

# What can we learn from landscape eco(toxico)logy for regulatory ecological risk assessment?

Ralf B. Schäfer Quantitative Landscape Ecology

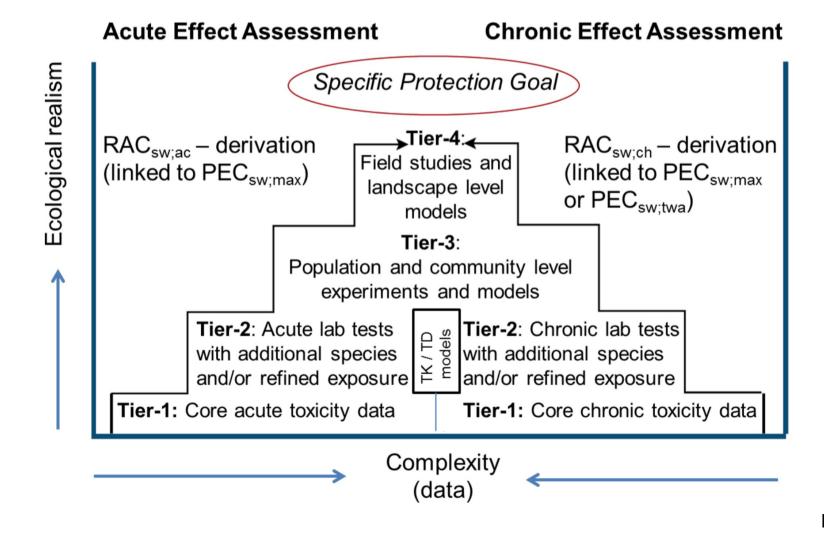


## Aims of pesticide risk assessment

 Regulation EC 1107/2009 repealing EC 91/414: no unacceptable effects on environment, ecosystems and biodiversity

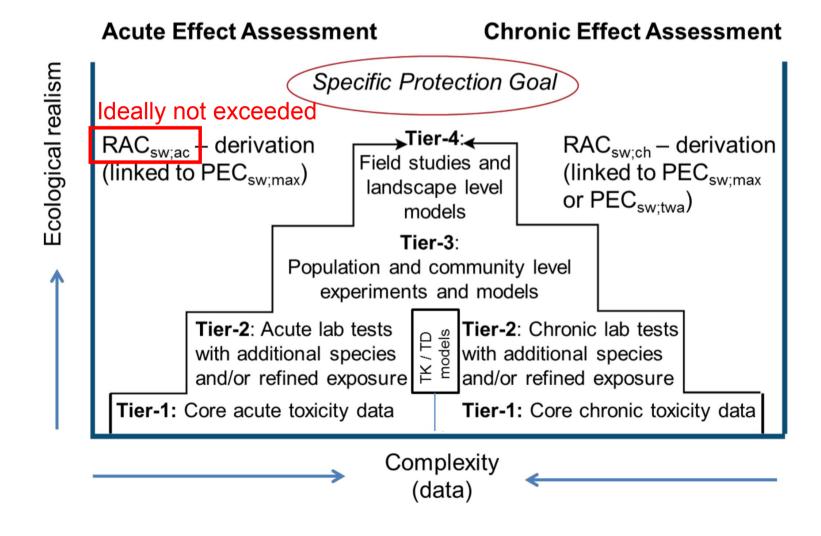
## Aims of pesticide risk assessment

- Regulation EC 1107/2009 repealing EC 91/414: no unacceptable effects on environment, ecosystems and biodiversity
- Tiered assessment



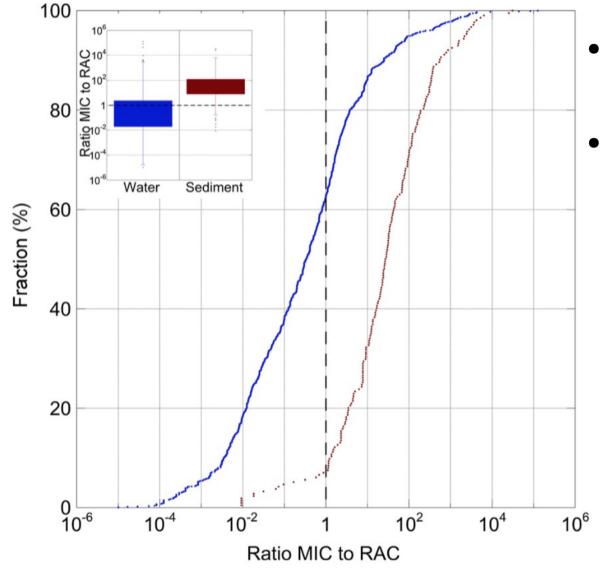
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- Tiered assessment



