

Statistical Eco(-toxico)logy

Improving the Utilisation of Data for
Environmental Risk Assessment

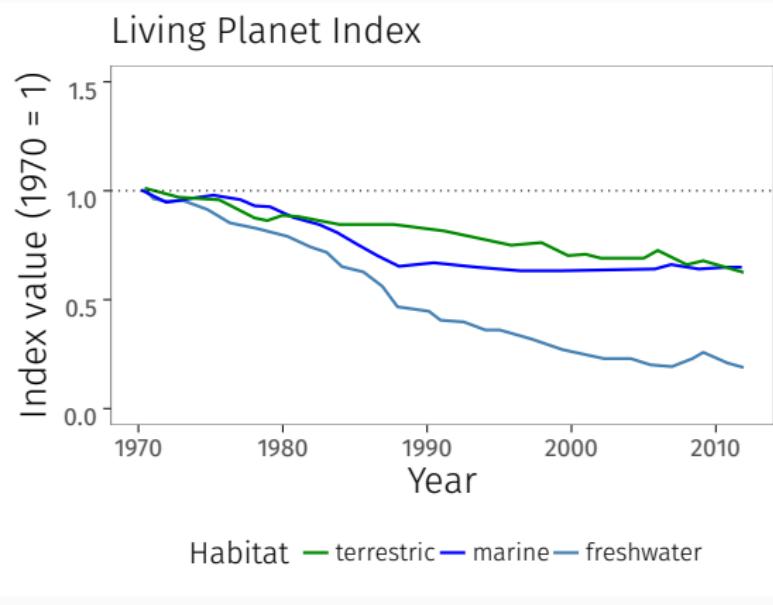
Eduard Szöcs

25th January 2017

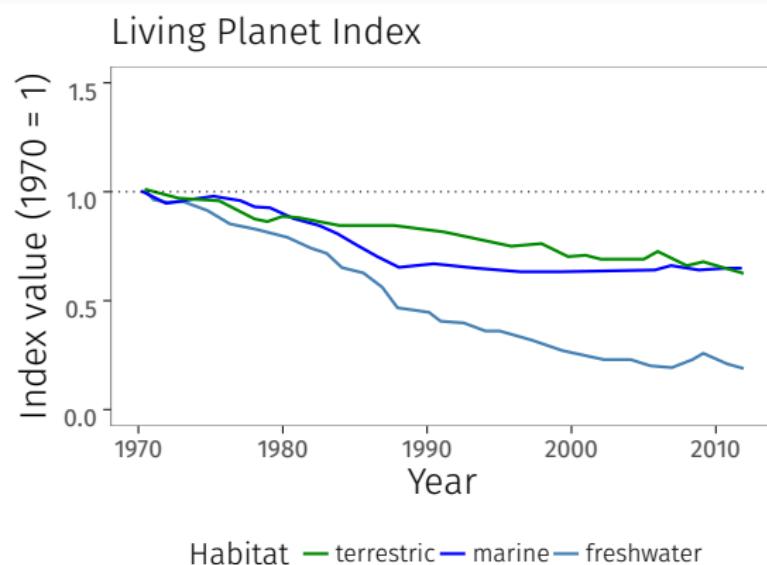
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Freshwater biodiversity is strongly declining



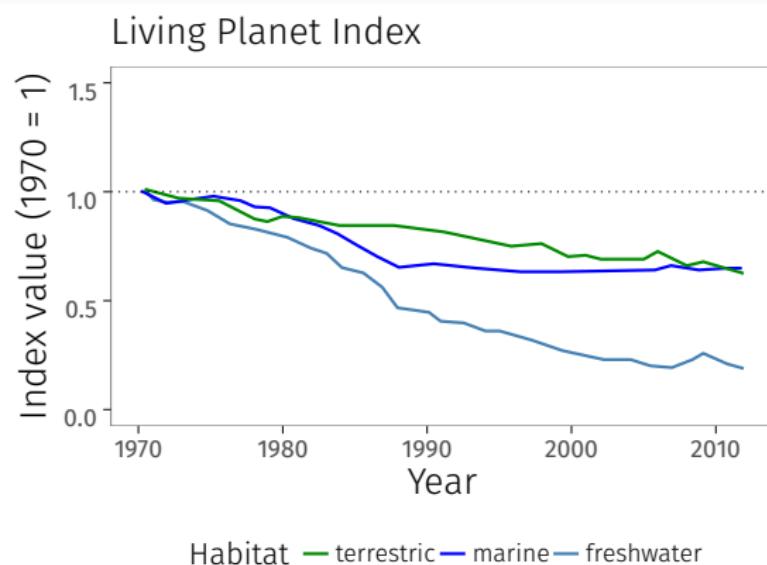
Freshwater biodiversity is strongly declining



Reasons

- Habitat loss
- Overexploitation
- Pollution
- Invasive species

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- **Pollution**
- Invasive species

Environmental Risk Assessment and Environmental Monitoring

Environmental Risk Assessment and Monitoring

Plant Protection Products
Regulation 1107/2009

Water Framework Directive
2000/60/EC

Environmental
Risk
Assessment

Environmental
Monitoring

Environmental Risk Assessment and Monitoring

Plant Protection Products
Regulation 1107/2009

Water Framework Directive
2000/60/EC

Effects

Environmental
Risk
Assessment

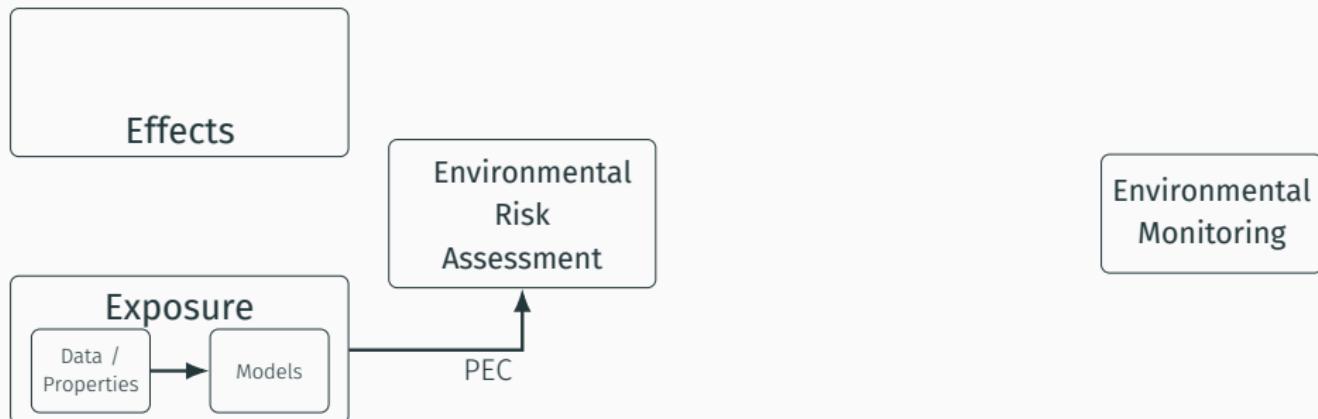
Exposure

Environmental
Monitoring

Environmental Risk Assessment and Monitoring

Plant Protection Products
Regulation 1107/2009

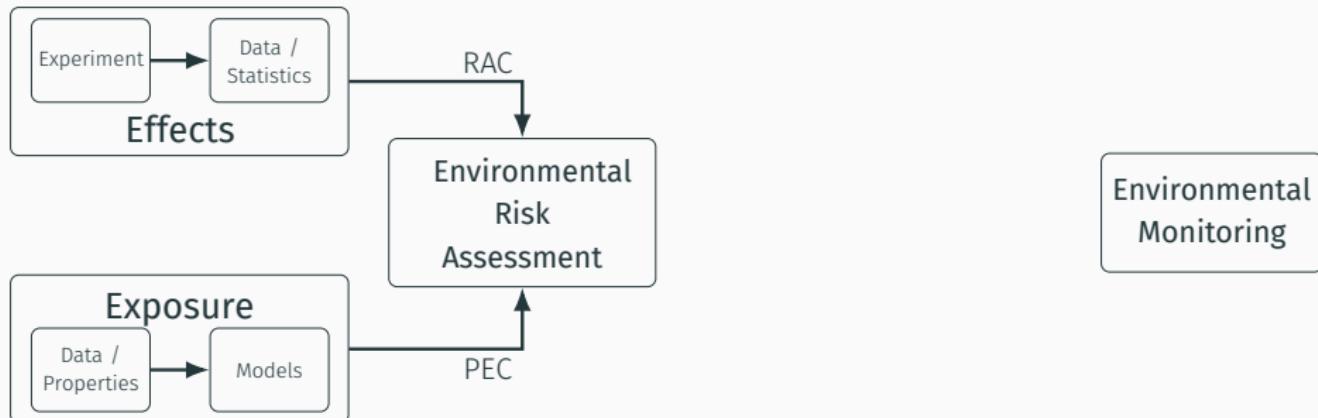
Water Framework Directive
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Environmental Risk Assessment and Monitoring

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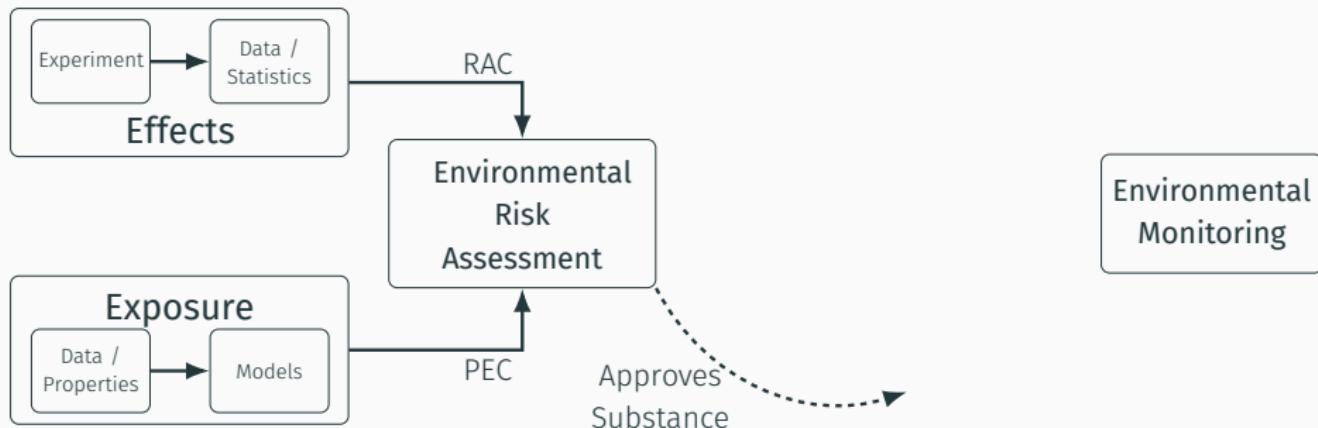
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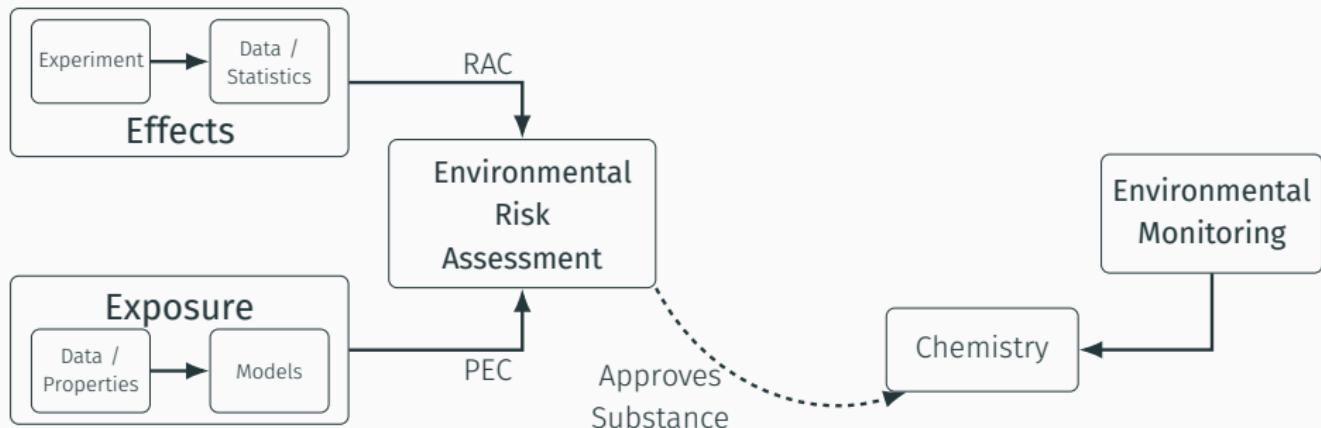
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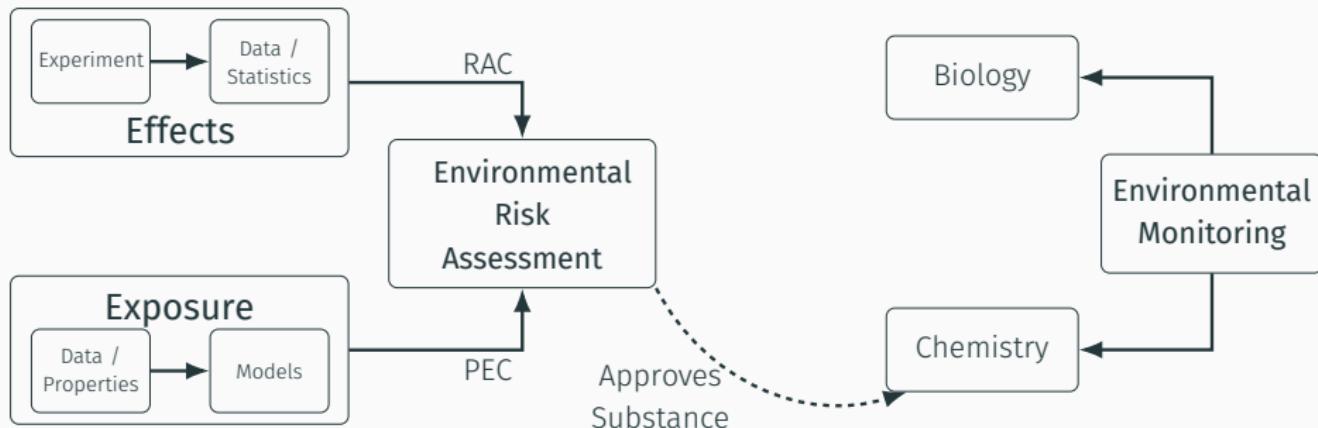
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Environmental Risk Assessment and Monitoring

