



Extending the SSD Concept to Explore Some Foundational Model Limitations: A Bayesian Hierarchical Approach

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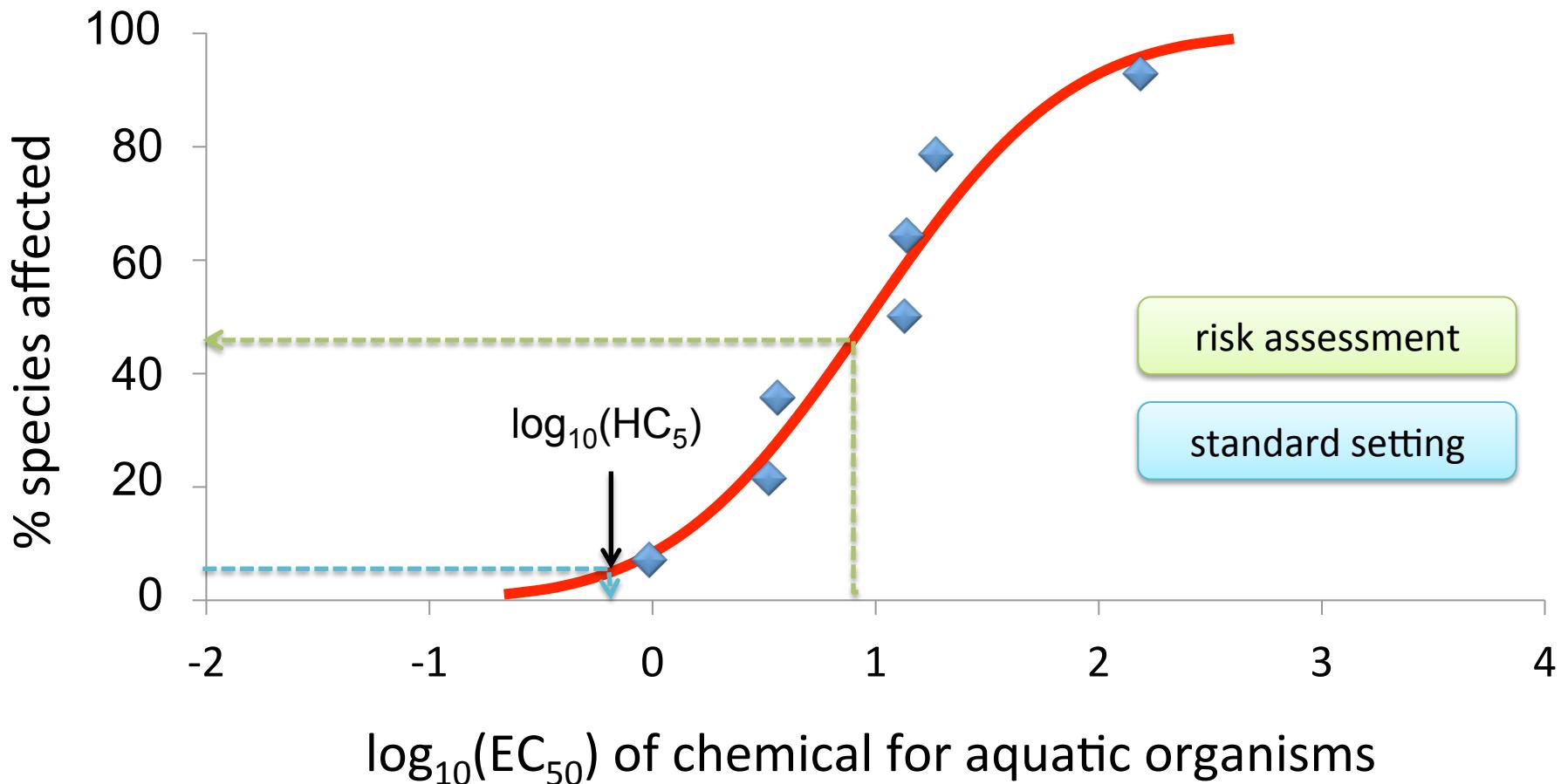
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Species Sensitivity Distributions (SSDs)



All Models Are Wrong...

