

Lira_combined_Statistical

Asger_Wretlind

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```
library(here)
library(tidyverse)
library(vroom)
library(tableone)

#Color Palette inspired by Kandinsky - Circles in a circle
color_palette <- c("yellow" = "#F2D479",
                  "pink" = "#BF4E83",
                  "green" = "#36BFA6",
                  "red" = "#C50B42",
                  "blue" = "#1262B2",
                  "sand" = "#F9E2C0",
                  "white" = "#E7E1D1")

#Import LirAlbu preprocessed data
data_albu <- vroom(here("data/0097_liralbu_4cer_data_preprocessed.csv"))

#Convert factors
data_albu <- data_albu %>%
  mutate(across(which(
    apply(., 2, function(x) {
      tmp <- unique(x)
      length(tmp) - sum(is.na(tmp)) == 2L})),
    ~ factor(.))) %>%
  mutate_at(c("PrePost", "Time_point"), ~ factor(.))

#Import LiraFlame preprocessed data
data_flame <- vroom(here("data/0083_liraflame_4cer_data_preprocessed.csv"))

#Convert factors
data_flame <- data_flame %>%
  mutate(across(which(
    apply(., 2, function(x) {
      tmp <- unique(x)
      length(tmp) - sum(is.na(tmp)) == 2L})),
    ~ factor(.)))

#Vector of clinical characteristics to for CTable_albu
Ctable_vars_albu <- c("Age", "Sex", "Weight", "Diabetes_Duration", "HbA1C_mmol_mol",
                    "HbA1c_percent", "Total_Cholesterol", "LDL", "Triglyceride",
                    "OfficeSBP", "OfficeDBP",
                    "Plasma_Creatinine", "eGFR_CKDEPI", "Log10MeanUAER")
```

```

#Clinical characteristics table albu at baseline
CCtable_albu <- data_albu %>%
  filter(Time_point == 1) %>%
  mutate(Sex = factor(Sex, levels = c(1, 0), labels = c("Man", "Woman"))) %>%
  rename(HbA1C_mmol_mol = HbA1c) %>%
  mutate(HbA1c_percent = (HbA1C_mmol_mol/10.929)+2.15) %>%
  CreateTableOne(data = ., vars = CCtable_vars_albu)

#Clinical characteristics table albu before each treatment
CCtable_albu_stratified <- data_albu %>%
  filter(Visit == 1) %>%
  mutate(Sex = factor(Sex, levels = c(1, 0), labels = c("Man", "Woman"))) %>%
  rename(HbA1C_mmol_mol = HbA1c) %>%
  mutate(HbA1c_percent = (HbA1C_mmol_mol/10.929)+2.15) %>%
  CreateTableOne(data = ., vars = CCtable_vars_albu, strata = "Treatment")

# #Save CCtable_albu
# write.csv(print(CCtable_albu, printToggle = FALSE), here("data/CCtable_albu.csv"))
# write.csv(print(CCtable_albu_stratified, printToggle = FALSE), here("data/CCtable_albu_stratified.csv"))

#UAER median and IQR
#Start of trial total
quantile(exp(data_albu$Log10MeanUAER)[data_albu$Visit == 1], na.rm = TRUE)

```

```

##          0%          25%          50%          75%          100%
##  3.506440  6.858487  8.467922 12.782060 32.186905

```

```

#Before Liraglutide
quantile(exp(data_albu$Log10MeanUAER[data_albu$Visit == 1 & data_albu$Treatment == "Liraglutide"]),
          na.rm = TRUE)

```

```

##          0%          25%          50%          75%          100%
##  3.506440  6.718327  8.749661 14.453193 30.569415

```

```

#Before Placebo
quantile(exp(data_albu$Log10MeanUAER[data_albu$Visit == 1 & data_albu$Treatment == "Placebo"]),
          na.rm = TRUE)

```

```

##          0%          25%          50%          75%          100%
##  4.660147  6.901838  8.115836 12.176272 32.186905

```

```

##Now for LiraFlame

#Vector of clinical characteristics to for CCtable_flame
CCtable_vars_flame <- c("age", "Sex", "V2_B_VAEGT", "dm_var", "HbA1C_mmol_mol",
                        "HbA1c_percent", "CHOL_v2", "LDL_v2", "TRIG_v2",
                        "Sys_gen_v2_beregnet", "Dia_gen_v2_beregnet",
                        "V2_B_CREA", "GFRapi_V2", "V2_U_LogMeanUAER")

#Clinical characteristics table flame, stratified treatment type
CCtable_flame <- data_flame %>%

```

```

filter(Time_point == 1) %>%
mutate(Sex = factor(Sex, levels = c(1, 2), labels = c("Man", "Woman"))) %>%
rename(HbA1C_mmol_mol = V2_B_HBA1C) %>%
mutate(HbA1c_percent = (HbA1C_mmol_mol/10.929)+2.15) %>%
mutate(Treatment = factor(Treatment,
                           levels = c(1, 2),
                           labels = c("Liraglutide", "Placebo"))) %>%
  CreateTableOne(data = ., vars = CTable_vars_flame, strata = "Treatment", addOverall = TRUE)

# #Save CTable_flame
# write.csv(print(CTable_flame, printToggle = FALSE), here("data/CTable_flame.csv"))

#UAER median and IQR
#Total
quantile(exp(data_flame$V2_U_LogMeanUAER[data_flame$Time_point == 1]), na.rm = TRUE)

##          0%          25%          50%          75%          100%
## 1.351250  1.723002  2.328206  3.460542  26.144427

#Liraglutide
quantile(exp(data_flame$V2_U_LogMeanUAER[data_flame$Time_point == 1 & data_flame$Treatment == 1]), na.rm = TRUE)

##          0%          25%          50%          75%          100%
## 1.351250  1.849833  2.533037  3.693273  26.144427

#Placebo
quantile(exp(data_flame$V2_U_LogMeanUAER[data_flame$Time_point == 1 & data_flame$Treatment == 2]), na.rm = TRUE)

##          0%          25%          50%          75%          100%
## 1.351250  1.611429  1.921704  3.194300  7.547296

rm(CTable_vars_albu, CTable_vars_flame)
rm(CTable_albu, CTable_albu_stratified, CTable_flame)

#Table of ceramide means at different stages and treatment
CerTable_means_albu <- data_albu %>%
  select(PrePost, Sample_ID, starts_with("Cer")) %>%
  pivot_wider(names_from = c(PrePost, Sample_ID), values_from = starts_with("Cer")) %>%
  pivot_longer(starts_with("Cer")) %>%
  separate(name, sep = "_", c("Ceramide", "PrePost", "Sample_ID")) %>%
  pivot_wider(names_from = c(Ceramide, PrePost), values_from = value) %>%
  summarise(across(starts_with("Cer"), list(Mean = ~ mean(., na.rm = TRUE),
                                           SD = ~ sd(., na.rm = TRUE)))) %>%

  pivot_longer(everything()) %>%
  separate(name, sep = "_", c("Ceramide", "PrePost", "Measure")) %>%
  pivot_wider(names_from = c(PrePost, Measure)) %>%
  relocate(starts_with("PlaceboPost"), .after = PlaceboPre_SD)

#t-test change between post and pre.
CerTable_ttestChange_albu <- data_albu %>%

```

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select(PrePost, Sample_ID, starts_with("Cer")) %>%
pivot_wider(names_from = c(PrePost, Sample_ID), values_from = starts_with("Cer")) %>%
pivot_longer(starts_with("Cer")) %>%
separate(name, sep = "_", c("Ceramide", "PrePost", "Sample_ID")) %>%
pivot_wider(names_from = c(Ceramide, PrePost), values_from = value) %>%
#filter(!if_any(everything(), ~ is.na(.x))) %>%
summarise(across(
  ends_with("Post"),
  list(meandif = ~ t.test(x = .,
    y = get(glue::glue(str_replace(
      cur_column(), "Post", "Pre"))),
    paired = TRUE)$estimate[[1]],
    CIlower = ~ t.test(x = .,
    y = get(glue::glue(str_replace(
      cur_column(), "Post", "Pre"))),
    paired = TRUE)$conf.int[[1]],
    CIupper = ~ t.test(x = .,
    y = get(glue::glue(str_replace(
      cur_column(), "Post", "Pre"))),
    paired = TRUE)$conf.int[[2]],
    pval = ~ t.test(x = .,
    y = get(glue::glue(str_replace(
      cur_column(), "Post", "Pre"))),
    paired = TRUE)$p.value))) %>%
rename_with(~str_replace(., "Post", ""), everything()) %>%
pivot_longer(everything()) %>%
separate(name, sep = "_", c("Ceramide", "Treatment", "ttest_output")) %>%
pivot_wider(names_from = c(Treatment, ttest_output), values_from = value) %>%
rename_with(~ paste0("Change_", .), !contains("Ceramide"))

# t.test(data_albu$`Cer24:1`[data_albu$PrePost == "PlaceboPost"],
#       data_albu$`Cer24:1`[data_albu$PrePost == "PlaceboPre"], paired = TRUE)

#Paired sample t-test between post placebo and post liraglutide from full data
CerTable_ttestEndpoints_albu <- data_albu %>%
select(PrePost, Sample_ID, starts_with("Cer")) %>%
pivot_wider(names_from = c(PrePost, Sample_ID), values_from = starts_with("Cer")) %>%
pivot_longer(starts_with("Cer")) %>%
separate(name, sep = "_", c("Ceramide", "PrePost", "Sample_ID")) %>%
pivot_wider(names_from = c(Ceramide, PrePost), values_from = value) %>%
summarise(across(
  contains("LiraPost"),
  list(meandif = ~ t.test(x = .,
    y = get(glue::glue(
      str_replace(cur_column(), "Lira", "Placebo"))),
    paired = TRUE)$estimate[[1]],
    CIlower = ~ t.test(x = .,
    y = get(glue::glue(
      str_replace(cur_column(), "Lira", "Placebo"))),
    paired = TRUE)$conf.int[[1]],
    CIupper = ~ t.test(x = .,
    y = get(glue::glue(

```

```

        str_replace(cur_column(), "Lira", "Placebo"))),
        paired = TRUE)$conf.int[[2]],
    pval = ~ t.test(x = .,
        y = get(glue::glue(
            str_replace(cur_column(), "Lira", "Placebo"))),
            paired = TRUE)$p.value))) %>%
rename_with(~str_replace(., "_LiraPost", ""), everything()) %>%
pivot_longer(everything()) %>%
separate(name, sep = "_", c("Ceramide", "ttest_output")) %>%
pivot_wider(names_from = ttest_output, values_from = value) %>%
rename_with(~ paste0("EndVsEnd_", .), !contains("Ceramide"))

#Combine CerTable_albu (Similar style to table 2 in von Scholten (2017))
CerTable_albu <- CerTable_means_albu %>%
    left_join(., CerTable_ttestChange_albu) %>%
    left_join(., CerTable_ttestEndpoints_albu) %>%
    relocate(starts_with("Change_Lira_"), .after = LiraPost_SD)

CerTable_albu

## # A tibble: 6 x 21
##   Ceramide LiraPre_Mean LiraP~1 LiraP~2 LiraP~3 Change~4 Chang~5 Chang~6 Chang~7
##   <chr>      <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 Cer16      0.748    0.162    0.732    0.149   -0.00975 -0.0558  0.0364  0.665
## 2 Cer18      0.546    0.164    0.534    0.153   -0.0121  -0.0680  0.0437  0.657
## 3 Cer20      0.644    0.230    0.620    0.227   -0.0215  -0.106   0.0630  0.604
## 4 Cer22      2.62     1.09     2.48     1.18    -0.194   -0.607   0.219   0.341
## 5 Cer24:0     8.77     4.82     7.98     4.33    -1.15    -3.10    0.794   0.233
## 6 Cer24:1     4.16     1.36     4.37     1.83     0.0762  -0.488   0.640   0.782
## # ... with 12 more variables: PlaceboPre_Mean <dbl>, PlaceboPre_SD <dbl>,
## #   PlaceboPost_Mean <dbl>, PlaceboPost_SD <dbl>, Change_Placebo_meandif <dbl>,
## #   Change_Placebo_CIlower <dbl>, Change_Placebo_CIupper <dbl>,
## #   Change_Placebo_pval <dbl>, EndVsEnd_meandif <dbl>, EndVsEnd_CIlower <dbl>,
## #   EndVsEnd_CIupper <dbl>, EndVsEnd_pval <dbl>, and abbreviated variable names
## #   1: LiraPre_SD, 2: LiraPost_Mean, 3: LiraPost_SD, 4: Change_Lira_meandif,
## #   5: Change_Lira_CIlower, 6: Change_Lira_CIupper, 7: Change_Lira_pval

# #Save CerTable_albu
# write.csv(CerTable_albu, here("data/CerTable_albu.csv"))

# #Combine all CerTable_albu
# CerTable_albu_full <- CerTable_albu %>%
#   left_join(CerTable_quartiles_albu, .) %>%
#   left_join(., CerTable_PercentChange_albu) %>%
#   relocate(Percent_change_Lira, .after = Change_Lira_pval) %>%
#   relocate(Percent_change_Placebo, .after = Change_Placebo_pval) %>%
#   left_join(., CerTable_EndVsEndDifference_albu) %>%
#   left_join(., CerTable_EffectSize_albu) %>%
#   pivot_longer(!starts_with("Cer")) %>%
#   pivot_wider(names_from = Ceramide, values_from = value)
#
# CerTable_albu_full

```

```

# #Save CerTable_albu_full
# write.csv(CerTable_albu_full, here("data/CerTable_albu_full.csv"))

rm(CerTable_means_albu, CerTable_ttestChange_albu, CerTable_ttestEndpoints_albu,
   CerTable_quartiles_albu, CerTable_PercentChange_albu, CerTable_EffectSize_albu,
   CerTable_EndVsEndDifference_albu)
rm(CerTable_albu, CerTable_albu_full)

library(lme4)
library(lmerTest)

#Function for extracting Coefficient (estimate), Standard error and p-value from LMM
LMM_extract <- function(Data, Formula){

  #Extract response variable
  response_var <- word(as.character(c(Formula)), 1)

  #Fit Linear mixed model
  tmp_LMM <- lmer(formula = as.formula(Formula), data = Data)

  #Summary object
  tmp_LMM_sum <- summary(tmp_LMM)

  #Extract explanatory variable(s)
  explanatory_var <- rownames(tmp_LMM_sum$coefficients)[nrow(tmp_LMM_sum$coefficients)]

  #Output
  out_df <- tibble(Resp_var = response_var,
                   Expl_var = explanatory_var,
                   LMM_Coeff = tmp_LMM_sum$coefficients[explanatory_var, "Estimate"],
                   Std_error = tmp_LMM_sum$coefficients[explanatory_var, "Std. Error"],
                   pval = tmp_LMM_sum$coefficients[explanatory_var, "Pr(>|t|)"])

  return(out_df)
}

# #Example
# data_albu %>%
#   lmer(formula = Cer16 ~ Treatment+Visit + Treatment:Visit + (1|Sample_ID)) %>%
#   summary()
#
# LMM_extract(Data = data_albu,
#             Formula = Cer16 ~ Treatment+Visit + Treatment:Visit + (1|Sample_ID) )

#Change the factor order
data_albu <- data_albu %>%
  mutate(Treatment = factor(Treatment, levels = c("Placebo", "Liraglutide")))

#Linear mixed models for each lipid in LirAlbu
LMM_overview_albu <- data_albu %>%
  summarise(across(starts_with("Cer"),
                   ~ LMM_extract(Data = data_albu,

