

Occurrence distributions

Using the 'ctmm' R package

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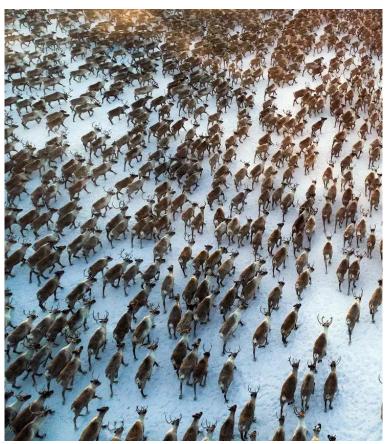












Billions of animals migrate worldwide to exploit **seasonal resources**, **escape severe weather**, **breed**, or to **avoid predation**.



Which areas of a landscape contain high-priority resources (e.g., migratory corridors/stopovers)?

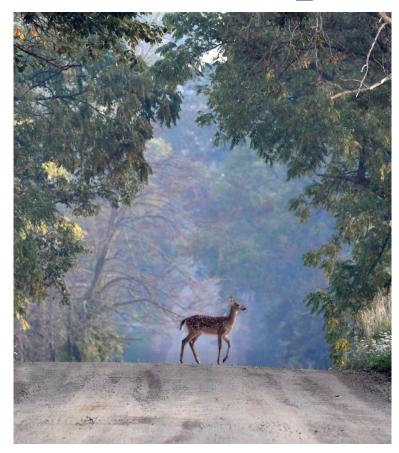


Nick Upton

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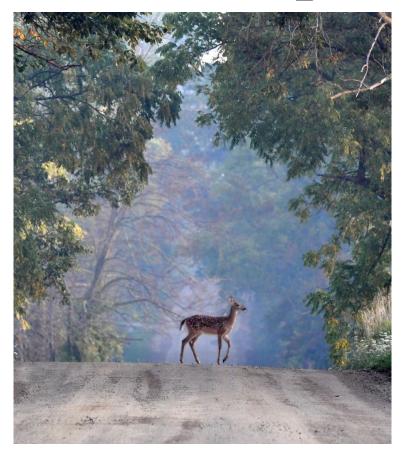
Which areas of a landscape contain high-priority resources (e.g., migratory corridors/stopovers)?

How likely is it to visit a location of interest (e.g., wildlife crossing sites, wind farms)?

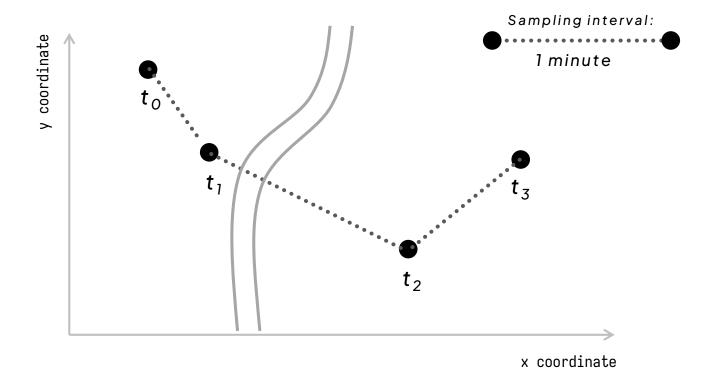


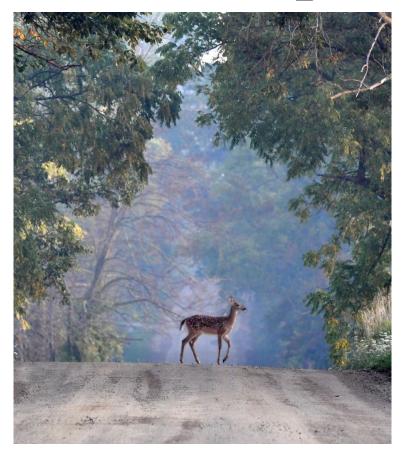
Roads have significant impacts on ecological systems, increasing the *extinction risk* of threatened species.

