ncepts

Again, al test hen it lly be

nance about

ound

ental ction tions

ng to nsive

other

data-

ure δ to be that

show Juen-

ntical

t size ason then and

and sh in ondi- $= x_2$

Careful consideration of this example will make the difference between the conditional and frequentist viewpoints clear. The overall performance of δ , in any type of repeated use, would indeed be 75%, but this arises because half the time the *actual* performance will be 100% and half the time the *actual* performance will be 50%. And, for any given application, one *knows* whether one is in the 100% or 50% case. It clearly would make little sense to conduct an experiment, use $\delta(x)$, and actually report 75% as the measure of accuracy, yet the frequentist viewpoint suggests doing so. Here is another standard example.

Example 13. Suppose X is 1, 2, or 3 and θ is 0 or 1, with X having the following probability density in each case:

	x		
-	1	2	3
f(x 0)	0.005	0.005	0.99
f(x 1)	0.0051	0.9849	0.01

The classical most powerful test of H_0 : $\theta=0$ versus H_1 : $\theta=1$, at level $\alpha=0.01$, concludes H_1 when X=1 or 2, and this test also has a Type II error probability of 0.01. Hence, a standard frequentist, upon observing x=1, would report that the decision is H_1 and that the test had error probabilities of 0.01. This certainly gives the impression that one can place a great deal of confidence in the conclusion, but is this the case? Conditional reasoning shows that the answer is sometimes no! When x=1 is observed, the likelihood ratio between $\theta=0$ and $\theta=1$ is (0.005)/(0.0051) which is very close to one. To a conditionalist (and to most other statisticians also), a likelihood ratio near one means that the data does very little to distinguish between $\theta=0$ and $\theta=1$. Hence the conditional "confidence" in the decision to conclude H_1 , when x=1 is observed, would be only about 50%. (Of course x=1 is unlikely to occur, but, when it does, should not a sensible answer be given?)

The next example is included for historical reasons, and also because it turns out to be a key example for development of the important Likelihood Principle in the next subsection. This example is a variant of the famous Cox (1958) conditioning example.

Example 14. Suppose a substance to be analyzed can be sent either to a laboratory in New York or a laboratory in California. The two labs seem equally good, so a fair coin is flipped to choose between them, with "heads" denoting that the lab in New York will be chosen. The coin is flipped and