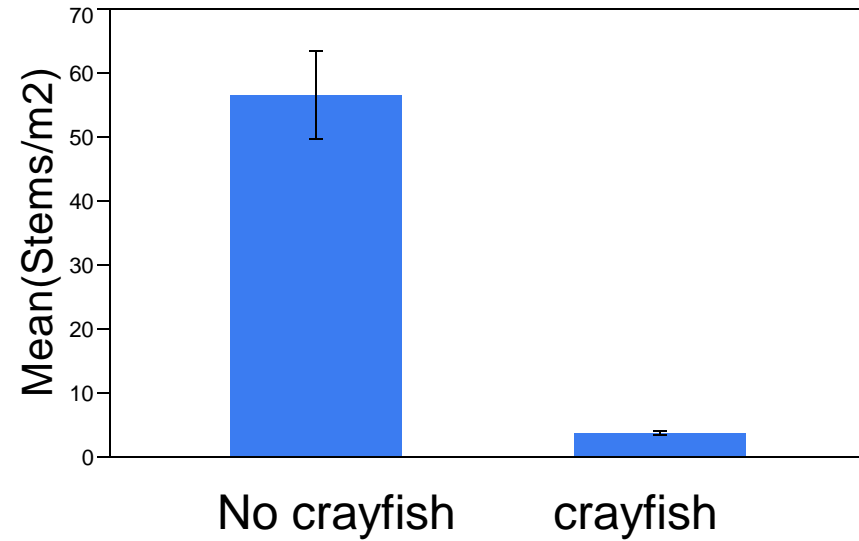
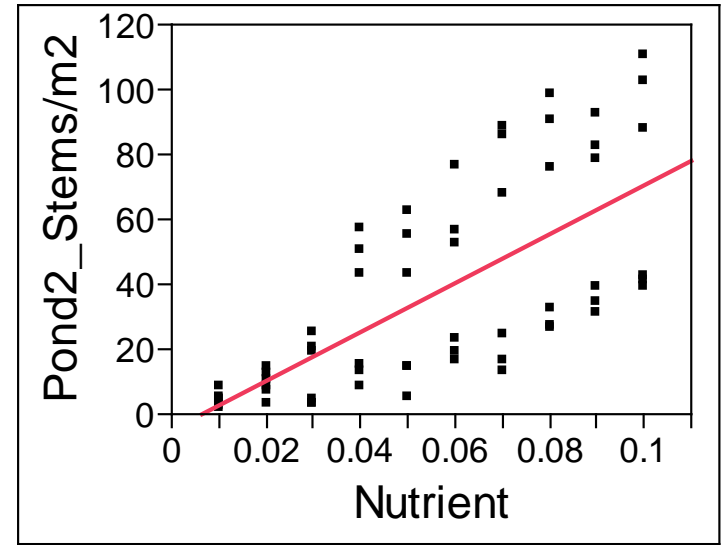


Analysis of Covariance (ANCOVA)

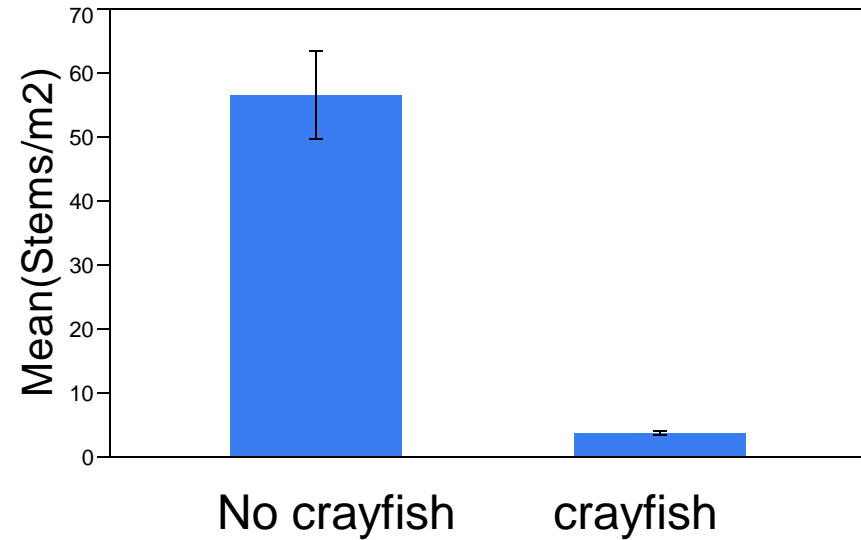
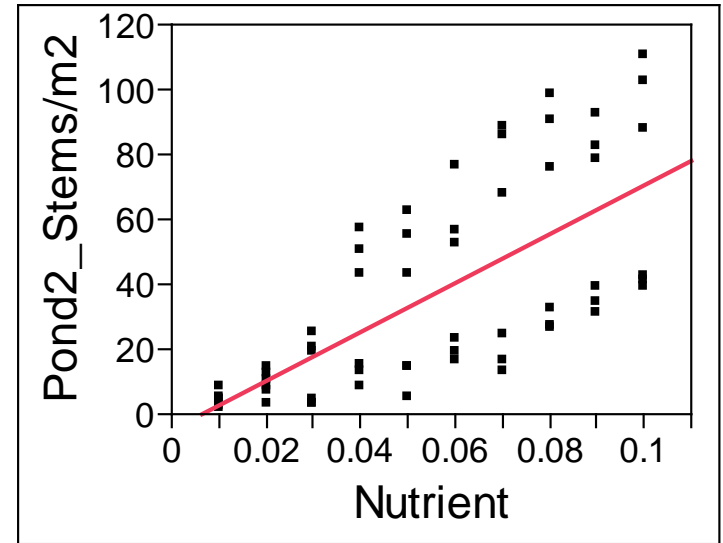
Hybrid of ANOVA and regression



Analysis of Covariance (ANCOVA)

Hybrid of ANOVA and regression

You have both categorical and continuous X variables

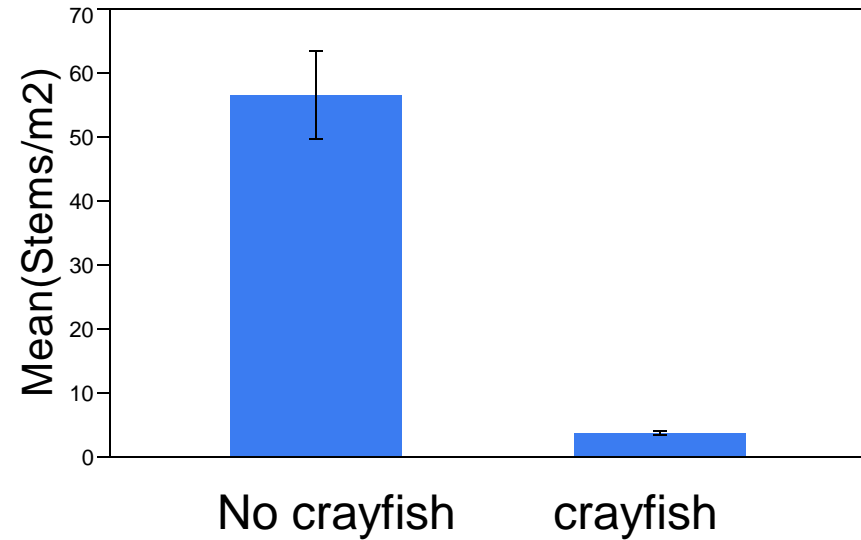
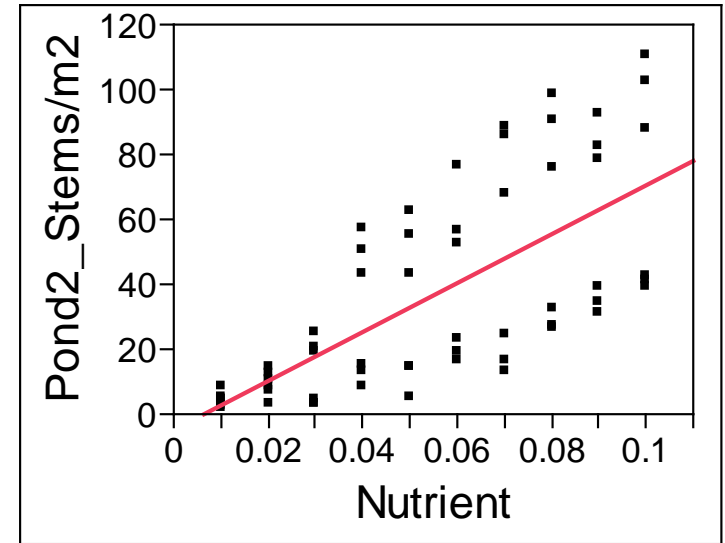


Analysis of Covariance (ANCOVA)

Hybrid of ANOVA and
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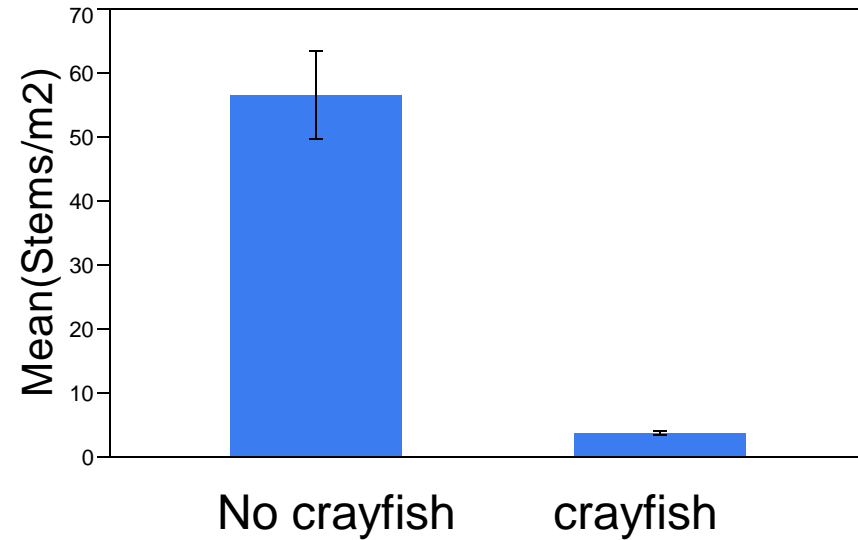
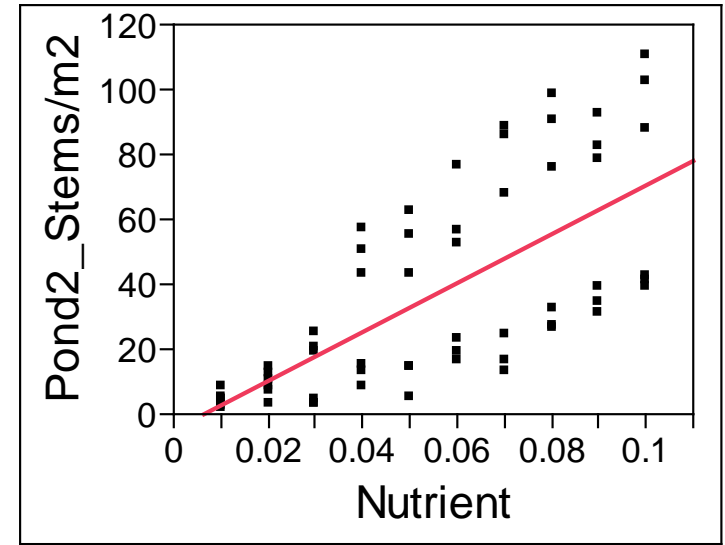
ANOVA design, with an
additional continuous
variable (covariate)



