

# 1 Introduction

Temperature and light cue numerous biological processes including....*list here*(). These cues often interact in complex ways and a major goal of biology is to quantify their both individual and interactive effects on biological activities (). This effort has only become more critical in recent decades as this such measurements have become essential for accurately predicting organisms' response to anthropogenic global change and informing numerous mitigation and adaptation strategies ().

It is extremely difficult to partition the individual effects of temperature and light cues and their interactions based on *in situ* observational studies because temperature and light variables tend to covary in nature (). However, well designed experiments in controlled environments (e.g. growth chambers, greenhouses, mesocosm) can be used to manipulate temperature and and light variable independently allowing for robust comparisons of their individual and interactive contributions to biological processes. Indeed this approach has provided many fruitful insights regarding temperature and light signaling over the last century () and has great potential to continue to advance fundamental and applied biological inquiries in the decades to come ().

However, controlled environment studies have their own challenges. Experimentalists must balance biological realism with statistical inference (), experimental effort with statistical power () and account for the effects of unmanipulated or unmeasured variables (). Below we highlight a particular challenge that can arise when experiments seeking to estimate the interactive effects of temperature and daylength on biological outcomes. Our example deals with the phenology (timing of recurring life cycle events, e.g. leaf budburst, flowering) in temperate woody plants, but the issues and solutions we present should be broadly applicable to many other organisms and biological processes that utilize temperature and daylength signals () (For a brief background overview of how temperature and light interact to influence the spring phenology of woody plants see box).

We begin below by reviewing the minimum experimental elements required to robustly test interactions between two or more variables. We then detail the problem of inference that can arise manipulating the both temperature and light in experiments and demonstrate the extent of this issue with mathematical and experimental examples. Finally, we conclude by outlining several possible solutions for overcoming these issues.

## 2 Testing interactions in controlled environmental

In order to have the statistical power to partition the individual and interactive effect of two or more variables in an experiment, one must:

1. Have at minimum two treatment levels of at least two variables of interest ().

2. Treatment levels must be full factorial (Fig. ??a.). Full factorial design are balanced (Fig. ??b.) and orthogonal (Fig. ??c.); which is to say that all possible treatment combinations are applied the treatments and the treatments are independent of each other (). ??.

These two critical elements may seem obvious but can be conspicuously absent from many published studies. In the case of woody plant phenology, despite a well established understanding that both temperature and light cues interactively influence phenological progress (), we found that in recently published database of Z controlled environment studies (), only X of them manipulated both light and temperature cues in the same experiment and only Y of those X did so with a design that was both balanced and orthogonal (see Supplement for details). This notable dearth of robust tests of light and temperature interactions may be related to the common limitations of time, space, and resources that experimentalists face (), but it may equally relate to a fundamental issue that arises from these variables themselves comprising of multiple axes of variability.

### 3 Axes of environmental variation in experiments and their implications

For most environmental variables, including light and temperature, there are two axes of variation that can be manipulated in an experiment.

1. Intensity: The amount or quality of a variable. We define temperature intensity as the amount of heat present in the system (measured in degrees). In the phenology literature this measurement is generally referred to as forcing. We define light intensity as the luminosity or irradiance present in the system (measured in lumens or watts).
2. Periodicity: The interval at which the intensity of the variable is applied. Hereafter, we refer to the periodicity of light as photoperiod (often used synonymously with ‘daylength’) and the periodicity of temperature as thermoperiod.

For spring of woody plants phenology, it is generally accepted that photoperiodicity is the primary light cue to which plants respond (( ) though see ( ) regarding light intensity and phenology). For temperature, conventionally both intensity and periodicity drive phenological activity ( ) and several metrics, (e.g. growing degree hours, thermal sums, growing degree days) that combine these two axes have been developed ( ). This assumption is well supported; under natural conditions diurnal temperature fluctuations temperate regions can be quite large in the field, and several studies have found that diurnal temperature variation strongly influences plant phenology ( ). In fact, several studies suggest that the even if thermoperiodicity is not an explicit treatment variable (manipulated systematically), incorporating it in experiments is essential for translating experimental results into real world predictions ( ).

