Trait Sexual Desire-Linked Subjective Sexual Arousal to Erotic and Non-Erotic Stimuli: Gender, Relationship Status, and Gender-Specificity

Code and analyses

Milena Vásquez-Amézquita [1,2,3,4,*] Juan David Leongómez [2,4]
Marina Begoña Martínez-González [1] Meredith L. Chivers [5]

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[1] Departamento de Ciencias Sociales, Universidad de la Costa, Barranquilla 080002, Colombia.
[2] Facultad de Psicología, Universidad El Bosque, Bogotá 110121, Colombia.
[3] Grupo de Investigación en Cultura, Educación y Sociedad, Universidad de la Costa, Barranquilla 080002, Colombia.
[4] CODEC: Ciencias Cognitivas y del Comportamiento, Universidad El Bosque, Bogotá 110121, Colombia.
[5] Department of Psychology, Queen's University, Kingston ON K7L 3N6, Canada.

[*] Correspondence: mvasquezam@unbosque.edu.co

Description

This document contains all code, and step by step explanations for all analyses, figures and tables (including supplementary figures and tables) for:

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Data available from the Open Science Framework (OSF): https://doi.org/10.17605/OSF.IO/3V2E7. All analyses were planned by Milena Vásquez-Amézquita and Juan David Leongómez. This document and its underlying code were created in `R Markdown` by Juan David Leongómez using LaTeX.

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1 Preliminaries

1.1 Load packages

This file was created using `knitr` (Xie, 2014), mostly using `tidyverse` (Wickham et al., 2019) syntax. As such, data wrangling was mainly done using packages such as `dplyr` (Wickham et al., 2023), and most figures were created or modified using `ggplot2` (Wickham, 2016). Tables were created using `knitr::kable` and `kableExtra` (Zhu, 2021).

Linear mixed models were fitted using `lmerTest` (Kuznetsova et al., 2017), assumptions were performed using `performance` (Lüdecke et al., 2021), contrasts and interactions were explored using `emmeans` (Lenth, 2023), and interactions were investigated using the package `interactions` (Long, 2019).

All packages used in this file can be directly installed from the Comprehensive R Archive Network (CRAN). For a complete list of packages used to create this file, and their versions, see section 4, at the end of the document.

```r
library(readxl)
library(ltm)
library(car)
library(tidyverse)
library(ggpubr)
library(tidyquant)
library(performance)
library(kableExtra)
library(psych)
library(scales)
library(lmerTest)
library(emmeans)
library(berryFunctions)
library(bestNormalize)
library(rstatix)
library(effectsize)
library(ggeffects)
#library(gtsummary)
#library(gt)
#library(interactions)
#library(MetBrewer)
#library(ggpmisc)
```

1.2 Define color palettes

Individual color palettes for figures by gender, stimuli sex, or relationship type.

```
# Palette to color figures by gender
color.Gender <- c("red","black")
# Palette to color figures by stimuli sex
color.StimuliSex <- c("#54278F","#FC4E2A")
# Palette to color figures by relationship type
color.Relationship <- c("#2171B5","#DD3497")



# Palette to color figures by dimension type
#color.Dimension <- c("#54278F","#41AB5D","#0570B0")
```

1.3 Custom functions

1.3.1 `pval.lev` and `pe2.lev`

This functions take p-values and epsilon squared effect sizes and formats them in LaTeX, highlighting significant p-values in bold and representing all in an appropriate level.

```
# Version 1 for LaTeX format
pval.lev <- function(pvals) {
  ifelse(pvals < 0.0001, "\\textbf{< 0.0001}",
         ifelse(pvals < 0.001, "\\textbf{< 0.001}",
                ifelse(pvals < 0.05, paste0("\\textbf{", round(pvals, 4), "}"),
                       round(pvals, 2))))
}


# Version for partial epsilon squared
pe2.lev <- function(pvals) {
  ifelse(pvals < 0.0001, "< 0.0001",
         ifelse(pvals < 0.001, "< 0.001",
                ifelse(pvals < 0.05, round(pvals, 4),
                       round(pvals, 2))))
}
```

1.3.2 `pval.stars`

This function takes p-values and adds starts to represent significance levels.

```
pval.stars <- function(pvals) {
  ifelse(pvals < 0.0001, "****",
         ifelse(pvals < 0.001, "***",
                ifelse(pvals < 0.01, "**",
                       ifelse(pvals < 0.05, "*", NA))))
}
```

1.3.3 `corr.stars`

This function creates a correlation matrix, and displays significance (function `corr.stars` modified from http://myowelt.blogspot.com/2008/04/beautiful-correlation-tables-in-r.html).

```
corr.stars <- function(x) {
  require(Hmisc)
  x <- as.matrix(x)
  R <- rcorr(x)$r
```

```
  p <- rcorr(x)$P
  # define notions for significance levels; spacing is important.
  mystars <- ifelse(p < .001,
                    paste0("\\textbf{", round(R, 2), "***}"),
                    ifelse(p < .01,
                           paste0("\\textbf{", round(R, 2), "**}"),
                           ifelse(p < .05,
                                  paste0("\\textbf{", round(R, 2), "*}"),
                                  ifelse(p < .10,
                                         paste0(round(R, 2), "$^{\\dagger}$"),
                                         format(round(R, 2), nsmall = 2)))))
  # build a new matrix that includes the correlations with their appropriate stars
  Rnew <- matrix(mystars,
                 ncol = ncol(x))
  diag(Rnew) <- paste(diag(R), " ",
                      sep = "")
  rownames(Rnew) <- colnames(x)
  colnames(Rnew) <- paste(colnames(x), "",
                          sep = "")
  # remove upper triangle
  Rnew <- as.matrix(Rnew)
  Rnew[upper.tri(Rnew, diag = TRUE)] <- ""
  Rnew <- as.data.frame(Rnew)
  # remove last column and return the matrix (which is now a data frame)
  Rnew <- cbind(Rnew[1:length(Rnew) - 1])
  return(Rnew)
}
```

1.3.4 `anova.sig.lm` and `anova.sig.lmer`

Functions to bold significant p values from summary model tables. It highlights significant p values, and formats the output in LaTeX, ready to be used with `kable`.

```
# Version 1 for linear models (lm)
anova.sig.lm <- function(model, custom_caption) {
  aovTab <- bind_cols(anova_summary(Anova(model, type = 3)),
                      epsilon_squared(model)) |>
    unite(col = "df", DFn:DFd, sep = ", ") |>
    select(Effect, df, F, p, Epsilon2_partial) |>
    mutate(p = pval.lev(p),
           Epsilon2_partial = pe2.lev(Epsilon2_partial)) |>
    mutate_at("Effect", str_replace_all, ":", " × ") |>
    kable(digits = 2,
          booktabs = TRUE,
          align = c("l", rep("c", 4)),
          linesep = "",
          caption = custom_caption,
          col.names = c("Effect", "$df$", "$F$", "$p$", "$\\epsilon^2_p$"),
          escape = FALSE) |>
    kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
    footnote(general = paste0("Sexual desire was transformed using an ordered quantile
                              normalization
                              (\\\\cite{petersonOrderedQuantileNormalization2020a}).
                              Results are type III ANOVA.
                              $R^2$ = ",
                              round(r2(model)$R2, 3),
                              ", $R^2_{adjusted}$ = ",
```

```
                              round(r2(model)$R2_adjusted, 3),
                              ". Gender = participants gender (women, men);
                              Relationship = relationship type (stable, single).
                              As effect size, we report partial epsilon squared
                              ($\\\\epsilon^2_p$), which provides a less biases
                              estimate than $\\\\eta^2$ (see
                              \\\\cite{albersWhenPowerAnalyses2018}).
                              Significant effects are in bold."),
             escape = FALSE,
             threeparttable = TRUE,
             footnote_as_chunk = TRUE)
  return(aovTab)
}

# Version 2 for linear mixed models (lmer)
anova.sig.lmer <- function(model, custom_caption) {
  aovTab <- bind_cols(anova(model),
                      epsilon_squared(model)) |>
    mutate(DenDF = round(DenDF, 2)) |>
    unite(col = "df", NumDF:DenDF, sep = ", ") |>
    rownames_to_column(var = "Effect") |>
    rename("F" = "F value",
           "p" = "Pr(>F)") |>
    select(Effect, df, F, p, Epsilon2_partial) |>
    mutate(p = pval.lev(p),
           Epsilon2_partial = pe2.lev(Epsilon2_partial)) |>
    mutate_at("Effect", str_replace_all, ":", " × ") |>
    mutate_at("Effect", str_replace_all, "`", "") |>
    kable(digits = 2,
          booktabs = TRUE,
          align = c("l", rep("c", 4)),
          linesep = "",
          caption = custom_caption,
          col.names = c("Effect", "$df$", "$F$", "$p$", "$\\epsilon^2_p$"),
          escape = FALSE) |>
    kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
    footnote(general = paste0("Results are type III ANOVA.
                              $R^2_{conditional}$ = ",
                              round(r2_nakagawa(model)$R2_conditional, 3),
                              ", $R^2_{marginal}$ = ",
                              round(r2_nakagawa(model)$R2_marginal, 3),
                              ". Gender = participants gender (women, men);
                              Stimuli sex = sex of each stimulus (male, female);
                              Stimuli content = content of each stimulus (erotic, non-erotic).
                              As effect size, we report partial epsilon squared
                              ($\\\\epsilon^2_p$), which provides a less biases
                              estimate than $\\\\eta^2$ (see
                              \\\\cite{albersWhenPowerAnalyses2018}).
                              Significant effects are in bold."),
             escape = FALSE,
             threeparttable = TRUE,
             footnote_as_chunk = TRUE)
  return(aovTab)
}
```

1.3.5 emms.sig

Function to create a table of estimated marginal means and contrasts at three levels of a covariate, representing significance levels from `emmeans::emmeans` outputs. The function highlights significant p values, and formats the output in LaTeX, ready to be used with `kable`.

```
# Version 1, for interactions
emms.sig <- function(low.i, mid.i, hi.i) {
  emm.low <- data.frame(low.i[[1]])
  emm.mid <- data.frame(mid.i[[1]])
  emm.hi <- data.frame(hi.i[[1]])
  con.low <- data.frame(low.i[[2]])
  con.mid <- data.frame(mid.i[[2]])
  con.hi <- data.frame(hi.i[[2]])

  low.tab <- merge(emm.low, con.low, by = 0, all = TRUE)
  mid.tab <- merge(emm.mid, con.mid, by = 0, all = TRUE)
  hi.tab <- merge(emm.hi, con.hi, by = 0, all = TRUE)

  tab <- bind_rows(low.tab, mid.tab, hi.tab) |>
    select(-c(1,3,6,10:13)) |>
    mutate(p.value = pval.lev(p.value)) |>
    kable(digits = 2,
          booktabs = TRUE,
          align = c("l", rep("c", 4), "l", rep("c", 2)),
          linesep = "",
          caption = paste0("Estimated marginal means and contrasts for ",
                           low.i[[1]]@misc$pri.vars[1],
                           " at different levels of ",
                           low.i[[1]]@misc$by.vars),
          col.names = c(low.i[[1]]@misc$pri.vars[1],
                        "EMM", "$SE$", "$2.5\\% CI$", "$97.5\\% CI$", "Contrast", "$z$", "$p$"),
          escape = FALSE) |>
  pack_rows(group_label = paste0(low.i[[1]]@misc$by.vars, " = Mean - SD"),
            start_row = 1,
            end_row = 2,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = paste0(low.i[[1]]@misc$by.vars, " = Mean"),
            start_row = 3,
            end_row = 4,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = paste0(low.i[[1]]@misc$by.vars, " = Mean + SD"),
            start_row = 5,
            end_row = 6,
            bold = FALSE,
            background = "lightgray") |>
  add_header_above(c(" " = 5, "Contrasts" = 3)) |>
  kable_styling(latex_options = "HOLD_position") |>
  footnote(general = paste0("EMM = estimated marginal mean.
           Significant effects are in bold.
           Continuous variables were centered and scaled (in this case, ",
           low.i[[1]]@misc$by.vars, ").
           An asymptotic method was used to avoid extreme computation
           times (hence, no degrees of freedom are included, and
           $z$ rather than $t$ statistics are reported).
```

```
           For contrasts, Tukey adjustment was used."),
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)

  return(tab)
}

# Version 2, for triple interactions
emms.sig2 <- function(low.i, mid.i, hi.i) {
  emm.low <- data.frame(low.i[[1]])
  emm.mid <- data.frame(mid.i[[1]])
  emm.hi <- data.frame(hi.i[[1]])
  con.low <- data.frame(low.i[[2]])
  con.mid <- data.frame(mid.i[[2]])
  con.hi <- data.frame(hi.i[[2]])

  low.tab <- merge(emm.low, con.low, by = 0, all = TRUE)
  mid.tab <- merge(emm.mid, con.mid, by = 0, all = TRUE)
  hi.tab <- merge(emm.hi, con.hi, by = 0, all = TRUE)

  tab <- bind_rows(low.tab, mid.tab, hi.tab) |>
    select(-c(1,4,7,11:14)) |>
    mutate(p.value = pval.lev(p.value)) |>
    kable(digits = 2,
          booktabs = TRUE,
          align = c("l", "l", rep("c", 4), "l", rep("c", 2)),
          linesep = "",
          caption = paste0("Estimated marginal means and contrasts for ",
                           low.i[[1]]@misc$pri.vars[1], " and ",
                           low.i[[1]]@misc$pri.vars[2],
                           " at different levels of ",
                           low.i[[1]]@misc$by.vars),
          col.names = c(low.i[[1]]@misc$pri.vars[1],
                        low.i[[1]]@misc$pri.vars[2],
                        "EMM", "$SE$", "$2.5\\% CI$", "$97.5\\% CI$", "Contrast", "$z$", "$p$"),
          escape = FALSE) |>
  pack_rows(group_label = paste0(low.i[[1]]@misc$by.vars, " = Mean - SD"),
            start_row = 1,
            end_row = 6,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = paste0(low.i[[1]]@misc$by.vars, " = Mean"),
            start_row = 7,
            end_row = 12,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = paste0(low.i[[1]]@misc$by.vars, " = Mean + SD"),
            start_row = 13,
            end_row = 18,
            bold = FALSE,
            background = "lightgray") |>
  add_header_above(c(" " = 6, "Contrasts" = 3)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = paste0("EMM = estimated marginal mean.
           Significant effects are in bold.
```

```
          Continuous variables were centered and scaled (in this case, ",
          low.i[[1]]@misc$by.vars, ")
          An asymptotic method was used to avoid extreme computation
          times (hence, no degrees of freedom are included, and
          $z$ rather than $t$ statistics are reported).
          For contrasts, Tukey adjustment was used."),
          threeparttable = TRUE,
          footnote_as_chunk = TRUE,
          escape = FALSE)

  return(tab)
}
```

1.3.6 `contr.stars`

Function to create a data frame of model contrasts, representing significance levels from an `emmeans::emmeans` output. These data frames are formatted to be called by the `ggpubr::stat_pvalue_manual` function used in model figures.

```
contr.stars <- function(emms){
  require(emmeans)
  x <- as.data.frame(contrast(emms, interaction = "pairwise"))
  x <- separate(x,
                col = 1,
                into = c("group1", "group2"),
                sep = " - ",
                remove = TRUE)
  x$p.signif <- ifelse(x$p.value < 0.0001, "****",
                       ifelse(x$p.value < 0.001, "***",
                              ifelse(x$p.value < 0.01, "**",
                                     ifelse(x$p.value < 0.05, "*", NA))))
  x <- x |>
    mutate_at("group1", str_replace_all, "[()]", "") |>
    mutate_at("group2", str_replace_all, "[()]", "")
  return(x)
}
```

1.3.7 `prob.dist.tab`

Function to create a table of the probability of a model for each distribution family, using the `check_distribution` function, from the `performance` package (Lüdecke et al., 2021). Values are sorted descending, first for probabilities according to the residual distribution, and then for probabilities according to the response variable. While 18 distribution families are tested, only families with at least one probability (either residual or response variable) higher than 10% are shown in the table.

```
prob.dist.tab <- function(mod){
  # Calculate probabilities for each distribution family
  tibble(check_distribution(mod)) |>
    arrange(desc(p_Response)) |>
    arrange(desc(p_Residuals)) |>
  # Select only distribution families with at leat a 10% probability
  filter(p_Residuals > 0.1 | p_Response > 0.1) |>
  # Transform probabilities to percentages
  mutate(p_Residuals = paste0(round(p_Residuals*100, 2), "\\%")) |>
  mutate(p_Response = paste0(round(p_Response*100, 2), "\\%")) |>
  # Capitalise first letter of each family distribution
  mutate(Distribution = sub("(.)", "\\U\\1", Distribution, perl = TRUE)) |>
```

```
  # Create table
  kable(booktabs = TRUE,
        align = c("l", "c", "c"),
        row.names = FALSE,
        caption = "Distributional family for the model",
        col.names = c("Family",
                      "Residuals",
                      "Response"),
        escape = FALSE) |>
  kable_styling(latex_options = "HOLD_position") |>
  # Bold highest probability
  row_spec(1, background = "#c4c4c4") |>
  footnote(general = "Only families with at least one probability higher than
  10\\\\% are shown, but a total of 18 distribution families were tested.
  The most likely distribution is highlighted.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
}
```

