# Week 6 Problem Based Learning and Practical Solutions

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### Problem based learning workshop

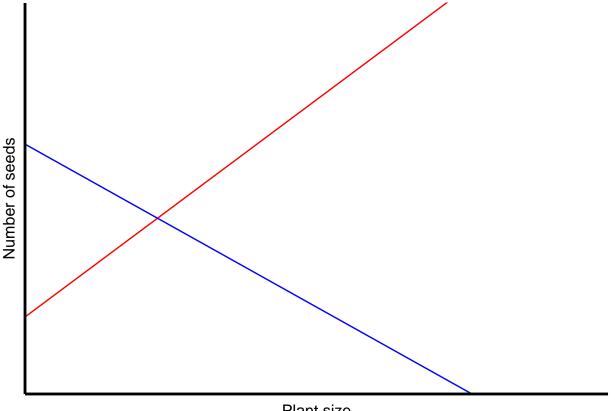
#### Exercise 1

A student wished to know whether seed production in an invasive plant depends on whether flowers are self-pollinated or cross-pollinated. She knew that seed production increases with plant size, so she measured the sizes of 50 plants before randomly assigning them to selfed and crossed treatments. She applied the treatment (pollen from self vs others) to the stigmas of 5 flowers per plant, and later counted how many seeds each flower produced.

- 1. Draw out the experimental design, indicating the number of replicates at each level.
- 25 plants per treatment.
- 5 flowers per plant.
- 2. How many independent data points are there?
- 50. Not 250 because flowers on the same plant are not independent.
- 3. How would you deal with the data to ensure that all data points used in the analysis are independent?
- I would use a plant-level random effect in the analysis.

- 4. In the general format (response-predictor variables), write down the full linear model that includes an interaction between the categorical variable and the covariate (i.e., with separate slopes and intercepts fitted for each treatment group). Make sure you can distinguish between these different types of predictor variable.
- number.of.seeds.in.a.flower ~ treatment \* plant.size + (1|plant.id)
- 5. Write down a simplified model that includes the main effects of the predictor variables only.
- number.of.seeds.in.a.flower ~ treatment + plant.size + (1|plant.id)
- 6. Sketch a graph (with labelled axes) showing a possible outcome if the full model best explains the data (i.e., if the interaction is significant).

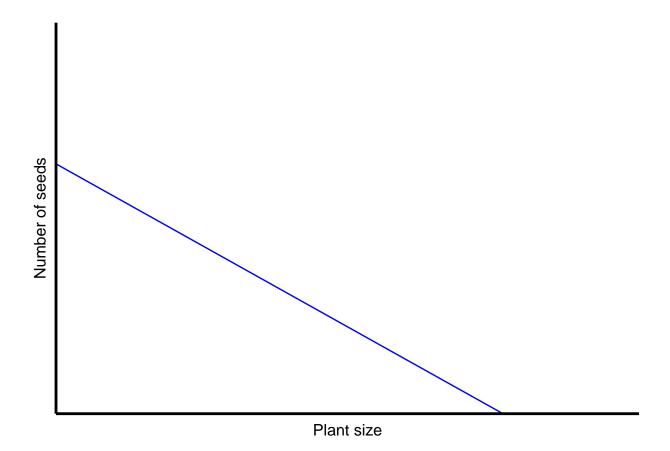
```
library(ggplot2)
ggplot() +
    geom_abline(intercept=1, slope=2, color='red') +
    geom_abline(intercept=8, slope=-1.5, color='blue') +
    xlim(1,6) + ylim(1,10) +
    theme classic() +
    theme(axis.ticks=element_blank(), axis.text=element_blank(),
        axis.line=element_line(size=1)) +
    xlab('Plant size') + ylab('Number of seeds')
```



Plant size

7. Sketch a graph (with labelled axes) showing a possible outcome if the simplified model best explains the data (i.e., if only the main effects are significant).

