

## No evidence that preferences for facial masculinity track changes in women's hormonal status

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### Abstract

Although widely cited as strong evidence that sexual selection has shaped human facial attractiveness judgments, evidence that preferences for masculine characteristics in men's faces are related to women's hormonal status is equivocal and controversial. Consequently, we conducted the largest ever longitudinal study of the hormonal correlates of women's preferences for facial masculinity (N=584). Analyses showed no evidence that preferences for facial masculinity were related to changes in women's salivary steroid hormone levels. Furthermore, both within-subject and between-subject comparisons showed no evidence that oral contraceptive use decreased masculinity preferences. However, women generally preferred masculinized over feminized versions of men's faces, particularly when assessing men's attractiveness for short-term, rather than long-term, relationships. Our results do not support the hypothesized link between women's preferences for facial masculinity and their hormonal status.

## Introduction

Exaggerated male sex-typical (i.e., masculine) characteristics in men have been proposed as cues of a strong immune system that would be inherited by offspring, but also linked to reduced willingness to invest time and other resources in personal relationships (Gangestad et al., 2004; Gildersleeve et al., 2014; Little et al., 2011; Perrett et al., 1998; Penton-Voak et al., 1999). Given this proposed trade off between the benefits and costs of choosing a masculine mate, researchers have hypothesized that women could maximize the benefits of their mate choices by mating with masculine men when fertile, while forming long-term relationships with relatively feminine men (Gangestad et al., 2004; Gildersleeve et al., 2014; Little et al., 2011; Penton-Voak et al., 1999).

Consistent with this hypothesis, some studies have reported that women show stronger preferences for masculine characteristics in men's faces when in hormonal states associated with high fertility (e.g., during the ovulatory phase of the menstrual cycle and/or when not using hormonal contraceptives, Ditzen et al., 2017; Johnston et al., 2001; Little & Jones, 2012; Little et al., 2002; Little et al., 2013; Penton-Voak et al., 1999; Penton-Voak & Perrett, 2000; Roney & Simmons, 2008; Roney et al., 2011; Vaughn et al., 2010; Welling et al., 2007). These effects are widely cited as evidence that sexual selection has shaped women's judgments of men's facial attractiveness (Gangestad & Simpson, 2000; Grammer et al., 2003; Fink & Penton-Voak, 2002; Thornhill & Gangestad, 1999).

The claim that women's preferences for facial masculinity are related to their hormonal status has been influential. However, it is also highly controversial (see Gildersleeve et al., 2014 and Wood et al., 2014 for meta-analyses drawing opposite conclusions about the robustness of hypothesized links between women's masculinity preferences and hormonal status). In particular, recent work has highlighted four potentially serious methodological problems with research on the hormonal correlates of masculinity preferences.

First, sample sizes are usually small, meaning that studies are badly underpowered (Gangestad et al., 2016). For example, the mean sample size of within-subject studies reporting significant effects of hormonal status on facial masculinity preferences is 40 women (median = 34). Consequently, results from previous studies are difficult to interpret (Blake et al., 2016; Gangestad et al., 2016).

Second, hormonal status is typically assessed using self-reported menstrual cycle data (e.g., number of days since onset of last menses or number of days until expected date of next menses, Harris, 2013; Johnston et al., 2001; Little & Jones, 2012; Munoz-Reyes et al., 2014; Penton-Voak et al., 1999; Penton-Voak & Perrett, 2000; Scott et al., 2014; Zietsch et al., 2015). This method is imprecise and prone to bias (Blake et al., 2016; Gangestad et al., 2016; Harris, 2013).

Third, many studies use between-subject designs. Use of between-subject designs in this research is potentially problematic because, even with large samples, the substantial genetic contribution to individual differences in facial masculinity preferences (Zietsch et al., 2015) could obscure subtle effects of hormonal status. Thus, although several recent studies testing for possible effects of hormonal status on facial masculinity preferences have reported null results (Harris, 2013; Marcinkowska et al., 2016; Munoz-Reyes et al., 2014; Scott et al., 2014; Zietsch et al., 2015), it is noteworthy that these studies all used between-subject designs.

Fourth, studies using within-subject designs typically test women on only two occasions (Johnston et al., 2001; Little & Jones, 2012; Little et al., 2013; Penton-Voak et al., 1999; Roney et al., 2011). This limited approach may not adequately capture complex changes in hormonal status (see, e.g., Roney & Simmons, 2013).

The current study directly addressed all of these potentially serious methodological problems by recruiting 584 heterosexual women for a longitudinal (i.e., within-subject) study in which both women's hormonal status and preferences for masculinity in men's faces were repeatedly assessed (519 women completed at least 5 test sessions, 176 women completed at least 10 test sessions). Changes in women's hormonal status were assessed by measuring steroid hormones from saliva samples and also by tracking within-subject changes in hormonal contraceptive use.

## Methods

### ***Participants***

Five hundred and ninety-eight heterosexual white women who reported that they were either not using any form of hormonal contraceptive (i.e., had natural menstrual cycles) or were using the combined oral contraceptive pill were recruited for the study. Data from 14 of these women were excluded from the dataset because they reported hormonal contraceptive use inconsistently within a single block of test sessions. Thus, the final data set was 584 women (mean age=21.46 years, SD=3.09 years). Participants completed up to three blocks of test sessions (mean time between Block 1 and Block 2 = 230 days; mean time between Block 2 and Block 3 = 487 days). Each of the three blocks of test sessions consisted of five weekly test sessions. Table 1 shows how many women completed one, two, three, four, or five test sessions in Block 1, Block 2, and Block 3.

**Table 1.** The number of women who completed five, four, three, two, or one weekly test sessions in Block 1, Block 2, and Block 3.

	<b>5 sessions</b>	<b>4 sessions</b>	<b>3 sessions</b>	<b>2 sessions</b>	<b>1 session</b>
<b>Block 1</b>	508	22	6	14	26
<b>Block 2</b>	184	3	3	1	4
<b>Block 3</b>	18	0	0	0	0

Forty-five women reported changing their hormonal contraceptive status between blocks during the study. Fifteen women reported changing from using the combined oral contraceptive pill to not using the combined oral

contraceptive pill and 30 women reported changing from not using the combined oral contraceptive pill to using the combined oral contraceptive pill.

### ***Stimuli***

The methods we used to manufacture stimuli to test women's preferences for facial masculinity have been used in many previous studies (e.g., Harris, 2013; Johnston et al., 2001; Little & Jones, 2012; Marcinkowska et al., 2016; Munoz-Reyes et al., 2014; Penton-Voak et al., 1999; Penton-Voak & Perrett, 2000; Scott et al., 2014; Welling et al., 2007; Zietsch et al., 2015). Responses to stimuli manufactured using these methods predict women's actual partner choices (DeBruine et al., 2006). They have also been shown to be very similar to responses to stimuli manufactured using other methods for manipulating sexually dimorphic characteristics in face images (DeBruine et al., 2006). We have made the stimuli from this study publicly available at [osf.io/9b4y7](https://osf.io/9b4y7).

First, we manufactured a female prototype (i.e., average) face by using specialist software (Tiddeman et al., 2001) to average the shape, color, and texture information from images of 50 young white women's faces. A male prototype face was also manufactured in this way by averaging the shape, color, and texture information from images of 50 young white men's faces.



**Figure 1.** Examples of masculinized (left) and feminized (right) versions of men's faces used to assess facial masculinity preferences in our study.

Next, we randomly selected 10 images from the set of 50 individual male faces. We then created a feminized and a masculinized version of each of these 10 male images by adding or subtracting 50% of the linear (i.e., vector) differences in 2D shape between symmetrized versions of the female and male prototypes to (or from) each individual image. This process created 10 pairs of face images in total, with each pair consisting of a feminized and a masculinized version of one of the individual face images. Examples of these stimuli are shown in Figure 1. Note that our feminized and masculinized versions of faces differed in sexually dimorphic shape characteristics only (i.e., were matched in other regards, such as identity, color, and texture, Tiddeman et al., 2001).

### **Procedure**

In each test session, women reported their current romantic partnership status (partnered or unpartnered), reported their hormonal contraceptive use status (using the combined oral contraceptive pill, not using any form of hormonal contraceptive), reported whether they were currently taking a scheduled break from the pill (and, if so, how many days into this scheduled break they were), provided a saliva sample, and completed two face preference tests (one assessing men's attractiveness for a short-term relationship, the other assessing men's attractiveness for a long-term relationship). Attractiveness of men for short-term relationships and long-term relationships were measured separately because hormonal status has previously been shown to influence women's masculinity preferences when assessing men's attractiveness for short-term, but not long-term, relationships (Little & Jones, 2012; Penton-Voak et al., 1999).

In the two face preference tests, women were shown the 10 pairs of male faces, each pair consisting of a masculinized and feminized version of a given individual. Women were instructed to select the more attractive face in each pair and to indicate the strength of that preference by choosing from the options "slightly more attractive", "somewhat more attractive", "more attractive", and "much more attractive". This procedure has been used to assess masculinity preferences in previous studies (e.g., Zietsch et al., 2015).

In the short-term attractiveness test, women were told: "You are looking for the type of person who would be attractive in a short-term relationship. This

implies that the relationship may not last a long time. Examples of this type of relationship would include a single date accepted on the spur of the moment, an affair within a long-term relationship, and possibility of a one-night stand.”

In the long-term attractiveness test, women were told: “You are looking for the type of person who would be attractive in a long-term relationship. Examples of this type of relationship would include someone you may want to move in with, someone you may consider leaving a current partner to be with, and someone you may, at some point, wish to marry (or enter into a relationship on similar grounds as marriage).”

Trial order within each test was fully randomized and the order in which the two face preference tests were completed in each test session was also fully randomized. Definitions of short-term and long-term relationships were taken from previous studies (Little & Jones, 2012; Penton-Voak et al., 2003).

Responses on the face preference test were coded using the following scale (higher scores indicate stronger masculinity preferences and the scale is centered on chance, i.e., zero):

0.5 to 3.5: masculinized face rated ‘slightly more attractive’ (=0.5), ‘somewhat more attractive’ (=1.5), ‘more attractive’ (=2.5) or ‘much more attractive’ (=3.5) than feminized face.

-0.5 to -3.5: feminized face rated ‘slightly more attractive’ (= -0.5), ‘somewhat more attractive’ (= -1.5), ‘more attractive’ (= -2.5) or ‘much more attractive’ (= -3.5) than masculinized face.

Each woman’s average masculinity preference score was calculated separately for the short-term and long-term judgments for each test session. Higher scores indicate stronger masculinity preferences.

In each face preference test, the 10 trials assessing preferences for sexually dimorphic shape characteristics were interspersed among 30 filler trials assessing preferences for other facial traits.

### ***Saliva samples***

Participants provided a saliva sample via passive drool (Papacosta & Nassis, 2011) in each test session. Participants were instructed to avoid consuming alcohol and coffee in the 12 hours prior to participation and avoid eating, smoking, drinking, chewing gum, or brushing their teeth in the 60 minutes prior to participation. Each woman’s test sessions took place at approximately the same time of day to minimize effects of diurnal changes in hormone levels (Veldhuis et al., 1988; Bao et al., 2003).

Saliva samples were frozen immediately and stored at -32°C until being shipped, on dry ice, to the Salimetrics Lab (Suffolk, UK) for analysis, where they were assayed using the Salivary 17 $\beta$ -Estradiol Enzyme Immunoassay Kit 1-3702 (M=3.30 pg/mL, SD=1.27 pg/mL, sensitivity=0.1 pg/mL, intra-assay

CV=7.13%, inter-assay CV=7.45%), Salivary Progesterone Enzyme Immunoassay Kit 1-1502 (M=148.55 pg/mL, SD=96.13 pg/mL, sensitivity=5 pg/mL, intra-assay CV=6.20%, inter-assay CV=7.55%), Salivary Testosterone Enzyme Immunoassay Kit 1-2402 (M=87.66 pg/mL, SD=27.19 pg/mL, sensitivity<1.0 pg/mL, intra-assay CV=4.60%, inter-assay CV=9.83%), and Salivary Cortisol Enzyme Immunoassay Kit 1-3002 (M=0.23 µg/dL, SD=0.16 µg/dL, sensitivity<0.003 µg/dL, intra-assay CV=3.50%, inter-assay CV=5.08%). Only hormone levels from women not using hormonal contraceptives were used in analyses (values given above are for these women only).

Hormone levels more than three standard deviations from the sample mean for that hormone or where Salimetrics indicated levels were outside their sensitivity range were excluded from the dataset (~1% of hormone measures were excluded for these reasons). The descriptive statistics given above do not include these excluded values. Values for each hormone were centered on their subject-specific means to isolate effects of within-subject changes in hormones. They were then scaled so the majority of the distribution for each hormone varied from -.5 to .5 to facilitate calculations in the linear mixed models. Since hormone levels were centered on their subject-specific means, women with only one value for a hormone could not be included in analyses considering hormone levels.

## ***Analyses***

Linear mixed models were used to test for possible effects of hormonal status on women's facial masculinity preferences. Analyses were conducted using R version 3.3.2 (R Core Team, 2016), with lme4 version 1.1-13 (Bates et al., 2014) and lmerTest version 2.0-33 (Kuznetsova et al., 2013). The dependent variable was masculinity preference score, which was centered on chance. The relationship context for which women had judged men's attractiveness was effect-coded (short-term=+0.5 and long-term=-0.5) and included as an independent variable in all analyses. Random slopes were specified maximally following Barr et al. (2013) and Barr (2013). Full model specifications and full results for each analysis are given in our Supplemental Information. Data files and analysis scripts are publicly available at [osf.io/9b4y7](https://osf.io/9b4y7).

## Results

### ***General preferences and relationship-context effect***

Significant intercepts in all analyses indicated that women generally preferred masculinized to feminized versions of men's faces. Masculinity preferences were also significantly stronger in the short-term than long-term relationship context in all analyses. The one exception was in the analyses described under Hypothesis 4. There, the relationship context effect was not significant, probably because these analyses were less powerful than our other analyses. Full results for these effects are given in our Supplemental Information.