1.4 Load and wrangle data

Change necessary variables to factor, sort levels, and rename variables

```
# Load data
dat <- read.csv("Data/BD_Heterosexuales_Vertical_BIG.csv") |>
  # Remove rows with missing values for Solitary sexual desire (SD_solitario)
  drop_na(SD_solitario) |>
  # Change variables to factor and sort their levels
  mutate_at(c("Contenido_Estimulo", "Sexo", "Sexo_Estimulo", "PrefSex", "EstRel", "Escolaridad",
              "Religion","TiempoRP"), as.factor) |>
  # Rename variables to English
  rename(Participant = Participante,
         Age = EdadParticipante,
         `Preferred sex` = PrefSex,
         Gender = Sexo,
         `Contraceptive uso` = Anticoncep,
         `Last period` = UltimoPer,
         `Period day` = Dia_ciclo,
         Education = Escolaridad,
         Location = Residencia,
         `Location (other)` = Residencia_3_TEXT,
         `Medical history` = AntMed,
         `Sexual orientation` = OS,
         `Relationship status` = EstRel,
         `Relationship duration` = TiempoRP,
         `Partner gender` = SexPareja,
         `Relationship type` = TipoRel,
         `Age at first intercourse` = Primera.ExpSex,
         `Consented to first intercourse` = ConExpSex,
         `Number of sexual partners` = Numero.Parejas,
         `Pornography consumed last month` = Pornografia_ultimo_mes,
         Relationship = TieneRelacion,
         `MGH-SFQ (total)` = MGH.SFQ_Total,
         `Dyadic sexual desire (Partner)` = SD_Diadico_pareja,
         `Solitary sexual desire` = SD_solitario,
         `Dyadic sexual desire (Attractive person)` = SD_Diadico_p_atractiva,
```

```
       `MGSS sexual satisfaction (General)` = Satisfaccion.Sexual..MGSS_general.,
       `MGSS sexual satisfaction (Partner)` = Satisfaccion.Sexual..MGSS_Pareja.,
       `Stimuli code` = Codigo_Estimulo,
       `Stimuli sex` = Sexo_Estimulo,
       `Stimuli content` = Contenido_Estimulo,
       `Subjective sexual attractiveness` = Atractivo,
       `Subjective sexual arousal` = Excitacion) |>
# Recode factor levels
mutate(`Stimuli content` = recode_factor(`Stimuli content`,
                                          Erotico = "Erotic",
                                          No_erotico = "Non-erotic")) |>
mutate(Gender = recode_factor(Gender,
                              Femenino = "Women",
                              Masculino = "Men")) |>
mutate(`Stimuli sex` = recode_factor(`Stimuli sex`,
                                     Femenino = "Female",
                                     Masculino = "Male")) |>
mutate(`Preferred sex` = recode_factor(`Preferred sex`,
                                       Hombre = "Male",
                                       Mujer = "Female")) |>
mutate(Education = recode(Education,
                          "Bachillerato" = "High school",
                          "Universitario" = "University",
                          "Postgrado" = "Postgraduate")) |>
mutate(Religion = recode(Religion,
                         "1" = "Religious",
                         "0" = "Non-religious")) |>
mutate(`Pornography consumed last month` = recode(`Pornography consumed last month`,
                                                  "Nunca" = "None",
                                                  "Una o dos veces" = "1-2 times",
                                                  "Tres a cinco veces" = "3-5 times",
                                                  "Mas de 5 veces" = "5 times or more")) |>
# Recode relationship duration
#mutate(`Relationship duration` = replace_na(`Relationship duration`, "Single"))
mutate(`Relationship duration` = recode(`Relationship duration`,
                              "Sin pareja actual"  = "Single",
                              "Menor a 6 meses" = "Less that 6 months",
                              "Entre 6 meses y 2 anos" = "Between 6 months and 2 years",
                              "Entre 2 y 5 anos" = "Between 2 and 5 years",
                              "MÃ¡s de 5 anos" = "More than 5 years"),
       `Relationship duration` = replace_na(`Relationship duration`, "Single")) |>
# Recode relationship type
mutate(Relationship = recode(`Relationship status`,
                              "Exclusiva/No viven juntos" = "Stable",
                              "Exclusiva/Matrimonio" = "Stable",
                              "No exclusiva" = "Non-stable",
                              "Soltero/sin contactos sexuales en un ano" = "Single",
                              "Soltero/contactos sexuales en un ano" = "Single")) |>
# Relevel factors
mutate(Education = fct_relevel(Education,
                                c("High school", "University", "Postgraduate")),
       `Pornography consumed last month` = fct_relevel(`Pornography consumed last month`,
                                c("None", "1-2 times",
                                  "3-5 times", "5 times or more")),
       `Relationship duration` = fct_relevel(`Relationship duration`,
                                c("Single", "Less that 6 months",
```

```
                                   "Between 6 months and 2 years",
                                   "Between 2 and 5 years",
                                   "More than 5 years"))) |>
  mutate(`Stimuli content` = as.factor(`Stimuli content`),
         `Stimuli sex` = as.factor(`Stimuli sex`)) |>
  # Filter participants in non-stable relationships
  filter(Relationship != "Non-stable") |>
  droplevels()
```

---

2 Descriptives

2.0.1 Figure S1. Demographic chacarteristics of the sample

Number of participants by demographic category.

```
# Get number of participant for each combination of demographic chacarteristic
dat.demog <- dat |>
  select(Participant, Gender, Relationship, Education, Religion,
         `Pornography consumed last month`) |>
  group_by(Participant) |>
  filter(row_number() == 1) |>
  ungroup() |>
  group_by(Gender, Relationship, Education, Religion,
         `Pornography consumed last month`) |>
  rename(Porn = `Pornography consumed last month`) |>
  tally() |>
  drop_na(Religion) |>
  ungroup()

# Create separate tables by gender
dat.demog.W <- filter(dat.demog, Gender == "Women")
dat.demog.M <- filter(dat.demog, Gender == "Men")

# Women
samp.w <- ggballoonplot(dat.demog.W, x = "Education", y = "Porn", size = "n",
              fill = "n",
              facet.by = c("Relationship", "Religion")) +
  scale_fill_viridis_c(option = "C", limits = c(1, max(dat.demog$n))) +
  scale_size_continuous(range = c(1, 7), limits = c(1, max(dat.demog$n))) +
  guides(fill = guide_legend(face = "italic"),
         size = guide_legend(face = "italic")) +
  labs(title = "Women", y = "Pornography consumed last month") +
  geom_text(aes(label = n),
            size = 3, nudge_x = 0.3, nudge_y = 0.1) +
  geom_text(aes(label = paste0("\n(",
                               percent(n/sum(dat.demog$n), accuracy = 0.1),
                               ")")),
            size = 2.5, nudge_x = 0.3, nudge_y = -0.05) +
  theme_tq() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
        axis.text.y = element_text(angle = 45, vjust = 0.5))

# Men
samp.m <- ggballoonplot(dat.demog.M, x = "Education", y = "Porn", size = "n",
              fill = "n",
```

```
facet.by = c("Relationship", "Religion")) +
  scale_fill_viridis_c(option = "C", limits = c(1, max(dat.demog$n))) +
  scale_size_continuous(range = c(1, 7), limits = c(1, max(dat.demog$n))) +
  guides(fill = guide_legend(face = "italic"),
         size = guide_legend(face = "italic")) +
  labs(title = "Men", y = NULL) +
  geom_text(aes(label = n),
            size = 3, nudge_x = 0.3, nudge_y = 0.1) +
  geom_text(aes(label = paste0("\n(",
                               percent(n/sum(dat.demog$n), accuracy = 0.1),
                               ")")),
            size = 2.5, nudge_x = 0.3, nudge_y = -0.05) +
  theme_tq() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
       axis.text.y = element_text(angle = 45, vjust = 0.5))

# Full plot
ggarrange(samp.w, samp.m,
          widths = c(1.1, 1),
          common.legend = TRUE,
          legend = "bottom")
```

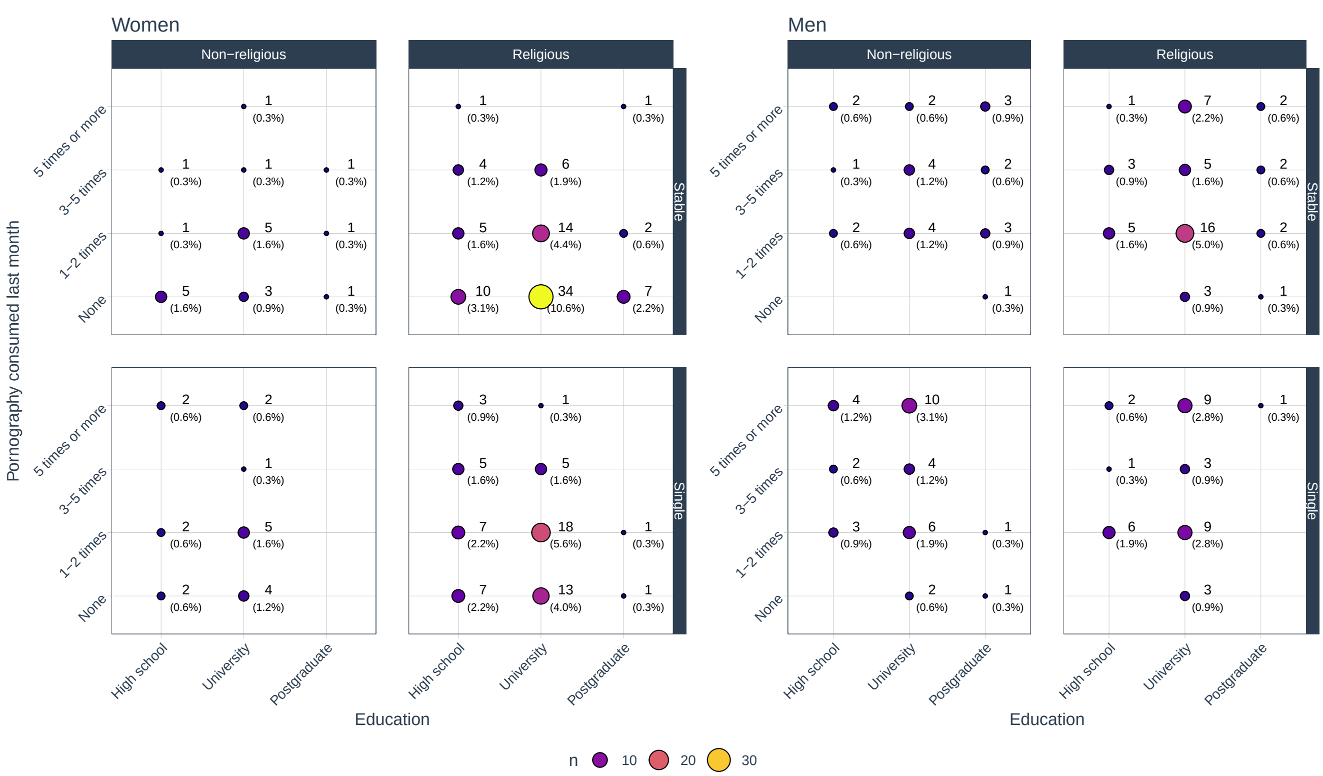

Figure S1. Number of participants by gender (left = women, rigth = men), Relationship (stable = top panels, single = bottom panels), Religion (non-religious = left panels by gender, religious = right panels by gender), Education (X axis), and pornography consumed during the last month (Y axis). The number of participants for each combination of these five variables is displayed as numbers (percentage in brackets), as well as by the color and size of the bubbles .

2.1 Descriptive statistics of the participants by gender

Calculate mean values per participant for relevant, numeric variables.

```r
# Summarize relevant variables by participant
dat.desc <- dat |>
  select(Participant, Gender, Age, Relationship, `Number of sexual partners`,
         `MGH-SFQ (total)`,
         `MGSS sexual satisfaction (General)`, `MGSS sexual satisfaction (Partner)`,
         `Subjective sexual attractiveness`, `Subjective sexual arousal`,
         `Solitary sexual desire`,
         `Dyadic sexual desire (Attractive person)`, `Dyadic sexual desire (Partner)`) |>
  group_by(Participant, Gender, Relationship) |>
  summarize_if(is.numeric, mean, na.rm = TRUE)
```

2.1.1 Table S1. Descriptive statistics of the participants by gender

Table of descriptives by gender.

```r
# Table of descriptives by gender and relationship status
describeBy(dat.desc ~ Relationship + Gender,
           mat=TRUE,
           digits=2)  |>
  rownames_to_column("Measured characteristic") |>
  select(1,3:4,6:9,12:13) |>
  slice(-(1:12)) |>
  select(1,3,2,4:9) |>
  # Remove numbers included to differentiate repeated row names (now on column 1)
  mutate("Measured characteristic" = str_replace_all(`Measured characteristic`,
                                                     c("1" = "", "2" = "", "3" = "", "4" = ""))) |>
  # Create table
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", "l", rep("c", 7)),
        linesep = "",
        caption = "Descriptive statistics the participants by gender
        and relationship status",
        col.names = c("Measured characteristic", "Gender", "Relationship status",
                      "$n$", "Mean", "$SD$", "Median", "Min", "Max"),
        longtable = TRUE,
        escape = FALSE) |>
  kable_styling(latex_options = c("HOLD_position"),
                font_size = 8.2) |>
  collapse_rows(columns = 1:3, valign = "middle") |>
  footnote(general = "Because for \\\\textit{Subjective sexual attractiveness} and
           \\\\textit{Subjective sexual arousal} there are are multiple within-subject
           observations, descriptives are calculated from mean values per participant.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S1. *Descriptive statistics the participants by gender and relationship status*

Measured characteristic	Gender	Relationship status	*n*	Mean	*SD*	Median	Min	Max
Age	Women	Stable	105	24.51	5.58	23.00	18.00	40.00
		Single	79	22.27	3.84	21.00	18.00	36.00
	Men	Stable	72	26.72	5.64	25.00	19.00	40.00
		Single	67	24.24	4.58	23.00	18.00	39.00
	Women	Stable	103	4.41	3.77	3.00	1.00	22.00
		Single	76	5.74	8.85	3.00	0.00	63.00

Number of sexual partners	Men	Stable	72	8.72	11.36	5.00	1.00	70.00
		Single	66	7.30	8.06	4.00	0.00	40.00
MGH-SFQ (total)	Women	Stable	104	3.31	0.96	3.75	0.00	4.00
		Single	79	2.80	1.23	3.50	0.00	4.00
	Men	Stable	72	3.59	0.62	3.90	0.60	4.00
		Single	67	3.38	0.83	3.80	0.60	4.00
MGSS sexual satisfaction (General)	Women	Stable	100	25.88	5.67	28.00	6.00	30.00
		Single	10	26.90	3.11	27.00	22.00	30.00
	Men	Stable	70	26.43	4.54	29.00	12.00	30.00
		Single	12	23.58	5.14	24.50	14.00	29.00
MGSS sexual satisfaction (Partner)	Women	Stable	100	28.13	4.20	30.00	8.00	30.00
		Single	10	28.10	2.13	29.00	25.00	30.00
	Men	Stable	70	28.49	3.48	30.00	6.00	30.00
		Single	12	26.08	4.85	27.50	15.00	30.00
Subjective sexual attractiveness	Women	Stable	105	2.94	1.11	2.78	1.00	5.49
		Single	79	3.19	1.06	3.11	1.44	6.77
	Men	Stable	72	3.27	0.94	3.24	1.11	6.20
		Single	67	3.20	0.90	3.18	1.09	5.72
Subjective sexual arousal	Women	Stable	105	1.59	0.68	1.39	1.00	4.21
		Single	79	1.75	0.71	1.52	1.00	4.39
	Men	Stable	72	2.24	0.83	2.07	1.00	4.57
		Single	67	2.16	0.78	2.05	1.00	4.09
Solitary sexual desire	Women	Stable	105	11.53	8.59	12.00	0.00	29.00
		Single	79	16.03	8.35	17.00	0.00	31.00
	Men	Stable	72	17.47	7.51	17.50	0.00	31.00
		Single	67	18.25	7.10	19.00	1.00	31.00
Dyadic sexual desire (Attractive person)	Women	Stable	105	10.55	7.64	10.00	0.00	30.00
		Single	79	14.06	7.39	15.00	0.00	32.00
	Men	Stable	72	16.21	7.44	15.50	0.00	32.00
		Single	67	17.57	6.66	17.00	2.00	30.00
Dyadic sexual desire (Partner)	Women	Stable	105	27.53	8.50	30.00	0.00	38.00
		Single	76	21.33	10.91	23.00	0.00	38.00
	Men	Stable	72	31.35	5.33	32.00	15.00	38.00
		Single	67	25.81	9.40	28.00	0.00	38.00

Note: Because for *Subjective sexual attractiveness* and *Subjective sexual arousal* there are are multiple within-subject observations, descriptives are calculated from mean values per participant.

2.1.2 Figure S2. Distribution of participants' measured variables by gender

Kernel density distributions by gender.

