

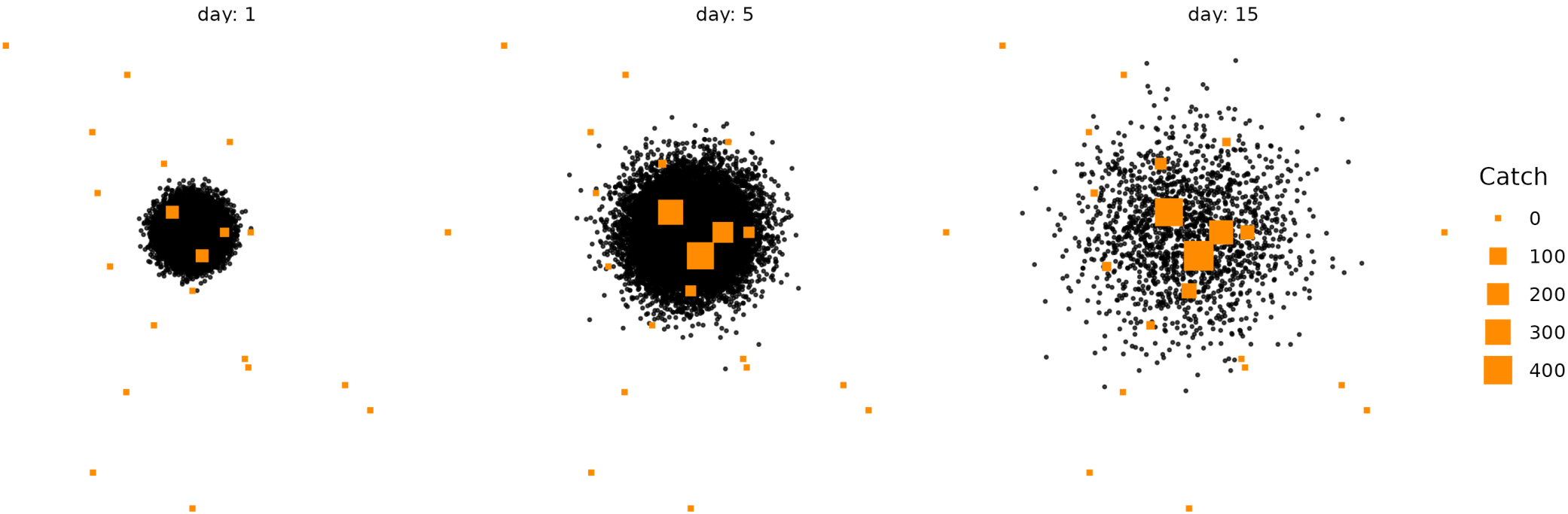
# Modelling dispersal, survival and trapping in SIT trials

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# Simulation of the process



## Speaker notes

- Thousands of insects (black dots) released from a single point
- They disperse independently from each other following a Brownian motion:
  - Changing direction continuously, with no preferential direction
  - Same model of fluid diffusion at thermal equilibrium
- A few traps (orange squares) scattered over the area catch a fraction of the nearby individuals.
- As time progresses, individuals either eventually get caught or die (naturally, or from other causes such as predators)

# Target quantities

Diffusivity

The spread rate of individuals

Survival rate

The fraction of individuals still alive at a given time

## Speaker notes

- Diffusivity: the *rate of diffusion*. I.e. The rate at which the particles spread.
- Survival rate: the fraction of individuals still alive at a given time

# Observations

Counts of individuals caught during trap activation periods

Example structure

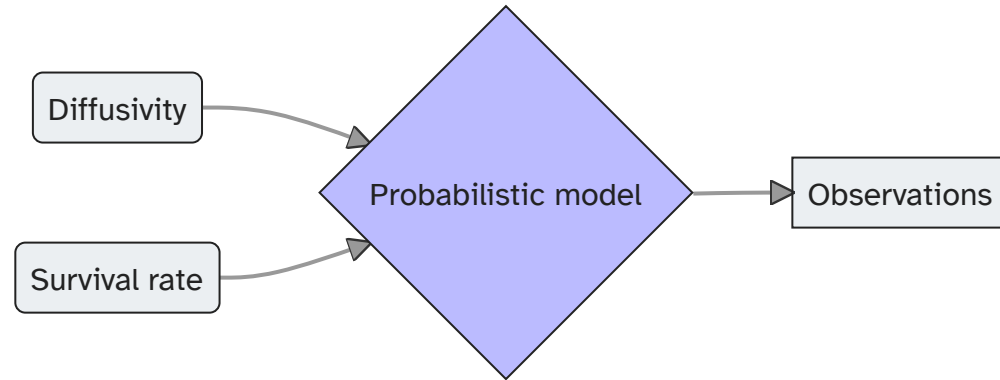
trap	ini_day	end_day	catch
1	0	7	31
2	1	8	15
1	10	15	12
...	...	...	...

