

# Class09: Candy Analysis Mini project

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## Import Data

```
candy_file <- "candy-data.csv"
candy <- read.csv(candy_file, 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

## Data Exploration

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

There are 85 in this dataset.

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

```
fruity_candy <- candy$fruity  
n_fruity_candy <- sum(fruity_candy == 1)  
n_fruity_candy
```

[1] 38

```
twix_winpercent <- candy["Twix", "winpercent"]  
twix_winpercent
```

[1] 81.64291

How many chocolate candy are in the dataset?

### My favorite candy

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

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

[1] 76.67378

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

[1] 39.0119

```
candy["Welch's Fruit Snacks",]$winpercent
```

[1] 44.37552

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
```

```
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

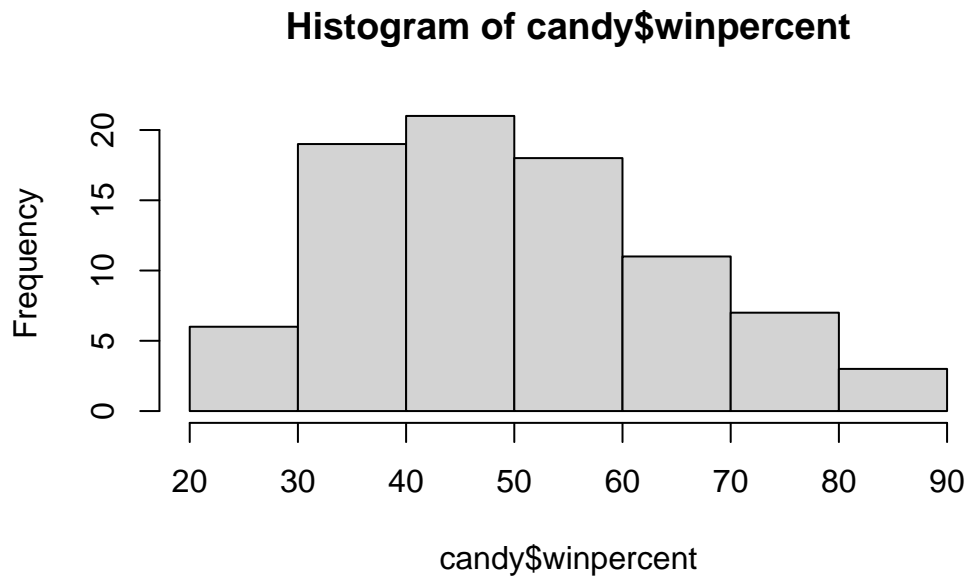
skim_variable	n_missing	complete	ratio	mean	sd	p0	p25	p50	p75	p100	hist
chocolate	0	1	0.44	0.50	0.00	0.00	0.00	0.00	1.00	1.00	
fruity	0	1	0.45	0.50	0.00	0.00	0.00	0.00	1.00	1.00	
caramel	0	1	0.16	0.37	0.00	0.00	0.00	0.00	0.00	1.00	
peanutyalmondy	0	1	0.16	0.37	0.00	0.00	0.00	0.00	0.00	1.00	
nougat	0	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		

Q6. Is there any variable/column that looks to be on a different scale to the majority of the other columns in the dataset?

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

Q8. Plot a histogram of winpercent values

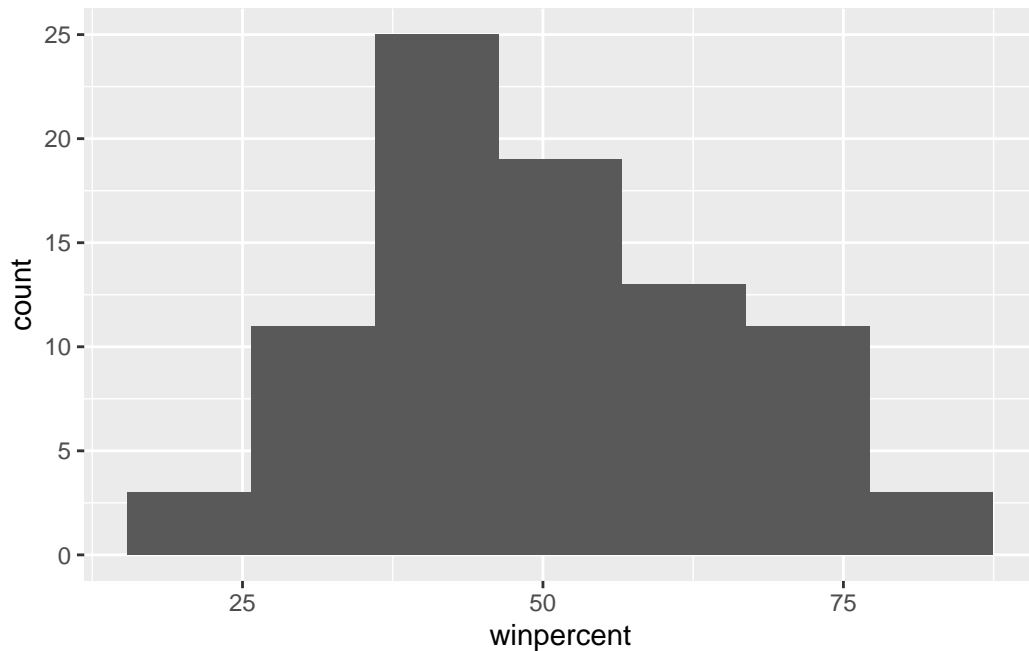
```
hist(candy$winpercent)
```



Q8. Plot a histogram of winpercent values using ggplot

```
library(ggplot2)
```

```
ggplot(candy) +  
  aes(winpercent) +  
  geom_histogram(bins = 7)
```



Q9. Is the distribution of winpercent values symmetrical?

They are not symmetrical

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

The center of the distribution is below 50%

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

```
[1] 50.31676
```

```
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?