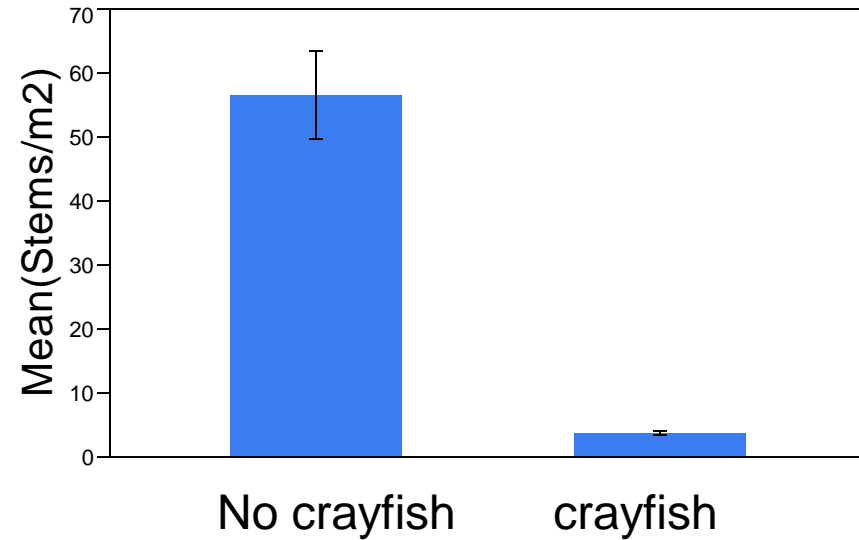
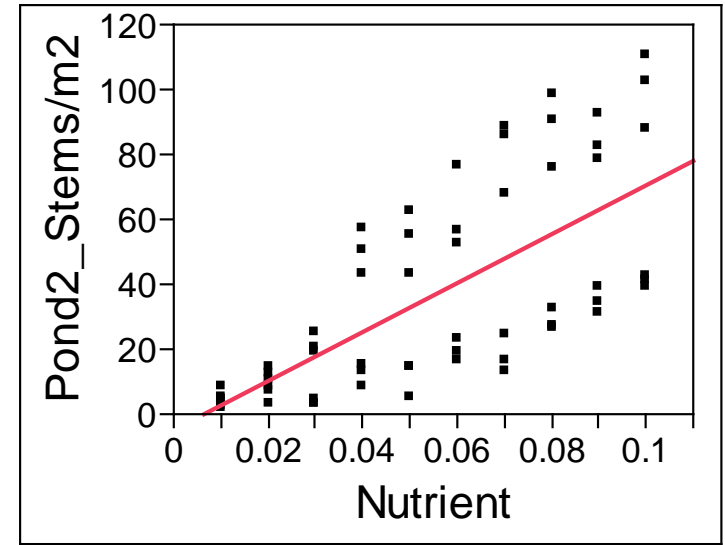
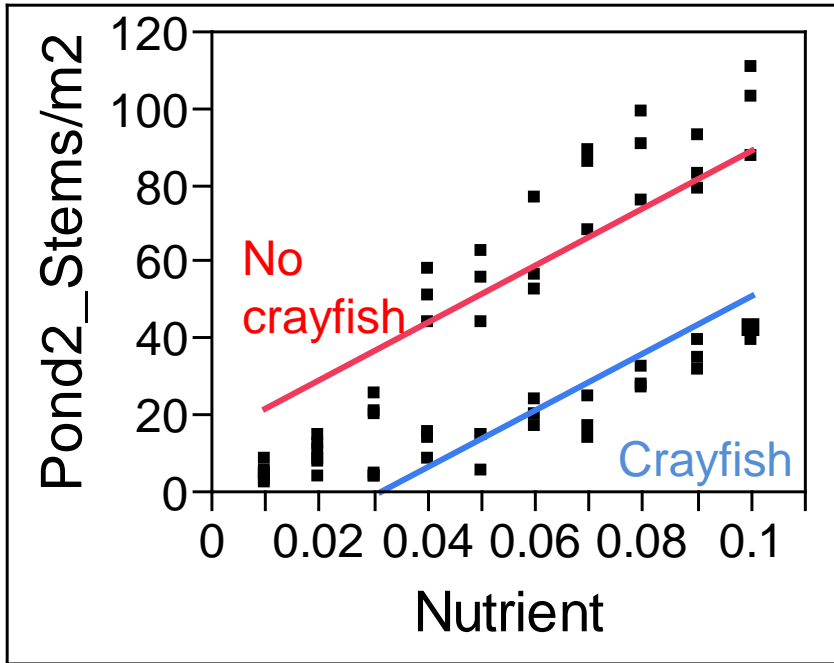
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Analysis of Covariance (ANCOVA)



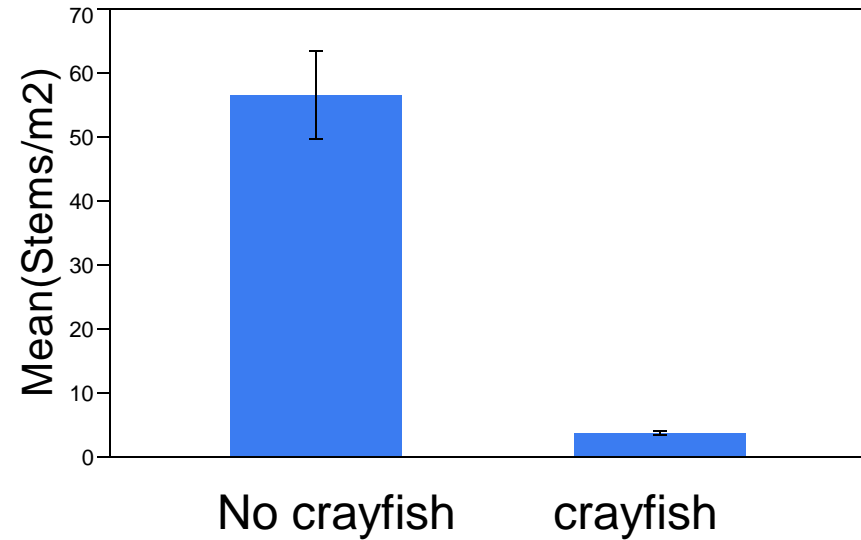
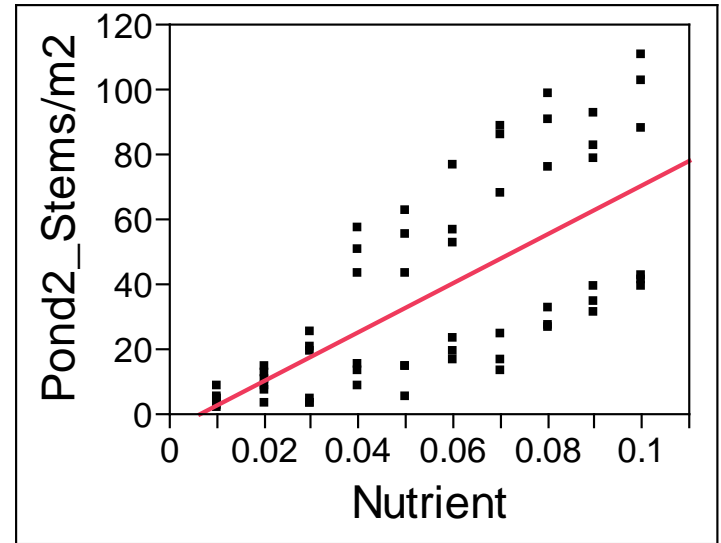
ANOVA design, with an additional continuous variable (covariate)

The modification of ANOVA that allows ANCOVA

ANOVA

$$Y_{ij} = \mu + A_{ij} + \varepsilon_{ij}$$

Observations Y are a combo of some mean level μ + the positive or negative average effect A of each level i plus error ε



