# Math 170S Homework #2: Exploratory Data Analysis

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<pre>suppressWarnings(library(ggplot2)) df &lt;- read.csv("HW2data.csv") head(df)</pre>	
## ID Sex Age Salary Expense ## 1 12 Male 22 2311 1050 ## 2 13 Male 24 3231 1265 ## 3 14 Male 27 2423 1109 ## 4 15 Male 19 3343 1511 ## 5 16 Female 20 2535 1147 ## 6 17 Female 24 3455 1564	
tail(df)	
## 46 39 Male 24 4877 2190 ## 47 39 Female 28 4069 1830 ## 48 39 Male 30 4989 2246 ## 49 39 Female 33 4181 1884 ## 50 39 Male 25 5101 2292 ## 51 39 Male 26 4293 1929	
Here is a function for calculating the mode	

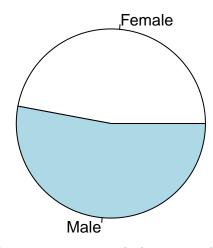
```
getmode <- function(v) {
   uniqv <- unique(v)
   uniqv[which.max(tabulate(match(v, uniqv)))]
}</pre>
```

### Univariate Analysis

#### Sex

```
pie(table(df$Sex), main = "Sex of Indidividuals in Data")
```

### Sex of Indidividuals in Data

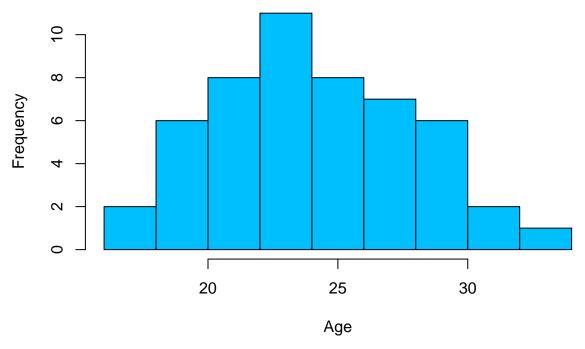


There is a near even split between male and female in this data, but with slightly more male

#### Age

```
summary(df$Age)
##
                     Median
                               Mean 3rd Qu.
      Min. 1st Qu.
                                                 Max.
     16.00
             22.00
                      24.00
                               24.51
                                                33.00
                                       27.00
getmode(df$Age)
## [1] 24
So the mean age is 24.51, the median age is 24, and the mode/most common age is also 24.
hist(df$Age, xlab = "Age", main = "Histogram of Ages", col = "deepskyblue")
```

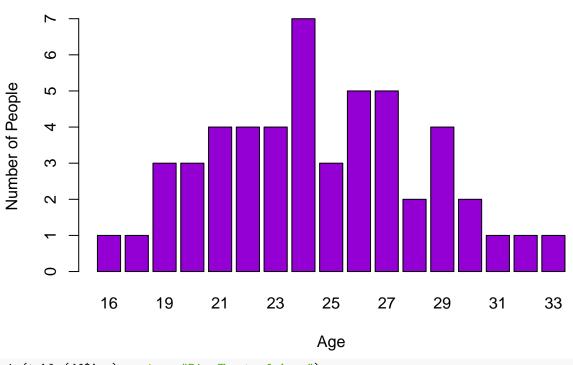
# **Histogram of Ages**



Age looks to be distributed fairly normally

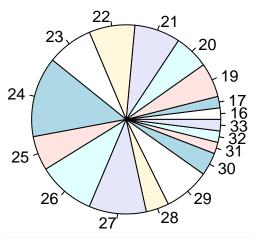
barplot(table(df\$Age), xlab = "Age", ylab = "Number of People", main = "Bar Plot of Ages", col = "darkv

# **Bar Plot of Ages**



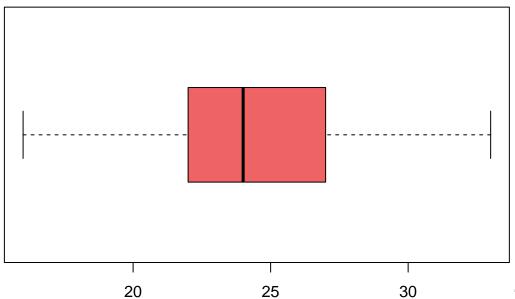
pie(table(df\$Age), main = "Pie Chart of Ages")

