

Lab 1: Exploratory Data Analysis (EDA)

Statistics 139 (special thanks to Julie Vu!)

September 08, 2023

Topics

- Basic Data Cleaning
- Numerical and Graphical Summaries
- Subsetting Data

The data in the file ‘dds_discrimination.csv’ represent a sample of 1,000 residents of California who receive funds from the California Department of Developmental Services (DDS); individuals receiving funds are referred to as ‘consumers’.

A study team examined the mean annual expenditure on consumers by ethnicity and found that the mean annual expenditures on Hispanic consumers was approximately one-third of the mean expenditures on White non-Hispanic consumers.

As a result, an allegation of ethnic discrimination was brought against the California DDS. Does this finding represent sufficient evidence of ethnic discrimination, or might there be more to the story?

The following variables are included in the dataset.

- **ID:** consumer ID number
- **Age.Cohort:** age group, where 1 refers to 0 - 5 years, 2 refers to 51+ years, 3 refers to 13 - 17 years, 4 refers to 18 - 21 years, 5 refers to 22 - 50 years, and 6 refers to 6 - 12 years.
- **Age:** age in years
- **Gender:** gender, recorded as 1 for female and 2 for male
- **Expenditures:** annual expenditure in dollars
- **Ethnicity:** ethnicity, recorded as either 1 for American Indian, 2 for Asian, 3 for Black, 4 for Hispanic, 5 for Multi Race, 6 for Native Hawaiian, 7 for Other, and 8 for White not Hispanic.

Problem 1: A little data clean-up

First, a bit of data cleaning will be helpful.

- a) Read the file into R as the `dds.discr` dataframe, and examine the first few observations.

```
#read file into R
dds.discr <- read.csv("data/dds_discrimination.csv")
head(dds.discr)
```

```
##   X    ID Age.Cohort Age Gender Expenditures Ethnicity
## 1 1 10210         3  17      1         2113          8
## 2 2 10409         5  37      2        41924          8
## 3 3 10486         1   3      2         1454          4
## 4 4 10538         4  19      1         6400          4
## 5 5 10568         3  13      2         4412          8
## 6 6 10690         3  15      1         4566          4
```

- b) The first column contains a 'variable' X that is just the row number as carried over from the CSV file. Run the following to eliminate X.

```
#remove first column
dds.discr[,1] <- NULL

#alternatively, we can give an extra argument into the read.csv function
dds.discr2 <- read.csv("data/dds_discrimination.csv", row.names=1)
head(dds.discr2)
```

```
##      ID Age.Cohort Age Gender Expenditures Ethnicity
## 1 10210         3  17      1         2113          8
## 2 10409         5  37      2        41924          8
## 3 10486         1   3      2         1454          4
## 4 10538         4  19      1         6400          4
## 5 10568         3  13      2         4412          8
## 6 10690         3  15      1         4566          4
```

- c) Datasets can sometimes have variables with long or messy names. For the sake of practice, read the documentation for the `colnames()` function and change the names of the variables to ones you find more convenient. To access the R help files for a function, type `?` before the function name.

```
# your work here

names(dds.discr) = c("id","age.cohort","age","gender","expenditures","ethnicity")
names(dds.discr)

## [1] "id"           "age.cohort"   "age"          "gender"       "expenditures"
## [6] "ethnicity"
```

- d) Let's look again at the dataset, this time using the `str()` function. What looks strange about the `gender` and `expenditures` variables? Hint: You've seen this in class already...

```
#maybe better to use str()  
str(dds.discr)
```

```
## 'data.frame':    1000 obs. of  6 variables:  
## $ id          : int  10210 10409 10486 10538 10568 10690 10711 10778 10820 10823 ...  
## $ age.cohort   : int   3 5 1 4 3 3 3 3 3 3 ...  
## $ age         : int  17 37 3 19 13 15 13 17 14 13 ...  
## $ gender      : chr   "1" "2" "2" "1" ...  
## $ expenditures: chr  "2113" "41924" "1454" "6400" ...  
## $ ethnicity   : int   8 8 4 4 8 4 8 3 8 4 ...
```