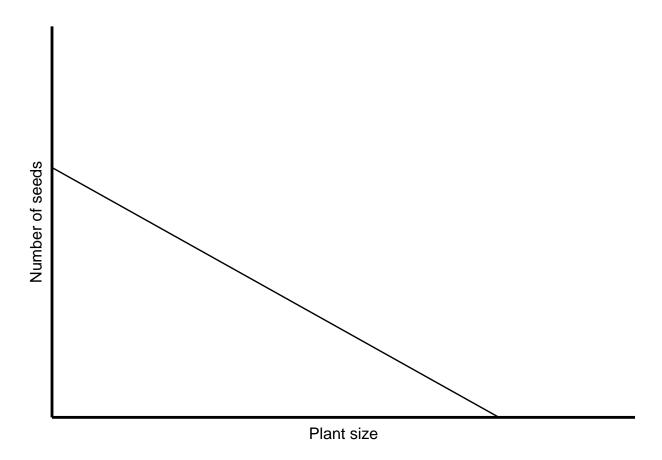
```
library(ggplot2)
ggplot() +
    geom_abline(intercept=1, slope=-1.5, color='red') +
    geom_abline(intercept=8, slope=-1.5, color='blue') +
    xlim(1,6) + ylim(1,10) +
    theme_classic() +
    theme(axis.ticks=element_blank(), axis.text=element_blank(),
        axis.line=element_line(size=1)) +
    xlab('Plant size') + ylab('Number of seeds')
```



8. Sketch 2 graphs (with labelled axes) showing a possible outcome if only one of the main effects is significant (do this for each main effect).

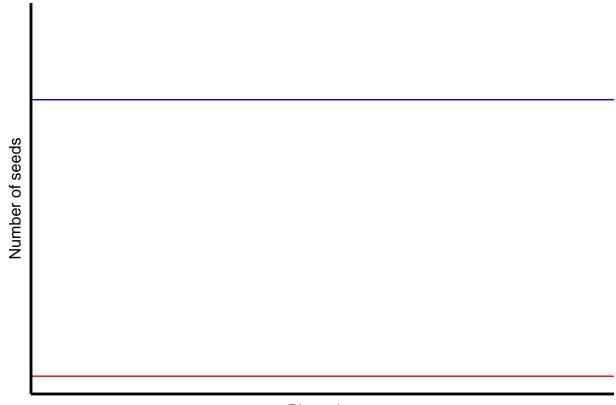
The plot below shows if there is an effect of size.

```
library(ggplot2)
ggplot() +
    geom_abline(intercept=8, slope=-1.5, color='black') +
```



The plot below shows if there is an effect of treatment.

```
library(ggplot2)
ggplot() +
    geom_abline(intercept=1, slope=0, color='red') +
    geom_abline(intercept=8, slope=0, color='blue') +
    xlim(1,6) + ylim(1,10) +
    theme_classic() +
    theme(axis.ticks=element_blank(), axis.text=element_blank(),
        axis.line=element_line(size=1)) +
    xlab('Plant size') + ylab('Number of seeds')
```



#### Plant size

#### Exercise 2: Exploring the specific assumptions of ANCOVA models

Are regression slopes similar across groups?

- 1. Which of the models that you wrote down in Exercise 1 tests this assumption?
- The model that includes the interaction.
- 2. Which of your sketched graphs from Exercise 1 represents a violation of this assumption? Note that this violation does not mean that you cannot fit an ANCOVA-like model, but simply means that you cannot simplify it to the "classic" ANCOVA form (i.e., an ANOVA adjusted for the regression of the response variable on the covariate).
- The presence of non-parallel lines suggests this.

Are covariate values similar across groups (i.e., are the covariate and categorical predictor independent)?

3. State whether the experimental design described in Exercise 1 is likely to meet this assumption and why.

- We want overlap between the size ranges for each treatment, otherwise comparisons are extrapolated. Yes, because the plants are randomly assigned to treatments and because the size is measured before the treatments are assigned.
- 4. A student was investigating the effects of grazing and soil nutrients on the growth of an invasive grass species. He established 20 plots containing the grass across Queensland. In each plot, he measured grass growth, soil nutrients and grazing intensity (scored as high or low). State whether the experimental design is likely to meet this assumption and why.
- No. There is a correlation between grazing history and grass growth. This is because cows won't eat grass in places where there is no grass. This is called collinearity.
- 5. How would you check this assumption in each case?
- Graph the data. Visualise it.

#### Exercise 3: Biological interpretation of ANCOVA models

#### Case study 1

Two researchers wished to know whether if increased reproduction reduces the longevity of male fruit-flies (it does in female fruit-flies). They manipulated reproduction by supplying each of 25 male flies with 1 virgin (receptive) female per day until he died, and each of 25 males flies with 1 newly-inseminated (unreceptive) female per day until he died. They also measured the thorax length of each male, in case it explained some of the variation in longevity.

