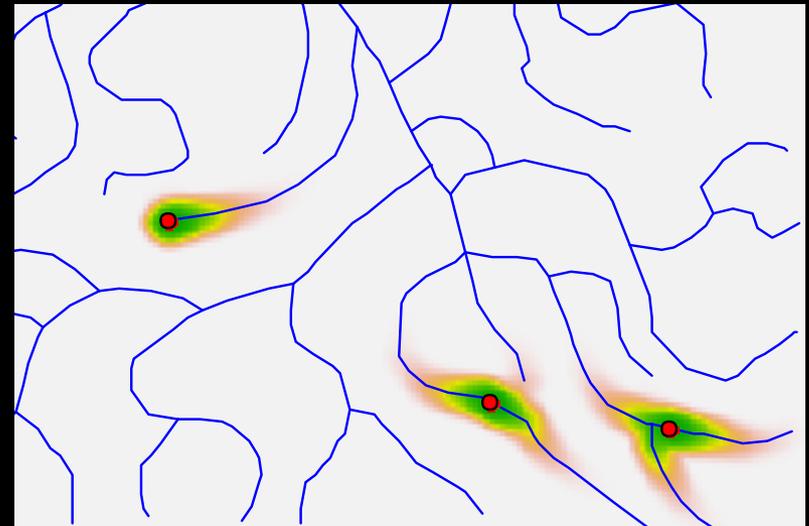


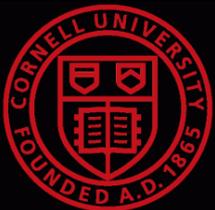
# Measuring 'ecological' distance in spatial capture-recapture models

*Chris Sutherland, Angela Fuller & Andy Royle*

American mink in riparian corridors



*The views expressed in these slides are that  
of the author only.*



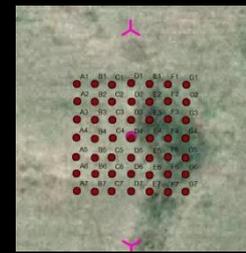
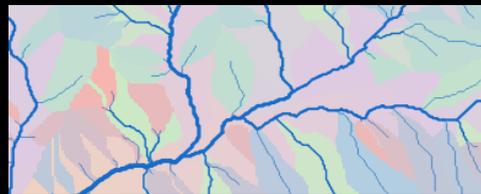
# Spatial Capture-Recapture (SCR)

Acknowledges the inherent spatially nature of ecological processes and observation processes:

- effective area sampled  $\sim$  *absolute* density
- heterogeneous encounter probabilities

but, useful for many other challenges in ecological research:

- movement, space-use, resource selection, survival, recruitment



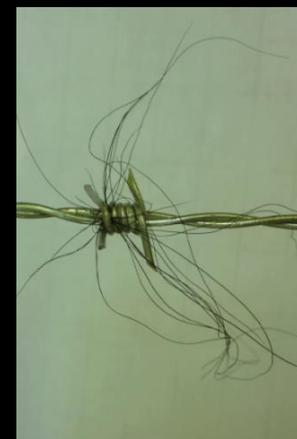
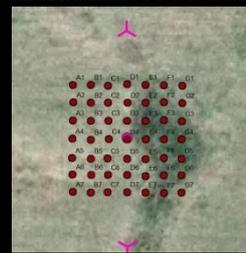
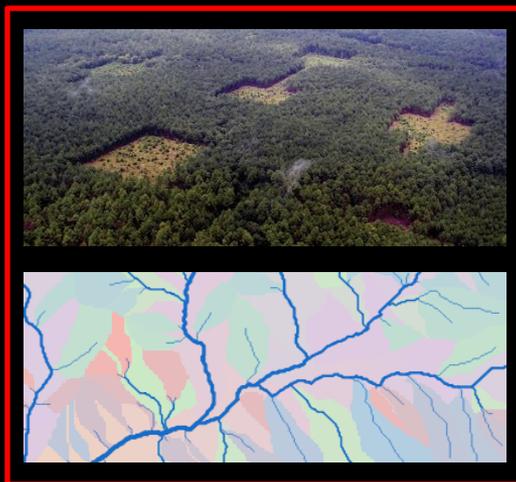
# Spatial Capture-Recapture (SCR)

Acknowledges the inherent spatially nature of ecological processes and observation processes:

- effective area sampled  $\sim$  *absolute* density
- heterogeneous encounter probabilities

but, useful for many other challenges in ecological research:

- movement, space-use, resource selection, survival, recruitment, *connectivity/landscape resistance* (Royle et al., 2013)



# Spatial Capture-Recapture (SCR)

Retaining the spatial information about trap locations and therefore about individual encounter locations:

Non-spatial encounter history

$$y_{i,k} = [y_{i,1,k} \ y_{i,2,k} \ \dots \ y_{i,J,k}]$$

Y =

		visit (k)		
			1	1
		1	0	1
		0	0	1
		0	0	0
		1	0	1
		0	1	
		1		
individual (i)	↓			

Caught in any trap

Spatial encounter history

$$y_{i,k} = [y_{i,1,k} \ y_{i,2,k} \ \dots \ y_{i,J,k}]$$

Y =

		visit (k)															
			1	1	1	0	0	0	0	0	0	0	0	0	0	0	
			1	1	1	0	0	0	0	0	0	0	0	0	0	0	0
		1	0	1	0	0	0	0	0	0	0	0	0	0	0	0	1
		0	0	0	0	0	0	0	0	0	1	1	0	1	0	1	0
		0	0	0	0	1	1	1	0	1	0	0	0	0	0	0	0
		0	1	1	0	1	0	1	0	0	0	0	0	0	0	0	1
		0	0	0	0	1	1	0	0	0	0	0	0	0	0	0	0
		1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
individual (i)	↓																
		trap (j)															

# Spatial Capture-Recapture (SCR)

A typical SCR data set therefore consists of:

- spatial locations of each trap -  $\mathbf{x}_j$
- individual-by-trap/location spatial encounter histories –  $\mathbf{y}_{i,j}$

Biologically: Observations  $\mathbf{y}_{i,j}$  occur as a result of movement around an home range center and frequency of detection decreases with distance between trap and activity center  $\mathbf{s}_i$

Statistically: Observations  $\mathbf{y}_{i,j}$  are realizations of a probability distribution whose mean is a latent variable  $\mathbf{s}_i$  i.e. a random-effects model

Model for encounter probability (many exist):

$$y_{i,j} \sim \text{Binomial}(K, p_{i,j})$$

$$p_{i,j} = f(d[\mathbf{s}_i, \mathbf{x}_j])$$

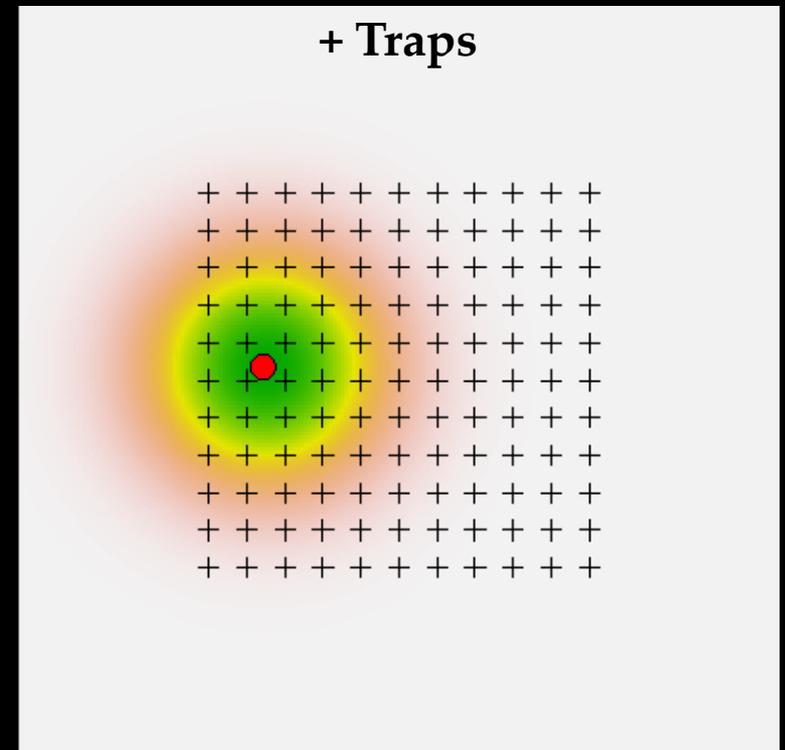
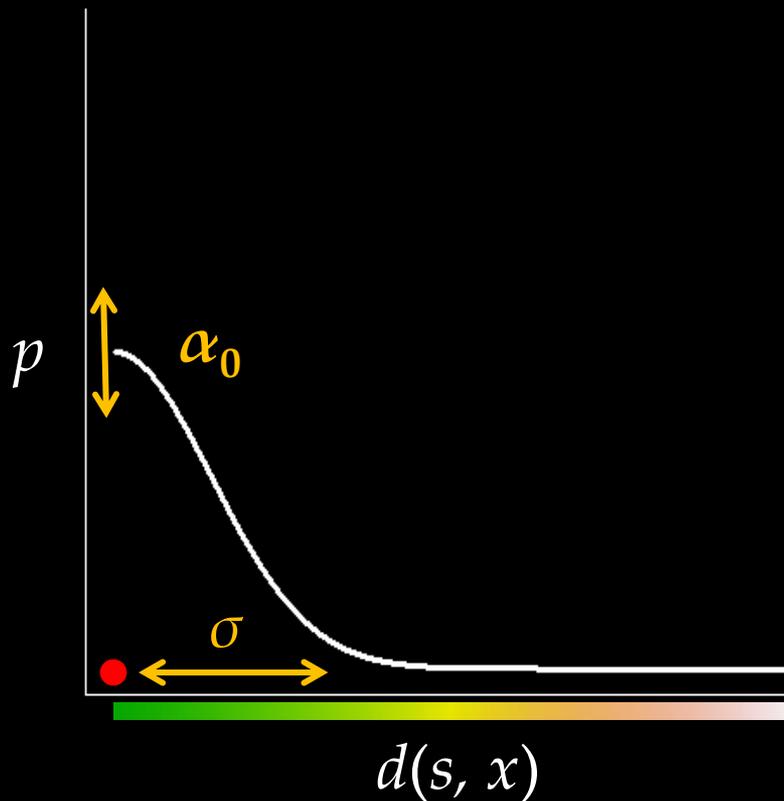
# Modeling movement using SCR

