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## Abstract

Common approximations for the minimum description length  $P$  (

Local Asymptotics and  
the Minimum Description Length

*Key Phrases:* *BIC*, hypothesis test, model selection, two-part code, universal code.



for  $\theta$  in a compact subset of  $\mathbf{R}^k$ , with the exception of a small set of vanishing measure. In the one-dimensional case, we show that the cost of coding a nonzero parameter from the exceptional set near zero is considerably less than (12) would suggest. Thus, adding such a parameter is "easier" than the approximation (3) would suggest. The disagreement follows from a lack of uniform convergence in the asymptotics which produce (3).

The example in the next section gives the explicit correspondence between the de-

Figure 1: TPe estimator which UiniUizes tPe description lengtP (1) is shown as a function Wf tPe Uean Wf tPe input data, on a standardQzed scale. This estiUator Wfiers no shrinSage attPe origin.



For this univariate probleU, tPe leadQng terU in the criterion siUpTifles since (1) = 1 and by using (4) we have

$$Y_{i, \cdot}) = \log$$

The MDL estimator  $\hat{\lambda}$  defined in equation (15) above is shrunk to zero for  $P_{\mathcal{V}} \bar{Y} < 2.4$ . In contrast, the BIC criterion produces an estimate of zero for data with mean satisfying  $P_{\mathcal{V}} \bar{Y} < P_{\mathcal{V}}$ . Our arguments require a very close accounting of the message length obtained in a two-part code for the data, and we turn to these issues.

Although (4) implies that coding the data using  $\lambda =$

Thus, one can obtain a shorter message by rounding to a more coarse grid. Such details have been discussed elsewhere (e.g., Wallace and Freeman 1987), and for our purposes any such rounding provides the shortest code length. The need to encode the parameter does not impede simple rounding

$\bar{Y}$

to minimize the excess length. Dependence on the universal code being used, such as round-off occasionally shifts the estimator because of changes in the code for

in Table 1; all three are optimal in the sense of Elias (1975) who proposed and named

adding one for tPe sign bit)

$$L'_p(j, j-1) = 2 + (1 + b \log(j, j-1)) + (b \log^{(2)}(j, j-1) + c \log^{(3)}(j, j-1) + \dots + \log^{(k)}(j, j-1)) \log^{(k)}(j, j-1), \quad (12)$$

where terms are included in tPe sum so Tong as tPe-fWld iterated Tog (e.g.,  $\log^{(2)}(j, j-1) = \log \log(j, j-1)$ ) is at least one. *TPUs* resembles a discretized version of  $\log^{(2)}$ . However,

$L'_p(j)$  is not a uniform approximation because it jumps by several bits at integers of tPe form  $j = 2^k - 1$ , with tPe jump equal to tPe number of Togarithmic sums in

(12). Table 1 shows tPe jumps (sin)-333 (comparing)  $\lfloor \log_2(j) \rfloor$  /  $\lfloor \log_2(j-1) \rfloor$  = 20.553 0 TD (L)  $\lfloor \log_2(j) \rfloor$  /  $\lfloor \log_2(j-1) \rfloor$  = 7.97 0



Table 1: *Examples of three optimal universal codes for nonnegative integers* Spaces are for the reader and are not needed in the actual codes. A sign bit would be appended for  $j \neq 0$ . The doubly compound and penultimate codes are from Elias (1975); the third is an arithmetic coder for the probabilities  $Q(jj) = Q(j) + Q(j|j)$

### **3 Model Selection via**



change might alter the critical value in the decision rule (16), most likely increasing the threshold slightly.

Universal Length functions like  $L_a$  are therefore (nearly) manipulable

$$\log R < L$$

$$\log R < \log R + r(R) \quad (17)$$

where  $r(R) = \log R \rightarrow 0$  as  $R \rightarrow \infty$ . In this sense, the lengths of an optimal universal code are logarithmic. This property, together with the ease of manipulating  $\log$  rather than  $\log$ , has led to the most common approximation to the code length  $L_{n,k}$ . It is this approximation, rather than intrinsic property of MDL principle itself, that leads to (logarithmic) universal codes.

$$L_{n,k} = \log n + \log k + 1 \quad (18)$$

If we fix  $n$  and let  $k$  vary, then (18) is a linear function of  $\log k$ . This fixed code length for representing a parameter rather than the varying length implied by  $L$

In particular, one can show that the code length obtained by this representation (over

provides a Tower asymptotic bound for the excess length. For example in the mean coding problem we have been discussing, let  $T$  denote a compact subset of  $\mathbb{R}$  and let  $A_n$  denote a set whose measure tends to zero as  $n \rightarrow \infty$ . Then for all  $\epsilon > 0$  and any  $\delta > 0$  there exists  $n$  such that

$$E_n \leq \epsilon + \delta n; \quad (20)$$

where  $E_n$  is the expected excess length. The perspective of using asymptotics on a fixed standard deviation scale (so that

$n$  is large) , its size remains fixed on a standard scale. The perspective of using asymptotics on a fixed standard deviation scale (so that

is about 2.4, the description length for the model is shorter when this parameter is included than when it is forced to zero. The use of MDL for testing a single parameter thus leads to a decision rule that resembles a traditional hypothesis test: there is a fixed threshold lying about 2 standard errors from the origin rather than a threshold which grows with the logarithm of the sample size.

This discrepancy from a logarithmic penalty arises because standard approximations for MDL

Rissanen, J. (1983). A universal prior for integers and estimation by minimum description length. *Annals of Statistics*, **11**

Figure 2: *The penultimate codebook.* Quadratics indicate the excess Uessage length above  $\log \frac{1}{P(Y)}$  for estimating  $\bar{n}$  when the parameter is encoded using the penultimate code.





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*when the parameter is encoded using the arithmetic  
code for*