- first find all chocolate candy
- find their winpercent values

- calculate the mean of these values
- then do the same for fruity candy and compare with the mean for chocolate candy

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

```
[1] 60.92153
```

```
fruit_inds <- as.logical(candy$fruity)
fruit.win <- candy[fruit_inds,]$winpercent
mean(fruit.win)
```

```
[1] 44.11974
```

Q12. Is this difference statistically significant?

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

Welch Two Sample t-test

```
data: chocolate.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
```

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

```
x <- c(5,6,4)
sort(x)
```

```
[1] 4 5 6
```

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

```
[1] 4 5 6
```

The order function returns the indices that make the input sorted.

```
inds <- order(candy$winpercent)
head(candy[inds,],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?

```
top <- order(candy$winpercent)
tail(candy[inds,],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

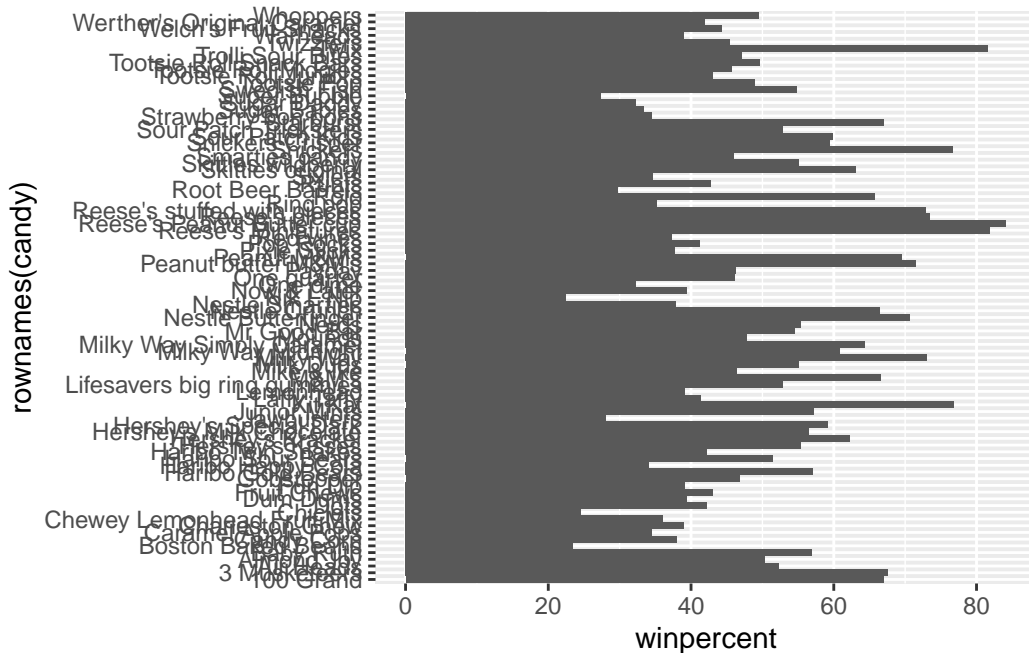
	crisped	rice	wafer	hard	bar	pluribus	sugar	percent
--	---------	------	-------	------	-----	----------	-------	---------

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

	pricepercent	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.

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

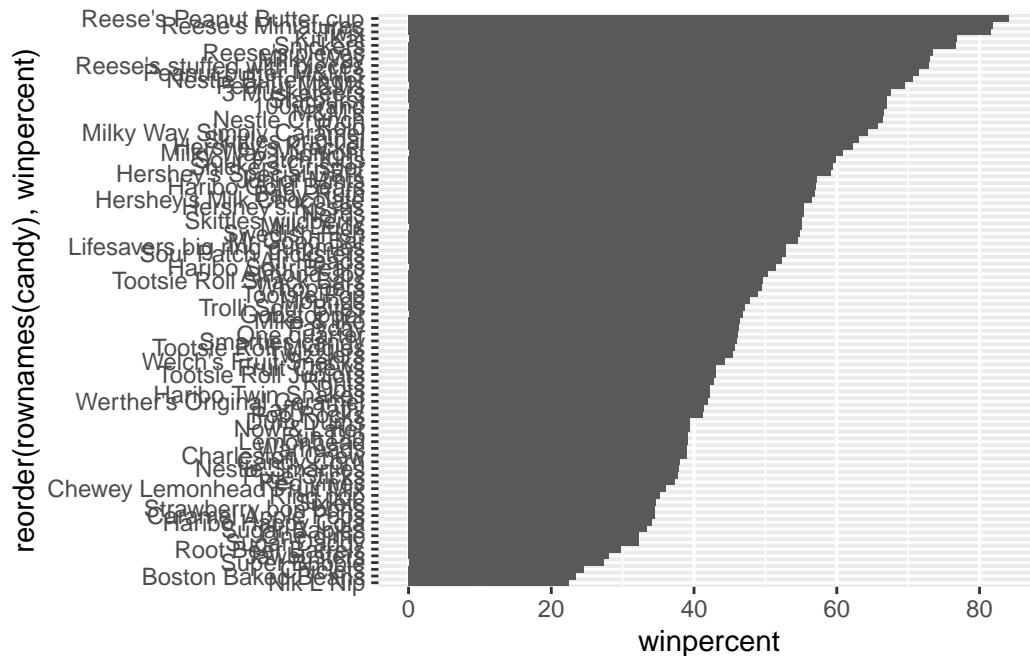


Q16. This is quite ugly, use the reorder() function to get the bars sorted by winpercent?



```
# |fig-height : 10
# |fig-width : 7
```