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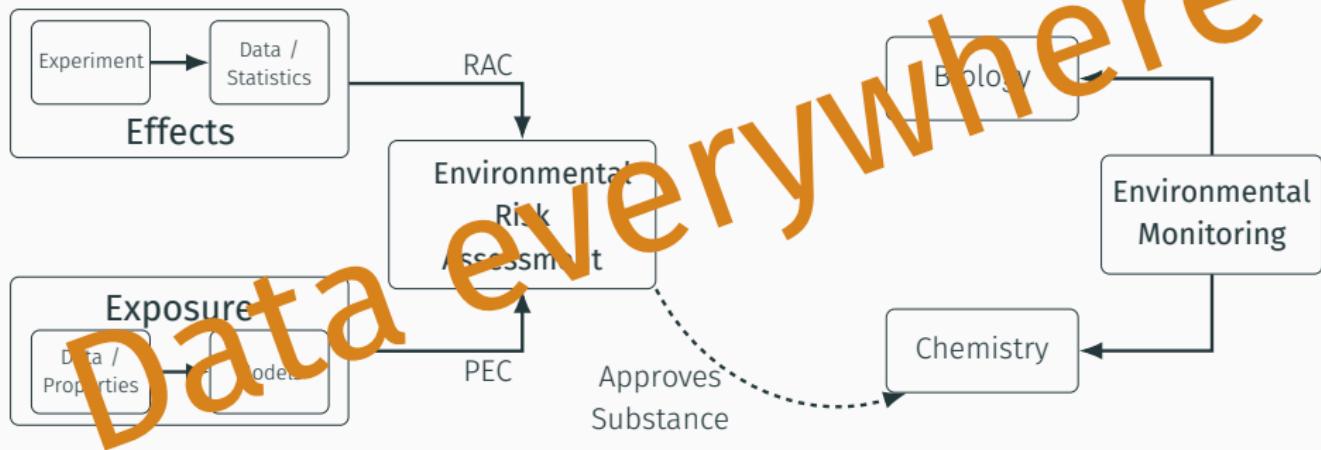
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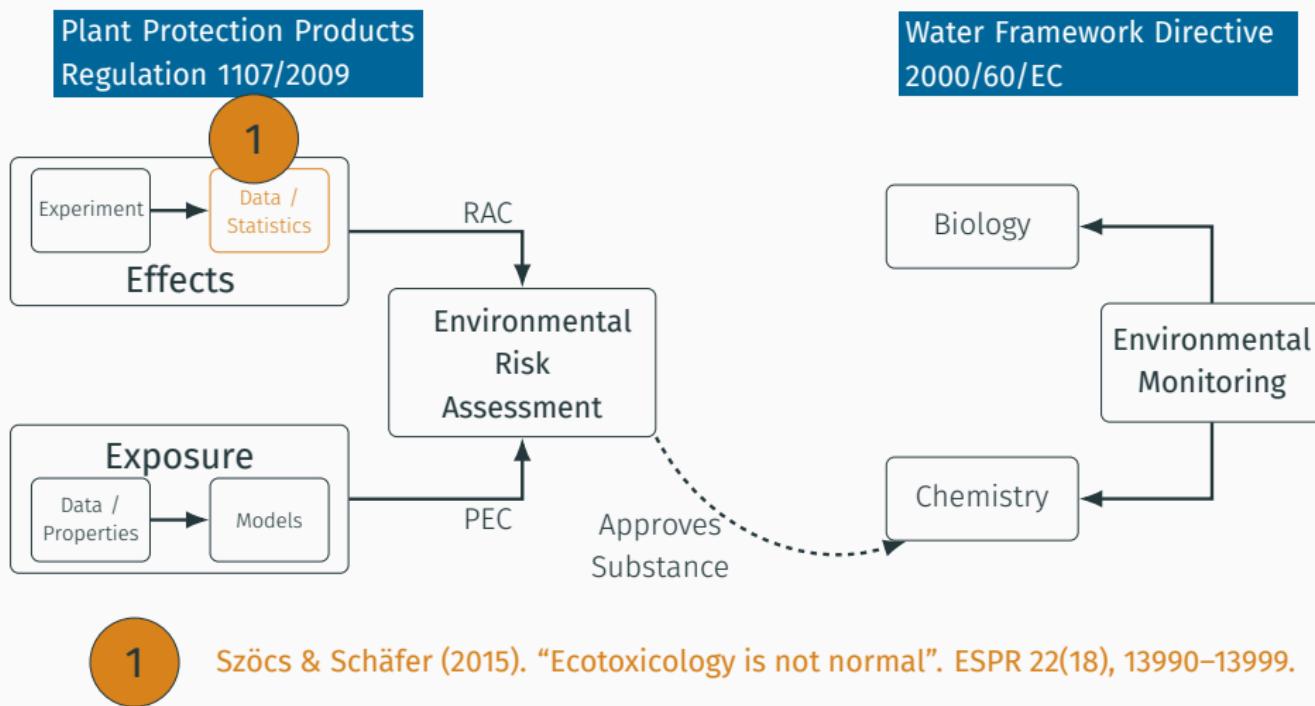
Environmental Risk Assessment and Monitoring

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Water Framework Directive
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Improving Statistics in ERA



1

Szöcs & Schäfer (2015). "Ecotoxicology is not normal". ESPR 22(18), 13990–13999.

Improving Statistics in ERA

Experiments in Effect Assessment



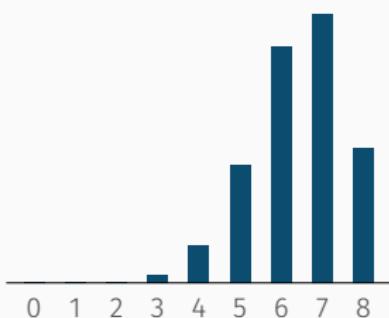
- Daphnia Test
- " x out of n survived"

Experiments in Effect Assessment

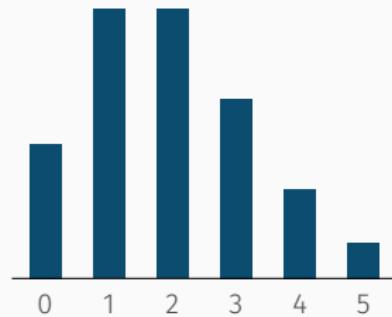


- Daphnia Test
- "*x out of n survived*"
- Mesocosm
- "*number of animals*"

Ecotoxicology is not normal

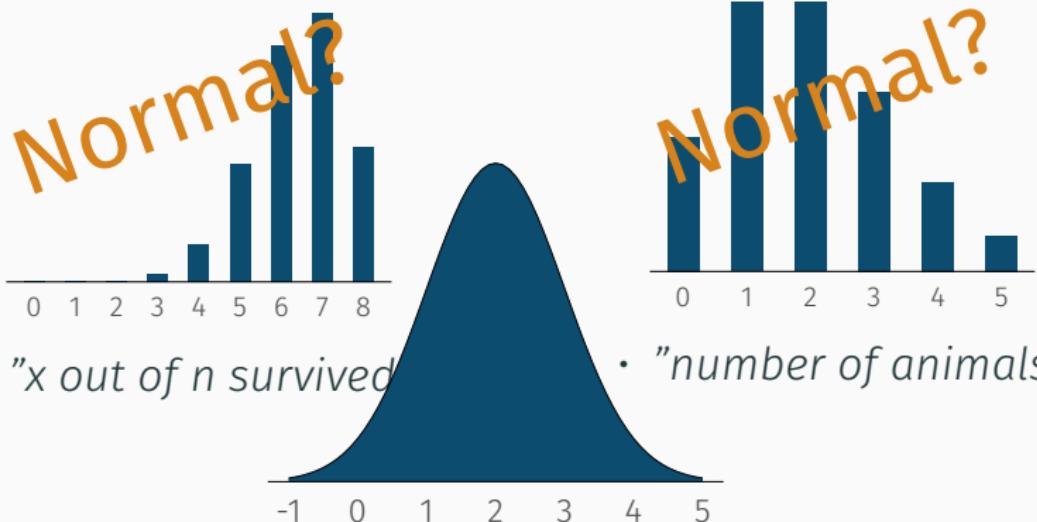


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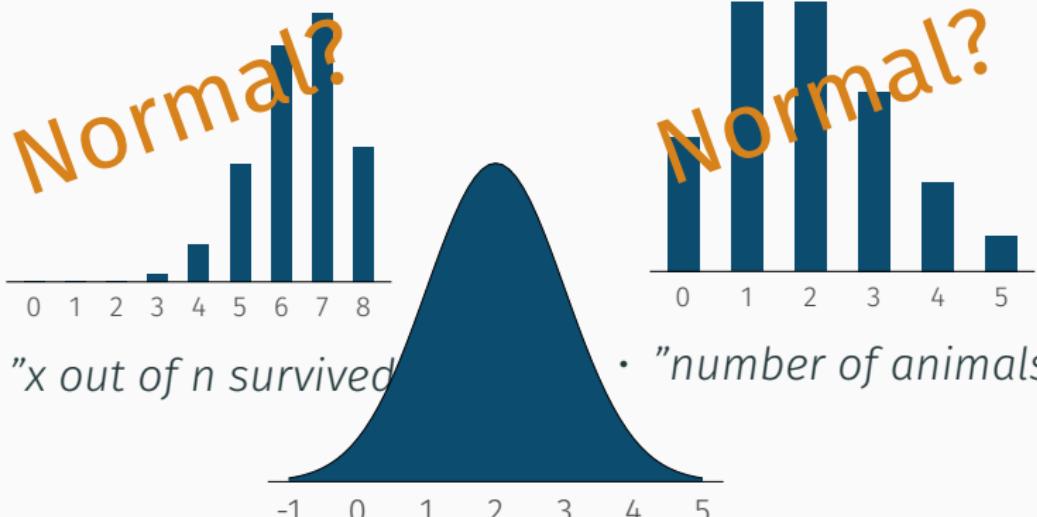


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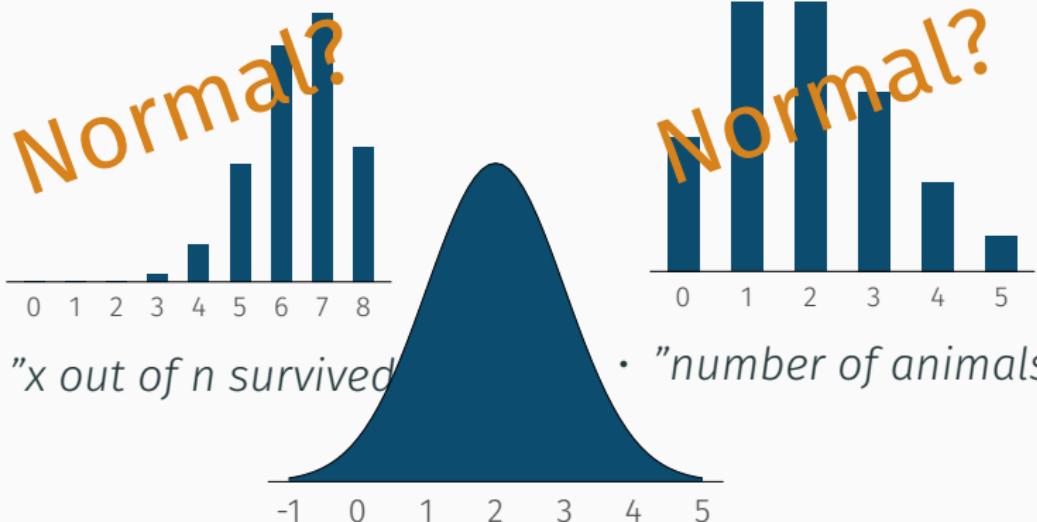


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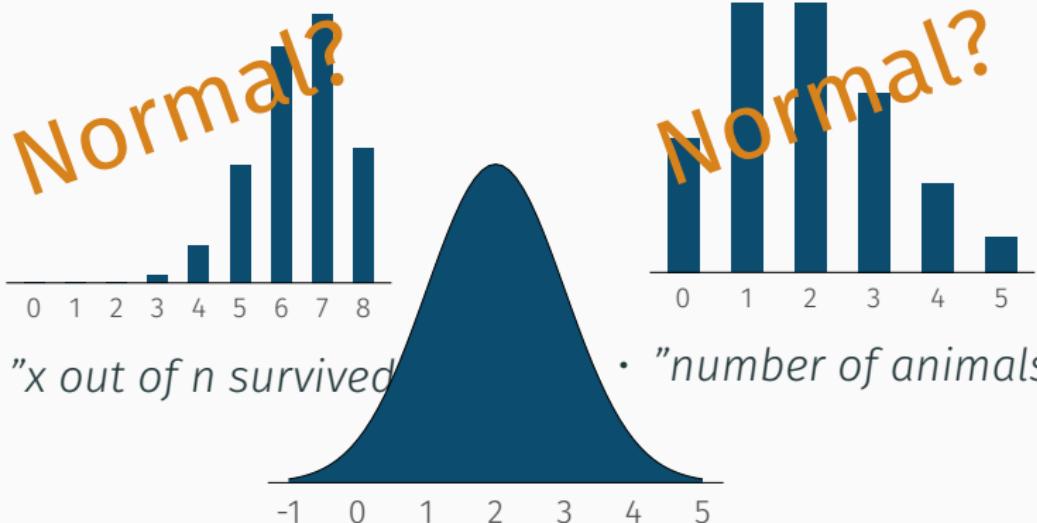
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- ignore?

