

Trying to understand albatross growth with Dynamic Energy Budget Theory



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Thanks to

**Carlos Teixeira,
Sarah Burgan,
BAS Archives,
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and the
GRS chairs**



PLR-1341649 (Leah Johnson PI, Sadie Ryan CoI)
Quantifying how bioenergetics and foraging determine
population dynamics in threatened Antarctic albatrosses

9 TIPS FOR LIVING WITH LESS PLASTIC



1

Bring your own shopping bag



2

Carry a reusable water bottle



3

Bring your own cup



4

Pack your lunch in reusable containers



5

Say no to disposable straws & cutlery



6

Skip the plastic produce bags



7

Slow down and dine in



8

Store leftovers in glass jars



9

Share these tips with your friends

How can we untangle intrinsic and extrinsic drivers of individual growth, reproduction, and survival in a changing environment?

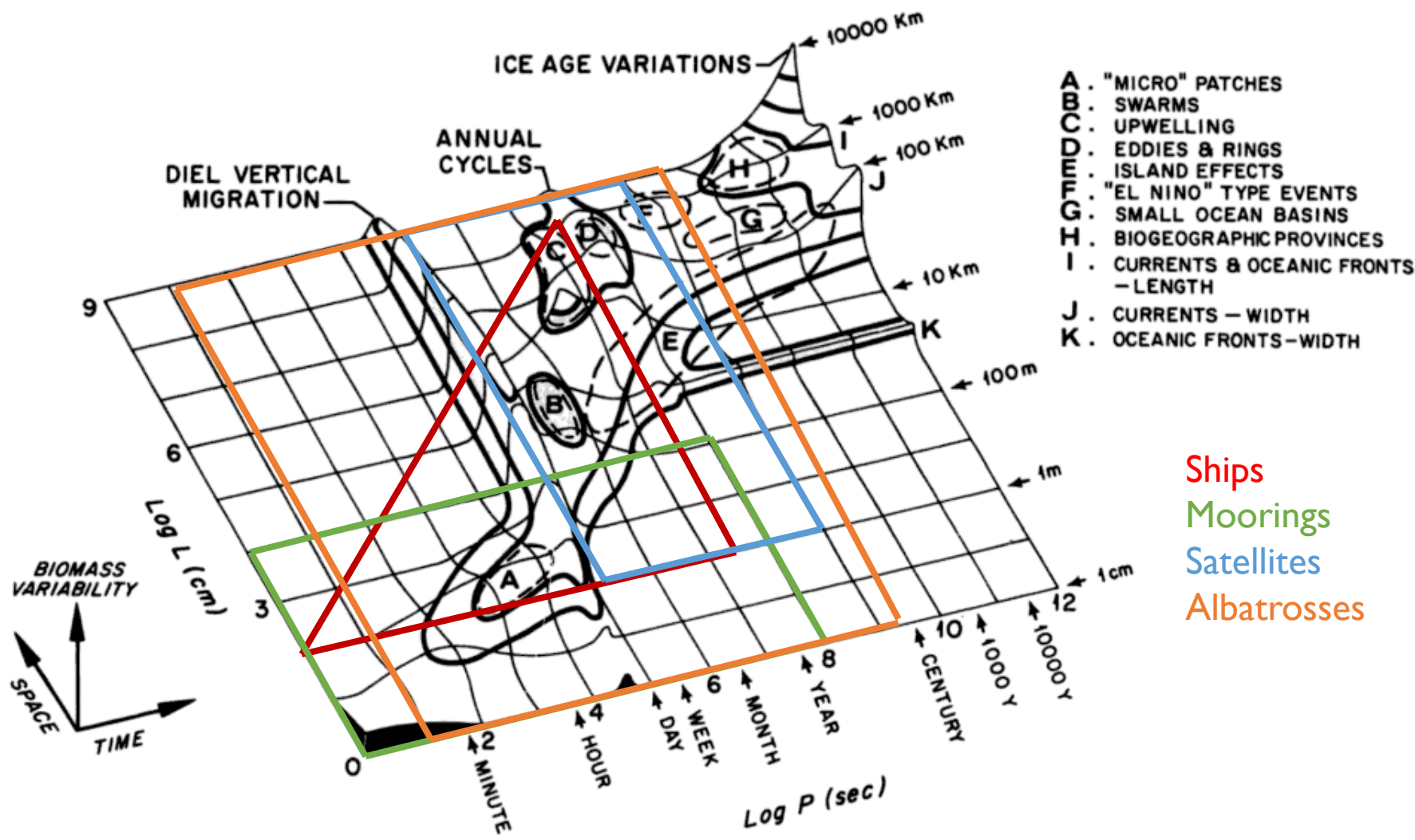


Albatrosses as a model system

- Large, long-lived seabirds
- Efficient fliers, travel very large distances
- 22 species threatened or endangered
- Many populations declining
 - Fisheries bycatch
 - Invasive predators
 - Extreme weather events
 - Climate variation
- We know much about drivers of population dynamics from past demography, but what about the future?



Diversion I: Scales of open ocean ecology





Goal:

Leverage DEB as a **mechanistic model** of energy acquisition and allocation (and thus survival).

Potential drivers:

- « Food quality
- « Food quantity
- « Feeding rate
- « Climate/Weather
- « Species-specific physiology

Data rich species:

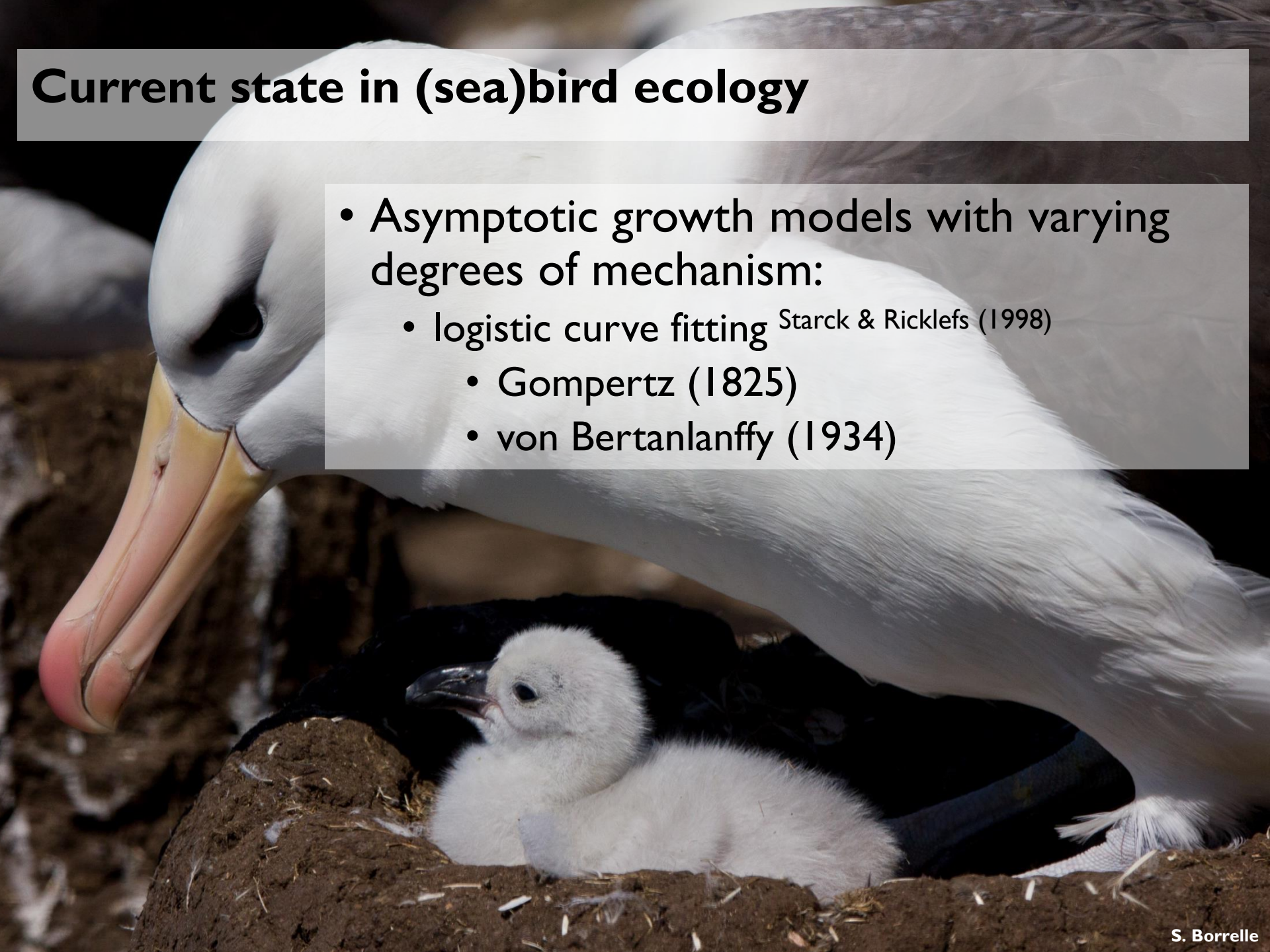
Wandering albatross *D. exulans*

Black-browed albatross *T. melanophris*

Grey-headed albatross *T. chrysostoma*

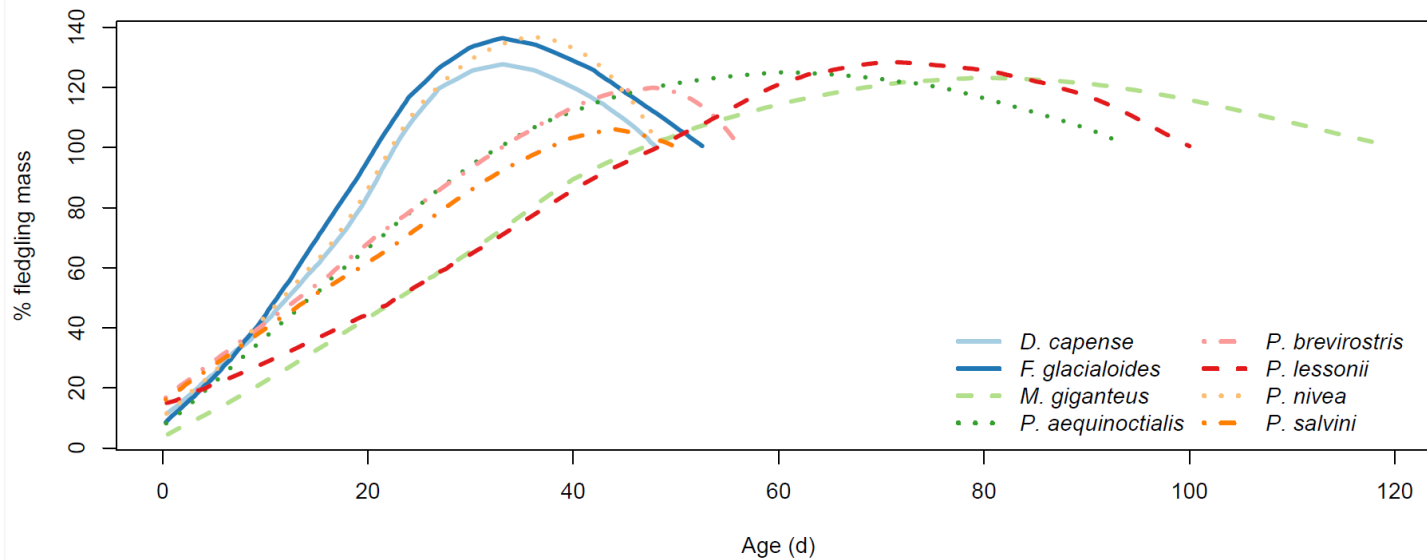
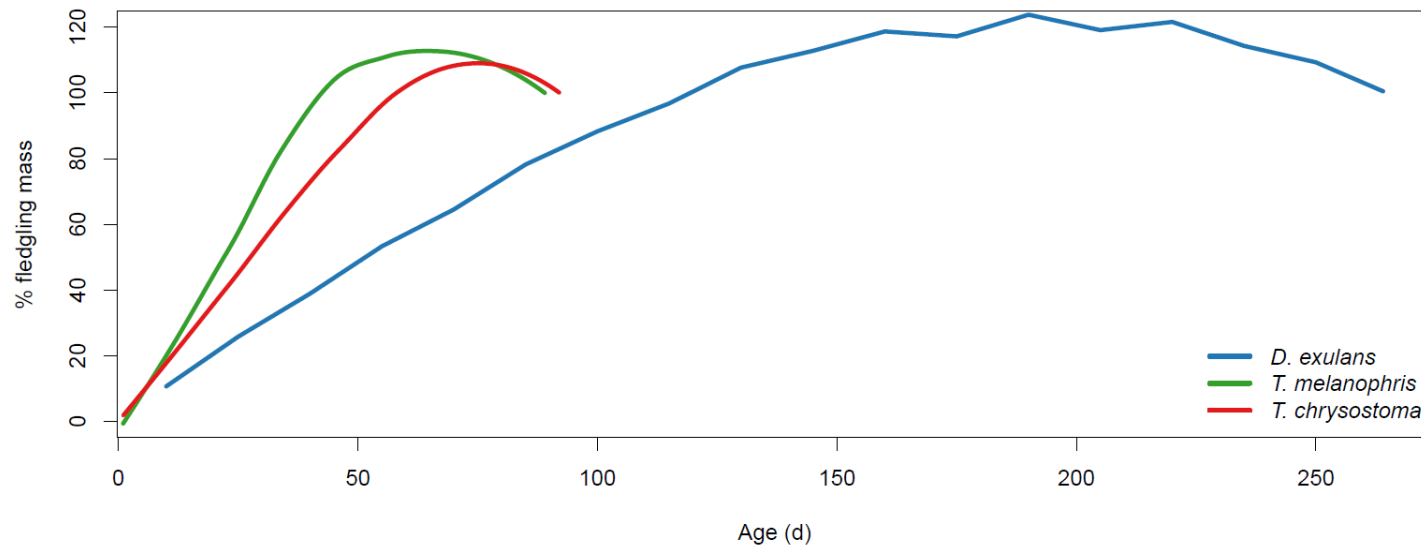
Light-mantled albatross *P. palpebrata*

Current state in (sea)bird ecology

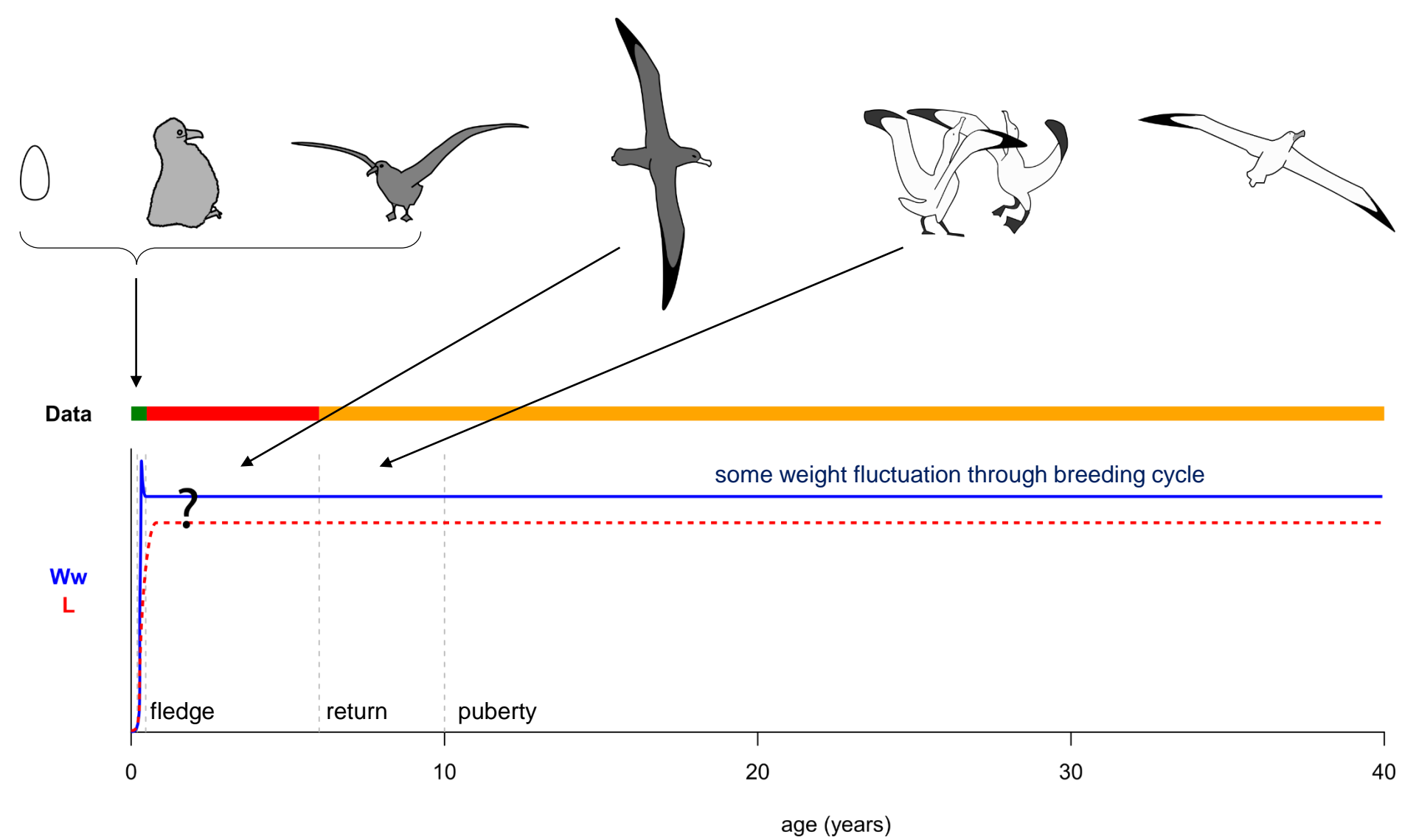


- Asymptotic growth models with varying degrees of mechanism:
 - logistic curve fitting Starck & Ricklefs (1998)
 - Gompertz (1825)
 - von Bertalanffy (1934)

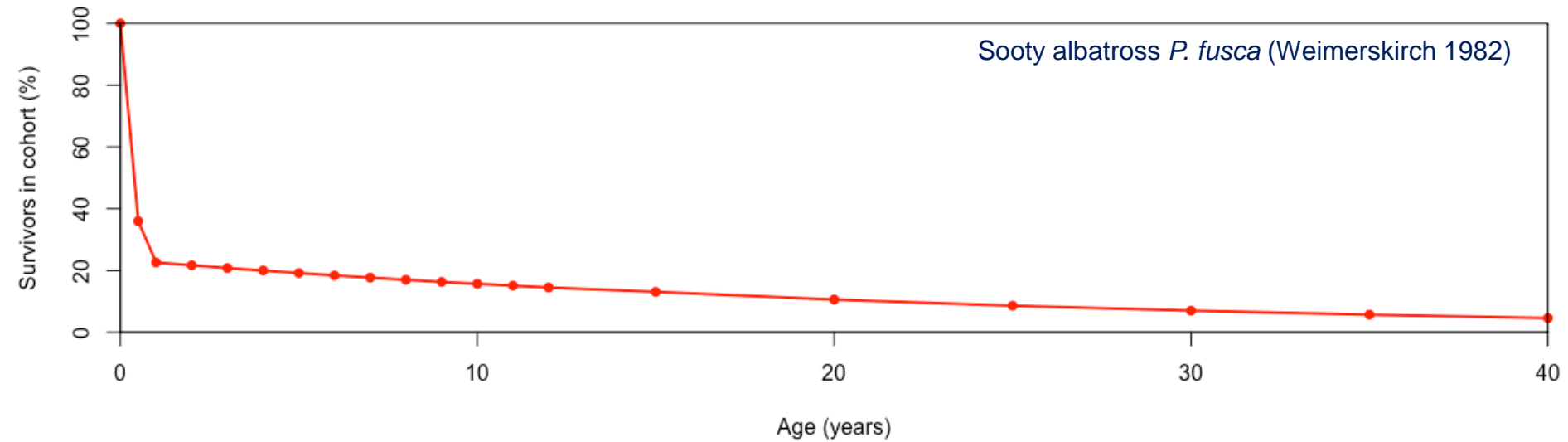
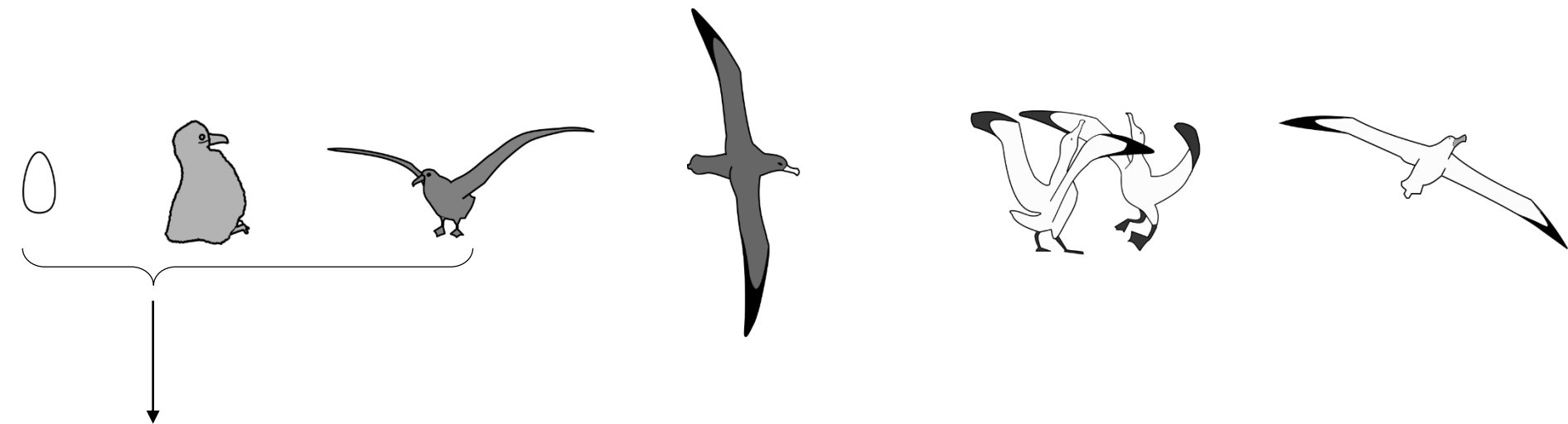
Seabird mass ontogenies do not follow logistic growth



