Class10 Halloween

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Explore the dataset

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

0 0

0 1

	choco	olate	fruity	caramel	peanut	yalmondy	nougat	crispedricewa	fer
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
	${\tt hard}$	bar	pluribus	sugarpe	ercent	priceper	cent wi	npercent	
100 Grand	0	1	()	0.732	0	.860	66.97173	
3 Musketeers	0	1	()	0.604	0	.511	67.60294	
One dime	0	0	()	0.011	0	.116	32.26109	
One quarter	0	0	()	0.011	0	.511	16.11650	

0

0.906

0.465

0.511

0.767

52.34146

50.34755

Q1

Almond Joy

Air Heads

```
nrow(candy)
```

[1] 85

There are 85 types of candies.

[1] 49.6535

library(skimr)
skim(candy)

inspect dataset with skim
install.packages('skim')

```
Q2
  sum(candy['fruity']==1)
[1] 38
There are 38 types of fruity candies.
Q3
  candy['Air Heads',]$winpercent
[1] 52.34146
Q4
  candy['Kit Kat',]$winpercent
[1] 76.7686
Q5
  candy['Tootsie Roll Snack Bars',]$winpercent
```

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	ntmenean	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

The variable winpercent is on a different scale.

Q7

<code>n_missing</code> is the number of NA entries, and <code>complete_rate</code> is how many of the entries are not NA.

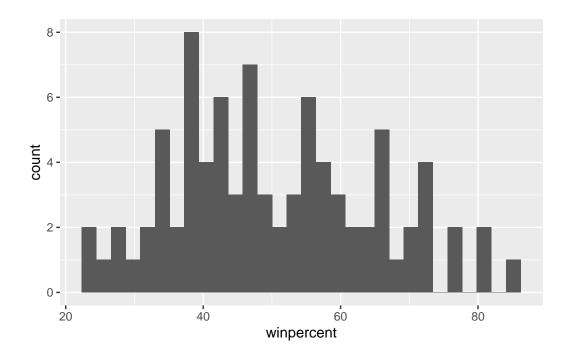
Examine the winpercent variable

Q8

```
library(ggplot2)

ggplot(candy)+
  aes(x=winpercent)+
  geom_histogram()
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



Q9

The distribution is not symmetrical.

Q10

The center of distribution is below 50%.