***Hypothesis 1. Do facial masculinity preferences track changes in measured steroid hormone levels in women not using hormonal contraceptives?***

The fertile phase of the menstrual cycle is characterized by the combination of high estradiol and low progesterone (Gangestad & Haselton, 2015; Puts et al., 2013). Additionally, some previous studies have suggested that changes in women's masculinity preferences are positively correlated with changes in estradiol (Roney & Simmons, 2008; Roney et al., 2011) and negatively correlated with changes in progesterone (Jones et al., 2005; Puts, 2006). We therefore used linear mixed models to test for possible effects of estradiol, progesterone, and their interaction on women's facial masculinity preferences. Masculinity preference scores could range from -3.5 to 3.5 (0 indicated no preference; higher scores indicated stronger masculinity preferences). This analysis included all women who were not using any form of hormonal contraceptive when tested (N=351). The specific models we used to test for hormonal correlates of within-woman changes in masculinity preferences are identical to those that we have used elsewhere to test for hormonal correlates of disgust sensitivity (Jones et al., 2017a) and sexual desire (Jones et al., 2017b). No effects involving hormone levels were significant in this analysis (all  $t_s < 0.88$ , all  $p_s > .38$ ), suggesting that women's preferences for facial masculinity are not related to their hormonal status.

We conducted additional analyses to test for previously reported effects of testosterone (Welling et al., 2007) and cortisol (Ditzen et al., 2017) on masculinity preferences, and for hypothesized effects of estradiol-to-

progesterone ratio on mating-related behavior (Eisenbruch et al., 2015).

These analyses also showed no evidence that women's preferences for masculine men were related to their hormone levels (see Supplemental Information).

At the suggestion of a reviewer, we also tested for an interaction between the effects of testosterone and cortisol (see Supplemental Information). The rationale for testing this interaction was that some research suggests that behavioral effects of testosterone are more pronounced when cortisol is low (the Dual Hormone Hypothesis, see Mehta & Prasad, 2015). Although there was a significant interaction between testosterone and cortisol ( $\beta=0.51$ ,  $SE=0.21$ ,  $t=2.39$ ,  $p=.018$ , 95% CIs=0.09, 0.93), it indicated that women's masculinity preferences were strongest when both testosterone and cortisol were high. Since this is not the pattern of results predicted by the Dual Hormone Hypothesis, was not an a priori prediction, and was the only significant hormone effect in multiple tests for possible effects of endogenous hormones on masculinity preferences, we suggest that it is likely to be a false positive.

A reviewer also asked that we repeat each of the analyses described above controlling for effects of test session order on masculinity preferences. Doing so did not alter the patterns of results (i.e., no non-significant effects became significant and no significant effects became non-significant). These analyses are reported in our Supplemental Information.

***Hypothesis 2. Do women not using hormonal contraceptives show stronger facial masculinity preferences than women using the combined oral contraceptive pill?***

Studies reporting that women not using hormonal contraceptives show stronger facial masculinity preferences than do women using hormonal contraceptives have been interpreted as converging evidence that women's hormonal status influences their facial masculinity preferences (Little et al., 2013). To investigate this issue in our data set, we first used linear mixed models to compare the facial masculinity preferences of women using the combined oral contraceptive pill (N=212) and women not using any form of hormonal contraceptive (N=326). This analysis included all women who had reported either no use of hormonal contraceptives throughout the study or use of the combined oral contraceptive pill throughout the study (responses from women who changed contraceptive status during the study are reported under Hypothesis 4). Although there was a significant effect of oral contraceptive use in this analysis ( $\beta=0.12$ ,  $SE=0.04$ ,  $t=2.75$ ,  $p=.006$ , 95% CIs= 0.03, 0.20), the effect was such that women using the combined oral contraceptive pill showed *stronger* masculinity preferences ( $M=0.47$ ,  $SEM=0.03$ ) than did women not using any form of hormonal contraceptive ( $M=0.35$ ,  $SEM=0.03$ ). Note that stronger masculinity preferences in women using the combined oral contraceptive pill is the opposite pattern of results to what would be expected if fertility had the hypothesized positive effect on women's masculinity preferences.

Stronger masculinity preferences in women using hormonal contraceptives have been reported in one other study (Cobey et al., 2015). We suggest that these between-group differences reflect effects of lifestyle and/or personality factors that are correlated with contraceptive use, rather than hormonal effects.

***Hypothesis 3. Do facial masculinity preferences of women using the combined oral contraceptive pill change when they are taking inactive pills?***

In women using the combined oral contraceptive pill, fertility-linked hormone levels are affected when women are not taking active pills (i.e., the scheduled 'hormone-free interval' or 'break') during their monthly cycle of oral contraceptive use (van Heusden & Fauser, 2002). If women's masculinity preferences are influenced by their hormonal status, one would then expect women's facial masculinity preferences to change during this scheduled break. To investigate this possibility, we used linear mixed models to compare the facial masculinity preferences of women ( $N=173$ ) using the combined oral contraceptive pill when they were taking active pills versus when they were taking a scheduled break from taking active pills. Note that not all women using the combined oral contraceptive pill were tested during a scheduled break. No effects involving the scheduled break were significant (both absolute  $ts<0.64$ , both  $ps>.52$ ).

***Possible moderating role of partnership status***

Some previous research has suggested that the magnitude of hormone-linked changes in women's masculinity preferences is moderated by their partnership status (i.e., whether or not they had a romantic partner, Penton-Voak et al., 1999). Thus, we repeated each of the analyses described above including partnership status and all possible interactions between partnership status and the other predictors (see Supplemental Information). These additional analyses also showed no evidence that women's salivary steroid hormone levels were related to their facial masculinity preferences or that oral contraceptive use decreased masculinity preferences.

***Hypothesis 4. Do facial masculinity preferences change when women start or stop using the combined oral contraceptive pill?***

During the course of the current study, 45 women changed their hormonal contraceptive use by either switching from using no hormonal contraceptive to using the combined oral contraceptive pill, or vice versa. There was a mean time of 360 days (SD=282 days, range=56 to 1113 days) between test sessions where women were using no hormonal contraceptives and those where they were using the combined oral contraceptive pill. A previous study of 18 women's facial masculinity preferences reported that women's preferences for masculinity in men's faces decreased when women started using oral contraceptives (Little et al., 2013). We therefore used linear mixed models to compare the facial masculinity preferences of these women when they were using the combined oral contraceptive pill and when they were using no form of hormonal contraceptive. Our analysis controlled for the direction of change in women's oral contraceptive use (i.e., whether they

changed from using no form of hormonal contraceptive to using the combined oral contraceptive pill, N=30; or vice versa, N=15). The effect of oral contraceptive use was not significant ( $\beta=0.08$ ,  $SE=0.05$ ,  $t=1.57$ ,  $p=.12$ , 95% CIs = -0.02, 0.17). Note that women's masculinity preferences tended to be stronger when they were using the combined oral contraceptive pill (although not significantly so), suggesting that a lack of power did not prevent detection of the hypothesized weaker masculinity preferences when women are using the combined oral contraceptive pill.

Because changes in oral contraceptive use could be associated with a change in partnership status, we repeated this analysis controlling for possible effects of changes in women's partnership status (see Supplemental Information). This additional analysis also did not show any evidence that using the combined oral contraceptive pill weakened women's masculinity preferences.

### ***Preferences for additional facial traits***

Some previous studies have tested for effects of hormonal status on other aspects of women's face preferences, such as preferences for femininity in women's faces, facial symmetry, facial averageness, and apparent facial health (reviewed in Jones et al., 2008). Consequently, we also tested for effects of hormonal status on women's preferences for these facial characteristics.

All male face preferences were assessed in the same short-term and long-term blocks with trial order fully randomized. All female face preferences were tested in a separate block, again with trial order fully randomized. The order in which women completed the short-term male attractiveness, long-term male attractiveness, and female attractiveness preference tasks in each test session was fully randomized. Femininity in women's faces was manipulated using identical methods to those that were used to manipulate masculinity in men's faces. Methods used to manipulate facial symmetry, facial averageness, and apparent facial health are reported in Quist et al. (2012), Jones et al. (2007), and Wincenciak et al. (2015), respectively.

Analyses of these preferences using the same type of models we used to test for effects of hormonal status on masculinity preferences also showed no clear evidence that face preferences were consistently related to women's hormonal status. Notably, we did not replicate putative effects of ovarian hormones on women's preferences for symmetry or apparent health previously reported for women not using hormonal contraceptives (reviewed in Jones et al., 2008). Full results, along with the data, analysis files, and stimuli, are publicly available at osf.io/9b4y7. These full results include a significant negative effect of cortisol on preferences for male facial symmetry and a significant negative effect of progesterone on preferences for male facial averageness. Neither of these results were *a priori* predictions, so we suggest they should be treated as preliminary findings.

## Discussion

Collectively, our analyses showed no evidence that changes in women's salivary hormone levels are associated with their facial masculinity preferences or that the combined oral contraceptive pill decreases women's masculinity preferences<sup>1</sup>. This was despite having a much larger sample size, having tested participants more often, and having used more reliable measures of hormonal status (e.g., measurements of multiple steroid hormones from saliva samples) than previous studies. Thus, the current study presents strong evidence against the popular and influential hypothesis that changes in women's facial masculinity preferences track changes in their hormonal status (Ditzen et al., 2017; Johnston et al., 2001; Little & Jones, 2012; Little et al., 2013; Penton-Voak et al., 1999; Penton-Voak & Perrett, 2000). Analyses of preferences for other facial traits (symmetry, averageness, apparent health) that some previous research had suggested may track changes in hormonal status also showed no evidence for consistent effects of hormonal status on face preferences.

A crucial piece of the rationale for predicting hormone-linked changes in women's preferences for facial masculinity is the claim that facial masculinity is a cue of men's heritable immunocompetence (Penton-Voak et al., 1999). Our null results for hormonal status and facial masculinity preferences add to a growing body of evidence calling this assumption into question (Lee et al., 2014; Scott et al., 2014). Rather than functioning as a cue of men's

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<sup>1</sup> The only significant effect of salivary hormones on masculinity preferences that we observed was an interaction between testosterone and cortisol that appeared to be a false positive.

immunocompetence, men's facial masculinity may primarily function as a cue of their intrasexual competitiveness (reviewed in Puts, 2010).

Although we find no evidence that women's masculinity preferences are linked to their hormonal status, our analyses do suggest that women show stronger preferences for masculine facial characteristics when assessing men's attractiveness for short-term relationships than when assessing men's attractiveness for long-term relationships. Although this pattern of results is consistent with the proposal that perceived costs associated with choosing a masculine mate cause women's preferences for masculinity in long-term partners to be weaker than preferences for masculinity in short-term partners (Little et al., 2011), we emphasize here that the effect of relationship context on masculinity preference was small.

In summary, and by contrast with previous research using smaller samples and less precise measures of hormonal status, our analyses show no evidence for links between women's hormonal status and preferences for facial masculinity. These results highlight the importance of employing large sample sizes and rigorous assessments of hormonal status (e.g., measures of salivary hormone levels) to test hypotheses concerning links between hormonal status and mate preferences.

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# Male Facial Masculinity

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# Overview

This supplemental information contains the R code for data analysis of **male facial masculinity** preferences reported in the manuscript (data are publicly available at <https://osf.io/9b4y7/> (<https://osf.io/9b4y7/>)).

*Hypothesis 1.* Do preferences track changes in measured steroid hormone levels in women not using hormonal contraceptives?

Women reporting no use of hormonal contraceptives and for whom at least two test sessions with valid hormone levels are available.

*Hypothesis 2.* Do women not using hormonal contraceptives show stronger preferences than women using the combined oral contraceptive pill?

Women reporting use of the combined oral contraceptive pill or no use of hormonal contraceptives across all test sessions.

*Hypothesis 3.* Do preferences of women using the combined oral contraceptive pill change when they are taking inactive pills?

Women who were tested during a scheduled break from use of the combined oral contraceptive pill.

*Hypothesis 4.* Do preferences change when women start or stop using the combined oral contraceptive pill?

Women who switched from using no hormonal contraceptive to using the combined oral contraceptive pill (or vice versa) between blocks of test sessions.

```
## Loading tidyverse: ggplot2
## Loading tidyverse: tibble
## Loading tidyverse: tidyr
## Loading tidyverse: readr
## Loading tidyverse: purrr
## Loading tidyverse: dplyr
```

```
## Conflicts with tidy packages -----
```

```
## filter(): dplyr, stats
## lag():    dplyr, stats
```

```
## Loading required package: Matrix
```

```
##
## Attaching package: 'Matrix'
```

```
## The following object is masked from 'package:tidyঃ':
```

```
##
```

```
##     expand
```

```
##
```

```
## Attaching package: 'lmerTest'
```

```
## The following object is masked from 'package:lme4':
```

```
##
```

```
##     lmer
```

```
## The following object is masked from 'package:stats':
```

```
##
```

```
##     step
```

```
##
```

```
## Attaching package: 'lubridate'
```

```
## The following object is masked from 'package:base':
```

```
##
```

```
##     date
```

```
# load data frames from Jones_hormones_data.Rmd
data_hormones <- readRDS("data_hormones.Rda")
data_hormones_partner <- readRDS("data_hormones_partner.Rda")
data_between <- readRDS("data_between.Rda")
data_between_partner <- readRDS("data_between_partner.Rda")
data_pillbreak <- readRDS("data_pillbreak.Rda")
data_pillbreak_partner <- readRDS("data_pillbreak_partner.Rda")
data_pill_switchers <- readRDS("data_pill_switchers.Rda")
data_pill_switchers_partner <- readRDS("data_pill_switchers_partner.Rda")
```

```
# filter only sexual dimorphism manipulations for male faces in the analyses below
```

```
theManip <- "sexdim"
```

```
theFaceSex <- "men"
```

```
confint_method <- "Wald" # c("profile", "Wald", "boot")
```

```
# calculate standard errors
se <- function(x, na.rm = FALSE) {
  if (na.rm) {
    the.SE <- sqrt(var(x,na.rm=TRUE)/length(na.omit(x)))
  } else {
    the.SE <- sqrt(var(x,na.rm=FALSE)/length(x))
  }

  return(the.SE)
}

# short summaries for lmerTest
mySummary <- function(lmer_summary) {

  coefTable <- lmer_summary$coefficients %>%
    round(3) %>%
    as.data.frame() %>%
    rownames_to_column()

  if (ncol(coefTable)>5) {
    coefTable <- coefTable %>%
      mutate(
        sig = ifelse(.[6]<.001, "***",
                     ifelse(.[6]<.01, "**",
                           ifelse(.[6]<.05, "*",
                                 ifelse(.[6]<.10, "+", "")))))
  }

  return(list(lmer_summary$ngrps, kable(coefTable)))
}
```

# Hypothesis 1

Do preferences track changes in measured steroid hormone levels in women not using hormonal contraceptives?