Bivariate normal encounter model:

$$y_{i,j} \sim \text{Binomial}(K, p_{i,j})$$

$$p_{i,j} = \alpha_0 \exp(-\alpha_1 d[s_i, x_j]^2)$$

$$\alpha_1 = 1/(2\sigma^2)$$



# Modeling movement using SCR [*poorly in some cases?*]

Bivariate normal encounter model:

$$y_{i,j} \sim \text{Binomial}(K, p_{i,j})$$

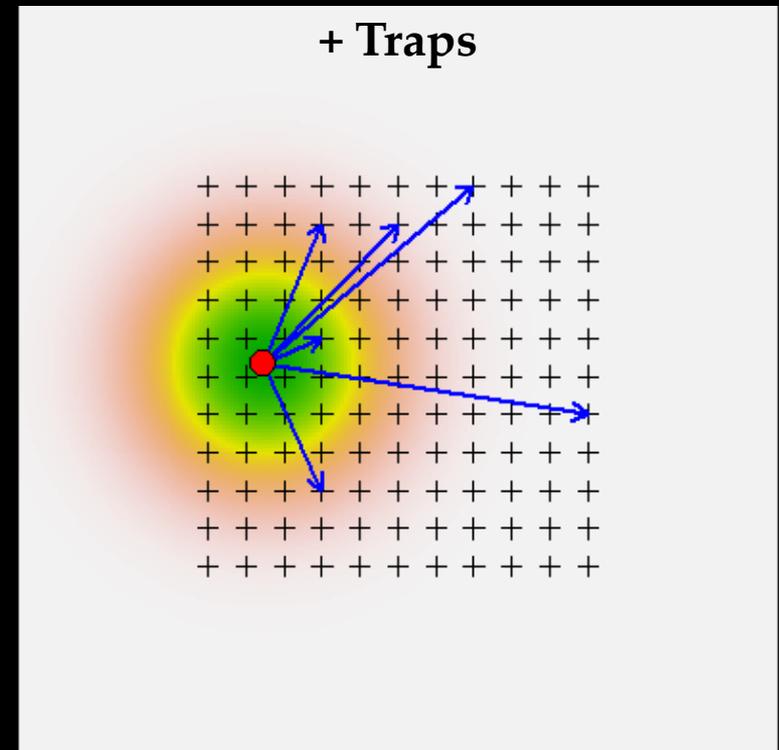
$$p_{i,j} = \alpha_0 \exp(-\alpha_1 d[s_i, x_j]^2)$$

$$\alpha_1 = 1/(2\sigma^2)$$

Euclidean distance:

- circular home range
- uniform landscape [use]

⇒ Biologically unrealistic



# Modeling movement using SCR [*poorly in some cases?*]

Bivariate normal encounter model:

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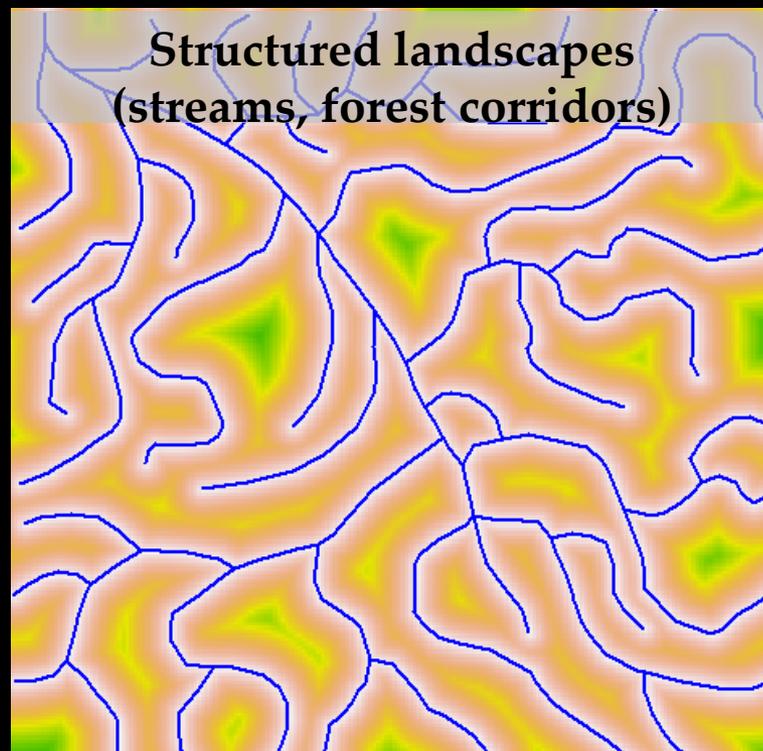
Euclidean distance:

- circular home range
- uniform landscape [use]

⇒ **Biologically unrealistic**

Movement is not equally likely  
through all habitats:

- cost/resistance



# Stream networks & riparian species – my motivation!

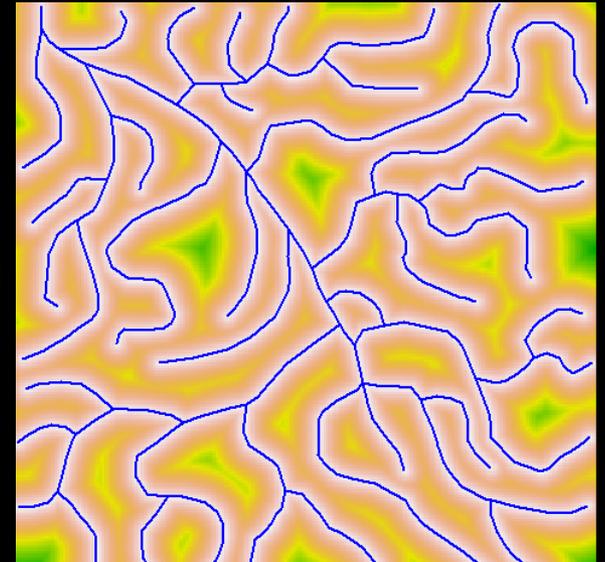
The American mink *Neovison vison*:

- economically important
- community regulator
- top predator (aquatic)
- indicators of ecosystem health



A perfect model system:

- semi-aquatic (∴ non-Euclid. movement)
- habitat specialist (∴ non-Euclid. movement)
- landscape (dist. to water) 'easy' to define

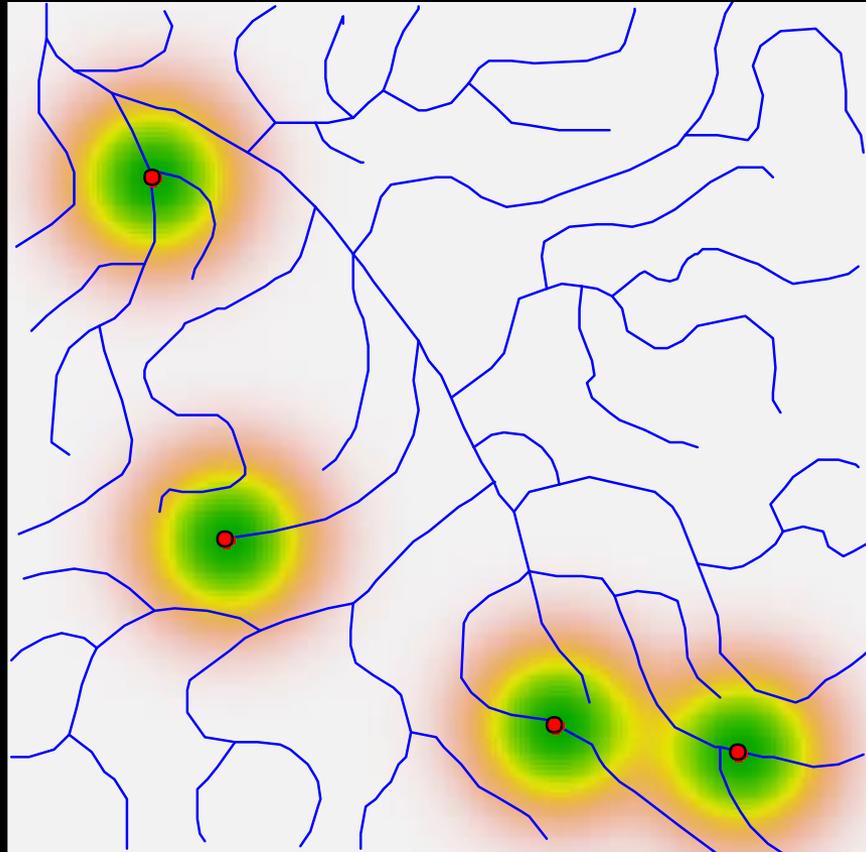


So, how should we measure distance?

# Distance as the *what?* moves...

Three ways to measure distance:

- Euclidean distance - distance as the 'crow flies'

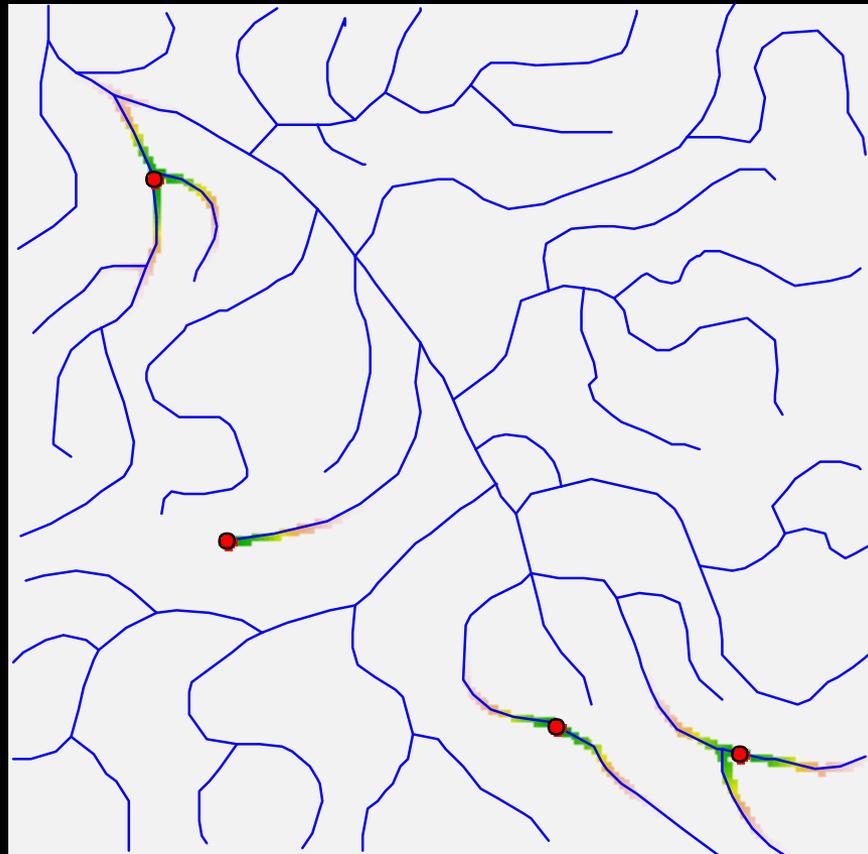
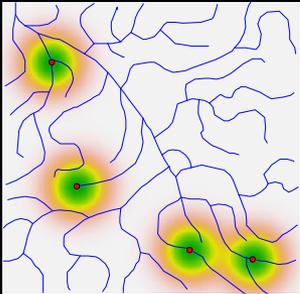