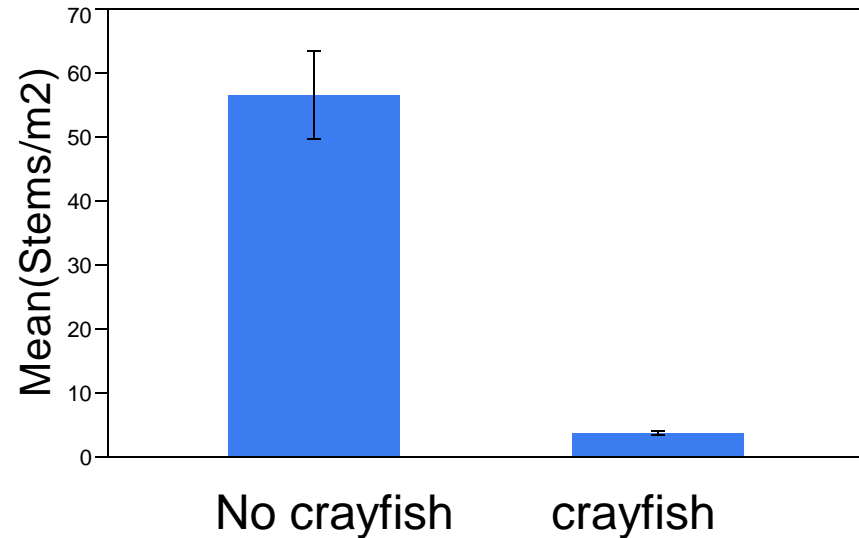
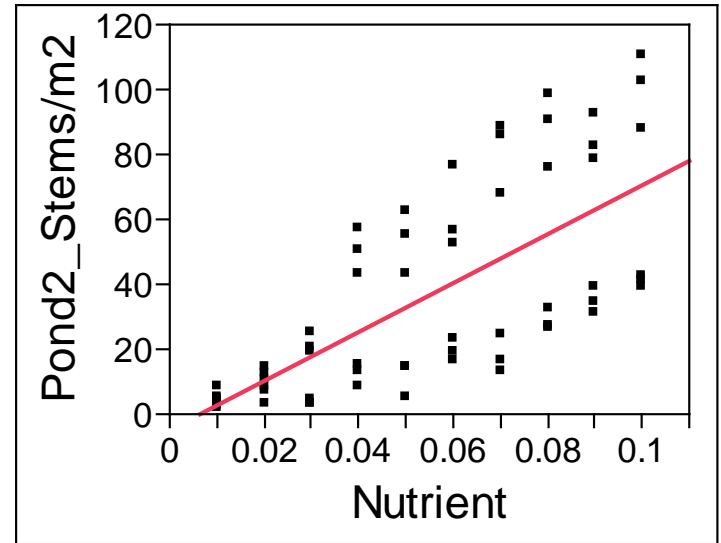
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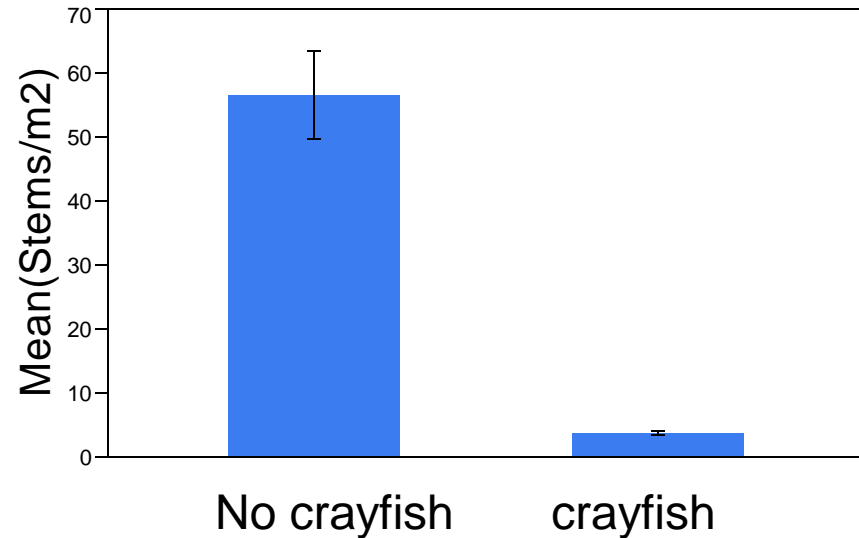
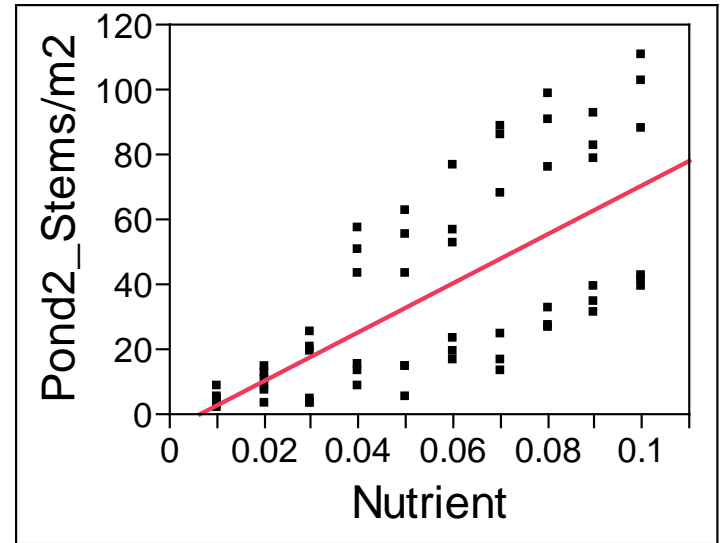
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So... correlate Y with the continuous variable first, and then use ANOVA to analyze residuals



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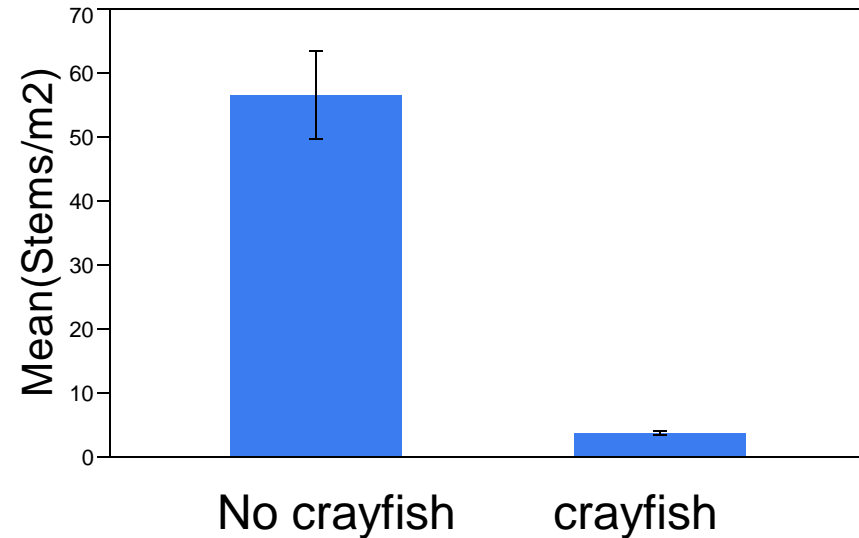
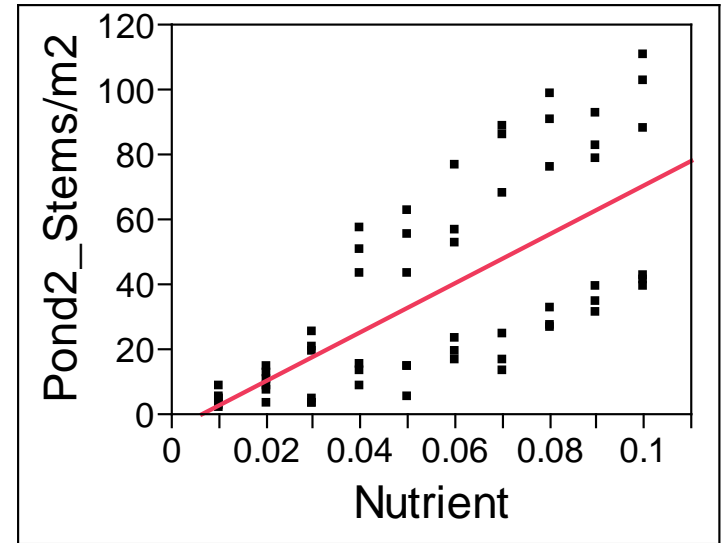
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ANCOVA

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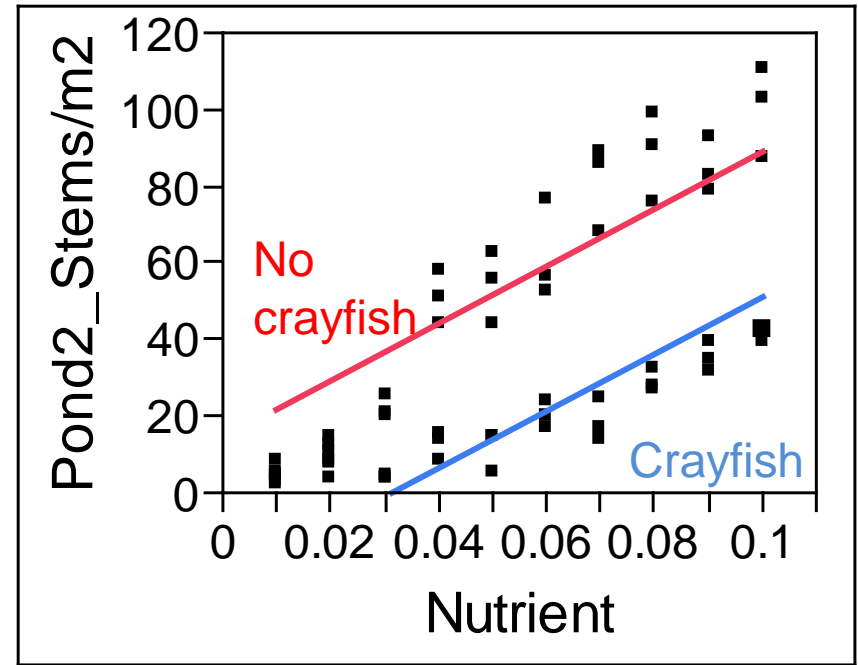
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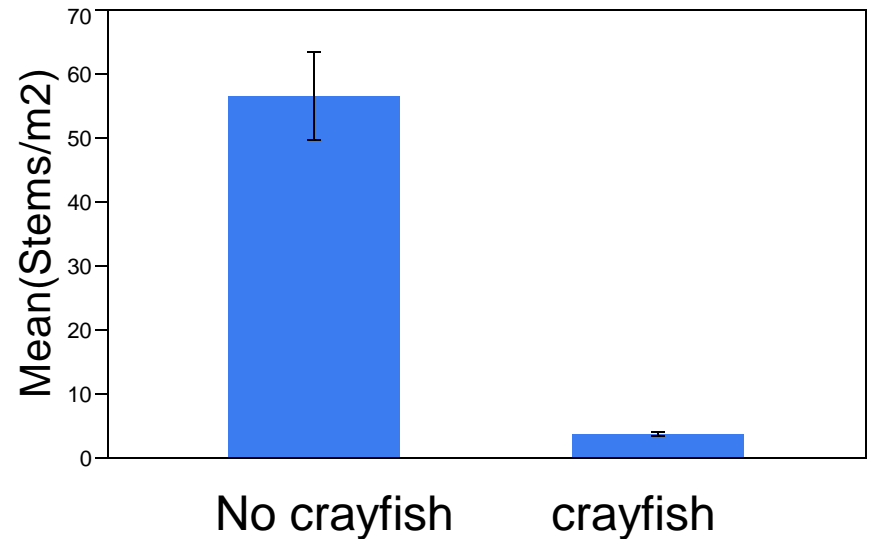
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If β is non-significant, it is zero, so the equation reduces to a 1-way ANOVA



The modification of ANOVA that allows ANCOVA

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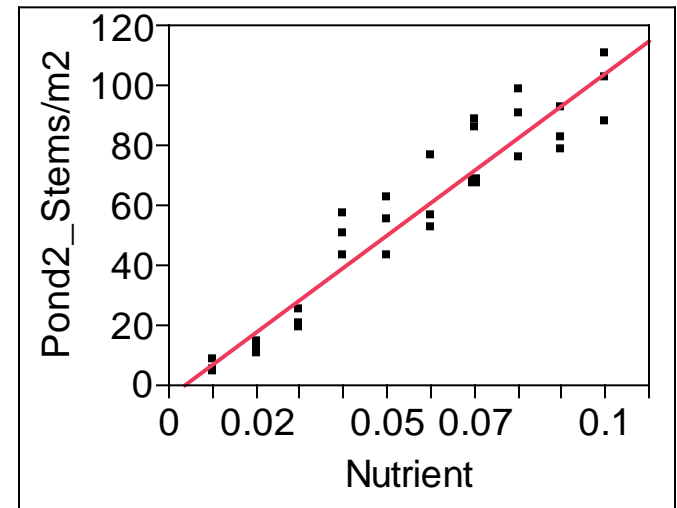
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If β is non-significant, it is zero, so the equation reduces to a 1-way ANOVA