For tests of effects of endogenous hormones. Women reporting no use of hormonal contraceptives and for whom at least two test sessions with valid hormone levels are available (all relationship statuses).

## Descriptive stats: data\_hormones

### Preferences

```
# create mean DV for all ratings by oc_id
stats_overall <- filter(data_hormones, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id) %>%
  summarise(
    overall_rating.c = mean(rating.c)
  ) %>%
  ungroup() %>%
  group_by() %>%
  summarise(
    context="overall",
    n = n_distinct(oc_id),
    mean_dv = mean(overall_rating.c),
    sd_dv = sd(overall_rating.c),
    se_dv = se(overall_rating.c)
  )

# create mean DV splitting by context
stats_context <- filter(data_hormones, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id, context) %>%
  summarise(
    context_rating.c = mean(rating.c)
  ) %>%
  group_by(context) %>%
  summarise(
    n = n_distinct(oc_id),
    mean_dv = mean(context_rating.c),
    sd_dv = sd(context_rating.c),
    se_dv = se(context_rating.c)
  )

)
rbind(stats_overall, stats_context)
```

```
## # A tibble: 3 x 5
##   context     n   mean_dv     sd_dv     se_dv
## * <chr> <int>     <dbl>     <dbl>     <dbl>
## 1 overall    351  0.3461916  0.4794891  0.02559323
## 2 LT         351  0.3214278  0.5017172  0.02677968
## 3 ST         351  0.3708437  0.5076745  0.02709766
```

## Hormones

```
filter(data_hormones, face_sex==theFaceSex, manip==theManip) %>%
  group_by() %>%
  summarise(
    mean_prog = mean(prog, na.rm = TRUE),
    sd_prog =sd(prog, na.rm = TRUE),
    se_prog =se(prog, na.rm = TRUE),
    mean_estr = mean(estr, na.rm = TRUE),
    sd_estr =sd(estr, na.rm = TRUE),
    se_estr =se(estr, na.rm = TRUE),
    mean_test = mean(test, na.rm = TRUE),
    sd_test =sd(test, na.rm = TRUE),
    se_test =se(test, na.rm = TRUE),
    mean_cort = mean(cort, na.rm = TRUE),
    sd_cort =sd(cort, na.rm = TRUE),
    se_cort =se(cort, na.rm = TRUE)
  ) %>% gather("stat", "value", 1:length(.)) %>%
  mutate(value = round(value, 4)) %>%
  separate(stat, c("stat", "hormone")) %>%
  spread(stat, value)
```

```
## # A tibble: 4 x 4
##   hormone      mean       sd     se
## * <chr>     <dbl>     <dbl>  <dbl>
## 1 cort      0.2288  0.1645  0.0025
## 2 estr      3.2945  1.2718  0.0195
## 3 prog     148.5481 96.1287 1.4812
## 4 test      87.6616 27.1943 0.4172
```

# Analyses H1: Hormones

## E + P + E\*P:

Testing for effects of estradiol, progesterone, and their interaction on preferences

```
model_h1_EP <- lmer(rating.c ~ estr.s * prog.s * context.e +
  (1 | oc_id) +
  (0 + estr.s:prog.s:context.e || oc_id) +
  (1 | block:oc_id) +
  (0 + estr.s:prog.s:context.e || block:oc_id) +
  (1 | block:oc_id:date) +
  (0 + context.e || block:oc_id:date),
  data = filter(data_hormones,
    face_sex==theFaceSex,
    manip==theManip),
  REML = FALSE)
summary_h1_EP <- summary(model_h1_EP)
mySummary(summary_h1_EP)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           2092              444              351
##
## [[2]]
##
##
##   rowname       Estimate    Std. Error      df   t value Pr(>|t| )
##   sig
##   -----
##   ----
##   ## (Intercept)     0.347      0.026  362.110  13.467  0.000
##   *** 
##   ## estr.s        -0.005      0.051 1733.893 -0.104  0.917
##   ## prog.s        -0.010      0.043 1735.838 -0.221  0.825
##   ## context.e      0.052      0.013 2081.180  4.095  0.000
##   *** 
##   ## estr.s:prog.s -0.099      0.253 1784.753 -0.393  0.695
##   ## estr.s:context.e 0.047      0.082 1723.701  0.569  0.569
##   ## prog.s:context.e 0.025      0.073 1560.698  0.344  0.731
##   ## estr.s:prog.s:context.e 0.362      0.412   72.798  0.879  0.382
```

```
confint(model_h1_EP, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##           rowname    2.5 %    97.5 %
## 1 (Intercept) 0.29686369 0.39799451
## 2 estr.s     -0.10517476 0.09454539
## 3 prog.s     -0.09478402 0.07553965
## 4 context.e   0.02718444 0.07710258
## 5 estr.s:prog.s -0.59503289 0.39635879
## 6 estr.s:context.e -0.11356992 0.20657516
## 7 prog.s:context.e -0.11742755 0.16748868
## 8 estr.s:prog.s:context.e -0.44529767 1.16952870
```

## E + P + EPratio:

Testing for effects of estradiol, progesterone, and their ratio on preferences

```
model_h1_EP_ep_ratio <- lmer(rating.c ~ estr.s * context.e +
                                prog.s * context.e +
                                ep_ratio.s * context.e +
                                (1 | oc_id) +
                                (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || oc_id) +
                                (1 | block:oc_id) +
                                (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || block:oc_id) +
                                (1 | block:oc_id:date) +
                                (0 + context.e || block:oc_id:date),
                                data = filter(data_hormones,
                                              face_sex==theFaceSex,
                                              manip==theManip),
                                REML = FALSE)
summary_h1_EP_ep_ratio <- summary(model_h1_EP_ep_ratio)
mySummary(summary_h1_EP_ep_ratio)
```

```
## [[1]]
## block:oc_id:date      block:oc_id      oc_id
##          2092           444            351
##
## [[2]]
##
##
##   rowname      Estimate   Std. Error      df   t value Pr(>|t|)    s
##   ig
##   -----
##   ---
##   ## (Intercept) 0.347     0.026    356.819 13.482  0.000 *
##   ** 
##   ## estr.s      0.002     0.052   1740.901  0.031  0.975
## 
##   ## context.e    0.056     0.012   2087.545  4.567  0.000 *
##   ** 
##   ## prog.s      -0.029    0.049   1721.956 -0.594  0.552
## 
##   ## ep_ratio.s  -0.018    0.025   1691.625 -0.717  0.473
## 
##   ## estr.s:context.e 0.054     0.084   2092.809  0.647  0.518
## 
##   ## context.e:prog.s 0.026     0.083   2090.185  0.317  0.751
## 
##   ## context.e:ep_ratio.s -0.006    0.043   2084.905 -0.132  0.895
```

```
confint(model_h1_EP_ep_ratio, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##          rowname      2.5 %     97.5 %
## 1      (Intercept)  0.29619719  0.39696665
## 2           estr.s -0.10019500  0.10345275
## 3        context.e  0.03190113  0.07986220
## 4           prog.s -0.12594936  0.06734871
## 5       ep_ratio.s -0.06779680  0.03147571
## 6 estr.s:context.e -0.10968230  0.21781394
## 7   context.e:prog.s -0.13601061  0.18844409
## 8 context.e:ep_ratio.s -0.08969433  0.07841501
```

## T + C:

Testing for effects of testosterone and cortisol on preferences

```
# Converges when we model random slopes for individual predictors and the interaction on oc_id
model_h1_TC <- lmer(rating.c ~ test.s * context.e + cort.s * context.e +
                      (1 | oc_id) +
                      (0 + test.s*context.e + cort.s*context.e || oc_id) +
                      (1 | block:oc_id) +
                      (0 + test.s*context.e + cort.s*context.e || block:oc_id) +
                      (1 | block:oc_id:date) +
                      (0 + context.e || block:oc_id:date),
                      data = filter(data_hormones,
                                    face_sex==theFaceSex,
                                    manip==theManip),
                      REML = FALSE)
summary_h1_TC <- summary(model_h1_TC)
mySummary(summary_h1_TC)
```

```
## [[1]]
## block:oc_id:date      block:oc_id      oc_id
##          2117            443            350
##
## [[2]]
##
##
##          rowname    Estimate   Std. Error      df   t value Pr(>|t|)    sig
##  -----
##  (Intercept)  0.348      0.026  351.997  13.558  0.000 *** 
##  test.s      -0.013      0.051 1517.989 -0.261  0.794  
##  context.e    0.054      0.016  342.392  3.302  0.001 ** 
##  cort.s      0.042      0.043   77.255  0.987  0.327  
##  test.s:context.e -0.036      0.077 1799.430 -0.472  0.637  
##  context.e:cort.s -0.003      0.057 1780.370 -0.055  0.956
```

```
confint(model_h1_TC, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##          rowname      2.5 %     97.5 %
## 1      (Intercept) 0.29755599 0.39812117
## 2           test.s -0.11289060 0.08638938
## 3        context.e  0.02209064 0.08660596
## 4           cort.s -0.04153990 0.12583238
## 5 test.s:context.e -0.18602187 0.11387628
## 6 context.e:cort.s -0.11546374 0.10919083
```

## T + C + T\*C:

Testing for effects of testosterone and cortisol plus their interaction on preferences

```
# Converges when we model random slopes for individual predictors and the interaction on oc_id
model_h1_TCi <- lmer(rating.c ~ test.s * context.e * cort.s +
                      (1 | oc_id) +
                      (0 + test.s*cort.s*context.e || oc_id) +
                      (1 | block:oc_id) +
                      (0 + test.s*cort.s*context.e|| block:oc_id) +
                      (1 | block:oc_id:date) +
                      (0 + context.e || block:oc_id:date),
                      data = filter(data_hormones,
                                    face_sex==theFaceSex,
                                    manip==theManip),
                      REML = FALSE)
summary_h1_TCi <- summary(model_h1_TCi)
mySummary(summary_h1_TCi)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           2117              443              350
##
## [[2]]
##
##
##   rowname       Estimate    Std. Error      df   t value Pr(>|t| )
##   sig
##   -----
##   ----
##   ## (Intercept)     0.340      0.026  363.861  13.180 0.000
##   *** 
##   ## test.s        -0.008      0.051 1466.856 -0.166 0.868
##   ## context.e      0.056      0.017  391.077  3.257 0.001
##   ** 
##   ## cort.s        0.017      0.043   65.367  0.399 0.692
##   ## test.s:context.e -0.036      0.077 1762.675 -0.476 0.634
##   ## test.s:cort.s    0.510      0.213  173.893  2.393 0.018
##   * 
##   ## context.e:cort.s 0.003      0.059 1383.311  0.055 0.956
##   ## test.s:context.e:cort.s -0.081      0.290  118.611 -0.277 0.782
```

```
confint(model_h1_TCi, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##           rowname    2.5 %    97.5 %
## 1 (Intercept) 0.28958949 0.39076303
## 2 test.s     -0.10808618 0.09122954
## 3 context.e   0.02211887 0.08896728
## 4 cort.s     -0.06644062 0.10035490
## 5 test.s:context.e -0.18660194 0.11367372
## 6 test.s:cort.s  0.09240440 0.92831933
## 7 context.e:cort.s -0.11299765 0.11951841
## 8 test.s:context.e:cort.s -0.64980293 0.48869086
```

## Analyses H1s: Hormones (+ session order)

### E + P + E\*P: (+ session order)

Testing for effects of estradiol, progesterone, and their interaction on preferences

```
model_h1_EP_s <- lmer(rating.c ~ estr.s * prog.s * context.e + session_n +
                         (1 | oc_id) +
                         (0 + estr.s:prog.s:context.e || oc_id) +
                         (1 | block:oc_id) +
                         (0 + estr.s:prog.s:context.e || block:oc_id) +
                         (1 | block:oc_id:date) +
                         (0 + context.e || block:oc_id:date),
                         data = filter(data_hormones,
                                       face_sex==theFaceSex,
                                       manip==theManip),
                         REML = FALSE)
summary_h1_EP_s <- summary(model_h1_EP_s)
mySummary(summary_h1_EP_s)
```

```
## [[1]]
## block:oc_id:date      block:oc_id      oc_id
##          2092            444            351
##
## [[2]]
##
##
##   rowname           Estimate    Std. Error      df   t value Pr(>|t| )
##   sig
##   -----
##   ----
##   ## (Intercept)     0.403       0.028  499.963  14.287  0.000
##   *** 
##   ## estr.s        -0.025       0.051 1749.410  -0.502  0.616
##   ## prog.s         0.007       0.043 1726.724   0.160  0.873
##   ## context.e      0.052       0.013 2081.200   4.093  0.000
##   *** 
##   ## session_n     -0.020       0.004  888.856  -4.924  0.000
##   *** 
##   ## estr.s:prog.s -0.110       0.251 1780.634  -0.440  0.660
##   ## estr.s:context.e 0.046       0.082 1731.563   0.567  0.571
##   ## prog.s:context.e 0.025       0.073 1564.314   0.346  0.730
##   ## estr.s:prog.s:context.e 0.357       0.413  73.123   0.866  0.389
```

```
confint(model_h1_EP_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                 rowname      2.5 %     97.5 %
## 1          (Intercept)  0.34790582  0.45853369
## 2             estr.s -0.12495796  0.07401377
## 3             prog.s -0.07784856  0.09169048
## 4        context.e  0.02716474  0.07708351
## 5       session_n -0.02823991 -0.01215962
## 6 estr.s:prog.s -0.60234695  0.38151110
## 7 estr.s:context.e -0.11380008  0.20635152
## 8   prog.s:context.e -0.11735560  0.16759816
## 9 estr.s:prog.s:context.e -0.45159613  1.16612625
```

