## ANM\_Lipid\_Mediation\_Analysis

## Asger Wretlind

2024-05-29

```
#Load libraries
library(tidyverse)
## Warning: pakke 'ggplot2' blev bygget under R version 4.3.1
## Warning: pakke 'purrr' blev bygget under R version 4.3.1
## Warning: pakke 'dplyr' blev bygget under R version 4.3.1
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
             1.1.3
## v dplyr
                       v readr
                                    2.1.4
## v forcats 1.0.0
                                    1.5.0
                        v stringr
## v ggplot2 3.4.3 v tibble
                                    3.2.1
## v lubridate 1.9.2
                                    1.3.0
                      v tidyr
## v purrr
              1.0.2
## -- Conflicts ------ tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(here)
## here() starts at H:/Desktop/ANM_PRS/ANM_Data_Analysis
library(vroom)
## Warning: pakke 'vroom' blev bygget under R version 4.3.1
## Vedhæfter pakke: 'vroom'
## De følgende objekter er maskerede fra 'package:readr':
##
##
      as.col_spec, col_character, col_date, col_datetime, col_double,
##
      col_factor, col_guess, col_integer, col_logical, col_number,
      col_skip, col_time, cols, cols_condense, cols_only, date_names,
##
##
      date_names_lang, date_names_langs, default_locale, fwf_cols,
##
      fwf_empty, fwf_positions, fwf_widths, locale, output_column,
```

##

problems, spec

```
#Set color palette
color_palette <- c("#11A1B7", "#FF660C", "#0CA61E", "#FE3C1A",</pre>
                   "#9966CC", "#4DDF2C", "#FE5387", "#85DOAB",
                   "#18548A", "#FCBBOB", "#FD908F", "#DF56BD", "#F0E4AD")
#Setting seed, since bootstrapping include some level of randomness
set.seed(123)
#Use resdidual of adjusting for Site
use_residuals <- TRUE</pre>
#Load data
data <- vroom(here("data/ANM_Lipid_Preprocessed_v4.csv"))</pre>
## Rows: 841 Columns: 293
## -- Column specification ------
## Delimiter: "\t"
         (8): ID, Site, Date, Status, Sex, DOB, Accommodation, Marital_Status
## dbl (285): Visit, Order, Label, Age, Fulltime_Education_Years, apoe, e4_p, e...
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
#Load in linear regression data
reg_data <- vroom(here("data/sup_table_lipid_regression_v1.6.csv"))</pre>
## Rows: 8576 Columns: 7
## -- Column specification ------
## Delimiter: "\t"
## chr (3): Model, Lipid, Outcome
## dbl (4): Estimate, StdError, Pval, PvalFDR
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
#Function to clear special characters that may create confusion
Clean_names <- function(lipid_names){</pre>
   tmp_lipid_names <- gsub("\\(", "", lipid_names)</pre>
   tmp_lipid_names <- gsub("\\:", "", tmp_lipid_names)</pre>
   tmp_lipid_names <- gsub("\\)", "", tmp_lipid_names)</pre>
   tmp_lipid_names <- gsub("\\-", "", tmp_lipid_names)</pre>
   tmp_lipid_names <- gsub("\\/", "", tmp_lipid_names)</pre>
   return(tmp_lipid_names)
#Run this chunk if residuals are needed
if(use_residuals){
#temporary data set
tmp data <- data
```

```
colnames(tmp_data) <- Clean_names(colnames(tmp_data))</pre>
#Loop iterating over each lipid and overwrite with residuals
for (i in colnames(tmp_data)[(which(colnames(tmp_data) %in% "Cerd420")):length(tmp_data)]){
    tmp_formula <- as.formula(paste0(i, "~ Site"))</pre>
    tmp_model <- glm(tmp_formula, data = tmp_data)</pre>
    tmp_data[, i] <- residuals(tmp_model)</pre>
}
#Set names back to original
colnames(tmp_data) <- colnames(data)</pre>
#Use residual data
data <- tmp_data
}
rm(use_residuals)
library(mediation)
## Indlæser krævet pakke: MASS
## Warning: pakke 'MASS' blev bygget under R version 4.3.1
##
## Vedhæfter pakke: 'MASS'
## Det følgende objekt er maskeret fra 'package:dplyr':
##
##
       select
## Indlæser krævet pakke: Matrix
## Warning: pakke 'Matrix' blev bygget under R version 4.3.1
##
## Vedhæfter pakke: 'Matrix'
## De følgende objekter er maskerede fra 'package:tidyr':
##
##
       expand, pack, unpack
## Indlæser krævet pakke: mvtnorm
## Warning: pakke 'mvtnorm' blev bygget under R version 4.3.1
```

```
## Indlæser krævet pakke: sandwich
## mediation: Causal Mediation Analysis
## Version: 4.5.0
Mediation_extract <- function(independent, dependent, mediator, data_med){</pre>
    #Remove missing variables from data
