

# Model complexity and model choice for animal movement models

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# Outline

- 1 Animal movement
- 2 Florida panthers
- 3 Hidden Markov models
- 4 Basic analysis (van de Kerk et al., 2015)
- 5 Incorporating diurnal variation (Li, 2015)
- 6 Broader issues/outlook

# Acknowledgements

**People** Michael Li, Madelon van de Kerk, Dave Onorato,  
Madan Oli

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National Park Service

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# To do

- pix!
- general REFS: Turchin, Morales et al, Langrock ... ?

# Animal movement: data

- observations:  
e.g. mass mark-recapture,  
longitudinal density, direct  
observation, telemetry (sequence pix)  
(VHF, GPS)
- most methods provide a  
sequence of times and  
locations for each individual

## ● summaries:

- home range  
(convex hull, kernel density estimate, etc.)
- root-mean-squared displacement
- step length and turning angle

(step length/turning angle pix)

## ● covariates:

e.g. habitat map,  
individual characteristics  
(sex, age, weight ...)

# Animal movement: questions

- simple description
- how do animals' movements change as a function of their (internal or external) environment?  
what does that tell us about their biology?
- how might animals' distributions, etc. change when conditions (density, habitat, ...) change?

# Biological/conservation issues

- Florida panther: *Puma concolor coryi*
- endangered subspecies
- severely reduced habitat
- small, isolated population
- currently recovering



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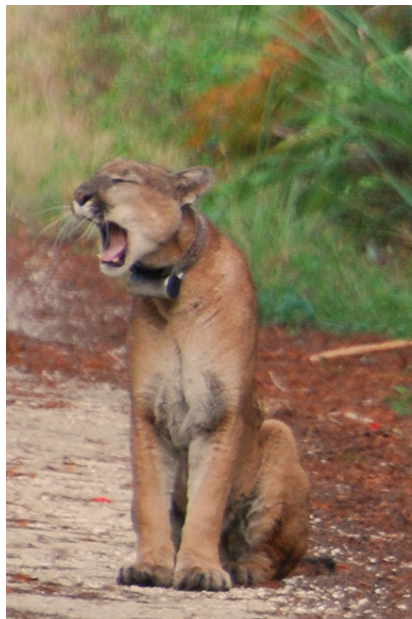
# Panther movement questions

- movement variation by sex and life history stage (juvenile, adult, mom with kittens . . . )
- effects of movement on threats (intraspecific aggression, roadkill) ?
- predicting the effects of future changes in population density / population structure / habitat

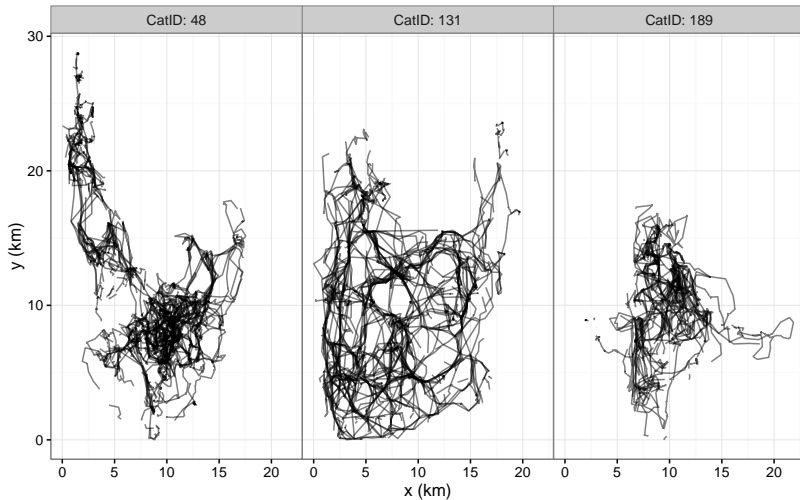


# Panther movement data

- panthers tracked, captured
- GPS collars
- 18 males (13 male, 5 female, 1-15 years old)
- 3200 panther days, hourly/bihourly; 49000 locations
- ?? per panther

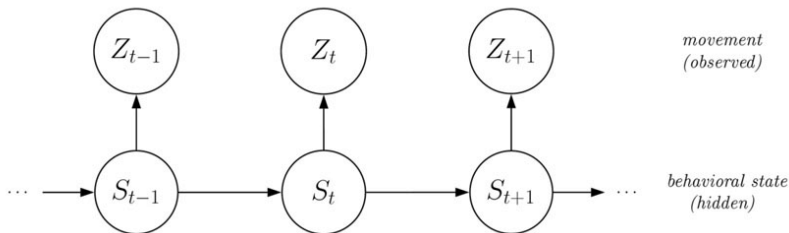


# example movement tracks



# Hidden Markov models

- finite mixture model with temporal dependence
- discrete time steps
- discrete latent state; *transition matrix*
- observations from *emission distributions*  
(continuous or discrete, univariate or multivariate)
- **multiphasic movement** (Fryxell et al., 2008; Langrock et al., 2012)