Why do we care about this?

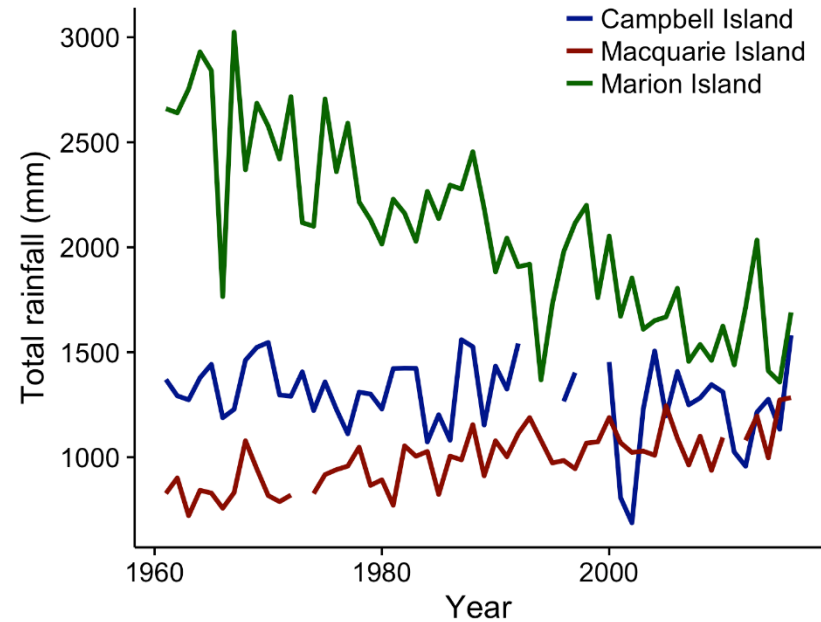
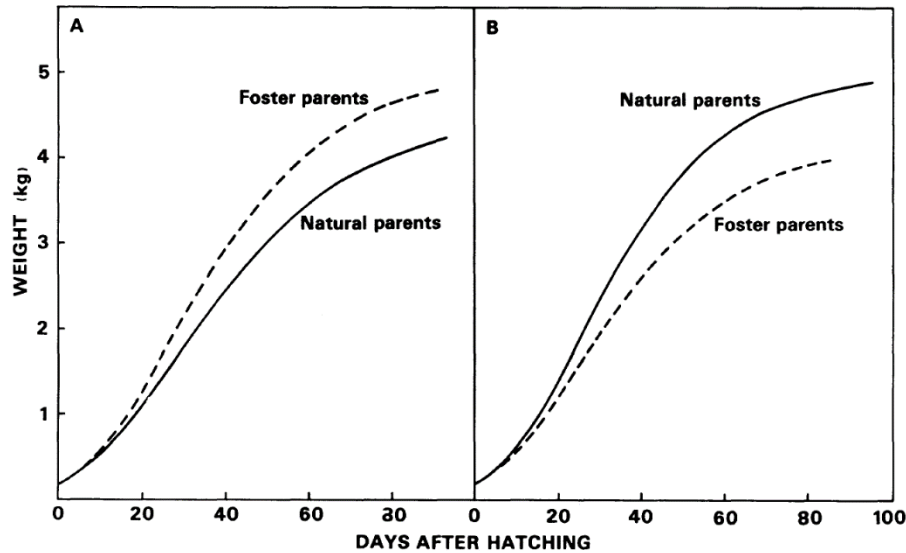


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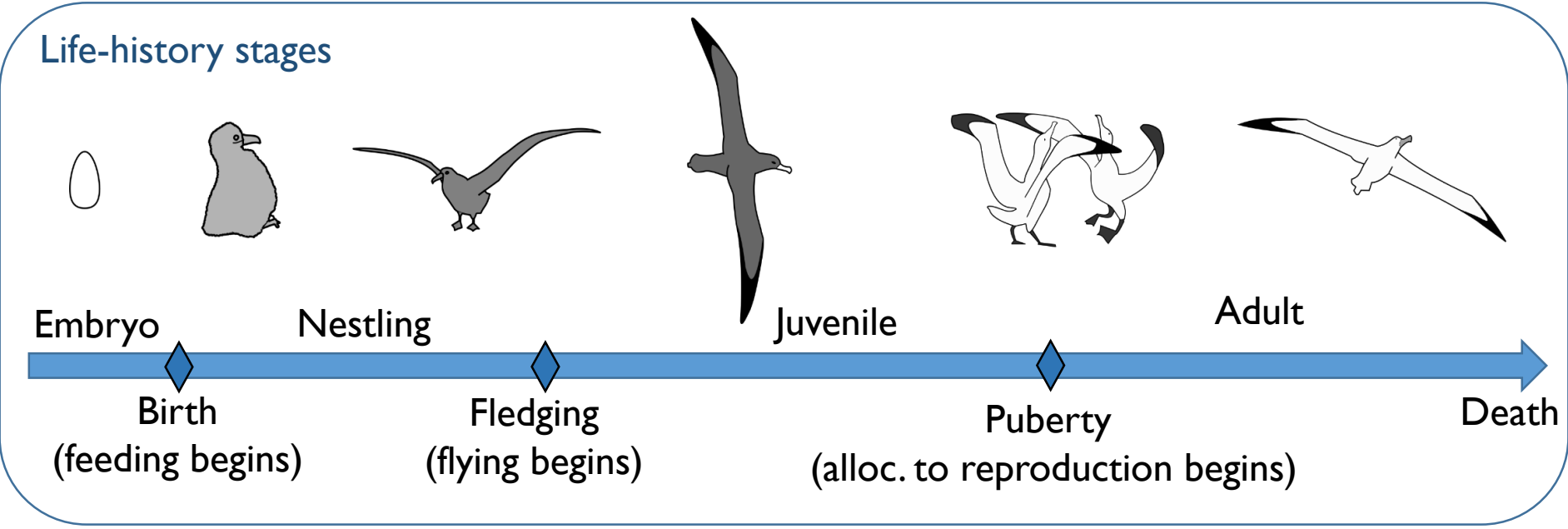
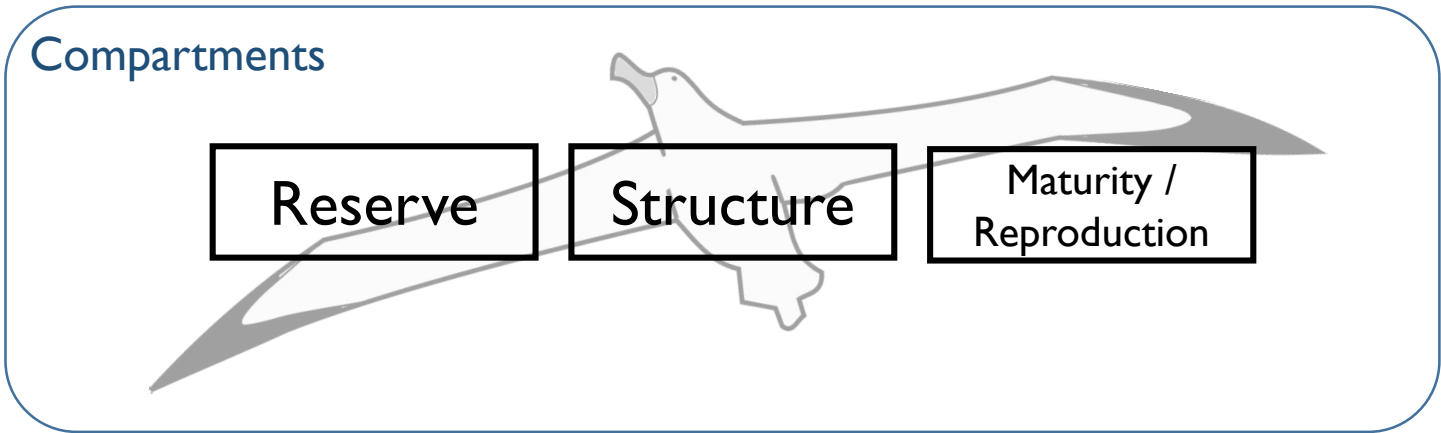
Early life survival is crucial for population dynamics

