# Class 10: Halloween Mini-Project

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### Import candy data

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

	choco	nlate	fruitw	caramel	neanut	yalmondy	nougat	crisnedr	icewafer
100 Grand	CHOCC	1	114109	1	pcanac	^	nougut	cribpear	1
100 Grand		т	U	1		0	U		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 j	pluribus	sugarpe	ercent	priceper	cent wi	npercent	
100 Grand	0	1	O	)	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. How many fruity candy types are in the dataset?

### sum(candy\$fruity)

[1] 38

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

```
candy["Haribo Gold Bears", ]$winpercent
```

#### [1] 57.11974

```
# candy %>%
# filter(rownames(candy)=="Haribo Gold Bears") %>%
# select(winpercent)
```

Q3a. Find fruity candy with a winpercent above 50%

### 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(winpercent >= 50) %>%
  filter(fruity==1)
```

	chocolate	iruity	caramel	peanutyalmondy	nougat
Air Heads	0	1	0	0	0
Haribo Gold Bears	0	1	0	0	0
Haribo Sour Bears	0	1	0	0	0
Lifesavers big ring gummies	0	1	0	0	0
Nerds	0	1	0	0	0
Skittles original	0	1	0	0	0
Skittles wildberry	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
Swedish Fish	0	1		0		0	0
	crispedricewa	afer	hard	bar	pluribus	sugarp	ercent
Air Heads		0	0	0	0		0.906
Haribo Gold Bears		0	0	0	1		0.465
Haribo Sour Bears		0	0	0	1		0.465
Lifesavers big ring gummies		0	0	0	0		0.267
Nerds		0	1	0	1		0.848
Skittles original		0	0	0	1		0.941
Skittles wildberry		0	0	0	1		0.941
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
Swedish Fish		0	0	0	1		0.604
	pricepercent	winj	percer	nt			
Air Heads	0.511	53	2.3414	<del>1</del> 6			
Haribo Gold Bears	0.465	5	7.1197	74			
Haribo Sour Bears	0.465	5:	1.4124	13			
Lifesavers big ring gummies	0.279	52	2.9113	39			
Nerds	0.325	5!	5.3540	)5			
Skittles original	0.220	63	3.0851	L4			
Skittles wildberry	0.220	5!	5.1037	70			
Sour Patch Kids	0.116	59	9.8640	00			
Sour Patch Tricksters	0.116	52	2.8259	95			
Starburst	0.220	6	7.0376	33			
Swedish Fish	0.755	54	4.8611	L1			

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

Side-note: the skimr::skim() function

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_	_missingcom	plete_ra	ntanean	$\operatorname{sd}$	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?

Yes, winpercent looks to be on a different scale to the majority of the other columns in the dataset. This requires me to scale my data before doing any analysis like PCA etc.

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

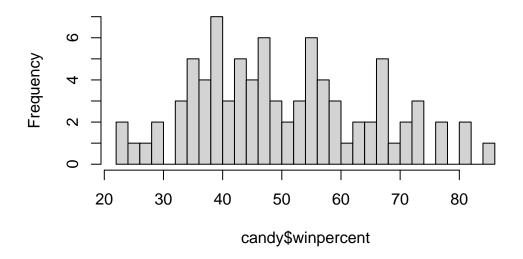
### candy\$chocolate

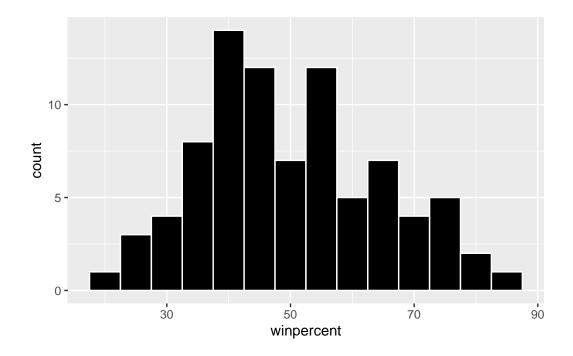
It represents T or F in a logical variable.

Q8. Plot a histogram of winpercent values

hist(candy\$winpercent, breaks=30)

## **Histogram of candy\$winpercent**





Q9. Is the distribution of winpercent values symmetrical?

NO

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
```

The center of the distribution, also shown as the median, is below 50%.

