

project2_test

Group_01

2021/7/7

```
library(tidyverse)
library(moderndiver)
library(skimr)
library(kableExtra)
library(gridExtra)
library(broom)
library(plotly)
library(GGally)
library(sjPlot)
```

```
data.score <- data %>%
  dplyr::select(Total.Number.of.Family.members, Total.Household.Income,
                Total.Food.Expenditure, Household.Head.Age, House.Floor.Area, House.Age,
                Number.of.bedrooms, Electricity)
```

```
data.score %>%
  cor()
```

	Total.Number.of.Family.members		
Total.Number.of.Family.members	1.00000000		
Total.Household.Income	0.19228742		
Total.Food.Expenditure	0.46924215		
Household.Head.Age	-0.06541636		
House.Floor.Area	-0.01415702		
House.Age	-0.07003586		
Number.of.bedrooms	0.07207630		
Electricity	0.09193871		
	Total.Household.Income	Total.Food.Expenditure	
Total.Number.of.Family.members	0.19228742	0.469242145	
Total.Household.Income	1.00000000	0.611494530	
Total.Food.Expenditure	0.61149453	1.000000000	
Household.Head.Age	0.06280405	-0.051724735	
House.Floor.Area	0.23413840	0.124320633	
House.Age	0.02471720	0.006725185	
Number.of.bedrooms	0.44137375	0.355734454	
Electricity	0.14866655	0.198610366	
	Household.Head.Age	House.Floor.Area	House.Age
Total.Number.of.Family.members	-0.06541636	-0.01415702	-0.070035856
Total.Household.Income	0.06280405	0.23413840	0.024717197
Total.Food.Expenditure	-0.05172474	0.12432063	0.006725185
Household.Head.Age	1.00000000	0.09057216	0.218079293

House.Floor.Area	0.09057216	1.00000000	0.074265080
House.Age	0.21807929	0.07426508	1.000000000
Number.of.bedrooms	0.15415511	0.37399081	0.123180471
Electricity	-0.01304412	0.10693465	0.085327324
	Number.of.bedrooms	Electricity	
Total.Number.of.Family.members	0.0720763	0.09193871	
Total.Household.Income	0.4413738	0.14866655	
Total.Food.Expenditure	0.3557345	0.19861037	
Household.Head.Age	0.1541551	-0.01304412	
House.Floor.Area	0.3739908	0.10693465	
House.Age	0.1231805	0.08532732	
Number.of.bedrooms	1.0000000	0.21376315	
Electricity	0.2137632	1.00000000	

```
my_skim <- skim_with(numeric = sfl(hist = NULL))
my_skim(data.score) %>%
  dplyr::select(-skim_type) %>%
  as_tibble() %>%
  kable(col.names = c("Variable", "Missing", "Complete", "Mean", "SD", "Min.", "1st Q.",
    "Median", "3rd Q.", "Max."),
    caption = "Summary statistics",
    booktabs = TRUE, digits = 2) %>%
  kable_styling(font_size = 10, latex_options = "hold_position") #create a summarized statistics table.
```