```
# Convert dat.desc to long format
datp <- dat.desc |>
  pivot_longer(cols = Age:`Dyadic sexual desire (Partner)`,
               names_to = "Variable",
               values_to = "Value") |>
  mutate(Variable = str_wrap(Variable, width = 30))
```

```
# Figure created as 3 separate panels (to use a different number of panels per row)
fs2a <- ggplot(datp |>
                 filter(Variable %in% c("Age",
                                         "Number of sexual partners",
                                         "Subjective sexual\nattractiveness",
                                         "Subjective sexual arousal")),
             aes(Value,
                 fill = Gender,
                 colour = Gender)) +
        geom_density(alpha = 0.3) +
        geom_vline(data = datp |>
                     filter(Variable %in% c("Age",
                                             "Number of sexual partners",
                                             "Subjective sexual\nattractiveness",
                                             "Subjective sexual arousal")) |>
                     group_by(Variable, Gender) |>
                     summarise(mean = mean(Value, na.rm =TRUE)),
                   size = 1,
                   aes(xintercept = mean, color = Gender, linetype = Gender)) +
        scale_color_manual(values = color.Gender) +
        scale_fill_manual(values = color.Gender) +
        facet_wrap(~ Variable,
                   scales = "free",
                   ncol = 4) +
        labs(y = "Density",
             x = NULL) +
        theme_tq()

fs2b <- ggplot(datp |>
                 filter(Variable %in% c("MGH-SFQ (total)",
                                         "MGSS sexual satisfaction\n(General)",
                                         "MGSS sexual satisfaction\n(Partner)")),
             aes(Value,
                 fill = Gender,
                 colour = Gender)) +
        geom_density(alpha = 0.3) +
        geom_vline(data = datp |>
                     filter(Variable %in% c("MGH-SFQ (total)",
                                             "MGSS sexual satisfaction\n(General)",
                                             "MGSS sexual satisfaction\n(Partner)")) |>
                     group_by(Variable, Gender) |>
                     summarise(mean = mean(Value, na.rm =TRUE)),
                   size = 1,
                   aes(xintercept = mean, color = Gender, linetype = Gender)) +
        scale_color_manual(values = color.Gender) +
        scale_fill_manual(values = color.Gender) +
        facet_wrap(~ Variable,
                   scales = "free",
                   ncol = 3) +
        labs(y = "Density",
             x = NULL) +
        theme_tq()

fs2c <- ggplot(datp |>
                 filter(Variable %in% c("Solitary sexual desire",
                                         "Dyadic sexual desire\n(Attractive person)",
```

```
                                        "Dyadic sexual desire (Partner)")),
             aes(Value,
                 fill = Gender,
                 colour = Gender)) +
        geom_density(alpha = 0.3) +
        geom_vline(data = datp |>
                    filter(Variable %in% c("Solitary sexual desire",
                                            "Dyadic sexual desire\n(Attractive person)",
                                            "Dyadic sexual desire (Partner)")) |>
                    group_by(Variable, Gender) |>
                    summarise(mean = mean(Value, na.rm =TRUE)),
                  size = 1,
                  aes(xintercept = mean, color = Gender, linetype = Gender)) +
        scale_color_manual(values = color.Gender) +
        scale_fill_manual(values = color.Gender) +
        facet_wrap(~ Variable,
                   scales = "free",
                   ncol = 3) +
        labs(y = "Density",
             x = NULL) +
        theme_tq()

# Full plot
ggarrange(fs2a, fs2b, fs2c,
          common.legend = TRUE,
          legend = "bottom",
          nrow = 3)
```

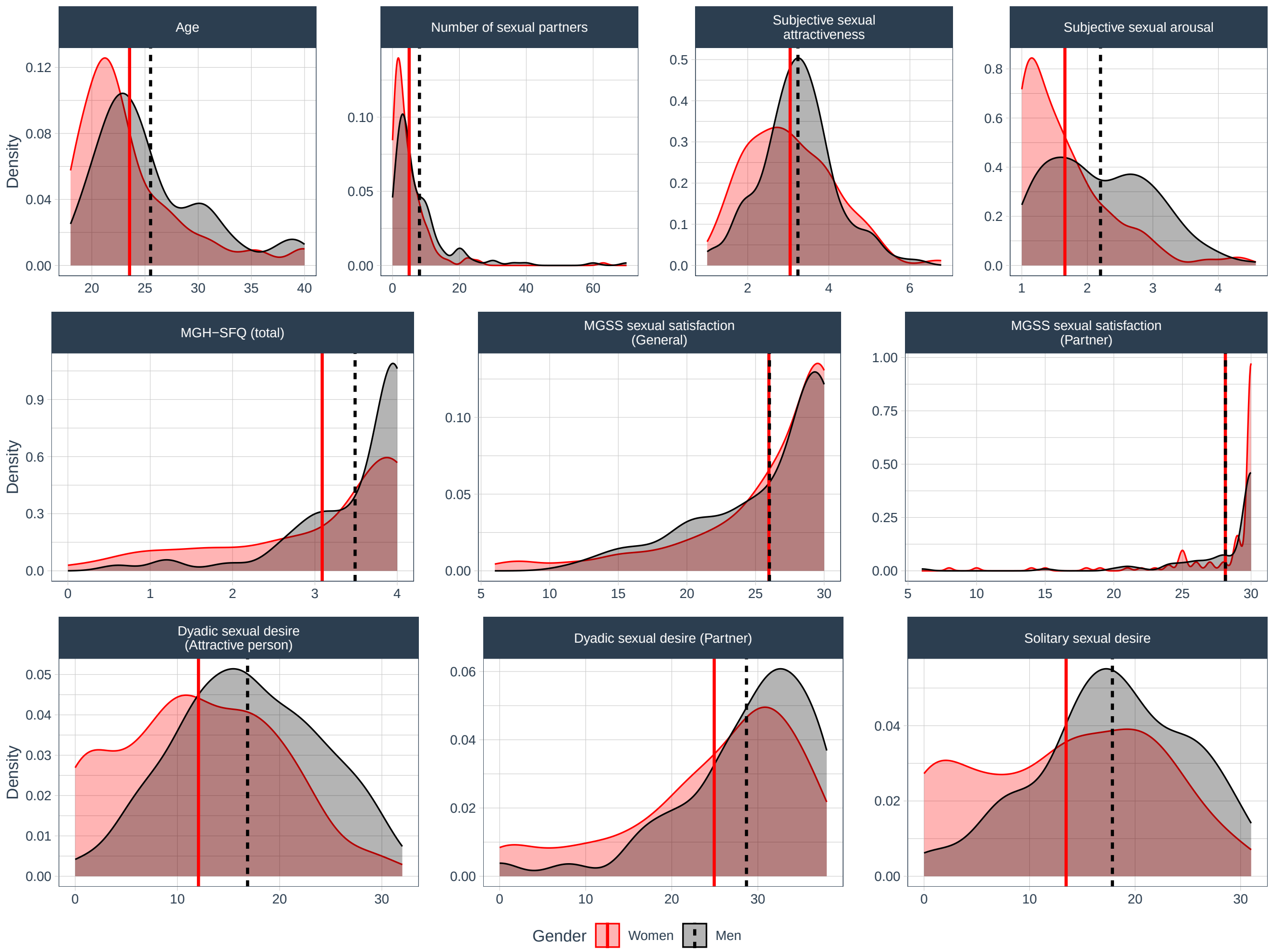

Figure S2. Distribution of measured variables by gender. Coloured vertical lines represent mean values by gender. Detailed descriptives are found in Table S1. Because for *Subjective sexual attractiveness* and *Subjective sexual arousal* there are are multiple within-subject observations, densities calculated from mean values per participant.

2.2 Correlations between measured variables

Correlation between numeric variables for women, men, and all participants combined, are reported in Table S2.

2.2.1 Table S2. Correlations between measured variables

Correlation matrix table.

```
# Correlations for women
dat.corr.W <- dat.desc |>
  ungroup() |>
  filter(Gender == "Women") |>
  select(Age:`Dyadic sexual desire (Partner)`) |>
  corr.stars() |>
  rownames_to_column(var = " ")

# Correlations for men
dat.corr.M <- dat.desc |>
  ungroup() |>
  filter(Gender == "Men") |>
  select(Age:`Dyadic sexual desire (Partner)`) |>
```

```
  corr.stars() |>
  rownames_to_column(var = " ")

# Correlations for all participants combined
dat.corr.All <- dat.desc |>
  ungroup() |>
  select(Age:`Dyadic sexual desire (Partner)`) |>
  corr.stars() |>
  rownames_to_column(var = " ")

# Full formated table
bind_rows(dat.corr.W, dat.corr.M, dat.corr.All) |>
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", rep("c", 9)),
        linesep = "",
        caption = "Correlations between measured variables",
        escape = FALSE) |>
  pack_rows(group_label = "Women",
            start_row = 1, end_row = 10,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "Men",
            start_row = 11, end_row = 20,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "All participants",
            start_row = 21, end_row = 30,
            bold = FALSE,
            background = "lightgray") |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  column_spec(2:10, width = "2.2cm") |>
  footnote(general = paste0("Values represent Pearson correlation coefficients ($r$). ",
                            "For significance, $^{\\\\dagger}p$ < 0.1, *$p$ < 0.05, ",
                            "**$p$ < 0.01, ***$p$ < 0.001. ",
                            "Significant correlations are in bold."),
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE) |>
  landscape()
```

Table S2. *Correlations between measured variables*

	Age	Number of sexual partners	MGH-SFQ (total)	MGSS sexual satisfaction (General)	MGSS sexual satisfaction (Partner)	Subjective sexual attractiveness	Subjective sexual arousal	Solitary sexual desire	Dyadic sexual desire (Attractive person)
Women									
Age									
Number of sexual partners	**0.24****								
MGH-SFQ (total)	-0.05	-0.07							
MGSS sexual satisfaction (General)	**-0.21***	0.02	**0.46*****						
MGSS sexual satisfaction (Partner)	-0.16†	-0.14	**0.32*****	**0.73*****					
Subjective sexual attractiveness	0.11	**0.18***	-0.04	**-0.22***	-0.18†				
Subjective sexual arousal	0.00	**0.17***	-0.13†	-0.18†	-0.16†	**0.54*****			
Solitary sexual desire	-0.14†	**0.28*****	0.05	-0.06	-0.18†	**0.31*****	**0.33*****		
Dyadic sexual desire (Attractive person)	0.06	**0.32*****	**-0.17***	-0.04	-0.17†	**0.34*****	**0.36*****	**0.44*****	
Dyadic sexual desire (Partner)	0.00	**0.21****	**0.43*****	**0.44*****	**0.27****	0.13†	0.04	**0.31*****	0.13†
Men									
Age									
Number of sexual partners	**0.23****								
MGH-SFQ (total)	0.04	0.02							
MGSS sexual satisfaction (General)	**-0.24***	-0.08	**0.36*****						
MGSS sexual satisfaction (Partner)	-0.13	-0.01	0.10	**0.63*****					
Subjective sexual attractiveness	0.10	-0.05	-0.08	-0.10	-0.02				
Subjective sexual arousal	**0.2***	0.07	0.05	-0.14	-0.09	**0.46*****			
Solitary sexual desire	-0.16†	0.00	0.09	0.10	0.17	**0.26****	0.11		
Dyadic sexual desire (Attractive person)	0.12	**0.29*****	0.03	-0.13	-0.08	**0.25****	**0.43*****	**0.25****	
Dyadic sexual desire (Partner)	0.11	0.07	**0.36*****	**0.55*****	**0.22***	0.14	**0.24****	**0.17***	**0.2***
All participants									
Age									
Number of sexual partners	**0.26*****								
MGH-SFQ (total)	0.02	0.01							
MGSS sexual satisfaction (General)	**-0.22****	-0.03	**0.42*****						
MGSS sexual satisfaction (Partner)	**-0.14***	-0.07	**0.24*****	**0.69*****					
Subjective sexual attractiveness	**0.12***	0.08	-0.03	**-0.18***	-0.12				
Subjective sexual arousal	**0.15****	**0.17****	0.01	**-0.15***	-0.12†	**0.5*****			
Solitary sexual desire	-0.09	**0.17****	0.11†	0.00	-0.05	**0.31*****	**0.3*****		
Dyadic sexual desire (Attractive person)	**0.14***	**0.33*****	-0.04	-0.07	-0.12†	**0.32*****	**0.45*****	**0.42*****	
Dyadic sexual desire (Partner)	0.08	**0.16****	**0.43*****	**0.46*****	**0.25*****	**0.15****	**0.18****	**0.3*****	**0.21*****

Note: Values represent Pearson correlation coefficients (r). For significance, $^{\dagger}p < 0.1$, $^{*}p < 0.05$, $^{**}p < 0.01$, $^{***}p < 0.001$. Significant correlations are in bold.

2.3 Internal consistency

Six variables were calculated from multiple items (1. MGH-SFQ, 2. Dyadic sexual desire (Partner), 3. Solitary sexual desire, 4. Dyadic sexual desire (Attractive person), 5. MGSS sexual satisfaction (General) and 6. MGSS sexual satisfaction (Partner)).

Data by item, for each participant, is included in the following data base, loaded as `dat.reli`:

```
dat.reli <- read_excel("Data/BD_ConsistenciaInterna.xlsx")  |>
  mutate(Sex = recode_factor(Sex,
                             "2" = "Women",
                             "1" = "Men")) |>
  rename(Gender = Sex) |>
  filter(Participante != 122)
```

Participant 122 was excluded because they did not respond the psychological scales.

To measure the internal consistency of these tests, we used standardized Cronbach's alpha (α or Tau-equivalent reliability: ρ_T) coefficients, using the function `cronbach.alpha` from the package `ltm` (Rizopoulos, 2006).

Importantly, given that for MGH-SFQ one item was answered only by men, the internal consistency of this variable was measured independently for each gender.

```
# MGH-SFQ for men
MGH.m <- dat.reli |>
  filter(Gender == "Men" ) |>
  select(3:7) |>
  drop_na() |>
  cronbach.alpha(CI = TRUE, standardized = TRUE)

# MGH-SFQ for women
MGH.w <- dat.reli |>
  filter(Gender == "Women" ) |>
  select(3:5,7) |>
  drop_na() |>
  cronbach.alpha(CI = TRUE, standardized = TRUE)

# Dyadic sexual desire (Partner)
DSD.p <- dat.reli |>
  select(9:13) |>
  drop_na() |>
  cronbach.alpha(CI = TRUE, standardized = TRUE)

# Solitary sexual desire
SSD.p <- dat.reli |>
  select(15:18) |>
  drop_na() |>
  cronbach.alpha(CI = TRUE, standardized = TRUE)

# Dyadic sexual desire (Attractive person)
DSD.a <- dat.reli |>
  select(20:23) |>
  drop_na() |>
  cronbach.alpha(CI = TRUE, standardized = TRUE)

# MGSS sexual satisfaction (General)
MGSS.g <- dat.reli |>
  select(26:30) |>
  drop_na() |>
  cronbach.alpha(CI = TRUE, standardized = TRUE)
```

```r
# MGSS sexual satisfaction (Partner)
MGSS.p <- dat.reli |>
  select(32:36) |>
  drop_na() |>
  cronbach.alpha(CI = TRUE, standardized = TRUE)
```

2.3.1 Table S3. Internal consistency of construct variables

Table of Cronbach's α for construct variables.