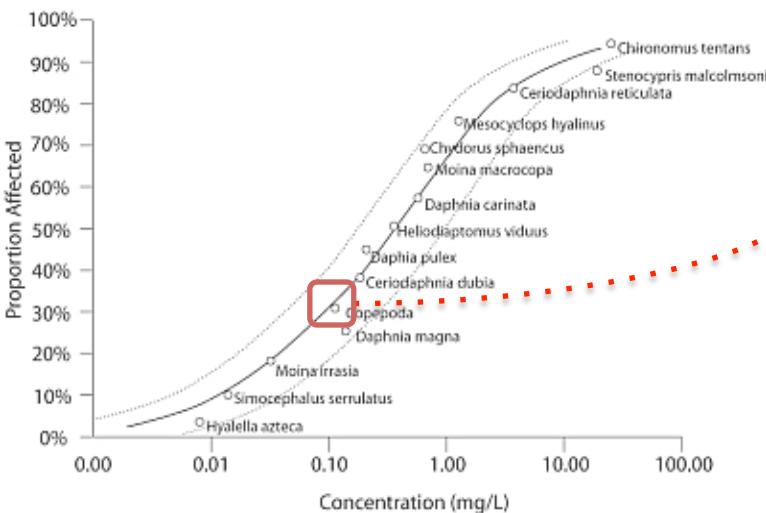
(the SSD is no exception!)... but some are better than others.

ASSUMPTION	PRACTICAL CONSEQUENCE
SSDs are independent	Information gained about each chemical risk assessment will not strengthen learning of future assessments.
Interspecies variation is attributable to chemical effects only	Observation of some species being more/less sensitive not accounted for.
Other sources of variation (e.g. inter-laboratory and intra-species variation) are ignorable or captured by an arbitrary assessment factor ($1 \leq AF \leq 5$)	Confounding of the HC ₅ interpretation – should it just represent interspecies variation?
Representative of all ecosystems	No account of specific assemblages and differing diversities.

Hierarchical Modelling

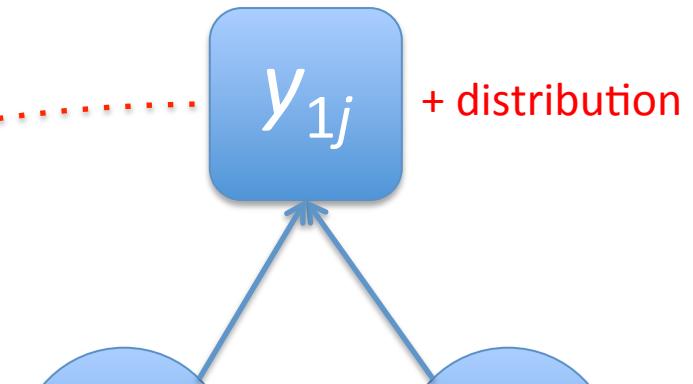
The standard SSD model can be written as a stochastic network

Chemical 1



log transformed toxicity datum
for chemical 1 and species j

=

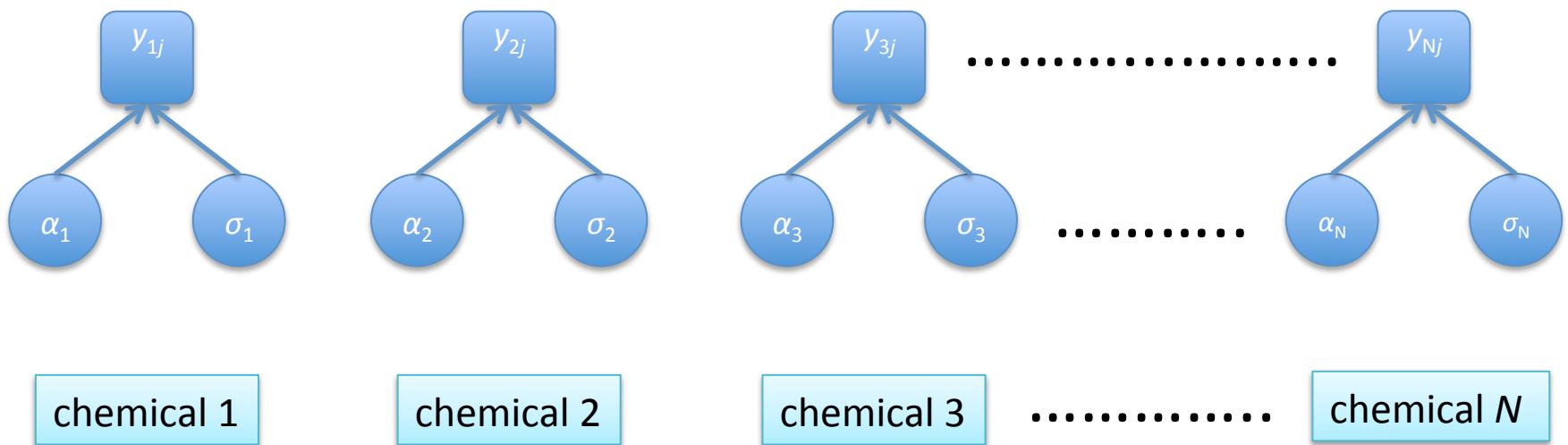


central
tendency

\propto dispersion
(interspecies
standard deviation)

Hierarchical Modelling

If we have N chemical risk assessments, the usual SSD model is a special (independence) case of a hierarchical model.