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

```
fruit_candy <- candy %>%
  filter(fruity==1)

summary(fruit_candy$winpercent)
```

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

```
chocolate_candy <- candy %>%
  filter(chocolate==1)
summary(chocolate_candy$winpercent)
```

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

```
# base R code for the one above
# summary(candy[as.logical(candy$chocolate),]$winpercent)
```

Chocolate cady is higher ranked than fruit candy, on average.

Q12. Is this difference statistically significant?

```
t.test(chocolate_candy$winpercent, fruit_candy$winpercent)
```

```
Welch Two Sample t-test
```

```
data: chocolate_candy$winpercent and fruit_candy$winpercent
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?

```
play <- c("d", "a", "c")
sort(play)</pre>
```

```
[1] "a" "c" "d"
```

### order(play)

### [1] 2 3 1

sort() organises the output in order, but order() is more useful because it specifically shows
you the ranking position of the variable (candy)

### head( candy[order(candy\$winpercent), ], 5)

	chocolate	fruity	caran	ו [בת	naaniitisalr	nondsz	nougat	
Nik L Nip	0	114169	caran	0	peanucyari	0	nougat 0	
-	-	1		_		4	_	
Boston Baked Beans		0		0		1	0	
Chiclets	0	1		0		0	0	
Super Bubble	0	1		0		0	0	
Jawbusters	0	1		0		0	0	
	crispedrio	cewafer	${\tt hard}$	bar	pluribus	sugai	rpercent	pricepercent
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	<del>l</del>						
Boston Baked Beans	23.41782	2						
Chiclets	24.52499	)						
Super Bubble	27.30386	3						
Jawbusters	28.12744	l.						

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

### tail( candy[order(candy\$winpercent), ], 5)

	${\tt chocolate}$	fruity	cara	nel	peanutyalm	nondy	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
	crispedrio	cewafer	${\tt hard}$	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
                                          0
Reese's Miniatures
                                               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
```

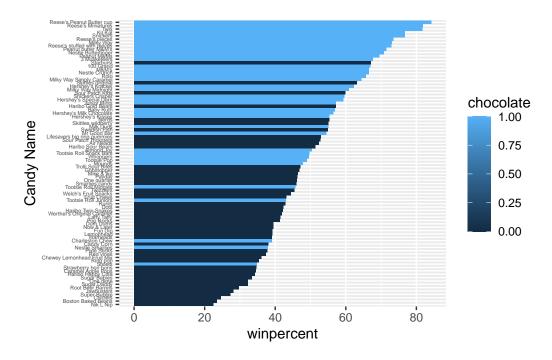
```
head( candy[order(candy$winpercent, decreasing = T), ], 5)
```

		chocolate	fruity	carar	nel	peanutyaln	nondy	nougat
Reese's Peanut Butter o	up	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	${\tt hard}$	bar	pluribus	sugai	rpercent
Reese's Peanut Butter o	up		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
		priceperce	ent win	percer	nt			
Reese's Peanut Butter o	up	0.6	351 8 <sup>4</sup>	1.1802	29			
Reese's Miniatures		0.2	279 8:	1.8662	26			
Twix		0.9	906 8:	1.6429	91			
Kit Kat		0.5	511 76	5.7686	30			
Snickers		0.6	351 76	6.6737	78			

Let's do a barplot of winpercent values

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

```
ggplot(candy, aes(x = winpercent, y = reorder(row.names(candy), winpercent), fill = chocolate
geom_col() +
labs(y="Candy Name") +
theme(axis.text.y = element_text(size = 4))
```



