

Global population dynamics and climate change: Comparing species-level impacts on two contrasting large mammals

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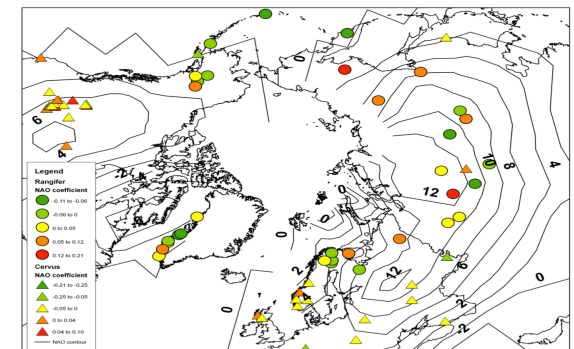
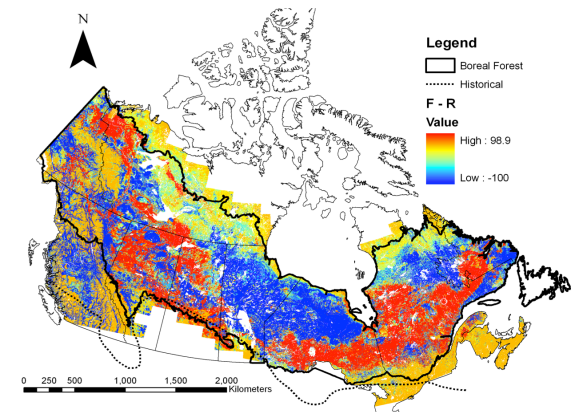
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Q1: How do landuse and climatic change affect species' niche throughout their ranges?

Q2: How does climatic influence vary across a species' range and globally after accounting for biotic interactions?

Q3: How does the spatial scale of population synchrony vary with climate change?





Cervus vs. Rangifer



Similarities

- Generalists
- Circumpolar
- Ruminants dependent on spring phenology

- Winter snow accumulation negatively affects elk (Creel & Creel 2009)

- Fall and winter - prefer areas of high snowfall
- Summer - prefer cooler areas corresponding to lower prevalence of insects
- Trophic miss-match contributing to Caribou populations declines (Sharma et al. 2009; Vors & Boyce 2009)

State-space models of globally distributed *Cervus* and *Rangifer* populations offer insights into predictors of error

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Large herbivore population time-series

- Accurately counting open populations of large herbivores often impossible
- Estimates of abundance have inherent noise
- Two major sources of noise:
 - Observer error
 - Process variation

Bayesian state-space models

- MCMC hierarchical models
- Decompose time-series:
 - Growth parameters
 - Observation error
 - Process variation

Absence of studies analyzing predictors of process & observation error

Goals

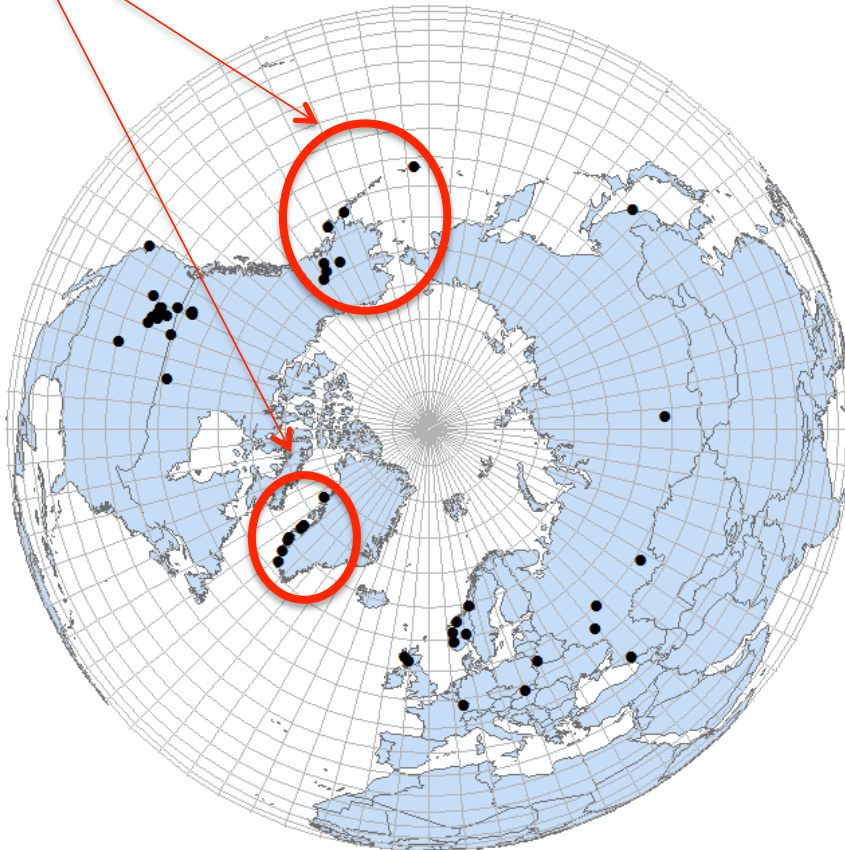
1. Examine variation in magnitude of process and observation error among and between *Cervus* and *Rangifer* populations
2. Quantify contributions by ecological and methodological variables to process variation and observation error within and among *Cervus* and *Rangifer*

