#### CORRECTION NOTES

#### CORRECTION TO

## "THE INDIVIDUAL ERGODIC THEOREM OF INFORMATION THEORY"

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Mr. James Abbott has pointed out that the argument on Page 811 of the above-cited work, *Ann. Math. Stat.*, Vol. 28, No. 3 (1957), pp. 809–811, is incorrect. The results of the paper are valid, however, and Page 811 may be replaced by the following discussion.

Note that

$$E\left(\frac{p(x_{-k}, \dots, x_{-1})}{p(x_{-k}, \dots, x_0)} \middle| x_0, \dots, x_{-k+1}\right) \leq \frac{p(x_{-k+1}, \dots, x_{-1})}{p(x_{-k+1}, \dots, x_0)}$$

with probability one. By the concavity of log, it follows that the  $g_k$  sequence,

$$g_k = -\log_2\left(\frac{p(x_{-k}, \cdots, x_0)}{p(x_{-k}, \cdots, x_{-1})}\right)$$

satisfies

$$E(g_k \mid x_0, \dots, x_{-k+1}) \leq g_{k-1}.$$

Since  $g_k \ge 0$ , and  $Eg_0 < \infty$ , the  $g_k$  sequence forms a non-negative lower semi-martingale and hence converges a.s. Actually, the convergence of the  $g_k$  sequence has been previously established by McMillan in [2].

Now consider  $P(\sup_{k \le n} g_k > \lambda)$ , and define the disjoint sets

$$E_j = \{g_j > \lambda, \sup_{k < j} g_k \leq \lambda\},\$$

whence  $P(\sup_{k \leq n} g_k > \lambda) = \sum_{j=1}^n P(E_j)$ . Let  $Z_i$  be the cylinder sets  $\{x_0 = a_i\}$  and  $f_k^{(i)}$  the functions  $-\log_2 P(x_0 = a_i | x_{-1}, \dots, x_{-k})$ . If  $\sum_A f(x_0, x_{-1}, \dots)$  indicates the sum of  $f(x_0, x_{-1}, \dots)$  over all sequences  $(x_0, x_{-1}, \dots)$   $\varepsilon A$ , then

$$P(E_j) = \sum_{E_j} p(x_{-j}, \dots, x_0) = \sum_{i} \sum_{E_j \cap Z_i} \frac{p(x_{-j}, \dots, x_0)}{p(x_{-j}, \dots, x_{-1})} p(x_{-j}, \dots, x_{-1}).$$

But on  $E_i$  we have the inequality

$$\frac{p(x_{-j}, \cdots, x_0)}{p(x_{-j}, \cdots, x_{-1})} = 2^{-g_j} \le 2^{-\lambda},$$

leading to

$$P(E_j) \leq 2^{-\lambda} \sum_{i} \sum_{E_j \cap Z_i} p(x_{-j}, \dots, x_{-1}) = 2^{-\lambda} \sum_{i} P(f_j^{(i)} > \lambda, \sup_{k < j} f_k^{(i)} \leq \lambda).$$

Finally, then,

$$P\left(\sup_{k\leq n}g_k>\lambda\right)\leq 2^{-\lambda}\sum_{i}P\left(\sup_{k\leq n}f_k^{(i)}>\lambda\right)\leq s\cdot 2^{-\lambda},$$

where s is the number of values that the process ranges over. This last inequality gives  $P(\sup_k g_k > \lambda) \leq s \cdot 2^{-\lambda}$ , which quickly leads to  $E(\sup_k g_k) < \infty$ .

### CORRECTION TO

# "BOUNDS ON NORMAL APPROXIMATIONS TO STUDENT'S AND THE CHI-SQUARE DISTRIBUTIONS"

By DAVID L. WALLACE

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The following correction should be made on p. 1127 of the above-titled article (Ann. Math. Stat., Vol. 30 (1959), pp. 1121-1130): In the conclusion of Corollary 2 to Theorem 4.2, the exponent of n should be  $-\frac{1}{2}$  and not  $\frac{1}{2}$ .