```
[1] 60.92153
           mean(candy$winpercent[as.logical(candy$fruity)])
 [1] 44.11974
 On average chocolate candies rank higher.
 Q12
           \verb|t.test(x=candy\$winpercent[as.logical(candy\$chocolate)]|, | y=candy\$winpercent[as.logical(candy\$chocolate)]| | y=candy\$winpercent[as.logical(candy§chocolate)]| | y=candy\$winpercent[as.logical(candy§chocolate
                  Welch Two Sample t-test
data: candy$winpercent[as.logical(candy$chocolate)] and candy$winpercent[as.logical(candy$f
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
           library(dplyr)
```

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

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 %>% arrange(winpercent) %>% head(5)

	cho	colate	fruity	caram	nel j	peanutyaln	nondy	nougat	
Nik L Nip		0	1		0		0	0	
Boston Baked Bea	ans	0	0		0		1	0	
Chiclets		0	1		0		0	0	
Super Bubble		0	1		0		0	0	
Jawbusters		0	1		0		0	0	
	cri	spedrio	cewafer	hard	bar	pluribus	sugar	percent	pricepercent
Nik L Nip			0	0	0	1		0.197	0.976
Boston Baked Bea	ans		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
	win	percent	t						

 Nik L Nip
 22.44534

 Boston Baked Beans
 23.41782

 Chiclets
 24.52499

 Super Bubble
 27.30386

 Jawbusters
 28.12744

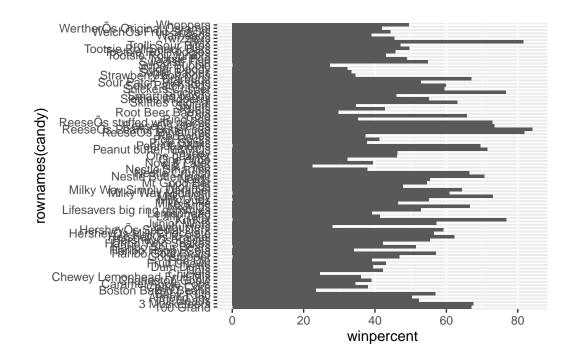
Q14

candy %>% arrange(winpercent) %>% tail(5)

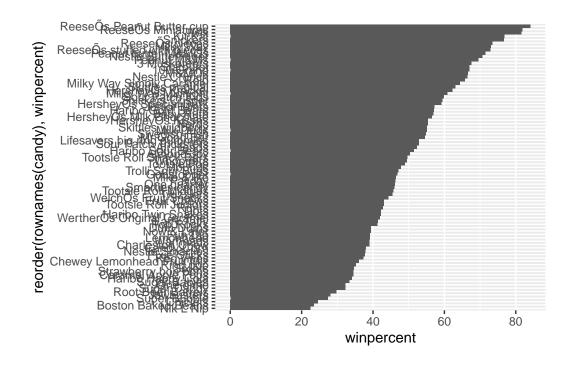
	chocolate	fruity	caramel	peanutyalmondy	nougat
Snickers	1	0	1	1	1
Kit Kat	1	0	0	0	0
Twix	1	0	1	0	0
ReeseÕs Miniatures	1	0	0	1	0
ReeseÕs Peanut Butter cup	1	0	0	1	0

	crispedricewa	afer	hard	bar	pluribus	sugarpercent
Snickers		0	0	1	0	0.546
Kit Kat		1	0	1	0	0.313
Twix		1	0	1	0	0.546
ReeseÕs Miniatures		0	0	0	0	0.034
ReeseÕs Peanut Butter cup		0	0	0	0	0.720
	pricepercent	win	percer	nt		
Snickers	0.651	76	6.6737	78		
Kit Kat	0.511	76	6.7686	30		
Twix	0.906	8:	1.6429	91		
ReeseÕs Miniatures	0.279	8:	1.8662	26		
ReeseÕs Peanut Butter cup	0.651	84	4.1802	29		

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

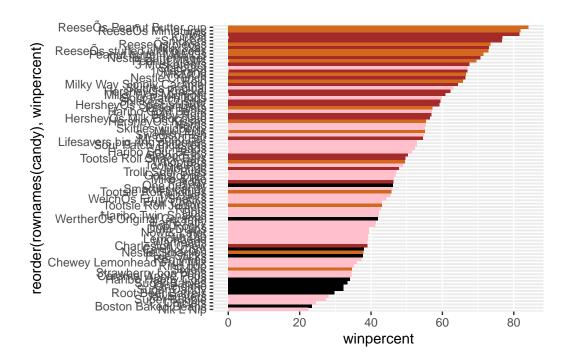


```
ggplot(candy)+
  aes(winpercent, reorder(rownames(candy), winpercent))+
  geom_col()
```



```
# prepare a sequence of colors by candy type
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)
```



The worst chocolate candy is Charleston Chew.

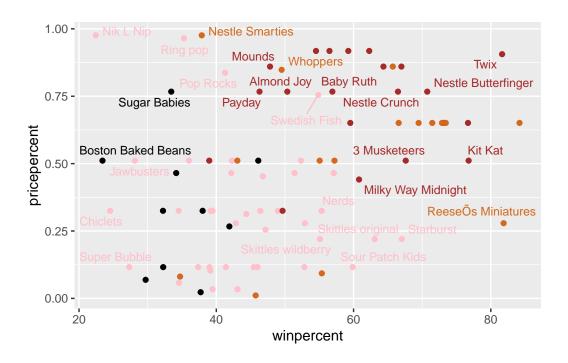
Q18

The worst fruity candy is Nik L Nip.

Examine the pricepercent variable

