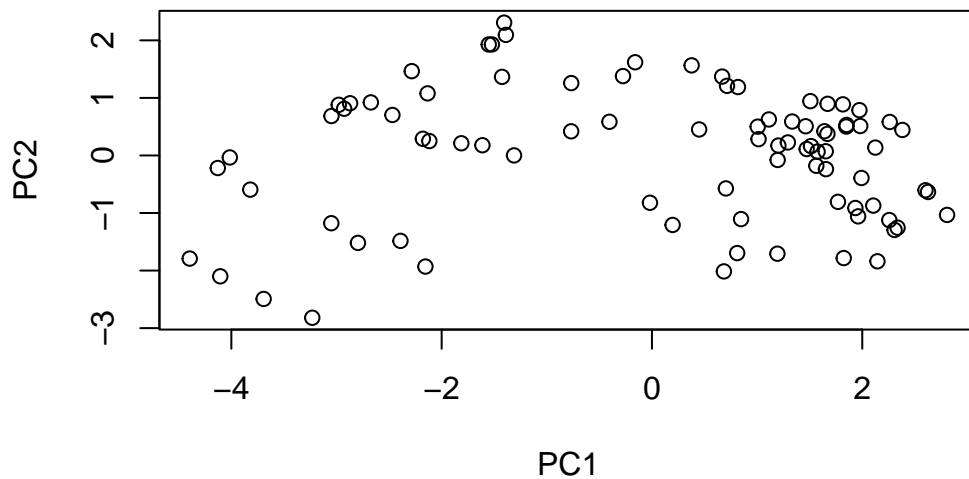
```
pca <- prcomp(candy, scale = T)
summary(pca)
```

Importance of components:

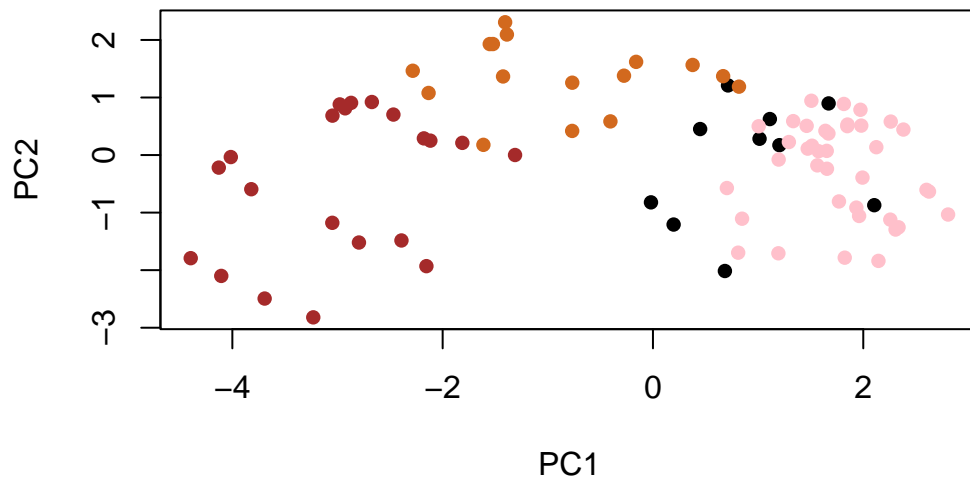
	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	2.0788	1.1378	1.1092	1.07533	0.9518	0.81923	0.81530
Proportion of Variance	0.3601	0.1079	0.1025	0.09636	0.0755	0.05593	0.05539
Cumulative Proportion	0.3601	0.4680	0.5705	0.66688	0.7424	0.79830	0.85369

	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(pca$x[,1:2])
```



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

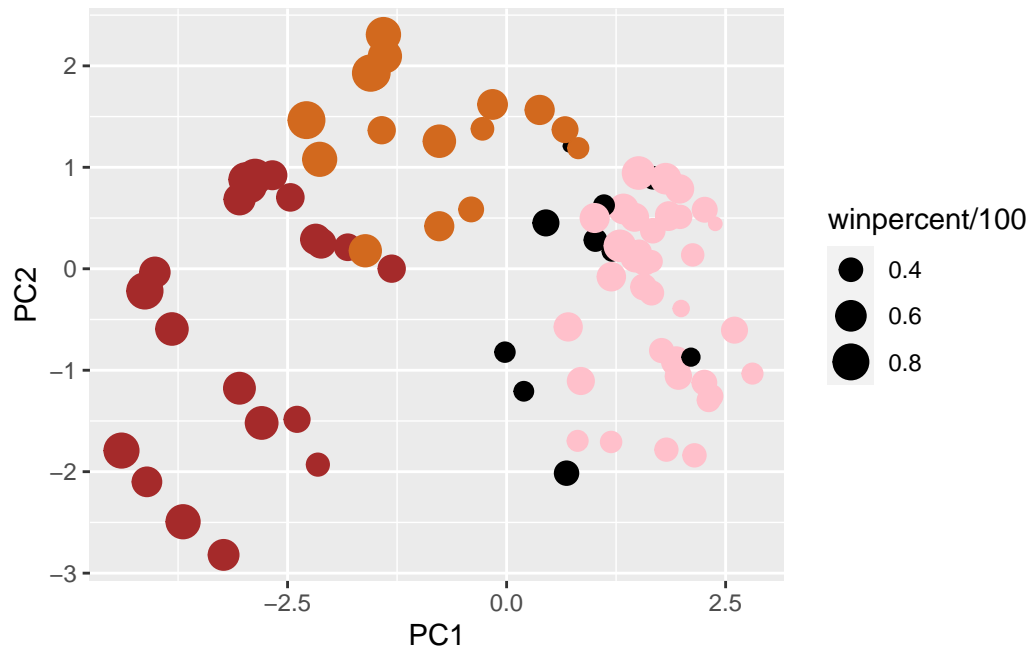


Improve the plot.