If the treatment effect (A) is non-significant, the equation may reduce to a linear regression

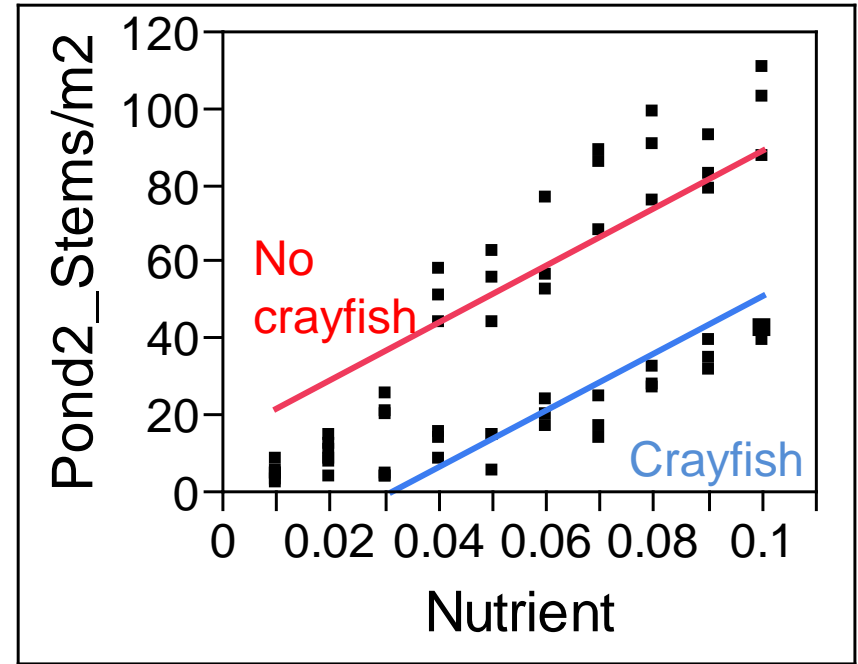


ANCOVA

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Generally, the algorithm first tests to see whether slopes differ between/among treatments

If not, a common slope is used – the simplest case



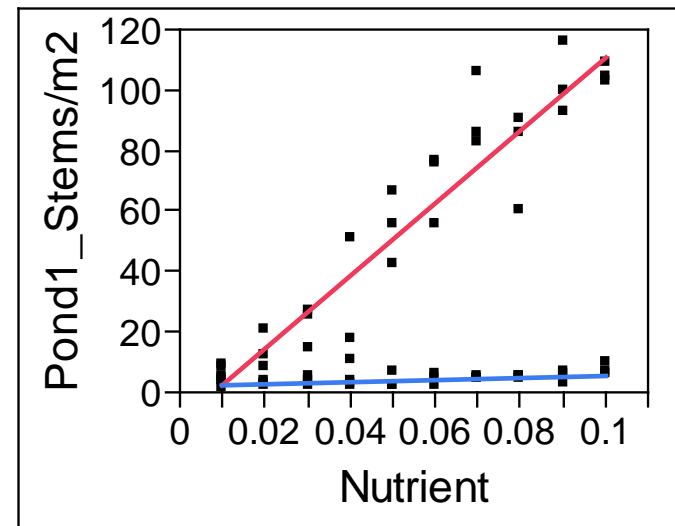
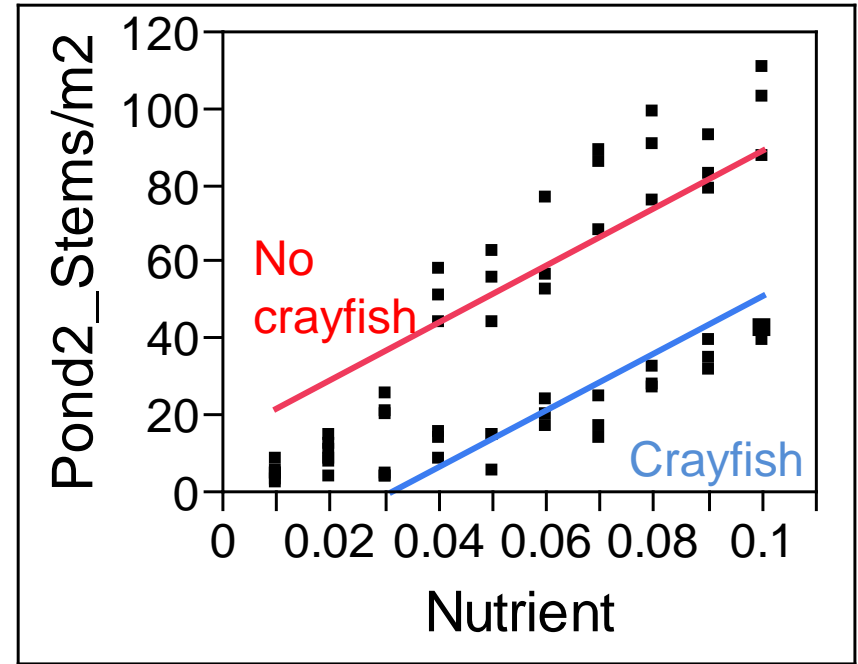
ANCOVA

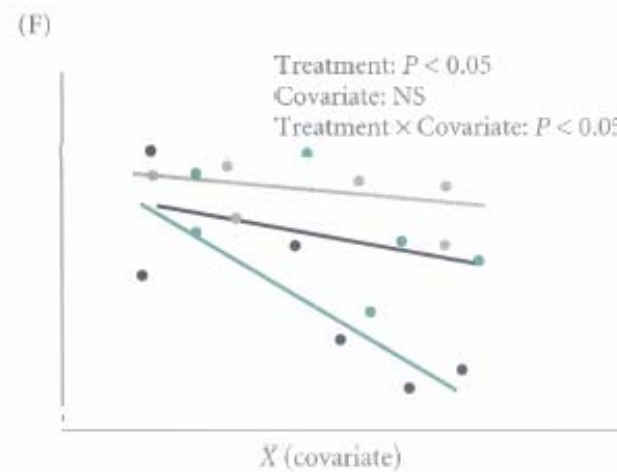
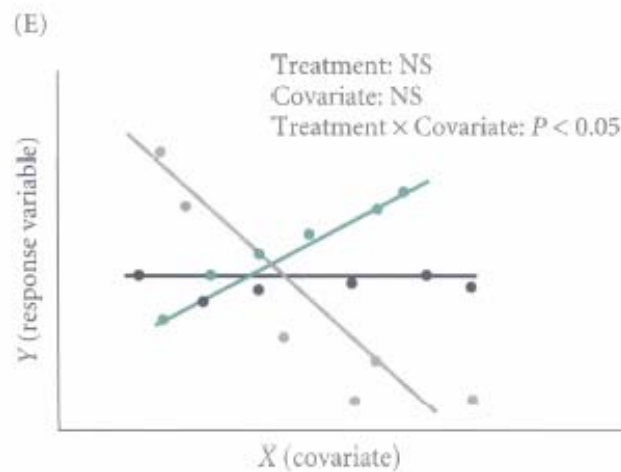
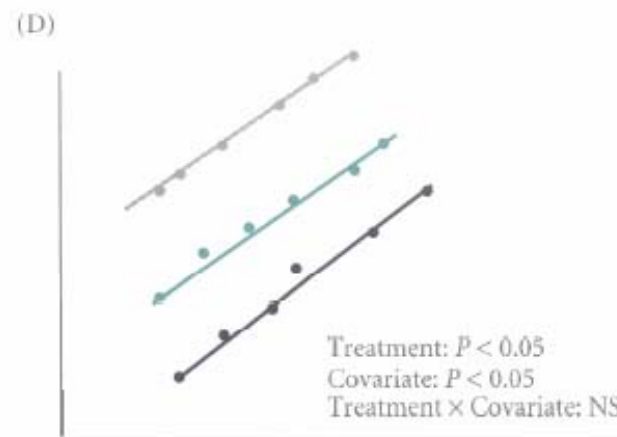
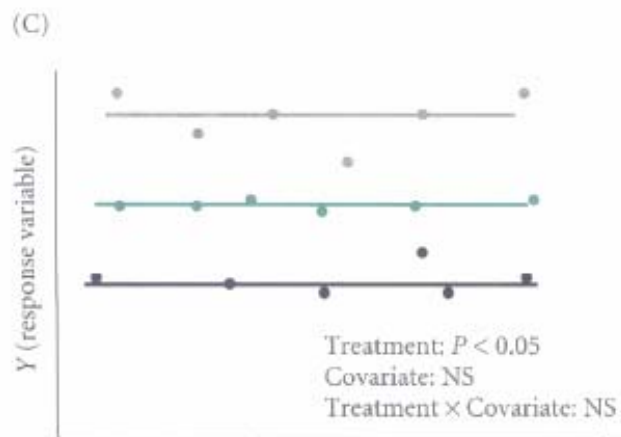
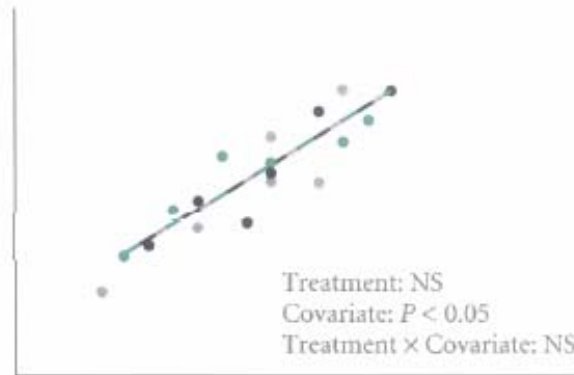
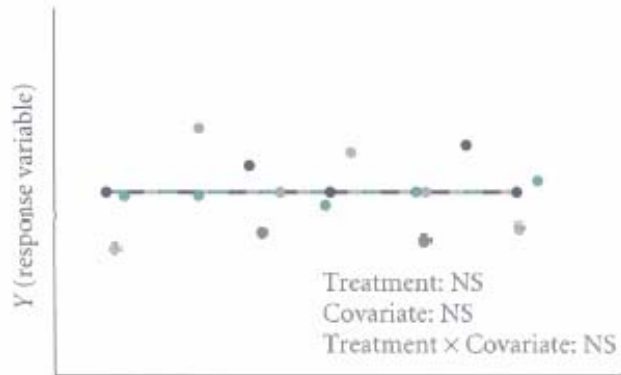
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Interaction term is significant when slopes differ significantly



