

class10

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

```
library(dplyr)
```

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

```
dim(candy)
```

```
[1] 85 12
```

13 types of candy!

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?

For M&M's:

```
candy$winpercent [34]
```

```
[1] 66.57458
```

Q4. What is the winpercent value for "Kit Kat"?

```
candy$winpercent [29]
```

[1] 76.7686

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

```
candy$winpercent [78]
```

[1] 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?

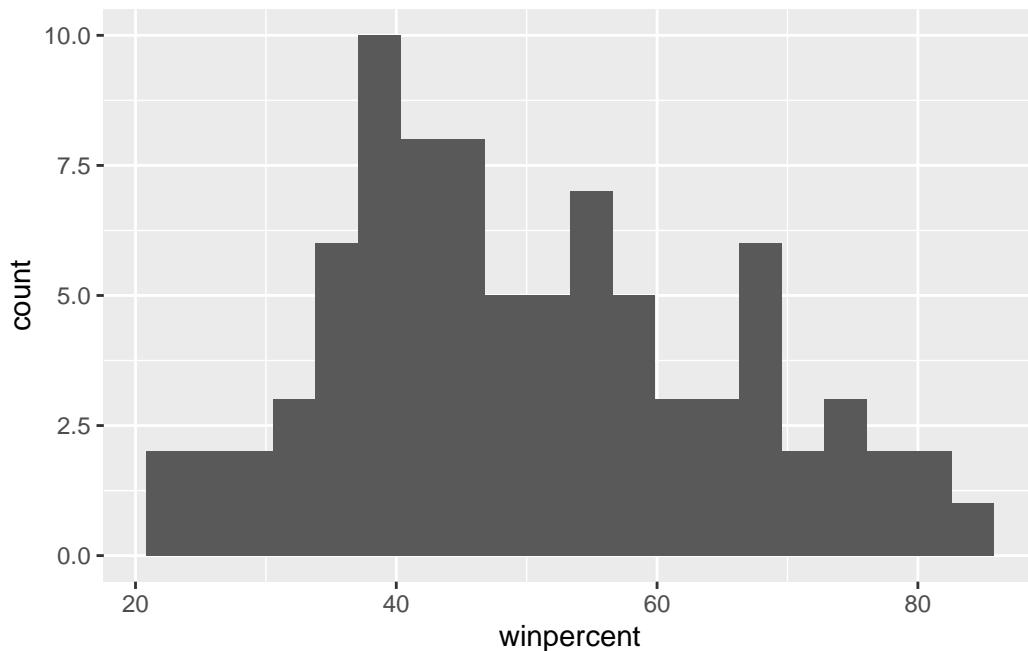
winpercent seems to be the odd column out in terms of scale and measurement.

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

It represents whether or not the candy has chocolate in it or not

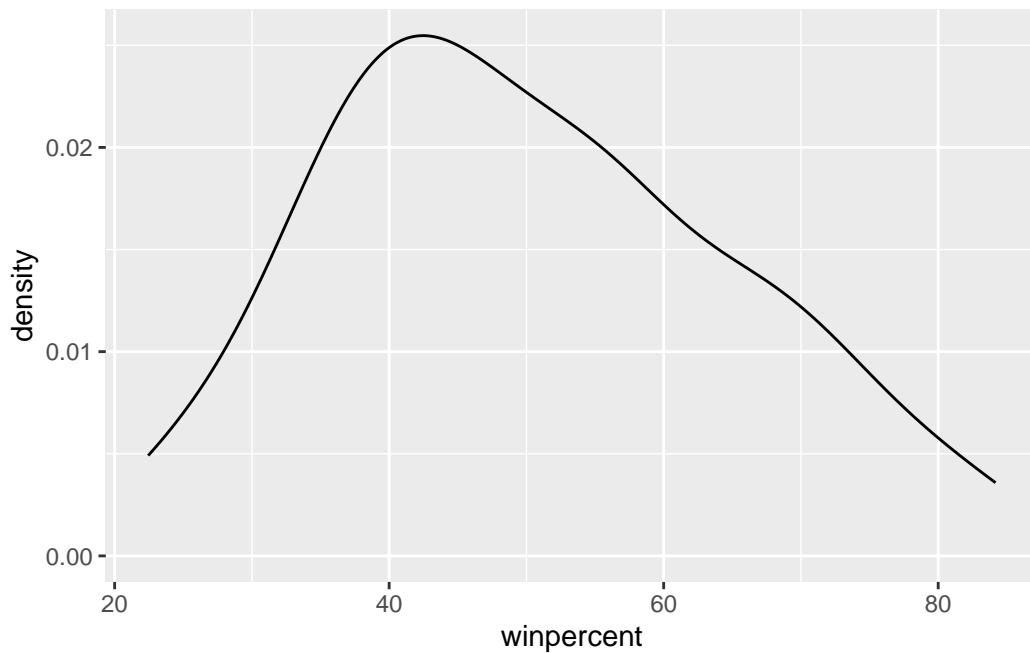
Q8. Plot a histogram of winpercent values

```
library(ggplot2)
ggplot(candy) +
  aes(winpercent) +
  geom_histogram(bins=20)
```



