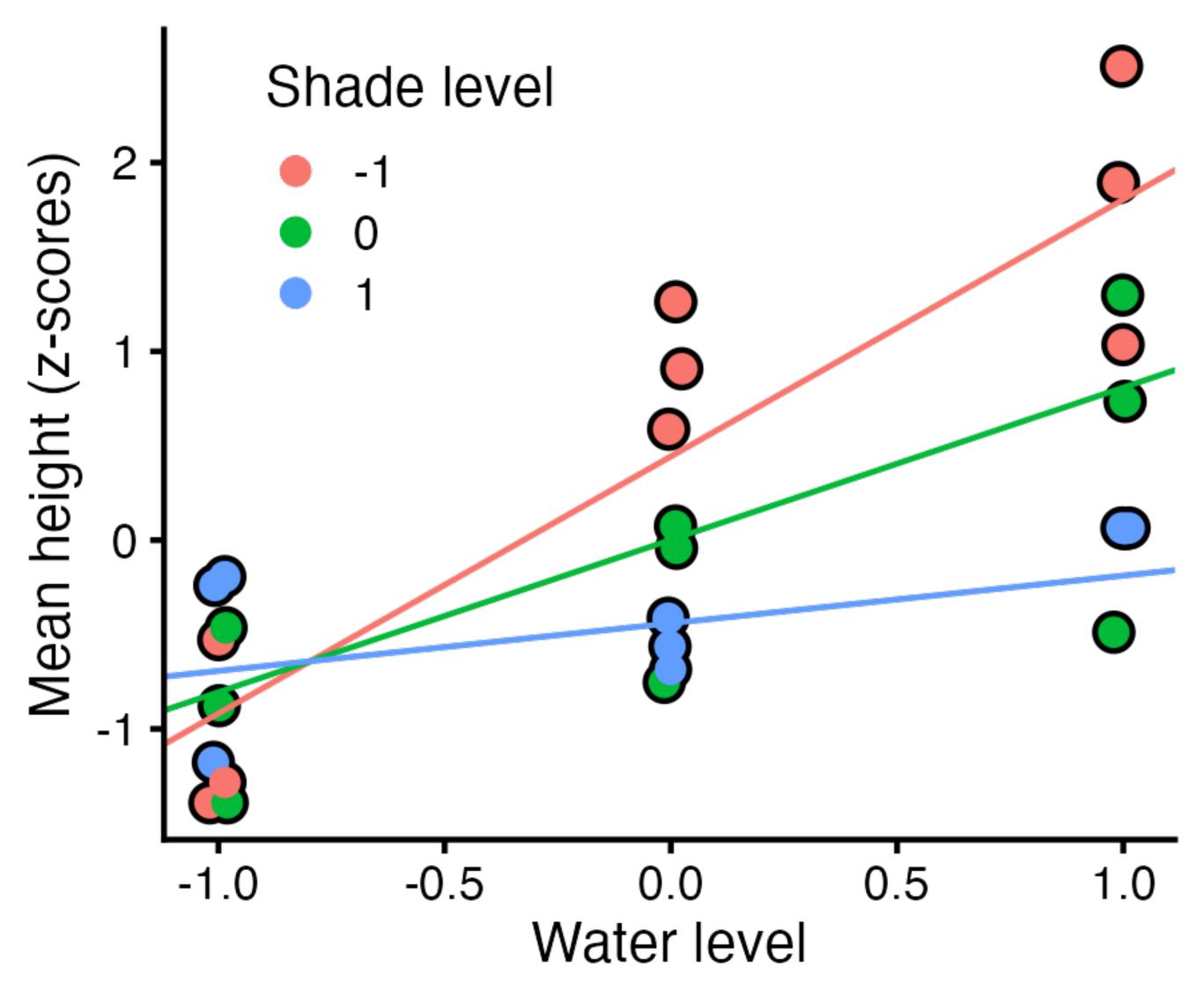
## Prediction lines



```
cf = coef(rt_fit)
a = cf["a"]; b = cf["b"]; c = cf["c"]; d = cf["d"]
 col = scales::hue_pal()(3)
 # Model: a + b*w + c*s + d*w*s
p = ggplot(df, aes(x = water, y = blooms, color = factor(shade))) + properties a graph of the state of the 
                          geom_point(size = 2) +
                          geom_abline(intercept = a - c, slope = b - d, col = col[1]) +
                          geom_abline(intercept = a + c, slope = b + d, col = col[3])
```

## Prediction lines

```
Shade level
cf = coef(rt_fit)
a = cf["a"]; b = cf["b"]; c = cf["c"]; d = cf["d"]
col = scales::hue_pal()(3)
# Model: a + b*w + c*s + d*w*s
p = ggplot(df, aes(x = water, y = blooms, color = factor(shade))) +
   geom_point(size = 2) +
   geom_abline(intercept = a - c, slope = b - d, col = col[1]) +
   geom_abline(intercept = a + c, slope = b + d, col = col[3])
                 -0.5
                                    0.5
                           0.0
        -1.0
                                              1.0
                      Water level
```

## Summary

- Multiple linear models allow us to use more than one predictor in a linear model
- These models do a form of automatic
   stratification
  - Ex: difference in size for individuals of the same age, effect of treatment for individuals of the same size
- The objetive is to compare like-to-like

- Coefficients can and do change with the inclusion of more predictors
- Coefficient interpretation is hard, use plots, predictions, scaling and transformations to make models easier to interpret
- Next week, we talk about principled ways of choosing if a variable should be added to a model, stay tuned!