

11.0 Appendix A

Table A1. Summary of the [assumptions](#) and pros/cons of the different [modelling approaches](#) (adapted from Wearn & Glover-Kapfer [2017] and Clarke et al. [2022]).

| Objective | Approach | Assumptions | Pros | Cons | References |
|--|---|---|---|---|---|
| Species inventory | Species inventory | <ul style="list-style-type: none"> No formal assumptions¹ | <ul style="list-style-type: none"> Maximum flexibility for study design (e.g., camera days per camera location or use of lure²)¹ | <ul style="list-style-type: none"> Not reliable estimates for inference ("considered as unfinished, working drafts")¹ | ¹ Wearn & Glover-Kapfer, 2017 |
| Species richness | Species richness | <ul style="list-style-type: none"> Cameras are randomly placed¹ Cameras are independent¹ detection probability of different species is equal¹ ("True" species richness estimation involves attempting to correct for "imperfect detection"¹) | <ul style="list-style-type: none"> Fundamental to ecological theory and often a key metric used in management¹ Simple to analyze, interpret and communicate¹ Models exist to estimate asymptotic species richness, including unseen species (simple versions of these models - EstimateS and the "vegan" R-packages)¹ | <ul style="list-style-type: none"> Dependent on the scale (as captured in the species-area relationship)¹ All species have equal weight in calculations, and community evenness is disregarded¹ Insensitive to changes in abundance, community structure and community composition¹ | ² Rovero et al., 2013 ³ MacKenzie et al., 2002 ⁴ MacKenzie et al., 2006 ⁵ Lambert, 1992 ⁶ Mullahy, 1986 ⁷ McCullagh & Nelder, 1989 |
| Species diversity | Species diversity | <ul style="list-style-type: none"> Cameras are randomly placed¹ Cameras are independent¹ The detection probability of different species remains the same¹ | <ul style="list-style-type: none"> Captures evenness and richness (although some indices only reflect evenness)¹ Most indices are easy to calculate and widely implemented in software packages (e.g., EstimateS and "vegan" in R)¹ | <ul style="list-style-type: none"> Many diversity indices exist, and it can be difficult to choose the most appropriate¹ Interpretation/communication not always straightforward¹ Insensitive to changes in community composition¹ (though this may be conditional on study design) | ⁸ Zorn, 1998 ⁹ Royle & Nichols, 2003 ¹⁰ MacKenzie et al., 2006 ¹¹ Karanth & Nichols, 1998 ¹² Karanth, 1995 ¹³ Clarke et al., 2023 ¹⁴ Noss et al., 2003 |
| Species diversity | β -diversity | <ul style="list-style-type: none"> Can be used to track changes in community composition¹ Plays a critical role in effective conservation prioritization (e.g., designing reserve networks)¹ Important for detecting changes in the fundamental processes¹ | <ul style="list-style-type: none"> Many measures; no single best measure for all purposes¹ Comparing measures across space, time and studies can be very difficult¹ Scale-dependent (i.e., the size of the communities that are being included)¹ | | ¹⁵ Kelly et al., 2008 ¹⁶ Moeller et al., 2018 |
| Occupancy ³ | Occupancy models ³ | <ul style="list-style-type: none"> Closed to changes in occupancy^[3] (abundance is constant)⁴ Sites and detections are independent⁴ The probability of occupancy and detection are constant across all sites | <ul style="list-style-type: none"> Does not require individual identification⁴ Just requires detection/non-detection data for each site¹ | <ul style="list-style-type: none"> Occupancy^[3] only measures distribution; it may be a misleading indicator of changes in abundance¹ Interpretation/communication of results may not be straightforward (if the scale | |