Q9. Is the distribution of winpercent values symmetrical?

```
ggplot(candy) +  
  aes(winpercent) +  
  geom_density()
```



No the distribution is not symmetrical

Q10. Is the center of the distribution above or 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?

```
#find all chocolate candy in dataset, extract their winpercent values, find mean of values, +  
  
choc inds <- as.logical(candy$chocolate)  
choc.candy <- candy[choc inds,]  
choc.win <- choc.candy$winpercent  
choc.mean <- mean(choc.win)  
choc.mean
```

```
[1] 60.92153
```

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

```
[1] 44.11974
```

```
choc.mean>=fruit.mean
```

```
[1] TRUE
```

Chocolate is ranked higher than fruit

Q12. Is this difference statistically significant?

```
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
```

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

```
ord.ind <- order(candy$winpercent)
head(candy[ord.ind, ], 5)
```

| | chocolate | fruity | caramel | peanuty | almondy | 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 |
| | crispedrice | wafer | hard | bar | pluribus | sugarpercent |
| Nik L Nip | 0 | 0 | 0 | 1 | | 0.197 |
| Boston Baked Beans | 0 | 0 | 0 | 1 | | 0.313 |
| Chiclets | 0 | 0 | 0 | 1 | | 0.046 |
| Super Bubble | 0 | 0 | 0 | 0 | | 0.162 |
| Jawbusters | 0 | 1 | 0 | 1 | | 0.093 |
| | pricepercent | 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?

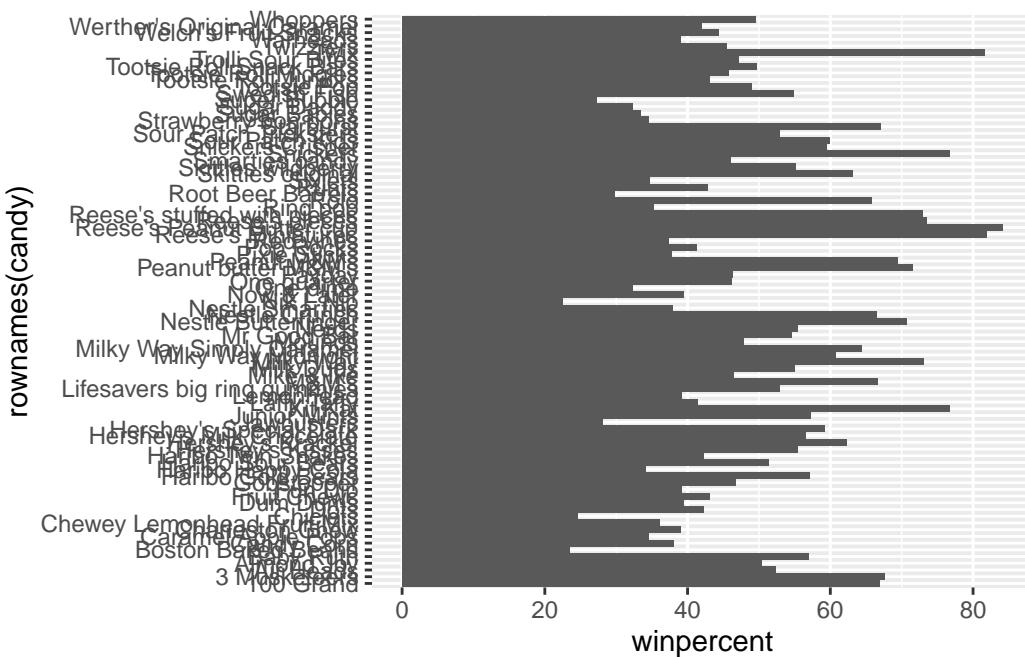
```
tail(candy[ord.ind, ], 5)
```

| | chocolate | fruity | caramel | peanuty | almondy | 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 |
| | crispedrice | wafer | 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 |
| 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 |