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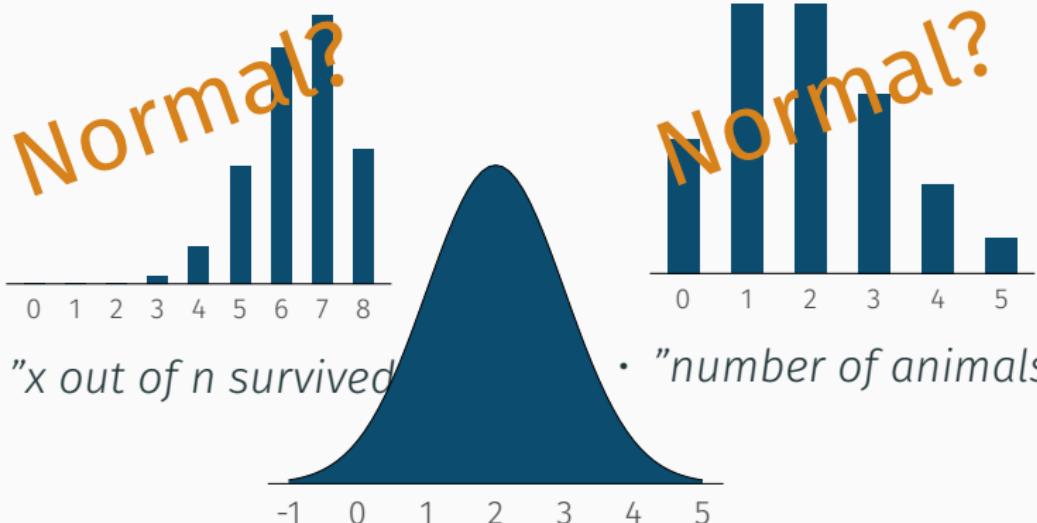
- " x out of n survived"
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- ignore?
- transform?

Ecotoxicology is not normal



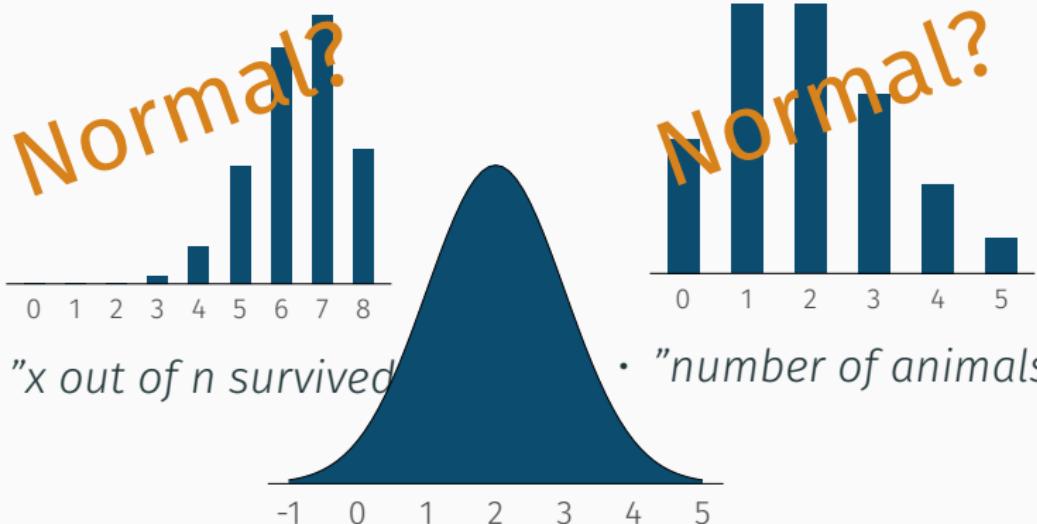
- " x out of n survived"
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- ignore?
- transform?
- non-parametric?

Ecotoxicology is not normal



- "x out of n survived"
- "number of animals"
- ignore?
- transform?
- non-parametric?
- Generalised Linear Model (GLM)?

Ecotoxicology is not normal



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- Generalised Linear Model (GLM)?

A recent history (uncomprehensive) of GLM in ecology

J. R. Statist. Soc. A,
(1972), **135**, Part 3, p. 370

370

Generalized Linear Models

By J. A. NELDER and R. W. M. WEDDERBURN

Rothamsted Experimental Station, Harpenden, Herts



A recent history (uncomprehensive) of GLM in ecology

Methods in Ecology and Evolution



Methods in Ecology & Evolution

doi: 10.1111/j.2041-210X.2010.00021.x

Do not log-transform count data

Robert B. O'Hara^{1*} and D. Johan Kotze²

¹Biodiversity and Climate Research Centre, Senckenberganlage 25, D-60325 Frankfurt am Main, Germany and

²Department of Environmental Sciences, PO Box 65, University of Helsinki, Helsinki FI-00014, Finland



A recent history (uncomprehensive) of GLM in ecology

Methods in Ecology and Evolution



Methods in Ecology and Evolution

doi: 10.1111/j.2041-210X.2011.00127.x

Distance-based multivariate analyses confound location and dispersion effects

David I. Warton^{1*}, Stephen T. Wright¹ and Yi Wang^{1,2}

¹School of Mathematics and Statistics and Evolution & Ecology Research Centre; and ²School of Computer Science and Engineering, The University of New South Wales, NSW 2052, Australia



A recent history (uncomprehensive) of GLM in ecology

Ecotoxicology

DOI 10.1007/s10646-015-1421-0

Analysing chemical-induced changes in macroinvertebrate communities in aquatic mesocosm experiments: a comparison of methods

Eduard Szöcs · Paul J. Van den Brink · Laurent Lagadic · Thierry Caquet ·
Marc Roucaute · Arnaud Auber · Yannick Bayona · Matthias Liess ·
Peter Ebke · Alessio Ippolito · Cajo J. F. ter Braak · Theo C. M. Brock ·
Ralf B. Schäfer



A recent history (uncomprehensive) of GLM in ecology

Environ Sci Pollut Res
DOI 10.1007/s11356-015-4579-3

RESEARCH ARTICLE

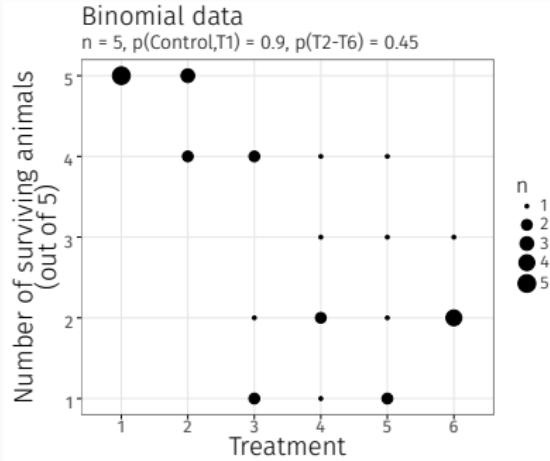
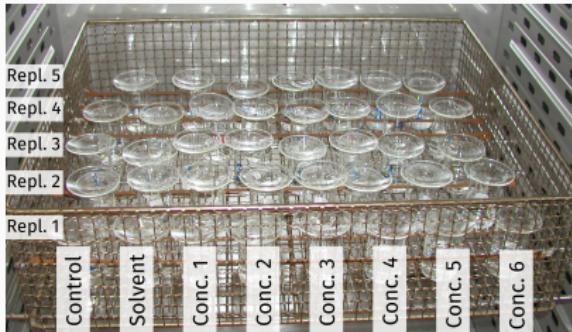
Ecotoxicology is not normal

A comparison of statistical approaches for analysis of count and proportion data in ecotoxicology

Eduard Szöcs¹ · Ralf B. Schäfer¹



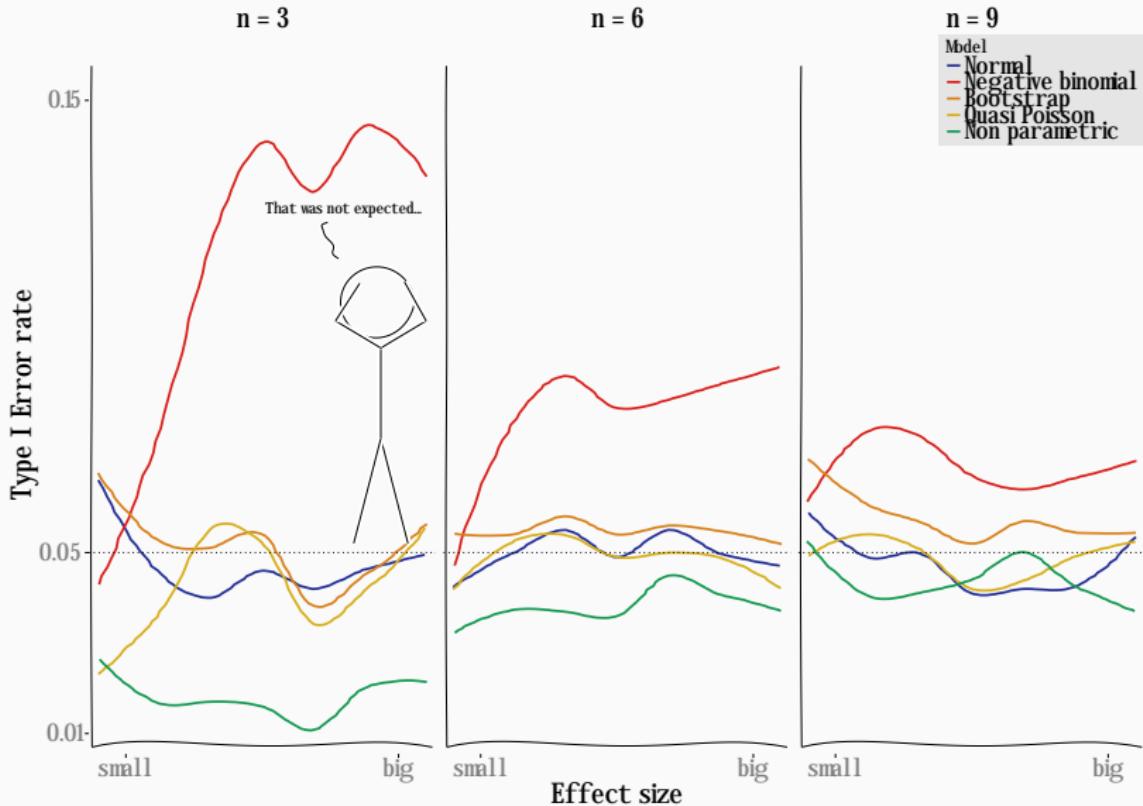
A simulation study



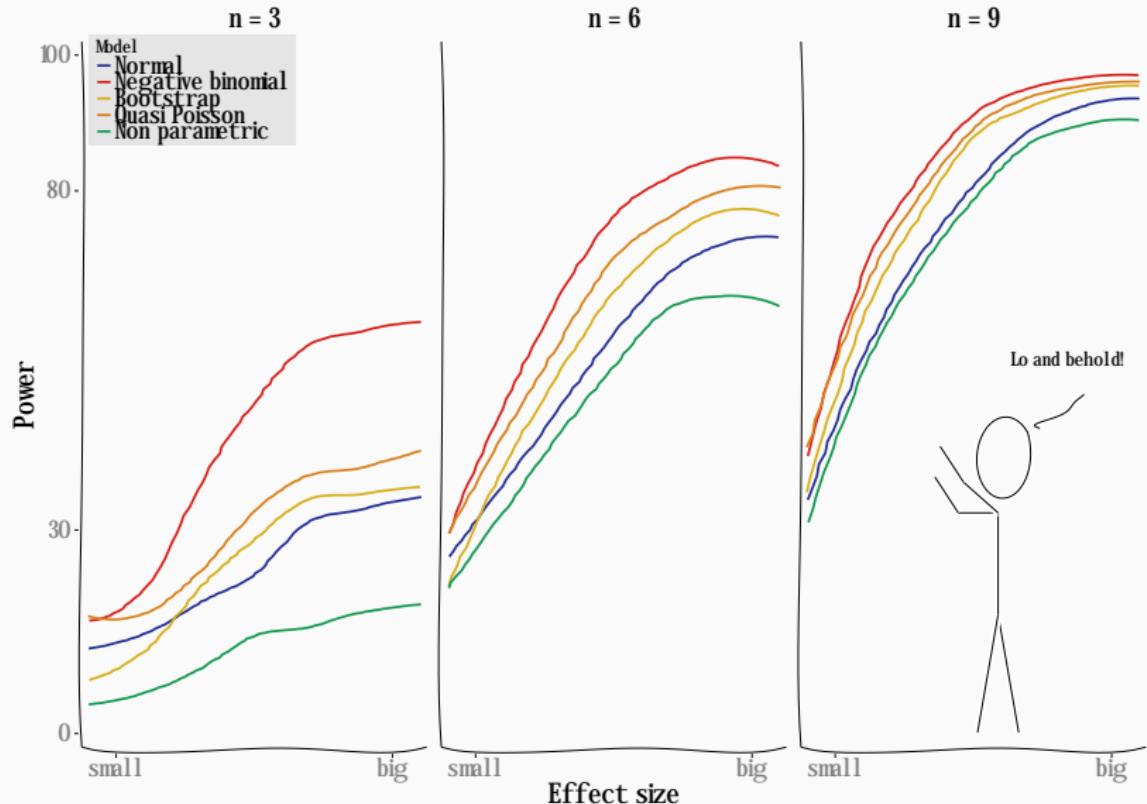
Simulation:

- Count & Binomial data
- Vary replicates & effect sizes
- LM, GLMs,
Non-parametric

Type I Errors: GLMs can fail



Power: But GLMs can do also better



What we learned from this study

1. Negative-binomial GLM show increased Type I errors
2. Can be fixed via bootstrap
3. Ecotoxicological experiments commonly low power
4. GLMs can increase this power

Where are we today?