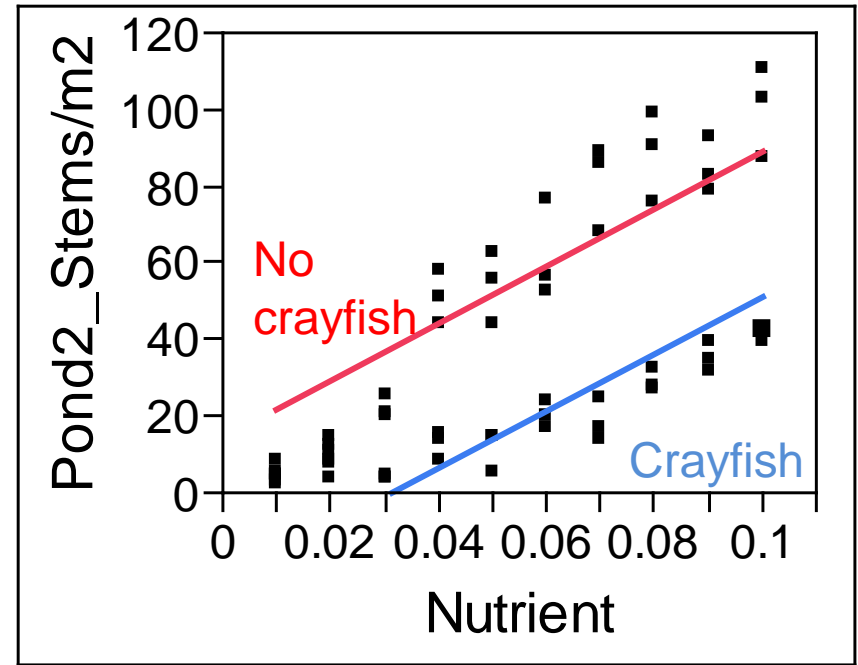


Gotelli &
Ellison Fig.
10.4

What are important assumptions in ANCOVA?

Normality/homoscedasticity

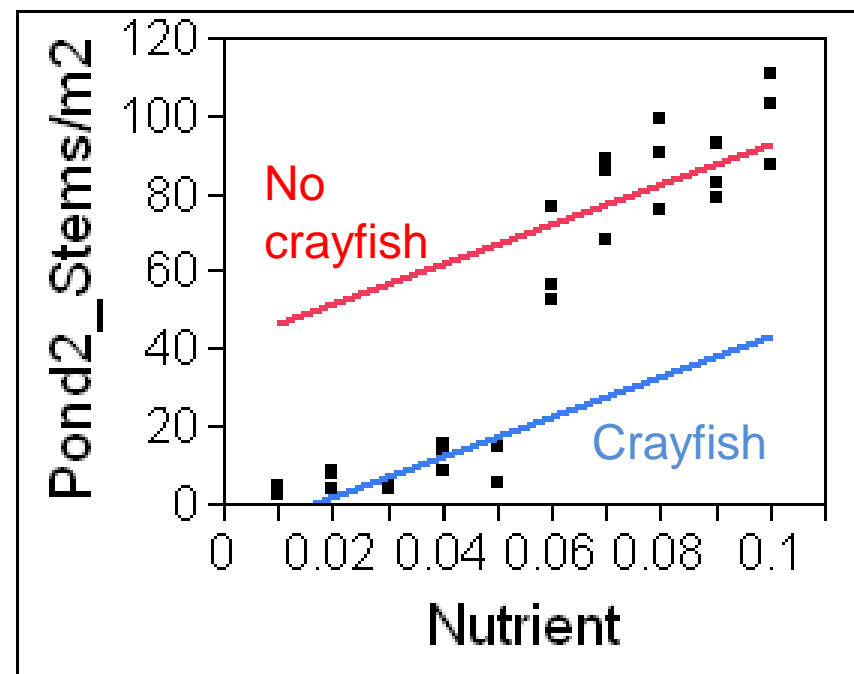
Independent samples



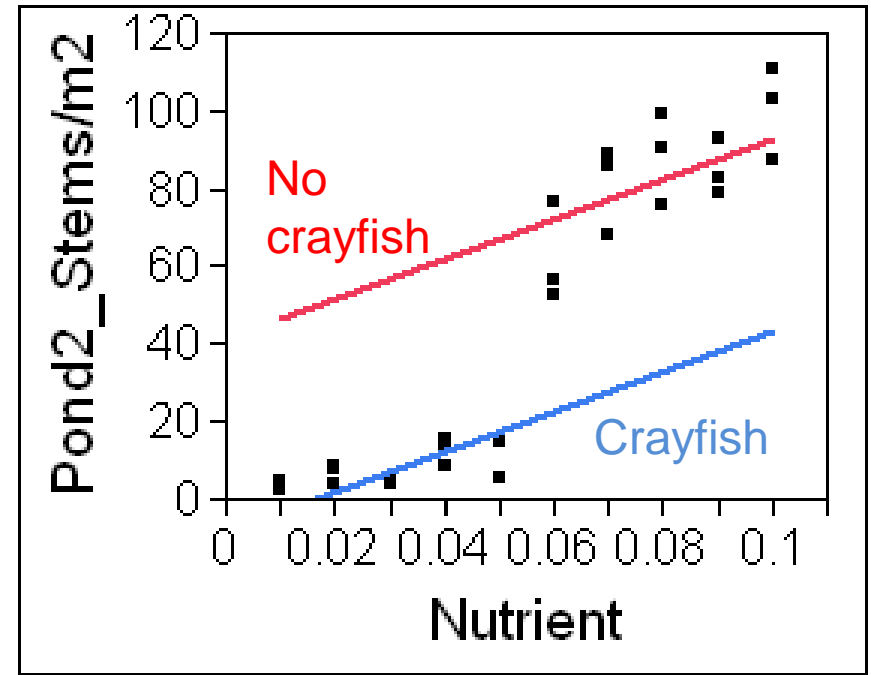
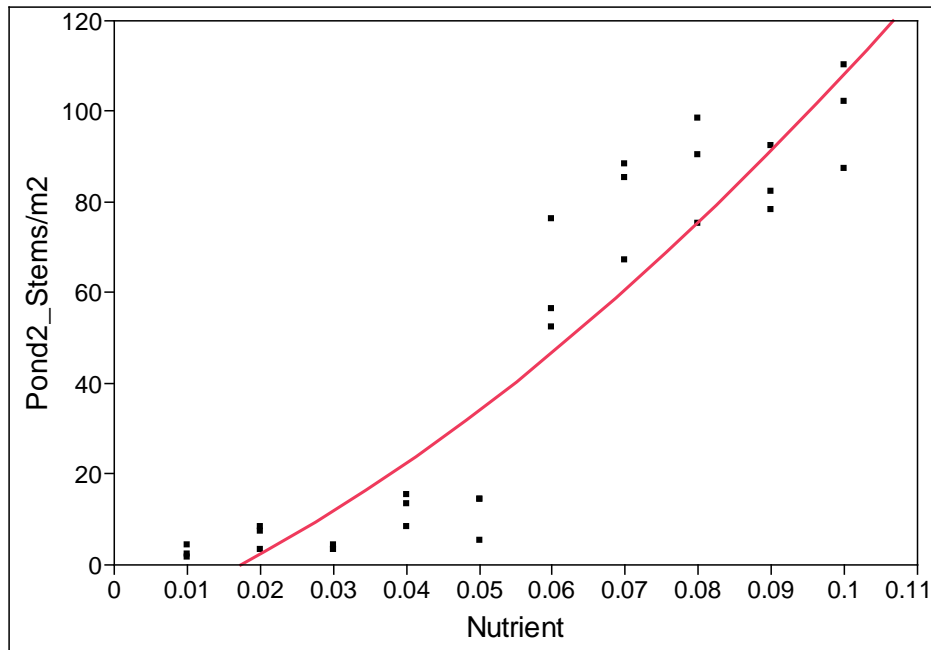
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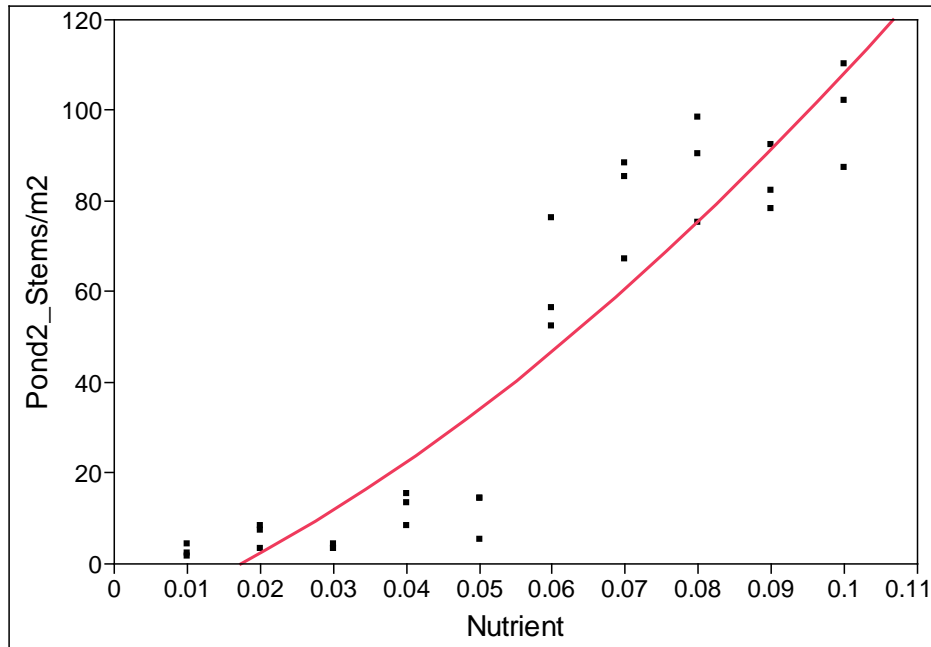
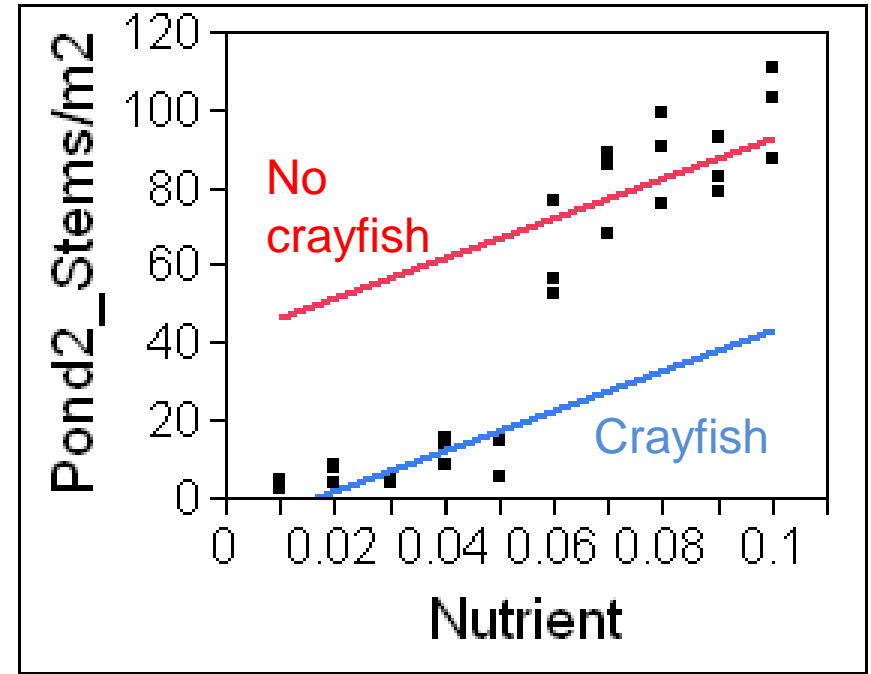
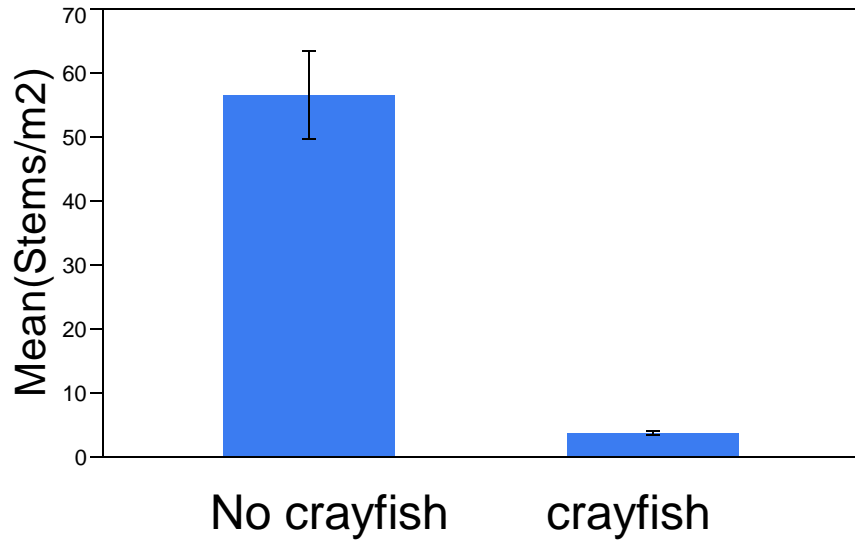
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Careful that effects are not confounded!