Where did an animal cross a linear feature (e.g., road- or railways)?

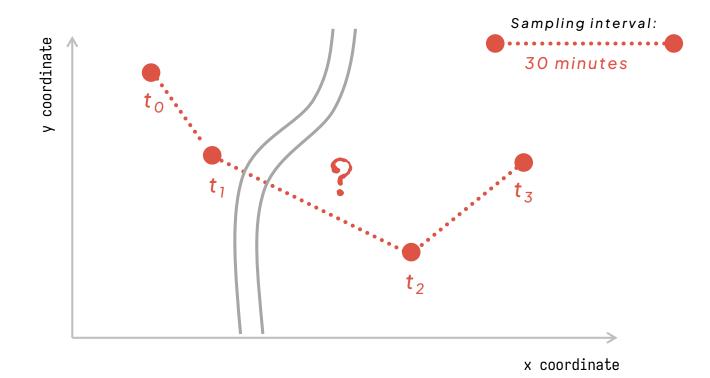


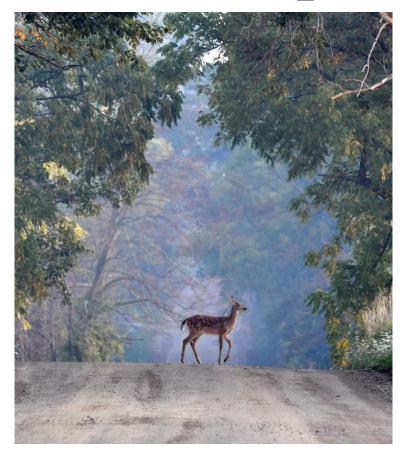
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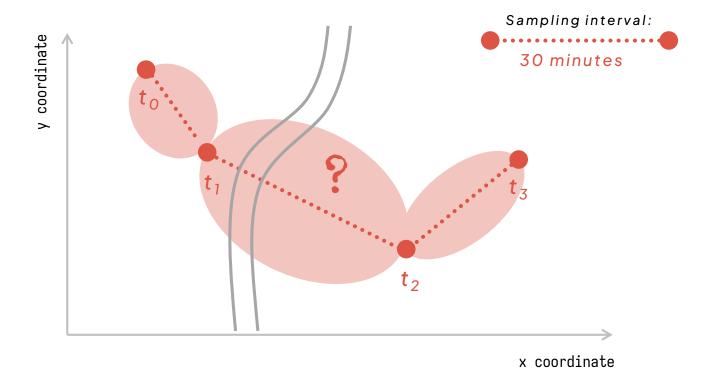


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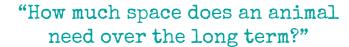
Utilization distribution

represents an animal's distribution and the probability of use throughout an area



Range distribution

extrapolate space use into the future



What is an animal's home range area? What is their the population range area? Are protected areas sufficiently large?



Occurrence distribution

interpolate between data points in the past

"Where did an animal go during a period of observation?"

Where did an animal cross a linear feature? How likely is it to visit a location of interest? How much time did it spend in a specific habitat?

 \mathcal{P} Alston et al. (2022)

Utilization distribution

represents an animal's distribution and the probability of use throughout an area

Range distribution extrapolate space use into the future Points in the past

Estimates home ranges

Considers all realizations of the movement process that could occur.

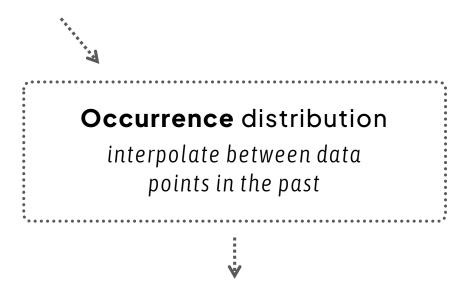
Estimates occurrence regions

Pinpoints the one realization of the movement process that did occur.