- Gender should be factor
- Expenditures should be numeric

Explain how the following two lines are designed to find the problems with the `gender` and `expenditures` variables, and explain what those problems are. You might have to read the documentation by typing `?` before any function names you don't know.

```
# see what values the var gender takes on  
table(dds.discr$gender)
```

```
##  
##      1      2 female FEMALE   male   Male   MALE  
##    500    494      2      1      1      1      1
```

```
# see what non-numeric values the var expenditures takes on  
dds.discr$expenditures[which(is.na(as.numeric(dds.discr$expenditures)))]
```

```
## Warning in which(is.na(as.numeric(dds.discr$expenditures))): NAs introduced by  
## coercion
```

```
## [1] "$46,571 " "$42,192 " "$54,616 " "$60,871 " "$3,673 "
```

Challenge: Explain how the following four lines of code fix the problem with the `Expenditures` variable. It's ok if you can't figure it out, this one is hard. You can just run the code for now.

```
# change gender var to uppercase  
dds.discr.uppercase <- toupper(dds.discr$gender)  
  
# if the value is MALE, change it to 2  
dds.discr.uppercase[which(dds.discr.uppercase == "MALE")] <- "2"
```

```
# if the value is FEMALE, change it to 1
dds.discr.uppercase[which(dds.discr.uppercase == "FEMALE")] <- "1"

# change gender var to type numeric
dds.discr$gender <- as.numeric(dds.discr.uppercase)
```

Challenge: Explain how the following two lines of code fix the problem with the Expenditures variable. It's ok if you can't figure it out, this one is hard. You can just run the code for now.

```
# substitute $ for non-numeric values of Expenditure with blank character
dds.discr$expenditures[which(is.na(as.numeric(dds.discr$expenditures)))] <- gsub("\\$", "", dds.discr$expenditures)
```

```
## Warning in which(is.na(as.numeric(dds.discr$expenditures))): NAs introduced by coercion
```

```
## Warning in which(is.na(as.numeric(dds.discr$expenditures))): NAs introduced by coercion
```

```
# substitute , for non-numeric values of Expenditure with blank character
dds.discr$expenditures[which(is.na(as.numeric(dds.discr$expenditures)))] <- gsub(",", "", dds.discr$expenditures)
```

```
## Warning in which(is.na(as.numeric(dds.discr$expenditures))): NAs introduced by coercion
```

```
## Warning in which(is.na(as.numeric(dds.discr$expenditures))): NAs introduced by coercion
```

```
# change Expenditures var to type numeric
dds.discr$expenditures <- as.numeric(dds.discr$expenditures)
```

- e) The categorical variables (age cohort, gender, and ethnicity) should be converted to factor variables. Read the documentation for `factor()` and recode these three variables. Note that age cohort is an ordered categorical variable.

```
summary(dds.discr)
```

```
##           id           age.cohort           age           gender           expenditures
##  Min.      :10210    Min.      :1.000    Min.      : 0.0    Min.      :1.000    Min.      : 222
##  1st Qu.:31809    1st Qu.:3.000    1st Qu.:12.0    1st Qu.:1.000    1st Qu.: 2899
##  Median :55384    Median :4.000    Median :18.0    Median :1.000    Median : 7026
##  Mean   :54663    Mean   :3.906    Mean   :22.8    Mean   :1.497    Mean   :18066
##  3rd Qu.:76135    3rd Qu.:5.000    3rd Qu.:26.0    3rd Qu.:2.000    3rd Qu.:37713
##  Max.   :99898    Max.   :6.000    Max.   :95.0    Max.   :2.000    Max.   :75098
##  ethnicity
```

```
## Min.    :1.000
## 1st Qu.:4.000
## Median :4.000
## Mean    :5.313
## 3rd Qu.:8.000
## Max.    :8.000
```

```
dds.discr$gender = factor(dds.discr$gender, levels=1:2, labels=c("F","M"))

dds.discr$age.cohort = factor(dds.discr$age.cohort, levels = c(1,6,3,4,5,2),
                             labels=c("0-5","6-12","13-17","18-21","22-50","51+"))

dds.discr$ethnicity = factor(dds.discr$ethnicity, levels=1:8, labels=c("American Indian","Asian",
```