I want a more custom color scheme where I can see both chocolate and bar and fruity etc. all from the one plot. To do this we can roll our own color vector...

```
library(wesanderson)
# Place holder color vector
# rep()
mycols <- rep("black", nrow(candy))
# I want 2nd entry of my dataset to be blue

mycols[as.logical(candy$bar)] <- "brown"
mycols[as.logical(candy$chocolate)] <- "chocolate"
mycols[as.logical(candy$fruity)] <- "pink"

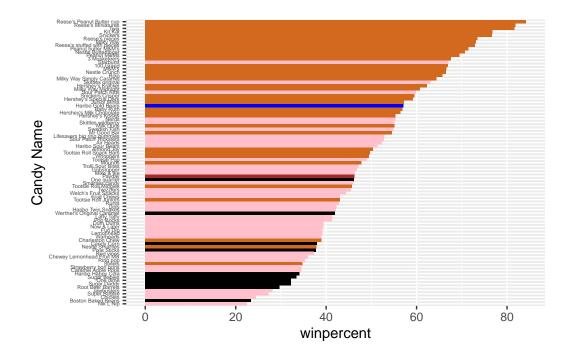
# Use blue for your favorite candy
rownames(candy) == "Haribo Gold Bears"</pre>
```

[1] FALSE FA

[73] FALSE F

```
mycols[rownames(candy)=="Haribo Gold Bears"] <- "blue"</pre>
```

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



Q17. What is the worst ranked chocolate candy?

Sixlets

Q18. What is the best ranked fruity candy?

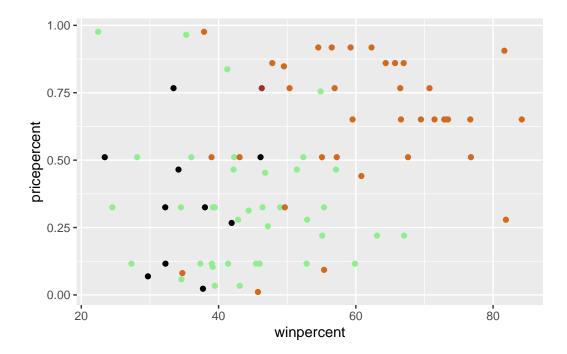
Starbursts

### **Price Percent**

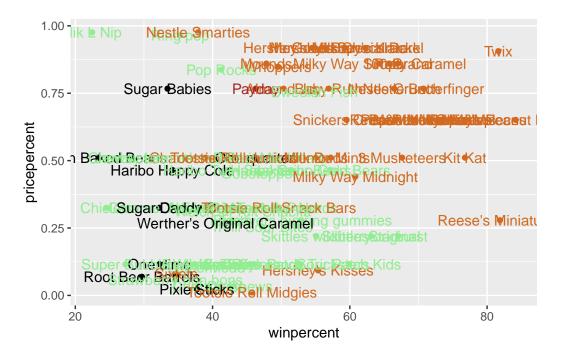
Plot a winpercent vs pricepercent to see what would be the best candy to buy...

```
mycols[as.logical(candy$fruity)] <- "lightgreen"</pre>
```

```
ggplot(candy,
    aes(x = winpercent,
        y = pricepercent)) +
    geom_point(col = mycols)
```