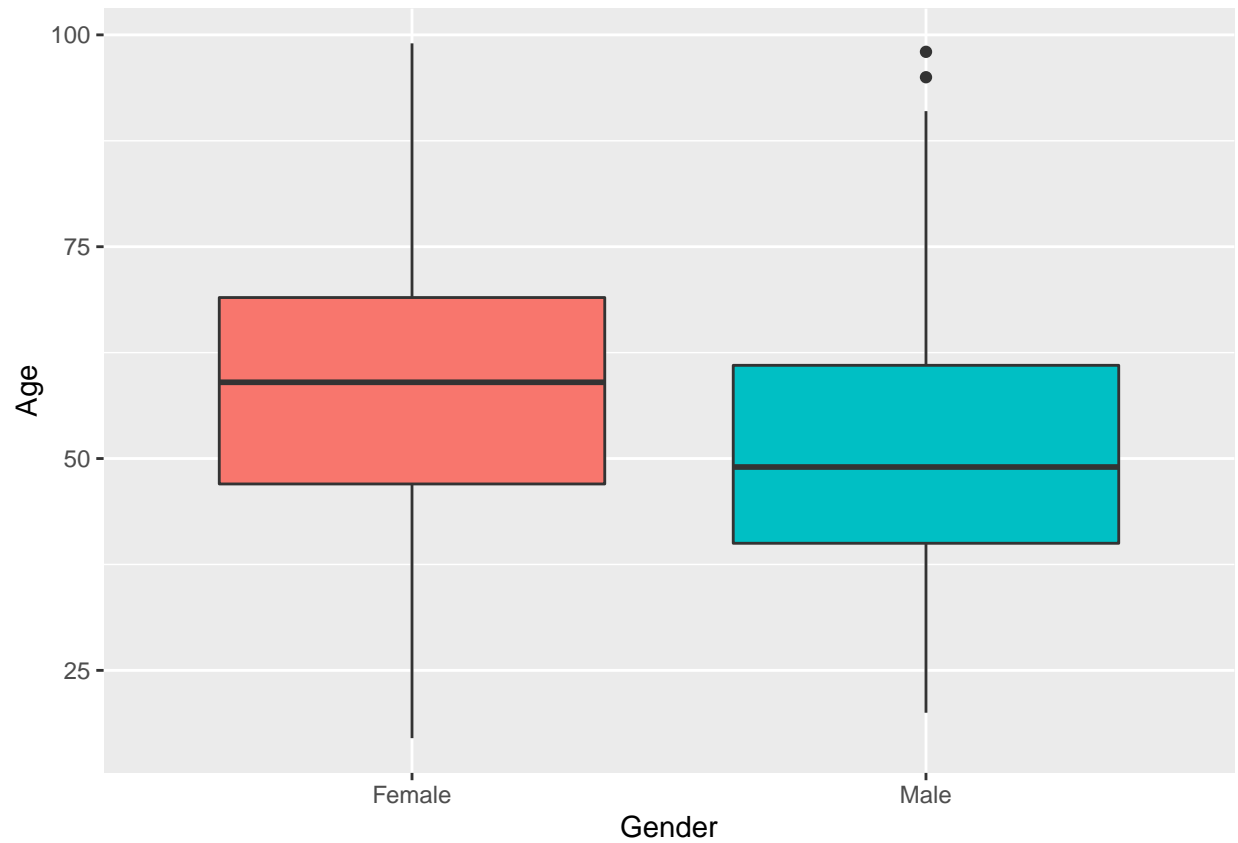
Table 1: Summary statistics

Variable	Missing	Complete	Mean	SD	Min.	1st Q.	Median	3rd Q.
Total.Number.of.Family.members	0	1	4.67	2.33	1	3	4	6
Total.Household.Income	0	1	269540.48	274564.17	11988	118565	188580	328335
Total.Food.Expenditure	0	1	80352.78	41194.36	6781	51922	73578	98493
Household.Head.Age	0	1	52.23	14.52	17	41	52	63
House.Floor.Area	0	1	90.92	99.20	5	32	54	102
House.Age	0	1	22.98	15.32	0	12	20	31
Number.of.bedrooms	0	1	2.26	1.44	0	1	2	3
Electricity	0	1	0.93	0.26	0	1	1	1