## E + P + EPratio: (+ session order)

Testing for effects of estradiol, progesterone, and their ratio on preferences

```
model_h1_EP_ep_ratio_s <- lmer(rating.c ~ estr.s * context.e +
                                    prog.s * context.e +
                                    ep_ratio.s * context.e + session_n +
                                    (1 | oc_id) +
                                    (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || oc_id) +
                                    (1 | block:oc_id) +
                                    (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || block:oc_id) +
                                    (1 | block:oc_id:date) +
                                    (0 + context.e || block:oc_id:date),
                                    data = filter(data_hormones,
                                                  face_sex==theFaceSex,
                                                  manip==theManip),
                                    REML = FALSE)
summary_h1_EP_ep_ratio_s <- summary(model_h1_EP_ep_ratio_s)
mySummary(summary_h1_EP_ep_ratio_s)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           2092              444             351
##
## [[2]]
##
##
## rowname            Estimate   Std. Error    df   t value Pr(>|t|) s
ig
## -----
## (Intercept)       0.402     0.028 493.868 14.302 0.000 *
## estr.s        -0.018     0.052 1752.360 -0.347 0.729
## context.e       0.056     0.012 2087.488  4.563 0.000 *
## prog.s        -0.015     0.049 1714.073 -0.298 0.766
## ep_ratio.s     -0.020     0.025 1687.992 -0.791 0.429
## session_n      -0.020     0.004 888.515 -4.930 0.000 *
## estr.s:context.e 0.054     0.084 2092.773  0.645 0.519
## context.e:prog.s 0.026     0.083 2090.138  0.316 0.752
## context.e:ep_ratio.s -0.006    0.043 2084.836 -0.131 0.895
```

```
confint(model_h1_EP_ep_ratio_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##           rowname    2.5 %    97.5 %
## 1 (Intercept) 0.34719714 0.45746899
## 2 estr.s     -0.11933383 0.08347172
## 3 context.e   0.03184974 0.07981145
## 4 prog.s     -0.11069007 0.08146607
## 5 ep_ratio.s -0.06911229 0.02937832
## 6 session_n  -0.02825967 -0.01218148
## 7 estr.s:context.e -0.10983204 0.21766519
## 8 context.e:prog.s -0.13608465 0.18837273
## 9 context.e:ep_ratio.s -0.08968998 0.07842241
```

## T + C: (+ session order)

Testing for effects of testosterone and coritsol on preferences

```
# Converges when we model random slopes for individual predictors and the interaction on on oc_id
model_h1_TC_s <- lmer(rating.c ~ test.s * context.e + cort.s * context.e + session_n +
                        (1 | oc_id) +
                        (0 + test.s*context.e + cort.s*context.e || oc_id) +
                        (1 | block:oc_id) +
                        (0 + test.s*context.e + cort.s*context.e|| block:oc_id) +
                        (1 | block:oc_id:date) +
                        (0 + context.e || block:oc_id:date),
                        data = filter(data_hormones,
                                      face_sex==theFaceSex,
                                      manip==theManip),
                        REML = FALSE)
summary_h1_TC_s <- summary(model_h1_TC_s)
mySummary(summary_h1_TC_s)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           2117              443             350
##
## [[2]]
##
##
##   rowname    Estimate   Std. Error      df   t value Pr(>|t|) sig
##   -----  -----
##   (Intercept)  0.402     0.028  484.840  14.351  0.000 *** 
##   test.s      -0.014     0.051  1578.123 -0.278  0.781  
##   context.e    0.054     0.016  342.404   3.299  0.001 ** 
##   cort.s       0.047     0.043   85.025   1.099  0.275  
##   session_n   -0.020     0.004  777.129  -4.849  0.000 *** 
##   test.s:context.e -0.036     0.077  1799.454 -0.472  0.637  
##   context.e:cort.s -0.003     0.057  1780.335 -0.054  0.957
```

```
confint(model_h1_TC_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##           rowname    2.5 %    97.5 %
## 1 (Intercept) 0.34678595 0.45649347
## 2 test.s      -0.11314404 0.08501393
## 3 context.e    0.02204139 0.08656692
## 4 cort.s       -0.03701695 0.13144677
## 5 session_n   -0.02738266 -0.01161808
## 6 test.s:context.e -0.18604885 0.11384184
## 7 context.e:cort.s -0.11540624 0.10924340
```

## T + C + T\*C: (+ session order)

Testing for effects of testosterone and cortisol plus their interaction on preferences

```
# Converges when we model random slopes for individual predictors and the interaction on on oc_id
model_h1_TCi_s <- lmer(rating.c ~ test.s * context.e * cort.s + session_n +
                         (1 | oc_id) +
                         (0 + test.s*cort.s*context.e || oc_id) +
                         (1 | block:oc_id) +
                         (0 + test.s*cort.s*context.e|| block:oc_id) +
                         (1 | block:oc_id:date) +
                         (0 + context.e || block:oc_id:date),
                         data = filter(data_hormones,
                                       face_sex==theFaceSex,
                                       manip==theManip),
                         REML = FALSE)
summary_h1_TCi_s <- summary(model_h1_TCi_s)
mySummary(summary_h1_TCi_s)
```

```
## [[1]]
## block:oc_id:date      block:oc_id      oc_id
##          2117            443           350
##
## [[2]]
##
##
##   rowname             Estimate    Std. Error      df   t value Pr(>|t|)
##   sig
##   -----
##   ----
##   ## (Intercept)       0.394      0.028   496.821  13.988  0.000
##   *** 
##   ## test.s        -0.010      0.051 1466.836 -0.196  0.845
## 
##   ## context.e       0.055      0.017   391.063  3.254  0.001
##   ** 
##   ## cort.s         0.023      0.043    61.456  0.540  0.591
## 
##   ## session_n      -0.019      0.004   787.814 -4.812  0.000
##   *** 
##   ## test.s:context.e -0.036      0.077 1762.260 -0.476  0.634
## 
##   ## test.s:cort.s     0.497      0.212   170.717  2.338  0.021
##   * 
##   ## context.e:cort.s  0.003      0.059 1379.452  0.056  0.955
## 
##   ## test.s:context.e:cort.s -0.081      0.290   117.349 -0.277  0.782
```

```
confint(model_h1_TCi_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                 rowname    2.5 %    97.5 %
## 1          (Intercept) 0.33864422 0.44900805
## 2             test.s -0.10897840 0.08918440
## 3        context.e  0.02207067 0.08892861
## 4            cort.s -0.06051673 0.10660158
## 5       session_n -0.02724470 -0.01147466
## 6 test.s:context.e -0.18663173 0.11363595
## 7      test.s:cort.s  0.08031600 0.91303187
## 8    context.e:cort.s -0.11294262 0.11956736
## 9 test.s:context.e:cort.s -0.64976485 0.48869776
```

## Descriptive stats: data\_hormones\_partner

```
data_hormones_partner %>%
  filter(face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id, partner.e) %>%
  summarise(overall_rating.c = mean(rating.c)) %>%
  group_by(partner.e) %>%
  summarise(
    n= n_distinct(oc_id),
    mean_dv = mean(overall_rating.c),
    sd_dv = sd(overall_rating.c),
    se_dv = se(overall_rating.c)
  )
```

```
## # A tibble: 2 x 5
##   partner.e     n   mean_dv     sd_dv     se_dv
##   <dbl> <int>    <dbl>    <dbl>    <dbl>
## 1     -0.5    210  0.3063844  0.4461867  0.03078981
## 2      0.5    111  0.4483861  0.5233885  0.04967784
```

## Analyses H1p: Hormones (+ partnership status)

### E + P + E\*P: (+ partnership status)

Testing for effects of estradiol, progesterone, and their interaction on preferences

```
model_h1p_EP <- lmer(rating.c ~ estr.s * prog.s * context.e * partner.e + session_n
+
(1 | oc_id) +
(0 + estr.s:prog.s:context.e || oc_id) +
(1 | block:oc_id) +
(0 + estr.s:prog.s:context.e|| block:oc_id) +
(1 | block:oc_id:date) +
(0 + context.e || block:oc_id:date),
data = filter(data_hormones_partner,
              face_sex==theFaceSex,
              manip==theManip),
REML = FALSE,
# use nelder-mead optimziation (as in lmer v < 1)
control = lmerControl(optimizer ="Nelder_Mead"))
summary_h1p_EP <- summary(model_h1p_EP)
mySummary(summary_h1p_EP)
```

```
## [[1]]
## block:oc_id:date          block:oc_id          oc_id
##           1829                  389                 321
##
## [[2]]
##
##
## rowname                         Estimate   Std. Error    df   t value
## Pr(>|t|)  sig
## -----
## -----
## ## (Intercept)             0.426      0.031  448.353  13.942
## 0.000 *** 
## ## estr.s                -0.055      0.057 1524.518 -0.951
## 0.342
## ## prog.s                 0.005      0.046 1500.036  0.111
## 0.912
## ## context.e              0.052      0.014 1824.334  3.656
## 0.000 ***
## ## partner.e              0.145      0.056  331.663  2.585
## 0.010 *
## ## session_n              -0.017      0.004  781.721 -3.924
## 0.000 ***
## ## estr.s:prog.s          -0.012      0.276 1560.202 -0.043
## 0.966
## ## estr.s:context.e       0.031      0.092 1828.769  0.334
## 0.738
## ## prog.s:context.e       0.026      0.078 1827.753  0.336
## 0.737
## ## estr.s:partner.e       -0.255      0.115 1520.185 -2.224
## 0.026 *
## ## prog.s:partner.e       0.068      0.092 1503.985  0.741
## 0.459
## ## context.e:partner.e    0.015      0.028 1824.334  0.512
## 0.608
## ## estr.s:prog.s:context.e 0.158      0.429 1843.423  0.367
## 0.713
## ## estr.s:prog.s:partner.e 0.611      0.552 1559.622  1.106
## 0.269
## ## estr.s:context.e:partner.e -0.029      0.185 1828.769 -0.154
## 0.877
## ## prog.s:context.e:partner.e -0.092      0.156 1827.753 -0.589
## 0.556
## ## estr.s:prog.s:context.e:partner.e 0.043      0.858 1843.426  0.050
## 0.960
```

```
confint(model_hlp_EP, method = confint_method) %>% as.data.frame() %>% rownames_to_
column() %>% filter(!is.na(`2.5 %`))
```

```
##                                     rowname    2.5 %    97.5 %
## 1                               (Intercept) 0.36606332 0.485817121
## 2                               estr.s -0.16710558 0.057956296
## 3                               prog.s -0.08524019 0.095487276
## 4                               context.e 0.02416887 0.080021388
## 5                               partner.e 0.03509114 0.255149907
## 6                               session_n -0.02564537 -0.008561301
## 7                               estr.s:prog.s -0.55299537 0.529445071
## 8                               estr.s:context.e -0.15037506 0.212213002
## 9                               prog.s:context.e -0.12662359 0.178998182
## 10                              estr.s:partner.e -0.47973940 -0.030248324
## 11                              prog.s:partner.e -0.11228872 0.248742773
## 12                              context.e:partner.e -0.04124848 0.070456548
## 13                              estr.s:prog.s:context.e -0.68329355 0.998416024
## 14                              estr.s:prog.s:partner.e -0.47164324 1.693129701
## 15                              estr.s:context.e:partner.e -0.39115893 0.334017108
## 16                              prog.s:context.e:partner.e -0.39743528 0.213808085
## 17 estr.s:prog.s:context.e:partner.e -1.63840366 1.725011241
```

## E + P + E\*P: (single women only to interpret interaction)

```
model_h1p_EP_single <- lmer(rating.c ~ estr.s * prog.s * context.e + session_n +
                                (1 | oc_id) +
                                (0 + estr.s:prog.s:context.e || oc_id) +
                                (1 | block:oc_id) +
                                (0 + estr.s:prog.s:context.e || block:oc_id) +
                                (1 | block:oc_id:date) +
                                (0 + context.e || block:oc_id:date),
                                data = filter(data_hormones_partner,
                                              face_sex==theFaceSex,
                                              manip==theManip,
                                              partner.e==0.5),
                                REML = FALSE)
```

```
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : M
odel is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
```

```
summary_h1p_EP_single <- summary(model_h1p_EP_single)
mySummary(summary_h1p_EP_single)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           1213            257              210
##
## [[2]]
##
##
##   rowname       Estimate    Std. Error      df   t value Pr(>|t| )
##   sig
## -----
##   ----
##   ## (Intercept)     0.365      0.034  306.784  10.772  0.000
##   *** 
##   ## estr.s        0.068      0.065 1014.804   1.051  0.294
##   ## prog.s       -0.025      0.059 1001.386  -0.416  0.677
##   ## context.e     0.045      0.017 1194.962   2.694  0.007
##   ** 
##   ## session_n    -0.021      0.005  470.549  -4.094  0.000
##   *** 
##   ## estr.s:prog.s -0.322      0.333 1033.324  -0.967  0.334
##   ## estr.s:context.e 0.044      0.109 1183.279   0.405  0.686
##   ## prog.s:context.e 0.088      0.101 1012.040   0.876  0.381
##   ## estr.s:prog.s:context.e 0.177      0.610   46.506   0.290  0.773
```

```
confint(model_hlp_EP_single, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`^2.5 %`))
```

```
##           rowname    2.5 %    97.5 %
## 1 (Intercept) 0.29837192 0.43109858
## 2 estr.s     -0.05920280 0.19611554
## 3 prog.s     -0.13997796 0.09094831
## 4 context.e   0.01213906 0.07700589
## 5 session_n  -0.03144196 -0.01108290
## 6 estr.s:prog.s -0.97470177 0.33077355
## 7 estr.s:context.e -0.16981893 0.25819050
## 8 prog.s:context.e -0.10915732 0.28576098
## 9 estr.s:prog.s:context.e -1.01941670 1.37332738
```

