

Class09

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Importing candy data

```
candy_file <- read.csv("candy-data.csv",row.names=1)
candy = data.frame(candy_file)
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 is in the dataset?

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

```
num.candy.types <- ncol(candy)
num.candy.types
```

```
[1] 12
```

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

```
num.fruity <- sum(candy$fruity)
num.fruity
```

```
[1] 38
```

What is your favorite candy?

Q3. What is your favorite candy (other than Twix) in the dataset and what is its winpercent value?

```
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 |>
  filter(row.names(candy)=="Air Heads") |>
  select(winpercent)
```

```
      winpercent
Air Heads  52.34146
```

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

```
library(dplyr)

candy |>
  filter(row.names(candy)=="Kit Kat") |>
  select(winpercent)
```

```
      winpercent
Kit Kat   76.7686
```

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

```
library(dplyr)

candy |>
  filter(row.names(candy)=="Tootsie Roll Snack Bars") |>
  select(winpercent)
```

```
      winpercent
Tootsie Roll Snack Bars  49.6535
```

Q6. Is there any variable/column that looks to be on a different scale to the majority of the other columns in the 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_vari- able	n_miss- ing	com- plete_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	
peanutyal- mondy	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	
crispedrice- wafer	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	

Yes, the `winpercent`.

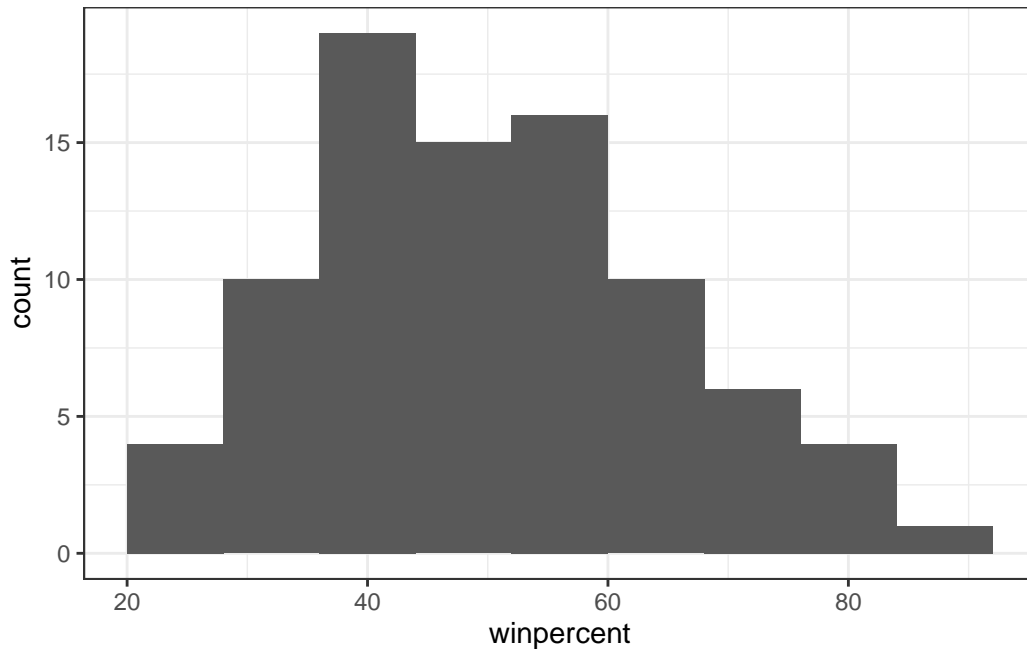
Q7. What do you think a zero and one represent for the `candy$chocolate` column?

0 in missing value means that there is no missing value in `chocolate`. 1 means that all the rows in `chocolate` has a value.

Exploratory analysis

Q8. Plot a histogram of `winpercent` values

```
library(ggplot2)
ggplot(candy)+
  aes(winpercent)+
  geom_histogram(binwidth = 8)+
  theme_bw()
```



Q9. Is the distribution of winpercent values symmetrical?

The distribution of winpercent is not symmetrical and it is right skewed.

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

```
summary(candy$winpercent)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
22.45	39.14	47.83	50.32	59.86	84.18

Based on the median, the the center of the distribution is below 50%.