 \mathcal{L} Alston et al. (2022)

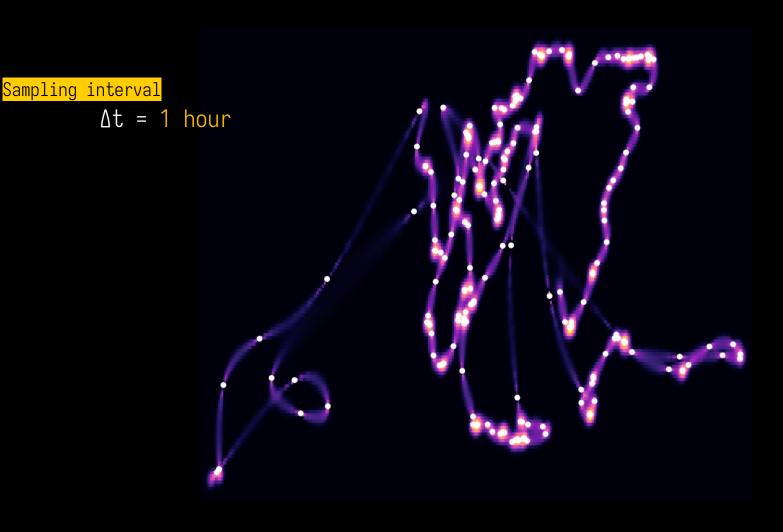
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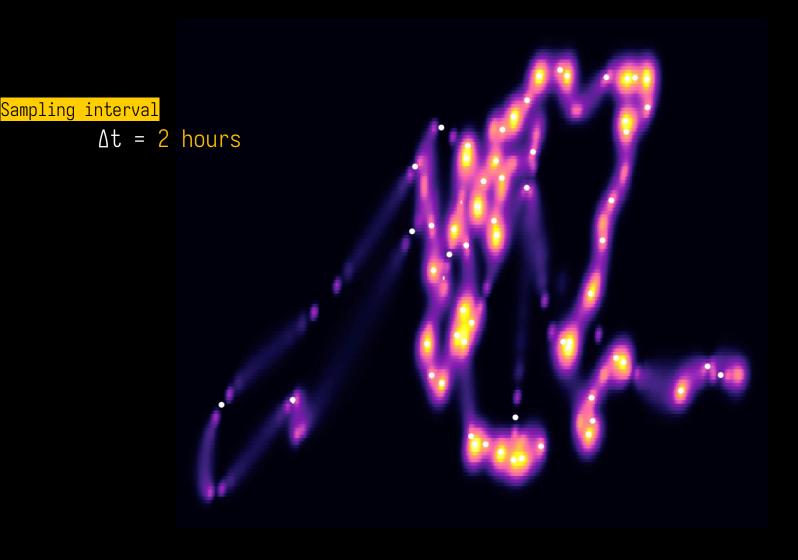
Pinpoints the one realization of the movement process that did occur.