- f) Save the clean version of the dataframe as an .Rdata file. (Alternatively, you could use `write.csv()` to write the dataframe to a CSV file.)

```
#save the file
save(dds.discr, file = "dds_discr.Rdata")
```

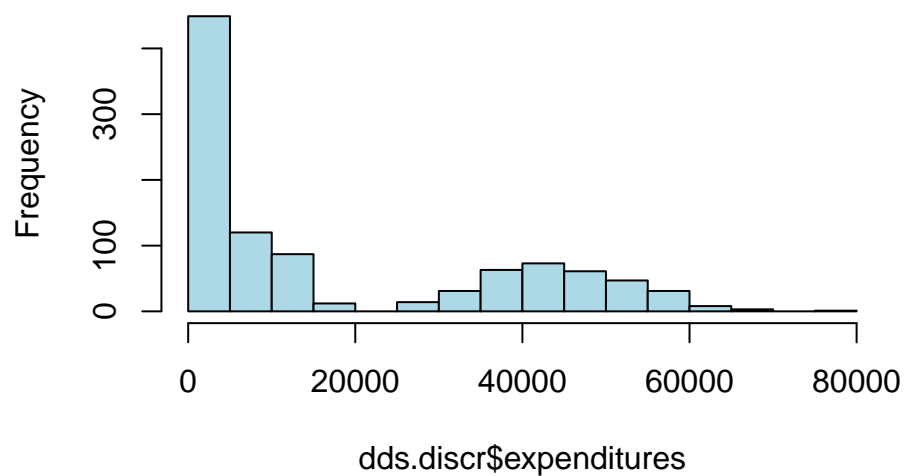
Problem 2: Univariate Explorations (aka, distributions)

Let's start by examining the distributions of single variables on their own. Create numerical and graphical summaries as appropriate.

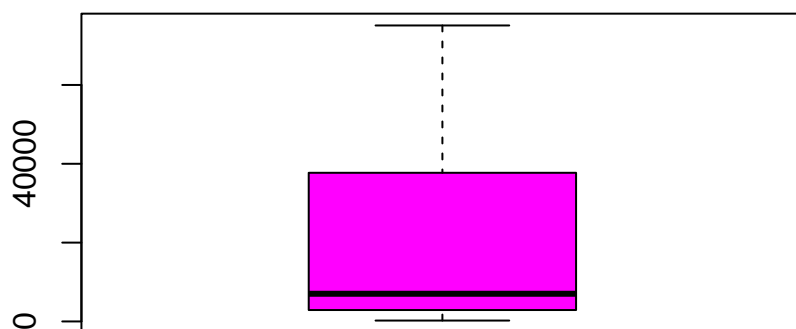
- a) Describe the distribution of annual expenditures. For most consumers, is the amount of financial support provided by the DDS relatively high or low?

```
#graphical summaries
hist(dds.discr$expenditures,col="lightblue")
```

Histogram of dds.discr\$expenditures



```
boxplot(dds.discr$expenditures, col="magenta")
```



```
#numerical summaries
```

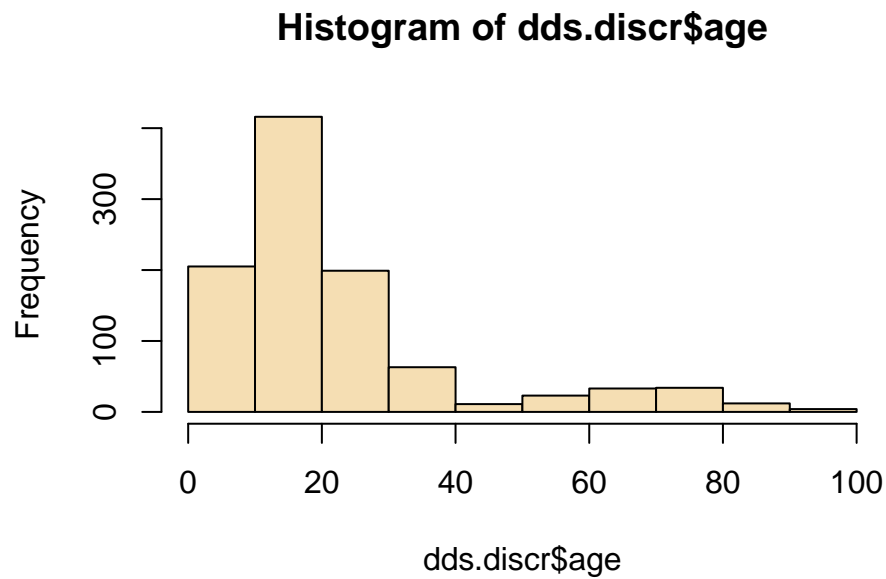
```
summary(dds.discr$expenditures)
```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	222	2899	7026	18066	37713	75098