(Fake data)

## Speaker notes

- Each trap is set up and remains active during a few days
- At the end of the period, operators collect and count their catch

# Goal





## Speaker notes

- Develop a probabilistic model for the observations, given the target quantities (forward)
- Use observations to conduct statistical inference on the parameters (backward)

We can only observe dispersal and survival

indirectly

through captures



## Speaker notes

Intuitively:

- Captures concentrated in the inner traps for a long time indicate small diffusivity and the converse
- Captures dropping quickly over time indicate low survival rate and the converse

# Dispersal

Brownian motion → Diffusion equation:

$\rho(\mathbf{x}, t)$ : population density at position  $\mathbf{x}$  and time  $t$ .

$$\frac{\partial \rho}{\partial t} = D \nabla^2 \rho$$

✓  $D$ : diffusivity coefficient

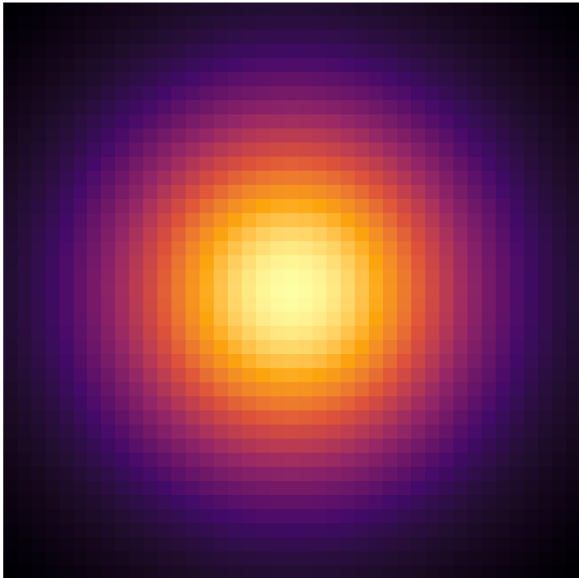
## Speaker notes

- Hypothesis: Brownian motion implies that the dispersion follows the diffusion equation, where
- $\nabla^2$ : Laplacian operator (divergence of gradient)
- Where we find one of the parameters of interest: the diffusivity coefficient