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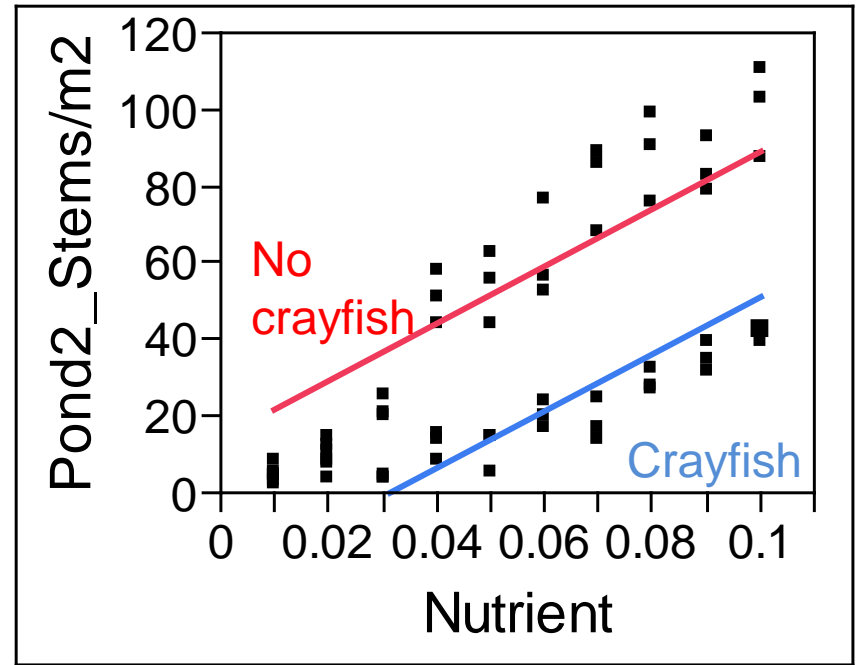
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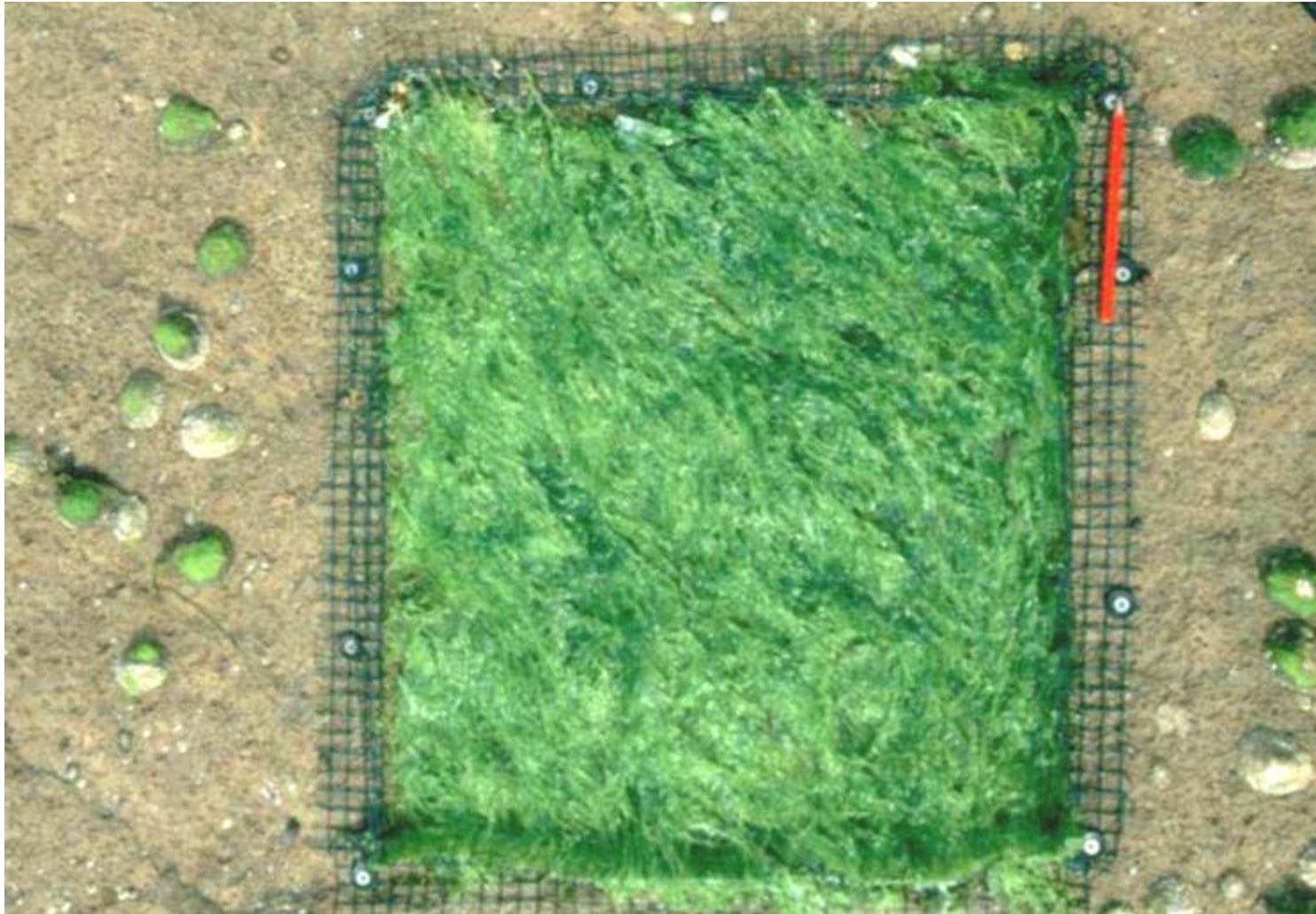
May be used in a range of situations, e.g.,

-to explain "extra" variation in experiment where the ANOVA design is of greatest interest



e.g.,

Presence/Absence of grazer in field enclosures
+ naturally varying temperature, slope, or nutrients....