# Distance as the *what?* moves...

Three ways to measure distance:

- River/stream distance - distance as the 'fish swims'

Distance as the  
'crow flies'

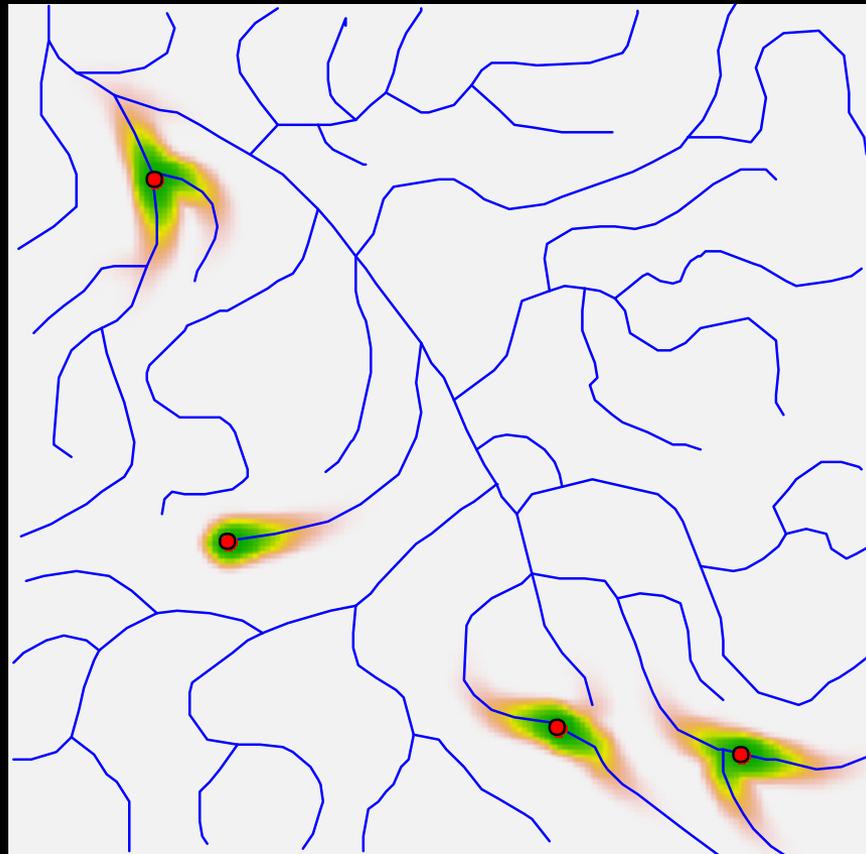
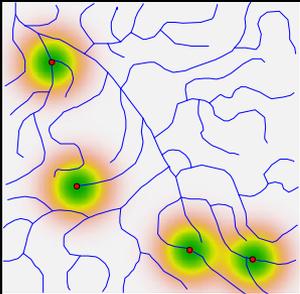


# Distance as the *what?* moves...

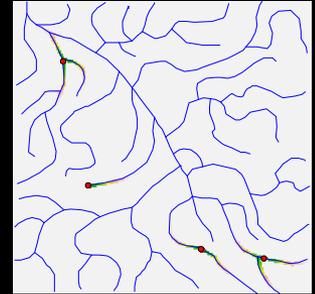
Three ways to measure distance:

- 'Cost weighted' distance - distance as the 'mink moves'

Distance as the  
'crow flies'

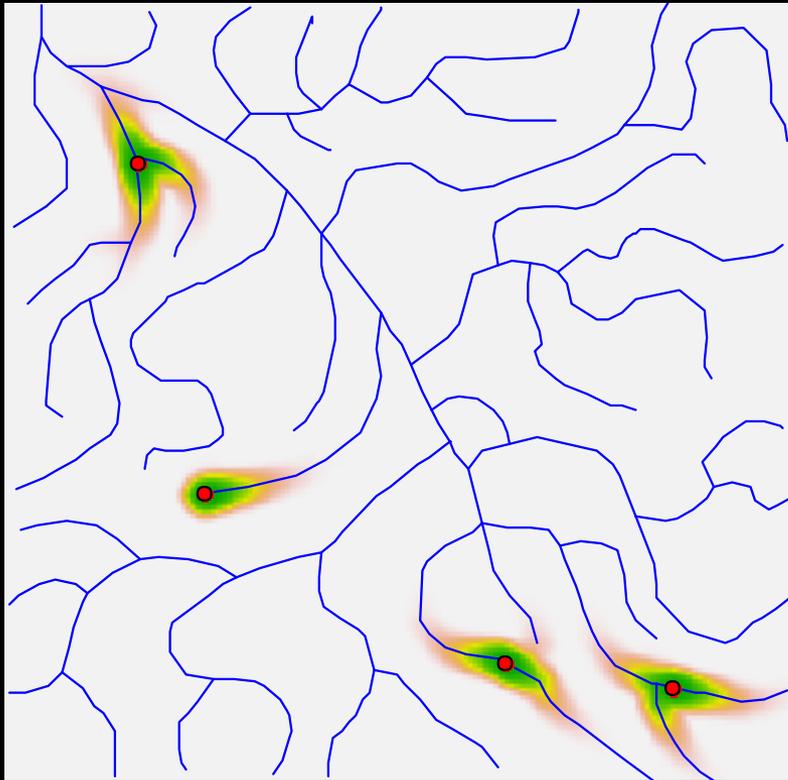


Distance as the  
'fish swims'



# Distance as the *MINK* moves...

Cost weighted 'mink moves' distance:



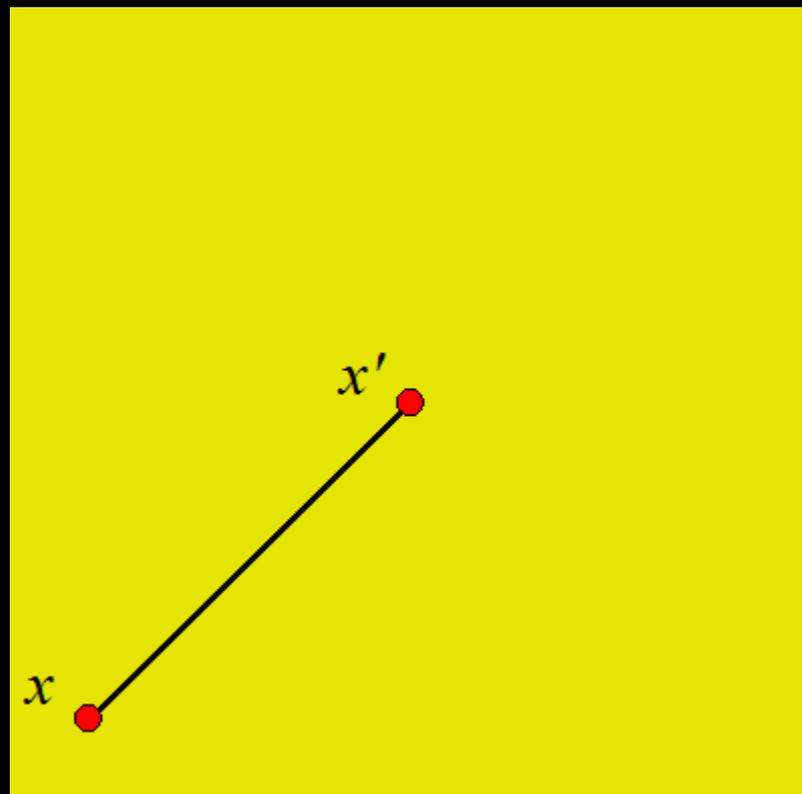
- Spatially varying landscape resistance (*distance to water*)
- Usually arbitrarily defined
- Use observations of movements to *estimate* resistance, *r*  
[*within SCR framework!*]

# Estimating landscape resistance using SCR

Gaussian encounter model:

$$y_{x,x'} \sim \text{Binomial}(K, p_{x,x'})$$

$$p_{x,x'} = \alpha_0 \exp(-\alpha_1 d[x, x']^2)$$



Royle *et al.*, 2013

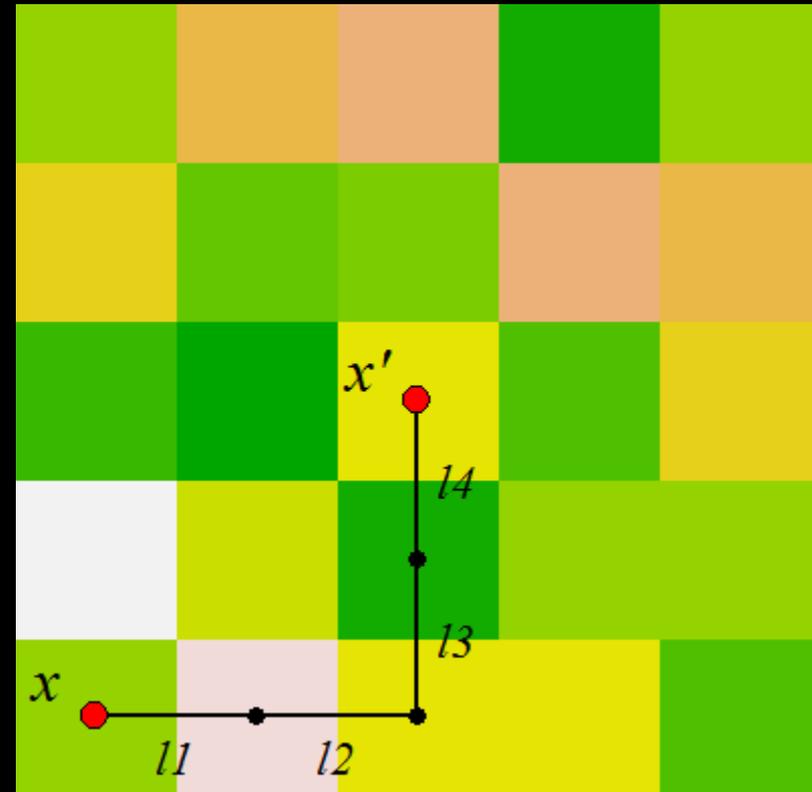
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Gaussian encounter model:

$$y_{x,x'} \sim \text{Binomial}(K, p_{x,x'})$$

$$p_{x,x'} = \alpha_0 \exp(-\alpha_1 d_{lcp}[x, x']^2)$$

$$d(x, x') = \sum_{i=1}^{m-1} \text{cost}(\mathbf{l}_i, \mathbf{l}_{i+1}) \|\mathbf{l}_i - \mathbf{l}_{i+1}\|$$