```

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        Formula = . ~ Treatment+Visit + Treatment:Visit + (1|Sample_ID))) %>%
pivot_longer(cols = everything()) %>%
unnest(cols = everything()) %>%
mutate("Project" = "LirAlbu")

#Change the factor order back
data_albu <- data_albu %>%
  mutate(Treatment = factor(Treatment, levels = c("Liraglutide", "Placebo")))

#Change the factor order
data_flame <- data_flame %>%
  mutate(Treatment = factor(Treatment, levels = c(2, 1)))

#Linear mixed models for each lipid in LiraFlame
LMM_overview_flame <- data_flame %>%
  rename(Visit = Time_point) %>%
  summarise(across(starts_with("Cer"),
    ~ LMM_extract(Data = data_flame,
      Formula = . ~ Treatment+Visit + Treatment:Visit + (1|Sample_ID))) %>%
pivot_longer(cols = everything()) %>%
unnest(cols = everything()) %>%
mutate("Project" = "LiraFlame")

#Change the factor order
data_flame <- data_flame %>%
  mutate(Treatment = factor(Treatment, levels = c(1, 2)))

#Combine tables
LMM_overview <- rbind(LMM_overview_albu, LMM_overview_flame) %>%
  select(-c(Resp_var, Expl_var)) %>%
  pivot_wider(names_from = Project,
    values_from = c(LMM_Coeff, Std_error, pval)) %>%
  rename(Ceramide = name) %>%
  relocate(contains("LirAlbu"), .after = Ceramide)

# #Save LMM_overview as CSV
# write.csv(LMM_overview, here("data/LMM_overview.csv"))

rm(LMM_overview_albu, LMM_overview_flame, LMM_extract)
rm(LMM_overview)

library(lme4)
library(ggeffects)

#Dirty fix on for error in names with ":"
colnames(data_albu) <- gsub("Cer24:", "Cer24_", colnames(data_albu))

#Ceramides to plot
tmp_cer <- data_albu %>%
  select(starts_with("Cer")) %>%
  colnames()

##LirAlbu

```

```

#Empty list to populate in loop
LMM_albu_list <- list()

#Loop through each ceramide and save LMM model, prediction and plot into LMM_albu_list
for(i in 1:length(tmp_cer)){

  #LMM model
  model <- data_albu %>%
    #mutate(Visit = recode(Visit, `1` = 0, `2` = 12)) %>%
    lmer(formula = paste0("`", tmp_cer[i], "`",
      " ~ Treatment+Visit + Treatment:Visit + (1|Sample_ID)"))

  #LMM prediction
  prediction <- ggpredict(model, terms = c("Visit", "Treatment"))
  prediction$x <- c(0, 0, 12, 12)

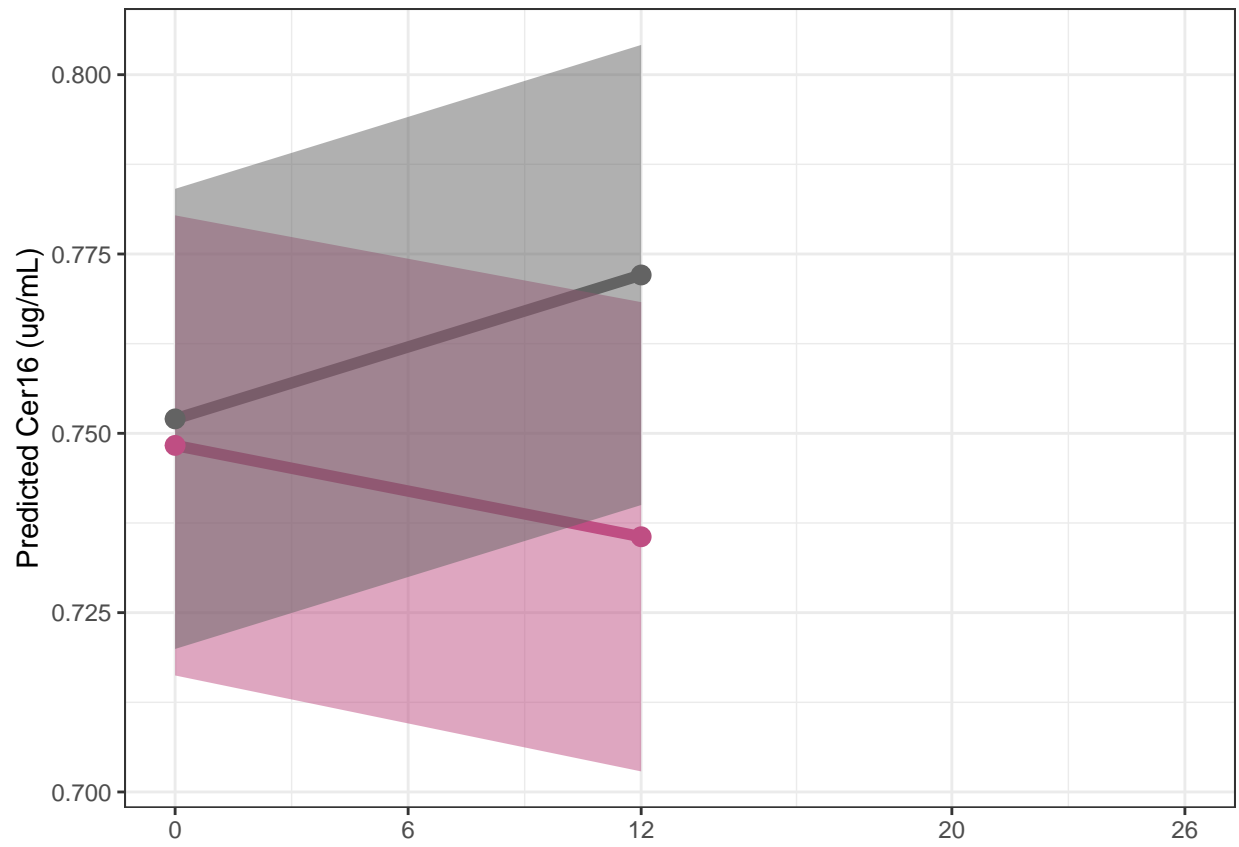
  #LMM plot
  plot <- ggplot(prediction)+
    geom_line(aes(x = x, y = predicted, color = group), lwd = 2, alpha = 1) +
    geom_ribbon(aes(x = x, ymin = predicted - std.error,
      ymax = predicted + std.error, fill = group),
      alpha = 0.5) +
    geom_point(aes(x = x, y = predicted, color = group), size = 3)+

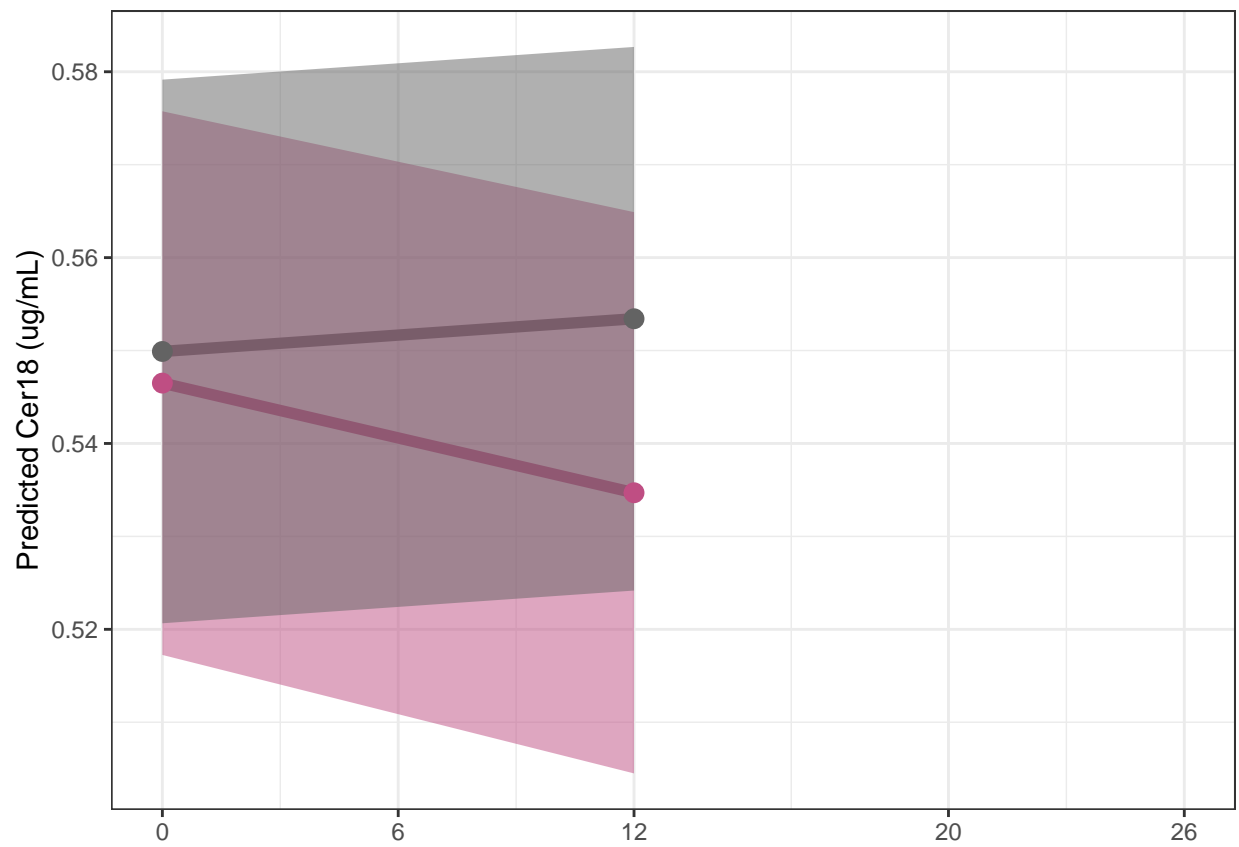
    scale_colour_manual(values = c("#BF4E83", "#636363"))+
    scale_fill_manual(values = c("#BF4E83", "#636363"))+
    scale_x_continuous(breaks = c(0, 6, 12, 20, 26), limits = c(0, 26))+
    #ylim(-2.65, -2.55)+
    ylab(paste0("Predicted ", tmp_cer[i], " (ug/mL)"))+
    xlab("Weeks") +
    #ggtitle("Trial 1") +
    theme_bw()+
    theme(legend.position= "none", #c(0.15,0.1),
      legend.background = element_rect(fill = NA),
      axis.title.x = element_blank(),
      rect = element_rect(fill = "transparent"),
      panel.background = element_rect(fill = "transparent"))

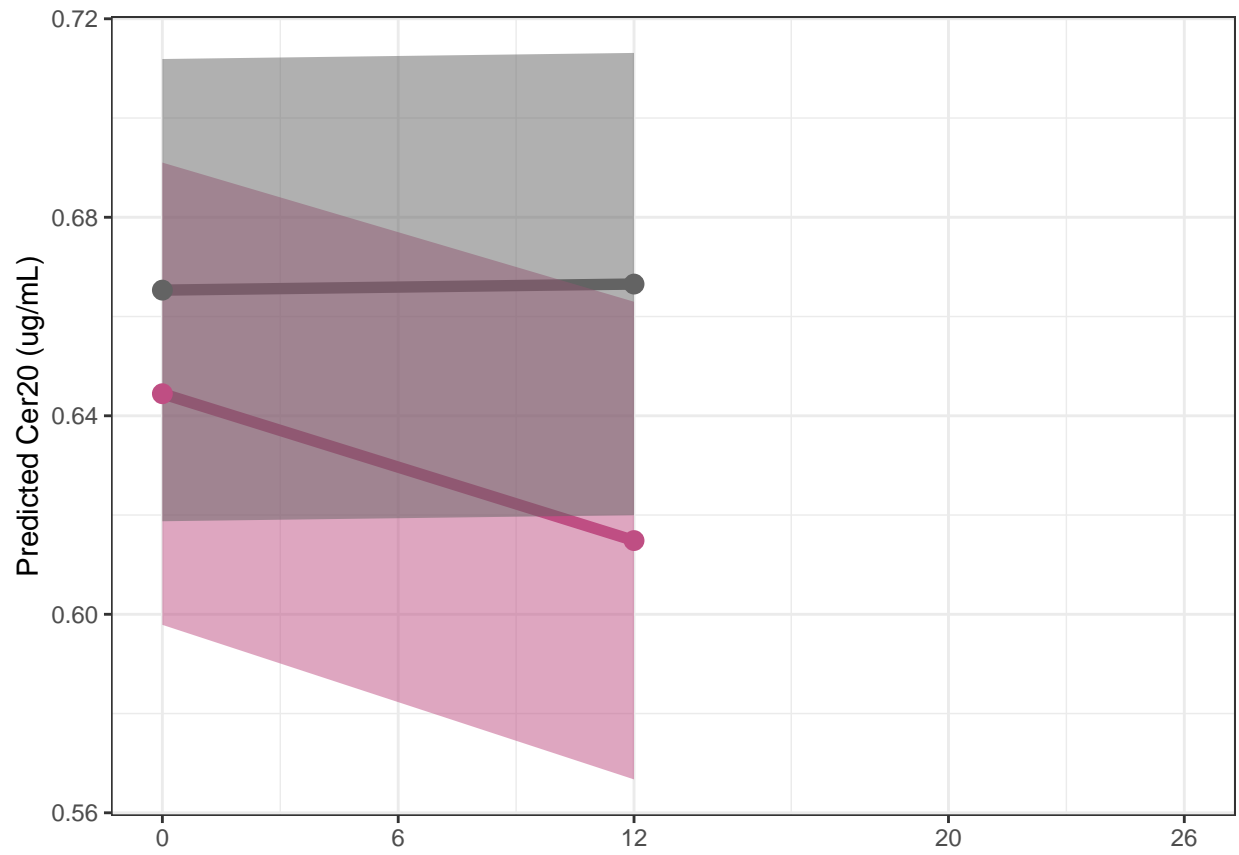
  #Populate LMM_albu_list with more lists
  LMM_albu_list[tmp_cer[i]] <- list(c(Model = list(model),
    Prediction = list(prediction),
    Plot = list(print(plot))))

