

# Statistical Eco(-toxico)logy

Improving the Utilisation of Data for  
Environmental Risk Assessment

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Eduard Szöcs

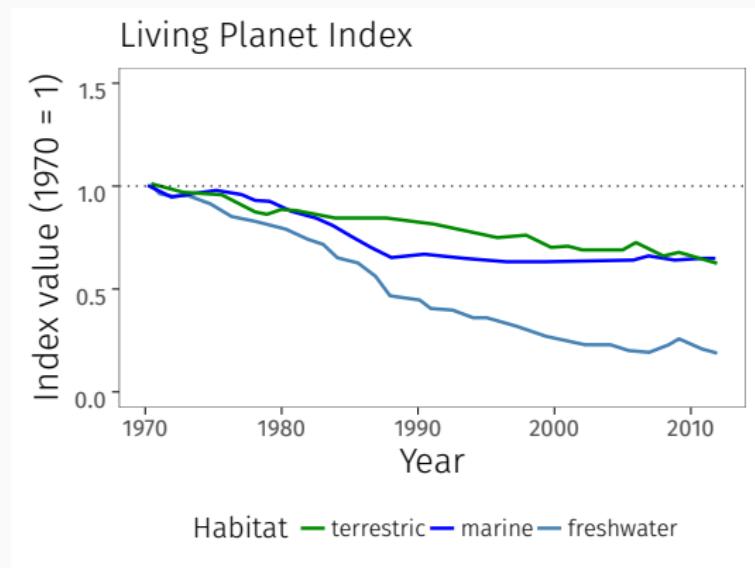
25<sup>th</sup> January 2017

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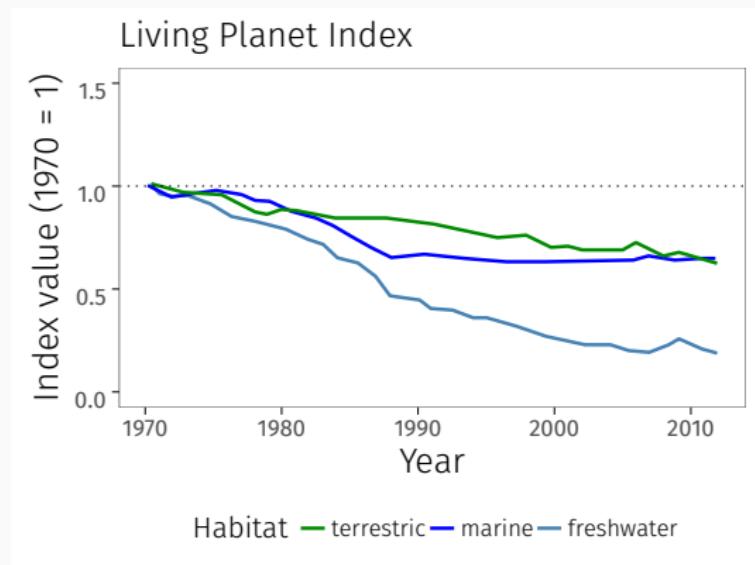
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1. Environmental Risk Assessment (ERA) and Monitoring
2. Improving Statistics in ERA
3. Exploring Monitoring Data for ERA
4. Solutions for Data Handling in ERA

# Freshwater biodiversity is strongly declining



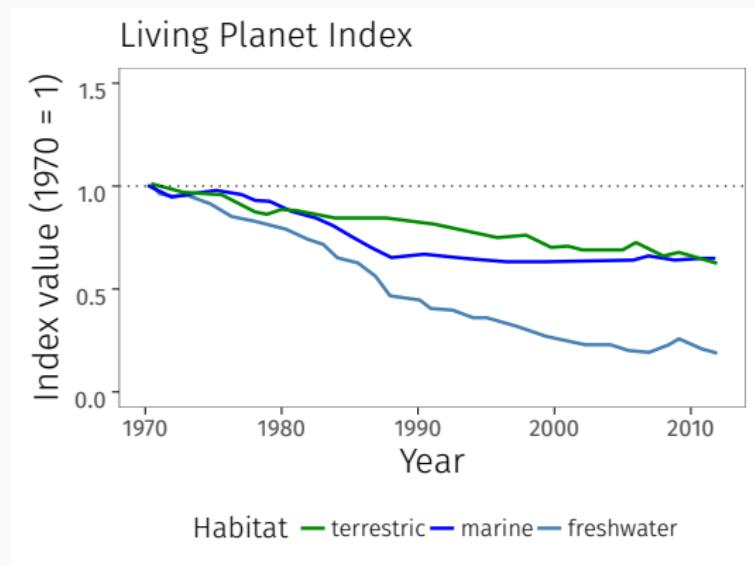
# Freshwater biodiversity is strongly declining



## Reasons

- Habitat loss
- Overexploitation
- Pollution
- Invasive species

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# Environmental Risk Assessment and Environmental Monitoring

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# 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

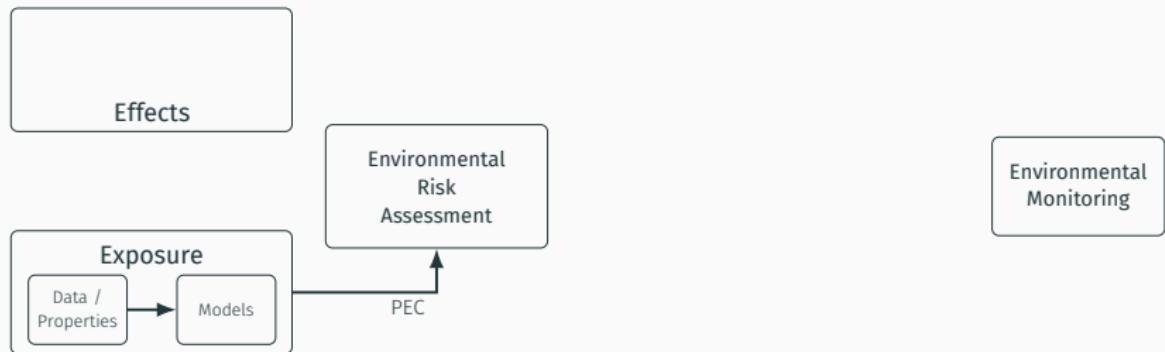
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# Environmental Risk Assessment and Monitoring

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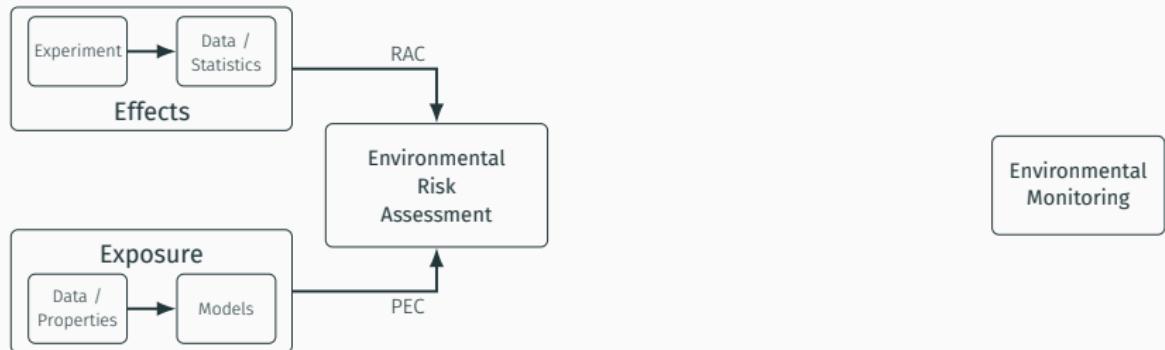
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# Environmental Risk Assessment and Monitoring

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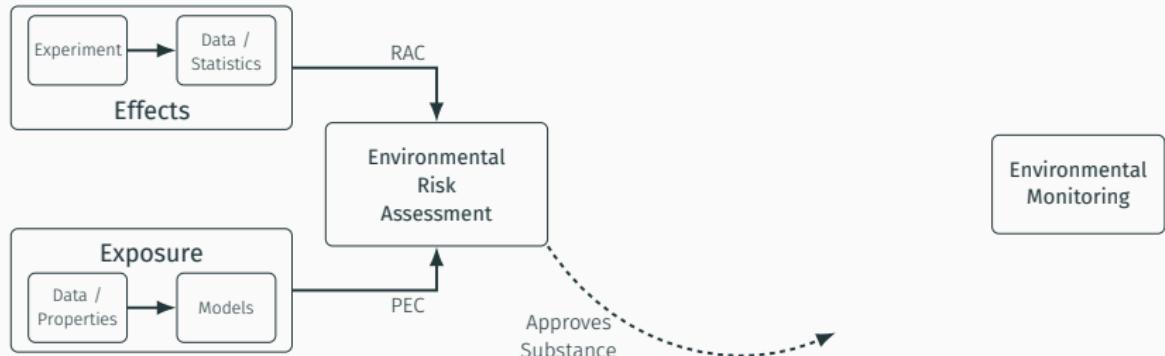
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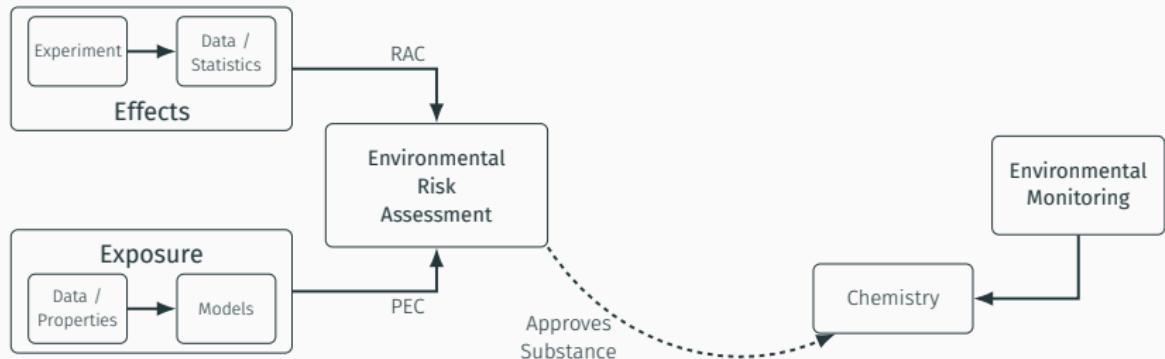
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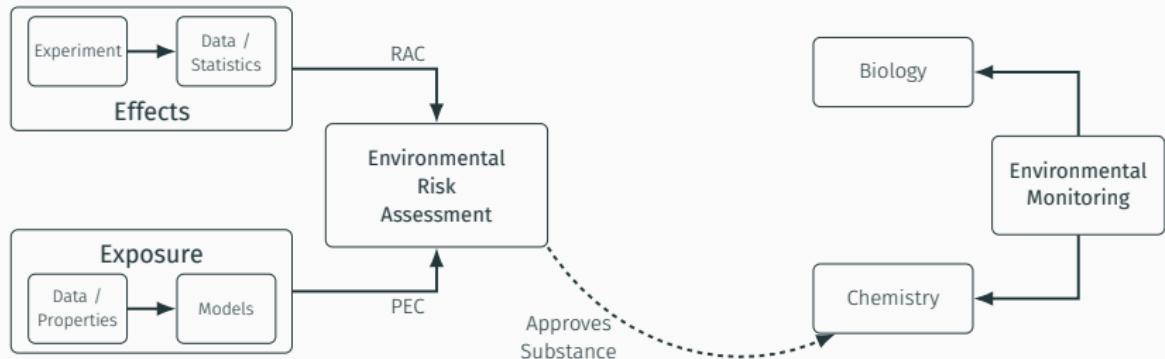
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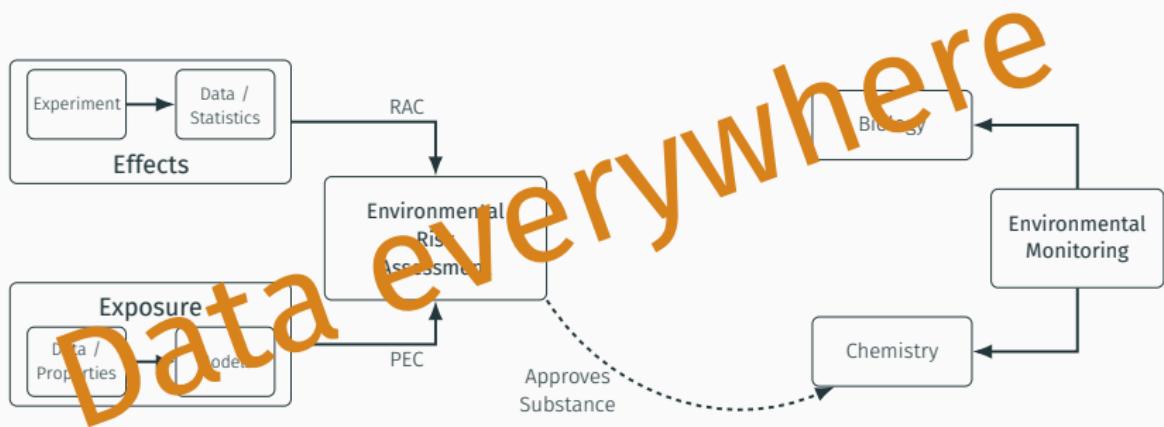
Water Framework Directive  
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# Environmental Risk Assessment and Monitoring