0.75

0.50

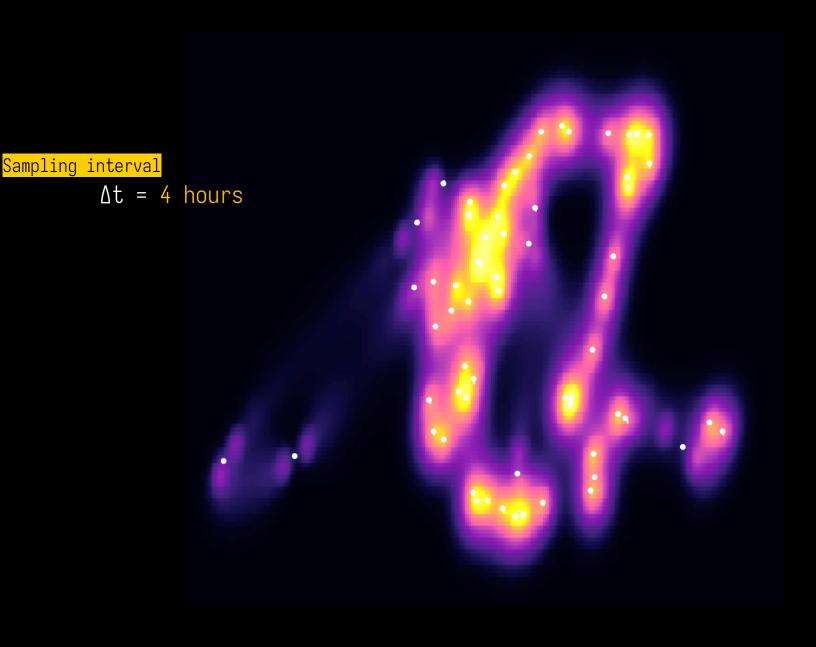
0.25



0.75

0.50

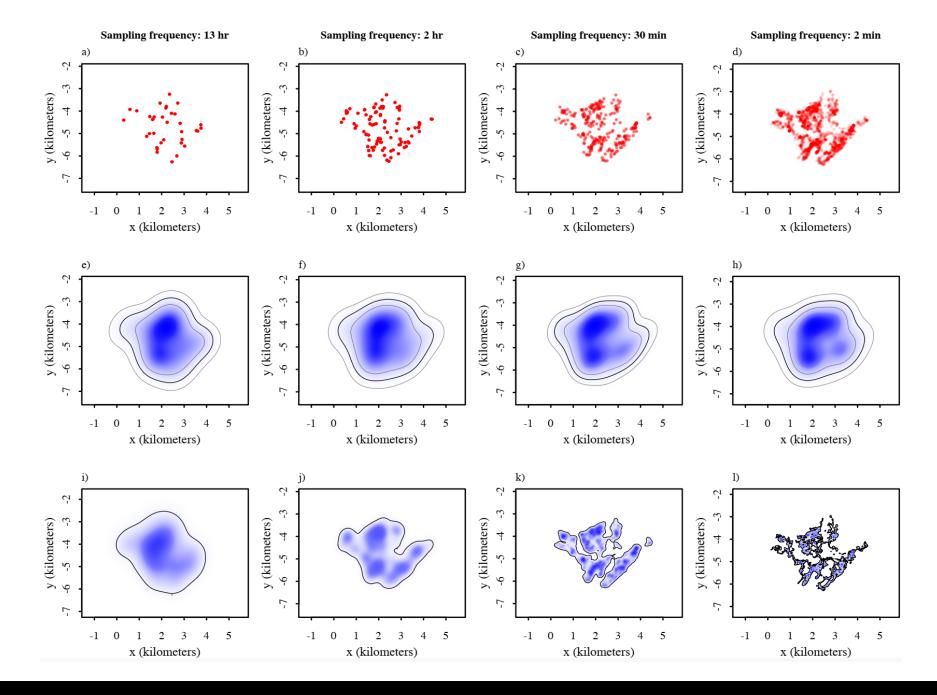
0.25

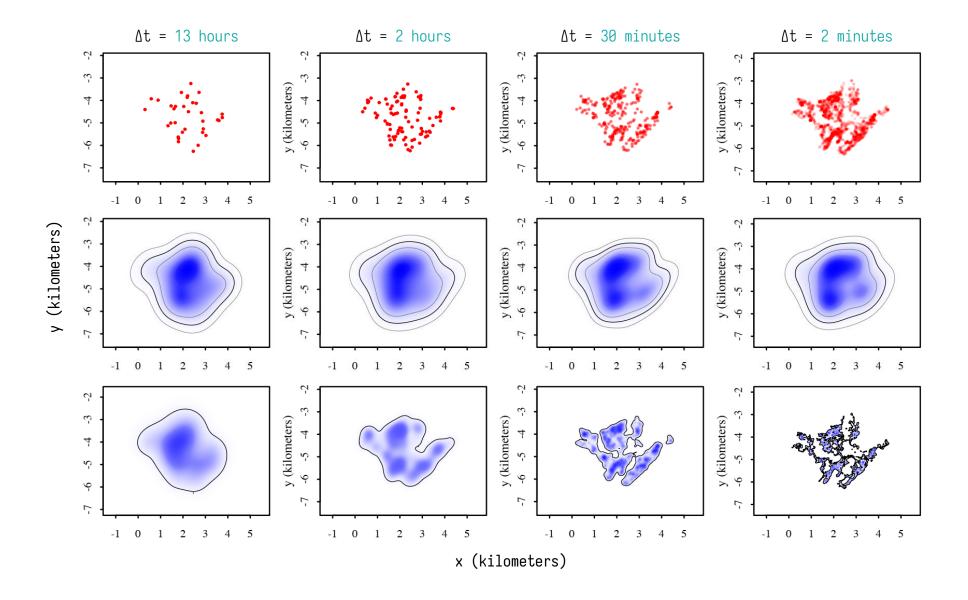


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