

Class 10: Halloween Mini-Project

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Importing the Data

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

candy = read.csv(candy_file, row.names=1)
head(candy)
```

	chocolate	fruity	caramel	peanut	almond	nougat	crisp	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	66.97173	
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	

```
nrow(candy)
```

[1] 85

Q1. How many different candy types are in this dataset?

There are 85 different brands of candies within this dataset. >Q2. How many fruity candy types are in the dataset?

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

```
[1] 38
```

There are 38 fruity candy types.

```
candy["Warheads", ]$winpercent
```

```
[1] 39.0119
```

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

```
[1] 76.7686
```

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

```
[1] 49.6535
```

Q3. What is your favorite candy in the dataset and what is its winpercent value?

My favorite candy in the dataset is Warheads, and its winpercent is 39.01. >Q4. What is the winpercent value for “Kit Kat”?

76.77% >Q5. What is the winpercent value for “Tootsie Roll Snack Bars”?

49.65%

```
library("skimr")
skim(candy)
```

Table 1: Data summary

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

Table 1: Data summary

Group variables	None
-----------------	------

Variable type: numeric

skim_variable	n_missing	complete	rat	mean	sd	p0	p25	p50	p75	p100	hist
chocolate	0	1	0.44	0.50	0.00	0.00	0.00	0.00	1.00	1.00	
fruity	0	1	0.45	0.50	0.00	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	0.00	1.00	
peanutyalmondy	0	1	0.16	0.37	0.00	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	0.00	1.00	
crispedricewafer	0	1	0.08	0.28	0.00	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	0.00	1.00	
bar	0	1	0.25	0.43	0.00	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	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 row has values that go above 1 whereas the other rows range from only 0 to 1.

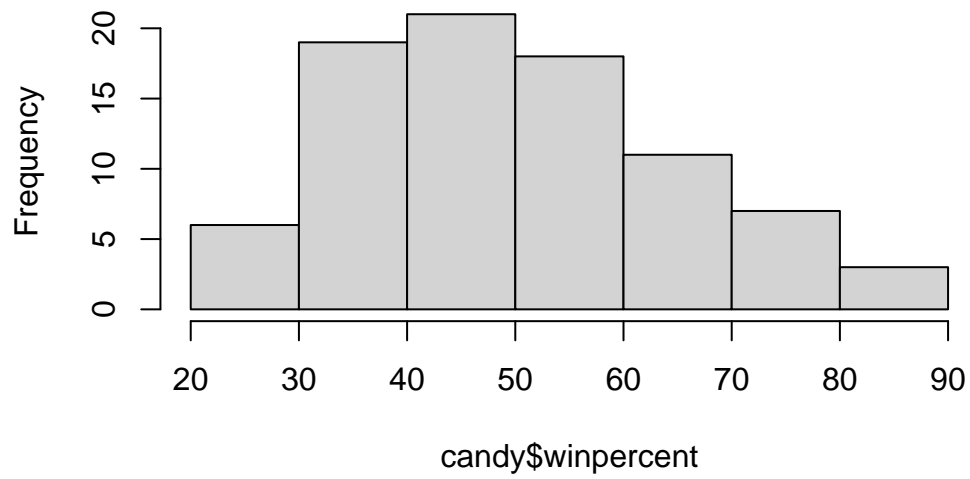
Q7. What do you think a zero and one represent for the candy\$chocolate column?

I believe a 0 represents a FALSE value where the candy isn't chocolate. A 1 represents TRUE where the candy is chocolate.

Q8. Plot a histogram of winpercent values

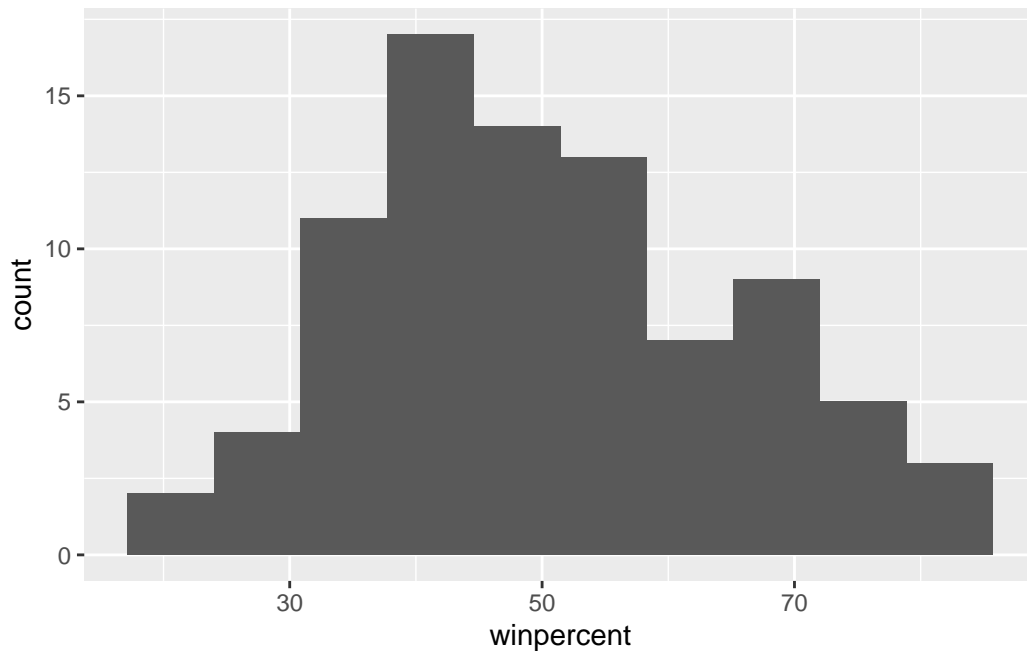
```
hist(candy$winpercent)
```

Histogram of candy\$winpercent