## RACs are frequently exceeded

Meta-analysis of EU insecticide concentrations (165 studies)

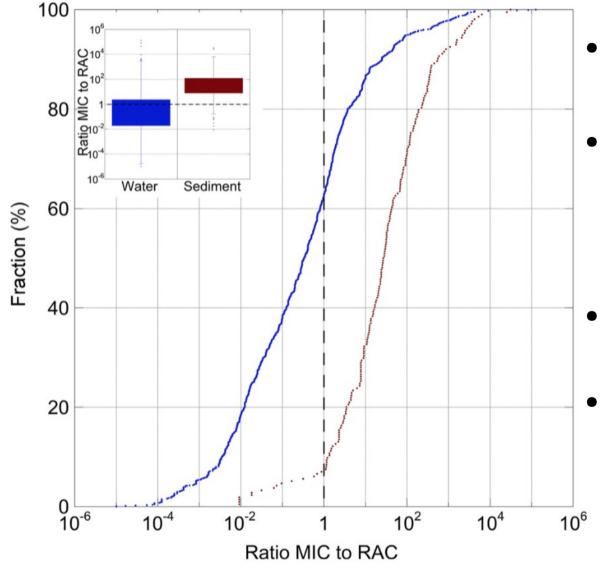


- ~ 40% exceedance of RAC<sub>SW</sub>
- Exceedance in 55% of sites (n = 385)

Stehle & Schulz 2015 Environ Sci Pollut Res

## RACs are frequently exceeded

Meta-analysis of EU insecticide concentrations (165 studies)



- ~ 40% exceedance of RAC<sub>SW</sub>
- Exceedance in 55% of sites (n = 385)

- Analysis of German monitoring data
  - Exceedance in 25% of sites (n = 3049)

Stehle & Schulz 2015 Environ Sci Pollut Res

### Reasons for RAC exceedance

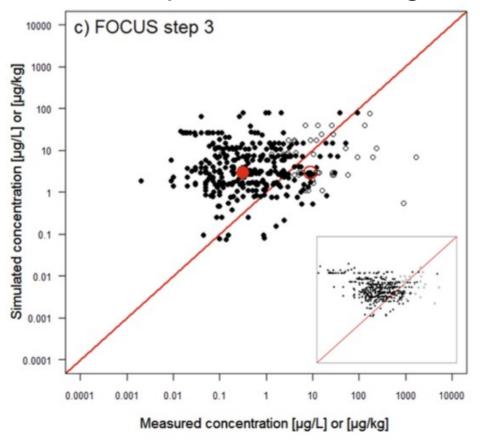
- Non-compliance with risk mitigation conditions
  - → Unlikely given the magnitude of exceedances?

### Reasons for RAC exceedance

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   → Unlikely given the magnitude of exceedances?
- Unreliable exposure characterisation?

42% underprediction for insecticides

e) FOCUS step 3 (realistic) Simulated concentration [µg/L] or [µg/kg] 0.001 Measured concentration [µg/L] or [µg/kg] 13% underprediction for fungicides