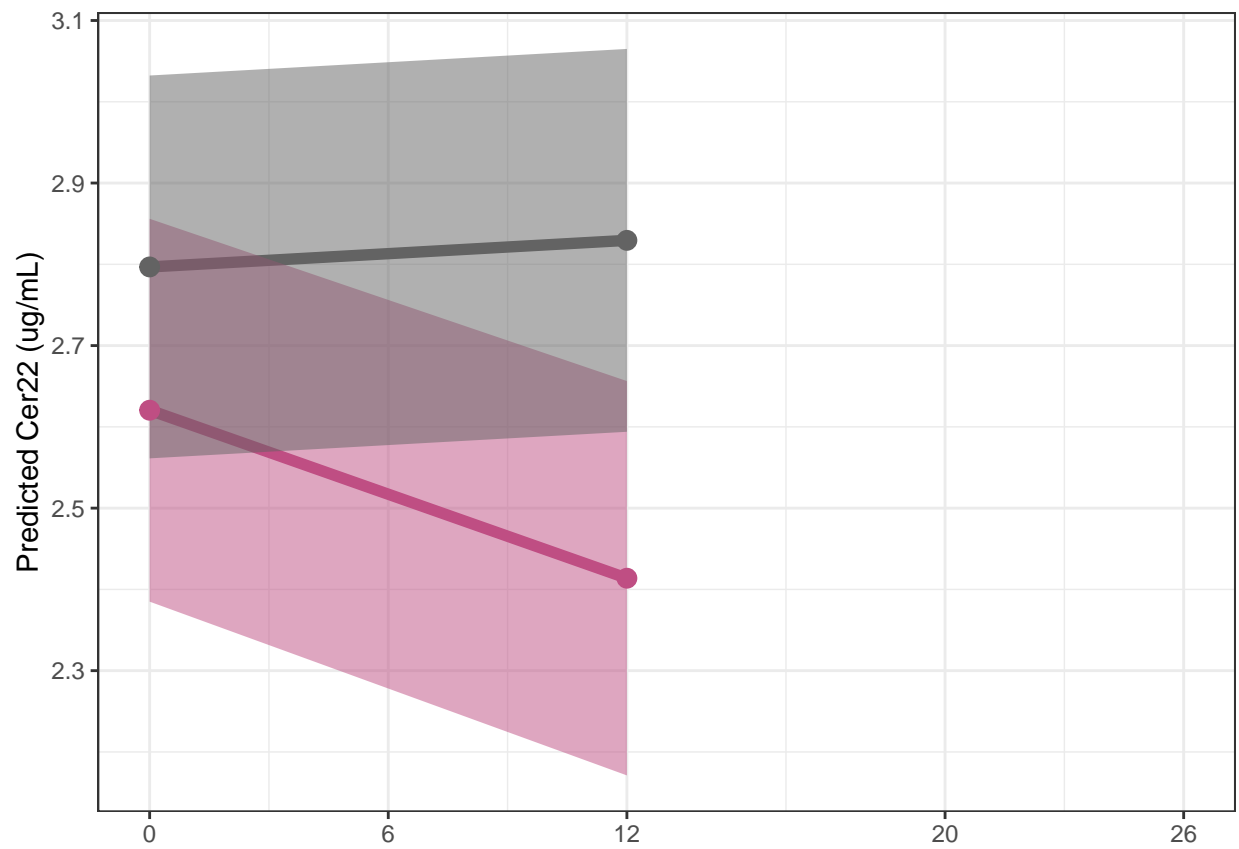
  rm(model, prediction, plot)
}

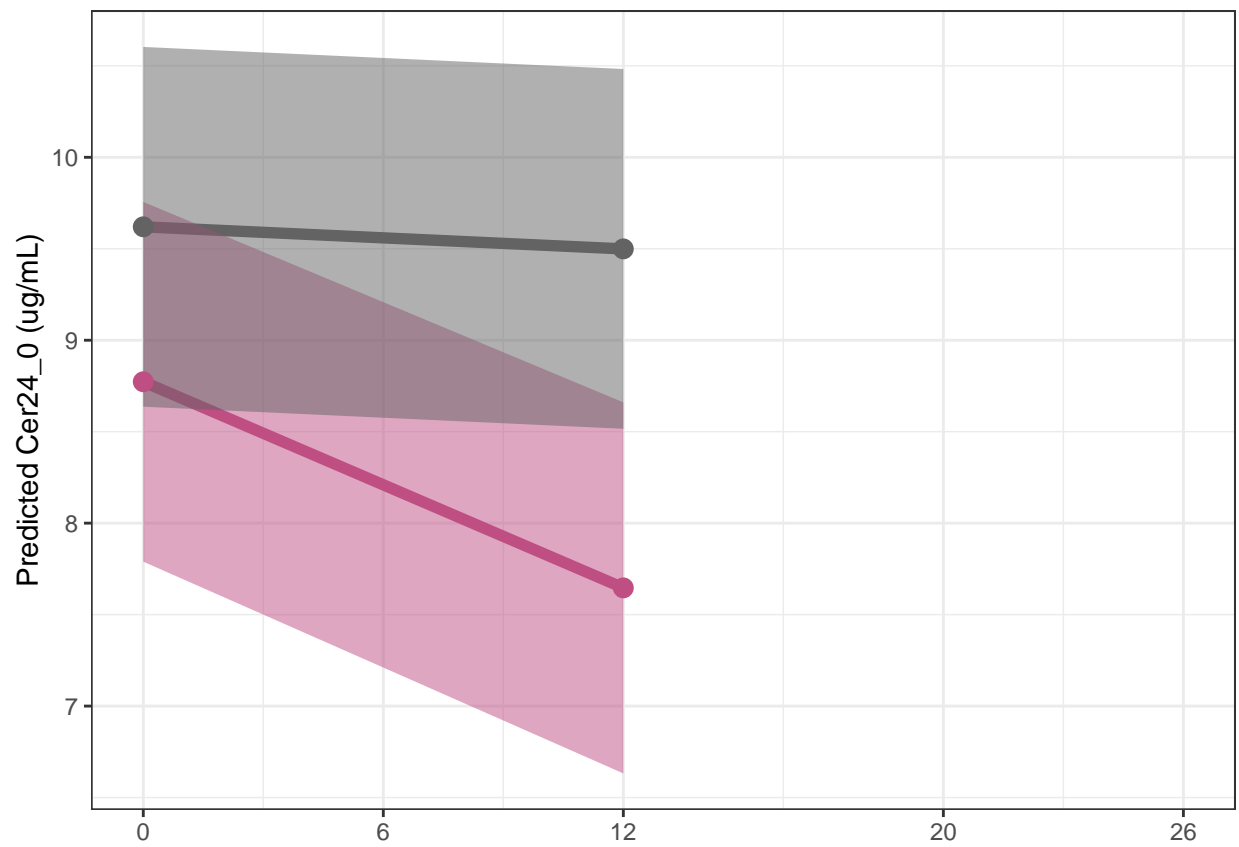
```

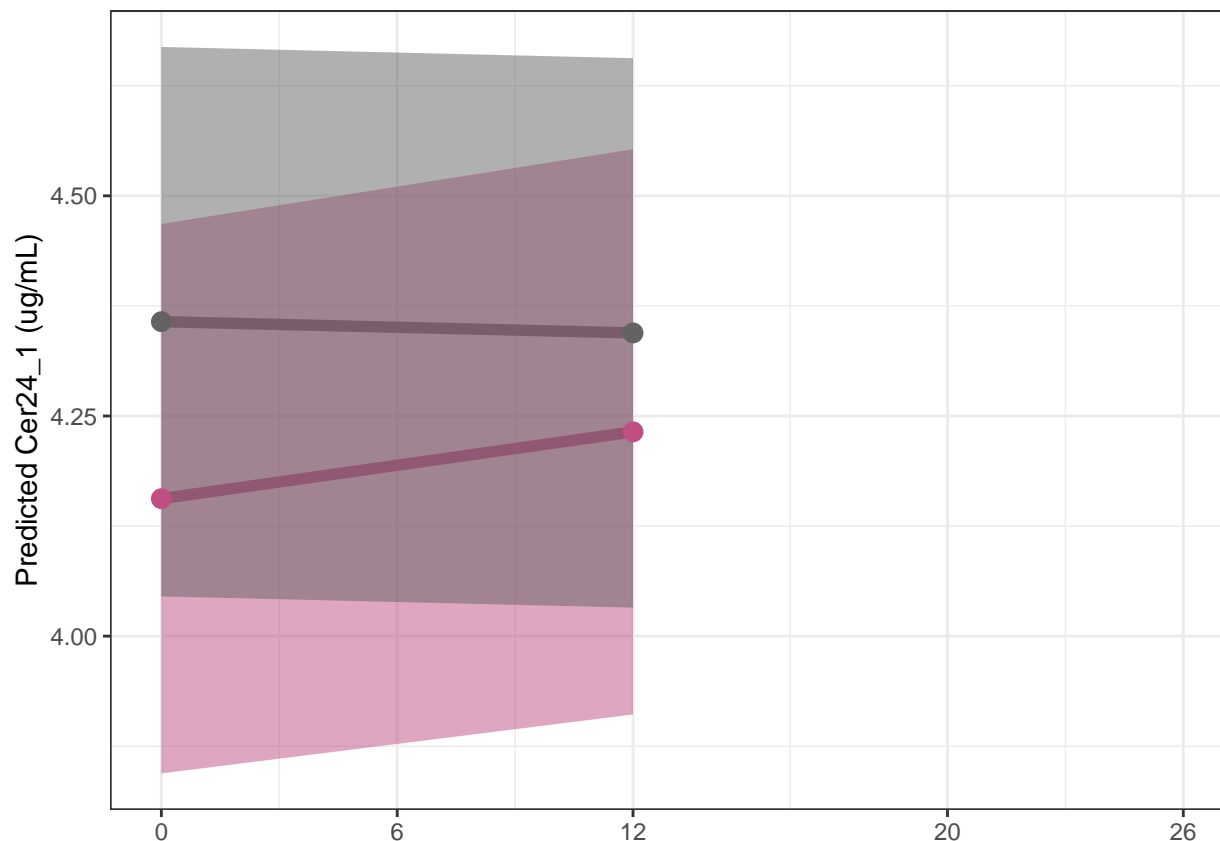













```
#change names back to contain ":"
colnames(data_albu) <- gsub("Cer24_", "Cer24:", colnames(data_albu))

##LiraFlame

#Dirty fix on for error in names with ":"
colnames(data_flame) <- gsub("Cer24:", "Cer24_", colnames(data_flame))

#Empty list to populate in loop
LMM_flame_list <- list()

#Loop through each ceramide and save LMM model, prediction and plot into LMM_flame_list
for(i in 1:length(tmp_cer)){

  #LMM model
  model <- data_flame %>%
    mutate(Treatment = recode(Treatment, `1` = "Liraglutide", `2` = "Placebo")) %>%
    rename(Visit = Time_point) %>%
    #mutate(Visit = recode(Visit, `1` = 0, `2` = 26)) %>%
    lmer(formula = paste0("`", tmp_cer[i], "`",
      " ~ Treatment+Visit + Treatment:Visit + (1|Sample_ID)"))

  #LMM prediction
  prediction <- ggpredict(model, terms = c("Visit", "Treatment"))
  prediction$x <- c(0, 0, 26, 26)
```

```

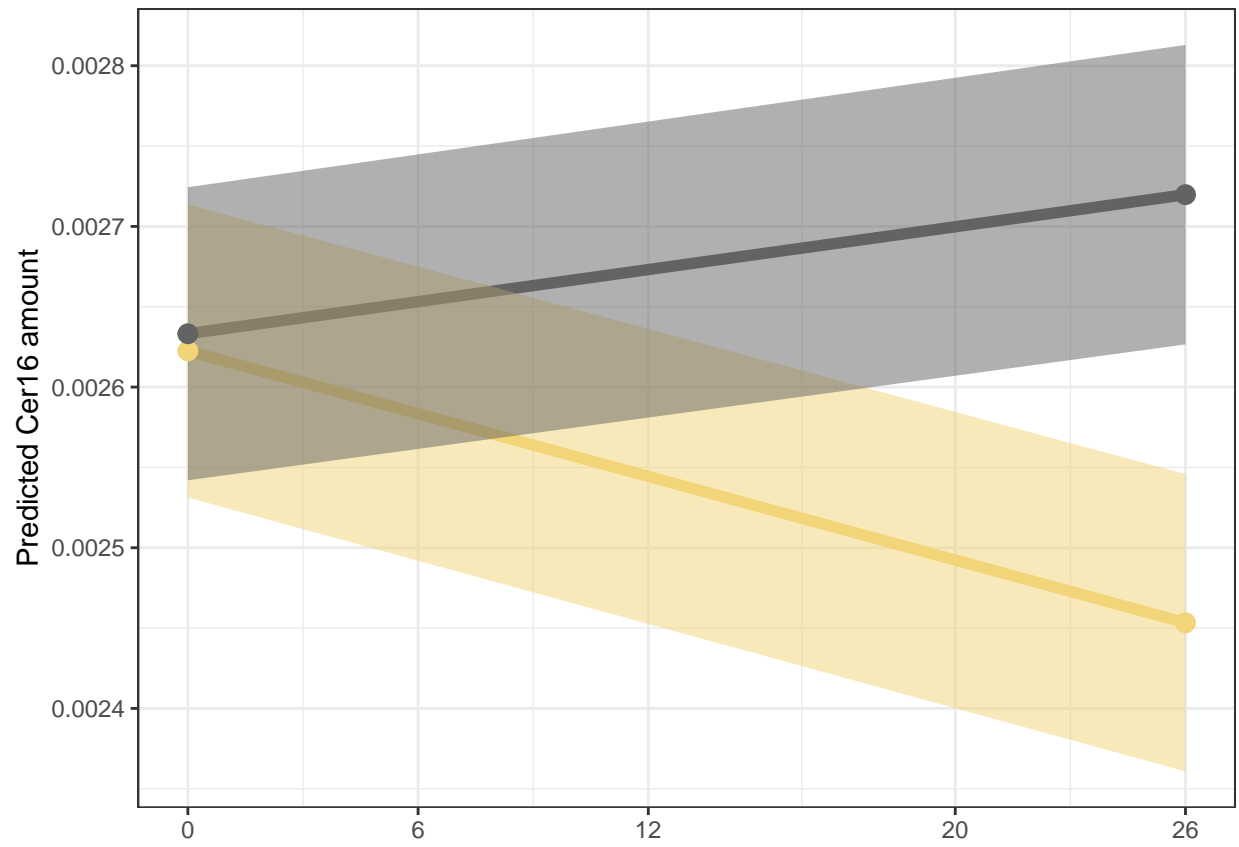
#LMM plot
plot <- ggplot(prediction)+
  geom_line(aes(x = x, y = predicted, color = group), lwd = 2, alpha = 1) +
  geom_ribbon(aes(x = x, ymin = predicted - std.error,
                ymax = predicted + std.error, fill = group),
            alpha = 0.5) +
  geom_point(aes(x = x, y = predicted, color = group), size = 3)+