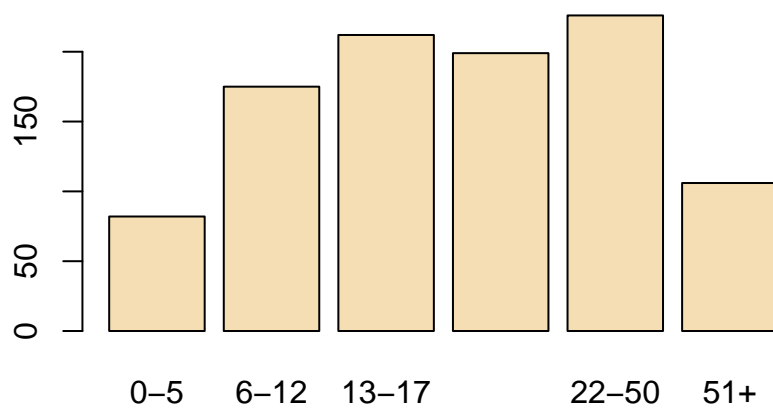
- For most consumers, the amount of financial support tends to be low (right skewed distribution)

b) Do consumers tend to be older or younger?

```
#graphical summaries  
hist(dds.discr$age, col = "wheat")
```



```
plot(dds.discr$age.cohort, col = "wheat")
```



```
#numerical summaries
summary(dds.discr$age)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.0   12.0   18.0   22.8   26.0   95.0
```

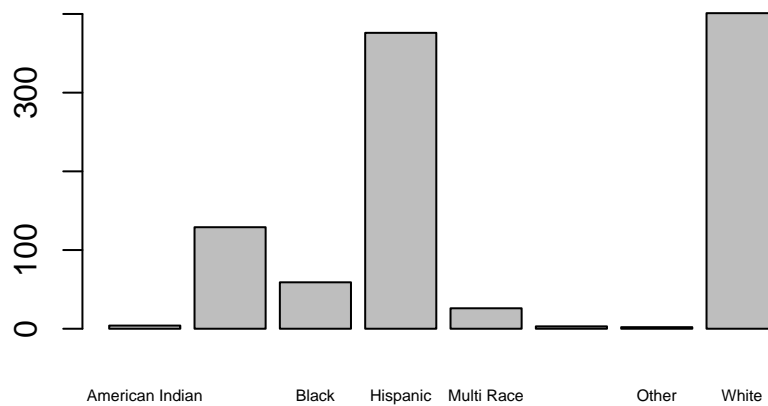
```
table(dds.discr$age.cohort)
```

```
##
##      0-5  6-12 13-17 18-21 22-50  51+
##      82   175   212   199   226   106
```

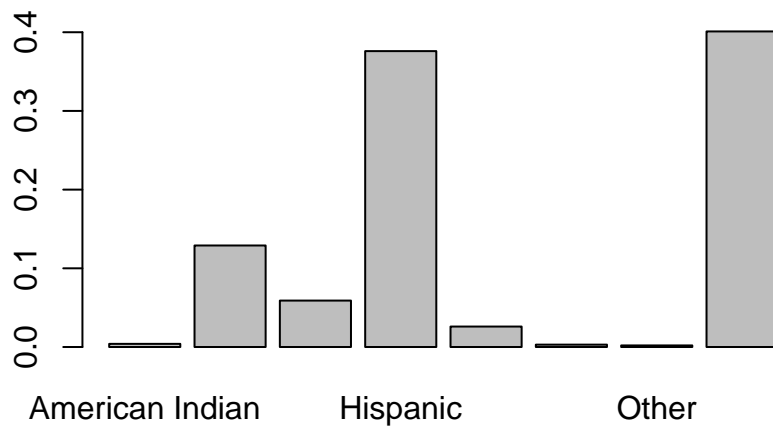
- Consumers tend to be younger (also right skewed distribution)