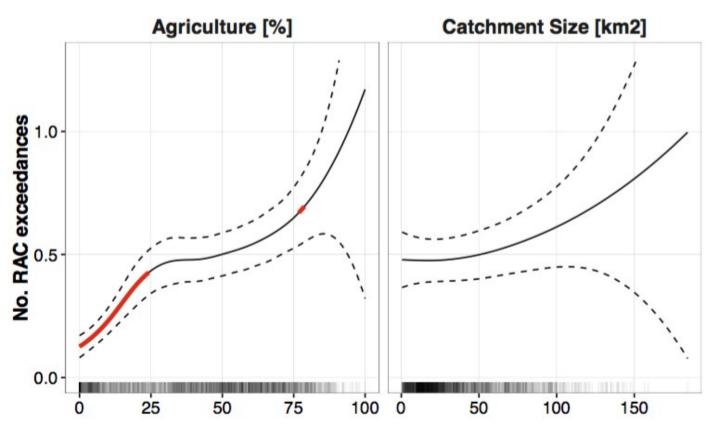


Knäbel et al. 2012 Environ Sci Technol

Knäbel et al. 2014 Environ Sci Technol

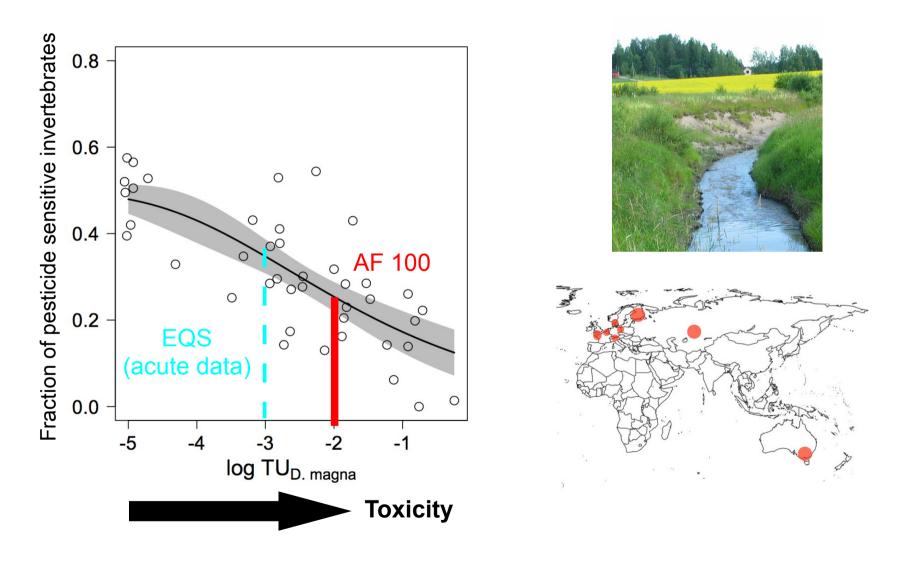
### Reasons for RAC exceedance

- Non-compliance with risk mitigation conditions
   → Unlikely given the magnitude of exceedances?
- Unreliable exposure characterisation?
- Landscape influences level of RAC exceedance
- Analysis of German monitoring data
- 42,200 grab samples, all-season