```
library(ggplot2)

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



Q9. Is the distribution of winpercent values symmetrical?

No it is skewed right towards higher winpercents.

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

```
median(candy$winpercent)
```

```
[1] 47.82975
```

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

```
[1] 50.31676
```

The center of the distribution is above 50%.

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

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

```
[1] 60.92153
```

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

```
[1] 44.11974
```

The average for chocolate candy is higher.

Q12. Is this difference statistically significant?

```
t.test(candy$winpercent[as.logical(candy$fruity)], candy$winpercent[as.logical(candy$choco
```

Welch Two Sample t-test

```
data: candy$winpercent[as.logical(candy$fruity)] and candy$winpercent[as.logical(candy$choco
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:
 -22.15795 -11.44563
sample estimates:
mean of x mean of y
 44.11974  60.92153
```

The difference between the averages of chocolate and fruity candy is statistically significant since the p-value obtained is 2.871e-08 which is much smaller than 0.05.

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

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

	chocolate	fruity	caramel	peanutyalmondy	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	
	crispedricewafer	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	
	winpercent						
Nik L Nip	22.44534						
Boston Baked Beans	23.41782						
Chiclets	24.52499						
Super Bubble	27.30386						
Jawbusters	28.12744						

The five least liked candies by winpercent are Nik L 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?

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

	chocolate	fruity	caramel	peanutyalmondy	nougat		
ReeseÕs Peanut Butter cup	1	0	0		1	0	
ReeseÕs 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	
	crispedricewafer	hard	bar	pluribus	sugarpercent		
ReeseÕs Peanut Butter cup		0	0	0	0	0.720	
ReeseÕs 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	
	pricepercent	winpercent					