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



```
ggsave("mybarplot.png", height =10)
```

Saving 5.5 x 10 in image

Add my custom colors to the barplot

```
my_cols=rep("grey", nrow(candy))
my_cols[candy$fruity ==1] <- "pink"
my_cols[candy$chocolate ==1] <- "chocolate"
my_cols[candy$bar ==1] <- "brown"
my_cols
```

```
[1] "brown" "brown" "grey" "grey" "pink" "brown"
```

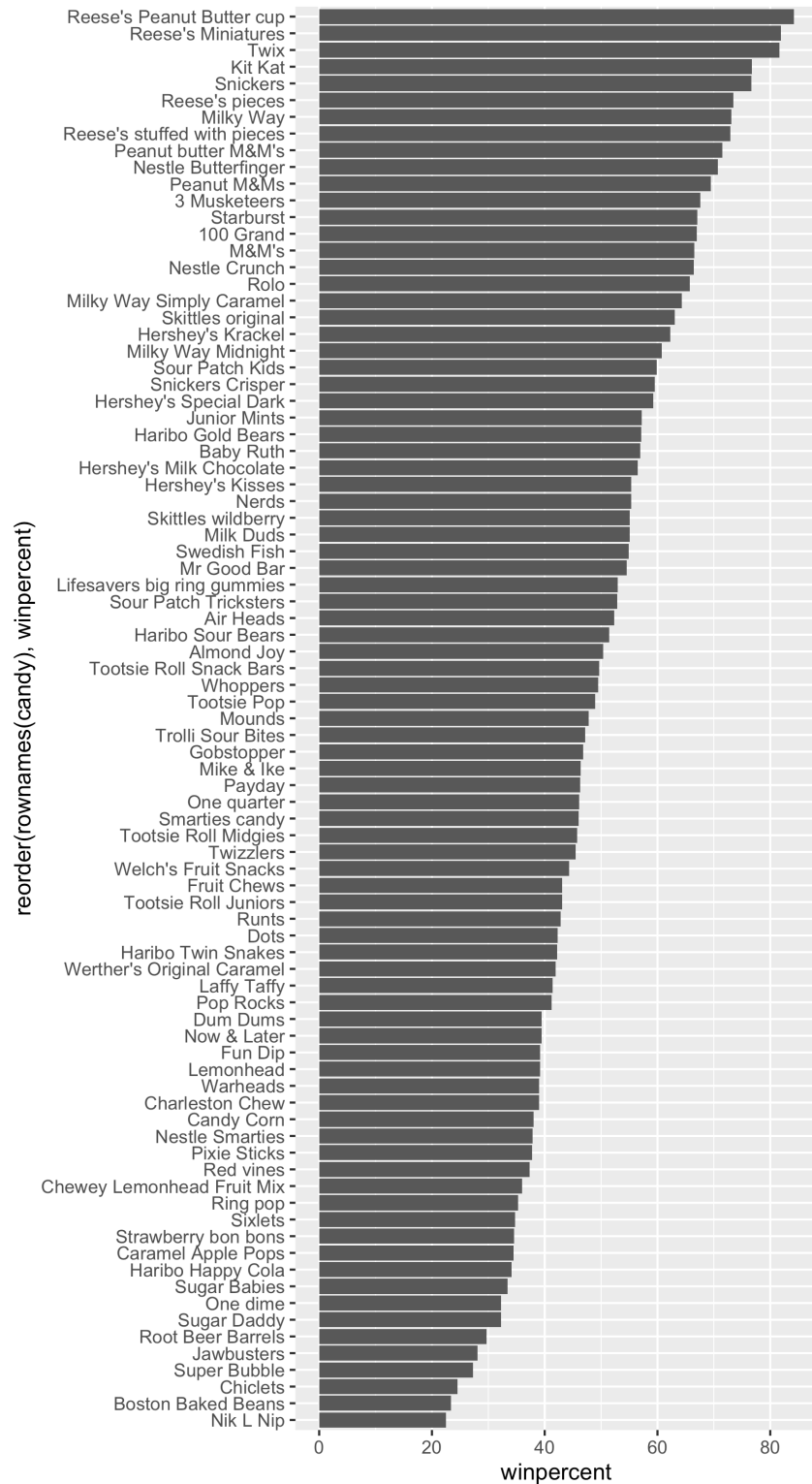


Figure 1: Exported image that is a bit bigger so I can read it

```

[7] "brown"      "grey"      "grey"      "pink"      "brown"      "pink"
[13] "pink"       "pink"      "pink"      "pink"      "pink"      "pink"
[19] "pink"       "grey"      "pink"      "pink"      "chocolate" "brown"
[25] "brown"      "brown"      "pink"      "chocolate" "brown"      "pink"
[31] "pink"       "pink"      "chocolate" "chocolate" "pink"      "chocolate"
[37] "brown"      "brown"      "brown"      "brown"      "brown"      "pink"
