

Class 09 Halloween Candy Mini Proj

Dylan Mullaney (A16869792)

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

Importing candy data

For a txt file, could use read.table

```
url <- "https://raw.githubusercontent.com/fivethirtyeight/data/master/candy-power-ranking/candy.csv"
candy <- read.csv(url, 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

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

```
nrow(candy)
```

[1] 85

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

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

[1] 38

```
#This is a count value, not a percent
```

Winpercent refers to the likelihood a candy will be chosen as the winner.

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

```
candy["Haribo Happy Cola",] $winpercent
```

[1] 34.15896

```
#This is a percent value, not a count
```

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

How many candies are chocolate?

```
sum(candy$chocolate)
```

[1] 37

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 different scale.

```
#Install package in R brain
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	ratio	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		

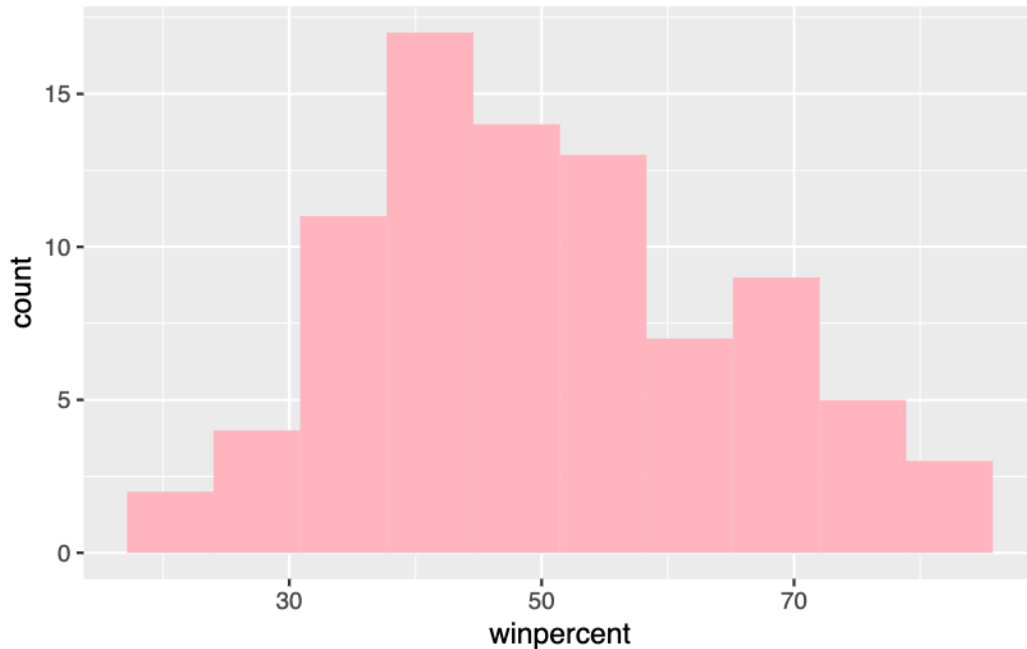
Q7. What do you think a zero and one represent for the candy\$chocolate column?

This represents a candy that doesn't match this identity, ie it isn't chocolate.

Q8. Plot a histogram of winpercent values

```
library(ggplot2)

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



Q9. Is the distribution of winpercent values symmetrical?

No, based on the shape of the histogram.

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

It appears to be below 50%. Mean is above 50, but the median is below.

```
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?

-Step 1: find all chocolate candy -Step 2: find their winpercent values -Step 3: summarize these values (mean, median) -Step 4: repeat with fruit candies -Step 5: compare the two summart values

Step 1

```
choc.inds <- candy$chocolate == 1
candy[choc.inds,]
```

	chocolate	fruity	caramel	peanut	almond	nougat
100 Grand	1	0	1		0	0
3 Musketeers	1	0	0		0	1
Almond Joy	1	0	0		1	0
Baby Ruth	1	0	1		1	1
Charleston Chew	1	0	0		0	1
Hershey's Kisses	1	0	0		0	0
Hershey's Krackel	1	0	0		0	0
Hershey's Milk Chocolate	1	0	0		0	0
Hershey's Special Dark	1	0	0		0	0
Junior Mints	1	0	0		0	0
Kit Kat	1	0	0		0	0
Peanut butter M&M's	1	0	0		1	0
M&M's	1	0	0		0	0
Milk Duds	1	0	1		0	0
Milky Way	1	0	1		0	1
Milky Way Midnight	1	0	1		0	1
Milky Way Simply Caramel	1	0	1		0	0
Mounds	1	0	0		0	0
Mr Good Bar	1	0	0		1	0
Nestle Butterfinger	1	0	0		1	0
Nestle Crunch	1	0	0		0	0
Peanut M&Ms	1	0	0		1	0
Reese's Miniatures	1	0	0		1	0
Reese's Peanut Butter cup	1	0	0		1	0
Reese's pieces	1	0	0		1	0
Reese's stuffed with pieces	1	0	0		1	0
Rolo	1	0	1		0	0
Sixlets	1	0	0		0	0
Nestle Smarties	1	0	0		0	0
Snickers	1	0	1		1	1
Snickers Crisper	1	0	1		1	0
Tootsie Pop	1	1	0		0	0
Tootsie Roll Juniors	1	0	0		0	0
Tootsie Roll Midgies	1	0	0		0	0
Tootsie Roll Snack Bars	1	0	0		0	0
Twix	1	0	1		0	0
Whoppers	1	0	0		0	0
	crisped	rice	wafer	hard bar	pluribus	sugar percent
100 Grand		1	0	1	0	0.732
3 Musketeers		0	0	1	0	0.604
Almond Joy		0	0	1	0	0.465
Baby Ruth		0	0	1	0	0.604

