

Some Performance Tests of "quicksort" and Descendants

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Detailed performance evaluations are presented for six ACM algorithms: *quicksort* (No. 64), *Shellsort* (No. 201), *stringsort* (No. 207), "*TREESORT3*" (No. 245), *quickersort* (No. 271), and *qsort* (No. 402). Algorithms 271 and 402 are refinements of algorithm 64, and all three are discussed in some detail. The evidence given here demonstrates that *qsort* (No. 402) requires many more comparisons than its author claims. Of all these algorithms, *quickersort* requires the fewest comparisons to sort random arrays.

Key Words and Phrases: sorting, in-place sorting, sorting efficiency, sorting performance tests, quicksort, quickersort, qsort, Shellsort, stringsort, TREESORT3, utility sort algorithm, general-purpose sort algorithm, sorting algorithm documentation

CR Categories: 4.49, 5.31

1. Introduction

Recently, I had to obtain an in-place sort routine requiring the fewest possible comparisons. Thinking that the latest sort procedure among the ACM algorithms would likely be best, I prepared a Fortran version of van Emden's *qsort*, which appeared satisfactory. But thorough testing showed that it does not perform as claimed. Its grandfather, Hoare's *quicksort*, and especially its father, Scowen's *quickersort*, required fewer comparisons. One purpose of this report is to compare the performance of these three algorithms in some detail.

A close look at the published data pertaining to performance tests of these and other ACM algorithms reveals that these data are incomplete. The actual numbers of comparisons required to sort particular arrays are never stated; sometimes the routine actually tested is only vaguely described, and its listing not given. Consequently, reliable conclusions about the relative merits of the algorithms cannot be reached. At least a listing of the actual code tested should be provided, the test arrays should be precisely specified, and the number of comparisons should be explicitly stated; such information is more useful than the timing figures usually reported. The other purpose of this report is to provide this kind of information for the following six ACM sort algorithms:

No. 64	<i>quicksort</i>	by C.A.R. Hoare	[1]
No. 201	<i>Shellsort</i>	by J. Boothroyd	[2]
No. 207	<i>stringsort</i>	by J. Boothroyd	[3]
No. 245	<i>TREESORT3</i>	by R.W. Floyd	[4]
No. 271	<i>quickersort</i>	by R.S. Scowen	[5]
No. 402	<i>qsort</i>	by M.H. van Emden	[6]

The report is arranged in the following sections: Section 2 briefly describes *quicksort*, *quickersort*, and *qsort*; Section 3 outlines the performance tests; and Section 4 discusses the results.

2. quicksort, quickersort, and qsort

Hoare [7] has provided an excellent account of how *quicksort* works. A brief summary may suffice as background for subsequent discussions. Given an array $a(j)$, $i \leq j \leq k$, to be sorted, *quicksort* first chooses an element Y of this array at random. It then rearranges the elements until the array has been partitioned into three parts: (a) a middle subarray, consisting of Y ; (b) a low subarray, none of whose elements is larger than Y ; and (c) a high subarray, none of whose elements is less than Y . While Y is now in its proper sorted position and does not require further processing, the low and the high subarrays are not necessarily yet in proper order. *quicksort* continues to partition the subarrays until low or high subarrays of length one are obtained; such subarrays need not be sorted further. Though seemingly complex, this procedure can be easily coded and is very efficient. Its correctness has been proved by Foley and

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Hoare [8]. The number of comparisons required to sort an array of length n is proportional to $n \log_2(n)$.

After an array has been partitioned and either the low or the high subarray chosen for immediate further processing, the index pair (i, k) , specifying the subarray whose processing is deferred, is saved in a pushdown stack. The process can be coded recursively to use a stack administered by a compiler/operating system, as Hoare did in *quicksort*, or a stack can be simulated explicitly, as Scowen did in *quickersort* and as was done in the versions shown here.

Hillmore [9] improved on Hoare's published version of *quicksort* by pointing out that it is better not to split subarrays of length two, but just to sort them instead. Sorting such arrays requires only one comparison, whereas splitting may require more.

In *quickersort* [5], Scowen improved on *quicksort* by selecting the middle element of the array as Y , rather than choosing an element at random. He pointed out [5, p. 670] that "the best possible value of Y would be one which splits the segment into two halves of equal size, thus if the array (segment) is roughly sorted, the middle element is an excellent choice. If the array is completely random, the middle element is as good as any other. If however, the array $a[1:j]$ is such that the parts $a[1:j \div 2]$ and $a[j \div 2 + 1:j]$ are both sorted the middle element could be very bad." In such circumstances, Y should be chosen at random, as in *quicksort*.

In addition to choosing Y differently and using Hillmore's modification, Scowen's *quickersort* contains a third improvement over *quicksort*, concerning the length of the stack. The number of index pairs to be saved is minimal ($\leq \log_2(n)$) if those that define the larger of the two subarrays are stored, while immediate processing continues with the smaller.

While picking the middle element is better than choosing Y at random when the array is already almost in sort order (a not unusual situation), *quicksort* and *quickersort* are expected to perform identically with random arrays. Hoare [7] predicted that the expected average number of comparisons required for *quicksort* to sort n unequal randomly ordered items is $\sim 2n \log_e(n) = 1.386 n \log_2(n)$, and stated that the theoretical minimum number of comparisons in this situation, which will be achieved if the splitting always yields subarrays of equal length, is $\log_2(n!) \sim n \log_2(n)$. This is discussed further in the Appendix.

After noting that the predicted average expected number of comparisons exceeds the predicted theoretical minimum by a factor of ~ 1.4 , Hoare suggested that the number of comparisons could be reduced by choosing Y on the basis of a random sample of the array to be partitioned. Van Emde [10] shows theoretically how the average number of comparisons for the entire array decreases as the size of the random sample increases. This approach has also been investigated by Frazer and McKellar [11].

In *qsort* [6], however, van Emde introduces another method for choosing Y . Briefly, instead of a single

element, he chooses two, $X = a(i)$ and $Z = a(j)$, $X \leq Z$, defining a bounding interval. The current values of X and Z are continually updated to allow for a proper partition, which is complete when the elements have been rearranged such that $X = a(i)$ and $Z = a(i + 1)$. Now there are a low subarray, none of whose elements is greater than X , a high subarray, none of whose elements is less than Z , and a middle subarray, which now contains two elements. The low and high segments obtained by *qsort* are expected to contain more nearly the same number of elements than they would with *quicksort* or *quickersort*. The average number of comparisons for random arrays is therefore expected to be closer to the theoretical minimum. Van Emde predicts that this number, for large n , is $1.140 n \log_2(n)$, less than the predicted value for *quicksort*, and reports timing data intended to verify this prediction.