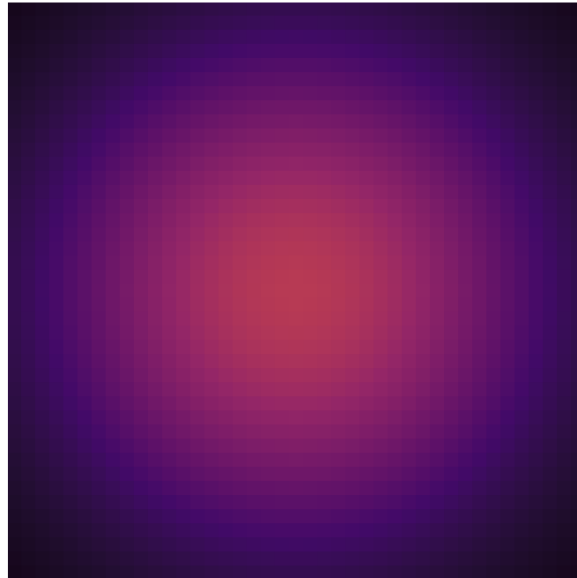
# Dispersal: Closed-form solution

$$X_t \mid t, D \sim \mathcal{N}(\mathbf{0}, 2tD\mathbf{I}_2)$$

day: 1



day: 2



day: 3

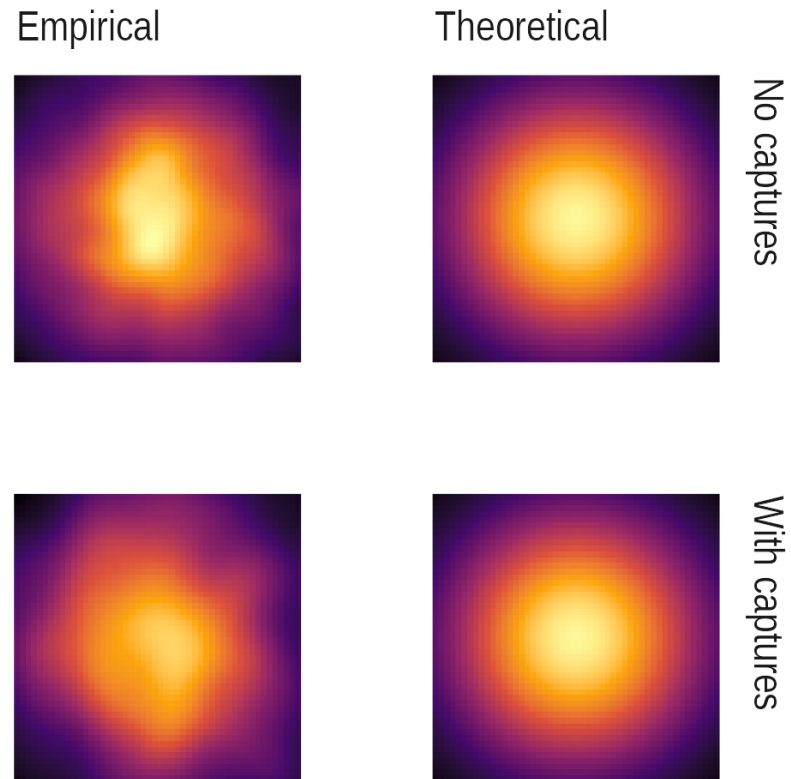


## Speaker notes

- This simple dispersal model has a very simple analytical expression in terms of a bivariate Gaussian distribution centred at the release point and with dispersion proportional to  $D$  and  $t$



# Effect of traps on density



Kernel density estimate of the empirical and theoretical distributions at day 15 with or without traps

## Speaker notes

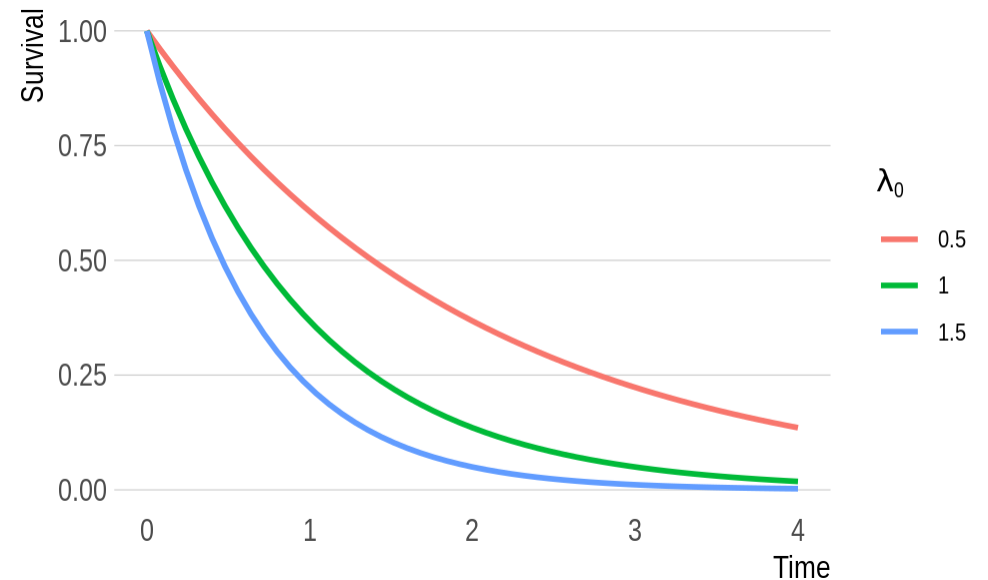
- The capture of insects in traps cause a slight reduction in the density near the origin, with respect to the theoretical density (which ignores catch)
- In practice this is typically irrelevant. Only a problem if there is a significant share of insects caught after many days of capture to make it noticeable (15, in the example)
- However, this is a source of bias in the estimate of the diffusivity coefficient (wich is proportional to the variance of this density)

# Survival

Let  $T_0$  be the time of death of individuals.