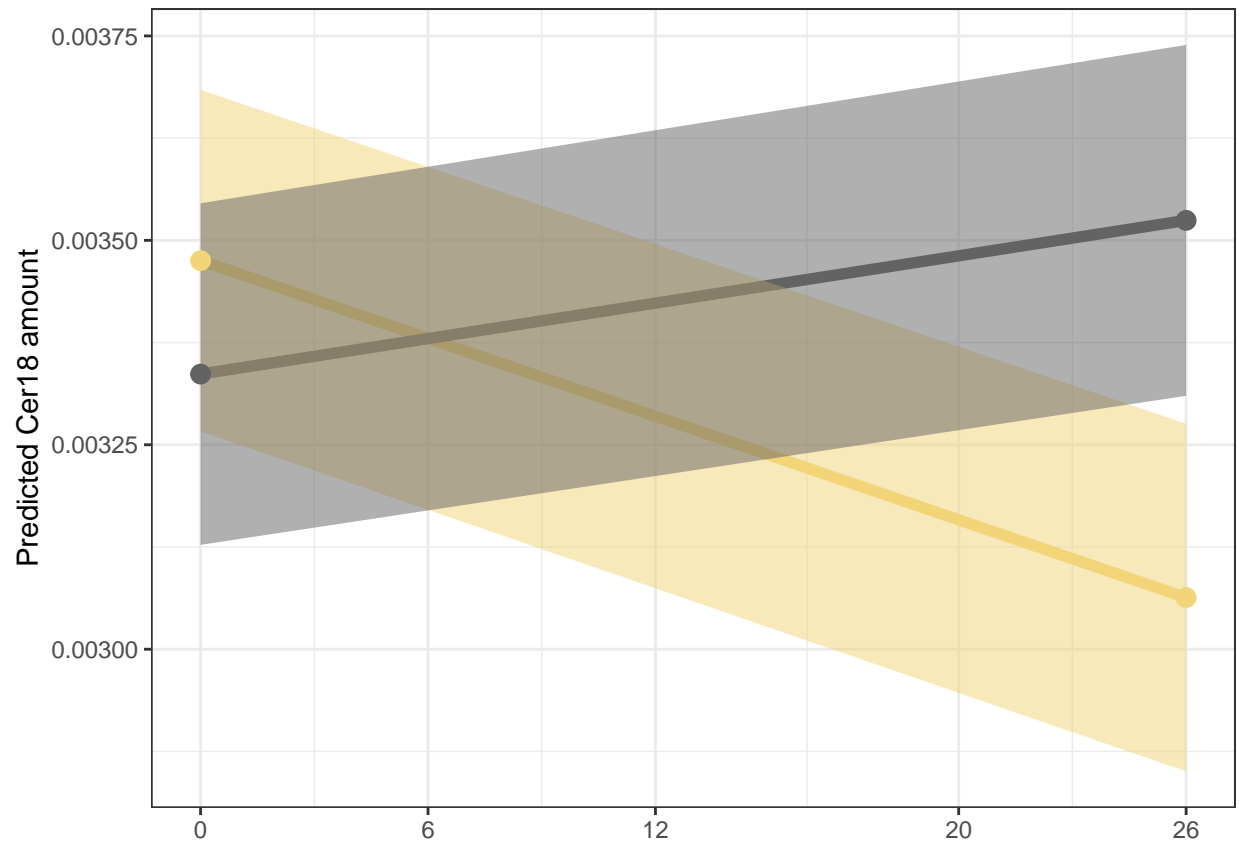
  scale_colour_manual(values = c("#F2D479", "#636363"))+
  scale_fill_manual(values = c("#F2D479", "#636363"))+
  scale_x_continuous(breaks = c(0, 6, 12, 20, 26), limits = c(0, 26))+
  #ylim(-2.65, -2.55)+
  ylab(paste0("Predicted ", tmp_cer[i], " amount"))+
  xlab("Weeks") +
  #ggtitle("Trial 1") +
  theme_bw()+
  theme(legend.position= "none",#c(0.15,0.1),
        legend.background = element_rect(fill = NA),
        axis.title.x = element_blank(),
        #axis.title.y = element_blank(),
        rect = element_rect(fill = "transparent"),
        panel.background = element_rect(fill = "transparent"))

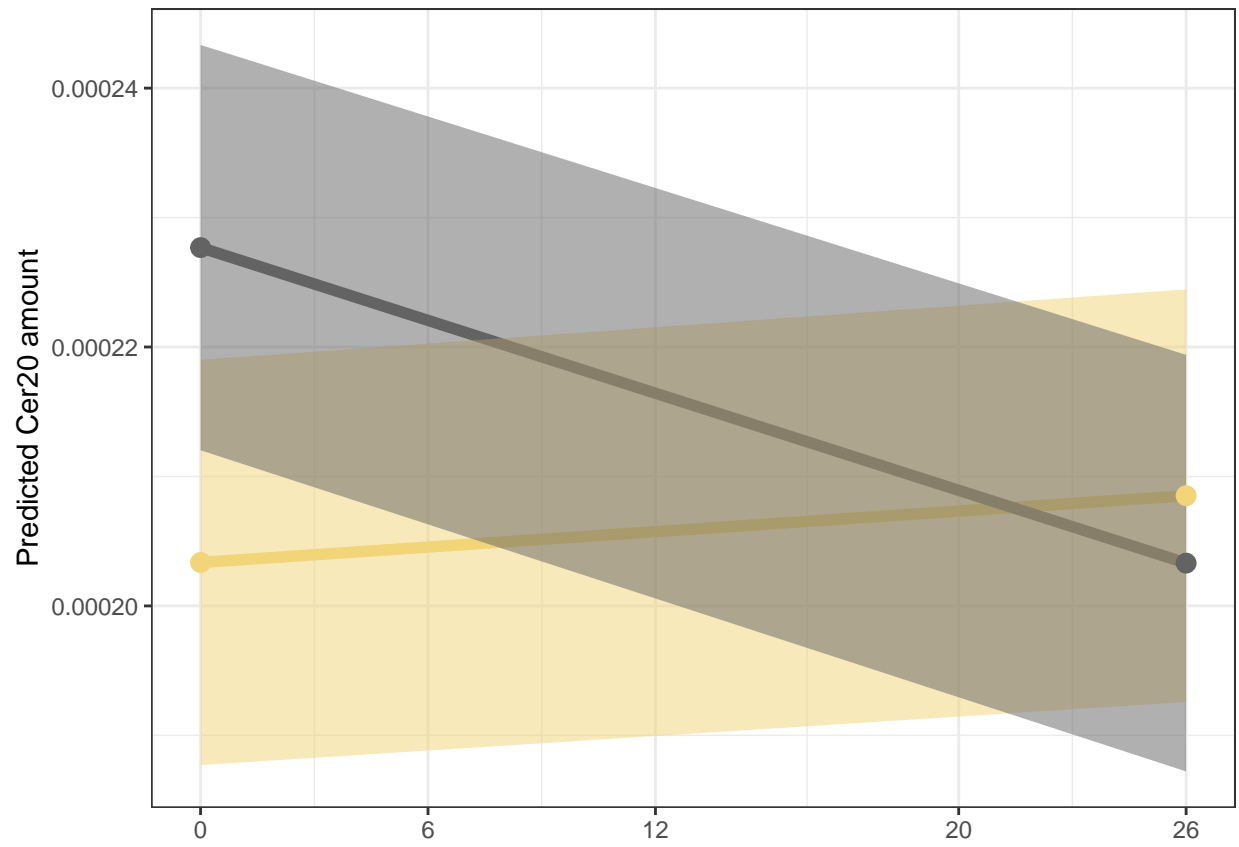
#Populate LMM_flame_list with more lists
LMM_flame_list[tmp_cer[i]] <- list(c(Model = list(model),
                                     Prediction = list(prediction),
                                     Plot = list(print(plot))))

