

# Implementing realistic biological variability in an individual-based Dynamic Energy Budget model

Josef Koch and Karel De Schampelaere

Laboratory of Environmental Toxicology and Aquatic Ecology, Environmental Toxicology Unit (GhEnToxLab), Ghent University, Ghent, Belgium

## Introduction

### DEB-IBM

- Allows for extrapolation of individual-level energetic effects to populations



DEB[1] + IBM[2]

### Species: *Nitocra spinipes*

- Brackish water copepod
- Worldwide distribution
- Ecotoxicological test species\*
- DEB parameters available<sup>[3]</sup>

\* - OECD ENV/JM/MONO(2014)17  
- ISO 14669:1999  
- ISO/TS 18220:2016




### Inter-individual variability

- Integral in population resilience
- Data available for *N. spinipes*:



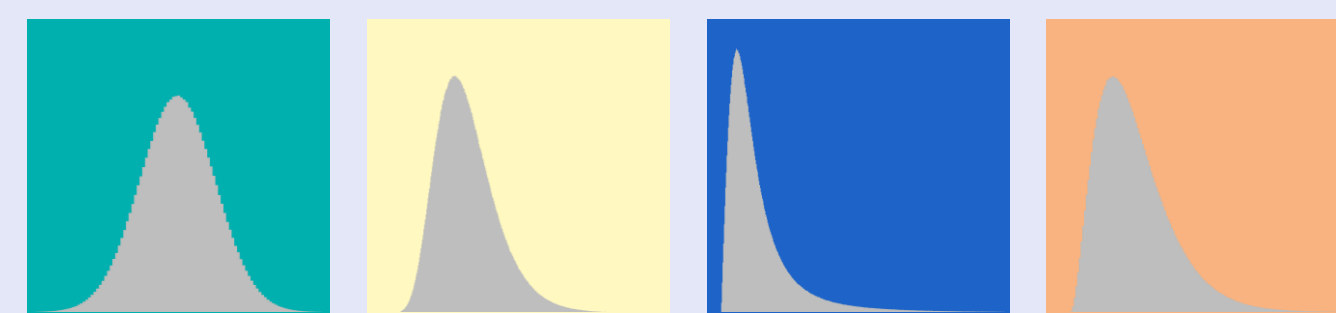
**Challenge: How can variability in DEB parameters be estimated from variation in data?**



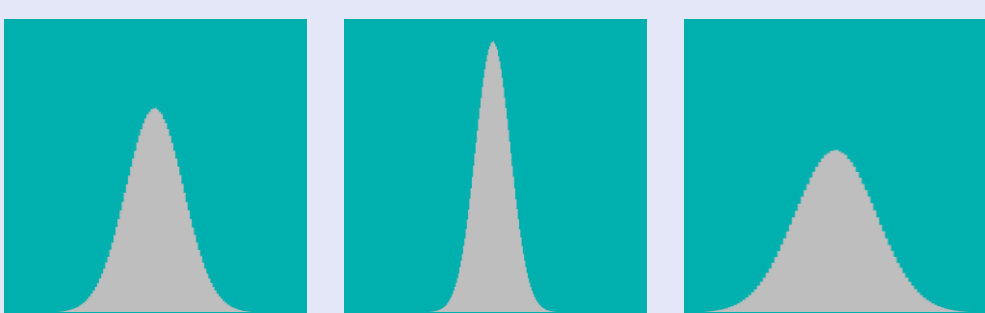
### What parameters?

$$\begin{matrix} \{\dot{F}_m\} & \kappa_X & [E_G] & \kappa & \dot{k}_J \\ \{\dot{p}_{Am}\} & [\dot{p}_M] & \dot{v} & \{\dot{p}_T\} & E_H^p \\ & \kappa_R & E_H^b & E_H^j & E_H^p \end{matrix}$$

