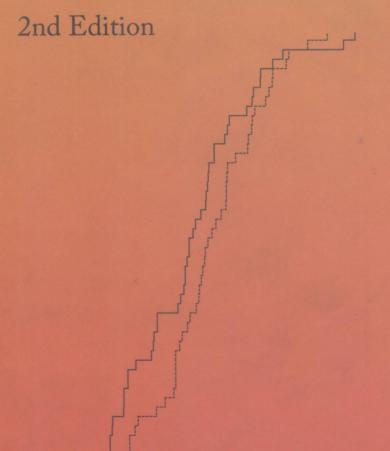
Frederick James

Statistical Methods in Experimental Physics



04-34 527 E-2

Frederick James

CERN, Switzerland

Statistical Methods in Experimental Physics



Published by

World Scientific Publishing Co. Pte. Ltd.

5 Toh Tuck Link, Singapore 596224

USA office: 27 Warren Street, Suite 401-402, Hackensack, NJ 07601 UK office: 57 Shelton Street, Covent Garden, London WC2H 9HE

British Library Cataloguing-in-Publication Data

A catalogue record for this book is available from the British Library.

The first edition:

Statistical Methods in Experimental Physics, by W. T. Eadie, D. Drijard, F. E. James, M. Roos and B. Sadoulet was published in 1971 by North-Holland Publishing Co., Amsterdam, New York and Oxford, and reprinted in 1977, 1982 and 1988

STATISTICAL METHODS IN EXPERIMENTAL PHYSICS (2nd Edition)

Copyright © 2006 by World Scientific Publishing Co. Pte. Ltd.

All rights reserved. This book, or parts thereof, may not be reproduced in any form or by any means, electronic or mechanical, including photocopying, recording or any information storage and retrieval system now known or to be invented, without written permission from the Publisher.

For photocopying of material in this volume, please pay a copying fee through the Copyright Clearance Center, Inc., 222 Rosewood Drive, Danvers, MA 01923, USA. In this case permission to photocopy is not required from the publisher.

ISBN 981-256-795-X ISBN 981-270-527-9 (pbk) Statistical Methods in Experimental Physics 2nd Edition

PREFACE TO THE SECOND EDITION

Thirty years after the first publication of the first edition, it was decided that the continued demand justified the production of a new edition. The five original authors agreed that the new edition should reproduce as much as possible the complete text of the first edition, of course corrected for mistakes and modified to take into account recent developments. The original publisher ceded his rights to World Scientific Publishing Co. who kindly agreed to set the entire first edition in LaTeX to initiate the preparation of the new edition. Unfortunately, only one of the five original authors was ready to do this preparation, but the four others agreed to yield their rights in the interest of allowing the renewed availability of the book.

The new edition has required a considerable amount of work. For example, there are over 1000 formulae in the book, about half of which required modification, mostly for improved and consistent notation, but also to correct all the mistakes that have been reported over the years. In addition, the author of the second edition has had thirty years' additional experience in statistical data analysis, which necessarily translates into a better understanding of some problems and requires more than cosmetic changes in a few chapters. The overall result is that most of the text comes from the first edition, but the modifications are sufficiently important that the author of the second edition must take all the responsibility for the final text.

For the reader, the most striking difference between the editions will certainly be the improved typesetting. All the other benefits of computer preparation should make this edition much easier to read and more reliable than its predecessor.

F. E. James July 2006, Geneva

PREFACE TO THE FIRST EDITION

This course in statistics, written by one statistician (W.T.E.) and four highenergy physicists, addresses itself to physicists (and experimenters in related sciences) in their task of extracting information from experimental data. Physicists often lack elementary knowledge of statistics, yet find themselves with problems requiring advanced methods, if adequate methods at all exist. To meet their needs, a sufficient course would have to be very long. Such courses do indeed exist [e.g. Kendall], only the physicists usually do not take the time to read them.

We attempt to give a course which is reasonably short, and yet sufficient for experimental physics. This obviously requires a compromise between theoretical rigour and amount of useful methods described.

Thus we are obliged to state many results without any rigorous proof (or with no proof at all); still we have the ambition to present more than just a cook-book of prescriptions and formulae. We omit the mention of many techniques which, in our judgement, seem to be of lesser importance to experimental physics.

On the other hand, we do introduce many theoretical concepts which may not seem immediately useful to the experimenter. This we think is necessary for two reasons. Firstly, the experimenter may need to know some theory or some "generalized methods" in order to design his own methods, experimental physics posing always novel questions. This is a justification for the stress on Information theory (Chapter 5), and for the attempt in Chapter 7 to define a "general" method of estimation. We hope that although the method the reader will arrive at may not be optimal, still it will be useful.