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| | | <p>within a stratum or can be modelled using covariates⁴</p> <ul style="list-style-type: none"> Species are not misidentified⁴ | <ul style="list-style-type: none"> Relatively easy-to-use software exists for fitting models (PRESENCE, MARK, and the “unmarked” R package)¹ “Open” models exist that allow for the estimation of site colonization and extinction rates^{1,4} Multi-species occupancy models^[3] allow the inclusion of interactions among species while controlling for imperfect detection¹ | <p>of movement is much larger than the camera spacing the results should be interpreted as “probability of use” rather than occupancy¹</p> | ¹⁷ Chandler & Royle, 2013 ¹⁸ Borchers & Efford, 2008 ¹⁹ Efford, 2004 ²⁰ Royle & Young, 2008 ²¹ Royle et al., 2009 |
| Relative abundance indices | Poisson | <ul style="list-style-type: none"> Since used for many approaches, many assumptions exist¹ | <ul style="list-style-type: none"> Simple to calculate and technically possible (even with small sample sizes when robust methods might fail)¹ Relative abundance indices often do correlate with abundance¹ Calibration with independent density estimates is possible¹ | <ul style="list-style-type: none"> Difficult to draw inferences (a large number of assumptions); comparisons across space, time, species, and studies are difficult¹ Requires stringent study design (e.g., random sampling, standardized methods)¹ | ²² O'Brien et al., 2011 ²³ Doran-Myers, 2018 ²⁴ Morin et al., 2022 ²⁵ Green et al., 2020 ²⁶ Parmenter et al., 2003 ²⁷ Noss et al., 2012 |
| | Zero-inflated Poisson (ZIP) ⁵ | | | | |
| | Negative binomial (NB) ⁶ | | | | |
| | Zero-inflated negative binomial (ZINB) ⁷ | | | | |
| | Hurdle models ⁸ | | | | |
| | Other | | | | |
| Absolute abundance; Unmarked population | Royle-Nichols model ^{9,10} | <ul style="list-style-type: none"> Individual detection probability is constant¹ | <ul style="list-style-type: none"> Can relax assumption of constant abundance¹ Abundance is a fundamental parameter in wildlife research and monitoring¹ Can be applied to unmarked species¹ Only requires detection/non-detection data for each site (not counts)¹ May be used in models with relative abundance to control for imperfect detection¹ | <ul style="list-style-type: none"> Assumes a relatively specific relationship between local abundance and species-level detection probability¹ Depends on sampling area¹ Requires all or some of a population to be marked¹ No dedicated, simple software for this model (but can be implemented in MARK and the “unmarked” package in R)¹ | ²⁸ Sollmann et al., 2013a ²⁹ Sollmann et al., 2013b ³⁰ Rich et al., 2014 ³¹ Whittington et al., 2018 ³² Efford et al., 2009b |
| Population size / Absolute abundance / vital rates / | Capture-recapture (CR) / capture-mark-recapture (CMR) ^{11,12} | <ul style="list-style-type: none"> Demographically closed (i.e., no births or deaths)¹ Geographically closed (i.e., no immigration or emigration)¹ | <ul style="list-style-type: none"> May be used as a relative abundance index that controls for imperfect detection¹ Easy-to-use software exists to implement (e.g., CAPTURE); MARK | <ul style="list-style-type: none"> Requires that individuals are distinguishable.¹ (However, CR^[11,12] has also been used to estimate abundance of species that lack natural markers but that have phenotypic | ³³ Royle et al., 2014 |

