

Class 8: Halloween Mini-Project

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Importing the Candy Dataset

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

	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	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

What's in the dataset?

```
nrow(candy) #rows in the dataset (candy types)
```

[1] 85

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

There are 85 different candy types in the dataset

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

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

```
[1] 38
```

There are 38 fruity candy types in the dataset.

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

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

```
[1] 69.48379
```

Win % of Peanut M&Ms is 69.5%

Q4: What is the winpercent value for “Kit Kat”?

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

```
[1] 76.7686
```

Win % of Kit Kat is 76.8%.

Q5: What is the winpercent value for “Tootsie Roll Snack Bars”?

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

```
[1] 49.6535
```

Win % of Tootsie Roll Snack Bars is 49.7%.

Using the **skim()** function to get an overview of a given dataset...

```
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	
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 values are in whole percentages, while others are in proportions.

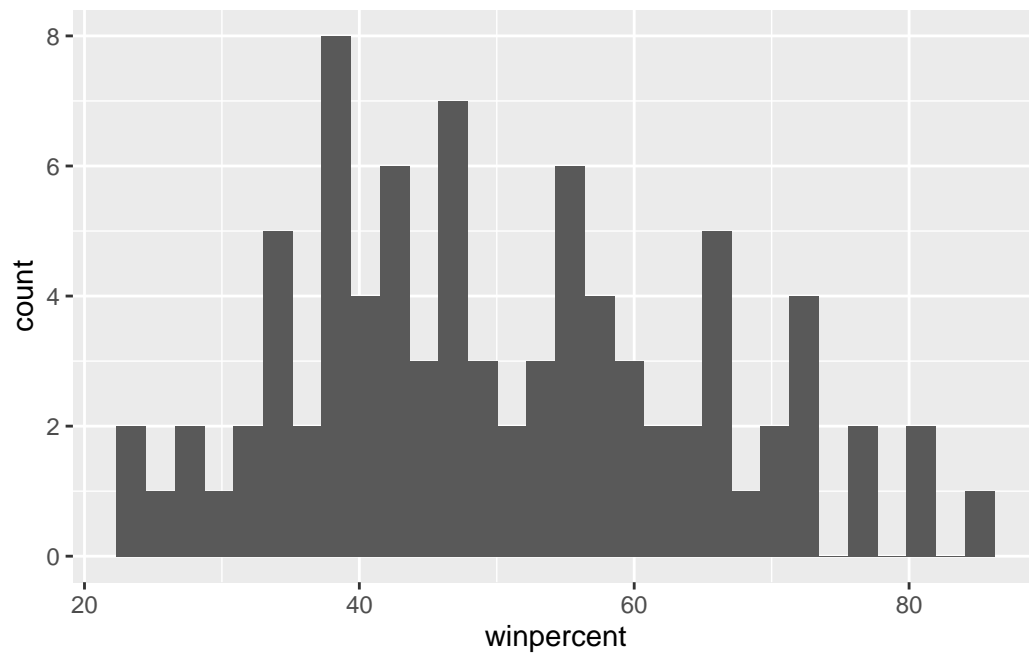
Q7: What do you think a zero and one represent for the candy\$chocolate column?

Whether or not the given candy from a given row is (one) and isn't (zero) under the chocolate category.

Q8: Plot a histogram of winpercent values

```
library(ggplot2)
ggplot(candy, aes(x = winpercent)) +
  geom_histogram()
```

``stat_bin()`` using ``bins = 30``. Pick better value with ``binwidth``.



Q9: Is the distribution of winpercent values symmetrical?

No, it appears to have a right-skew.

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

Below 50%.

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

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

```
[1] 60.92153
```

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

```
[1] 44.11974
```

Chocolate candy is ranked higher than fruit candy!

Q12: Is this difference statistically significant?