## E + P + E\*P: (partnered women only to interpret interaction)

```
model_h1p_EP_partnered<- lmer(rating.c ~ estr.s * prog.s * context.e + session_n +
  (1 | oc_id) +
  (0 + estr.s:prog.s:context.e || oc_id) +
  (1 | block:oc_id) +
  (0 + estr.s:prog.s:context.e|| block:oc_id) +
  (1 | block:oc_id:date) +
  (0 + context.e || block:oc_id:date),
  data = filter(data_hormones_partner,
                face_sex==theFaceSex,
                manip==theManip,
                partner.e==0.5),
  REML = FALSE)
summary_h1p_EP_partnered <- summary(model_h1p_EP_partnered)
mySummary(summary_h1p_EP_partnered)
```

```
## [[1]]
## block:oc_id:date      block:oc_id      oc_id
##       616              132            111
##
## [[2]]
##
##
##   rowname           Estimate   Std. Error      df   t value Pr(>|t| )
##   sig
##   -----
##   -----
##   ## (Intercept)     0.476     0.055  156.368   8.605 0.000
##   *** 
##   ## estr.s        -0.184     0.097  509.662  -1.895 0.059
##   +
##   ## prog.s         0.037     0.072  499.467   0.514 0.608
##   ## context.e      0.059     0.023  611.615   2.546 0.011
##   *
##   ## session_n     -0.009     0.008  310.422  -1.128 0.260
##   ## estr.s:prog.s 0.314     0.449  519.964   0.700 0.484
##   ## estr.s:context.e 0.017     0.151  611.925   0.115 0.909
##   ## prog.s:context.e -0.020    0.121  613.555  -0.162 0.871
##   ## estr.s:prog.s:context.e 0.187     0.685  622.045   0.272 0.785
```

```
confint(model_h1p_EP_partnered, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                 rowname      2.5 %     97.5 %
## 1          (Intercept)  0.36739476  0.584110357
## 2             estr.s -0.37352261  0.006266024
## 3             prog.s -0.10441485  0.178561485
## 4            context.e  0.01368495  0.105252094
## 5            session_n -0.02454124  0.006608364
## 6 estr.s:prog.s -0.56618316  1.194798314
## 7 estr.s:context.e -0.27828146  0.312884884
## 8 prog.s:context.e -0.25652139  0.217263246
## 9 estr.s:prog.s:context.e -1.15517448  1.528229097
```

Note that the non-significant effect of estradiol (estimate = -0.18, p = .064) is in the opposite direction to what would be predicted from previous work reporting positive effects of estradiol on preferences.

## E + P + EPratio: (+ partnership status)

```
model_h1p_EP_EPratio <- lmer(rating.c ~ estr.s * context.e * partner.e +
                                prog.s * context.e * partner.e +
                                ep_ratio.s * context.e * partner.e +
                                session_n +
                                (1 | oc_id) +
                                (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || oc_id) +
                                (1 | block:oc_id) +
                                (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || block:oc_id) +
                                (1 | block:oc_id:date) +
                                (0 + context.e || block:oc_id:date),
                                data = filter(data_hormones_partner,
                                              face_sex==theFaceSex,
                                              manip==theManip),
                                REML = FALSE)
summary_h1p_EP_EPratio <- summary(model_h1p_EP_EPratio)
mySummary(summary_h1p_EP_EPratio)
```

```
## [[1]]
## block:oc_id:date          block:oc_id          oc_id
##           1829                  389                 321
##
## [[2]]
##
##
## rowname                         Estimate   Std. Error      df   t value    P
r(>|t|)  sig
## -----
## -----
## (Intercept)                   0.426     0.030    442.881  13.990
0.000 *** 
## estr.s                         -0.047    0.058   1525.690 -0.797
0.426
## context.e                      0.053     0.014   1825.327  3.901
0.000 ***
## partner.e                      0.151     0.056   327.020   2.692
0.007 **
## prog.s                          -0.012    0.052   1490.862 -0.233
0.816
## ep_ratio.s                     -0.006    0.027   1487.415 -0.240
0.811
## session_n                      -0.017    0.004   785.285 -3.943
0.000 ***
## estr.s:context.e               0.040     0.094   1827.758  0.423
0.672
## estr.s:partner.e              -0.211    0.117   1522.741 -1.805
0.071 +
## context.e:partner.e            0.015     0.027   1825.327  0.540
0.589
## context.e:prog.s              0.002     0.089   1827.927  0.027
0.979
## partner.e:prog.s              -0.013    0.105   1493.486 -0.122
0.903
## context.e:ep_ratio.s          -0.031    0.046   1822.168 -0.671
0.502
## partner.e:ep_ratio.s          -0.094    0.054   1487.687 -1.730
0.084 +
## estr.s:context.e:partner.e    -0.053    0.189   1827.758 -0.283
0.777
## context.e:partner.e:prog.s    -0.036    0.179   1827.927 -0.202
0.840
## context.e:partner.e:ep_ratio.s 0.068     0.093   1822.167  0.734
0.463
```

```
confint(model_h1p_EP_EPratio, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`^2.5 %`))
```

```
##          rowname      2.5 %     97.5 %
## 1          (Intercept) 0.36631251 0.485675874
## 2             estr.s -0.16091873 0.067898350
## 3            context.e  0.02660868 0.080334099
## 4            partner.e  0.04098187 0.260297814
## 5             prog.s -0.11479394 0.090443965
## 6            ep_ratio.s -0.05955255 0.046574593
## 7            session_n -0.02571215 -0.008637281
## 8        estr.s:context.e -0.14512286 0.225040882
## 9        estr.s:partner.e -0.43908057 0.018015728
## 10       context.e:partner.e -0.03891104 0.068539748
## 11       context.e:prog.s -0.17276633 0.177511198
## 12       partner.e:prog.s -0.21787043 0.192430750
## 13       context.e:ep_ratio.s -0.12187774 0.059697382
## 14       partner.e:ep_ratio.s -0.19979771 0.012442808
## 15   estr.s:context.e:partner.e -0.42362610 0.316701350
## 16   context.e:partner.e:prog.s -0.38635090 0.314204150
## 17 context.e:partner.e:ep_ratio.s -0.11360621 0.249544098
```

## T + C: (+ partnership status)

Testing for effects of testosterone and cortisol on preferences

```
model_h1p_TC <- lmer(rating.c ~ test.s*context.e*partner.e +
                      cort.s*context.e*partner.e + session_n +
                      (1 | oc_id) +
                      (0 + test.s:context.e + cort.s:context.e || oc_id) +
                      (1 | block:oc_id) +
                      (0 + test.s:context.e + cort.s:context.e || block:oc_id) +
                      (1 | block:oc_id:date) +
                      (0 + context.e || block:oc_id:date),
                      data = filter(data_hormones_partner,
                                    face_sex==theFaceSex,
                                    manip==theManip),
                      REML = FALSE)
summary_h1p_TC <- summary(model_h1p_TC)
mySummary(summary_h1p_TC)
```

```
## [[1]]
## block:oc_id:date          block:oc_id          oc_id
##           1846                  388                 320
##
## [[2]]
##
##
## rowname             Estimate   Std. Error      df   t value Pr(>|
t|) sig
## -----
## (Intercept)       0.424     0.030    432.502   14.022   0.
000 ***  
873  
## test.s        -0.009     0.054    1532.447  -0.160   0.  
000 ***  
## context.e      0.054     0.014    1842.428   3.989   0.  
000 ***  
## partner.e      0.146     0.056    323.959   2.619   0.  
009 **  
## cort.s         0.039     0.042    1528.308   0.925   0.  
355  
## session_n      -0.016    0.004    612.761  -3.823   0.  
000 ***  
## test.s:context.e 0.037     0.090    1840.655   0.408   0.  
684  
## test.s:partner.e -0.061    0.108    1532.345  -0.568   0.  
570  
## context.e:partner.e 0.009     0.027    1842.428   0.312   0.  
755  
## context.e:cort.s   -0.016    0.070    1839.047  -0.235   0.  
814  
## partner.e:cort.s    0.037     0.085    1527.107   0.440   0.  
660  
## test.s:context.e:partner.e 0.007     0.181    1840.654   0.038   0.  
970  
## context.e:partner.e:cort.s 0.004     0.139    1839.048   0.032   0.  
974
```

```
confint(model_hlp_TC, method = confint_method) %>% as.data.frame() %>% rownames_to_
column() %>% filter(!is.na(`2.5 %`))
```

```
##          rowname    2.5 %    97.5 %
## 1      (Intercept) 0.36475511 0.483289751
## 2             test.s -0.11476123 0.097468978
## 3            context.e  0.02765686 0.081094792
## 4           partner.e  0.03673752 0.255209685
## 5             cort.s -0.04385607 0.122297045
## 6        session_n -0.02450491 -0.007893281
## 7 test.s:context.e -0.14046155 0.214214589
## 8 test.s:partner.e -0.27367579 0.150766690
## 9 context.e:partner.e -0.04493114 0.061944661
## 10 context.e:cort.s -0.15283413 0.120149187
## 11 partner.e:cort.s -0.12882631 0.203503403
## 12 test.s:context.e:partner.e -0.34783333 0.361519241
## 13 context.e:partner.e:cort.s -0.26852653 0.277440092
```

## T + C + T\*C: (+ partnership status)

Testing for effects of testosterone and cortisol plus their interaction on preferences

```
# Converges when we model random slopes for individual predictors and the interaction on oc_id
model_hlp_TCi <- lmer(rating.c ~ test.s * context.e * cort.s *partner.e +
  (1 | oc_id) +
  (0 + test.s*cort.s*context.e || oc_id) +
  (1 | block:oc_id) +
  (0 + test.s*cort.s*context.e|| block:oc_id) +
  (1 | block:oc_id:date) +
  (0 + context.e || block:oc_id:date),
  data = filter(data_hormones_partner,
                face_sex==theFaceSex,
                manip==theManip),
  REML = FALSE)
summary_hlp_TCi <- summary(model_hlp_TCi)
mySummary(summary_hlp_TCi)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           1846            388             320
##
## [[2]]
##
##
## rowname                         Estimate   Std. Error    df   t value
## Pr(>|t|)  sig
## -----
## -----
## ## (Intercept)           0.373     0.028  331.790  13.362
## 0.000 *** 
## ## test.s              -0.003     0.056  911.345 -0.052
## 0.959
## ## context.e            0.053     0.019  349.311  2.819
## 0.005 ** 
## ## cort.s              0.025     0.049  19.986  0.504
## 0.620
## ## partner.e            0.143     0.056  331.790  2.555
## 0.011 *
## ## test.s:context.e     0.029     0.084  1500.097 0.345
## 0.730
## ## test.s:cort.s        0.467     0.248  128.086  1.882
## 0.062 +
## ## context.e:cort.s     -0.010    0.067  1058.835 -0.147
## 0.883
## ## test.s:partner.e     -0.034    0.111  911.345 -0.307
## 0.759
## ## context.e:partner.e  0.005     0.038  349.311  0.137
## 0.891
## ## cort.s:partner.e    0.008     0.098  19.986  0.080
## 0.937
## ## test.s:context.e:cort.s -0.079    0.344  119.267 -0.229
## 0.819
## ## test.s:context.e:partner.e 0.028     0.169  1500.097 0.168
## 0.867
## ## test.s:cort.s:partner.e  0.503     0.496  128.086  1.014
## 0.312
## ## context.e:cort.s:partner.e -0.013    0.135  1058.835 -0.097
## 0.923
## ## test.s:context.e:cort.s:partner.e 0.201     0.687  119.267  0.292
## 0.771
```

```
confint(model_hlp_TCi, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`^2.5 %`))
```

```
##                      rowname      2.5 %     97.5 %
## 1                  (Intercept) 0.31861762 0.42815523
## 2                     test.s -0.11170927 0.10597994
## 3                   context.e  0.01613966 0.08982415
## 4                     cort.s -0.07146086 0.12096978
## 5                  partner.e  0.03328075 0.25235597
## 6    test.s:context.e -0.13619310 0.19437599
## 7      test.s:cort.s -0.01938666 0.95301568
## 8    context.e:cort.s -0.14188202 0.12209433
## 9    test.s:partner.e -0.25175183 0.18362660
## 10   context.e:partner.e -0.06851595 0.07885303
## 11      cort.s:partner.e -0.18459377 0.20026750
## 12    test.s:context.e:cort.s -0.75224120 0.59501725
## 13    test.s:context.e:partner.e -0.30227308 0.35886510
## 14      test.s:cort.s:partner.e -0.46928547 1.47551921
## 15      context.e:cort.s:partner.e -0.27705667 0.25089603
## 16 test.s:context.e:cort.s:partner.e -1.14666416 1.54785273
```