# **Pie Chart of Ages**



boxplot(df\$Age, col="indianred2", plot=T, horizontal=T, main="Boxplot of Age")

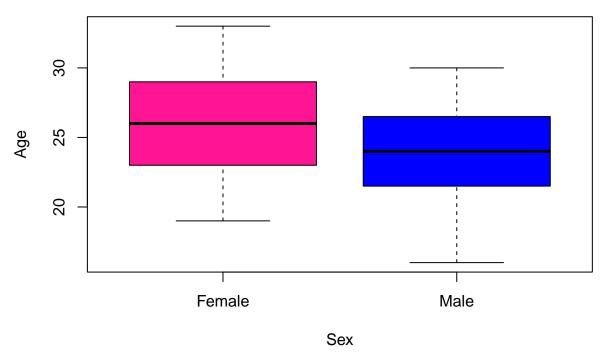
# **Boxplot of Age**



20 25 30 There are fewer people at the oldest and youngest ends of the spectrum, i.e. fewer 16 and 17 year olds and 32 or 33 year olds than people in their 20s

boxplot(df\$Age ~ df\$Sex, col=c("deeppink","blue"), xlab="Sex", ylab="Age", main="Boxplot of Age by Sex"

# **Boxplot of Age by Sex**



There seem to be more older female indidviduals in the data compared to males

#### Salary

```
summary(df$Salary)

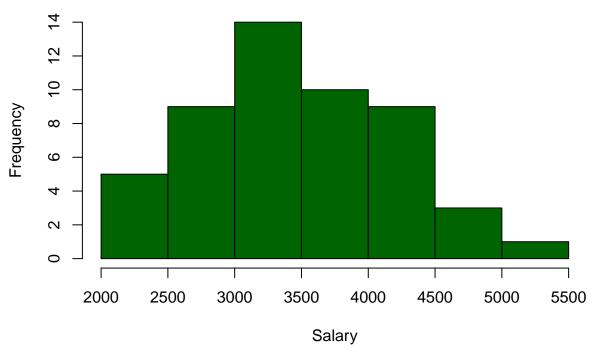
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 2191 2991 3447 3472 4002 5101
getmode(df$Salary)
```

#### ## [1] 2311

So the average salary is 3472, the median salary is 3447, and the most common salary is 2311. There is a large jump from the mode to the mean and median; larger values/possible outliers may have some influence on the mean. We can see that the max value is much higher than the 3rd quartile

```
hist(df$Salary, xlab = "Salary", main = "Histogram of Salaries", col = "darkgreen")
```

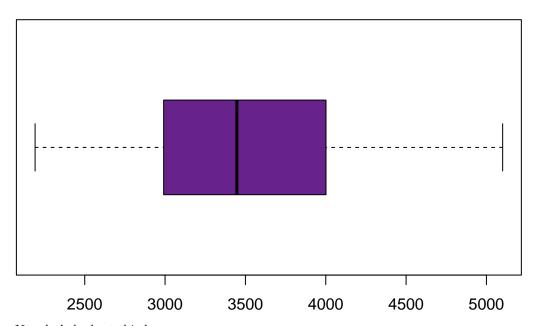
# **Histogram of Salaries**



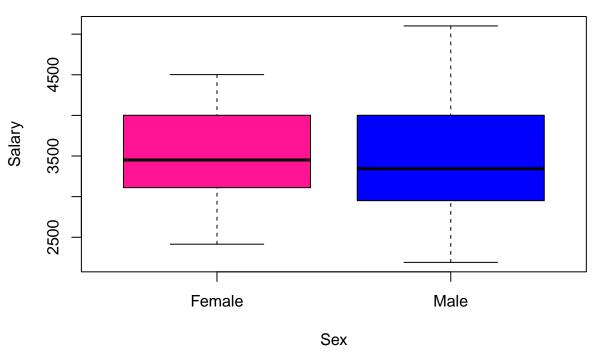
Salary looks slightly right skewed, which supports what we saw in our previous observation.