Gender&age

```
data.gender <- data %>%
  select(Household.Head.Sex, Household.Head.Age)

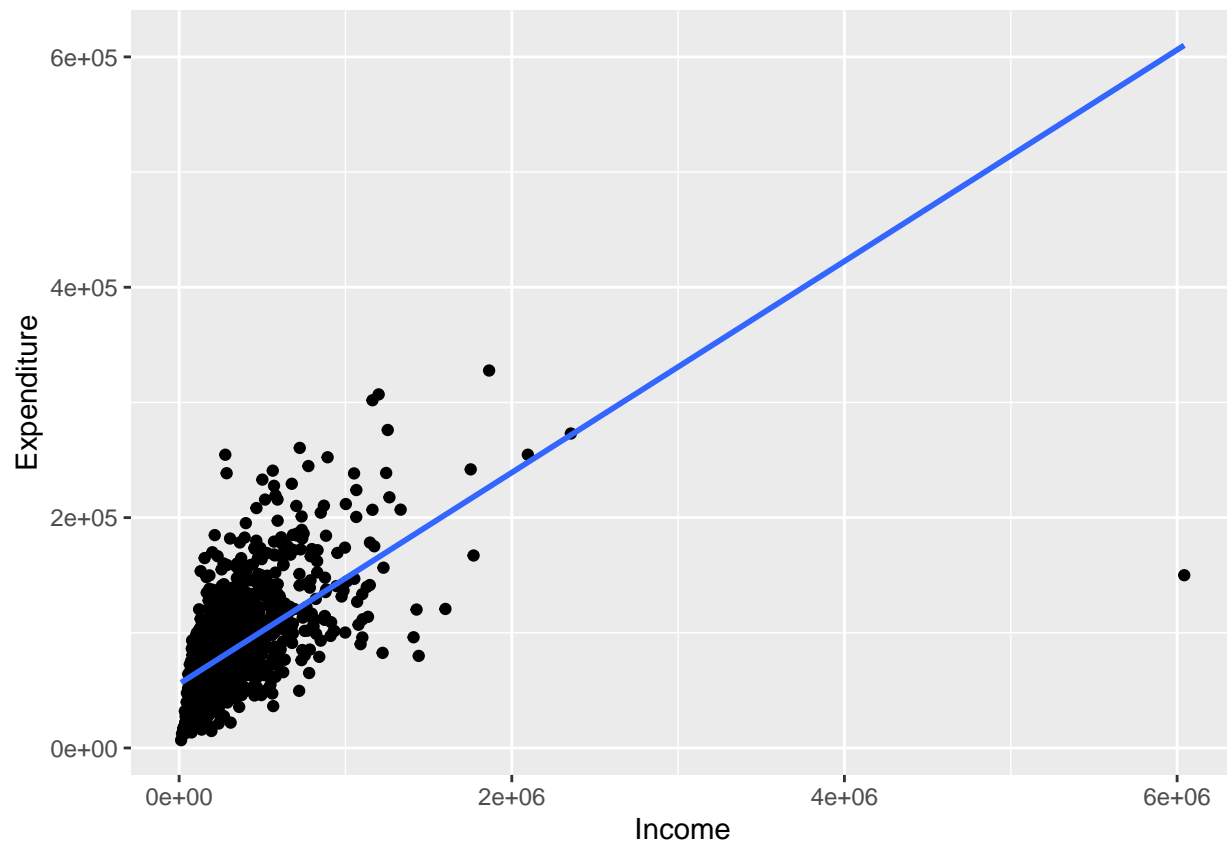
ggplot(data = data.gender, aes(x = Household.Head.Sex, y = Household.Head.Age, fill = Household.Head.Sex)) +
  geom_boxplot() +
  labs(x = "Gender", y = "Age")+
  theme(legend.position = "none")
```



balance

```
data.balance <- data %>%  
  select(Total.Household.Income, Total.Food.Expenditure)
```

```
ggplot(data = data.balance, aes(x = Total.Household.Income, y = Total.Food.Expenditure)) +  
  geom_point() +  
  labs(x = "Income", y = "Expenditure") +  
  geom_smooth(method = glm, se = FALSE) +  
  theme(legend.position = "none")
```



Model

```
model_full <- glm(Total.Number.of.Family.members ~ Total.Household.Income +
  Total.Food.Expenditure + Household.Head.Age + House.Floor.Area + House.Age +
  Number.of.bedrooms + Electricity, data = data)
```

```
model_full %>%
  summary()
```