c) Is there an equal representation among ethnic groups?

```
#graphical summaries
barplot(table(dds.discr$ethnicity), cex.names = 0.5)
```



```
barplot(prop.table(table(dds.discr$ethnicity)))
```

```
#numerical summaries
table(dds.discr$ethnicity)
```

```
##
## American Indian      Asian      Black      Hispanic      Multi Race
##           4          129         59          376          26
## Native Hawaiian      Other      White
##           3           2         401
```

```
prop.table(table(dds.discr$ethnicity))
```

```
##
## American Indian      Asian      Black      Hispanic      Multi Race
##           0.004      0.129      0.059      0.376      0.026
## Native Hawaiian      Other      White
##           0.003      0.002      0.401
```

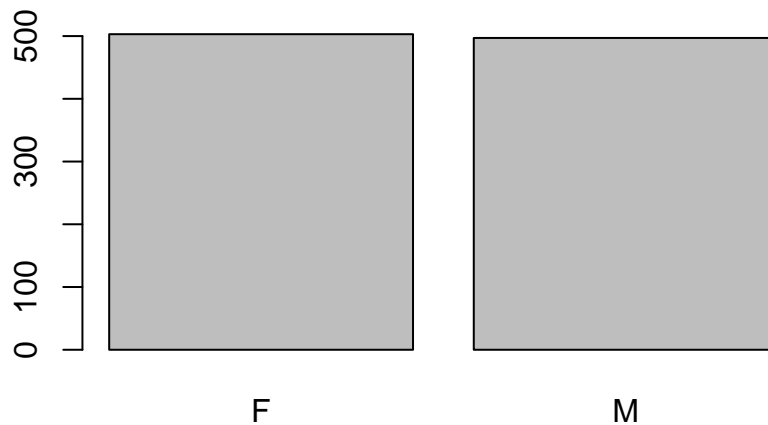
```
mean(dds.discr$ethnicity == "White" | dds.discr$ethnicity == "Hispanic")
```

```
## [1] 0.777
```

- can't tell bc we don't know the actual % breakdown of ethnicity of people in the US
- not equal because White and Hispanic consumers make up 77% of the group

d) Does gender appear to be balanced?

```
#graphical summaries  
plot(dds.discr$gender)
```



```
table(dds.discr$gender)
```

```
##  
##    F    M  
## 503 497
```

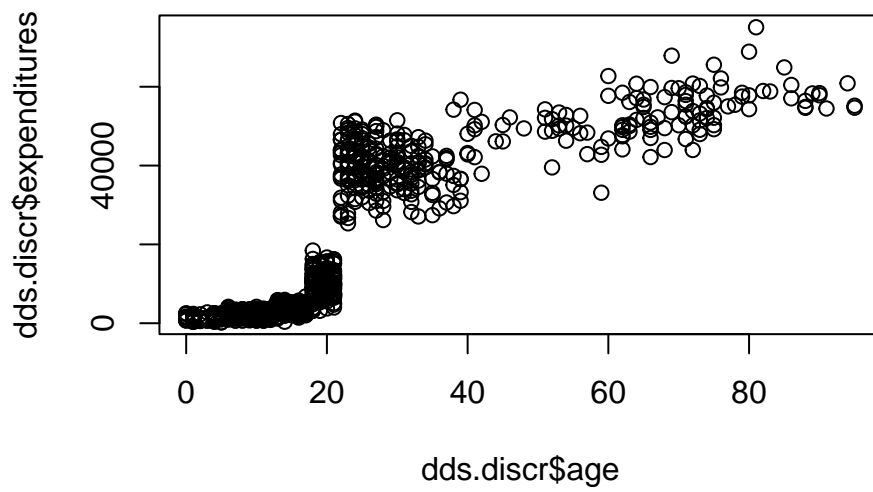
```
* balanced!
```

Problem 3: Bivariate Explorations (aka, relationships)