boxplot(df\$Salary, col="darkorchid4", plot=T, horizontal=T, main="Boxplot of Salary Values")

# **Boxplot of Salary Values**



### **Boxplot of Salary Values by Sex**



Males in this data have a wider range of salary than females

#### Expense

```
summary(df$Expense)

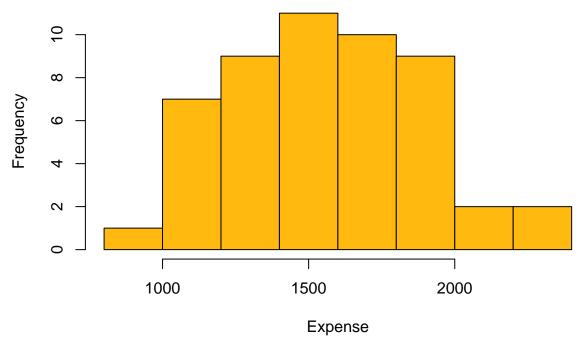
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 990 1324 1548 1559 1794 2292
getmode(df$Expense)
```

## [1] 1351

So the mean is 1559, the median is 1548, and the mode is 1351. Again, large expense values may be increasing the mean

hist(df\$Expense, xlab = "Expense", main = "Histogram of Expense Values", col="darkgoldenrod1")

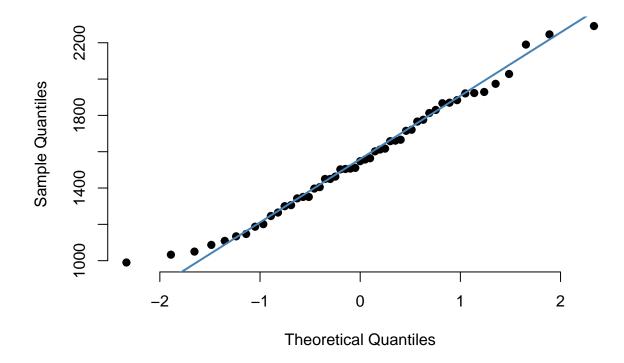
# **Histogram of Expense Values**



Expense is not right skewed like salary, but the larger values around 2000 do seem to lie a bit more outside the rest of the data

```
qqnorm(df$Expense, pch = 19, frame = FALSE)
qqline(df$Expense, col = "steelblue", lwd = 2)
```

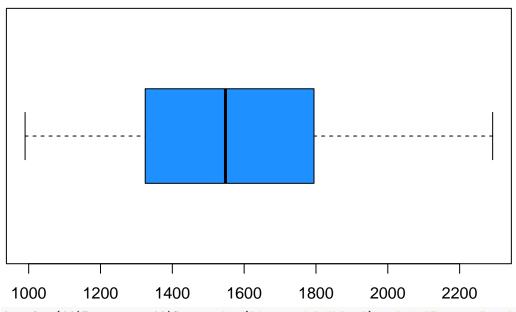
# Normal Q-Q Plot



From the qqplot, expense is fairly normally distributed. The points on the plot seem to fall on a fairly straight line and don't seem to have some other pattern that would imply skewedness

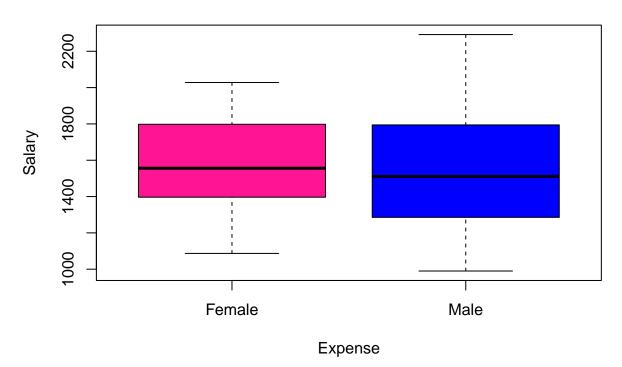
boxplot(df\$Expense, col="dodgerblue", plot=T, horizontal=T, main="Boxplot of Expense Values")