Assuming a constant hazard rate<sup>1</sup>  $\lambda_0$ ,

$$\begin{aligned} S_0(t) &= P(T_0 \leq t) \\ &= \exp\{-\lambda_0 t\} \end{aligned}$$



1. A common, albeit unrealistic simplifying hypothesis.

## Speaker notes

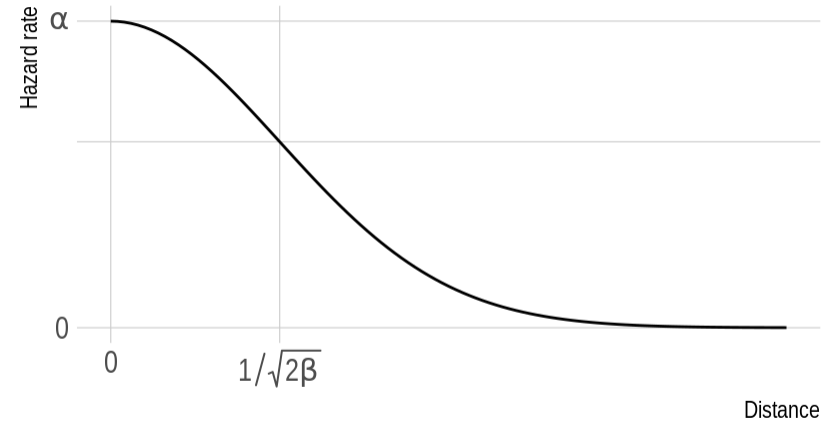
- Hazard rate: instantaneous probability of death, given survival to that time.
- A simple (and common, although maybe not realistic) assumption is that the hazard rate is constant: if an individual survived until  $t$ , the probability of death in the next instant is independent of  $t$
- Insects die, eventually, we don't know at which rate
- Furthermore, we can't **observe** their death directly
- When an individual gets caught, we know that they have survived until then

# Capture

Assumption: the hazard rate of capture decreases exponentially with the squared distance to the trap

$$h_i(t \mid X_t) = \alpha e^{-\beta \|X_t - x_i\|^2}$$

$$\alpha, \beta > 0$$



## Speaker notes

- We need to model the capture process, even though its parameters are not our target quantities
- For trap  $i$  at  $x_i$ , conditional to position  $X_t$ .

# Notation

Random variables associated to a released individual

End Time

$$T > 0$$

Time at which an individual ceases its activity

End Cause

$$C \in \{0, 1, \dots, I\}$$

Either death ( $C = 0$ ) or capture in trap  $i$  ( $C = i$ )

## Speaker notes

- Use concepts and tools from time-to-event (survival) models with competing risks



# Specific hazard rates

$$h_i(t) = \mathbb{E}_{X_t} \left[ \alpha e^{-\beta \|X_t - x_i\|^2} \right], \quad \alpha, \beta > 0$$

