Class10: Halloween Mini Project

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1. Importing Candy Data

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

candy = read.csv(candy_file, row.names=1)
head(candy)</pre>
```

	choco	olate	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	O	0	0
Air Heads		0	1	0	O	0	0
Almond Joy		1	0	0	1	. 0	0
	hard	bar j	pluribus	sugarpe	ercent priceper	cent wi	npercent
100 Grand	0	1	C)	0.732	.860	66.97173
3 Musketeers	0	1	C)	0.604	.511	67.60294
One dime	0	0	C)	0.011	.116	32.26109
One quarter	0	0	C)	0.011	.511	46.11650
Air Heads	0	0	C)	0.906	.511	52.34146
Almond Joy	0	1	C)	0.465	.767	50.34755

Q1. How many different candy types are in this dataset? A. There are 85 different candy types.

```
nrow(candy)
```

[1] 85

Q2. How many fruity candy types are in the dataset? A. There are 38 fruity candy types.

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

[1] 38

2. What is your favorite candy?

Q3. What is your favorite candy in the dataset and what is it's winpercent value? A. Snickers; 76.67378

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

[1] 76.67378

Q4. What is the winpercent value for "Kit Kat"? A. 76.7686 Q5. What is the winpercent value for "Tootsie Roll Snack Bars"? A. 49.6535

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

[1] 76.7686

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

[1] 49.6535

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(rownames(candy) %in% c("Kit Kat", "Tootsie Roll Snack Bars")) |>
  select(winpercent)
```

winpercent

Kit Kat 76.7686 Tootsie Roll Snack Bars 49.6535

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

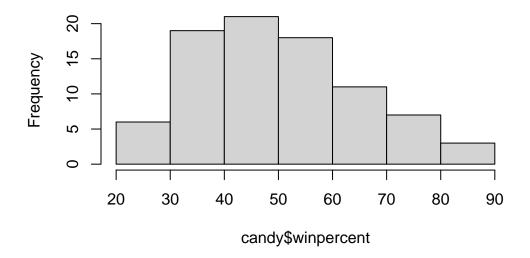
skim_variable n_	_missingcomp	olete_ra	atmenean	sd	p0	p25	p50	p75	p100	hist
chocolate	0	1	0.44	0.50	0.00	0.00	0.00	1.00	1.00	
fruity	0	1	0.45	0.50	0.00	0.00	0.00	1.00	1.00	
caramel	0	1	0.16	0.37	0.00	0.00	0.00	0.00	1.00	
peanutyalmondy	0	1	0.16	0.37	0.00	0.00	0.00	0.00	1.00	
nougat	0	1	0.08	0.28	0.00	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? A. winpercent is on a different scale.

Q7. What do you think a zero and one represent for the candy\$\chocolate \column? A. The 1 means that the candy has chocolate in it.

```
inds <- candy$chocolate== 1</pre>
choc.win <- candy[inds,]$winpercent</pre>
inds <- candy$fruity== 1</pre>
fruit.win <- candy[inds,]$winpercent</pre>
summary(choc.win)
   Min. 1st Qu. Median
                           Mean 3rd Qu.
                                            Max.
  34.72
          50.35 60.80
                                           84.18
                          60.92 70.74
summary(fruit.win)
  Min. 1st Qu. Median
                          Mean 3rd Qu.
                                            Max.
          39.04 42.97
  22.45
                          44.12 52.11
                                           67.04
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
     Q8. Plot a histogram of winpercent values
hist(candy$winpercent)
```

Histogram of candy\$winpercent



Q9. Is the distribution of winpercent values symmetrical? A. No it is not symmetrical.

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

Q11. On average is chocolate candy higher or lower ranked than fruit candy? A. The chocolate (60) is ranked higher than the candy (44).

Q12. Is this difference statistically significant? A. Yes, the difference is statistically significant.

