# Stat 410 | Salary Project Report ¶

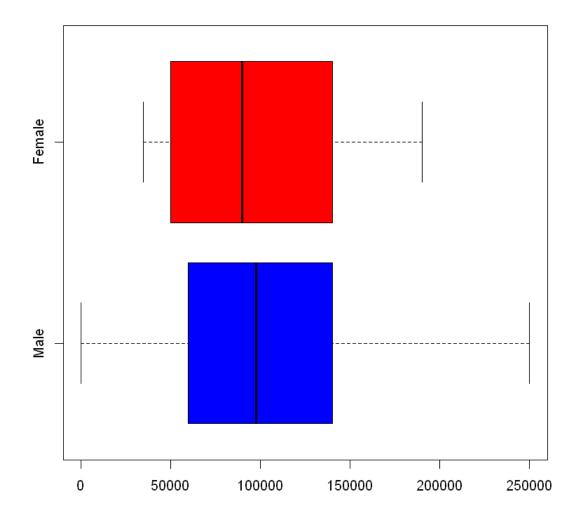
# Rachel, Abdullah, Gustavo

#### **Importing Data & Data Cleaning**

```
salary = read.csv('Salary Data.csv')
In [1]:
            str(salary)
            'data.frame': 375 obs. of 6 variables:
                                 : int 32 28 45 36 52 29 42 31 26 38 ...
             $ ï..Age
                                  : Factor w/ 3 levels "", "Female", "Male": 3 2 3 2 3 3 2 3
             $ Gender
            2 3 ...
             $ Education.Level : Factor w/ 4 levels "","Bachelor's",..: 2 3 4 2 3 2 3 2
            2 4 ...
            $ Job.Title
                                 : Factor w/ 175 levels "", "Account Manager", ...: 161 19 1
            32 103 24 83 95 106 84 152 ...
             $ Years.of.Experience: num 5 3 15 7 20 2 12 4 1 10 ...
             $ Salary
                                 : int 90000 65000 150000 60000 200000 55000 120000 8000
            0 45000 110000 ...
In [2]: ► salary$Age = salary$i..Age
            salary = subset(salary, select = -c(i..Age))
            str(salary)
            'data.frame':
                           375 obs. of 6 variables:
             $ Gender
                          : Factor w/ 3 levels "", "Female", "Male": 3 2 3 2 3 3 2 3
            2 3 ...
             $ Education.Level : Factor w/ 4 levels "", "Bachelor's",..: 2 3 4 2 3 2 3 2
            2 4 ...
             $ Job.Title
                                : Factor w/ 175 levels "", "Account Manager", ...: 161 19 1
            32 103 24 83 95 106 84 152 ...
            $ Years.of.Experience: num 5 3 15 7 20 2 12 4 1 10 ...
             $ Salary
                                 : int 90000 65000 150000 60000 200000 55000 120000 8000
            0 45000 110000 ...
                                  : int 32 28 45 36 52 29 42 31 26 38 ...
             $ Age
```

# **Exploratory Data Analysis**

#### **Results For Male And Female Salaries**



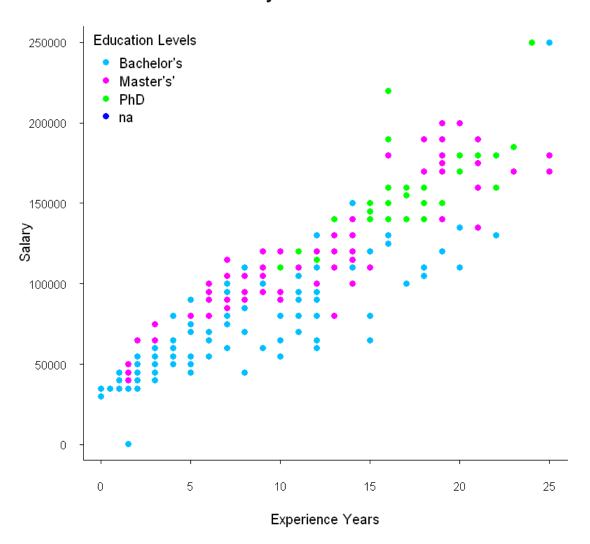
```
In [5]: ▶ levels(salary$Education.Level[])
```