    data_complete <- data_med %>%
        filter(!is.na(get(dependent))) %>%
        filter(!is.na(get(independent))) %>%
        filter(!is.na(get(mediator)))
    #Clean special characters
    colnames(data_complete) <- Clean_names(colnames(data_complete))</pre>
    #Save variables to global environment, as bootstrap loop will forget them otherwise
    assign("independent clean", Clean names(independent), envir = globalenv())
    assign("dependent_clean", Clean_names(dependent), envir = globalenv())
    assign("mediator_clean", Clean_names(mediator), envir = globalenv())
    \#Model\ 1 - Total\ effect\ (X -> Y)
    model_1 <- glm(</pre>
        formula = as.formula(paste0(dependent_clean, " ~ ", independent_clean)),
        data = data_complete)
    #Model 2 - Indirect effect (X -> M)
    model_2 <- glm(</pre>
        formula = as.formula(paste0(mediator_clean, " ~ ", independent_clean)),
        data = data_complete)
    #Model 3 - Indirect and direct effect (X + M -> Y)
    model_3 <- glm(</pre>
        formula = as.formula(paste0(dependent_clean, " ~ ", independent_clean, " + ", mediator_clean)),
        data = data complete)
    #Model 4 - (M -> Y)
    model_4 <- glm(</pre>
        formula = as.formula(paste0(dependent_clean, " ~ ", mediator_clean)),
        data = data_complete)
    #Mediation
    mediation_results <- mediate(model.m = model_2, model.y = model_3,</pre>
               treat = independent_clean,
               mediator = mediator_clean,
               boot = TRUE, sims = 500)
    #Output table
    table_out <- data.frame("tmp" = c())</pre>
    table_out[mediator, "independent"] <- independent</pre>
    table_out[mediator, "dependent"] <- dependent</pre>
    table_out[mediator, "mediator"] <- mediator</pre>
```

```
#Model 1
#Estimate
table_out[mediator, "model1_estimate"] <-</pre>
    summary(model 1)$coefficients[independent clean,
                                    which(grepl("Estimate",
                  colnames(summary(model_1)$coefficients)))]
#p-value
table_out[mediator, "model1_pvalue"] <-</pre>
    summary(model_1)$coefficients[independent_clean,
                                    which(grepl("Pr\\(>",
                  colnames(summary(model_1)$coefficients)))]
#Model 2
#Estimate
table_out[mediator, "model2_estimate"] <-</pre>
    summary(model_2)$coefficients[independent_clean,
                                    which(grepl("Estimate",
                                                                                          colnames (summa
#p-value
table_out[mediator, "model2_pvalue"] <-</pre>
    summary(model 2)$coefficients[independent clean,
                                    which(grepl("Pr\\(>",
                                                                                        colnames(summary
#Model 3
#Estimate
table_out[mediator, "model3_estimate"] <-</pre>
    summary(model_3)$coefficients[independent_clean,
                                    which(grepl("Estimate",
                                                                                           colnames(summ
#p-value
table_out[mediator, "model3_pvalue"] <-</pre>
    summary(model_3)$coefficients[independent_clean,
                                    which(grepl("Pr\\(>",
                                                                                           colnames(summ
#Model 4
#Estimate
table_out[mediator, "model4_estimate"] <-</pre>
    summary(model 4)$coefficients[mediator clean,
                                    which(grepl("Estimate",
                                                                                           colnames(summ
#p-value
table_out[mediator, "model4_pvalue"] <-</pre>
    summary(model_4)$coefficients[mediator_clean,
                                    which(grepl("Pr\\(>",
                                                                                          colnames (summa
#Mediate ADE Estimate
table_out[mediator, "ADE_estimate"] <- mediation_results$z0</pre>
#Mediate ADE p-value
table_out[mediator, "ADE_pvalue"] <- summary(mediation_results) %>%
    capture.output() %>%
    .[grepl("ADE", .)] %>%
```

```
.[length(.)] %>%
        str_split(" ") %>%
        unlist() %>%
        .[nzchar(.)] %>%
        .[!grepl("\\*", .)] %>%
        .[!nchar(.) == 1] %>%
        .[length(.)]