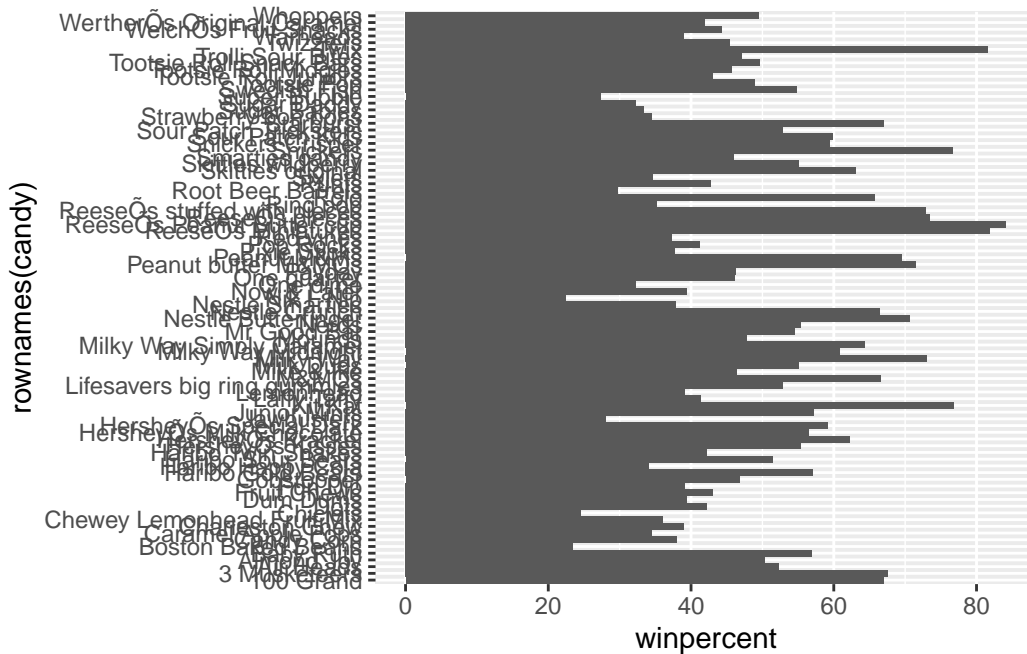
Reese's Peanut Butter cup	0.651	84.18029
Reese's Miniatures	0.279	81.86626
Twix	0.906	81.64291
Kit Kat	0.511	76.76860
Snickers	0.651	76.67378

The 5 most liked candies by winpercent 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.

```
library(ggplot2)

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

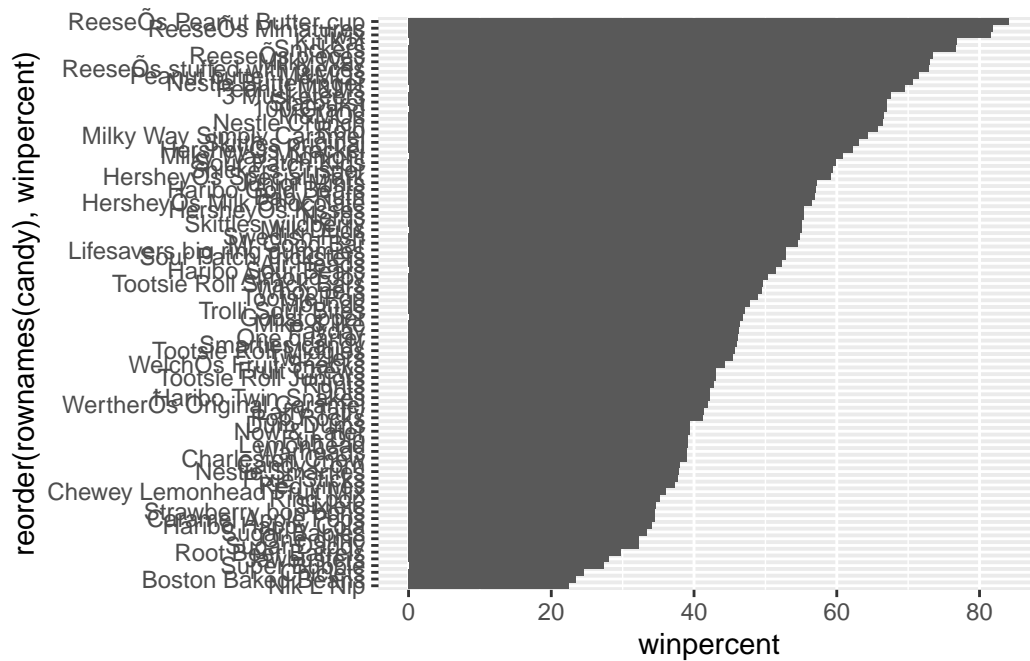


Q.16 This is quite ugly, use the reorder() function to get the bars sorted by win-percent?