add labels

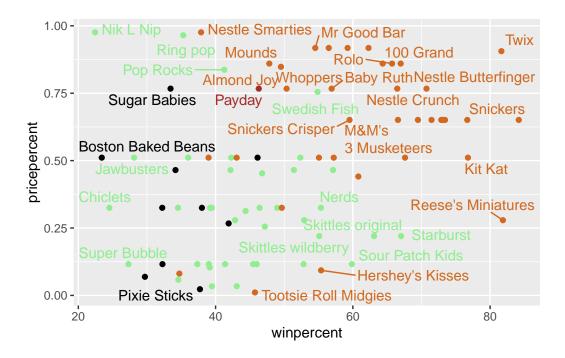


Make the labels non-overlapping

```
library(ggrepel)

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

Warning: ggrepel: 50 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 minatures

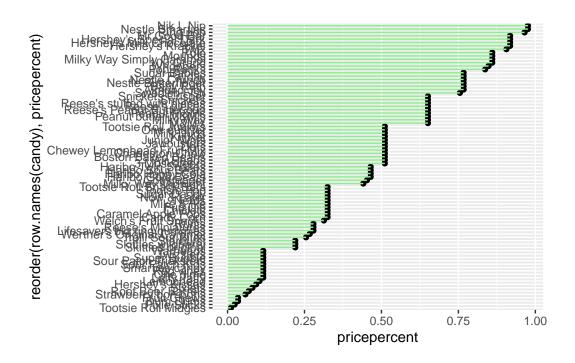
Q20. What are the top 5 most expensive candy types in the dataset and of these which is the least popular?

Nik L Nip, Nestle Smarties, Ring Pop, Hershey's Krackel, Hershey's Milk Chocolate

```
ord <- order(candy$pricepercent, decreasing = T)
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().



### Correlation

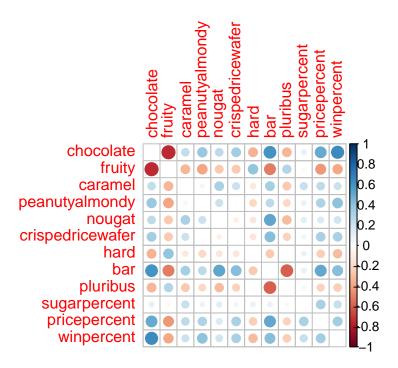
```
library(corrplot)
```

corrplot 0.95 loaded

```
cij <- cor(candy)
cij</pre>
```

```
chocolate
                                fruity
                                           caramel peanutyalmondy
                                                                       nougat
chocolate
                 1.0000000 -0.74172106 0.24987535
                                                       0.37782357 0.25489183
                -0.7417211 1.00000000 -0.33548538
fruity
                                                      -0.39928014 -0.26936712
                 0.2498753 -0.33548538 1.00000000
                                                       0.05935614 0.32849280
caramel
peanutyalmondy
                 0.3778236 -0.39928014 0.05935614
                                                       1.00000000 0.21311310
```

```
0.2548918 -0.26936712 0.32849280
                                                    0.21311310 1.00000000
nougat
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
               -0.3396752 0.29972522 -0.26958501
pluribus
                                                   -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
                    0.21311310 -0.12235513 0.33396002 -0.26958501
caramel
peanutyalmondy
                    -0.01764631 -0.20555661 0.26041960 -0.20610932
nougat
                    -0.08974359 -0.13867505 0.52297636 -0.31033884
crispedricewafer
                     hard
                    -0.13867505
                               1.00000000 -0.26516504 0.01453172
bar
                     0.42375093 -0.26516504 1.00000000 -0.59340892
pluribus
                    0.06994969 0.09180975
                                           0.09998516 0.04552282
sugarpercent
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
                -0.03439296
                             -0.4309685 -0.3809381
fruity
caramel
                 0.22193335
                              0.2543271 0.2134163
peanutyalmondy
                 0.08788927
                              0.3091532 0.4061922
                              0.1531964 0.1993753
nougat
                 0.12308135
crispedricewafer
                 0.06994969
                              0.3282654 0.3246797
hard
                             -0.2443653 -0.3103816
                 0.09180975
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, diag = F)
```



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?

