

HW#3

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```
# Load the data  
library(faraway)  
tg <- faraway::teengamb
```

Part a

```
#Fit the first model  
lm1 <- lm(gamble ~ sex + status + income + verbal, data = tg)  
# Create the design matrices and hat matrices  
X <- model.matrix(lm1)  
X1 <- X[,1]  
H <- X%*%solve(t(X)%*%X)%*%t(X)  
H1 <- X1%*%solve(t(X1)%*%X1)%*%t(X1)  
# Define y  
y = tg$gamble  
# Compute yhat  
yhat <- H%*%y  
# Compute ybar  
ybar <- H1%*%y  
# Compute SSR and SST  
SSR <- t(yhat - ybar)%*%(yhat - ybar)  
SST <- t(y - ybar)%*%(y - ybar)  
# Compute r squared  
rsqrt <- SSR/SST  
rsqrt
```

```
##           [,1]  
## [1,] 0.5267234
```

Part b

```
# Compute the residual  
tg$r = y - yhat  
# Find what is the maximal residual value  
rmax = max(tg$r)  
rmax
```

```
## [1] 94.25222
```

```
# Find which case has largest residual  
which(tg$r == rmax)
```

```
## [1] 24
```

Part c

```
# Find the mean of the residuals  
mean(tg$r)
```

```
## [1] -1.359206e-14
```

```
mean(lm1$residuals)
```

```
## [1] -3.065293e-17
```

```
# Find the merdian of the residuals  
median(tg$r)
```

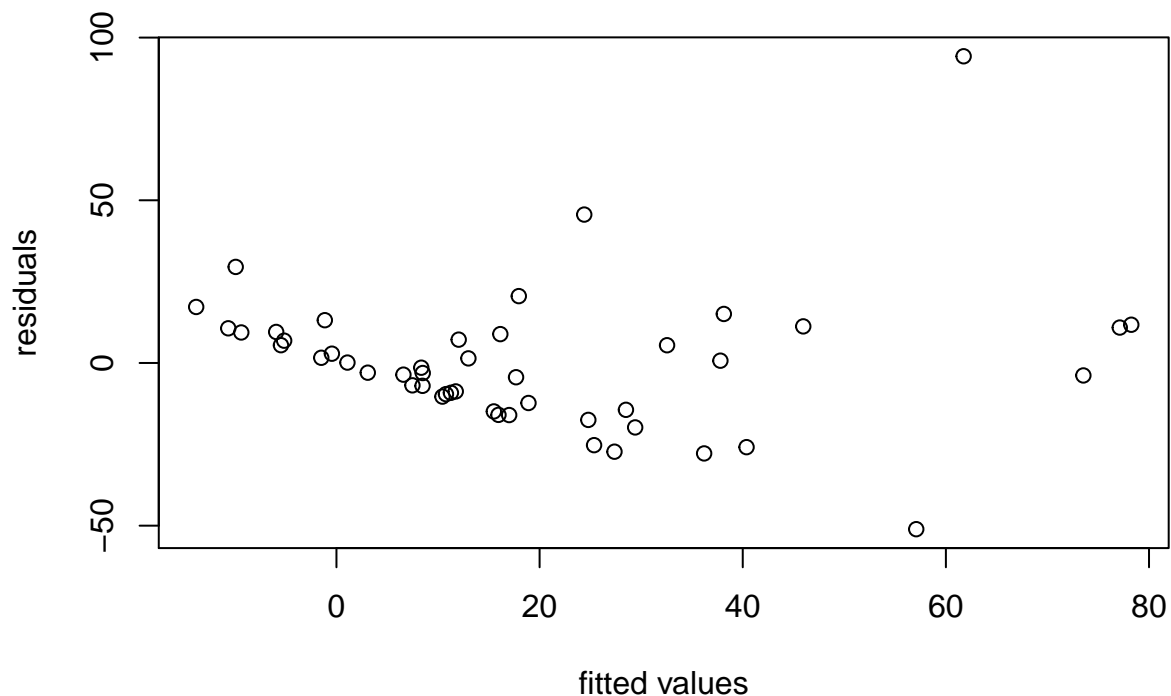
```
## [1] -1.451392
```