3. Performance Tests

Sorting algorithms were tested with five different types of array: (a) sorted arrays: $a(i) = i$, $1 \leq i \leq n$; (b) arrays sorted in reverse order: $a(i) = n + 1 - i$, $1 \leq i \leq n$; and (c) random arrays: $a(i) = r(i)$, $1 \leq i \leq n$, where each $r(i)$ is one of a sequence of normally distributed random numbers in the interval $(0, 1)$, generated by a procedure written by Murphy [12]; (d) arrays almost in sort, generated as follows: first set $a_m(i) = i$, $1 \leq i \leq n$, then pick m elements at random and set each equal to a different random number, i.e. $a_m[n \cdot r(k)] = n \cdot r'(k)$, $1 \leq k \leq m$, $m < n$, $n \cdot r(k) \geq 1$, where $r(k)$ and $r'(k)$ are random numbers as in (c) above; (e) arrays of equal-length sorted blocks, generated as follows: first set $a_m(i) = r(i)$, $1 \leq i \leq n$, where $r(i)$ are random numbers as in (c) above, then sort all m subarrays of length n/m of $a_m(i)$ in place, such that $a_m(i)$ contains m adjacent sorted sequences.

After every test, the array was checked to verify that it had been sorted properly. All tests involving random numbers were repeated R times, the measurements were averaged, and the standard deviations of the averages were computed.

The following parameters were measured: (a) N_c , the number of comparisons; (b) N_f , the number of fetches from the array; (c) N_s , the number of stores into the array; and (d) N_r , the number of partitions required (or the number of segments of length greater than one) (for *quicksort*, *quickersort*, and *qsort* only). The results are shown in Tables I–VII and are discussed briefly in Section 4. All counts in Tables I–VI have been divided by $n \log_2(n)$ (i.e. a reported count K_r was computed from the observed count K_0 according to $K_r = (0.69315 K_0) / [n \log_e(n)]$). A value followed by a second one in parentheses is the average of the values from R repeat runs (each repeat uses different random numbers); the value in parentheses is the standard deviation of the average (expressed in percent, truncated).

Table I.

QUICKSORT : SORTING PERFORMANCE TEST RESULTS

LINE	N	M	R	COMPARES	FETCHES	STORES	PARTITIONS
SORTED ARRAYS							
1	32		96	1,231(9)	1,363(8)	0,000(0)	.131(4)
2	128		48	1,249(8)	1,344(7)	0,000(0)	.095(2)
3	512		24	1,267(5)	1,341(4)	0,000(0)	.074(0)
4	2048		12	1,277(4)	1,338(4)	0,000(0)	.061(0)
5	8192		6	1,311(5)	1,362(5)	0,000(0)	.051(0)
6	32768		3	1,321(2)	1,366(2)	0,000(0)	.044(0)
ARRAYS SORTED IN REVERSE ORDER							
7	32		96	1,194(7)	1,525(6)	.200(0)	.131(4)
8	128		48	1,239(6)	1,478(5)	.143(0)	.095(2)
9	512		24	1,281(6)	1,467(5)	.111(0)	.074(1)
10	2048		12	1,307(3)	1,459(2)	.091(0)	.060(0)
11	8192		6	1,282(3)	1,410(3)	.077(0)	.051(0)
12	32768		3	1,273(0)	1,384(0)	.067(0)	.044(0)
RANDOM ARRAYS							
13	32		96	1,150(9)	1,719(6)	.431(7)	.131(4)
14	128		48	1,204(7)	1,739(4)	.440(3)	.094(2)
15	512		24	1,222(5)	1,741(3)	.444(1)	.074(1)
16	2048		12	1,229(2)	1,739(1)	.449(0)	.060(0)
17	8192		6	1,283(3)	1,783(2)	.449(0)	.051(0)
18	32768		3	1,282(1)	1,779(0)	.452(0)	.044(0)
ARRAYS ALMOST IN SORT							
19	32	4	96	1,175(10)	1,538(8)	.225(36)	.131(4)
20	512	64	24	1,227(5)	1,557(4)	.256(8)	.074(1)
21	512	8	24	1,242(6)	1,484(4)	.168(23)	.074(1)
22	512	1	24	1,294(5)	1,394(5)	.066(78)	.074(1)
23	8192	1024	6	1,319(4)	1,614(4)	.243(8)	.051(0)
24	8192	128	6	1,298(3)	1,532(2)	.183(8)	.051(0)
25	8192	16	6	1,266(2)	1,449(1)	.131(23)	.051(0)
26	8192	2	6	1,272(1)	1,393(1)	.070(48)	.051(0)
ARRAYS OF EQUAL-LENGTH SORTED BLOCKS							
27	32	2	96	1,150(8)	1,694(6)	.413(10)	.131(4)
28	32	4	96	1,138(7)	1,706(5)	.438(8)	.125(5)
29	32	16	96	1,144(8)	1,719(5)	.444(7)	.131(4)
30	512	2	24	1,253(6)	1,771(4)	.444(2)	.074(1)
31	512	4	24	1,253(5)	1,756(4)	.442(2)	.074(1)
32	512	16	24	1,215(5)	1,734(3)	.446(2)	.073(1)
33	512	64	24	1,223(6)	1,740(3)	.443(2)	.074(1)
34	8192	2	6	1,274(4)	1,771(3)	.444(0)	.051(0)
35	8192	4	6	1,295(3)	1,795(2)	.449(1)	.051(0)
36	8192	16	6	1,283(3)	1,780(2)	.446(0)	.051(0)
37	8192	64	6	1,308(4)	1,805(3)	.446(1)	.051(0)

ALL COUNTS HAVE BEEN DIVIDED BY
N*LOG2(N)

Table III.