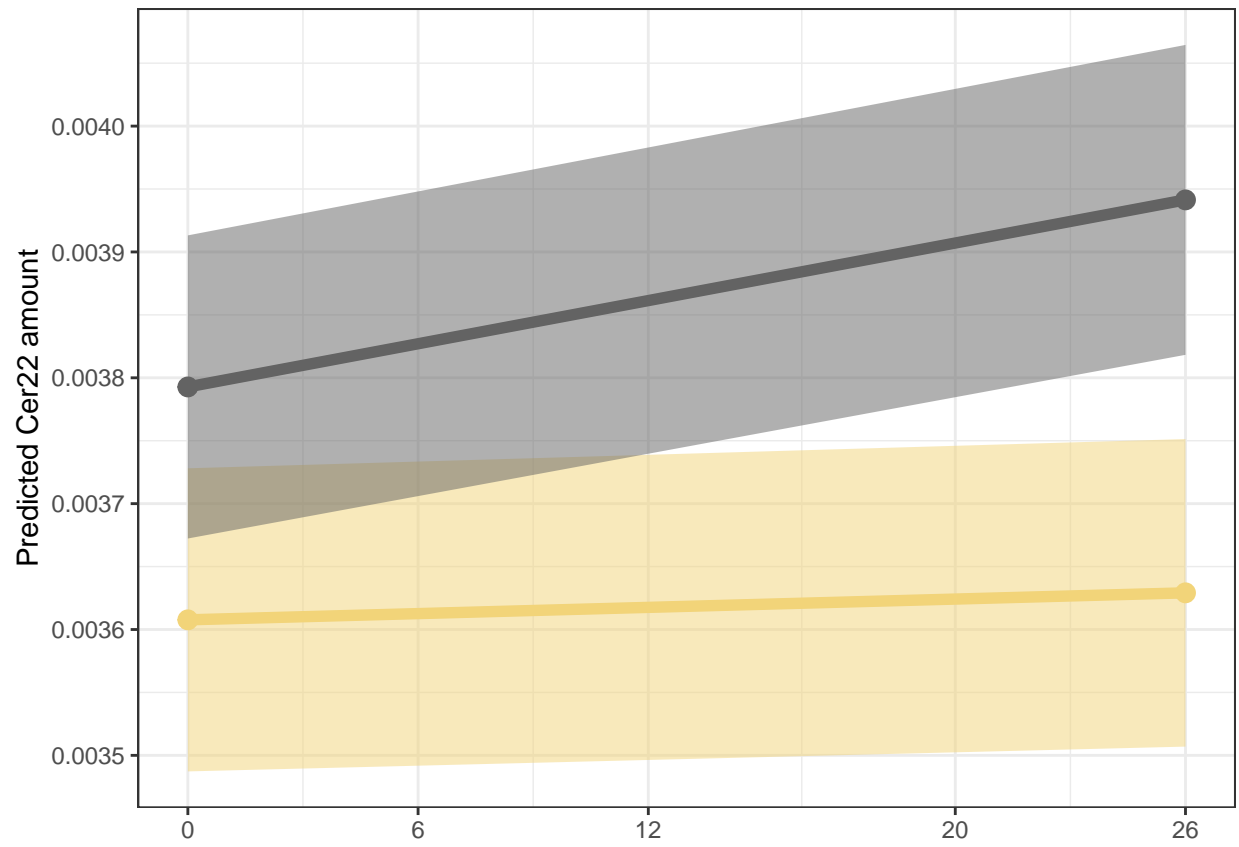
rm(model, prediction, plot)
}

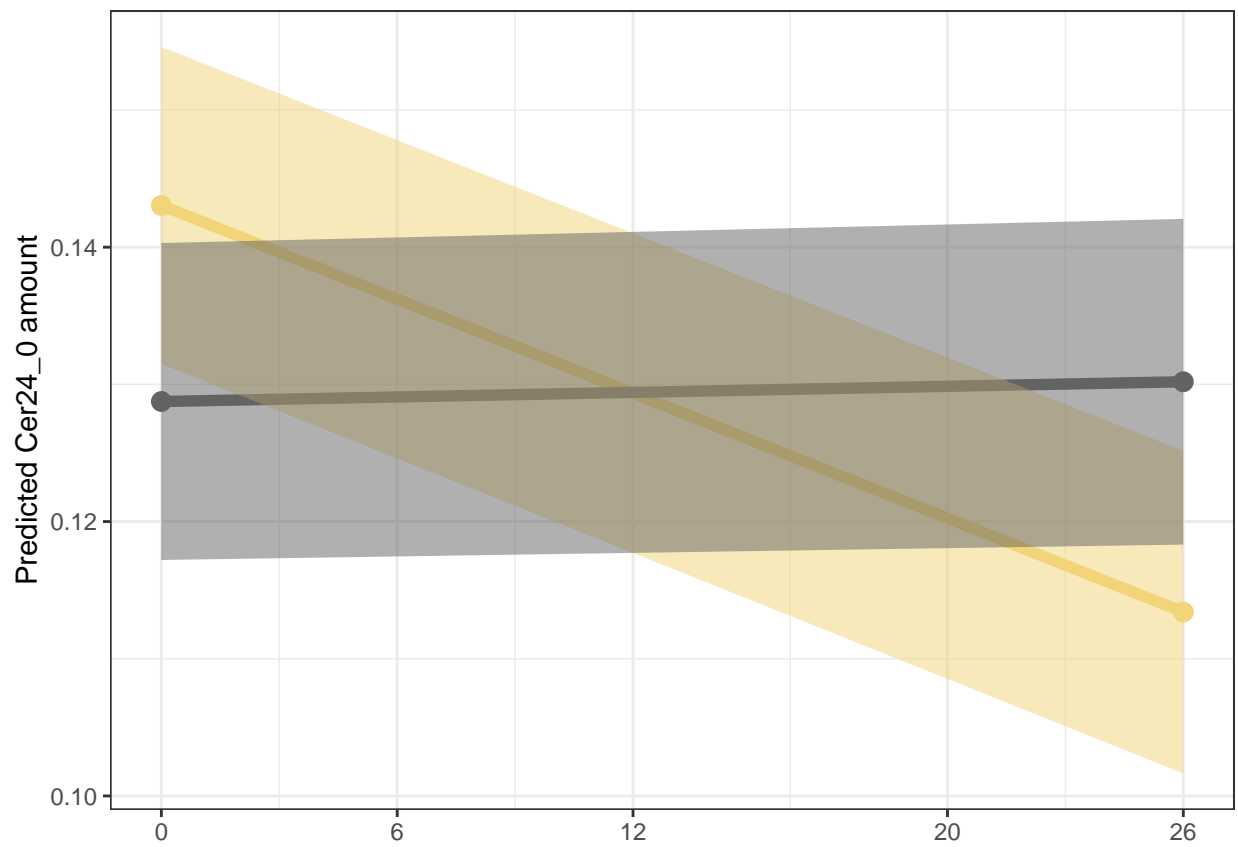
```

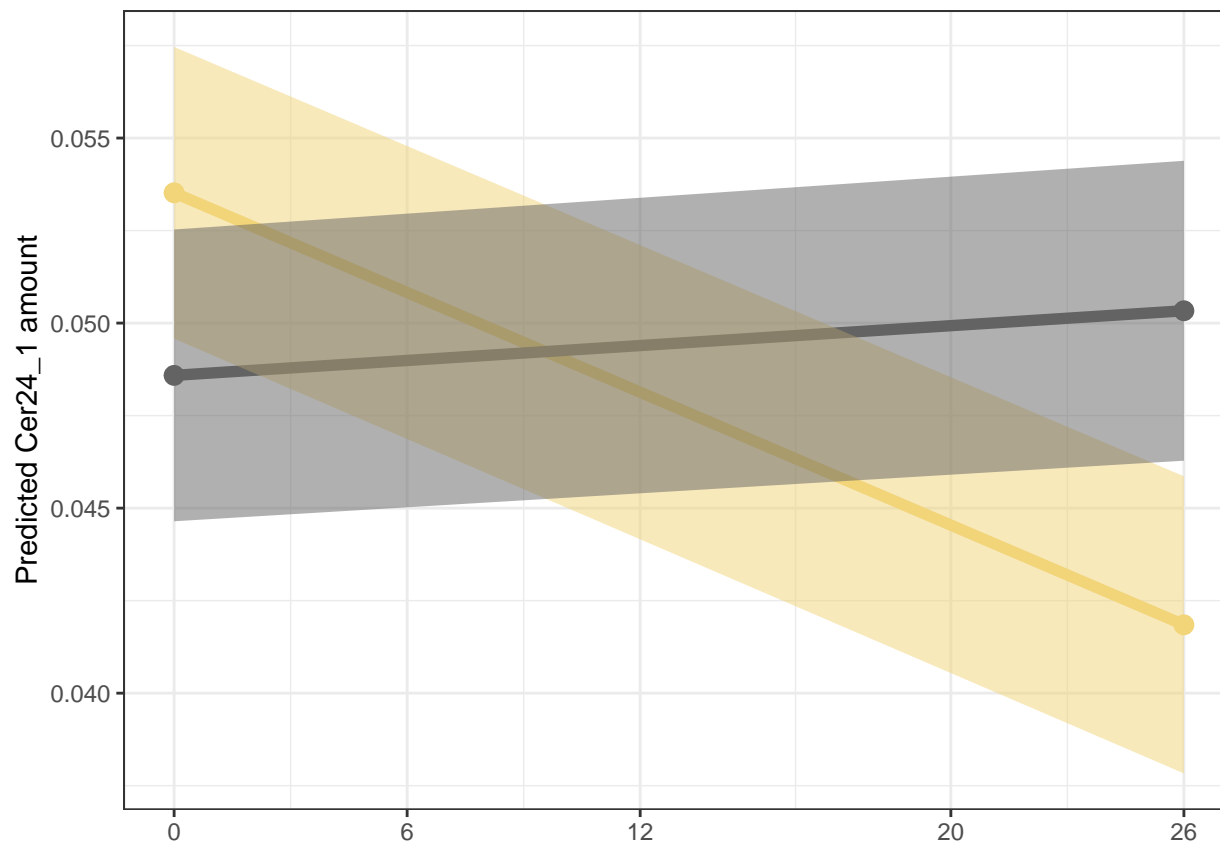








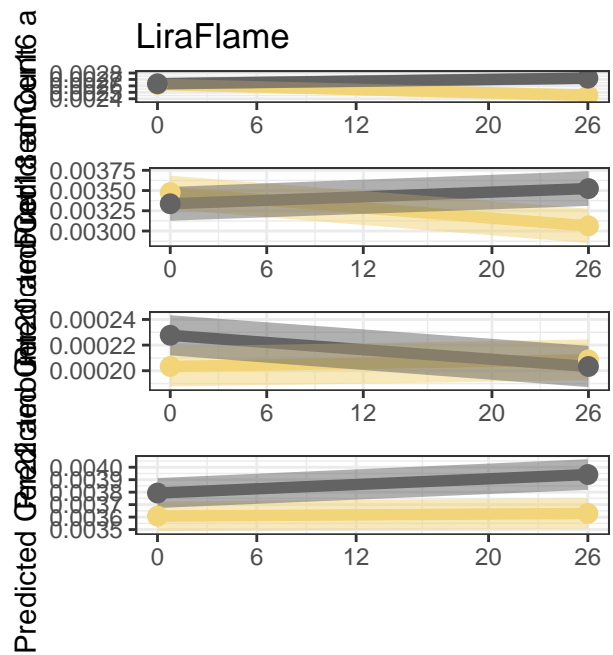
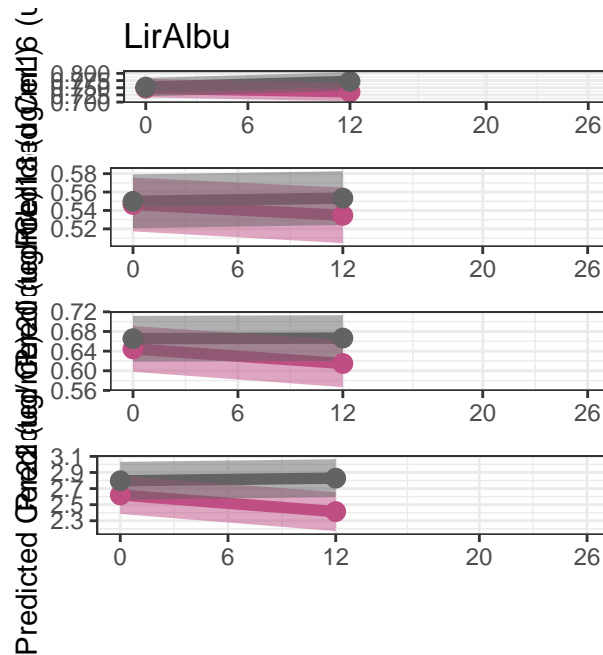




```
#change names back to contain ":"
colnames(data_flame) <- gsub("Cer24_", "Cer24:", colnames(data_flame))

## Combine plots
library(ggpubr)

#Combining LMM plots from LiraFlame and LirAlbu
ggarrange(LMM_albu_list$Cer16$Plot + ggtitle("LirAlbu"),
  LMM_flame_list$Cer16$Plot + ggtitle("LiraFlame"),
  LMM_albu_list$Cer18$Plot, LMM_flame_list$Cer18$Plot,
  LMM_albu_list$Cer20$Plot, LMM_flame_list$Cer20$Plot,
  LMM_albu_list$Cer22$Plot, LMM_flame_list$Cer22$Plot,
  LMM_albu_list$`Cer24:0`$Plot, LMM_flame_list$`Cer24:0`$Plot,
  LMM_albu_list$`Cer24:1`$Plot + theme(legend.position= c(0.8,0.2),
    legend.title=element_blank(),
    axis.title.x = element_text()),
  LMM_flame_list$`Cer24:1`$Plot + theme(legend.position= c(0.15,0.2),
    legend.title=element_blank(),
    axis.title.x = element_text()),
  ncol = 2, nrow = 6)
```



```
#export 12x12

#rm(LMM_albu_list, LMM_flame_list)
rm(tmp_cer, i)

#changes names from CerX to CX Cer
CerX_to_CX_Cer <- function(Cer_name){
  if (is.vector(Cer_name)) {
    Cer_name[grepl("Cer", Cer_name)] <- gsub("Cer", "C", Cer_name[grepl("Cer", Cer_name)])
    Cer_name[grepl("Cer", Cer_name)] <- gsub(":0", "", Cer_name[grepl("Cer", Cer_name)])
    Cer_name[grepl("C\\d{2}", Cer_name)] <- paste0(Cer_name[grepl("C\\d{2}", Cer_name)], " Cer")
  } else {
    if (grepl("Cer", Cer_name)) {
      Cer_name <- gsub("Cer", "C", Cer_name)
      Cer_name <- gsub(":0", "", Cer_name)
      Cer_name <- paste0(Cer_name, " Cer")
    } else {
      Cer_name <- Cer_name
    }
  }
  return(Cer_name)
}

# #Example
# CerX_to_CX_Cer("Cer24:0")
# CerX_to_CX_Cer("test")
```

```

# CerX_to_CX_Cer(c("Cer24:0", "test"))
# CerX_to_CX_Cer(c("Cer24:0", "test", "Computer"))

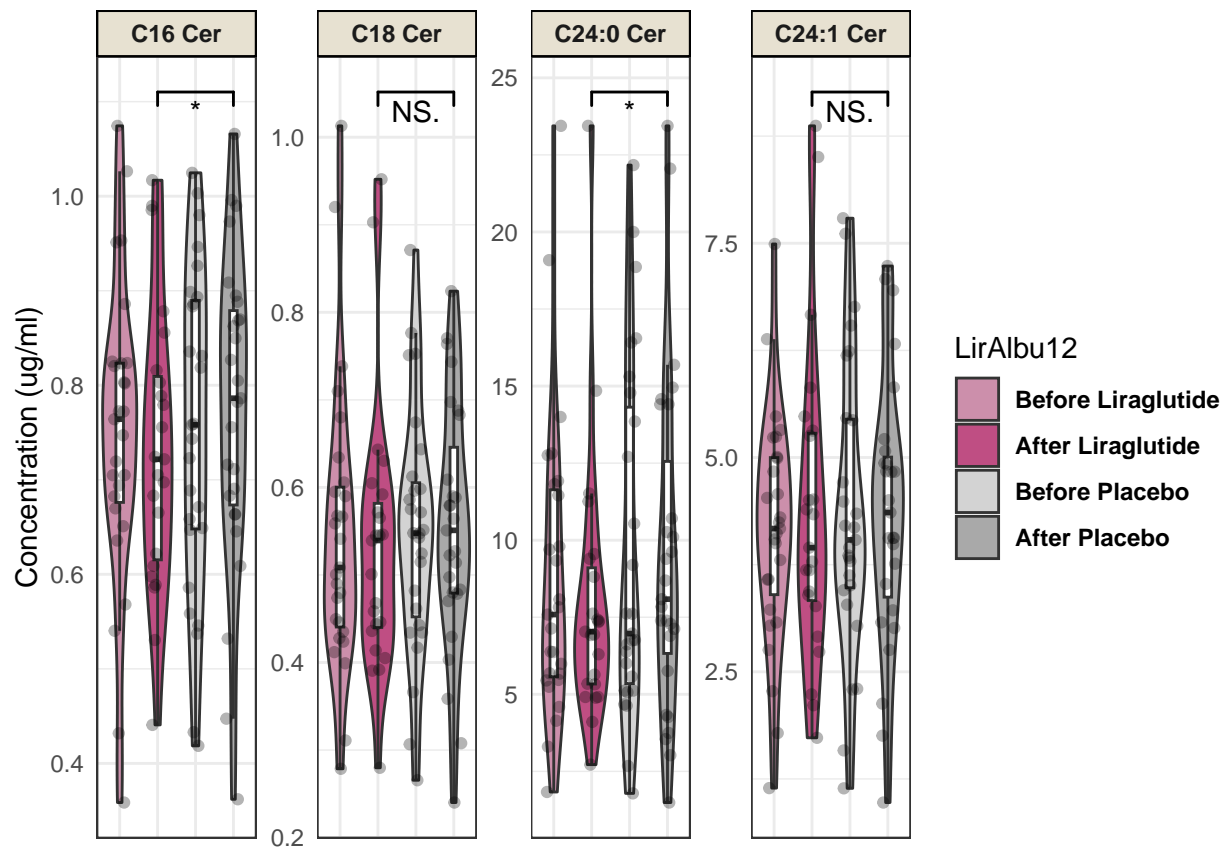
library(ggpubr)

#Boxplot LirAlbu
plot_A <- data_albu %>%
  select(PrePost, Sample_ID, starts_with("Cer")) %>%
  select(-c(Cer20, Cer22)) %>%
  pivot_wider(names_from = c(PrePost, Sample_ID), values_from = starts_with("Cer")) %>%
  pivot_longer(starts_with("Cer")) %>%
  separate(name, sep = "_", c("Ceramide", "PrePost", "Sample_ID")) %>%
  pivot_wider(names_from = c(Ceramide, PrePost), values_from = value) %>%
  pivot_longer(starts_with("Cer")) %>%
  separate(name, sep = "_", c("Ceramide", "PrePost")) %>%
  mutate(PrePost = factor(PrePost,
    levels = c("LiraPre", "LiraPost", "PlaceboPre", "PlaceboPost"),
    labels = c("Before Liraglutide",
      "After Liraglutide",
      "Before Placebo",
      "After Placebo"))) %>%

  mutate(Ceramide = CerX_to_CX_Cer(Ceramide)) %>%
  ggplot(aes(x = PrePost, y = value)) +
  geom_violin(aes(fill = PrePost))+
  scale_fill_manual(values = c("#cc8fab", "#BF4E83", "lightgrey", "darkgrey"))+
  guides(fill = guide_legend(title = "LirAlbu12"))+
  geom_boxplot(outlier.shape = NA,
    width = 0.2)+
  geom_jitter(alpha = 0.3,
    position = position_jitter(width = 0.2))+
  # stat_compare_means(comparisons = list(c("After Liraglutide", "After Placebo")),
  #   method = "t.test", paired = TRUE,
  #   label = "p.signif", hide.ns = TRUE)+
  geom_signif(comparisons = list(c("After Liraglutide", "After Placebo")),
    test = "t.test", test.args = list(paired = TRUE),
    tip_length = 0.02, map_signif_level = TRUE, vjust = 1.8
  )+
  ylab("Concentration (ug/ml)")+
  theme_bw()+
  facet_wrap(~ Ceramide, scales = "free_y", nrow = 1)+
  theme(
    #axis.title.y = element_blank(),
    axis.title.x = element_blank(),
    #legend.title = element_blank(),
    legend.position = "right",
    legend.text = element_text(face = "bold"),
    axis.text.x = element_blank(),
    axis.ticks = element_blank(),
    strip.text = element_text(face = "bold"),
    strip.background = element_rect(colour="black",
      fill="#E7E1D1"))

plot_A

```



```
plot_B <- data_flame %>%
  mutate(PrePost = factor(paste0(Treatment, "_", Time_point),
    levels = c("1_1", "1_2", "2_1", "2_2"),
    label = c("Before Liraglutide",
      "After Liraglutide",
      "Before Placebo",
      "After Placebo"))) %>%