    #Mediate Total_effect Estimate
   table_out[mediator, "TotE_estimate"] <- mediation_results$tau.coef</pre>
    #Mediate Total_effect p-value
    table_out[mediator, "TotE_pvalue"] <- summary(mediation_results) %>%
        capture.output() %>%
        .[grepl("Total Effect", .)] %>%
        .[length(.)] %>%
        str_split(" ") %>%
        unlist() %>%
        .[nzchar(.)] %>%
        .[!grepl("\\*", .)] %>%
        .[!nchar(.) == 1] %>%
        .[length(.)]
    #Mediate ACME Estimate
   table out[mediator, "ACME estimate"] <- mediation results$d0
    #Mediate ACME p-value
    table_out[mediator, "ACME_pvalue"] <- summary(mediation_results) %>%
        capture.output() %>%
        .[grepl("ACME", .)] %>%
        .[length(.)] %>%
        str_split(" ") %>%
        unlist() %>%
        .[nzchar(.)] %>%
        .[!grepl("\\*", .)] %>%
        .[!nchar(.) == 1] %>%
        .[length(.)]
    #Mediate proportion Estimate
   table_out[mediator, "Prop_estimate"] <- mediation_results$n.avg</pre>
    #Mediate proportion p-value
   table_out[mediator, "Prop_pvalue"] <- mediation_results$n.avg.p</pre>
    #Clean global environment variables
   rm(independent_clean, dependent_clean, mediator_clean, envir = globalenv())
   return(table_out)
}
tmp_lipids <- tibble(reg_data) %>%
   filter(Outcome == "AD Female not Adjusted for APOE" |
              Outcome == "AD Male not Adjusted for APOE") %>%
```

```
filter(PvalFDR < 0.05) %>%
pull(Lipid)
```

```
#Vector of independent variables (lipids)
vec_independent <- tmp_lipids</pre>
#Vector of mediators (Confounders)
vec_mediators <- c("Total_C", "Total_TG", "HDL_C", "LDL_C", "ApoB")</pre>
#Note that variables with missing values or "factor names" will cause trouble
data_med <- data %>%
   mutate(AD_CTL = if_else(Status == "ADC", 1, NA)) %>%
   mutate(AD_CTL = if_else(Status == "CTL", 0, AD_CTL)) %>%
   relocate(AD_CTL, .after = Status) %>%
   filter(!is.na(AD_CTL)) %>%
   filter(Sex == "Female")
#Table to populate
mediate_summary <- data.frame("tmp" = c())</pre>
#Outer loop through independent variables
for (j in vec_independent){
    #Inner loop trough mediators
   for (i in vec_mediators){
        #Mediation
        mediate_summary <- mediate_summary %>%
            rbind(., Mediation_extract(independent = j,
                                        dependent = "AD_CTL",
                                        mediator = i,
                                        data_med = data_med))
   }
}
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
```

```
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
```

```
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
```

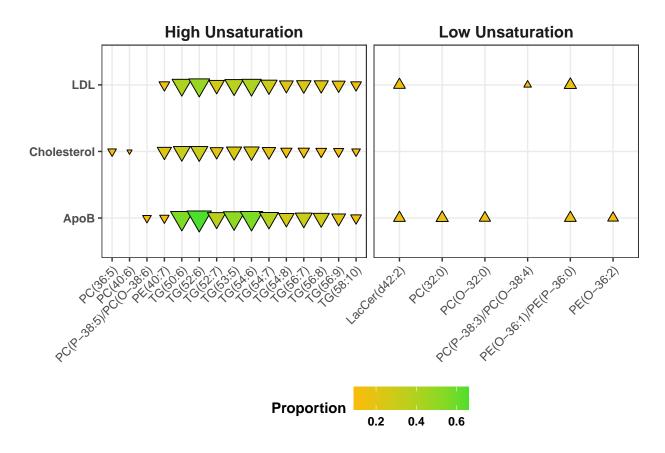
```
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
```

```
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
```

```
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
```

```
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
  Running nonparametric bootstrap
## Running nonparametric bootstrap
## Running nonparametric bootstrap
##
## Running nonparametric bootstrap
#Set rownames to numbers
rownames(mediate_summary) <- 1:nrow(mediate_summary)</pre>
rm(i, j, vec_mediators, vec_independent)
#Get regression direction (AD adjusted up or down regulation)