Call:

```
glm(formula = Total.Number.of.Family.members ~ Total.Household.Income +
  Total.Food.Expenditure + Household.Head.Age + House.Floor.Area +
  House.Age + Number.of.bedrooms + Electricity, data = data)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-5.5671	-1.4626	-0.3084	1.2037	10.7417

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.746e+00	2.667e-01	10.298	< 2e-16 ***
Total.Household.Income	-1.022e-06	2.384e-07	-4.287	1.91e-05 ***
Total.Food.Expenditure	3.197e-05	1.540e-06	20.759	< 2e-16 ***

Household.Head.Age	-4.491e-04	3.520e-03	-0.128	0.89850
House.Floor.Area	-7.261e-04	5.357e-04	-1.355	0.17550
House.Age	-9.472e-03	3.301e-03	-2.870	0.00416 **
Number.of.bedrooms	-9.756e-02	4.121e-02	-2.367	0.01802 *
Electricity	1.696e-01	1.929e-01	0.879	0.37955

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 4.130968)

Null deviance: 9384.0 on 1724 degrees of freedom
 Residual deviance: 7092.9 on 1717 degrees of freedom
 AIC: 7352.3

Number of Fisher Scoring iterations: 2

```
model_significant <- glm(Total.Number.of.Family.members ~ Total.Household.Income +
  Total.Food.Expenditure + House.Age +
  Number.of.bedrooms, data = data)
```

```
model_significant %>%
  summary()
```

Call:

```
glm(formula = Total.Number.of.Family.members ~ Total.Household.Income +
  Total.Food.Expenditure + House.Age + Number.of.bedrooms,
  data = data)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-5.5796	-1.4561	-0.3048	1.1778	10.6187

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.828e+00	1.375e-01	20.560	< 2e-16 ***
Total.Household.Income	-1.061e-06	2.364e-07	-4.487	7.71e-06 ***
Total.Food.Expenditure	3.229e-05	1.513e-06	21.340	< 2e-16 ***
House.Age	-9.507e-03	3.223e-03	-2.950	0.00322 **
Number.of.bedrooms	-1.103e-01	3.855e-02	-2.862	0.00425 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 4.129948)

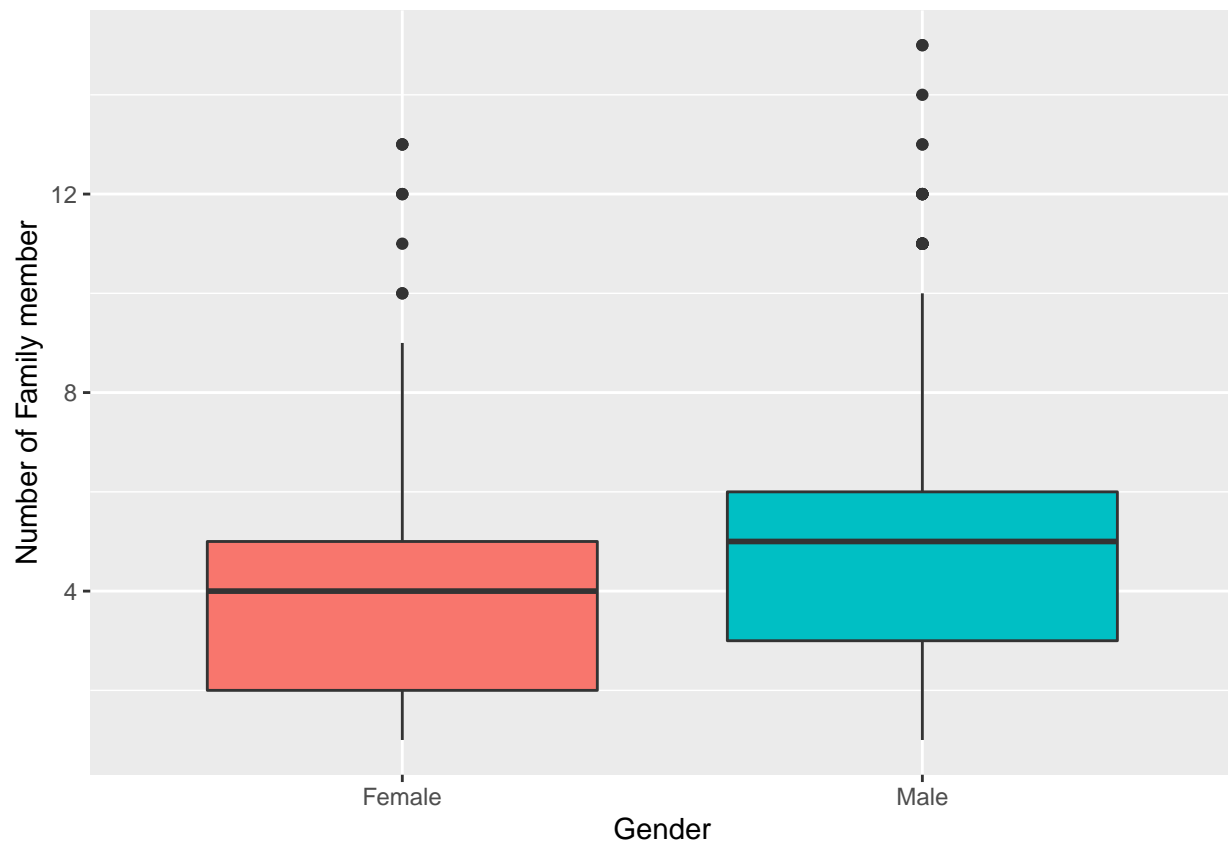
Null deviance: 9384.0 on 1724 degrees of freedom
 Residual deviance: 7103.5 on 1720 degrees of freedom
 AIC: 7348.8