SSD interspecies variance parameters are heterogeneous *between* chemicals i .

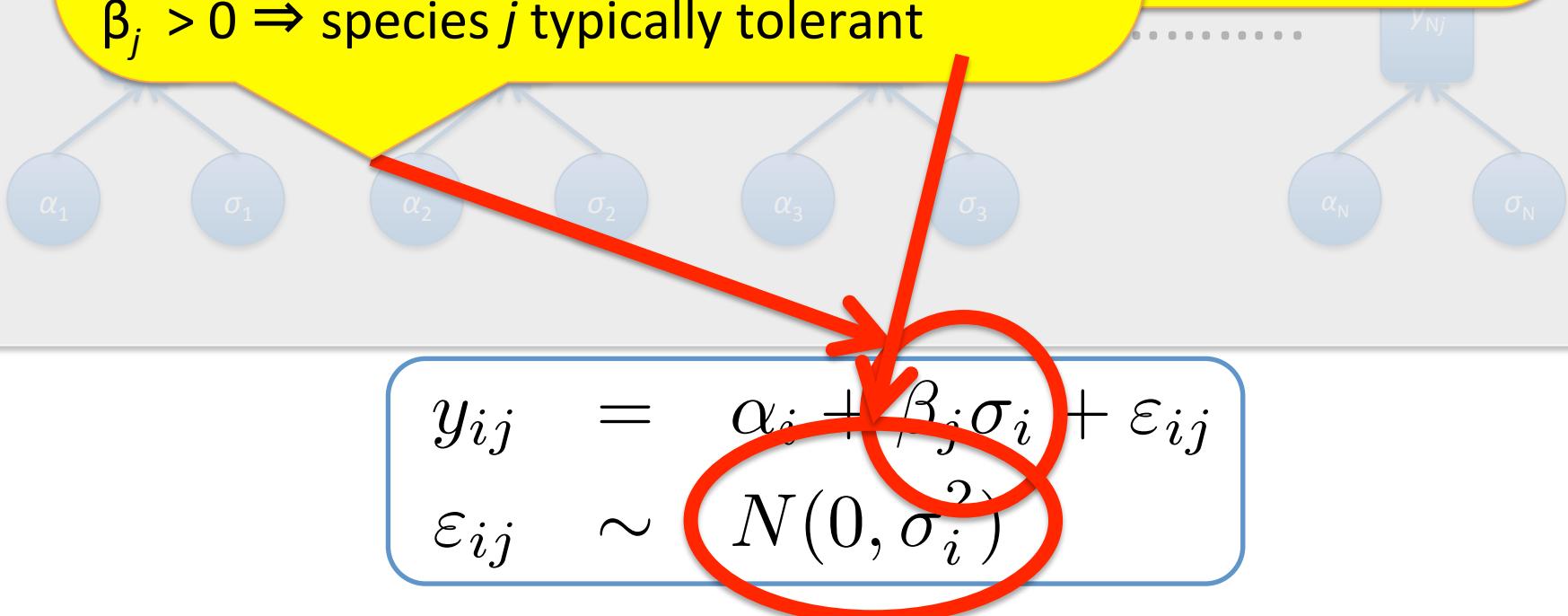
Therefore β_j measures species position as number of standard deviations from mean (log-)toxicity.

