

1 **Progressing statistical analysis for regulatory** 2 **ecotoxicology: developments, processes, and opinions**

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12 **Editor's Note:** This article is part of the special series “Statistical Analysis of
13 Ecotoxicology Data for Regulatory Purposes.” The series provides an overview of
14 approaches in statistical ecotoxicology and reflects recent developments, processes
15 and debates. The papers represent viewpoints from academia, industry and
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4 **Abstract**

5 Statistical methods and computational tools applicable to ecotoxicology have developed
6 and improved over recent decades, but regulatory risk assessments haven't kept pace,
7 instead relying on outdated statistical methods and tools. Sessions in recent SETAC
8 meetings dedicated to the development of statistical methods and tools for
9 ecotoxicology have generated a high level of interest and discussion. Ecotoxicologists
10 have expressed interest in the revision of the 2006 OECD document No. 54, "Current
11 approaches in the statistical analysis of ecotoxicity data: a guidance to application"
12 because they believe that the guidance is no longer reflective of contemporary statistical
13 methods and tools; its revision is ongoing. Against this background, a call for papers for
14 Special Series in IEAM was launched in autumn 2024 with manuscript submissions
15 accepted until summer 2025. The purpose of this Special Series was to survey the
16 statistical ecotoxicology landscape and reflect on recent developments, processes, and
17 opinions. The editors of the Special Series welcomed and encouraged contributions and
18 viewpoints from all employment sectors. The resulting series comprises eleven papers
19 (seven original articles, three brief communications, and a workshop synthesis), with
20 authors from academia, industry, government, and other research organizations. The
21 series covers topics such as: recent progress in concentration-response modelling and
22 hypothesis testing, including use of generalized linear models; strengths and
23 weaknesses of established toxicity metrics (EC_x , NOEC, BMD); application of toxicity
24 metrics for species sensitivity distribution (SSD) modelling; added values of mechanistic
25 effect modelling; and case studies highlighting opportunities and challenges related to
26 the various statistical methods. This introductory paper aims to give an overview of the
27 papers in the Special Series, summarizes their main topics and methods, and
28 addresses further challenges to progress statistical analysis for regulatory
29 ecotoxicology.

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31 **Keywords:** concentration-response modelling, generalized linear and additive models,
32 mixed-effect models, hypothesis testing, regulatory statistics

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34 **Key points:**

- 35 1) The ongoing revision of guidance for statistical analysis of ecotoxicity data has
36 inspired scientific developments and debates involving industry, government,
37 academia and other research organizations.
- 38 2) This paper introduces a Special Series of eleven papers, which cover novel
39 approaches, applications, and assessments of methods related to concentration-
40 response modelling, hypothesis testing, and derived toxicity metrics for regulatory
41 use.
- 42 3) Updated guidance on statistical analysis should be followed by revised
43 recommendations for experimental design for ecotoxicity studies.

44

45 Introduction

46 Statistical methods and computational tools applicable to ecotoxicology have developed
47 and improved over recent decades, but risk assessments for regulatory purposes have
48 not kept pace and rely on outdated statistical methods and tools that are no longer
49 considered state of the art (Laskowski, 1995; Moe et al., 2025; OECD, 2002;
50 Wasserstein & Lazar, 2016). OECD document No. 54 (“Current approaches in the
51 statistical analysis of ecotoxicity data: a guidance to application”) was published in 2006
52 and contained practical recommendations for statistical analysis of experimental test
53 results in ecotoxicology (OECD, 2006). In the 20 years since its release, statistical best
54 practices have evolved, also within ecotoxicology.

55 Since 2024, the revision of OECD document No. 54 is ongoing, and a revised version is
56 expected for publication in 2027. The initial phase of this revision process, led by the
57 German Environment Agency (*Umweltbundesamt*; UBA), was publicly presented as a

58 poster during the SETAC Europe 34th Annual Meeting (Daniels et al., 2024). In
59 September 2024, the UBA also organized a workshop with a hybrid part to present the
60 initiative and an in-person part to identify and discuss priority topics, as well as an
61 additional online recap event for this workshop (Daniels et al., 2025a). Inspired by this
62 initiative, the guest editors of this Special Series have expressed both short-term wishes
63 and long-term visions for the revision of OECD document No. 54 and beyond (Moe et
64 al., 2025).

65 Set against this background, the purpose of this Special Series is to provide an
66 overview of the statistical ecotoxicology landscape and reflect on recent developments.
67 Contributions to this Special Series were invited from all employment sectors, i.e.,
68 academia, industry, government, and other research organizations. The Special Series
69 was announced through the SETAC website, LinkedIn posts by the guest editors, the
70 online recap event for the UBA workshop, and SETAC Annual Meetings in 2024 (North
71 America) and 2025 (Europe).