tmp_direction <- reg_data %>%
   filter(Outcome == "AD Female not Adjusted for APOE") %>%
   mutate(Direction = ifelse(Estimate > 0, "Positive", "Negative")) %>%
    dplyr::select(Lipid, Direction) %>%
   rename(independent = Lipid)
#Data wrangling for plot
data_tmp <- mediate_summary %>%
   filter(ACME_pvalue < 0.05) %>%
   filter(model1_pvalue < 0.05 &</pre>
           model2_pvalue < 0.05 &
           model3 pvalue < 0.05 &
           model4_pvalue < 0.05) %>%
```

```
#filter(independent != "e4_c") %>%
    mutate(Proportion = round(abs(Prop_estimate), 2)) %>%
   left_join(., tmp_direction, by = "independent") %>%
   mutate(independent = gsub("_A", "", independent)) %>%
   mutate(independent = gsub("_B", "", independent)) %>%
   mutate(mediator = gsub("_C", "", mediator)) %>%
   mutate(mediator = gsub("Total", "Cholesterol", mediator)) %>%
   mutate(Saturation = if_else(as.numeric(str_extract(independent, "(?<=:)[0-9]+(?=\\))")) >= 5, "High
#Correct specific lipid
data_tmp$Saturation[data_tmp$independent == "PC(P-36:4)/PC(0-36:5)"] <- "High Unsaturation"
#Create the plot
Mediation_plot <- ggplot(data_tmp, aes(x = independent,</pre>
                                       y = mediator,
                   bg = Proportion,
                   size = Proportion,
                   shape = Direction))+
    geom_point()+
   scale_shape_manual(values = c(25, 24), guide = "none")+
    scale_size_continuous(guide = "none") +
    scale_color_gradient(high = "#4DDF2C", low = "#FCBB0B",
                         aesthetics = "bg",
                         breaks = c(0.2, 0.4, 0.6))+
   facet_wrap(~ Saturation, scales = "free_x", nrow = 1)+
   theme_bw()+
    theme(
        axis.text.x = element_text(angle = 45, hjust = 1),
        axis.title = element_blank(),
       legend.position = "bottom",
        strip.background = element_blank(),
        strip.text = element_text(face = "bold", size = 12),
        axis.text.y = element_text(face = "bold"),
        legend.text = element_text(face = "bold"),
        legend.title = element_text(face = "bold"))
# #Save figure
# pdf(here("figures/Mediation_plot_v1.4.pdf"), width = 8, height = 3)
Mediation_plot
```



```
# dev.off()

# #Save plot object
# saveRDS(Mediation_plot, here("data/Mediation_plot_v1.4.rds"))

# #Save supplementary table
# vroom_write(mediate_summary, here("data/sup_table_mediation_v1.0.csv"))

rm(data_med, data_tmp, tmp_data, tmp_direction, tmp_model, tmp_formula, tmp_lipids)

library(ggdag)
```

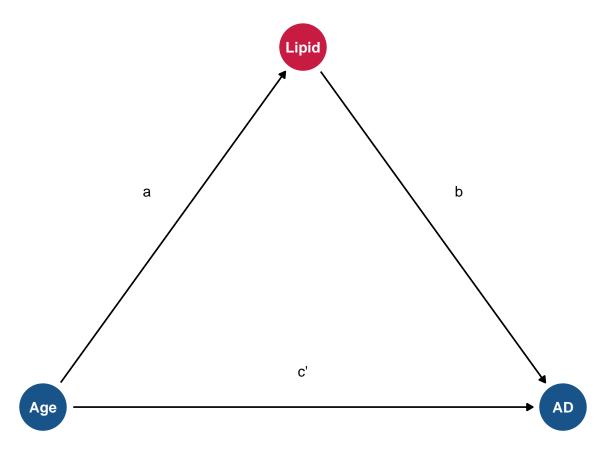
```
## Warning: pakke 'ggdag' blev bygget under R version 4.3.3
##
## Vedhæfter pakke: 'ggdag'
## Det følgende objekt er maskeret fra 'package:stats':
##
## filter
```

```
#Work in progress
# #Select mediation analysis to plot
# tmp_mediate <- mediate_summary[mediate_summary$mediator == "PC(0-32:0)" &
                                  mediate summary$independent == "Sex",]
#
#
# tmp_mediate$mediator <- Clean_names(tmp_mediate$mediator)</pre>
# #Coordinates
\# x\_coord \leftarrow c(1, 2, 3)
\# names(x_coord) <- c(tmp_mediate\#independent, tmp_mediate\#mediator, tmp_mediate\#dependent)
# y_coord <- c(0, 1, 0)
# names(y_coord) <- c(tmp_mediate$independent, tmp_mediate$mediator, tmp_mediate$dependent)
\# coord\_dag \leftarrow list(x = x\_coord, y = y\_coord)