' 'Bachelor\'s' 'Master\'s' 'PhD'

```
    | x = salary$Years.of.Experience

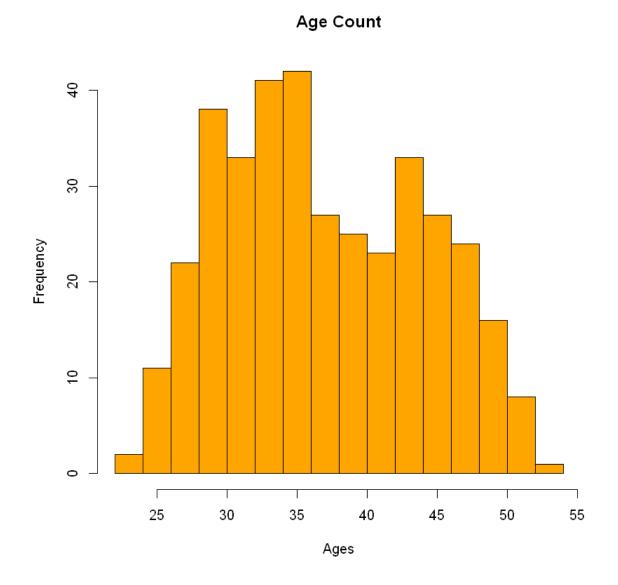
In [6]:
            y = salary$Salary
            plot(x,y, type = 'n',
                 main = 'Salary v Education Level',
                 xlab = 'Experience Years', ylab = 'Salary',
                 bty = '1', las = 1, cex.axis = .8,
                 tc1 = -0.2)
            points(x = salary$Years.of.Experience[salary$Education.Level == "Bachelor's"],
                   y = salary$Salary[salary$Education.Level == "Bachelor's"],
                   pch = 16, col = 'deepskyblue')
            points(x = salary$Years.of.Experience[salary$Education.Level == "Master\'s"],
                   y = salary$Salary[salary$Education.Level == "Master\'s"],
                   pch = 16, col = 'magenta')
            points(x = salary$Years.of.Experience[salary$Education.Level == "PhD"],
                   y = salary$Salary[salary$Education.Level == "PhD"],
                   pch = 16, col = 'green')
            points(x = salary$Years.of.Experience[salary$Education.Level == ''],
                   y = salary$Salary[salary$Education.Level == ''],
                   pch = 16, col = 'blue')
            leg_cols = c("deepskyblue", "magenta", "green", "blue")
            leg_sym = c(16, 16, 16, 16)
            leg_lab = c("Bachelor's", "Master\'s'", "PhD", "na")
            legend('topleft',
                   col = leg_cols, pch = leg_sym,
                   legend = leg lab, bty = "n",
                   title = "Education Levels")
```

# Salary v Education Level



```
In [7]: N layout(matrix(1:2, 1, 2))
hist(fem, main = "Salary for Females", col = "lightblue", xlab = "Salaries", brown freq = TRUE)
hist(male, main = "Salary for Males", col = "red", xlab = "Salaries", breaks = Ifreq = TRUE)
```





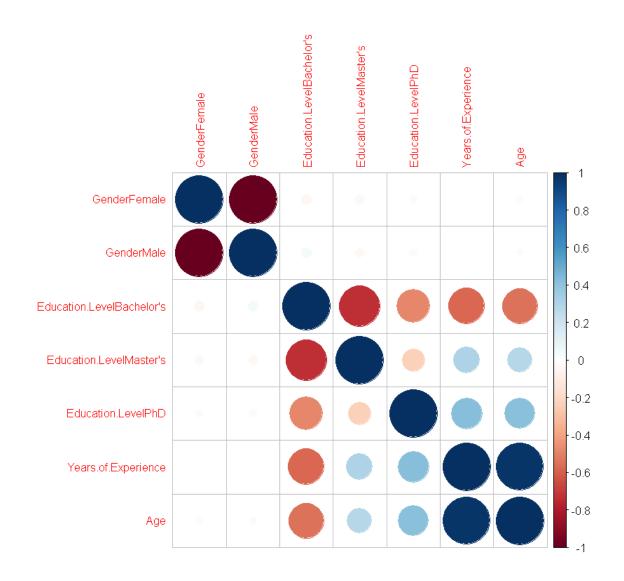
# **Statistical Analysis**

```
In [9]: 
■ age_form = Salary ~ Gender + Education.Level + Years.of.Experience + Age
form = Salary ~ Gender + Education.Level + Years.of.Experience
```

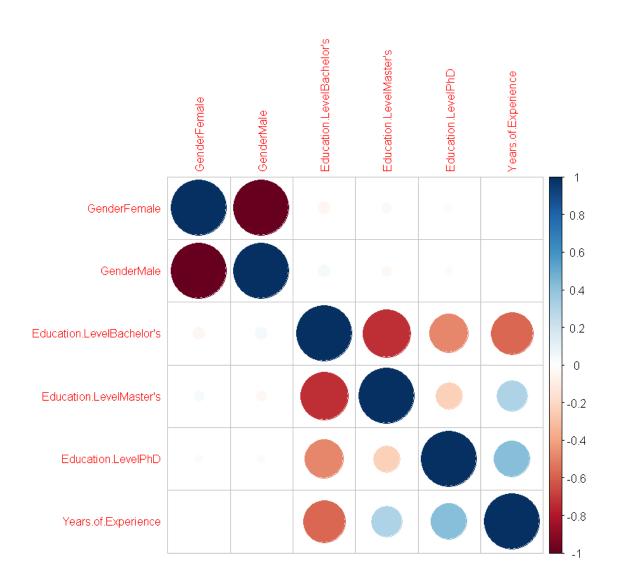
```
In [10]: N X = model.matrix(form, data = salary)
X[1,]
age_X = model.matrix(age_form, data = salary)
```

(Intercept) 1
GenderFemale 0
GenderMale 1
Education.LevelBach... 1
Education.LevelMast... 0
Education.LevelPhD 0
Years.of.Experience 5