## Analyses H1ps: Hormones (+ session order, + partnership status)

### E + P + E\*P: (+ session order, + partnership status)

Testing for effects of estradiol, progesterone, and their interaction on preferences

```
model_h1_EP_partner_s <- lmer(rating.c ~ estr.s * prog.s * context.e * partner.e +
session_n +
  (1 | oc_id) +
  (0 + estr.s:prog.s:context.e || oc_id) +
  (1 | block:oc_id) +
  (0 + estr.s:prog.s:context.e|| block:oc_id) +
  (1 | block:oc_id:date) +
  (0 + context.e || block:oc_id:date),
  data = filter(data_hormones_partner,
                face_sex==theFaceSex,
                manip==theManip),
  REML = FALSE,
  # use nelder-mead optimziation (as in lmer v < 1)
  control = lmerControl(optimizer ="Nelder_Mead"))
summary_h1_EP_partner_s <- summary(model_h1_EP_partner_s)
mySummary(summary_h1_EP_partner_s)
```

```
## [[1]]
## block:oc_id:date          block:oc_id          oc_id
##           1829                  389                 321
##
## [[2]]
##
##
## rowname                         Estimate   Std. Error    df   t value
## Pr(>|t|)  sig
## -----
## -----
## ## (Intercept)             0.426      0.031  448.353  13.942
## 0.000 *** 
## ## estr.s                -0.055      0.057 1524.518 -0.951
## 0.342
## ## prog.s                 0.005      0.046 1500.036  0.111
## 0.912
## ## context.e              0.052      0.014 1824.334  3.656
## 0.000 ***
## ## partner.e              0.145      0.056  331.663  2.585
## 0.010 *
## ## session_n              -0.017      0.004  781.721 -3.924
## 0.000 ***
## ## estr.s:prog.s          -0.012      0.276 1560.202 -0.043
## 0.966
## ## estr.s:context.e       0.031      0.092 1828.769  0.334
## 0.738
## ## prog.s:context.e       0.026      0.078 1827.753  0.336
## 0.737
## ## estr.s:partner.e       -0.255      0.115 1520.185 -2.224
## 0.026 *
## ## prog.s:partner.e       0.068      0.092 1503.985  0.741
## 0.459
## ## context.e:partner.e    0.015      0.028 1824.334  0.512
## 0.608
## ## estr.s:prog.s:context.e 0.158      0.429 1843.423  0.367
## 0.713
## ## estr.s:prog.s:partner.e 0.611      0.552 1559.622  1.106
## 0.269
## ## estr.s:context.e:partner.e -0.029      0.185 1828.769 -0.154
## 0.877
## ## prog.s:context.e:partner.e -0.092      0.156 1827.753 -0.589
## 0.556
## ## estr.s:prog.s:context.e:partner.e 0.043      0.858 1843.426  0.050
## 0.960
```

```
confint(model_h1_EP_partner_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                                     rowname    2.5 %    97.5 %
## 1                               (Intercept) 0.36606332 0.485817121
## 2                               estr.s -0.16710558 0.057956296
## 3                               prog.s -0.08524019 0.095487276
## 4                               context.e 0.02416887 0.080021388
## 5                               partner.e 0.03509114 0.255149907
## 6                               session_n -0.02564537 -0.008561301
## 7                               estr.s:prog.s -0.55299537 0.529445071
## 8                               estr.s:context.e -0.15037506 0.212213002
## 9                               prog.s:context.e -0.12662359 0.178998182
## 10                              estr.s:partner.e -0.47973940 -0.030248324
## 11                              prog.s:partner.e -0.11228872 0.248742773
## 12                              context.e:partner.e -0.04124848 0.070456548
## 13                              estr.s:prog.s:context.e -0.68329355 0.998416024
## 14                              estr.s:prog.s:partner.e -0.47164324 1.693129701
## 15                              estr.s:context.e:partner.e -0.39115893 0.334017108
## 16                              prog.s:context.e:partner.e -0.39743528 0.213808085
## 17 estr.s:prog.s:context.e:partner.e -1.63840366 1.725011241
```

## E + P + E\*P: (single women only to interpret interaction)

```
model_h1_EP_partner_single_s <- lmer(rating.c ~ estr.s * prog.s * context.e + session_n +
                                         (1 | oc_id) +
                                         (0 + estr.s:prog.s:context.e || oc_id) +
                                         (1 | block:oc_id) +
                                         (0 + estr.s:prog.s:context.e|| block:oc_id) +
                                         (1 | block:oc_id:date) +
                                         (0 + context.e || block:oc_id:date),
                                         data = filter(data_hormones_partner,
                                                       face_sex==theFaceSex,
                                                       manip==theManip,
                                                       partner.e===-0.5),
                                         REML = FALSE)
```

```
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, : Model is nearly unidentifiable: large eigenvalue ratio
## - Rescale variables?
```

```
summary_h1_EP_partner_single_s <- summary(model_h1_EP_partner_single_s)
mySummary(summary_h1_EP_partner_single_s)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           1213            257              210
##
## [[2]]
##
##
##   rowname       Estimate    Std. Error      df   t value Pr(>|t| )
##   sig
##   -----
##   ----
##   ## (Intercept)     0.365      0.034  306.784  10.772  0.000
##   *** 
##   ## estr.s        0.068      0.065 1014.804   1.051  0.294
##   ## prog.s       -0.025      0.059 1001.386  -0.416  0.677
##   ## context.e     0.045      0.017 1194.962   2.694  0.007
##   ** 
##   ## session_n    -0.021      0.005  470.549  -4.094  0.000
##   *** 
##   ## estr.s:prog.s -0.322      0.333 1033.324  -0.967  0.334
##   ## estr.s:context.e  0.044      0.109 1183.279   0.405  0.686
##   ## prog.s:context.e  0.088      0.101 1012.040   0.876  0.381
##   ## estr.s:prog.s:context.e  0.177      0.610   46.506   0.290  0.773
```

```
confint(model_h1_EP_partner_single_s, method = confint_method) %>% as.data.frame()
%>% rownames_to_column() %>% filter(!is.na(`^2.5 %`))
```

```
##           rowname    2.5 %    97.5 %
## 1 (Intercept) 0.29837192 0.43109858
## 2 estr.s     -0.05920280 0.19611554
## 3 prog.s     -0.13997796 0.09094831
## 4 context.e   0.01213906 0.07700589
## 5 session_n  -0.03144196 -0.01108290
## 6 estr.s:prog.s -0.97470177 0.33077355
## 7 estr.s:context.e -0.16981893 0.25819050
## 8 prog.s:context.e -0.10915732 0.28576098
## 9 estr.s:prog.s:context.e -1.01941670 1.37332738
```

## E + P + E\*P: (partnered women only to interpret interaction)

```
model_h1_EP_partner_partnered_s <- lmer(rating.c ~ estr.s * prog.s * context.e + session_n +
                                         (1 | oc_id) +
                                         (0 + estr.s:prog.s:context.e || oc_id) +
                                         (1 | block:oc_id) +
                                         (0 + estr.s:prog.s:context.e || block:oc_id) +
                                         (1 | block:oc_id:date) +
                                         (0 + context.e || block:oc_id:date),
                                         data = filter(data_hormones_partner,
                                                       face_sex==theFaceSex,
                                                       manip==theManip,
                                                       partner.e==0.5),
                                         REML = FALSE)
summary_h1_EP_partner_partnered_s <- summary(model_h1_EP_partner_partnered_s)
mySummary(summary_h1_EP_partner_partnered_s)
```

```
## [[1]]
## block:oc_id:date      block:oc_id          oc_id
##           616             132            111
##
## 
## [[2]]
## 
## 
##   rowname           Estimate    Std. Error       df   t value Pr(>|t| )
##   sig
##   -----
##   -----
##   ## (Intercept)     0.476      0.055  156.368   8.605  0.000
##   *** 
##   ## estr.s        -0.184      0.097  509.662  -1.895  0.059
##   +
##   ## prog.s         0.037      0.072  499.467   0.514  0.608
## 
##   ## context.e      0.059      0.023  611.615   2.546  0.011
##   *
##   ## session_n     -0.009      0.008  310.422  -1.128  0.260
## 
##   ## estr.s:prog.s  0.314      0.449  519.964   0.700  0.484
## 
##   ## estr.s:context.e 0.017      0.151  611.925   0.115  0.909
## 
##   ## prog.s:context.e -0.020     0.121  613.555  -0.162  0.871
## 
##   ## estr.s:prog.s:context.e 0.187      0.685  622.045   0.272  0.785
```

```
confint(model_h1_EP_partner_partnered_s, method = confint_method) %>% as.data.frame()
() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                 rowname      2.5 %     97.5 %
## 1          (Intercept)  0.36739476  0.584110357
## 2             estr.s -0.37352261  0.006266024
## 3             prog.s -0.10441485  0.178561485
## 4            context.e  0.01368495  0.105252094
## 5            session_n -0.02454124  0.006608364
## 6 estr.s:prog.s -0.56618316  1.194798314
## 7 estr.s:context.e -0.27828146  0.312884884
## 8 prog.s:context.e -0.25652139  0.217263246
## 9 estr.s:prog.s:context.e -1.15517448  1.528229097
```

Note that the non-significant effect of estradiol (estimate = -0.18, p = .064) is in the opposite direction to what would be predicted from previous work reporting positive effects of estradiol.

## E + P + EPratio: (+ session order, + partnership status)

```
model_h1p_EP_EPratio_s <- lmer(rating.c ~ estr.s * context.e * partner.e +
                                    prog.s * context.e * partner.e +
                                    ep_ratio.s * context.e * partner.e +
                                    session_n +
                                    (1 | oc_id) +
                                    (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || oc_id) +
                                    (1 | block:oc_id) +
                                    (0 + estr.s:context.e + prog.s:context.e + ep_ratio.s:context.e || block:oc_id) +
                                    (1 | block:oc_id:date) +
                                    (0 + context.e || block:oc_id:date),
                                    data = filter(data_hormones_partner,
                                                  face_sex==theFaceSex,
                                                  manip==theManip),
                                    REML = FALSE)
summary_h1p_EP_EPratio_s <- summary(model_h1p_EP_EPratio_s)
mySummary(summary_h1p_EP_EPratio_s)
```

```
## [[1]]
## block:oc_id:date          block:oc_id          oc_id
##           1829                  389                 321
##
## [[2]]
##
##
## rowname                         Estimate   Std. Error      df   t value    P
r(>|t|)  sig
## -----
## -----
## (Intercept)                   0.426     0.030    442.881  13.990
## estr.s                         -0.047    0.058   1525.690 -0.797
## context.e                      0.053     0.014   1825.327  3.901
## partner.e                      0.151     0.056   327.020   2.692
## prog.s                         -0.012    0.052   1490.862 -0.233
## ep_ratio.s                     0.816     0.811   1487.415 -0.240
## session_n                      0.000     0.000   785.285 -3.943
## estr.s:context.e               0.672     0.040   1827.758  0.423
## estr.s:partner.e               0.071     -0.211   1522.741 -1.805
## context.e:partner.e            0.589     0.015   1825.327  0.540
## context.e:prog.s               0.979     0.002   1827.927  0.027
## partner.e:prog.s              0.903     -0.013   1493.486 -0.122
## context.e:ep_ratio.s           0.502     -0.031   1822.168 -0.671
## partner.e:ep_ratio.s           0.084     -0.094   1487.687 -1.730
## estr.s:context.e:partner.e    0.777     -0.053   1827.758 -0.283
## context.e:partner.e:prog.s    0.840     -0.036   1827.927 -0.202
## context.e:partner.e:ep_ratio.s 0.463     0.068   1822.167  0.734
```

```
confint(model_hlp_EP_EPratio_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`^2.5 %`))
```

```
##          rowname      2.5 %     97.5 %
## 1          (Intercept) 0.36631251 0.485675874
## 2             estr.s -0.16091873 0.067898350
## 3            context.e  0.02660868 0.080334099
## 4            partner.e  0.04098187 0.260297814
## 5             prog.s -0.11479394 0.090443965
## 6            ep_ratio.s -0.05955255 0.046574593
## 7            session_n -0.02571215 -0.008637281
## 8        estr.s:context.e -0.14512286 0.225040882
## 9        estr.s:partner.e -0.43908057 0.018015728
## 10       context.e:partner.e -0.03891104 0.068539748
## 11       context.e:prog.s -0.17276633 0.177511198
## 12       partner.e:prog.s -0.21787043 0.192430750
## 13       context.e:ep_ratio.s -0.12187774 0.059697382
## 14       partner.e:ep_ratio.s -0.19979771 0.012442808
## 15   estr.s:context.e:partner.e -0.42362610 0.316701350
## 16   context.e:partner.e:prog.s -0.38635090 0.314204150
## 17 context.e:partner.e:ep_ratio.s -0.11360621 0.249544098
```

## T + C: (+ session order, + partnership status)

Testing for effects of testosterone and cortisol on preferences

```
model_h1p_TC_s <- lmer(rating.c ~ test.s*context.e*partner.e +
                           cort.s*context.e*partner.e + session_n +
                           (1 | oc_id) +
                           (0 + test.s:context.e + cort.s:context.e || oc_id) +
                           (1 | block:oc_id) +
                           (0 + test.s:context.e + cort.s:context.e || block:oc_id) +
                           (1 | block:oc_id:date) +
                           (0 + context.e || block:oc_id:date),
                           data = filter(data_hormones_partner,
                                         face_sex==theFaceSex,
                                         manip==theManip),
                           REML = FALSE)
summary_h1p_TC_s <- summary(model_h1p_TC_s)
mySummary(summary_h1p_TC_s)
```

```
## [[1]]
## block:oc_id:date          block:oc_id          oc_id
##           1846                  388                 320
##
## [[2]]
##
##
## rowname             Estimate   Std. Error      df   t value Pr(>|
t|) sig
## -----
## (Intercept)       0.424     0.030    432.502   14.022   0.
000 ***  
873  
## test.s        -0.009     0.054   1532.447  -0.160   0.  
000 ***  
## context.e      0.054     0.014   1842.428   3.989   0.  
000 ***  
## partner.e      0.146     0.056   323.959   2.619   0.  
009 **  
## cort.s         0.039     0.042   1528.308   0.925   0.  
355  
## session_n      -0.016    0.004    612.761  -3.823   0.  
000 ***  
## test.s:context.e 0.037     0.090   1840.655   0.408   0.  
684  
## test.s:partner.e -0.061    0.108   1532.345  -0.568   0.  
570  
## context.e:partner.e 0.009     0.027   1842.428   0.312   0.  
755  
## context.e:cort.s   -0.016    0.070   1839.047  -0.235   0.  
814  
## partner.e:cort.s    0.037     0.085   1527.107   0.440   0.  
660  
## test.s:context.e:partner.e 0.007     0.181   1840.654   0.038   0.  
970  
## context.e:partner.e:cort.s 0.004     0.139   1839.048   0.032   0.  
974
```

```
confint(model_hlp_TC_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##          rowname    2.5 %    97.5 %
## 1      (Intercept) 0.36475511 0.483289751
## 2             test.s -0.11476123 0.097468978
## 3            context.e  0.02765686 0.081094792
## 4           partner.e  0.03673752 0.255209685
## 5             cort.s -0.04385607 0.122297045
## 6        session_n -0.02450491 -0.007893281
## 7 test.s:context.e -0.14046155 0.214214589
## 8 test.s:partner.e -0.27367579 0.150766690
## 9 context.e:partner.e -0.04493114 0.061944661
## 10 context.e:cort.s -0.15283413 0.120149187
## 11 partner.e:cort.s -0.12882631 0.203503403
## 12 test.s:context.e:partner.e -0.34783333 0.361519241
## 13 context.e:partner.e:cort.s -0.26852653 0.277440092
```