```
# plot winpercent against pricepercent to assess the candies
# install.packages('ggrepel')

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=7)
```



It's the ReeseOs Miniatures.

Q20

```
head(candy[order(candy$pricepercent, decreasing = T), c(11, 12)], n=5)
```

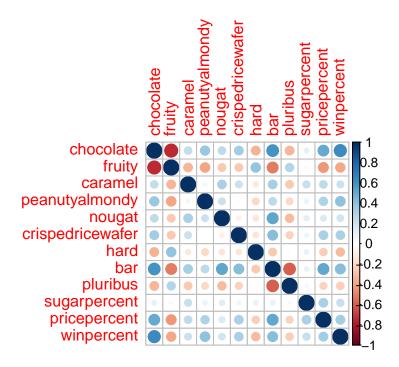
	pricepercent	winpercent
Nik L Nip	0.976	22.44534
Nestle Smarties	0.976	37.88719
Ring pop	0.965	35.29076
HersheyÕs Krackel	0.918	62.28448
HersheyÕs Milk Chocolate	0.918	56.49050

Correlation Structure

```
# examine correlations between the variables with corrplot
# install.packages('corrplot')
library(corrplot)
```

corrplot 0.92 loaded

```
pwc <- cor(candy)
corrplot(pwc)</pre>
```



Q22

chocolate and fruity, bar and pluribus, bar and fruity are evidently anticorrelated.

Q23

chocolate and winpercent seem to be most postively correlated.

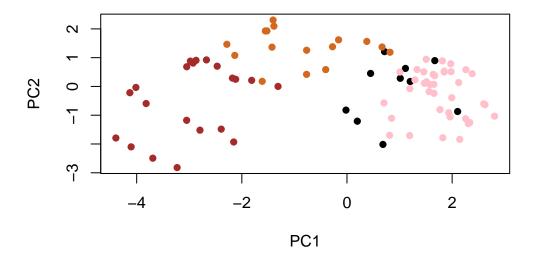
PCA

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

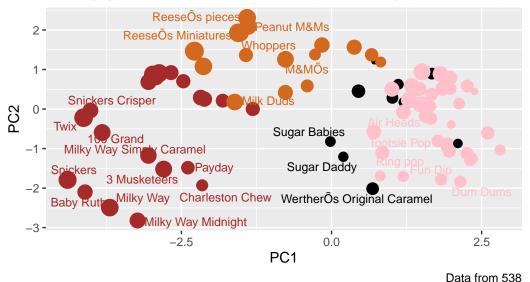
```
plot(pca$x[,1:2], col=my_cols, pch=16)
```



```
# plot PCs and see how they relate to variables of interest
# include text and label for ggrepel
my_data <- cbind(candy, pca$x[, 1:3])</pre>
```

Halloween Candy PCA

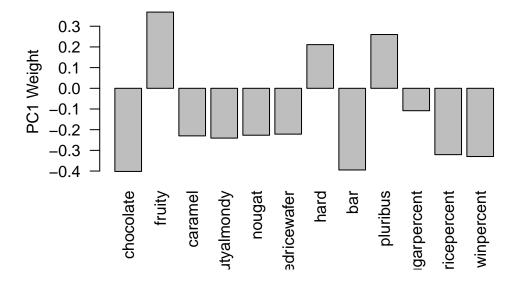
Color by type: chocolate (chocolate), bar (brown), fruity (pink), others (black



interactive plot with plotly
install.packages('plotly')
#library(plotly)
#ggplotly(p)
variable contribution to PC1

par(margin=c(8,4,2,2))

Warning in par(margin = c(8, 4, 2, 2)): "margin" is not a graphical parameter



Fruity, hard and pluribus are shown to have positive contributions. Since PC1 largely separates the 3 types fruity, chocolate and bar, it makes sense that fruity shows strong positive contribution. Pluribus seems to be positively correlated with fruity and makes similar contribution as fruity candy are usually pluribus, whereas chocolates are usually packed as bars.