Charleston Chew	0	0	1	0	0.604
Hershey's Kisses	0	0	0	1	0.127
Hershey's Krackel	1	0	1	0	0.430
Hershey's Milk Chocolate	0	0	1	0	0.430
Hershey's Special Dark	0	0	1	0	0.430
Junior Mints	0	0	0	1	0.197
Kit Kat	1	0	1	0	0.313
Peanut butter M&M's	0	0	0	1	0.825
M&M's	0	0	0	1	0.825
Milk Duds	0	0	0	1	0.302
Milky Way	0	0	1	0	0.604
Milky Way Midnight	0	0	1	0	0.732
Milky Way Simply Caramel	0	0	1	0	0.965
Mounds	0	0	1	0	0.313
Mr Good Bar	0	0	1	0	0.313
Nestle Butterfinger	0	0	1	0	0.604
Nestle Crunch	1	0	1	0	0.313
Peanut M&Ms	0	0	0	1	0.593
Reese's Miniatures	0	0	0	0	0.034
Reese's Peanut Butter cup	0	0	0	0	0.720
Reese's pieces	0	0	0	1	0.406
Reese's stuffed with pieces	0	0	0	0	0.988
Rolo	0	0	0	1	0.860
Sixlets	0	0	0	1	0.220
Nestle Smarties	0	0	0	1	0.267
Snickers	0	0	1	0	0.546
Snickers Crisper	1	0	1	0	0.604
Tootsie Pop	0	1	0	0	0.604
Tootsie Roll Juniors	0	0	0	0	0.313
Tootsie Roll Midgies	0	0	0	1	0.174
Tootsie Roll Snack Bars	0	0	1	0	0.465
Twix	1	0	1	0	0.546
Whoppers	1	0	0	1	0.872

	pricepercent	winpercent
100 Grand	0.860	66.97173
3 Musketeers	0.511	67.60294
Almond Joy	0.767	50.34755
Baby Ruth	0.767	56.91455
Charleston Chew	0.511	38.97504
Hershey's Kisses	0.093	55.37545
Hershey's Krackel	0.918	62.28448
Hershey's Milk Chocolate	0.918	56.49050
Hershey's Special Dark	0.918	59.23612

Junior Mints	0.511	57.21925
Kit Kat	0.511	76.76860
Peanut butter M&M's	0.651	71.46505
M&M's	0.651	66.57458
Milk Duds	0.511	55.06407
Milky Way	0.651	73.09956
Milky Way Midnight	0.441	60.80070
Milky Way Simply Caramel	0.860	64.35334
Mounds	0.860	47.82975
Mr Good Bar	0.918	54.52645
Nestle Butterfinger	0.767	70.73564
Nestle Crunch	0.767	66.47068
Peanut M&Ms	0.651	69.48379
Reese's Miniatures	0.279	81.86626
Reese's Peanut Butter cup	0.651	84.18029
Reese's pieces	0.651	73.43499
Reese's stuffed with pieces	0.651	72.88790
Rolo	0.860	65.71629
Sixlets	0.081	34.72200
Nestle Smarties	0.976	37.88719
Snickers	0.651	76.67378
Snickers Crisper	0.651	59.52925
Tootsie Pop	0.325	48.98265
Tootsie Roll Juniors	0.511	43.06890
Tootsie Roll Midgies	0.011	45.73675
Tootsie Roll Snack Bars	0.325	49.65350
Twix	0.906	81.64291
Whoppers	0.848	49.52411

Step 2

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

```
[1] 66.97173 67.60294 50.34755 56.91455 38.97504 55.37545 62.28448 56.49050
[9] 59.23612 57.21925 76.76860 71.46505 66.57458 55.06407 73.09956 60.80070
[17] 64.35334 47.82975 54.52645 70.73564 66.47068 69.48379 81.86626 84.18029
[25] 73.43499 72.88790 65.71629 34.72200 37.88719 76.67378 59.52925 48.98265
[33] 43.06890 45.73675 49.65350 81.64291 49.52411
```