$\beta_j < 0 \Rightarrow$ species j typically sensitive

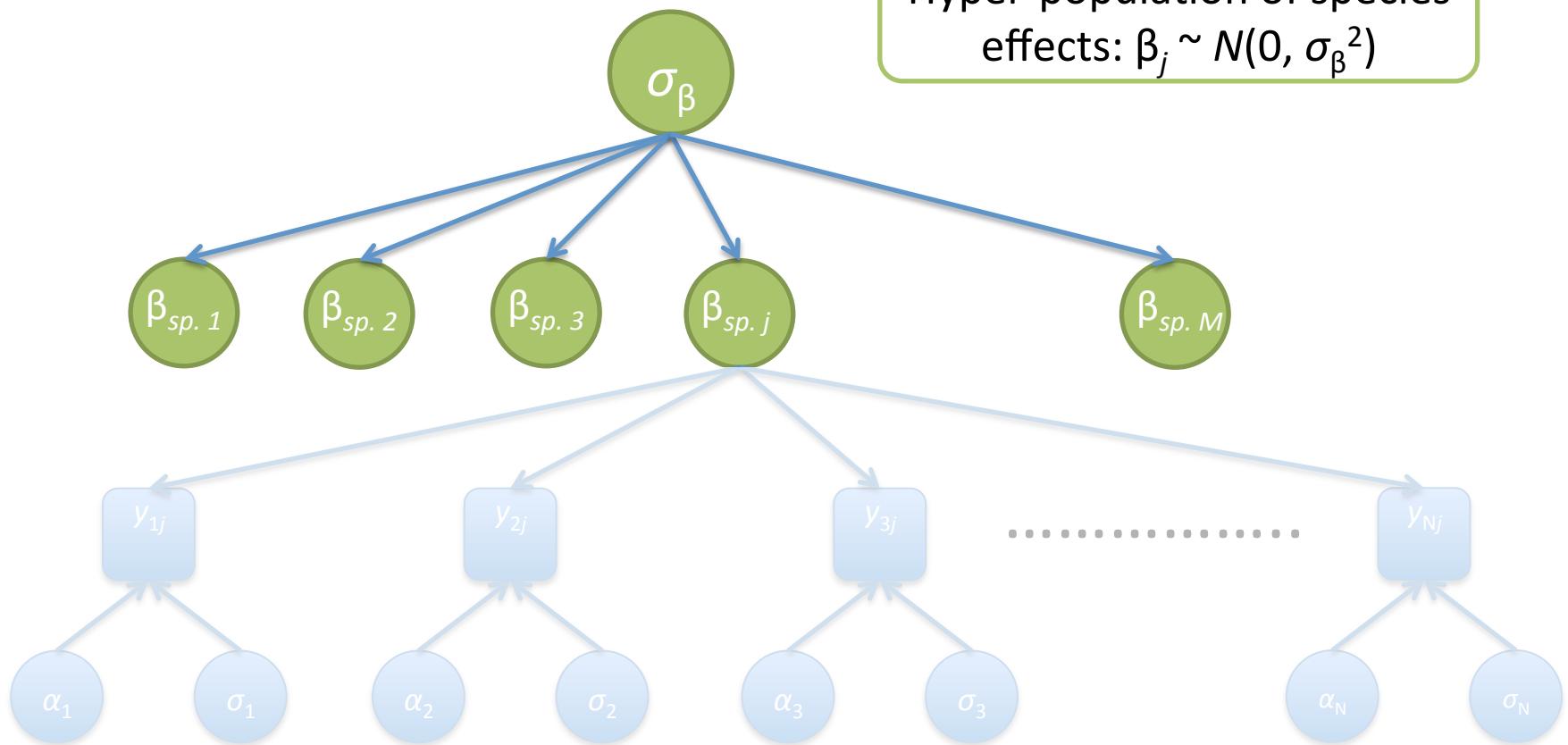
$\beta_j > 0 \Rightarrow$ species j typically tolerant

.....
and an error term

generally accepted
parameters; although
can be used with
suitable.