```
#assign 'choc' object as vector of win percentages of chocolate candies
choc <- candy$winpercent[as.logical(candy$chocolate)]
#assign 'fruit' object as vector of win percentages of fruity candies
fruity <- candy$winpercent[as.logical(candy$fruity)]

t.test(choc, fruity)
```

Welch Two Sample t-test

```
data:  choc and 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
```

The difference is statistically significant!

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	peanut	almondy	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

	crisp	rice	wafer	hard	bar	pluribus	sugar	percent	price	percent
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

Nik L Nip, Boston Baked Beans, Chiclets, Super Bubble, Jawbusters are the 5 least liked candies!

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

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

	chocolate	fruity	caramel	peanut	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

	crisp	rice	wafer	hard	bar	pluribus	sugar	percent
Reese's Peanut Butter cup				0	0	0		0.720
Reese's Miniatures				0	0	0		0.034

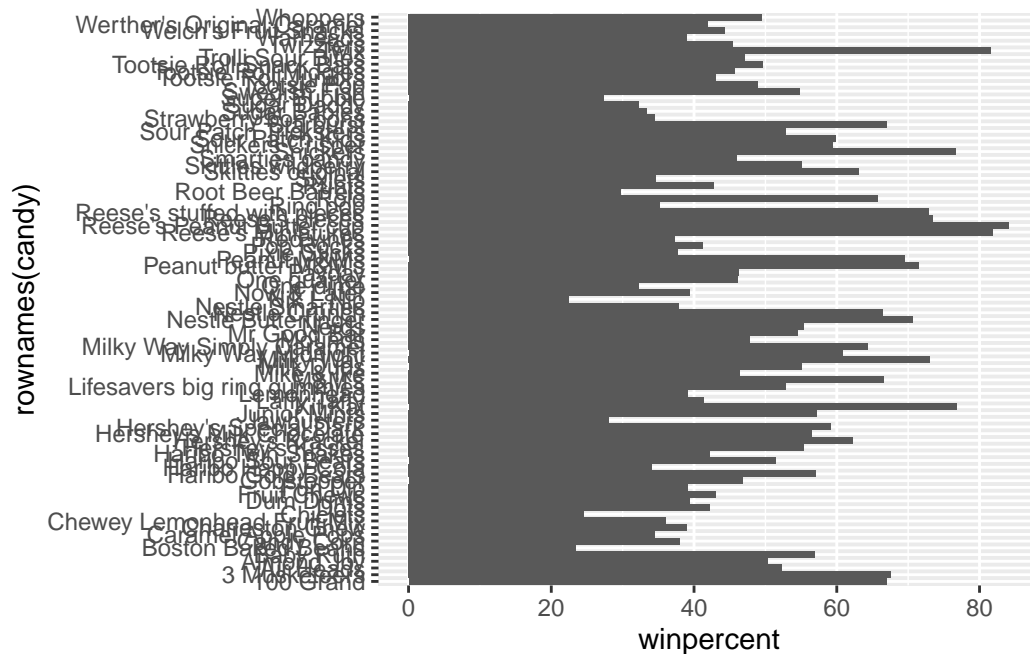
Twix	1	0	1	0	0.546
Kit Kat	1	0	1	0	0.313
Snickers	0	0	1	0	0.546

	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

Reese's Peanut Butter cup, Reese's Miniatures, Twix, Kit Kat, Snickers are the top 5 favorites!

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

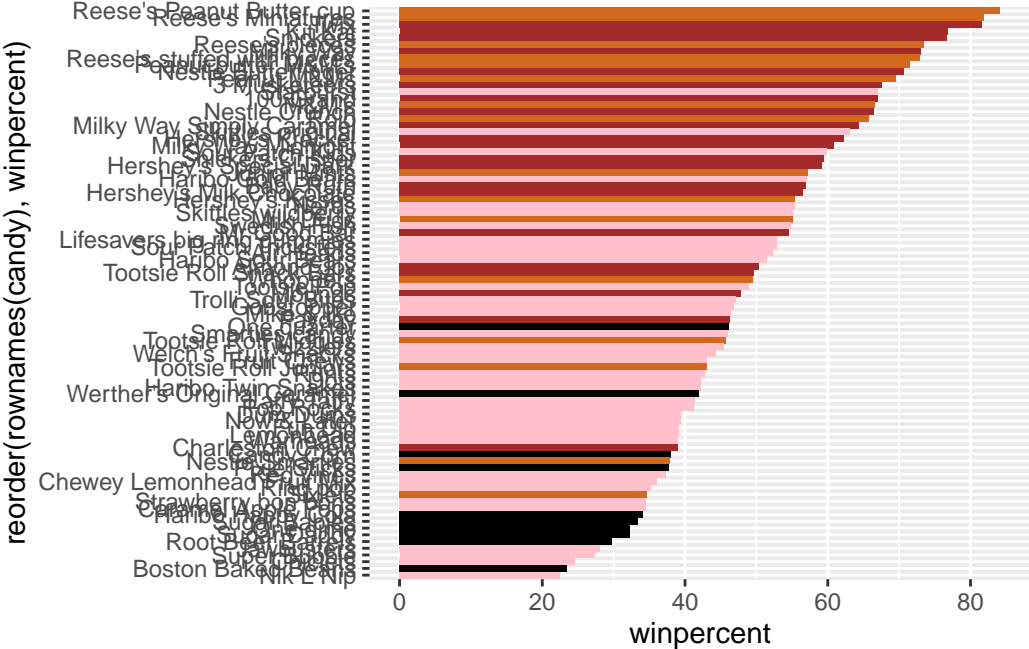
```
ggplot(candy, aes(winpercent, rownames(candy))) +  
  geom_col()
```



Reordering by winpercent:

```
#setting color parameters
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)
```



Q17: What is the worst ranked chocolate candy?

Nik L Nip

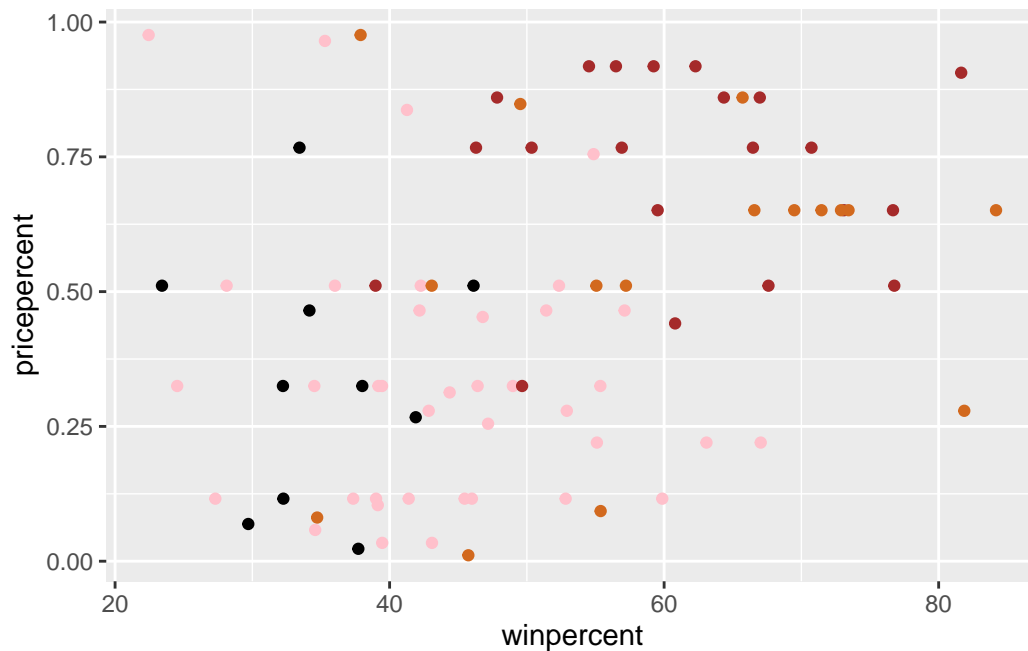
Q18: What is the best ranked fruity candy?

Starburst

Exploring 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)
```

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?

```
#ordering pricepercent to see most/least expensive candy
ord <- order(candy$pricepercent, decreasing = FALSE)
head(candy[ord, c(11,12)], n = 5)
```

	pricepercent	winpercent
Tootsie Roll Midgies	0.011	45.73675

Pixie Sticks	0.023	37.72234
Dum Dums	0.034	39.46056
Fruit Chews	0.034	43.08892
Strawberry bon bons	0.058	34.57899