# Hidden Markov models (cont.)

**state:**

$$S_t \sim \text{Multinomial}(S_{t-1}, \mu_{S,t})$$
$$\mu_{S,t} = \text{multi-logistic}(\mathbf{X}_{S,t} \boldsymbol{\beta}_S)$$

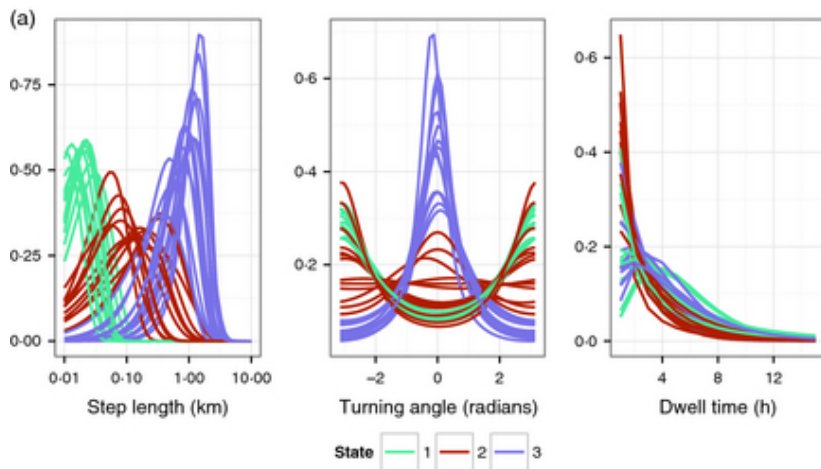
**emission:**

$$\mathbf{Z}_t \sim \{\text{Dist}_1(\mu_{Z_1,S_t}), \dots, \text{Dist}_n(\mu_{Z_n,S_t})\}$$
$$\mu_{Z_i,S_t} = g^{-1}(\mathbf{X}_{Z_i,t} \boldsymbol{\beta}_{Z_i,S_t})$$

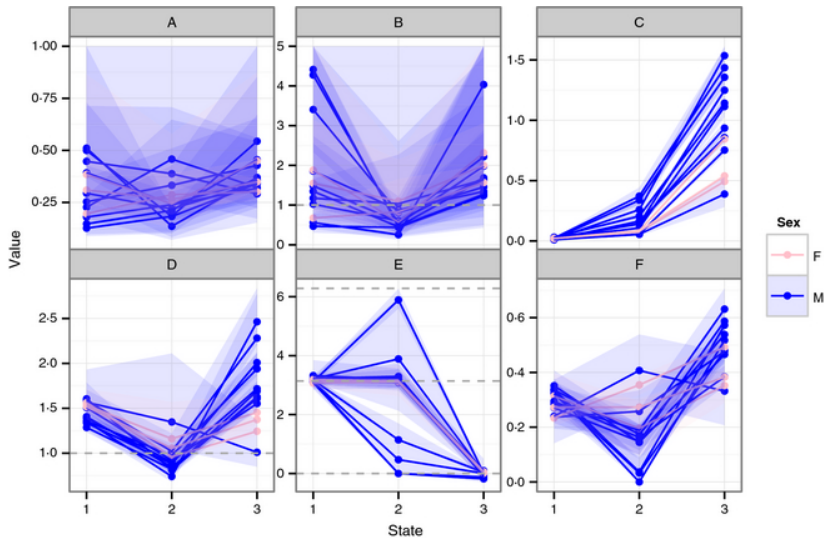
# Hidden Markov models (part 3)

- *forward-backward algorithm* for estimating parameters
- *Viterbi algorithm* for estimating most probable state sequences
- depmixS4 package (Visser and Speekenbrink, 2010) (also moveHMM (Michelot et al., 2016))
- hidden *semi-Markov* models: allow for non-geometric *dwell distributions* (Langrock, 2011; Augustine, 2016): move.HMM