[43] "brown"      "brown"      "pink"      "pink"      "brown"      "chocolate"
[49] "grey"       "pink"      "pink"      "chocolate" "chocolate" "chocolate"
[55] "chocolate" "pink"      "chocolate" "grey"      "pink"      "chocolate"
[61] "pink"       "pink"      "chocolate" "pink"      "brown"      "brown"
[67] "pink"       "pink"      "pink"      "pink"      "grey"      "grey"
[73] "pink"       "pink"      "chocolate" "chocolate" "chocolate" "brown"
[79] "pink"       "brown"      "pink"      "pink"      "pink"      "grey"
[85] "chocolate"

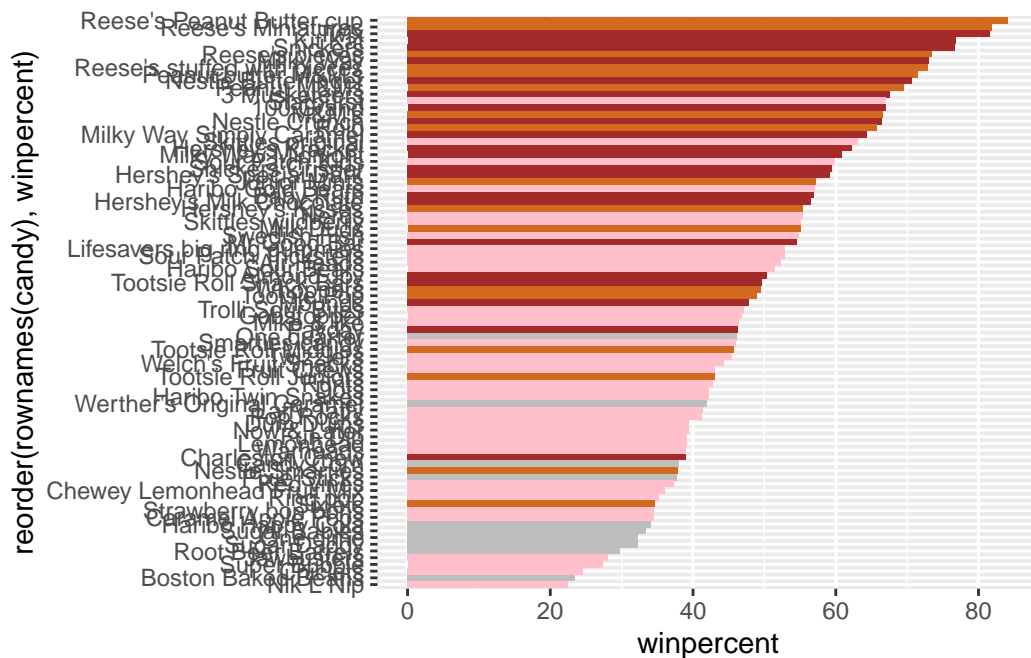
```

```

# |fig-height : 10
# |fig-width : 7

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

```



Q17. What is the worst ranked chocolate candy?

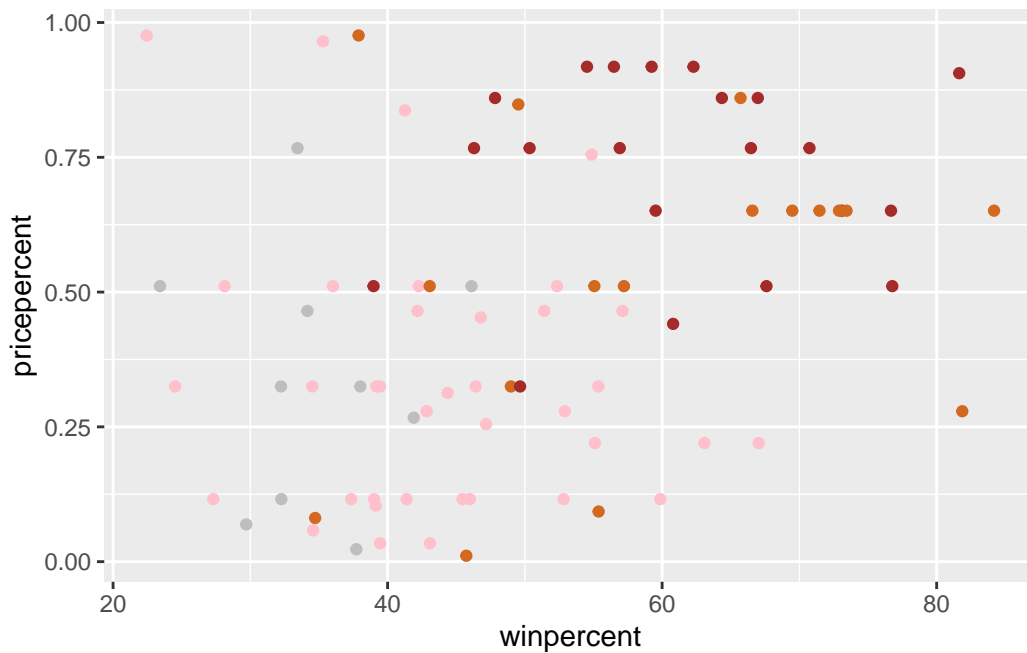
The worst ranked chocolate candy is Sixlets

Q18. What is the best ranked fruity candy?