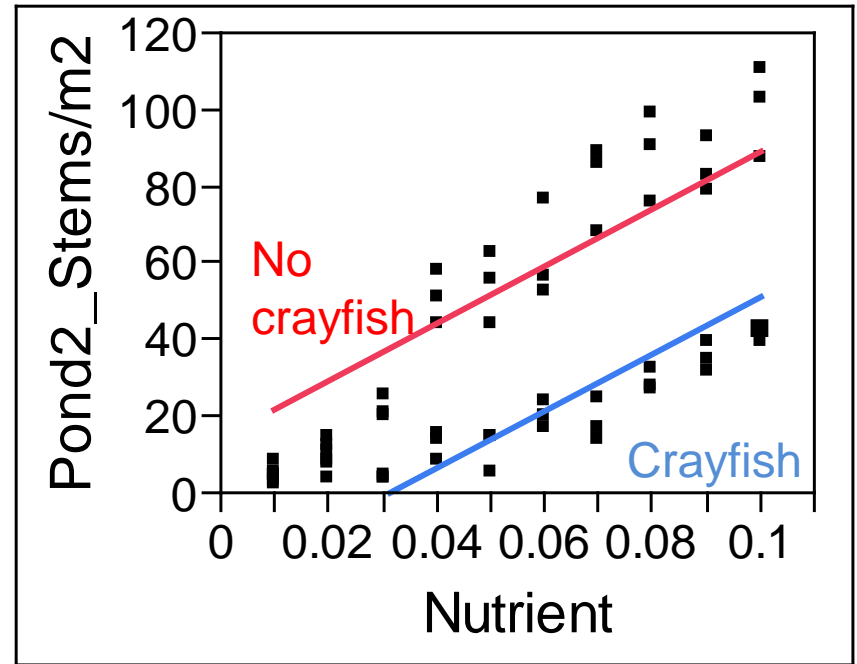
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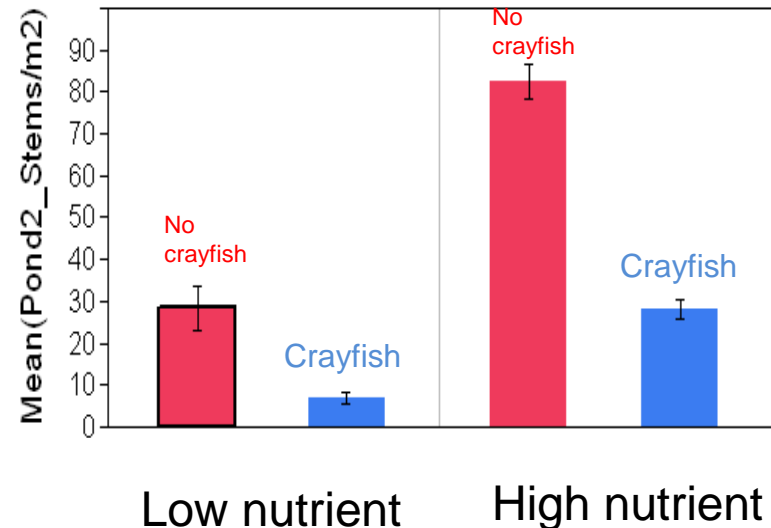
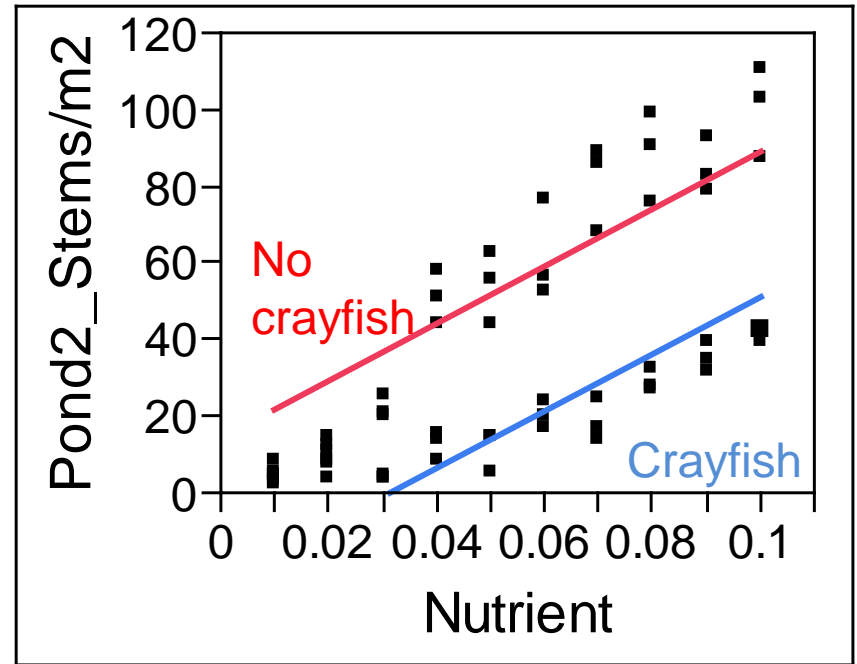
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If you try to “shoehorn” a continuous variable into discrete categories for a 2-way ANOVA, you will need to treat that factor as **random** rather than **fixed**



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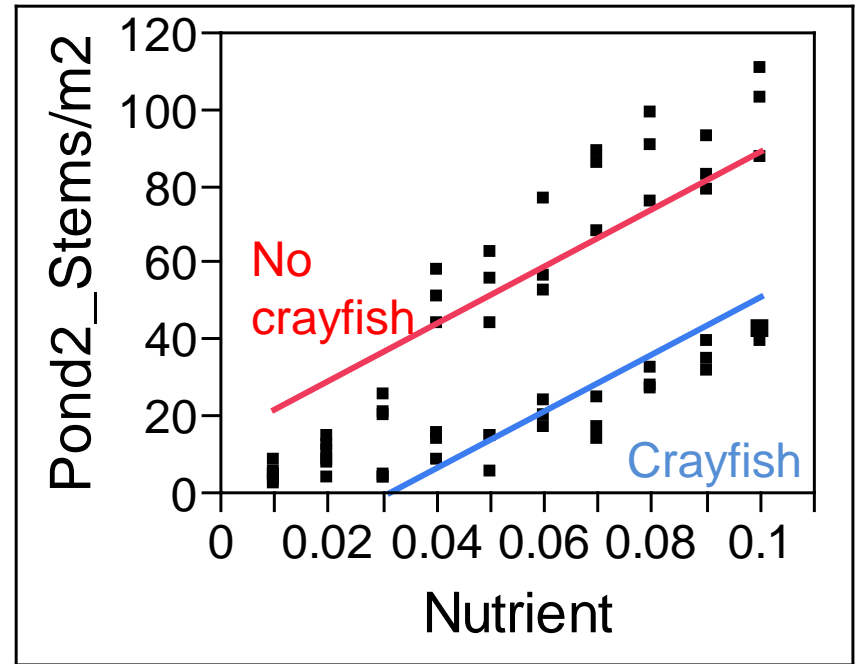
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Random – the treatment levels represent a random sample of all possible levels – inferences valid across other levels

Fixed – the levels included are the **ONLY** levels of interest

ANCOVA

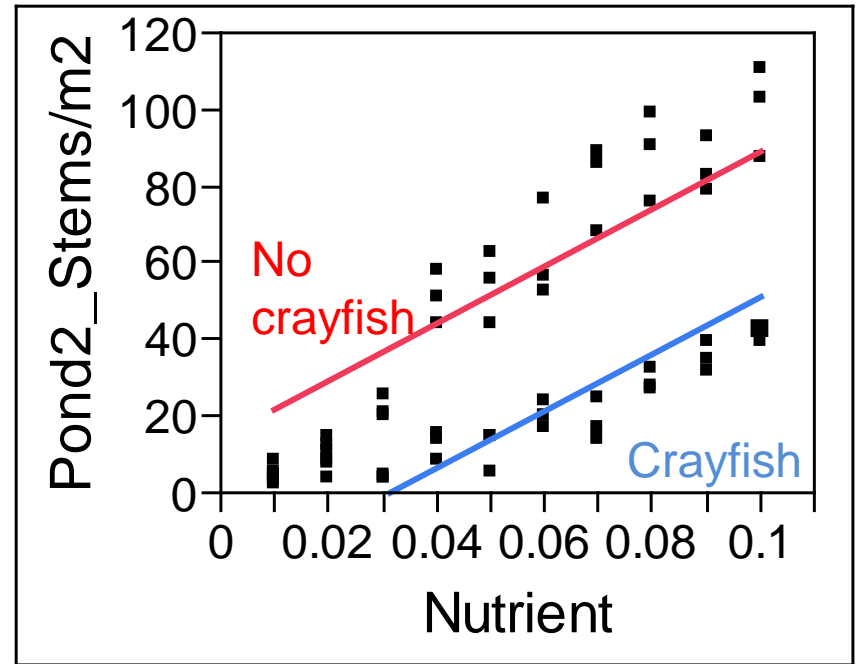
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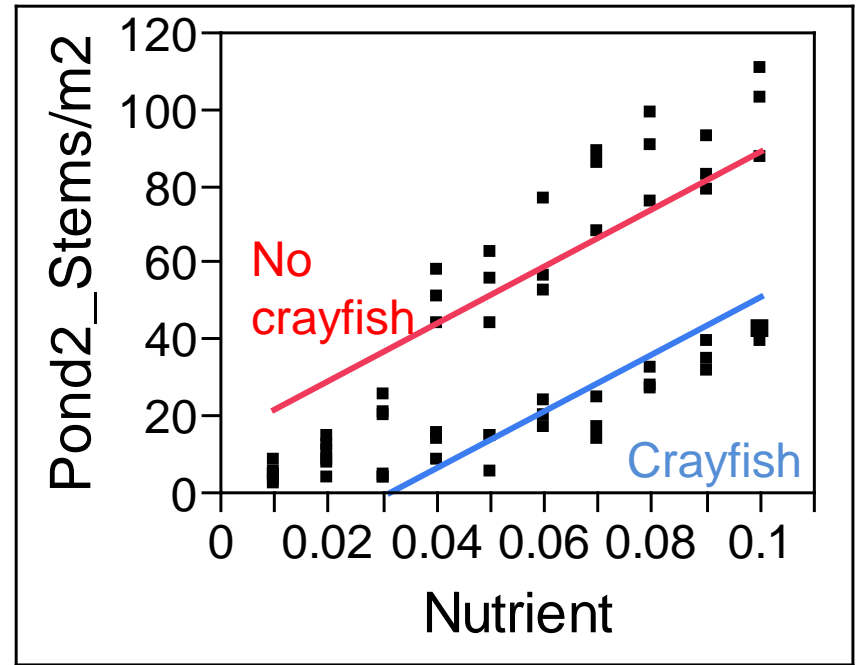
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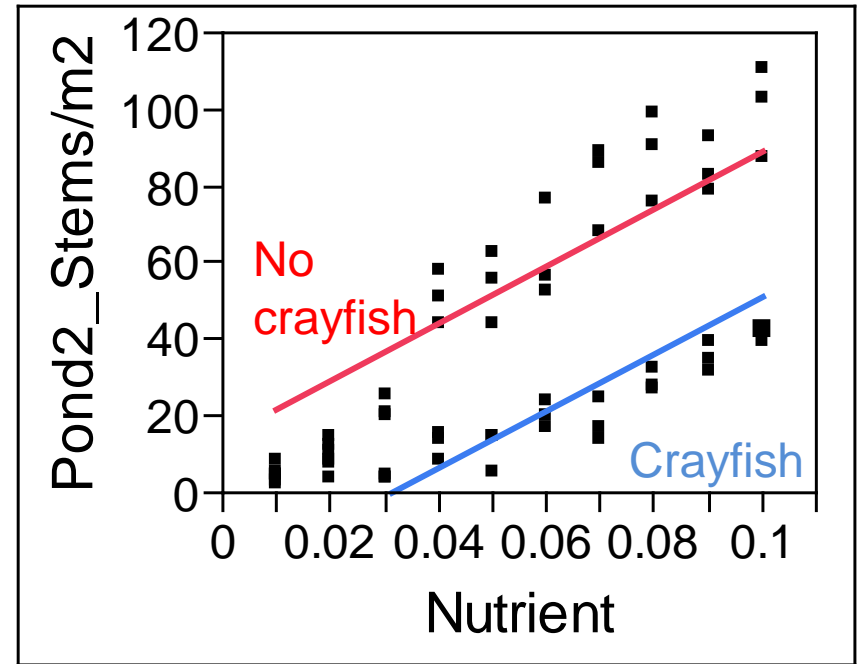
Who cares?! Calculations differ, significance values change dramatically

Why is it important to identify fixed vs. random factors?