# State distributions

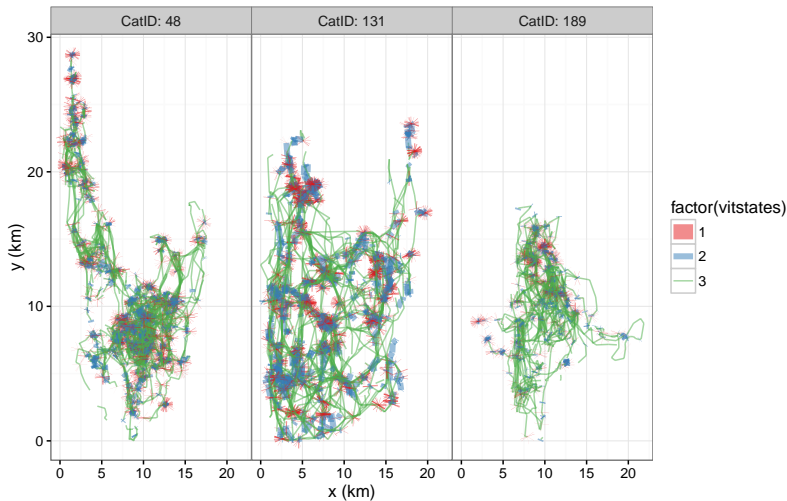


# Parameter estimates





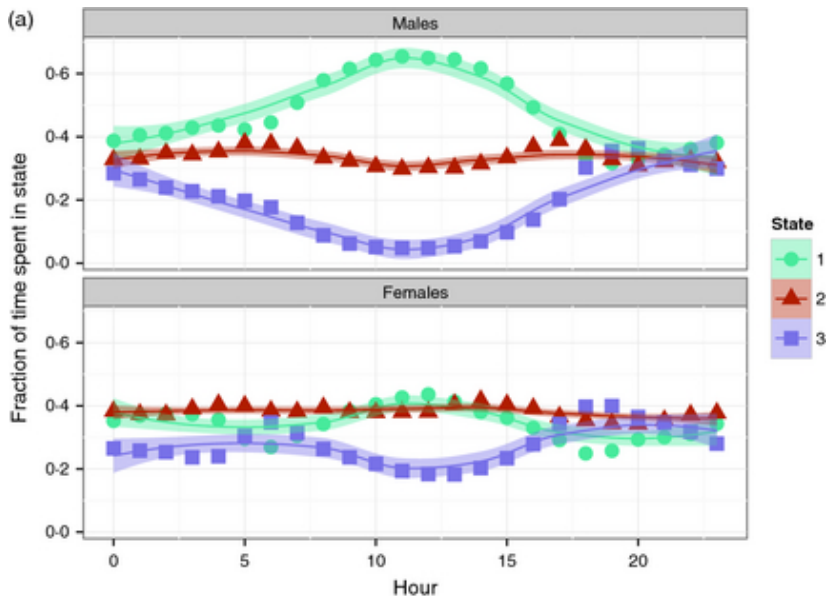
# Tracks with Viterbi estimates



# Transition parameters

picture/table here of transition parameters (network diagram??)

# Diurnal variation



# what can we conclude so far?

## good news

- basic biology: males move faster, farther
- three states are identifiable, sensible
- ...

## bad news

- diurnal variation in Viterbi results - but it's not in the model!
- estimates of model complexity are too high

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Animal movement  
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Panthers  
○○○○

HMM  
○○○

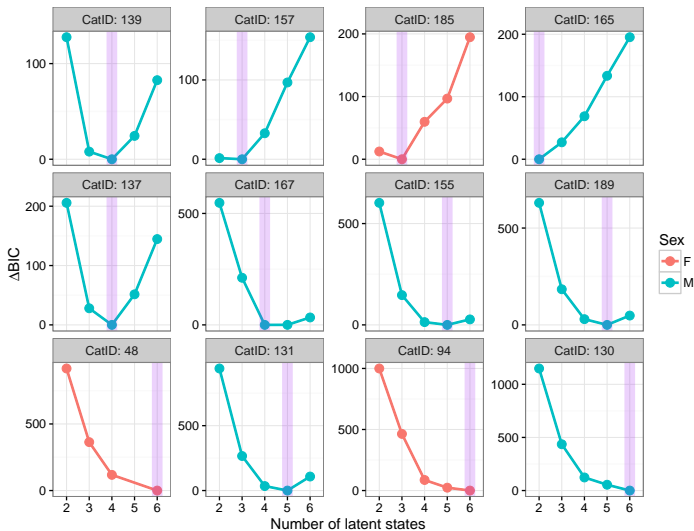
Basic analysis  
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Diurnal model  
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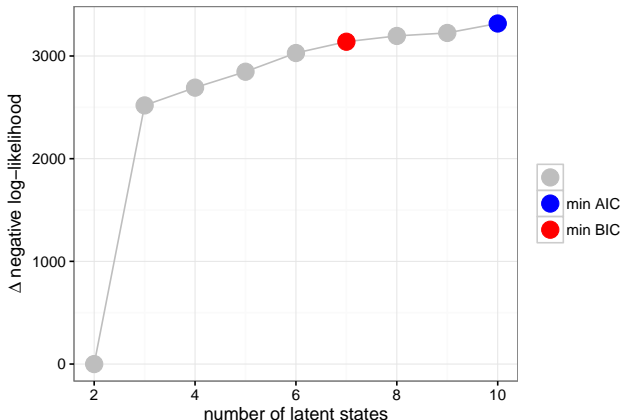
Broader issues/outlook  
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References