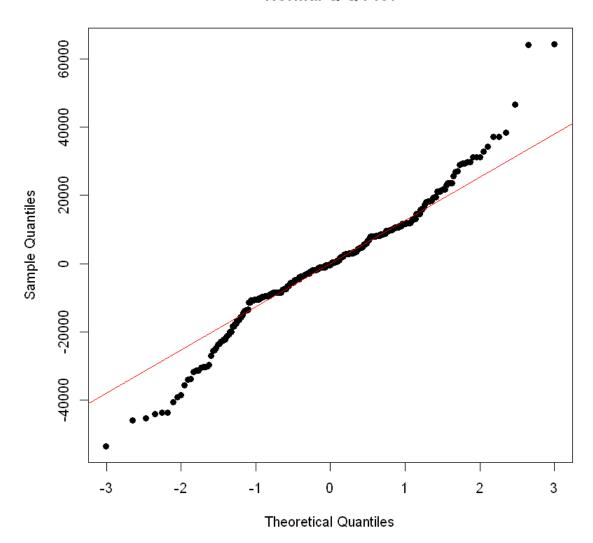
corrplot 0.92 loaded



```
In [12]: | library(corrplot)
    corrplot(cor(X[,-1]), tl.cex = .8)
```

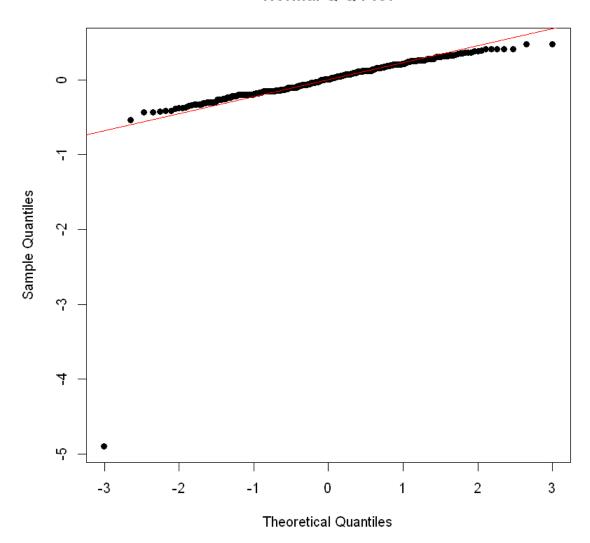


#### **Normal Q-Q Plot**

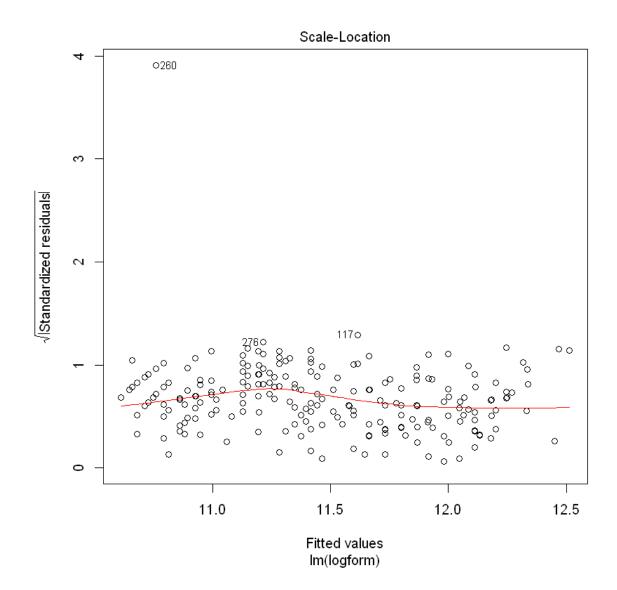


• We can see that the data is not normal enough to contain normality of residuals so we will attempt to Transform the data by taking the Log() of our salary response

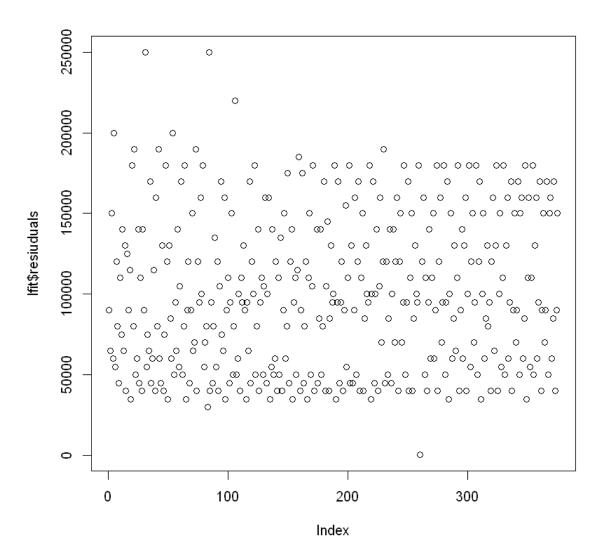
#### Normal Q-Q Plot



In [18]: ▶ plot(lfit, which = 3)