QSORT : SORTING PERFORMANCE TEST RESULTS

LINE	N	M	R	COMPARES	FETCHES	STORES	PARTITIONS
SORTED ARRAYS							
1	32			1,338	.700	0,000	.094
2	128			1,560	.741	0,000	.070
3	512			1,737	.785	0,000	.055
4	2048			1,867	.820	0,000	.045
5	8192			1,963	.847	0,000	.038
6	32768			2,034	.867	0,000	.033
ARRAYS SORTED IN REVERSE ORDER							
7	32			1,338	.700	.200	.094
8	128			1,560	.741	.143	.070
9	512			1,737	.785	.111	.055
10	2048			1,867	.820	.091	.045
11	8192			1,963	.847	.077	.038
12	32768			2,034	.867	.067	.033
RANDOM ARRAYS							
13	32		96	1,156(3)	.738(4)	.463(8)	.081(0)
14	128		48	1,302(1)	.815(3)	.467(2)	.059(1)
15	512		24	1,421(1)	.883(1)	.469(0)	.047(0)
16	2048		12	1,508(1)	.940(2)	.467(0)	.038(0)
17	8192		6	1,576(0)	.989(1)	.467(0)	.033(0)
18	32768		3	1,617(0)	1,011(1)	.470(0)	.028(0)
ARRAYS ALMOST IN SORT							
19	32	4	96	1,225(7)	.838(15)	.250(32)	.081(7)
20	512	64	24	2,074(12)	1,795(13)	.264(13)	.047(1)
21	512	8	24	2,305(24)	1,902(31)	.194(19)	.048(4)
22	512	1	24	1,904(13)	1,132(32)	.065(72)	.051(4)
23	8192	1024	6	2,447(16)	2,346(18)	.280(5)	.033(0)
24	8192	128	6	9,219(19)	8,986(20)	.196(14)	.033(1)
25	8192	16	6	25,940(40)	25,625(41)	.739(17)	.033(4)
26	8192	2	6	5,902(50)	5,214(60)	.073(60)	.033(4)
ARRAYS OF EQUAL-LENGTH SORTED BLOCKS							
27	32	2	96	1,306(1)	.769(2)	.463(9)	.075(8)
28	32	4	96	1,206(3)	.725(4)	.463(9)	.081(0)
29	32	16	96	1,163(3)	.725(5)	.475(7)	.081(7)
30	512	2	24	1,989(3)	1,379(4)	.473(1)	.047(1)
31	512	4	24	1,639(2)	1,071(4)	.461(2)	.047(0)
32	512	16	24	1,466(2)	.923(4)	.465(1)	.047(1)
33	512	64	24	1,411(1)	.871(2)	.465(1)	.047(1)
34	8192	2	6	2,649(2)	1,999(3)	.479(1)	.033(0)
35	8192	4	6	2,035(3)	1,419(4)	.466(1)	.033(0)
36	8192	16	6	1,702(1)	1,104(2)	.466(1)	.032(0)
37	8192	64	6	1,616(2)	1,026(5)	.467(0)	.032(0)

ALL COUNTS HAVE BEEN DIVIDED BY
N*LOG2(N)

Table II.

QUICKSORT : SORTING PERFORMANCE TEST RESULTS

LINE	N	M	R	COMPARES	FETCHES	STORES	PARTITIONS
SORTED ARRAYS							
1	32			.738	1,119	.281	.100
2	128			.795	1,077	.211	.071
3	512			.836	1,057	.166	.056
4	2048			.864	1,046	.136	.045
5	8192			.885	1,039	.115	.038
6	32768			.900	1,033	.100	.033
ARRAYS SORTED IN REVERSE ORDER							
7	32			.813	1,513	.563	.138
8	128			.845	1,358	.413	.100
9	512			.875	1,274	.322	.078
10	2048			.896	1,223	.264	.064
11	8192			.912	1,189	.223	.054
12	32768			.923	1,163	.193	.047
RANDOM ARRAYS							
13	32		96	.925(11)	1,631(6)	.569(6)	.131(4)
14	128		48	1,052(9)	1,675(5)	.528(3)	.094(2)
15	512		24	1,099(6)	1,690(3)	.518(1)	.074(1)
16	2048		12	1,169(5)	1,736(3)	.506(0)	.060(0)
17	8192		6	1,178(3)	1,730(2)	.500(0)	.051(0)
18	32768		3	1,241(3)	1,780(2)	.495(0)	.044(0)
ARRAYS ALMOST IN SORT							
19	32	4	96	.775(5)	1,294(5)	.388(9)	.125(5)
20	512	64	24	.929(3)	1,311(3)	.308(7)	.074(1)
21	512	8	24	.870(2)	1,182(3)	.238(10)	.073(2)
22	512	1	24	.843(1)	1,091(2)	.186(9)	.062(6)
23	8192	1024	6	1,034(2)	1,392(3)	.306(6)	.051(0)
24	8192	128	6	.948(2)	1,212(3)	.214(9)	.051(1)
25	8192	16	6	.909(1)	1,129(3)	.170(16)	.050(2)
26	8192	2	6	.896(0)	1,074(2)	.133(7)	.045(9)
ARRAYS OF EQUAL-LENGTH SORTED BLOCKS							
27	32	2	96	1,425(16)	2,100(11)	.538(9)	.131(4)
28	32	4	96	1,063(15)	1,763(9)	.563(5)	.131(4)
29	32	16	96	.925(13)	1,631(7)	.569(5)	.125(5)
30	512	2	24	3,373(24)	3,919(20)	.471(5)	.074(1)
31	512	4	24	1,602(12)	2,184(9)	.508(1)	.074(1)
32	512	16	24	1,209(8)	1,798(5)	.514(1)	.074(1)
33	512	64	24	1,113(5)	1,704(3)	.516(1)	.074(0)
34	8192	2	6	7,851(36)	8,343(34)	.441(3)	.051(0)
35	8192	4	6	1,880(8)	2,422(6)	.490(1)	.051(0)
36	8192	16	6	1,332(5)	1,880(3)	.497(0)	.051(0)
37	8192	64	6	1,234(3)	1,785(2)	.500(0)	.051(0)

ALL COUNTS HAVE BEEN DIVIDED BY
N*LOG2(N)

Table IV.

SHELLSORT : SORTING PERFORMANCE TEST RESULTS

LINE	N	M	R	COMPARES	FETCHES	STORES
SORTED ARRAYS						
1	32			.644	1.288	0.000
2	128			.724	1.449	0.000
3	512			.780	1.560	0.000
4	2048			.819	1.638	0.000
5	8192			.846	1.693	0.000
6	32768			.867	1.733	0.000
ARRAYS SORTED IN REVERSE ORDER						
7	32			.938	2.750	.875
8	128			1.051	3.009	.906
9	512			1.137	3.200	.926
10	2048			1.199	3.337	.939
11	8192			1.244	3.437	.949
12	32768			1.278	3.511	.956
RANDOM ARRAYS						
13	32		96	1,000(4)	2,900(6)	.900(11)
14	128		48	1,200(2)	3,506(4)	1,104(6)
15	512		24	1,379(2)	4,083(2)	1,325(4)
16	2048		12	1,567(3)	4,733(4)	1,599(6)
17	8192		6	1,779(3)	5,510(4)	1,952(6)
18	32768		3	2,143(3)	6,914(4)	2,628(6)
ARRAYS ALMOST IN SORT						
19	32	4	96	.800(6)	1,963(10)	.350(30)
20	512	64	24	1,166(2)	3,125(3)	.792(7)
21	512	8	24	.945(2)	2,224(4)	.333(14)
22	512	1	24	.813(2)	1,690(5)	.065(72)
23	8192	1024	6	1,514(1)	4,379(2)	1,350(3)
24	8192	128	6	1,225(0)	3,211(1)	.760(2)
25	8192	16	6	1,013(2)	2,361(3)	.334(13)
26	8192	2	6	.900(2)	1,907(5)	.107(48)
ARRAYS OF EQUAL-LENGTH SORTED BLOCKS						
27	32	2	96	.869(2)	2,331(5)	.594(12)
28	32	4	96	1,013(4)	2,975(4)	.950(9)
29	32	16	96	.975(3)	2,806(5)	.850(10)
30	512	2	24	1,079(3)	2,840(5)	.681(10)
31	512	4	24	1,508(3)	4,605(5)	1,589(7)
32	512	16	24	2,492(2)	8,634(3)	3,650(4)
33	512	64	24	1,740(2)	5,579(3)	2,098(3)
34	8192	2	6	1,139(1)	2,911(2)	.633(5)
35	8192	4	6	1,365(2)	4,664(4)	1,510(7)
36	8192	16	6	4,205(1)	15,264(1)	6,855(1)
37	8192	64	6	7,362(0)	28,005(0)	13,281(0)