# Model complexity



# Model complexity (Manx shearwaters, (Dean et al., 2013))



# Broadening the model

Attempting to fix these problems:

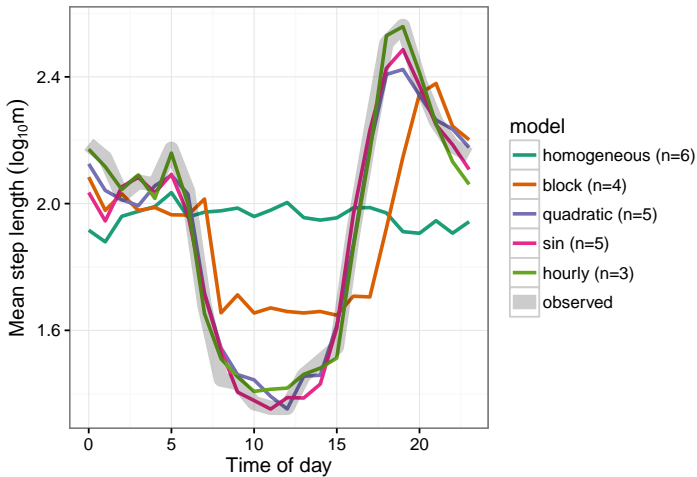
- extend the model to allow covariates
- specifically, allow for diurnal variation
  - simplify model (log-Normal step length only)
  - *fixed* state-specific emissions parameters (step length mean and std dev)
  - time-varying transition parameters
  - also try *finite mixture models* (independent occupancy)
- how much does this help?



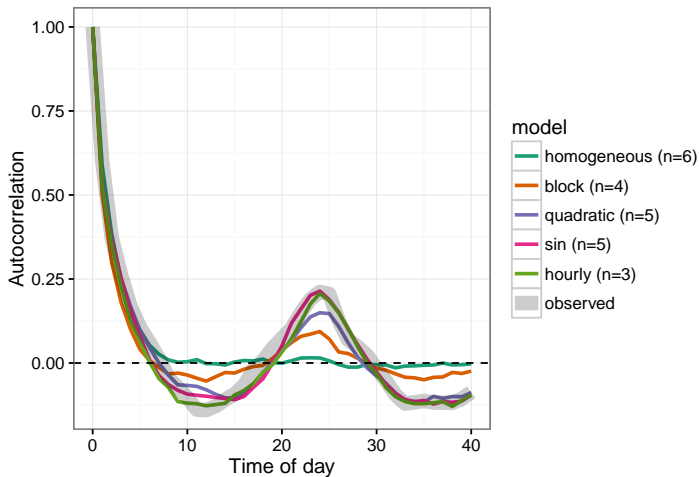
# Temporal models

figure showing alternative temporal models

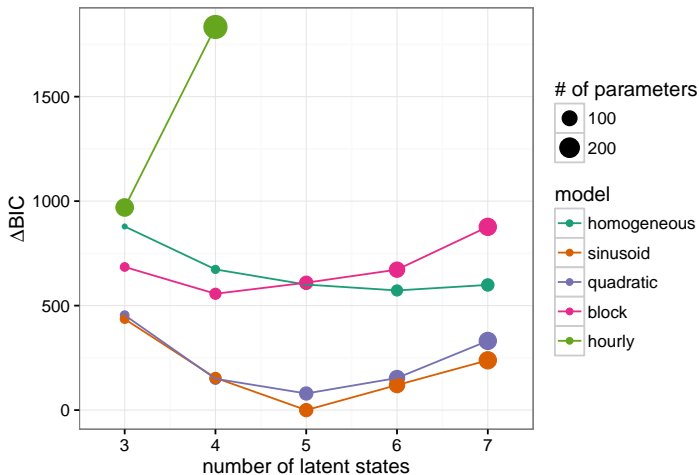
# Temporal patterns (step length)



# Temporal patterns (autocorrelation)



# Goodness of fit/model complexity



# Model complexity

Figure showing BIC plots for all cats tried

Animal movement  
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Panthers  
○○○○

HMM  
○○○

Basic analysis  
○○○○○○○○

**Diurnal model**  
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Broader issues/outlook  
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References

# Model complexity

Simulation results

# Animal movement: open challenges

- Cognition
- Intraspecific interaction/collective movement
- Continuous-time movement models
- Edges, barriers, and corridors
- Efficient (big-data) approaches

# Big data and small models

- simple model families +  
model misspecification →  
overparameterization



Animal movement  
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Panthers  
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HMM  
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Basic analysis  
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Diurnal model  
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**Broader issues/outlook**  
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References

# an aside on AIC vs BIC

# Tools needed

- cross-validation (Wenger and Olden, 2012)
- diagnostic plots
- score tests?

