

After dark, all CATs are leopards

Casting Light on the Use of CPCAT (Closure Principle Computational Approach Test) in Non-Target Soil Organism Assessments

Meet us booth 242

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Background

Recently, the 'Closure Principle Computational Approach Test' (CPCAT) was developed as a method that should overcome the shortcomings of NOEC/LOEC application in ecotoxicological pesticide risk assessments¹. CPCAT is supposed to handle abundance data characterized by low abundances, Poisson distributed, and overdispersion without restrictions. Assuming that abundance data is Poisson

distributed, the method is therefore recommended as a valuable tool e.g. for non-target arthropod and soil organism assessments under field conditions in the draft soil guidance document. In this context, a common approach for guidance conform earthworm and mesofauna data collection is to transform low-resolution sampling data to a square meter scale before performing the statistical analysis^{2,3}.

Aim

Investigating the sensitivity of CPCAT to data transformation and its implications for the reliability of results exemplified for abundance data of non-target soil organisms under field conditions.

Methods

1. Simulate dataset

Create datasets to mimic a situation with no differences between control and treatments from which no statistical significances can be expected and false positives can easily be identified.

- Simulate a dataset with control + two treatment levels (to be suitable for multiple comparisons)
- Draw random samples (n=5) for control and treatment levels from the same Poisson distribution with λ=6.
 => no differences between control and treatment!
- Multiply nominal dataset with factor 10 to 200 in steps of 10 to simulate upscaling from a sample scale to an arbitrary analysis scale
- · Simulate 100 replicates to cover random variation

2. Calculate p-values with CPCAT

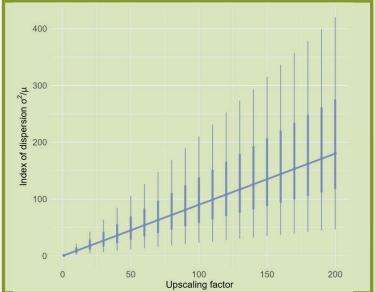


Figure 1: Index of dispersion (ratio of variance over mean) of the simulated data. An index of 1 represent equality of variance and mean, which is expected from a Poisson distribution; index values > 1 represent over-dispersed data. Here the upscaling leads to a linear increase of the ratio since the variance increases quadratically and the mean linearly.

Results

- High sensitivity of CPCAT to upscaled Poisson sampling data, leading to decreased p-values and consequently an inflation in the false-positive rate which is correlated with the extent of the transformation applied.
- Upscaling particularly magnified the dispersion of the data in violation of CPCAT's assumption regarding Poisson-distributed data.
- The CPCAT algorithm uses Poisson-distributed data to evaluate the observed data and lacks a variance stabilizing mechanism to deal with over-dispersed data.

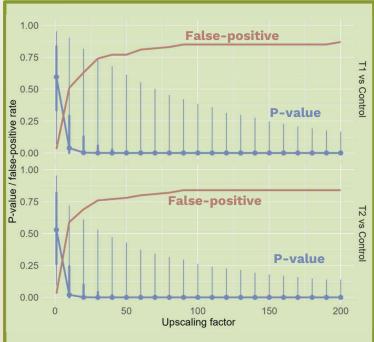


Figure 2: CPCAT p-values for a range of simulated datasets comparing two treatments (T1 and T2) to a control with no differences in their means along a gradient of different multiplication factors upscaling the Poisson data. The blue line denotes the resulting p-values with 50% quantile, 50% and 90% interquartile ranges (25%-75% and 5%-95% quantiles, respectively). The red line indicated the false-positive rate.

Conclusion

- · Robustness and reliability of CPCAT results under the current recommendations cannot be guaranteed
- Careful application of CPCAT to (scaled) abundance data (validation of method's suitability is required)
- CPCAT should be extended to be able to handle over-dispersed abundance data
- Guidelines should consider the current limitations of CPCAT in their suggested procedures