### What probability distributions?

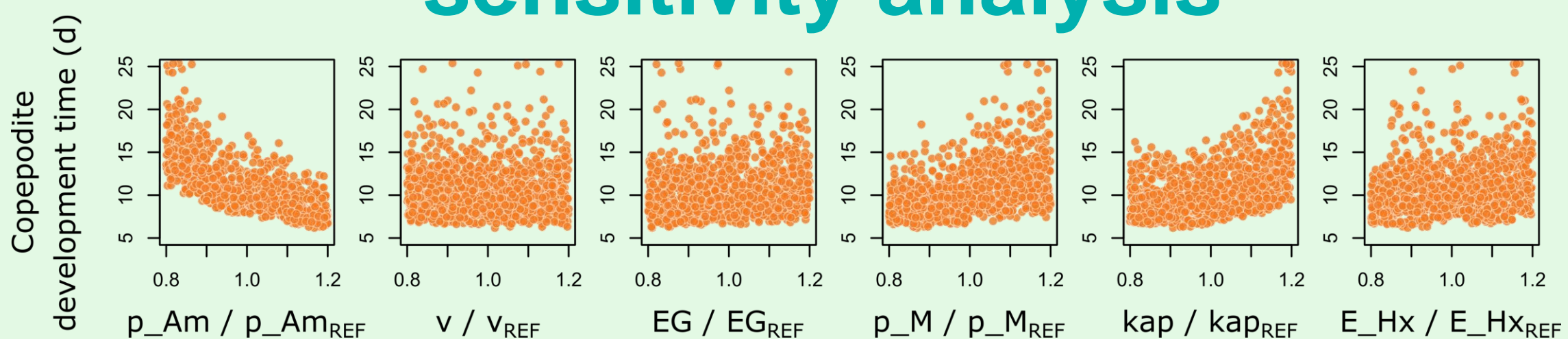


### What scale of variation?



## Methods

### Variance based global sensitivity analysis



Monte Carlo simulations with all parameters uniformly distributed (+/- 20%) around their respective reference value from the preceding parameter estimation<sup>[3]</sup>

$$S_i = \frac{V_{X_i}(E_{X \sim i}(Y|X_i))}{V(Y)}$$

The first-order effect indices  $S_i$ <sup>[4]</sup> (always between 0 and 1) indicate how much of the output variance  $V(Y)$  can be directly attributed to each respective input parameter  $X_i$

### Compare common distribution types

Gaussian  
VS.  
Log-normal  
VS.  
Log-logistic  
VS.  
Gamma

### Optimisation algorithm for scale parameter

DEB parameter distributed around mean with scale parameter

Run Monte Carlo simulations of 10<sup>6</sup> individuals

Distribution of simulated data

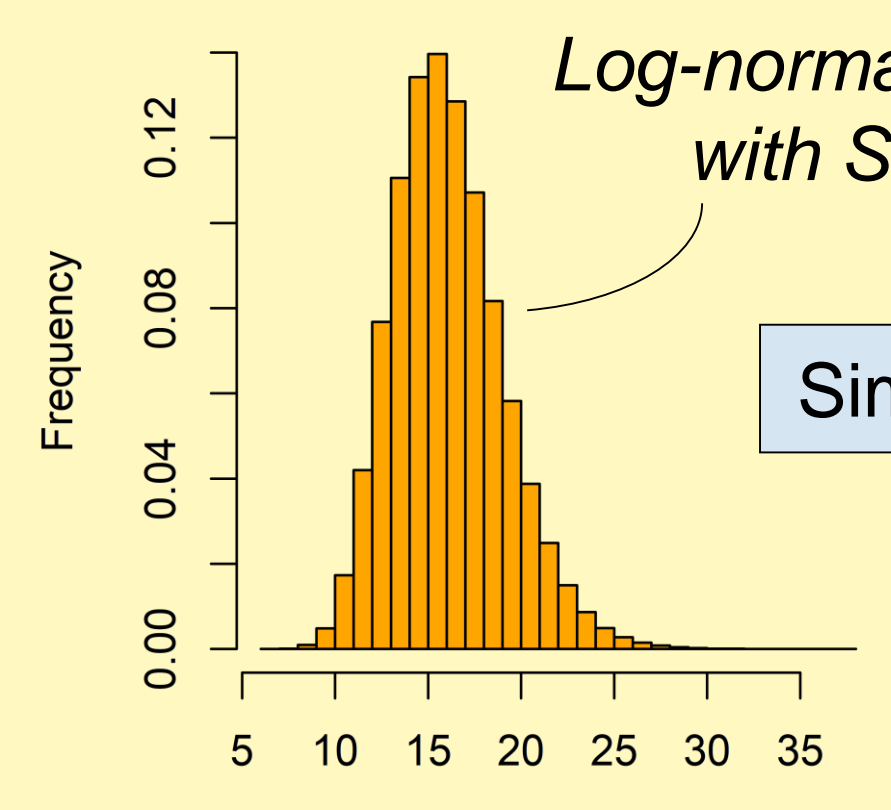
Adjust scale parameter to minimize loss function