| <u>Objective</u> | <u>Approach</u> | <u>Assumptions</u> | <u>Pros</u> | <u>Cons</u> | <u>References</u> |
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| Density: Marked population | | <ul style="list-style-type: none"> Each individual has at least some probability of being captured² Overall sampled area should encompass full extent of individuals movements^{2,11} Activity centres are randomly dispersed and stationary¹³ | <p>Implements more complicated models with covariates (and must be used for mark-resight modelling)¹</p> <ul style="list-style-type: none"> Can use the robust design with “open” models to obtain recruitment and survival rate estimates¹ | <p>and/or environment-induced characteristics^{2,14,15} When the sample size is large enough to reliably estimate density with CR,^[11,12] individuals are unlikely to have a unique marker^{2,14,15})</p> <ul style="list-style-type: none"> Dependent on the surveyed area, which is difficult to track and calculate¹ Requires a minimum number of captures and recaptures¹ Relatively stringent requirements for study design (e.g., no “holes” in the trapping grid)¹ Geographic closure at the plot level, which is often unrealistic¹⁶ Assumes a specific relationship between abundance and detection¹ Density cannot be explicitly estimated because the true area animals occupy is never measured (only approximated)¹⁷ | <p>³⁴ Augustine et al., 2019</p> <p>³⁵ Bugar et al., 2018</p> <p>³⁶ Sun et al., 2022</p> <p>³⁷ Sollmann, 2018</p> <p>³⁸ Augustine et al., 2018</p> <p>³⁹ Davis et al., 2021</p> <p>⁴⁰ Rowcliffe et al., 2008</p> <p>⁴¹ Rowcliffe et al., 2013</p> <p>⁴² Rowcliffe et al., 2014</p> <p>⁴³ Rowcliffe et al., 2016</p> |
| Density / population size; Marked population | Spatially explicit capture recapture (SECR) ¹⁸⁻²¹ (also referred to as Spatial capture-recapture [SCR]) | <ul style="list-style-type: none"> Individuals do not lose marks or are misidentified¹ All animals have an equal probability of capture (or, for spatially explicit models, an equal probability of capture for a given distance from the centre of their home range)¹ Captures of different individuals are independent¹ No behavioural response to being trapped or marked¹ Sampling occasions are independent¹ Population is demographically closed (i.e., no births or deaths)¹ For conventional models, geographically closed, i.e., no immigration or emigration)¹ | <ul style="list-style-type: none"> Produces direct estimates of density or population size for explicit spatial regions¹⁷ Allows researchers to mark a subset of the population/to take advantage of natural markings¹ Estimates are fully comparable across space, time, species and studies¹ Density estimates obtained in a single model, fully incorporate spatial information of locations and individuals¹ Both likelihood-based and Bayesian versions of the model have been implemented in relatively easy-to-use software (DENSITY and SPACECAP, respectively, as well as associated R packages)¹ | <ul style="list-style-type: none"> Requires that individuals are identifiable¹ Requires that a minimum number of individuals are trapped (each recaptured multiple times ideally)¹ Requires that each individual is captured at a number of camera locations¹ Multiple cameras per station may be required to identify individuals; difficult to implement at large spatial scales as it requires a high density of cameras^{13,24} May not be precise enough for long-term monitoring²⁵ Cameras must be close enough that individuals come into contact with multiple cameras^{1,17} | <p>⁴⁴ Rowcliffe et al., 2011</p> <p>⁴⁵ Cusack et al., 2015</p> <p>⁴⁶ Nakashima et al., 2017</p> <p>⁴⁷ Meek et al., 2016</p> <p>⁴⁸ Anile & Devillard, 2016</p> <p>⁴⁹ Huggard, 2018</p> <p>⁵⁰ Becker et al., 2022</p> <p>⁵¹ Warbington & Boyce, 2020</p> |

| Objective | Approach | Assumptions | Pros | Cons | References |
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| | | <ul style="list-style-type: none"> Spatially explicit models have further assumption about animal movement.^{1,18,21,22} These include: Home ranges do not change during the survey¹ Captures does not affect movement patterns¹ Random placement with respect to the distribution and orientation of home ranges¹ Distribution of home range centres follows a defined distribution (Poisson, or other, e.g.,)¹ | <ul style="list-style-type: none"> Flexibility in study design (e.g., "holes" in the trapping grid)¹ "Open" SECR^[18-21] models exist that allow for estimation of recruitment and survival rates¹ "Avoid ad-hoc definitions of study area and edge effects"²³ SECR^[18-21] accounts for variation in individual detection probability; can produce spatial variation in density; SECR^[18-21] more sensitive "to detect moderate-to-major populations changes" (+/-20-80%)^{13,24} | <ul style="list-style-type: none"> ½ MMDM (Mean Maximum Distance Moved) will usually lead to an under - estimation of home range size and thus overestimation of density^{1,26,27} | ⁵² Howe et al., 2017 ⁵³ Palencia et al., 2021 ⁵⁴ Gilbert et al., 2020 ⁵⁵ Twining et al., 2022 ⁵⁶ Bessone et al., 2020 ⁵⁷ Moeller et al., 2018 |
| Density: Marked population | Spatial mark-resight (SMR) (type of SCR model) ^{17,28,29} | <ul style="list-style-type: none"> Demographic and geographic closure of the population during the survey¹ Detections are independent²⁹ Detection probability decays with increasing distance of the camera from the activity centre^{29,30} Animals have stable activity centres³⁰ Individual marks are not lost (for maximum precision), but SMR^[17,28,29] does allow for inclusion of marked but unidentified resighting detections^{28,31} The number of marked animals present is known before resightings^{28,30} Animals are ungrouped²⁹ Counts of unmarked animals are modeled with a Poisson distribution³⁰ Cameras randomly placed with respect to activity centres²⁸ Marked animals are a random sample of the population with home ranges located inside the state space^{29,30} All animals have stable activity centres within home ranges where detection probability is greatest^{28,32} | <ul style="list-style-type: none"> Estimates are fully comparable to SECR^[18-21] of marked species¹ Can be applied to a broader range of species than SECR^[18-21]¹ Allows researcher to take advantage of natural markings¹ Allows researcher to mark a subset of the population (note - precision is dependent on number of marked individuals in a population)¹ | <ul style="list-style-type: none"> Animals may have to be physically captured and marked if natural marks do not exist on enough individuals¹ All individuals must be identifiable¹ Allows for density estimation for a unmarked population, but the precision of the density estimates are likely to be very low value¹ Remains poorly tested with camera data, although it offers promise¹ Density estimates are likely less precise than with SECR^[18-21] or REM, unless a large proportion of the population have marks¹ Requires sampling points to be close enough that individuals encounter multiple cameras¹ | ⁵⁸ Loonam et al., 2021 ⁵⁹ Bridges & Noss, 2011 ⁶⁰ Rovero & Zimmermann, 2016 ⁶¹ Borchers & Marques, 2017 |