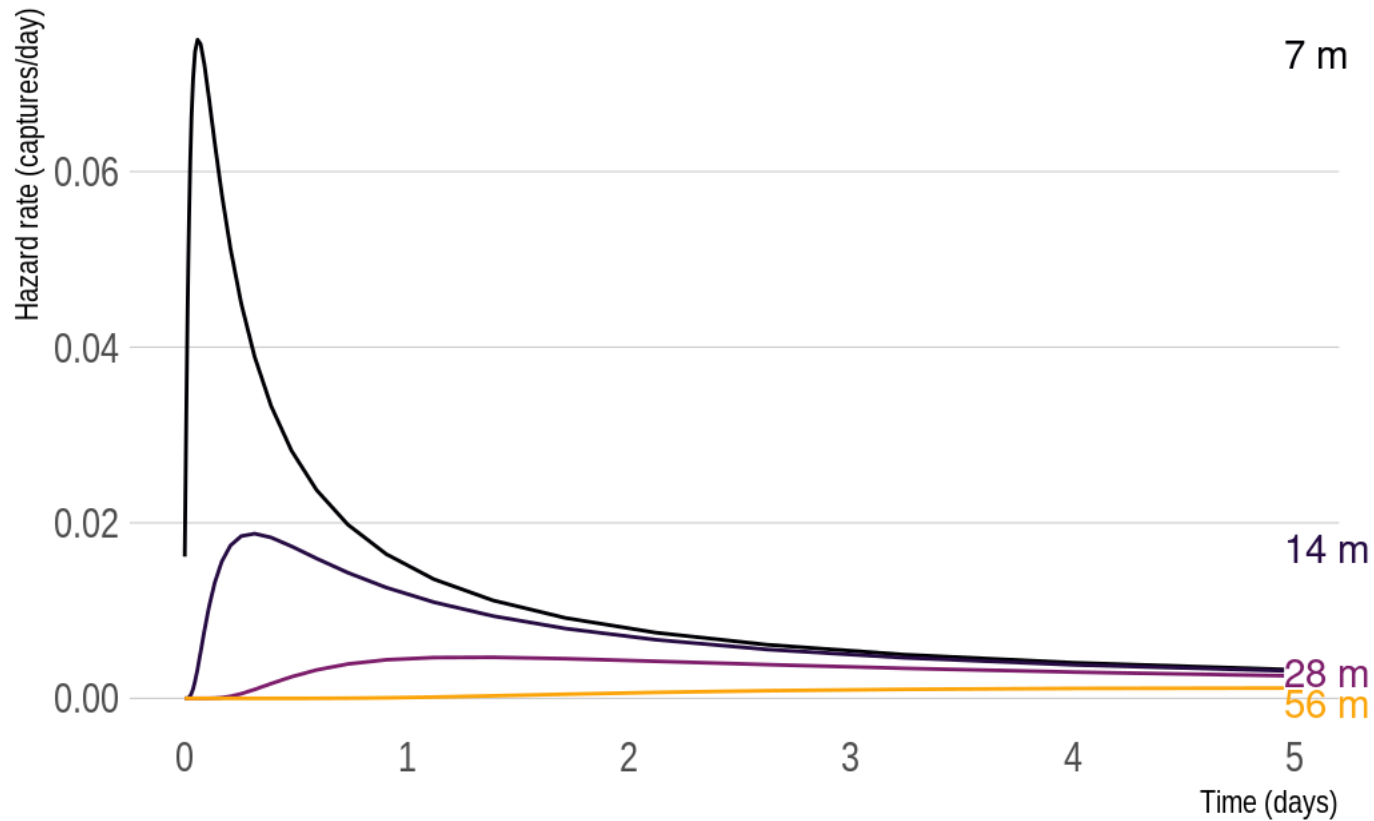
✓ Result

$$h_i(t) = \frac{\alpha}{1 + 4\beta t D} e^{\frac{-\beta \|x_i\|^2}{1 + 4\beta t D}}$$

## Speaker notes

- We don't know the position of every individual at time  $t$ , so we consider the expectation over its (known) probability density
- The calculation yields a closed-form expression for the specific hazard rates