```
median(lm1$residuals)
```

```
## [1] -1.451392
```

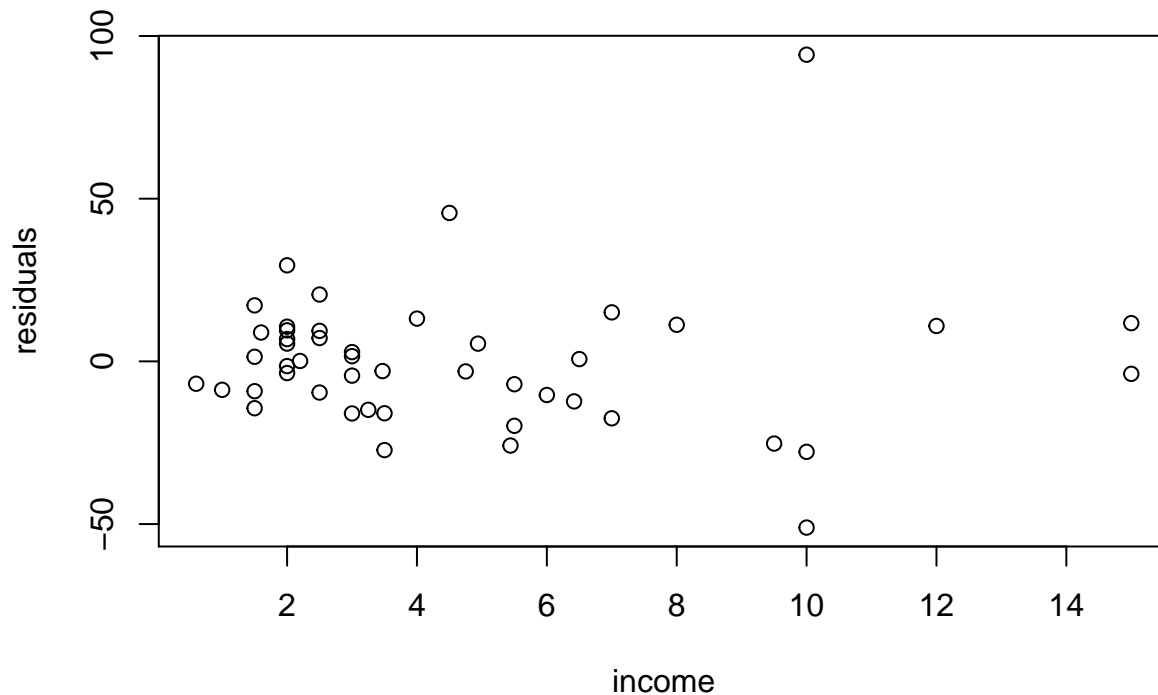
Part d

```
# Plot the residuals against the fitted values  
plot(lm1$fitted,tg$r,xlab = "fitted values", ylab = "residuals")
```



Part e

```
# Plot the residuals against the variable income  
plot(tg$income,tg$r,xlab = "income", ylab = "residuals")
```



Part d

```
lm1
```

```
##
## Call:
## lm(formula = gamble ~ sex + status + income + verbal, data = tg)
##
## Coefficients:
## (Intercept)      sex      status      income      verbal
##  22.55565    -22.11833     0.05223     4.96198    -2.95949
```

```
tg
```

```
##   sex status income verbal gamble      r
## 1   1    51   2.00      8    0.00 10.6507430
## 2   1    28   2.50      8    0.00  9.3711318
## 3   1    37   2.00      6    0.00  5.4630298
## 4   1    28   7.00      4    7.30 -17.4957487
## 5   1    65   2.00      8   19.60 29.5194692
## 6   1    61   3.47      6    0.10 -2.9846919
## 7   1    28   5.50      7    1.45 -7.0242994
## 8   1    27   6.42      5    6.60 -12.3060734
## 9   1    43   2.00      6    1.70  6.8496267
## 10  1    18   6.00      7    0.10 -10.3329505
## 11  1    18   3.00      6    0.10  1.5934936
## 12  1    43   4.75      6    5.40 -3.0958161
## 13  1    30   2.20      4    1.20  0.1172839
## 14  1    28   2.00      6    3.60  9.5331344
## 15  1    38   3.00      6    2.40  2.8488167
## 16  1    38   1.50      8    3.40 17.2107726
## 17  1    28   9.50      8    0.10 -25.2627227
```

```
## 18 1 18 10.00 5 8.40 -27.7998544
## 19 1 43 4.00 8 12.00 13.1446553
## 20 0 51 3.50 9 0.00 -15.9510624
## 21 0 62 3.00 8 1.00 -16.0041386
## 22 0 47 2.50 9 1.20 -9.5801478
## 23 0 43 3.50 5 0.10 -27.2711657
## 24 0 27 10.00 4 156.00 94.2522174
## 25 0 71 6.50 7 38.50 0.6993361
## 26 0 38 1.50 7 2.10 -9.1670510
## 27 0 51 5.44 4 14.50 -25.8747696
## 28 0 38 1.00 6 3.00 -8.7455549
## 29 0 51 0.60 7 0.60 -6.8803097
## 30 0 62 5.50 8 9.60 -19.8090866
## 31 0 18 12.00 2 88.00 10.8793766
## 32 0 30 7.00 7 53.20 15.0599340
## 33 0 38 15.00 7 90.00 11.7462296
## 34 0 71 2.00 10 3.00 -3.5932770
## 35 0 28 1.50 1 14.10 -14.4016736
## 36 0 61 4.50 8 70.00 45.6051264
## 37 0 71 2.50 7 38.50 20.5472529
## 38 0 28 8.00 6 57.20 11.2429290
## 39 0 51 10.00 6 6.00 -51.0824078
## 40 0 65 1.60 6 25.00 8.8669438
## 41 0 48 2.00 9 6.90 -1.4513921
## 42 0 61 15.00 9 69.70 -3.8361619
## 43 0 75 3.00 8 13.30 -4.3831786
## 44 0 66 3.25 9 0.60 -14.8940753
## 45 0 62 4.94 6 38.00 5.4506347
## 46 0 71 1.50 7 14.40 1.4092321
## 47 0 71 2.50 9 19.20 7.1662399
```

```
summary(lm1)
```

```
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) 22.555651 17.196803  1.3116  0.19677
## sex        -22.118330  8.211115 -2.6937  0.01011
## status      0.052234  0.281112  0.1858  0.85349
## income      4.961979  1.025392  4.8391 1.792e-05
## verbal     -2.959493  2.172150 -1.3625  0.18031
##
## n = 47, p = 5, Residual SE = 22.69034, R-Squared = 0.53
```