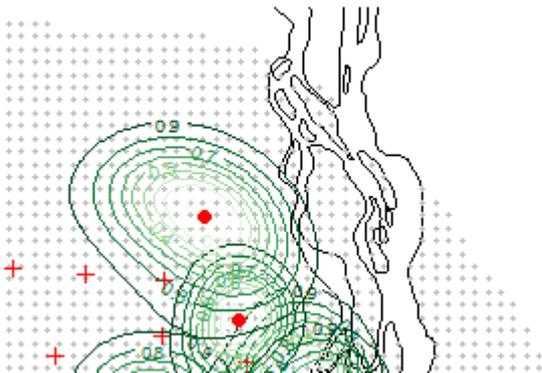


# SECR – *Spatially Explicit Capture-Recapture*

## Modelo Espacialmente Explícito de Capture e Recaptura

- Funções para estimar a densidade de uma população fechada amostrada por diferentes tipo de detectores
- Os modelos populacionais incorporam a detecção dependente da distância e são ajustados pela máxima verossimilhança



# SECR – Spatially Explicit Capture-Recapture

<http://cran.r-project.org/web/packages/secr/index.html>

<http://www.otago.ac.nz/density/SECRinR.html>

Desenvolvido por:

BIOMETRICS 64, 377–385  
June 2008

DOI: 10.1111/j.1541-0420.2007.00927.x



The screenshot shows the University of Otago Zoology Department website. The header features the department's logo and a scenic image of a lake and mountains. The main content area is titled 'Dr Murray Efford' and includes contact information: 'Email [murray\\_efford@otago.ac.nz](mailto:murray_efford@otago.ac.nz)' and 'Phone 64-3-479-8403; 64-3-476-4668'. A search bar and a 'Share this page...' button are also visible. A sidebar on the left lists navigation options such as 'About us', 'Studying Zoology', 'Research', 'Our people', 'News & events', 'Sustainability', 'Contact us', and 'Zoology Intranet'. A small portrait of Dr Murray Efford is shown in the bottom right of the profile section.

<http://www.otago.ac.nz/zoology/staff/otago008940.html>

## Spatially Explicit Maximum Likelihood Methods for Capture-Recapture Studies

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and

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*email:* [murray.efford@stonebow.otago.ac.nz](mailto:murray.efford@stonebow.otago.ac.nz)

**SUMMARY.** Live-trapping capture-recapture studies of animal populations with fixed trap locations inevitably have a spatial component: animals close to traps are more likely to be caught than those far away. This is not addressed in conventional closed-population estimates of abundance and without the spatial com-

*Ecology*, 92(12), 2011, pp. 2202–2207  
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## Estimation of population density by spatially explicit capture-recapture analysis of data from area searches

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**Abstract.** The recent development of capture-recapture methods for estimating animal population density has focused on passive detection using devices such as traps or automatic cameras. Some species lend themselves more to active searching: a polygonal plot may be searched repeatedly and the locations of detected individuals recorded, or a plot may be searched just once and multiple cues (feces or other sign) identified as belonging to particular individuals. This report presents new likelihood-based spatially explicit capture-recapture (SECR) methods for such data. The methods are shown to be at least as robust in simulations as an equivalent Bayesian analysis, and to have negligible bias and near-nominal confidence interval coverage with parameter values from a lizard data set. It is recommended on the basis of simulation that plots for SECR should be at least as large as the home range of the target

# SCR – *Spatial Capture-Recapture*

Desenvolvido por:



## USGS Patuxent Wildlife Research Center Staff Profile

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### Jeffrey A. (Andy) Royle

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**Fax:** 301-497-5545

**Email:** [aroyle@usgs.gov](mailto:aroyle@usgs.gov)

### Research Statistician

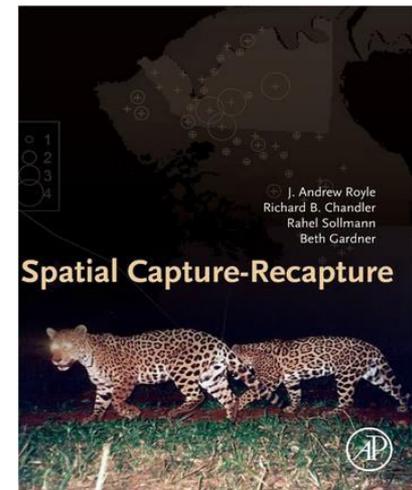
Dr. Royle is a Research Statistician engaged in the development of statistical methods and analytic tools for animal demographic modeling, statistical inference and sampling wildlife populations and communities. His current research is focused on hierarchical models of animal abundance and occurrence, Bayesian analysis in ecology, spatial modeling, and the development of spatial capture-recapture models.

### Education/Training:

Ph.D. (Statistics), North Carolina State University, 1996.

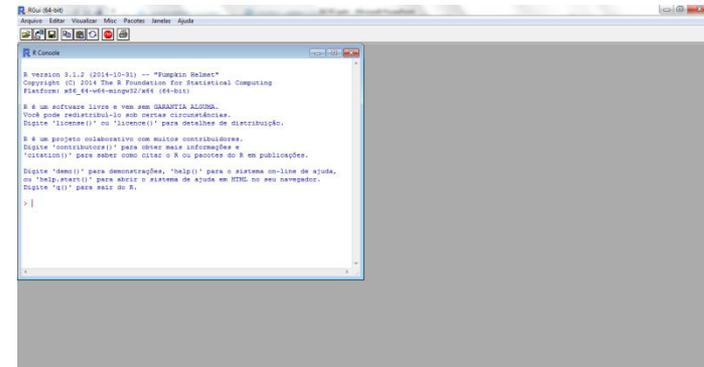
M.St. (Statistics), North Carolina State University, 1993

B.S. (Fisheries and Wildlife Management), Michigan State University, 1990



# SECR – Spatially Explicit Capture-Recapture

## DENSITY X R



### O Density:

- Funciona apenas no Windows
- Algoritmos não bem documentados
- Ajusta somente o processo homogêneo de Poisson
- Omite avanços recentes do SECR

### O R possui a mais:

- Funciona em outros sistemas operacionais
- Modelagem de superfícies de densidade
- Especificação do modelo baseado em fórmula flexível
- Modelo médio
- Detecção acústica
- Processo paralelo
- Modelos híbridos (inferência Bayesiana)
- a0 parametrização para heterogeneidade compensatória
- telemetria integrada com captura-recaptação
- Distâncias não euclidianas (fornecida pelo usuário)