In [19]: ▶ plot(salary\$Salary, lfit\$resiuduals)



```
In [20]:

■ summary(age_lfit)
             Call:
             lm(formula = age_log_form, data = salary)
             Residuals:
                 Min
                          10 Median
                                          3Q
                                                 Max
             -4.9327 -0.1280 -0.0031 0.1393 0.4445
             Coefficients:
                                     Estimate Std. Error t value Pr(>|t|)
             (Intercept)
                                      9.47930
                                                0.31552 30.043 < 2e-16 ***
             GenderMale
                                      0.06118
                                                0.03316
                                                          1.845 0.065838 .
                                                          4.605 5.7e-06 ***
             Education.LevelMaster's 0.20142
                                                0.04374
             Education.LevelPhD
                                                          3.431 0.000670 ***
                                      0.20106
                                                0.05860
             Years.of.Experience
                                      0.02182
                                                0.01286
                                                          1.697 0.090512 .
             Age
                                      0.04198
                                                0.01161
                                                          3.616 0.000341 ***
             Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
             Residual standard error: 0.3169 on 367 degrees of freedom
               (2 observations deleted due to missingness)
             Multiple R-squared: 0.7159,
                                            Adjusted R-squared: 0.7121
                           185 on 5 and 367 DF, p-value: < 2.2e-16
             F-statistic:
In [21]:

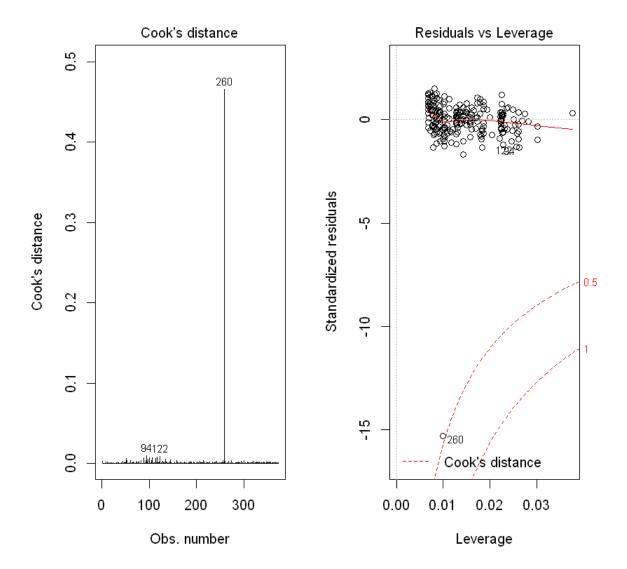
■ summary(lfit)
             Call:
             lm(formula = logform, data = salary)
             Residuals:
                 Min
                          10 Median
                                          3Q
                                                Max
             -4.9023 -0.1462 0.0057 0.1603 0.4795
             Coefficients:
                                      Estimate Std. Error t value Pr(>|t|)
                                                0.035331 300.396 < 2e-16 ***
             (Intercept)
                                     10.613363
             GenderMale
                                      0.046497
                                                0.033443
                                                           1.390 0.16526
             Education.LevelMaster's 0.181446
                                                0.044096
                                                           4.115 4.79e-05 ***
             Education.LevelPhD
                                      0.183966
                                                0.059363
                                                           3.099 0.00209 **
                                                0.003177 21.065 < 2e-16 ***
             Years.of.Experience
                                      0.066933
             Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
             Residual standard error: 0.322 on 368 degrees of freedom
               (2 observations deleted due to missingness)
             Multiple R-squared: 0.7058,
                                            Adjusted R-squared: 0.7026
```

F-statistic: 220.7 on 4 and 368 DF, p-value: < 2.2e-16

	2.5 %	97.5 %
(Intercept)	8.858846279	10.09976267
GenderMale	-0.004025752	0.12637640
Education.LevelMaster's	0.115405674	0.28743033
Education.LevelPhD	0.085815481	0.31629535
Years.of.Experience	-0.003462775	0.04711255
Age	0.019151193	0.06480510

### 

	2.5 %	97.5 %
(Intercept)	10.54388645	10.68283944
GenderMale	-0.01926598	0.11226081
Education.LevelMaster's	0.09473426	0.26815819
Education.LevelPhD	0.06723237	0.30069877
Years.of.Experience	0.06068464	0.07318103