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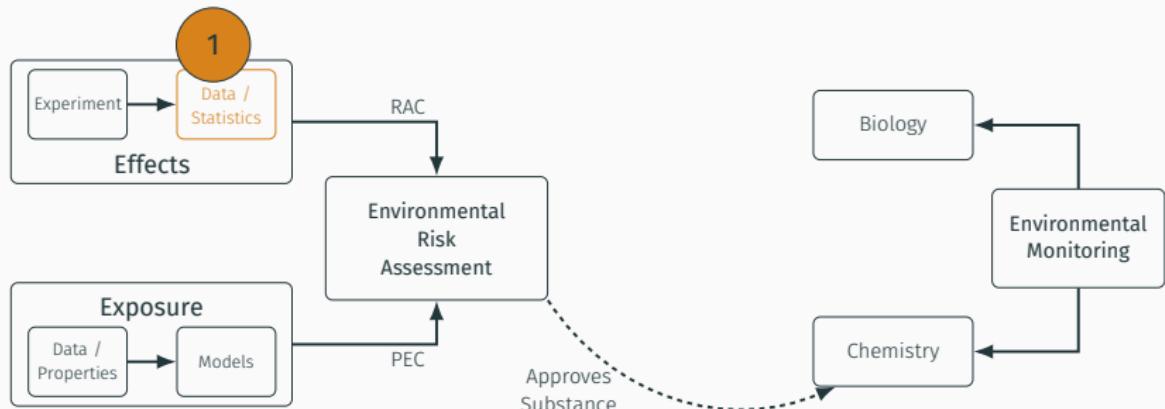
Water Framework Directive  
2000/60/EC



# Improving Statistics in ERA

Plant Protection Products  
Regulation 1107/2009

Water Framework Directive  
2000/60/EC



1

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

# Improving Statistics in ERA

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# Experiments in Effect Assessment



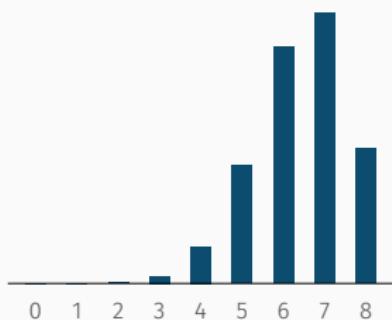
- Daphnia Test
- " $x$  out of  $n$  survived"

# Experiments in Effect Assessment

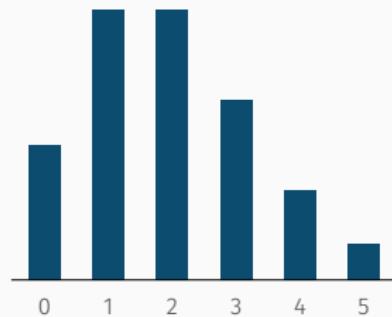


- Daphnia Test
- “*x out of n survived*”
- Mesocosm
- “*number of animals*”

# Ecotoxicology is not normal

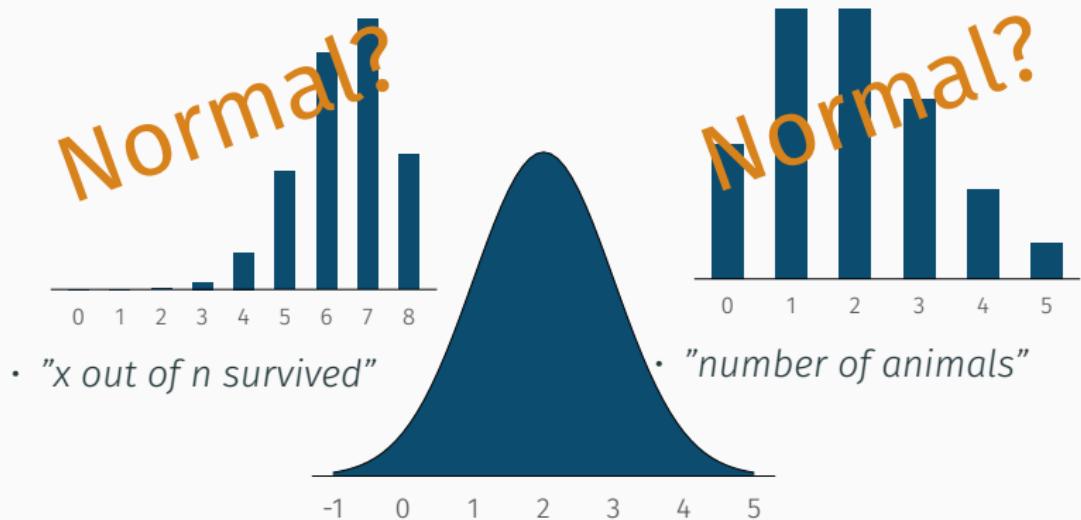


- " $x$  out of  $n$  survived"

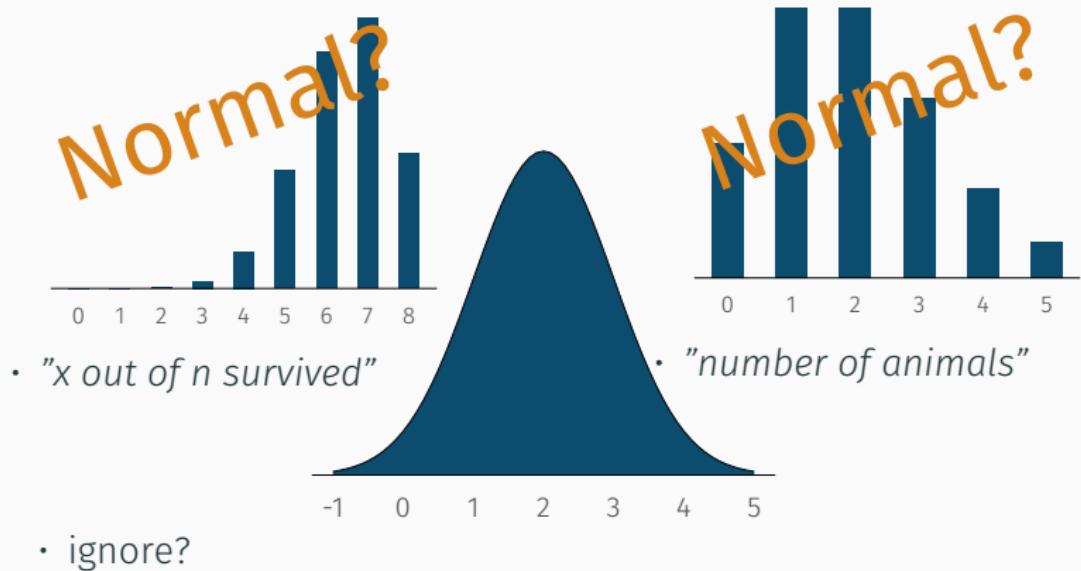


- "number of animals"

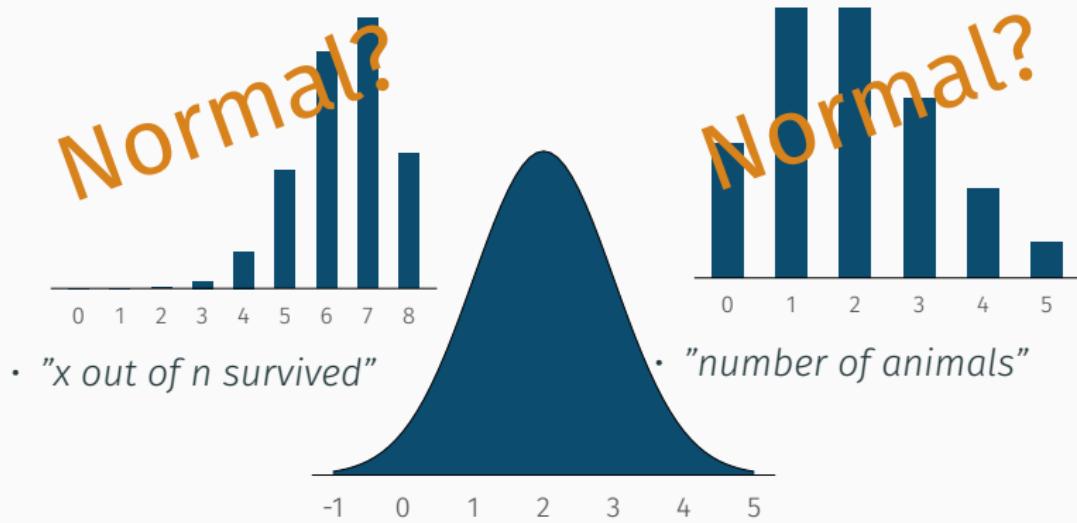
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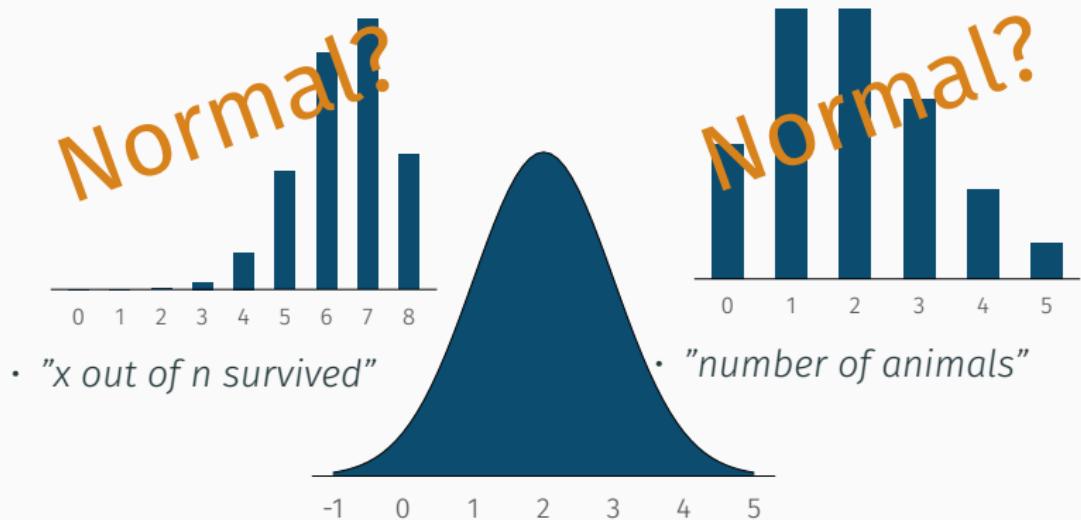