The best ranked fruity candy is Starburst

Plot of winpercent vs pripercent

```
ggplot(candy) +  
  aes(winpercent, pricepercent) +  
  geom_point(col=my_cols)
```



```
my_cols=rep("black", nrow(candy))  
my_cols[candy$fruity ==1] <- "pink"  
my_cols[candy$chocolate ==1] <- "chocolate"  
my_cols[candy$bar ==1] <- "brown"  
my_cols
```

```
[1] "brown"    "brown"    "black"    "black"    "pink"     "brown"  
[7] "brown"    "black"    "black"    "pink"     "brown"     "pink"
```

```

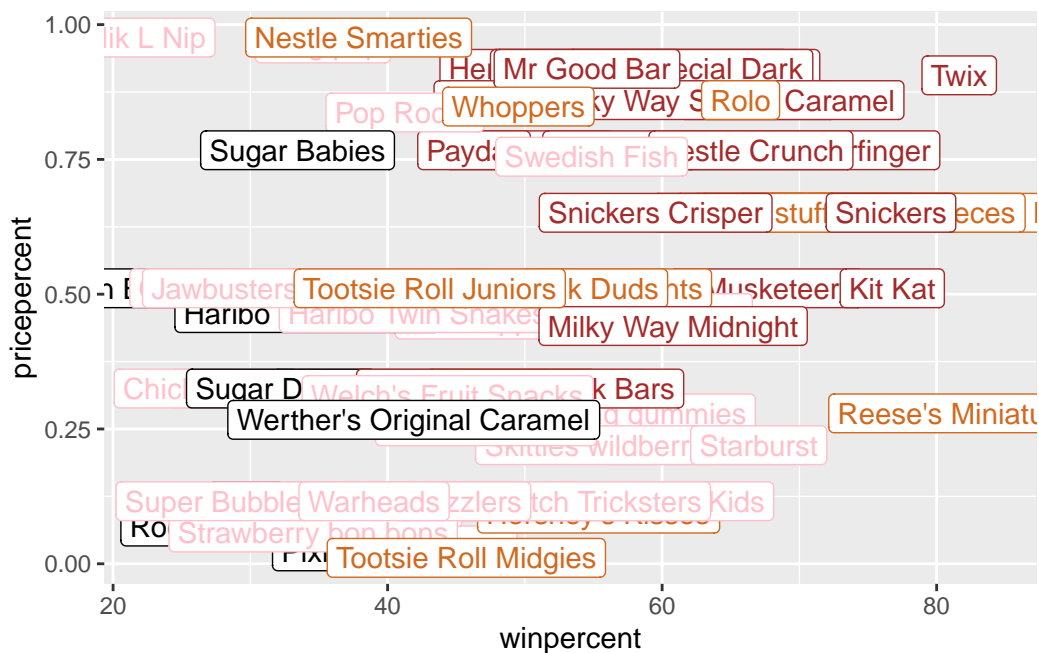
[13] "pink"      "pink"      "pink"      "pink"      "pink"      "pink"
[19] "pink"      "black"     "pink"      "pink"      "chocolate" "brown"
[25] "brown"     "brown"     "pink"      "chocolate" "brown"     "pink"
[31] "pink"      "pink"      "chocolate" "chocolate" "pink"      "chocolate"
[37] "brown"     "brown"     "brown"     "brown"     "brown"     "pink"
[43] "brown"     "brown"     "pink"      "pink"      "brown"     "chocolate"
[49] "black"     "pink"      "pink"      "chocolate" "chocolate" "chocolate"
[55] "chocolate" "pink"      "chocolate" "black"     "pink"      "chocolate"
[61] "pink"      "pink"      "chocolate" "pink"      "brown"     "brown"
[67] "pink"      "pink"      "pink"      "pink"      "black"     "black"
[73] "pink"      "pink"      "chocolate" "chocolate" "chocolate" "brown"
[79] "pink"      "brown"     "pink"      "pink"      "pink"      "black"
[85] "chocolate"

```

```

ggplot(candy) +
  aes(winpercent, pricepercent, label=rownames(candy)) +
  geom_point(col=my_cols) +
  geom_label(col=my_cols)