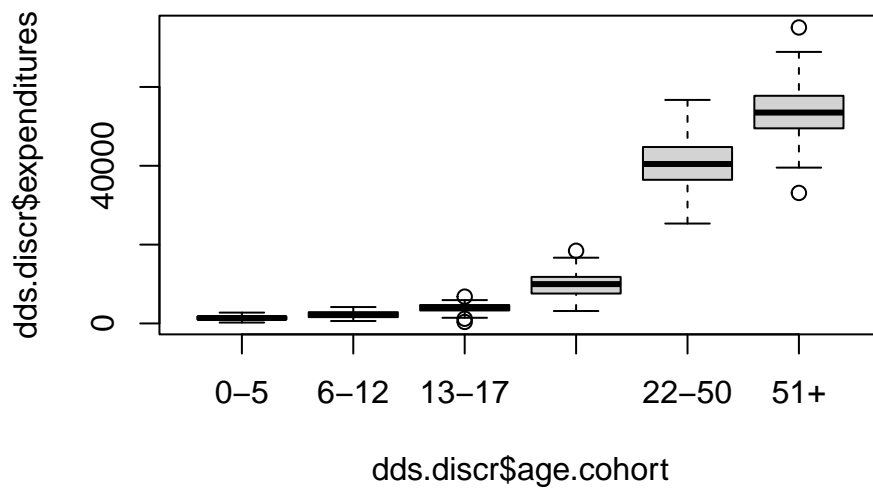
Let's explore how variables are related to each other.

- a) How do annual expenditures vary by age? Explore this using the quantitative and categorical versions of age separately. Conjecture a reason for the trend in the data.

```
# graphical  
plot(dds.discr$expenditures ~ dds.discr$age)
```



```
plot(dds.discr$expenditures ~ dds.discr$age.cohort)
```



```
# numerical
tapply(dds.discr$expenditures, dds.discr$age.cohor, summary)
```

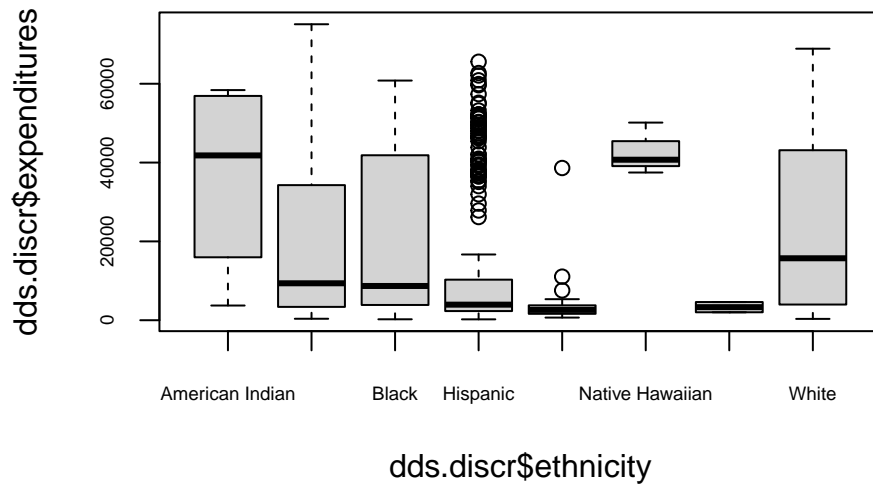
```
## $'0-5'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
##      222      1034      1380      1415      1739      2750
##
## $'6-12'
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      620     1602     2191     2227     2846     4163
##
## $'13-17'
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      386     3306     3952     3923     4666     6798
##
## $'18-21'
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      3153     7588     9979     9889     11806    18435
##
## $'22-50'
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      25348    36447    40456    40209    44721    56716
##
## $'51+'
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      33110    49515    53509    53522    57746    75098
```