Step 3

```
choc.mean <- mean(choc.win)
choc.mean
```

```
[1] 60.92153
```

Step 4

```
fruit.inds <- candy$fruity == 1
candy[fruit.inds,]
```

	chocolate	fruity	caramel	peanutyalmondy	nougat
Air Heads	0	1	0	0	0
Caramel Apple Pops	0	1	1	0	0
Chewy Lemonhead Fruit Mix	0	1	0	0	0
Chiclets	0	1	0	0	0
Dots	0	1	0	0	0
Dum Dums	0	1	0	0	0
Fruit Chews	0	1	0	0	0
Fun Dip	0	1	0	0	0
Gobstopper	0	1	0	0	0
Haribo Gold Bears	0	1	0	0	0
Haribo Sour Bears	0	1	0	0	0
Haribo Twin Snakes	0	1	0	0	0
Jawbusters	0	1	0	0	0
Laffy Taffy	0	1	0	0	0
Lemonhead	0	1	0	0	0
Lifesavers big ring gummies	0	1	0	0	0
Mike & Ike	0	1	0	0	0
Nerds	0	1	0	0	0
Nik L Nip	0	1	0	0	0
Now & Later	0	1	0	0	0
Pop Rocks	0	1	0	0	0
Red vines	0	1	0	0	0
Ring pop	0	1	0	0	0
Runts	0	1	0	0	0
Skittles original	0	1	0	0	0
Skittles wildberry	0	1	0	0	0
Smarties candy	0	1	0	0	0
Sour Patch Kids	0	1	0	0	0
Sour Patch Tricksters	0	1	0	0	0
Starburst	0	1	0	0	0

Strawberry bon bons	0	1	0	0	0
Super Bubble	0	1	0	0	0
Swedish Fish	0	1	0	0	0
Tootsie Pop	1	1	0	0	0
Trolli Sour Bites	0	1	0	0	0
Twizzlers	0	1	0	0	0
Warheads	0	1	0	0	0
Welch's Fruit Snacks	0	1	0	0	0
	crisped	rice	wafer	hard bar	pluribus sugarpercent
Air Heads		0	0	0	0.906
Caramel Apple Pops		0	0	0	0.604
Chewey Lemonhead Fruit Mix		0	0	0	1 0.732
Chiclets		0	0	0	1 0.046
Dots		0	0	0	1 0.732
Dum Dums		0	1	0	0 0.732
Fruit Chews		0	0	0	1 0.127
Fun Dip		0	1	0	0 0.732
Gobstopper		0	1	0	1 0.906
Haribo Gold Bears		0	0	0	1 0.465
Haribo Sour Bears		0	0	0	1 0.465
Haribo Twin Snakes		0	0	0	1 0.465
Jawbusters		0	1	0	1 0.093
Laffy Taffy		0	0	0	0 0.220
Lemonhead		0	1	0	0 0.046
Lifesavers big ring gummies		0	0	0	0 0.267
Mike & Ike		0	0	0	1 0.872
Nerds		0	1	0	1 0.848
Nik L Nip		0	0	0	1 0.197
Now & Later		0	0	0	1 0.220
Pop Rocks		0	1	0	1 0.604
Red vines		0	0	0	1 0.581
Ring pop		0	1	0	0 0.732
Runts		0	1	0	1 0.872
Skittles original		0	0	0	1 0.941
Skittles wildberry		0	0	0	1 0.941
Smarties candy		0	1	0	1 0.267
Sour Patch Kids		0	0	0	1 0.069
Sour Patch Tricksters		0	0	0	1 0.069
Starburst		0	0	0	1 0.151
Strawberry bon bons		0	1	0	1 0.569
Super Bubble		0	0	0	0 0.162
Swedish Fish		0	0	0	1 0.604
Tootsie Pop		0	1	0	0 0.604

Trolli Sour Bites	0	0	0	1	0.313
Twizzlers	0	0	0	0	0.220
Warheads	0	1	0	0	0.093
Welch's Fruit Snacks	0	0	0	1	0.313