```r
# Create table
tibble(Variable = c("MGH-SFQ", "MGH-SFQ",
                    "MGSS sexual satisfaction (General)",
                    "MGSS sexual satisfaction (Partner)",
                    "Dyadic sexual desire (Partner)",
                    "Solitary sexual desire",
                    "Dyadic sexual desire (Attractive person)"),
       Gender = c("Men", "Women", rep(" ", 5)),
       p = c(MGH.m$p,
             MGH.w$p,
             MGSS.g$p,
             MGSS.p$p,
             DSD.p$p,
             SSD.p$p,
             DSD.a$p),
       n = c(MGH.m$n,
             MGH.w$n,
             MGSS.g$n,
             MGSS.p$n,
             DSD.p$n,
             SSD.p$n,
             DSD.a$n),
       alpha = c(MGH.m$alpha,
                 MGH.w$alpha,
                 MGSS.g$alpha,
                 MGSS.p$alpha,
                 DSD.p$alpha,
                 SSD.p$alpha,
                 DSD.a$alpha),
       ci2.5 = c(MGH.m$ci[1],
                 MGH.w$ci[1],
                 MGSS.g$ci[1],
                 MGSS.p$ci[1],
                 DSD.p$ci[1],
                 SSD.p$ci[1],
                 DSD.a$ci[1]),
       ci97.5 = c(MGH.m$ci[2],
                  MGH.w$ci[2],
                  MGSS.g$ci[2],
                  MGSS.p$ci[2],
                  DSD.p$ci[2],
                  SSD.p$ci[2],
                  DSD.a$ci[2])) |>
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", "l", rep("c", 5)),
```

```
        linesep = "",
        caption = "Internal consistency of measured variables",
        escape = FALSE,
        col.names = c("Variable", "Gender",
                      "Items",
                      "$n$",
                      "$\\alpha$",
                      "$2.5\\% CI$",
                      "$97.5\\% CI$")) |>
  collapse_rows(columns = 1, valign = "middle") |>
  kable_styling(latex_options = "HOLD_position") |>
  footnote(general = "95\\\\% confidence intervals were calculated with 1,000 bootstrap samples.
           Standardized Cronbach's alpha ($\\\\alpha$) coefficients were computed.
           MGH-SFQ is reported by gender, because one item was answered only by men.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S3. *Internal consistency of measured variables*

Variable	Gender	Items	n	α	$2.5\% CI$	$97.5\% CI$
MGH-SFQ	Men	5	139	0.82	0.72	0.88
	Women	4	181	0.86	0.82	0.90
MGSS sexual satisfaction (General)		5	188	0.92	0.89	0.94
MGSS sexual satisfaction (Partner)		5	187	0.91	0.84	0.95
Dyadic sexual desire (Partner)		5	309	0.90	0.88	0.92
Solitary sexual desire		4	314	0.91	0.89	0.93
Dyadic sexual desire (Attractive person)		4	320	0.89	0.86	0.91

Note: 95% confidence intervals were calculated with 1,000 bootstrap samples. Standardized Cronbach's alpha (α) coefficients were computed. MGH-SFQ is reported by gender, because one item was answered only by men.

2.4 Controlling for Relationship Duration and MGSS Sexual Satisfaction (Partner) in Sexual Desire Dimensions

To **ensure that the three sexual desire dimensions were not influenced by Relationship Duration or MGSS sexual satisfaction (Partner)**, we applied a three-step adjustment process:

1. **Estimating the effects:**
   - We performed separate **linear regressions** where each sexual desire dimension was predicted by **Relationship Duration** and **MGSS sexual satisfaction (Partner)**.
   - This allowed us to quantify how much these external factors influence each dimension.
2. **Evaluating statistical significance:**
   - We conducted **Type III ANOVA** to determine which predictors had a significant effect on each sexual desire dimension.
   - Only **MGSS sexual satisfaction (Partner) significantly predicted Dyadic Sexual Desire (Partner)**.
3. **Removing the effects:**
   - We adjusted **only Dyadic Sexual Desire (Partner)** by extracting the **residuals** from the regression model.
   - These residuals represent the **variation independent of MGSS sexual satisfaction (Partner)** and were then standardized for comparability.

Additionally, **MGSS sexual satisfaction (Partner) was mean-centered** before analysis.

Step 1: Estimating the Effects of Relationship Duration & Partner Satisfaction

```
# Select only participants in stable relationships and ensure required variables are available
dat_ctl <- dat |>
  group_by(Participant) |>
  slice_head() |>
  filter(Relationship == "Stable") |>
  ungroup()

# Fit linear models predicting each dimension of sexual desire
ctl_SSD <- lm(`Solitary sexual desire` ~
                `Relationship duration` + `MGSS sexual satisfaction (Partner)`,
              data = dat_ctl)
ctl_PD <- lm(`Dyadic sexual desire (Partner)` ~
               `Relationship duration` + `MGSS sexual satisfaction (Partner)`,
             data = dat_ctl)
ctl_APD <- lm(`Dyadic sexual desire (Attractive person)` ~
                `Relationship duration` + `MGSS sexual satisfaction (Partner)`,
              data = dat_ctl)
```

Step 2: Displaying ANOVA Results for Each Model

The table below presents Type III ANOVA results for each model. Significant effects indicate that Relationship Duration or Partner Satisfaction meaningfully predict the corresponding sexual desire dimension.

```
# Combine ANOVA results for all models
anova_results <- bind_cols(bind_cols(anova_summary(Anova(ctl_SSD, type = 3)),
                                     epsilon_squared(ctl_SSD)) |>
                             unite(col = "df", DFn:DFd, sep = ", "),
                           bind_cols(anova_summary(Anova(ctl_PD, type = 3)),
                                     epsilon_squared(ctl_PD)) |>
                             unite(col = "df", DFn:DFd, sep = ", "),
                           bind_cols(anova_summary(Anova(ctl_APD, type = 3)),
                                     epsilon_squared(ctl_APD)) |>
                             unite(col = "df", DFn:DFd, sep = ", ")) |>
  select(-starts_with(c("p<.05", "ges...", "Parameter...", "CI"))) |>  # Remove Sum of Squares columns
  mutate(across(starts_with("p..."), pval.lev)) |> # Format p-values
  rename(Effect = Effect...1) |>
  select(-starts_with("Effect...")) |>
  mutate_at("Effect", str_replace_all, "`", "")

# Create a formatted table
anova_results |>
  kable(booktabs = TRUE,
        align = c("l", rep("c", 9)),  # Align columns (left for first, center for the rest)
        digits = 3,
        caption = "Effects of relationship duration and MGSS sexual satisfaction (Partner) in
                   sexual desire dimensions",
        col.names = c("Effect", rep(c("$df$", "$F$", "$p$", "$\\epsilon^2_p$"), times = 3)),
        escape = FALSE) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  add_header_above(c(" " = 1,
                     "Solitary sexual desire" = 4,
                     "Dyadic sexual desire\n(Partner)" = 4,
                     "Dyadic sexual desire\n(Attractive person)" = 4)) |>
  footnote(general = "As effect size, we report partial epsilon squared
                     ($\\\\epsilon^2_p$), which provides a less biases
```

```
                    estimate than $\\\\eta^2$ (see
                    \\\\cite{albersWhenPowerAnalyses2018}).
                    Significant effects are in bold.",
          threeparttable = TRUE,
          footnote_as_chunk = TRUE,
          escape = FALSE)
```

Table S4. *Effects of relationship duration and MGSS sexual satisfaction (Partner) in sexual desire dimensions*

	Solitary sexual desire				Dyadic sexual desire (Partner)				Dyadic sexual desire (Attractive person)			
Effect	*df*	*F*	*p*	ϵ_p^2	*df*	*F*	*p*	ϵ_p^2	*df*	*F*	*p*	ϵ_p^2
Relationship duration	3, 165	0.482	0.70	0	3, 165	2.081	0.1	0.041	3, 165	0.095	0.96	0
MGSS sexual satisfaction (Partner)	1, 165	0.029	0.86	0	1, 165	8.875	**0.003**	0.045	1, 165	0.884	0.35	0

Note: As effect size, we report partial epsilon squared (ϵ_p^2), which provides a less biases estimate than η^2 (see Albers and Lakens, 2018). Significant effects are in bold.

Step 3: Controlling Scores Based on ANOVA Results

From the ANOVA results, only the effect of MGSS sexual satisfaction (Partner) on Dyadic sexual desire (Partner) was significant. Thus, only Dyadic Sexual Desire (Partner) scores were adjusted, while the other dimensions remained unchanged.

```
# Prepare dataset with necessary variables, removing missing values
dat_tl_PD_fin <- dat_ctl |>
  select(Participant, `Dyadic sexual desire (Partner)`,
         `MGSS sexual satisfaction (Partner)`) |>
  drop_na()

# Fit a model to predict Dyadic Sexual Desire (Partner) using partner satisfaction only
ctl_PD_fin <- lm(`Dyadic sexual desire (Partner)` ~ `MGSS sexual satisfaction (Partner)`,
                 data = dat_tl_PD_fin)

# Adjust the Dyadic Sexual Desire (Partner) scores by replacing them with their residuals
dat_ctl <- dat_tl_PD_fin |>
  mutate(`Dyadic sexual desire (Partner)` =
           mean(`Dyadic sexual desire (Partner)`) + resid(ctl_PD_fin))

# Update the original dataset with the adjusted scores
dat <- dat |>
  mutate(`Dyadic sexual desire (Partner)` = as.numeric(`Dyadic sexual desire (Partner)`)) |>
  rows_update(dat_ctl |> select(-`MGSS sexual satisfaction (Partner)`),
              by = "Participant", unmatched = "ignore")
```

---

3 Hypothesis tests

3.1 Hypothesis 1: All dimensions of trait sexual desire (TSD) will be higher in men than in women, and the differences will be stronger or weaker according to relationship status

We tested whether relationship type and gender interact as predictors of sexual desire (H1a: Solitary TSD; H1b: Dyadic TSD toward an attractive person; H1c: Dyadic TSD toward a partner). To examine this hypothesis, we modeled the effects of relationship type and gender on each of the three TSD scores.

However, models using the original TSD scores did not meet the assumption of normally distributed residuals. To address this, we applied an ordered normalization transformation to each TSD variable. We then fitted and compared models predicting both the original (as a proportion, to make scores comparable) and transformed (normalized) TSD dimensions. In all three cases, models using the normalized variables provided a better fit, so all inferences are based on these models.

3.1.1 Data

A data frame was created with one row per participant, where sexual desire variables were normalized as proportions. An ordered quantile normalization transformation (Peterson & Cavanaugh, 2020) was then applied using the `orderNorm` function from the `bestNormalize` package (Peterson, 2021), and the transformed values were added as new variables.

```
# Process the dataset and create transformed variables
dat_m1 <- dat |>
  # Group the data by participant
  group_by(Participant) |>
  # Select only the first (top) observation for each participant
  slice_head() |>
  # Remove the grouping structure to avoid unintended behavior in later operations
  ungroup() |>
  # Create new proportion variables by normalizing each sexual desire measure
  mutate("Solitary sexual desire (proportion)" =
           `Solitary sexual desire` / 31,
         "Dyadic sexual desire: Attractive person (proportion)" =
           `Dyadic sexual desire (Attractive person)` / 32,
         "Dyadic sexual desire: Partner (proportion)" =
           `Dyadic sexual desire (Partner)` / 38)

# Apply ordered normalization transformations to the proportion variables
trs_SSD <- orderNorm(dat_m1$`Solitary sexual desire (proportion)`)
trs_DSDat <- orderNorm(dat_m1$`Dyadic sexual desire: Attractive person (proportion)`)
trs_DSDpt <- orderNorm(dat_m1$`Dyadic sexual desire: Partner (proportion)`)

# Add the transformed variables back into the dataset
dat_m1 <- dat_m1 |>
  mutate("Solitary sexual desire (normalized)" =
          predict(trs_SSD),  # Transformed solitary sexual desire
         "Dyadic sexual desire: Attractive person (normalized)" =
           predict(trs_DSDat),  # Transformed dyadic sexual desire (attractive person)
         "Dyadic sexual desire: Partner (normalized)" =
           predict(trs_DSDpt))  # Transformed dyadic sexual desire (partner)
```

3.1.2 Hypothesis 1a: Solitary TSD

3.1.2.1 Model the effects of relationship type and gender on Solitary TSD We fitted models with both the original (proportion; `m1a_prop`) and transformed (normalized; `m1a_norm`) TSD scores, and performed posterior predictive checks (PPCs). As shown elsewhere (e.g., Gabry et al., 2019), if simulated data from one model are more similar to the observed outcome, that model is likely to be preferred.

```
m1a_prop <- lm(`Solitary sexual desire (proportion)` ~ Gender * Relationship,
            data = dat_m1)

m1a_norm <- lm(`Solitary sexual desire (normalized)` ~ Gender * Relationship,
           data = dat_m1)
```

3.1.2.1.1 Figure S3: Posterior predictive checks (PPCs) for Hypothesis 1a. PPCs were performed using the `check_model` function from the `performance` package (Lüdecke et al., 2021), and reported in Fig. S3.

Simulated data from the normalized Solitary TSD model (Fig. S3b) are more similar to the observed outcome, so this model is preferred.

```r
ppc_m1a <- ggarrange(plot(check_model(m1a_prop,
                                 panel = FALSE,
                                 check = "pp_check")$PP_CHECK,
                          colors = c("red", "grey30")) +
                       labs(title = NULL, subtitle = NULL) +
                       theme_tq() +
                       facet_wrap(~ 1, labeller = as_labeller(c(
                         "1" = "Original (proportion) Solitary TSD"))),
                     plot(check_model(m1a_norm,
                                 panel = FALSE,
                                 check = "pp_check")$PP_CHECK,
                          colors = c("red", "grey30")) +
                       labs(title = NULL, subtitle = NULL) +
                       theme_tq() +
                       facet_wrap(~ 1, labeller = as_labeller(c(
                         "1" = "Transformed (normalized) Solitary TSD"))),
                     labels = "auto",
                     common.legend = TRUE,
                     legend = "bottom")
ppc_m1a
```

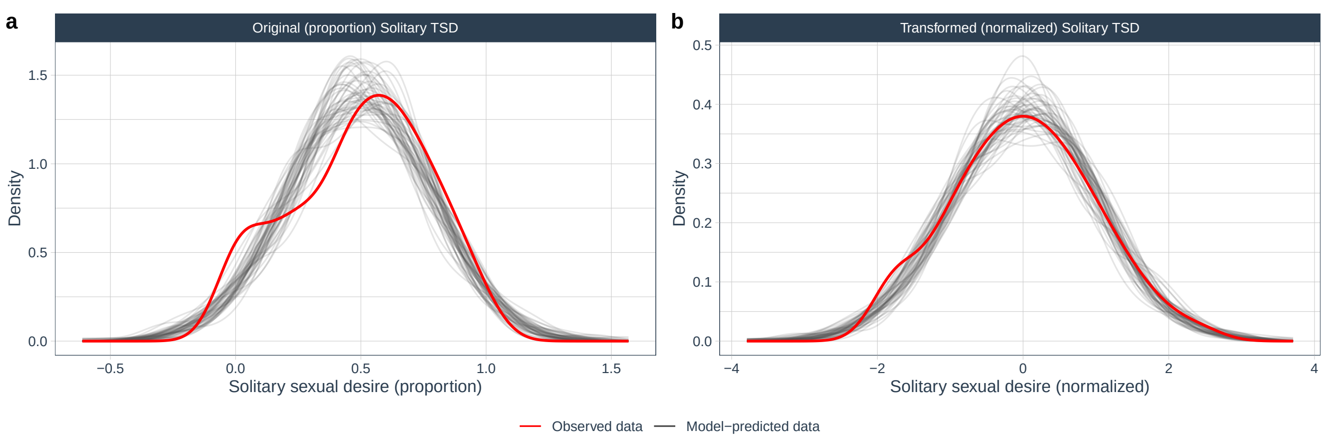

Figure S3. Posterior predictive check. **(a)** Original (proportion) Solitary TSD; **(b)** Transformed (normalized) Solitary TSD. In both panels, red lines represent the observed data, and thin black lines represent 50 iterations of simulated data from each model.

3.1.2.2 Table S5. ANOVA-type table for the interaction between `Relationship type`, and `Gender`

This tables summarizes the results of the model.

```r
anova.sig.lm(model = m1a_norm, custom_caption = "Effects of relationship type and gender on
          solitary sexual desire")
```

Table S5. *Effects of relationship type and gender on solitary sexual desire*

Effect	*df*	*F*	*p*	ϵ_p^2
Gender	1, 319	22.42	$<$ **0.0001**	0.06
Relationship	1, 319	14.07	$<$ **0.001**	0.03
Gender × Relationship	1, 319	4.23	**0.04**	0.01

Note:
Sexual desire was transformed using an ordered quantile normalization (Peterson and Cavanaugh, 2020). Results are type III ANOVA. $R^2 = 0.103$, $R^2_{adjusted} = 0.095$. Gender = participants gender (women, men); Relationship = relationship type (stable, single). As effect size, we report partial epsilon squared (ϵ_p^2), which provides a less biases estimate than η^2 (see Albers and Lakens, 2018). Significant effects are in bold.

3.1.2.3 *Post-hoc* comparisons

Because the main effects of gender, relationship type, and their interaction are significant, we explored these effects using estimated marginal means.