Relevant simulation scenarios

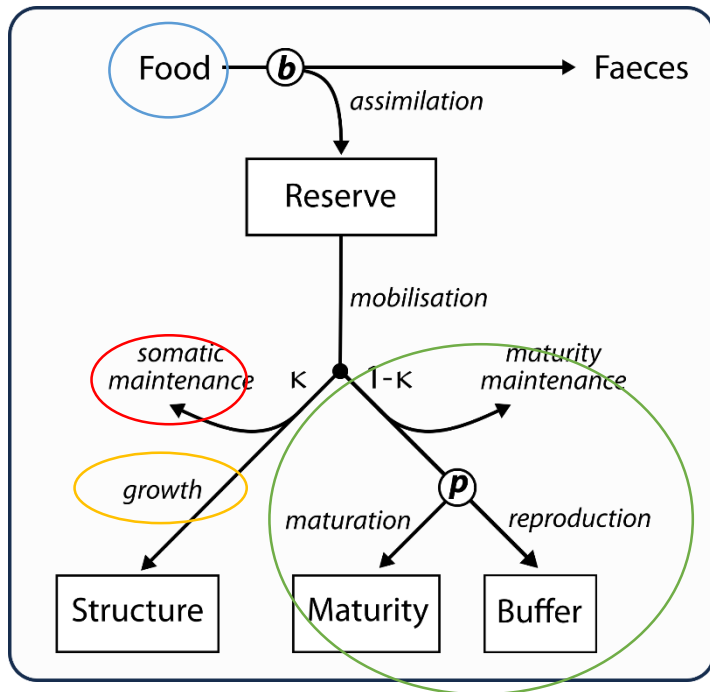


- Re-analysis of cross fostering experiments of sister-species (Prince 1981)
- weather related nestling survival under climate change
 - Rainfall impacts nestling survival (Wheeler et al 2013, GHA Marion Island)
 - Changing precipitation on subantarctic islands (Adams et al 2009)

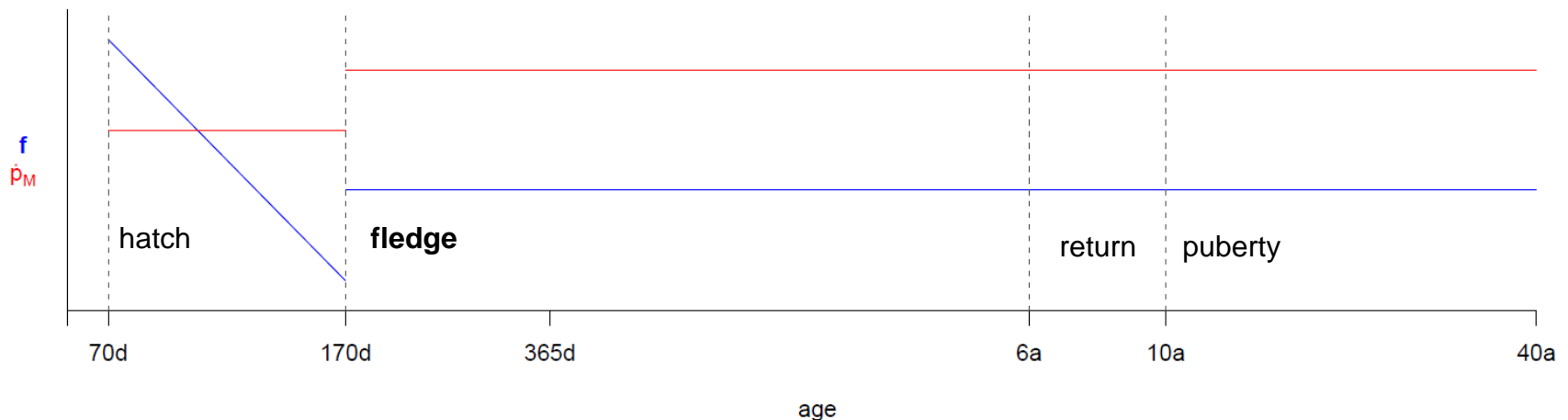
Wandering albatross DEB model (Teixeira et al. 2014)



Wandering albatross DEB model (Teixeira et al. 2014)

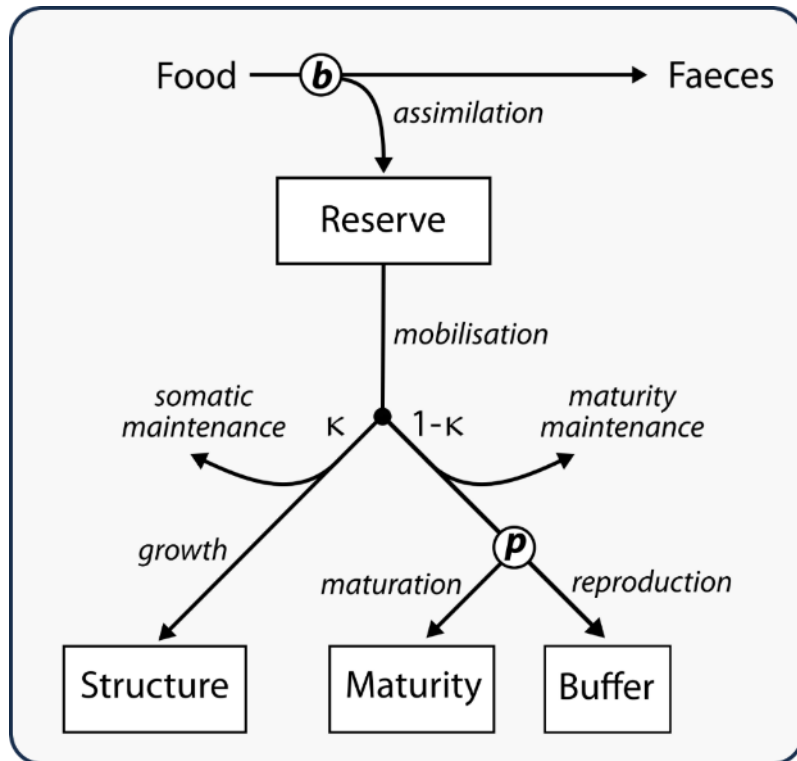


- food composition constant
- functional response declines linearly until fledging
- *constant $f = 0.8$ post-fledging*
- **stepwise constant**
- **pre- < post-fledging**
- **poorly constrained by data**
- **humped body mass trajectory**
- **slightly humped structural growth**
- **total body water content drops**

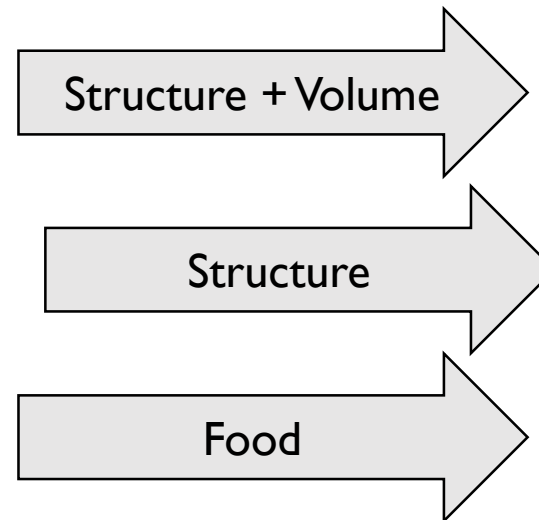


Confronting the model with data

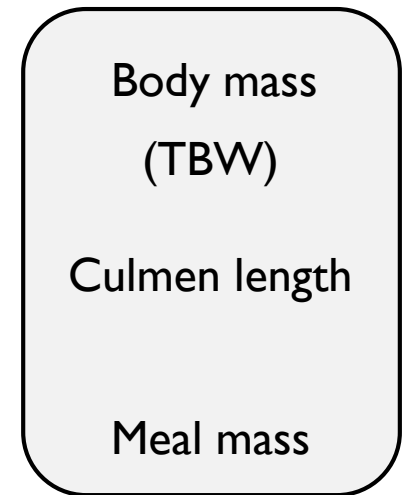
Process model



Observation model

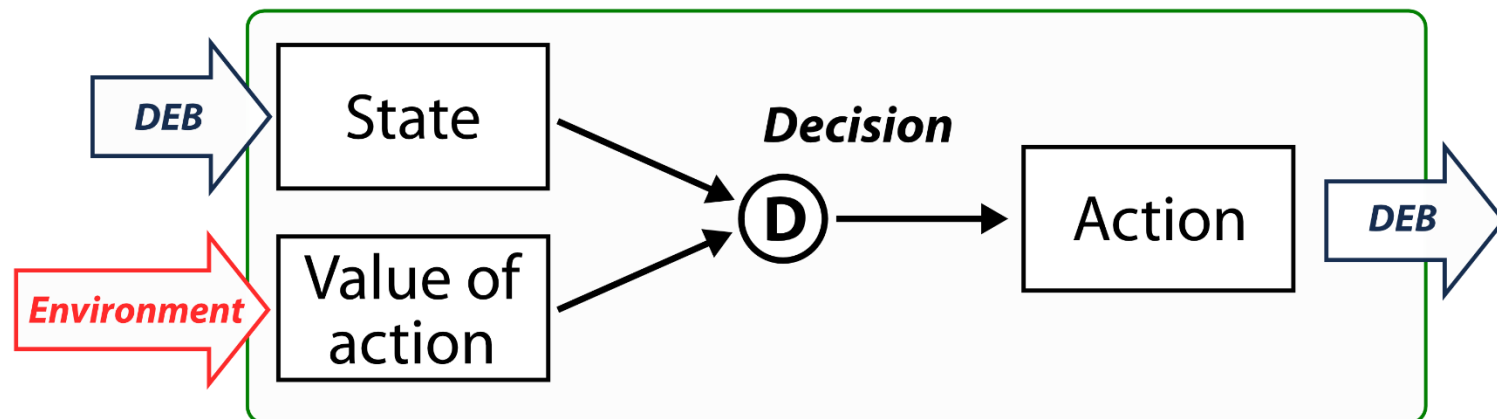


Data



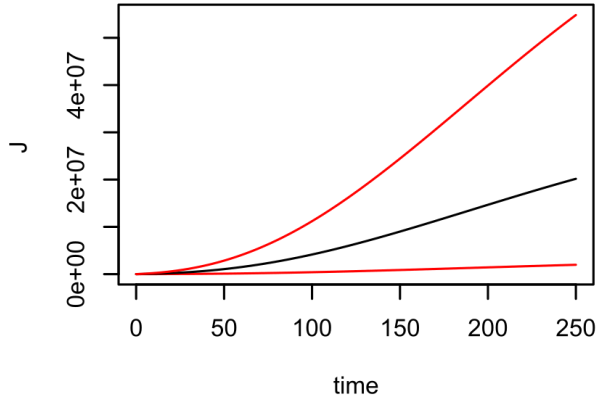
Fitting can be achieved with the “covariation method”, a weighted least-squares approach. We propose a Bayesian approach to better handle uncertainty.