Table V.

STRINGSORT : SORTING PERFORMANCE TEST RESULTS

LINE	N	M	R	COMPARES	FETCHES	STORES
SORTED ARRAYS						
1	32			1.175	2.750	.400
2	128			.853	1.991	.286
3	512			.666	1.554	.222
4	2048			.545	1.272	.182
5	8192			.462	1.077	.154
6	32768					
ARRAYS SORTED IN REVERSE ORDER						
7	32			1.175	2.750	.400
8	128			.853	1.991	.286
9	512			.666	1.554	.222
10	2048			.545	1.272	.182
11	8192			.462	1.077	.154
12	32768					
RANDOM ARRAYS						
13	32	96		2.819(19)	6.675(19)	1.025(18)
14	128	48		2.868(14)	6.796(14)	1.018(14)
15	512	24		2.776(11)	6.524(11)	.972(11)
16	2048	12		2.659(5)	6.242(5)	.924(5)
17	8192	6		2.672(0)	6.268(0)	.923(0)
18	32768					
ARRAYS ALMOST IN SORT						
19	32	4	96	2.281(4)	5.356(5)	.794(4)
20	512	64	24	2.605(0)	6.100(0)	.889(0)
21	512	8	24	1.797(17)	4.196(17)	.602(17)
22	512	1	24	.665(0)	1.553(0)	.222(0)
23	8192	1024	6	2.724(0)	6.371(0)	.923(0)
24	8192	128	6	1.839(0)	4.293(0)	.615(0)
25	8192	16	6	1.384(0)	3.229(0)	.462(0)
26	8192	2	6	.923(0)	2.154(0)	.308(0)
ARRAYS OF EQUAL-LENGTH SORTED BLOCKS						
27	32	2	96	1.156(0)	2.719(0)	.400(0)
28	32	4	96	2.263(1)	5.331(1)	.800(0)
29	32	16	96	2.788(20)	6.588(20)	1.013(19)
30	512	2	24	.665(0)	1.553(0)	.222(0)
31	512	4	24	1.329(0)	3.102(0)	.444(0)
32	512	16	24	1.941(0)	4.549(0)	.667(0)
33	512	64	24	2.546(0)	5.980(0)	.889(0)
34	8192	2	6	.461(0)	1.077(0)	.154(0)
35	8192	4	6	.923(0)	2.153(0)	.308(0)
36	8192	16	6	1.382(0)	3.226(0)	.462(0)
37	8192	64	6	1.808(0)	4.232(0)	.615(0)
ALL COUNTS HAVE BEEN DIVIDED BY N*LOG2(N)						

Table VI.

TREESORT3 SORTING PERFORMANCE TEST RESULTS

LINE	N	M	R	COMPARES	FETCHES	STORES
SORTED ARRAYS						
1	32			1.444	3.544	1.394
2	128			1.628	3.738	1.302
3	512			1.727	3.832	1.246
4	2048			1.785	3.874	1.198
5	8192			1.825	3.917	1.181
6	32768			1.849	3.924	1.151
ARRAYS SORTED IN REVERSE ORDER						
7	32			1.263	3.056	1.181
8	128			1.444	3.298	1.137
9	512			1.563	3.439	1.096
10	2048			1.641	3.543	1.082
11	8192			1.697	3.613	1.067
12	32768			1.739	3.671	1.062
RANDOM ARRAYS						
13	32	96		1.388(1)	3.381(1)	1.313(1)
14	128	48		1.556(0)	3.558(0)	1.228(0)
15	512	24		1.656(0)	3.662(0)	1.178(0)
16	2048	12		1.720(0)	3.726(0)	1.146(0)
17	8192	6		1.762(0)	3.767(0)	1.123(0)
18	32768	3		1.794(0)	3.798(0)	1.107(0)
ARRAYS ALMOST IN SORT						
19	32	4	96	1.425(1)	3.500(1)	1.381(0)
20	512	64	24	1.695(1)	3.763(1)	1.224(1)
21	512	8	24	1.722(0)	3.823(0)	1.243(0)
22	512	1	24	1.728(0)	3.835(0)	1.246(0)
23	8192	1024	6	1.777(2)	3.813(2)	1.148(2)
24	8192	128	6	1.804(1)	3.871(1)	1.166(1)
25	8192	16	6	1.825(0)	3.913(0)	1.177(0)
26	8192	2	6	1.824(0)	3.913(0)	1.178(0)
ARRAYS OF EQUAL-LENGTH SORTED BLOCKS						
27	32	2	96	1.381(1)	3.363(1)	1.319(1)
28	32	4	96	1.388(1)	3.388(1)	1.325(1)
29	32	16	96	1.400(1)	3.413(1)	1.325(1)
30	512	2	24	1.635(0)	3.603(0)	1.152(0)
31	512	4	24	1.619(0)	3.581(0)	1.154(0)
32	512	16	24	1.641(0)	3.628(0)	1.168(0)
33	512	64	24	1.655(0)	3.661(0)	1.179(0)
34	8192	2	6	1.744(0)	3.717(0)	1.102(0)
35	8192	4	6	1.731(0)	3.697(0)	1.101(0)
36	8192	16	6	1.737(0)	3.710(0)	1.105(0)
37	8192	64	6	1.745(0)	3.728(0)	1.110(0)
ALL COUNTS HAVE BEEN DIVIDED BY N*LOG2(N)						