```

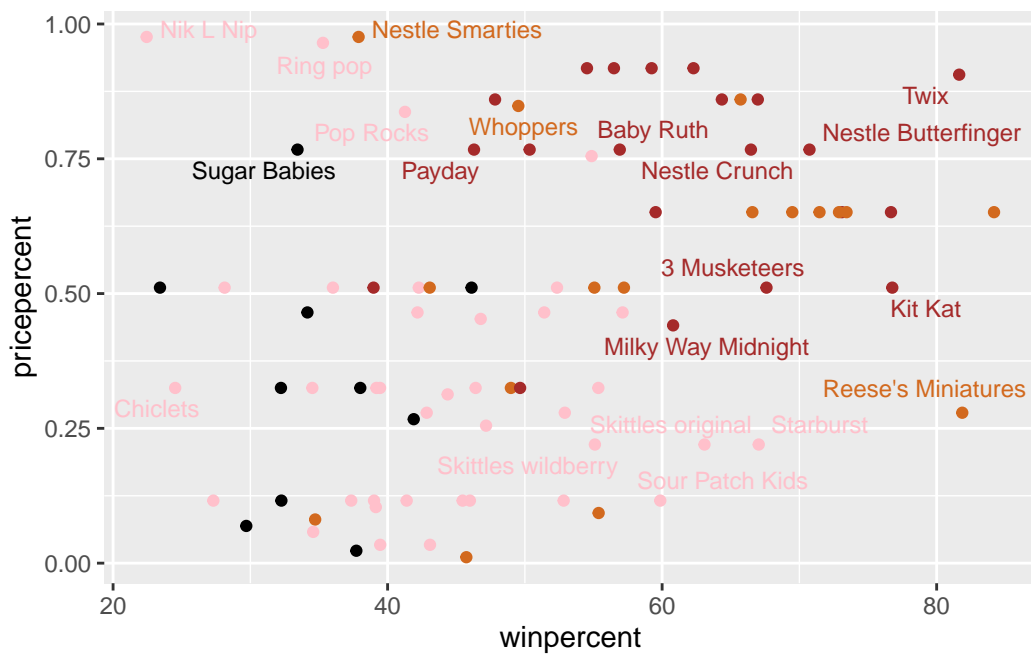


There are just too many labels in this above plot to be readable. We can use `ggrepel()` package to do a better job of placing these labels

```
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



## 5 Exploring the correlation structure

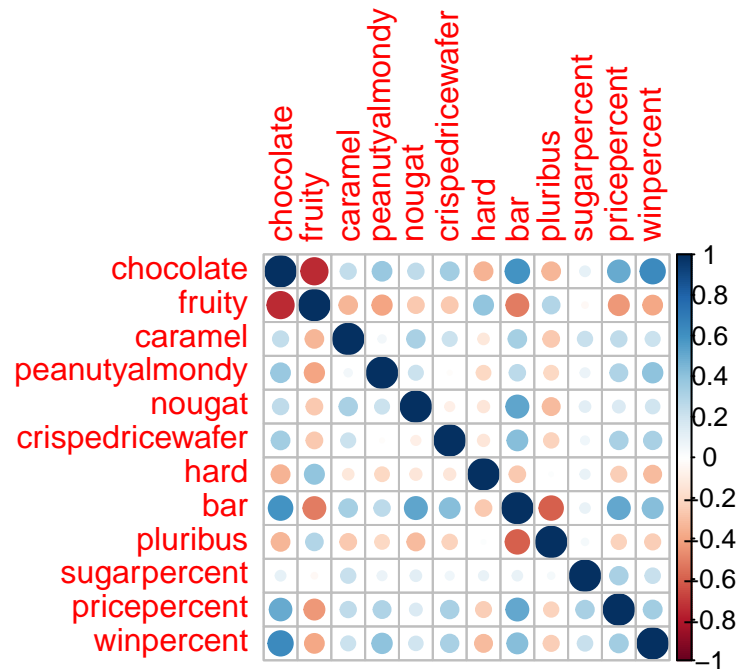
```
library(corrplot)
```

corrplot 0.92 loaded

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

	chocolate	fruity	caramel	peanutyalmondy	nougat
chocolate	1.0000000	-0.74172106	0.24987535	0.37782357	0.25489183
fruity	-0.7417211	1.00000000	-0.33548538	-0.39928014	-0.26936712
caramel	0.2498753	-0.33548538	1.00000000	0.05935614	0.32849280
peanutyalmondy	0.3778236	-0.39928014	0.05935614	1.00000000	0.21311310
nougat	0.2548918	-0.26936712	0.32849280	0.21311310	1.00000000
crispedricewafer	0.3412098	-0.26936712	0.21311310	-0.01764631	-0.08974359
hard	-0.3441769	0.39067750	-0.12235513	-0.20555661	-0.13867505
bar	0.5974211	-0.51506558	0.33396002	0.26041960	0.52297636
pluribus	-0.3396752	0.29972522	-0.26958501	-0.20610932	-0.31033884
sugarpercent	0.1041691	-0.03439296	0.22193335	0.08788927	0.12308135
pricepercent	0.5046754	-0.43096853	0.25432709	0.30915323	0.15319643
winpercent	0.6365167	-0.38093814	0.21341630	0.40619220	0.19937530
	crispedricewafer	hard	bar	pluribus	
chocolate	0.34120978	-0.34417691	0.59742114	-0.33967519	
fruity	-0.26936712	0.39067750	-0.51506558	0.29972522	
caramel	0.21311310	-0.12235513	0.33396002	-0.26958501	
peanutyalmondy	-0.01764631	-0.20555661	0.26041960	-0.20610932	
nougat	-0.08974359	-0.13867505	0.52297636	-0.31033884	
crispedricewafer	1.00000000	-0.13867505	0.42375093	-0.22469338	
hard	-0.13867505	1.00000000	-0.26516504	0.01453172	
bar	0.42375093	-0.26516504	1.00000000	-0.59340892	
pluribus	-0.22469338	0.01453172	-0.59340892	1.00000000	
sugarpercent	0.06994969	0.09180975	0.09998516	0.04552282	
pricepercent	0.32826539	-0.24436534	0.51840654	-0.22079363	
winpercent	0.32467965	-0.31038158	0.42992933	-0.24744787	
	sugarpercent	pricepercent	winpercent		
chocolate	0.10416906	0.5046754	0.6365167		
fruity	-0.03439296	-0.4309685	-0.3809381		
caramel	0.22193335	0.2543271	0.2134163		
peanutyalmondy	0.08788927	0.3091532	0.4061922		
nougat	0.12308135	0.1531964	0.1993753		
crispedricewafer	0.06994969	0.3282654	0.3246797		
hard	0.09180975	-0.2443653	-0.3103816		
bar	0.09998516	0.5184065	0.4299293		
pluribus	0.04552282	-0.2207936	-0.2474479		
sugarpercent	1.00000000	0.3297064	0.2291507		
pricepercent	0.32970639	1.0000000	0.3453254		
winpercent	0.22915066	0.3453254	1.0000000		