  select(Sample_ID, PrePost, starts_with("Cer")) %>%
  select(-c(Cer20, Cer22)) %>%
  pivot_longer(cols = starts_with("Cer"), names_to = "Ceramide", values_to = "Cer_val") %>%
  mutate(Ceramide = CerX_to_CX_Cer(Ceramide)) %>%
  ggplot(aes(x = PrePost, y = Cer_val)) +
  geom_violin(aes(fill = PrePost)) +
  scale_fill_manual(values = c("#f2e3b6", "#f2d479", "lightgrey", "darkgrey")) +
  guides(fill = guide_legend(title = "LiraFlame26")) +
  geom_boxplot(outlier.shape = NA,
    width = 0.2) +
  geom_jitter(alpha = 0.3,
    position = position_jitter(width = 0.2)) +
  # stat_compare_means(comparisons = list(c("After Liraglutide", "After Placebo")),
  #   method = "t.test", label = "p.signif", hide.ns = TRUE) +
  ylab("Amount") +
  theme_bw() +
  facet_wrap(~ Ceramide, scales = "free_y", nrow = 1) +
  theme(
    #axis.title.y = element_blank(),
```

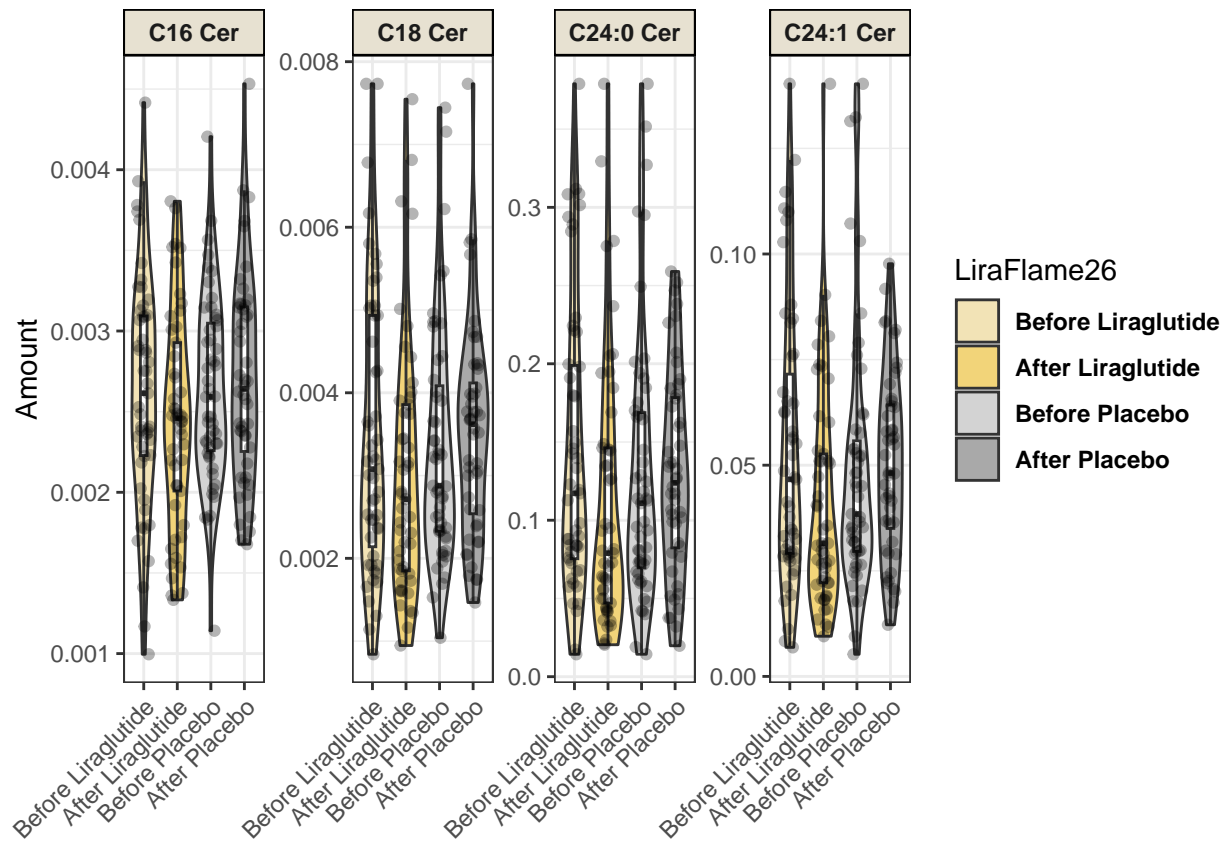


```

axis.title.x = element_blank(),
#legend.title = element_blank(),
legend.position = "right",
axis.text.x = element_text(angle = 45, vjust = 1, hjust = 1),
legend.text = element_text(face = "bold"),
strip.text = element_text(face = "bold"),
strip.background = element_rect(colour="black",
                                fill="#E7E1D1"))

```

plot_B



```

#Combined plot A + B
plot_AB <- ggarrange(plot_A, plot_B,
                      labels = c("A", "B"),
                      heights = c(1, 1.3),
                      ncol = 1, align = "v")

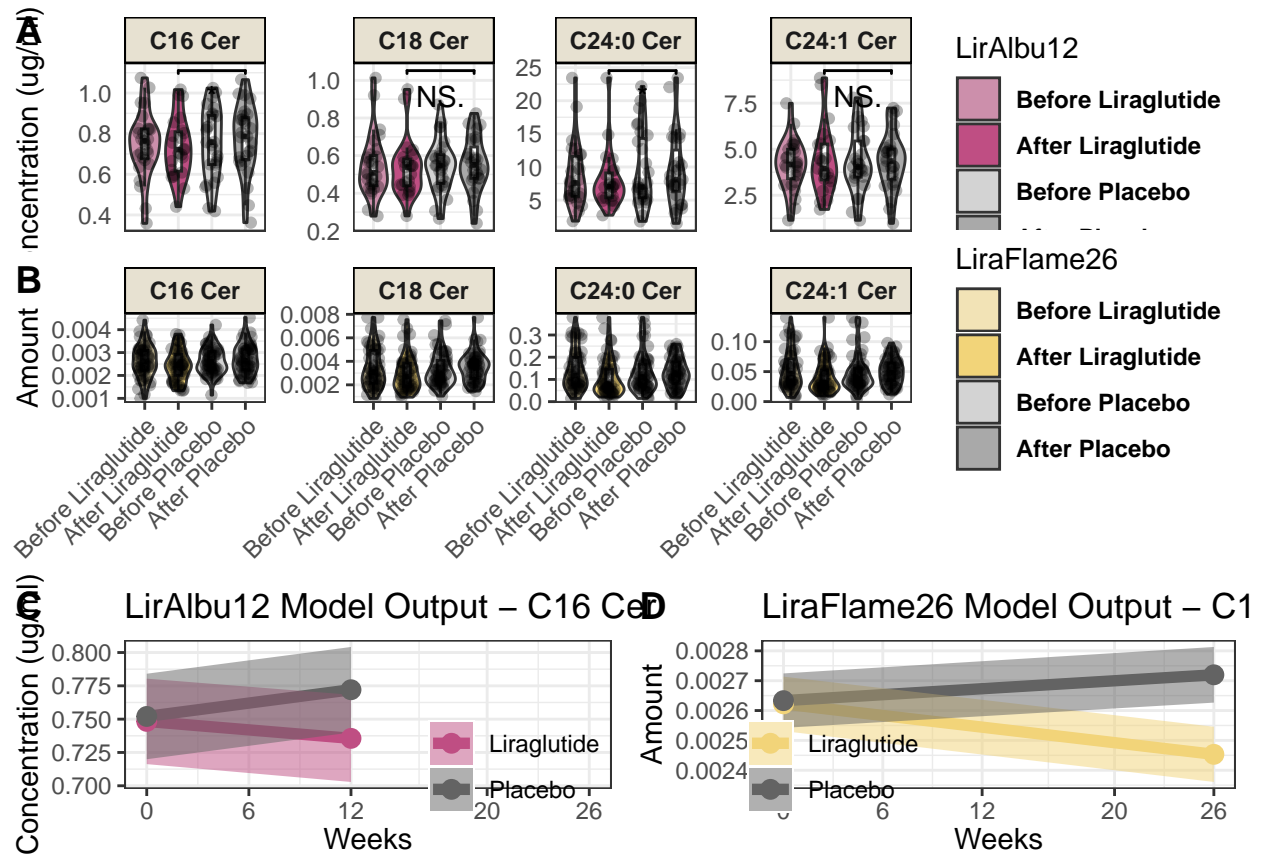
#Combined plot LMM 1 + 2
plot_LMM12 <- ggarrange(
  LMM_albu_list$Cer16$Plot +
  ggtitle("LirAlbu12 Model Output - C16 Cer") +
  ylab("Concentration (ug/ml)") +
  theme(legend.position = c(0.8, 0.2),
        legend.title = element_blank(),
        axis.title.x = element_text()),

```

```
LMM_flame_list$Cer16$Plot +
  ggtitle("LiraFlame26 Model Output - C16 Cer") +
  ylab("Amount")+
  theme(legend.position = c(0.15, 0.2),
        legend.title = element_blank(),
        axis.title.x = element_text()),
  labels = c("C", "D"), ncol = 2)
```

```
#Combine plot_AB and plot_LMM12
```

```
ggarrange(plot_AB, plot_LMM12, ncol = 1, heights = c(2,1), align = "v")
```



```
#export 10x12
```

```
rm(plot_A, plot_B, plot_AB, LMM_albu_list, LMM_flame_list, plot_LMM12)
```

```
library(ggcorrplot)
library(Hmisc)
library(ggpubr)
```

```
##LirAlbu
```

```
#Variables of interest
```

```
tmp_voi <- c("Age", "Sex", "Weight", "Diabetes_Duration", "HbA1c",
             "Total_Cholesterol", "LDL", "Triglyceride", "OfficeSBP",
             "eGFR_CKDEPI", "Log10MeanUAER")
```

```

#Prepare data for heatmap, select variables of interest, remove non-numeric, impute missing values
tmp_cor <- data_albu %>%
  select(rev(starts_with("Cer")), all_of(tmp_voi)) %>%
  select_if(., is.numeric) %>%
  mutate(across(everything(), ~ if_else(is.na(.x), median(., na.rm = TRUE), .)))

#Fix names
tmp_names <- c("Age", "Weight", "Diabetes duration", "HbA1c", "Total cholesterol",
               "LDL", "Triglyceride", "Systolic blood pressure", "eGFR", "Log10(UAER)")

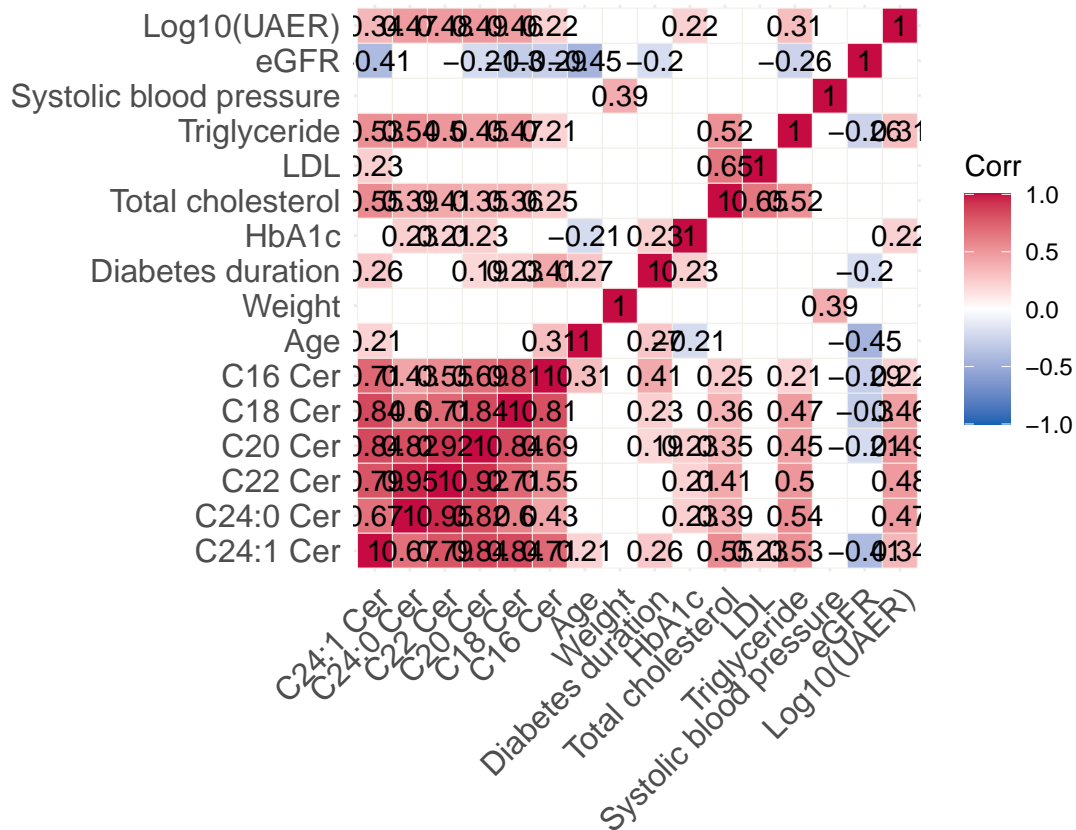
colnames(tmp_cor)[!grepl("Cer", colnames(tmp_cor))] <- tmp_names
colnames(tmp_cor) <- CerX_to_CX_Cer(colnames(tmp_cor))

#Calculate correlation matrix
tmp_cor <- Hmisc::rcorr(as.matrix(tmp_cor), type = "pearson")

#set NA p-values to 0
tmp_cor$P[is.na(tmp_cor$P)] <- 0

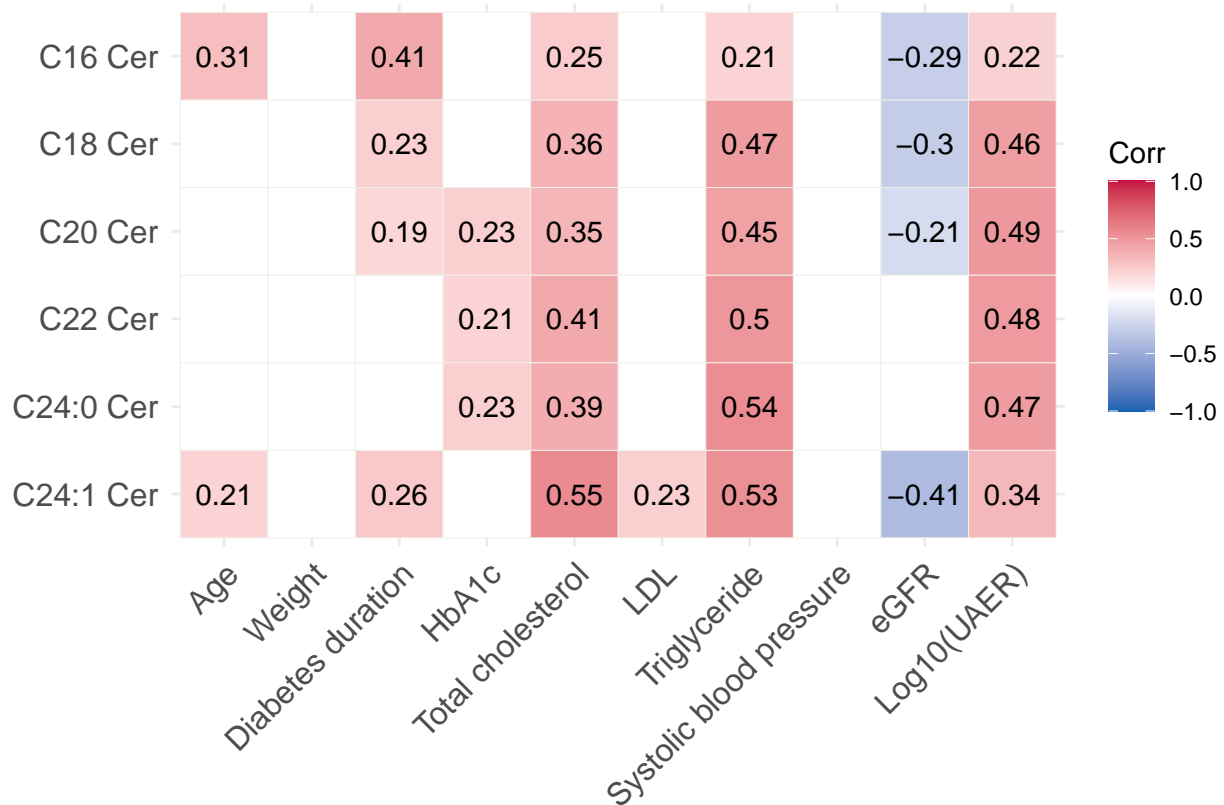
#Full
ggcorrplot::ggcorrplot(tmp_cor$r,
                        hc.order = FALSE,
                        outline.col = "#F2EFE9",
                        p.mat = tmp_cor$P,
                        sig.level = 0.05,
                        insig = "blank",
                        lab = TRUE,
                        ggtheme = ggplot2::theme_minimal,
                        colors = c("#1262B2", "white", "#C50B42"))

```