Listing 1.

```

FUNCTION SHORT (A,N,NC,NF,NS,NR,PUSH,LPUSH)
C RUDOLF LOESER, 1972 OCT 10.
C--THIS IS A SORTING ROUTINE, SPECIFICALLY, IT IS A DRIVER
C FOR ARRAY-SPLITTING SORTING ALGORITHMS - FUNCTION KSORT.
C--THE CONTENTS OF THE ARRAY A, COMPRISING N ELEMENTS IN
C N SUCCESSIVE WORDS, WILL BE SORTED INTO NON-DESCENDING
C ORDER.
C NORMAL RETURN IS WITH SHORT=0.
C--UPON RETURN, NC = NUMBER OF COMPARISONS; NF = NUMBER OF
C FETCHES FROM A; NS = NUMBER OF STORES IN TO A.
C--ALSO UPON RETURN, NR = NUMBER OF SUBARRAYS OF LENGTH > 1
C (I.E. THE NUMBER OF CALLS TO KSORT).
C--PUSH, OF LENGTH LPUSH, IS AN ARRAY OF WORKING STORAGE,
C FOR SIMULATING A PUSHDOWN STACK.
C LPUSH=2*LOG2(N) SHOULD BE AMPLE.
C IF, AS SORTING PROCEEDS, MORE STORAGE IS REQUIRED THAN
C WAS SUPPLIED, SHORT WILL STOP SORTING AND IMMEDIATELY
C RETURN WITH SHORT=0.
      INTEGER PUSH(1),U,U1,SHORT
      NC=0
      NF=0
      NS=0
      NR=0
      SHORT=1
      IF (N=1) 109,109,101
101  J=LPUSH-2
      M=0
      L1=1
      U1=N
102  IF (U1-L1) 107,107,103
103  K=KSORT(A,L1,U1,L,U,NC,MF,MS)
      NR=NR+1
      NC=NC+MC
      NF=NF+MF
      NS=NS+MS
      IF (K) 107,107,104
      IF (M=J) 106,106,105
104  SHORT=0
      GO TO 109
105  M=M+2
      PUSH(M-1)=L
      PUSH(M)=U
      GO TO 102
107  IF (M) 109,109,108
108  L1=PUSH(M-1)
      U1=PUSH(M)
      M=M-2
      GO TO 102
109  RETURN
      END

```

The Fortran listings are exactly those of the routines tested. Since Fortran does not allow the recursive calls of the Algol versions of *quicksort* and *qsort*, I provided a driver that saves and restores index pairs explicitly (Listing 1). This routine drives not only *quicksort* and *qsort*; but *quicksort* as well (even though Scowen did not use recursive calls). Thus:

quicksort Listing 1 and Listing 2,
quicksort Listing 1 and Listing 3,
qsort Listing 1 and Listing 4,
Shellsort Listing 5,
stringsort Listing 6,
TREESORT3 Listing 7.

Tests of all routines except *quicksort* with sorted arrays, reverse-sorted arrays, and random arrays, for $n = 128$ and 2048, were repeated in Algol with a different random-number generator [13]. (The compiler used [14] would not compile *quicksort*.) The Algol versions were exactly as published (with small exceptions in *qsort* and *TREESORT3*, specified later). They were modified for determining N_c , N_s , and N_r . The observed values of N_c , N_s , and N_r from the Algol tests were identical to those in Tables I–VI for sorted and reverse-sorted arrays; for random arrays, the Algol values were well within the standard deviations of the Fortran results.

All the tests reported here involved arrays whose length n is even and equal to an integral power of 2.

Listing 2.

```

      FUNCTION KSCRT (A,L1,U1,L,U,NC,NF,NS)
C--THIS IS A SPLITTING ROUTINE FOR SCRT.
C IT IS AN ADAPTATION OF
C -QUICKSORT- (ACM ALGORITHMS 63-64) BY C.A.R. HOARE.
C (TRANSLATED FROM ALGOL.
C RUDOLF LOESER, 1971 SEP 10).
C--UPON ENTRY, (L1,U1) ARE THE INCLUSIVE LIMITING INDICES
C OF THE ARRAY TO BE SPLIT. IF U1-L1<1: CASE A:
C IF U1-L1=1: CASE B: IF U1-L1>1: CASE C.
C--UPON RETURN,
C IF CASE A: KSORT=0 AND NOTHING WAS DONE;
C IF CASE B: KSORT=0 AND THE ARRAY WAS SORTED;
C IF CASE C: KSORT=1 AND THE ARRAY WAS SPLIT, SUCH THAT
C (L1,U1) ARE THE LIMITING INDICES OF THE SMALLER SEGMENT,
C AND (L,U) ARE THE LIMITING INDICES OF THE LARGER ONE.
      DIMENSION A(1)
      INTEGER L1,U,P
      NC=0
      NF=0
      NS=0
      IF (U1-L1-1) 100,102,104
100  KSORT=0
101  CONTINUE
      RETURN
102  NC=NC+1
      NF=NF+2
      IF (A(L1)=A(U1)) 100,100,103
103  NF=NF+2
      NS=NS+2
      X=A(L1)
      A(L1)=A(U1)
      A(U1)=X
      GO TO 100
104  KSORT=1
      D=U1-L1+1
      P=D*RNRF(0)
      P=P+L1
      IF (P=L1) 105,106,106
      IF (P=U1) 107,107,105
106  NF=NF+1
      T=A(P)
      I=L1
      J=U1
108  IF (I=U1) 110,110,109
109  I=U1
      GO TO 112
110  NC=NC+1
      NF=NF+1
      IF (A(I)=T) 111,111,112
111  I=I+1
      GO TO 108
112  IF (J=L1) 113,114,114
113  J=L1
      GO TO 116
114  NC=NC+1
      NF=NF+1
      IF (A(J)=T) 116,115,115
115  J=J-1
      GO TO 112
116  IF (I=J) 117,119,119
117  NF=NF+2
      NS=NS+2
      X=A(I)
      A(I)=A(J)
      A(J)=X
      I=I+1
      J=J-1
      GO TO 108
119  IF (I=P) 120,122,122
120  NF=NF+2
      NS=NS+2
      X=A(I)
      A(I)=A(P)
      A(P)=X
      I=I+1
      GO TO 125
122  IF (P=J) 123,125,125
123  NF=NF+2
      NS=NS+2
      X=A(P)
      A(P)=A(J)
      A(J)=X
      J=J-1
125  IF ((J=L1)-(U1-I)) 127,126,126
126  L=L1
      U=J
      L1=I
      GO TO 101
127  L=J
      U=U1
      U1=J
      GO TO 101
      END