- Individual heterogeneity is important in albatross populations
 - Ultimately we want to understand population distributions of DEB parameters
- Exploit scaling relationships from DEB theory to propagate parameters from data-rich to data poor species
- Propagate parameter uncertainty into coupled models
 - e.g. state-space foraging models

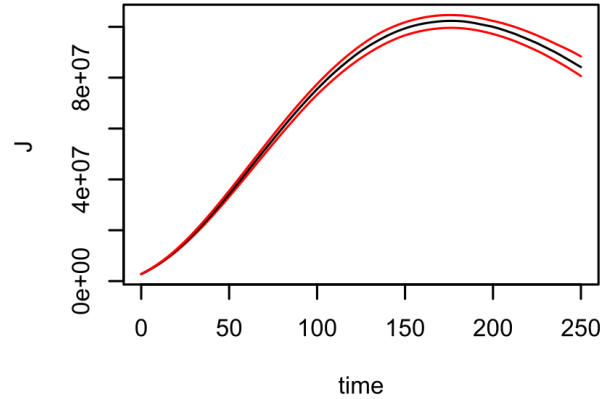


Model fits: Wandering Albatross (strong priors)

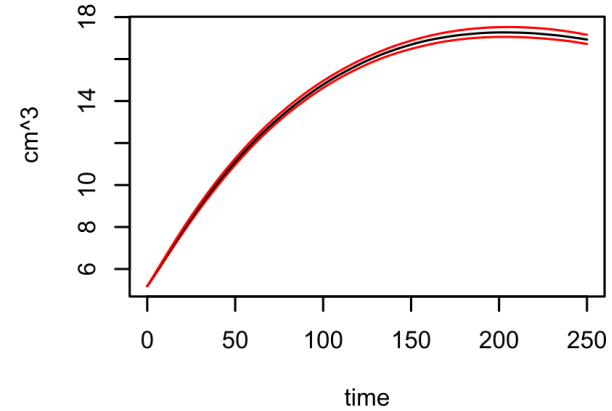
Maturity



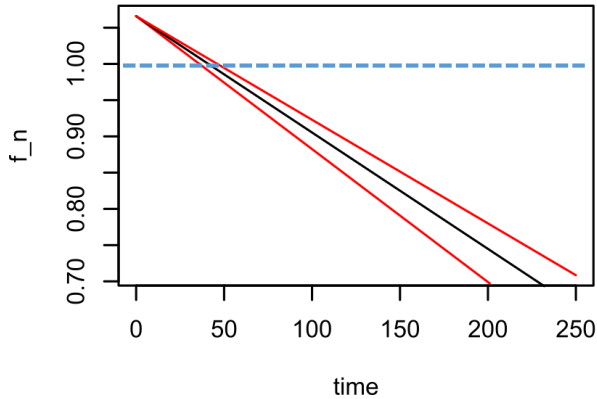
Reserve



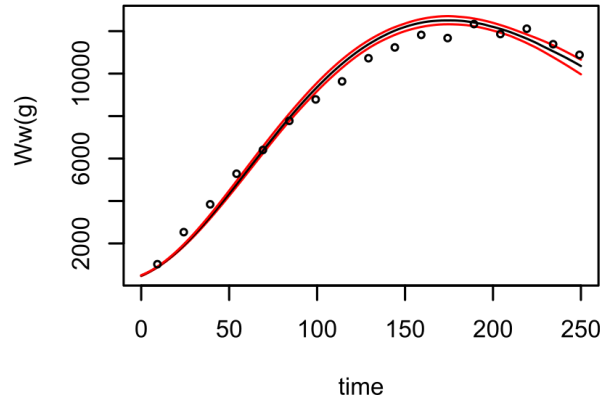
Structure



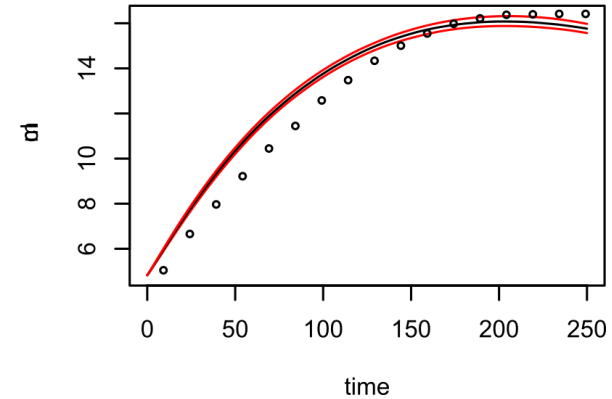
Food



Wet weight



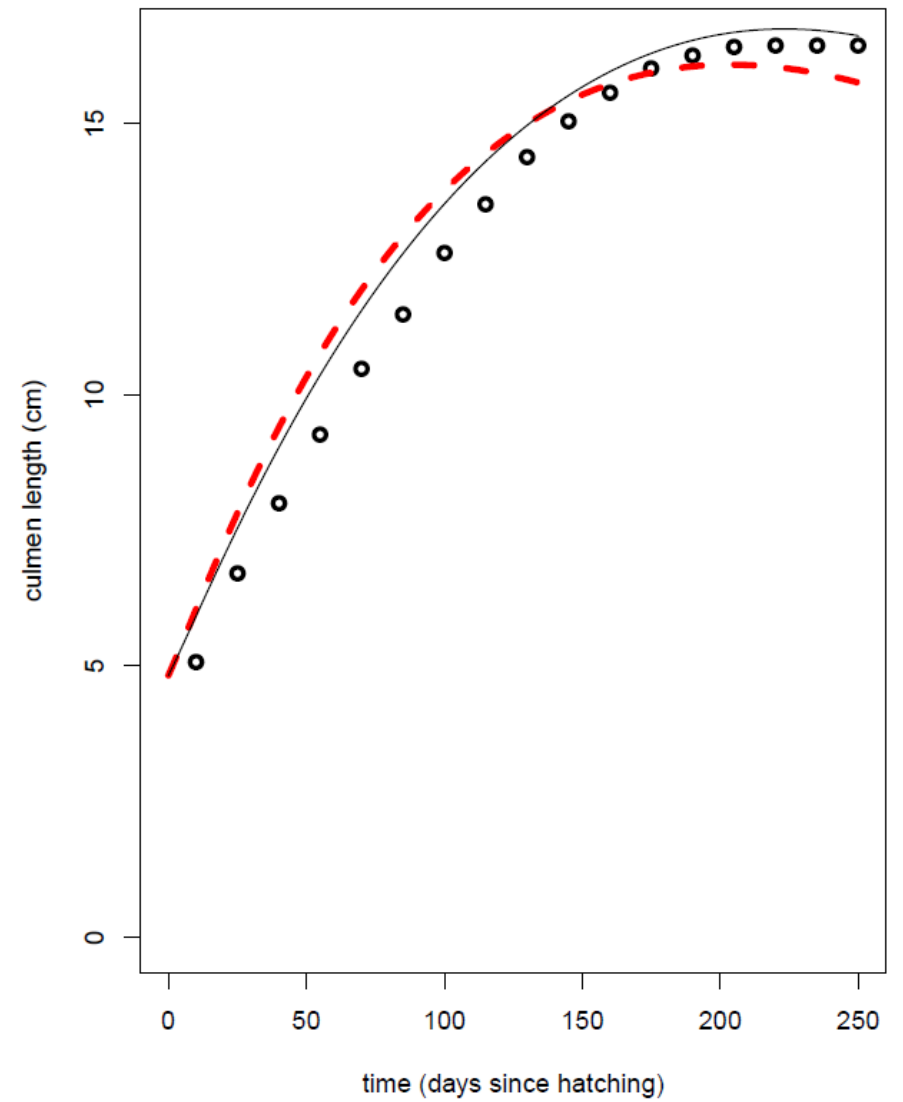
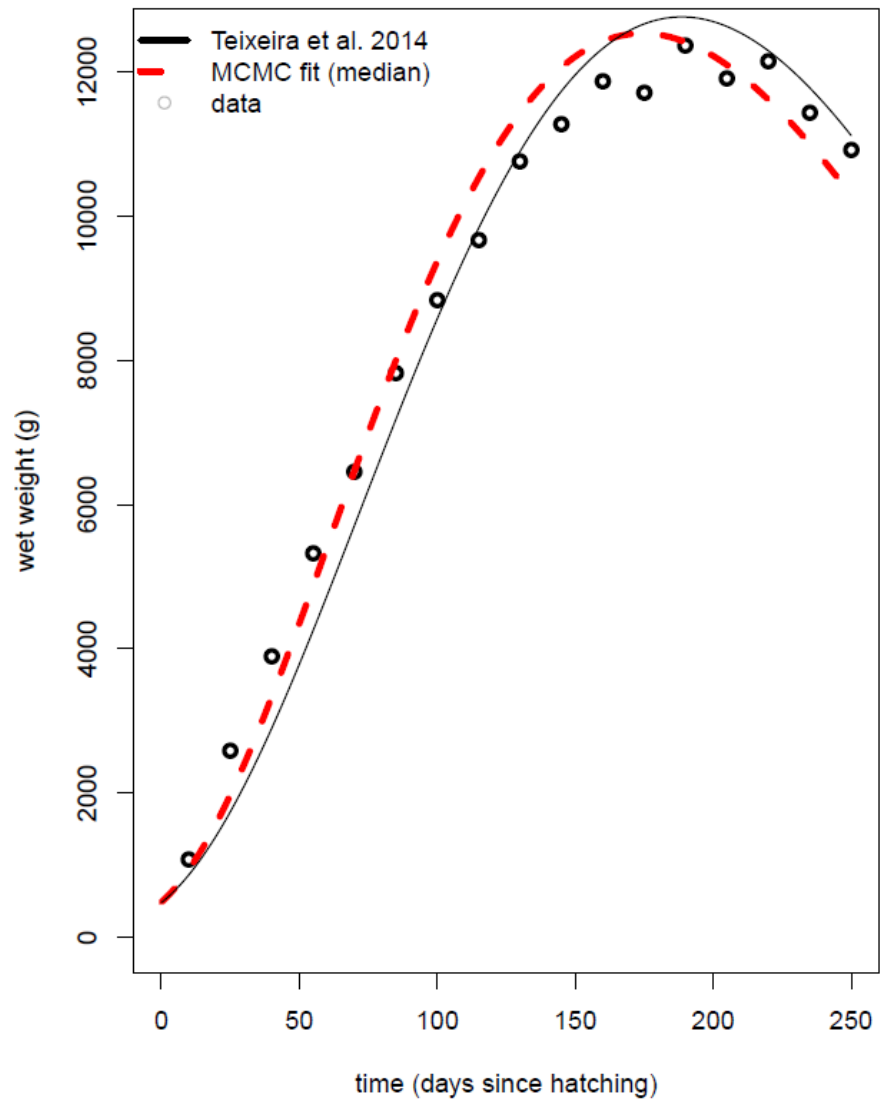
Culmen