Goals

1. Examine variation in magnitude of process and observation error among and between *Cervus* and *Rangifer* populations

Globally distributed populations

34 *Cervus* populations
13 *Rangifer* populations



Time-series	<i>Cervus</i>	<i>Rangifer</i>
No. of years	12-46 ($\mu = 30$)	15-74 ($\mu = 45$)
% of missing years	0.03 - 0.22% (12 pops.)	0.04 - 0.43% (6 pops.)

State-space models

$$X_{i,t} = \beta_{i,0} + (1 + \beta_{i,1})X_{i,t-1} + \varepsilon_{i,t} \quad (1)$$

$$Y_{i,t} \sim N(X_{i,t}, \tau_{i,t}) \quad (2)$$

$X_{i,t}$: time-series of \log_e true abundances

$\beta_{i,0}$: rate of intrinsic population growth

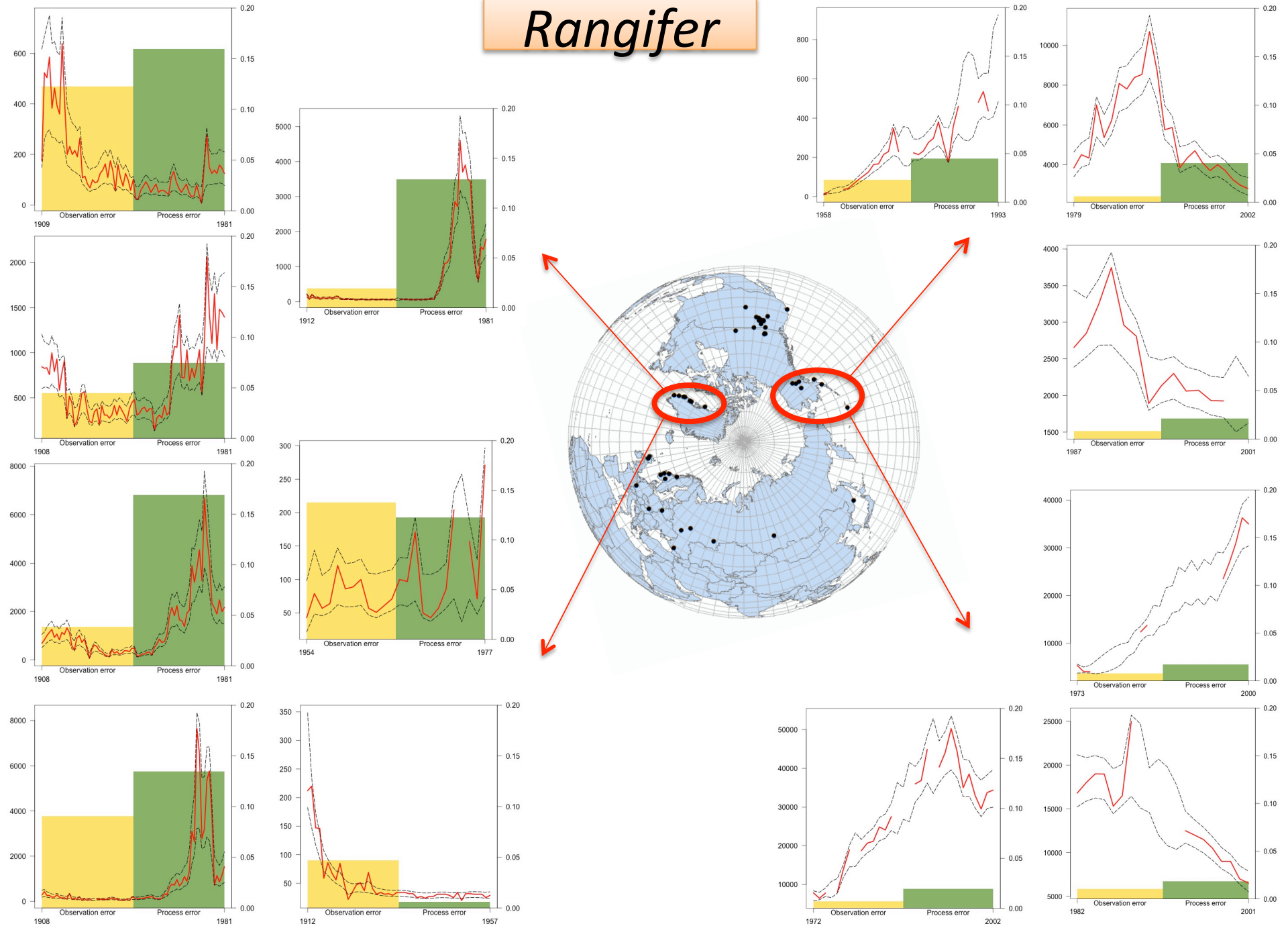
$\beta_{i,1}$: strength of direct density dependence