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

[1] 60.92153

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

[1] 44.11974

3. Overall Candy Rankings

Q13. What are the five least liked candy types in this set? A. Nik L Nip, Boston Baked Beans, Chiclets, Super Bubble, Jawbusters

```
inds <- order((candy$winpercent))
head(candy[inds,], n=5)</pre>
```

				_				
	chocolate	fruity	cara	nel j	peanutyaln	nondy :	nougat	
Nik L Nip	0	1		0		0	0	
Boston Baked Bean	в 0	0		0		1	0	
Chiclets	0	1		0		0	0	
Super Bubble	0	1		0		0	0	
Jawbusters	0	1		0		0	0	
	crispedri	cewafer	hard	bar	pluribus	sugar	percent	pricepercent
Nik L Nip		0	0	0	1		0.197	0.976
Boston Baked Bean	3	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	t						
Nik L Nip	22.4453	4						
Boston Baked Bean	23.41782	2						
Chiclets	24.52499	9						
Super Bubble	27.30386	3						
Jawbusters	28.1274	1						

Q14. What are the top 5 all time favorite candy types out of this set? A. Reese's Peanut Butter cup, Reese's Miniatures, Twix, Kit Kat, and Snickers.

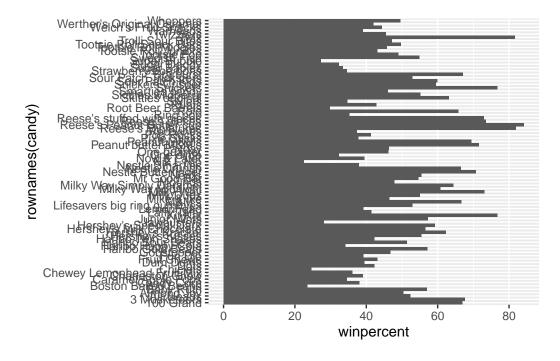
```
inds2 <- order(desc(candy$winpercent))
head(candy[inds2,], n= 5)</pre>
```

	chocolate	fruity	caran	nel	peanutyalm	nondy	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
	crispedrio	cewafer	hard	bar	pluribus	sugai	rpercent
Reese's Peanut Butter cup		0	0	0	0		0.720

Reese's 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	winpe	rcent	;		
Reese's Peanut Butter cup	0.651	84.	18029)		
Reese's Miniatures	0.279	81.	86626	3		
Twix	0.906	81.	64291	-		
Kit Kat	0.511	76.	76860)		
Snickers	0.651	76.	67378	}		

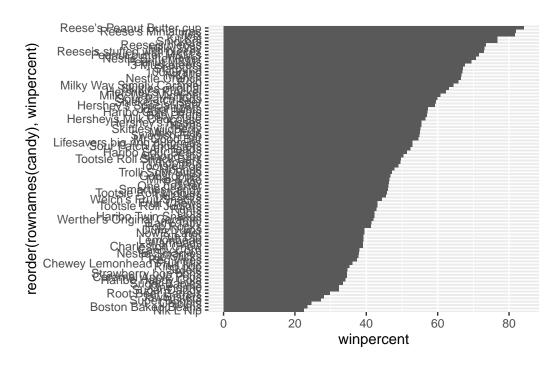
Q15. Make a first barplot of candy ranking based on winpercent values.

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



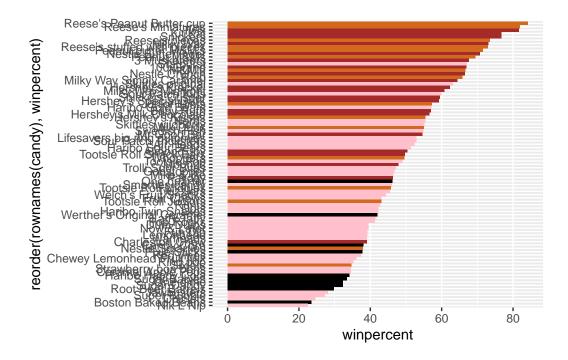
Q16. This is quite ugly, use the reorder() function to get the bars sorted by winpercent?