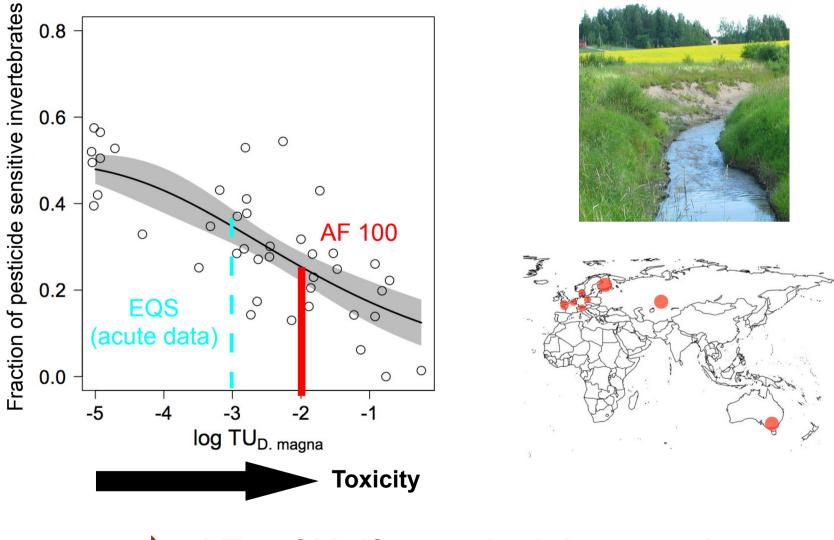


Szöcs et al. 2016 Environ Sci Technol under review

Meta-analysis: 8 field studies with 111 water bodies

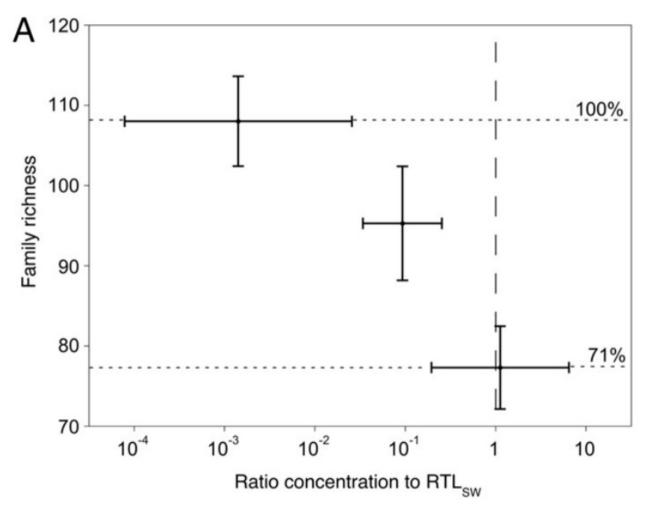


Meta-analysis: 8 field studies with 111 water bodies

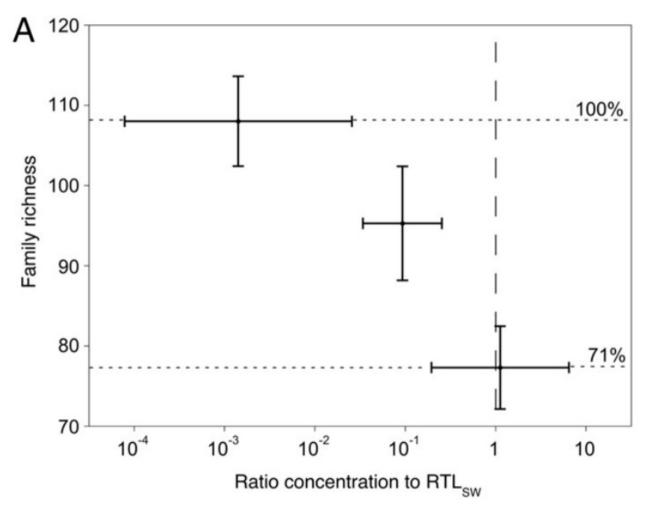


AFs of Uniform principles not always protective

Re-analysis of field study data from South-East Australia and France (Beketov et al. 2013 PNAS)



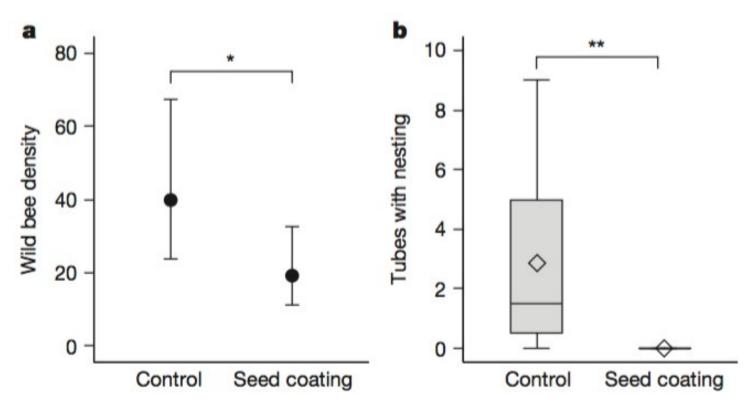
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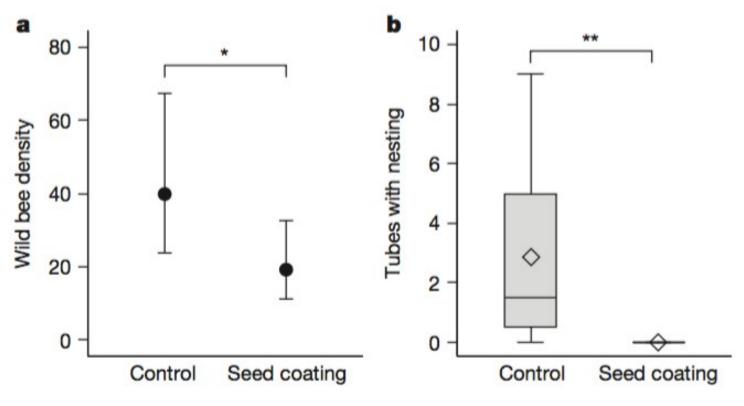
30% loss of invertebrate biodiversity at RAC

Blind field study with 8 paired sites (control – treatment with pyrethroid and neonicotinoid)



Rundlöf et al. 2015 Nature

Blind field study with 8 paired sites (control – treatment with pyrethroid and neonicotinoid)



Rundlöf et al. 2015 Nature



Pollinators not sufficiently protected?

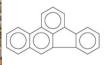
- Discrepancy between effect thresholds of model systems and field studies found for pesticides, ionizing chemicals, and metals (Schäfer 2014 ET&C, Liess and Beketov 2011 Ecotoxicology)
  - → pond mesocosms unlikely protective for streams

- Discrepancy between effect thresholds of model systems and field studies found for pesticides, ionizing chemicals, and metals (Schäfer 2014 ET&C, Liess and Beketov 2011 Ecotoxicology)
- Multiple stressors