```
library(ggplot2)
```



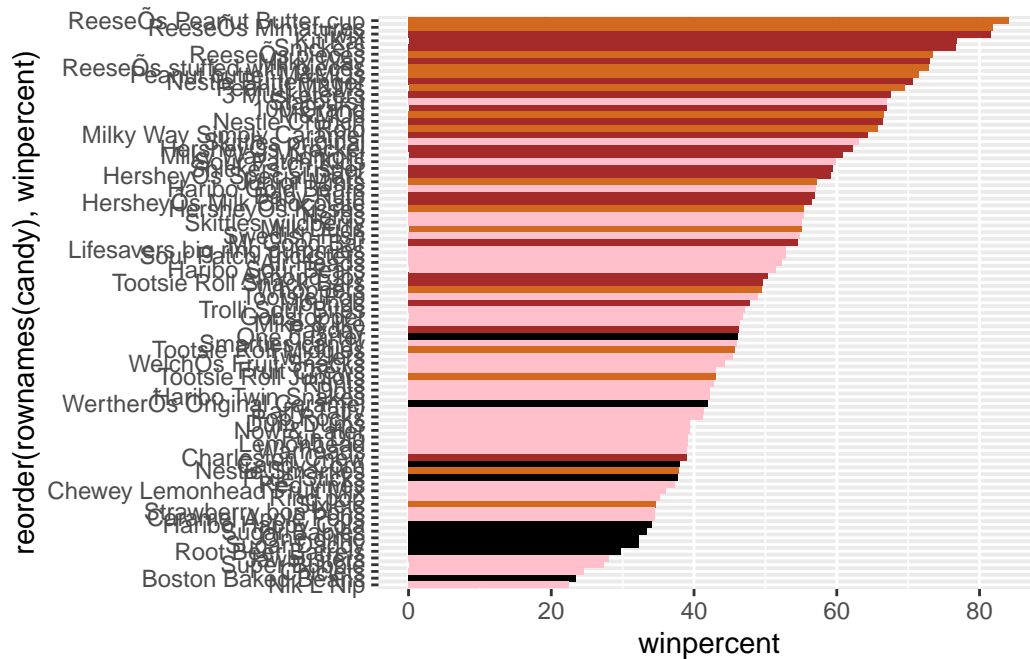
```
ggplot(candy) +
  aes(winpercent, reorder(rownames(candy), winpercent)) + geom_col()
```



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"

library(ggplot2)
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?

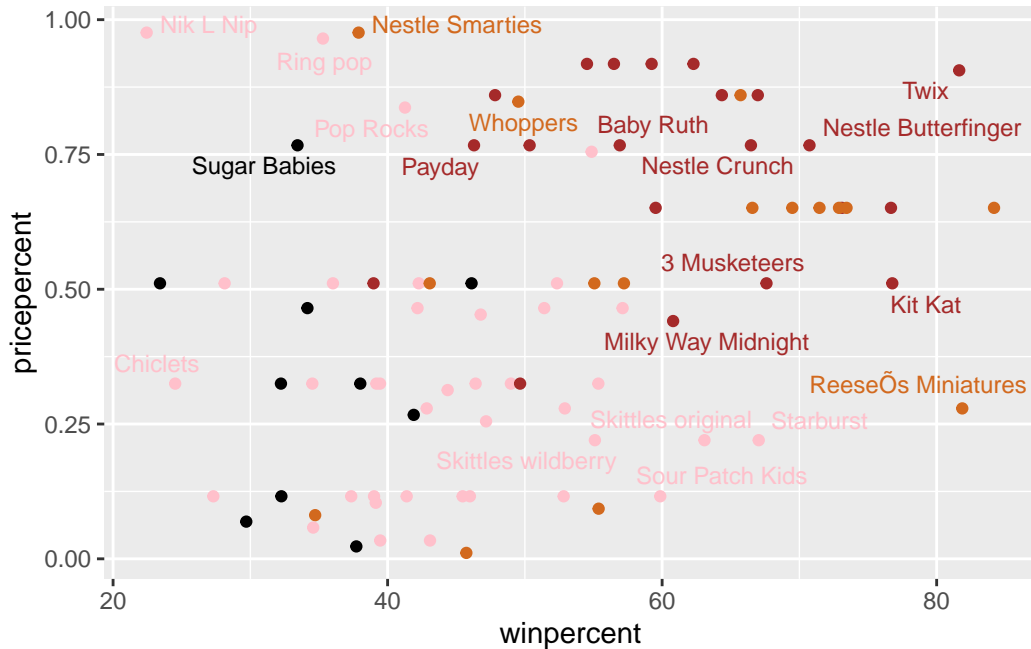
Sixlets > Q18. What is the best ranked fruity candy?

Starburst