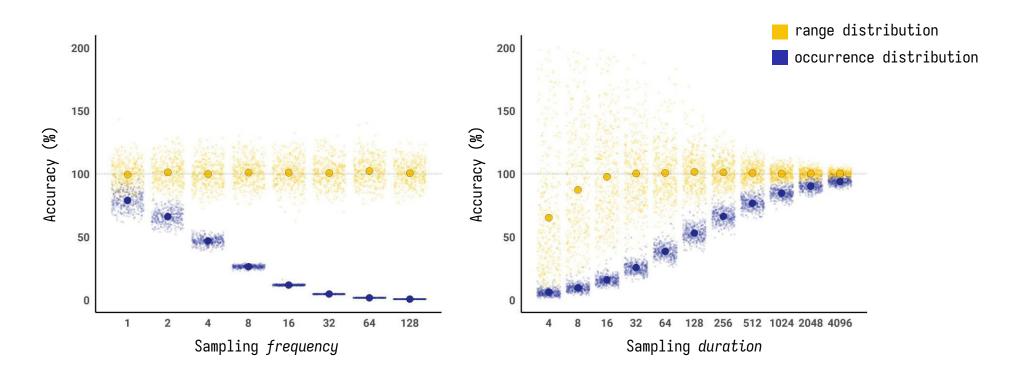


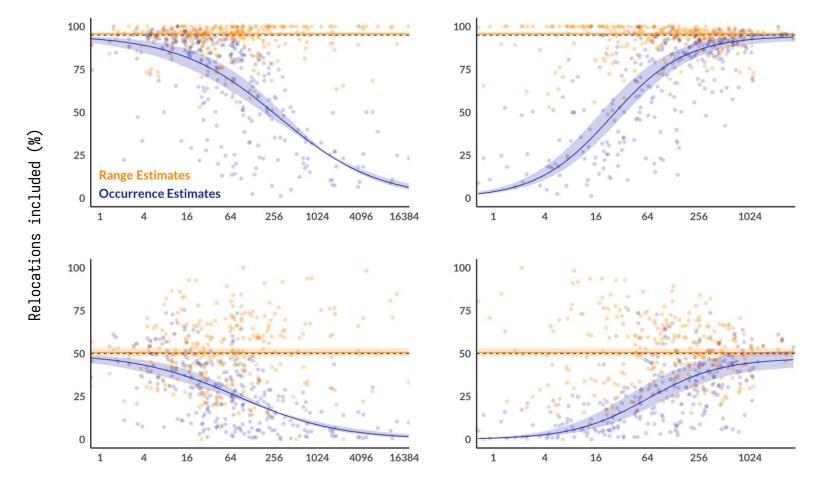




Simulated examples

Size of occurrence estimates are a function of sampling interval, while size of range estimates are not.