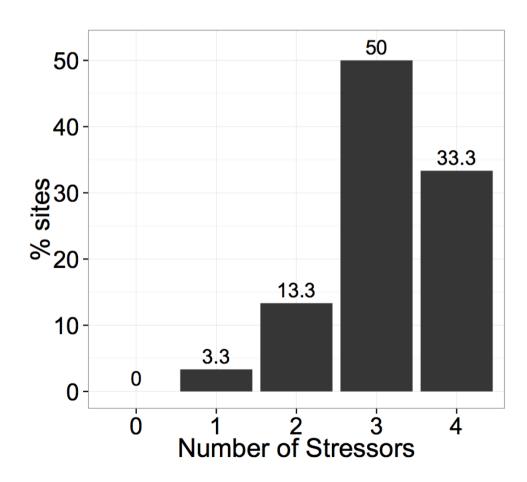
 Analysis of German monitoring data for 4 stressors





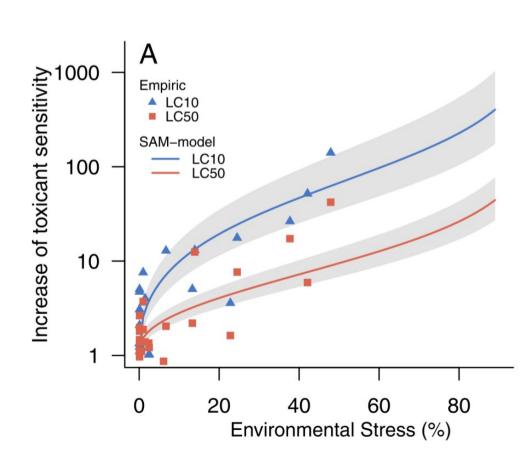




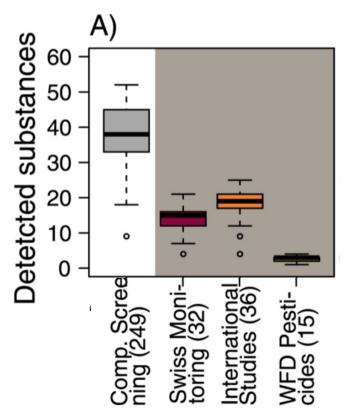


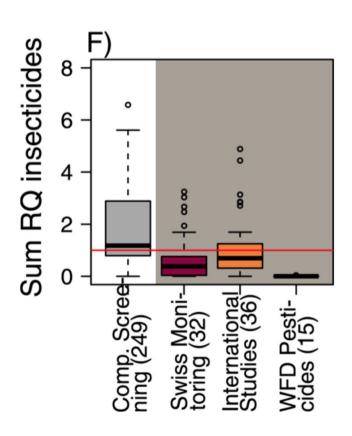
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- Multiple stressors

 Meta-Analysis of stressor pairs (env. stressor & chemical) with complete dose-response curves (n = 23)



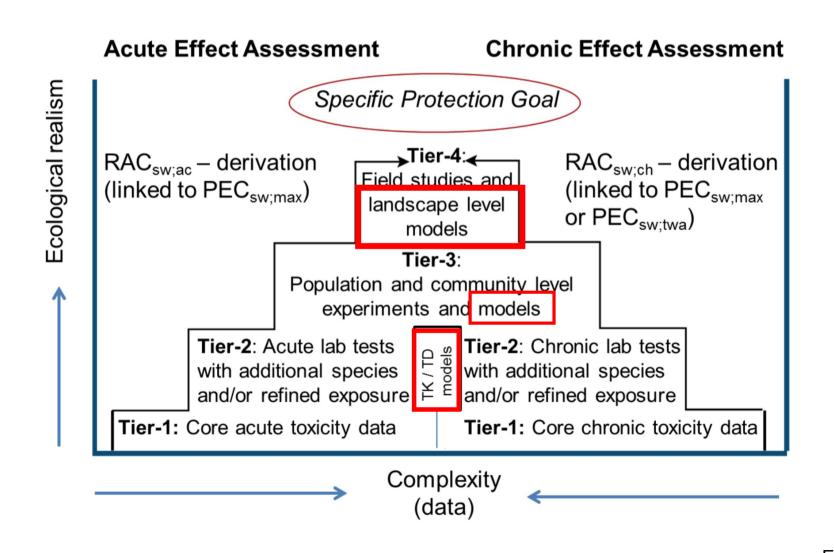
- Discrepancy between effect thresholds of model systems and field studies found for pesticides, ionizing chemicals, and metals (Schäfer 2014 ET&C, Liess and Beketov 2011 Ecotoxicology)
- Multiple stressors
- Mixture toxicity