$\varepsilon_{i,t}$: process stochasticity

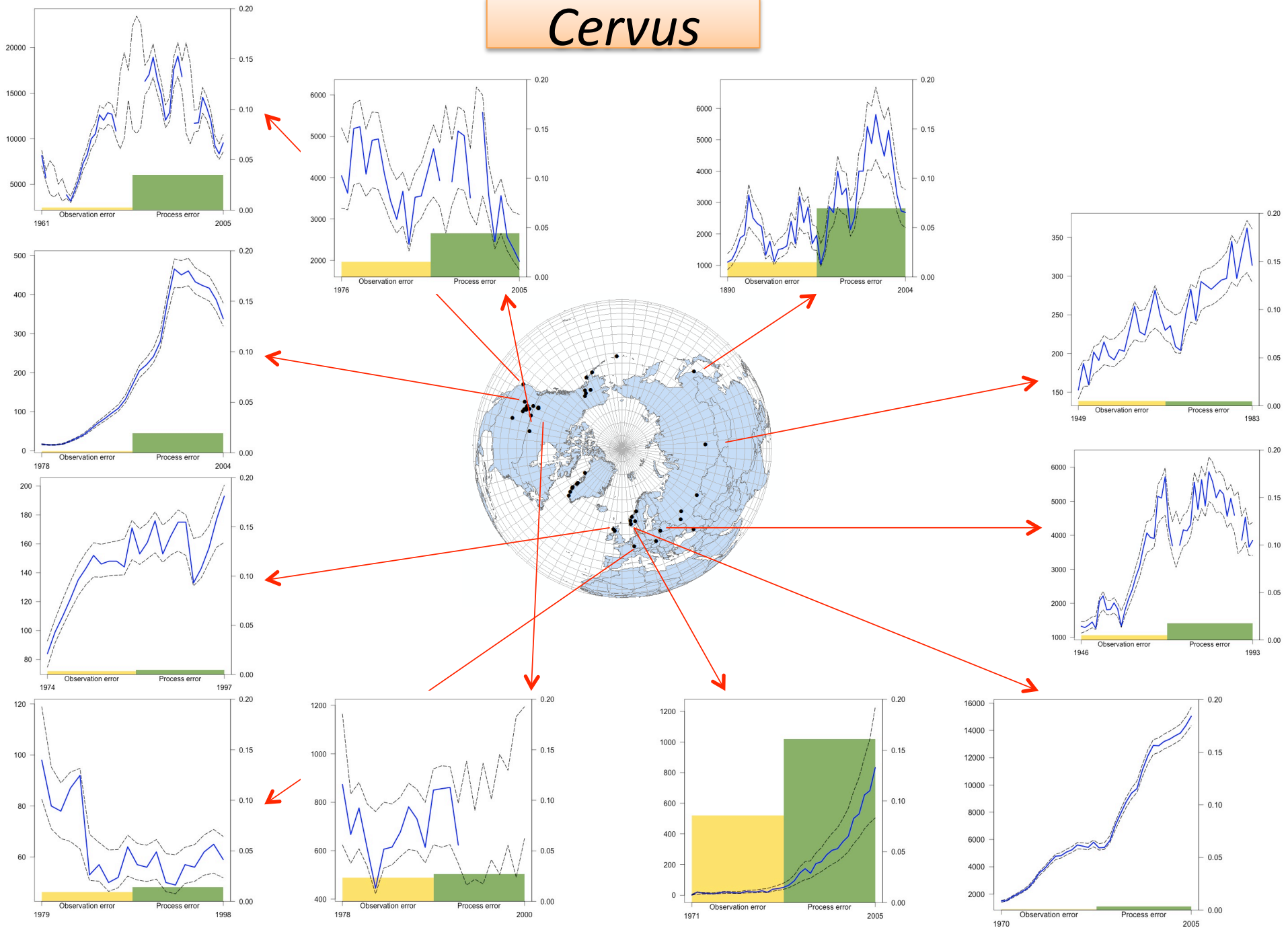
$Y_{i,t}$: observations

$\tau_{i,t}$: observation error

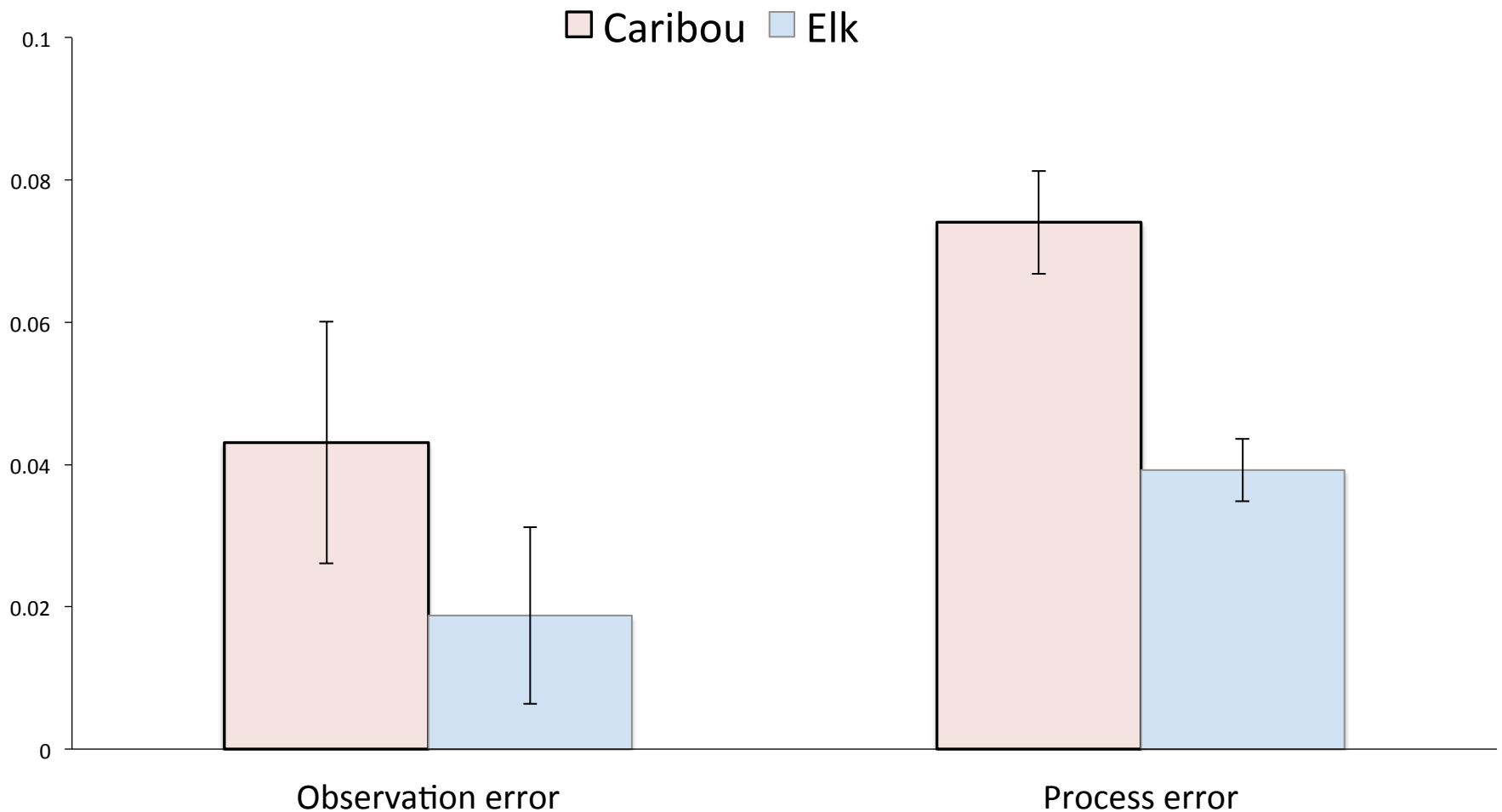