# #Create graph
# dag <- dagify(formula(paste(tmp_mediate$dependent, " ~ ", tmp_mediate$independent)),
                formula(paste(tmp_mediate$mediator, " ~ ", tmp_mediate$independent)),
#
                formula(paste(tmp_mediate$dependent, " ~ ", tmp_mediate$mediator)),
#
                coords = coord_dag) %>%
#
      tidy_dagitty() %>%
      mutate(colour = ifelse(name == "lipid", "farve1", "farve2"))
#
#### #Working
\#Coordinates
coord_dag <- list(</pre>
 x = c(Age = 1, Lipid = 2, AD = 3),
 y = c(Age = 0, Lipid = 1, AD = 0))
#Create graph
dag <- dagify(AD ~ Age,
              Lipid ~ Age,
              AD ~ Lipid,
              coords = coord_dag) %>%
    tidy_dagitty() %>%
    mutate(colour = ifelse(name == "Lipid", "farve1", "farve2"))
#Visualize
dag_plot <- dag %>%
    ggplot(aes(x = x, y = y, xend = xend, yend = yend)) +
    geom_dag_point(aes(colour = colour))+
    scale_color_manual(values = c("#C71B42", "#18548A"))+
    geom_dag_edges() +
    geom_dag_text() +
    annotate("text", x = c(1.4, 2.0, 2.6),
                     y = c(0.6, 0.1, 0.6),
             label = c("a", "c'", "b"))+
```

```
theme_void()+
    theme(legend.position = "none")

# #Save figure
# pdf(here("figures/Directed_acylic_graph_v1.1.pdf"), width = 5, height = 3)

dag_plot
```



```
# dev.off()

# #Save plot object
# saveRDS(dag_plot, here("data/dag_plot_v1.0.rds"))

#Clean lipid names of all lipids

tmp_lipid_names <- colnames(data[which(colnames(data) == "Cer(d42:0)"):length(data)])

tmp_lipid_names_clean <- Clean_names(tmp_lipid_names)

#Create a new data subset and clean the lipid names
data_AD <- data
colnames(data_AD) <- Clean_names(colnames(data_AD))

#Subset only AD and control samples
data_AD <- data_AD %>%
    mutate(AD_CTL = if_else(Status == "ADC", 1, NA)) %>%
    mutate(AD_CTL = if_else(Status == "CTL", 0, AD_CTL)) %>%
```

```
relocate(AD_CTL, .after = Status) %>%
    filter(!is.na(AD_CTL)) %>%
    mutate(Sex_bin = if_else(Sex == "Female", 1, NA)) %>%
    mutate(Sex_bin = if_else(Sex == "Male", 0, Sex_bin)) %>%
    relocate(Sex_bin, .after = Sex) %>%
    data.frame()
#TG5811
#PC0320
#Step 1 - Total effect
#Sex ---> AD
Model_1 <- glm(AD_CTL ~ Sex_bin, data = data_AD, family = "binomial")</pre>
summary(Model_1)
##
## Call:
## glm(formula = AD_CTL ~ Sex_bin, family = "binomial", data = data_AD)
## Coefficients:
##
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.19845
                           0.12164 -1.631
                                              0.103
                           0.15750 0.091
                                              0.928
## Sex_bin
               0.01431
##
## (Dispersion parameter for binomial family taken to be 1)
       Null deviance: 931.07 on 675 degrees of freedom
## Residual deviance: 931.06 on 674 degrees of freedom
## AIC: 935.06
##
## Number of Fisher Scoring iterations: 3
#Step 2 - Indirect effect
#Sex ---> Mediator (Lipid)
Model_2 <- glm(PCO340 ~ Sex_bin, data = data_AD)</pre>
summary(Model_2)
##
## Call:
## glm(formula = PCO340 ~ Sex_bin, data = data_AD)
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.028281
                           0.008209 -3.445 0.000607 ***
                           0.010633