3.1.2.3.1 Table S6. Estimated marginal means and contrasts between participants' gender.

Table of estimated marginal means and contrasts between genders. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1a1 <- emmeans(m1a_norm, ~ Gender)

emms.m1a1.tab <- tibble(data.frame(emms.m1a1))

t.m1a1 <- contr.stars(emms.m1a1) |>
  mutate(p.value = pval.lev(p.value))

merge(emms.m1a1.tab, t.m1a1, by = 0, all = TRUE) |>
  select(-c(1,15)) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", " ") |>
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", rep("c", 5), "l", rep("c", 5)),
        linesep = "",
        caption = "Estimated marginal means and contrasts between participants' gender",
        col.names = c("Gender",
                      "EMM",
                      "$SE$",
                      "$df$",
                      "$2.5\\% CI$",
                      "$97.5\\% CI$",
                      "Contrast",
                      "Difference",
                      "$SE$",
                      "$df$",
                      "$t$",
                      "$p$"),
        escape = FALSE) |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
```

```
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S6. *Estimated marginal means and contrasts between participants' gender*

						Contrasts					
Gender	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Women	-0.17	0.07	319	-0.30	-0.03	Women - Men	-0.46	0.1	319	-4.36	$<$ **0.0001**
Men	0.29	0.08	319	0.13	0.44						

Note: Significant effects are in bold.

3.1.2.3.2 Table S7. Estimated marginal means and contrasts between relationship status. Table of estimated marginal means and contrasts between relationship status. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1a2 <- emmeans(m1a_norm, ~ Relationship)

emms.m1a2.tab <- tibble(data.frame(emms.m1a2))

t.m1a2 <- contr.stars(emms.m1a2) |>
  mutate(p.value = pval.lev(p.value))

merge(emms.m1a2.tab, t.m1a2, by = 0, all = TRUE) |>
  select(-c(1,15)) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", " ") |>
  kable(digits = 2,
          booktabs = TRUE,
          align = c("l", rep("c", 5), "l", rep("c", 5)),
          linesep = "",
          caption = "Estimated marginal means and contrasts between relationship status",
          col.names = c("Relationship type",
                         "EMM",
                         "$SE$",
                         "$df$",
                         "$2.5\\% CI$",
                         "$97.5\\% CI$",
                         "Contrast",
                         "Difference",
                         "$SE$",
                         "$df$",
                         "$t$",
                         "$p$"),
          escape = FALSE) |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S7. *Estimated marginal means and contrasts between relationship status*

						Contrasts					
Relationship type	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Stable	-0.09	0.07	319	-0.23	0.05	Stable - Single	-0.3	0.1	319	-2.89	**0.0041**
Single	0.21	0.08	319	0.06	0.36						

Note: Significant effects are in bold.

3.1.2.3.3 Table S8. Estimated marginal means and contrasts between gender by relationship status. Table of estimated marginal means and contrasts between gender by relationship status. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1a3 <- emmeans(m1a_norm, ~ Gender | Relationship)

emms.m1a3.tab <- tibble(data.frame(emms.m1a3))

t.m1a3 <- contr.stars(emms.m1a3) |>
  mutate(p.value = pval.lev(p.value))

t.m1a3.f <- t.m1a3 |>
  insertRows(2, new = NA) |>
insertRows(4, new = NA)

merge(emms.m1a3.tab, t.m1a3.f, by = 0, all = TRUE) |>
  select(-c(1,3,11,17)) |>
  drop_na(Gender) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", "") |>
  kable(digits = 2,
          booktabs = TRUE,
          align = c("l", "l", rep("c", 5), "l", rep("c", 5)),
          linesep = "",
          caption = "Estimated marginal means and contrasts between gender by
                    relationship status",
          col.names = c("Gender",
                        #"Relationship",
                        "EMM",
                        "$SE$",
                        "$df$",
                        "$2.5\\% CI$",
                        "$97.5\\% CI$",
                        "Contrast",
                        "Difference",
                        "$SE$",
                        "$df$",
                        "$t$",
                        "$p$"),
          escape = FALSE) |>
  pack_rows(group_label = "Relationship status: Stable",
            start_row = 1,
            end_row = 2,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "Relationship status: Single",
            start_row = 3,
            end_row = 4,
```

```
          bold = FALSE,
          background = "lightgray") |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S8. *Estimated marginal means and contrasts between gender by relationship status*

						Contrasts					
Gender	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Relationship status: Stable											
Women	-0.43	0.09	319	-0.61	-0.25	Women - Men	-0.67	0.14	319	-4.74	**< 0.0001**
Men	0.24	0.11	319	0.03	0.46						
Relationship status: Single											
Women	0.09	0.10	319	-0.11	0.30	Women - Men	-0.24	0.15	319	-1.57	0.12
Men	0.33	0.11	319	0.11	0.55						

Note: Significant effects are in bold.

3.1.2.4 Figure S4. Effects of gender and relationship type on solitary sexual desire

This figure summarizes the results of hypothesis 1a.

```
# Gender main effect
h1a1 <- ggplot(dat_m1, aes(x = Gender, y = `Solitary sexual desire (normalized)`,
                          color = Gender)) +
  scale_color_manual(values = color.Gender) +
  scale_fill_manual(values = color.Gender) +
  geom_linerange(data = emms.m1a1.tab |>
                   rename("Solitary sexual desire (normalized)" = emmean),
                 mapping = aes(ymin = lower.CL, ymax = upper.CL)) +
  geom_point(data = emms.m1a1.tab |>
                   rename("Solitary sexual desire (normalized)" = emmean),
              position = position_dodge(0.1),
              size = 3) +
  stat_pvalue_manual(t.m1a1,
                     label = "p.signif",
                     y.position = 0.55,
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Relationship main effect
h1a2 <- ggplot(dat_m1, aes(x = Relationship, y = `Solitary sexual desire (normalized)`,
                          color = Relationship)) +
  scale_color_manual(values = color.Relationship) +
  scale_fill_manual(values = color.Relationship) +
  geom_linerange(data = emms.m1a2.tab |>
                   rename("Solitary sexual desire (normalized)" = emmean),
                 mapping = aes(ymin = lower.CL, ymax = upper.CL)) +
  geom_point(data = emms.m1a2.tab |>
                   rename("Solitary sexual desire (normalized)" = emmean),
              position = position_dodge(0.1),
              size = 3) +
```

```
  stat_pvalue_manual(t.m1a2,
                     label = "p.signif",
                     y.position = 0.45,
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Relationship × Gender interaction
h1a3 <- ggplot(dat_m1, aes(x = Gender, y = `Solitary sexual desire (normalized)`,
                           color = Gender)) +
  scale_color_manual(values = color.Gender) +
  scale_fill_manual(values = color.Gender) +
  facet_wrap(~Relationship) +
  geom_linerange(data = emms.m1a3.tab |>
                  rename("Solitary sexual desire (normalized)" = emmean),
                mapping = aes(ymin = lower.CL, ymax = upper.CL)) +

  geom_point(data = emms.m1a3.tab |>
                  rename("Solitary sexual desire (normalized)" = emmean),
             position = position_dodge(0.1),
             size = 3) +

  stat_pvalue_manual(t.m1a3,
                     label = "p.signif",
                     y.position = 0.7,
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Full figure for hypothesis 1 (a, b and c)
p1a <- ggarrange(h1a1, h1a2, h1a3,
                 ncol = 3,
                 labels = "auto",
                 widths = c(1,1,1.5))
p1a
```

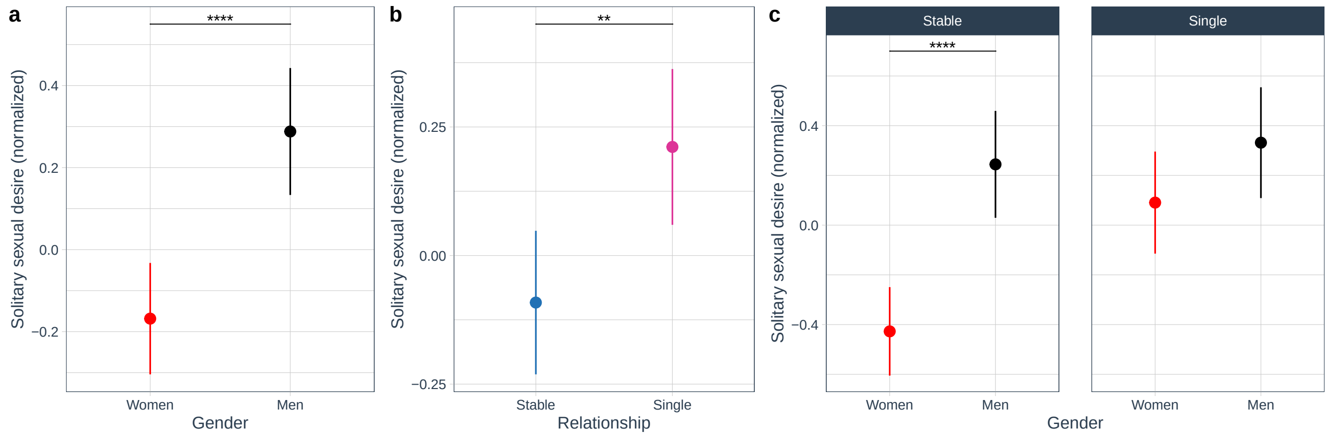

Figure S4. Effects of gender and relationship type on solitary sexual desire. Solitary sexual desire was transformed using ordered quantile normalization (Peterson & Cavanaugh, 2020). **(a)** Simple comparison between sexual desire by gender (for detailed results, see Table S6); **(b)** Simple comparison between relationship status levels (for detailed results, see Table S7); **(c)** Interaction between relationship type and relationship status (see Table S5; for detailed results, see Table S8). Dots and bars represent estimated marginal means and 95% CI. In all cases, significant effects are represented with lines and stars: *$p < 0.05$, **$p < 0.01$, ***$p < 0.001$, ****$p < 0.0001$.

3.1.3 Hypothesis 1b: Dyadic TSD (Attractive person)

3.1.3.1 Model the effects of relationship type and gender on Dyadic TSD: Attractive person We fitted models with both the original (proportion; `m1b_prop`) and transformed (normalized; `m1b_norm`) TSD scores, and performed posterior predictive checks (PPCs). As shown elsewhere (e.g., Gabry et al., 2019), if simulated data from one model are more similar to the observed outcome, that model is likely to be preferred.

```
options(contrasts = c("contr.sum","contr.poly"))
m1b_prop <- lm(`Dyadic sexual desire: Attractive person (proportion)` ~ Gender * Relationship,
            data = dat_m1)

m1b_norm <- lm(`Dyadic sexual desire: Attractive person (normalized)` ~ Gender * Relationship,
           data = dat_m1)
```

3.1.3.1.1 Figure S5: Posterior predictive checks (PPCs) for Hypothesis 1b. PPCs were performed using the `check_model` function from the `performance` package (Lüdecke et al., 2021), and reported in Fig. S5. Simulated data from the normalized Solitary TSD model (Fig. S5b) are more similar to the observed outcome, so this model is preferred.

```
ppc_m1b <- ggarrange(plot(check_model(m1b_prop,
                                      panel = FALSE,
                                      check = "pp_check")$PP_CHECK,
                          colors = c("red", "grey30")) +
                       labs(title = NULL, subtitle = NULL) +
                       theme_tq() +
                       facet_wrap(~ 1, labeller = as_labeller(c(
                         "1" = "Original (proportion) Dyadic TSD: Attractive person"))),
                     plot(check_model(m1b_norm,
                                      panel = FALSE,
                                      check = "pp_check")$PP_CHECK,
                          colors = c("red", "grey30")) +
                       labs(title = NULL, subtitle = NULL) +
                       theme_tq() +
                       facet_wrap(~ 1, labeller = as_labeller(c(
                         "1" = "Transformed (normalized) Dyadic TSD: Attractive person"))),
                     labels = "auto",
```

```
                            common.legend = TRUE,
                            legend = "bottom")
ppc_m1b
```

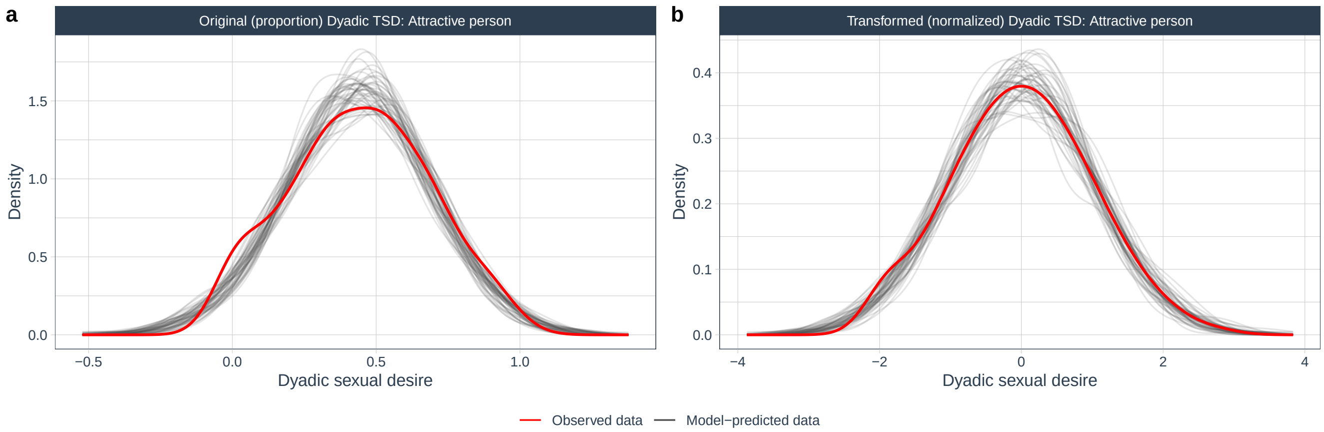

Figure S5. Posterior predictive check. **(a)** Original (proportion) Solitary TSD; **(b)** Transformed (normalized) Solitary TSD. In both panels, red lines represent the observed data, and thin black lines represent 50 iterations of simulated data from each model.

3.1.3.2 Table S9. ANOVA-type table for the interaction between `Relationship type`, and `Gender`

This tables summarizes the results of the model.

```
anova.sig.lm(model = m1b_norm, custom_caption = "Effects of relationship type and gender on
         Dyadic sexual desire: Attractive person")
```

Table S9. *Effects of relationship type and gender on Dyadic sexual desire: Attractive person*

Effect	*df*	*F*	*p*	ϵ_p^2
Gender	1, 319	29.85	**< 0.0001**	0.09
Relationship	1, 319	8.20	**0.004**	0.03
Gender × Relationship	1, 319	1.73	0.19	0.00

Note:
Sexual desire was transformed using an ordered quantile normalization (Peterson and Cavanaugh, 2020). Results are type III ANOVA. $R^2 = 0.122$, $R^2_{adjusted} = 0.114$. Gender = participants gender (women, men); Relationship = relationship type (stable, single). As effect size, we report partial epsilon squared (ϵ_p^2), which provides a less biases estimate than η^2 (see Albers and Lakens, 2018). Significant effects are in bold.

3.1.3.3 *Post-hoc* comparisons

Because the main effects of gender and relationship type, but not their interaction, are significant, we explored these effects using estimated marginal means.

3.1.3.3.1 Table S10. Estimated marginal means and contrasts between participants' gender.

Table of estimated marginal means and contrasts between genders. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1b1 <- emmeans(m1b_norm, ~ Gender)

emms.m1b1.tab <- tibble(data.frame(emms.m1b1))
```

```
t.m1b1 <- contr.stars(emms.m1b1) |>
  mutate(p.value = pval.lev(p.value))

merge(emms.m1b1.tab, t.m1b1, by = 0, all = TRUE) |>
  select(-c(1,15)) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", " ") |>
  kable(digits = 2,
          booktabs = TRUE,
          align = c("l", rep("c", 5), "l", rep("c", 5)),
          linesep = "",
          caption = "Estimated marginal means and contrasts between participants' gender",
          col.names = c("Gender",
                        "EMM",
                        "$SE$",
                        "$df$",
                        "$2.5\\% CI$",
                        "$97.5\\% CI$",
                        "Contrast",
                        "Difference",
                        "$SE$",
                        "$df$",
                        "$t$",
                        "$p$"),
          escape = FALSE) |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S10. *Estimated marginal means and contrasts between participants' gender*

						Contrasts					
Gender	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Women	-0.22	0.07	319	-0.36	-0.09	Women - Men	-0.57	0.1	319	-5.46	$<$ **0.0001**
Men	0.35	0.08	319	0.19	0.50						

Note: Significant effects are in bold.

3.1.3.3.2 Table S11. Estimated marginal means and contrasts between relationship status.

Table of estimated marginal means and contrasts between relationship status. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1b2 <- emmeans(m1b_norm, ~ Relationship)

emms.m1b2.tab <- tibble(data.frame(emms.m1b2))

t.m1b2 <- contr.stars(emms.m1b2) |>
  mutate(p.value = pval.lev(p.value))

merge(emms.m1b2.tab, t.m1b2, by = 0, all = TRUE) |>
  select(-c(1,15)) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", " ") |>
```

```
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", rep("c", 5), "l", rep("c", 5)),
        linesep = "",
        caption = "Estimated marginal means and contrasts between relationship status",
        col.names = c("Relationship type",
                      "EMM",
                      "$SE$",
                      "$df$",
                      "$2.5\\% CI$",
                      "$97.5\\% CI$",
                      "Contrast",
                      "Difference",
                      "$SE$",
                      "$df$",
                      "$t$",
                      "$p$"),
        escape = FALSE) |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S11. *Estimated marginal means and contrasts between relationship status*

						Contrasts					
Relationship type	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Stable	-0.09	0.07	319	-0.22	0.05	Stable - Single	-0.3	0.1	319	-2.86	**0.0045**
Single	0.21	0.08	319	0.06	0.36						

Note: Significant effects are in bold.