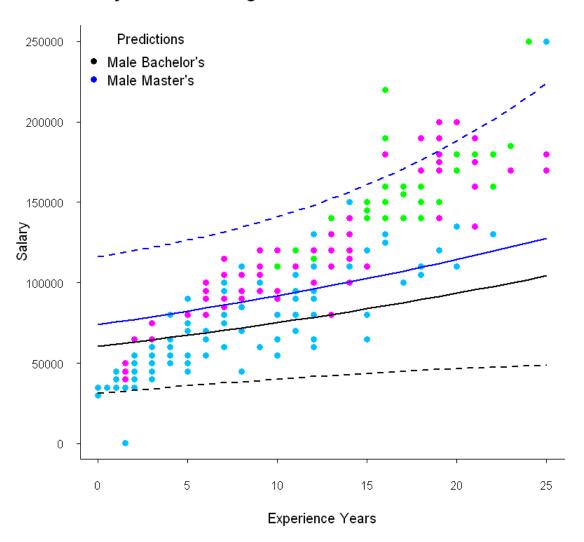
```
In [25]:
          male_master = data.frame(Gender = "Male",
                                            Education.Level = "Master's",
                                            Years.of.Experience = seq(0,25))
             male bachelor = data.frame(Gender = "Male",
                                            Education.Level = "Bachelor's",
                                            Years.of.Experience = seq(0,25))
             age male master = data.frame(Gender = "Male",
                                                                          #new data frame with
                                      Education.Level = "Master's",
                                      Years.of.Experience = seq(0,25),
                                      Age = 35)
             age_male_bachelor = data.frame(Gender = "Male",
                                                                          #new data framewith
                                            Education.Level = "Bachelor's",
                                            Years.of.Experience = seq(0,25),
                                            Age = 35)
             #prediction with Age variable
             male_bach_pred = exp(predict(lfit, newdata = male_bachelor, interval = "predict")
             male mast pred = exp(predict(lfit, newdata = male master, interval = "prediction")
             #predictions with age variable included for bachelors and masters
             age_male_bach_pred = exp(predict(age_lfit, newdata = age_male_bachelor, interval
             age male mast pred = exp(predict(age lfit, newdata = age male master, interval
```

```
In [26]:

    | x = salary$Years.of.Experience

             y = salary$Salary
             ExpSeq = seq(0,25)
             plot(x,y, type = 'n',
                  main = "Salary Prediction w/Age for Males with Bachelor's vs Master's",
                  xlab = 'Experience Years', ylab = 'Salary',
                  bty = 'l', las = 1, cex.axis = .8,
                  tc1 = -0.2)
             points(x = salary$Years.of.Experience[salary$Education.Level == "Bachelor's"],
                    y = salary$Salary[salary$Education.Level == "Bachelor's"],
                    pch = 16, col = 'deepskyblue')
             points(x = salary$Years.of.Experience[salary$Education.Level == "Master\'s"],
                    y = salary$Salary[salary$Education.Level == "Master\'s"],
                    pch = 16, col = 'magenta')
             points(x = salary$Years.of.Experience[salary$Education.Level == "PhD"],
                    y = salary$Salary[salary$Education.Level == "PhD"],
                    pch = 16, col = 'green')
             lines(x = ExpSeq, y = age_male_bach_pred[,'fit'], col = 'black', lwd = 2)
             lines(x = ExpSeq, y = age_male_bach_pred[,'upr'], col = 'black', lwd = 2, lty =
             lines(x = ExpSeq, y = age_male_bach_pred[,'lwr'], col = 'black', lwd = 2, lty =
             lines(x = ExpSeq, y = age_male_mast_pred[,'fit'], col = 'blue', lwd = 2)
             lines(x = ExpSeq, y = age_male_bach_pred[,'upr'], col = 'blue', lwd = 2, lty =
             lines(x = ExpSeq, y = age_male_bach_pred[,'lwr'], col = 'black', lwd = 2, lty =
             leg cols = c("black", "blue")
             leg_sym = c(16, 16)
             leg lab = c("Male Bachelor's", "Male Master's")
             legend('topleft',
                    col = leg_cols, pch = leg_sym,
                    legend = leg_lab, bty = "n",
                    title = "Predictions")
```