Empirical examples
Holds up in real world data!

Effective sample size (N)

estimating occurrence region

Generalized time-series **Kriging** framework:

Example 1 Kriging is a widely used tool in geostatistics, engineering, and computer science. It is a statistically optimal framework for interpolating between discrete locations, with well-known statistical properties.

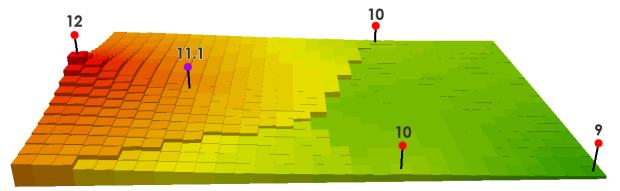
estimating occurrence region

Generalized time-series **Kriging** framework:

Fleming et al. (2016)

Kriging is a widely used tool in geostatistics, engineering, and computer science. It is a statistically optimal framework for **interpolating between discrete locations**, with well-known statistical properties.

Fig. spatially interpolated temperature from weather stations.



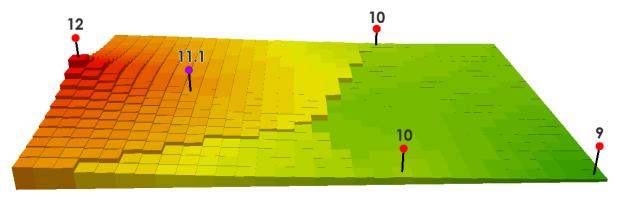
Estimating occurrence region

Generalized time-series **Kriging** framework:

 \mathcal{L} Fleming et al. (2016)

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Fig. spatially interpolated temperature from weather stations.



Encompasses both Brownian Bridge Movement Model (BBMM; Horne et al. 2007) and Correlated Random Walk library (CRAWL; Johnson et al. 2008).

Generalized time-series **Kriging** framework:

Q Fleming *et al.* (2016)

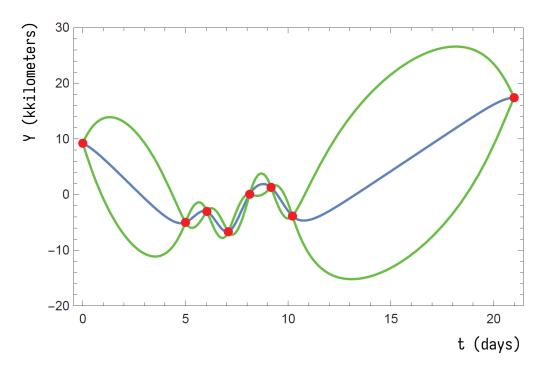


Fig. Time series of Mongolia Gazelle locations (red), with Krige interpolated locations (blue) and 95% contour intervals (green).

Not only does *Kriging* provide an optimal prediction surface, but it also delivers a measure of confidence of how likely that prediction will be true.

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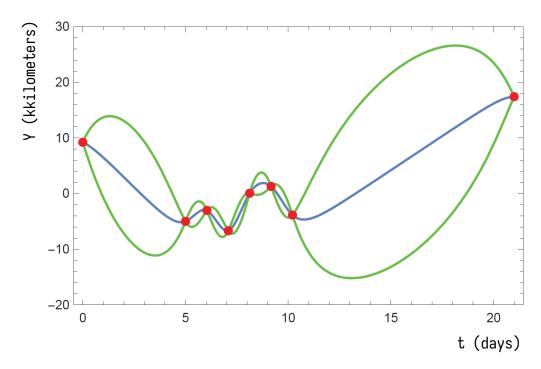