3.1.3.3.3 Table S12. Estimated marginal means and contrasts between gender by relationship status.

Table of estimated marginal means and contrasts between gender by relationship status. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1b3 <- emmeans(m1b_norm, ~ Gender | Relationship)

emms.m1b3.tab <- tibble(data.frame(emms.m1b3))

t.m1b3 <- contr.stars(emms.m1b3) |>
  mutate(p.value = pval.lev(p.value))

t.m1b3.f <- t.m1b3 |>
  insertRows(2, new = NA) |>
insertRows(4, new = NA)

merge(emms.m1b3.tab, t.m1b3.f, by = 0, all = TRUE) |>
  select(-c(1,3,11,17)) |>
  drop_na(Gender) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", "") |>
  kable(digits = 2,
        booktabs = TRUE,
```

```
        align = c("l", "l", rep("c", 5), "l", rep("c", 5)),
        linesep = "",
        caption = "Estimated marginal means and contrasts between gender by
                  relationship status",
        col.names = c("Gender",
                      #"Relationship",
                      "EMM",
                      "$SE$",
                      "$df$",
                      "$2.5\\% CI$",
                      "$97.5\\% CI$",
                      "Contrast",
                      "Difference",
                      "$SE$",
                      "$df$",
                      "$t$",
                      "$p$"),
        escape = FALSE) |>
  pack_rows(group_label = "Relationship status: Stable",
            start_row = 1,
            end_row = 2,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "Relationship status: Single",
            start_row = 3,
            end_row = 4,
            bold = FALSE,
            background = "lightgray") |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S12. *Estimated marginal means and contrasts between gender by relationship status*

						Contrasts					
Gender	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Relationship status: Stable											
Women	-0.44	0.09	319	-0.62	-0.26	Women - Men	-0.71	0.14	319	-5.00	**< 0.0001**
Men	0.27	0.11	319	0.05	0.48						
Relationship status: Single											
Women	0.00	0.10	319	-0.21	0.20	Women - Men	-0.43	0.15	319	-2.82	**0.0051**
Men	0.43	0.11	319	0.21	0.65						

Note: Significant effects are in bold.

3.1.3.4 Figure S6. Effects of gender and relationship type on Dyadic sexual desire: Attractive person

This figure summarizes the results of hypothesis 1b.

```
# Gender main effect
h1b1 <- ggplot(dat_m1, aes(x = Gender, y = `Dyadic sexual desire: Attractive person (normalized)`,
                          color = Gender)) +
  scale_color_manual(values = color.Gender) +
  scale_fill_manual(values = color.Gender) +
```

```
  geom_linerange(data = emms.m1b1.tab |>
                rename("Dyadic sexual desire: Attractive person (normalized)" = emmean),
              mapping = aes(ymin = lower.CL, ymax = upper.CL)) +

  geom_point(data = emms.m1b1.tab |>
                rename("Dyadic sexual desire: Attractive person (normalized)" = emmean),
             position = position_dodge(0.1),
             size = 3) +

  stat_pvalue_manual(t.m1b1,
                     label = "p.signif",
                     y.position = 0.6,
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Relationship main effect
h1b2 <- ggplot(dat_m1, aes(x = Relationship, y = `Dyadic sexual desire: Attractive person (normalized)`,
                          color = Relationship)) +
  scale_color_manual(values = color.Relationship) +
  scale_fill_manual(values = color.Relationship) +
  geom_linerange(data = emms.m1b2.tab |>
                rename("Dyadic sexual desire: Attractive person (normalized)" = emmean),
              mapping = aes(ymin = lower.CL, ymax = upper.CL)) +

  geom_point(data = emms.m1b2.tab |>
                rename("Dyadic sexual desire: Attractive person (normalized)" = emmean),
             position = position_dodge(0.1),
             size = 3) +

  stat_pvalue_manual(t.m1b2,
                     label = "p.signif",
                     y.position = 0.45,
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Relationship × Gender interaction
h1b3 <- ggplot(dat_m1, aes(x = Gender, y = `Dyadic sexual desire: Attractive person (normalized)`,
                           color = Gender)) +
  scale_color_manual(values = color.Gender) +
  scale_fill_manual(values = color.Gender) +
  facet_wrap(~Relationship) +
  geom_linerange(data = emms.m1b3.tab |>
                rename("Dyadic sexual desire: Attractive person (normalized)" = emmean),
              mapping = aes(ymin = lower.CL, ymax = upper.CL)) +

  geom_point(data = emms.m1b3.tab |>
                rename("Dyadic sexual desire: Attractive person (normalized)" = emmean),
             position = position_dodge(0.1),
             size = 3) +

  stat_pvalue_manual(t.m1b3,
                     label = "p.signif",
                     y.position = c(0.6, 0.7),
                     tip.length = 0) +
```

```
  guides(color = "none") +
  theme_tq()

# Full figure for hypothesis 1 (a, b and c)
p1b <- ggarrange(h1b1, h1b2, h1b3,
                 ncol = 3,
                 labels = "auto",
                 widths = c(1,1,1.5))
p1b
```

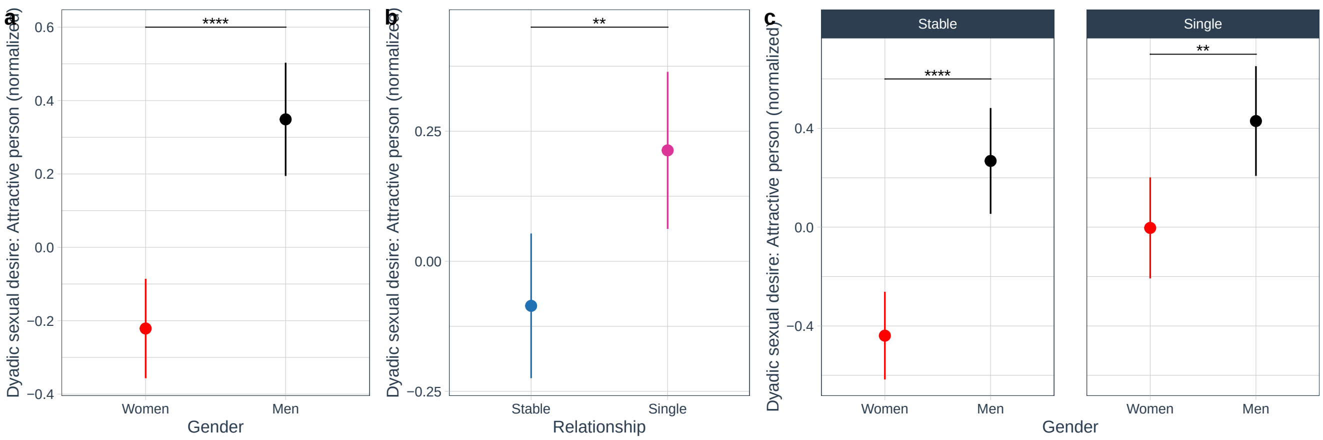

Figure S6. Effects of gender and relationship type on Dyadic sexual desire: Attractive person. Dyadic sexual desire: Attractive person was transformed using ordered quantile normalization (Peterson & Cavanaugh, 2020). **(a)** Simple comparison between sexual desire by gender (for detailed results, see Table S10); **(b)** Simple comparison between relationship status levels (for detailed results, see Table S11); **(c)** Interaction between relationship type and relationship status (see Table S9; for detailed results, see Table S12). Dots and bars represent estimated marginal means and 95% CI. In all cases, significant effects are represented with lines and stars: *$p < 0.05$, **$p < 0.01$, ***$p < 0.001$, ****$p < 0.0001$.

3.1.4 Hypothesis 1c: Dyadic TSD (Partner)

3.1.4.1 Model the effects of relationship type and gender on Dyadic TSD: Partner We fitted models with both the original (proportion; `m1c_prop`) and transformed (normalized; `m1c_norm`) TSD scores, and performed posterior predictive checks (PPCs). As shown elsewhere (e.g., Gabry et al., 2019), if simulated data from one model are more similar to the observed outcome, that model is likely to be preferred.

```
options(contrasts = c("contr.sum","contr.poly"))
m1c_prop <- lm(`Dyadic sexual desire: Partner (proportion)` ~ Gender * Relationship,
            data = dat_m1)

m1c_norm <- lm(`Dyadic sexual desire: Partner (normalized)` ~ Gender * Relationship,
           data = dat_m1)
```

3.1.4.1.1 Figure S7: Posterior predictive checks (PPCs) for Hypothesis 1c. PPCs were performed using the `check_model` function from the `performance` package (Lüdecke et al., 2021), and reported in Fig. S7. Simulated data from the normalized Solitary TSD model (Fig. S7b) are more similar to the observed outcome, so this model is preferred.

```
ppc_m1c <- ggarrange(plot(check_model(m1c_prop,
                                      panel = FALSE,
                                      check = "pp_check")$PP_CHECK,
                                colors = c("red", "grey30")) +
                             labs(title = NULL, subtitle = NULL) +
```

```
                          theme_tq() +
                          facet_wrap(~ 1, labeller = as_labeller(c(
                            "1" = "Original (proportion) Dyadic TSD: Partner"))),
                        plot(check_model(m1c_norm,
                                         panel = FALSE,
                                         check = "pp_check")$PP_CHECK,
                             colors = c("red", "grey30")) +
                          labs(title = NULL, subtitle = NULL) +
                          theme_tq() +
                          facet_wrap(~ 1, labeller = as_labeller(c(
                            "1" = "Transformed (normalized) Dyadic TSD: Partner"))),
                        labels = "auto",
                        common.legend = TRUE,
                        legend = "bottom")
ppc_m1c
```

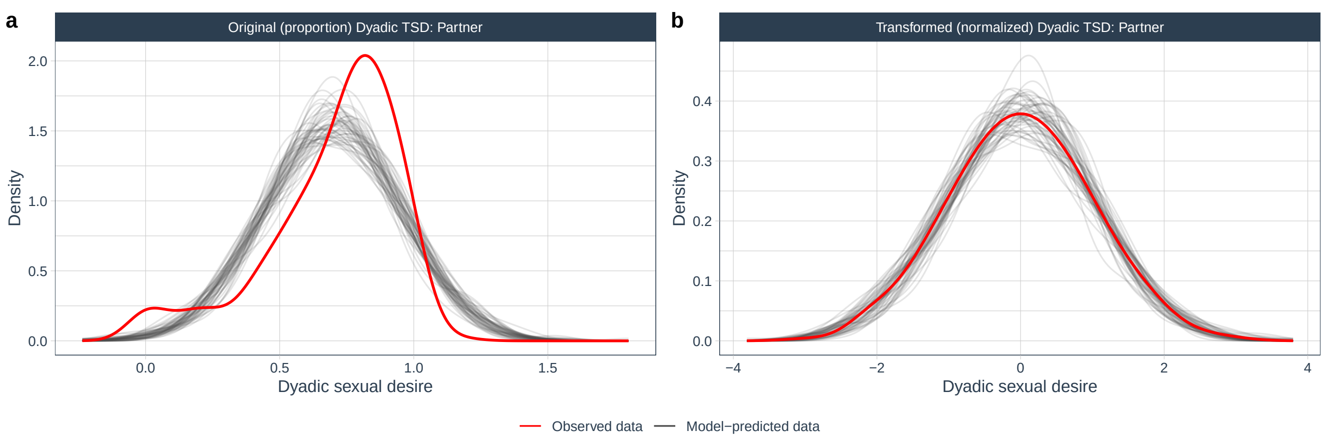

Figure S7. Posterior predictive check. **(a)** Original (proportion) Solitary TSD; **(b)** Transformed (normalized) Solitary TSD. In both panels, red lines represent the observed data, and thin black lines represent 50 iterations of simulated data from each model.

3.1.4.2 Table S13. ANOVA-type table for the interaction between `Relationship type`, and `Gender`

This tables summarizes the results of the model.

```
anova.sig.lm(model = m1c_norm, custom_caption = "Effects of relationship type and gender on
          Dyadic sexual desire: Partner")
```

Table S13. *Effects of relationship type and gender on Dyadic sexual desire: Partner*

Effect	*df*	*F*	*p*	ϵ_p^2
Gender	1, 316	15.49	**< 0.001**	0.0365
Relationship	1, 316	31.60	**< 0.0001**	0.09
Gender × Relationship	1, 316	0.00	0.98	< 0.0001

Note: Sexual desire was transformed using an ordered quantile normalization (Peterson and Cavanaugh, 2020). Results are type III ANOVA. $R^2 = 0.125$, $R^2_{adjusted} = 0.117$. Gender = participants gender (women, men); Relationship = relationship type (stable, single). As effect size, we report partial epsilon squared (ϵ_p^2), which provides a less biases estimate than η^2 (see Albers and Lakens, 2018). Significant effects are in bold.

3.1.4.3 *Post-hoc* comparisons Because the main effects of gender and relationship type, but not their interaction, are significant, we explored these effects using estimated marginal means.

3.1.4.3.1 Table S14. Estimated marginal means and contrasts between participants' gender. Table of estimated marginal means and contrasts between genders. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1c1 <- emmeans(m1c_norm, ~ Gender)

emms.m1c1.tab <- tibble(data.frame(emms.m1c1))

t.m1c1 <- contr.stars(emms.m1c1) |>
  mutate(p.value = pval.lev(p.value))

merge(emms.m1c1.tab, t.m1c1, by = 0, all = TRUE) |>
  select(-c(1,15)) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", " ") |>
  kable(digits = 2,
          booktabs = TRUE,
          align = c("l", rep("c", 5), "l", rep("c", 5)),
          linesep = "",
          caption = "Estimated marginal means and contrasts between participants' gender",
          col.names = c("Gender",
                        "EMM",
                        "$SE$",
                        "$df$",
                        "$2.5\\% CI$",
                        "$97.5\\% CI$",
                        "Contrast",
                        "Difference",
                        "$SE$",
                        "$df$",
                        "$t$",
                        "$p$"),
          escape = FALSE) |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S14. *Estimated marginal means and contrasts between participants' gender*

						Contrasts					
Gender	EMM	SE	df	$2.5\% CI$	$97.5\% CI$	Contrast	Difference	SE	df	t	p
Women	-0.21	0.07	316	-0.35	-0.07	Women - Men	-0.42	0.11	316	-3.94	$<$ **0.001**
Men	0.20	0.08	316	0.05	0.36						

Note: Significant effects are in bold.

3.1.4.3.2 Table S15. Estimated marginal means and contrasts between relationship status. Table of estimated marginal means and contrasts between relationship status. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1c2 <- emmeans(m1c_norm, ~ Relationship)

emms.m1c2.tab <- tibble(data.frame(emms.m1c2))

t.m1c2 <- contr.stars(emms.m1c2) |>
  mutate(p.value = pval.lev(p.value))

merge(emms.m1c2.tab, t.m1c2, by = 0, all = TRUE) |>
  select(-c(1,15)) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", " ") |>
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", rep("c", 5), "l", rep("c", 5)),
        linesep = "",
        caption = "Estimated marginal means and contrasts between relationship status",
        col.names = c("Relationship type",
                      "EMM",
                      "$SE$",
                      "$df$",
                      "$2.5\\% CI$",
                      "$97.5\\% CI$",
                      "Contrast",
                      "Difference",
                      "$SE$",
                      "$df$",
                      "$t$",
                      "$p$"),
        escape = FALSE) |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S15. *Estimated marginal means and contrasts between relationship status*

						Contrasts					
Relationship type	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Stable	0.29	0.07	316	0.15	0.43	Stable - Single	0.6	0.11	316	5.62	**$<$ 0.0001**
Single	-0.30	0.08	316	-0.46	-0.15						

Note: Significant effects are in bold.