Tootsie Roll Midgies

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

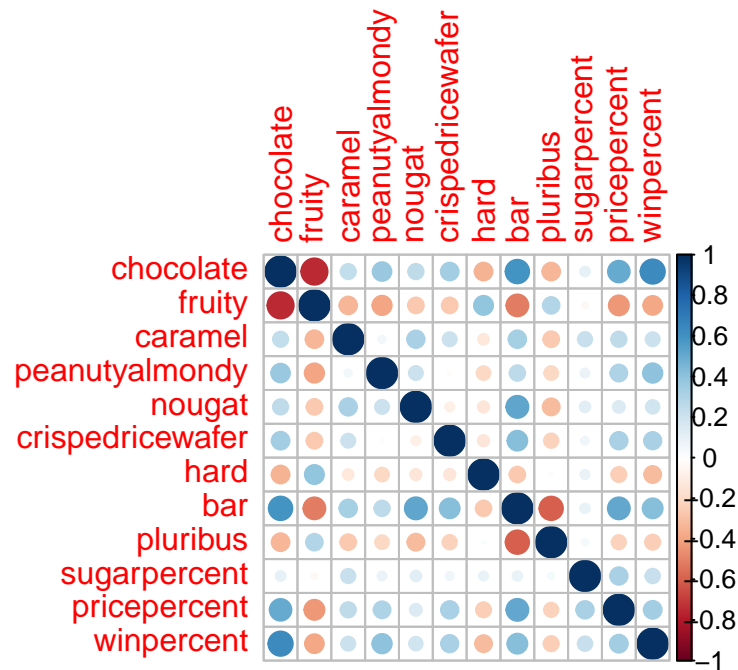
Nik L Nip is the most expensive and least popular of the dataset.