Constructing F-values:

Fixed effect:

$$\frac{\text{Mean square}_{\text{factor}}}{\text{Mean square error}}$$



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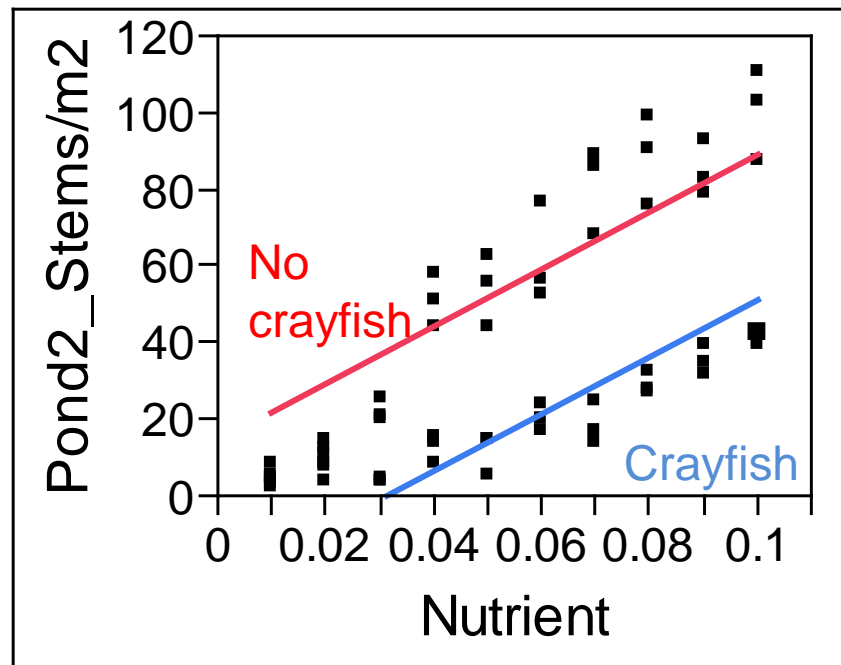
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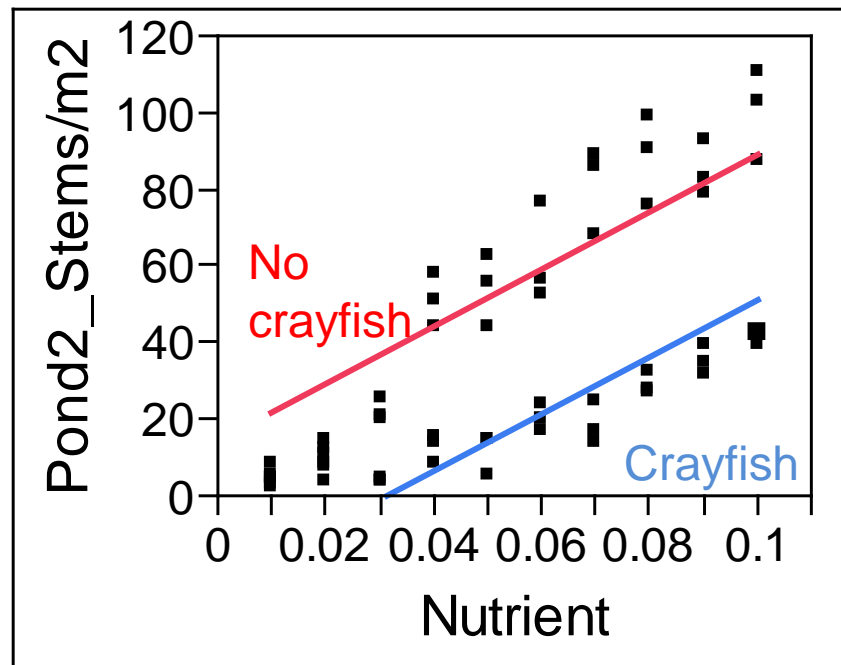
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Random effect:

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Mean square_{interaction term} is strongly affected by # levels and not by n – usually a bigger number, so F-value smaller (less likely to be significant)



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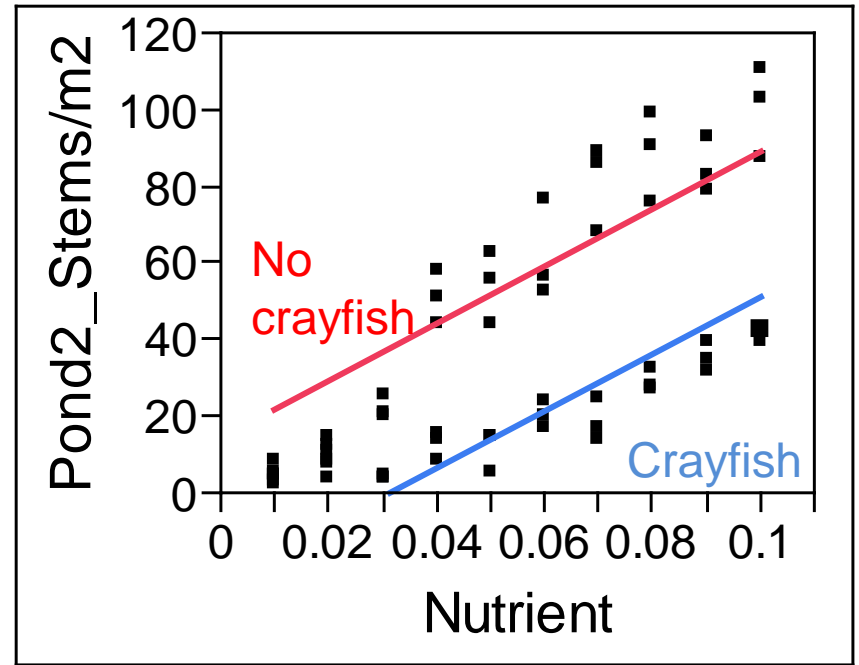
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Moral:

Do regression when regression makes sense

Do ANOVA when ANOVA makes sense

Do ANCOVA when ANCOVA makes sense



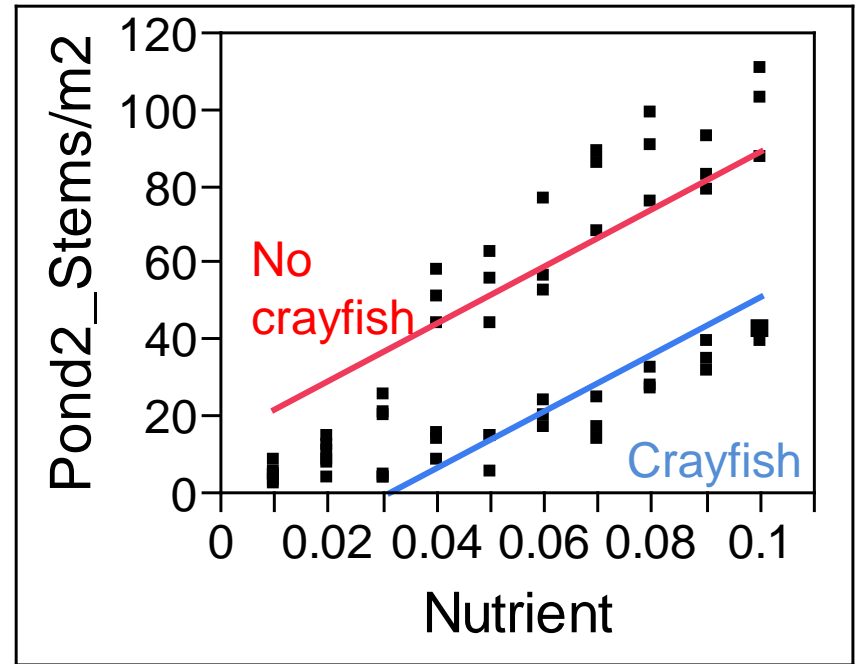
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Think about whether effects are random or fixed BEFORE you do your study – it may change your sampling approach



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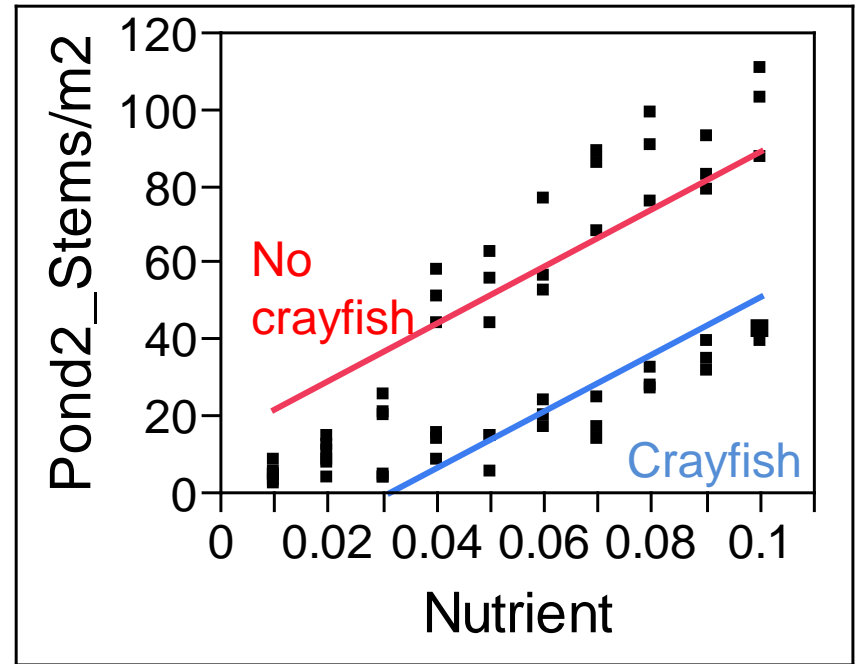
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More resources

ANCOVA, fixed/random effects:

Gotelli & Ellison. *A Primer of Ecological Statistics*. Sinauer.

Neter et al. *Applied Linear Statistical Models*. McGraw Hill.



Fixed/random effects in designing experiments with regression:

Cottingham et al (2005) Knowing when to draw the line: designing more informative ecological experiments. *Frontiers in Ecology and the Environment*. 3(3): 145–152

<http://microbes.kbs.msu.edu/Publications/Cottingham%20et%20al.%202005.pdf>

More about what fixed and random effects are, especially for ecologists:

Bennington & Thayne. 1994. Use and Misuse of Mixed Model Analysis of Variance in Ecological Studies. *Ecology* 75(3): 717-722.

<http://www.jstor.org/stable/1941729>