	pricepercent	winpercent
Air Heads	0.511	52.34146
Caramel Apple Pops	0.325	34.51768
Chewey Lemonhead Fruit Mix	0.511	36.01763
Chiclets	0.325	24.52499
Dots	0.511	42.27208
Dum Dums	0.034	39.46056
Fruit Chews	0.034	43.08892
Fun Dip	0.325	39.18550
Gobstopper	0.453	46.78335
Haribo Gold Bears	0.465	57.11974
Haribo Sour Bears	0.465	51.41243
Haribo Twin Snakes	0.465	42.17877
Jawbusters	0.511	28.12744
Laffy Taffy	0.116	41.38956
Lemonhead	0.104	39.14106
Lifesavers big ring gummies	0.279	52.91139
Mike & Ike	0.325	46.41172
Nerds	0.325	55.35405
Nik L Nip	0.976	22.44534
Now & Later	0.325	39.44680
Pop Rocks	0.837	41.26551
Red vines	0.116	37.34852
Ring pop	0.965	35.29076
Runts	0.279	42.84914
Skittles original	0.220	63.08514
Skittles wildberry	0.220	55.10370
Smarties candy	0.116	45.99583
Sour Patch Kids	0.116	59.86400
Sour Patch Tricksters	0.116	52.82595
Starburst	0.220	67.03763
Strawberry bon bons	0.058	34.57899
Super Bubble	0.116	27.30386
Swedish Fish	0.755	54.86111
Tootsie Pop	0.325	48.98265
Trolli Sour Bites	0.255	47.17323
Twizzlers	0.116	45.46628
Warheads	0.116	39.01190
Welch's Fruit Snacks	0.313	44.37552

```
fruit.win <- candy[fruit.inds,]$winpercent
fruit.win
```

```
[1] 52.34146 34.51768 36.01763 24.52499 42.27208 39.46056 43.08892 39.18550
[9] 46.78335 57.11974 51.41243 42.17877 28.12744 41.38956 39.14106 52.91139
[17] 46.41172 55.35405 22.44534 39.44680 41.26551 37.34852 35.29076 42.84914
[25] 63.08514 55.10370 45.99583 59.86400 52.82595 67.03763 34.57899 27.30386
[33] 54.86111 48.98265 47.17323 45.46628 39.01190 44.37552
```

```
fruit.mean<- mean(fruit.win)
fruit.mean
```

```
[1] 44.11974
```

It appears that the mean of fruity candy is much lower than the mean of chocolate candy, comparing 61% to 44%.

```
choc.mean
```

```
[1] 60.92153
```

```
fruit.mean
```

```
[1] 44.11974
```

Q12. Is this difference statistically significant?

Yes, this t-test states that the values are not equal and the difference is statistically different based on the p value.

```
t.test(choc.win, fruit.win)
```

Welch Two Sample t-test

```
data:  choc.win and fruit.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
```

Overall Candy Rankings

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

Nik L Nip, Boston Baked Beans, Chiclets, Super Bubble, Jawbusters

```
# Not that useful, sorts values numerically
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)
sort(x)
```

```
[1] 1 10 100
```

```
x[order(x)]
```

```
[1] 1 10 100
```

The `order()` function tells us how to 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.

```
ord.inds <- order(candy$winpercent)
#tells which number candy is lowest to highest
head(candy[ord.inds, ])
```

	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	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
Root Beer Barrels				0	1	0	1	0.732		0.069

	win	percent
Nik L Nip	22.44	534
Boston Baked Beans	23.41	782
Chiclets	24.52	499
Super Bubble	27.30	386
Jawbusters	28.12	744
Root Beer Barrels	29.70	369

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

Reese's pieces, Snickers, Kit Kat, Twix, Reese's miniatures

```
tail(candy[ord.inds,])
```

	chocolate	fruity	caramel	peanut	almond	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

	crisped	rice	wafer	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

	price	percent	win	percent
--	-------	---------	-----	---------

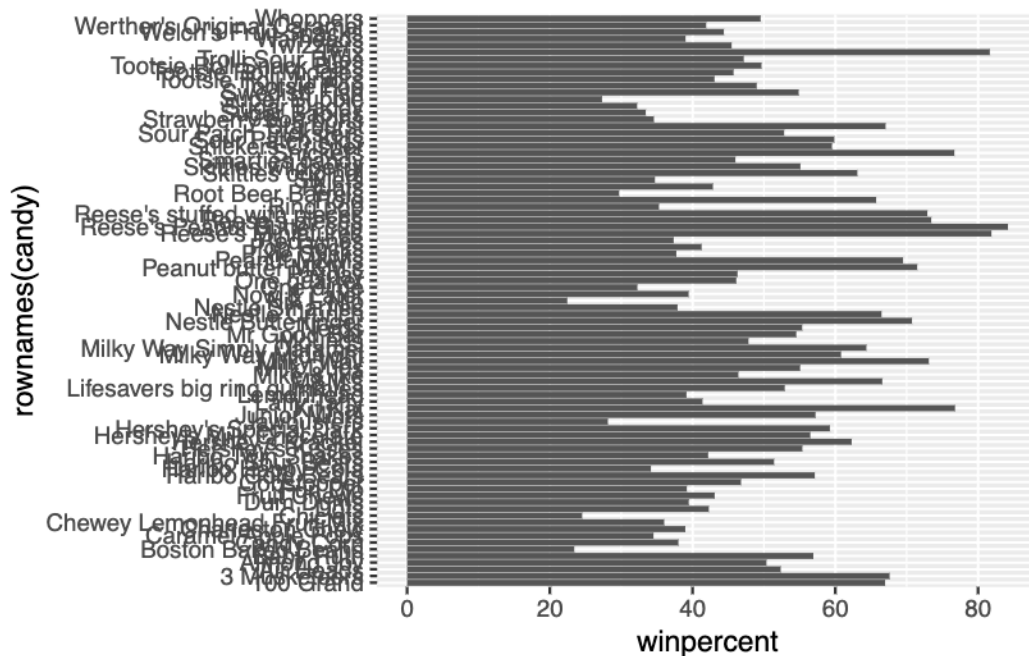
Reese's pieces	0.651	73.43499
Snickers	0.651	76.67378
Kit Kat	0.511	76.76860
Twix	0.906	81.64291
Reese's Miniatures	0.279	81.86626
Reese's Peanut Butter cup	0.651	84.18029