```
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

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 )
```

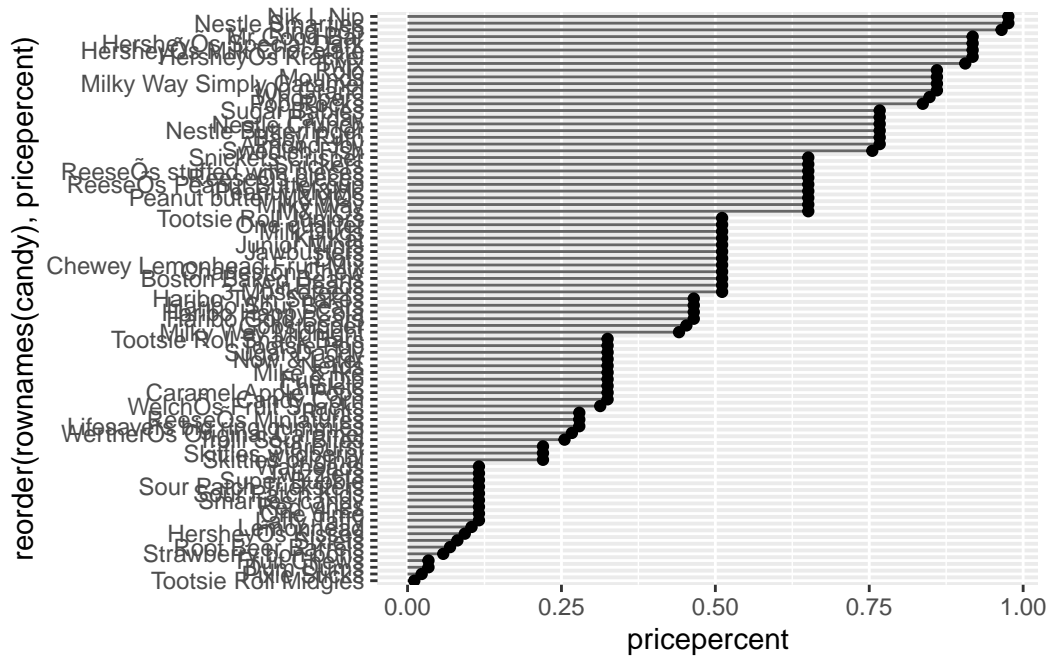
	pricepercent	winpercent
Nik L Nip	0.976	22.44534
Nestle Smarties	0.976	37.88719
Ring pop	0.965	35.29076
Hershey's Krackel	0.918	62.28448
Hershey's Milk Chocolate	0.918	56.49050

Nik L Nip is in the top 5 most expensive and is the least liked at about 22.45%.

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()`.

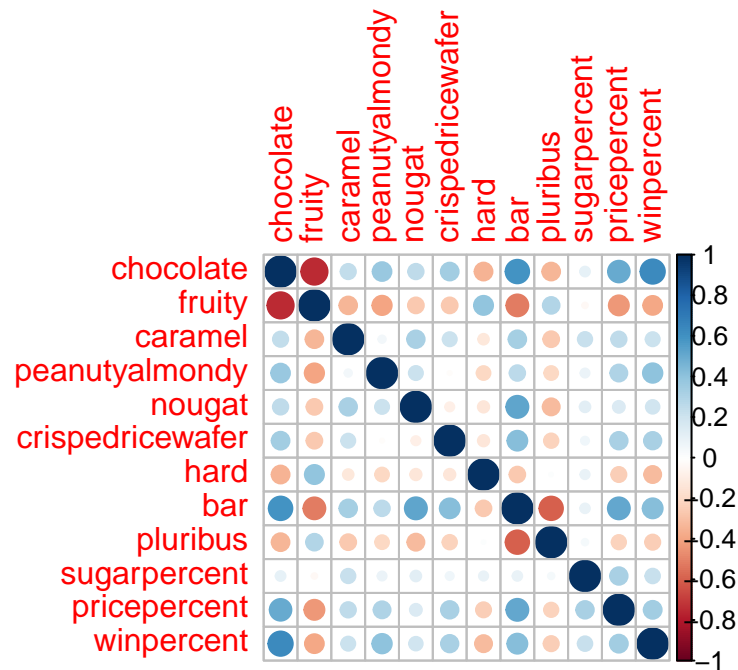
```
# Make a lollipop chart of pricepercent
ggplot(candy) +
  aes(pricepercent, reorder(rownames(candy), pricepercent)) +
  geom_segment(aes(yend = reorder(rownames(candy), pricepercent),
                    xend = 0), col="gray40") +
  geom_point()
```



