

*Using Statistical Methods
for Water Quality Management
Issues, Problems and Solutions*

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*Using Statistical Methods
for Water Quality Management*

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*Using Statistical Methods
for Water Quality Management
Issues, Problems and Solutions*

Graham B. McBride

National Institute of Water & Atmospheric Research
Hamilton, New Zealand



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This book is dedicated to

Robyn

Stephen and Christine

Richard, Chris

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Preface

When managing water resources we often seek to separate pattern from randomness. For example, is the quality of water in a lake really deteriorating? Is that because the land-use patterns in its watershed are intensifying? Just how sure can we be of a conclusion about that? What are the health risks from swimming in this river or at that beach? Is there an important change in stream benthic communities downstream of mine waste discharges? These are issues that statistical methods can shed light on. I hope this book will help water resources managers and scientists to formulate, implement and interpret better and more appropriate methods to address such matters.

To do so, I have not sought to repeat material readily available in other texts.¹ Rather than being a “how to” compendium of the many procedures to be found in them, I have focused on material not generally available elsewhere, both in general (such as in Bayesian modes of inference) and in particular (such as percentile standards, water-related human health risk modeling, and MPN methodology for microbiological enumerations).

This book’s origins lie in an invitation from my colleague, Dr Bryan Manly, to present a paper to the Joint Statistical Meetings held in 2000 at Indianapolis. The title of the paper (as suggested by him) was “Statistical Methods Helping and Hindering Environmental Science and Management.” It drew on experience I had gained over some years in using statistical methods to address practical questions that commonly

¹Green (1979 [127]), Berthouex & Brown (1980 [23]), Snedecor & Cochran (1980 [291]), Sokal & Rohlf (1981 [292]), Iman & Conover (1983 [161]), Gilbert (1987 [118]), Krebs (1989 [176]), Ward *et al.* (1980 [338]), Gibbons (1994 [117]), Zar (1996 [349]), Manly (2001 [198]), Millard & Neerchal (2001 [226]), Helsel & Hirsch (2002 [147]), Townend (2002 [319]), Bolstad (2004 [28]).

arise in managing water resources. For example: Just what does the analyst mean when stating that a result is “statistically significant”? How should one devise and document percentile standards for effluent discharges, drinking water, or environmental standards? How many samples should be required in drinking water standards in order to provide a satisfactory level of assurance that the waters’ quality sufficiently protects public health? When and how can one take account of what we think we already know when framing compliance rules or when analyzing experiments?

The Indianapolis paper, with its somewhat provocative title, was subsequently published (McBride 2002 [206]), and it resulted in an invitation to write this book. In doing so, I have attempted to give “lines of approach” (rather than “answers”) to a range of issues that arise in water management (and in water science), such as those given above. I say “lines of approach” very deliberately because I have so often been asked, “What is the statistically correct way of analyzing these data?” In fact, there is seldom (if ever) *one* correct way, either for analysis of data or for design of sampling programs. We make judgments, not rules (Stewart-Oaten 1995 [300]). There may be a number of alternative approaches, each with its merits and drawbacks, and awareness of these can only be beneficial. There are of course many more inappropriate and incorrect lines of approach. So a function of this book is to clarify the appropriateness of such methods for various issues. By doing so, I hope to alert scientists and water resources managers to the wide variety of fruitful statistical methods that can be used. For example, it can come as a surprise that statistical methods do not just deal with numbers; incorporation of nominal data can greatly enhance their utility (e.g., using water color, wind direction, and octants of cloud cover in models of near-shore contamination by fecal bacteria).