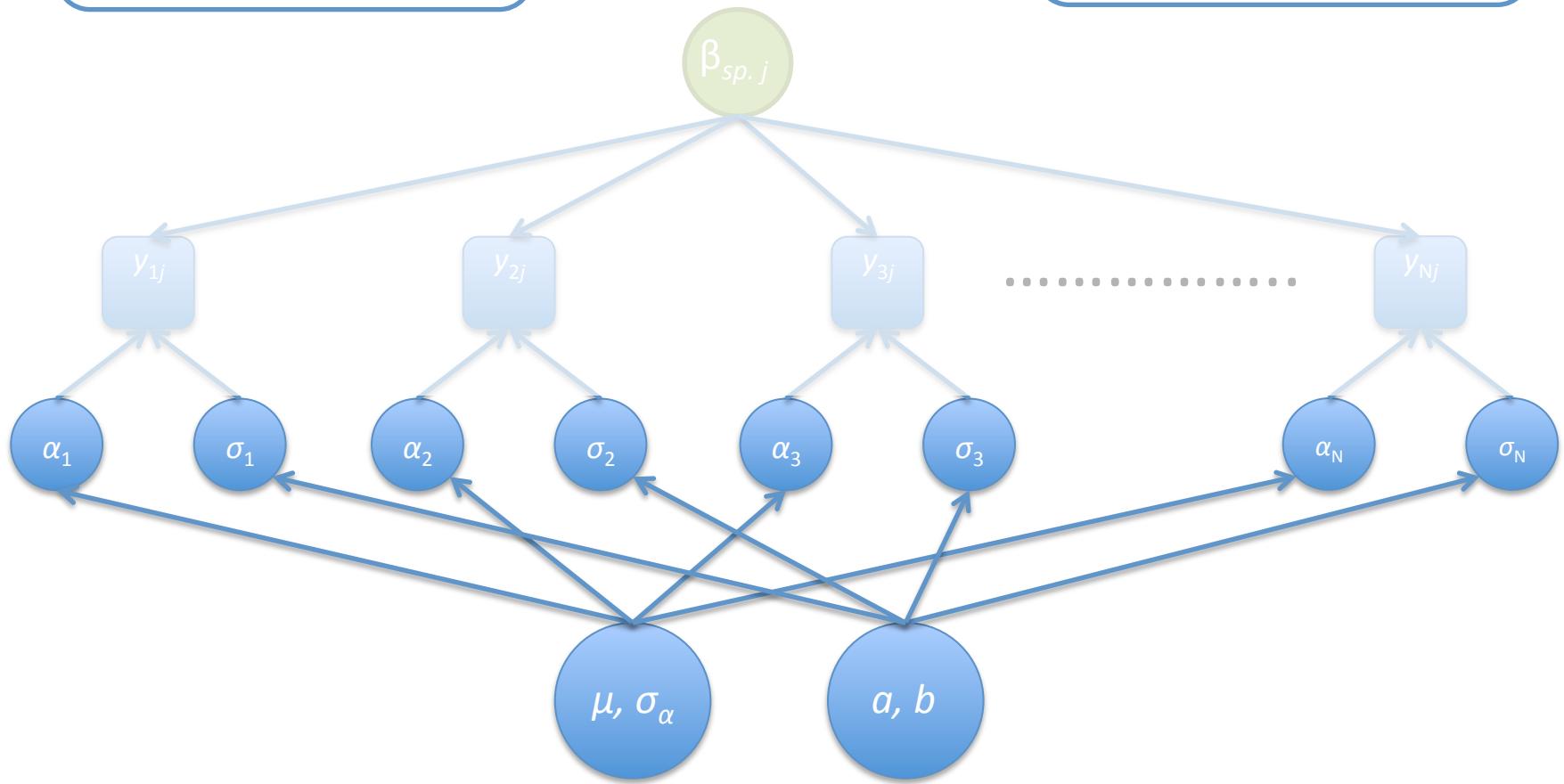


Hyper-population of species
effects: $\beta_j \sim N(0, \sigma_\beta^2)$



Hyper-population of
chemical effects:
 $\alpha_i \sim N(\mu, \sigma_\alpha^2)$

Hyper-population of
interspecies variances:
 $\sigma_i^{-2} \sim \Gamma(a, b)$



Bayesian Analysis

- Need to ensure propagation of $> 1^{\text{st}}$ level uncertainty.
- Update prior distributions about the hyper-parameters using observed data to retrieve posterior distributions.
- Use posterior distributions to make hazard assessment inferences for *retrospective* and *prospective* chemical assessments.