# Salary Prediction w/Age for Males with Bachelor's vs Master's

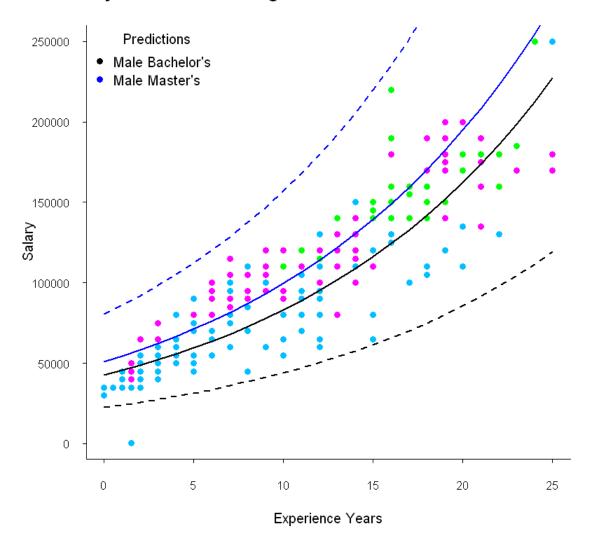


```
In [27]:

    | x = salary$Years.of.Experience

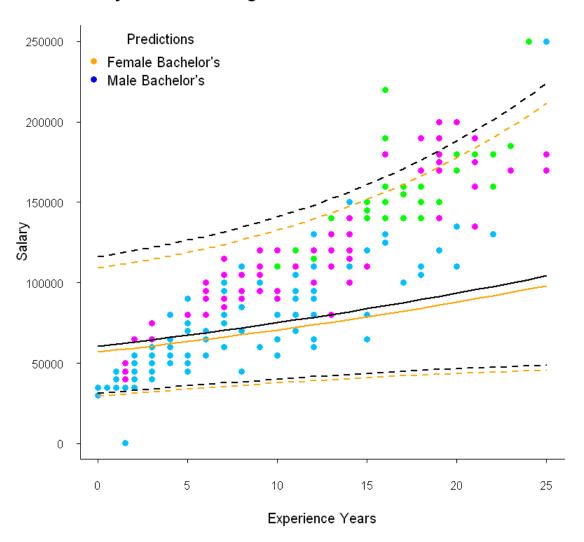
             y = salary$Salary
             ExpSeq = seq(0,25)
             plot(x,y, type = 'n',
                  main = "Salary Prediction w/out Age for Males with Bachelor's vs Master's"
                  xlab = 'Experience Years', ylab = 'Salary',
                  bty = 'l', las = 1, cex.axis = .8,
                  tc1 = -0.2)
             points(x = salary$Years.of.Experience[salary$Education.Level == "Bachelor's"],
                    y = salary$Salary[salary$Education.Level == "Bachelor's"],
                    pch = 16, col = 'deepskyblue')
             points(x = salary$Years.of.Experience[salary$Education.Level == "Master\'s"],
                    y = salary$Salary[salary$Education.Level == "Master\'s"],
                    pch = 16, col = 'magenta')
             points(x = salary$Years.of.Experience[salary$Education.Level == "PhD"],
                    y = salary$Salary[salary$Education.Level == "PhD"],
                    pch = 16, col = 'green')
             lines(x = ExpSeq, y = male_bach_pred[,'fit'], col = 'black', lwd = 2)
             lines(x = ExpSeq, y = male_bach_pred[,'upr'], col = 'black', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = male_bach_pred[,'lwr'], col = 'black', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = male_mast_pred[,'fit'], col = 'blue', lwd = 2)
             lines(x = ExpSeq, y = male_bach_pred[,'upr'], col = 'blue', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = male_bach_pred[,'lwr'], col = 'black', lwd = 2, lty = 2)
             leg cols = c("black", "blue")
             leg_sym = c(16, 16)
             leg lab = c("Male Bachelor's", "Male Master's")
             legend('topleft',
                    col = leg_cols, pch = leg_sym,
                    legend = leg_lab, bty = "n",
                    title = "Predictions")
```

#### Salary Prediction w/out Age for Males with Bachelor's vs Master's



```
In [29]:
          x = salary$Years.of.Experience
             v = salary$Salary
             ExpSeq = seq(0,25)
             plot(x,y, type = 'n',
                  main = "Salary Prediction w/Age for Males vs Females with Bachelor's",
                  xlab = 'Experience Years', ylab = 'Salary',
                  bty = 'l', las = 1, cex.axis = .8,
                  tc1 = -0.2)
             points(x = salary$Years.of.Experience[salary$Education.Level == "Bachelor's"],
                    y = salary$Salary[salary$Education.Level == "Bachelor's"],
                    pch = 16, col = 'deepskyblue')
             points(x = salary$Years.of.Experience[salary$Education.Level == "Master\'s"],
                    y = salary$Salary[salary$Education.Level == "Master\'s"],
                    pch = 16, col = 'magenta')
             points(x = salary$Years.of.Experience[salary$Education.Level == "PhD"],
                    y = salary$Salary[salary$Education.Level == "PhD"],
                    pch = 16, col = 'green')
             points(x = salary$Years.of.Experience[salary$Education.Level == ''],
                    y = salary$Salary[salary$Education.Level == ''],
                    pch = 16, col = 'blue')
             lines(x = ExpSeq, y = age_male_bach_pred[,'fit'], col = 'black', lwd = 2)
             lines(x = ExpSeq, y = age_male_bach_pred[,'upr'], col = 'black', lwd = 2, lty =
             lines(x = ExpSeq, y = age_male_bach_pred[,'lwr'], col = 'black', lwd = 2, lty =
             lines(x = ExpSeq, y = age_fem_bach_pred[,'fit'], col = 'orange', lwd = 2)
             lines(x = ExpSeq, y = age fem bach pred[,'upr'], col = 'orange', lwd = 2, lty =
             lines(x = ExpSeq, y = age_fem_bach_pred[,'lwr'], col = 'orange', lwd = 2, lty =
             leg_cols = c("orange", "blue")
             leg sym = c(16, 16)
             leg_lab = c("Female Bachelor's", "Male Bachelor's")
             legend('topleft',
                    col = leg_cols, pch = leg_sym,
                    legend = leg lab, bty = "n",
                    title = "Predictions")
```