In planning this book, I had first thought that it would be wise to introduce statistical concepts gradually, by way of a series of practical examples. As an instance of that, one could introduce Bayesian approaches to data analysis when considering the framing of percentile compliance rules. However, on reflection and discussion with colleagues, it became apparent that some general and pervasive issues needed first to be discussed in their own right. Accordingly, the book is in two parts. The first, *Issues*, consisting of six chapters, presents the material I consider to be important to be understood when contemplating using statistical methods for water resources management. It includes a number of practical examples. The five chapters in the second part, *Problems and Solutions*, draw on material given in the first part. They cover a range of topics that often arise but are not covered in most texts (formulating environmental standards, using percentile standards, microbiological water quality, “most probable numbers” (MPNs) and human health, and a catch-all chapter including material on trends, impacts, concordance and detection limits). Each chapter contains a number of set problems, for which full “answers” are given in the final (12th) chapter. Some of these problems are more difficult than others; some are discursive, in which case “answers” are longer than is usual in a statistics text.

So what are those “general and pervasive issues”? Some are details—but important details (especially concerning correct usage of standard errors and explaining “error bars”). Others are substantial. They have to do with identifying management’s information needs, the role of various forms of hypothesis testing, and the types of

intervals that may be used to account for uncertainty (confidence intervals, tolerance intervals, and credible intervals). To set the scene for these issues, Chapter 1 discusses the use of statistical populations and samples in water management. Chapter 2 presents basic concepts of probability and statistics, including different modes of statistical inference, especially frequentist versus Bayesian (throughout the book I present the case for using both). Chapter 3 discusses and contrasts confidence intervals, tolerance intervals and credible intervals. Chapters 4 and 5 then cover general and specific issues to do with statistical hypothesis testing—these tests are widely used in water management, and not always appropriately. These chapters address questions such as: When are “one-sided” tests appropriate? What can we actually infer from a test of a single “null” hypothesis? When should we use nonparametric approaches? Is there a corresponding Bayesian test? What is the role of equivalence testing? How do we give effect to the precautionary approach? We also introduce, in Chapter 5, the “detection probability” in the context of one-sided and two-sided tests. Detailed (though not rigorous) mathematical arguments and calculation routines are presented in Chapter 6—finer mathematical details are admirably covered in papers and other texts (e.g., Ferguson 1967 [90], Lehmann 1986 [183], Freund 1992 [101], Lee 1997 [179], Gelman *et al.* 2000 [110], Casella & Berger 2002 [45], Wellek 2003 [340]). This chapter can be skipped without losing too much of the main information presented.

A number of these issues are seldom addressed in much detail in applied statistical texts. For example, most workers actually interpret confidence intervals in a Bayesian manner, yet the Bayesian view of probability and associated modes of inference are seldom presented. It is important for environmental professionals to grasp such matters and their consequences. I became aware of this in 1989 when, while browsing in the basement of the Colorado State University library, I stumbled across the 1970 text by Morrison & Henkel *The Significance Test Controversy* ([232]). My response was one of surprise: “What controversy?” I read avidly, from a text arising not from environmental science, but from psychology (from where so many statistical advances have emanated). Many colleagues involved in water science and management (in my home country and abroad) have had the same response when confronted by the notion that statistical methodology, especially hypothesis testing, is accompanied (to this day) in the statistical literature by a degree of controversy. Some statisticians have even advocated that tests be abandoned altogether [e.g., in chapters of a recent book, edited by Harlow, Muliak, and Steiger (1997 [136]): *What if there were no significance tests?*]. But, in water management at least, we cannot abandon tests—comparisons, hence tests, are often required (between sites, between treatment levels, with standards, . . .).

Throughout I have made use of footnotes, giving detail that readers can pass over. More substantial technical details are contained in Appendices to some chapters. These too can be passed over without compromising understanding of the text. References are contained in a single list at the book’s end, along with an author index and a subject index.

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Part I

Issues

‘Far better an approximate answer to the right question, which is often vague, than the exact answer to the wrong question, which can always be made precise.’

—John Tukey (1962 [321])

‘It’s all in the mind, you know.’