* People aged 22+ have more expenditures than 21 and below consumers. This could be due to high

b) How does the distribution of expenditures vary by ethnic group?

```
# graphical
boxplot(dds.discr$expenditures ~ dds.discr$ethnicity, cex.axis = 0.6)
```



```
# numerical
```

```
tapply(dds.discr$expenditures, dds.discr$ethnicity, summary)
```

```
## $'American Indian'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   3726  22085   41818   36438  56170   58392
##
## $Asian
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    374   3382   9369   18392  34274   75098
##
## $Black
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    240   3870   8687   20885  41857   60808
##
## $Hispanic
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    222   2331   3952   11066  10292   65581
##
## $'Multi Race'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    669   1690   2622   4457   3750   38619
##
## $'Native Hawaiian'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  37479  39103  40727  42782  45434   50141
##
## $Other
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      2018    2667    3316    3316    3966    4615
##
## $White
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      340    3977    15718   24698   43134   68890
```

* The Hispanic group has many outliers.
 * American Indian and Native Hawaiian groups have the highest average expenditures

Problem 4: Exploring Evidence of Discrimination

Hispanic and White non-Hispanic individuals comprise the majority of the data. The rest of this analysis will focus on comparing how expenditures vary between these two groups.

- a) Do Hispanic consumers, on average, seem to receive less financial support from the California DDS than a White non-Hispanic consumer?

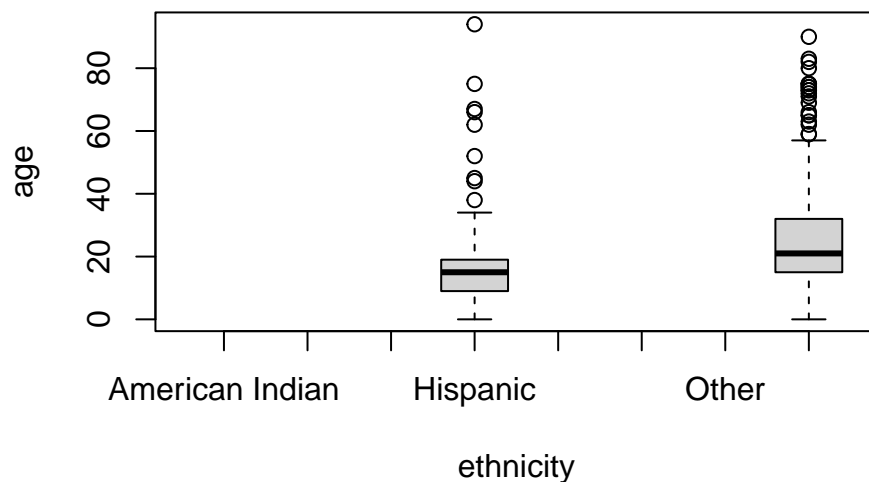
```
tapply(dds.discr$expenditures, dds.discr$ethnicity, mean)
```

```
## American Indian      Asian      Black      Hispanic      Multi Race
##      36438.250      18392.372      20884.593      11065.569      4456.731
## Native Hawaiian      Other      White
##      42782.333      3316.500      24697.549
```

* Hispanic consumers tend to receive ~\$11k on average, less than White consumers (\$25k on average)

- b) Recall that expenditures is strongly associated with age. Is there also an association between age and ethnicity, for these two ethnic groups?

```
plot(age~ethnicity, dds.discr[dds.discr$ethnicity == c("White", "Hispanic"),])
```



- c) For a closer look at the relationship between age, ethnicity, and expenditures, compare how average expenditures differ by ethnic group within each age cohort. Describe your findings.

```
with(dds.discr[dds.discr$ethnicity == c("White", "Hispanic"),], tapply(expenditures, list(ethn
```

##	0-5	6-12	13-17	18-21	22-50	51+
## American Indian	NA	NA	NA	NA	NA	NA
## Asian	NA	NA	NA	NA	NA	NA
## Black	NA	NA	NA	NA	NA	NA
## Hispanic	1478.864	2469.079	3890.200	9497.625	40586.39	56303.00
## Multi Race	NA	NA	NA	NA	NA	NA
## Native Hawaiian	NA	NA	NA	NA	NA	NA
## Other	NA	NA	NA	NA	NA	NA
## White	1220.000	2201.889	3727.879	9907.139	40344.14	51991.36

- d) Does there seem to be evidence of ethnic discrimination in the amount of financial support provided by the California DDS? Explain why the bivariate analysis conducted by the study team was misleading (bonus for remembering the specific term for the responsible phenomenon, which was covered in Stat 110!).

- Simpson's paradox