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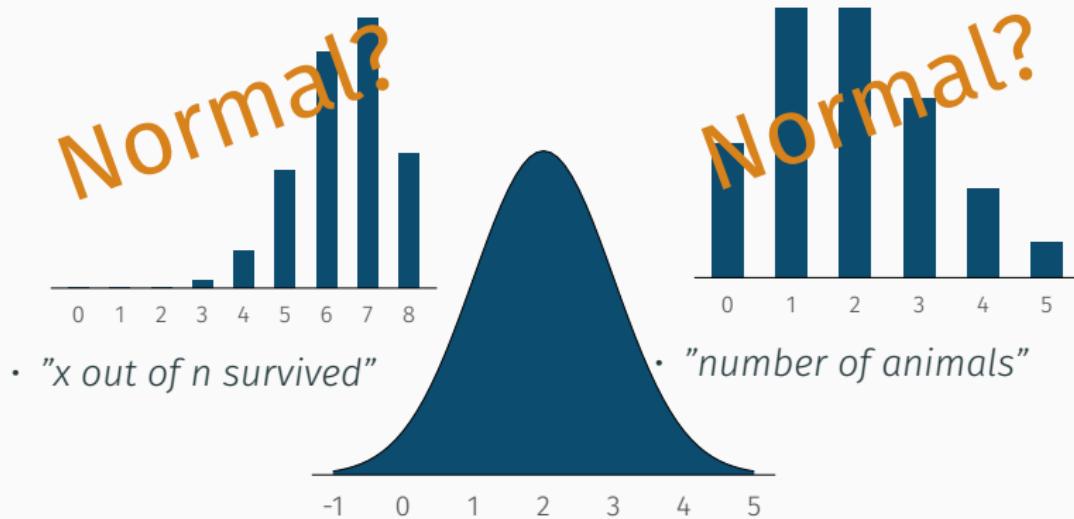
- ignore?
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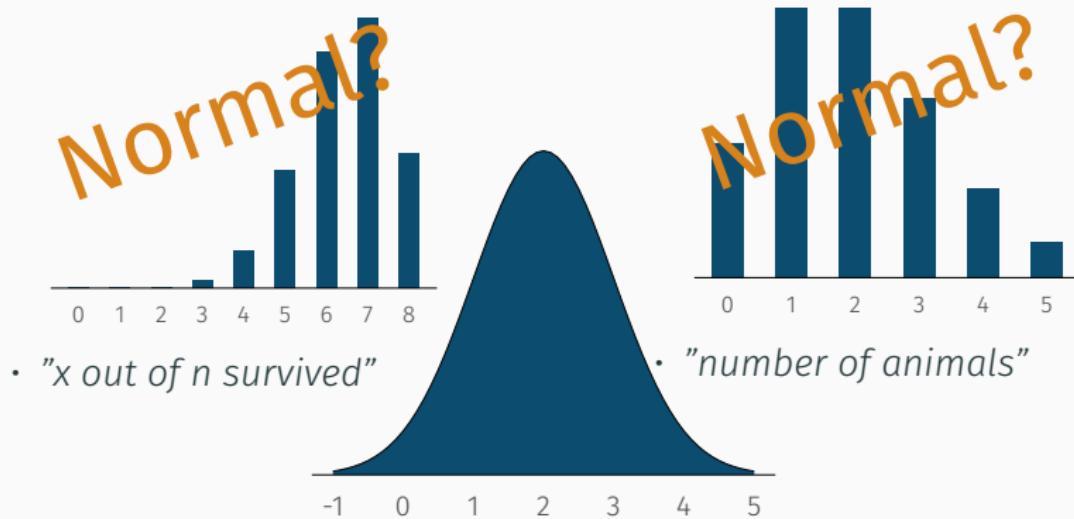
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# Ecotoxicology is not normal



- ignore?
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- Generalised Linear Model (GLM)?

# Ecotoxicology is not normal



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

# A recent history of GLM (uncomprehensive) 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*



## 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<sup>1\*</sup> and D. Johan Kotze<sup>2</sup>

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

<sup>2</sup>Department of Environmental Sciences, PO Box 65, University of Helsinki, Helsinki FI-00014, Finland



# A recent history of GLM (uncomprehensive) in ecology

*Ecology*, 92(1), 2011, pp. 3–10  
© 2011 by the Ecological Society of America

## The arcsine is asinine: the analysis of proportions in ecology

DAVID I. WARTON<sup>1,2,3</sup> AND FRANCIS K. C. HUI<sup>1</sup>

<sup>1</sup>School of Mathematics and Statistics, The University of New South Wales, Sydney, NSW 2052 Australia  
<sup>2</sup>Evolution and Ecology Research Centre, The University of New South Wales, Sydney, NSW 2052 Australia



## 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<sup>1\*</sup>, Stephen T. Wright<sup>1</sup> and Yi Wang<sup>1,2</sup>

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



# A recent history of GLM (uncomprehensive) 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 of GLM (uncomprehensive) 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 Szöcs<sup>1</sup> · Ralf B. Schäfer<sup>1</sup>

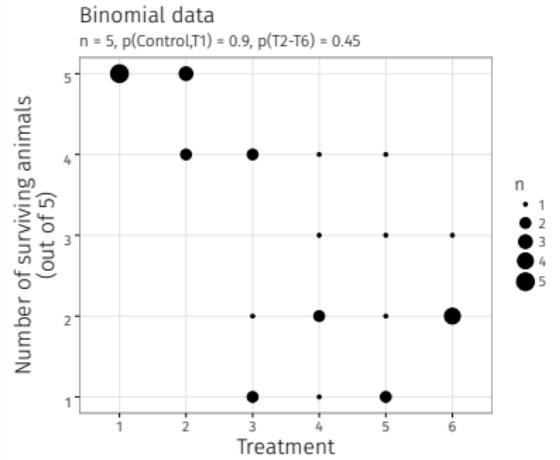


# A simulation study

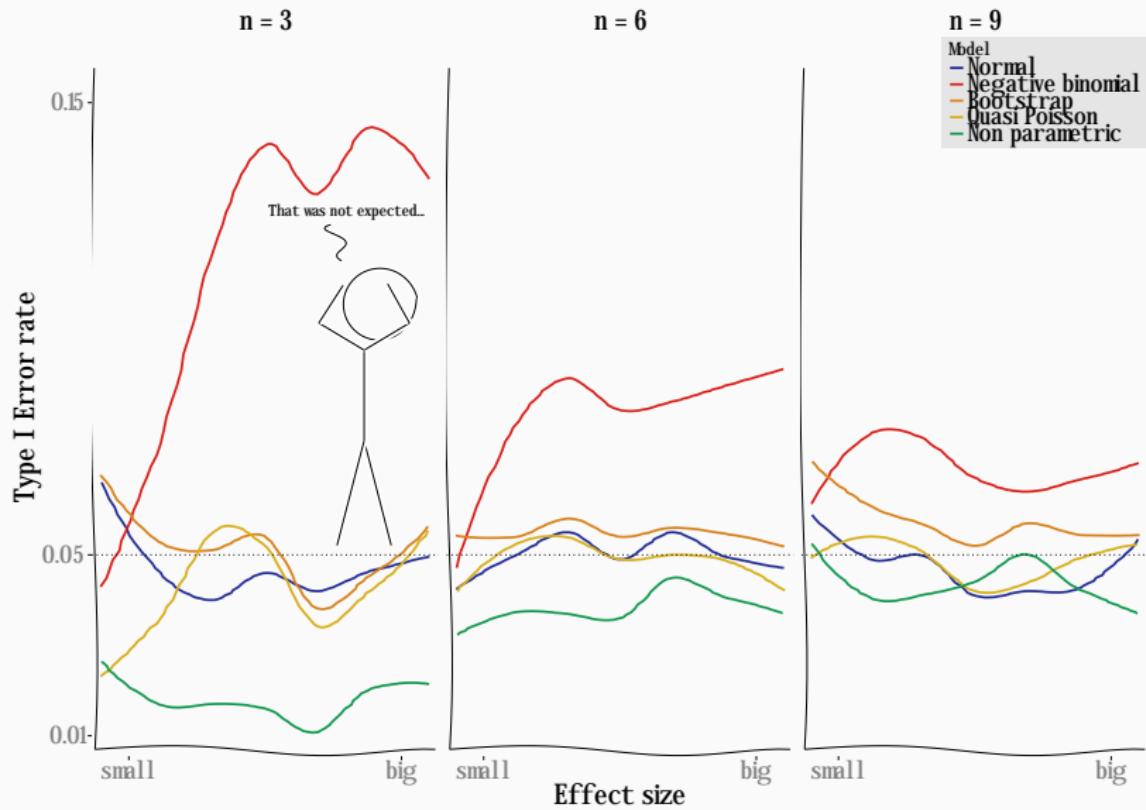


Simulation:

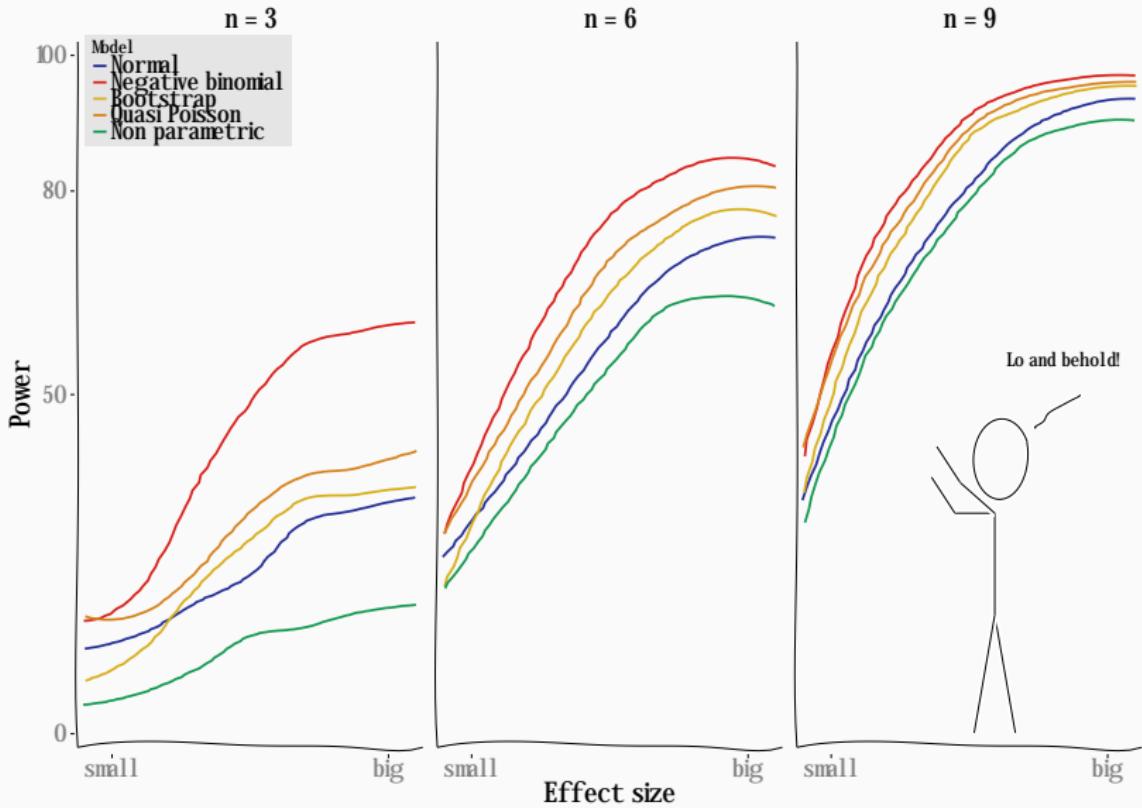
- Count & Binomial data
- Vary replicates & effect sizes
- LM, GLMs, Non-parametric
- Power & Type I Error



# 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. NOECs are not reliable
5. 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<sup>1\*</sup>, Mitchell Lyons<sup>2</sup>, Jakub Stoklosa<sup>1</sup> and Anthony R. Ives<sup>3</sup>

<sup>1</sup>School of Mathematics and Statistics and Evolution & Ecology Research Centre, University of New South Wales, NSW 2052, Australia; <sup>2</sup>School of Biological, Earth and Environmental Sciences, University of New South Wales, NSW 2052, Australia; and

<sup>3</sup>Department of Zoology, University of Wisconsin-Madison, Madison, WI 53706, USA



# Where are we today?

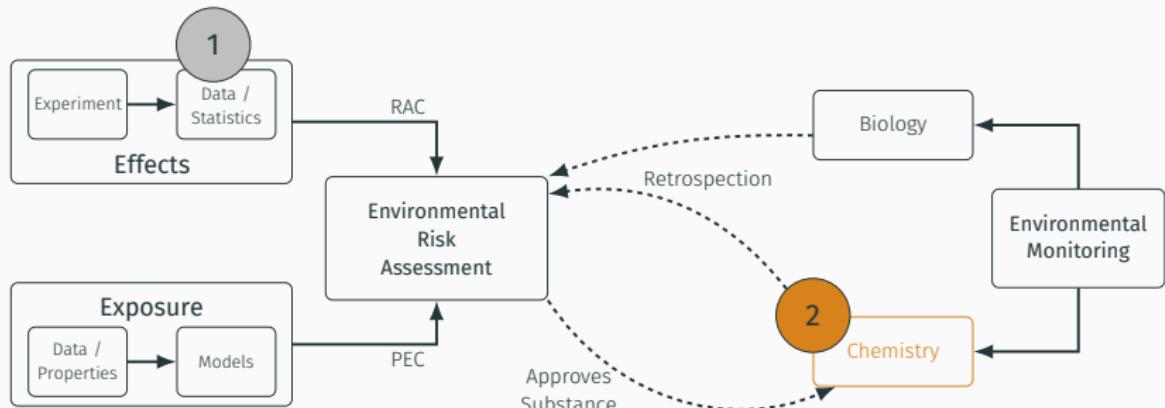
## Three points to consider ...