Royle *et al.*, 2013

# Estimating landscape resistance using SCR

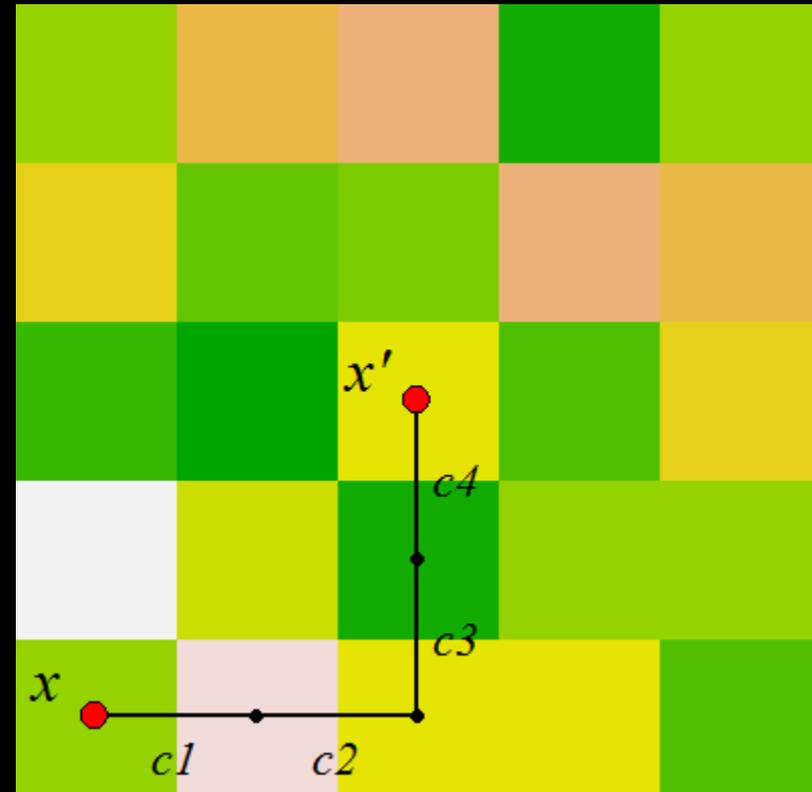
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$$\text{cost}(\mathbf{l}_i, \mathbf{l}_{i+1}) = f[\mathbf{r}, z(\mathbf{l}_i) - z(\mathbf{l}_{i+1})]$$



Royle *et al.*, 2013

# Estimating landscape resistance using SCR

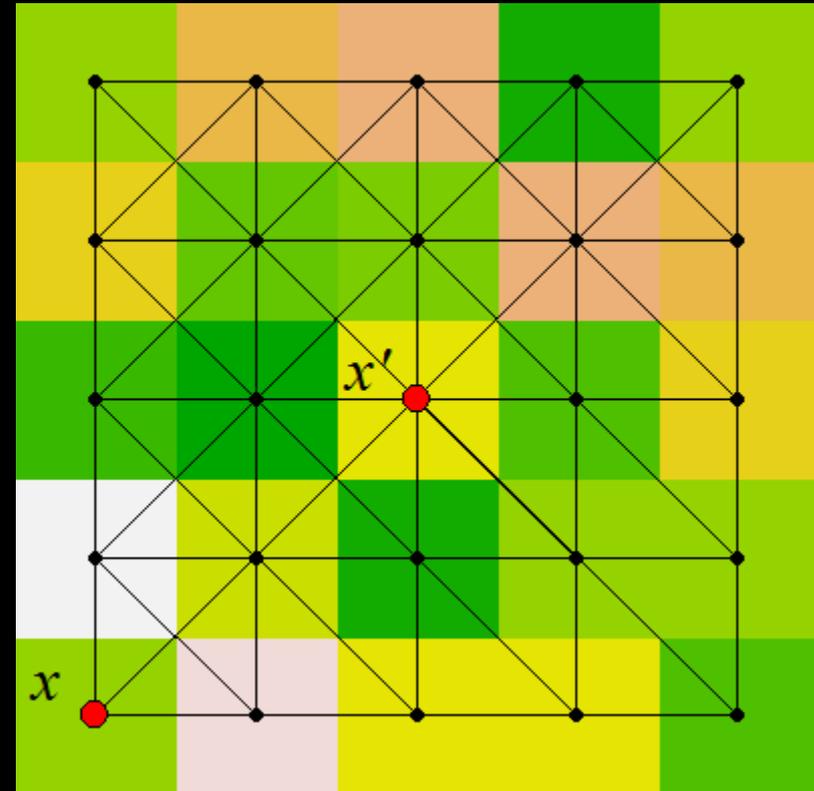
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$$\text{cost}(\mathbf{l}_i, \mathbf{l}_{i+1}) = f[\mathbf{r}, z(\mathbf{l}_i) - z(\mathbf{l}_{i+1})]$$



Royle *et al.*, 2013

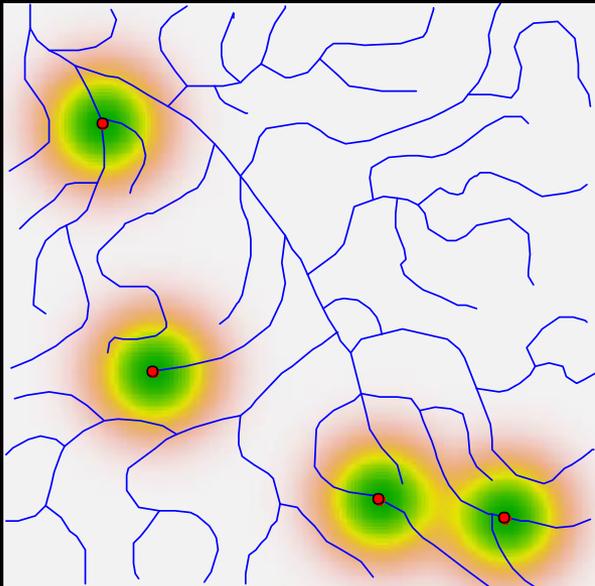
# The cost function - learning about the space-use

$$\log[\text{cost}(x, x')] = r \frac{z(x) + z(x')}{2}$$

$r$  provides information about space use patterns and behavior

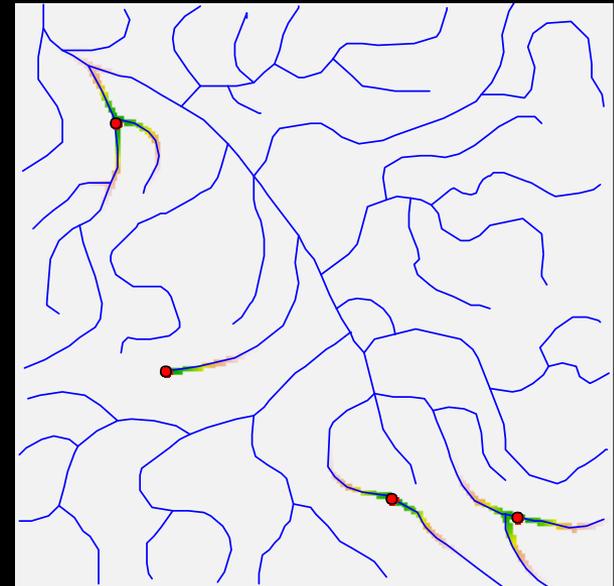
Euclidean distance  
(‘crow flies’)

$$r = 0$$



Stream distance  
(‘fish swims’)

$$r \rightarrow \infty$$



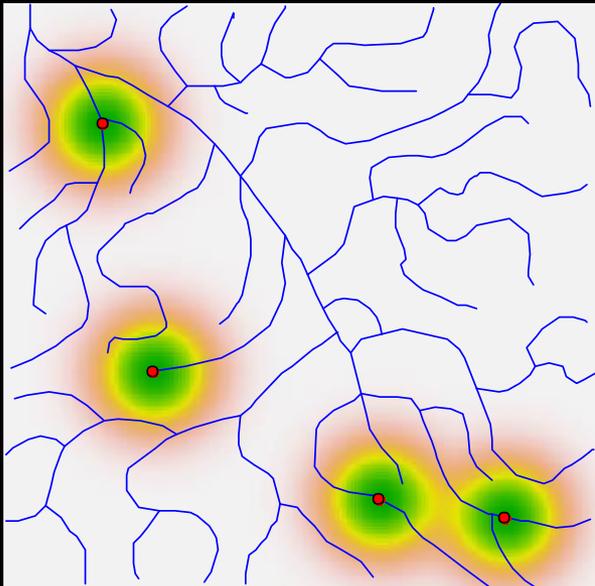
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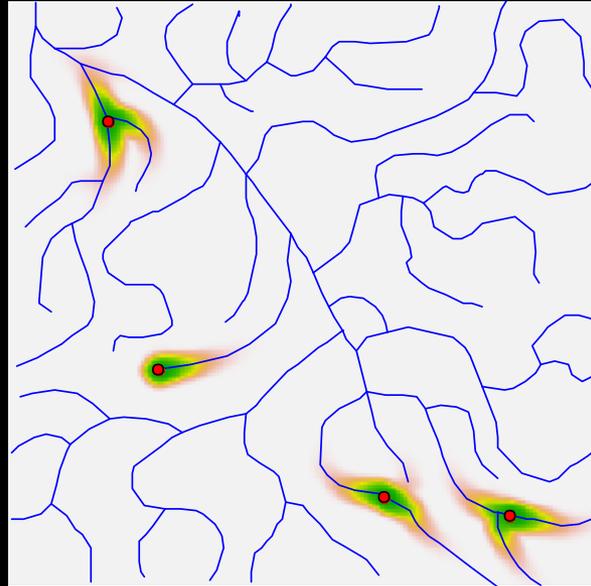
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$$r = 0$$