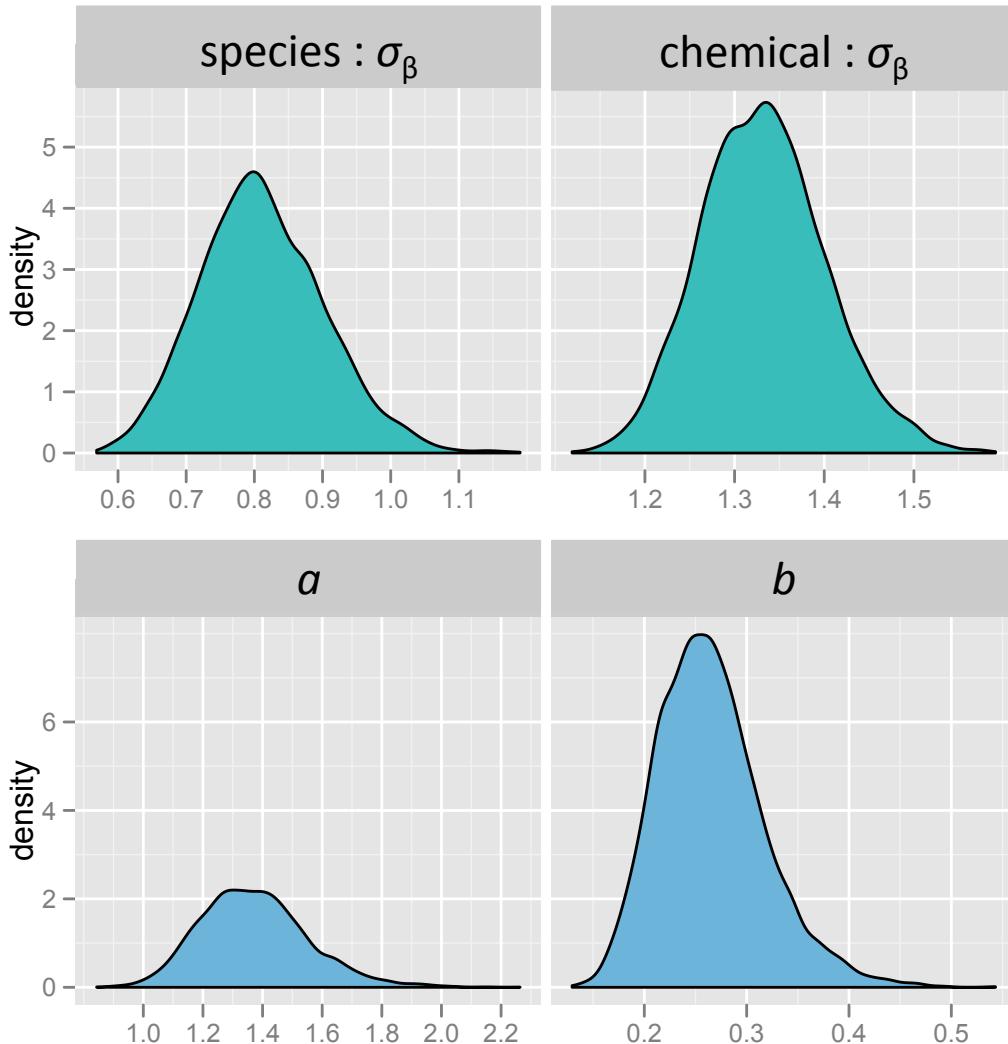
Example

- Ecotoxicity database extracted from the U.S. EPA Web-ICE database. 1600 $E(L)C_{50}$ values (lethality or immobility) spanning 201 chemicals (each with $n_i \geq 5$) and 77 species.

<http://www.epa.gov/ceampubl/fchain/webice/>

- Prior distributions chosen to closely represent ‘ignorance’ (so-called ‘*non-informative*’).

Hyper-parameter Posterior Distributions



median (+ 95% credible interval)

σ_β : 0.81 (0.65, 1.01)

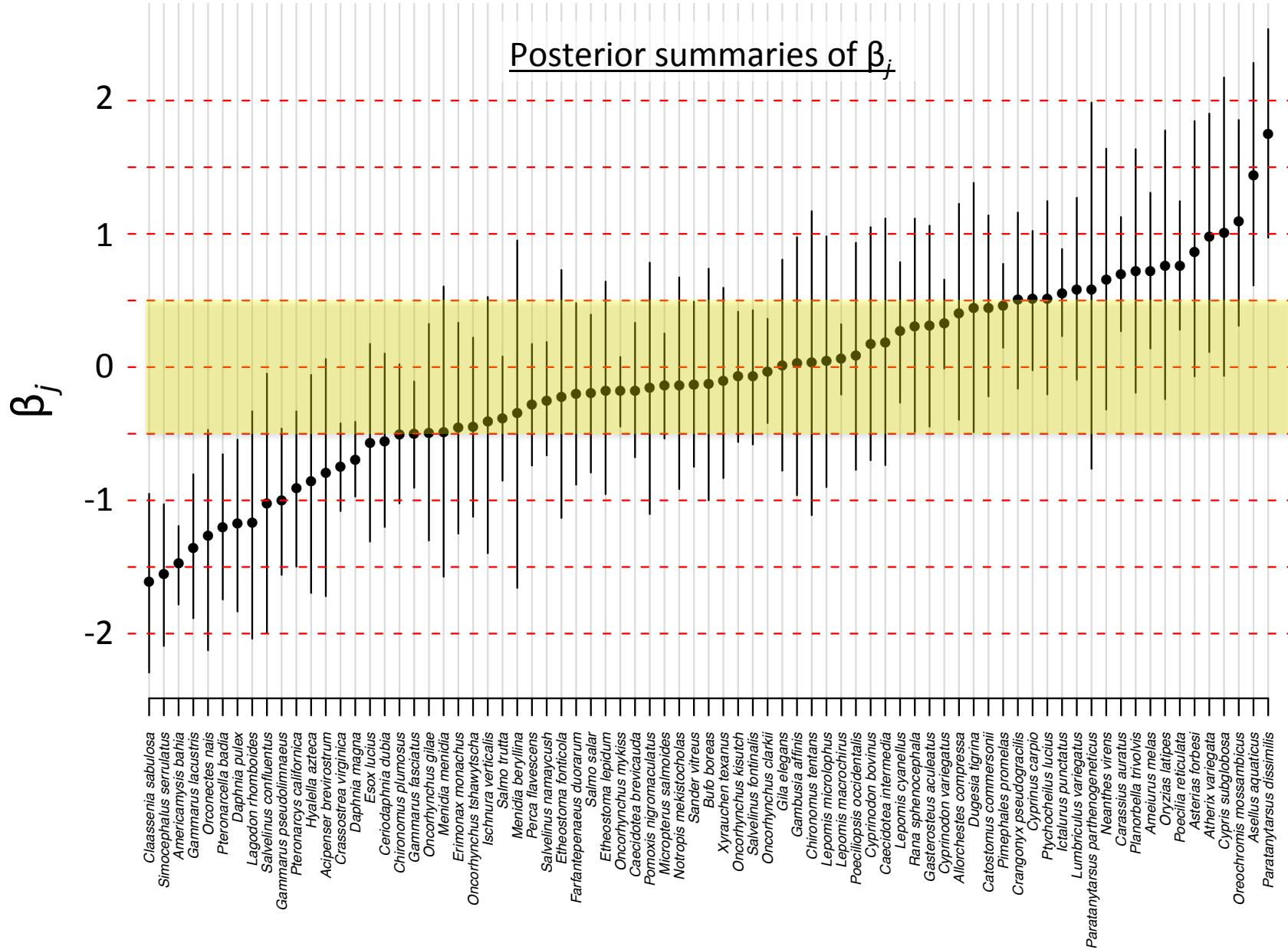
σ_α : 1.33 (1.21, 1.48)