# **Boxplot of Expense Values**



boxplot(df\$Expense ~ df\$Sex, col=c("deeppink","blue"), xlab="Expense", ylab="Salary", main="Boxplot of I

# **Boxplot of Expense Values by Sex**



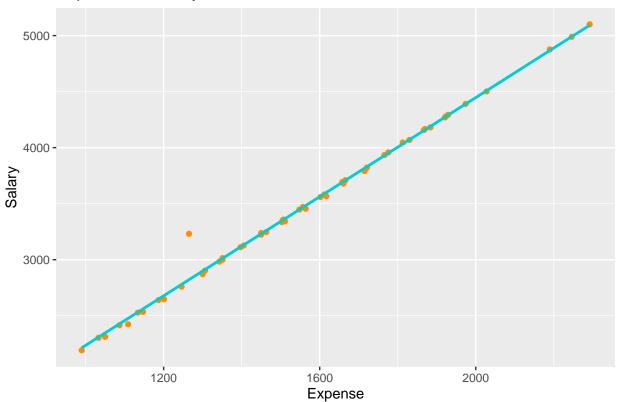
Males in this data also have a wider range of expenses than females

# Bivariate/Multivariate Analysis

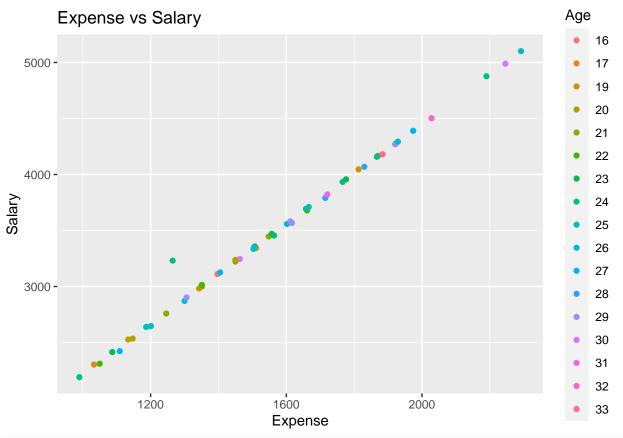
### Expense vs Salary

```
ggplot(df, aes(x=Expense, y=Salary)) + geom_point(colour="darkorange") + stat_smooth(method = "lm", col
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## `geom_smooth()` using formula = 'y ~ x'
```

### Expense vs Salary



ggplot(df, aes(x=Expense, y=Salary)) + geom\_point(aes(color = factor(Age))) + labs(colour = "Age", titl



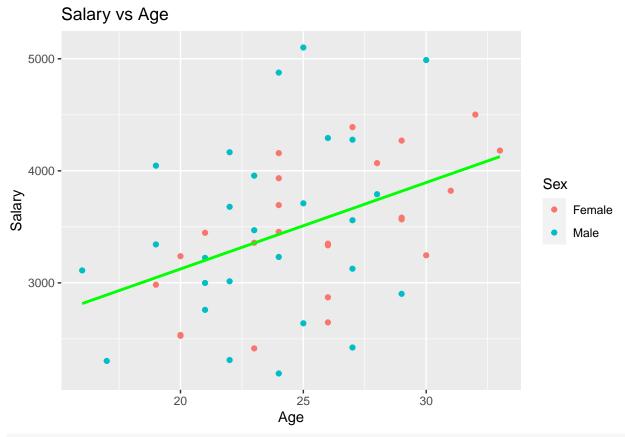
cor(df\$Expense, df\$Salary)

#### ## [1] 0.9965094

Expense and salary appear to be very strongly correlated, as evidenced by the correlation coefficient and the near perfect linear relationship seen in the scatterplot