Number of Fisher Scoring iterations: 2

Family number & Gender

```
data.sex_number <- data %>%  
  select(Household.Head.Sex, Total.Number.of.Family.members)
```

```
ggplot(data = data.sex_number, aes(x = Household.Head.Sex, y = Total.Number.of.Family.members, fill = H  
  geom_boxplot() +  
  labs(x = "Gender", y = "Number of Family member")+  
  theme(legend.position = "none")
```



Log-odds

```
data.sex_number$Household.Head.Sex <- as.factor(data.sex_number$Household.Head.Sex)
```

```
model_sex_number <- glm(Household.Head.Sex ~ Total.Number.of.Family.members, data = data.sex_number,  
  family = binomial(link = "logit"))
```

```
model_sex_number %>%  
  summary()
```

Call:

```
glm(formula = Household.Head.Sex ~ Total.Number.of.Family.members,
    family = binomial(link = "logit"), data = data.sex_number)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-2.4219   0.4705   0.6602   0.7163   0.9054

Coefficients:
                Estimate Std. Error z value Pr(>|z|)
(Intercept)      0.49674    0.13174   3.771 0.000163 ***
Total.Number.of.Family.members 0.18319    0.02844   6.442 1.18e-10 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1790.9  on 1724  degrees of freedom
Residual deviance: 1745.4  on 1723  degrees of freedom
AIC: 1749.4

Number of Fisher Scoring iterations: 4
```

```
levels(data.sex_number$Household.Head.Sex)
```

```
[1] "Female" "Male"
```

```
modelcoefs <- round(coef(model_sex_number),2)
```

$$\ln\left(\frac{p}{1-p}\right) = \alpha + \beta \cdot \text{number of family members} = 0.5 + 0.18 \cdot \text{number of family members},$$