Another option

```
# order.inds <- order(cnady$winpercent, decreasing =T)
# pull heading
```

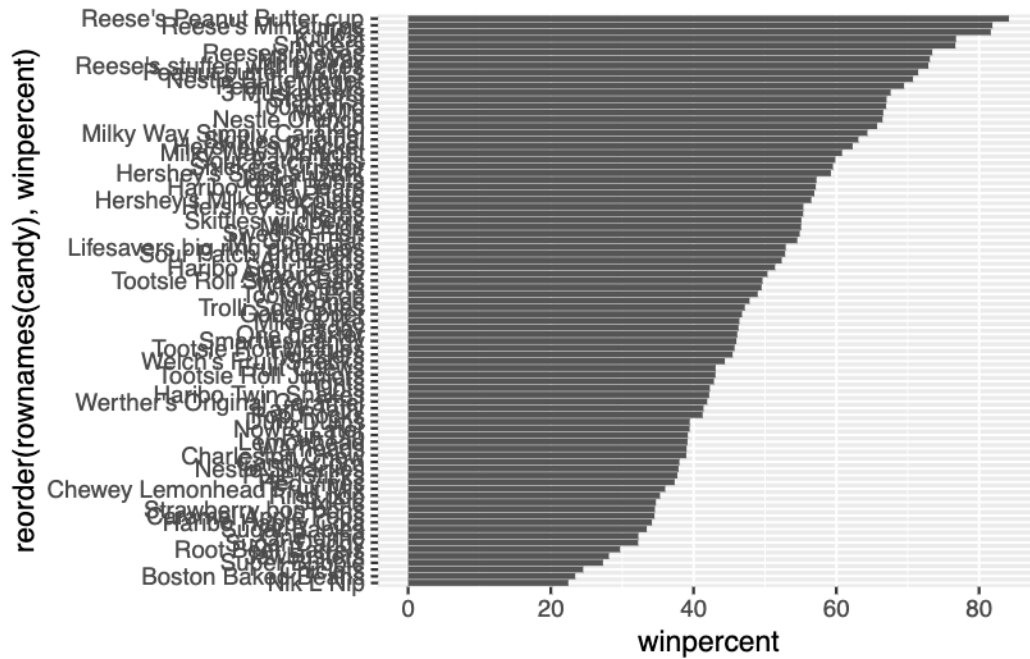
Q15. Make a first barplot of candy ranking based on winpercent values.

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



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

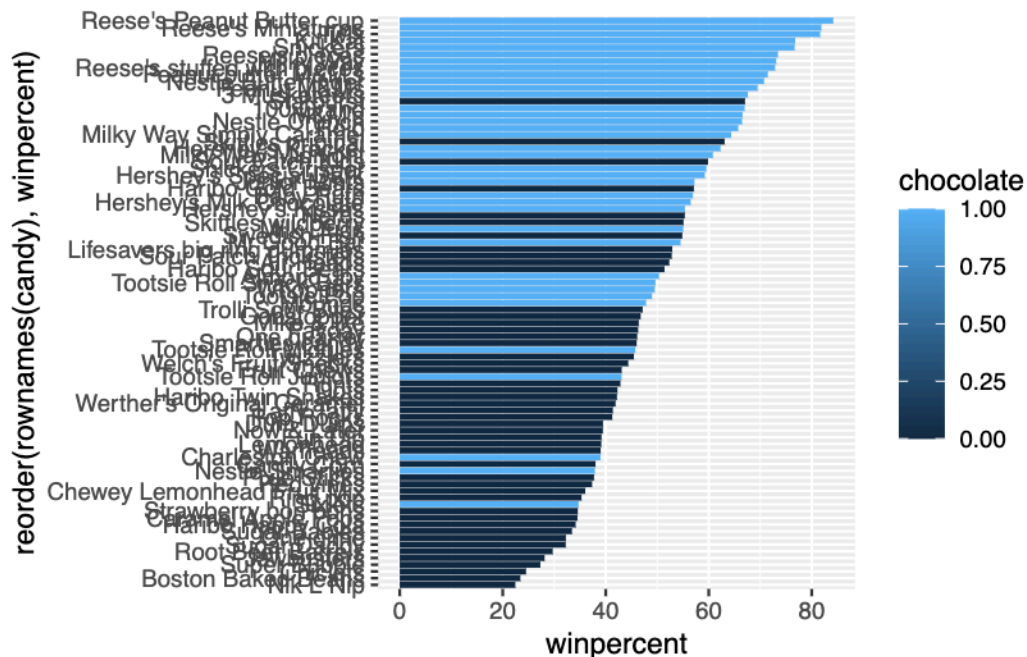
```
ggplot (candy) +
  aes(winpercent, reorder(rownames(candy), winpercent)) +
  geom_col()
```