# SECR – detectores

Exemplos

Table 3: Detector types

Sherman	←	single	traps that catch one animal at a time
Pitfall	←	multi	traps that may catch more than one animal at a time
AF/Pêlos	←	proximity	records presence at a point without restricting movement
	←	count	proximity detector allowing >1 detection per animal per time
	←	polygon	counts from searching one or more areas
	←	transect	counts from searching one or more transects
Quando há dependência	}	polygonX	binary data from mutually exclusive areas
		transectX	binary data from mutually exclusive transects
Vocalização	←	signal	detections and signal strengths at multiple microphones
Rádio colar	←	telemetry	locations from radiotelemetry

# SECR

- Permite:

- definir área de amostragem inicial com base no:

1- histórico de captura (caphist)

2- log da verossimilhança (mask.check) no ajuste do modelo\*

\*Não estima buffer com modelos híbridos (classe)

- histórico de captura do modelo ajustado (suggest.buffer)

- MMDM, dbar, RPSV, ARL e moves

# SECR

3 parâmetros primários compõem o modelo SECR

**1** - densidade populacional ( $D$ )

$(D)$  é um parâmetro derivado, computado de um objeto do secr.

**2** - probabilidade ( $g_0$ ) ou risco acumulativo ( $\lambda$ ) de detecção para um detector no centro da área de vida do indivíduo

**3** - Escala espacial ( $\sigma$ )

# SECR

Dois parâmetros adicionais podem compor todos os modelos em concorrência

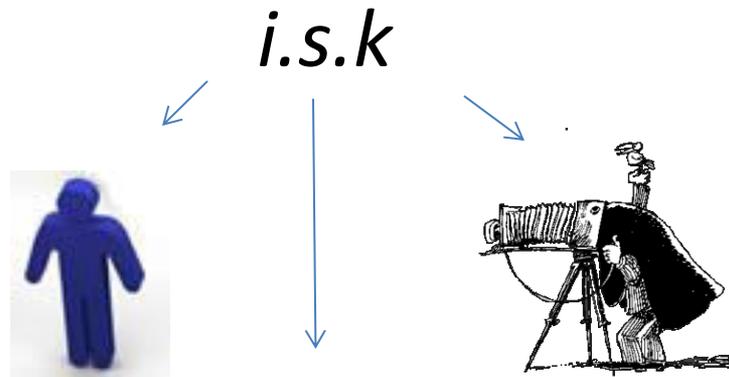
1 - parâmetro de forma ( $z$ )  $\longrightarrow$  função de detecção

2 - Parâmetro de classe, ( $hcov$ ) nos modelos *finite mixture models* e *hybrid mixture models*  $\longrightarrow$  estimativas por classe

Três parâmetros de detecção podem ser estimados diretamente quando o modelo é ajustado pela verossimilhança condicional (CL = TRUE no `secr.fit`)

# SECR - detecção

Estimativas populacionais de uma população fechada no SECR, modela-se a detecção do indivíduo  $i$  na ocasião  $s$  pelo detector  $k$ .



$i$  = indivíduo

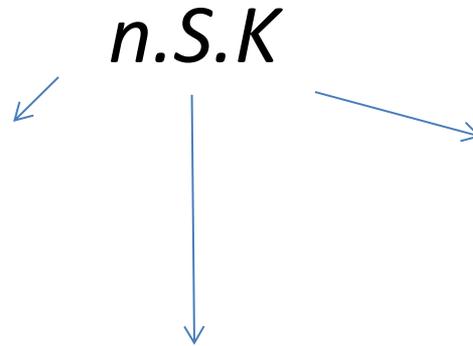
$S$  = ocasião

$k$  = detector



# SECR - detecção

Considerando  $n$  observações de  $i$  indivíduos em  $S$  ocasiões em  $K$  detectores:



$n$  = indivíduos observados  
 $S$  = ocasiões  
 $K$  = detectores



## SECR - detecção

Modelo nulo (constante), ***n.S.K*** tem a mesma probabilidade de detecção.

***n.S.K***

*n* = indivíduos observados

*S* = ocasiões

*K* = detectores

# SECR - detecção

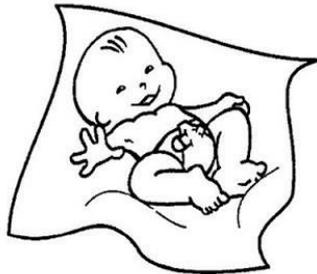
Segundo Otis et al (1978), as formas convencionais de variação na probabilidade de detecção (captura) incluem:

*n.S.K*



heterogeneidade  
Individual (h)

Ex: classe de idade



## SECR - detecção

Segundo Otis et al (1978), as formas convencionais de variação na probabilidade de detecção (captura) incluem:

*n.S.K*

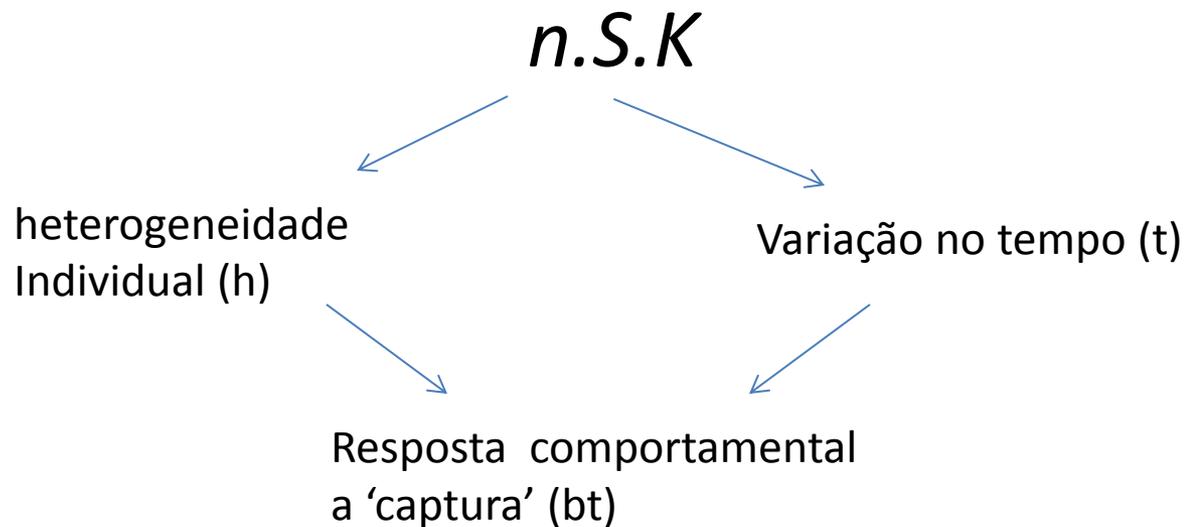


Variação no tempo (t)