It follows that a common approach in phenology experiments that seems to balance prior knowledge, biological realism and experimental inferences is to vary photoperiodicity, and thermal intensity and

periodicity (). For example, a basic experiment might include a long (12 hours) and short (8 hours) photoperiod treatment and a high (30/20 °C day/night) and low (20/10 °C day/night) temperature treatment. Note that in this case the thermoperiodicity is not an explicit treatment (both high and low temperature treatment employ a diurnal fluctuation of 10 °C), and is simply incorporated to enhance biological realism. At first glance, this design appears to meet the criteria of full factorial design, multiple treatment levels that are balanced and orthogonal, with mean high/low temperature treatments (25 and 15 °C respectively) and long/short photoperiod treatment applied in all possible combinations.

Yet the orthogonality of this design is based on the assumption of a 12 hour thermoperiod. If, rather the thermoperiod is coupled with the photoperiod, orthogonality is violated, in that the daily mean temperature of the long/high treatment will be higher than that of the short/high treatment, and the long/low treatment slightly warmer than the short low (Fig ??a), because the warmer day time temperature are applied for different durations across the high temperature treatments. Ultimately, this non-orthogonality introducing covariation among the photoperiod and temperature treatments, making it statistically impossible to differentiate their independent and interactive effects.

Of the X studies in the OSPREE database that manipulated both photoperiod and temperature, we found that at least Y may have this issue, suggesting that the true interactive effects of these cues on spring phenology is still quite poorly characterized. This may be in part why the relative contribution of temperature and photoperiod cues to spring phenology remains a contentious debate in the phenology literature ().

## Quantifying the uncertainty

To estimate the potential impact of this experimental artifact on estimations of cue effect sizes, we integrated the results of a large growth chamber phenology experiment by that employed this coupled photo-thermoperiod design. We used the simple geometry to a plane to calculate the approximate maximum amount of the forcing (temperature) and photoperiod effect estimates that could potentially be mis-attributed due to the latent non-orthogonality of this design. While the original model estimates of the forcing and photoperiod effects (units are “phenological sensitivity” or  $\Delta$  day of leaf out /  $\Delta$  cue level) were estimate -9.5 and -4.5 calculated that as much about 3.0 units of these effects could be mis attributed (for the full calculations see the Supplement.) Because forcing is expected to be the more dominant cue, we would expect the covariation of the variables may result in an over-estimation of the photoperiod effect and a weaker interaction estimate.

It is important to stress that this math is not meant to be rigid model correction. It is certainly possible that original model estimate approach the “true” values, but, due to the covariation of thermo and photoperiod there is no way of knowing for certain. (*Probably should think about how*

*to phase all this without making it seem like the Flynn paper is bad or wrong).*

Our estimate of “how much of each cue estimate could be misattributed due to the covariation of thermo- and photoperiod” can be rephrased as a maximum prediction of the expected difference in effect size estimate from a coupled photo- and thermo-period experiment and an uncoupled one in which diurnal photoperiod and thermoperiod are varied independently.

While we are aware of no experiments that explicitly compare these different designs, another later study by ? utilized many overlapping treatment levels and species from the same sampling site and several treatment levels with the ? study, but authors decoupled photo- and thermo-period, allowing for a reasonable comparison. We subset each dataset (publically available at HF and KNB ) to include only the species shared among the two studies, and re-analyzed the data, to compare difference in the photoperiod and forcing effect estimates (see supplemental method). As we predicted the un-coupled design estimated a weaker (less negative) photoperiod effect, and stronger forcing and interaction effects than the coupled experimental design (??). It is worth saying that there may be other factors driving the differences between these experiments as the we conducted in different years, used different methods for applying an addition temperature pre-treatment ,chilling (see Supplement), but this comparison is well matched to our mathematical predictions and prior knowledge about how temperature and photoperiod are expected to interacting in phenology.