Let's add some useful colors

This is not helpful, there isn't a scale from 0 to 1

```
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 exactly what candy is colored a particular color.

```
mycols <- rep("gray", nrow(candy))

mycols[candy$chocolate == 1] <- "chocolate"
mycols[candy$fruity == 1] <- "green"

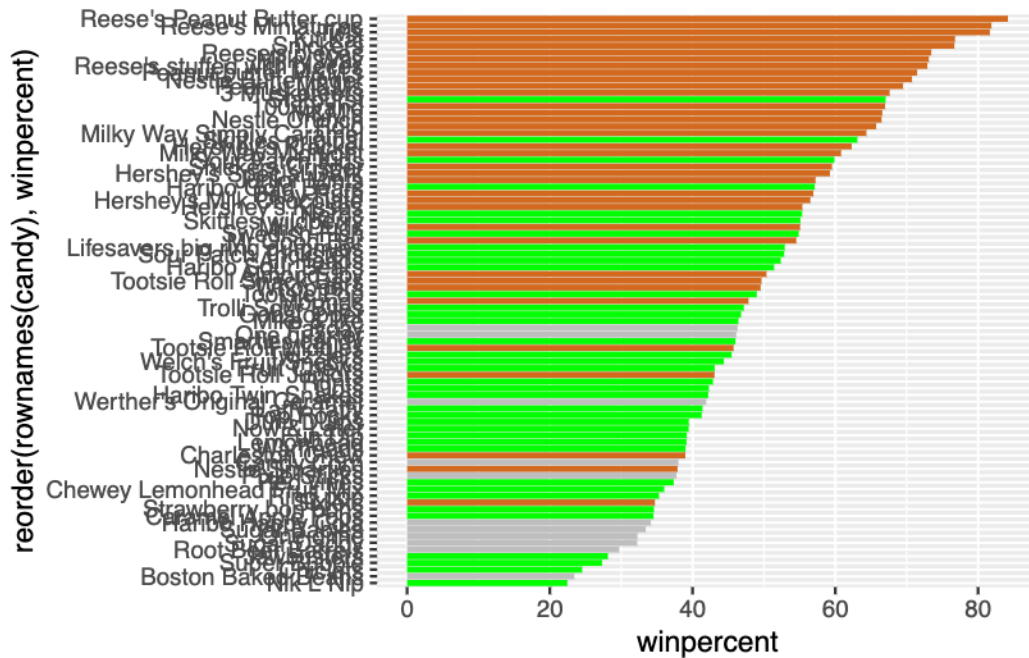
mycols
```

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



```
[79] "green"      "chocolate" "green"      "green"      "green"      "gray"
[85] "chocolate"
```

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



Both these functions work the same way to get out T/F values

```
c(1,0,1) ==1
```

```
[1] TRUE FALSE TRUE
```

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