Compute loss function

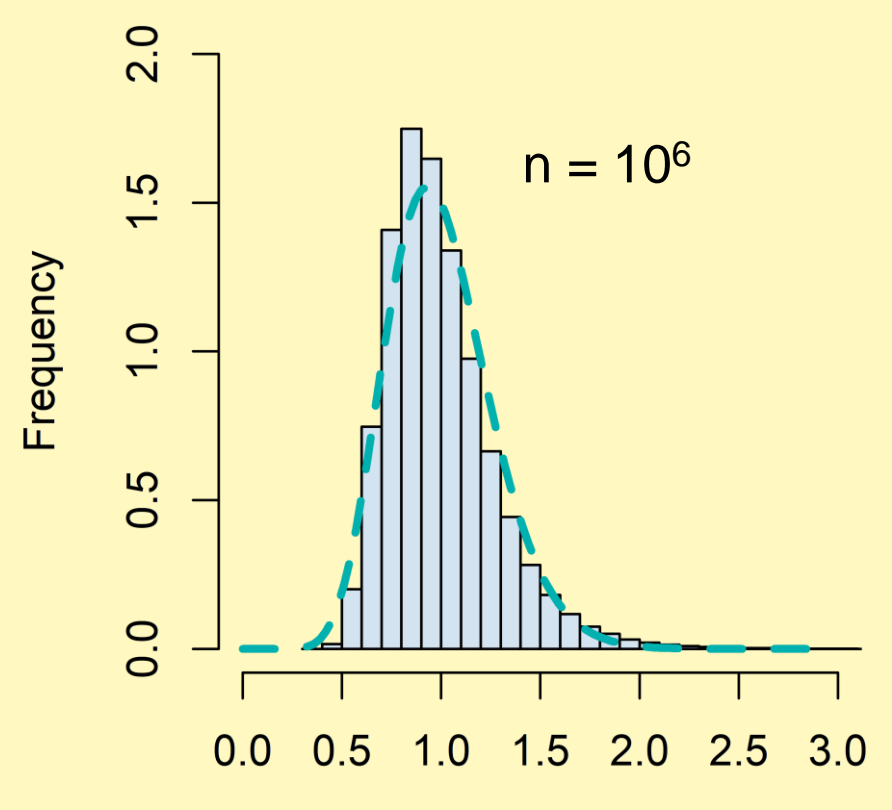
Distribution of measured data

## Results & Discussion

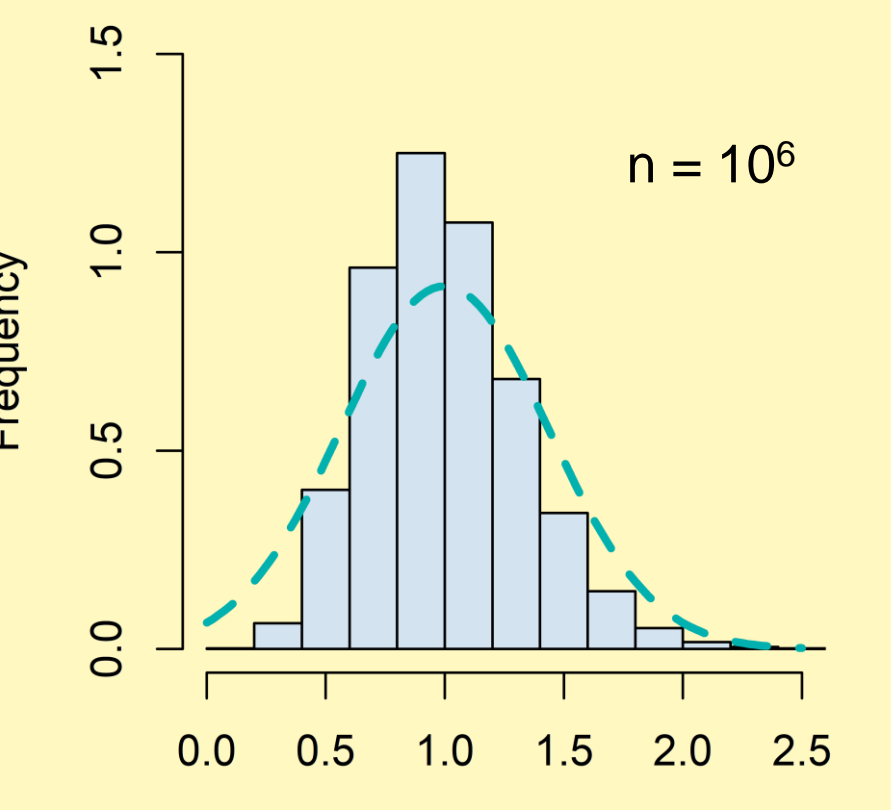
Parameter	$\{\dot{p}_{Am}\}$	$\dot{v}$	$[E_G]$	$[\dot{p}_M]$	$\kappa$	$E_H^x$
$S_i$ Dev. time	0.44	0.003	0.03	0.06	0.36	0.06
$S_i$ Brood size	0.42	0.001	0.01	0.11	0.39	0.05



Log-normal distribution with SD = 0.18



n = 10<sup>6</sup>

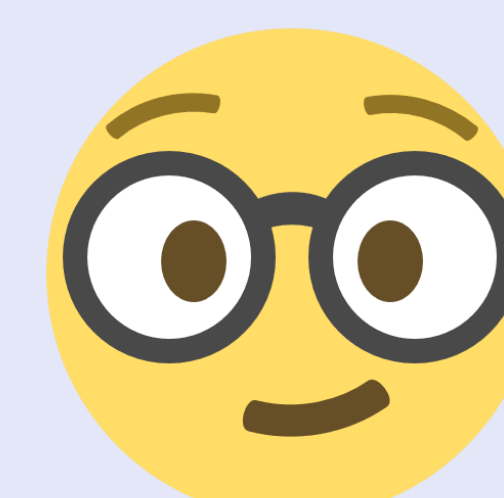


n = 10<sup>6</sup>

- Observed endpoints are most sensitive to the maximum assimilation rate  $\{\dot{p}_{Am}\}$
- Drawing  $\{\dot{p}_{Am}\}$  from a log-normal distribution with an optimised scale parameter (SD) led to the best possible approximation of the variation in real data

## Conclusions

- Variability in DEB parameters can be estimated from experimental data
- Adding variability to a single parameter provided a good approximation of observed variation in measured data and can easily be implemented in the IBM



josef.koch@ugent.be  
www.ecotox.ugent.be  
@GhEnToxLab @ugent  
Ghent University

## References

- [1] Kooijman, S.A.L.M., *Dynamic Energy Budget Theory For Metabolic Organisation*. 3rd ed. 2010, Cambridge: Cambridge University Press
- [2] Railsback, S.F. and V. Grimm, *Agent-based and Individual-based Modeling: A Practical Introduction*. 2nd ed. 2019, New Jersey: Princeton University Press
- [3] Koch, J. and K.A.C. De Schampelaere, *Two Dynamic Energy Budget Models for the Harpacticoid Copepod Nitocra spinipes*. *Journal of Sea Research*, 2018
- [4] Saltelli, A., et al., *Variance Based Sensitivity Analysis of Model Output. Design and Estimator for the Total Sensitivity Index*. *Computer Physics Communications*, 2010, 181(2): p. 259-270