⇒ more variation between chemicals

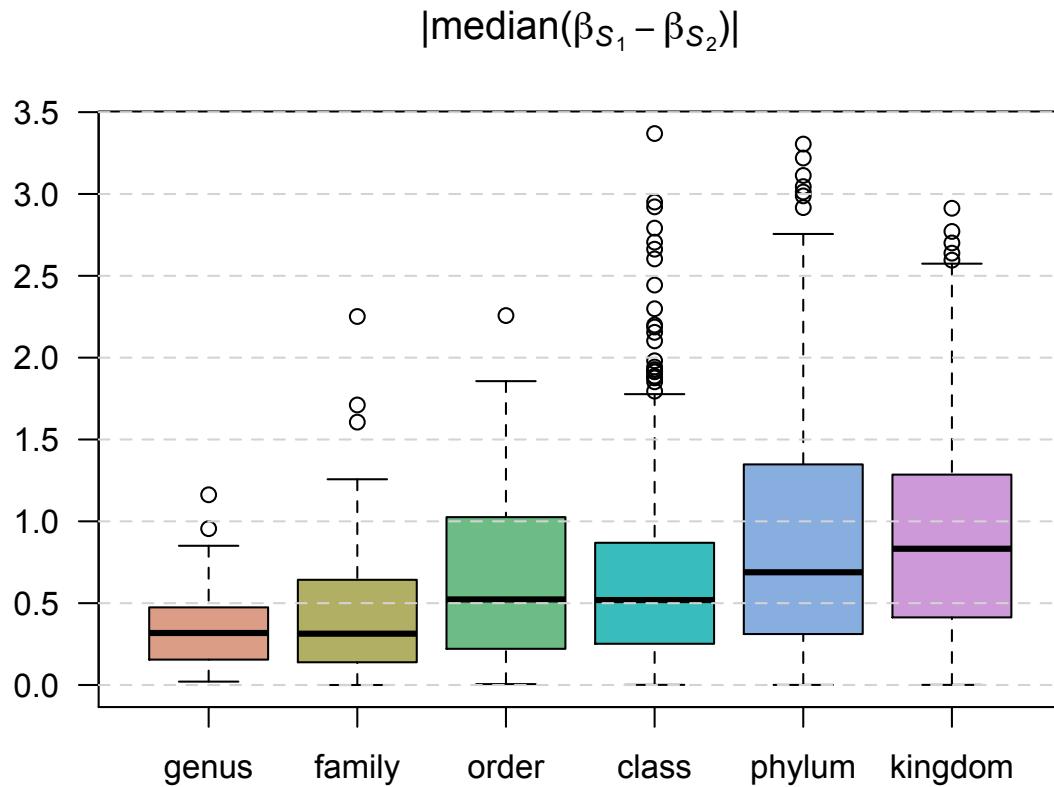
Estimates consistent with *European Food Safety Authority* (2005, *EFSA J. 301*, pp. 1-45) report.