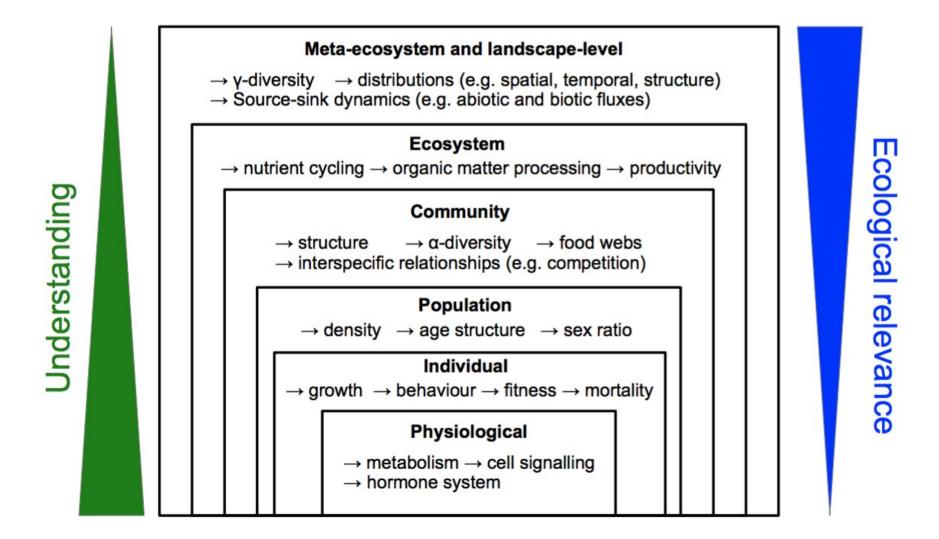


## Improving prospective risk assessment

Refining risk assessment via modelling?



Understanding of processes decreases with complexity
 e.g. 6 competing biodiversity theories McGill 2010 Ecol. Letters
 12 mathematical models for community assembly Presley et al. 2010 Oikos



Example: Models for effects of Bt maize on non-target butterflies





Flowering Maize

Peacock butterfly (Inachis)

#### **Journal of Applied Ecology**



Journal of Applied Ecology 2012, 49, 29-37

doi: 10.1111/j.1365-2664.2011.02083.x

Estimating the effects of Cry1F Bt-maize pollen on non-target Lepidoptera using a mathematical model of exposure

Example: Models for effects of Bt maize on non-target butterflies





Flowering Maize

eacock butterfly (Inachis)

#### **Journal of Applied Ecology**

British Ecological Society

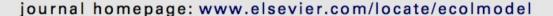
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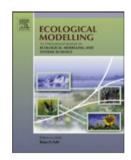
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#### **Ecological Modelling**





Increased mortality is predicted of *Inachis io* larvae caused by Bt-maize pollen in European farmland

Example: Models for effects of Bt maize on non-target butterflies





Flowering Maize

Peacock butterfly (Inachi

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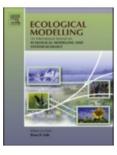
Ecological Modelling 268 (2013) 103-122



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#### **Ecological Modelling**





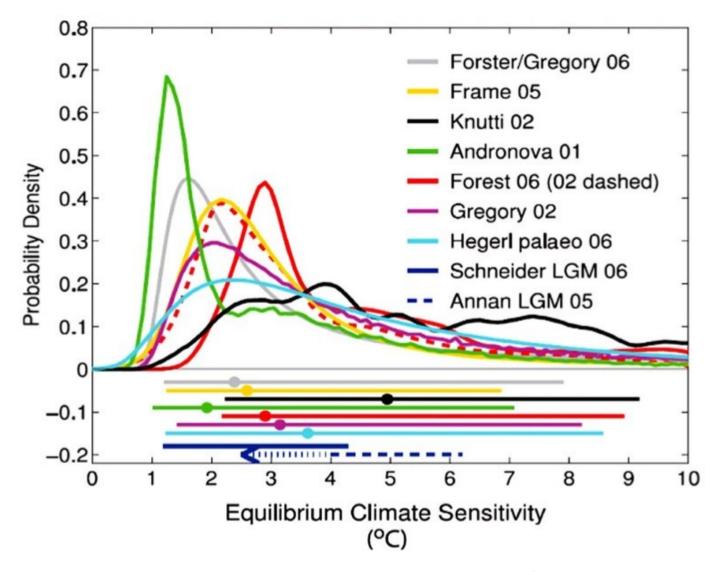
#### Letter to the Editor

No evidence requiring change in the risk assessment of *Inachis io* larvae



- 1. The use of data by Holst et al.
- 1.1. Pollen deposition on host plants

Example: Climate models prediction for doubling of CO<sub>2</sub>





 High risk of bias and uncertainty of models increases with level of complexity → Not suitable to derive precise regulatory thresholds for communities and biodiversity