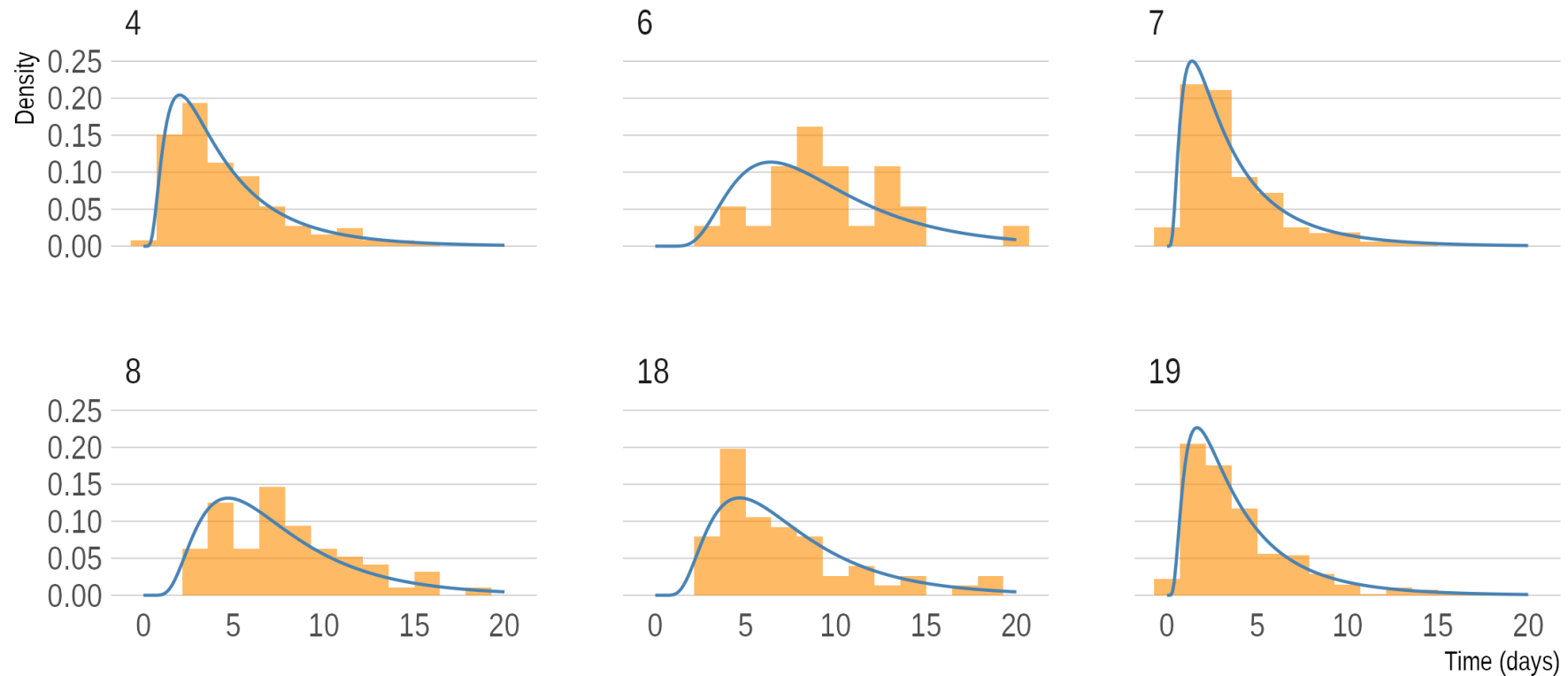
# Hazard functions of the capture times for traps at increasing distances from the release location



(Dispersion and capture parameters as in the simulation)

# Capture times

## Theoretical vs empirical distributions of capture times



(Traps with 10+ total catch)

# Mixed joint distribution

At any given time,

1. Additive hazard rates
2. Relative risks of death or capture are proportional to the specific hazard rates




✓ Result

$$\begin{aligned}\mathbb{P}(C = i, T = t) &= \mathbb{P}(C = i \mid T = t) f_T(t) \\ &= h_i(t) S(t)\end{aligned}$$