Correlation Plot

```
library(corrplot)
```

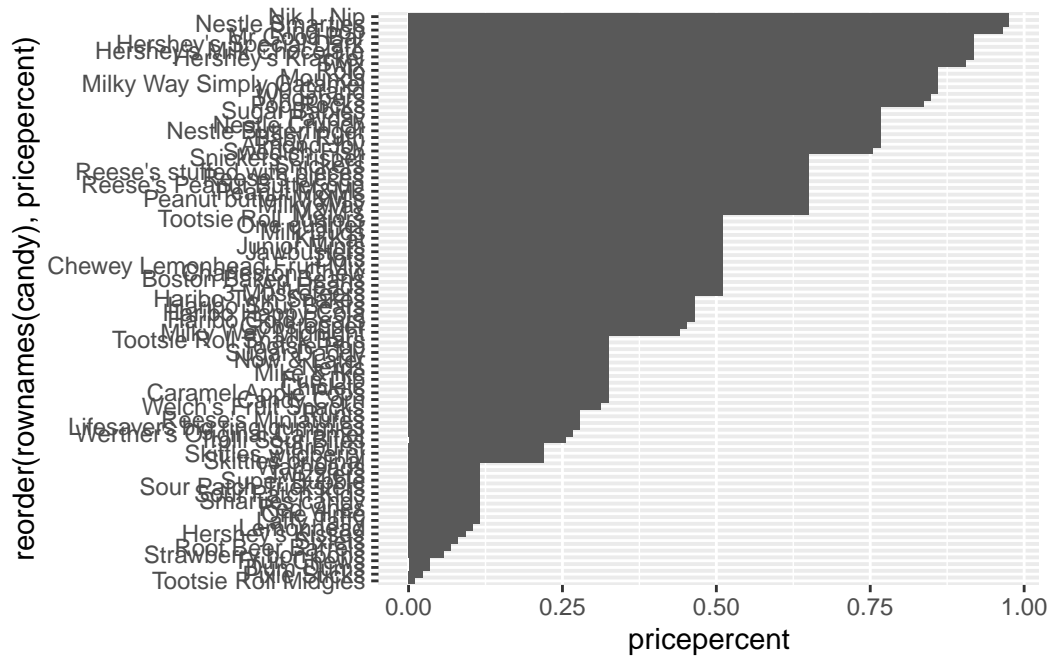
corrplot 0.92 loaded

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

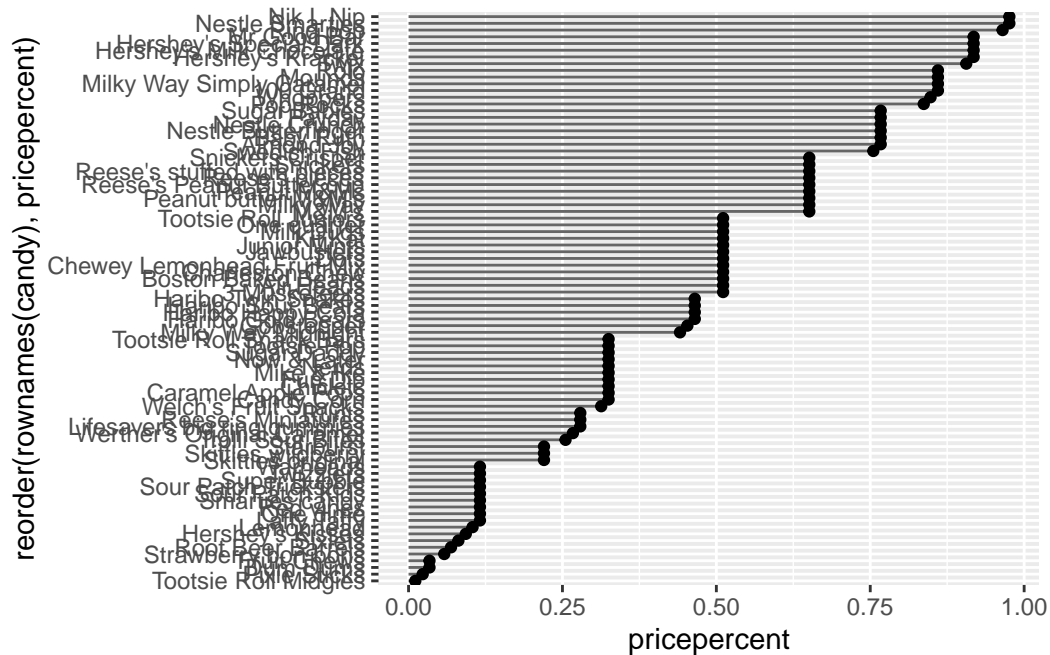


Q21: Make a barplot again with `geom_col()` this time using `pricepercent` and then improve this step by step, first ordering the x-axis by value and finally making a so called “dot chat” or “lollipop” chart by swapping `geom_col()` for `geom_point()` + `geom_segment()`.

```
#barplot of pricepercent
ggplot(candy) +
  aes(pricepercent, reorder(rownames(candy), pricepercent)) +
  geom_col()
```



```
#making it prettier: 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()
```



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 winpercent

PCA

```
#setting up PCA object
pca <- prcomp(candy, scale. = TRUE)
summary(pca)
```

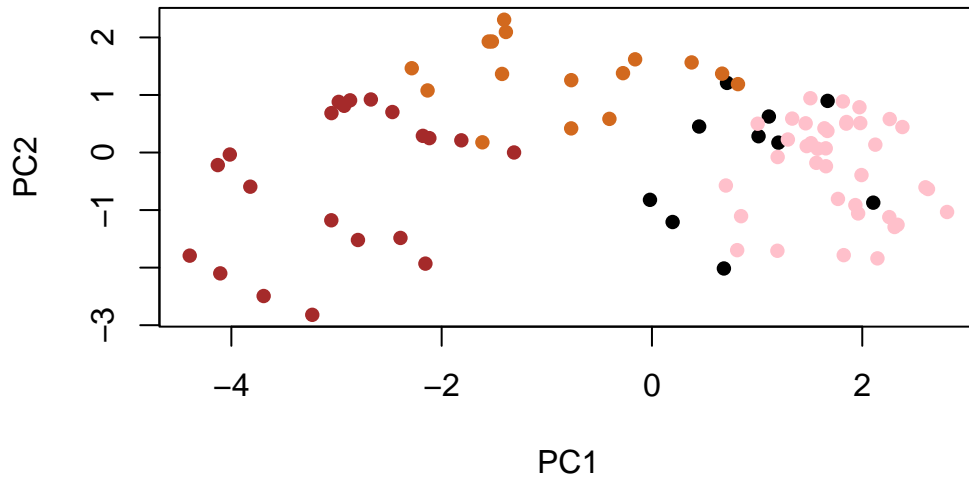
Importance of components:

	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

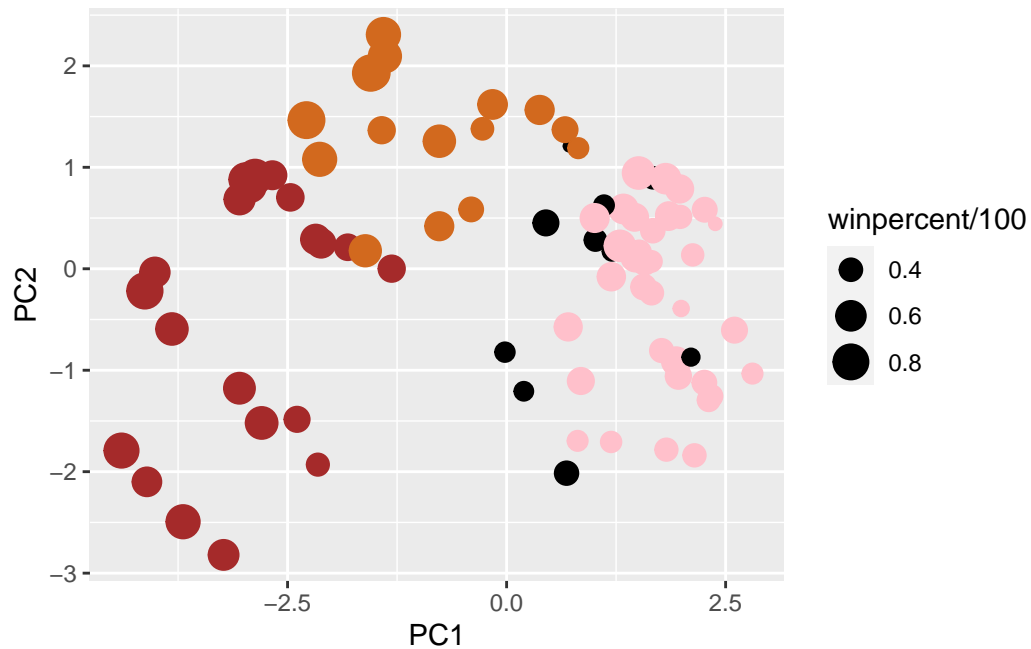
```
plot(pca$x[, 1:2], col = my_cols, pch = 16)
```



```
# 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
```



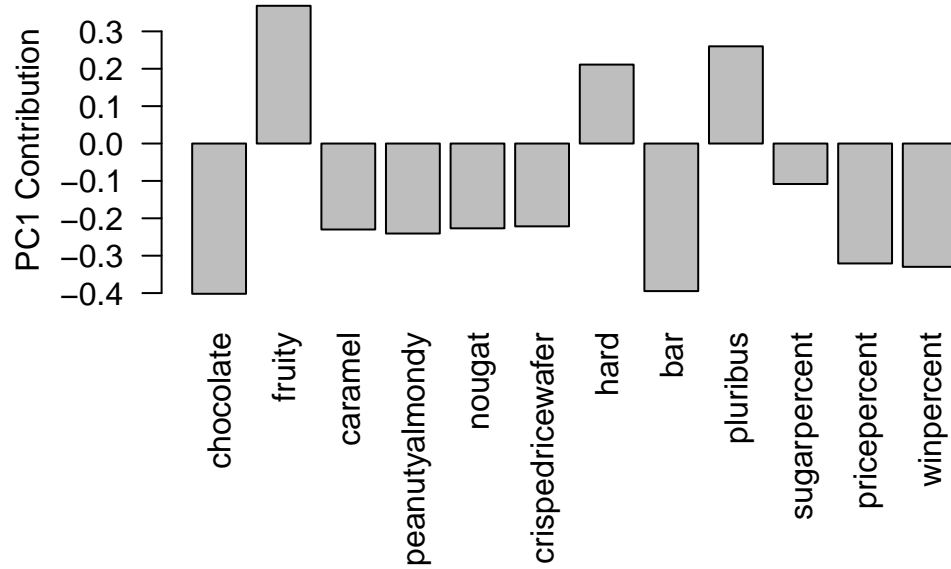
```
#putting labels on each of the dots; making it prettier
#library(ggrepel)

#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")

#passing plot 'p' to plotly to get interactive plot (removes overlap)
#library(plotly)
#ggplotly(p)
```

PC Loadings

```
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, pluribus - yes, it makes sense (they normally come in this form)