Secondly, the experimenter should be aware of the assumptions underlying a method, whether it be a standard method or his own. It is for this reason that we insist so much on the Central Limit Theorem, which is at the foundation of all "asymptotic" statistics (Chapters 3 and 7).

Quoting theorems, we also try to state their range of application, to avoid too careless use of some methods.

Among the underlying assumptions, especially important are the ones about the parent distributions of the data, since they will condition the results. In Chapter 4 we give a catalogue of useful ideal distributions; in real life they may have to be truncated (Sec. 4.3), experimental resolution may have to be folded in (Sec. 4.3), detection efficiency may have to be taken into account (Sec. 8.5). Moreover, the true distribution may not be known, in which case one is led to empirical distributions (Sec. 4.3), robust estimation (Sec. 8.7), and distribution-free tests (Chapter 11).

A very common tacit assumption in the everyday use of statistics is that the set of data is large enough for asymptotic conditions to apply. We try to distinguish clearly between asymptotic properties (usually simple whenever they are known) and finite sample properties (which are usually unknown). We also often give asymptotic expansions, in order to indicate how rapidly the asymptotic properties become true.

In general, we stress the various concepts of optimality. The justification for this is not only that this is the only way for a classical statistician to choose between different procedures, but also that experimental physicists handle ever increasing amounts of data, and therefore need increasingly optimal methods. However, there is an "optimal optimality", because the last bit of optimality can often be achieved only at great cost. This introduces the aspect of economy, which we try to stress on many occasions.

Facing the controversy between Bayesians and Anti-Bayesians ("classical" statisticians), we tend to favour the classical approach (because of professional bias), however keeping the reader partly informed about the Bayesian approach throughout. This attitude we justify as follows. In Chapter 6 we show how taking a decision from a limited amount of information leads to a fundamental indeterminacy: any decision depends on a priori assumptions. These assumptions being largely subjective by nature, we think that it is not the role of an experimenter to take decisions. His aim should be to summarize the results of his experiment for the rest of the physics community in such a way as to convey a maximum of information about the unknowns measured. In a certain sense this leaves to the general consensus the task to take decisions.

This is our motivation to the Information theory approach to Estimation. Logically, Test theory should then be Bayesian (since testing really is a decision). Our excuse for not being Bayesian in Test theory (Chapters 10, 11) is that physicists, as a matter of general practice, consider a confidence level as an objective measure of the "distance" of the experiment from the hypothesis tested.

A minor consequence of our professional bias is that in contrast to most (if not all) books on probability and statistics, we avoid using examples from gambling. Physicists often find it frustrating trying to convert such examples into physics; therefore, our examples are taken from physics (mainly high-energy physics). The theory is, of course, the same and gamblers should not be discouraged from converting our examples back into card games, dice, etc!

Let us finally point out that we do not discuss numerical optimization techniques, very important e.g. in the methods of maximum likelihood and least squares. The reasons are that there exist in our opinion excellent treatises of optimization, should the experimenter want to know the details of optimum-searching algorithms, and most physicists do have powerful optimization programs at their disposal (e.g. in the CERN Computer Program Library), which save them one more worry.

W. T. Eadie
D. Drijard
F. E. James
M. Roos
B. Sadoulet
December 1970, Geneva

CONTENTS

Preface to the Second Edition	V
Preface to the First Edition	vii
Chapter 1. Introduction	1
1.1. Outline	1
1.2. Language	2
1.3. Two Philosophies	3
1.4. Notation	4
Chapter 2. Basic Concepts in Probability	9
2.1. Definitions of Probability	9
2.1.1. Mathematical probability	10
2.1.2 Frequentist probability	10
2.1.3. Bayesian probability	11
2.2. Properties of Probability	12
2.2.1. Addition law for sets of elementary events	12
2.2.2. Conditional probability and independence	13
2.2.3. Example of the addition law: scanning efficiency	14
2.2.4. Bayes theorem for discrete events	15
2.2.5. Bayesian use of Bayes theorem	16
2.2.6. Random variable	17
2.3. Continuous Random Variables	18
2.3.1. Probability density function	19
2.3.2. Change of variable	20

