SEEC Stats Toolbox

Want to broaden your stats knowledge? Unsure of what you can do with your data? Still developing your proposal?

Join us for the monthly SEEC Stats Toolbox seminars where we introduce you to statistical methods that are useful for ecologists, environmental and conservation scientists.



Our next seminar:

 Topic:
 Intro to Multivariate Analyses

 Who:
 Dr Natasha Karenyi

 When:
 Thursday 20 April 2017 (1-2pm)

 Where:
 PD Hahn Lecture Theatre 3, PD Hahn Building Level 5, UCT

More details: www.seec.uct.ca.za







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Help 😡



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Our next seminar:

 Topic:
 Experimental and Survey Design

 Who:
 Prof Res Altwegg and Dr Birgit Erni

 When:
 Thursday 30 March 2017 (1-2pm)

 Where:
 PD Hahn Lecture Theatre 3, PD Hahn Building Level 5, UCT

More details: www.seec.uct.ca.za







Experimental and Survey Design

- ► Why?
- ► What?
- ► How?

Yoccoz, N. G., J. D. Nichols, and T. Boulinier. 2001. Monitoring of biological diversity in space and time. Trends in Ecology & Evolution 16:446 - 453.





Experimental and Survey Designs



from Mackenzie et al 2006, Academic Press





Experiments: what they are and why we need them

- carefully controlled environment
- everything stays constant
- except the factor of interest (treatment factor)
- measure change in response
- infer causal relationships





Observational vs experimental studies

Hypothetical example: Vegetation in relation to temperature and altitude





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Hypothetical example: Vegetation in relation to temperature and altitude



Summary







Three principles in experimental design

- 1. Randomization
- 2. Replication
- 3. Reduce (unexplained variation)





Three principles in experimental design

- 1. Randomization
- 2. Replication
- 3. Reduce (unexplained variation)

What is the experimental unit?





ED Principle 1: Randomization

Randomization

is the assignment of treatments to experimental units at random.





What happens if we don't randomize?







What happens if we don't randomize?







What happens if we don't randomize order?



time





What happens if we don't randomize order?







Summary: Why is randomization important?

- average out systematic effects (anticipated or not), extraneous factors not directly controlled (prevent confounding with unknown factors)
- ensures that treatment is independent of experimental unit's properties





ED Principle 2: Replication

each treatment is assigned to > 1 experimental unit, randomly and independently





Uncertainty – Variability

If all experimental units were exactly identical, it would be easy to establish cause and effect relationships.





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ED Principle 3: Reduce unexplained variation

- 1. keep all other factors constant
- 2. blocking

Blocking:

is the grouping of similar (homogeneous) experimental units into blocks.



- The experimental units in one block are similar.
- Each treatment occurs once in each block.
- Common blocking factors: age, spatial location, time (day or year)
- each block is a replicate



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Controls

- to measure the effect of a treatment, we need to compare response with to without treatment
- eliminate alternative explanations for results





Part 2: Constrained Design Studies

One or several principles of experimental design cannot be adhered to:

- impact studies
- observational studies





Sampling Designs for Impact Assessment



The decision key to the "main sequence" categories of environmental studies.

from Green 1979, John Wiley & Sons



Example: Impact of a new wind energy farm

Problem formulation:

- not an experiment, but impact assessment
- cannot randomly choose location
- only one farm (no replication of 'treatment')
- impact: effect of wind farm on bird behaviour around turbines
- hypothesis: change in the local density of raptors with presence of wind farm compared to abscence of wind farm
- sampling method: vantage point surveys











SEEC - Statistics in Ecology, Environment and Conservation

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o o o: temporal replication



000







o o o: temporal replication

spatial control









Random Sampling

Purpose: sample must be representative (of the population)

- where: where to place the vantage points
- when: times of day, season, weather conditions





No manipulation

- Cause and effect unclear
- Influence of unmeasured variables can never be ruled out (confounding)





No manipulation

- Cause and effect unclear
- Influence of unmeasured variables can never be ruled out (confounding)
- Two things to worry about:
 - 1. Observation process (Solution: distance sampling, repeated visits, multi-observer designs)
 - 2. Representativeness





1. Identify target population





- 1. Identify target population
- 2. Random sampling: no assumptions about the population needed





- 1. Identify target population
- 2. Random sampling: no assumptions about the population needed
 - every unit has a known probability of being in the sample
 - sample is drawn according to these probabilities
 - selection probabilities are used when making estimates





- 1. Identify target population
- 2. Random sampling: no assumptions about the population needed
 - every unit has a known probability of being in the sample
 - sample is drawn according to these probabilities
 - selection probabilities are used when making estimates
- **3.** Model-based sampling: unbiased if model describes population well





Random sampling

- Simple random sampling
- Stratified random sampling
- Cluster sampling
- Multistage sampling
- Adaptive sampling





Density of hadeda ibises





What is the average density of hadeda ibises in this landscape?





Simple random sampling



 $\mu = 1.98$ $\bar{x} = 1.35$ 95% CI : 0.21 - 2.49

Unbiased regardless of how densities are distributed.





Convenience sampling



 $\mu = 1.98$ $\bar{x} = 3.05$ 95% *CI* : 1.91 - 4.18

Preferentially sampling high-density areas leads to estimates that are biased high.





Stratified random sampling



 $\mu = 1.98$ $\bar{x} = \frac{1}{N} \sum_{h=1}^{L} N_h \bar{x}_h = 1.75$ $s^2 = \sum_{h=1}^{L} (\frac{N_h}{N})^2 (\frac{N_h - n_h}{N_h}) \frac{s_h^2}{n_h}$ 95%CI : 1.05 - 2.44

Strata should be as homogeneous as possible; blocking is a related idea in experimental design.





Model-based sampling



habitat gradient

Can be unbiased if the model is good; no random sampling needed.





Cluster sampling





- Clusters selected at random
- Most efficient if variability within cluster is high and variability among clusters is low





Multistage sampling





2. Select small units at random within selected large units





Adaptive sampling





- 1. Select units at random
- 2. If species detected, sample surrounding units





Random sampling

- Probability of being in the sample is known for each unit
- This probability is used when making inference
- Lots of designs to choose from
- Some designs are more effective than others for your situation
- Things to think about:
 - Scale of variability and spatial correlation (do a pilot)
 - Do you have informative covariates? (stratification; model-based sampling)
 - Costs of travelling to sites
- Things get complicated quickly





Summary: Nine principles of good study design

- 1. formulate problem (design, predictors, hypothesis ...)
- 2. replicate time, location
- 3. random sample
- 4. control (space and time)
- 5. conduct a pilot study
- 6. worry about observation process
- 7. subsampling, stratified sampling where necessary
- 8. appropriate sample unit size
- 9. use an appropriate statistical model

Adapted from: Green 1979, John Wiley & Sons





Key references

Experimental design:

- Underwood, A. J. 1997. Experiments in ecology: their logical design and interpretation using analysis of variance. Cambridge University Press.
- Montgomery, D. C. 2012. Design and Analysis of Experiments, 8th edition. Wiley.

Constrained designs / impact studies:

 Green, R. H. 1979. Sampling design and statistical methods for environmental biologists. John Wiley & Sons.

Survey design:

Thompson, S. K. 2012. Sampling. 3rd edition. John Wiley & Sons.





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