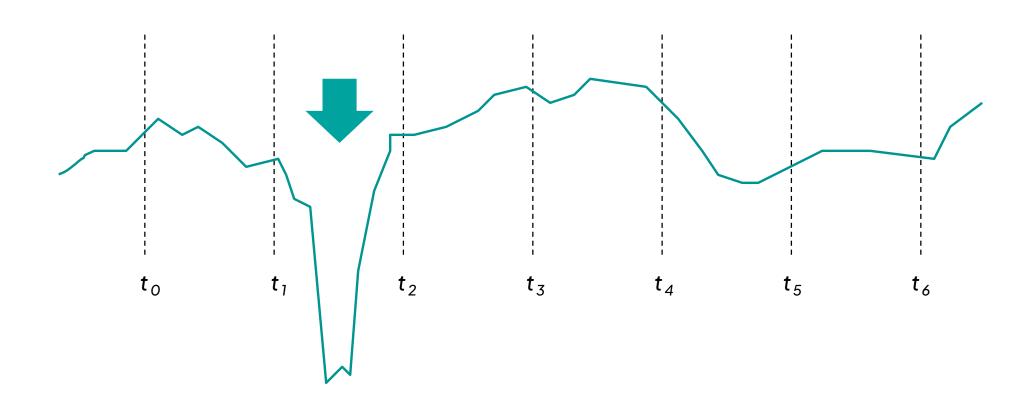
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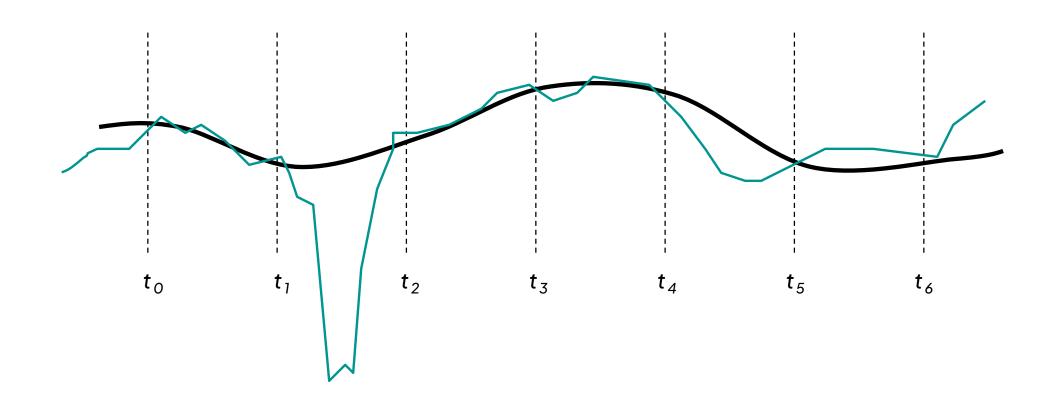
Two step process:

- 1. Select a movement model, (describing an animal's movements)
- 2. Solve for an animal's location at time t, (conditional upon the data and the fitted model)

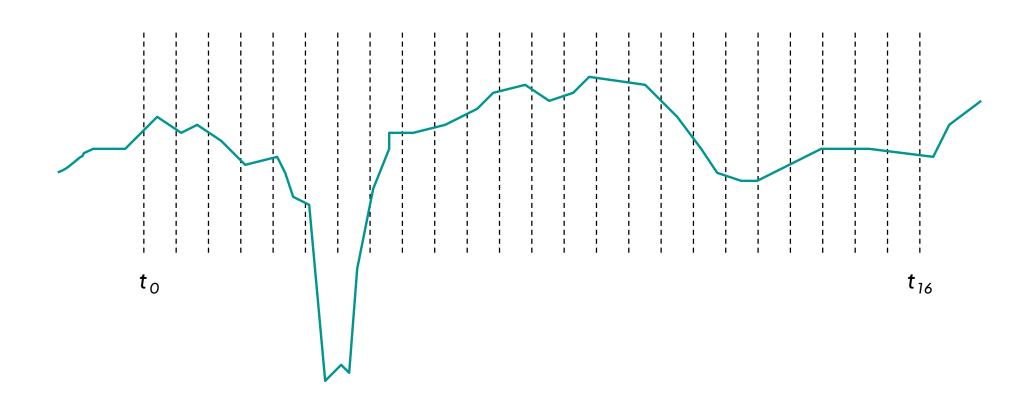
If each dashed line represented a sample point (in 1-D), this spacing would miss a major local source of variation.