72 Special Series articles summary

73 The Special Series contains seven original articles, three brief communications, and a
74 workshop synthesis, which showcase current progress in statistical analysis for
75 regulatory ecotoxicology, and exploratory research that will inform future progress, with
76 a strong link to the currently ongoing revision of OECD document No. 54. Article 1
77 (workshop synthesis) provides broad perspectives for the current developments within
78 the field. Articles 2–6 cover various aspects of concentration-response models for
79 deriving toxicity metrics (EC_x and BMDL). Article 7 addresses the use of such metrics in
80 species-sensitivity distribution (SSD) studies. Articles 8 and 9 focus on developments
81 for deriving no observed effect concentrations (NOECs). Article 10 discusses a method
82 for analyzing ecotoxicity data from field studies. Finally, Article 11 highlights mechanistic
83 effect modelling as a complementary approach to empirical data analysis. Below we
84 provide a summary of each of the eleven articles.

85 1. Daniels et al., 2025a: **“High time to update statistical guidance in**
86 **ecotoxicology—a workshop synthesis on the revision of OECD document No.**

87 **54**". This paper sets the scene for the Special Series, by providing background
88 information on the revision process and main steps and outcomes through summer
89 2025. It provides a thorough summary of the workshop activities, debates and main
90 outcomes. The workshop identified the following key topics for improved guidance,
91 as ranked by the number of notecards: (1) improved user-friendliness, (2) guidance
92 on model selection and test settings, (3) scope of OECD document No. 54, (4)
93 improved assessment concepts and statistical approaches, e.g., generalized linear
94 models (GLMs), equivalence testing, alternative toxicity metrics, Bayesian
95 approaches; (5) quality assessment and diagnostics of tests and methods, (6) TK-
96 TD (toxicokinetic-toxicodynamic) modelling and GUTS (General Unified Threshold
97 model of Survival); and (7) software use. The paper highlights the dilemma between
98 scientific drive for progress and regulatory need for standardization, which is a
99 common theme for the following papers.

- 100 2. Hill et al., 2025: "**Multilevel modeling of concentration-response data can**
101 **improve risk assessment: a case study of copper effects on fish**". This study
102 applies a multi-level or mixed generalized non-linear model to account for different
103 sources of variation in a concentration-response model. The authors highlight the
104 multi-level approach for making better use of available data from multiple
105 experiments. This paper also shows the efficient use of a generalized model for
106 fitting a concentration-response model by logistic regression, although with
107 constraints on the shape (symmetrical tails).
- 108 3. Erickson et al., 2026: "**Implications of asymmetric concentration-response**
109 **relationships on effect concentration estimation: A case study regarding the**
110 **chronic toxicity of NaCl to *Ceriodaphnia dubia***". The case study explores the
111 importance of asymmetry in concentration-response relationships using logistic
112 regression. The authors offer both symmetrical and asymmetrical versions of their
113 model. The symmetric version of their model is similar to the model used by Hill et
114 al. (2025); the asymmetric version allows for skewed tails. Toxicity metrics in the
115 lower tail of the concentration-response relationship, e.g., EC₁₀, were more sensitive
116 to the symmetry assumption than EC₅₀ estimates.

- 117 4. Daniels et al., 2025b: **“Four-parameter nonlinear regression and maximum**
118 **achievable effect in ecotoxicology: just visually appealing or relevant for risk**
119 **assessment?”**. This paper explores another aspect of concentration-response
120 models, namely a maximum achievable effect level, as represented by the fourth
121 parameter in a four-parameter regression (4PR). While 4PR models can be
122 analyzed in custom-built tools, the resulting EC_x value, i.e., relative EC_x , can deviate
123 from the absolute EC_x . The authors call for higher transparency regarding the
124 analysis and reporting of relative versus absolute EC_x .
- 125 5. Baalkilde & Jensen, 2025: **“Hybrid method benchmark dose estimation with a**
126 **heterogeneous variance structure”**. The benchmark dose (BMD) concept is now
127 established in ecotoxicology but was not covered in OECD document No. 54 in
128 2006. This paper discusses the importance of proper variance structure in a
129 concentration-response model and demonstrates how the variance structure
130 assumption can affect the resulting lower benchmark dose (BMDL). This approach
131 can be seen as a compromise between a non-linear concentration-response model
132 and a GLM.
- 133 6. Cedergreen et al., 2026: **“How normalization of reporter gene *in vitro* assays**
134 **affects the results”**. This paper highlights the importance of rigorous experimental
135 control validation and data normalization in *in vitro* bioassays, particularly when
136 applied to environmental samples, e.g., wastewater. The authors also call for more
137 investigations of how exposure duration affects the response of *in vitro* test systems.
- 138 7. Hu et al., 2026: **“Ecotoxicological Risk of Sulfate in Baltic Sea Brackish Water**
139 **Assessed Using Model-Averaged Species Sensitivity Distribution”**. This paper
140 applies a toxicity metric, i.e., EC_{10} , in SSD models to derive predicted no-effect
141 concentrations (PNECs) for environmental risk assessment. Key methodological
142 aspects in this study include the model-averaging of SSDs, and evaluation of
143 alternative assessment factor scenarios to account for different types of uncertainty.
- 144 8. Schimera et al., 2025: **“The role of background variability for interpreting**
145 **biological relevance and statistical significance in Collembola soil field**
146 **studies”**. Based on a field experiment, this study explores the use of closure
147 principal computational approach test (CPCAT) for increasing power in hypothesis

148 testing. Shortcomings of this approach are identified for field studies with high
149 background variability, e.g., over-dispersed data compared to a Poisson distribution.
150 The authors emphasize the importance of study- and sampling-specific control data
151 for the assessment of biological relevance.