## T + C + T\*C: (+ session order, + partnership status)

```
model_h1p_TCi_s <- lmer(rating.c ~ cort.s*test.s*context.e*partner.e + session_n +
                           (1 | oc_id) +
                           (0 + test.s:cort.s:context.e || oc_id) +
                           (1 | block:oc_id) +
                           (0 + test.s:cort.s:context.e || block:oc_id) +
                           (1 | block:oc_id:date) +
                           (0 + context.e || block:oc_id:date),
                           data = filter(data_hormones_partner,
                                         face_sex==theFaceSex,
                                         manip==theManip),
                           REML = FALSE)
summary_h1p_TCi_s <- summary(model_h1p_TCi_s)
mySummary(summary_h1p_TCi_s)
```

```
## [[1]]
## block:oc_id:date          block:oc_id          oc_id
##           1846                  388                 320
##
## [[2]]
##
##
## rowname                         Estimate   Std. Error    df   t value
## Pr(>|t|)  sig
## -----
## -----
## ## (Intercept)             0.419      0.030  439.928  13.801
## 0.000 *** 
## ## cort.s                0.025      0.043  1529.942   0.586
## 0.558
## ## test.s                -0.004     0.054  1532.392  -0.072
## 0.943
## ## context.e              0.056      0.014  1840.561   3.850
## 0.000 ***
## ## partner.e              0.141      0.056  331.147   2.518
## 0.012 *
## ## session_n              -0.016     0.004  612.014  -3.801
## 0.000 ***
## ## cort.s:test.s          0.375      0.221  1595.136   1.702
## 0.089 +
## ## cort.s:context.e       -0.010     0.072  1111.058  -0.138
## 0.890
## ## test.s:context.e       0.037      0.091  1756.212   0.401
## 0.688
## ## cort.s:partner.e       0.025      0.087  1529.013   0.289
## 0.773
## ## test.s:partner.e       -0.049     0.108  1532.276  -0.453
## 0.650
## ## context.e:partner.e    0.006      0.029  1840.561   0.209
## 0.835
## ## cort.s:test.s:context.e -0.139     0.374   67.938  -0.370
## 0.712
## ## cort.s:test.s:partner.e 0.483      0.441  1594.468   1.096
## 0.273
## ## cort.s:context.e:partner.e -0.003     0.145  1111.058  -0.019
## 0.985
## ## test.s:context.e:partner.e -0.015     0.182  1756.212  -0.081
## 0.936
## ## cort.s:test.s:context.e:partner.e 0.086      0.748   67.938   0.114
## 0.909
```

```
confint(model_hlp_TCi_s, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                                     rowname    2.5 %    97.5 %
## 1                               (Intercept) 0.35955593 0.478590412
## 2                               cort.s -0.05958307 0.110410402
## 3                               test.s -0.11012168 0.102341259
## 4                               context.e 0.02728460 0.083879674
## 5                               partner.e 0.03124191 0.250749902
## 6                               session_n -0.02439409 -0.007795043
## 7                               cort.s:test.s -0.05678580 0.807712891
## 8                               cort.s:context.e -0.15196333 0.131962503
## 9                               test.s:context.e -0.14199738 0.215049877
## 10                              cort.s:partner.e -0.14496449 0.195036104
## 11                              test.s:partner.e -0.26160416 0.163306340
## 12                              context.e:partner.e -0.05056463 0.062625491
## 13                              cort.s:test.s:context.e -0.87196470 0.594855233
## 14                              cort.s:test.s:partner.e -0.38126739 1.347856324
## 15                              cort.s:context.e:partner.e -0.28670343 0.281148225
## 16                              test.s:context.e:partner.e -0.37174946 0.342345359
## 17 cort.s:test.s:context.e:partner.e -1.38122168 1.552418569
```

## Hypothesis 2

Do women not using hormonal contraceptives show stronger preferences than women using the combined oral contraceptive pill?

One of these women was excluded from analyses because she did not complete any male face preference tests.

## Descriptive stats: data\_between

```
# create mean DV for all ratings by oc_id
stats_overall <- filter(data_between, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id) %>%
  summarise(
    overall_rating.c = mean(rating.c)
  ) %>%
  ungroup() %>%
  group_by() %>%
  summarise(
    context = "overall",
    pill.e = "overall",
    n= n_distinct(oc_id),
    mean_dv = mean(overall_rating.c),
    sd_dv = sd(overall_rating.c),
    se_dv = se(overall_rating.c)
  )

# create mean DV splitting by context
stats_context <- filter(data_between, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id, context) %>%
  summarise(
    context_rating.c = mean(rating.c)
  ) %>%
  group_by(context) %>%
  summarise(
    pill.e = "overall",
    n= n_distinct(oc_id),
    mean_dv = mean(context_rating.c),
    sd_dv = sd(context_rating.c),
    se_dv = se(context_rating.c)
  )

# stats by pill.e
stats_pill <- filter(data_between, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id, pill.e) %>%
  summarise(
    overall_rating.c = mean(rating.c)
  ) %>%
  ungroup() %>%
  group_by(pill.e) %>%
  summarise(
    context = "overall",
    n= n_distinct(oc_id),
    mean_dv = mean(overall_rating.c),
    sd_dv = sd(overall_rating.c),
    se_dv = se(overall_rating.c)
  ) %>% select(context, pill.e, n, mean_dv, sd_dv, se_dv)

rbind(stats_overall, rbind(stats_context, stats_pill))
```

```
## # A tibble: 5 x 6
##   context pill.e     n   mean_dv    sd_dv    se_dv
## * <chr>  <chr> <int>      <dbl>      <dbl>      <dbl>
## 1 overall overall    538  0.3976217  0.4851936  0.02091818
## 2       LT overall    538  0.3755005  0.5137469  0.02214920
## 3       ST overall    538  0.4198717  0.5229769  0.02254714
## 4 overall     -0.5    326  0.3508539  0.4837502  0.02679245
## 5 overall      0.5    212  0.4695382  0.4796947  0.03294557
```

## Analyses H2: Pill

```
model_h2 <- lmer(rating.c ~ pill.e * context.e +
  (1 | oc_id) +
  (0 + context.e || oc_id) +
  (1 | oc_id:block) +
  (0 + context.e || oc_id:block) +
  (1 | oc_id:block:date) +
  (0 + context.e || oc_id:block:date),
  data=filter(data_between, face_sex==theFaceSex, manip==theManip),
  REML=FALSE)
summary_h2 <- summary(model_h2)
mySummary(summary_h2)
```

```
## [[1]]
## oc_id:block:date          oc_id:block           oc_id
##            3200                  681                 538
##
## [[2]]
##
##
##   rowname    Estimate   Std. Error      df   t value Pr(>|t|)    sig
##   -----  -----  -----  -----  -----  -----
##   (Intercept)  0.407     0.021  538.264  19.221  0.000 *** 
##   pill.e      0.116     0.042  538.264   2.747  0.006 ** 
##   context.e    0.047     0.016  540.559   2.954  0.003 ** 
##   pill.e:context.e -0.027    0.032  540.559  -0.866  0.387
```

```
confint(model_h2, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`^2.5 %`))
```

```
##           rowname    2.5 %    97.5 %
## 1 (Intercept) 0.36507506 0.44798122
## 2     pill.e  0.03329955 0.19911186
## 3   context.e  0.01575287 0.07790052
## 4 pill.e:context.e -0.08959624 0.03469907
```

Note that the effect of OCP use (estimate = +0.116, p = .006) is in the opposite direction to what would be expected if fertility is positively associated with preference.

# Analyses H2p: Pill (+ partnership status)

```
model_h2p <- lmer(rating.c ~ pill.e * context.e * partner.e +
                    (1 | oc_id) +
                    (0 + context.e || oc_id) +
                    (1 | oc_id:block) +
                    (0 + context.e || oc_id:block) +
                    (1 | oc_id:block:date) +
                    (0 + context.e || oc_id:block:date),
                    data=filter(data_between_partner, face_sex==theFaceSex, manip=
                    =theManip),
                    REML=FALSE)
summary_h2p <- summary(model_h2p)
mySummary(summary_h2p)
```

```
## [[1]]
## oc_id:block:date      oc_id:block      oc_id
##          2798             600            492
##
## [[2]]
##
##
##   rowname           Estimate    Std. Error      df   t value Pr(>|t|
## |)  sig
## -----
## (Intercept)       0.436     0.023  491.341 18.884 0.0
## 00 ***          0.119     0.046  491.341  2.583 0.0
## 10 *            0.049     0.017  483.540  2.813 0.0
## 05 **           0.014     0.046  491.341  0.304 0.7
## 61               0.014     0.046  491.341  0.304 0.7
## 80               -0.019    0.035  483.540 -0.554 0.5
## 80               -0.247    0.092  491.341 -2.672 0.0
## 08 **           0.068     0.035  483.540  1.952 0.0
## 51 +            0.090     0.069  483.540  1.295 0.1
## 96
```

```
confint(model_h2p, method = confint_method) %>% as.data.frame() %>% rownames_to_col
umn() %>% filter(!is.na(`2.5 %`))
```

```
##                      rowname      2.5 %     97.5 %
## 1          (Intercept)  0.3903766689  0.48079293
## 2             pill.e   0.0287428868  0.20957542
## 3        context.e   0.0147775065  0.08269070
## 4       partner.e  -0.0763792100  0.10445332
## 5  pill.e:context.e -0.0871204643  0.04870591
## 6  pill.e:partner.e -0.4273560163 -0.06569095
## 7 context.e:partner.e -0.0002664157  0.13555996
## 8 pill.e:context.e:partner.e -0.0460820309  0.22557073
```

## Pill (single women only to interpret interaction)

```
model_h2p_single <- lmer(rating.c ~ pill.e * context.e +
                           (1 | oc_id) +
                           (0 + context.e || oc_id) +
                           (1 | oc_id:block) +
                           (0 + context.e || oc_id:block) +
                           (1 | oc_id:block:date) +
                           (0 + context.e || oc_id:block:date),
                           data=filter(data_between_partner, face_sex==theFaceSex, partner
                           .e===-0.5, manip==theManip),
                           REML=FALSE,
                           # use nelder-mead optimziation (as in lmer v < 1)
                           control = lmerControl(optimizer ="Nelder_Mead")
                           )
summary_h2p_single <- summary(model_h2p_single)
mySummary(summary_h2p_single)
```

```
## [[1]]
## oc_id:block:date      oc_id:block      oc_id
##           1489            319            259
##
## [[2]]
##
##
##   rowname    Estimate   Std. Error      df   t value Pr(>|t|)    sig
##   -----  -----  -----  -----  -----  -----
##   (Intercept)  0.428     0.032  251.741  13.304  0.000 *** 
##   pill.e      0.243     0.064  251.741   3.774  0.000 *** 
##   context.e    0.016     0.022  252.842   0.743  0.458    
##   pill.e:context.e -0.063     0.044  252.842  -1.439  0.151
```

```
confint(model_h2p_single, method = confint_method) %>% as.data.frame() %>% rownames
_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                      rowname      2.5 %     97.5 %
## 1          (Intercept)  0.36525917  0.49147636
## 2             pill.e   0.11678808  0.36922246
## 3        context.e  -0.02663534  0.05917798
## 4  pill.e:context.e -0.14883821  0.02278844
```

## Pill (partnered women only to interpret interaction)

```
model_h2p_partnered <- lmer(rating.c ~ pill.e * context.e +
                               (1 | oc_id) +
                               (0 + context.e || oc_id) +
                               (1 | oc_id:block) +
                               (0 + context.e || oc_id:block) +
                               (1 | oc_id:block:date) +
                               (0 + context.e || oc_id:block:date),
                               data=filter(data_between_partner,
                                           face_sex==theFaceSex,
                                           partner.e== 0.5,
                                           manip==theManip),
                               REML=FALSE)
summary_h2p_partnered <- summary(model_h2p_partnered)
mySummary(summary_h2p_partnered)
```

```
## [[1]]
## oc_id:block:date      oc_id:block          oc_id
##           1309             281              233
##
## [[2]]
##
##
##   rowname    Estimate   Std. Error     df   t value Pr(>|t|) sig
##   -----  -----
##   (Intercept) 0.443     0.033  236.351 13.463 0.000 *** 
##   pill.e      -0.004    0.066  236.351 -0.062 0.950  
##   context.e    0.082     0.027  231.207  3.066 0.002 ** 
##   pill.e:context.e 0.026     0.053  231.207  0.493 0.622
```

```
confint(model_h2p_partnered, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`^2.5 %`))
```

```
##       rowname    2.5 %   97.5 %
## 1 (Intercept) 0.37822773 0.5071121
## 2     pill.e -0.13297648 0.1247922
## 3   context.e  0.02954214 0.1342130
## 4 pill.e:context.e -0.07832164 0.1310201
```

## Hypothesis 3

Do preferences of women using the combined oral contraceptive pill change when they are taking inactive pills?

pill\_break.e coding: using active pill= 0.5, using inactive pill/being on a pill break = -0.5