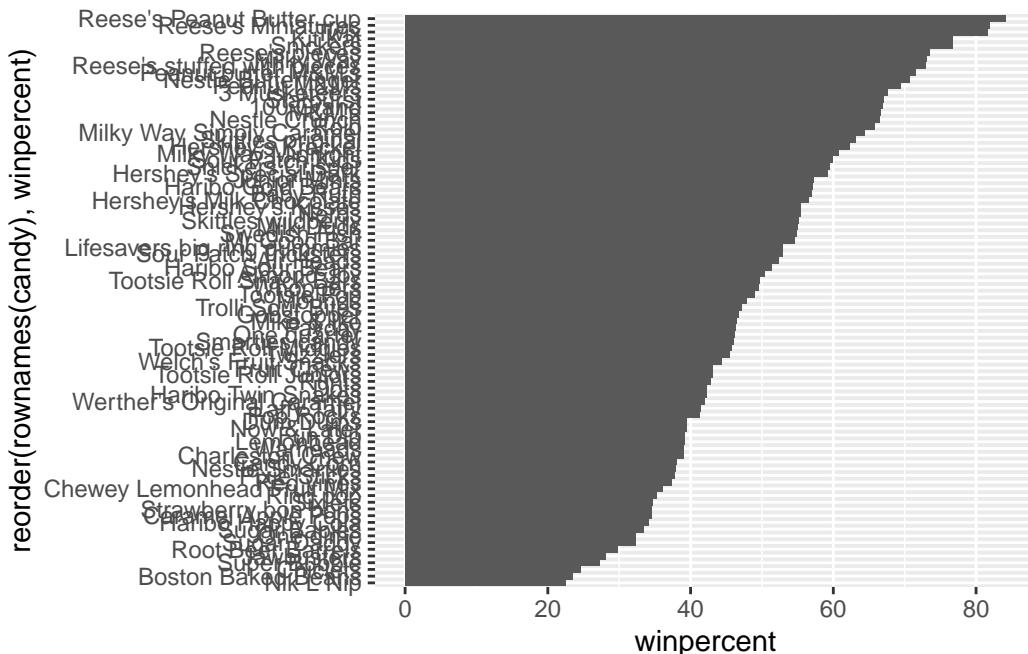
Q15. Make a first barplot of candy ranking based on winpercent values.

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



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

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

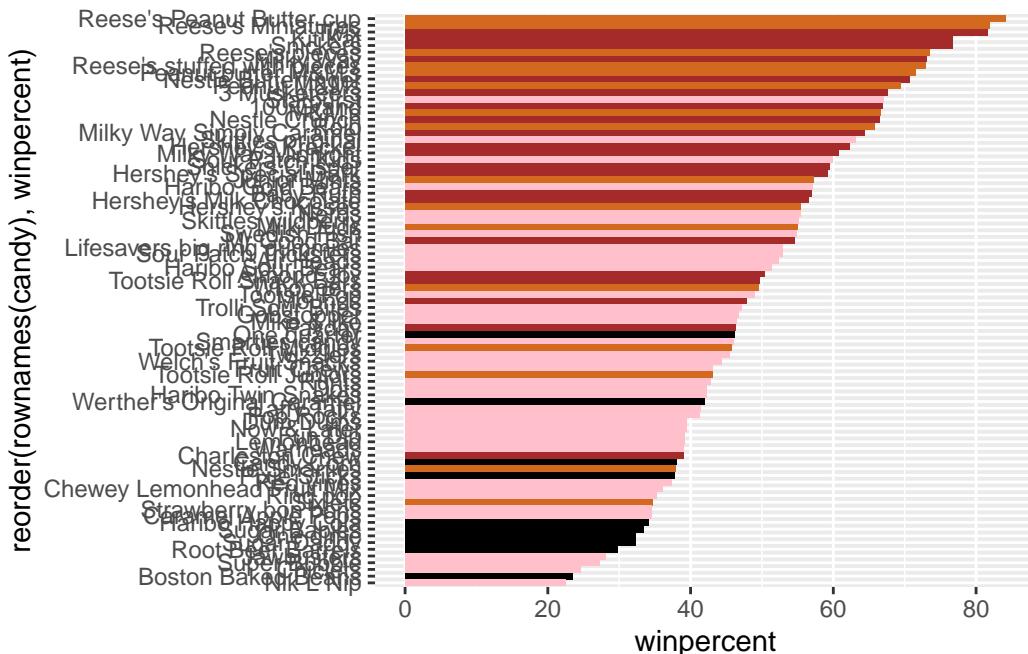


```

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

ggplot(candy) +
  aes(x=winpercent,
      y=reorder( rownames(candy), winpercent)) +
  geom_col(fill=mycols)

```



Q17. What is the worst ranked chocolate candy?