Rangifer



Cervus



Contrast in error between *Cervus* and *Rangifer* populations



Goals

2. Quantify contributions by ecological and methodological variables to process variation and observation error within and among *Cervus* and *Rangifer*

Predictors

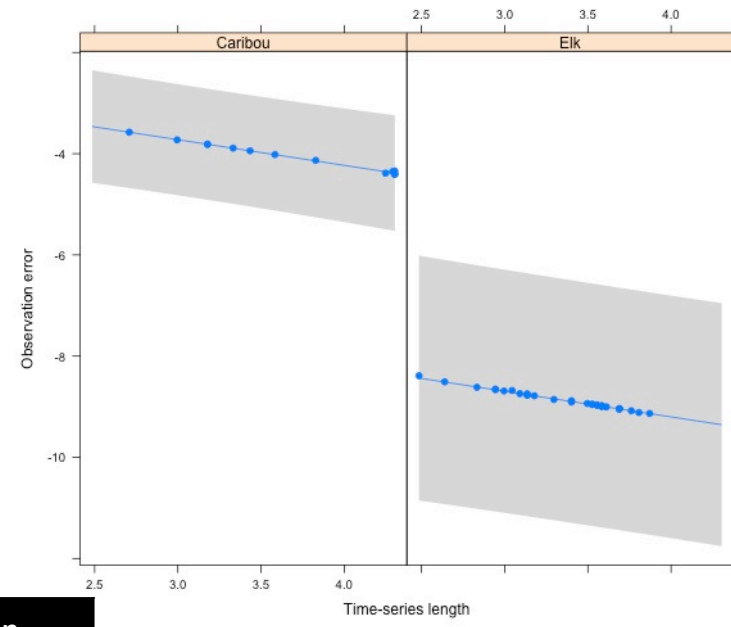
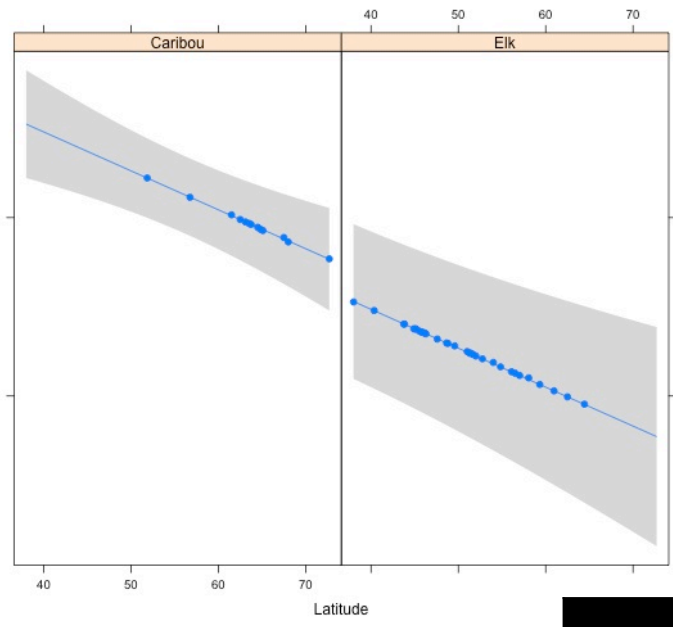
Methodological