	2.3.3. Cumulative, marginal and conditional distributions	21
	2.3.4. Bayes theorem for continuous variables	22
	$2.3.5.$ Bayesian use of Bayes theorem for continuous variables $\ . \ . \ .$	22
2.4.	Properties of Distributions	24
	2.4.1. Expectation, mean and variance	24
	2.4.2. Covariance and correlation	26
	2.4.3. Linear functions of random variables	28
	2.4.4. Ratio of random variables	30
	2.4.5. Approximate variance formulae	32
	2.4.6. Moments	33
2.5.	Characteristic Function	34
	2.5.1. Definition and properties	34
	2.5.2. Cumulants	37
	2.5.3. Probability generating function	38
	2.5.4. Sums of a random number of random variables	39
	2.5.5. Invariant measures	41
Cha	apter 3. Convergence and the Law of Large Numbers	43
	The Tchebycheff Theorem and Its Corollary	43
0.1.	3.1.1. Tchebycheff theorem	43
	3.1.2. Bienaymé–Tchebycheff inequality	44
3.2.	Convergence	45
0.2.	3.2.1. Convergence in distribution	45
	3.2.2. The Paul Levy theorem	46
	3.2.3. Convergence in probability	46
	3.2.4. Stronger types of convergence	47
3.3.	The Law of Large Numbers	47
	3.3.1. Monte Carlo integration	48
	3.3.2. The Central Limit theorem	49
	3.3.3. Example: Gaussian (Normal) random number generator	51
Cha	apter 4. Probability Distributions	53
4.1.	Discrete Distributions	53
	4.1.1. Binomial distribution	53
	4.1.2. Multinomial distribution	56
	4.1.3. Poisson distribution	57
	4.1.4. Compound Poisson distribution	60
	4.1.5. Geometric distribution	62

5.5. Example of Experimental Design	 109
Chapter 6. Decision Theory	111
6.1. Basic Concepts in Decision Theory	 112
6.1.1. Subjective probability, Bayesian approach	 112
6.1.2. Definitions and terminology	 113
6.2. Choice of Decision Rules	 114
6.2.1. Classical choice: pre-ordering rules	 114
6.2.2. Bayesian choice	115
6.2.3. Minimax decisions	 116
6.3. Decision-theoretic Approach to Classical Problems	 117
6.3.1. Point estimation	117
6.3.2. Interval estimation	 118
6.3.3. Tests of hypotheses	 118
6.4. Examples: Adjustment of an Apparatus	 121
6.4.1. Adjustment given an estimate of the apparatus perfo	121
6.4.2. Adjustment with estimation of the optimum adjustm	123
6.5. Conclusion: Indeterminacy in Classical and Bayesian Decisi	124
Chapter 7. Theory of Estimators	127
7.1. Basic Concepts in Estimation	127
7.1.1. Consistency and convergence	128
7.1.2. Bias and consistency	 129
7.2. Usual Methods of Constructing Consistent Estimators	 130
7.2.1. The moments method	 131
7.2.2. Implicitly defined estimators	132
7.2.3. The maximum likelihood method	135
7.2.4. Least squares methods	137
7.3. Asymptotic Distributions of Estimates	139
7.3.1. Asymptotic Normality	139
7.3.2. Asymptotic expansion of moments of estimates	 141
7.3.3. Asymptotic bias and variance of the usual estimators	 144
7.4. Information and the Precision of an Estimator	146
7.4.1. Lower bounds for the variance — Cramér-Rao inequa	147
7.4.2. Efficiency and minimum variance	149
7.4.3. Cramér–Rao inequality for several parameters	151
7.4.4. The Gauss–Markov theorem	 152
7.4.5. Asymptotic efficiency	153

7.5.	Bayesian Inference	154
	7.5.1. Choice of prior density	154
	7.5.2. Bayesian inference about the Poisson parameter	156
	7.5.3. Priors closed under sampling	157
	7.5.4. Bayesian inference about the mean, when the variance	
	is known	157
	7.5.5. Bayesian inference about the variance, when the mean	
	is known	159
	7.5.6. Bayesian inference about the mean and the variance	161
	7.5.7. Summary of Bayesian inference for Normal parameters \dots	162
Cha	apter 8. Point Estimation in Practice	163
8.1.	Choice of Estimator	163
	8.1.1. Desirable properties of estimators	164
	8.1.2. Compromise between statistical merits	165
	8.1.3. Cures to obtain simplicity	166
	8.1.4. Economic considerations	168
8.2.	The Method of Moments	170
	8.2.1. Orthogonal functions	170
	8.2.2. Comparison of likelihood and moments methods	172
8.3.	The Maximum Likelihood Method	173
	8.3.1. Summary of properties of maximum likelihood	173
	8.3.2. Example: determination of the lifetime of a particle in a	
	restricted volume	175
	8.3.3. Academic example of a poor maximum likelihood estimate .	177
	8.3.4. Constrained parameters in maximum likelihood	179
8.4.	The Least Squares Method (Chi-Square)	182
	8.4.1. The linear model	183
	8.4.2. The polynomial model	185
	8.4.3. Constrained parameters in the linear model	186
	8.4.4. Normally distributed data in nonlinear models	190
	8.4.5. Estimation from histograms; comparison of likelihood and	
	least squares methods	191
8.5.	Weights and Detection Efficiency	193
	8.5.1. Ideal method maximum likelihood	194
	8.5.2. Approximate method for handling weights	196
	8.5.3. Exclusion of events with large weight	199
	8.5.4. Least squares method	201