```
#Subset
heatmap_albu <- ggcorrplot::ggcorrplot(tmp_cor$r[!grepl("Cer", colnames(tmp_cor$r)),
grepl("Cer", colnames(tmp_cor$r))],
outline.col = "#F2EFE9",
p.mat = tmp_cor$P[!grepl("Cer", colnames(tmp_cor$P)),
grepl("Cer", colnames(tmp_cor$P))],
sig.level = 0.05,
insig = "blank",
lab = TRUE,
ggtheme = ggplot2::theme_minimal,
colors = c("#1262B2", "white", "#C50B42"))

heatmap_albu
```



```
#export 5x6

rm(tmp_cor, tmp_voi, tmp_names)

##LiraFlame
#Variables of interest
tmp_voi <- c("age", "Sex", "V2_B_VAEGT", "dm_var", "V2_B_HBA1C",
             "CHOL_v2", "LDL_v2", "TRIG_v2", "Sys_gen_v2_beregnet",
             "GFRapi_V2", "V2_U_LogMeanUAER")

#Prepare data for heatmap, select variables of interest, remove non-numeric, impute missing values
tmp_cor <- data_flame %>%
  select(rev(starts_with("Cer")), all_of(tmp_voi)) %>%
  select_if(., is.numeric) %>%
  mutate(across(everything(), ~ if_else(is.na(.x), median(., na.rm = TRUE), .)))

#Fix names
tmp_names <- c("Age", "Weight", "Diabetes duration", "HbA1c", "Total cholesterol",
               "LDL", "Triglyceride", "Systolic blood pressure", "eGFR", "Log10(UAER)")

colnames(tmp_cor)[!grepl("Cer", colnames(tmp_cor))] <- tmp_names
colnames(tmp_cor) <- CerX_to_CX_Cer(colnames(tmp_cor))

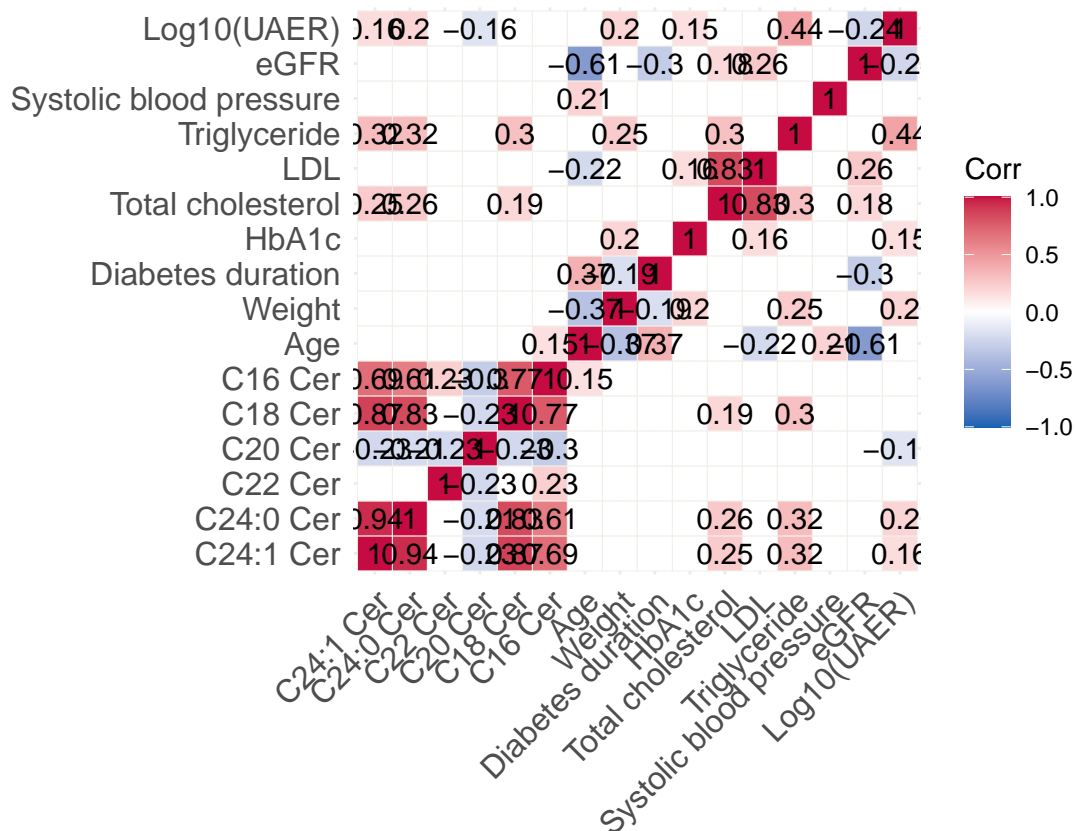
#Calculate correlation matrix
tmp_cor <- Hmisc::rcorr(as.matrix(tmp_cor), type = "pearson")
```

```

#set NA p-values to 0
tmp_cor$P[is.na(tmp_cor$P)] <- 0

#Full
ggcorrplot::ggcorrplot(tmp_cor$r,
  hc.order = FALSE,
  outline.col = "#F2EFE9",
  p.mat = tmp_cor$P,
  sig.level = 0.05,
  insig = "blank",
  lab = TRUE,
  ggtheme = ggplot2::theme_minimal,
  colors = c("#1262B2", "white", "#C50B42"))

```

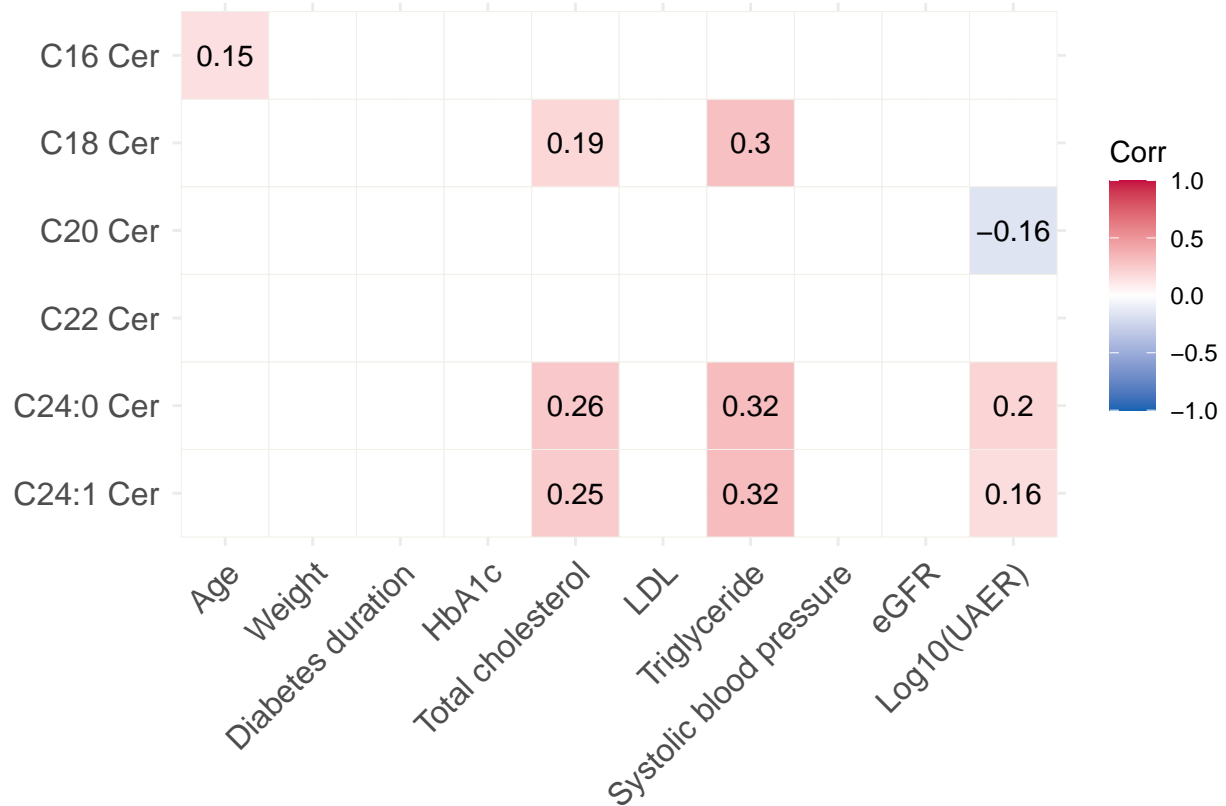


```

#Subset
heatmap_flame <- ggcorrplot::ggcorrplot(tmp_cor$r[!grepl("Cer", colnames(tmp_cor$r)),
  grepl("Cer", colnames(tmp_cor$r))],
  outline.col = "#F2EFE9",
  p.mat = tmp_cor$P[!grepl("Cer", colnames(tmp_cor$P)),
    grepl("Cer", colnames(tmp_cor$P))],
  sig.level = 0.05,
  insig = "blank",
  lab = TRUE,
  ggtheme = ggplot2::theme_minimal,
  colors = c("#1262B2", "white", "#C50B42"))

```

```
heatmap_flame
```



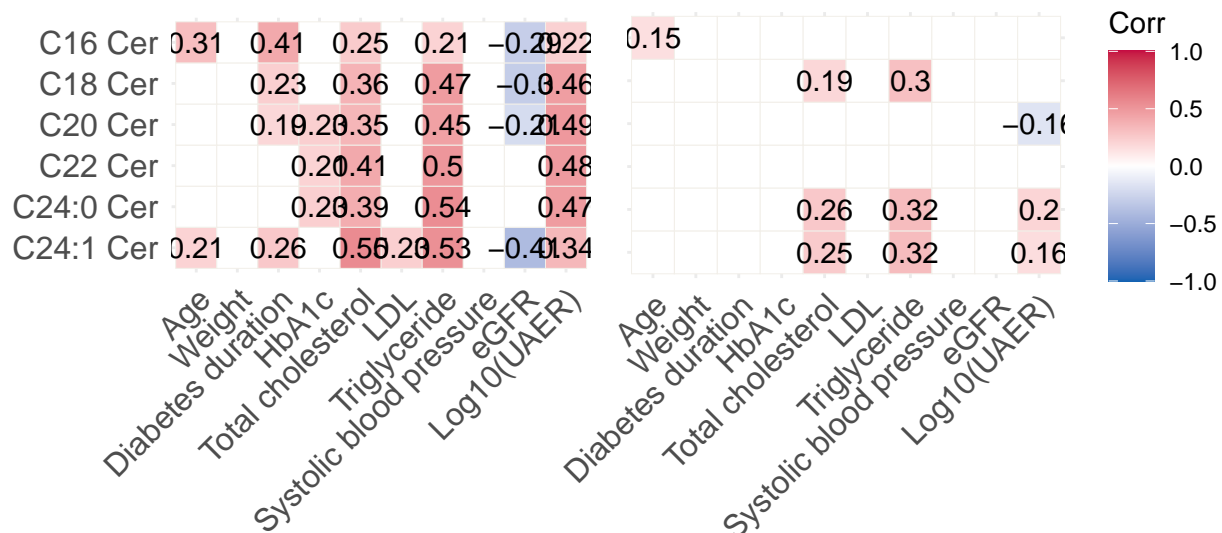
```
#export 5x6
```

```
#combine
```

```
ggarrange(heatmap_albu+
  theme(legend.position = "none"),
  heatmap_flame+
  theme(axis.text.y = element_blank()),
  labels = c("LirAlbu12", "LiraFlame26"),
  widths = c(1, 1.055), align = "h")
```

LirAlbu12

LiraFlame26



```
#export 4.5x10
```

```
rm(tmp_cor, tmp_voi, tmp_names, heatmap_albu, heatmap_flame)
```

```
#Calculate variable change
```

```
data_albu_change <- data_albu %>%
  select(PrePost, Sample_ID, starts_with("Cer"), `Log10MeanUAER`, Weight, HbA1c) %>%
  pivot_wider(names_from = c(PrePost, Sample_ID),
    values_from = c(starts_with("Cer"), Log10MeanUAER, Weight, HbA1c)) %>%
  pivot_longer(c(starts_with("Cer"),
    starts_with("Log10MeanUAER"),
    starts_with("Weight"),
    starts_with("HbA1c")) %>%
    separate(name, sep = "_", c("Variable", "PrePost", "Sample_ID")) %>%
    pivot_wider(names_from = c(Variable, PrePost), values_from = value) %>%
    mutate(across(ends_with("Post"),
      ~ . - get(glue::glue(str_replace(cur_column(), "Post", "Pre")))
    )) %>%
    select(-ends_with("Pre")) %>%
    rename_with(~str_replace(., "Post", ""), everything()) %>%
    pivot_longer(c(starts_with("Cer"),
      starts_with("Log10MeanUAER"),
      starts_with("Weight"),
      starts_with("HbA1c")) %>%
    separate(name, sep = "_", c("Variable", "Treatment")) %>%
    pivot_wider(names_from = c(Variable), values_from = value) %>%
```



```

rename_with(~str_replace(., ":", ""), everything())

#Loop through lipids of interest, and potential mediators
#Lipids of interest
Lipid_list <- data_albu_change %>%
  select(starts_with("Cer")) %>%
  colnames()

#Mediators of interest
Mediator_list <- data_albu_change %>%
  select(Weight, HbA1c, Log10MeanUAER) %>%
  colnames()

#Empty table to populate
Mediator_table <- data.frame()