The worst ranked candy is Sixlets.

Q18. What is the best ranked fruity candy?

The best ranked fruity candy is Starburst!

Winpercent vs pricepoint

```
library(ggrepel)
library(ggplot2)
ggplot(candy) +
  aes(winpercent, pricepercent, label=rownames(candy)) +
  geom_point(col=mycols) +
  geom_text(col=mycols) +
  geom_text_repel(col=mycols, size=3.3, max.overlaps=50)
```



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 miniatures offer the best bang for your buck!

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

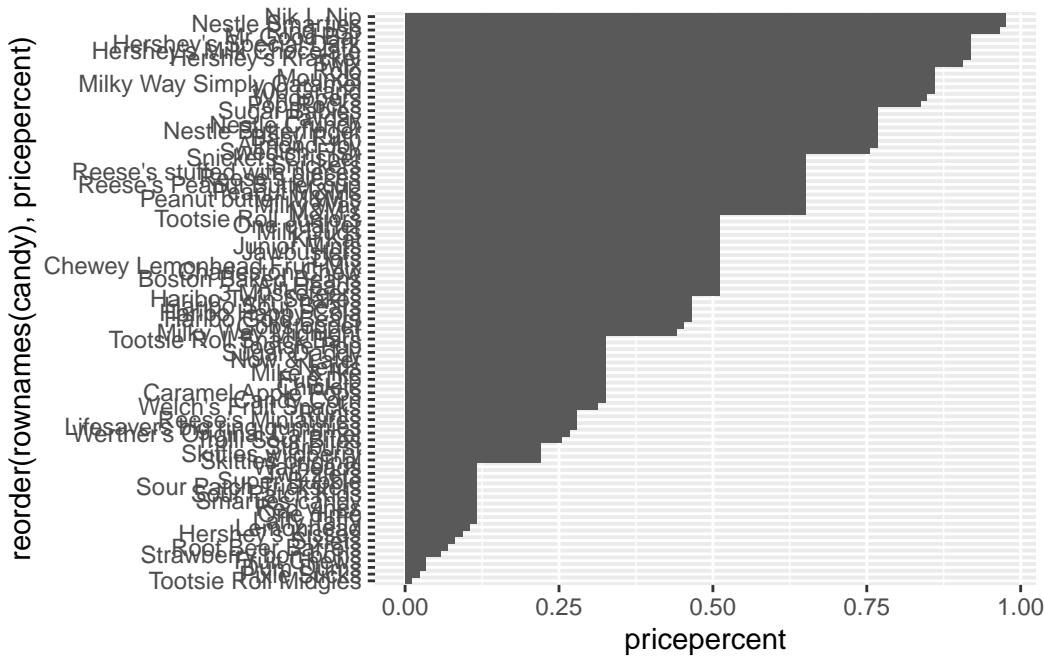
```
ord1 <- order(candy$pricepercent, decreasing = TRUE)
head(candy[ord1,c(11,12)], n=5)
```