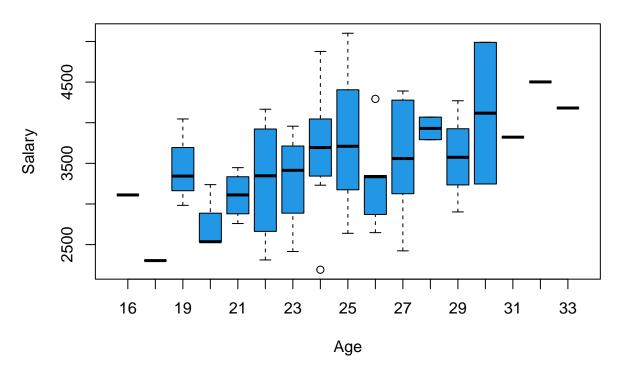
### Salary vs Age

```
ggplot(df, aes(x=Age, y=Salary)) + geom_point(aes(color = factor(Sex))) + stat_smooth(method = "lm", co
## `geom_smooth()` using formula = 'y ~ x'
```



boxplot(Salary~Age,data=df, main="Salary Data by Age", xlab="Age", ylab="Salary", col=4)

# Salary Data by Age



#### cor(df\$Salary, df\$Age)

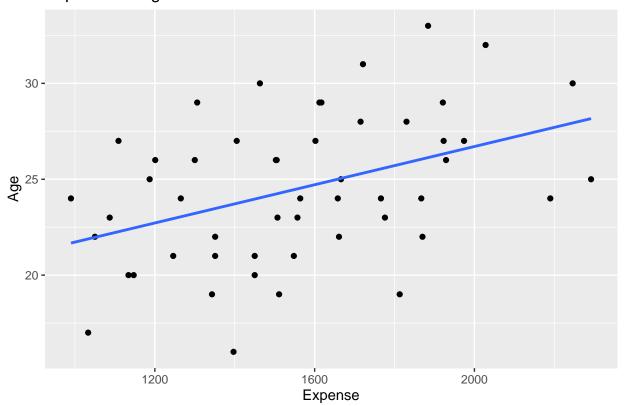
#### ## [1] 0.4136556

Salary and age are weakly, positively correlated.

### Expense vs Age

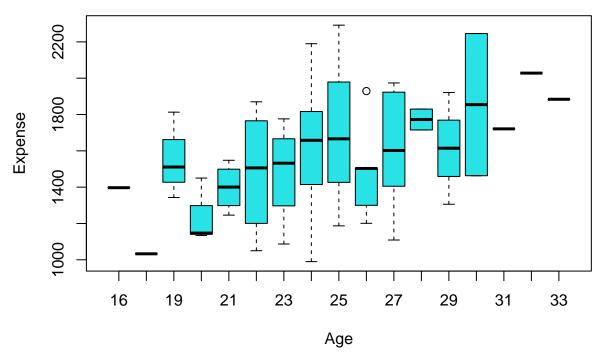
ggplot(df, aes(x=Expense, y=Age)) + geom\_point() + stat\_smooth(method = "lm", se= FALSE, size = 1) + lag
## `geom\_smooth()` using formula = 'y ~ x'

### Expense vs Age



boxplot(Expense~Age,data=df, main="Expense Data by Age", xlab="Age", ylab="Expense", col=5)

# **Expense Data by Age**



This boxplot looks very similar to the boxplot of salary by age, perhaps due to correlation between those variables.

#### cor(df\$Expense, df\$Age)

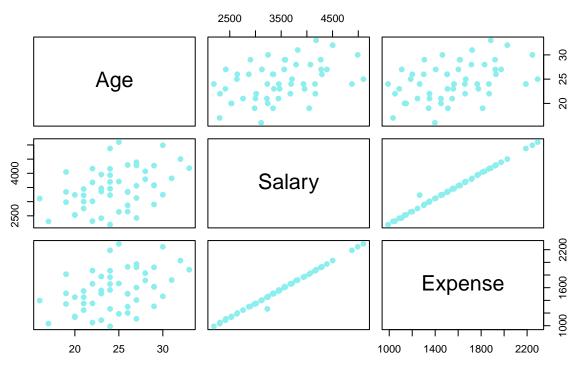
#### ## [1] 0.4189089

Expense and age have a somewhat weak, positive correlation, similar to salary and age

### Scatterplot Matrix of Age, Salary, and Expense

```
plot(df[3:5], col="darkslategray2", main = "Scatterplot Matrix of Data", pch=19)
```

# **Scatterplot Matrix of Data**



It's clear from the scatterplot matrix that expense and salary are very closely correlated, while age is not very strongly correlated with either salary or expense. All the relationships are positive