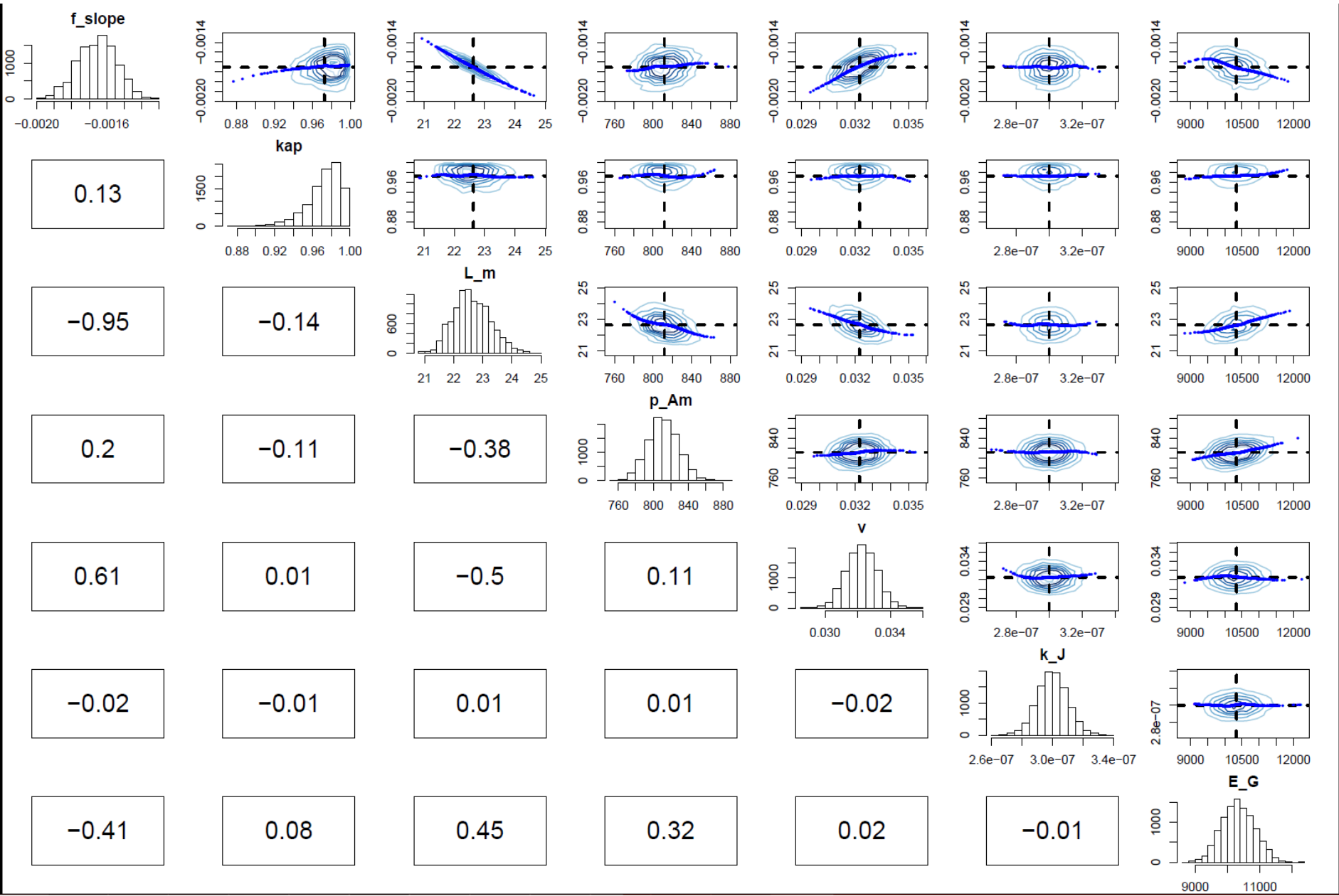
Early overfeeding → rapid build-up of reserve

Declining food input towards fledging → draw-down of reserve

Model fits: Wandering Albatross

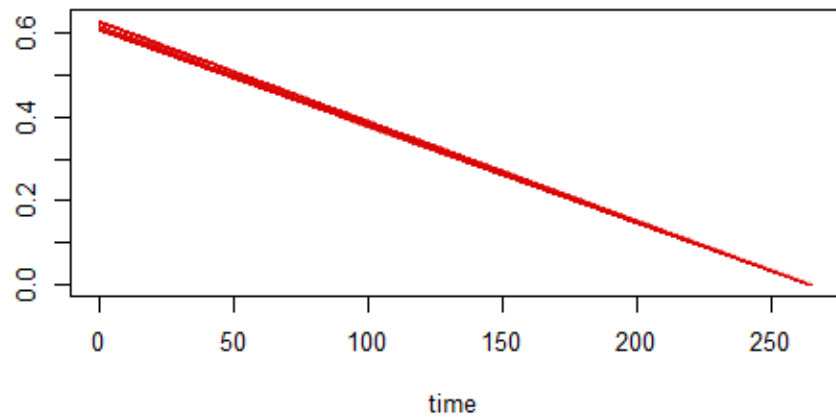


Parameter estimation results

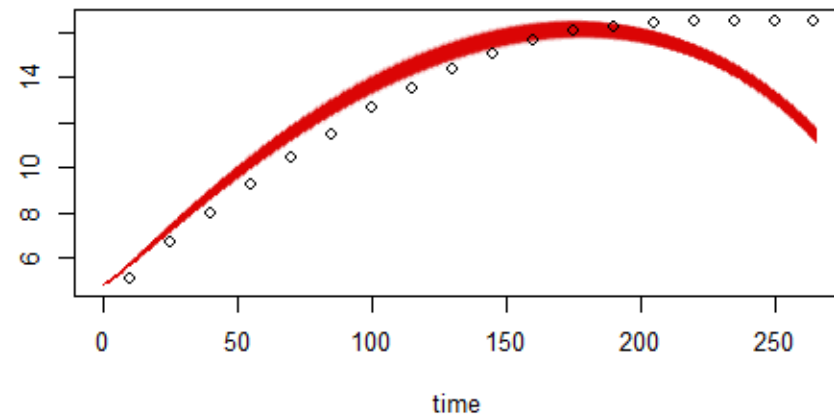


Parameter estimation results (weak priors)