```

Listing 3.

```

      FUNCTION KSORT (A,L1,U1,L,U,NC,NF,NS)
C--THIS IS A SPLITTING ROUTINE FOR SHORT.
C IT IS AN ADAPTATION OF
C -QUICKERSORT- (ACM ALGORITHM 271) BY R.A. SCOWEN.
C (TRANSLATED FROM ALGOL.
C RUDOLF LOESER, 1971 SEP 10).
C--UPON ENTRY, (L1,U1) ARE THE INCLUSIVE LIMITING INDICES
C OF THE ARRAY TO BE SPLIT. IF U1-L1<1: CASE A:
C IF U1-L1=1: CASE B: IF U1-L1>1: CASE C.
C--UPON RETURN,
C IF CASE A: KSORT=0 AND NOTHING WAS DONE;
C IF CASE B: KSORT=0 AND THE ARRAY WAS SORTED;
C IF CASE C: KSORT=1 AND THE ARRAY WAS SPLIT, SUCH THAT
C (L1,U1) ARE THE LIMITING INDICES OF THE SMALLER SEGMENT,
C AND (L,U) ARE THE LIMITING INDICES OF THE LARGER ONE.
      DIMENSION A(1)
      INTEGER U1,U,P,Q
      NC=0
      NF=0
      NS=0
      IF (U1-L1-1) 100,102,104
100  KSORT=0
101  CONTINUE
      RETURN
102  NC=NC+1
      NF=NF+2
      IF (A(L1)=A(U1)) 100,100,103
103  NF=NF+2
      NS=NS+2
      X=A(L1)
      A(L1)=A(U1)
      A(U1)=X
      GO TO 100
104  KSORT=1
      P=(L1+U1)/2
      NF=NF+1
      T=A(P)
      NF=NF+1
      NS=NS+1
      A(P)=A(L1)
      Q=U1
      K=L1
106  K=K+1
      IF (K=Q) 107,107,113
107  NC=NC+1
      NF=NF+1
      IF (A(K)=T) 106,106,108
108  IF (Q=K) 113,109,109
109  NC=NC+1
      NF=NF+1
      IF (A(Q)=T) 111,110,110
110  Q=Q-1
      GO TO 108
111  NF=NF+2
      NS=NS+2
      X=A(K)
      A(K)=A(Q)
      A(Q)=X
      Q=Q-1
      GO TO 106
113  NF=NF+2
      NS=NS+2
      A(L1)=A(Q)
      A(Q)=T
      IF ((Q=Q)-(L1+U1)) 116,116,115
115  L=L1
      U=Q-1
      L1=Q+1
      GO TO 101
116  L=Q+1
      U=U1
      U1=Q-1
      GO TO 101
      END

```

They were all repeated with arrays of odd length $n - 1$. (Arrays of length n were generated just as before, but the value of the array length transmitted as a calling argument was $n - 1$.) All the measured quantities were comparable; neither odd nor even array lengths gave anomalous results.

4. Results

The tests with sorted arrays, reverse-sorted arrays, and arrays of equal-length sorted blocks were intended to show how these routines perform under extreme conditions. They do not represent practical applications of sorting. Tests with random arrays and with arrays

Listing 4.

```

FUNCTION KSCRT (A,L1,U1,L,U,NC,NF,NS)
C--THIS IS A SPLITTING ROUTINE FOR S-CRT.
C IT IS AN ADAPTATION OF
C -SCRT- (ACM ALGORITHM 402) BY P.F. VAN EMDEN.
C (TRANSLATED FROM ALGOL.
C RUDOLF LOESER, 1971 JAN 13).
C--UPON ENTRY, (L1,U1) ARE THE INCLUSIVE LIMITING INDICES
C OF THE ARRAY TO BE SPLIT. IF U1-L1<1: CASE A:
C IF U1-L1=1: CASE B: IF U1-L1>1: CASE C.
C--UPON RETURN,
C IF CASE A: KSCRT=0 AND NOTHING WAS DONE;
C IF CASE B: KSCRT=0 AND THE ARRAY WAS SORTED;
C IF CASE C: KSCRT=1 AND THE ARRAY WAS SPLIT, SUCH THAT
C (L1,U1) ARE THE LIMITING INDICES OF THE SMALLER SEGMENT,
C AND (L,U) ARE THE LIMITING INDICES OF THE LARGER ONE.
      DIMENSION A(1), INDEX(1)
      INTEGER L,U1,P,Q
      KSCRT=1
      NC=0
      NS=0
      L=L1
      U=U1
      P=L
      Q=U
      NF=2
      X=A(P)
      Z=A(Q)
      NC=NC+1
      IF (X=Z) 102,102,101
101  NS=NS+2
      Y=X
      X=Z
      A(P)=Z
      Z=Y
      A(Q)=Y
102  IF (U-L-1) 103,103,105
103  KSCRT=0
104  RETURN
105  XX=X
      IX=P
      ZZ=Z
      IZ=Q
106  JP=P+1
      KP=Q-1
      IF (KP-JP) 110,107,107
107  DO 109 P=JP,KP
      NF=NF+1
      X=A(P)
      NC=NC+1
      IF (X=XX) 109,111,111
109  CONTINUE
110  P=KP
      GO TO 122
111  JP=P+1
      KP=Q-1
      IF (KP-JP) 115,112,112
112  DO 114 J=JP,KP
      Q=Q-1
      NF=NF+1
      Z=A(Q)
      NC=NC+1
      IF (Z=ZZ) 116,116,114
114  CONTINUE
115  Q=P
      P=P-1
      Z=X
      NF=NF+1
      X=A(P)
      NC=NC+1
      IF (X=Z) 118,118,117
117  NS=NS+2
      Y=X
      X=Z
      A(P)=Z
      Z=Y
      A(Q)=Y
      NC=NC+1
      IF (X=XX) 120,120,119
119  XX=X
      IX=P
      NC=NC+1
      IF (Z=ZZ) 121,106,106
121  ZZ=Z
      IZ=Q
      GO TO 106
122  IF (P-IX) 123,125,123
123  NC=NC+1
      IF (X=XX) 124,125,125
124  NS=NS+2
      A(P)=XX
      A(IX)=X
      IF (Q-IZ) 126,128,126
126  NC=NC+1
      IF (Z=ZZ) 127,128,127
127  NS=NS+2
      A(Q)=ZZ
      A(IZ)=Z
128  IF ((U-Q)-(P-L)) 130,130,129
129  L1=L
      U1=P-1
      L=Q+1
      GO TO 104
130  U1=U
      L1=Q+1
      U=P-1
      GO TO 104
      END