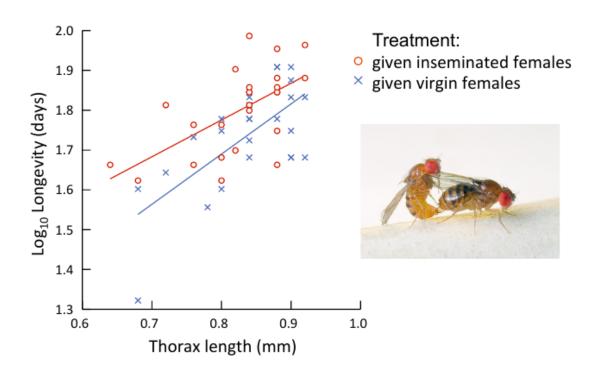


Figure 1. Longevities and thorax lengths of male fruit-flies supplied with either virgin (receptive) or newly-inseminated (unreceptive) females daily until death (data are log-transformed to meet the general assumptions of linear model

Table 1. Full ANCOVA model (response variable is  $log_{10}$  longevity).

Source	SS	DF	MS	F	Р
Treatment	0.011	1	0.011	1.250	0.269
Thorax	0.282	1	0.282	33.047	<0.001
Treatment*thorax	0.007	1	0.007	0.781	0.381
residuals	0.393	46	0.009		

Table 2. Simplified ANCOVA model (response variable is  $log_{10}$  longevity).

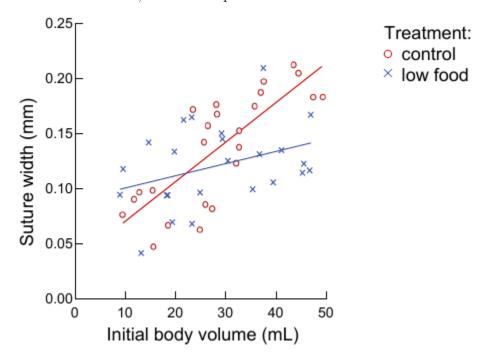
Source	SS	DF	MS	F	Р
Treatment	0.070	1	0.070	8.253	0.006
Thorax	0.283	1	0.282	33.302	<0.001
residuals	0.400	47	0.009		

<sup>1.</sup> Do the data meet the specific assumption of ANCOVA that covariate values are similar across groups?

- Yeah, the red and blue points overlap on the x-axis a lot.
- 2. Do the data meet the specific assumption of ANCOVA that regression slopes are similar across groups?
- The regression slopes are slimial across groups.
- 3. State which ANCOVA model best explains the data (the full model in Table 1, or the simplified model in Table 2) and why.
- The model without the interaction.
- 4. What is the biological interpretation of these results?
- There is a main effect of mating frequency on longevity. This effect is still significant even if we account for the effect of thorax.

#### Case study 2

Shrinking of the test (i.e., endoskeleton) is reported in some species of sea urchins when food is in short supply. These tests consist of calcareous plates joined by sutures, which might allow tests some degree of elasticity (much like human skulls) and explain this reduction in size. A researcher studied the role of suture width in the shrinking of sea urchin tests on different diets. He manipulated diet by maintaining 24 urchins in a low food treatment (with food depleted below natural levels) and 24 urchins in a control treatment (with food at natural levels). He also measured the initial body volume of each urchin, in case it explained some of the variation in suture width.



*Table 3. Full ANCOVA model (response variable is suture width).* 

Source	SS	DF	MS	F	Р
Treatment	0.005	1	0.005	4.501	0.040
Body volume	0.034	1	0.034	30.757	<0.001
Treatment*body volume	0.010	1	0.010	8.584	0.005
residuals	0.049	44	0.001		

Table 4. Simplified ANCOVA model (response variable is suture width).

Source	SS	DF	MS	F	Р
Treatment	0.003	1	0.003	2.218	0.143
Body volume	0.032	1	0.032	24.333	<0.001
residuals	0.059	45	0.001		

- 1. Do the data meet the specific assumption of ANCOVA that regression slopes are similar across groups?
- Yes. The data for each treatment overlaps on the x-axis.
- 2. State which ANCOVA model best explains the data (the full model in Table 3, or the simplified model in Table 4) and why.
- The model with the interaction (Table 3).
- 3. What is the biological interpretation of these results?
- The effect of treatment on suture width depends on initial body volume. You cannot predict what the effect of treatment is unless you know the initial body volume.

## R practical session

#### General notes