

## CHAPTER 1. INTRODUCTION

Among all prescriptions for statistical behavior, the Likelihood Principle (LP) stands out as the simplest and yet most farreaching. It essentially states that all evidence, which is obtained from an experiment, about an unknown quantity  $\theta$ , is contained in the likelihood function of  $\theta$  for the given data. The implications of this are profound, since most non-Bayesian approaches to statistics and indeed most standard statistical measures of evidence (such as coverage probability, error probabilities, significance level, frequentist risk, etc.) are then contraindicated.

The LP was always implicit in the Bayesian approach to statistics, but its development as a separate statistical principle was due in large part to ideas of R. A. Fisher and G. Barnard (see Section 3.2 for references). It received major notice when Birnbaum (1962a) showed it to be a consequence of the more commonly trusted Sufficiency Principle (that a sufficient statistic summarizes the evidence from an experiment) and Conditionality Principle (that experiments not actually performed should be irrelevant to conclusions). Since then the LP has been extensively debated by statisticians interested in foundations, but has been ignored by most statisticians. There are perhaps several reasons for this. First, the consequences of the LP seem so absurd to many classical statisticians that they feel it a waste of time to even study the issue. Second, a cursory investigation of the LP reveals certain oft-stated objections, foremost of which is the apparent dependence of the principle on assuming exact knowledge of the (parametric) model for the experiment (so that an exact likelihood function exists). Since the model is rarely true, (hasty)