```

Listing 5.

```

SUBROUTINE SHELL (A,N,NC,NF,NS)
C--THIS IS A SORTING ROUTINE.
C IT IS AN ADAPTATION OF
C -SHELLSORT- (ACM ALGORITHM 201) BY J. BOOTHROYD.
C (TRANSLATED FROM ALGOL.
C RUDOLF LOESER, 1971 OCT 11).
C--THE CONTENTS OF THE ARRAY A, COMPRISING N ELEMENTS IN
C N SUCCESSIVE WORDS, WILL BE SORTED INTO NON-DESCENDING
C ORDER.
C--UPON RETURN, NC = NUMBER OF COMPARISONS; NF = NUMBER OF
C FETCHES FROM A; NS = NUMBER OF STORES IN TO A.
      DIMENSION A(1)
      NC=0
      NF=0
      NS=0
      I=1
101  IF (I-N) 102,102,103
102  I=I+1
      GO TO 101
103  M=I-1
104  M=M/2
      IF (M) 110,110,105
105  K=N-M
      DO 109 J=1,K
      I=J+M
      I=I-M
      IF (I) 109,109,107
107  L=I+M
      NC=NC+1
      NF=NF+2
      IF (A(L)-A(I)) 108,109,109
108  NF=NF+2
      NS=NS+2
      W=A(I)
      A(I)=A(L)
      A(L)=W
      GO TO 106
109  CONTINUE
      GO TO 104
110  RETURN
      END

```

almost in sort, however, have practical relevance and may be useful to someone attempting to choose an algorithm for a particular sorting application.

Results for *quicksort* and *quickersort* appear in Tables I and II, respectively. Both the Fortran and the Algol versions of *quicksort* incorporate the improvement suggested by Hillmore [9]. For random arrays (lines 13–18), the two routines were expected to perform nearly alike. Indeed, their values of N_c are the same, which indicates that they split the arrays in the same way. The differences in the other measurements simply reflect their different scan-termination and array-element permutation techniques.

Overall, *quicksort* is a remarkably stable algorithm. Values of N_c and N_r seem to depend mainly on n and are essentially the same for the different types of arrays. Perhaps this stability results from choosing Y at random, no matter what the situation. In no case does N_c attain Hoare's theoretical minimum $N_c/[n \log_2(n)] = 1$. N_c always remains well below the expected average value $N_c/[n \log_2(n)] = 1.386$.

Algorithm *quickersort* does respond to the type of array. It uses fewer comparisons than *quicksort* when its choice of Y yields values nearer the median than *quicksort's* random choice. It performs badly on arrays of equal-length sorted blocks, where it picks an inappropriate value for Y . Both these effects were predicted. In many cases, N_c is considerably below $n \log_2(n)$ and even approaches the actual minimum value (cf. Appendix).

The results for *qsort* appear in Table III. The Algol version tested corresponds to the published one, with

Listing 6.

```

SUBROUTINE STRINGS (A,N,NC,NF,NS)
C--THIS IS A SORTING ROUTINE.
C IT IS AN ADAPTATION OF
C -STRINGSORT- (ACM ALGORITHM 207) BY J. BOOTHROYD.
C (TRANSLATED FROM ALGOL,
C RUDOLF LOESER, 1971 OCT 11).
C--THE CONTENTS OF THE ARRAY A, COMPRISING N ELEMENTS IN
C N SUCCESSIVE WORDS, WILL BE SORTED INTO NON-DESCENDING
C ORDER.
C--UPON RETURN, NC = NUMBER OF COMPARISONS; NF = NUMBER OF
C FETCHES FROM A; NS = NUMBER OF STORES IN TO A.
C>>USES A(N+1) THROUGH A(2*N) FOR SCRATCH STORAGE.<<
  DIMENSION A(1)
  INTEGER C,U,V,Z,C(3)
  NC=0
  NF=0
  NS=0
100 I=1
  J=N
  C(1)=N+1
  C(3)=N+N
101 D=1
  GO TO 114
102 NC=NC+1
  NF=NF+2
  IF (A(I)-A(Z)) 111,103,103
103 GO TO (104,108), V
104 NC=NC+1
  NF=NF+2
  IF (A(J)-A(Z)) 110,105,105
105 NC=NC+1
  NF=NF+2
  IF (A(I)-A(J)) 108,106,106
106 NF=NF+1
  NS=NS+1
  A(M)=A(J)
  J=J-1
  GO TO 115
108 NF=NF+1
  NS=NS+1
  A(M)=A(I)
  I=I+1
  GO TO 115
110 V=2
  GO TO 108
111 U=2
112 NC=NC+1
  NF=NF+2
  IF (A(J)-A(Z)) 113,106,106
113 D=-D
  C(D+2)=M
114 MD=-D
  M=C(MD+2)
  V=1
  U=1
  GO TO 105
115 Z=M
  M=M+D
  IF (J=I) 117,116,116
116 GO TO (102,112), U
117 IF (M-(N+1)) 100,119,118
118 I=N+1
  J=N+N
  C(1)=1
  C(3)=N
  GO TO 101
119 RETURN
END

```

two small modifications: the if statement labeled *out* and the next if statement as well were recoded as two if statements each, so that unnecessary comparisons could be avoided. The modified code reads:

```

out:  if p ≠ ix then
      begin if x ≠ xx then
        begin a[p] := xx; a[ix] := x end
      end;
      if q ≠ iz then
        begin if z ≠ zz then
          begin a[q] := zz; a[iz] := z end
        end;
mark: if u - q > p - l then      etc.

```

The Fortran version is equivalent to this modified Algol version.

Consider first the results for random arrays. The observed value of N_e is greater than *quicksort*'s and *quicksort*'s; in fact, it is almost 50 percent greater than the value van Emden predicted for large n . Nevertheless,

Listing 7.

```

SUBROUTINE TREES (A,N,NC,NF,NS)
C--THIS IS A SORTING ROUTINE.
C IT IS AN ADAPTATION OF
C -TREESORT3- (ACM ALGORITHM 245) BY R.W. FLOYD.
C (TRANSLATED FROM ALGOL,
C RUDOLF LOESER, 1971 OCT 11).
C--THE CONTENTS OF THE ARRAY A, COMPRISING N ELEMENTS IN
C N SUCCESSIVE WORDS, WILL BE SORTED INTO NON-DESCENDING
C ORDER.
C--UPON RETURN, NC = NUMBER OF COMPARISONS; NF = NUMBER OF
C FETCHES FROM A; NS = NUMBER OF STORES IN TO A.
  DIMENSION A(1)
  NC=0
  NF=0
  NS=0
  I=N/2+1
101 I=I-1
  IF (I=1) 103,103,102
102 CALL SIFTUP (A,I,N,NC,NF,NS)
  GO TO 101
103 I=N+1
104 I=I-1
  IF (I=1) 106,106,105
105 CALL SIFTUP (A,1,I,NC,NF,NS)
  NF=NF+2
  NS=NS+2
  X=A(1)
  A(1)=A(I)
  A(I)=X
  GO TO 104
106 RETURN
END

SUBROUTINE SIFTUP (A,L,N,NC,NF,NS)
G--A SUBROUTINE FOR TREES.
  DIMENSION A(1)
  I=L
  NF=NF+1
  COPY=A(I)
100 J=I+1
  IF (J=N) 101,101,107
101 IF (J=N) 102,104,104
102 NC=NC+1
  NF=NF+2
  IF (A(J+1)-A(J)) 104,104,103
103 J=J+1
104 NC=NC+1
  NF=NF+1
  IF (A(J)-COPY) 107,107,105
105 NF=NF+1
  NS=NS+1
  A(I)=A(J)
  I=J
  GO TO 100
107 NS=NS+1
  A(I)=COPY
  RETURN
END