Three days earlier...



Where are we today?

Three days earlier...

Methods in Ecology and Evolution



Methods in Ecology and Evolution 2015, **6**, 828–835

doi: 10.1111/2041-210X.12386

For testing the significance of regression coefficients, go ahead and log-transform count data

Anthony R. Ives*

Department of Zoology, University of Wisconsin-Madison, Madison, WI 53706, USA



Where are we today?

Current state of knowledge:

Methods in Ecology and Evolution



Methods in Ecology and Evolution 2016, 7, 882–890

doi: 10.1111/2041-210X.12552

FORUM

Three points to consider when choosing a LM or GLM test for count data

David I. Warton^{1*}, Mitchell Lyons², Jakub Stoklosa¹ and Anthony R. Ives³

¹School of Mathematics and Statistics and Evolution & Ecology Research Centre, University of New South Wales, NSW 2052, Australia; ²School of Biological, Earth and Environmental Sciences, University of New South Wales, NSW 2052, Australia; and
³Department of Zoology, University of Wisconsin-Madison, Madison, WI 53706, USA

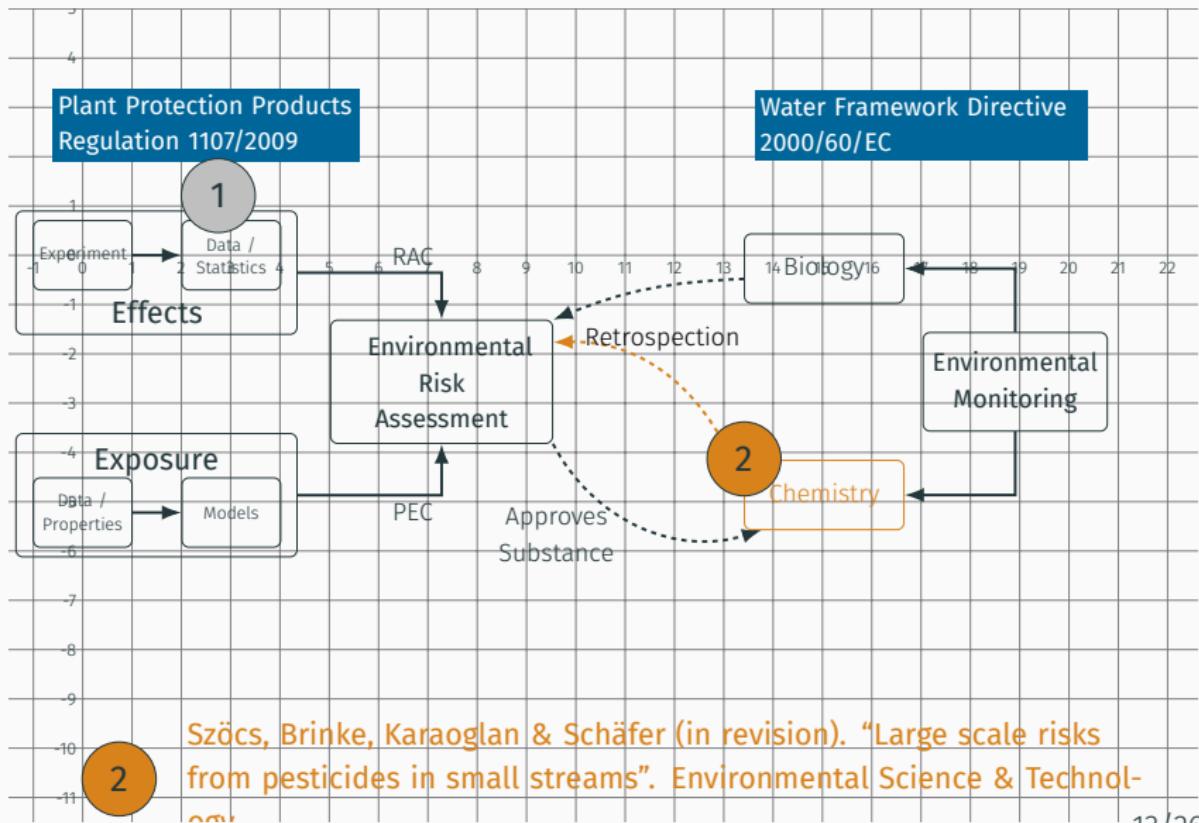


Where are we today?

Three points to consider ...

1. Choose your model based on data properties
2. Fix Type I errors by resampling
3. GLMs have greater power

Exploring Monitoring Data for ERA



Exploring Monitoring Data for ERA

Goals & Hypotheses

Goal: Combine monitoring and ERA

- Compile nation-wide monitoring data
- Focus on small streams
- Identify risks & influencing factors

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Hypotheses

1. Agriculture
2. Stream size
3. Precipitation
4. Annual dynamics

Analysing chemical concentrations

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- Concentrations < LOQ (96% of all measurements)

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$$y \sim ZAGA = \begin{cases} \text{Binomial GLM} & \text{if } y < LOQ \\ \text{Gamma GLM} & \text{if } y \geq LOQ \end{cases}$$

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- $RQ = \frac{C}{RAC}$

Analysing chemical concentrations

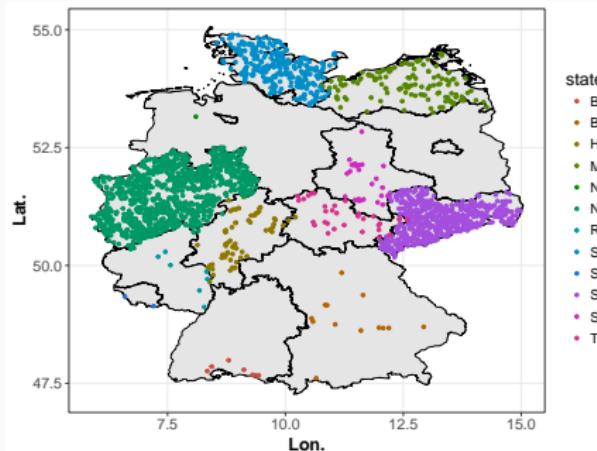
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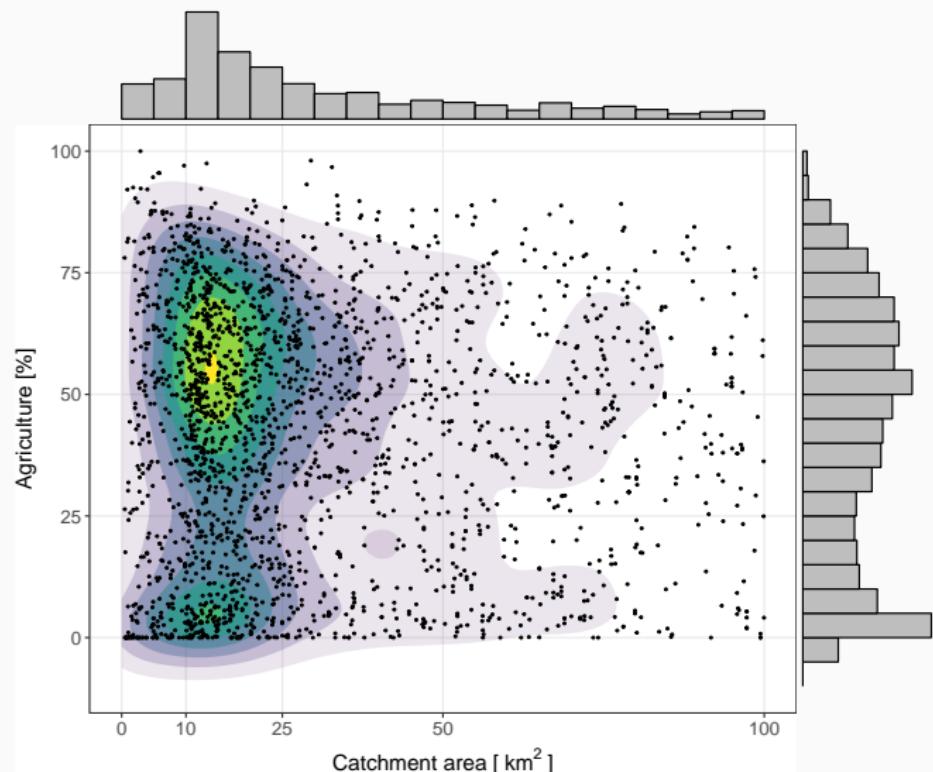
- Risk Quotient
 - $RQ = \frac{C}{RAC}$
- Predictors
 - Catchment Size
 - Agricultural land use
 - Precipitation