```
[1] TRUE FALSE TRUE
```

Q17. What is the worst ranked chocolate candy?

Sixlets

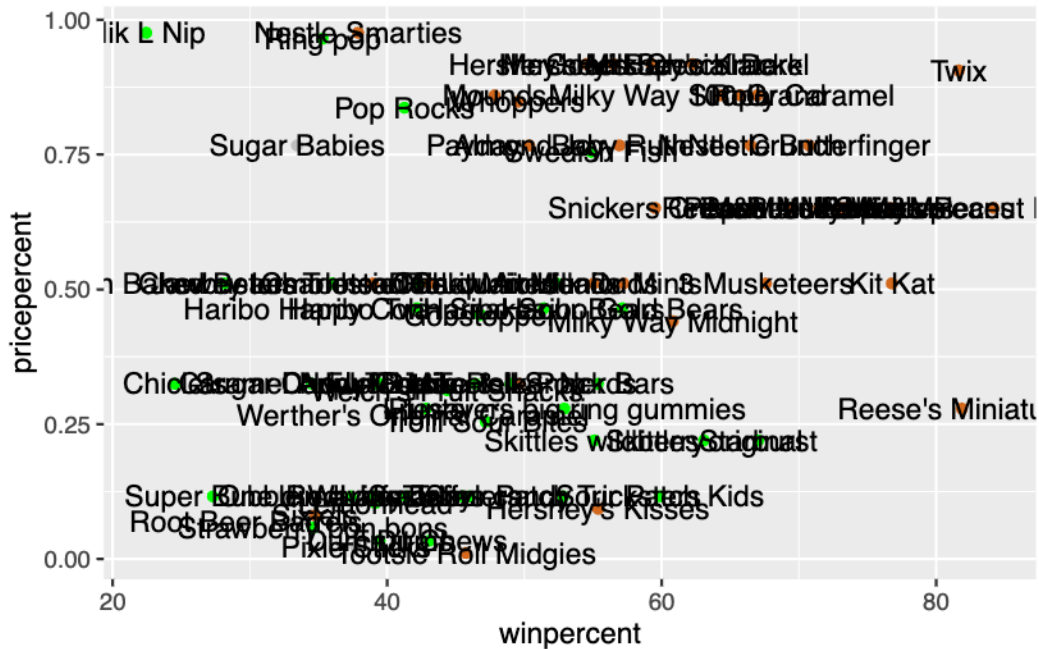
Q18. What is the best ranked fruity candy?

Starburst

Let's look at pricepercent

Making a plot of winpercent(x) versus pricepercent(y)

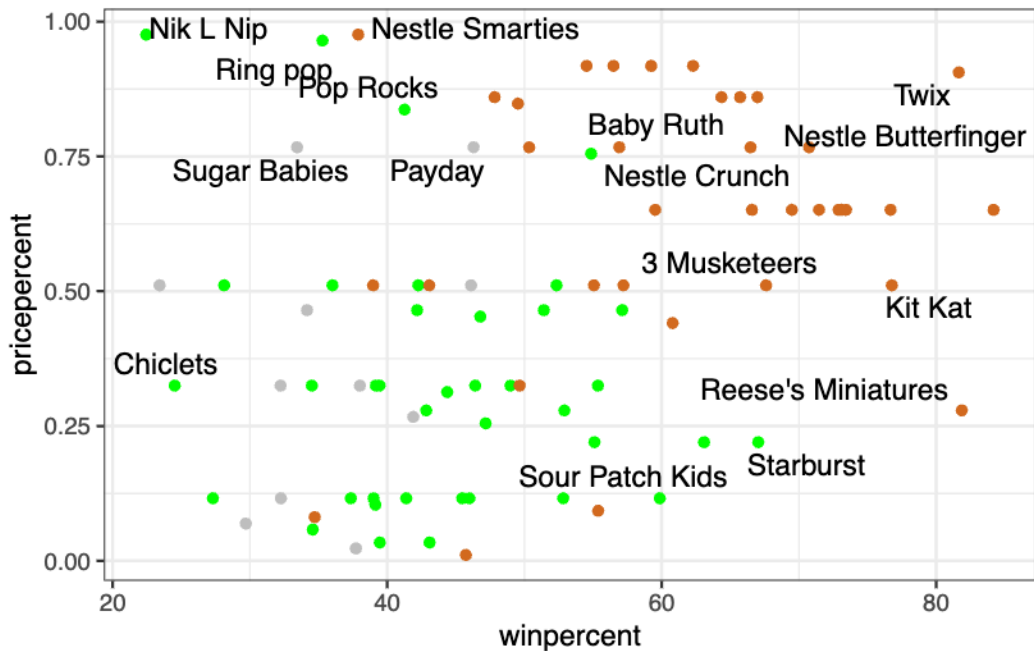
```
ggplot(candy) +  
  aes(winpercent, pricepercent, label=rownames(candy)) +  
  geom_point(col=mycols) +  
  geom_text()
```



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

```
library(ggrepel)  
  