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

```
choc.candy <- candy[candy$chocolate==1,]
choc.win <- choc.candy$winpercent
summary(choc.win)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
34.72	50.35	60.80	60.92	70.74	84.18

```
fruity.candy <- candy[candy$fruity==1,]
fruity.win <- fruity.candy$winpercent
summary(fruity.win)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
22.45	39.04	42.97	44.12	52.11	67.04

As the mean of chocolate winpercent is higher, on average chocolate candy is higher ranked than fruity candy.

Q12. Is this difference statistically significant?

```
t.result <- t.test(choc.win,fruity.win)
t.result
```

Welch Two Sample t-test

```
data:  choc.win and fruity.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
```

Since $p < 0.05$, this difference is statistically significant.

Overall Candy Rankings

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

```
head(candy[order(candy$winpercent),],n=5)
```

	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	

	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

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

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

```
tail(candy[order(candy$winpercent),],n=5)
```

	chocolate	fruity	caramel	peanut	almond	nougat	
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	

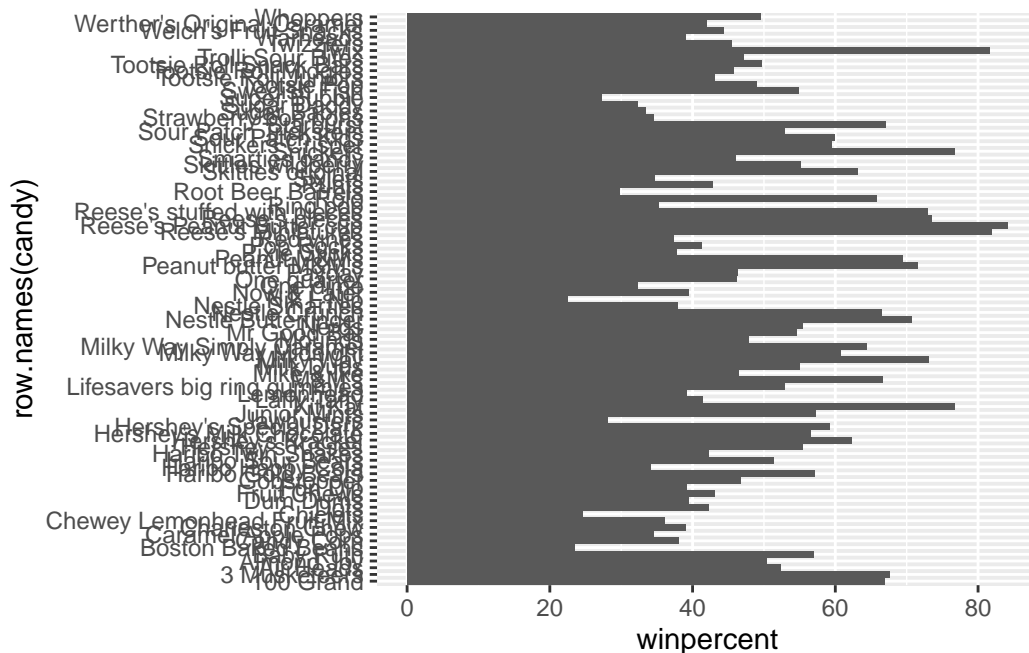
	crisped	rice	wafer	hard	bar	pluribus	sugar	percent
Snickers		0	0	1		0		0.546
Kit Kat		1	0	1		0		0.313
Twix		1	0	1		0		0.546
Reese's Miniatures		0	0	0		0		0.034
Reese's Peanut Butter cup		0	0	0		0		0.720