Compiled data: Big, but inhomogeneous

- 1,766,104 measurements
- 478 pesticides
- 24,743 samples
- 2,301 sites

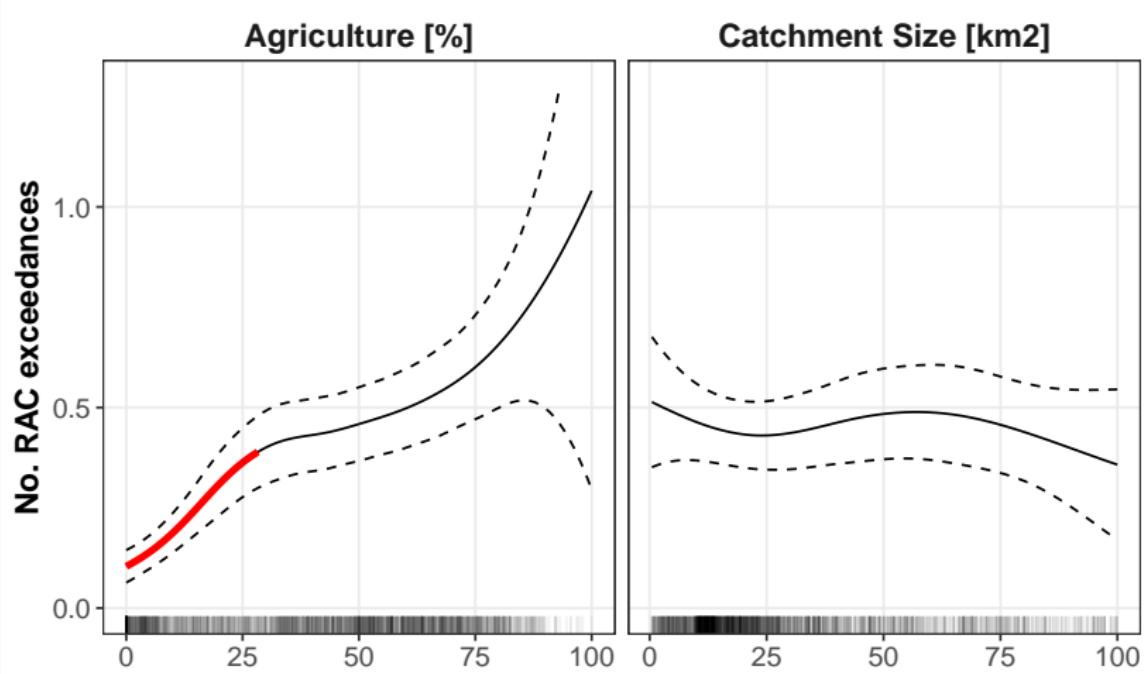


Monitoring: Small streams are underrepresented

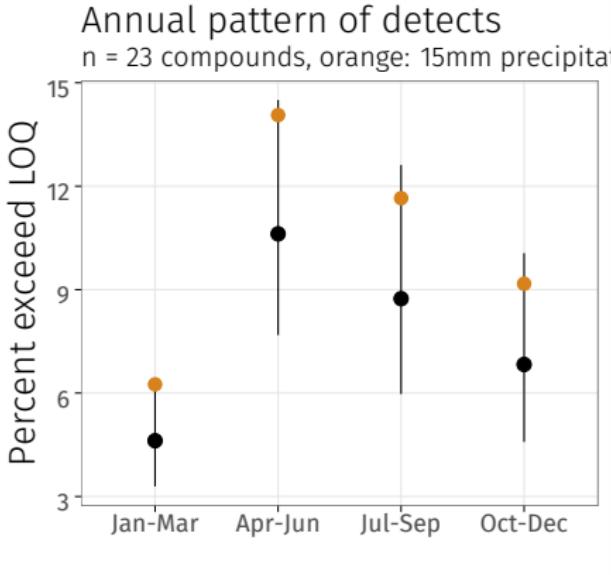


Landscape: Factors influencing risk

- 25% of sites with at least one concentration > RAC

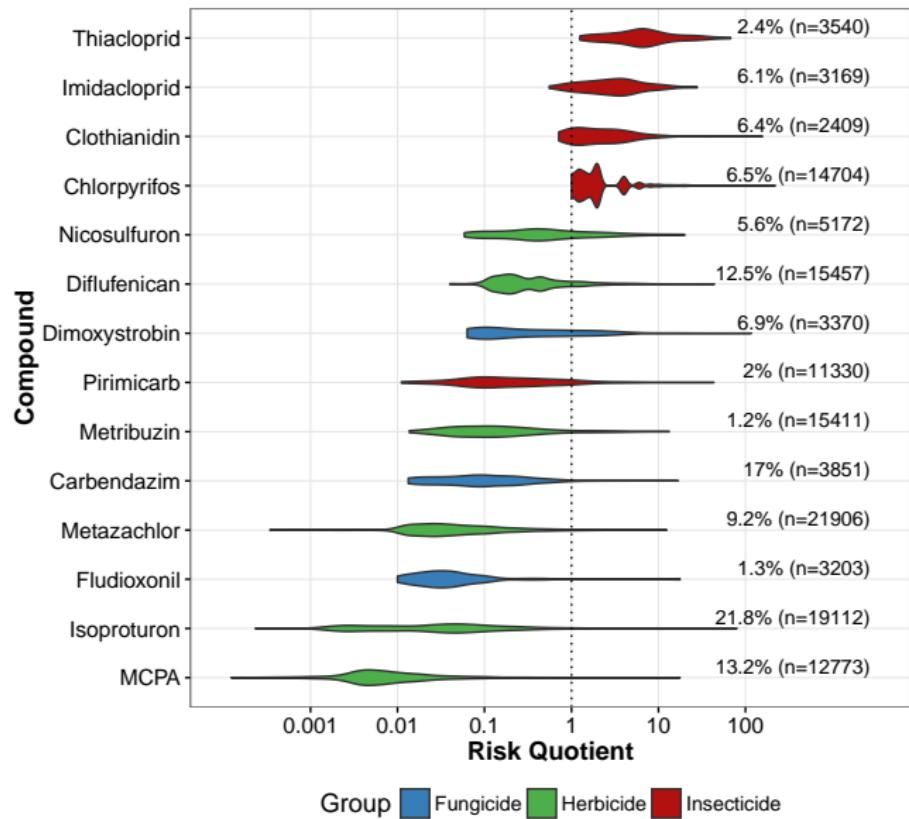


Sampling: Factors influencing risk



- Peak in summer
- Increase by precipitation
- absolute concentrations
>>variability

Risks: Compounds exceeding risk thresholds

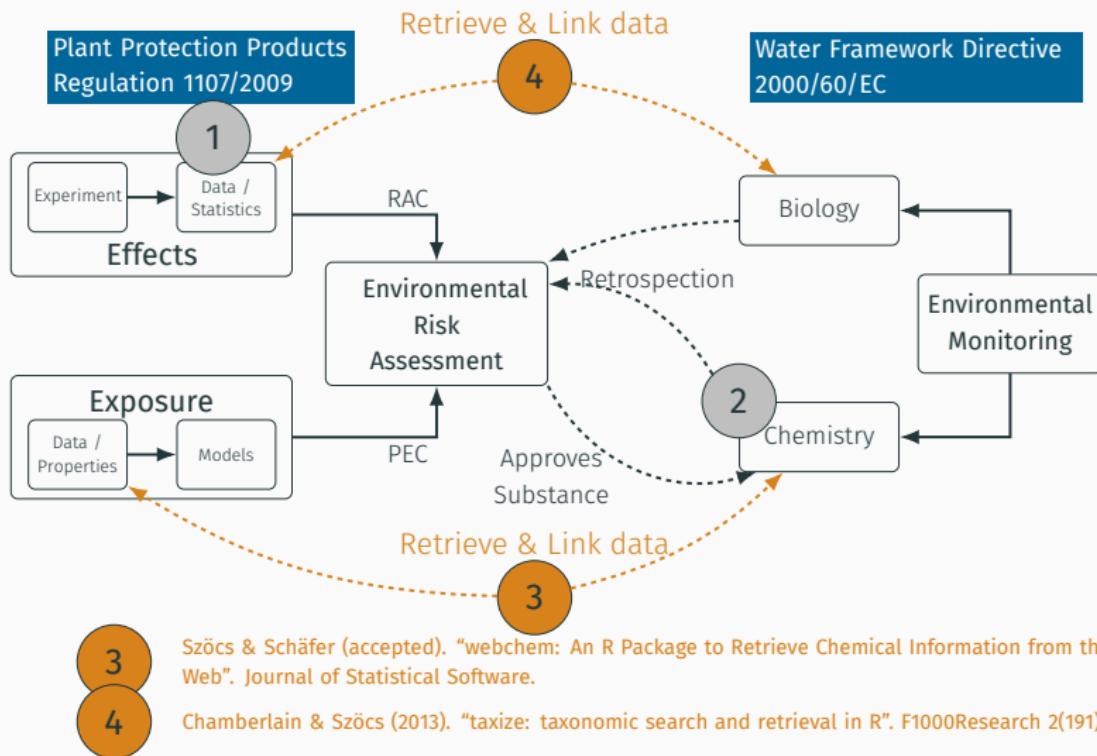


What we learned from this study

1. Differences between **states**
2. Small streams are **underrepresented**
3. **Agricultural** sources
4. **LOQ** gives additional insights
 - Annual **dynamics**
 - **Precipitation** increases concentrations
5. **Neonicotinoids**

Solutions for Data Handling in ERA

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Biologists & Chemists face the same problems

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Names

Osmia rufa, *Osmia bicornis*,

Chlorpyrifos, Chlorpyriphos,

Osmia ruffa, *Osmia*

Chlorphyrifos, Chlorpypifot

unilandauis

Biologists & Chemists face the same problems

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Hierarchies

Hymenoptera/ Apoidea/

Megachilidae/ *Osmia/ rufa*

organophosphate, ester,

insecticide

Biologists & Chemists face the same problems

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Traits / Properties

Wing length, Mass, Season	Mass, K_{OW} , LC_{50}
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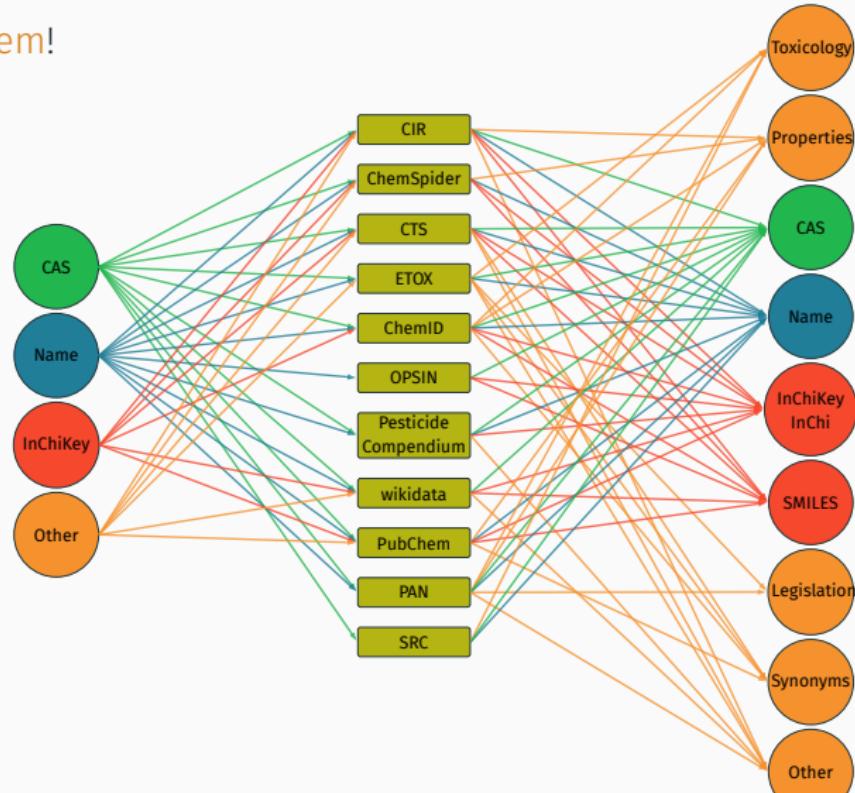
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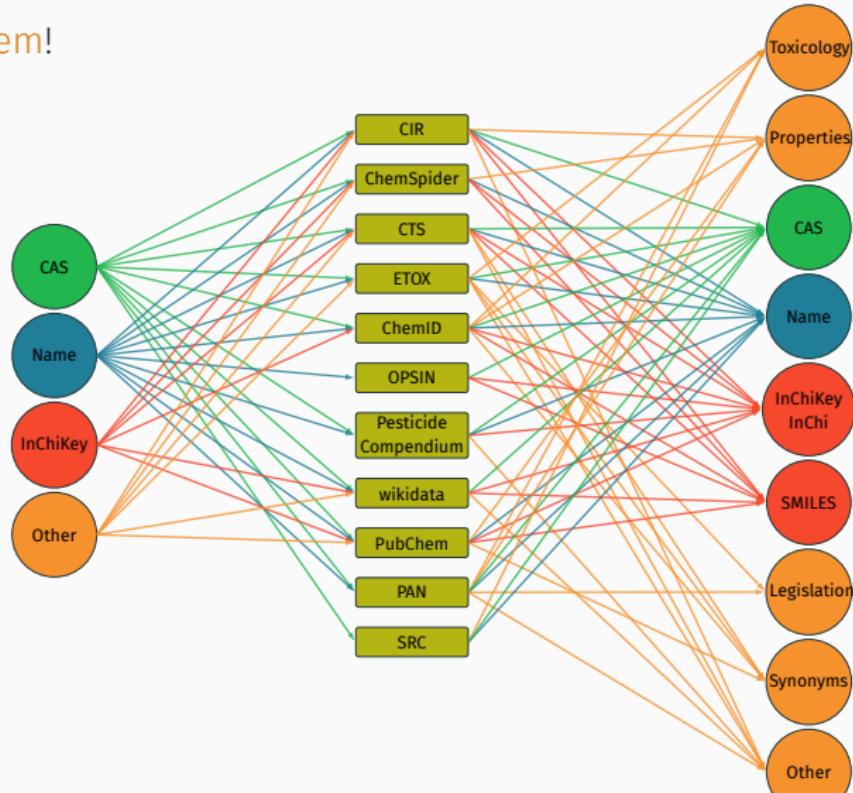
Instead of wasting time...

... use webchem!



Instead of wasting time...

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Instead of wasting time...

... use taxize!



Global Invasive Species
Database



Catalogue of Life



Plantminer



ThePlantList

Instead of wasting time...

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Global Invasive Species
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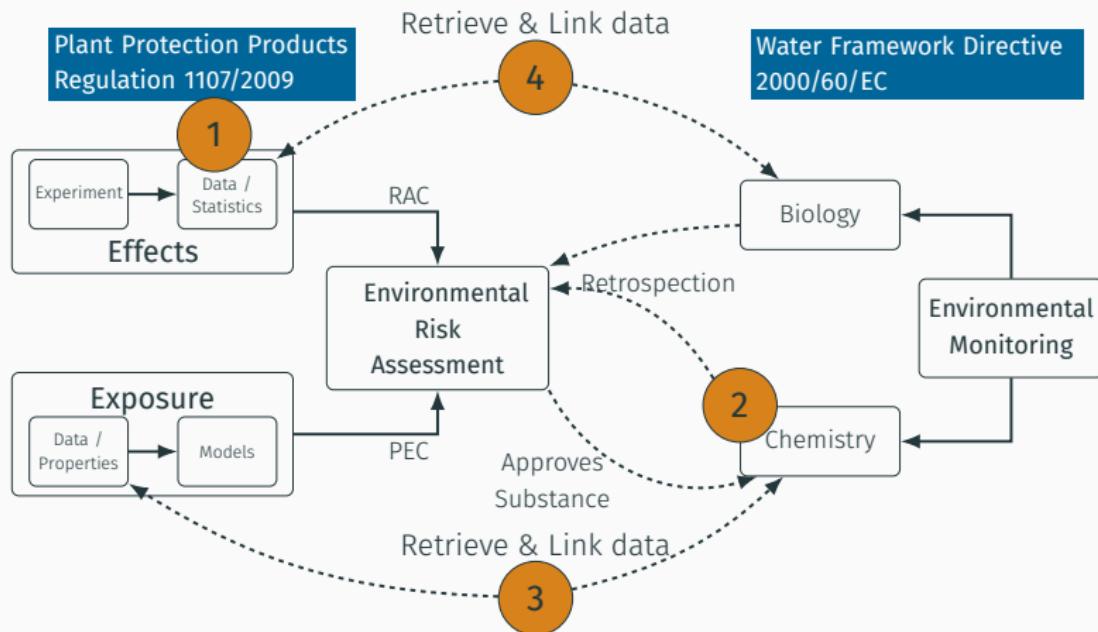


ThePlantList

*"Days of searching done during my morning coffee.
Dr. Susan E. Johnston, University of Edinburgh. On twitter.
Amazing. taxize"*

Recap

Recap: What did I look at?