1. Choose your model based on data properties
2. Fix Type I errors by resampling
3. Models that better fit the data have better power properties

# Exploring Monitoring Data for ERA

Plant Protection Products  
Regulation 1107/2009

Water Framework Directive  
2000/60/EC



2

Szöcs, Brinke, Karaoglan & Schäfer (submitted). "Large scale risks from pesticides in small streams". Environmental Science & Technology.

# Exploring Monitoring Data for ERA

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## Goal: Can monitoring inform ERA?

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

# Goals & Hypotheses

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

## Hypotheses

1. Agricultural streams show highest concentrations
2. Small streams show highest concentrations
3. Precipitation at/before sampling increases concentrations
4. Highest concentrations in summer

# Analysing chemical concentrations

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- Concentrations < LOQ (96% of all measurements)

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- Hurdle-model:

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

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

# Analysing chemical concentrations

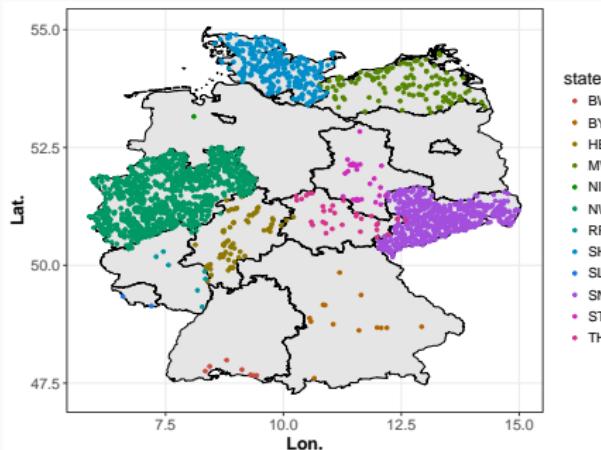
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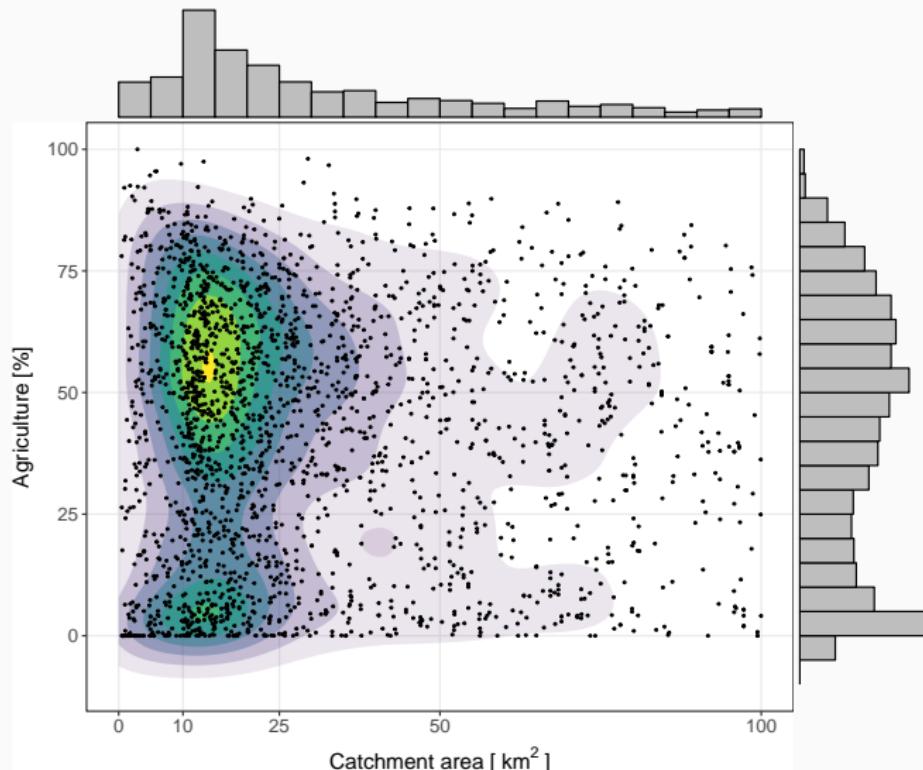
- Risk Quotient
  - $RQ = \frac{C}{RAC}$
- Predictors
  - Catchment Size
  - Agricultural land use
  - Precipitation

# Compiled data: Big, but inhomogeneous

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

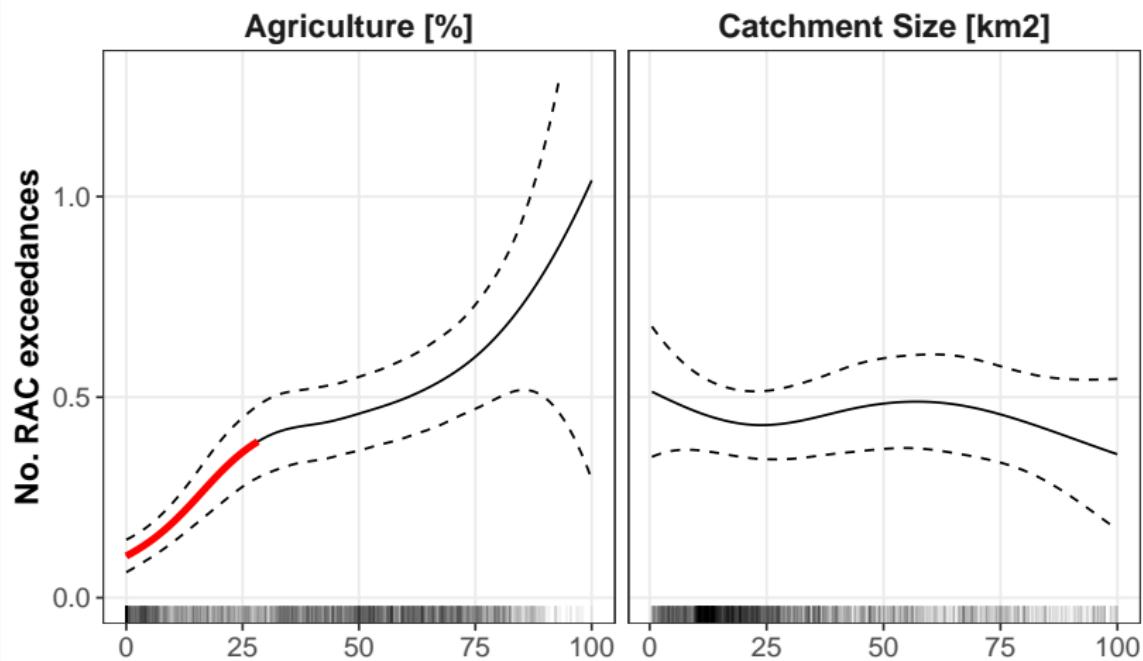


# Monitoring: Small streams are underrepresented

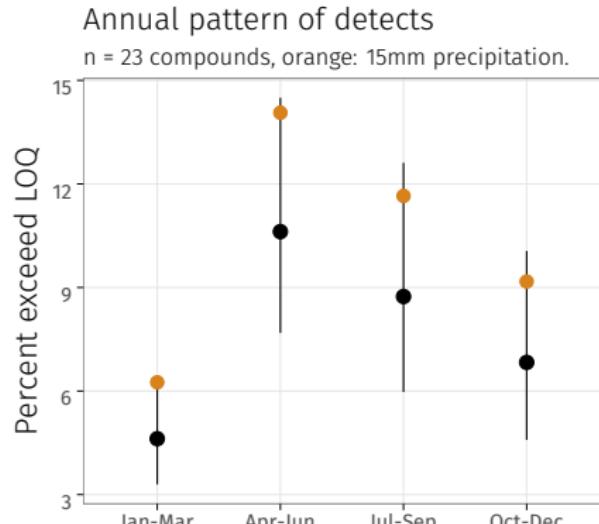


# Landscape: Factors influencing risk

- 25% of sites with at least one concentration > RAC
- 50% of sites with at least one concentration > RAC / 10

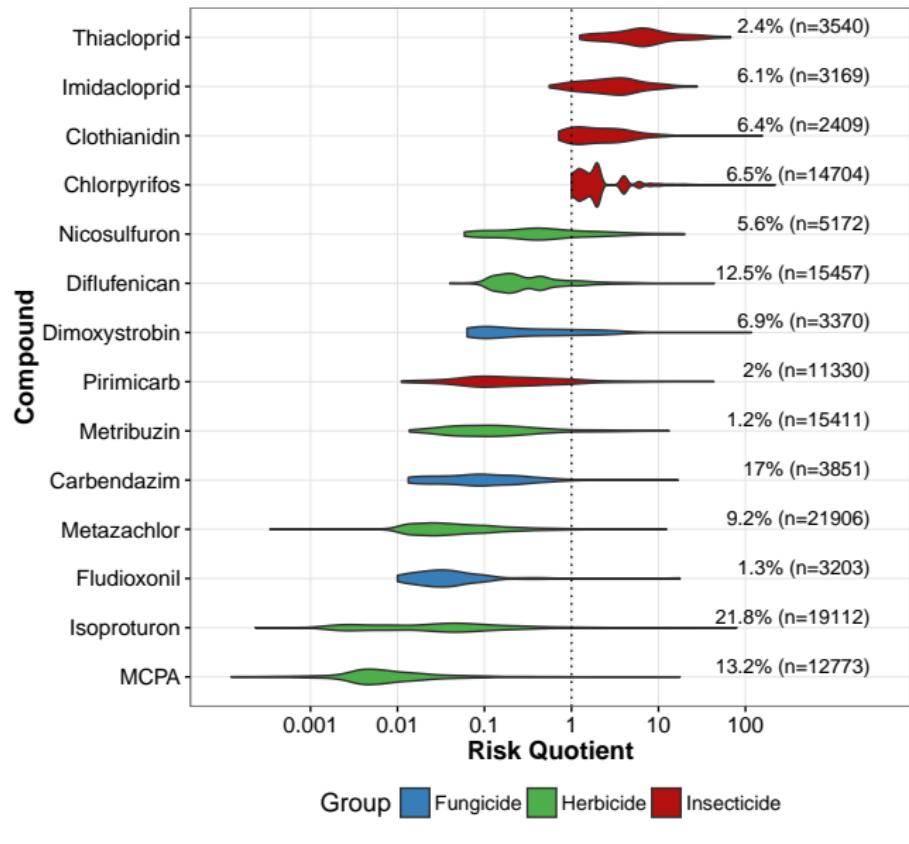


# Sampling: Factors influencing risk



- Peak in summer
  - compound specific
- Increase by **precipitation**
  - day before sampling
  - not at day of sampling
- absolute concentrations show much higher variability

# Risks: Compounds exceeding risk thresholds



# What we learned from this study

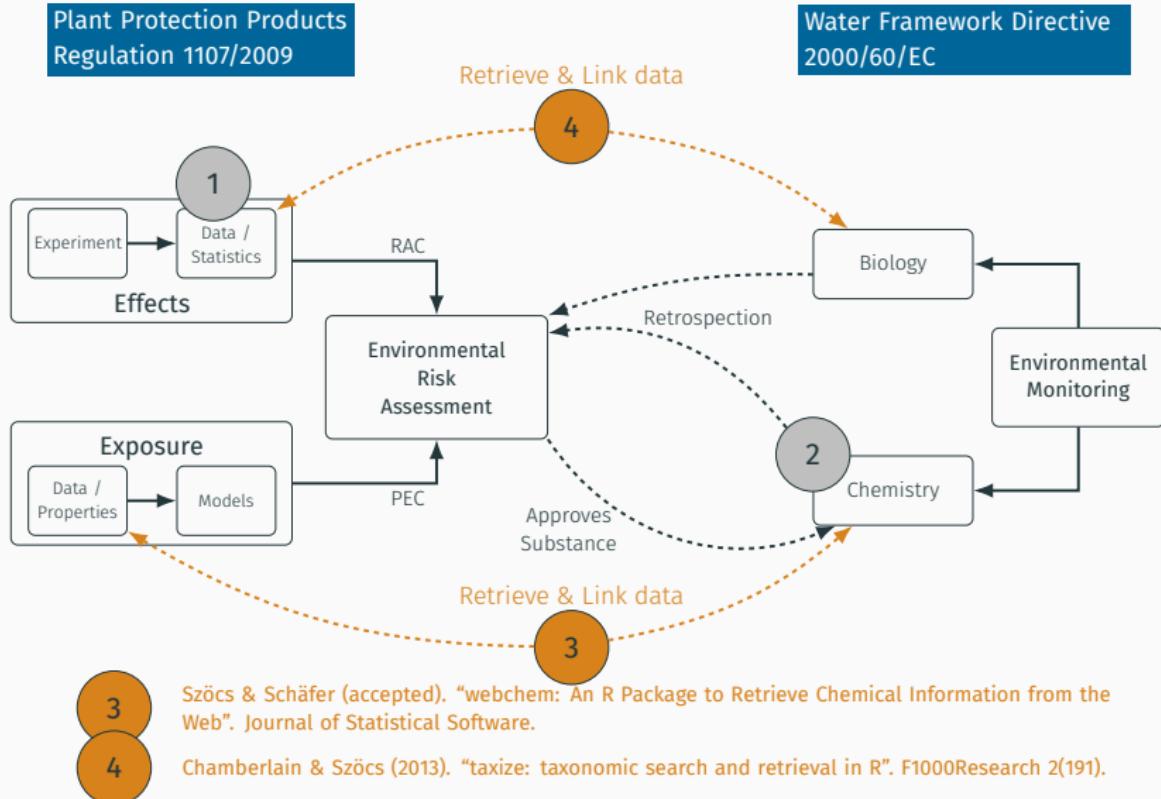
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1. Differences between states
2. Small streams are underrepresented
3. Agricultural sources
4. LOQ gives additional insights
  - Annual dynamics
  - Precipitation increases concentrations
5. Risk underestimated
6. Neonicotinoids