- species
- time series length
- proportion of missing data

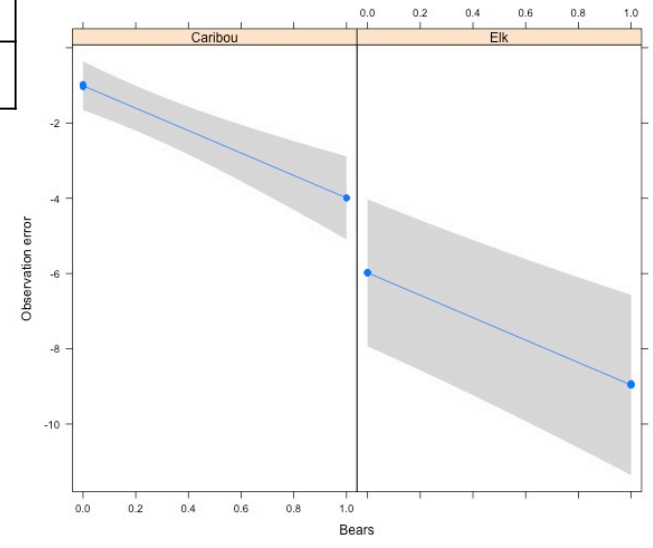
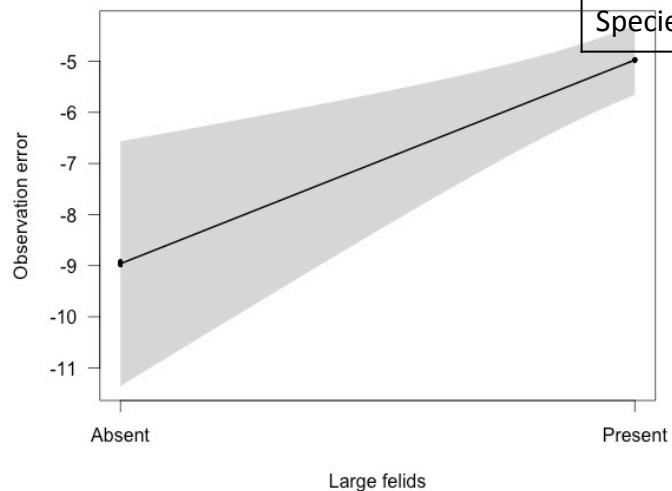
Ecological