Recap: What we learned from my PhD Thesis

✓ Improving Statistics in ERA

- Change your model, not your data
- Take LOQ into account

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- Risk drivers and dynamics
- Agricultural small streams neglected & at risk
- Neonicotinoids

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✓ Improving Statistics in ERA

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✓ Exploring Monitoring Data for ERA

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✓ Solutions for Linking Data in ERA

- Handling big eco(toxico-)logical data not easy
- Now easier

Statistical Ecotoxicology

Improving the Utilisation of Data for
Environmental Risk Assessment

Eduard Szöcs

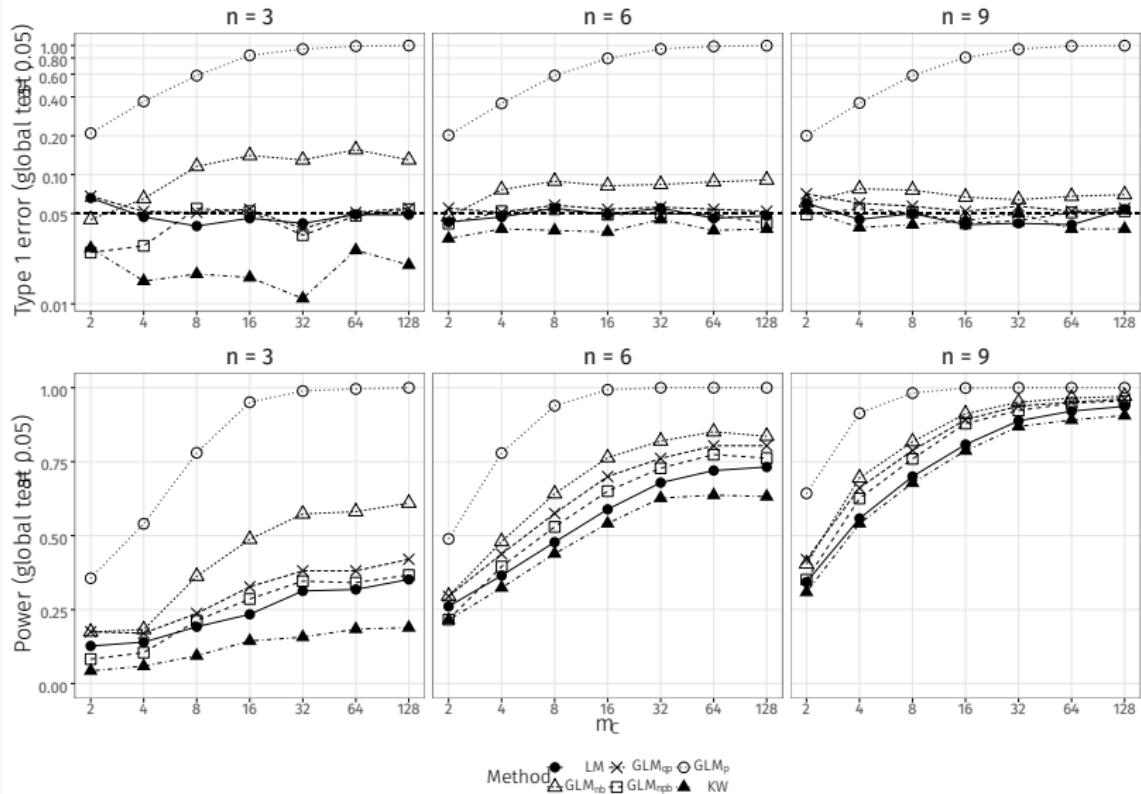
□ <http://edild.github.io/>

🐦 [@EduardSzoebs](https://twitter.com/EduardSzoebs)

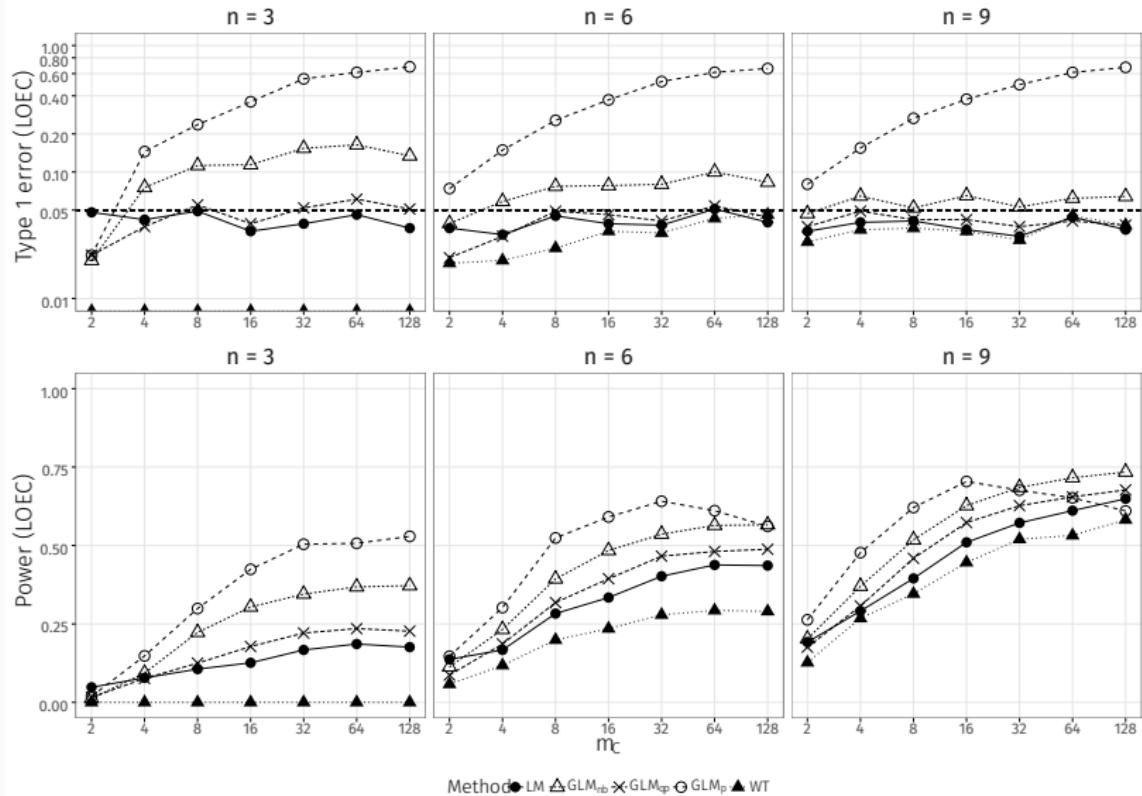
📄 https://github.com/edild/phd_defense

📄 https://github.com/edild/phd_thesis

Power en detail



For LOEC it is even worse



Comparison with Ives...

Szöcs (2015)

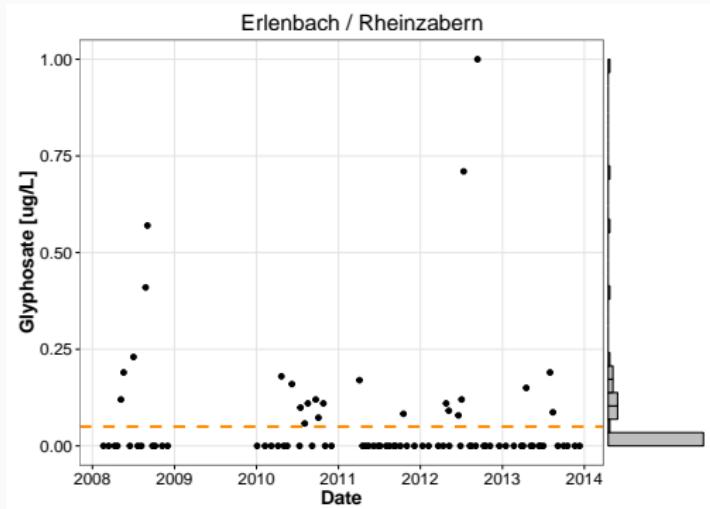
- factorial design
- one predictor
- low replicated
- LM, GLM, bootstrap
- High T1 error of NB
- Quasi-Poisson worked well
- Bootstrap fixes the problems

Ives (2015)

- continuous design
- two predictors
- well replicated
- LM, GLM
- High T1 error of NB
- Quasi-Poisson has problems with multiple predictors
-

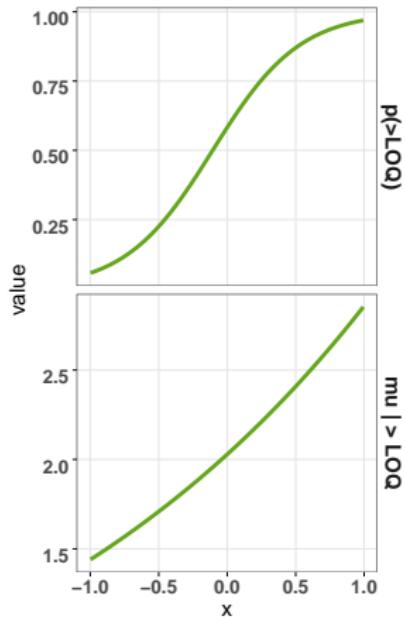
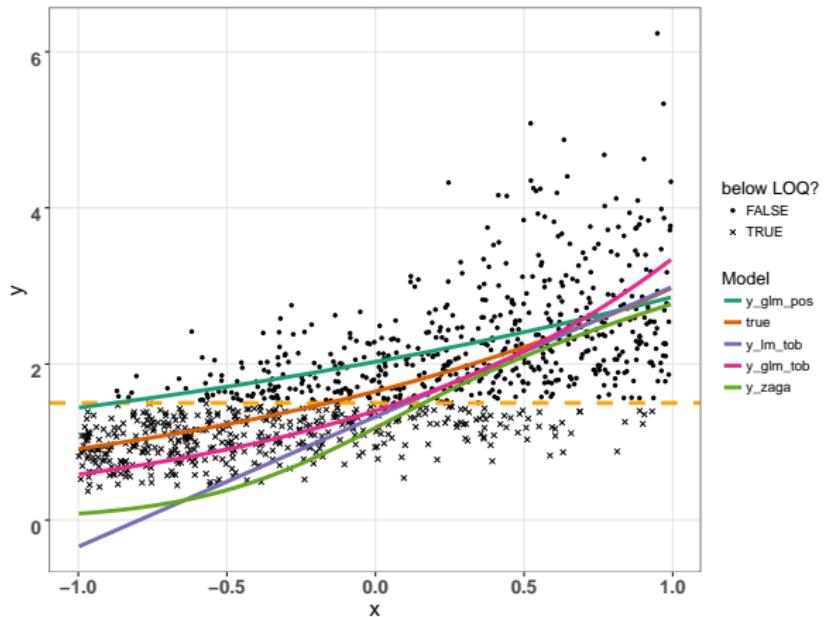
Idiosyncrasies of chemical concentrations

- continuous distribution in \mathbb{R}_0^+
 - censoring ($x < \text{LOQ}$)
 - non-linearity (season, trends)
 - dependency (spatial, temporal)
 - missing data

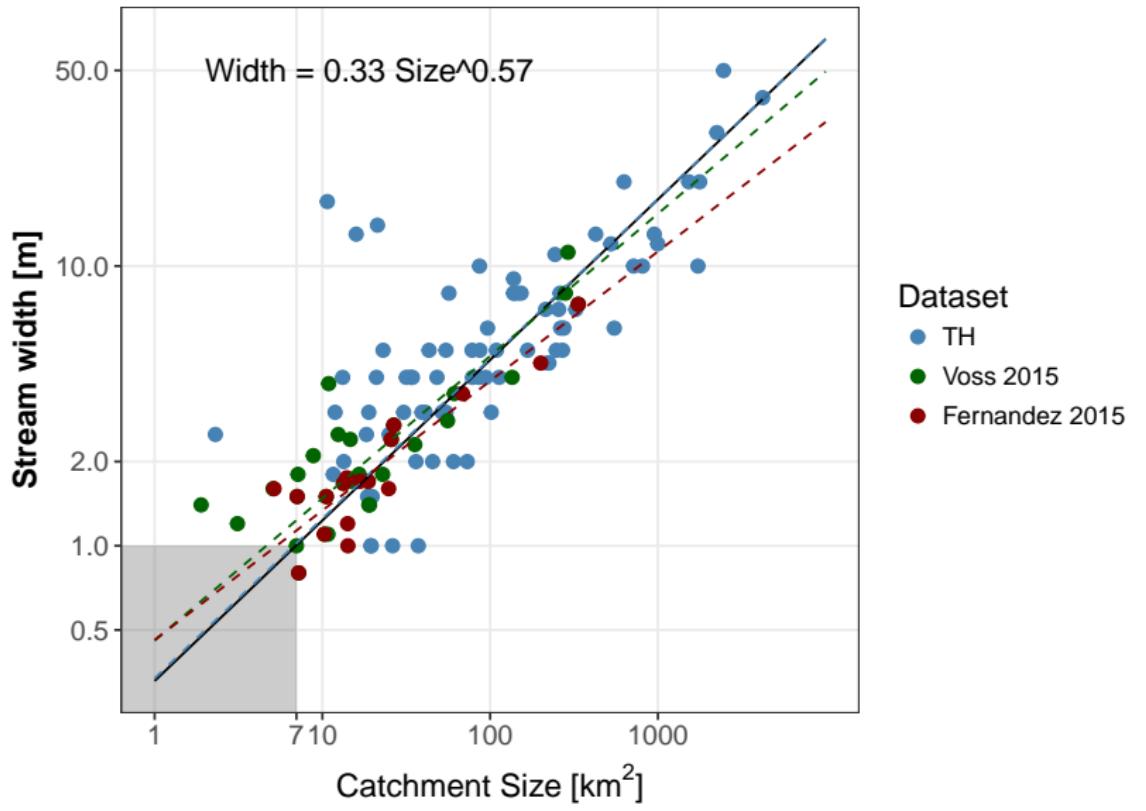


ZAGA what...?