# SECR - detecção

Segundo Otis et al (1978), as formas convencionais de variação na probabilidade de detecção (captura) incluem:



*Trap-happy*

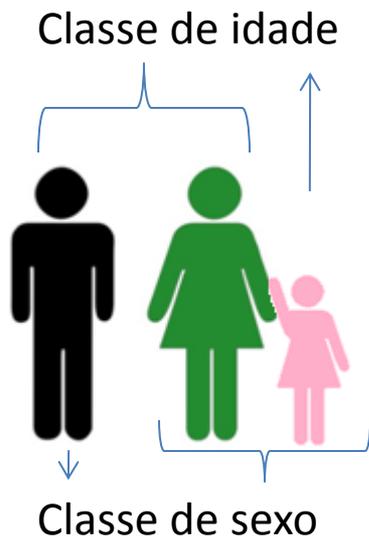


*Trap-shy*

# SECR - detecção

Diferenças entre indivíduos ( $i$ ) esforço ( $s$ ) e ambientes amostrados ( $k$ ), podem ser inseridos como **co-variáveis** do modelo.

*indivíduo, ocasião, detector*



<http://ultradownloads.com.br/papel-de-parede/Trilha-na-Floresta>



## SECR - detecção

No SECR, há duas adições de complexidade:

Podem ser inseridas

variações entre sessões (temporadas) –  $(1...R)$

e entre classes latentes –  $(1...M)$

*$n.S.K.M.R$*

# SECR - detecção

Estrutura de uma matriz

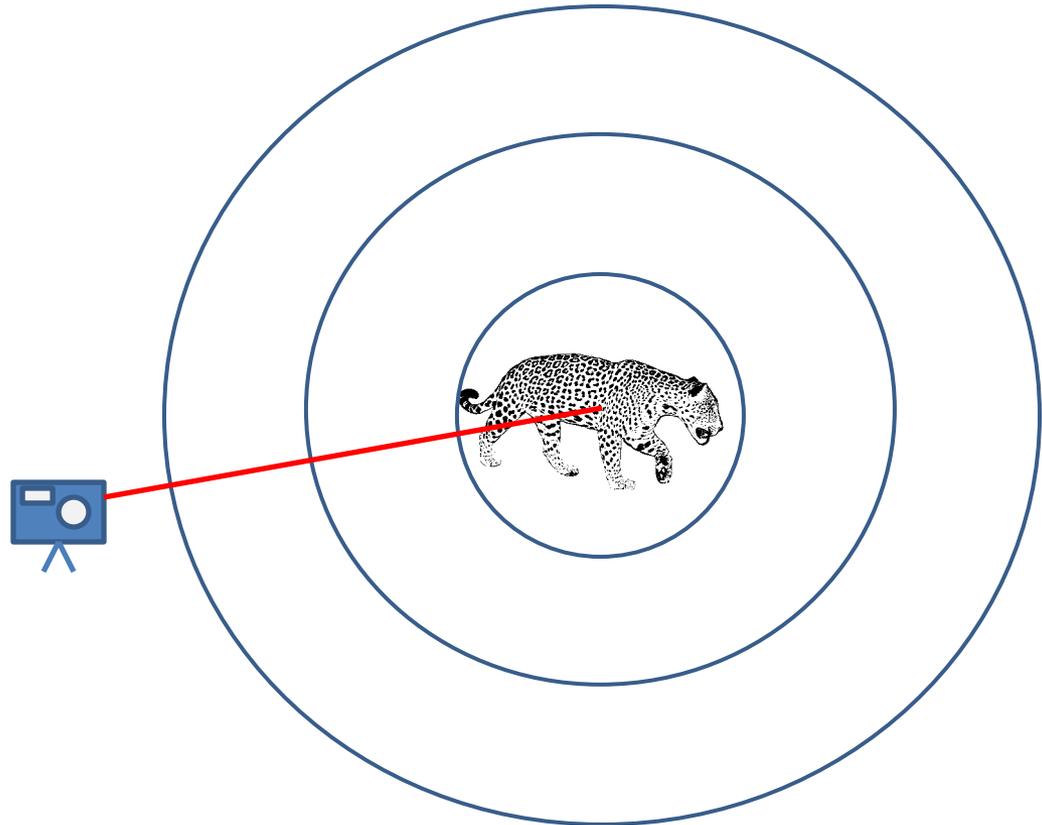
*n.S.K.M.R*

	A
1	id,detector,x,y,/1,2,3,4,5,6,7,8,9
2	1,TS1,358597,8914064,/216,241,221,287,262,330,175,224,272
3	2,TS2,360498,8913916,/216,241,218,289,262,314,191,238,257
4	3,TS4,362500,8912302,/216,241,146,311,262,312,193,38,254
5	4,TS5,361631,8910940,/237,215,215,311,264,313,191,237,241
6	5,TS6,359707,8911736,/219,233,215,316,260,313,192,238,241
7	6,TS7,357769,8910418,/220,214,232,314,262,313,192,238,241

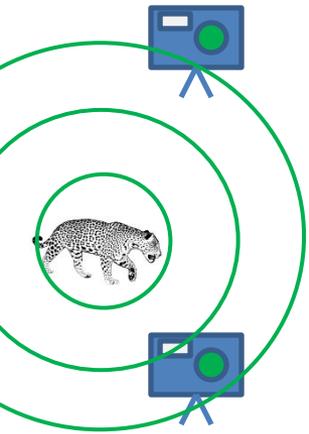
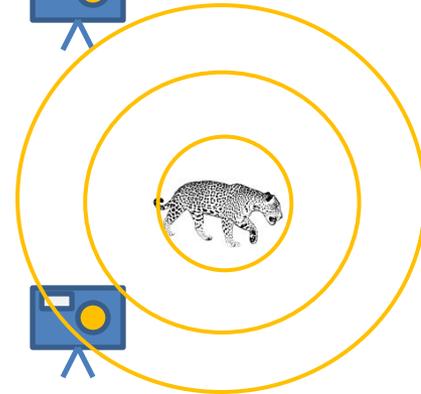
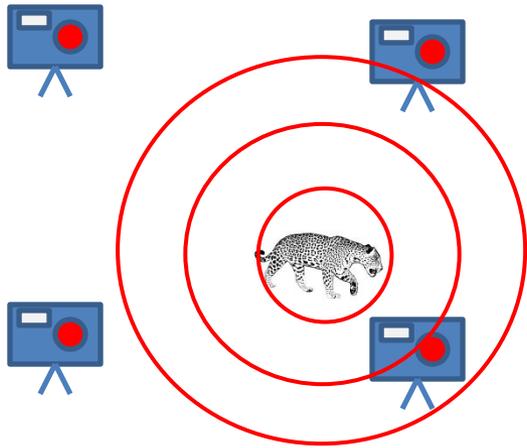
session	id	occasion	trapid	sex
F1	1F	3	2	f
F1	1F	3	3	f
F1	2M	1	2	m
F1	3F	6	2	f
F1	4M	5	2	m
F1	5F	1	8	f
F1	5F	4	9	f
F1	5F	6	8	f
F1	5F	9	9	f
F1	6M	3	5	m