Equivalent to ≈ 3 additional measurements → stabilizes interspecies variance estimate.

Evidence that some species are more sensitive than others

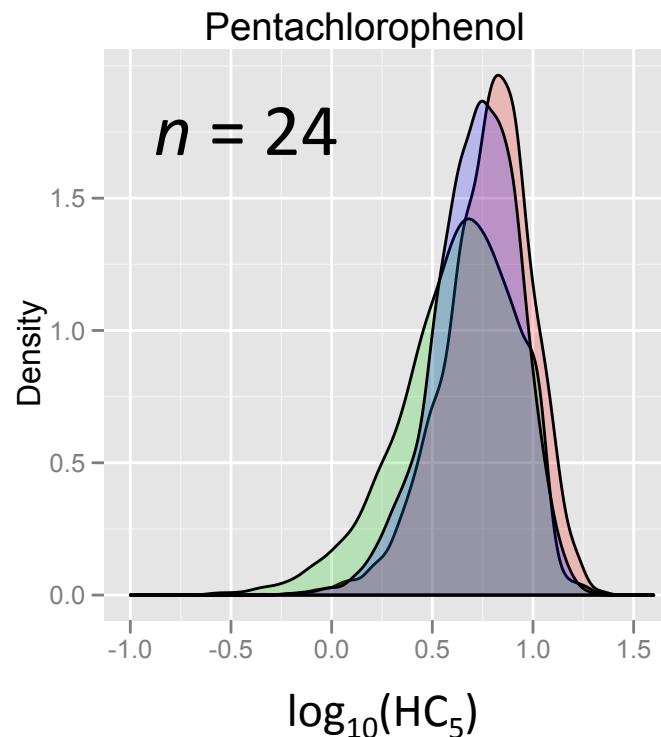
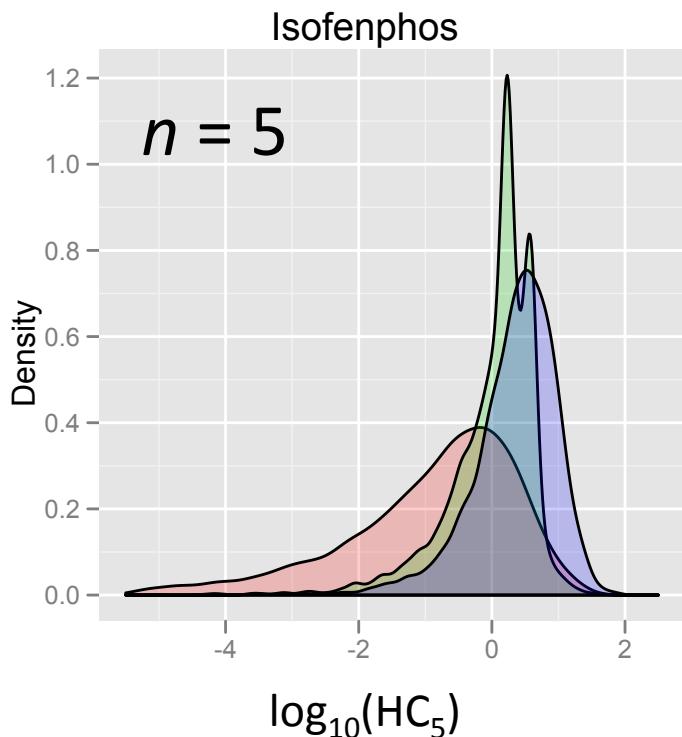


The Role of Taxonomy



The more taxonomically spread species are in an SSD, the larger the interspecies variance will be.

Posterior HC₅ Distributions



Model Assumption

- Status quo
- Extrapolate
- Interpolate

Aldenberg &
Jaworska (2000);
EES

Status quo = REACH Technical Guidance Document with log-normal SSD

Extrapolate = ecosystem is an infinitely large collection of species

Interpolate = ecosystem comprised of 77 species listed in database

} hierarchical
models

Conclusions

- The SSD concept is not defunct!
- Hierarchical modelling and Bayesian statistics open up the option for ‘better’ modelling with transparent uncertainty propagation.
- Useable for multiple-hypothesis testing and risk management.

Acknowledgements

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[DISCUSSION] Mick Hamer (Syngenta) and Malyka Galay-Burgos (ECETOC)

Existing Hierarchical Approaches

- Luttik & Aldenberg (1997), *ET&C*
- European Food Safety Authority (2005)
- Jager *et al.* (2007), *EES*
- Morton (2008), *Environmetrics*
- U.S. EPA Web-ICE Program(?)

Common theme: use data from multiple chemicals to improve future risk assessments