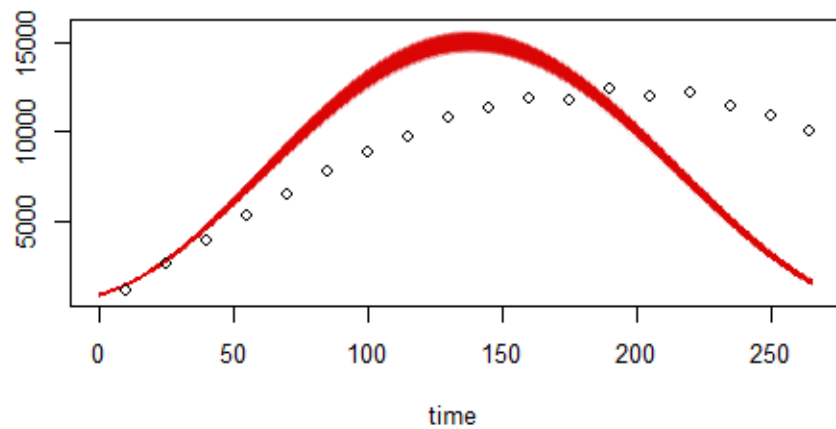
f_n



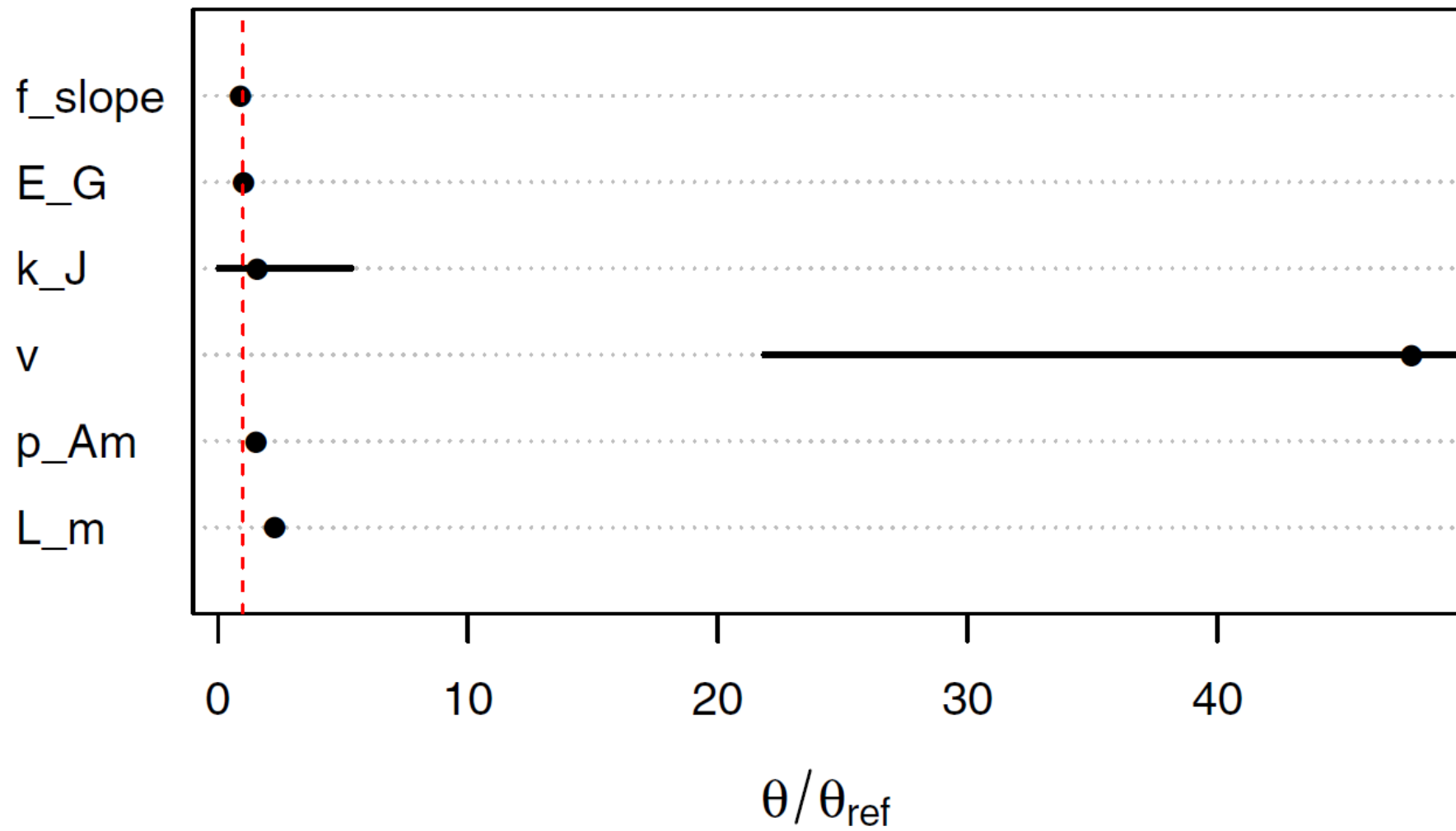
L_cul



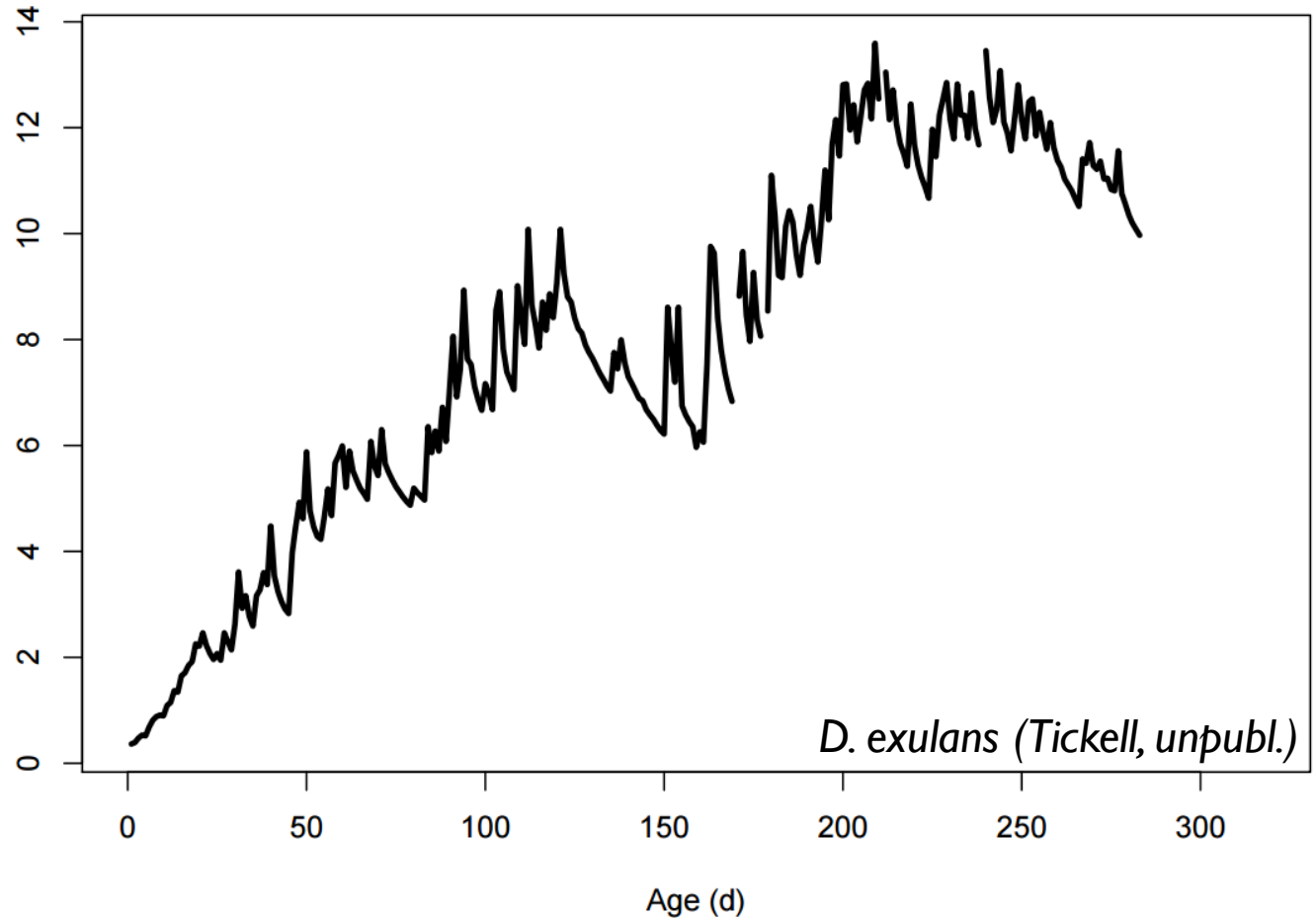
Ww



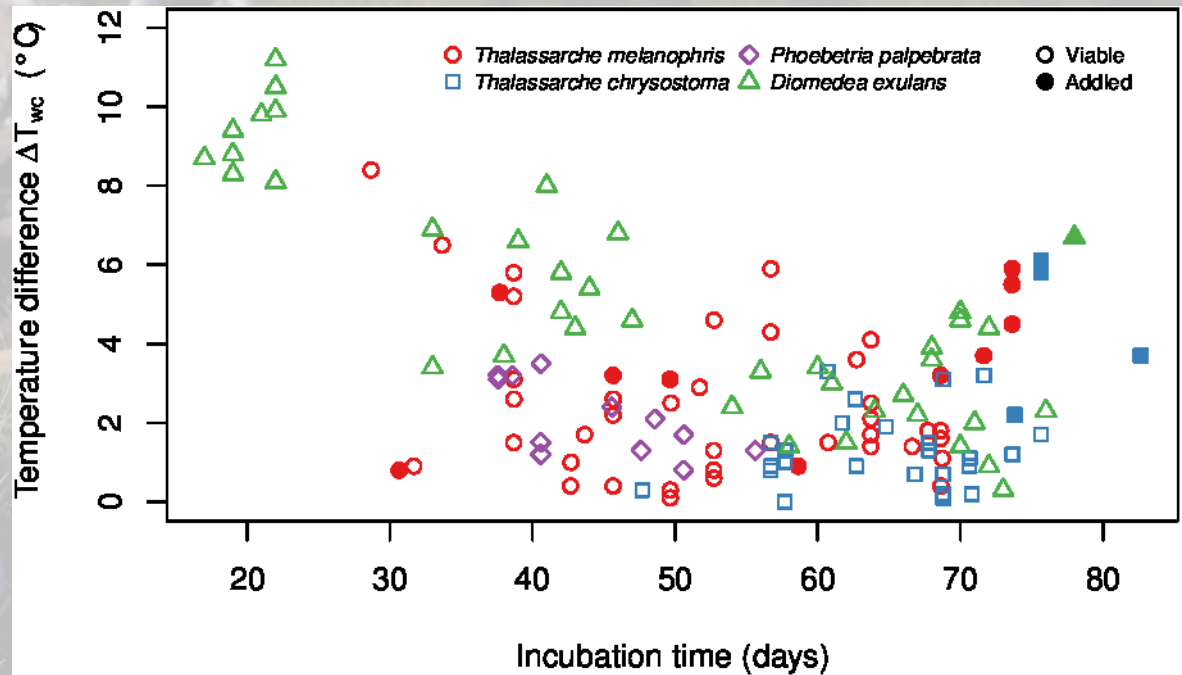
Parameter estimation results (weak priors)



Albatross mass ontogenies are not smooth



Albatross are not born as endotherms



Boersch-Supan, Johnson et al., submitted

- DEB provides a quantitative framework to investigate the effects of multiple drivers on individual growth, reproduction, and survival
- The Teixeira model provides a mechanistic explanation of humped mass ontogenies
- Bayesian parameter estimation has potential to link the model to data, incorporating parameter uncertainties and covariances
- More (smarter?) data is needed for model identifiability
 - Better constrained feeding model
 - Respiration and temperature data in egg phase

Thank you for listening!