## SECR - detecção

Relaciona a probabilidade de detecção “ $g$ ” ou o esperado número de detecções “ $\lambda$ ” de um animal à distância do detector ao “centro potencial de atividade do indivíduo”



# SECR - detecção



# SECR - detecção

No SECR a detecção é definida por uma função.

Dois parâmetros para *halfnormal* –  $g_0$  e  $\sigma$

$$g(d) = g_0 \exp\left(\frac{-d^2}{2\sigma^2}\right)$$

Três parâmetros para função *hazard-rate* –  $g_0$ ,  $\sigma$  e  $z$

$$g(d) = g_0 [1 - \exp\{-(d/\sigma)^{-z}\}]$$

Outras possibilidades são discutidas

?*detectfn*

$d$  = distância

$g_0$  = magnitude (intercepção)  
da função de detecção

$\sigma$  = escala espacial  
da função de detecção

$z$  = parâmetro de forma (área)

# SECR - detecção

Code	Name	Parameters	Function
0 HN	<u>halfnormal</u>	$g_0, \sigma$	$g(d) = g_0 \exp\left(\frac{-d^2}{2\sigma^2}\right)$
1 HR	hazard rate	$g_0, \sigma, z$	$g(d) = g_0 [1 - \exp\{-(d/\sigma)^{-z}\}]$
2 EX	exponential	$g_0, \sigma$	$g(d) = g_0 \exp\{-(d/\sigma)\}$
3 CHN	compound halfnormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - \{1 - \exp\left(\frac{-d^2}{2\sigma^2}\right)\}^z]$
4 UN	uniform	$g_0, \sigma$	$g(d) = g_0, d \leq \sigma; g(d) = 0, \text{ otherwise}$
5 WEX	w exponential	$g_0, \sigma, w$	$g(d) = g_0, d < w; g(d) = g_0 \exp\left(-\frac{d-w}{\sigma}\right), \text{ otherwise}$
6 ANN	annular normal	$g_0, \sigma, w$	$g(d) = g_0 \exp\left\{\frac{-(d-w)^2}{2\sigma^2}\right\}$
7 CLN	cumulative lognormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - F\{(d - \mu)/s\}]$
8 CG	cumulative gamma	$g_0, \sigma, z$	$g(d) = g_0 \{1 - G(d; k, \theta)\}$
9 BSS	binary signal strength	$b_0, b_1$	$g(d) = 1 - F\{-(b_0 + b_1 d)\}$
10 SS	signal strength	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1 d)]/s\}$
11 SSS	signal strength spherical	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1(d-1) - 10 \log_{10} d^2)]/s\}$
14 HHN	hazard halfnormal	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\left(\frac{-d^2}{2\sigma^2}\right); g(d) = 1 - \exp(-\lambda(d))$
15 HHR	hazard hazard rate	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 (1 - \exp\{-(d/\sigma)^{-z}\}); g(d) = 1 - \exp(-\lambda(d))$
16 HEX	hazard exponential	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\{-(d/\sigma)\}; g(d) = 1 - \exp(-\lambda(d))$
17 HAN	hazard annular normal	$\lambda_0, \sigma, w$	$\lambda(d) = \lambda_0 \exp\left\{\frac{-(d-w)^2}{2\sigma^2}\right\}; g(d) = 1 - \exp(-\lambda(d))$
18 HCG	hazard cumulative gamma	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 \{1 - G(d; k, \theta)\}; g(d) = 1 - \exp(-\lambda(d))$

0  Default - recomendada

0-3 e 5-8  Trabalham com a probabilidade de detecção “g”

# SECR - detecção

Code	Name	Parameters	Function
0 HN	halfnormal	$g_0, \sigma$	$g(d) = g_0 \exp\left(\frac{-d^2}{2\sigma^2}\right)$
1 HR	hazard rate	$g_0, \sigma, z$	$g(d) = g_0 [1 - \exp\{-(d/\sigma)^{-z}\}]$
2 EX	exponential	$g_0, \sigma$	$g(d) = g_0 \exp\{-(d/\sigma)\}$
3 CHN	compound halfnormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - \{1 - \exp\left(\frac{-d^2}{2\sigma^2}\right)\}^z]$
4 UN	uniform	$g_0, \sigma$	$g(d) = g_0, d \leq \sigma; g(d) = 0, \text{ otherwise}$
5 WEX	w exponential	$g_0, \sigma, w$	$g(d) = g_0, d < w; g(d) = g_0 \exp\left(-\frac{d-w}{\sigma}\right), \text{ otherwise}$
6 ANN	annular normal	$g_0, \sigma, w$	$g(d) = g_0 \exp\left\{\frac{-(d-w)^2}{2\sigma^2}\right\}$
7 CLN	cumulative lognormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - F\{(d - \mu)/s\}]$
8 CG	cumulative gamma	$g_0, \sigma, z$	$g(d) = g_0 \{1 - G(d; k, \theta)\}$
9 BSS	binary signal strength	$b_0, b_1$	$g(d) = 1 - F\{-(b_0 + b_1 d)\}$
10 SS	signal strength	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1 d)]/s\}$
11 SSS	signal strength spherical	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1(d-1) - 10 \log_{10} d^2)]/s\}$
14 HHN	hazard halfnormal	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\left(\frac{-d^2}{2\sigma^2}\right); g(d) = 1 - \exp(-\lambda(d))$
15 HHR	hazard hazard rate	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 (1 - \exp\{-(d/\sigma)^{-z}\}); g(d) = 1 - \exp(-\lambda(d))$
16 HEX	hazard exponential	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\{-(d/\sigma)\}; g(d) = 1 - \exp(-\lambda(d))$
17 HAN	hazard annular normal	$\lambda_0, \sigma, w$	$\lambda(d) = \lambda_0 \exp\left\{\frac{-(d-w)^2}{2\sigma^2}\right\}; g(d) = 1 - \exp(-\lambda(d))$
18 HCG	hazard cumulative gamma	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 \{1 - G(d; k, \theta)\}; g(d) = 1 - \exp(-\lambda(d))$