8.6.	Reduction of Bias	204
	8.6.1. Exact distribution of the estimate known	204
	8.6.2. Exact distribution of the estimate unknown	206
8.7.	Robust (Distribution-free) Estimation	207
	8.7.1. Robust estimation of the centre of a distribution	208
	8.7.2. Trimming and Winsorization	210
	8.7.3. Generalized p^{th} -power norms	211
	8.7.4. Estimates of location for asymmetric distributions	213
Cha	apter 9. Interval Estimation	215
9.1.	Normally distributed data	216
	9.1.1. Confidence intervals for the mean	216
	9.1.2. Confidence intervals for several parameters	218
	9.1.3. Interpretation of the covariance matrix	223
9.2.	The General Case in One Dimension	225
	9.2.1. Confidence intervals and belts	225
	9.2.2. Upper limits, lower limits and flip-flopping	227
	9.2.3. Unphysical values and empty intervals	229
	9.2.4. The unified approach	229
	9.2.5. Confidence intervals for discrete data	231
9.3.	Use of the Likelihood Function	233
	9.3.1. Parabolic log-likelihood function	233
	9.3.2. Non-parabolic log-likelihood functions	234
	9.3.3. Profile likelihood regions in many parameters	236
9.4.	Use of Asymptotic Approximations	238
	9.4.1. Asymptotic Normality of the maximum likelihood estimate .	238
	9.4.2. Asymptotic Normality of $\partial \ln L/\partial \theta$	238
	9.4.3. $\partial L/\partial \theta$ confidence regions in many parameters	240
	9.4.4. Finite sample behaviour of three general methods of interval	
	estimation	240
9.5.	Summary: Confidence Intervals and the Ensemble	246
	The Bayesian Approach	248
	9.6.1. Confidence intervals and credible intervals	249
	9.6.2. Summary: Bayesian or frequentist intervals?	250
Cha	apter 10. Test of Hypotheses	253
10.1	. Formulation of a Test	254
	10.1.1. Basic concepts in testing	

	10.1.2. Example: Separation of two classes of events	255
10.2.		257
		257
		259
		260
		261
10.3.		263
		63
		64
10.4.		266
	10.4.1. Existence of a uniformly most powerful test for the	
		267
		268
		269
10.5.		270
	Description —	270
	10.5.2. Asymptotic distribution for continuous families of	
		71
		73
	10.5.4. Examples	74
	10.5.5. Small sample behaviour	79
	10.5.6. Example of separate families of hypotheses	82
	10.5.7. General methods for testing separate families	85
10.6.		87
		87
		92
	10.6.3. Sequential probability ratio test for a continuous family	
	of hypotheses	97
10.7.	Summary of Optimal Tests	98
Chaj	oter 11. Goodness-of-Fit Tests	99
11.1	GOF Testing: From the Test Statistic to the P-value 2	99
		01
11.2.		02
		03
		03
11.3		08
11.0.	11.3.1. Runs test	
	11.0.1. 1001ID 0000	UU

$xviii \quad \textit{Statistical Methods in Experimental Physics - 2nd Edn}.$

	11.3.2. Empty cell test, order statistics						309
	11.3.3. Neyman–Barton smooth test						
11.4.	Tests Free of Binning						
	11.4.1. Smirnov-Cramér-von Mises test						
	11.4.2. Kolmogorov test						
	11.4.3. More refined tests based on the EDF .						
	11.4.4. Use of the likelihood function						317
11.5.	Applications						
	11.5.1. Observation of a fine structure						
	11.5.2. Combining independent estimates						323
	11.5.3. Comparing distributions						327
11.6.	Combining Independent Tests						330
	11.6.1. Independence of tests						330
	11.6.2. Significance level of the combined test						
Refer	ences						335
Subje	ect Index						341

Chapter 1

INTRODUCTION

1.1. Outline

The subject of the following ten chapters can be divided into two main parts:

• Theory of Probability: Chapters 2-4

• Statistics: Chapters 5–11

The theory of probability is needed only to provide the necessary tools for statistics, which forms the main body of the course.

Chapters 5 and 6 define two general approaches to the choice of estimators: the *information* approach and the *decision theory* approach. The former consists essentially in maximizing the amount of information in the estimate, whereas the latter is based on minimizing the loss involved in making the wrong decision about the parameter value. In the limit of large data samples, the two approaches are equivalent, but where they differ we will try to point out the distinction.

Estimation of parameters is divided into three chapters, 7 and 8 dealing with point estimation, theory and practice, and 9 dealing with interval estimation. Tests of hypotheses are divided into general testing, Chapter 10, and goodness-of-fit tests, Chapter 11.

Our reference policy is as follows. We quote literature when we have omitted the proof of an important result, or when we want to give hints for further