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



```
#Adding useful colors
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)
```

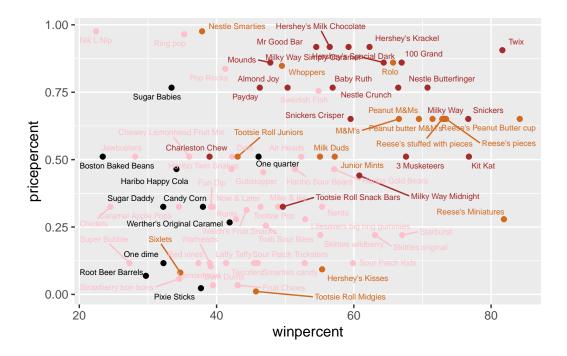


Q17. What is the worst ranked chocolate candy? A. The worst ranked chocolate candy is Sixlets. Q18. What is the best ranked fruity candy? A. The best ranked fruity candy is Starburst.

4. Taking a look at 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=2, max.overlaps = 80)
```



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? A. Tootsie Roll Midgies

```
order <- order((candy$winpercent/candy$pricepercent), decreasing = TRUE)
head(candy[order,c(11,12)], n=1)</pre>
```

```
pricepercent winpercent Tootsie Roll Midgies 0.011 45.73675
```

Q20. What are the top 5 most expensive candy types in the dataset and of these which is the least popular?

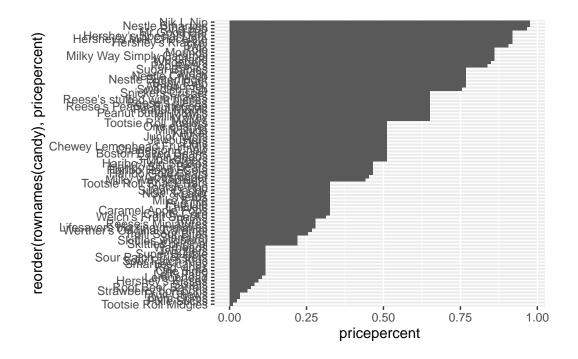
A. The 5 most expensive candies are Nik L Nip, Nestle Smarties, Ring pop, Hershey's Krackel, and Hershey's Milk Chocolate. The least popular is Nik L Nip.

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

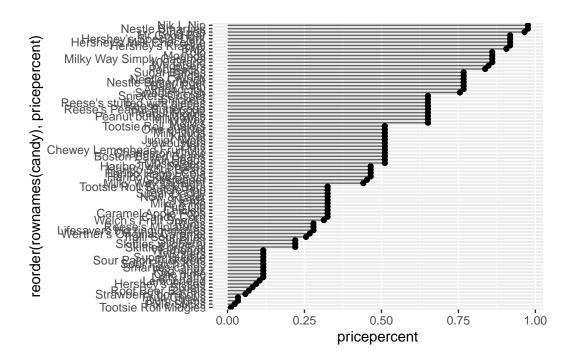
	pricepercent	winpercent
Nik L Nip	0.976	22.44534
Nestle Smarties	0.976	37.88719
Ring pop	0.965	35.29076
Hershey's Krackel	0.918	62.28448
Hershey's Milk Chocolate	0.918	56.49050

#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().

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



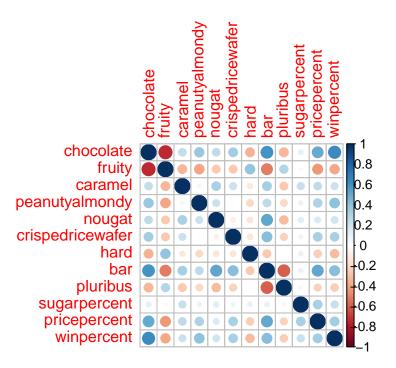
Make a lollipop chart of pricepercent



5. Exploring the Correlation structure

```
library(corrplot)
```

corrplot 0.92 loaded



Q22. Examining this plot what two variables are anti-correlated (i.e. have minus values)? A. The fruity and chocolate are anti-correlated. Q23. Similarly, what two variables are most positively correlated? A. Winpercent and bar with chocolate are the most positively correlated.

6. Principal Component Analysis