| | price | percent | 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 | |

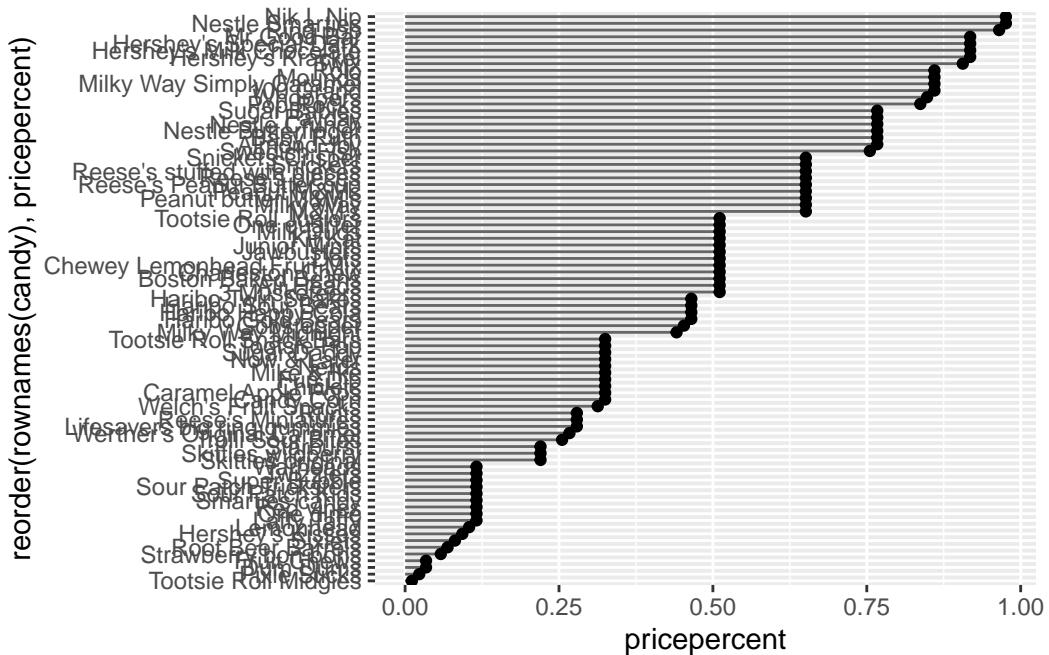
Nik L Nip is tied for most expensive but is somehow one of the least popular candies overall.

Q21. Optional

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



```
ggplot(candy) +
  aes(pricepercent, reorder(rownames(candy), pricepercent)) +
  geom_segment(aes(yend = reorder(rownames(candy), pricepercent),
                    xend = 0), col="gray40") +
  geom_point()
```



correlation

```
library(corrplot)
```

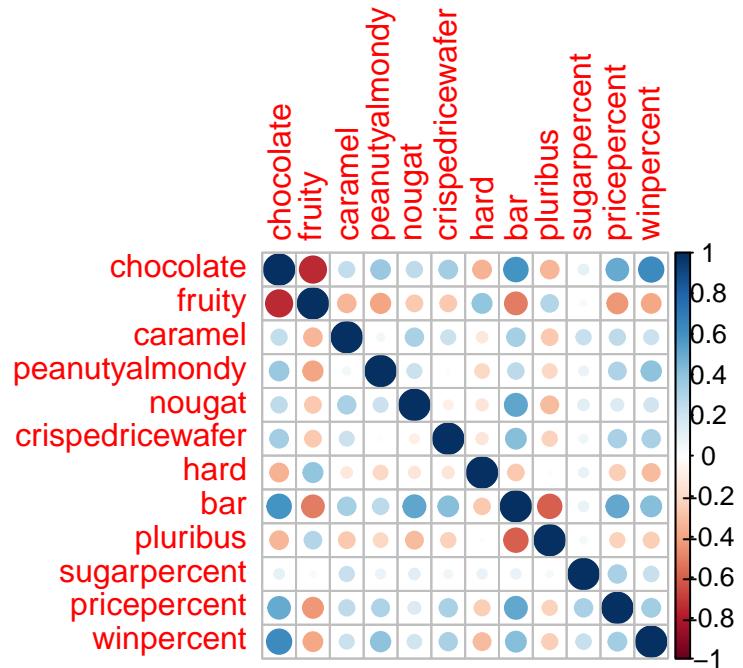
corrplot 0.95 loaded

```
cor(candy)
```

| | 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 | | | |

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



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

Chocolate and fruity are anti-correlated, as well as pluribus and bar.

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

Winpercent and chocolate, and bar and chocolate are all positively correlated.

principal component analysis

the main function in base R for this ‘prcomp()’ and we want to set ‘scale=TRUE’