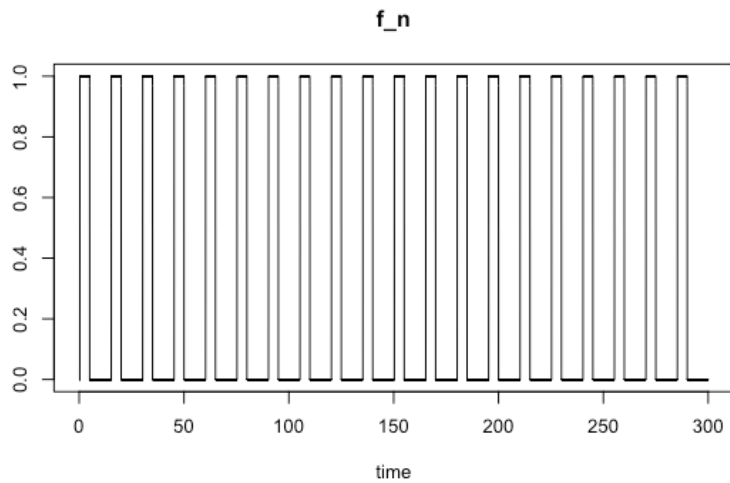
pboesu@gmail.com

[@pboesu](https://twitter.com/pboesu)

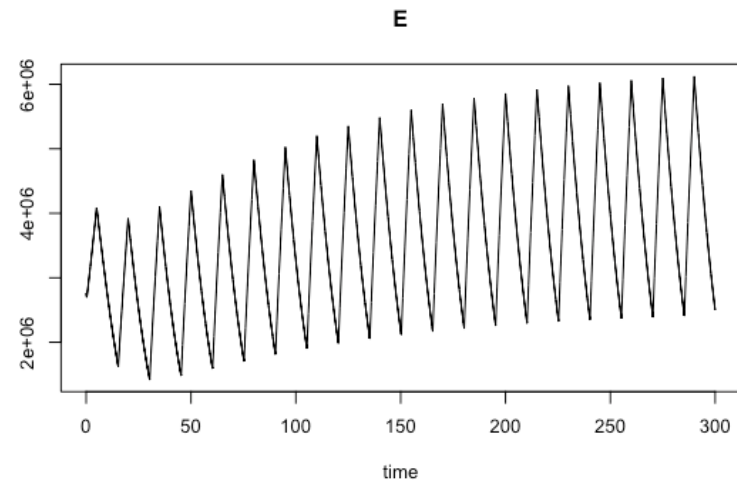
github.com/pboesu/seabirdeb

Model dynamics with pulsed food

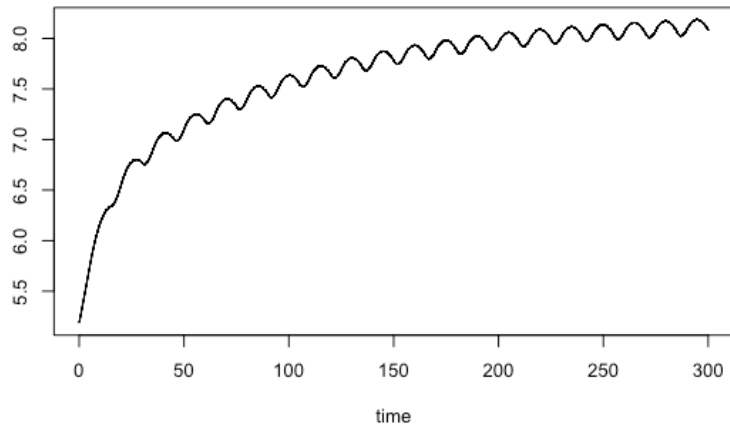
food



reserve density

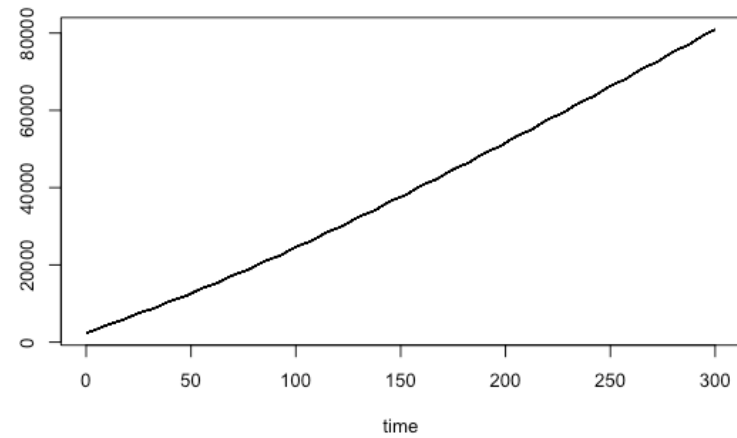


L



structure

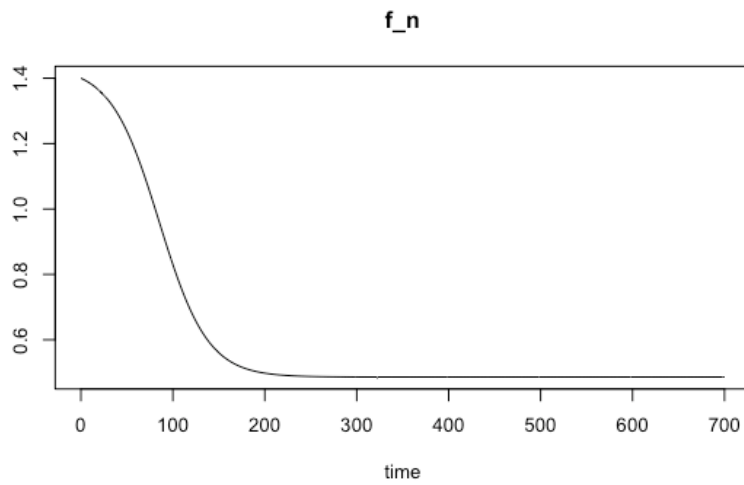
H



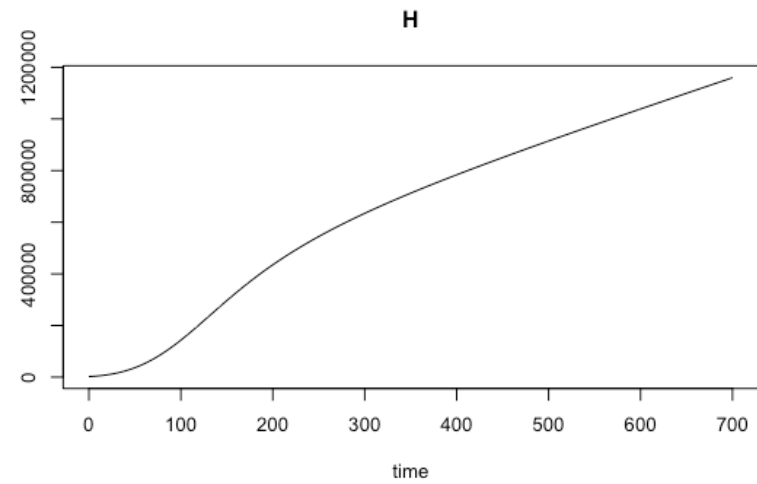
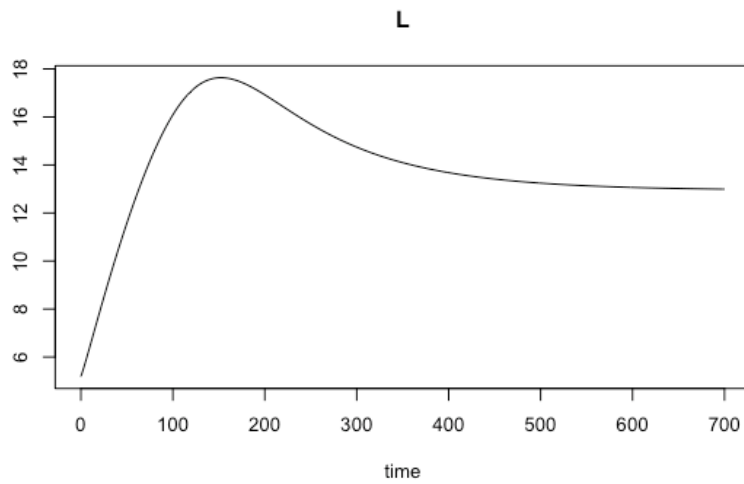
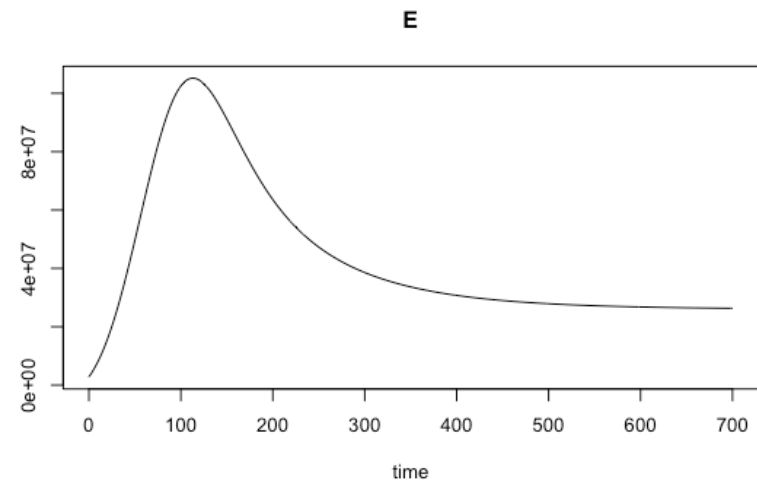
maturity

Model dynamics with logistic food

food



reserve density



structure

maturity

deBIInfer provides templates for:

1. the (D)DE model
2. the observation model and likelihood
3. the prior distributions of the parameters.

Additionally, the user must specify/input:

- Data
- Proposal variances

ODE Equations

$$\frac{dL}{dt} = \frac{\dot{r}L}{3}$$

$$\frac{dE}{dt} = \{\dot{p}_{Am}\}fL^2 - \dot{p}_C$$

$$\frac{dE_H}{dt} = (1 - \kappa)\dot{p}_C - \dot{k}_J E_H$$

$$\frac{df}{dt} = f_{slope}$$

Observation Equations

$$W_w = L^3 \left(1 + f \frac{\{\dot{p}_{Am}\}w_E}{\dot{v}d_V\mu_E} \right) d_V \delta_{wd}$$

$$L_{cul} = \delta_M L$$

$$\mathcal{L}(\mathcal{Y}|\boldsymbol{\theta}) = \prod_t \frac{1}{\tilde{L}_t \sigma_L \sqrt{2\pi}} \exp\left(-\frac{(\ln \tilde{L}_t - \ln L_t)^2}{2\sigma_L^2}\right) \times \frac{1}{\tilde{W}_{wt} \sigma_W \sqrt{2\pi}} \exp\left(-\frac{(\ln \tilde{W}_{wt} - \ln W_{wt})^2}{2\sigma_W^2}\right)$$

Once these are specified, we use a standard Metropolis within Gibbs MCMC algorithm to obtain draws from the posterior distribution.