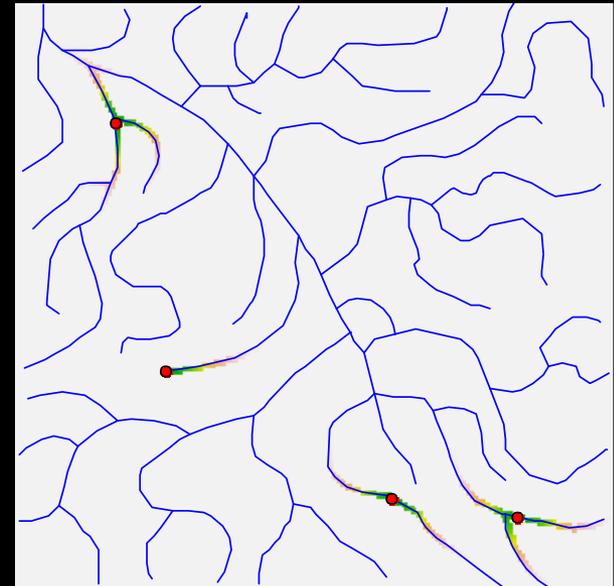
Ecological distance  
(‘mink moves’)

$$0 < r < \infty$$



Stream distance  
(‘fish swims’)

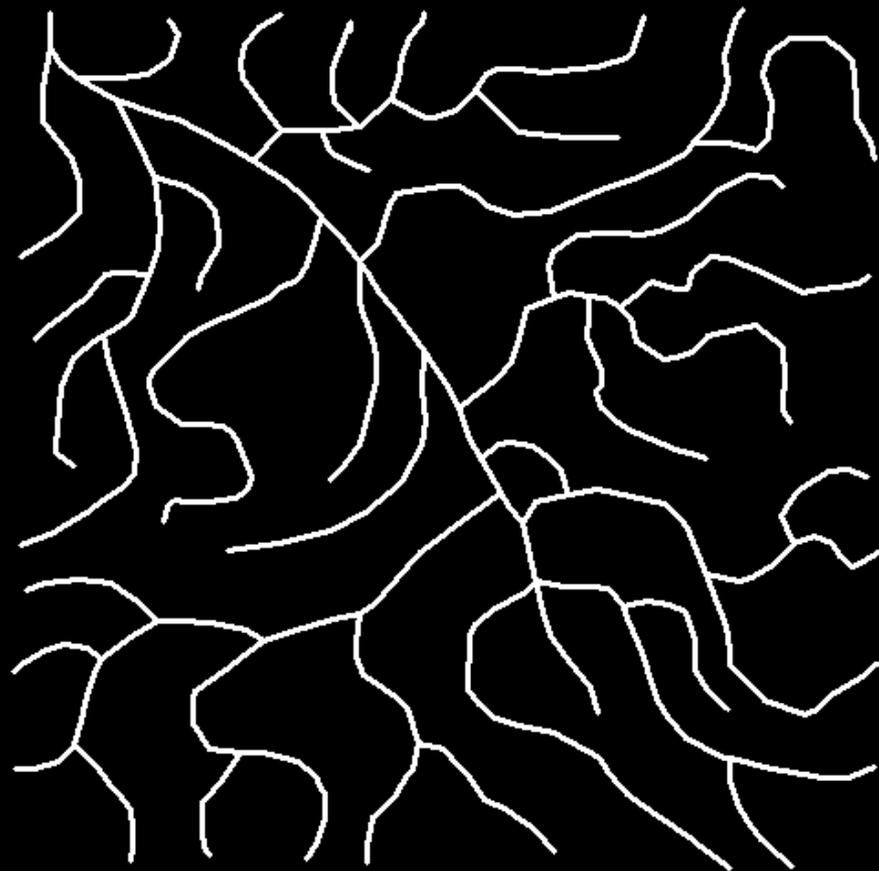
$$r \rightarrow \infty$$



Testing the theory:  
*Mink-like* simulation study

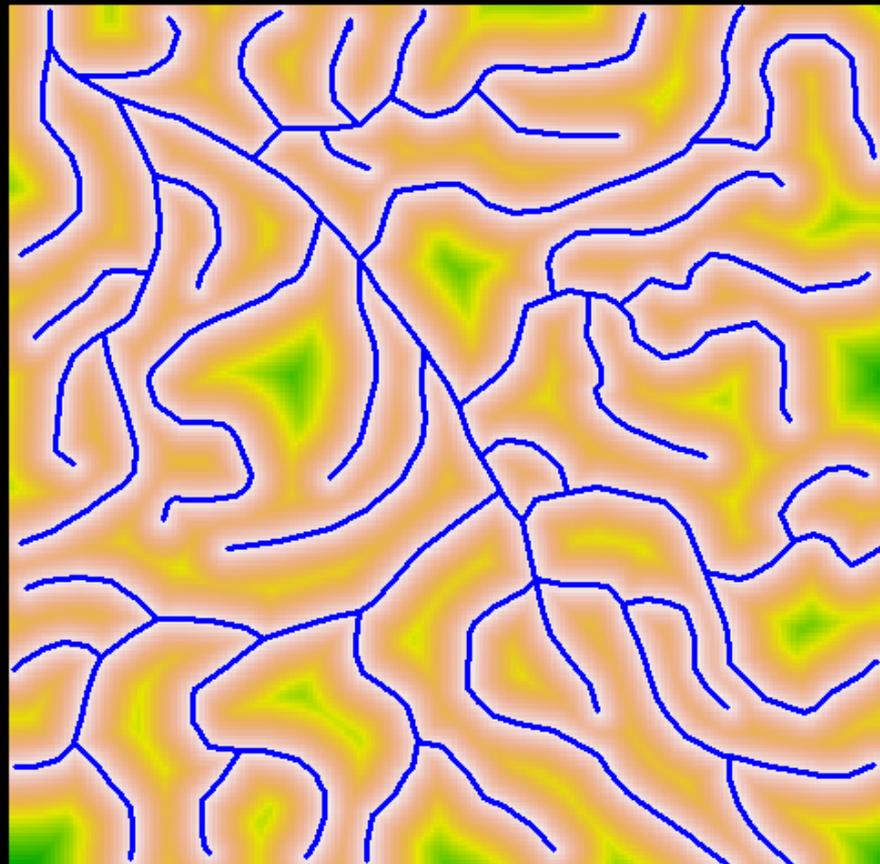
# Mink-like simulation study - Data

- water layer



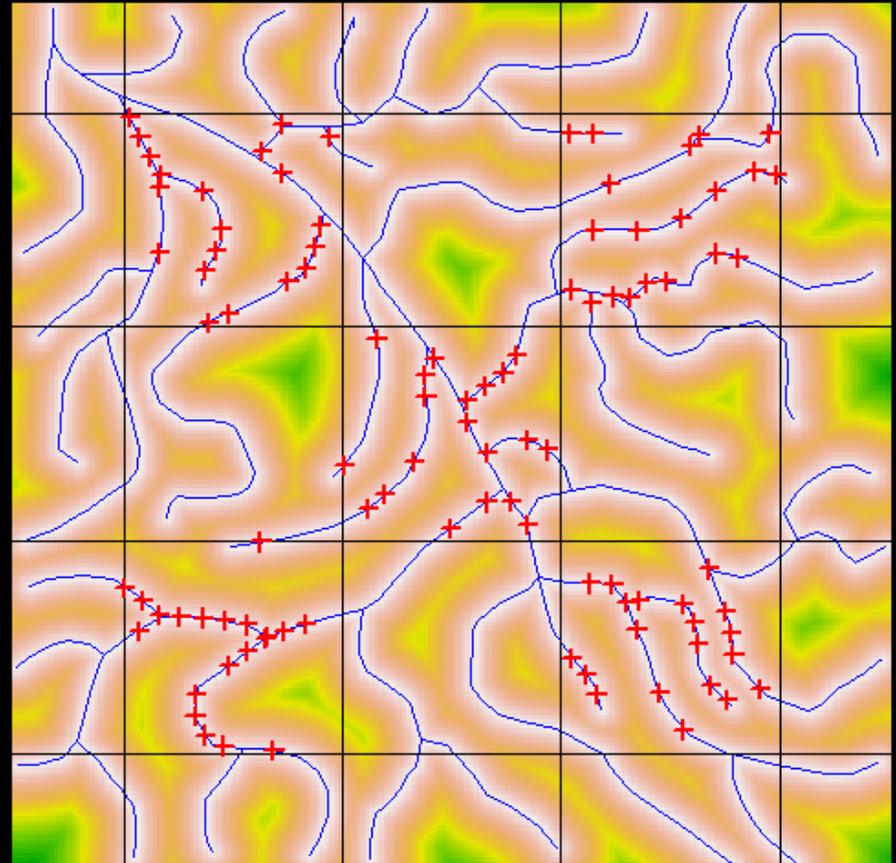
# Mink-like simulation study - Data

- water layer
- distance to water covariate  
(200m x 200m grid cells)



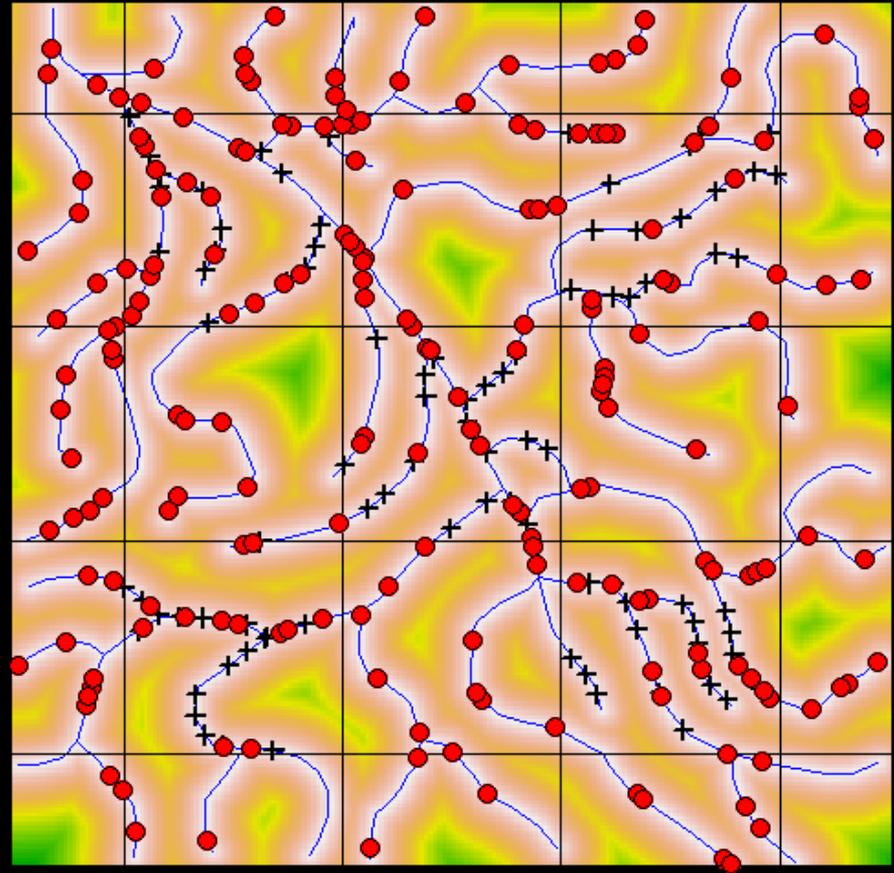
# Mink-like simulation study - Data

- water layer
- distance to water covariate  
(200m x 200m grid cells)
- 100 traps (5 clusters of 20)