## 4 Paths Forward

In the sections above we have systematically demonstrated that experiments which covary thermoperiod and photoperiod cannot robustly differentiate the individual effect of temperature and photoperiod on a spring phenology (or any other biological process) or quantify their interactive influence. Given the paucity of interactive studies in the literature, it is clear that more well designed studies will be needed to better characterize the effect of these cues. Below we offer several generalized experiment designs that improve statistical orthogonality of controlled environment experiment that could be further developed and adjusted to fit the needs of experimentalists across many subfields of biology.

1. **Manipulate photoperiod and temperature intensity with no thermoperiodicity.** This approach allows for the maintenance of statistical orthogonality across treatment combinations (??b.). The main drawback is that this design sacrifices the biological realism of diurnal temperature variation, which may make it more difficult to translate estimates from experiments to real world applications.
2. **Compensatory diurnal temperature fluctuations.** There are almost unlimited pairs of integers that can reduce to the same mean (e.g.  $24+26/2 = 30+20 = 25$ ) and the non-orthogonality of the mean daily temperature that arises in a coupled photo-thermoperiod design could be

corrected for by proportionately increasing the diurnal temperature fluctuation of the short photoperiod treatment relatively to the long treatments factor (??c.). However, if the differences between day and night temperature has a meaningful biological effect (), this introduces another confounding, non-orthogonal factor for interpreting temperature and photoperiod effects.

3. **Uncouple thermoperiod and photoperiod.** By varying thermoperiod and photoperiod independently (??d.), statistical orthogonality can be maintained across treatment. However, this approach it may introduce new artifacts that occur from the biological rather than statistical interactions that occur between light and temperature. For example, there is evidence that increasing temperatures in the first two hours of the photoperiod can be almost as effective for stimulating shoot elongation as similar temperature increases for the whole photoperiod (?), and that in phenology daytime warming can be as much as three times for effect that night time warming in cueing phenology (). With this design, treatments must inherently differ in the amount of time the warmer daytime temperature extend into the dark nighttime light regime, introducing a new axis of non-orthogonality.

In correcting one problem, each one of these designs introduces another, which may in fact be an intrinsic property of any experimental manipulation. This fact should caution experimentalists to continue to think carefully about our designs and perhaps most importantly, remind us to be humble in our inference, and think critically about what is, and isn't accounted for in our work.