14 a 18 

Parametrizadas em termos do número esperado de detecções “ $\lambda$ ” ao invés da probabilidade de detecção “ $g$ ”



Detectores tipo “count”, se a função é interpretada como uma distribuição de atividade

# SECR - detecção

	Code	Name	Parameters	Function
	0 HN	halfnormal	$g_0, \sigma$	$g(d) = g_0 \exp\left(\frac{-d^2}{2\sigma^2}\right)$
Hayes & Buckland, 1983	→ 1 HR	hazard rate	$g_0, \sigma, z$	$g(d) = g_0 [1 - \exp\{-(d/\sigma)^{-z}\}]$
	→ 2 EX	exponential	$g_0, \sigma$	$g(d) = g_0 \exp\{-(d/\sigma)\}$
Efford & Dawson, 2009	→ 3 CHN	compound halfnormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - \{1 - \exp\left(\frac{-d^2}{2\sigma^2}\right)\}^z]$
Simulações *	→ 4 UN	uniform	$g_0, \sigma$	$g(d) = g_0, d \leq \sigma; g(d) = 0, \text{ otherwise}$
	5 WEX	w exponential	$g_0, \sigma, w$	$g(d) = g_0, d < w; g(d) = g_0 \exp\left(-\frac{d-w}{\sigma}\right), \text{ otherwise}$
	6 ANN	annular normal	$g_0, \sigma, w$	$g(d) = g_0 \exp\left\{\frac{-(d-w)^2}{2\sigma^2}\right\}$
	7 CLN	cumulative lognormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - F\{(d - \mu)/s\}]$
	8 CG	cumulative gamma	$g_0, \sigma, z$	$g(d) = g_0 \{1 - G(d; k, \theta)\}$
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	10 SS	signal strength	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1 d)]/s\}$
	11 SSS	signal strength spherical	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1(d-1) - 10 \log_{10} d^2)]/s\}$
	14 HHN	hazard halfnormal	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\left(\frac{-d^2}{2\sigma^2}\right); g(d) = 1 - \exp(-\lambda(d))$
Hayes & Buckland, 1983	→ 15 HHR	hazard hazard rate	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 (1 - \exp\{-(d/\sigma)^{-z}\}); g(d) = 1 - \exp(-\lambda(d))$
	→ 16 HEX	hazard exponential	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\{-(d/\sigma)\}; g(d) = 1 - \exp(-\lambda(d))$
	17 HAN	hazard annular normal	$\lambda_0, \sigma, w$	$\lambda(d) = \lambda_0 \exp\left\{\frac{-(d-w)^2}{2\sigma^2}\right\}; g(d) = 1 - \exp(-\lambda(d))$
	18 HCG	hazard cumulative gamma	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 \{1 - G(d; k, \theta)\}; g(d) = 1 - \exp(-\lambda(d))$

\* Problemas de maximização da verossimilhança por métodos de gradiente.

Não recomendadas para uso no secr devido a longa cauda (buffer irracionalmente muito grande para estabilizar a estimativa de densidade)

# SECR - detecção

Code	Name	Parameters	Function
0 HN	halfnormal	$g_0, \sigma$	$g(d) = g_0 \exp\left(-\frac{d^2}{2\sigma^2}\right)$
1 HR	hazard rate	$g_0, \sigma, z$	$g(d) = g_0 [1 - \exp\{-(d/\sigma)^{-z}\}]$
2 EX	exponential	$g_0, \sigma$	$g(d) = g_0 \exp\{-(d/\sigma)\}$
3 CHN	compound halfnormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - \{1 - \exp\left(-\frac{d^2}{2\sigma^2}\right)\}^z]$
4 UN	uniform	$g_0, \sigma$	$g(d) = g_0, d \leq \sigma; g(d) = 0, \text{ otherwise}$
5 WEX	w exponential	$g_0, \sigma, w$	$g(d) = g_0, d < w; g(d) = g_0 \exp\left(-\frac{d-w}{\sigma}\right), \text{ otherwise}$
6 ANN	annular normal	$g_0, \sigma, w$	$g(d) = g_0 \exp\left\{-\frac{(d-w)^2}{2\sigma^2}\right\}$
7 CLN	cumulative lognormal	$g_0, \sigma, z$	$g(d) = g_0 [1 - F\{(d - \mu)/s\}]$
8 CG	cumulative gamma	$g_0, \sigma, z$	$g(d) = g_0 \{1 - G(d; k, \theta)\}$
9 BSS	binary signal strength	$b_0, b_1$	$g(d) = 1 - F\{-(b_0 + b_1 d)\}$
10 SS	signal strength	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1 d)]/s\}$
11 SSS	signal strength spherical	$\beta_0, \beta_1, \text{sdS}$	$g(d) = 1 - F\{[c - (\beta_0 + \beta_1(d-1) - 10 \log_{10} d^2)]/s\}$
14 HHN	hazard halfnormal	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\left(-\frac{d^2}{2\sigma^2}\right); g(d) = 1 - \exp(-\lambda(d))$
15 HHR	hazard hazard rate	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 (1 - \exp\{-(d/\sigma)^{-z}\}); g(d) = 1 - \exp(-\lambda(d))$
16 HEX	hazard exponential	$\lambda_0, \sigma$	$\lambda(d) = \lambda_0 \exp\{-(d/\sigma)\}; g(d) = 1 - \exp(-\lambda(d))$
17 HAN	hazard annular normal	$\lambda_0, \sigma, w$	$\lambda(d) = \lambda_0 \exp\left\{-\frac{(d-w)^2}{2\sigma^2}\right\}; g(d) = 1 - \exp(-\lambda(d))$
18 HCG	hazard cumulative gamma	$\lambda_0, \sigma, z$	$\lambda(d) = \lambda_0 \{1 - G(d; k, \theta)\}; g(d) = 1 - \exp(-\lambda(d))$

Similaridades com SCR (Royle & Gardner, 2011)