```
# 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



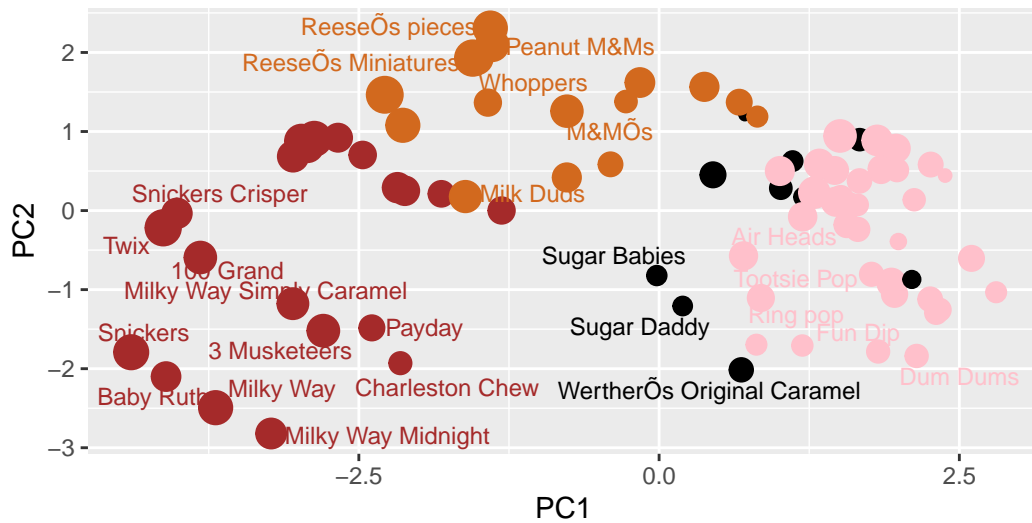
We can use ggrepel to label the ggplot.

```
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: 60 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

How about we make an interactive plot.

```
#install.packages("plotly")
library(plotly)
```

Attaching package: 'plotly'

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

last_plot

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

filter

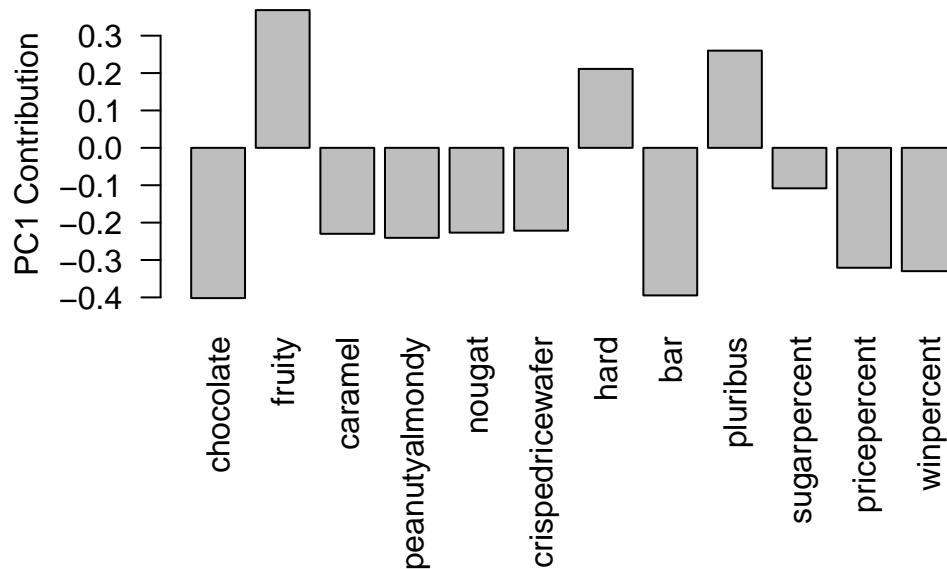
The following object is masked from 'package:graphics':

layout

```
#ggplotly(p) #interactive plot cant be rendered as pdf
```

PC1 contributions plot.

```
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. These make sense because these categories are the most distinct so they are the best attributes to cluster different candies because candies within these categories have the least overlap into other candy types.