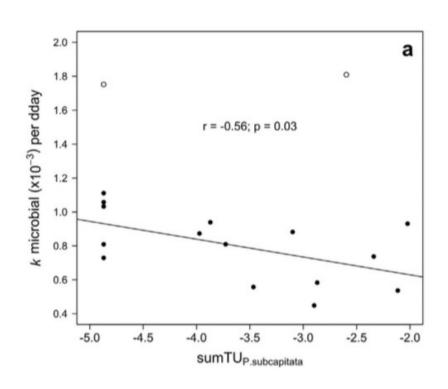
- High risk of bias and uncertainty of models increases with level of complexity → Not suitable to derive precise regulatory thresholds for communities and biodiversity
- Nevertheless, models can be useful to:
  - identify research gaps
  - explore scenarios (e.g. multiple stressors)
  - test hypotheses and understand patterns
  - estimate regulatory thresholds for individuals (e.g. TK-TD) and populations, where community-processes are largely irrelevant

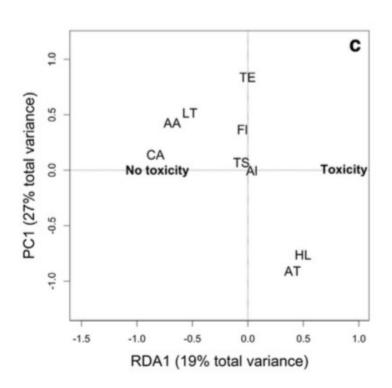
## Improving prospective risk assessment

#### Include all relevant receptors

Aquatic hyphomycetes?

Reduction in ecosystem functioning (microbial decomposition) along fungicide toxicity gradient, toxic effect confirmed in laboratory





Fernandez et al. 2015 Sci. Tot. Environ.

## Improving retrospective risk assessment

Post-registration monitoring

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 Wouldn't this be extremely costly?

## Improving retrospective risk assessment

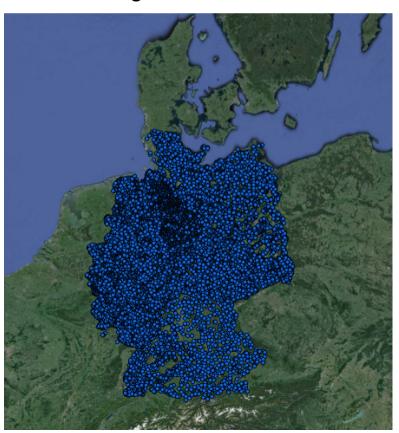
Post-registration monitoring
 Wouldn't this be extremely costly?

Join forces with other programs

Demonstration farms for integrated plant protection



Biological & chemical monitoring networks for WFD



## Integrating regulatory frameworks?

- Pesticides are managed prospectively in authorisation and retrospectively in Water Framework Directive
- Protection goals vary and could be more specific
- RACs and EQS (and pesticides considered) can differ between member states

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- Pesticides are managed prospectively in authorisation and retrospectively in Water Framework Directive
- Protection goals vary and could be more specific
- RACs and EQS (and pesticides considered) can differ between member states
  - Water body-type specific protection goals?
  - Harmonise and improve consistency

"To minimize risks of negative impacts on humans and the environment, the most stringent restriction in all legislations should be adopted."

Brack et al. 2016 Sci. Tot. Environ. in press

### Conclusions

- Risk assessment does not stand reality check (frequent RAC exceedances)
- Current approach not fully protective
- Models no panacea
- Field-relevance requires field observation
- Integration with WFD may fix several shortcomings

## Thank you!



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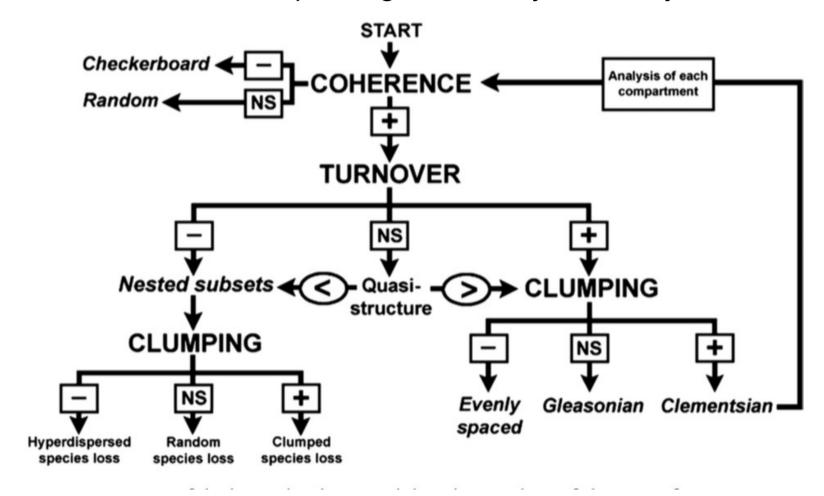
@LandscapEcology