```
corrplot(cij)
```



## 6. Principal Component Analysis

We will perform a PCA of the candy. Key question: Do we need to scale the data before PCA?

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

Importance of components:

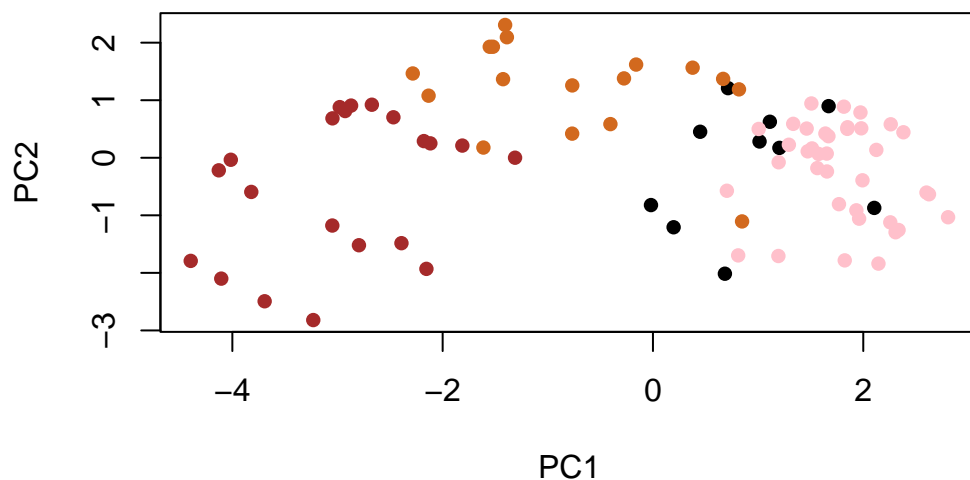
	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	2.0788	1.1378	1.1092	1.07533	0.9518	0.81923	0.81530
Proportion of Variance	0.3601	0.1079	0.1025	0.09636	0.0755	0.05593	0.05539
Cumulative Proportion	0.3601	0.4680	0.5705	0.66688	0.7424	0.79830	0.85369

	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



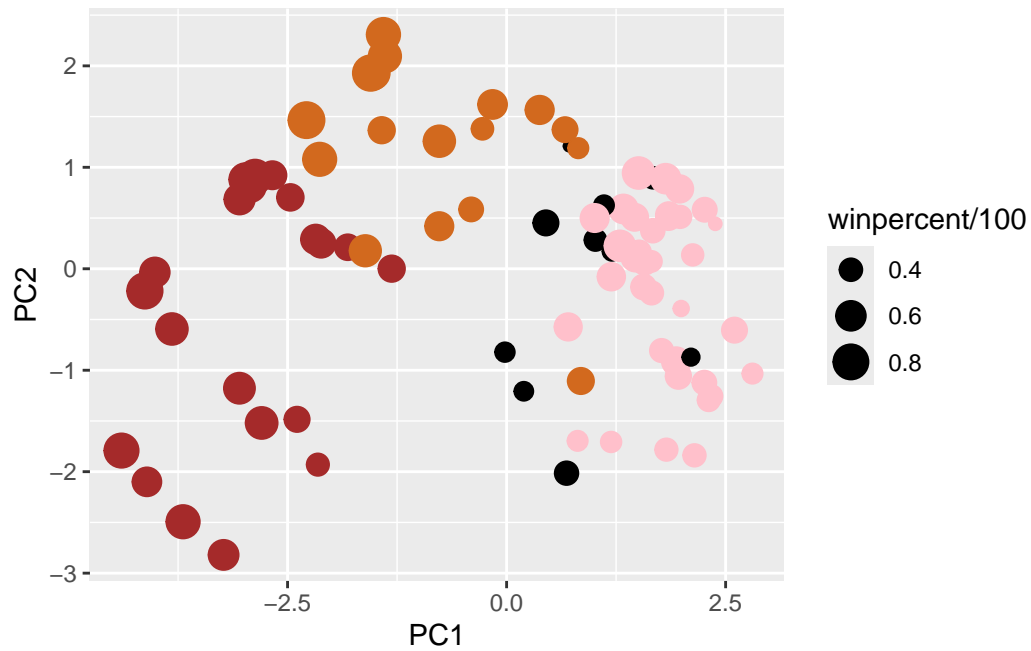
```
plot(pca$x[,1:2], col=my_cols, pch=16)
```



Make a ggplot version of this figure:

Make this more polished

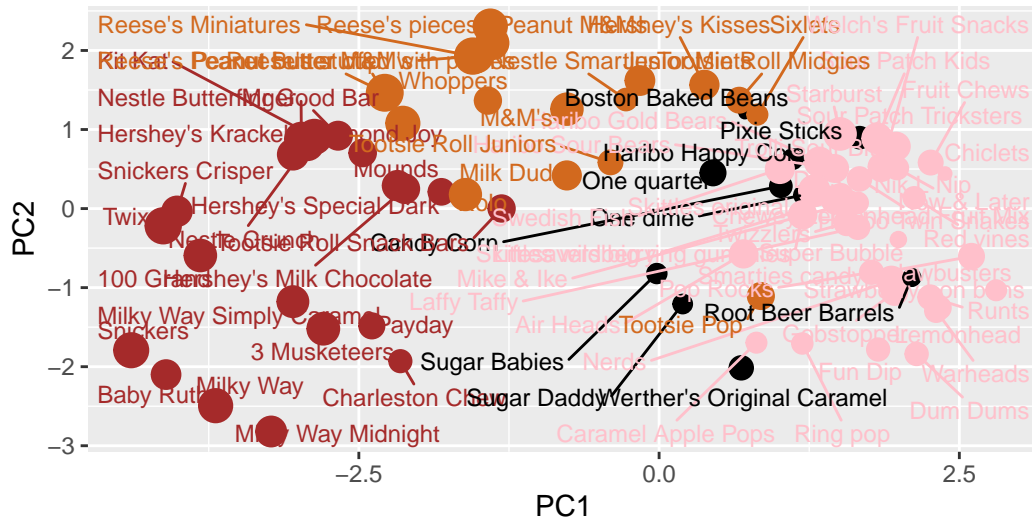
```
my_data <- cbind(candy, pca$x[,1:3])
p <- ggplot(my_data) +
  aes(x=PC1, y=PC2,
      size=winpercent/100,
      text=rownames(my_data),
      label=rownames(my_data)) +
  geom_point(col=my_cols)
p
```