Variables that are identical, variable compared to themselves (i.e. Chocolate=Chocolate, fruity=fruity) Other than that, Chocolate and bar or Chocolate and winpercent or bar and pricepercent or bar and nougat

### **PCA**

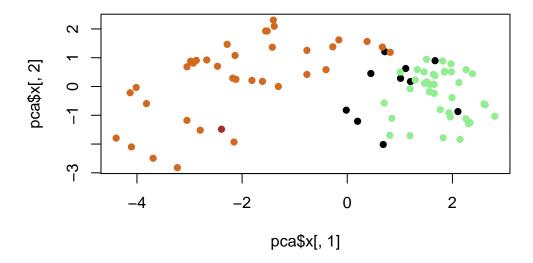
```
pca <- prcomp(candy, scale. = TRUE)
summary(pca)</pre>
```

#### Importance of components:

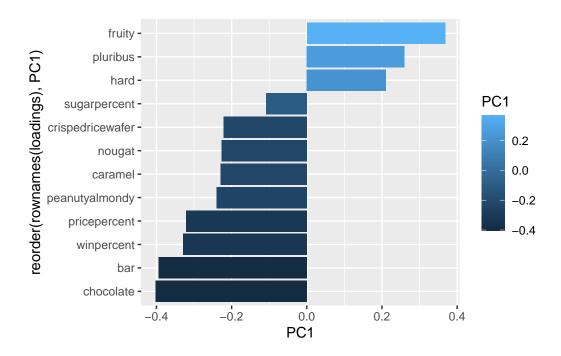
```
PC1
                                  PC2
                                         PC3
                                                 PC4
                                                         PC5
                                                                 PC6
                                                                         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
```

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

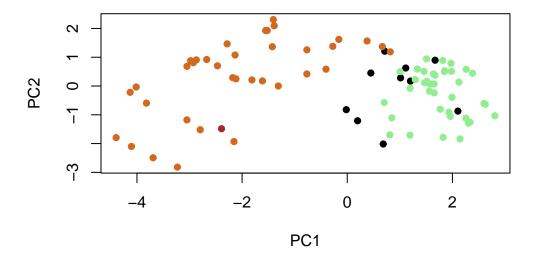


How do the original variable (columns) contribute to the new PCs? I will look at PC1 here:

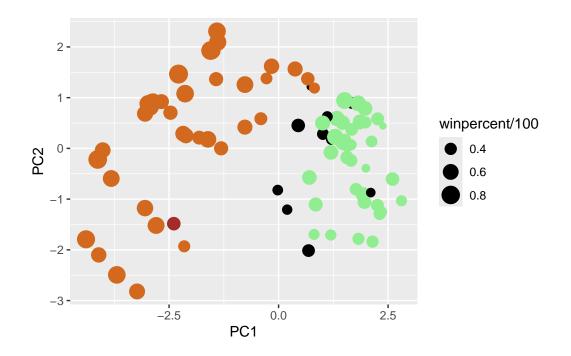


Q24. What original variables are picked up strongly by PC1 in the positive direction? Do these make sense to you? Fruity, pluribus, and hard variables are picked up strongly by PC1 in the positive direction. This makes sense becuase it correlates to the previous correlation graph that told us the three variable, shown in the positive direction, positively correlate with each other. Similarly, variables "chocolate" and "bar" being heavily in the negative direction makes sense since the correlation graph told us that those two variables correlate with each other well.

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



```
my_data <- cbind(candy, pca$x[,1:3])</pre>
```



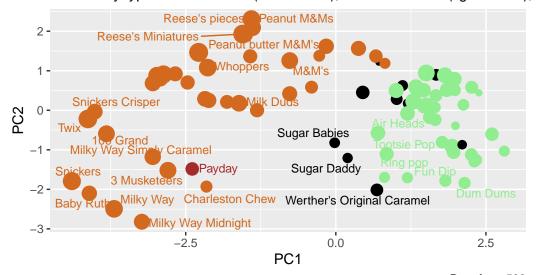
```
library(ggrepel)

p + geom_text_repel(size=3.3, col=mycols, 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)

```
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)
```