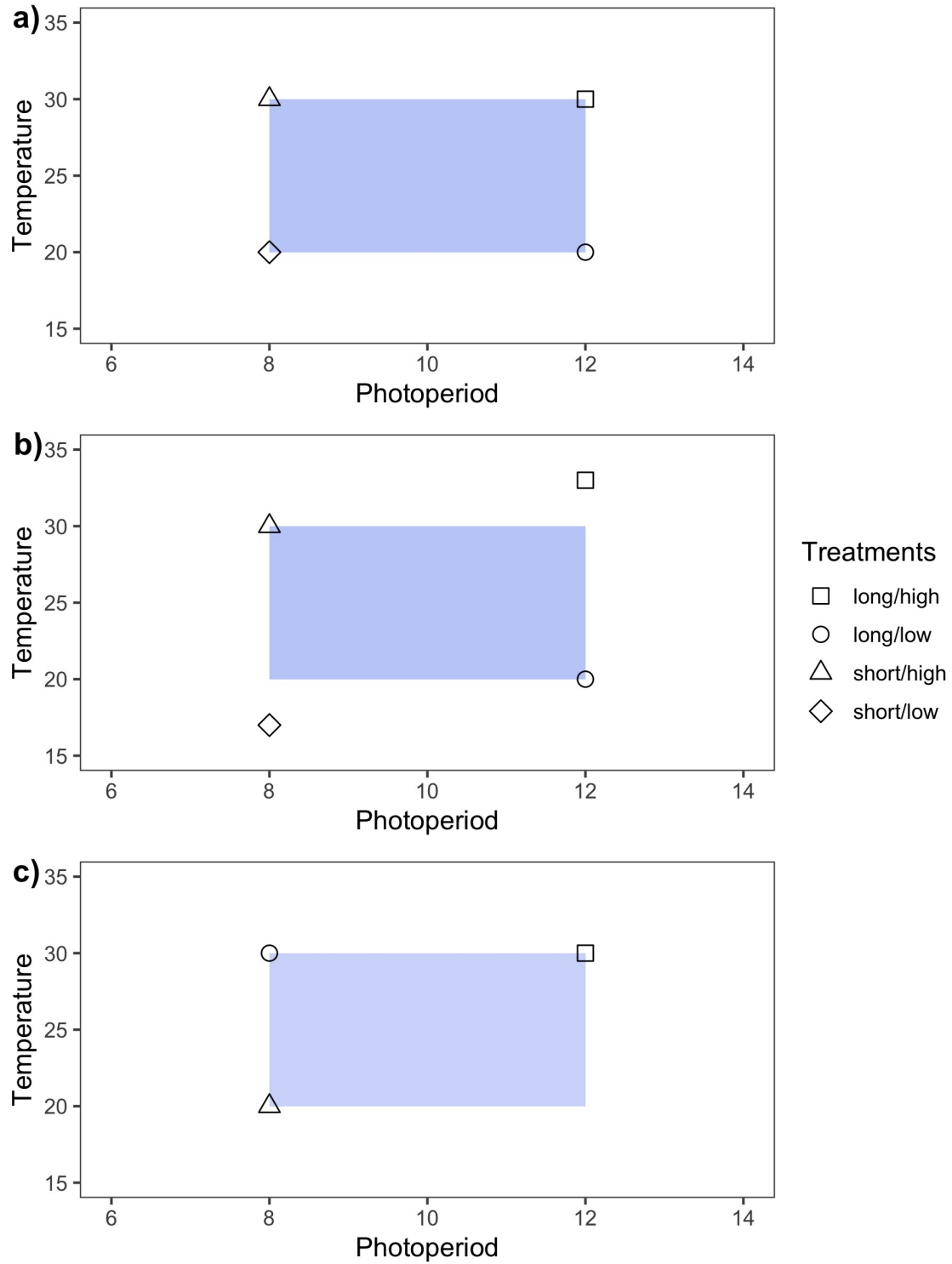


Figure 1: Idealized experimental designs demonstrate three approaches for varying temperature and light treatment level in controlled environment experiments. Design **a)** is fully factorial in that treatments levels are balanced and orthogonal. This design is appropriate for testing interactions between two or more variables. In **b)** the design is balanced both not orthogonal. Non-orthogonality in experiments often arises in experiments when there is covariation among the test variables is unaccounted for. In **c)**, the experimental design is orthogonal but unbalanced. Lack of balance in experiments often arises due to time, space or resource limitations.