## Solutions for Data Handling in ERA

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# Solutions for Data Handling in ERA



Biologists & Chemists face the same problems

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# Biologists & Chemists face the same problems

## Names

*Osmia rufa, Osmia bicornis, Osmia ruffa, Osmia unilandauis*

Chlorpyrifos, Chlorpyriphos,  
Chlorphyrifos, Chlorpyrifot

## Biologists & Chemists face the same problems

## Names

*Osmia rufa*, *Osmia bicornis*, *Osmia ruffa*, *Osmia unilandauis* Chlorpyrifos, Chlorpyriphos, Chlorphyrifos, Chlorpypifot

## Hierarchies

Biologists & Chemists face the same problems

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*Osmia rufa*, *Osmia bicornis*, *Osmia ruffa*, *Osmia unilandauis* Chlorpyrifos, Chlorpyriphos, Chlorphyrifos, Chlorpypifot

## Hierarchies

## Traits / Properties

## Wing length, Mass, Season

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Biologists & Chemists face the same problems

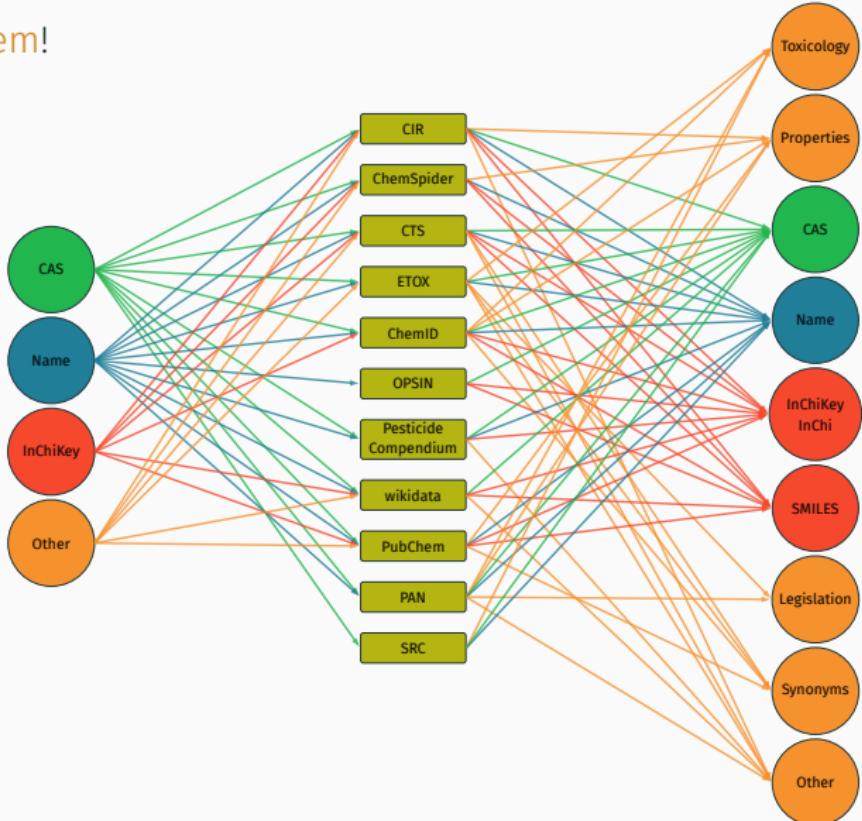
<u>Names</u>	
<i>Osmia rufa</i> , <i>Osmia bicornis</i> , <i>Osmia ruffa</i> , <i>Osmia unilandauis</i>	Chlorpyrifos, Chlorpyriphos, Chlorphyrifos, Chlorpypifot
<u>Hierarchies</u>	
Hymenoptera/ Apoidea/ Megachilidae/ <i>Osmia/ rufa</i>	organophosphate, ester, insecticide
<u>Traits / Properties</u>	
Wing length, Mass, Season	Mass, $K_{OW}$ , $LC_{50}$
<u>Identifiers</u>	
NCBI, ITIS, EOL, ...	2921-88-2, Clc1c(OP(=S)[...], InChI=1S/C9H11C[...], SBPBAQFW[...], CSID,...

Biologists & Chemists face the same problems

<u>Names</u>	
<i>Osmia rufa</i> , <i>Osmia bicornis</i> , <i>Osmia ruffa</i> , <i>Osmia unilandauis</i>	Chlorpyrifos, Chlorpyriphos, Chlorphyrifos, Chlorpyrifot
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NCBI, ITIS, EOL, ...	2921-88-2, Clc1c(OP(=S)[...], InChI=1S/C9H11C[...], SBPBAQFW[...], CSID,...
<u>Amount of data</u>	
2993 taxa	478 pesticides

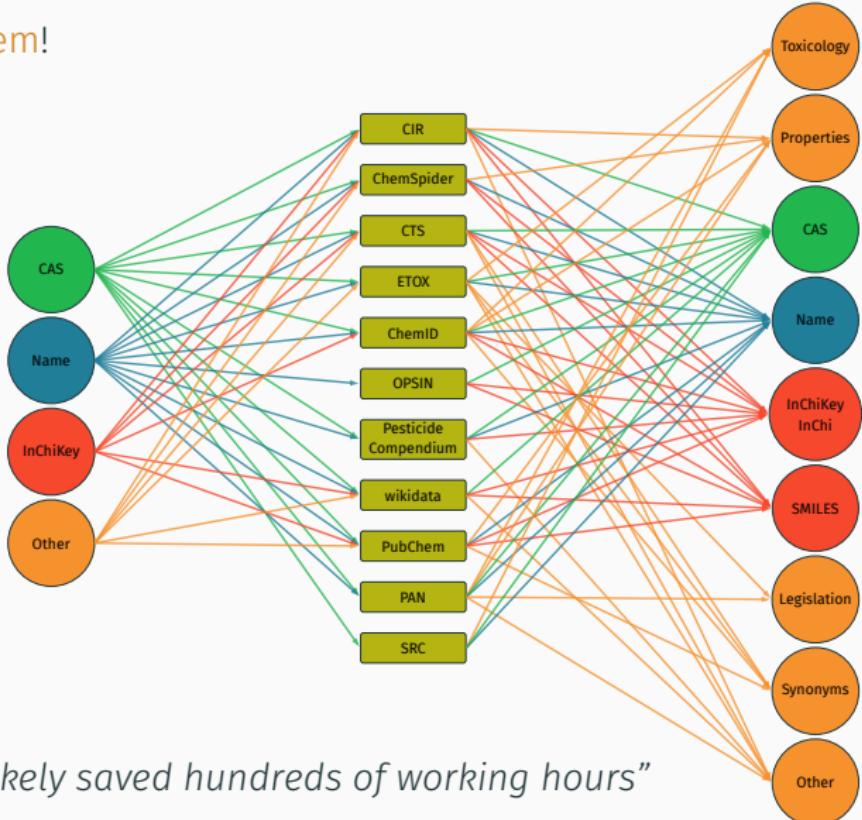
# Instead of wasting time...

... use webchem!



# Instead of wasting time...

... use webchem!



*"webchem ...likely saved hundreds of working hours"*

# Instead of wasting time...

... use taxize!

A grid of logos for various biodiversity databases and tools:

- ITIS**: Global Invasive Species Database
- iPlant Collaborative™**
- Plantminer**
- Tropicos®**
- GBIF**
- NCBI**
- eOL Encyclopedia of Life**
- gnl**
- RED LIST**
- ubio**
- Canadensys**
- ThePlantList**

# Instead of wasting time...

... use taxize!

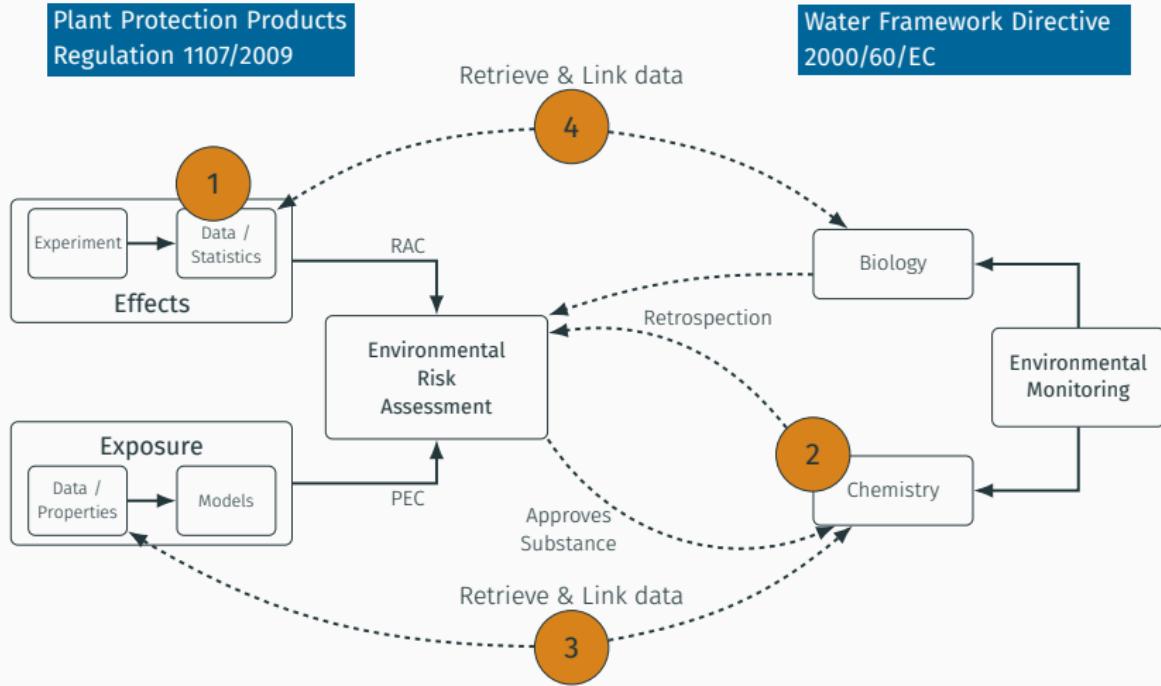
The image shows a collection of logos for different biological databases and projects, all related to taxonomy or biodiversity. The logos are arranged in two rows. The top row includes: ITIS (with a green circular icon containing a tree, bird, and flower), iPlant Collaborative (with a blue circular icon and the text "iPlant Collaborative"), and Plantminer (with a green circular icon containing dots and the text "P L A N T M I N E R"). The bottom row includes: Global Invasive Species Database (with a green circular icon containing a tree, bird, and flower), Catalogue of Life (with a colorful horizontal bar and the text "Catalogue of Life"), Tropicos (with a green circular icon containing a tree and the text "Tropicos"), GBIF (with a green circular icon containing a leaf and the text "GBIF"), NCBI (with a blue square icon containing a white stylized "N" and the text "NCBI"), eOL (with a green circular icon containing a leaf and the text "eOL Encyclopedia of Life"), gni (with a green circular icon containing colored dots and the text "gni"), RED LIST (with a red circular icon containing a red paw print and the text "RED LIST"), ubio (with a green circular icon containing a leaf and the text "ubio"), Canadensys (with a red circular icon containing a leaf and the text "Canadensys"), and ThePlantList (with a green circular icon containing a leaf and the text "ThePlantList").

"Days of searching done during my morning coffee. Amazing. **taxize**."

## Recap

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# Recap: What did I look at?