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

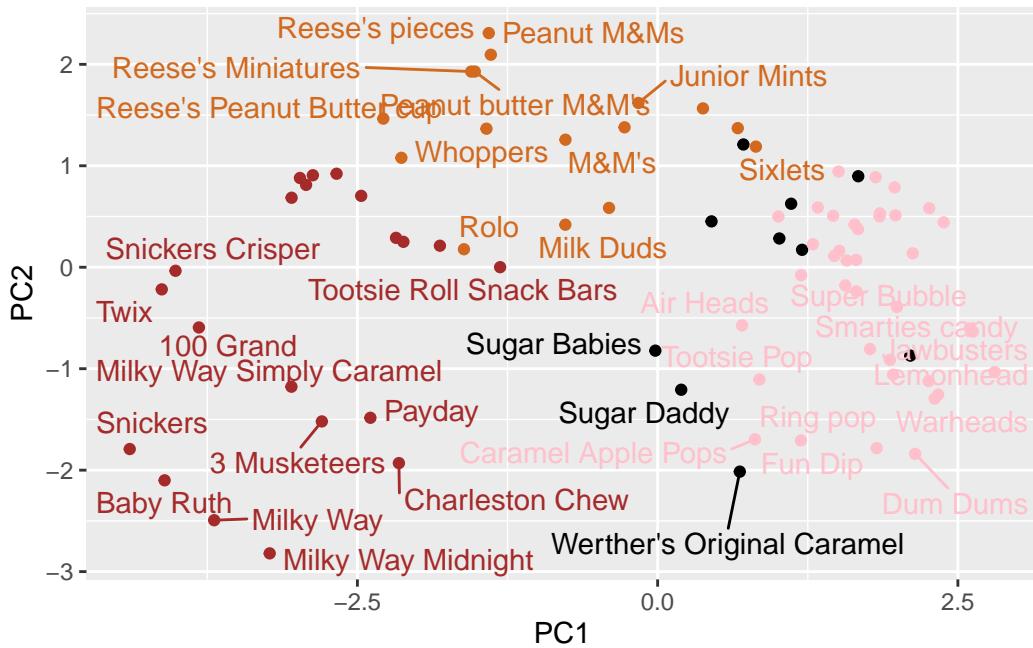
```

```

#pca$x
library(ggrepel)
ggplot(pca$x) +
  aes(PC1, PC2, label=rownames(pca$x)) +
  geom_point(col=mycols) +
  geom_text_repel(col=mycols)

```

Warning: ggrepel: 48 unlabeled data points (too many overlaps). Consider increasing max.overlaps

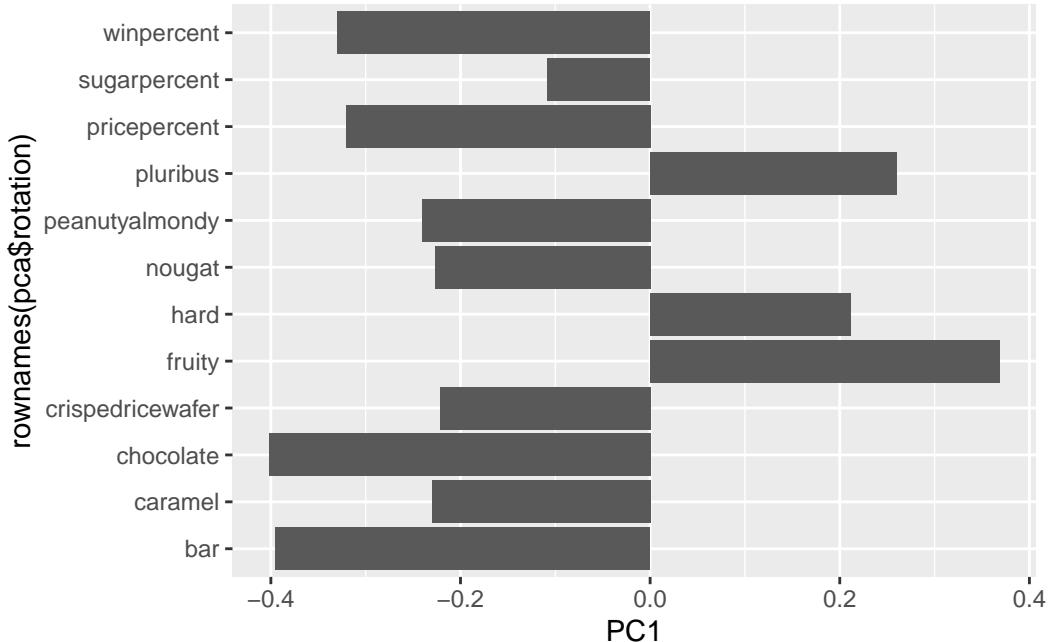


don't forget about your variable "loadings" - how the original variables contribute to your new PCs ...

```

ggplot(pca$rotation) +
  aes(PC1, rownames(pca$rotation)) +
  geom_col()

```



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

PC1 has the fruity and pluribus variables in a strong positive direction. This makes sense as many fruity candies are small pieces of candy that are packaged together in a bag.