ggplot(candy) +  
  aes(winpercent, pricepercent, label=rownames(candy)) +  
  geom_point(col=mycols) +  
  geom_text_repel(max.overlaps = 6) +  
  theme_bw()
```

Warning: ggrepel: 69 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?

Hershey's Krackel, Nik L Nip, Ring Pop, Nestle Smarties, and Hershey's Milk Chocolate. Nik L Nip is the least popular.

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

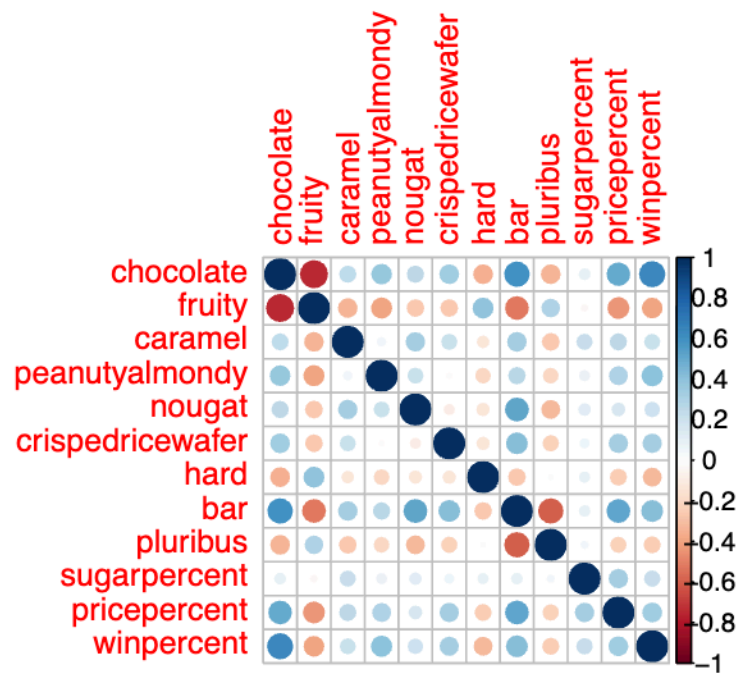
	pricepercent	winpercent
Nik L Nip	0.976	22.44534
Nestle Smarties	0.976	37.88719
Ring pop	0.965	35.29076
Hershey's Krackel	0.918	62.28448
Hershey's Milk Chocolate	0.918	56.49050

Exploring the correlation structure

```
library(corrplot)
```

corrplot 0.95 loaded

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



Down the diagonal there is a perfect 1 correlation because its comparing to itself. -1 is high negative correlation, +1 is high positive correlation.

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

Fruity and chocolate.

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

Winpercent and chocolate.

PCA

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

Importance of components:

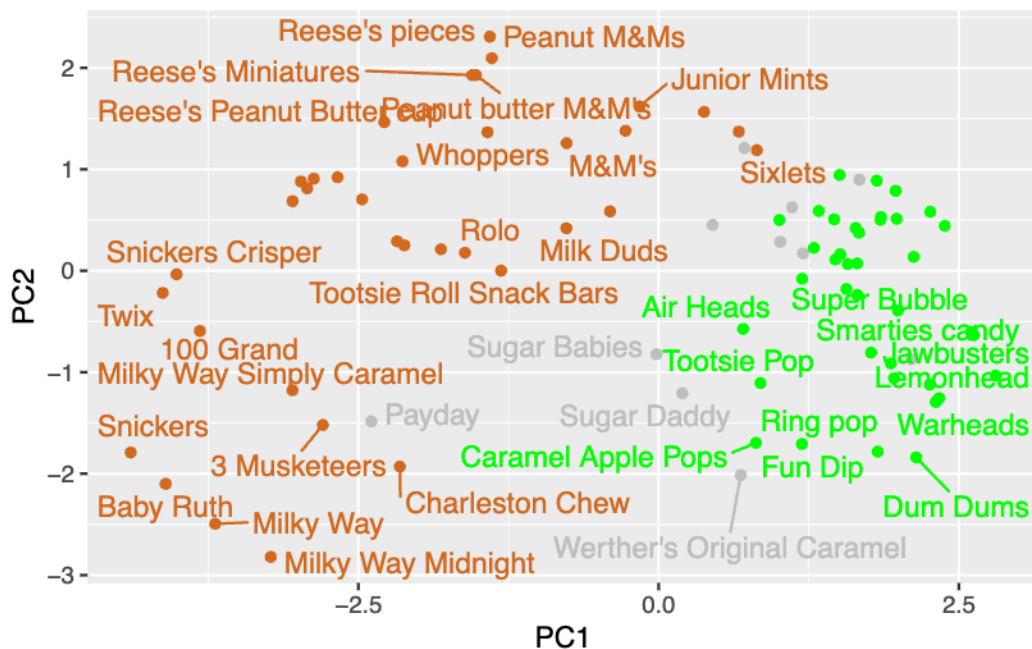
	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	2.0788	1.1378	1.1092	1.07533	0.9518	0.81923	0.81530
Proportion of Variance	0.3601	0.1079	0.1025	0.09636	0.0755	0.05593	0.05539
Cumulative Proportion	0.3601	0.4680	0.5705	0.66688	0.7424	0.79830	0.85369

	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

Let's plot our main result 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



```
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

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

PC1 has a strong positive correlation with our fruity candies, things that are hard and come with multiple items in one package. This makes sense because all these variables are connection, in that fruity candies are more often hard and come with multiple items in a package.

```
#barplot(pca$rotation[,1], las=2, ylab="PC1 Contribution")
```

```
ggplot(pca$rotation) +
```

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