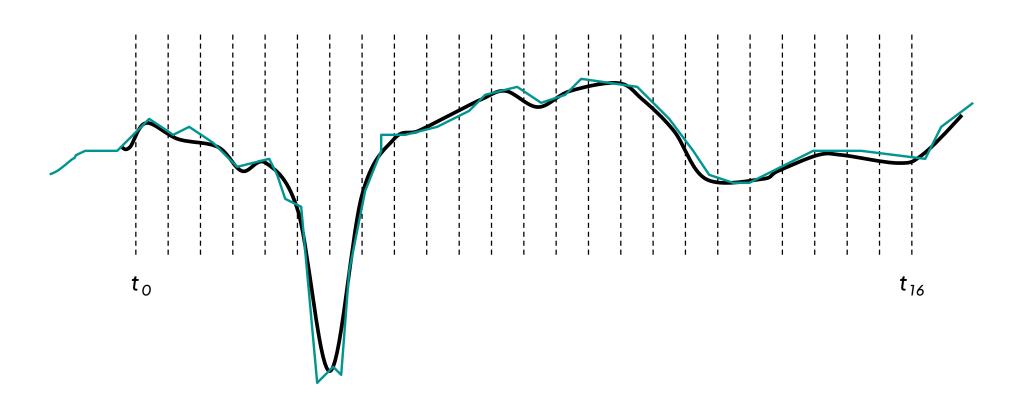
Our interpolated surface (represented in 1-D by the black line) would look like this.



Only by increasing the sampling frequency, would we pick up that local variation.



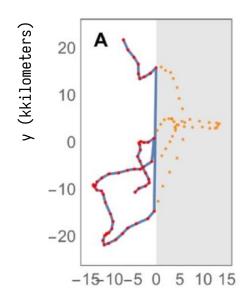
Here our interpolated surface is much closer to reality at the local level, but we pay for this in the form of higher data gathering cost.

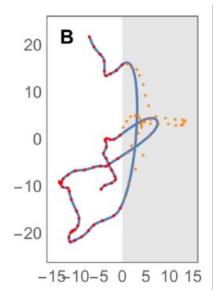




Now that we know how to interpolate, we can:

- ► Fill in missing gaps,
- Project beyond the sampling period,
- ► Identify areas of potential use, (e.g., related to human-wildlife conflict).



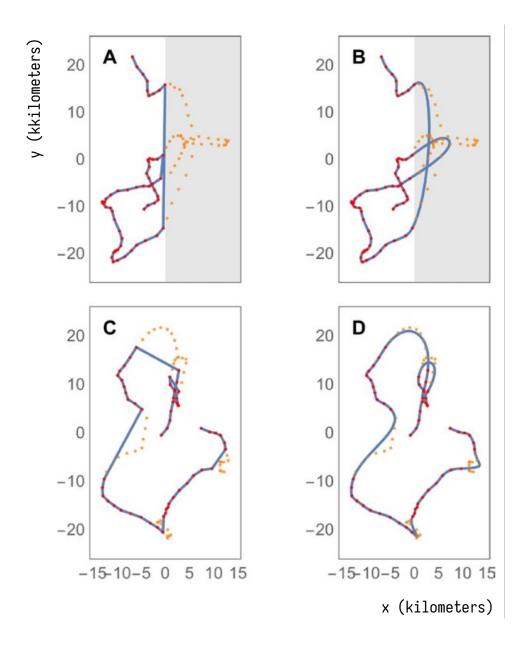


x (kilometers)



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Applications

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Can be very **memory-intensive**! (on large data sets, consider looping through each animal individually).