## Descriptive stats: data\_pillbreak

```
# create mean DV for all ratings by oc_id
stats_overall <- filter(data_pillbreak, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id) %>%
  summarise(
    overall_rating.c = mean(rating.c)
  ) %>%
  ungroup() %>%
  group_by() %>%
  summarise(
    context = "overall",
    n= n_distinct(oc_id),
    mean_dv = mean(overall_rating.c),
    sd_dv = sd(overall_rating.c),
    se_dv = se(overall_rating.c)
  )
)

# create mean DV splitting by context
stats_context <- filter(data_pillbreak, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id, context) %>%
  summarise(
    context_rating.c = mean(rating.c)
  ) %>%
  group_by(context) %>%
  summarise(
    n= n_distinct(oc_id),
    mean_dv = mean(context_rating.c),
    sd_dv = sd(context_rating.c),
    se_dv = se(context_rating.c)
  )
)

rbind(stats_overall, stats_context)
```

```
## # A tibble: 3 x 5
##   context     n   mean_dv    sd_dv    se_dv
## * <chr> <int>    <dbl>    <dbl>    <dbl>
## 1 overall    173  0.4380490  0.4488670  0.03412672
## 2       LT    173  0.4112107  0.4790548  0.03642186
## 3       ST    173  0.4650578  0.5110369  0.03885342
```

## Analyses H3: Pill-break

```
model_h3 <- lmer(rating.c ~ context.e*pill_break.e +
  (1 | oc_id) +
  (0 + pill_break.e:context.e || oc_id) +
  (1 | oc_id:block) +
  (0 + pill_break.e:context.e || oc_id:block ) +
  (1 | oc_id:block:date) +
  (0 + context.e || oc_id:block:date ),
  data=filter(data_pillbreak, face_sex==theFaceSex, manip==theManip)
'
REML=FALSE)
summary_h3 <- summary(model_h3)
mySummary(summary_h3)
```

```
## [[1]]
## oc_id:block:date      oc_id:block          oc_id
##           1019            206             173
##
## [[2]]
##
##
##   rowname       Estimate    Std. Error      df   t value  Pr(>|t|) 
sig
## -----
## (Intercept) 0.436        0.034     179.486 12.664 0.000
*** 
## context.e   0.057        0.021     935.351  2.726 0.007
** 
## pill_break.e 0.014        0.022     832.882  0.635 0.526
## context.e:pill_break.e 0.007        0.047     208.006  0.149 0.882
```

```
confint(model_h3, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##           rowname 2.5 % 97.5 %
## 1 (Intercept) 0.36874879 0.50378236
## 2 context.e   0.01598821 0.09781643
## 3 pill_break.e -0.02932669 0.05743556
## 4 context.e:pill_break.e -0.08536960 0.09937692
```

## Analyses H3p: Pill-break (+ partnership status)

```
model_h3p<-lmer(rating.c ~ context.e*pill_break.e * partner.e +
                    (1 | oc_id) +
                    (0 + pill_break.e*context.e || oc_id) +
                    (1 | oc_id:block) +
                    (0 + pill_break.e*context.e || oc_id:block ) +
                    (1 | oc_id:block:date) +
                    (0 + context.e || oc_id:block:date ),
                    data=filter(data_pillbreak_partner,
                                face_sex==theFaceSex,
                                manip==theManip),
                    REML=FALSE)
summary_h3p <- summary(model_h3p)
mySummary(summary_h3p)
```

```
## [[1]]
## oc_id:block:date      oc_id:block      oc_id
##          989             200            168
##
## [[2]]
##
##
## rowname                         Estimate   Std. Error    df   t value
Pr(>|t|)  sig
## -----
## (Intercept)                 0.431     0.035  180.905  12.298
  0.000 ***

## context.e                     0.039     0.036  187.100   1.098
  0.274

## pill_break.e                  0.019     0.024  147.433   0.805
  0.422

## partner.e                     0.016     0.059  163.947   0.277
  0.782

## context.e:pill_break.e       0.001     0.040  203.777   0.019
  0.985

## context.e:partner.e         0.145     0.070  205.922   2.087
  0.038 *

## pill_break.e:partner.e      -0.023     0.048  184.083  -0.488
  0.626

## context.e:pill_break.e:partner.e -0.091     0.079  203.849  -1.147
  0.253
```

```
confint(model_h3p, method = confint_method) %>% as.data.frame() %>% rownames_to_col
umn() %>% filter(!is.na(`2.5 %`))
```

```
##          rowname      2.5 %    97.5 %
## 1          (Intercept) 0.362297291 0.49966784
## 2          context.e -0.030972377 0.10982504
## 3          pill_break.e -0.027716868 0.06631501
## 4          partner.e -0.099015064 0.13157370
## 5          context.e:pill_break.e -0.076724321 0.07818687
## 6          context.e:partner.e  0.008806622 0.28148741
## 7          pill_break.e:partner.e -0.117070412 0.07035938
## 8 context.e:pill_break.e:partner.e -0.245577842 0.06429855
```

## Pill break (single women only to interpret interaction)

```
model_h3p_single <- lmer(rating.c ~ context.e*pill_break.e +
                           (1 | oc_id) +
                           (0 + pill_break.e*context.e || oc_id) +
                           (1 | oc_id:block) +
                           (0 + pill_break.e*context.e || oc_id:block ) +
                           (1 | oc_id:block:date) +
                           (0 + context.e || oc_id:block:date ),
                           data=filter(data_pillbreak_partner, partner.e=="-0.5",
                                       face_sex==theFaceSex, manip==theManip),
                           REML=FALSE)
summary_h3p_single <- summary(model_h3p_single)
mySummary(summary_h3p_single)
```

```
## [[1]]
## oc_id:block:date      oc_id:block      oc_id
##            335                  68                  61
##
## [[2]]
##
##
##   rowname      Estimate   Std. Error      df   t value Pr(>|t| )
## sig
##   -----
##   (Intercept) 0.439       0.059     62.169    7.409   0.000
## ***
##   context.e -0.038       0.039     67.475   -0.975   0.333
##
##   pill_break.e 0.033       0.038    275.893    0.864   0.388
##
##   context.e:pill_break.e 0.053       0.060     58.735    0.876   0.385
```

```
confint(model_h3p_single, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                 rowname      2.5 %     97.5 %
## 1          (Intercept)  0.32284883  0.55509562
## 2           context.e -0.11439962  0.03839673
## 3        pill_break.e -0.04168064  0.10746122
## 4 context.e:pill_break.e -0.06543218  0.17114999
```

## Pill break (partnered women only to interpret interaction)

```
model_h3p_partnered <- lmer(rating.c ~ context.e*pill_break.e + 
                               (1 | oc_id) +
                               (0 + pill_break.e*context.e || oc_id) +
                               (1 | oc_id:block) +
                               (0 + pill_break.e*context.e || oc_id:block ) +
                               (1 | oc_id:block:date) +
                               (0 + context.e || oc_id:block:date ),
                               data=filter(data_pillbreak_partner,
                                           partner.e=="0.5",
                                           face_sex==theFaceSex,
                                           manip==theManip),
                               REML=FALSE,
                               # use nelder-mead optimziation (as in lmer v < 1)
                               control = lmerControl(optimizer ="Nelder_Mead"))
summary_h3p_partnered <- summary(model_h3p_partnered)
mySummary(summary_h3p_partnered)
```

```
## [[1]]
## oc_id:block:date      oc_id:block      oc_id
##       654                  132            114
##
## [[2]]
##
##
##   rowname      Estimate    Std. Error      df   t value Pr(>|t| )
## sig
##   -----
##   (Intercept)  0.413       0.040    119.511  10.338 0.000
## ***
##   context.e    0.109       0.047    121.140   2.290 0.024
## *
##   pill_break.e 0.004       0.030    99.365   0.118 0.906
##   context.e:pill_break.e -0.049       0.047   130.407  -1.031 0.304
```

```
confint(model_h3p_partnered, method = confint_method) %>% as.data.frame() %>% rownames_to_column() %>% filter(!is.na(`2.5 %`))
```

```
##                 rowname      2.5 %     97.5 %
## 1          (Intercept)  0.33505338 0.49180959
## 2           context.e   0.01564686 0.20152335
## 3      pill_break.e -0.05493096 0.06198526
## 4 context.e:pill_break.e -0.14125492 0.04385052
```

# Hypothesis 4

Do preferences change when women start or stop using the combined oral contraceptive pill?

## Descriptive stats: data\_pill\_switchers

```
# create mean DV for all ratings by oc_id
stats_overall <- filter(data_pill_switchers, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id) %>%
  summarise(
    overall_rating.c = mean(rating.c)
  ) %>%
  ungroup() %>%
  group_by() %>%
  summarise(
    context = "overall",
    n= n_distinct(oc_id),
    mean_dv = mean(overall_rating.c),
    sd_dv = sd(overall_rating.c),
    se_dv = se(overall_rating.c)
  )

# create mean DV splitting by context
stats_context <- filter(data_pill_switchers, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id, context) %>%
  summarise(
    context_rating.c = mean(rating.c)
  ) %>%
  group_by(context) %>%
  summarise(
    n= n_distinct(oc_id),
    mean_dv = mean(context_rating.c),
    sd_dv = sd(context_rating.c),
    se_dv = se(context_rating.c)
  )

  rbind(stats_overall, stats_context)
```

```
## # A tibble: 3 x 5
##   context     n   mean_dv    sd_dv    se_dv
## * <chr> <int>     <dbl>     <dbl>     <dbl>
## 1 overall     45  0.3643366  0.4321600  0.06442261
## 2 LT         45  0.3365961  0.4811474  0.07172522
## 3 ST         45  0.3917319  0.4190430  0.06246725
```

## Interval between pill use and non-use testing blocks

```
switchers_date_diffs <- filter(data_pill_switchers, face_sex==theFaceSex, manip==theManip) %>%
  group_by(oc_id, block_pill, direction.e) %>%
  summarise(
    min_date = min(date),
    max_date = max(date),
    the_date = ifelse(
      mean(direction.e) == -.5,
      ifelse(block_pill == 0, max_date, min_date),
      ifelse(block_pill == 1, max_date, min_date)
    )
  ) %>%
  ungroup() %>%
  mutate(block_pill = paste0("hc", block_pill)) %>%
  select(oc_id, block_pill, the_date) %>%
  spread(block_pill, the_date) %>%
  mutate( date_diff = abs(interval(ymd(hc0), ymd(hc1)) / ddays(1)))

switchers_date_diffs %>%
  group_by() %>%
  summarise(
    mean = mean(date_diff),
    sd = sd(date_diff),
    se = se(date_diff),
    min = min(date_diff),
    max = max(date_diff)
  ) %>% gather("date_diff", "value", 1:length(.)) %>%
  mutate(value = round(value, 4))
```

```
## # A tibble: 5 x 2
##   date_diff     value
##       <chr>     <dbl>
## 1     mean  360.3111
## 2       sd  282.0838
## 3       se   42.0506
## 4       min  56.0000
## 5       max 1113.0000
```

## Analyses H4: Pill-switch

Within-subject change in pill use: (not considering partnership status)

```
model_h4 <- lmer(rating.c ~ pill.e*context.e + direction.e +
                  (1 | oc_id) +
                  (0 + context.e:pill.e || oc_id) +
                  (1 | oc_id:block) +
                  (0 + context.e || oc_id:block) +
                  (1 | oc_id:block:date) +
                  (0 + context.e || oc_id:block:date),
                  data=filter(data_pill_switchers, face_sex==theFaceSex, manip==
theManip),
                  REML=FALSE)
summary_h4 <- summary(model_h4)
mySummary(summary_h4)
```

```
## [[1]]
## oc_id:block:date      oc_id:block          oc_id
##           529             107                45
##
## 
## [[2]]
##
##
##   rowname    Estimate   Std. Error      df   t value  Pr(>|t|) sig
##   -----  -----  -----  -----  -----  -----  -----
##   (Intercept) 0.389     0.067  45.605  5.826  0.000 *** 
##   pill.e      0.077     0.049  63.381  1.570  0.121  
##   context.e    0.056     0.031  107.290 1.813  0.073 +  
##   direction.e  0.147     0.134  45.848  1.102  0.276  
##   pill.e:context.e 0.087     0.061  107.290 1.424  0.157
```

```
confint(model_h4, method = confint_method) %>% as.data.frame() %>% rownames_to_colu
mn() %>% filter(!is.na(`2.5 %`))
```

```
##           rowname    2.5 %   97.5 %
## 1 (Intercept) 0.258416339 0.5204390
## 2 pill.e     -0.019123218 0.1732065
## 3 context.e   -0.004497919 0.1155415
## 4 direction.e -0.114877104 0.4098350
## 5 pill.e:context.e -0.032826771 0.2072520
```

## Analyses H4p: Pill-switch (+ partnership status change)

Within-subject change in pill use: (considering possible effects of change in partnership status)

```
model_h4p <- lmer(rating.c ~ pill.e * context.e + partner.e + direction.e +
  (1 | oc_id) +
  (0 + context.e:pill.e + partner.e || oc_id) +
  (1 | oc_id:block) +
  (0 + context.e || oc_id:block) +
  (1 | oc_id:block:date) +
  (0 + context.e || oc_id:block:date),
  data=filter(data_pill_switchers_partner,
              face_sex==theFaceSex,
              manip==theManip),
  REML=FALSE)
summary_h4p <- summary(model_h4p)
mySummary(summary_h4p)
```

```
## [[1]]
## oc_id:block:date      oc_id:block      oc_id
##          399                  81                 35
##
## 
## [[2]]
##
##
##   rowname    Estimate   Std. Error      df   t value Pr(>|t|) sig
##   -----  -----
##   (Intercept) 0.453     0.078  35.645  5.822  0.000 *** 
##   pill.e      0.005     0.063  49.679  0.076  0.940  
##   context.e    0.027     0.027  81.046  0.975  0.332  
##   partner.e    0.123     0.088  35.706  1.392  0.172  
##   direction.e  0.179     0.156  35.712  1.149  0.258  
##   pill.e:context.e 0.021     0.055  81.046  0.389  0.698
```

```
confint(model_h4p, method = confint_method) %>% as.data.frame() %>% rownames_to_col
umn() %>% filter(!is.na(`2.5 %`))
```

```
##       rowname    2.5 %    97.5 %
## 1 (Intercept) 0.30043773 0.60539761
## 2   pill.e -0.11920153 0.12879195
## 3 context.e -0.02685847 0.08002384
## 4   partner.e -0.04994438 0.29496199
## 5 direction.e -0.12624575 0.48381275
## 6 pill.e:context.e -0.08568008 0.12808420
```