—Wallace Greenslade and Spike Milligan, *The Goon Show* (1952–1960)

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1

Introduction

1.1 CONVENTIONS

Key statistical terms are written in **bold type** when they first appear. Italicized upper case Roman letters are used to denote data and the statistics derived from them, making it easier to distinguish them from text. For example X_i refers to the i th numerical **datum** in a **sample**, and \bar{X} is their **arithmetic mean**—usually just stated as the **mean**.¹ Data do not have to be numbers, they can be **categories** (such as water color), but in that case a mean is not defined. Italicized lowercase Greek letters are used for **population parameters** such as the **true mean**, which is **estimated** by \bar{X} (so the true mean is denoted by μ). Samples are **drawn** from populations, and the parameters are a succinct way to summarize the shape of the populations. Lowercase italicized Roman letters (such as x and y) refer to the population **random variables** from which samples may be drawn. Particular values of those random variables are written in uppercase letters (e.g., X and Y). X_p represents the 100 p th percentile of the x distribution (so $X_{0.95}$ is the 95th percentile—hereafter written as “95%ile”). Overbars always denote a mean; for example, the difference between the means of two **independent** samples drawn from x and y is written as $\bar{X} - \bar{Y}$, whereas if those samples were **paired** we would be contemplating the mean difference $\bar{X} - \bar{Y}$.

Like any discipline, statistics has its own language conventions. While most can be explained as we go along, those to do with the words **sample** and **error**, the symbol p , and even the word **statistics** need some explanation (though the context should make the meaning of such phrases clear).

¹Other types of means may be used, especially the **geometric mean**.

To an environmental professional a sample is a volume or mass of material taken from the environment—for example, a container of stream water for subsequent phosphorus analysis in a laboratory, or a reading of pH from a probe. But to a statistician, a sample is a collection of results or **observations**—for example, phosphorus concentrations in a set of water “samples.” So to a statistician one **sample** contains **many data** and the **sample size** is the number of data in the sample—not the volume of the container. Similarly, an environmental professional would regard an **error** as just that—a mistake. But to a statistician, **sampling error** is the natural **variability** inherent among data taken from a **population** and is therefore always present and to be accounted for. If there are errors in measurement (nothing in this world is perfect), these are referred to as **measurement error**.

Then we have a **standard error**. This is not a recommended way to make a mistake! It is another measure of variability—not of the data, but of a parameter estimated from the data, such as the mean.² Next, because of conventions in wide use, we are forced to use the symbol *p* ambiguously (it can mean **p-value** (an **exceedance probability**) or **proportion**. Uppercase *P* generally denotes a **cumulative probability**, expressed as a percentage (e.g., the *P*th percentile).

Finally, the term “**statistics**” itself can have two meanings. The singular **statistic** is a number that characterizes some feature of a set of data—for example, the sample arithmetic mean as a measure of the **central tendency** of data. The plural **statistics** can also refer to the methods that use such numbers to make estimates and comparisons. In this text we generally use the term **statistic** for the former meaning and **statistical methods** for the latter.

1.2 THE ESSENTIALS

The measurements we make are taken from samples that are only a tiny proportion of the water body that we want information about. So any **inferences** drawn from the data are **uncertain** to some degree. Statistical methods are a means of handling this uncertainty, both in designing sampling programs and in obtaining useful information from the results of sampling.

Classical (frequentist) statistical methods allow us to draw inferences from sample data with, under certain assumptions, a known degree of uncertainty. Furthermore, they can permit different data analysts to follow the same procedures, thus reaching the same conclusions from a given set of data.³ Statistical methods also provide a means of specifying how monitoring data should be analyzed before they are collected, thus ensuring that the information needs of management will be met from a planned monitoring program.

²If that parameter is the mean the standard error is in fact the **standard deviation of the mean** of the data, *not* of the data themselves. Importantly, it tends to get smaller as the number of data increases. More on that later (Section 2.13.1).

³This is not the case for the **Bayesian** methods that we shall meet later—although these can have some definite advantages, as we shall see.