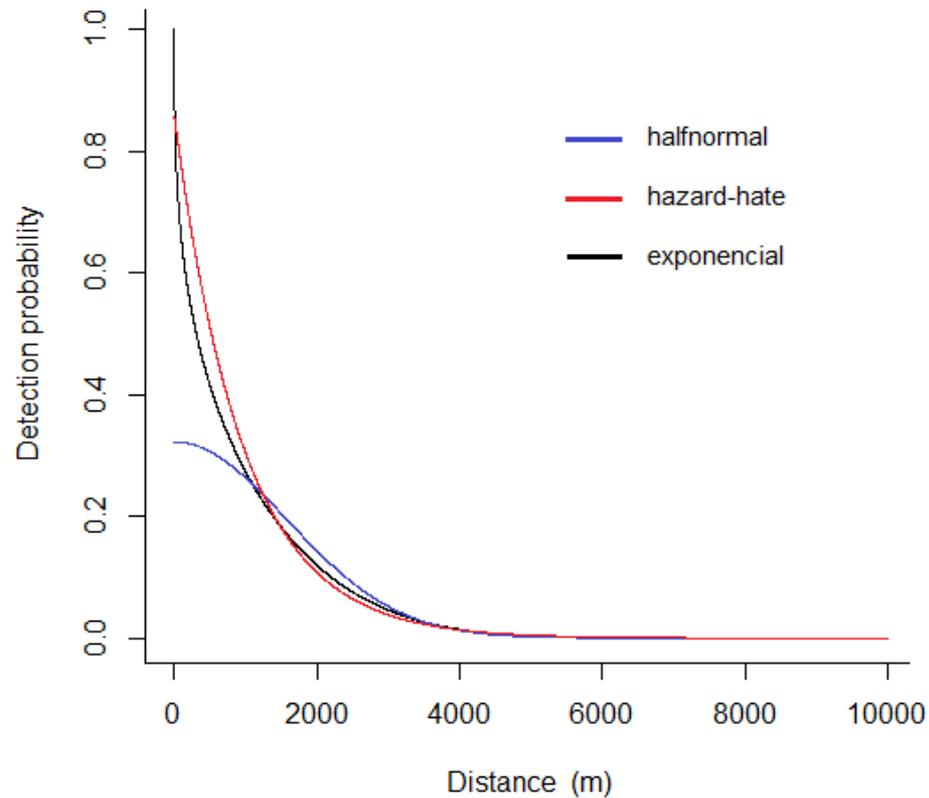
cumulative hazard = exposure function

hazard halfnormal = halfnormal exposure (- fator 2 no  $\sigma^2$ )

hazard exponential = exponential function

# SECR - detecção

Duas funções relacionando a probabilidade de detecção à distância ( $d$ ) e suas respectivas formas gráficas.



# SECR

## Ajusta o modelo “state” e o modelo observacional

a) Observacional  $\longrightarrow$  função de detecção dependente da distância ( $g_0$  ou  $\lambda$ ).

– Probabilidade de detecção de um indivíduo em um detector em relação à distância do centro de atividade

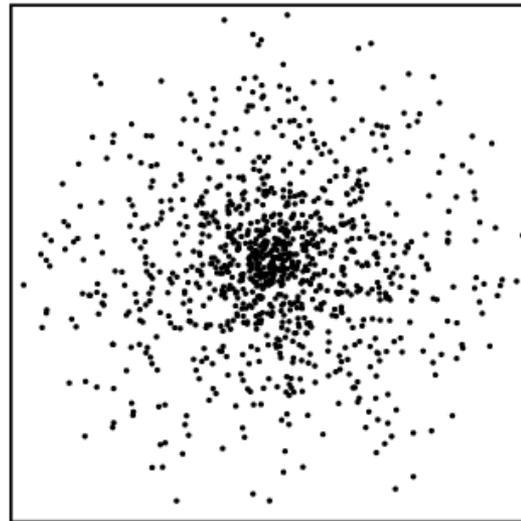
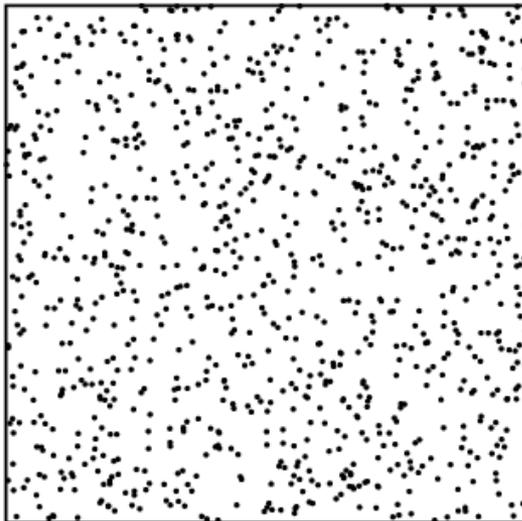
b) State  $\longrightarrow$  Processo de Poisson para distribuição espacial

-Homogêneo

(constante ao longo do espaço)

-Não uniforme

(variando ao longo do espaço)



# SECR - estrutura

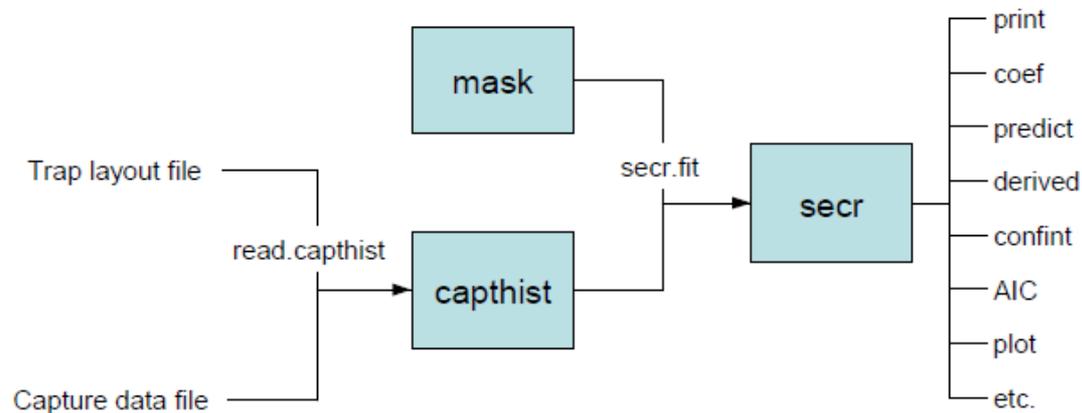


Figure 3: Essentials of the `secr` package. Each object class (shaded box) comes with methods to display and manipulate the data it contains (e.g. `print`, `summary`, `plot`, `rbind`, `subset`). The function `read.caphist` forms a `traps` object from the trap layout data and saves it as an attribute along with the capture data in a `caphist` object. If a habitat mask is not provided, one will be generated automatically by `secr.fit`. Any of the objects input to `secr.fit` may include a dataframe of covariates whose names may be used in a model formula. Fitted `secr` models may be further manipulated with the methods shown on the right. Additional functions (not shown) construct a regular detector array (e.g. `make.grid`, `make.circle`), form a mask from a `traps` object (`make.mask`), or simulate detection of a known population (`sim.caphist`).