# Recap: What we learned from my PhD Thesis

## ✓ Improving Statistics in ERA

- Change your model, not your data
- Ultimately ban NOEC
- Take LOQ into account

# Recap: What we learned from my PhD Thesis

## ✓ Improving Statistics in ERA

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## ✓ Exploring Monitoring Data for ERA

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

# Recap: What we learned from my PhD Thesis

## ✓ Improving Statistics in ERA

- Change your model, not your data
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## ✓ Exploring Monitoring Data for ERA

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- Feedback for ERA

## ✓ 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/>

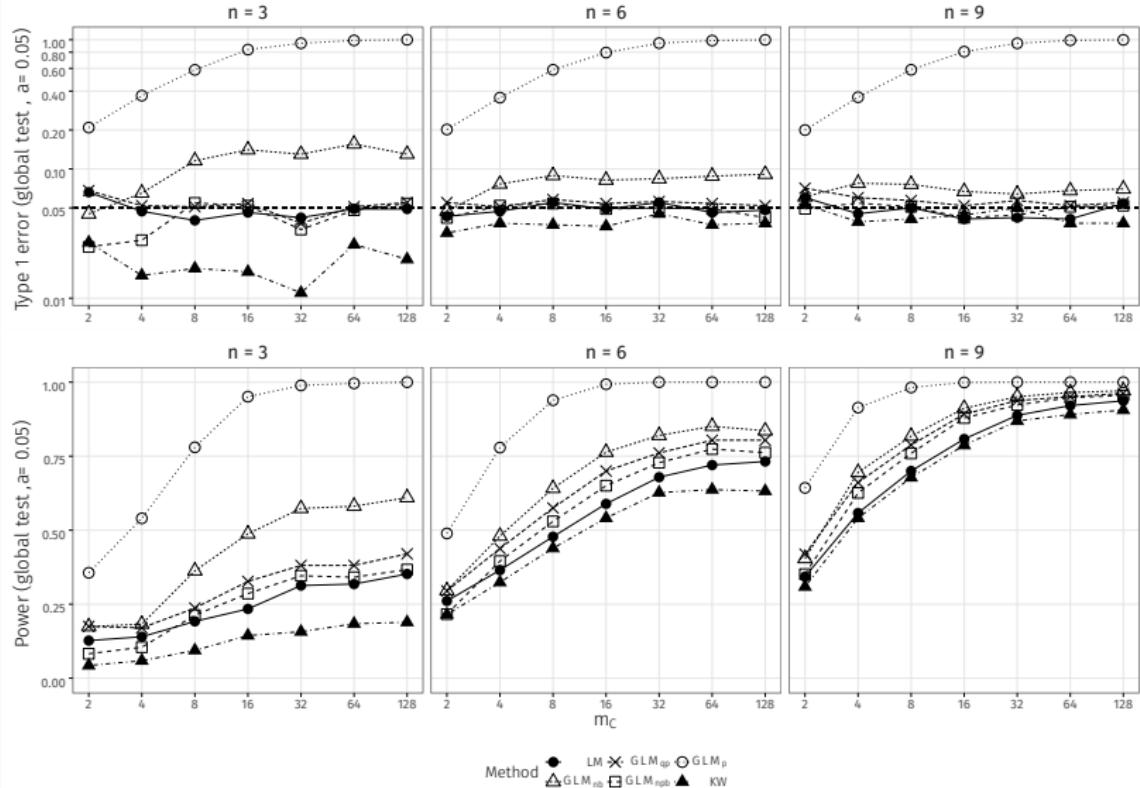
🐦 [@EduardSzoebs](https://twitter.com/EduardSzoebs)

📄 [https://github.com/edild/phd\\_defense](https://github.com/edild/phd_defense)

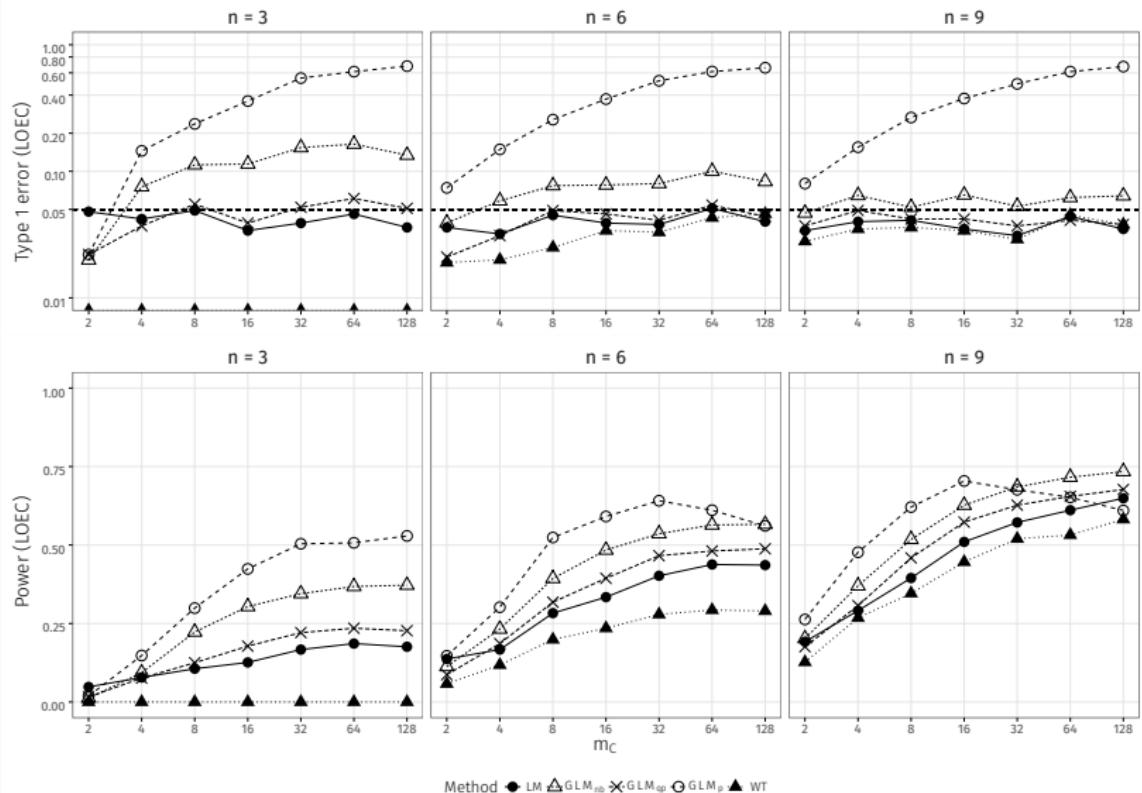
📄 [https://github.com/edild/phd\\_thesis](https://github.com/edild/phd_thesis)



# Power en detail



# For LOEC it is even worse



# Comparison with Ives...

## Szöcs (2015)

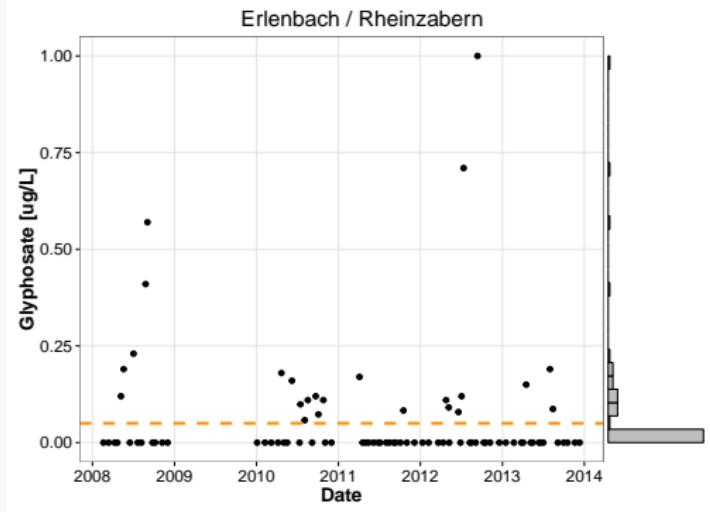
- factorial design
- one predictor
- low replicated
- LM, GLM, bootstrap
- High T1 error of NB
- Quasi-Poisson worked well
- Bootstrap fixes the problems

## Ives (2015)

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

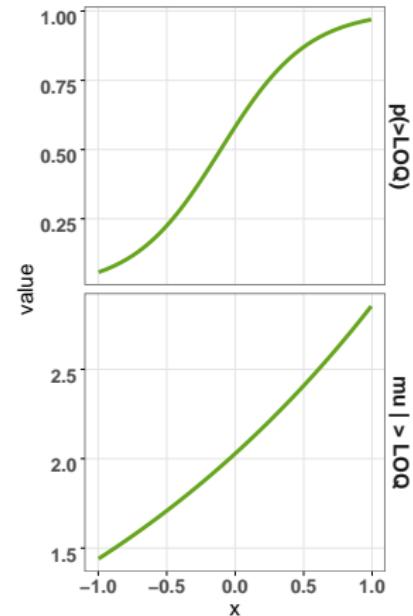
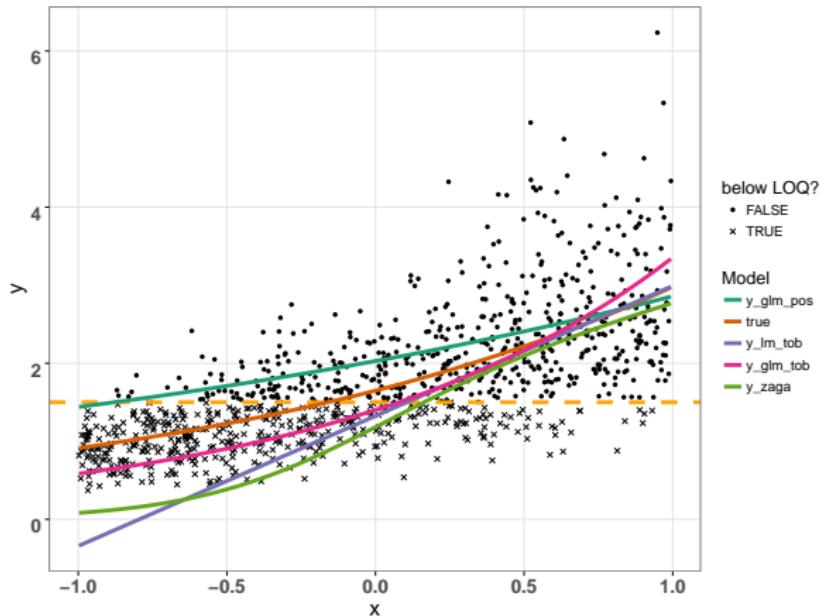
## Idiosyncrasies of chemical concentrations