# Salary Prediction w/Age for Males vs Females with Bachelor's

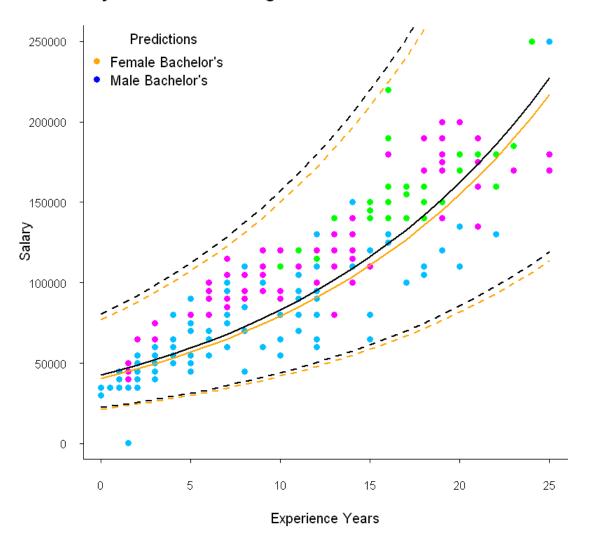


```
In [30]:

    | x = salary$Years.of.Experience

             y = salary$Salary
             ExpSeq = seq(0,25)
             plot(x,y, type = 'n',
                  main = "Salary Prediction w/out Age for Males vs Females with Bachelor's",
                  xlab = 'Experience Years', ylab = 'Salary',
                  bty = 'l', las = 1, cex.axis = .8,
                  tc1 = -0.2)
             points(x = salary$Years.of.Experience[salary$Education.Level == "Bachelor's"],
                    y = salary$Salary[salary$Education.Level == "Bachelor's"],
                    pch = 16, col = 'deepskyblue')
             points(x = salary$Years.of.Experience[salary$Education.Level == "Master\'s"],
                    y = salary$Salary[salary$Education.Level == "Master\'s"],
                    pch = 16, col = 'magenta')
             points(x = salary$Years.of.Experience[salary$Education.Level == "PhD"],
                    y = salary$Salary[salary$Education.Level == "PhD"],
                    pch = 16, col = 'green')
             points(x = salary$Years.of.Experience[salary$Education.Level == ''],
                    y = salary$Salary[salary$Education.Level == ''],
                    pch = 16, col = 'blue')
             lines(x = ExpSeq, y = male_bach_pred[,'fit'], col = 'black', lwd = 2)
             lines(x = ExpSeq, y = male_bach_pred[,'upr'], col = 'black', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = male_bach_pred[,'lwr'], col = 'black', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = fem_bach_pred[,'fit'], col = 'orange', lwd = 2)
             lines(x = ExpSeq, y = fem bach pred[,'upr'], col = 'orange', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = fem_bach_pred[,'lwr'], col = 'orange', lwd = 2, lty = 2)
             leg_cols = c("orange", "blue")
             leg sym = c(16, 16)
             leg_lab = c("Female Bachelor's", "Male Bachelor's")
             legend('topleft',
                    col = leg_cols, pch = leg_sym,
                    legend = leg lab, bty = "n",
                    title = "Predictions")
```

