

## A PROBABILISTIC REMARK ON ALGEBRAIC PROGRAM TESTING

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Until very recently, research in software reliability has divided quite neatly into two — usually warring — camps: methodologies with a mathematical basis and methodologies without such a basis. In the former view, “reliability” is identified with “correctness” and the principle tool has been formal and informal verification [1]. In the latter view, “reliability” is taken to mean the ability to meet overall functional goals to within some predefined limits [2,3]. We have argued in [4] that the latter view holds a great deal of promise for further development at both the practical and analytical levels. Howden [5] proposes a first step in this direction by describing a method for “testing” a certain restricted class of programs whose behavior can — in a sense Howden makes precise — be *algebraicized*. In this way, “testing” a program is reduced to an equivalence test, the major components of which become

- (i) a combinatorial identification of “equivalent” structures;
- (ii) an algebraic test

$$f_1 \equiv f_2,$$

where  $f_i$ ,  $i = 1, 2$  is a multivariable polynomial (multinomial) of degree specified by the program being considered.

In arriving at a method for exact solution of (ii), Howden derives an algorithm, that requires evaluation of multinomials  $f(x_1, \dots, x_m)$  of maximal degree  $d$  at  $O[(d+1)^m]$  points. For large values of  $m$  (a typical

case for realistic examples), this method becomes prohibitively expensive.

Since, however, a test for reliability rather than a certification of correctness is desired, a natural question is whether or not Howden's method can be improved by settling for less than an exact solution to (ii).

We are inspired by Rabin [6] and, less directly, by the many successes of Erdős and Spencer [7] to attempt a *probabilistic* solution to (ii). Using these methods, we show that (ii) can be tested with probability of error  $\epsilon^*$  with only  $O(g(\epsilon))$  evaluations of multinomials, where  $g$  is a slowly growing function of only  $\epsilon$ . In particular, 30 or so evaluations should give sufficiently small probability of error for most practical situations. The remainder of this note is devoted to proving this result.

Let us denote by  $P_{\neq 0}(m, d)$  the class of multinomials

$$f(x_1, \dots, x_m) \neq 0$$

(over some arbitrary but fixed integral domain) whose degree does not exceed  $d > 0$ . We define

$$P(m, d, r) = \min \text{Prob} \{1 \leq x_1 \leq r, f(x_1, \dots, x_m) \neq 0\}$$

$$f \in P_{\neq 0}(m, d).$$

\* See Rabin's account of algorithms that may err with fixed probability [6].

We think of  $P(m, d, r)$  as the minimal relative frequency with which witnesses to the non-nullity of a multinomial of the appropriate kind can occur in the chosen interval. We will derive a lower bound  $p$  for  $P(m, d, r)$ . Then  $(1 - p)$  is an upper bound on the error in selecting a random point from the  $m$ -cube. We then iterate the procedure by  $t$  independent random selections to obtain a small probability of error  $(1 - p)^t$ . Notice, in particular, that since a polynomial of degree  $d$  has at most  $d$  roots (ignoring multiplicity), the largest probability of finding a root must be at least the probability of finding a root by randomly sampling in the interval  $1 \leq x_1 \leq r$ ; thus

$$P(1, d, r) \geq 1 - d/r.$$

Now, consider some

$$f(x_1, \dots, x_m, y) \neq 0$$

of degree at most  $d$ . But there are then multinomials  $\{g_i\}_{i \leq d}$ , not all  $\neq 0$ , such that

$$f(x_1, \dots, x_m, y) = \sum_{i=0}^d g_i(x_1, \dots, x_m) y^i.$$

Let us suppose that  $g_k \in P_{\neq 0}(m, d)$ . Thus

$$\begin{aligned} \text{Prob}\{1 \leq x_i \leq r, f(x_1, \dots, x_m, y) \neq 0\} \\ \geq \text{Prob}\{g_k(x_1, \dots, x_m) \neq 0 \text{ and } y \text{ is not a root}\} \\ \geq P(m, d, r)(1 - d/r). \end{aligned}$$

Continuing inductively, we obtain a lower bound in  $P(m, d, r)$  as follows:

$$P(m, d, r) \geq (1 - d/r)^m. \quad (1)$$

But

$$\begin{aligned} \lim_{m \rightarrow \infty} (1 - d/r)^m &= \lim_{m \rightarrow \infty} \left[1 + \frac{1}{m} \left(\frac{-dm}{r}\right)\right]^m \\ &= \exp(-dm/r) \end{aligned} \quad (2)$$

Combining (1) and (2), we have for large  $m$ ,  $r = dm$ ,

$$P(m, d, dm) \geq e^{-1}.$$

Thus, with  $t$  evaluations of  $f$  for independent choices of points from the  $m$ -cube with sides  $r = dm$ , the probability of missing a witness to the non-nullity of  $f(x_1, \dots, x_m)$  is at most

$$(1 - e^{-1})^t.$$

Table 1 shows the probable error in testing  $f \equiv 0$  by  $t$  evaluations of  $f$  at randomly chosen points for some typical values of  $d, m, r, t$ . Notice that for  $dm = r$ ,  $t = 30$ , this is already  $< 10^{-5}$ .

## References

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Table 1  
Probable error in testing  $f(x_1, \dots, x_m) \equiv 0$  (degree  $\leq d$ ) by  $t$  random evaluations in  $\{1, \dots, r\}$

$[1 - P(m, d, r)]^t$						
$dm$	$r$	$t = 10$	$t = 20$	$t = 30$	$t = 50$	$t = 100$
10	10	$10 \times 10^{-3}$	$106 \times 10^{-6}$	$1 \times 10^{-6}$	$109 \times 10^{-12}$	$12 \times 10^{-21}$
20	10	$233 \times 10^{-3}$	$54 \times 10^{-3}$	$13 \times 10^{-3}$	$695 \times 10^{-6}$	$483 \times 10^{-9}$
50	10	$935 \times 10^{-3}$	$873 \times 10^{-3}$	$816 \times 10^{-3}$	$713 \times 10^{-3}$	$509 \times 10^{-3}$
$10^2$	10	$\sim 1$	$\sim 1$	$\sim 1$	$\sim 1$	$\sim 1$
10	$10^2$	$61 \times 10^{-12}$	$< 10^{-20}$	$< 10^{-20}$	$< 10^{-20}$	$\sim 0$
20	$10^2$	$38 \times 10^{-9}$	$1 \times 10^{-15}$	$< 10^{-20}$	$< 10^{-20}$	$\sim 0$
50	$10^2$	$88 \times 10^{-6}$	$8 \times 10^{-9}$	$704 \times 10^{-15}$	$< 10^{-20}$	$< 10^{-20}$
$10^3$	$10^2$	$\sim 1$	$\sim 1$	$\sim 1$	$\sim 1$	$\sim 1$
10	$10^3$	$< 10^{-20}$	$< 10^{-20}$	$< 10^{-20}$	$\sim 0$	$\sim 0$
20	$10^3$	$9 \times 10^{-18}$	$< 10^{-20}$	$< 10^{-20}$	$\sim 0$	$\sim 0$
50	$10^3$	$76 \times 10^{-15}$	$< 10^{-20}$	$< 10^{-20}$	$\sim 0$	$\sim 0$

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