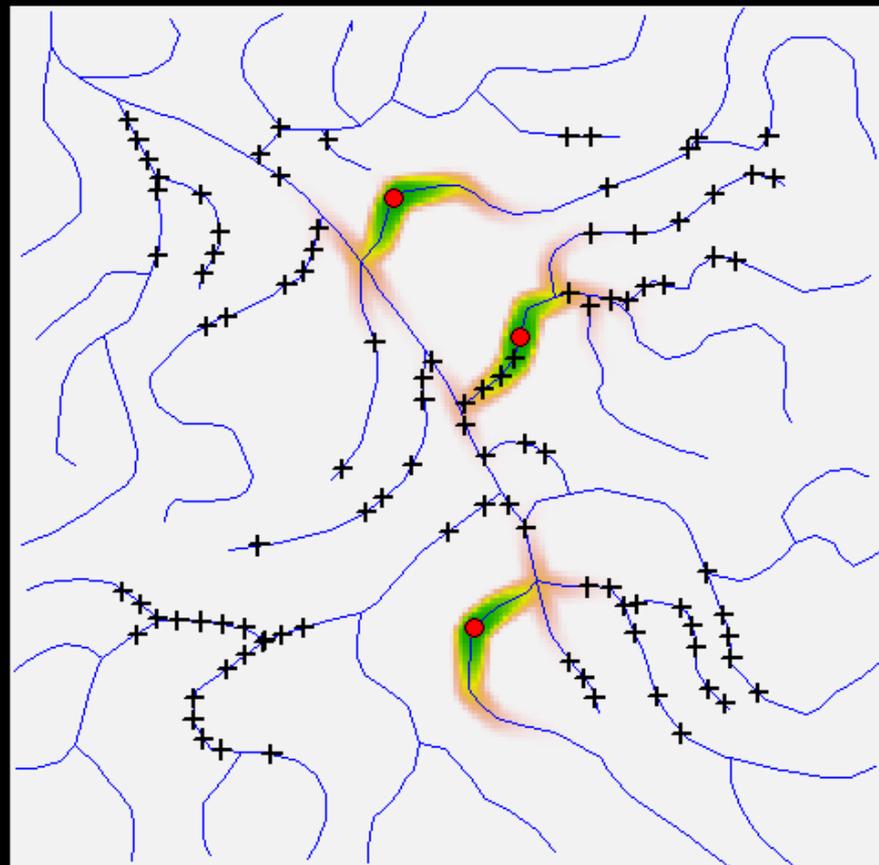
# Mink-like simulation study - Data

- water layer
- distance to water covariate  
(200m x 200m grid cells)
- 100 traps (5 clusters of 20)
- $N = 200$  mink located along the  
water way



# Mink-like simulation study - Data

- water layer
- distance to water covariate  
(200m x 200m grid cells)
- 100 traps (5 clusters of 20)
- $N = 200$  mink located along the water way
- generate spatial encounter histories using:  
 $\alpha_0 = 0.38; \sigma = 0.05; r = 2.5$

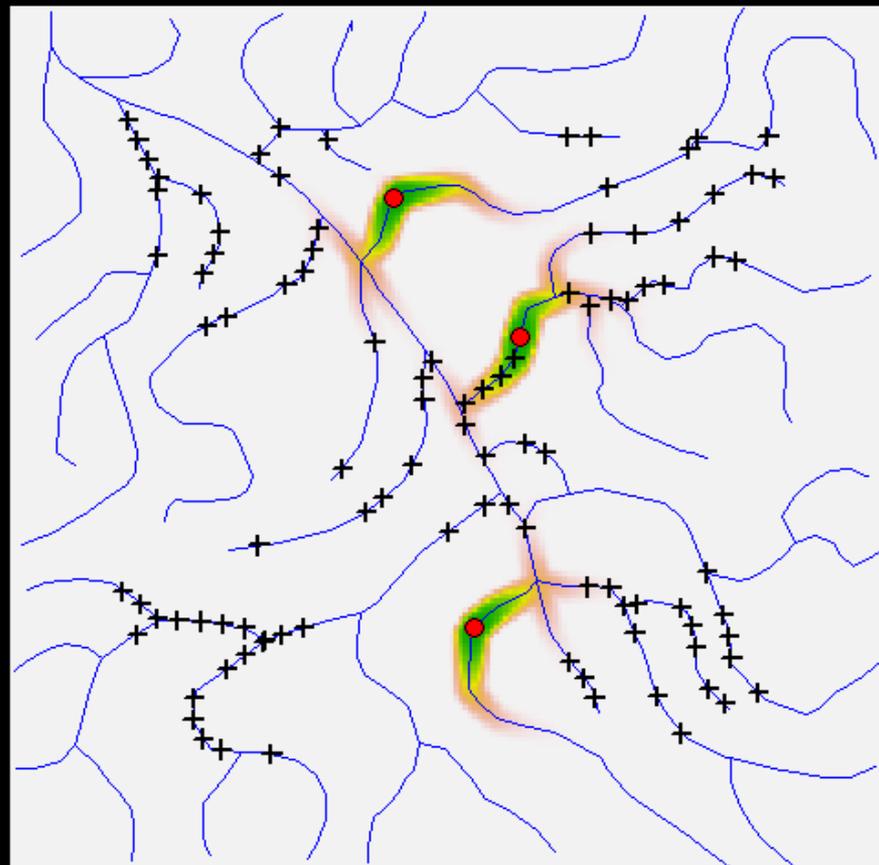


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(200m x 200m grid cells)
- 100 traps (5 clusters of 20)
- $N = 200$  mink located along the water way
- generate spatial encounter histories using:

$$\alpha_0 = 0.38; \sigma = 0.05; r = 2.5$$

- Fit the model (repeat 253 times)