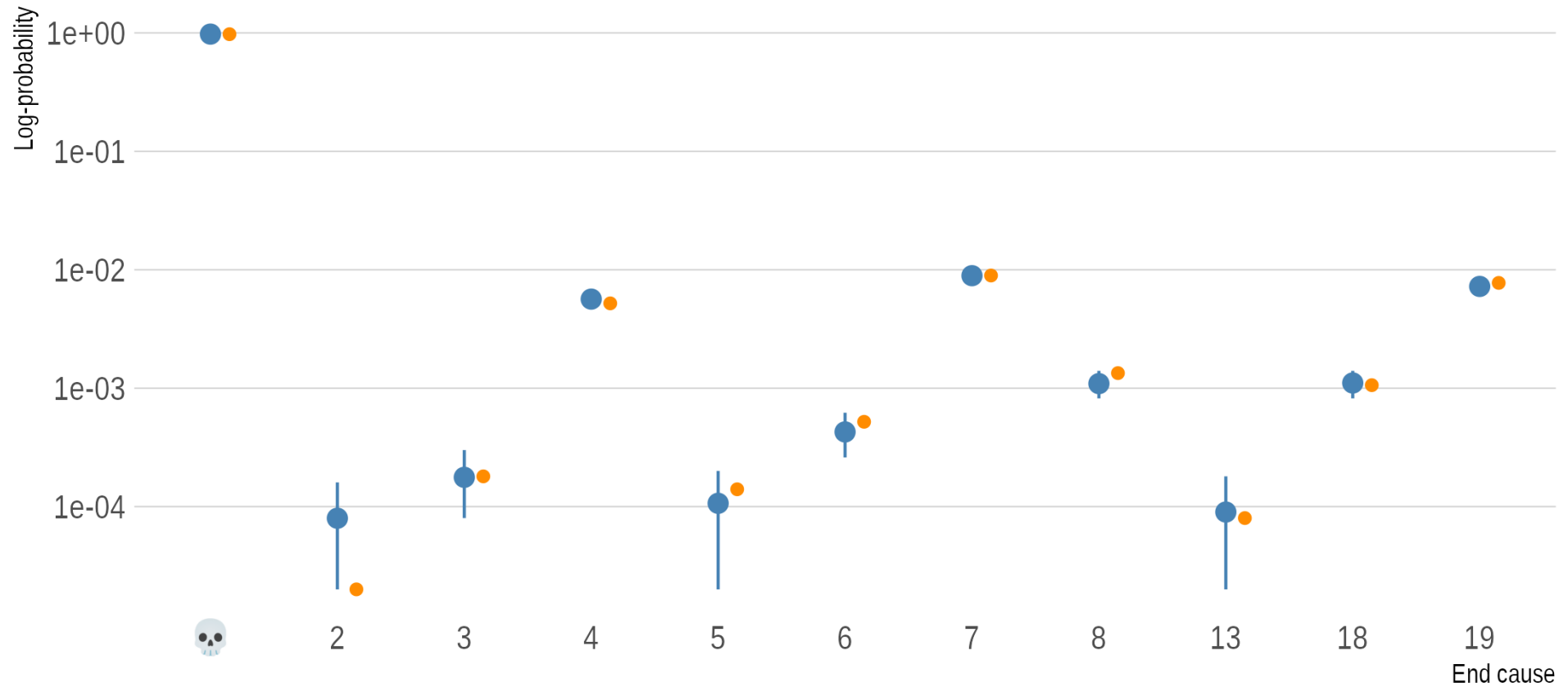
where  $S(t) = \exp \left( - \int_0^t \sum_{i=0}^I h_i(u) du \right)$

## Speaker notes

Borrowing ideas from survival models with competing risks,

- At any given time, the hazard rates are additive
- Enabling the (numerical) computation of the overall survival (i.e. fraction of individuals that are still active at time  $t$ )
-  We can integrate the distribution over the observation intervals for each trap to evaluate the likelihood!! 
-  Sadly, we lose the analytical expressions and need to evaluate the integrals numerically
- The exponential component multiplying the specific hazard rate is the overall survival function  $S(t)$