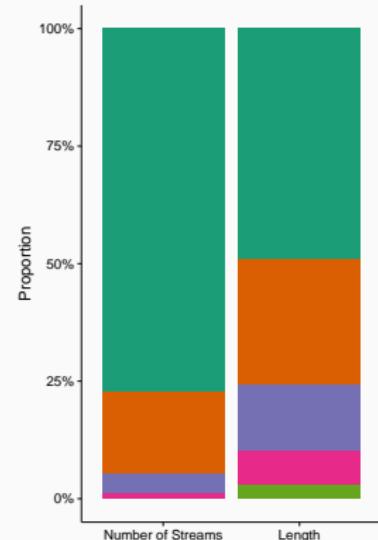
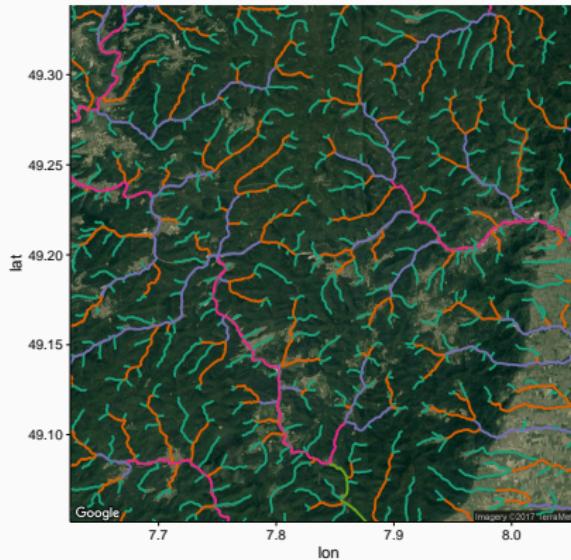
shiny app: <http://uni-ko-ld.de/g4>



Stream size / width



Importance of small streams

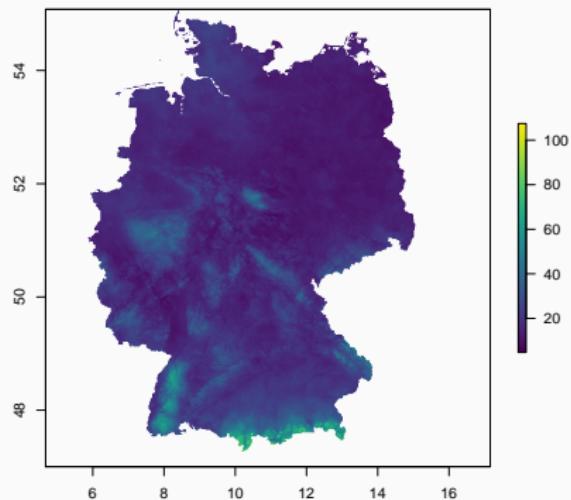


- Biodiversity
- Refuge for re-colonisation

Biggs et al. (2016). The importance of small waterbodies for biodiversity and ecosystem services: implications for policy makers. *Hydrobiologia*. DOI: 10.1007/s10750-016-3007-0.

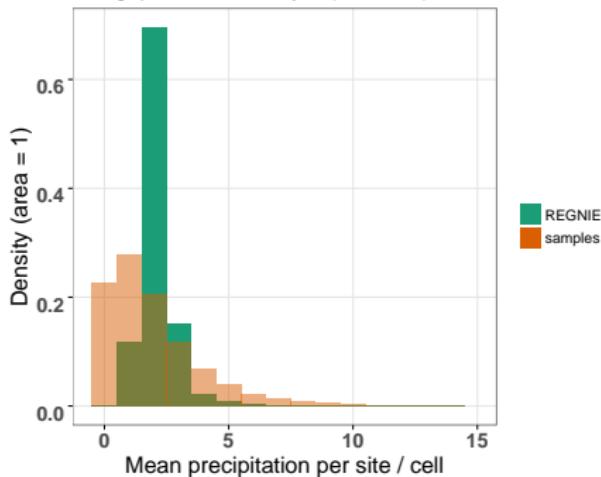
Precipitation in Germany and the samples

Mean # days > 10mm (REGNIE, 2005–2015)



Distribution of precipitation at sites / cell

Average per site / cell within year (2005–2015)



Comparison with other studies

Szöcs (2016)

- Germany
- Monitoring
- Grab sampling
- Pesticides
- Neonic + Chlorpyrifos (OP)
- ZAGA (<LOQ)

Stehle (2015)

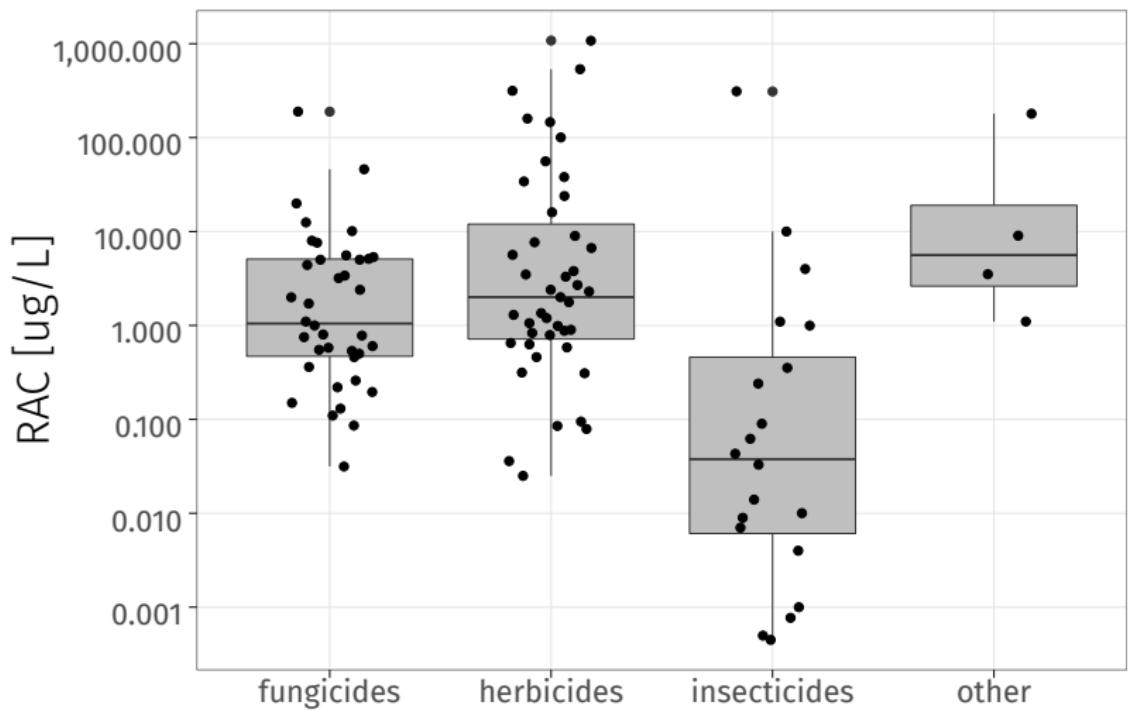
- Europe / Global
- Publications
- Grab & Event driven sampling
- Insecticides
- Organophos.+ Pyrethroids
- LM for >LOQ

Knauer (2016)

- Switzerland
- Monitoring
- Grab sampling
- Pesticides
- Chlorpyrifos + Herb + Fung
- no model

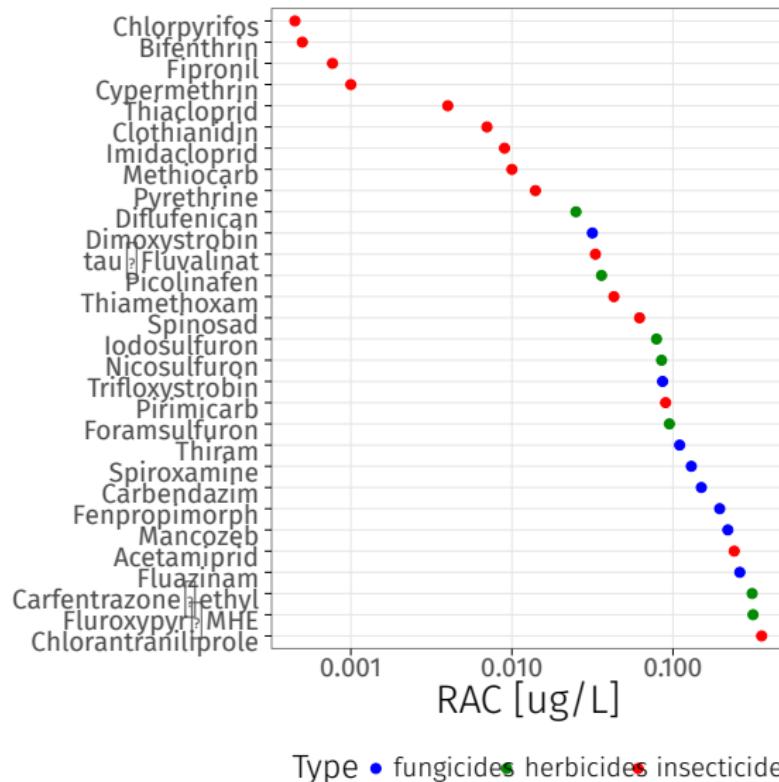
RACs by Type

105 RACs provided by UBA splitted by group

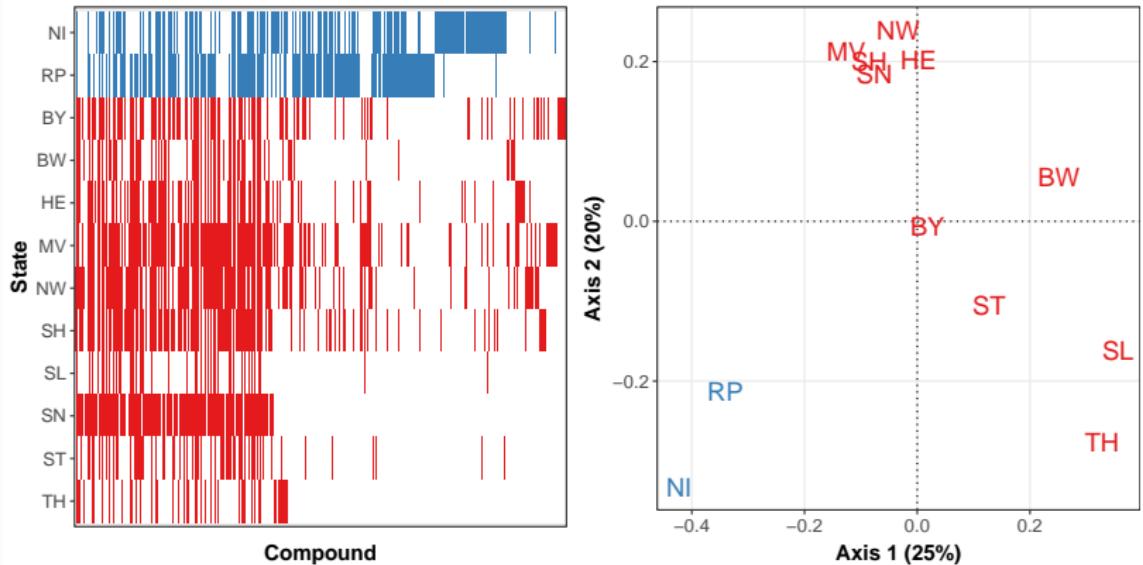


RACs by Compound

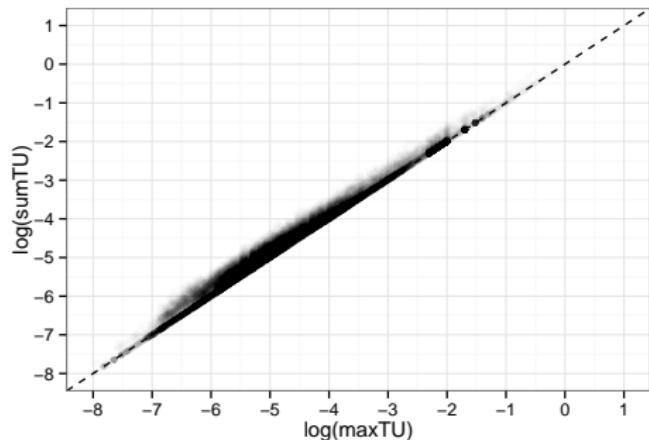
30 lowest RACs



Analysed compound spectra by state



Mixtures are common, but one compound dominates the risk

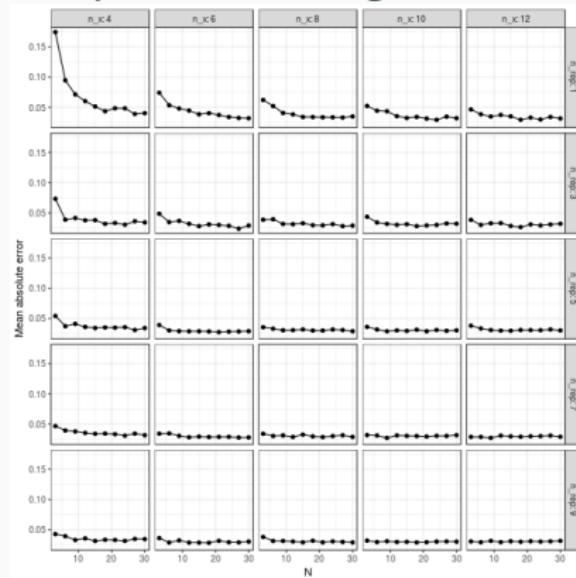


- up to 50 compounds in one sample
- high correlation
- ~ 0.5 TU increase
- mainly one compound responsible for risk

Simulations are worth their work, use them *a priori*!