3.1.4.3.3 Table S16. Estimated marginal means and contrasts between gender by relationship status. Table of estimated marginal means and contrasts between gender by relationship status. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.m1c3 <- emmeans(m1c_norm, ~ Gender | Relationship)

emms.m1c3.tab <- tibble(data.frame(emms.m1c3))

t.m1c3 <- contr.stars(emms.m1c3) |>
  mutate(p.value = pval.lev(p.value))

t.m1c3.f <- t.m1c3 |>
```

```
  insertRows(2, new = NA) |>
  insertRows(4, new = NA)

merge(emms.m1c3.tab, t.m1c3.f, by = 0, all = TRUE) |>
  select(-c(1,3,11,17)) |>
  drop_na(Gender) |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", "") |>
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", "l", rep("c", 5), "l", rep("c", 5)),
        linesep = "",
        caption = "Estimated marginal means and contrasts between gender by
                  relationship status",
        col.names = c("Gender",
                      #"Relationship",
                      "EMM",
                      "$SE$",
                      "$df$",
                      "$2.5\\% CI$",
                      "$97.5\\% CI$",
                      "Contrast",
                      "Difference",
                      "$SE$",
                      "$df$",
                      "$t$",
                      "$p$"),
        escape = FALSE) |>
  pack_rows(group_label = "Relationship status: Stable",
            start_row = 1,
            end_row = 2,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "Relationship status: Single",
            start_row = 3,
            end_row = 4,
            bold = FALSE,
            background = "lightgray") |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
  footnote(general = "Significant effects are in bold.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S16. *Estimated marginal means and contrasts between gender by relationship status*

						Contrasts					
Gender	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*t*	*p*
Relationship status: Stable											
Women	0.09	0.09	316	-0.09	0.27	Women - Men	-0.41	0.14	316	-2.90	**0.004**
Men	0.50	0.11	316	0.28	0.72						
Relationship status: Single											
Women	-0.51	0.11	316	-0.72	-0.30	Women - Men	-0.42	0.16	316	-2.68	**0.0077**
Men	-0.09	0.11	316	-0.32	0.13						

Note: Significant effects are in bold.

3.1.4.4 Figure S8. Effects of gender and relationship type on Dyadic sexual desire: Partner

This figure summarizes the results of hypothesis 1c.

```
# Gender main effect
h1c1 <- ggplot(dat_m1, aes(x = Gender, y = `Dyadic sexual desire: Partner (normalized)`,
                           color = Gender)) +
  scale_color_manual(values = color.Gender) +
  scale_fill_manual(values = color.Gender) +
  geom_linerange(data = emms.m1c1.tab |>
                   rename("Dyadic sexual desire: Partner (normalized)" = emmean),
                 mapping = aes(ymin = lower.CL, ymax = upper.CL)) +
  geom_point(data = emms.m1c1.tab |>
               rename("Dyadic sexual desire: Partner (normalized)" = emmean),
             position = position_dodge(0.1),
             size = 3) +
  stat_pvalue_manual(t.m1c1,
                     label = "p.signif",
                     y.position = 0.4,
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Relationship main effect
h1c2 <- ggplot(dat_m1, aes(x = Relationship, y = `Dyadic sexual desire: Partner (normalized)`,
                          color = Relationship)) +
  scale_color_manual(values = color.Relationship) +
  scale_fill_manual(values = color.Relationship) +
  geom_linerange(data = emms.m1c2.tab |>
                  rename("Dyadic sexual desire: Partner (normalized)" = emmean),
                mapping = aes(ymin = lower.CL, ymax = upper.CL)) +
  geom_point(data = emms.m1c2.tab |>
                  rename("Dyadic sexual desire: Partner (normalized)" = emmean),
             position = position_dodge(0.1),
             size = 3) +
  stat_pvalue_manual(t.m1c2,
                     label = "p.signif",
                     y.position = 0.5,
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Relationship × Gender interaction
h1c3 <- ggplot(dat_m1, aes(x = Gender, y = `Dyadic sexual desire: Partner (normalized)`,
```

```
                           color = Gender)) +
  scale_color_manual(values = color.Gender) +
  scale_fill_manual(values = color.Gender) +
  facet_wrap(~Relationship) +
  geom_linerange(data = emms.m1c3.tab |>
                   rename("Dyadic sexual desire: Partner (normalized)" = emmean),
                 mapping = aes(ymin = lower.CL, ymax = upper.CL)) +
  geom_point(data = emms.m1c3.tab |>
                   rename("Dyadic sexual desire: Partner (normalized)" = emmean),
              position = position_dodge(0.1),
              size = 3) +
  stat_pvalue_manual(t.m1c3,
                     label = "p.signif",
                     y.position = c(0.8, 0.2),
                     tip.length = 0) +
  guides(color = "none") +
  theme_tq()

# Full figure for hypothesis 1 (a, b and c)
p1c <- ggarrange(h1c1, h1c2, h1c3,
                 ncol = 3,
                 labels = "auto",
                 widths = c(1,1,1.5))
p1c
```

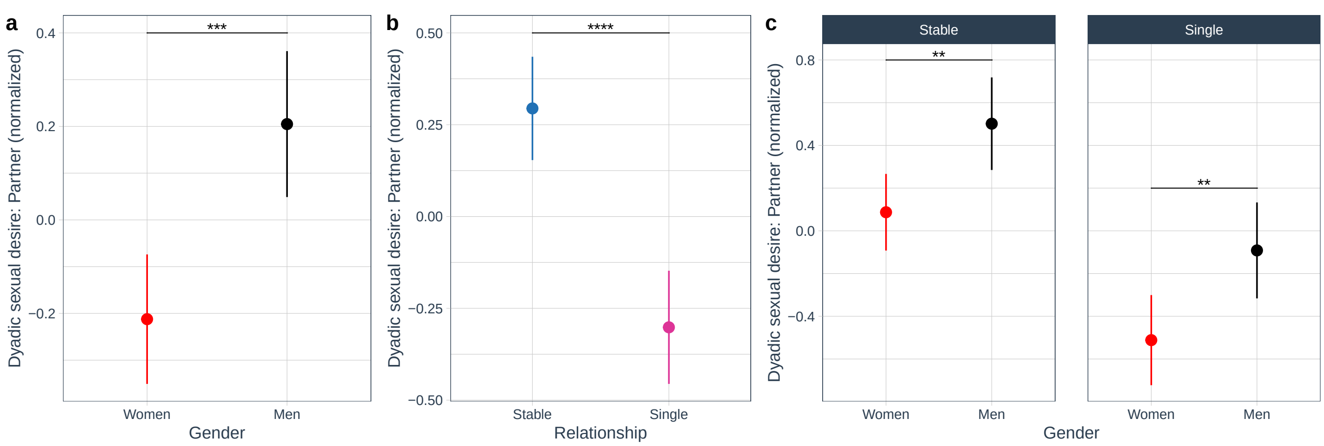

Figure S8. Effects of gender and relationship type on Dyadic sexual desire: Partner. Dyadic sexual desire: Partner was transformed using ordered quantile normalization (Peterson & Cavanaugh, 2020). **(a)** Simple comparison between sexual desire by gender (for detailed results, see Table S14); **(b)** Simple comparison between relationship status levels (for detailed results, see Table S15); **(c)** Interaction between relationship type and relationship status (see Table S13; for detailed results, see Table S16). Dots and bars represent estimated marginal means and 95% CI. In all cases, significant effects are represented with lines and stars: $*p < 0.05$, $**p < 0.01$, $***p < 0.001$, $****p < 0.0001$.

3.2 Data filtering for hypotheses 2 and 3.

To avoid over-complicating the models, first we tested whether the effects of stimuli on sexual arousal were stronger depending on the content of the stimuli (erotic versus non-erotic). This was, in fact, the case.

3.2.1 Table S17. ANOVA-type table for the effects of stimuli content, gender and stimuli content on Subjective sexual arousal

We fitted a linear mixed model with Gender, Stimuli sex, Stimuli content, and their interactions, as fixed effects for Subjective sexual arousal and including, as random effects, random intercepts per stimulus, as well as random intercepts and slopes for the effect of stimuli content.

```
m_stim_cont <- lmer(`Subjective sexual arousal` ~
             Gender * `Stimuli sex` * `Stimuli content` +
             (1 | `Stimuli code`) +
             (1 + `Stimuli content`*`Stimuli sex` | Participant),
           data = dat,
           control = lmerControl(optimizer = "bobyqa"))

anova.sig.lmer(model = m_stim_cont,
               custom_caption = "Effects of relationship type and gender
               on Dyadic sexual desire: Partner")
```

Table S17. *Effects of relationship type and gender on Dyadic sexual desire: Partner*

Effect	*df*	*F*	*p*	ϵ_p^2
Gender	1, 321	42.47	< **0.0001**	0.11
Stimuli sex	1, 447	96.15	< **0.0001**	0.18
Stimuli content	1, 363.12	86.50	< **0.0001**	0.19
Gender × Stimuli sex	1, 321	471.68	< **0.0001**	0.59
Gender × Stimuli content	1, 321	5.02	**0.0257**	0.01
Stimuli sex × Stimuli content	1, 286.22	21.51	< **0.0001**	0.07
Gender × Stimuli sex × Stimuli content	1, 321	116.42	< **0.0001**	0.26

Note: Results are type III ANOVA. $R^2_{conditional} = 0.734$, $R^2_{marginal} = 0.314$. Gender = participants gender (women, men); Stimuli sex = sex of each stimulus (male, female); Stimuli content = content of each stimulus (erotic, non-erotic). As effect size, we report partial epsilon squared (ϵ_p^2), which provides a less biases estimate than η^2 (see Albers and Lakens, 2018). Significant effects are in bold.

The effects of stimuli on sexual arousal were stronger for erotic compared to non-erotic stimuli; to illustrate this, we compared the (within-subject) difference in reported sexual arousal between stimuli sexes, for women and men. This difference was larger when viewing erotic than non-erotic stimuli in both women (erotic: 0.77, non-erotic: 0.57) but especially in men (erotic: 2.75, non-erotic: 1.60; see Table S18 and Fig. S9). Considering this, we tested all predictions of hypotheses 2 and 3 only on responses to erotic stimuli.

3.2.2 Table S18. Estimated marginal means and contrasts between subjective sexual arousal depending on stimuli sex, by stimuli content and participant gender.

Table of estimated marginal means and contrasts between between subjective sexual arousal depending on stimuli sex, by stimuli content and participant gender. All estimated marginal means and contrasts were calculated using the `emmeans` function from the `emmeans` package (Lenth, 2023).

```
emms.stim_cont <- emmeans(m_stim_cont, pairwise ~ `Stimuli sex` | `Stimuli content`  + Gender,
                    adjust = "bonferroni",
                    lmer.df = "satterthwaite")

emms.stim_cont.tab <- tibble(data.frame(emms.stim_cont$emmeans))|>
  rename("Subjective sexual arousal" = emmean,
         "Stimuli content" = Stimuli.content,
         "Stimuli sex" = Stimuli.sex)

t.stim_cont <- contr.stars(emms.stim_cont) |>
```

```
  mutate(p.value = pval.lev(p.value))

t.stim_cont.f <- t.stim_cont |>
  insertRows(2, new = NA) |>
  insertRows(4, new = NA) |>
  insertRows(6, new = NA) |>
  insertRows(8, new = NA)

merge(emms.stim_cont.tab, t.stim_cont.f, by = 0, all = TRUE) |>
  select(-c(1,3,4,12,13,19)) |>
  drop_na("Stimuli sex") |>
  unite(Contrast, group1, group2, sep = " - ") |>
  mutate_at("Contrast", str_replace_all, "NA - NA", " ") |>
  mutate(across(c(df.x, df.y), as.character)) |>
  mutate(across(c(df.x, df.y), str_replace_all, "Inf", "$\\\\infty$")) |>
  kable(digits = 2,
        booktabs = TRUE,
        align = c("l", "l", rep("c", 5), "l", rep("c", 5)),
        linesep = "",
        caption = "Estimated marginal means for the three dimensions of sexual desire by
      relationship status",
        col.names = c("Stimuli sex",
                      "EMM",
                      "$SE$",
                      "$df$",
                      "$2.5\\% CI$",
                      "$97.5\\% CI$",
                      "Contrast",
                      "Difference",
                      "$SE$",
                      "$df$",
                      "$z$",
                      "$p$"),
        escape = FALSE) |>
  pack_rows(group_label = "Gender: Women - Stimuli content: Erotic",
            start_row = 1,
            end_row = 2,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "Gender: Women - Stimuli content: Non-erotic",
            start_row = 3,
            end_row = 4,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "Gender: Men - Stimuli content: Erotic",
            start_row = 5,
            end_row = 6,
            bold = FALSE,
            background = "lightgray") |>
  pack_rows(group_label = "Gender: Men - Stimuli content: Non-erotic",
            start_row = 7,
            end_row = 8,
            bold = FALSE,
            background = "lightgray") |>
  add_header_above(c(" " = 6, "Contrasts" = 6)) |>
  kable_styling(latex_options = c("HOLD_position", "scale_down")) |>
```

```
  footnote(general = "EMM = estimated marginal mean.
           Degrees of freedom ($df$) are asymptotic.
           Bonferroni adjustment was used.",
           threeparttable = TRUE,
           footnote_as_chunk = TRUE,
           escape = FALSE)
```

Table S18. *Estimated marginal means for the three dimensions of sexual desire by relationship status*

						Contrasts					
Stimuli sex	EMM	*SE*	*df*	2.5%*CI*	97.5%*CI*	Contrast	Difference	*SE*	*df*	*z*	*p*
Gender: Women - Stimuli content: Erotic											
Female	1.46	0.10	∞	1.25	1.66	Female - Male	-0.77	0.11	∞	-6.80	$<$ **0.0001**
Male	2.23	0.08	∞	2.08	2.38						
Gender: Women - Stimuli content: Non-erotic											
Female	1.12	0.09	∞	0.94	1.30	Female - Male	-0.57	0.11	∞	-5.27	$<$ **0.0001**
Male	1.69	0.07	∞	1.56	1.82						
Gender: Men - Stimuli content: Erotic											
Female	3.84	0.12	∞	3.61	4.07	Female - Male	2.75	0.13	∞	21.60	$<$ **0.0001**
Male	1.09	0.09	∞	0.92	1.26						
Gender: Men - Stimuli content: Non-erotic											
Female	2.65	0.10	∞	2.45	2.85	Female - Male	1.60	0.12	∞	13.44	$<$ **0.0001**
Male	1.05	0.07	∞	0.91	1.19						

Note: EMM = estimated marginal mean. Degrees of freedom (*df*) are asymptotic. Bonferroni adjustment was used.

3.2.3 Figure S9. Effects of stimuli content (erotic, non-erotic) on subjective sexual arousal

This figure summarizes the results of the model to determine whether the effects of stimuli on sexual arousal were stronger depending on the content of the stimuli (erotic versus non-erotic).

```
# Prepare data for vertical comparison lines between Male and Female
diff_data <- emms.stim_cont.tab %>%
  select(`Stimuli sex`, Gender, `Stimuli content`, `Subjective sexual arousal`) %>%
  pivot_wider(names_from = `Stimuli sex`, values_from = `Subjective sexual arousal`) %>%
  mutate(
    ymin = Male,  # Start of line at Male's mean arousal
    ymax = Female  # End of line at Female's mean arousal
  ) %>%
  mutate(
    # Define custom x positions slightly offset for better readability
    x_pos = rep(c(as.numeric(as.factor(`Stimuli content`[1])) - 0.25,
                  as.numeric(as.factor(`Stimuli content`[2])) + 0.25), 2)
  )

# Create the plot
ggplot(emms.stim_cont.tab, aes(x = `Stimuli sex`, y = `Subjective sexual arousal`,
                               color = `Stimuli content`)) +
  # Separate plots for each Gender
  facet_wrap(~Gender) +
  # Set custom colors for Stimuli content
  scale_color_manual(values = color.Gender) +
  scale_fill_manual(values = color.Gender) +
  # Add confidence interval ranges
  geom_linerange(data = emms.stim_cont.tab,
                 mapping = aes(ymin = asymp.LCL, ymax = asymp.UCL),
```

```r
               position = position_dodge(0.5)) +
  # Add individual data points with position dodge to avoid overlap
  geom_point(data = emms.stim_cont.tab,
             position = position_dodge(0.5),
             size = 3) +
  # Add statistical significance annotations
  stat_pvalue_manual(t.stim_cont,
                     label = "p.signif",
                     y.position = c(2.7, 3, 4.2, 3),  # Adjusted y positions for clarity
                     tip.length = 0,
                     color = "Stimuli content",
                     position = position_dodge(0.5)) +
  # Add vertical dotted lines WITHOUT arrows
  geom_segment(data = diff_data,
               aes(x = x_pos, xend = x_pos,
                   y = ymin, yend = ymax,
                   color = `Stimuli content`),
               linewidth = 0.5,
               linetype = "dotted") +  # Dotted lines
  # Add SOLID arrows separately, with NO line
  geom_segment(data = diff_data,
               aes(x = x_pos, xend = x_pos,
                   y = ymin, yend = ymax,
                   color = `Stimuli content`),
               linetype = "solid",    # Make sure arrows are solid
               linewidth = 0,         # Hide the line itself
               arrow = arrow(length = unit(0.3, "cm"), type = "closed", ends = "both")) +
  # Rotated & Centered Difference Labels on the vertical lines
  geom_text(data = diff_data,
            aes(x = x_pos - 0.06, y = (ymin + ymax) / 2,
                label = abs(round(ymax - ymin, 2)),
                color = `Stimuli content`),
            angle = 90,       # Rotate text vertically
            hjust = 0.5,      # Center horizontally
            vjust = 0.5,      # Center vertically on the line
            size = 2.5) +
  theme_tq()
```