# Distribution of end causes



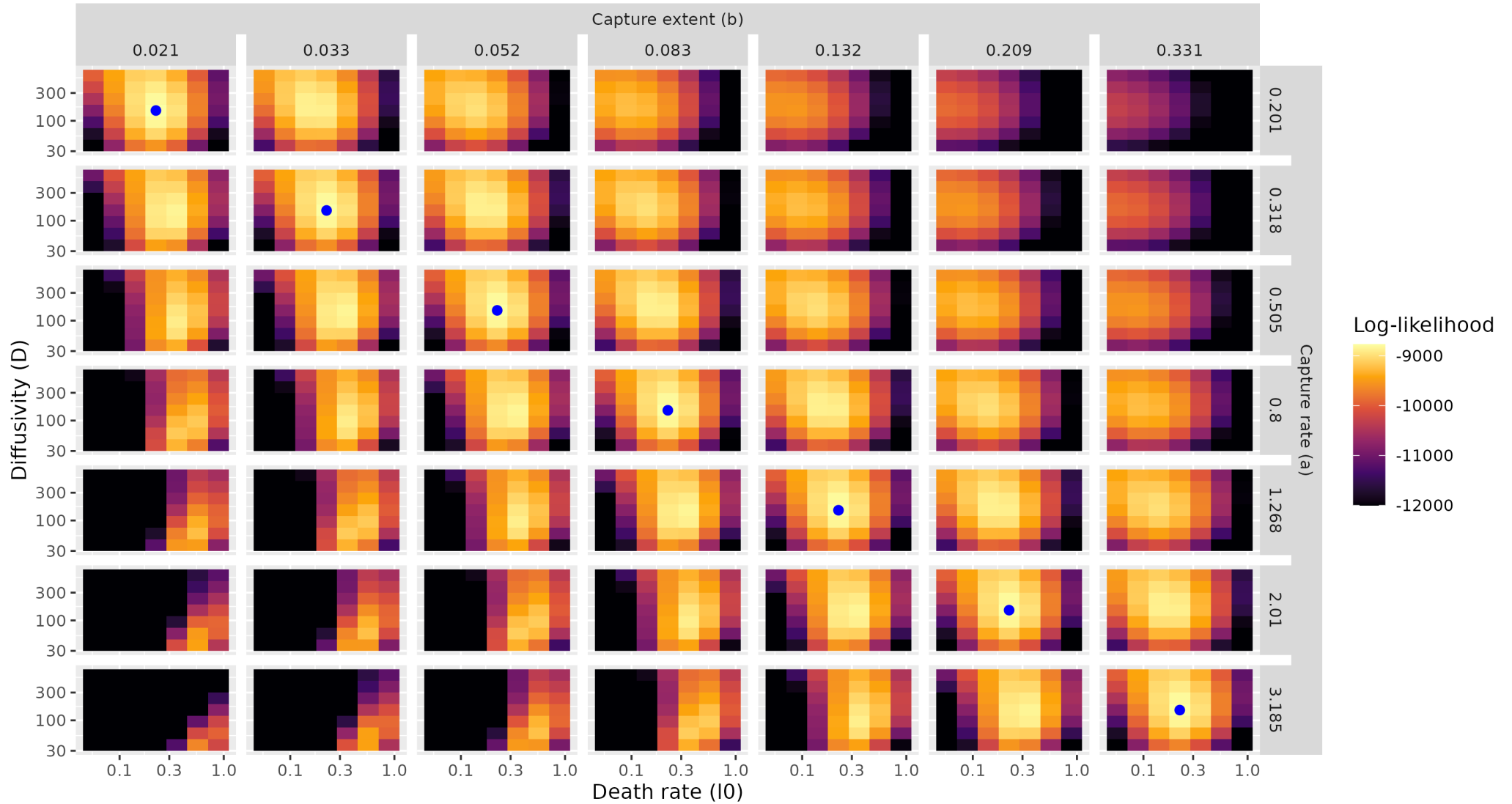
Predicted vs. **observed** proportion of end events by cause (death / trap)

## Speaker notes

- We evaluated the expected and 95% quantile interval for the proportion of dead and caught individuals in each of the traps over all the simulation period and compared with the realised frequency (only traps with 1+ captures shown)



# Full likelihood



## Speaker notes

- We borrowed ideas from survival models with competing risks to combine the specific hazard rates and compute the joint distribution of end times and causes
- We integrate numerically over the observation periods to obtain the likelihood function
- The figure shows the likelihood function of the simulated scenario.
- The blue points indicate the maximum likelihood (which is almost flat over the diagonal)
- The true value is in the center of the central panel.

# SIT: sterile insect technique

Sequence of releases of sterile insects for pest control

- Released males mate with wild females producing sterile eggs
- ✓ Avoid pesticides (which contaminate the environment)

[Introductory video presentation](#)

## Speaker notes

- Control disease transmission and reduce agricultural damage without insecticides
- Sterile males mate with wild females reducing offspring
- Need to quantify parameters such as number of individuals to release, temporal frequency, spatial extent, etc. to produce the desired effect of reduction or suppression of the wild population.
- These parameters depend, among other things on the **diffusivity** and the **survival** behaviour of the released sterile males in the context of the specific area.

# Issues

- Bias in dispersion (and diffusivity)
- Efficiency in numerical calculations (time consuming)
- Identifiability of capture parameters  $\alpha$  and  $\beta$   
(reparameterise!)

# Extensions

- Non-isotropy
- Varying *attractiveness* of traps
- Integrate further `sit` parameters into the model
- Optimal designs

## Speaker notes

- Further parameters such as egg hatching rates, competitiveness...


# Conclusions

- ✓ A formal probabilistic model for the dispersion, survival and capture of insects in `sit` experiments
- ✓ Enable inference with appropriate quantification of uncertainty in the target parameters

[WIP]



# Thank you

-  Press 's' to access speaker notes
- Slides available at <https://astre.gitlab.cirad.fr/presentations/202408-modah-modelling-sit/>
- An extended (23) version of the slides are at <https://astre.gitlab.cirad.fr/presentations/202408-modah-modelling-sit/extended.html>

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