## SECR – como trabalha

Grupo de classes do R (conjunto de dados e funções que devem ser manipulados – S3 classe)

Classes fundamentais são:

- ***traps*** (localização dos detectores e o tipo)
- ***capthist*** (histórico de detecção espacial, incluindo o objeto ***traps***)
- ***mask*** (pontos na área de máscara, área amostrada)
- ***secr*** (ajuste de modelo SECR)

Co-variáveis dos detectores, dos indivíduos e do habitat podem ser incluídas

## SECR - formato dos dados (*input*)

Duas tabelas (.txt ou .csv)

Capturas - no formato trapID

```
session,id,occasion,trapid, sex  
F1,1F,3,2,f  
F1,1F,3,3,f  
F1,2M,1,2,m  
F1,3F,6,2,f  
F1,4M,5,2,m  
F1,5F,1,8,f  
F1,5F,4,9,f  
F1,5F,6,8,f  
F1,5F,9,9,f  
F1,6M,3,5,m
```

(...)



Co-variável (hcov)

Detectores- (AF)

	A	B
1	detector,id,x,y	
2	1,TS1,358597,8914064	
3	2,TS2,360498,8913916	
4	3,TS3,362555,8913760	
5	4,TS4,362500,8912302	
6	5,TS6,359707,8911736	
7	6,TS7,357769,8910418	

(...)

# SECR - formato dos dados (*input*) – ARMADILHAS DE REGISTRO



Duas tabelas (.txt ou .csv)

Capturas - no formato trapID

Detectores- *proximity* (AF)

```

session,id,occasion,trapid, sex
F1,1F,3,2,f
F1,1F,3,3,f
F1,2M,1,2,m
F1,3F,6,2,f
F1,4M,5,2,m
F1,5F,1,8,f
F1,5F,4,9,f
F1,5F,6,8,f
F1,5F,9,9,f
F1,6M,3,5,m
(...)
    
```

	A	B
1	detector,id,x,y	
2	1,TS1,358597,8914064	
3	2,TS2,360498,8913916	
4	3,TS3,362555,8913760	
5	4,TS4,362500,8912302	
6	5,TS6,359707,8911736	
7	6,TS7,357769,8910418	

*make.caphist*

```

Session = F2
, , 1
    1 2 3 4 5 6 7 8
10M 0 0 0 0 0 0 0 0
11F 0 0 0 0 0 0 0 0
12M 0 0 0 0 0 0 0 0
13F 0 0 0 0 0 0 0 0
14M 0 0 0 0 0 0 0 0
15F 0 0 0 0 0 0 0 0
16F 0 0 0 0 0 0 0 0
17M 0 0 0 0 0 0 0 0
1F 0 0 0 0 0 0 0 0
2M 1 0 0 0 0 0 0 0
3F 0 0 1 0 1 0 0 0
6M 0 0 0 0 0 0 0 0
7F 0 0 0 0 0 0 0 0
8M 0 0 0 0 0 0 0 0
9F 0 0 0 0 0 0 0 0

, , 2
    1 2 3 4 5 6 7 8
10M 0 0 0 0 0 0 0 0
11F 0 0 0 0 0 0 0 0
12M 0 0 0 0 0 0 0 0
13F 0 0 0 0 0 0 0 0
14M 0 0 0 0 0 0 0 0
15F 0 0 0 0 0 0 0 0
16F 0 0 0 0 0 0 0 0
17M 0 0 0 0 0 0 0 0
1F 0 0 0 0 0 0 0 0
2M 1 0 1 0 0 0 0 0
3F 0 0 0 0 0 0 0 0
6M 0 0 0 0 0 0 0 0
7F 0 0 0 0 0 0 0 0
8M 0 0 1 0 0 0 0 0
9F 0 0 0 0 0 0 0 0

, , 3
    1 2 3 4 5 6 7 8
10M 0 1 0 0 1 1 0 0
11F 0 0 0 0 0 0 0 0
12M 0 0 0 0 0 0 0 0
13F 0 0 0 0 0 0 0 0
14M 0 0 0 0 0 0 0 0
15F 0 0 0 0 0 0 0 0
16F 0 0 0 0 0 0 1 1
17M 0 0 0 0 0 0 0 0
1F 0 1 0 0 0 0 0 0
2M 0 0 0 0 0 0 0 0
3F 0 0 0 0 0 0 0 0
6M 0 0 0 0 0 0 0 0
7F 0 0 0 0 0 0 0 0
8M 0 0 0 0 0 0 0 0
9F 0 0 0 0 0 0 0 0

, , 4
    1 2 3 4 5 6 7 8
10M 0 0 0 0 1 1 0 0
11F 0 0 0 0 0 0 0 0
12M 0 0 0 0 0 0 1 0
13F 0 0 0 0 0 0 0 0
14M 0 0 0 0 0 0 0 0
15F 0 0 0 0 0 0 0 0
16F 0 0 0 0 0 0 0 0
17M 0 0 0 0 0 0 0 0
1F 0 0 0 0 0 0 0 0
2M 0 0 0 0 0 0 0 0
3F 0 0 0 0 0 0 0 0
6M 0 0 0 0 0 0 0 0
7F 0 0 0 0 0 0 0 0
8M 0 0 0 0 0 0 0 0
9F 0 0 0 0 0 0 0 0

, , 10
    1 2 3 4 5 6 7 8
10M 0 0 1 0 0 0 0 1
11F 1 0 0 0 0 0 0 0
12M 0 0 0 0 0 0 0 0
13F 0 0 0 0 0 0 0 0
14M 0 0 0 0 0 0 0 0
15F 0 0 0 1 1 0 0 0
16F 0 0 0 0 0 0 0 0
17M 0 0 0 0 0 1 0 0
1F 0 0 0 0 0 0 0 0
2M 0 0 0 0 0 0 0 0
3F 0 0 0 0 0 0 0 0
6M 0 0 0 0 0 0 0 0
7F 1 0 0 0 1 0 0 0
8M 0 0 0 0 0 0 0 0
9F 1 1 0 0 1 0 1 0
    
```

5 tabelas omitidas  
(...)