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## Mathematical underpinning for pesticide effects on community assembly unclear

12 mathematical models for explaining community assembly

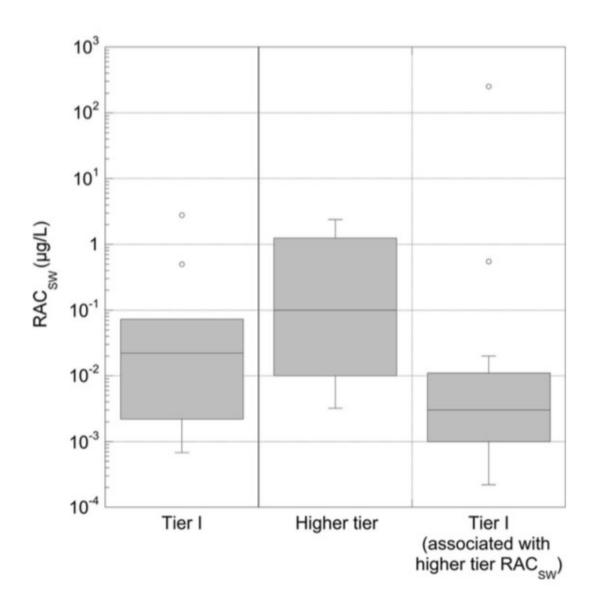


## Mathematical underpinning for pesticide effects on biodiversity unclear

#### Six prominent unified biodiversity theories

Unified theory	Key references	Input parameters	Math	Spatial model	Assertion 1 Intraspecific spatial	Assertion 2 Interspecific variation in global abundance	Assertion 3 Interspecific independence
Continuum	(Gauch & Whittaker 1972; Hengeveld <i>et al.</i> 1979; Coleman <i>et al.</i> 1982; McGill & Collins 2003)	$S$ , $N_{\text{max}}\sim$ , $\sigma\sim$ , $A$	Probability theory (analytical)	Density surface (aggregate)	Peak and tail	$N_{\rm max}$ and $\sigma$ are sampled from distribution (input)	Each peak is located according to a Poisson process (random wrt other species)
Neutral	(Caswell 1976; Hubbell & Foster 1986; Bell 2000, 2001; Hubbell 2001)	$S$ or $\Theta$ , $N$ , m	Birth-death (analytical) + Lattice (simulation)	Lattice (individual)	Dispersal-limited	Metacommunity processes create logseries regional abundances (derived)	Each lattice cell can be populated by any species
Metapopulation	(Hanski & Gyllenberg 1997)	S, A~, w~	Levins metapopulation differential equation (analytical)	Probability present   A (aggregate)	Incidence	Density of species $w_i$ is sampled from loguniform distribution (input)	Presence of one species on a patch is modelled independent of any other species allowing simple summation
Generalized Fractal	(Harte et al. 1999; Green et al. 2003; Storch et al. 2008)	S, l,∼, D;∼	Hierarchical division (simulation)	Spatially explicit (aggregate)	Hierchically clumped	Each species is modelled with four nested levels of multiplication of a uniformly distributed random variables approaching a central limit theorem like situation (derived)	Each species is modelled and placed in space independent of other species allowing simple summation
Clustered Poisson	(Plotkin & Muller-Landau 2002; Plotkin <i>et al.</i> 2002; Morlon <i>et al.</i> 2008)	S, N, aggregation parameters	Neyman–Scott process	Point process (individual)	Explicitly aggregated	Regional species abundance distribution is specified (various used) (input)	Each species is its own clustered Poisson process without reference to other species, allowing simple summation
MaxEnt	(Harte et al. 2005; Pueyo et al. 2007; Harte 2008)	S, A, N, E	MaxEnt (analytical)	Not spatially explicit (aggregate)	Derived exponential P	MaxEnt plus a constraint on total abundance gives a logseries SAD (derived)	$P_i(n A)$ is independent of other species, allowing simple summation

## Past mesocosm studies increased RAC more than an order of magnitude



Stehle & Schulz 2015 Environ Sci Pollut Res

# Mesocosm studies would require about 9 replicates for reliable power, but would still not capture rare taxa that are important for ecosystem functions

Mouillot et al. 2013 PLOS Biology