	price	percent	winpercent
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

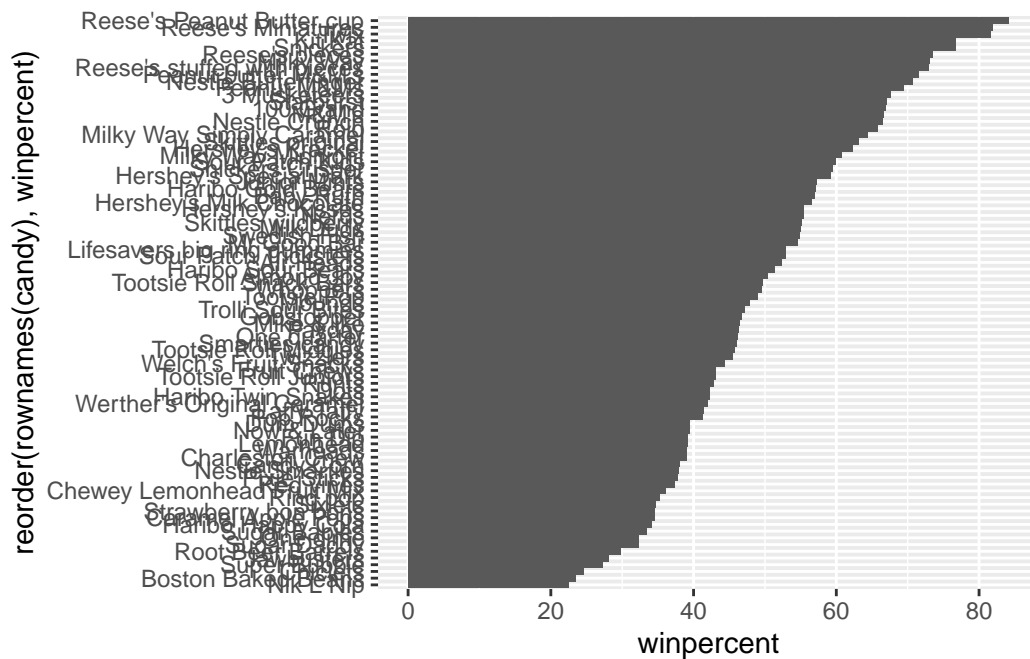
Q15. Make a first barplot of candy ranking based on winpercent values.

```
library(ggplot2)
ggplot(candy)+
  aes(winpercent,row.names(candy))+
  geom_col(position = position_dodge(width = 2))
```



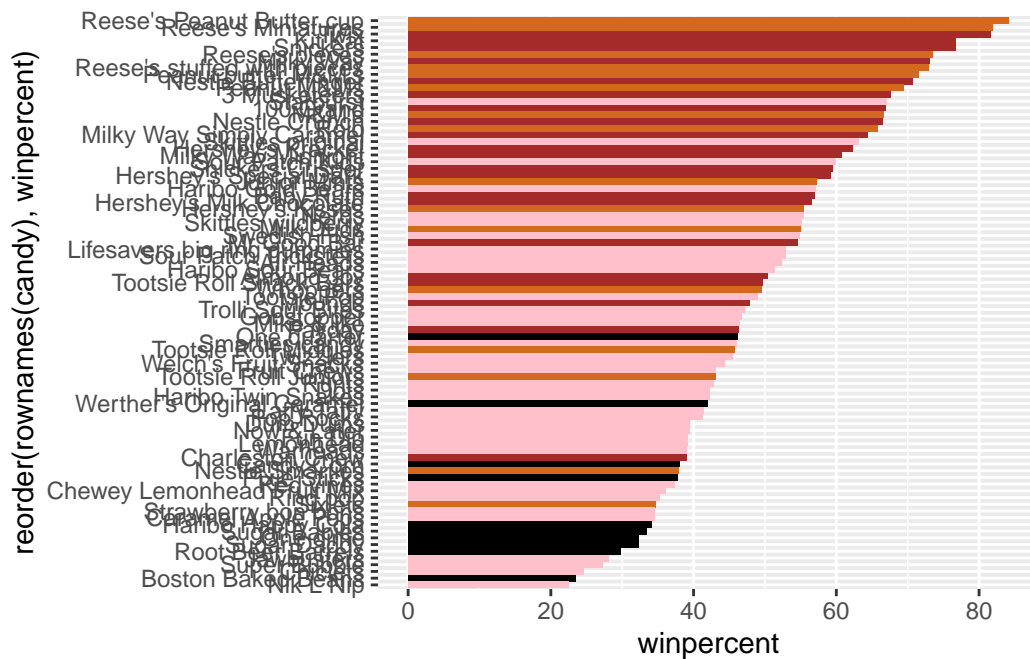
Q16. This is quite ugly, use the reorder() function to get the bars sorted by winpercent?

```
sort.candy <- candy[order(candy$winpercent),]
library(ggplot2)
ggplot(candy)+
  aes(winpercent,reorder(row.names(candy),winpercent))+
  geom_col(position = position_dodge(width = 2))
```