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

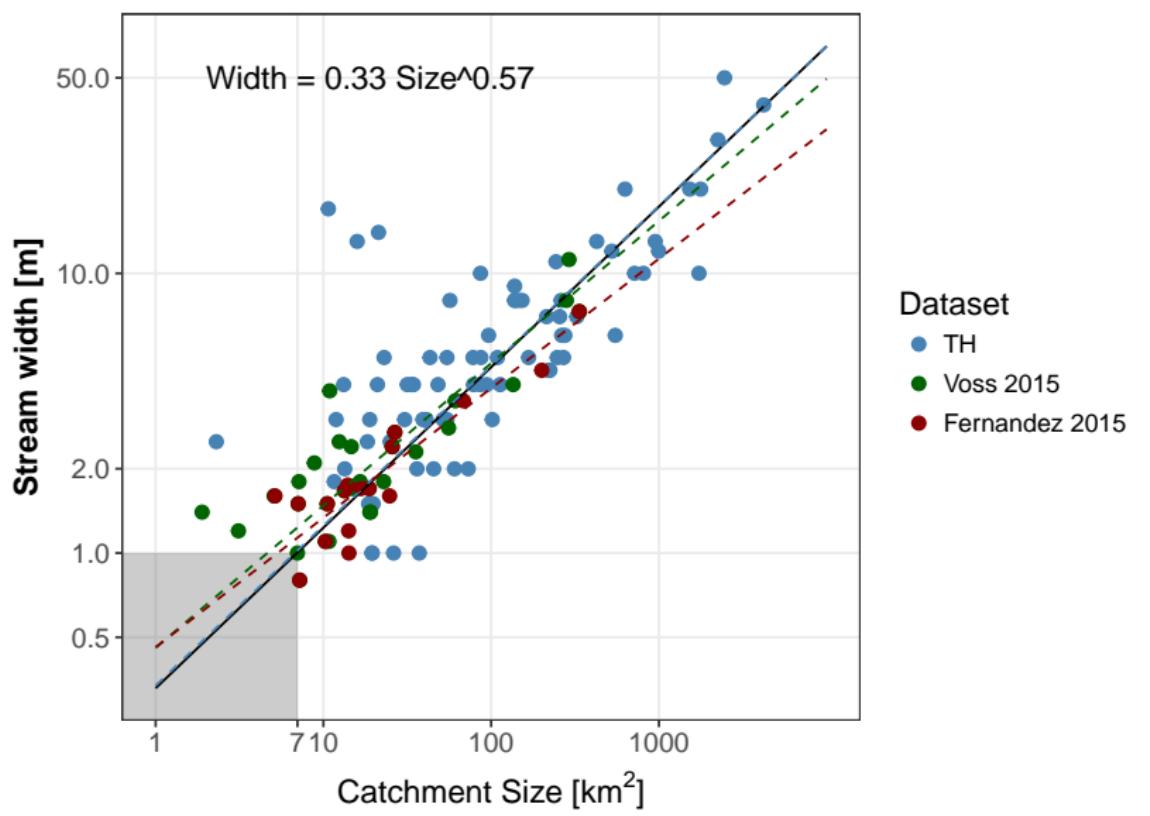


# ZAGA what...?

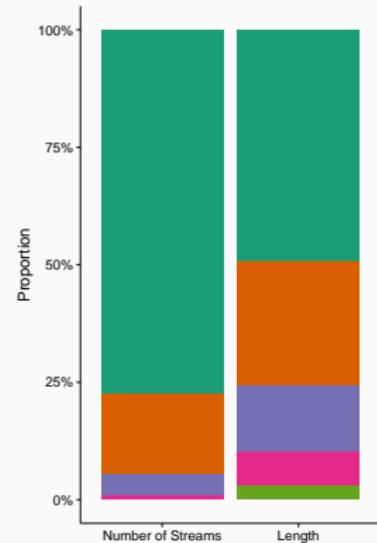
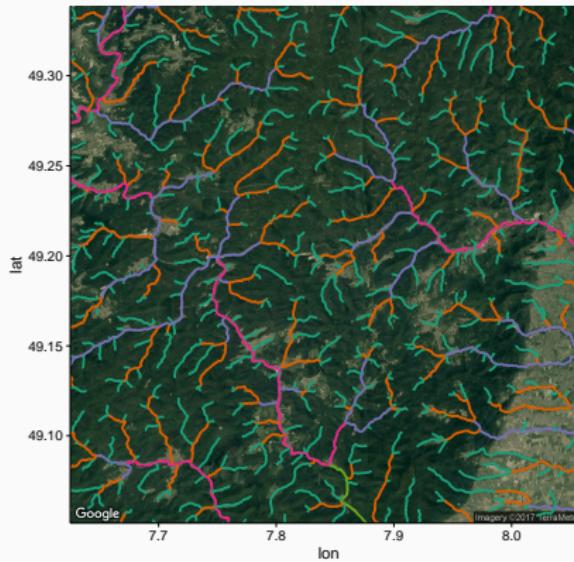
shiny app: <http://uni-ko-ld.de/g4>



# Stream size / width



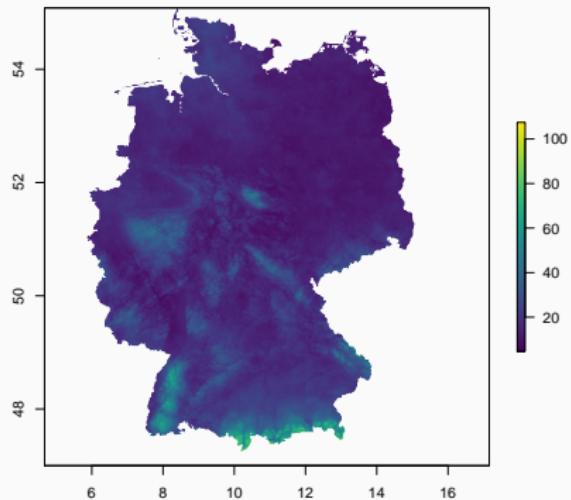
# Importance of small streams



- Biodiversity
- Refuge for re-colonisation

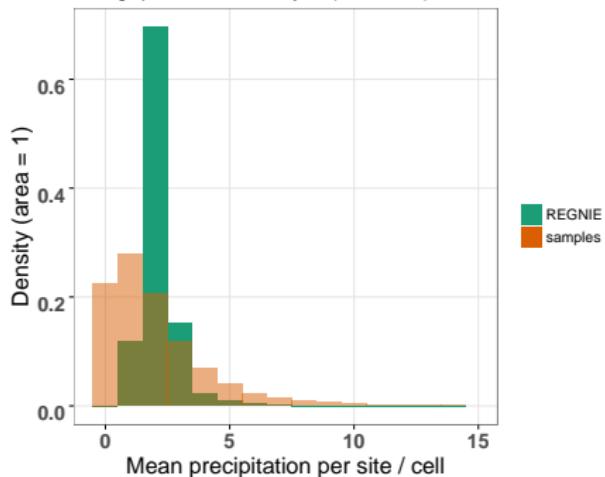
# Precipitation in Germany and the samples

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
- Neonics + Chlorpyrifos (OP)
- ZAGA (<LOQ)

## Stehle (2015)

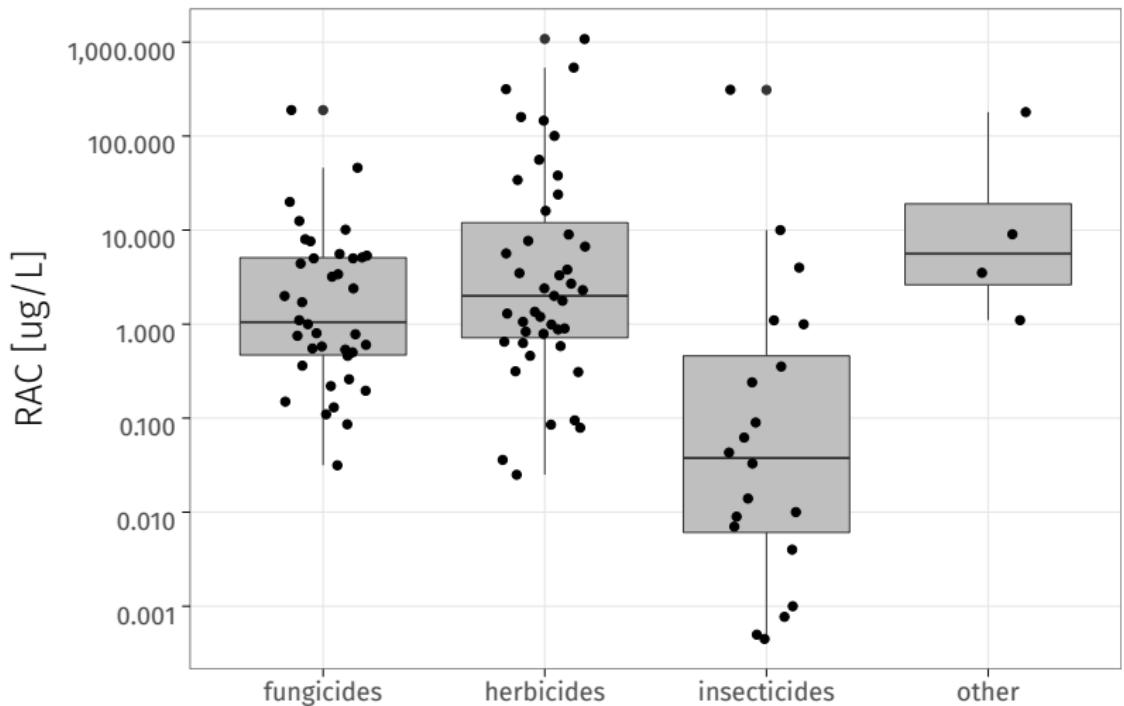
- Europe / Global
- Publications
- Grab & Event driven sampling
- Insecticides
- Organophos.+ Pyrethroids
- LM for >LOQ

## Knauer (2016)

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

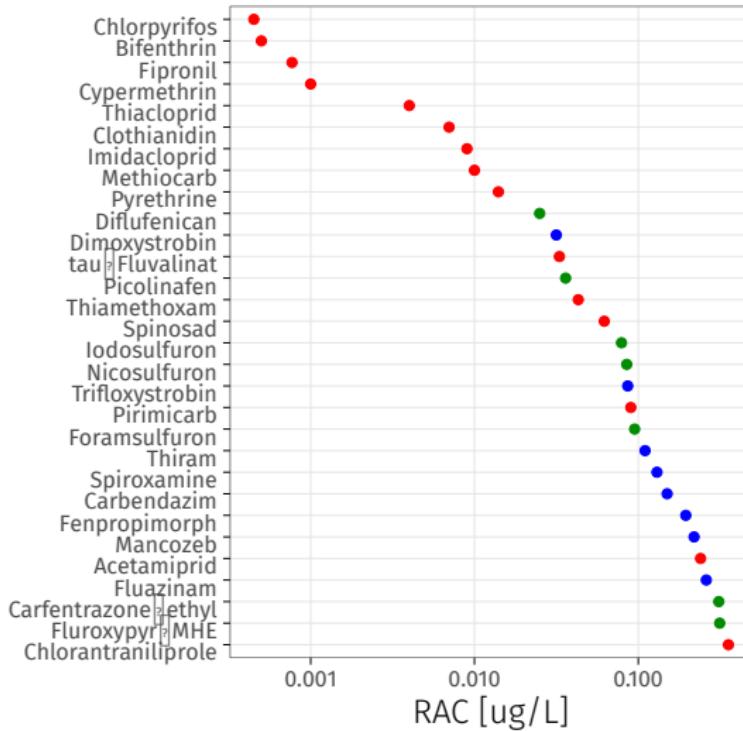
# RACs by Type

105 RACs provided by UBA splitted by group



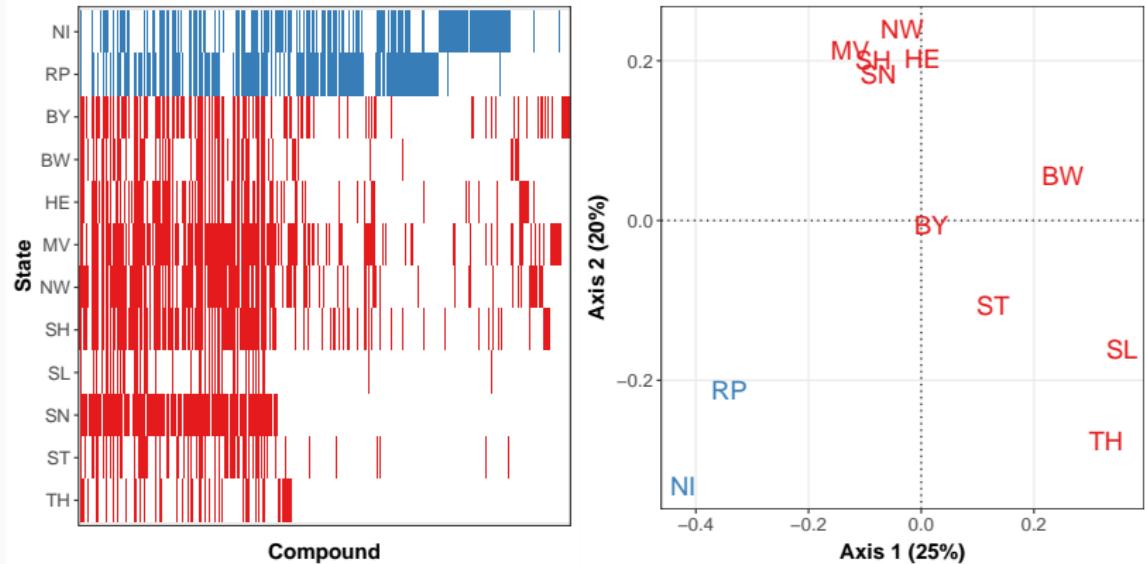
# RACs by Compound

30 lowest RACs

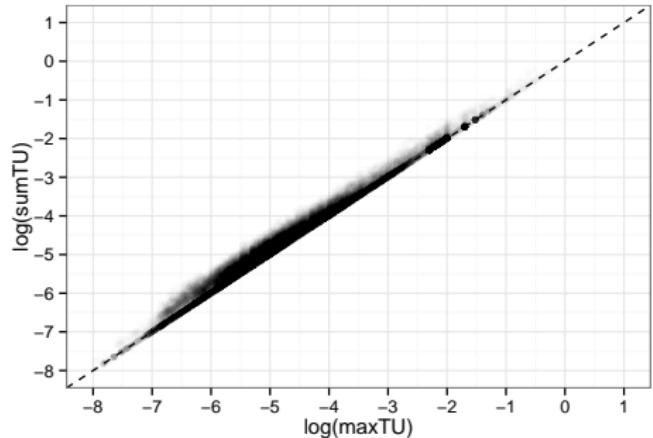


Type • fungicides ● herbicides • insecticides

# Analysed compound spectra by state



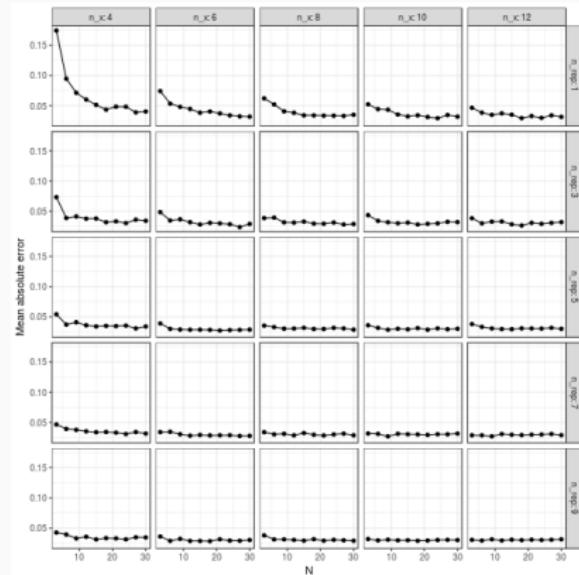
# Mixtures are common, but one compound dominates the risk