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| Density; Unmarked population | Spatial count (SC) (type of SCR model) ^{17,33} | <ul style="list-style-type: none"> • Camera must be close enough together that animals are detected at multiple cameras^{13,17} • Population closure^{13,17} • Independence of detections^{13,17} • Activity centres randomly dispersed^{13,17} • Activity centres are stationary^{13,17} | <ul style="list-style-type: none"> • Does not require individual identification¹³ | <ul style="list-style-type: none"> • Produces imprecise estimates even under ideal circumstances unless it is supplemented with auxiliary data (e.g., telemetry)^{17,23,28,29} • Precision decreases with an increasing number of individuals detected at a camera²⁴ (as overlap of home ranges of individuals' increases)^{13,34} • Not appropriate for low density or elusive species when recaptures too few to confidently infer the number and location of activity centres^{13,35} • Not appropriate for high-density populations with evenly spaced activity centres (camera[-specific] counts will be too similar and impair activity centre inference)¹³ • Ill-suited to populations that exhibit group-travelling behaviour^{13,36} • Study design (camera arrangement) can dramatically affect the accuracy and precision of density estimates^{13,37} • Cameras must be close enough that animals are detected at multiple camera locations (may be challenging to implement at large scales as many cameras are needed)^{13,17} | |
| Density / population size; Partially Marked population | Spatial Partial Identity Model (Categorical SPIM; catSPIM) ^{13,34,36} (Extension of SC model using animal traits (e.g. Sex Class , antler points) and model parameters) | <ul style="list-style-type: none"> • Same as SC^{13,34,36} • Each categorical identifier (e.g., Sex Class, collar, etc) has fixed number of possibilities (e.g., male/female, collared/not collared)³⁶ • All possible values of categorical identifiers occur in the population with probabilities that can be estimated^{13,34,36} • Every individual is assigned "full categorical identity" (i.e., "set of traits given all categorical identifiers and possibilities")^{13,34} | <ul style="list-style-type: none"> • May produce more precise and less biased density estimates than SC with less information^{13,36} | <ul style="list-style-type: none"> • Sensitive to non-independent movement (e.g., group-travel; can cause over-dispersion and bias estimates^{13,36}); may limit application to solitary species only^{13,36} • May produce be less reliable/accurate estimates for high-density populations^{13,36} • To few categorical identifiers/possibilities can result in mis-assignments and overestimating density^{13,34,26} | |