```
pca <- prcomp(candy, scale. = TRUE)
summary(pca)</pre>
```

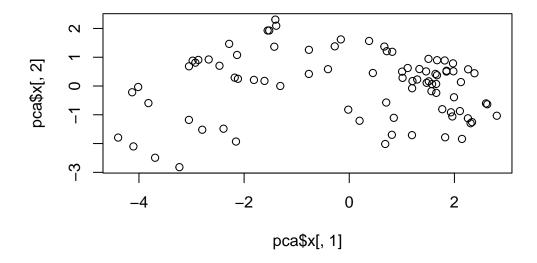
Importance of components:

```
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
```

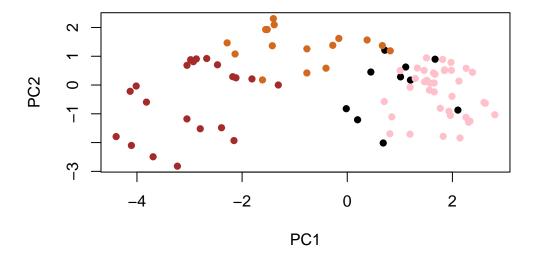
pca\$rotation[,1]

peanutyalmondy	caramel	fruity	chocolate
-0.2407155	-0.2299709	0.3683883	-0.4019466
bar	hard	crispedricewafer	nougat
-0.3947433	0.2111587	-0.2215182	-0.2268102
winpercent	pricepercent	sugarpercent	pluribus
-0.3298035	-0.3207361	-0.1083088	0.2600041

#Now we can plot our main PCA score plot of PC1 vs PC2.
plot(pca\$x[, 1], pca\$x[, 2])

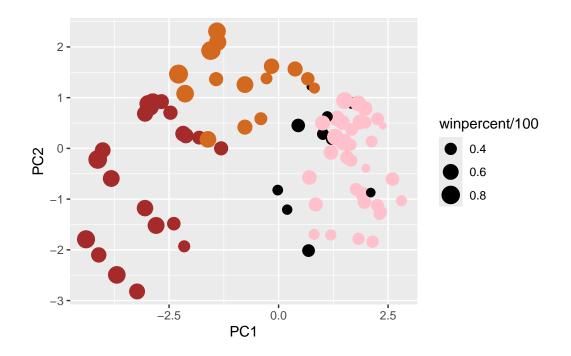


#We can change the plotting character and add some color: $plot(pca$x[,1:2], col=my_cols, pch=16)$



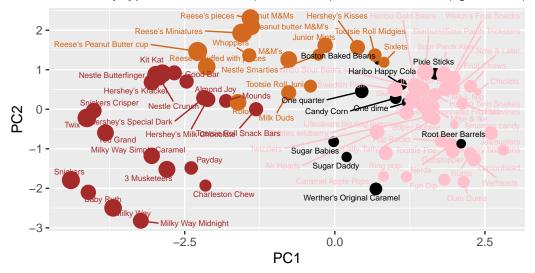
We can make a much nicer plot with the ggplot 2 $\,$

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



Halloween Candy PCA Space

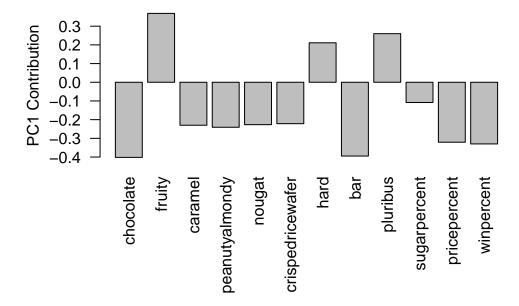
Colored by type: chocolate bar (dark brown), chocolate other (light brown),



Data from 538

```
#library(plotly)
#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? A. Fruity candy, hard candy, and pluribus candy(coming from a bag or box of multiple candy). Yes these variable make sense.