                                    5.193 2.74e-07 ***
## Sex_bin
               0.055215
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for gaussian family taken to be 0.0183989)
##
       Null deviance: 12.897 on 675 degrees of freedom
## Residual deviance: 12.401 on 674 degrees of freedom
## AIC: -778.53
```

```
##
## Number of Fisher Scoring iterations: 2
#Step 3 - Indirect and direct effect
     -->Mediator-->
#Sex ----> AD
Model_3 <- glm(AD_CTL ~ Sex_bin + TG5811, data = data_AD, family = "binomial")</pre>
summary(Model_3)
##
## Call:
## glm(formula = AD_CTL ~ Sex_bin + TG5811, family = "binomial",
##
       data = data_AD)
##
## Coefficients:
##
              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.25385
                           0.12406 -2.046
                                             0.0407 *
                           0.16042
                                   0.488
## Sex_bin
               0.07833
                                             0.6253
## TG5811
               -0.84223
                           0.19444 -4.332 1.48e-05 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 931.07 on 675 degrees of freedom
## Residual deviance: 911.23 on 673 degrees of freedom
## AIC: 917.23
##
## Number of Fisher Scoring iterations: 4
mediation_results <- mediate(model.m = Model_2, model.y = Model_3,</pre>
               treat = "Sex_bin", mediator = "TG5811",
               boot = TRUE, sims = 500)
## Running nonparametric bootstrap
summary(mediation_results)
##
## Causal Mediation Analysis
## Nonparametric Bootstrap Confidence Intervals with the Percentile Method
##
##
                            Estimate 95% CI Lower 95% CI Upper p-value
## ACME (control)
                            -0.01142
                                         -0.01860
                                                         -0.01 <2e-16 ***
## ACME (treated)
                            -0.01151
                                         -0.01902
                                                         -0.01 <2e-16 ***
## ADE (control)
                             0.01936
                                         -0.05511
                                                          0.10
                                                                  0.59
## ADE (treated)
                             0.01927
                                         -0.05467
                                                          0.10
                                                                  0.59
## Total Effect
                             0.00785
                                         -0.06790
                                                          0.09
                                                                  0.78
## Prop. Mediated (control) -1.45481
                                         -5.79272
                                                          3.42
                                                                  0.78
## Prop. Mediated (treated) -1.46634
                                                          3.43
                                                                  0.78
                                         -5.84196
```

-0.01884

-0.01146

-0.01 <2e-16 \*\*\*

## ACME (average)