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[candy$fruity==1] = "pink"
library(ggplot2)
ggplot(candy) +
  aes(winpercent, reorder(rownames(candy),winpercent)) +
  geom_col(fill=my_cols)
```



Q17. What is the worst ranked chocolate candy?

Sixlets

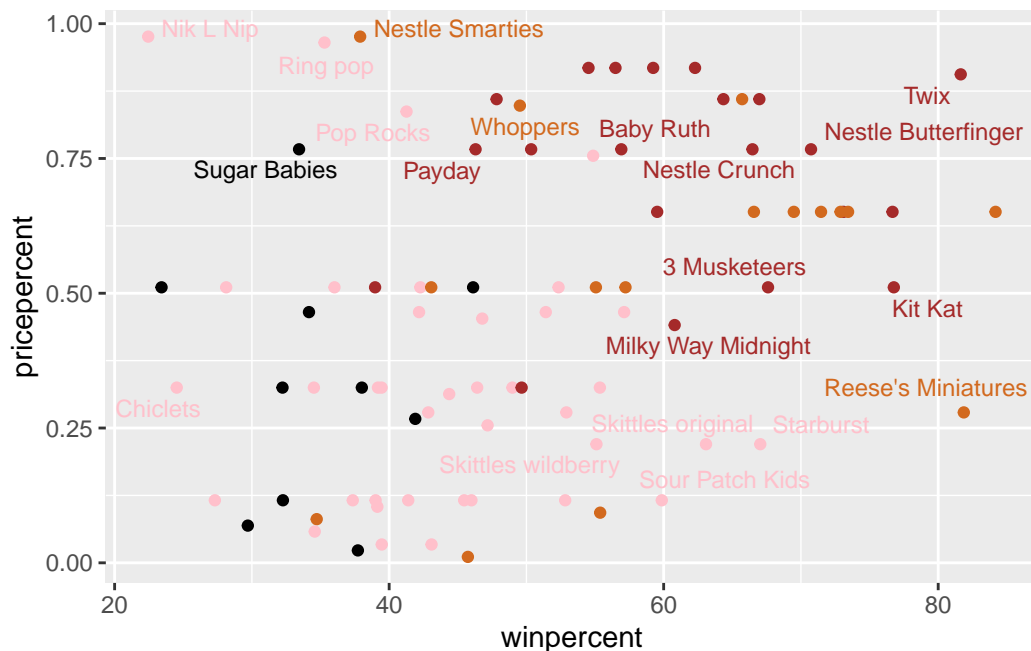
Q18. What is the best ranked fruity candy?

Starburst

Taking a look at pricepercent

```
library(ggrepel)
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?

```
head(candy[order(candy$pricepercent),],n=5)
```

	chocolate	fruity	caramel	peanut	almond	nougat
Tootsie Roll Midgies	1	0	0		0	0
Pixie Sticks	0	0	0		0	0
Dum Dums	0	1	0		0	0
Fruit Chews	0	1	0		0	0
Strawberry bon bons	0	1	0		0	0

	crispedrice	wafer	hard bar	pluribus	sugarpercent
Tootsie Roll Midgies	0	0	0	1	0.174
Pixie Sticks	0	0	0	1	0.093
Dum Dums	0	1	0	0	0.732
Fruit Chews	0	0	0	1	0.127
Strawberry bon bons	0	1	0	1	0.569

	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?

```
tail5 <- tail(candy[order(candy$pricepercent),],n=5)
t5xp <- tail5[nrow(tail5):1,]
t5xp
```

	chocolate	fruity	caramel	peanutyalmondy	nougat
Nestle Smarties	1	0	0	0	0
Nik L Nip	0	1	0	0	0
Ring pop	0	1	0	0	0
Mr Good Bar	1	0	0	1	0
Hershey's Special Dark	1	0	0	0	0

	crispedrice	wafer	hard bar	pluribus	sugarpercent
Nestle Smarties	0	0	0	1	0.267
Nik L Nip	0	0	0	1	0.197
Ring pop	0	1	0	0	0.732
Mr Good Bar	0	0	1	0	0.313
Hershey's Special Dark	0	0	1	0	0.430