# SECR - formato dos dados (*input*) – ARMADILHAS DE CAPTURA

Duas tabelas (.txt ou .csv)

Capturas - no formato trapID

Detectores- *multi* (*live traps*)



```
session,id,occasion,trapid, sex
F1,1F,3,2,f
F1,1F,3,3,f
F1,2M,1,2,m
F1,3F,6,2,f
F1,4M,5,2,m
F1,5F,1,8,f
F1,5F,4,9,f
F1,5F,6,8,f
F1,5F,9,9,f
F1,6M,3,5,m
```

(...)

	A	B
1	detector,id,x,y	
2	1,TS1,358597,8914064	
3	2,TS2,360498,8913916	
4	3,TS3,362555,8913760	
5	4,TS4,362500,8912302	
6	5,TS6,359707,8911736	
7	6,TS7,357769,8910418	

(...)

*make.caphist*

```
Session = F2
      1  2  3  4  5  6  7  8
10M  0  3  9  0  3  3  9 10
11F 10  0  0  0  0  0  0  0
12M  0  0  0  0  0  0  4  0
13F  8  0  0  0  0  0  0  0
14M  0  8  0  0  7  0  0  0
15F  0  0  0 10 10  0  0  0
16F  0  0  0  0  0  0  3  3
17M  0  0  0  0  0 10  0  0
1F   0  3  0  0  0  0  0  0
2M   1  0  2  0  0  0  0  0
3F   0  0  1  0  1  0  0  0
6M   0  0  5  0  5  0  0  0
7F  10  0  0  0 10  0  0  0
8M   0  0  2  0  0  0  0  7
9F   9 10  0  0 10  0 10  0
```



# SECR – preditores dos modelos

Table 4: Automatically generated predictor variables used in detection models

	Variable	Description	Notes
grupo	← g	group	interaction of the capthist individual covariates listed in argument <code>groups</code> of <code>secr.fit</code>
tempo	t	time factor	one level for each occasion
	T	time trend	linear trend over occasions on link scale
indivíduo	b	learned response	step change after first detection
	B	transient response	depends on detection at preceding occasion (Markovian response)
	bk	animal x site response	site-specific step change
	Bk	animal x site response	site-specific transient response
local	k	site learned response	site effectiveness changes once any animal caught
	K	site transient response	site effectiveness depends on preceding occasion
temporadas classe	← session	session factor	one level for each session
	Session	session trend	linear trend on link scale
	← h2	2-class mixture	finite mixture model with 2 latent classes

# SECR – argumentos dos modelos

Table 6: Some examples of the `model` argument in `secr.fit`

Model	Description
<code>g0 ~ 1</code>	<code>g0</code> is constant across animals, occasions and detectors
<code>g0 ~ b</code>	learned response affects <code>g0</code>
<code>list(g0~b, sigma~b)</code>	learned response affects both <code>g0</code> and <code>sigma</code>
<code>g0 ~ h2</code>	2-class finite mixture for heterogeneity in <code>g0</code>
<code>g0 ~ b + T</code>	learned response in <code>g0</code> combined with trend over occasions
<code>sigma ~ g</code>	detection scale <code>sigma</code> differs between groups
<code>sigma ~ g*T</code>	group-specific trend in <code>sigma</code>
<code>D ~ cover</code>	density varies with 'cover' given in <code>covariates(mask)</code>
<code>list(D~g, g0~g)</code>	both density and <code>g0</code> differ between groups
<code>D ~ session</code>	session-specific density

# SECR – concorrência entre os modelos

## Critério de Informação de Akaike – AIC

### Importante:

Quando utilizar modelos:

- ajustado pela *conditional likelihood* (CL=TRUE);
- de mistura de classes (*hybrid mixture models*) hcov;
- utilizando argumento de grupo;
- multi-session ;

Comparar com outros que possuem as mesmas informações.

## SECR – ajuste dos modelos

- Ajustados numericamente maximizando a estimativa da verossimilhança
- A probabilidade envolve a integração sobre os centros de atividade desconhecidos dos indivíduos (somatória dos pontos na área de máscara)
- Algoritmo de maximização utilizado é o Newton-Raphson (`state::nlm`)

# SECR - *output*

```
> fit.0
```

```
secr.fit(capthist = ocelot1.data, mask = HabitatMask, hcov = "sex",  
         trace = FALSE)  
secr 2.9.3, 09:20:59 26 fev 2015
```

```
Detector type      proximity  
Detector number    11  
Average spacing    2014.6 m  
x-range            357769 366270 m  
y-range            8908010 8914064 m  
N animals          : 10  
N detections       : 29  
N occasions        : 9  
Mask area          : 24592.48 ha  
  
Model              : D~1 g0~1 sigma~1 pmix~h2  
Mixture (hcov)    : sex  
Fixed (real)      : none  
Detection fn       : halfnormal  
Distribution       : poisson  
N parameters       : 4  
Log likelihood     : -102.9075  
AIC                : 213.8151  
AICc               : 221.8151
```

## Beta parameters (coefficients)

	beta	SE.beta	lcl	ucl
D	-7.146798e+00	0.3753632	-7.882497	-6.4111000
g0	-6.858587e-01	0.5190152	-1.703110	0.3313923
sigma	7.403547e+00	0.1684468	7.073397	7.7336963
pmix.h2m	-1.969780e-06	0.6324558	-1.239593	1.2395887

## Variance-covariance matrix of beta parameters

	D	g0	sigma	pmix.h2m
D	1.408975e-01	-3.114667e-02	-3.083344e-02	6.238620e-08
g0	-3.114667e-02	2.693767e-01	-1.203667e-02	1.354182e-07
sigma	-3.083344e-02	-1.203667e-02	2.837432e-02	-2.436886e-08
pmix.h2m	6.238620e-08	1.354182e-07	-2.436886e-08	4.000004e-01

## Fitted (real) parameters evaluated at base levels of covariates

session = F1, h2 = f

	link	estimate	SE.estimate	lcl	ucl
D	log	7.873809e-04	3.062767e-04	3.772899e-04	1.643216e-03
g0	logit	3.349550e-01	1.156159e-01	1.540596e-01	5.820981e-01
sigma	log	1.641797e+03	2.785289e+02	1.180150e+03	2.284029e+03
pmix	logit	5.000005e-01	1.581140e-01	2.245076e-01	7.754931e-01

session = F1, h2 = m

	link	estimate	SE.estimate	lcl	ucl
D	log	7.873809e-04	3.062767e-04	3.772899e-04	1.643216e-03
g0	logit	3.349550e-01	1.156159e-01	1.540596e-01	5.820981e-01
sigma	log	1.641797e+03	2.785289e+02	1.180150e+03	2.284029e+03
pmix	logit	4.999995e-01	1.581140e-01	2.245069e-01	7.754924e-01

# SECR