```
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, fruity and bar, etc. >Q23. Similarly, what two variables are most positively correlated?

Chocolate and winpercent.

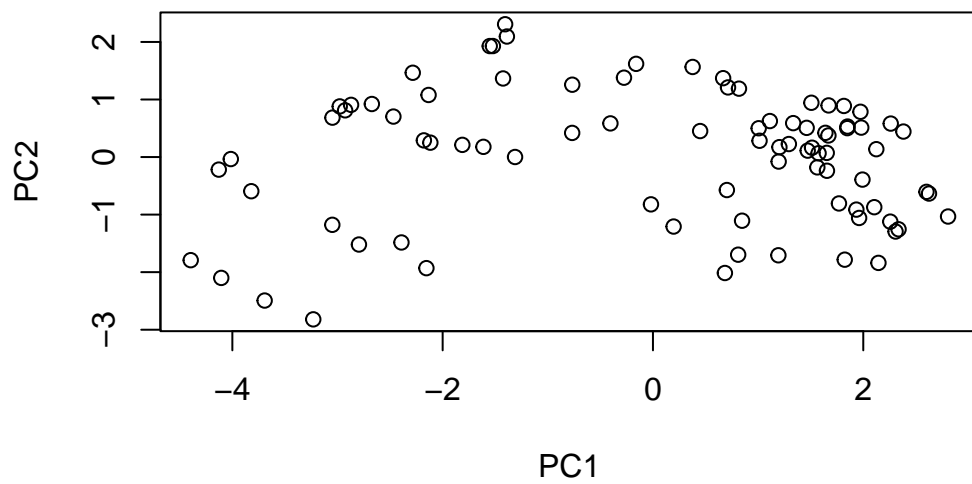
```
pca <- prcomp(candy, scale = TRUE)
summary(pca)
```

Importance of components:

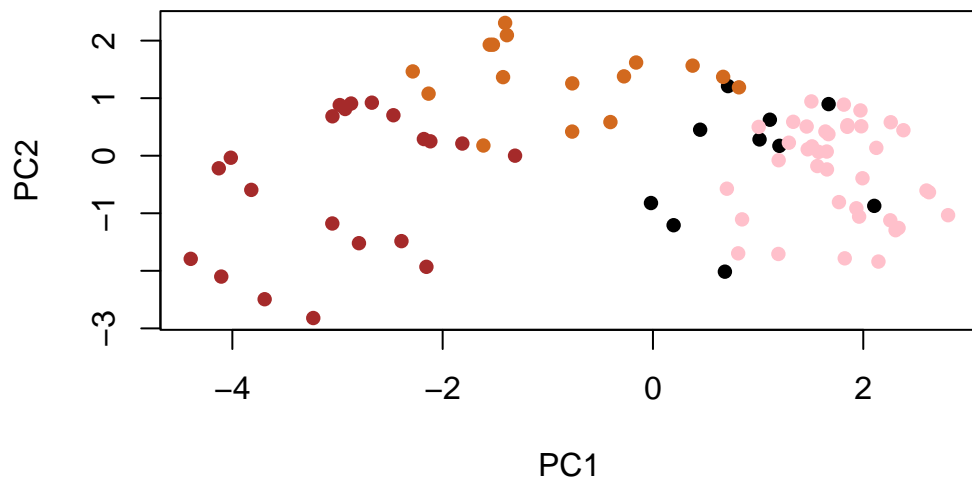
	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	2.0788	1.1378	1.1092	1.07533	0.9518	0.81923	0.81530
Proportion of Variance	0.3601	0.1079	0.1025	0.09636	0.0755	0.05593	0.05539
Cumulative Proportion	0.3601	0.4680	0.5705	0.66688	0.7424	0.79830	0.85369

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

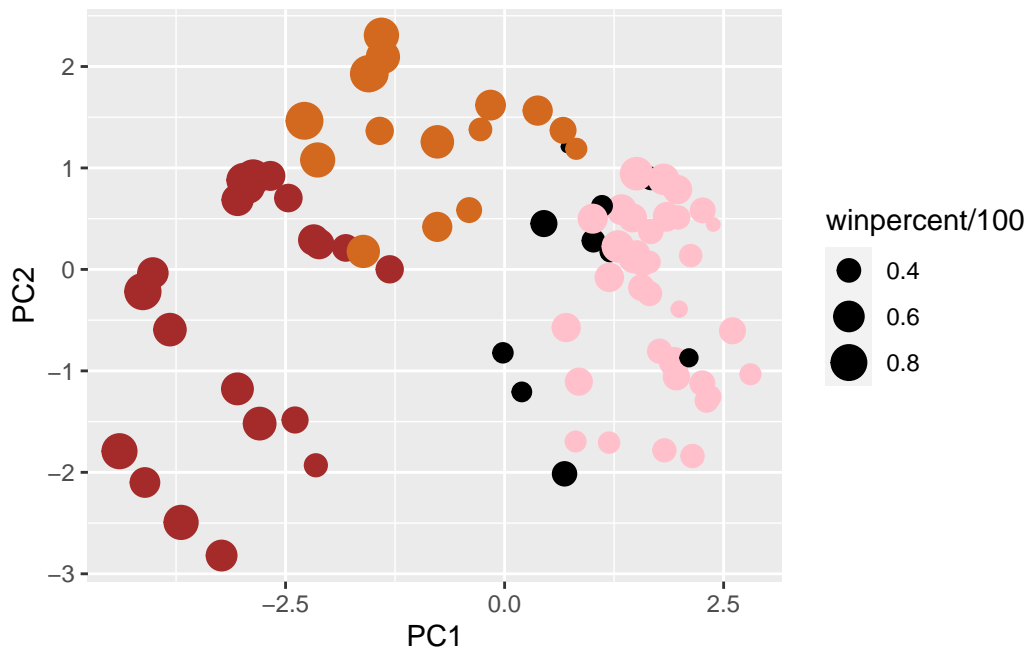


```
plot(pca$x[,1:2], col=my_cols, pch=16)
```