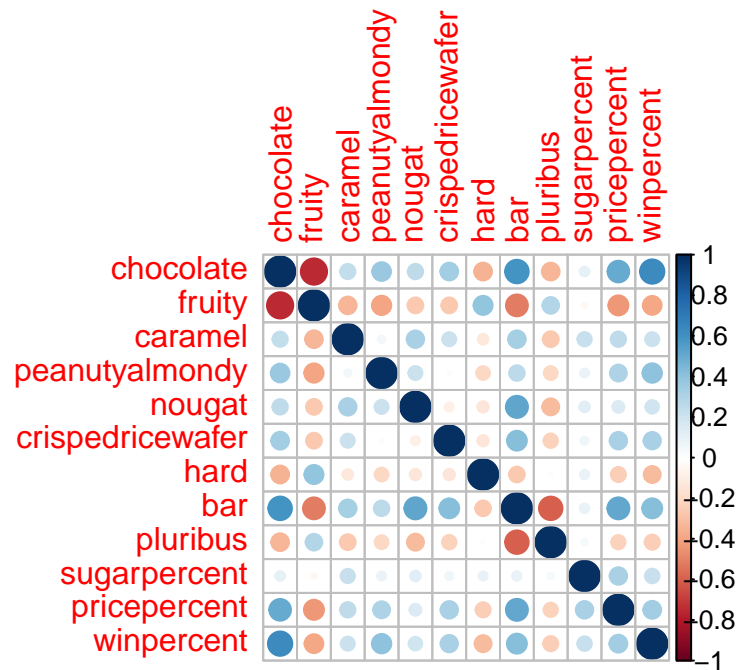
	pricepercent	winpercent
Nestle Smarties	0.976	37.88719
Nik L Nip	0.976	22.44534
Ring pop	0.965	35.29076
Mr Good Bar	0.918	54.52645
Hershey's Special Dark	0.918	59.23612

Exploring the correlation structure

```
library(corrplot)
```

corrplot 0.95 loaded

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



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

Fruity and chocolate are the most anti-correlated pair.

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

winpercent and chocolate.

Principal Component Analysis

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

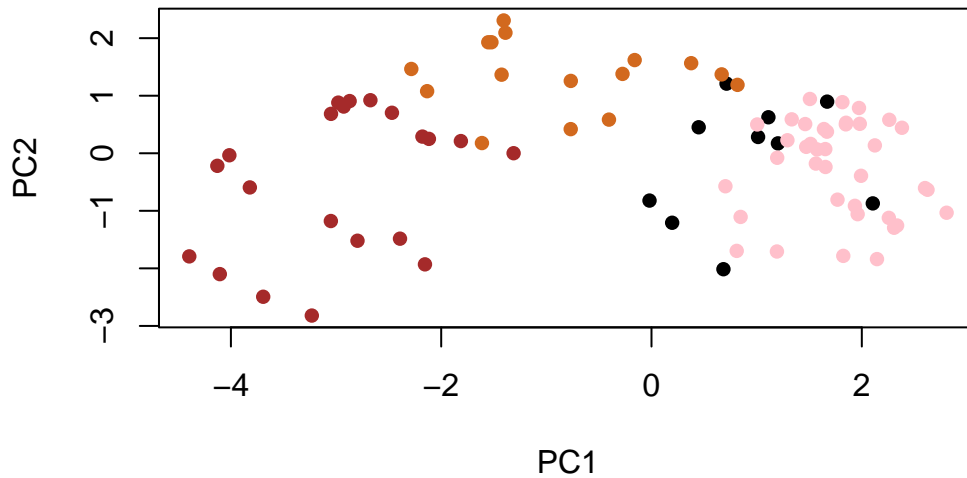
Importance of components:

	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

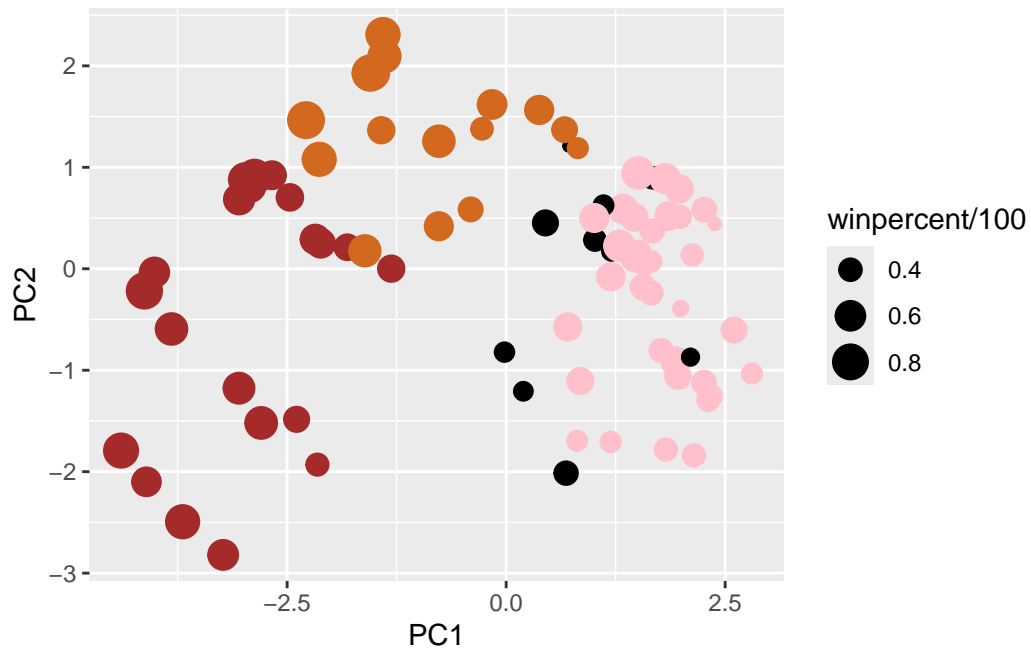
	PC8	PC9	PC10	PC11	PC12
Standard deviation	0.74530	0.67824	0.62349	0.43974	0.39760
Proportion of Variance	0.04629	0.03833	0.03239	0.01611	0.01317

Cumulative Proportion 0.89998 0.93832 0.97071 0.98683 1.00000

```
#my_cols=rep("black", nrow(candy))
#my_cols[as.logical(candy$chocolate)] = "chocolate"
#my_cols[as.logical(candy$bar)] = "brown"
#my_cols[candy$fruity==1] = "pink"
plot(pca$x[,1:2],col=my_cols,pch=16)
```



```
my_data <- cbind(candy, pca$x[,1:3])
p <- ggplot(my_data) +
  aes(PC1, PC2,
       size=winpercent/100,
       text=rownames(my_data),
       label=rownames(my_data)) +
  geom_point(col=my_cols)
p
```

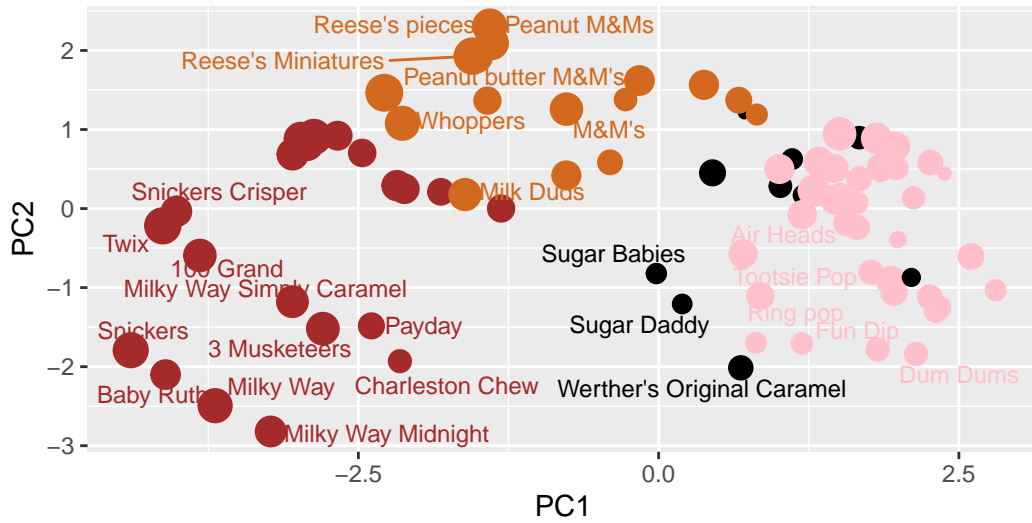


```
library(ggrepel)
p + geom_text_repel(size=3.3, col=my_cols, max.overlaps = 7) +
  theme(legend.position = "none") +
  labs(title="Halloween Candy PCA Space",
        subtitle="Colored by type: chocolate bar (dark brown), chocolate other (light brown),",
        caption="Data from 538")
```