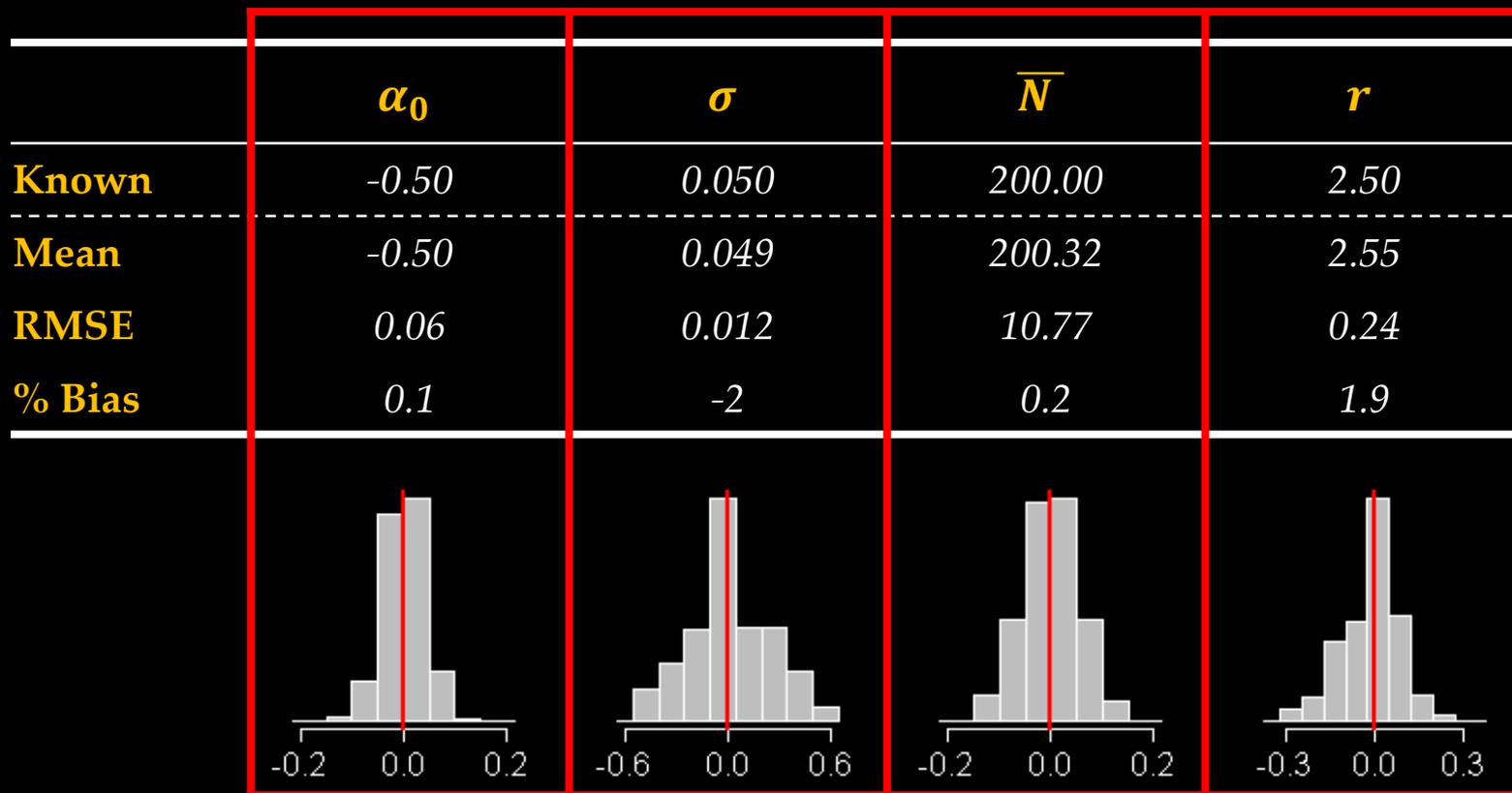
# Mink-like simulation study - Results

1. Retrieving known parameter values using the '*mink moves*' ecological distance model:

Statistical properties of  $\hat{\theta} = [\alpha_0, \sigma, \bar{N}, r]$

# Mink-like simulation study - Results

Estimating parameters of known values  $\theta = [\alpha_0, \sigma, \bar{N}, r]$



$$(\hat{\theta} - \theta) / \theta$$

# Mink-like simulation study - Results

1. Retrieving known parameter values using the '*mink moves*' ecological distance model:

✓  $\hat{\theta} = [\alpha_0, \sigma, \bar{N}, r]$  recovered with little bias

2. Compare the 'performance' of two competing models:

$M_{ecological}$  – '*mink moves*' model estimating landscape resistance

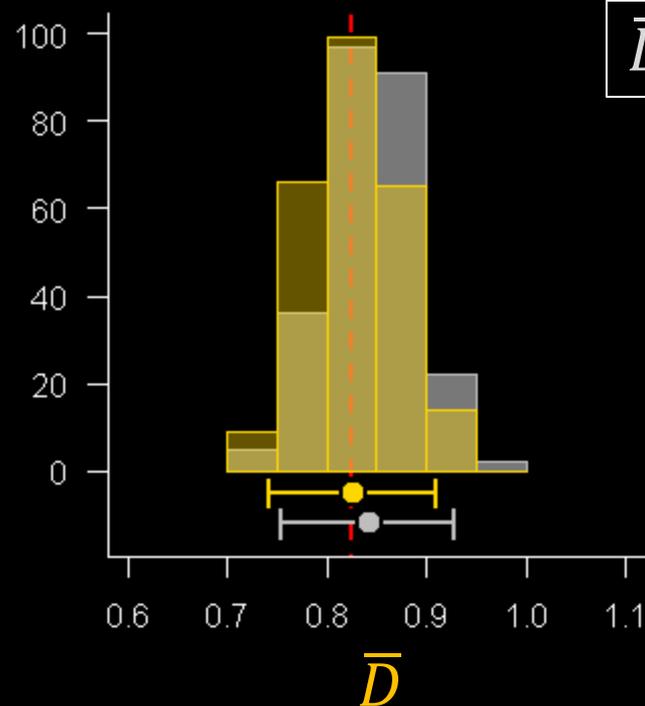
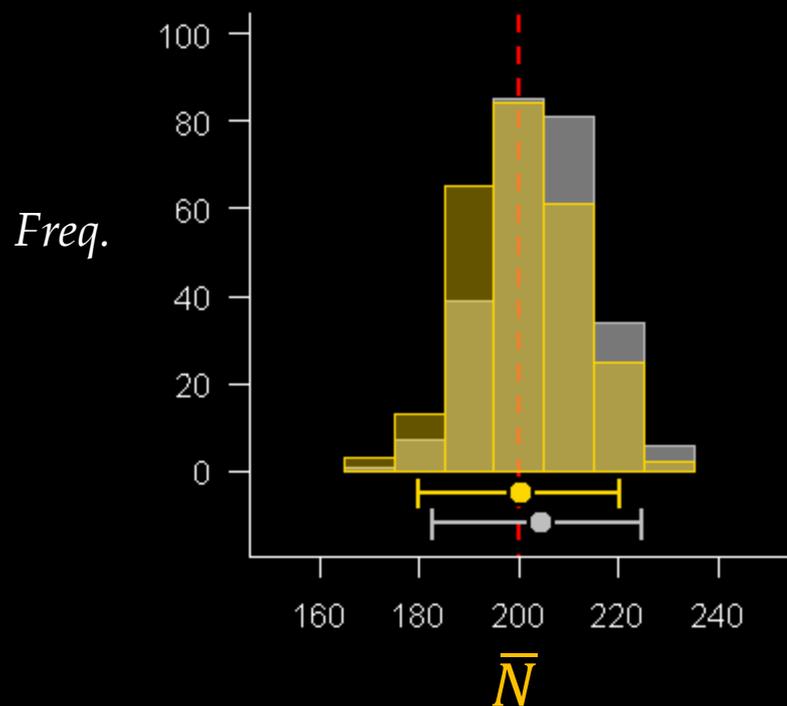
*vs.*

$M_{euclidean}$  – '*crow flies*' model assuming stationary home ranges

# Mink-like simulation study - Results

Comparing estimates of  $\bar{N}$  (and  $\bar{D}$ ) using Euclidean *vs.* ecological distance:

% bias in $N$ ( $D$ )	
Euclidean	2.2
Ecological	0.2

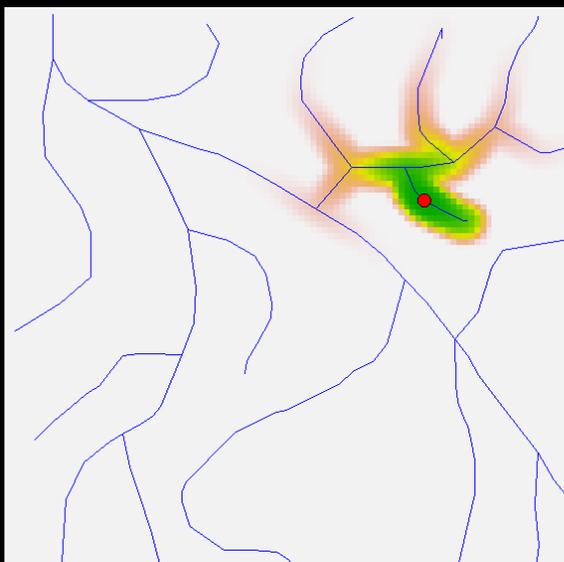


$$\bar{D} = \bar{N}/A$$

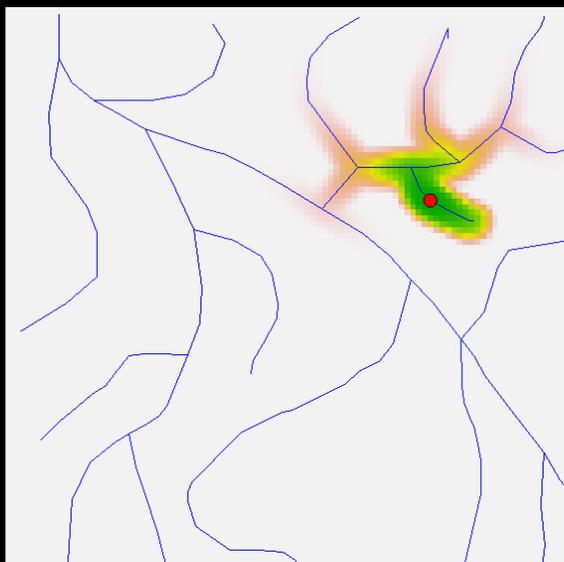
# Mink-like simulation study - Results

Comparing space-use patterns (home range *shapes*) assuming  
Euclidean *vs.* ecological distance:

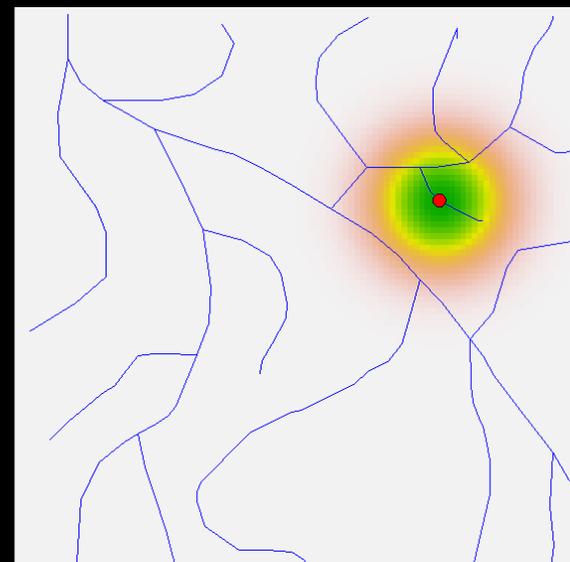
Simulated space-use  
data ('truth')



Estimated using  
'ecological distance'



Estimated using  
'Euclidean distance'



# Mink-like simulation study - Models

1. Retrieving known parameter values using the '*mink moves*' ecological distance model:

✓ Unbiased estimators of  $\hat{\theta} = [\alpha_0, \sigma, \bar{N}, r]$

2. Compare the 'performance' of two competing models:

✓  $M_{ecological}$  – '*mink moves*' model estimating landscape resistance  
vs.  
 $M_{euclidean}$  – '*crow flies*' model assuming stationary home ranges

Testing the theory:

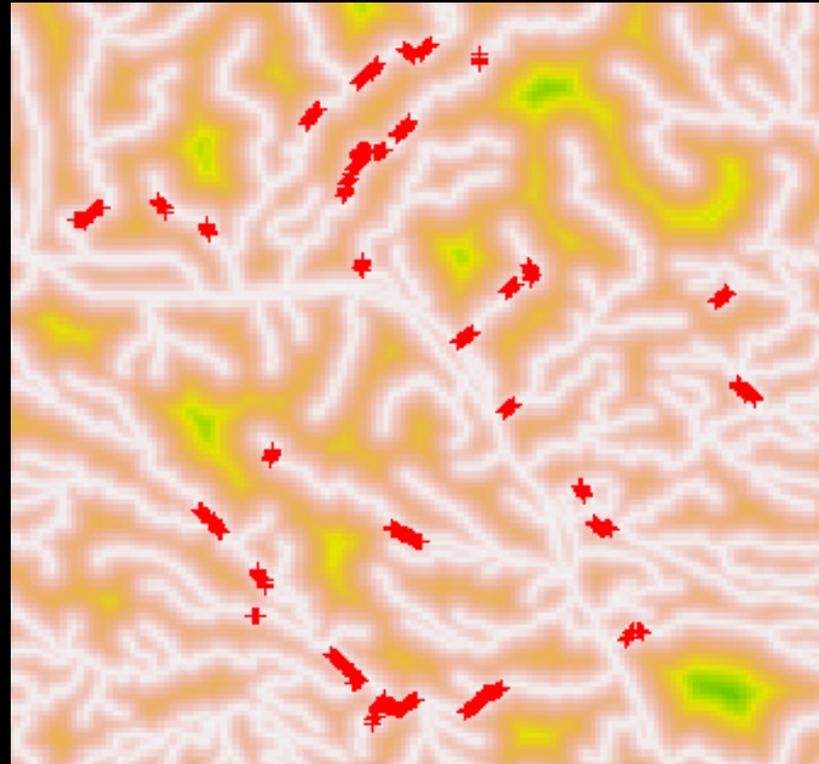
~~Mink-like simulation study~~

Application to a population of  
American mink

# Mink study - Data

- Study area = 293.04 km<sup>2</sup> (515km of stream)
- Scat detection dogs
- Genetic identification of individuals
- 25 transects = 255 'effective traps'
- 37 unique individuals