#Nested loop going through all the lipid and mediators
#Each combination is tested in 3 linear regression models:
#Model 1: Lipid ~ Treatment
#Model 2: Mediator ~ Treatment
#Model 3: Lipid ~ Treatment + Mediator
#Model 4: Causal Mediation Analysis of Model 2 and 3 (bootstrap 1000)
#Note: this step quite slow, get a cup of coffee
if(TRUE){ #Set this to TRUE in order to run
  for(l in Mediator_list){
    for(i in Lipid_list){

      #Step 1: Total effect - Treatment on lipid: i
      form1 <- paste(i, "~ Treatment")
      model1 <- lm(form1, data_albu_change)
      model1_sum <- summary(model1)

      #Step 2: Effect on mediator - Treatment on Mediator l
      form2 <- paste(l, "~ Treatment")
      model2 <- lm(form2, data_albu_change)
      model2_sum <- summary(model2)

      #Step 3: Effect of mediator on response - Medi_var on Resp_Var
      form3 <- paste(i, "~ Treatment +", l)
      model3 <- lm(form3, data_albu_change)
      model3_sum <- summary(model3)

      #Step 4: Casual mediation analysis
      model4 <- mediation::mediate(model.m = model2, model.y = model3,
                                   treat = "Treatment",
                                   mediator = l,
                                   boot = TRUE)

      #Extract explanatory variable(s)
      explanatory_var <- rownames(model1_sum$coefficients)[nrow(model1_sum$coefficients)]

      #Populate

```

```

#Model 1
Mediator_table[paste0("Model1_Estimate"), i] <-
  model1_sum$coefficients[explanatory_var, "Estimate"]
Mediator_table[paste0("Model1_StdError"), i] <-
  model1_sum$coefficients[explanatory_var, "Std. Error"]
Mediator_table[paste0("Model1_Pval"), i] <-
  model1_sum$coefficients[explanatory_var, "Pr(>|t|)"]
Mediator_table[paste0("Model1_R2"), i] <- model1_sum$adj.r.squared
#Model2
Mediator_table[paste0("Model2_", 1, "_Estimate"), i] <-
  model2_sum$coefficients[explanatory_var, "Estimate"]
Mediator_table[paste0("Model2_", 1, "_StdError"), i] <-
  model2_sum$coefficients[explanatory_var, "Std. Error"]
Mediator_table[paste0("Model2_", 1, "_Pval"), i] <-
  model2_sum$coefficients[explanatory_var, "Pr(>|t|)"]
Mediator_table[paste0("Model2_", 1, "_R2"), i] <- model2_sum$adj.r.squared
#Model3
Mediator_table[paste0("Model3_", 1, "_Estimate"), i] <-
  model3_sum$coefficients[1, "Estimate"]
Mediator_table[paste0("Model3_", 1, "_StdError"), i] <-
  model3_sum$coefficients[1, "Std. Error"]
Mediator_table[paste0("Model3_", 1, "_Pval"), i] <-
  model3_sum$coefficients[1, "Pr(>|t|)"]
Mediator_table[paste0("Model3_", 1, "_R2"), i] <- model3_sum$adj.r.squared
#Model4
Mediator_table[paste0("Model4_", 1, "_Total_Estimate"), i] <- model4$tau.coef
Mediator_table[paste0("Model4_", 1, "_Total_CIlower"), i] <- model4$tau.ci[["2.5%"]]
Mediator_table[paste0("Model4_", 1, "_Total_CIupper"), i] <- model4$tau.ci[["97.5%"]]
Mediator_table[paste0("Model4_", 1, "_Total_Pval"), i] <- model4$tau.p

Mediator_table[paste0("Model4_", 1, "_ADE_Estimate"), i] <- model4$z.avg
Mediator_table[paste0("Model4_", 1, "_ADE_CIlower"), i] <- model4$z.avg.ci[["2.5%"]]
Mediator_table[paste0("Model4_", 1, "_ADE_CIupper"), i] <- model4$z.avg.ci[["97.5%"]]
Mediator_table[paste0("Model4_", 1, "_ADE_Pval"), i] <- model4$z.avg.p

Mediator_table[paste0("Model4_", 1, "_ACME_Estimate"), i] <- model4$d.avg
Mediator_table[paste0("Model4_", 1, "_ACME_CIlower"), i] <- model4$d.avg.ci[["2.5%"]]
Mediator_table[paste0("Model4_", 1, "_ACME_CIupper"), i] <- model4$d.avg.ci[["97.5%"]]
Mediator_table[paste0("Model4_", 1, "_ACME_Pval"), i] <- model4$d.avg.p
}
}
}
# # Save Mediator_table as CSV
# write.csv(Mediator_table, here("data/Mediator_table.csv"))

rm(model1, model1_sum, model2, model2_sum, model3, model3_sum, model4,
  data_albu_change, explanatory_var, form1, form2, form3, i, 1,
  Lipid_list, Mediator_list)

# #Step 1: Total effect - Treatment on Cer16
# model1 <- lm(Cer16 ~ Treatment, data = data_albu_change)
# summary(model1)
#

```

```

# #Step 2: Effect on mediator - Treatment on weight
# model2 <- lm(Weight ~ Treatment, data = data_albu_change)
# summary(model2)
#
# #Step 3: Effect of mediator on dependent - Weight on Cer
# model3 <- lm(Cer16 ~ Treatment+Weight, data = data_albu_change)
# summary(model3)
#
# #Step 4: Casual mediation analysis
# model4 <- mediate(model.m = model2, model.y = model3,
#                   treat = "Treatment", mediator = "Weight", boot = TRUE)
# summary(model4)

#Formatting mediator table
Mediator_table_compact <- Mediator_table %>%
  mutate(Model = rownames(.)) %>%
  as_tibble(.) %>%
  filter(grepl("Model4", Model)) %>%
  separate(Model, sep = "_", c("Model", "Mediator", "Measure", "Output")) %>%
  select(-Model) %>%
  filter(!grepl("ADE", Measure)) %>%
  pivot_longer(contains("Cer"), names_to = "Ceramide") %>%
  pivot_wider(names_from = c(Mediator, Measure, Output), values_from = value) %>%
  mutate(across(ends_with("_Estimate"),
    ~ paste0(round(., 3), " (",
      round(get(glue::glue(str_replace(cur_column(), "_Estimate", "_CIlower"))), 3), ";",
      round(get(glue::glue(str_replace(cur_column(), "_Estimate", "_CIupper"))), 3), ")")
    ))) %>%
  select(-ends_with("_CIlower"), -ends_with("_CIupper"))

#NOTE: There might be small variations in the values due the randomness of the bootstrapping
# #Save Mediator_table_compact as CSV
# write.csv(Mediator_table_compact, here("data/Mediator_table_compact.csv"))

rm(Mediator_table, Mediator_table_compact)

```

```

#Calculate variable change
data_flame_change <- data_flame %>%
  select(Time_point, Treatment, Sample_ID, starts_with("Cer"),
    V2_U_LogMeanUAER, V5_U_LogMeanUAER, V2_B_VAEGT, V5_B_VAEGT, V2_B_HBA1C, V5_B_HBA1C) %>%
  mutate(Time_point = factor(Time_point,
    labels = c("Pre", "Post"))) %>%
  mutate(Treatment = factor(Treatment,
    labels = c("Lira", "Placebo"))) %>%
  rename_with(~ gsub("_.", "", .)) %>%
  pivot_wider(names_from = c(Treatment, Time_point, Sample_ID),
    values_from = c(starts_with("Cer"), V2LogMeanUAER, V5LogMeanUAER,
      V2VAEGT, V5VAEGT, V2HBA1C, V5HBA1C)) %>%
  pivot_longer(everything()) %>%
  separate(name, sep = "_", c("Variable", "Treatment", "Time_point", "Sample_ID")) %>%
  pivot_wider(names_from = c(Variable, Time_point), values_from = value) %>%

```

```

select(-c(V2LogMeanUAER_Post, V5LogMeanUAER_Pre,
          V2VAEGT_Post, V5VAEGT_Pre,
          V2HBA1C_Post, V5HBA1C_Pre)) %>%
rename_with(~ gsub("V2", "", .)) %>%
rename_with(~ gsub("V5", "", .)) %>%
mutate(across(ends_with("_Post"),
              ~ . - get(glue::glue(str_replace(cur_column(), "_Post", "_Pre"))))
        ) %>%
select(-ends_with("_Pre")) %>%
rename_with(~str_replace(., "_Post", ""), everything()) %>%
rename_with(~str_replace(., ":", ""), everything())

#Loop through lipids of interest, and potential mediators
#Lipids of interest
Lipid_list_flame <- data_flame_change %>%
  select(starts_with("Cer")) %>%
  colnames()

#Mediators of interest
Mediator_list_flame <- data_flame_change %>%
  select(VAEGT, HBA1C, LogMeanUAER) %>%
  colnames()

#Empty table to populate
Mediator_table_flame <- data.frame()

#Nested loop going through all the lipid and mediators
#Each combination is tested in 3 linear regression models:
#Model 1: Lipid ~ Treatment
#Model 2: Mediator ~ Treatment
#Model 3: Lipid ~ Treatment + Mediator
#Model 4: Causal Mediation Analysis of Model 2 and 3 (bootstrap 1000)
#Note: this step quite slow, get a cup of coffee
if(TRUE){ #Set this to TRUE in order to run
for(l in Mediator_list_flame){
  for(i in Lipid_list_flame){

    #Step 1: Total effect - Treatment on lipid: i
    form1 <- paste(i, "~ Treatment")
    model1 <- lm(form1, data_flame_change)
    model1_sum <- summary(model1)

    #Step 2: Effect on mediator - Treatment on Mediator l
    form2 <- paste(l, "~ Treatment")
    model2 <- lm(form2, data_flame_change)
    model2_sum <- summary(model2)

    #Step 3: Effect of mediator on response - Medi_var on Resp_Var
    form3 <- paste(i, "~ Treatment +", l)
    model3 <- lm(form3, data_flame_change)
    model3_sum <- summary(model3)
  }
}
}

```

```

#Step 4: Casual mediation analysis
model4 <- mediation::mediate(model.m = model2, model.y = model3,
                             treat = "Treatment",
                             mediator = 1,
                             boot = TRUE)

#Extract explanatory variable(s)
explanatory_var <- rownames(model1_sum$coefficients)[nrow(model1_sum$coefficients)]

#Populate
#Model 1
Mediator_table_flame[paste0("Model1_Estimate"), i] <-
  model1_sum$coefficients[explanatory_var, "Estimate"]
Mediator_table_flame[paste0("Model1_StdError"), i] <-
  model1_sum$coefficients[explanatory_var, "Std. Error"]
Mediator_table_flame[paste0("Model1_Pval"), i] <-
  model1_sum$coefficients[explanatory_var, "Pr(>|t|)"]
Mediator_table_flame[paste0("Model1_R2"), i] <- model1_sum$adj.r.squared

#Model2
Mediator_table_flame[paste0("Model2_", 1, "_Estimate"), i] <-
  model2_sum$coefficients[explanatory_var, "Estimate"]
Mediator_table_flame[paste0("Model2_", 1, "_StdError"), i] <-
  model2_sum$coefficients[explanatory_var, "Std. Error"]
Mediator_table_flame[paste0("Model2_", 1, "_Pval"), i] <-
  model2_sum$coefficients[explanatory_var, "Pr(>|t|)"]
Mediator_table_flame[paste0("Model2_", 1, "_R2"), i] <- model2_sum$adj.r.squared

#Model3
Mediator_table_flame[paste0("Model3_", 1, "_Estimate"), i] <-
  model3_sum$coefficients[1, "Estimate"]
Mediator_table_flame[paste0("Model3_", 1, "_StdError"), i] <-
  model3_sum$coefficients[1, "Std. Error"]
Mediator_table_flame[paste0("Model3_", 1, "_Pval"), i] <-
  model3_sum$coefficients[1, "Pr(>|t|)"]
Mediator_table_flame[paste0("Model3_", 1, "_R2"), i] <- model3_sum$adj.r.squared

#Model4
Mediator_table_flame[paste0("Model4_", 1, "_Total_Estimate"), i] <- model4$tau.coef
Mediator_table_flame[paste0("Model4_", 1, "_Total_CIlower"), i] <- model4$tau.ci[["2.5%"]]
Mediator_table_flame[paste0("Model4_", 1, "_Total_CIupper"), i] <- model4$tau.ci[["97.5%"]]
Mediator_table_flame[paste0("Model4_", 1, "_Total_Pval"), i] <- model4$tau.p

Mediator_table_flame[paste0("Model4_", 1, "_ADE_Estimate"), i] <- model4$z.avg
Mediator_table_flame[paste0("Model4_", 1, "_ADE_CIlower"), i] <- model4$z.avg.ci[["2.5%"]]
Mediator_table_flame[paste0("Model4_", 1, "_ADE_CIupper"), i] <- model4$z.avg.ci[["97.5%"]]
Mediator_table_flame[paste0("Model4_", 1, "_ADE_Pval"), i] <- model4$z.avg.p

Mediator_table_flame[paste0("Model4_", 1, "_ACME_Estimate"), i] <- model4$d.avg
Mediator_table_flame[paste0("Model4_", 1, "_ACME_CIlower"), i] <- model4$d.avg.ci[["2.5%"]]
Mediator_table_flame[paste0("Model4_", 1, "_ACME_CIupper"), i] <- model4$d.avg.ci[["97.5%"]]
Mediator_table_flame[paste0("Model4_", 1, "_ACME_Pval"), i] <- model4$d.avg.p
}
}
}
# #Save Mediator_table as CSV

```

```

# write.csv(Mediator_table_flame, here("data/Mediator_table_flame.csv"))

rm(model1, model1_sum, model2, model2_sum, model3, model3_sum, model4,
    data_flame_change, explanatory_var, form1, form2, form3, i, l,
    Lipid_list_flame, Mediator_list_flame)

#Formatting mediator table
Mediator_table_flame_compact <- Mediator_table_flame %>%
  mutate(Model = rownames()) %>%
  as_tibble(.) %>%
  filter(grepl("Model4", Model)) %>%
  separate(Model, sep = "_", c("Model", "Mediator", "Measure", "Output")) %>%
  select(-Model) %>%
  filter(!grepl("ADE", Measure)) %>%
  pivot_longer(contains("Cer"), names_to = "Ceramide") %>%
  pivot_wider(names_from = c(Mediator, Measure, Output), values_from = value) %>%
  mutate(across(ends_with("_Estimate"),
    ~ paste0(format(., digits = 4), " (",
      format(get(glue::glue(str_replace(cur_column(),
        "_Estimate", "_CIlower"))), digits = 4), ";",
      format(get(glue::glue(str_replace(cur_column(),
        "_Estimate", "_CIupper"))), digits = 4), ")")
    ))) %>%
  select(-ends_with("_CIlower"), -ends_with("_CIupper"))

#NOTE: There might be small variations in the values due the randomness of the bootstrapping
# #Save Mediator_table_flame_compact as CSV
# write.csv(Mediator_table_flame_compact, here("data/Mediator_table_flame_compact.csv"))

rm(Mediator_table_flame, Mediator_table_flame_compact)

```