152 9. Gao et al., 2026: “**The role of statistical power in context: implications for**
153 **regulatory practices**”. This paper contains a critical review of current statistical
154 practices for hypothesis testing in ecotoxicology, followed by data examples and
155 power comparisons for various statistical approaches. Focusing on NOECs, the
156 authors stress the need for robust methodologies that can accommodate both the
157 complexities of ecological data and the practical requirements of regulatory
158 frameworks.

159 10. Vermeiren et al., 2026: “**Getting most out of ecotoxicological field data with**
160 **generalized linear and additive mixed models: a 6-step analysis framework**”.
161 The authors demonstrate the advantages of GLMs and generalized additive models
162 (GAMs) for analyzing responses with different data types, e.g., counts, proportions
163 and continuous data, as well covariates and other potential factors from field studies
164 within a single analysis while considering the repeated sampling. An analysis
165 framework is proposed to guide the use of GLMs and GAMs in ecotoxicology,
166 supported by the long tradition of these methods in ecology.

167 11. Jager, 2025: “**It’s about time: moving away from statistical analysis of**
168 **ecotoxicity data**”. This brief communication points out a general shortcoming for
169 statistical methods for ecotoxicology: that toxicity metrics derived from
170 concentration-response models represent a fixed exposure duration, rather than the
171 full exposure pattern as a function of time and argues that mechanistic effect models
172 can better support a more comprehensive concentration-time-response analysis.

173

174 Conclusion and recommendations

175 Statistical practices in ecotoxicology are in flux, and the revision of OECD document No.
176 54 serves as a catalyst for a critical review of current practices. The contributions to this

177 Special Series cover many priority topics that were presented during the UBA online
178 recap event, including, for example:

- 179 • Parametrization of concentration-response models (Daniels et al., 2025b; Hill et al.,
180 2025; Erickson et al., 2026).
- 181 • Use of GLMs – in particular multilevel, i.e., GLMMs –, both for concentration-
182 response models (Hill et al., 2025) and for hypothesis testing (Vermeiren et al.,
183 2026).
- 184 • Various data treatment-related issues, e.g., variability (Baalkilde & Jensen, 2025)
185 and scaling (Cedergreen et al., 2026).
- 186 • Additional challenges related to analysis and interpretation of field studies
187 (Vermeiren et al., 2026; Schimera et al., 2025).
- 188 • Discussions on the appropriateness and relevance of NOECs (Gao et al., 2026;
189 Schimera et al., 2025).
- 190 • The use of derived toxicity metrics in downstream assessments (Hu et al., 2026).
- 191 • Shortcomings of traditional statistical approaches in ecotoxicology, compared to
192 mechanistic effects models (Jager, 2025).

193 Several other topics related to statistical analysis have been discussed at the UBA
194 workshops (Daniels et al., 2025a) as well as in other fora. In our view, the following
195 topics merit more in-depth coverage: more extensive use of generalized and multi-level
196 principles for concentration-response modelling; added value of probability-based
197 Bayesian approaches; and equivalence testing as an alternative to traditional
198 hypothesis testing.

199 Moreover, in the wider context of regulatory risk assessments, the type of toxicity
200 metrics addressed in this series (e.g., EC_x, BMD, and NOEC) are only steppingstones
201 on the path from data to decisions. The conceptual linkage between uncertainties in
202 ecotoxicological studies and regulatory risk assessments would also benefit from
203 statistical scrutiny. Furthermore, updated statistical methods and best practices should
204 influence and optimize guidance on experimental designs.

205 Two decades after the first guidance documents were released by OECD (2006) and
206 ISO (2006), momentum is finally being built up to improve the use of statistical
207 approaches for regulatory ecotoxicology. Important landmarks, such as the revision of
208 OECD document No. 54, will contribute to improved practices and will advance data
209 literacy within ecotoxicology, both inside and outside of the regulatory context. Further
210 long-term improvements will require continued efforts by the statistics community in
211 ecotoxicology and beyond: better collaboration with regulatory authorities; the crucial
212 role of statistics in supporting the 3 Rs (replace, reduce, and refine the use of animals
213 for toxicity testing); robust and transparent computational solutions; and investment in
214 educational possibilities. Statistical practices have come a long way since OECD
215 document No. 54 was first released 20 years ago, and we are looking forward to the
216 exciting new chapters in the field of statistics for regulatory ecotoxicology.

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