# Salary Prediction w/out Age for Males vs Females with Bachelor's



```
In [31]:

  | forSection = log(Salary) ~ Age + Years.of.Experience + Education.Level

             forSect = step(lm(forSection, data = salary),
                             scope = list(upper = log(Salary) ~ Age + Years.of.Experience + E
                                          lower = log(Salary) ~ 1), direction = "both")
             Start: AIC=-849.9
             log(Salary) ~ Age + Years.of.Experience + Education.Level
                                     Df Sum of Sq
                                                      RSS
                                                              AIC
             + Job.Title
                                          24.4318 12.764 -902.85
                                    173
             + Gender
                                      1
                                           0.3418 36.854 -851.35
             <none>
                                                  37.196 -849.90
             - Years.of.Experience
                                      1
                                           0.3817 37.578 -848.10
             - Age
                                      1
                                           1.1719 38.368 -840.33

    Education.Level

                                      2
                                           2.2597 39.456 -831.91
             Step: AIC=-902.85
             log(Salary) ~ Age + Years.of.Experience + Education.Level + Job.Title
                                     Df Sum of Sq
                                                      RSS
                                                              AIC
                                           0.0415 12.806 -905.64
             - Education.Level
             - Years.of.Experience
                                           0.0006 12.765 -904.83
                                      1
             <none>
                                                  12.764 -902.85
             + Gender
                                      1
                                           0.0387 12.726 -901.98
                                           0.4268 13.191 -892.58
             - Age
                                      1
                                          24.4318 37.196 -849.90

    Job.Title

                                    173
             Step: AIC=-905.64
             log(Salary) ~ Age + Years.of.Experience + Job.Title
                                     Df Sum of Sa
                                                      RSS
                                           0.0101 12.816 -907.34
             - Years.of.Experience
                                      1
             <none>
                                                  12.806 -905.64
             + Gender
                                      1
                                           0.0282 12.778 -904.46
                                           0.0415 12.764 -902.85
             + Education.Level
                                      2
                                           0.4027 13.209 -896.09
             - Age
                                      1
             - Job.Title
                                          26.6501 39.456 -831.91
                                    173
             Step: AIC=-907.34
             log(Salary) ~ Age + Job.Title
                                     Df Sum of Sq
                                                      RSS
                                                              AIC
             <none>
                                                   12.816 -907.34
                                           0.0316 12.784 -906.26
             + Gender
                                      1
             + Years.of.Experience
                                           0.0101 12.806 -905.64
                                      1
             + Education.Level
                                      2
                                           0.0510 12.765 -904.83
             - Age
                                      1
                                           2.8522 15.668 -834.39
             - Job.Title
                                          27.7076 40.524 -823.95
                                    173
             addmodel = log(Salary) ~ Gender + Education.Level + Years.of.Experience + Gender
In [32]:
             addfit = lm(addmodel, data = salary)
In [33]:
```

```
In [34]:

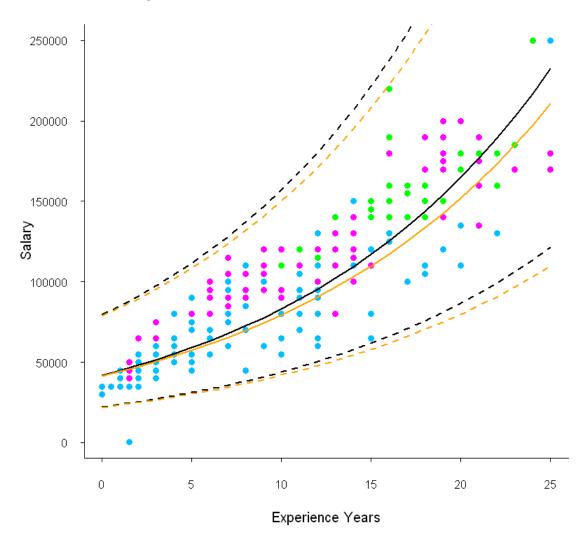
    summary(addfit)

             Call:
             lm(formula = addmodel, data = salary)
             Residuals:
                 Min
                          10 Median
                                          3Q
                                                 Max
             -4.8891 -0.1442 0.0015 0.1639 0.4776
             Coefficients:
                                             Estimate Std. Error t value Pr(>|t|)
             (Intercept)
                                            10.632737
                                                        0.045440 233.994 < 2e-16 ***
             GenderMale
                                             0.011598
                                                        0.061347
                                                                   0.189 0.85015
             Education.LevelMaster's
                                                        0.044130
                                             0.181213
                                                                   4.106 4.96e-05 ***
             Education.LevelPhD
                                                                   3.117 0.00197 **
                                             0.185293
                                                        0.059439
             Years.of.Experience
                                             0.064985
                                                        0.004283 15.172 < 2e-16 ***
             GenderMale:Years.of.Experience 0.003482
                                                        0.005129
                                                                   0.679 0.49770
             Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
             Residual standard error: 0.3223 on 367 degrees of freedom
               (2 observations deleted due to missingness)
             Multiple R-squared: 0.7062,
                                             Adjusted R-squared: 0.7022
             F-statistic: 176.4 on 5 and 367 DF, p-value: < 2.2e-16
In [35]:
          ▶ beta = coef(addfit)
             beta
                         (Intercept)
                                    10.6327372465597
                       GenderMale
                                    0.0115980400473509
             Education.LevelMast...
                                    0.181213011323071
```

```
| Content | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/246559/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/24659/ | 10.632/3/2467/ | 10.632/3/2467/ | 10.632/3/2467/ | 10.632/3/2467/ | 10.632/3/2467/ | 10.632/3/2467/ | 10.6
```

```
In [37]:
          ▶ plot(x,y, type = 'n',
                  main = "Salary Prediction for Males vs Females with Bachelor's",
                  xlab = 'Experience Years', ylab = 'Salary',
                  bty = '1', las = 1, cex.axis = .8,
                  tc1 = -0.2)
             points(x = salary$Years.of.Experience[salary$Education.Level == "Bachelor's"],
                    y = salary$Salary[salary$Education.Level == "Bachelor's"],
                    pch = 16, col = 'deepskyblue')
             points(x = salary$Years.of.Experience[salary$Education.Level == "Master\'s"],
                    y = salary$Salary[salary$Education.Level == "Master\'s"],
                    pch = 16, col = 'magenta')
             points(x = salary$Years.of.Experience[salary$Education.Level == "PhD"],
                    y = salary$Salary[salary$Education.Level == "PhD"],
                    pch = 16, col = 'green')
             points(x = salary$Years.of.Experience[salary$Education.Level == ''],
                    y = salary$Salary[salary$Education.Level == ''],
                    pch = 16, col = 'blue')
             lines(x = ExpSeq, y = male_addpred[,'fit'], col = 'black', lwd = 2)
             lines(x = ExpSeq, y = male_addpred[,'upr'], col = 'black', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = male addpred[,'lwr'], col = 'black', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = fem_addpred[,'fit'], col = 'orange', lwd = 2)
             lines(x = ExpSeq, y = fem_addpred[,'upr'], col = 'orange', lwd = 2, lty = 2)
             lines(x = ExpSeq, y = fem_addpred[,'lwr'], col = 'orange', lwd = 2, lty = 2)
```

# Salary Prediction for Males vs Females with Bachelor's



In [ ]: **M**