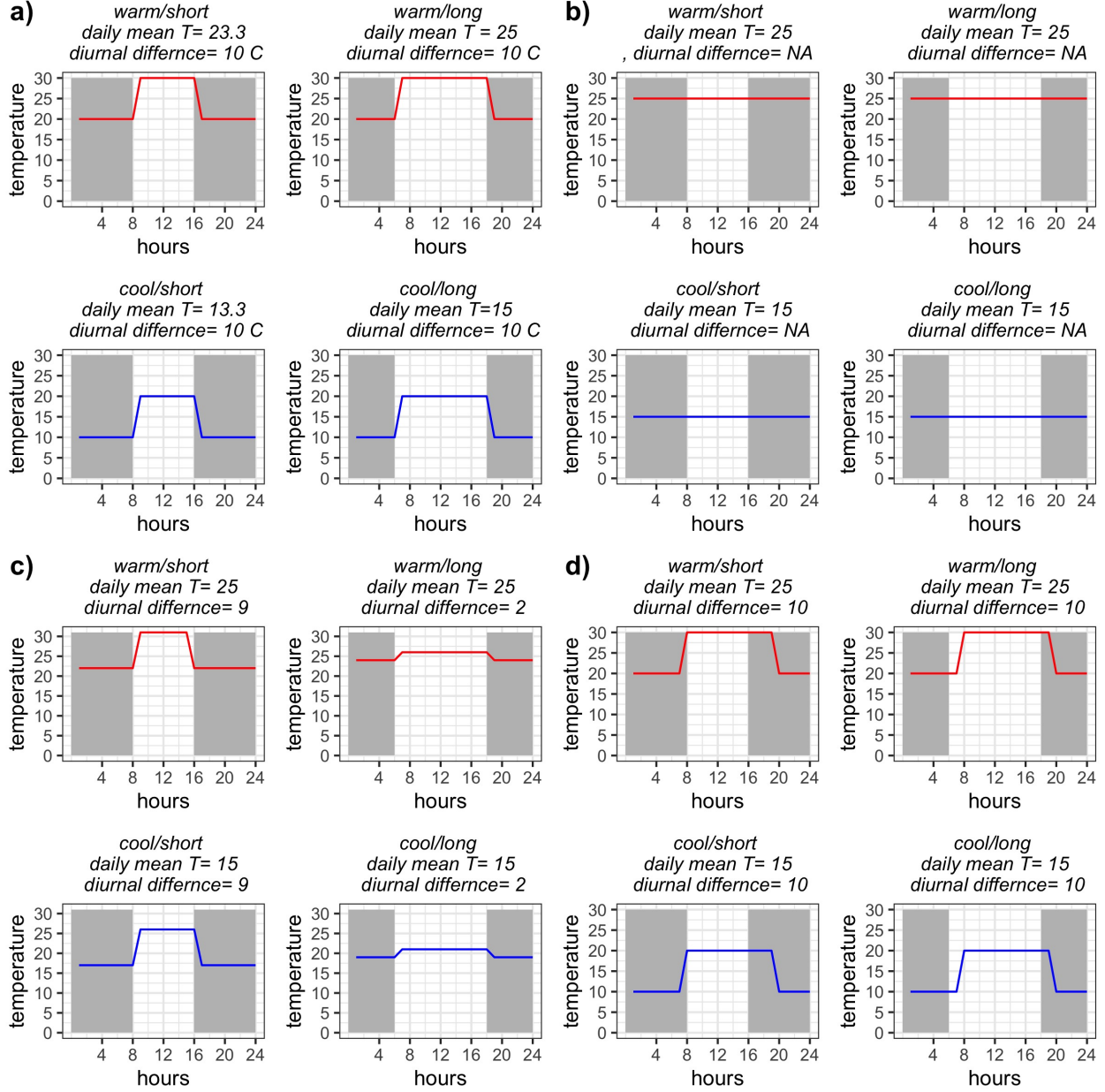


Figure 2: Conceptualized experimental designs to test temperature and daylength interactions on a biological response. In **a)** the design incorporates a standardize diurnal temperature fluctuation across all treatment. Because this thermoperiod is coupled with the photoperiod, while the same day and night temperatures are applied for the high and low temperature treatments respectively, the mean daily temperatures differ across each photoperiod treatment generating non-orthogonality. Designs **b)**, **c)** and **d)** are all designs that can correct this non-orthogonality. Design **b)** manipulated temperature intensity only (no thermoperiodicity). In **c)** photo- and thermo- periods are still coupled but the orthogonality of mean daily temperature is maintained by proportionately varying the diurnal temperature fluctuations across treatments. In design **d)** standard diurnal temperature fluctuations are maintained but, thermoperiod and photoperiod are decoupled and varied independently, maintaining orthogonality daily mean temperatures.