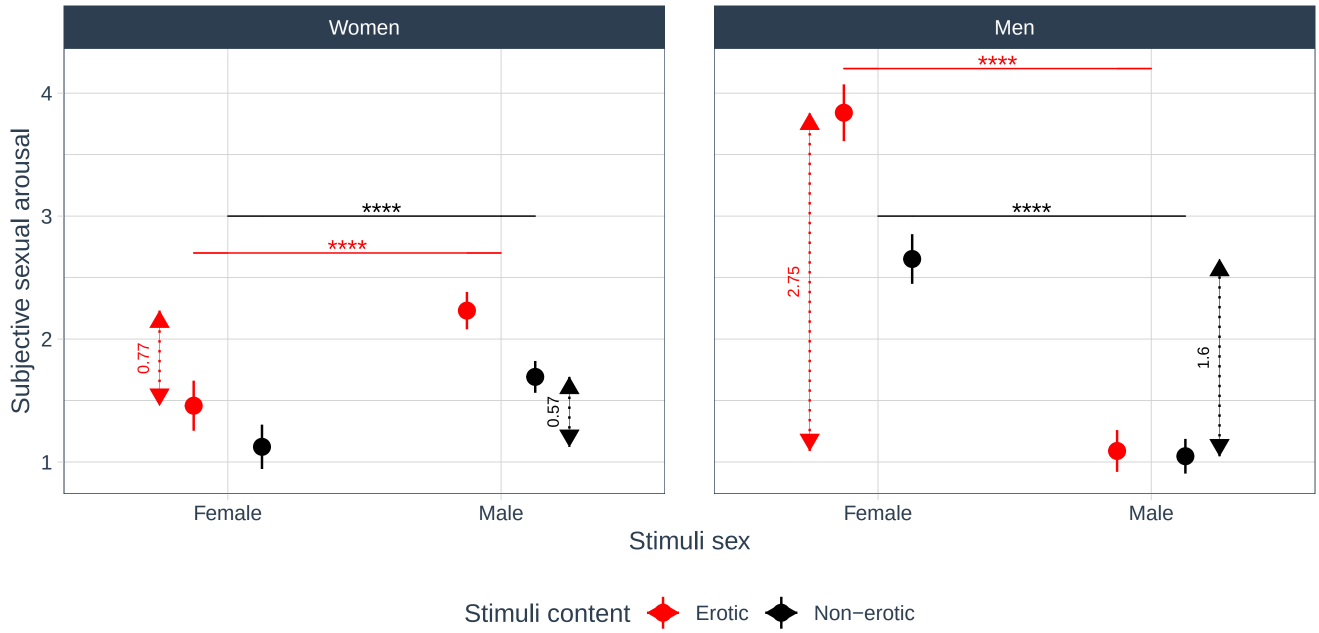

Figure S9. Effects of stimuli content (erotic, non-erotic) on subjective sexual arousal for women's (left panel) and men's (right panel) scores of male and female stimuli (see Table S17; for detailed results, see Table S18). Dots and bars represent estimated marginal means and 95% CI. Vertical lines with arrow heads represent the (absolute) difference in reported subjective sexual arousal for male and femnale stimuli, by stimuli content and gender. In all cases, significant effects are represented with lines and stars: $*p < 0.05$, $**p < 0.01$, $***p < 0.001$, $****p < 0.0001$.

3.3 Hypothesis 2: The association between trait sexual desire (TSD) and subjective sexual arousal (SSA) will vary by TSD dimension, with these associations being gender-specific in men and gender-non-specific in women.

We tested whether the slope of SSA as predictor of sexual desire varies for each of the three dimensions of TSD (H1a: Solitary TSD; H1b: Dyadic TSD toward an attractive person; H1c: Dyadic TSD toward a partner), depending on the gender of the participants and the sex of the stimuli. The prediction was specific for each TSD dimension:

- **H2a**: A significant association between solitary TSD and SSA toward erotic stimuli.
- **H2b**: A significant association between dyadic TSD toward an attractive person and SSA toward erotic stimuli.
- **H2c**: No significant association between dyadic TSD toward a partner and SSA toward erotic stimuli.

To examine this hypothesis, we modeled the effects of each of the three TSD dimension scores, gender, stimulus sex, and their interactions, on SSA. We included random intercepts for each stimulus, as well as random intercepts and slopes between stimuli sex for each participant.

SSA is, however, a discrete variable with only 7 ordered levels. Because of this, we modelled iech hipothesis as a Cumulative Link Mixed Model (CLMM), a Generalized Mixed Model (GLMM) with a Poisson family, and as a linear mixed model (LMM). Because the predictions of the three types of model largely agree and are therefore robust (as shown below), we made inferences from the LMM, as it is the easiest to interpret and the one with the most useful functions available in R to extract information from.

3.3.1 Data

A data frame was created with one row per participant, where sexual desire variables were normalized as proportions. An ordered quantile normalization transformation (Peterson & Cavanaugh, 2020) was then applied using the `orderNorm` function from the `bestNormalize` package (Peterson, 2021), and the transformed values were added as new variables.

```
# Process the dataset and create transformed variables
dat_m2 <- dat |>
  filter(`Stimuli content` == "Erotic")
```

3.3.2 Hypothesis 2a: Solitary TSD

3.3.2.1 Model the effects of relationship type and gender on Solitary TSD We fitted models with both the original (proportion; `m1a_prop`) and transformed (normalized; `m1a_norm`) TSD scores, and performed posterior predictive checks (PPCs) to select the best model. As shown elsewhere (e.g., Gabry et al., 2019), if simulated data from one model are more similar to the observed outcome, that model is likely to be preferred.

```
m2a_lmer <- lmer(`Subjective sexual arousal` ~
               `Solitary sexual desire` * Gender * `Stimuli sex` +
               (1 | `Stimuli code`) +
               (1 + `Solitary sexual desire` | Participant),
             data = dat_m2,
             control = lmerControl(optimizer = "bobyqa"))
```

```
anova.sig.lmer(model = m2a_lmer,
               custom_caption = "Effects of Solitary sexual desire, Stimuli sex, and gender
               on Subjective sexual arousal")
```

```
## Warning: Looks like you are using syntactically invalid variable names, quoted in
##   backticks: `Stimuli sex`. This may result in unexpected behaviour.
##   Please rename your variables (e.g., `Stimuli.sex` instead of `Stimuli
##   sex`) and fit the model again.
## Warning: Looks like you are using syntactically invalid variable names, quoted in
##   backticks: `Stimuli sex`. This may result in unexpected behaviour.
##   Please rename your variables (e.g., `Stimuli.sex` instead of `Stimuli
##   sex`) and fit the model again.
```

Table S19. *Effects of Solitary sexual desire, Stimuli sex, and gender on Subjective sexual arousal*

Effect	*df*	*F*	*p*	ϵ_p^2
Solitary sexual desire	1, 184.93	14.87	**< 0.001**	0.07
Gender	1, 139.14	13.38	**< 0.001**	0.08
Stimuli sex	1, 125.55	208.69	**< 0.0001**	0.62
Solitary sexual desire × Gender	1, 184.93	1.71	0.19	0.0038
Solitary sexual desire × Stimuli sex	1, 25435	0.73	0.39	< 0.0001
Gender × Stimuli sex	1, 25435	2224.73	**< 0.0001**	0.08
Solitary sexual desire × Gender × Stimuli sex	1, 25435	52.89	**< 0.0001**	0.002

Note: Results are type III ANOVA. $R^2_{conditional} = 0.597$, $R^2_{marginal} = 0.334$. Gender = participants gender (women, men); Stimuli sex = sex of each stimulus (male, female); Stimuli content = content of each stimulus (erotic, non-erotic). As effect size, we report partial epsilon squared (ϵ_p^2), which provides a less biases estimate than η^2 (see Albers and Lakens, 2018). Significant effects are in bold.

3.3.2.1.1 Figure S3: Posterior predictive checks (PPCs) for Hypothesis 1a. PPCs were performed using the `check_model` function from the `performance` package (Lüdecke et al., 2021), and reported in Fig. S3. Simulated data from the normalized Solitary TSD model (Fig. S3b) are more similar to the observed outcome, so this model is preferred.

```
ppc_m2a <- ggarrange(plot(check_model(m1a_prop,
                                      panel = FALSE,
                                      check = "pp_check")$PP_CHECK,
                          colors = c("red", "grey30")) +
```

```
                              labs(title = NULL, subtitle = NULL) +
                              theme_tq() +
                              facet_wrap(~ 1, labeller = as_labeller(c(
                                "1" = "Original (proportion) Solitary TSD"))),
                            plot(check_model(m1a_norm,
                                             panel = FALSE,
                                             check = "pp_check")$PP_CHECK,
                                 colors = c("red", "grey30")) +
                              labs(title = NULL, subtitle = NULL) +
                              theme_tq() +
                              facet_wrap(~ 1, labeller = as_labeller(c(
                                "1" = "Transformed (normalized) Solitary TSD"))),
                            labels = "auto",
                            common.legend = TRUE,
                            legend = "bottom")
ppc_m1a
```

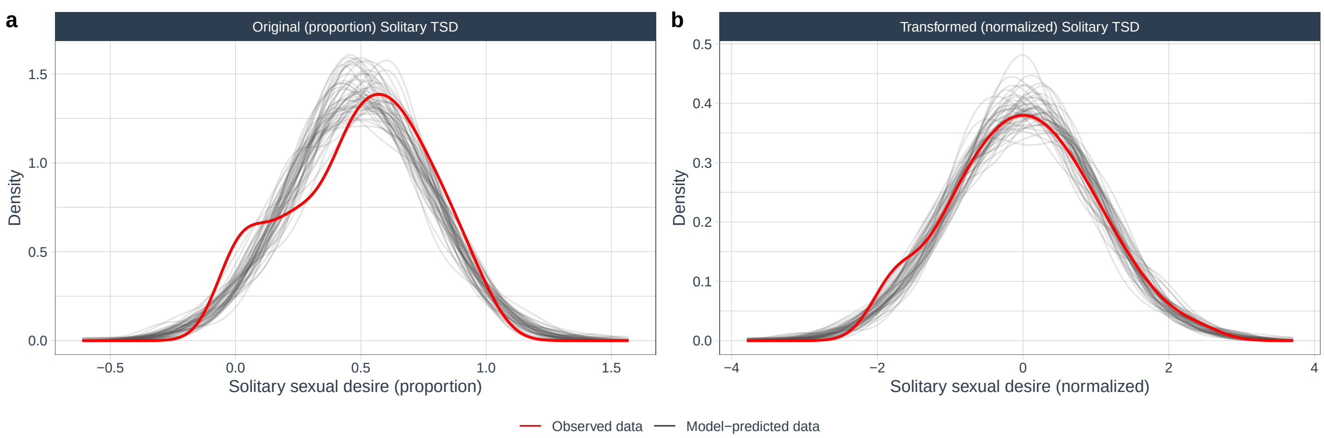

Figure S10. Posterior predictive check. **(a)** Original (proportion) Solitary TSD; **(b)** Transformed (normalized) Solitary TSD. In both panels, red lines represent the observed data, and thin black lines represent 50 iterations of simulated data from each model.

---

4 Session info (for reproducibility)

```
library(pander)
pander(sessionInfo(), locale = FALSE)
```

R version 4.4.2 (2024-10-31)

Platform: x86_64-pc-linux-gnu

attached base packages: *stats*, *graphics*, *grDevices*, *utils*, *datasets*, *methods* and *base*

other attached packages: *pander(v.0.6.5)*, *Hmisc(v.5.2-2)*, *ggeffects(v.2.2.0)*, *effectsize(v.1.0.0)*, *rstatix(v.0.7.2)*, *bestNormalize(v.1.9.1)*, *berryFunctions(v.1.22.5)*, *emmeans(v.1.10.7)*, *lmerTest(v.3.1-3)*, *lme4(v.1.1-36)*, *Matrix(v.1.7-2)*, *scales(v.1.3.0)*, *psych(v.2.4.12)*, *kableExtra(v.1.4.0)*, *performance(v.0.13.0)*, *PerformanceAnalytics(v.2.0.8)*, *quantmod(v.0.4.26)*, *TTR(v.0.24.4)*, *xts(v.0.14.1)*, *zoo(v.1.8-12)*, *tidyquant(v.1.0.10)*, *ggpubr(v.0.6.0)*, *lubridate(v.1.9.4)*, *forcats(v.1.0.0)*, *stringr(v.1.5.1)*, *dplyr(v.1.1.4)*, *purrr(v.1.0.4)*, *readr(v.2.1.5)*, *tidyr(v.1.3.1)*, *tibble(v.3.2.1)*, *ggplot2(v.3.5.1)*, *tidyverse(v.2.0.0)*, *car(v.3.1-3)*, *carData(v.3.0-5)*, *ltm(v.1.2-0)*, *polycor(v.0.8-1)*, *msm(v.1.8.2)*, *MASS(v.7.3-64)*, *readxl(v.1.4.3)* and *knitr(v.1.49)*

loaded via a namespace (and not attached): *rstudioapi(v.0.17.1)*, *datawizard(v.1.0.0)*, *magrittr(v.2.0.3)*, *TH.data(v.1.1-3)*, *estimability(v.1.5.1)*, *farver(v.2.1.2)*, *nloptr(v.2.1.1)*, *rmarkdown(v.2.29)*, *vctrs(v.0.6.5)*, *minqa(v.1.2.8)*, *base64enc(v.0.1-3)*, *butcher(v.0.3.4)*, *htmltools(v.0.5.8.1)*, *curl(v.6.2.0)*, *broom(v.1.0.7)*, *cellranger(v.1.1.0)*, *Formula(v.1.2-5)*, *parallelly(v.1.42.0)*, *htmlwidgets(v.1.6.4)*, *sandwich(v.3.1-1)*, *admisc(v.0.37)*, *lifecycle(v.1.0.4)*, *iterators(v.1.0.14)*, *pkgconfig(v.2.0.3)*, *R6(v.2.5.1)*, *fastmap(v.1.2.0)*, *rbibutils(v.2.3)*, *future(v.1.34.0)*, *digest(v.0.6.37)*, *numDeriv(v.2016.8-1.1)*, *colorspace(v.2.1-1)*, *labeling(v.0.4.3)*, *timechange(v.0.3.0)*, *abind(v.1.4-8)*, *compiler(v.4.4.2)*, *rngtools(v.1.5.2)*, *withr(v.3.0.2)*, *doParallel(v.1.0.17)*, *htmlTable(v.2.4.3)*, *backports(v.1.5.0)*, *ggsignif(v.0.6.4)*, *lava(v.1.8.1)*, *tools(v.4.4.2)*, *foreign(v.0.8-88)*, *RobStatTM(v.1.0.11)*, *future.apply(v.1.11.3)*, *nnet(v.7.3-20)*, *glue(v.1.8.0)*, *quadprog(v.1.5-8)*, *nlme(v.3.1-167)*, *grid(v.4.4.2)*, *checkmate(v.2.3.2)*, *cluster(v.2.1.8)*, *see(v.0.10.0)*, *generics(v.0.1.3)*, *recipes(v.1.1.0)*, *gtable(v.0.3.6)*, *nortest(v.1.0-4)*, *tzdb(v.0.4.0)*, *class(v.7.3-23)*, *data.table(v.1.16.4)*, *hms(v.1.1.3)*, *xml2(v.1.3.6)*, *foreach(v.1.5.2)*, *pillar(v.1.10.1)*, *splines(v.4.4.2)*, *lattice(v.0.22-6)*, *survival(v.3.8-3)*, *tidyselect(v.1.2.1)*, *gridExtra(v.2.3)*, *reformulas(v.0.4.0)*, *bookdown(v.0.42)*, *svglite(v.2.1.3)*, *xfun(v.0.50)*, *expm(v.1.0-0)*, *hardhat(v.1.4.1)*, *timeDate(v.4041.110)*, *stringi(v.1.8.4)*, *yaml(v.2.3.10)*, *boot(v.1.3-31)*, *evaluate(v.1.0.3)*, *codetools(v.0.2-20)*, *cli(v.3.6.3)*, *rpart(v.4.1.24)*, *parameters(v.0.24.1)*, *xtable(v.1.8-4)*, *systemfonts(v.1.2.1)*, *Rdpack(v.2.6.2)*, *munsell(v.0.5.1)*, *Rcpp(v.1.0.14)*, *globals(v.0.16.3)*, *coda(v.0.19-4.1)*, *parallel(v.4.4.2)*, *gower(v.1.0.2)*, *bayestestR(v.0.15.2)*, *doRNG(v.1.8.6.1)*, *listenv(v.0.9.1)*, *viridisLite(v.0.4.2)*, *mvtnorm(v.1.3-3)*, *ipred(v.0.9-15)*, *prodlim(v.2024.06.25)*, *insight(v.1.0.2)*, *rlang(v.1.1.5)*, *cowplot(v.1.1.3)*, *multcomp(v.1.4-28)* and *mnormt(v.2.1.1)*

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