| Objective | Approach | Assumptions | Pros | Cons | References |
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| | | <ul style="list-style-type: none"> There is no change in an individual's identity trait during the survey period (e.g., antlers present/absent)³⁴ | | | |
| Density / population size; Partially Marked population | Spatial Partial Identity Model (2-flank SPIM) ^{13,38} (extension of SCR model augmented with data from partially-identifying images) | <ul style="list-style-type: none"> Same as SCR^{13,38} Capture processes for left-side, right-side and both-side images are independent^{13,38} | <ul style="list-style-type: none"> Same as SCR^{13,38} Improved precision of density estimates relative to SCR^{13,38,39} Many study designs can be used (paired sample stations, single camera locations, and hybrids of both paired- and single camera locations)^{13,38,39} Can be used with single-camera and hybrid sampling designs, and therefore requires fewer cameras (or sample more area) than SCR^{13,38} May be more robust to non-independence than SC^{34,38} | <ul style="list-style-type: none"> Computationally intensive^{13,38} Increased precision is less pronounced in high-density populations^{13,38} | |
| Density; Unmarked | Random encounter models (REM) ^{40,41} | <ul style="list-style-type: none"> Demographic and geographic closure^{23,40} Random with respect to animal movement^{1,40} Animal movement is not affected by the cameras^{1,40} Independent "contacts" between camera locations can be accurately counted^{1,40} Unbiased estimates of animal activity levels and animal speed can be obtained^{1,42,43} Camera's detection zone can be approximated well using a 2D cone shape, defined by the radius and angle parameters⁴⁴ If activity and speed are to be estimated from camera data, two additional assumption: <ul style="list-style-type: none"> All animals are active during the peak daily activity⁴² Animals moving quickly past a camera are not missed⁴³ | <ul style="list-style-type: none"> Flexible study design (e.g., "holes" in grids allowed, camera spacing less important)¹ Can be applied to unmarked species¹ Allows community-wide density estimation¹ Outputs also include informative parameter estimates (i.e., animal speed and activity levels, and detection zone parameters)¹ Comparable estimates to SECR^[18-21]¹ Does not require marked animals or identification of individuals^{23,40} Can use camera spacing without regard to population home range size^{23,40} Direct estimation of density; avoids ad-hoc definitions of study area⁴⁰ | <ul style="list-style-type: none"> Requires relatively stringent study design, particularly (e.g., random sampling and use of bait or lure)¹ Requires independent estimates of animal speed or measurement of animal speed within videos¹ No dedicated, simple software¹ Random relative to animal movement, grid preferred, avoid multiple captures of same individual, area coverage important for abundance estimation² Possible sources of error include inaccurate measurement of detection zone and movement rate^{41,45} | |

| Objective | Approach | Assumptions | Pros | Cons | References |
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| Density; Unmarked | Random encounter and staying time (REST) ⁴⁶ | <ul style="list-style-type: none"> The population is closed (animal density is constant during the survey)⁴⁰ The detection probability is perfect¹ ($p = 1$) unless otherwise modelled⁴⁶ Camera locations are representative of the available habitat⁴⁶ Cameras are randomly placement with respect to the spatial distribution of animals⁴⁶ Animal movement and behaviour are not affected by cameras⁴⁶ Observations are independent⁴⁶ The observed distribution of staying time in the focal area fits the distribution of movement⁴⁶ The observed staying time must follow a given parametric distribution⁴⁶ | <ul style="list-style-type: none"> Provides unbiased estimates of animal density, even when animal movement speed varies, and animals travel in pairs⁴⁶ | <ul style="list-style-type: none"> Attraction or aversion to cameras is exhibited in some species⁴⁷ and could affect the time within the detection zone and subsequently affect estimates of density²³ Requires accurate measurements of the area of the camera detection zone, which has been a challenge in previous studies^{23,44-46,48} Mathematically challenging⁴⁵ | |
| Density; Unmarked | Time in front of the camera (TIFC) ⁴⁹⁻⁵¹ | <ul style="list-style-type: none"> Cameras are placed randomly, or representative relative to animal movement⁵⁰ No influence of cameras on animal movement⁵⁰ Reliable detection of animals in part of the camera's FOV (at least)⁵⁰ | <ul style="list-style-type: none"> Does not require individual identification⁵¹ Makes no assumption about home range⁵¹ Comparable to estimates from SECR^{[18-21]51} | <ul style="list-style-type: none"> Requires careful calculation of the effective area of detection⁵¹ A high level of measurement error⁵⁰ | |
| Density; Unmarked | Distance sampling (DS) ^{52,61} | <ul style="list-style-type: none"> Random or systematic random placements consistent with the assumption that points are placed independently of animal locations⁵² Placed randomly with respect to animal movement⁵³ Certain detection at distance 0⁵³ Certain detection at focal area⁵³ Closed population⁵³ Animal movement and behaviour not affected by the cameras⁵³ | <ul style="list-style-type: none"> A shortcut to controlling for variation in detection distances by only counting individuals within a short distance with an unobstructed view, and well sampled across cameras and species¹ Density estimates are unbiased by animal movement "since camera-animal distance is measured at a certain instant in time (intervals of duration t apart)"^{13,52} Can be applied to low-density populations^{13,53} | <ul style="list-style-type: none"> May require discarding a portion of the dataset (when the best fitting model truncates the dataset)¹ Biased by movement speed⁵³ Best suited to larger animals; the smaller the focal species, the lower [wildlife] cameras must be set, which reduces the depth of the viewshed, and thus sampling size and the flexibility of the model^{13,52} Does not permit inference about spatial variation in abundance (unless using hierarchical distance which can model | |