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

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

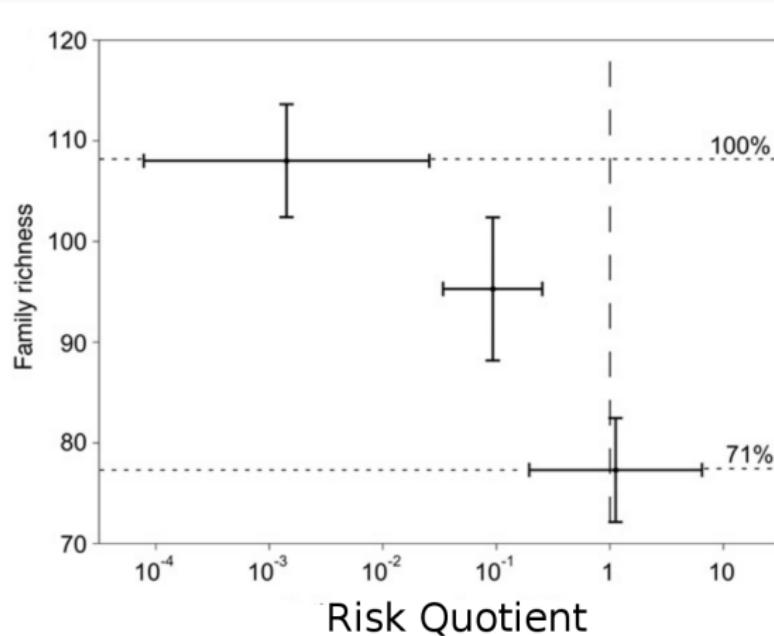
Experimental design for dose-response experiments - a simulation  
[http://edild.github.io/lc50\\_bias\\_sim/](http://edild.github.io/lc50_bias_sim/)



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

# Effects of RAC exceedances

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

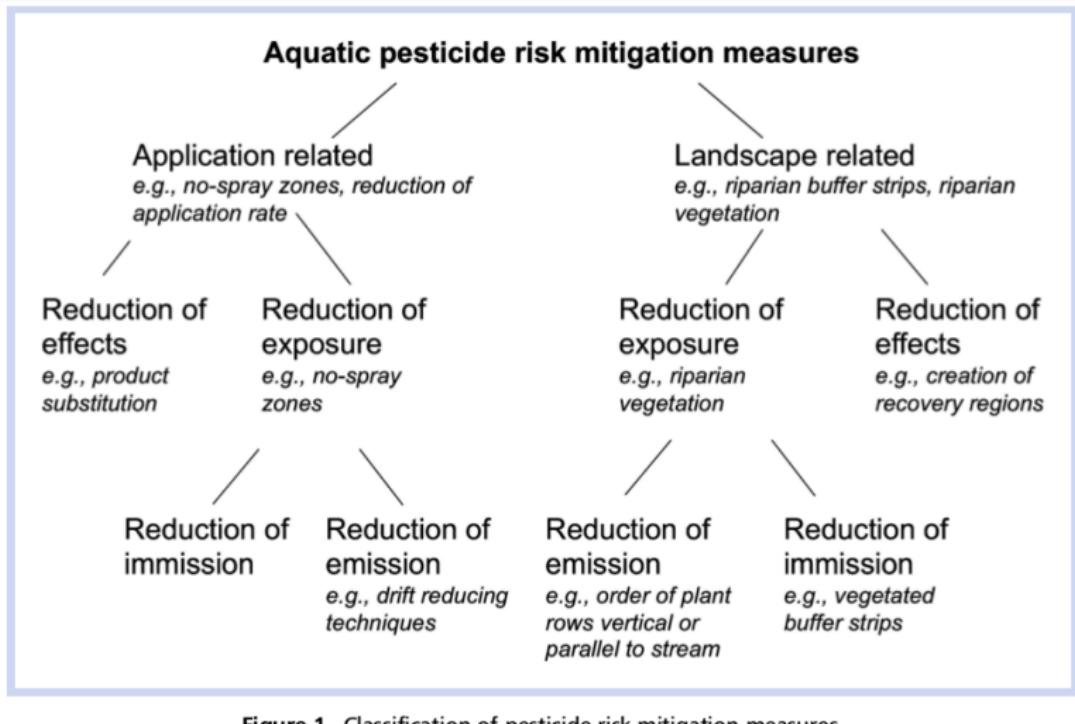


Modified from: Stehle, Schulz (2015). Agricultural insecticides threaten surface waters at the global scale. PNAS 112, 5750–5755.

# 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

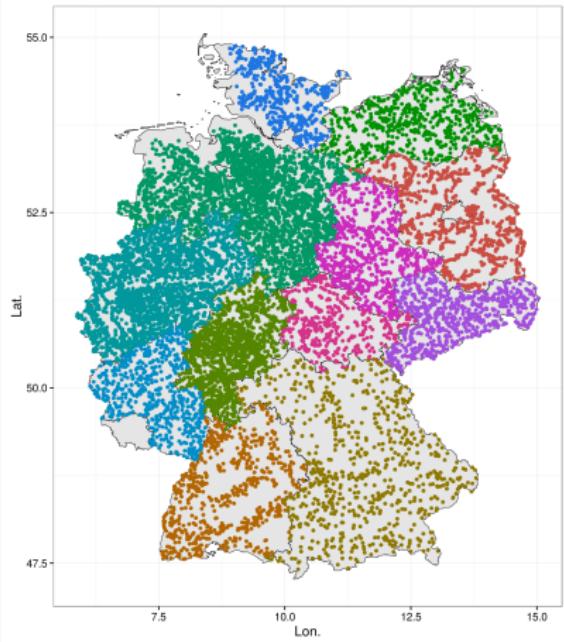
# Risk Mitigation Measures



**Figure 1.** Classification of pesticide risk mitigation measures.

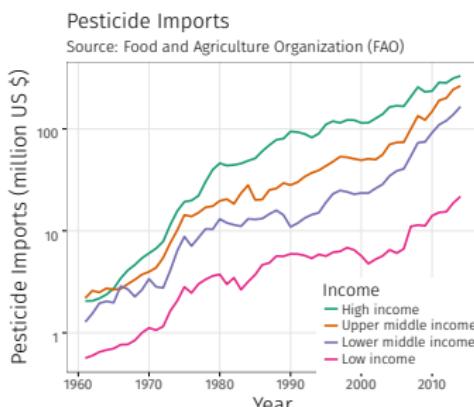
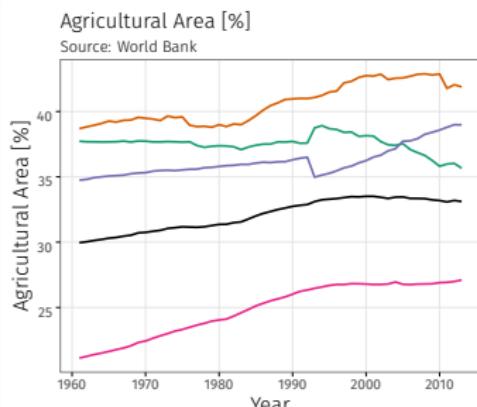
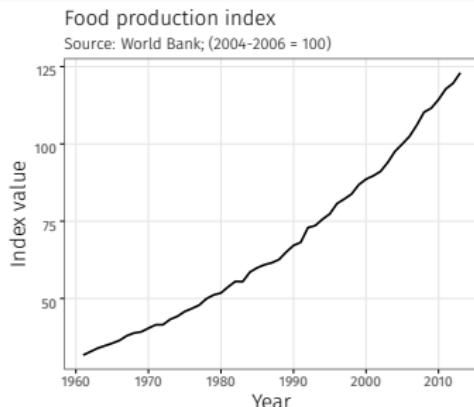
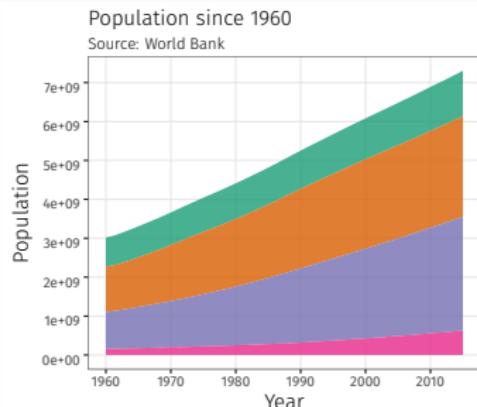
Bereswill, Streloke, Schulz (2014). Risk mitigation measures for diffuse pesticide entry into aquatic ecosystems: Proposal of a guide to identify appropriate measures on a catchment scale: Guide to Identify Pesticide Risk Mitigation Measures. IEAM 10, 286–298.

# Biotic field effects



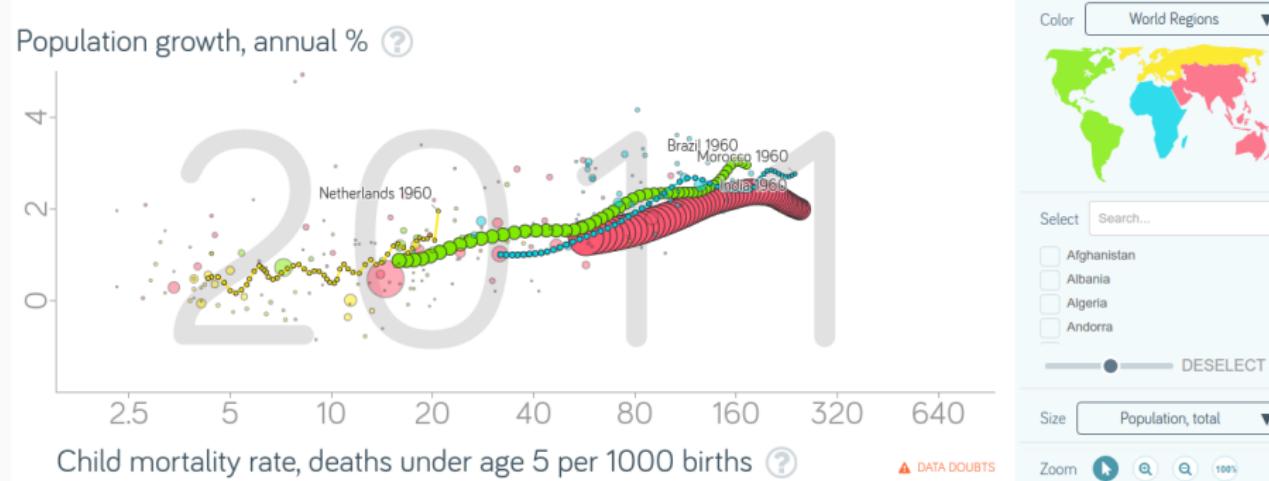
- biological data with good spatial coverage
- 60% of spatial congruence
- Large scale effects largely unknown.
- Some work left...
- Future....

# A global perspective (I)



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

## A global perspective (II)



# Software availability

Stable versions on CRAN, dev versions on github.

`webchem` [github.com/ropensci/webchem](https://github.com/ropensci/webchem)

`taxize` [github.com/ropensci/taxize](https://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)
- My colleagues & collaborators (too many to list here)
- German Environment Agency (for funding & collab)
- My parents Anca & Helmut (for their support)
- My girlfriend Anja (for everything)