Warning: ggrepel: 59 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),

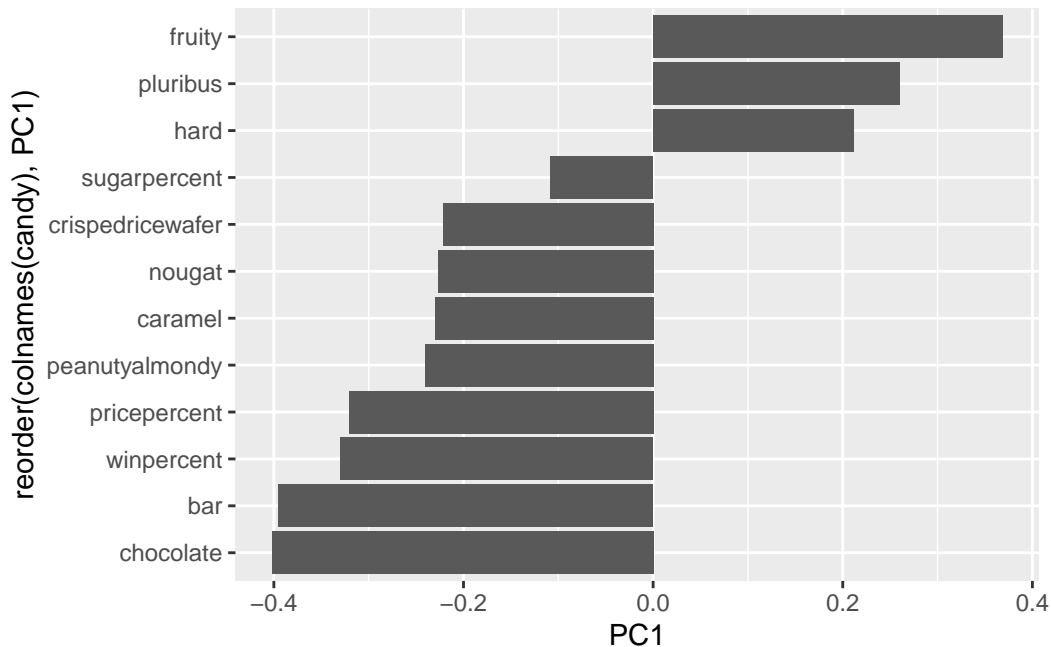


Data from 538

```
#library(plotly)
#ggplotly(p)
```

Q24. Complete the code to generate the loadings plot above. What original variables are picked up strongly by PC1 in the positive direction? Do these make sense to you? Where did you see this relationship highlighted previously?

```
ggplot(pca$rotation) +
  aes(PC1, reorder(colnames(candy), PC1)) +
  geom_col()
```

The variables **fruity**, **pluribus** and **hard** are picked up strongly by PC1 in the positive direction. This make sense as fruity candies are hard, pluribus candies. This relationship was also shown in previous correlation structure plot, where these three are positively correlated with each other.

Summary

Q25. Based on your exploratory analysis, correlation findings, and PCA results, what combination of characteristics appears to make a “winning” candy? How do these different analyses (visualization, correlation, PCA) support or complement each other in reaching this conclusion?

A combination of **chocolate** and **bar** seems to make a “winning” candy based on. First, based on the “Overall Candy Rankings” figure from Q17-18, both **chocolate** and **bar** candies have relatively higher **winpercent** compared to most of the other candies. Second, based on the correlation map from “Exploring the correlation structure”, both **chocolate** and **bar** are strongly positively correlated with **winpercent**. Finally, based on the PCA, in PC1, **chocolate**, **bar** and **winpercent** are all picked up strongly by PC1 in the negative direction, suggesting that these characteristics tend to occur at the same time and cover a large portion of the variance. Combinely, these analyses support and complement each other and suggest that chocolate bar candies seems to make a “winning” candy.