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| | | <ul style="list-style-type: none"> Animals detected at initial locations (e.g., they do not change course in response to the camera prior to detection)⁵³ Distances are measured exactly (however if the data from different distances will be grouped ("binned") for analysis later, an accuracy of +/- 1m may suffice)⁵³ Observations are independent events⁵³ Snapshot moments selected independently of animal locations⁵³ | <ul style="list-style-type: none"> Does not require individual identification⁵² | <ul style="list-style-type: none"> spatial variation as a function of covariates)^{13,54} "Calculating camera-animal distances can be labour-intensive and time-consuming (However, recently developed techniques (e.g., Johanns et al., 2022) show promise for simplifying and automating the process)"¹³ Requires good understanding of the focal populations' activity patterns; density estimates can be biased (e.g., under-estimated) when regular periods of inactivity are not accounted for (using detection times to infer periods of activity may help overcome this limitation)"^{13,52,53} Tends to underestimate density^{13,52,55} Low population density and reactivity to cameras may be major sources of bias"^{13,56} | |
| Density; Unmarked | Time-to-event (TTE) model ⁵⁷ | <ul style="list-style-type: none"> Demographic and geographic closure^{57,58} Locations randomly placed, systematic, systematic random⁵⁷ Independent detections in space and time⁵⁷ Spatial counts of animals (or counts in equal subsets of the landscape) are Poisson-distributed⁵⁸ Accurate estimate of movement speed⁵⁸ | <ul style="list-style-type: none"> Can be efficient for estimating abundance of common species (with a lot of images)⁵⁷ | <ul style="list-style-type: none"> Requires independent estimates of movement rate (difficult to attain without telemetry data)⁵⁷ Assumes that detection probability is 1 (or apply extension to account for imperfect detection)⁵⁷ | |
| Density; Unmarked | Space-to-event (STE) models ⁵⁷ | <ul style="list-style-type: none"> Demographic and geographic closure⁵⁷ Locations randomly placed^v Independent detections in space and time⁵⁷ Spatial counts of animals in a small area (or counts in equal subsets of the landscape) are Poisson-distributed⁵⁸ | <ul style="list-style-type: none"> Can be efficient for estimating abundance of common species (with a lot of images)⁵⁷ Does not require estimate of movement rate⁵⁷ | <ul style="list-style-type: none"> Assumes that detection probability is 1⁵⁷ | |

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|---|---|--|--|--|-------------------|
| Density; Unmarked | Instantaneous sampling (IS) ⁵⁷ | <ul style="list-style-type: none"> • Demographic and geographic closure⁵⁷ • Locations randomly placed⁵⁷ • Independent detections in space and time⁵⁷ | <ul style="list-style-type: none"> • Can be efficient for estimating abundance of common species (with a lot of images)⁵⁷ • Flexible assumption of animals' distribution⁵⁷ | <ul style="list-style-type: none"> • Requires accurate counts of animals⁵⁷ • Assumes that detection probability is 1⁵⁷ • Reduced precision⁵⁷ | |
| Behaviour (Diel activity patterns, mating, boldness, etc.) | | <ul style="list-style-type: none"> • Assumptions vary depending on the behavioural metric¹ • For studies of activity patterns and temporal interactions of species: activity level is the only factor determining detection rates; animals are active when camera detection rate reaches its maximum in daily cycle^{33,60} | <ul style="list-style-type: none"> • Can detect difficult to observe behaviours (i.e., boldness, or mating)⁵⁹ • Long-term data on behavioural changes that would be difficult to obtain otherwise (i.e., time-limited human observers, or costly GPS collars)⁵⁹ • Can monitor behaviour in response to specific locations (i.e., compost sites, which might be more difficult using GPS collars for example)⁶⁰ • Can evaluate interactions between species⁶⁰ | <ul style="list-style-type: none"> • Behavioural metrics may not reflect the behavioural state (inferred)⁶⁰ • Biases associated with equipment (i.e., presence of the camera itself may change behaviour studied)⁶⁰ • Difficult to consider individual variation⁶⁰ | |