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



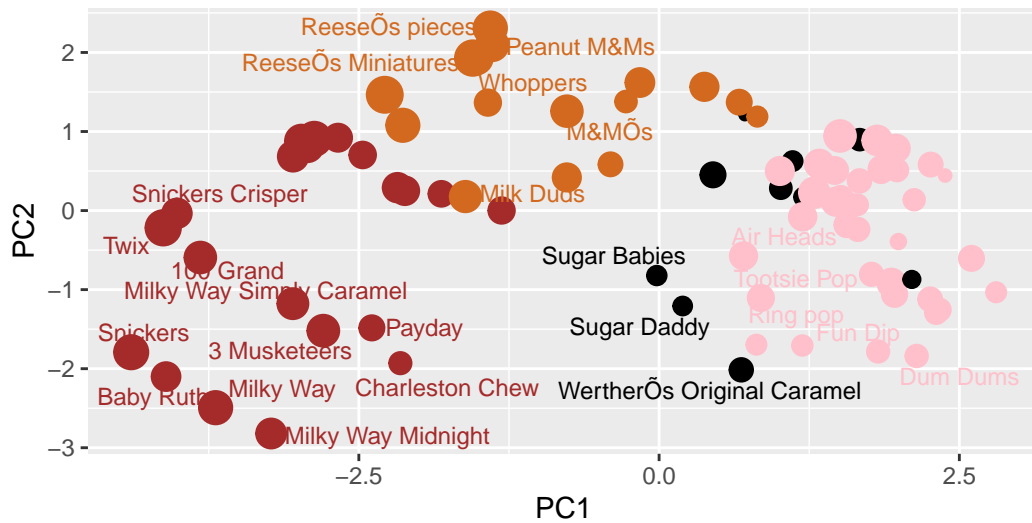
```
library(ggrepel)

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: 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

```
library(plotly)
```

Attaching package: 'plotly'

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

last_plot

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

filter

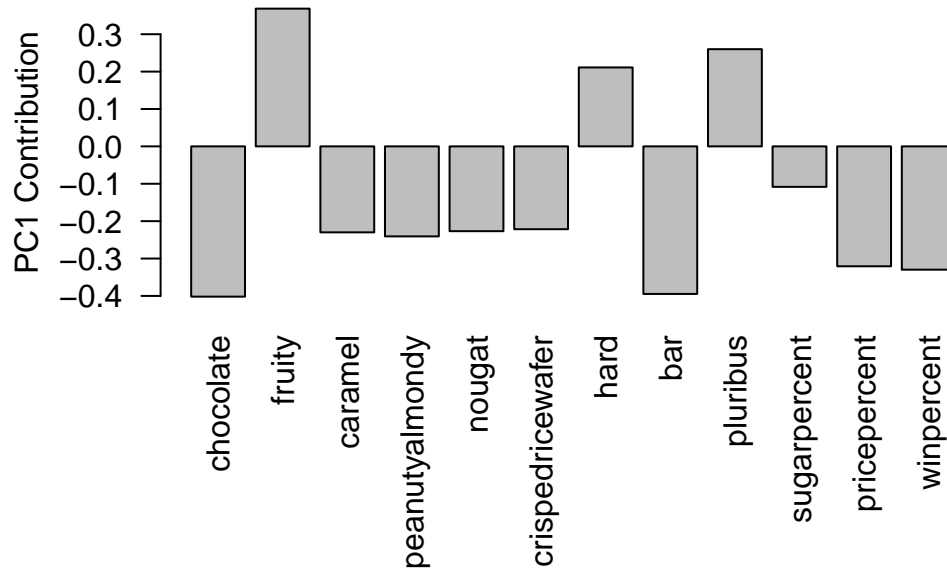
The following object is masked from 'package:graphics':

layout

```
#ggplotly was not rendering properly in pdf format, only works in html format  
#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. These make sense since these groups are the most unique in comparison to the other groups. Fruity, hard, and pluribus don't share as much with the other groups as others do for example chocolate which shares a lot with many groups.