

Estimating animal abundance using automatically derived observation distances

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1. Why do we need observation distances?

- Counting animals is difficult for most species, because we cannot identify individuals
- Observations at random locations → representative sample
- Camera trapping has become an increasingly popular approach



The lynx is rather an exception...

1. Why do we need observation distances?

- The probability to observe an animal decreases with its distance from a camera trap
- The form of this decrease depends on the properties of the animals, the vegetation and terrain, as well as the weather





Example: Red deer observations in the Bavarian Forest National Park in different distances from the camera trap (in metres)

$$D = \frac{2t \sum_{k=1}^{K} n_k}{\Theta \omega^2 \sum_{k=1}^{K} T_k \hat{P}_k} \frac{1}{\hat{A}}$$





K= Number of camera trap locations

 n_k = Number of observed animals at camera trap location \boldsymbol{k}

 Θ = Angle of view

 ω = Truncation distance

 T_k = Deployment time



 \hat{P}_k = Estimated probability of obtaining an image of an animal within Θ and ω at a snapshot moment

 \hat{A} = Activity level

2. How to obtain observation distances?



2. How to obtain observation distances?

 Manual estimation of observation distances based on distance markers in 1,2,..., 15 m distance



3. Semi-automatic distance estimation

- Animal detection via MegaDetector (Beery et al. 2019)
- Relative depth estimation by a deep learning algorithm (DPT, Ranftl et al. 2021)
- Transformation to absolute distance estimates based on at least two reference images with an object in a known distance
- The 20th percentile inside the bounding box around an animal is extracted → estimated distance to the animal

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Overcoming abundance w	the distance estimation bottleneck in estimating animal rith camera traps	Check for updates
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5 m



10 m

25

20

5

0 Depth [m]



4. Example datasets







Distance from camera trap (manual) [m]



Distance from camera trap (manual) [m]

Bavaria



Distance from camera trap (manual) [m]





5



Semi-automatically estimated distance [m]

Brandenburg



Semi-automatically estimated distance [m]



Semi-automatically estimated distance [m]

7. Comparison of the detection probabilities



Species (per area)

8. Comparison of the population density estimates



9. Conclusions

- Semi-automatic distance estimates can reduce the time and effort that are needed for the population density estimation of unmarked species
- The number of false negatives is generally not related to the distance from the camera trap
- The agreement of manual and semi-automatic distance estimates is best at ca. 4 m
- Population density estimates are often robust, but problematic cases require further attention



Thank you for your attention