Frequency	1	2	3	4	5	6	7
# individuals (in $n$ traps)	24	8	3	1	-	-	1



# Mink study - Results

1. Compare the two competing models:

$M_{ecological}$  – '*mink moves*' model estimating landscape resistance

*vs.*

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# Mink study - Results

1. Compare the two competing models:

$M_{ecological}$  – ‘mink moves’ model estimating landscape resistance

*vs.*

$M_{euclidean}$  – ‘crow flies’ model assuming stationary home ranges

## *Model fit (AIC)*

	$N_{params}$	AIC	$\Delta$ AIC	Density (se)
$M_{ecological}$	4	366.59	-	1.06 (0.50)
$M_{euclidean}$	3	372.70	6.11	1.08 (0.54)

# Mink study - Results

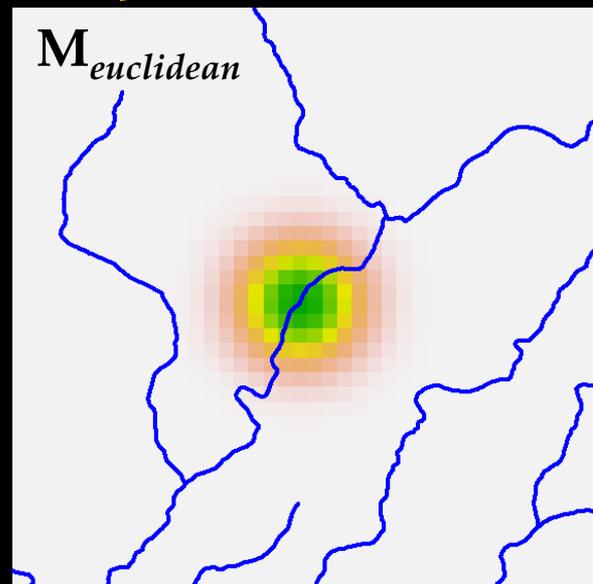
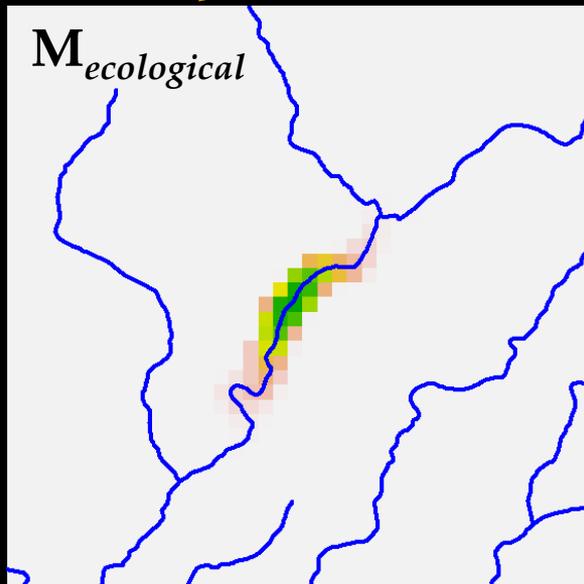
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**Space-use/home range shape:  $r = 20.35$**



# Mink study - Results

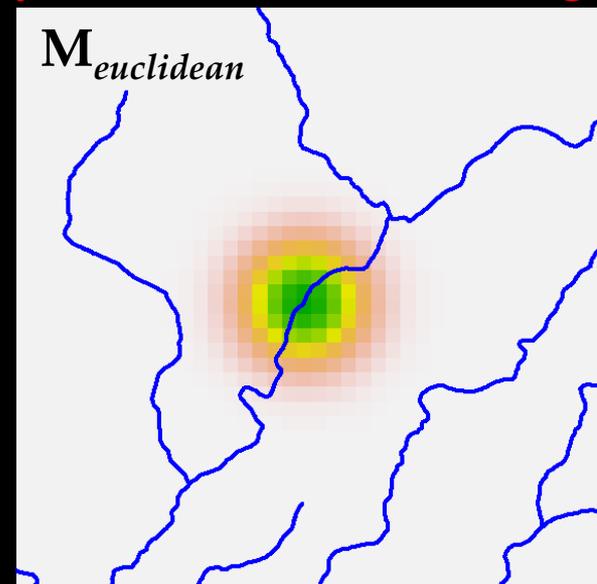
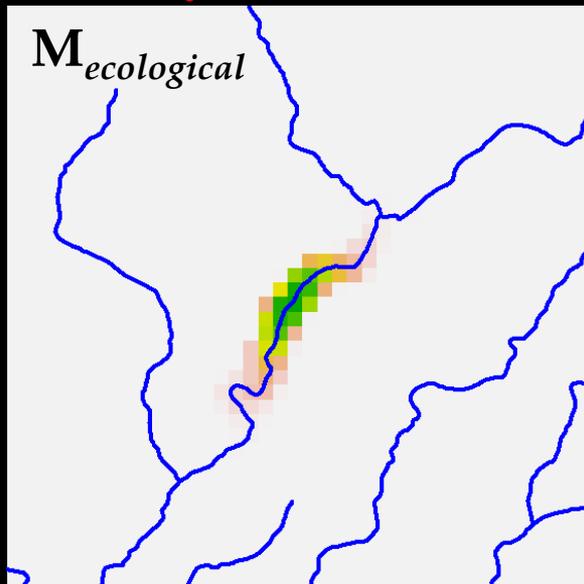
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vs.

$M_{euclidean}$  – ‘crow flies’ model assuming stationary home ranges

7 x more ‘costly’ to move 100m away from water than along water



# Mink study - Results

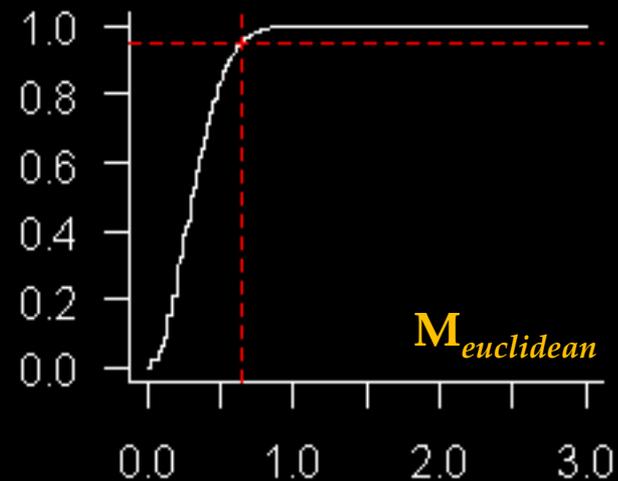
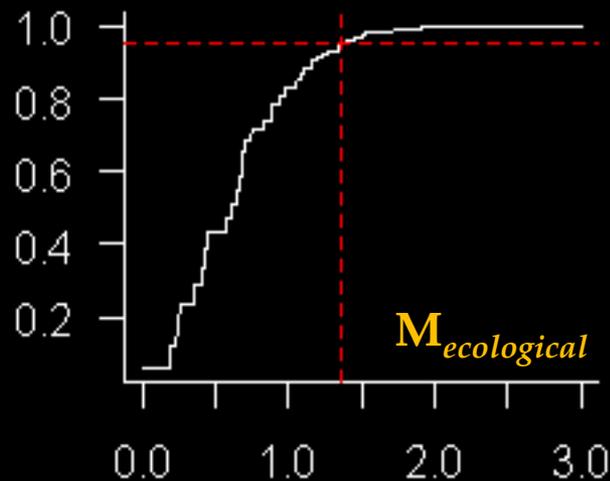
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**95% Home range size**



# Mink study - Results

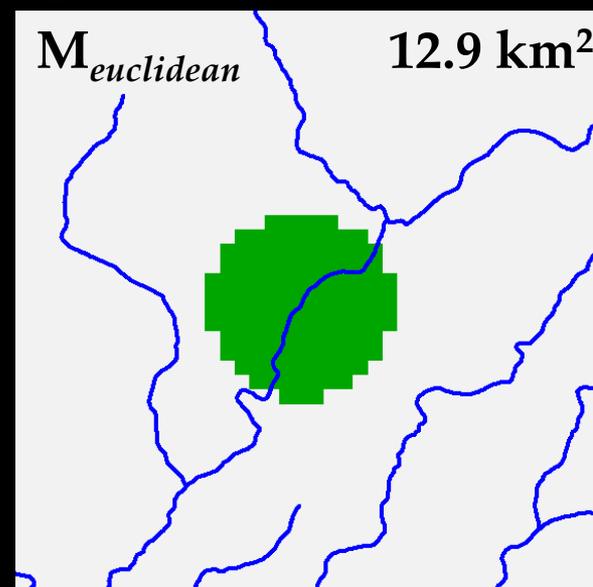
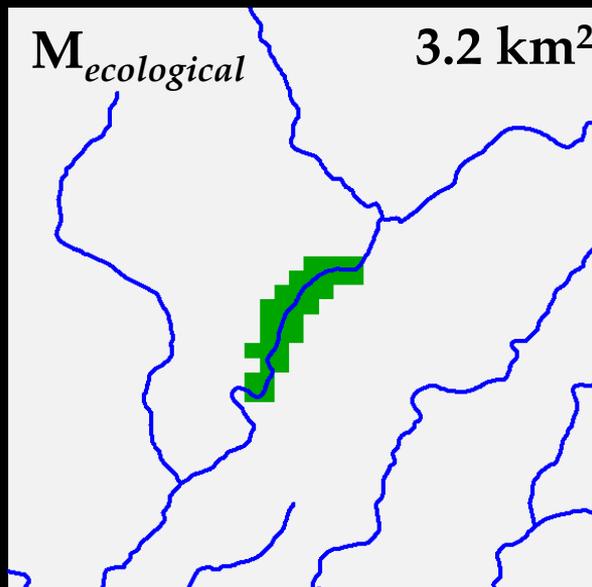
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*vs.*

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**95% Home range size**



# Alternative distance measures and SCR...

Advantages of measuring *ecological distance*:

- relax the Euclidean assumptions of SCR methods
- no bias in estimators of abundance/density BUT
  - estimation of landscape resistance parameter
  - shape/size of irregular home-ranges/space-use