Experimental design for dose-response experiments - a simulation

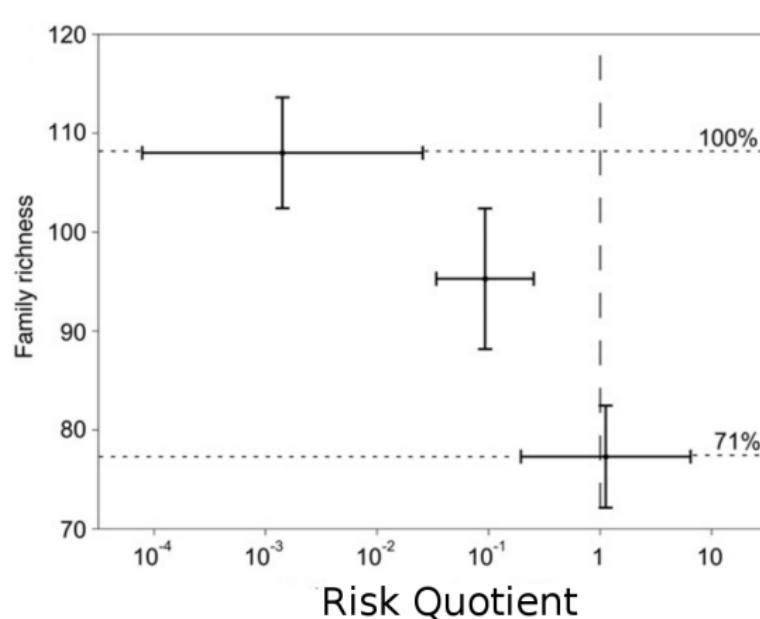
http://edild.github.io/lc50_bias_sim/



GLM-Explorer: <http://uni-ko-lid.de/g3>

Effects of RAC exceedances

- RACs should never be exceeded (=protection goal)
- If so, biological effects likely



Modified from: Stehle, Schulz (2015). Agricultural insecticides threaten surface waters at the global scale. PNAS 112, 5750–5755.

Reasons for observed RAC exceedances

- Risk Mitigation fails
 - Risk mitigation measures (erosion rills, wind)
 - Farmer do not adhere (GAP, no spray zones)
- Risk Assessment fails
 - Exposure Assessment
 - Models not working (Knäbel et al.)
 - Effect Assessment
 - Missed sensitive species
 - New document asks also for insects

Risk Mitigation Measures

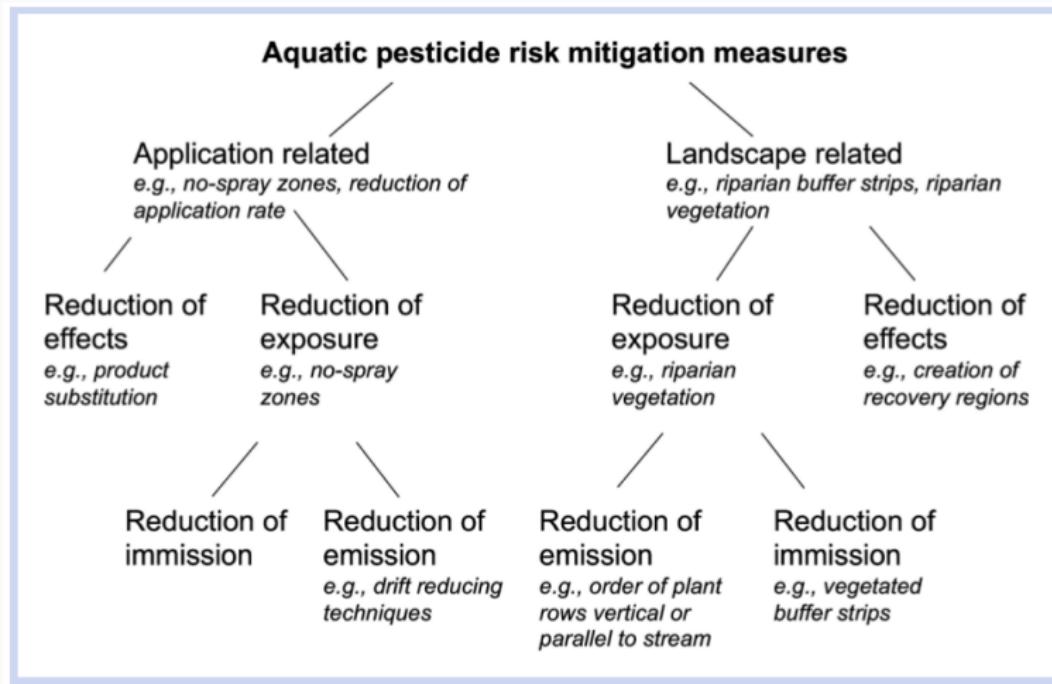
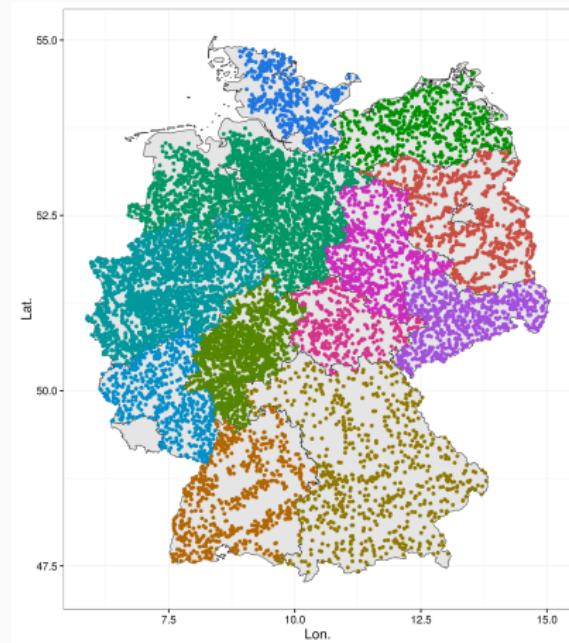


Figure 1. Classification of pesticide risk mitigation measures.

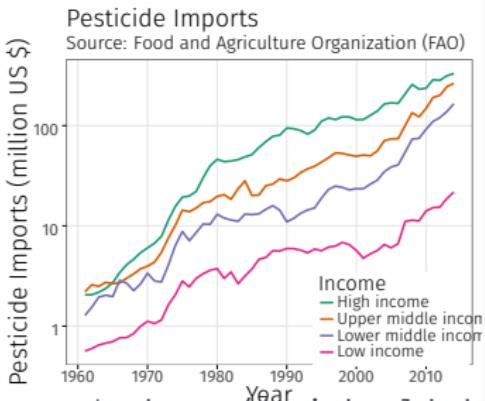
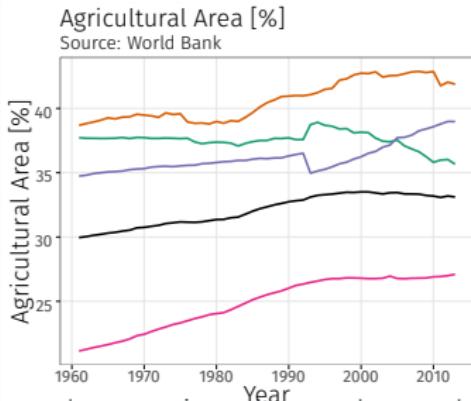
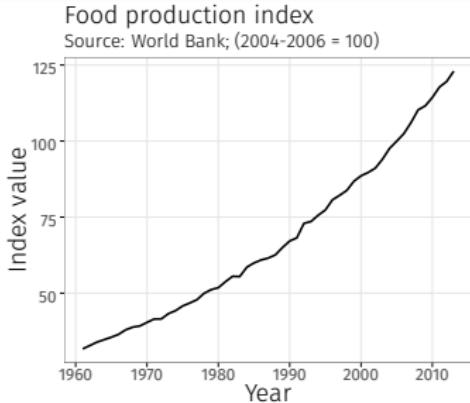
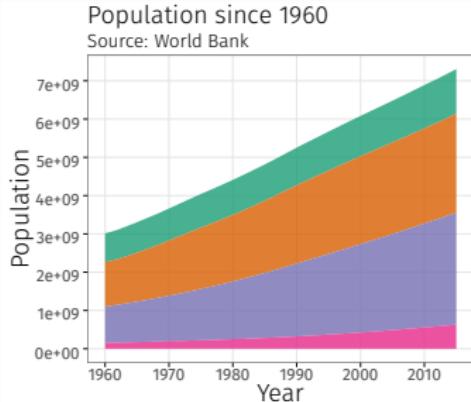
Bereswill, Strelcok, Schulz (2014). Risk mitigation measures for diffuse pesticide entry into aquatic ecosystems: Proposal of a guide to identify appropriate measures on a catchment scale: Guide to Identify Pesticide Risk Mitigation Measures. *IEAM* 10, 286–298.

Biotic field effects



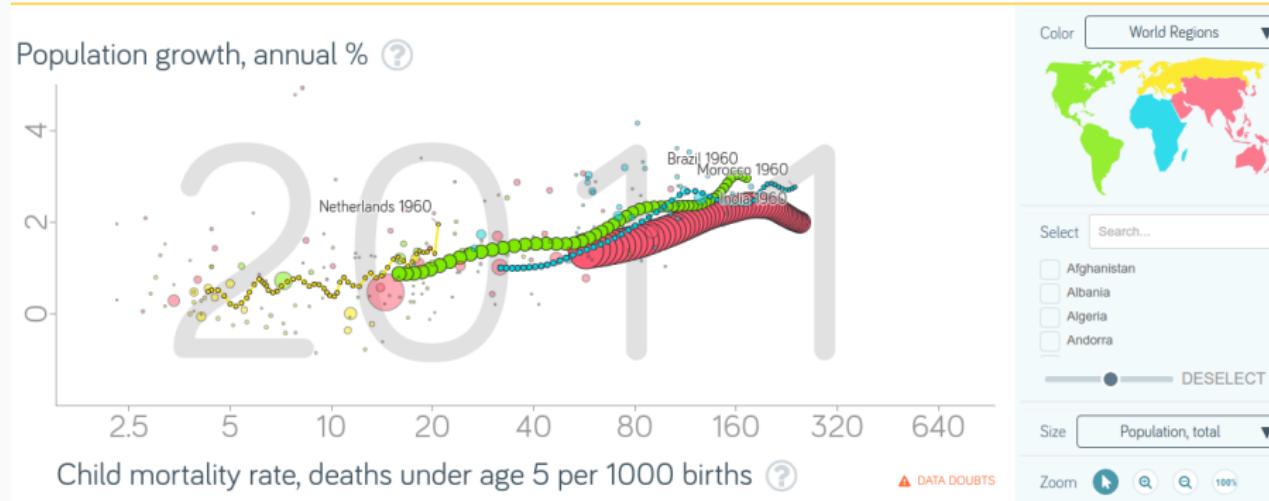
- biological data with good spatial coverage
- 60% of spatial congruence
- Large scale effects largely unknown.
- Some work left...
- Future....

A global perspective (I)



Source code to retrieve data and reproduce results: <http://uni-ko-l.de/g7>

A global perspective (II)



Source: <https://www.gapminder.org/tools>

Software availability

Stable versions on CRAN, dev versions on github.

`webchem` github.com/ropensci/webchem

`taxize` github.com/ropensci/taxize

Best practices for Software:

- open source (permissive MIT License)
- version control (git)
- automated tests (Travis-CI)
- in source documentation (roxygen)

Many Thanks To

- My supervisor Prof. Dr. Ralf. B. Schäfer (for his support, openness, opportunities & discussions)
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- My parents Anca & Helmut (for their support)
- My girlfriend Anja (for everything)