```
p + geom_text_repel(size=3.3, col=my_cols, max.overlaps = 100) +
  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")
```

## Halloween Candy PCA Space

Colored by type: chocolate bar (dark brown), chocolate other (light brown),



Data from 538

Make this interactive with 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)
```

How do the original variables contribute to our PCs? For this we look at the loadings component of our results object ie. the `pca$rotation` object.

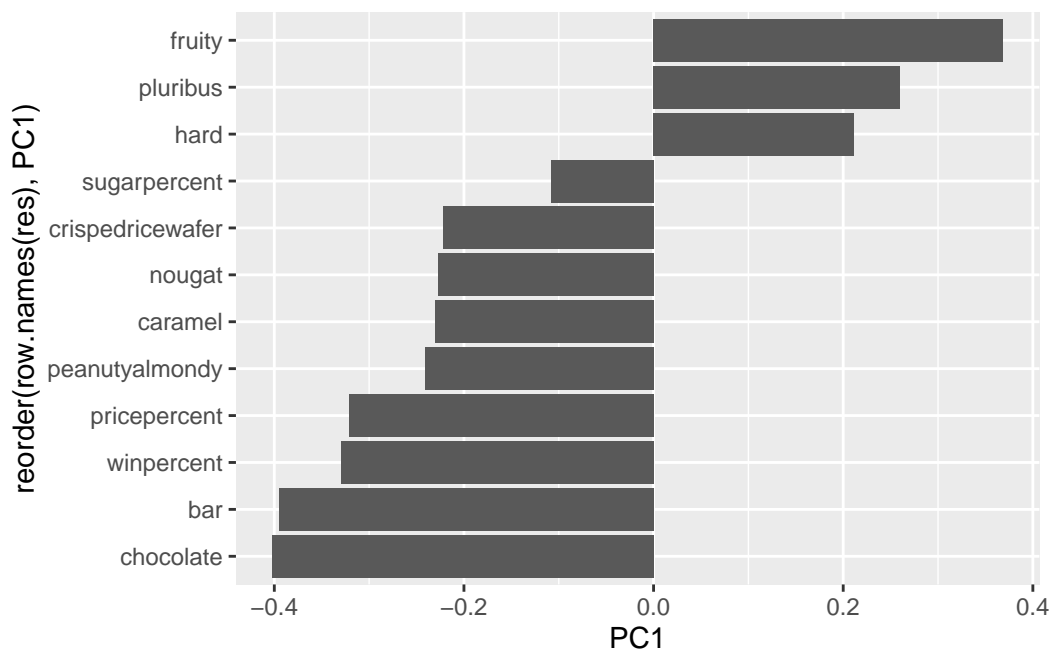
Make a barplot with ggplot and order the bars by their value. Recall that you need a data.frame as input for ggplot

```
res <- pca$rotation
```

```
row.names(res)
```

```
[1] "chocolate"      "fruity"          "caramel"         "peanutyalmondy"  
[5] "nougat"         "crispedricewafer" "hard"            "bar"  
[9] "pluribus"       "sugarpercent"    "pricepercent"    "winpercent"
```

```
ggplot(res) +  
  aes(PC1, reorder(row.names(res), PC1)) +  
  geom_col()
```



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

Fruit, Pluribus and hard are all picked up in the +ve direction. these make sense based on the correlation structure in the dataset. If you are fruity candy, you will tend to be hard and come in a pack of multiple candies in it.