- latitude
- hunting
- predation: bears, large felids and wolves

Observation error

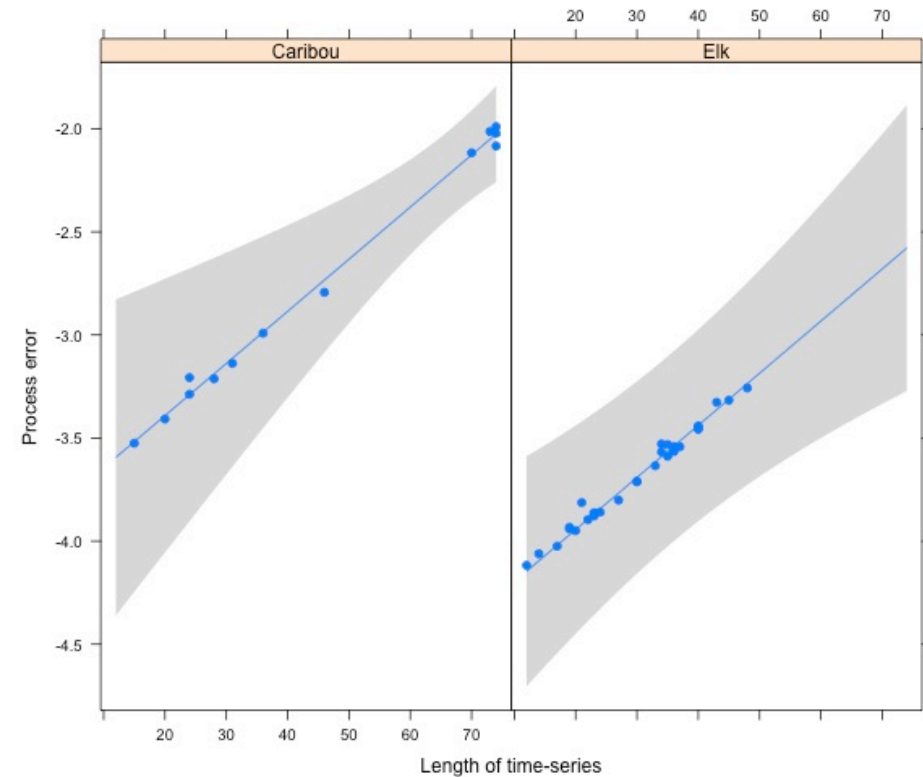
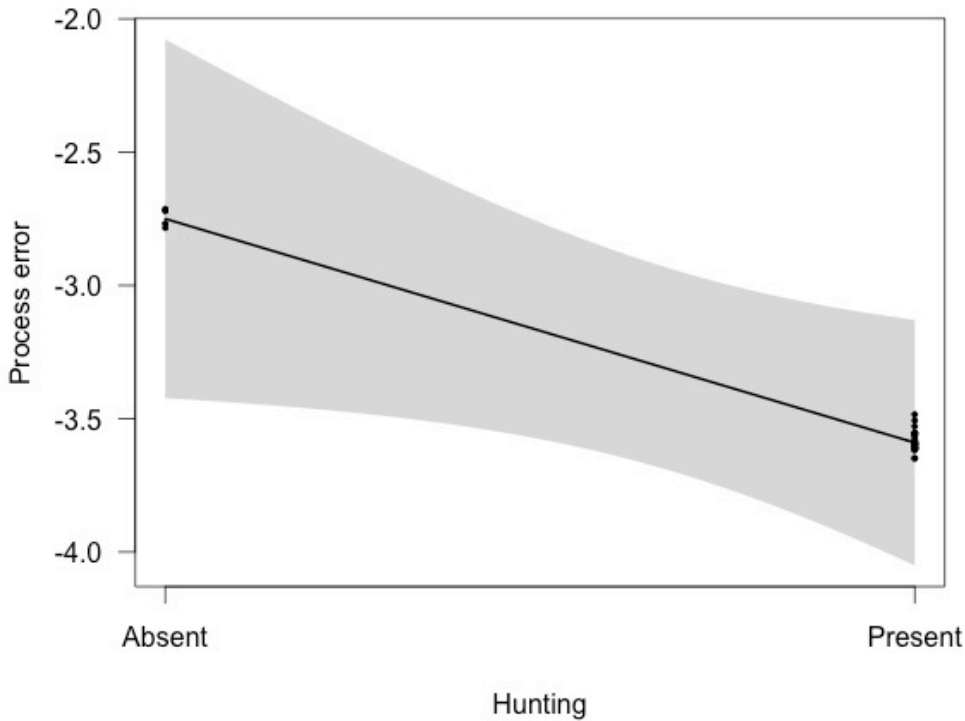


	Estimate	Std. Error	p
Latitude	-0.1	0.03	< 0.001
log _e (Years)	-0.51	0.14	< 0.001
Large felids	3.98	1.16	0.001
Bears	-2.98	0.6	< 0.001
Species	-4.97	1.14	< 0.001



Process error

	Estimate	Std. Error	p
Length of time-series	0.03	0.01	< 0.001
Species	-0.55	0.34	0.11
Hunting	-0.84	0.42	0.05



Conclusions

- Process variation higher than observation error for most populations
- Error higher in *Rangifer* than *Cervus*
- Higher number of significant predictors of observation error

Open questions

- Does high process variation imply non-optimal population models?
- How do you correct/treat time-series with high observation error?