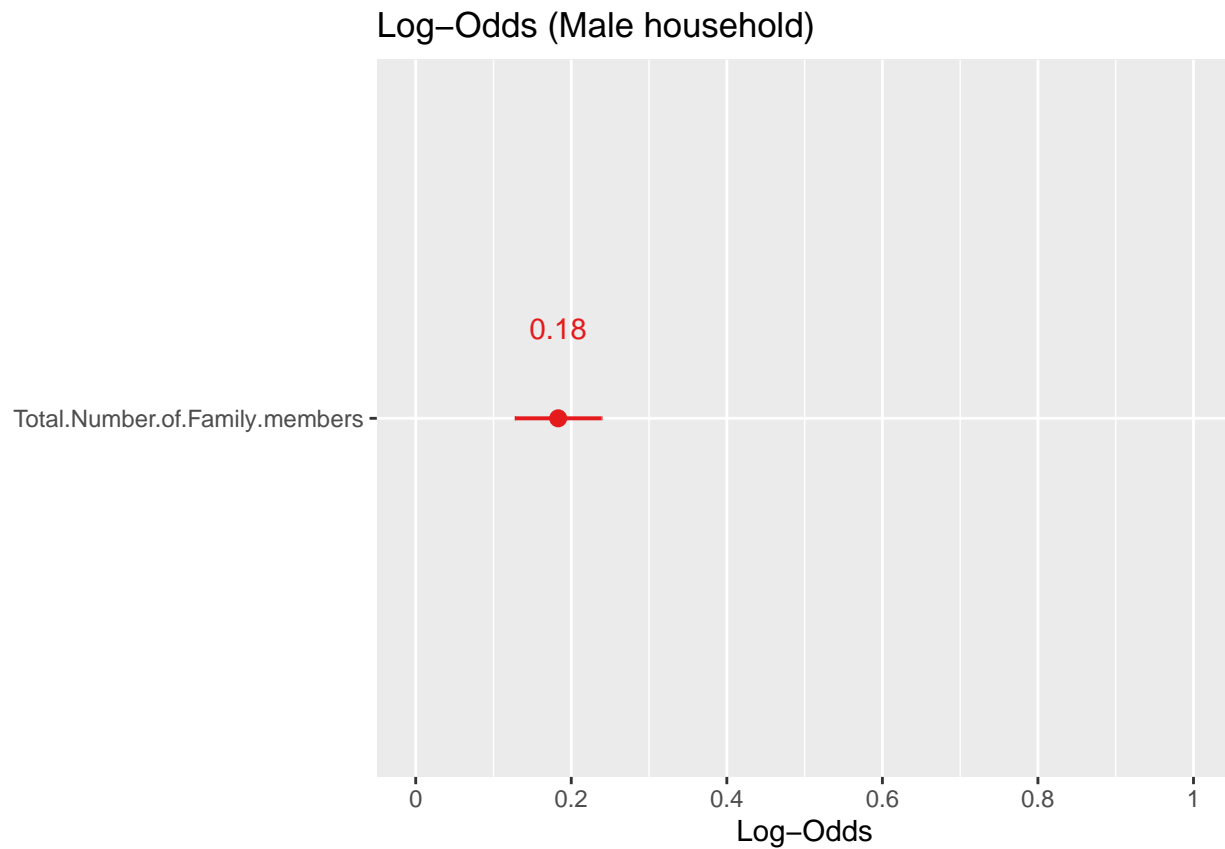
Where $p = \text{Prob}(\text{Male})$ and $1 - p = \text{Prob}(\text{Female})$. Hence, the log-odds of the household being male increase by 0.18 for every one unit increase in number of family members. This provides us with a point estimate of how the log-odds changes with age.

However, we are also interested in producing a 95% confidence interval for these log-odds.

```
confint(model_sex_number) %>%
  kable()
```

	2.5 %	97.5 %
(Intercept)	0.2388990	0.7555347
Total.Number.of.Family.members	0.1282353	0.2397474

```
plot_model(model_sex_number, show.values = TRUE, transform = NULL,
           title = "Log-Odds (Male household)", show.p = FALSE)
```