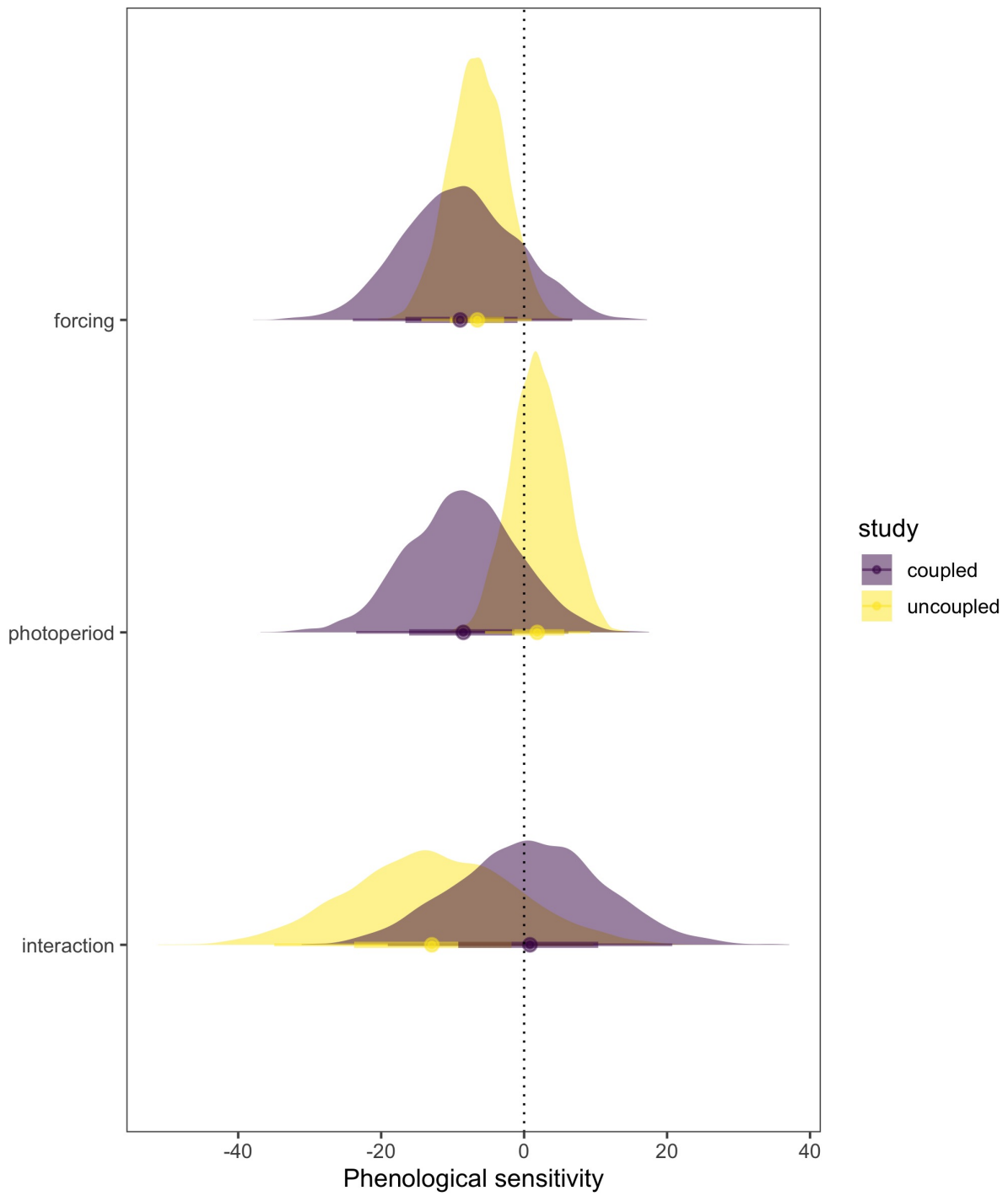


Figure 3: Need a caption but basically, estimates differs in expected ways