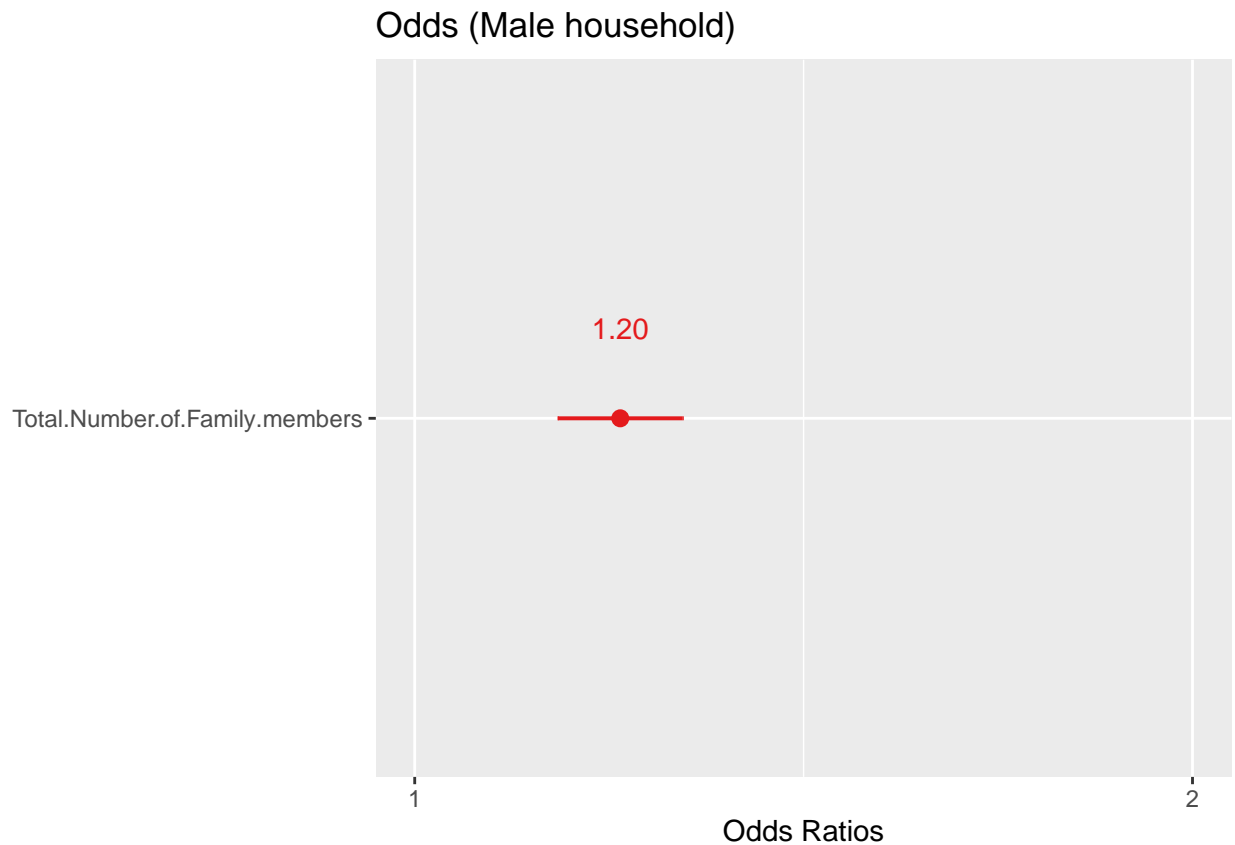


Now, let's add the estimates of the log-odds to our data set:

```
data.sex_number <- data.sex_number %>%  
  mutate(logodds.male = predict(model_sex_number))
```

Odds

```
plot_model(model_sex_number, show.values = TRUE, axis.lim = c(1,1.5),  
           title = "Odds (Male household)", show.p = FALSE)
```

Now, let's add the estimates of the odds to our data set:

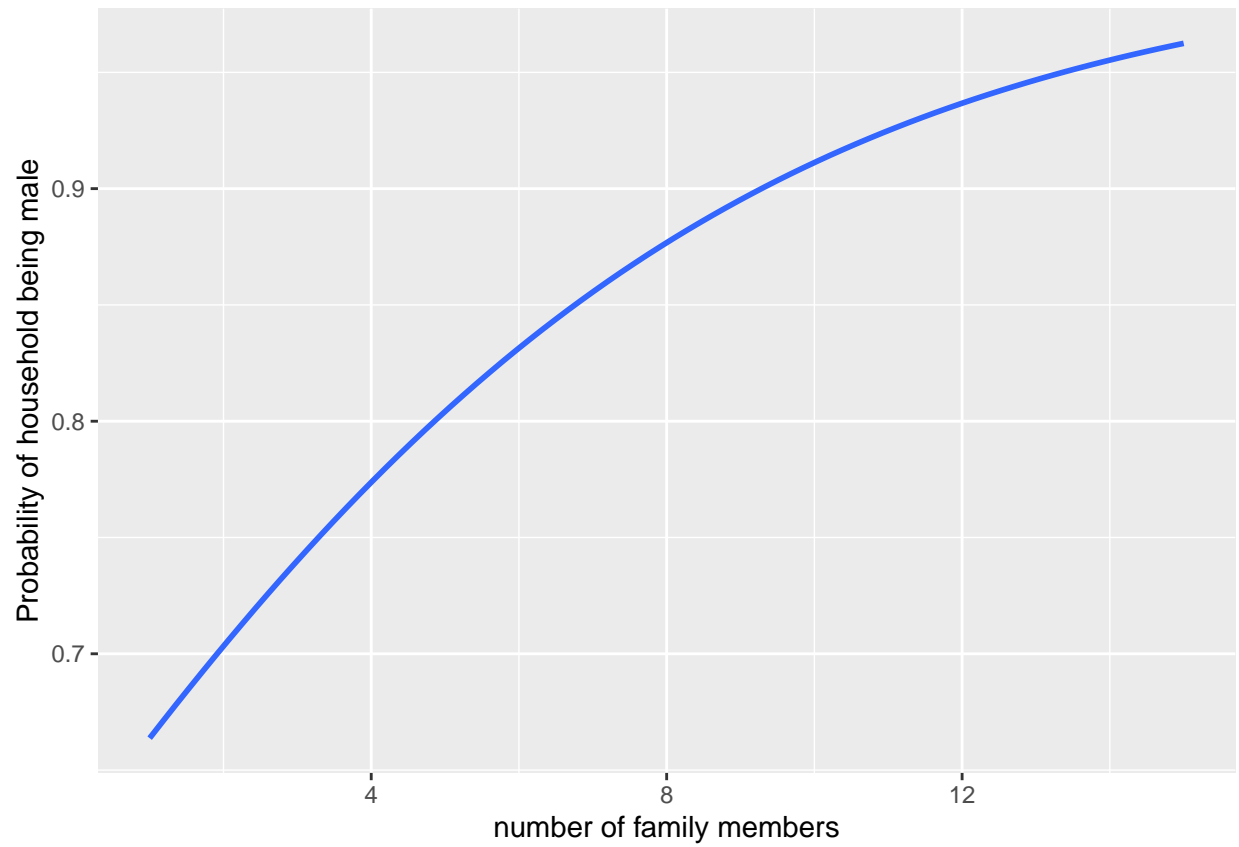
```
data.sex_number <- data.sex_number %>%
  mutate(odds.male = exp(logodds.male))
```

Probabilities

```
data.sex_number <- data.sex_number %>%
  mutate(probs.male = fitted(model_sex_number))
```

Plot the probability of being male

```
ggplot(data = data.sex_number, aes(x = Total.Number.of.Family.members, y = probs.male)) +
  geom_smooth(method="glm",
             method.args = list(family="binomial"),
             se = FALSE) +
  labs(x = "number of family members", y = "Probability of household being male")
```



```
plot_model(model_sex_number, type = "pred", title = "",  
           axis.title = c("number of family members", "Prob. of household being male"))
```

```
$Total.Number.of.Family.members
```