```

there are two indications that *qsort* partitions arrays more effectively than *quicksort*: (1) *qsort*'s standard deviation of N_e is consistently less than half that for *quicksort*; and (2) *qsort*'s N_r is significantly smaller than *quicksort*'s. But only part of this decrease is due to better partitions; some of it comes about because *qsort* makes middle subarrays containing two, rather than just one, elements. To test how *qsort* would perform if it gave middle subarrays containing just one array element (like *quicksort*), I modified it: just before the if statement labeled *mark*, I randomly set either $p = q$ or $q = p$. The results, labeled *qsortx*, are shown in Table VIII. About 15 percent of the decrease of N_r is due to the larger middle subarrays; the remaining 85 percent must be due to better partitions.

In a second test, to determine the reason for the large values of N_e , the comparisons occurring in the bounding-interval adjustment section of the algorithm (beginning with the statement labeled *dist* and ending just before

the statement labeled *out*) were counted separately. The results appear in Figure 1. When the comparisons in question are excluded from N_c , the remaining number of comparisons does not exceed the predicted limit $N_c/[n \log_2(n)] = 1.140$.

It is not clear why van Emden found that *qsort* ran faster than his version of *quicksort*. The effect of the greater N_c must have been more than compensated by some other effect; most likely, the smaller N_r caused the smaller running time.

I tested the routines *Shellsort*, *stringsort*, and *TREESORT3* because Blair [15] reported on them in his certification of *quicksort*. He gives only timing data, which show *quicksort* running the fastest. I wanted to see whether the machine-independent measurements of my tests corroborate his results; they do. The results appear in Tables IV-VI. The Algol versions tested were exactly as published, except that in *TREESORT3*, following Abrams' [16] suggestion, *exchange* was not coded as a separate procedure.

Boothroyd's *Shellsort* has the advantage of being a very compact algorithm. His *stringsort*, however, is quite complex and, more important, requires auxiliary storage as large as the array to be sorted. Floyd's *TREESORT3*, whose correctness has been proved by London [17], is much like *quicksort* in its stability, but not in its performance.

All the test results are summarized in Table VII, where, at the position for each measurement, the name of the algorithm producing the smallest value of that measurement is given. (Of course, the column labeled

"Partitions" applies only for *quicksort*, *quicksort*, and *qsort*.)

In order to choose an in-place sorting algorithm for a particular application in a particular computing environment, one should consider not only the values of N_c , N_r , N_s , and N_t reported here, but also the times t_c , t_f , t_s , and t_r for comparing, fetching, storing, and initiating a splitting pass, respectively. If relative values of these times are known, then data such as those re-

Table VII.

SORTING PERFORMANCE TESTS : SUMMARY ROUTINES HAVING SMALLEST VALUES OF THE MEASURED COUNTS.							
LINE	N	M	R	COMPARES	FETCHES	STORES	PARTITIONS
SORTED ARRAYS							
1	32			SHELLSORT	QSORT	QUICKSORT	QSORT
2	128			SHELLSORT	QSORT	QUICKSORT	QSORT
3	512			STRINGSORT	QSORT	QUICKSORT	QSORT
4	2048			STRINGSORT	QSORT	QUICKSORT	QUICKERSORT
5	8192			STRINGSORT	QSORT	QUICKSORT	QUICKERSORT
6	32768			SHELLSORT	QSORT	QUICKSORT	QUICKERSORT
ARRAYS SORTED IN REVERSE ORDER							
7	32			QUICKERSORT	QSORT	QUICKSORT	QSORT
8	128			QUICKERSORT	QSORT	QUICKSORT	QSORT
9	512			STRINGSORT	QSORT	QUICKSORT	QSORT
10	2048			STRINGSORT	QSORT	QUICKSORT	QSORT
11	8192			STRINGSORT	QSORT	QUICKSORT	QSORT
12	32768			QUICKERSORT	QSORT	QUICKSORT	QSORT
RANDOM ARRAYS							
13	32			QUICKERSORT	QSORT	QUICKSORT	QSORT
14	128			QUICKERSORT	QSORT	QUICKSORT	QSORT
15	512			QUICKERSORT	QSORT	QUICKSORT	QSORT
16	2048			QUICKERSORT	QSORT	QUICKSORT	QSORT
17	8192			QUICKERSORT	QSORT	QUICKSORT	QSORT
18	32768			QUICKERSORT	QSORT	QUICKSORT	QSORT
ARRAYS ALMOST IN SORT							
19	32	4		QUICKERSORT	QSORT	QUICKSORT	QSORT
20	512	64		QUICKERSORT	QUICKERSORT	QUICKSORT	QSORT
21	512	8		QUICKERSORT	QUICKERSORT	QUICKSORT	QSORT
22	512	1		STRINGSORT	QUICKERSORT	QSORT	QSORT
23	8192	1024		QUICKERSORT	QUICKERSORT	QUICKSORT	QSORT
24	8192	128		QUICKERSORT	QUICKERSORT	QUICKSORT	QSORT
25	8192	16		QUICKERSORT	QUICKERSORT	QUICKSORT	QSORT
26	8192	2		QUICKERSORT	QUICKERSORT	QUICKSORT	QSORT
ARRAYS OF EQUAL-LENGTH SORTED BLOCKS							
27	32	2		SHELLSORT	QSORT	STRINGSORT	QSORT
28	32	4		SHELLSORT	QSORT	QUICKSORT	QSORT
29	32	16		QUICKERSORT	QSORT	QUICKSORT	QSORT
30	512	2		STRINGSORT	QSORT	STRINGSORT	QSORT
31	512	4		QUICKERSORT	QSORT	QUICKSORT	QSORT
32	512	16		QUICKERSORT	QSORT	QUICKSORT	QSORT
33	512	64		QUICKERSORT	QSORT	QUICKSORT	QSORT
34	8192	2		STRINGSORT	STRINGSORT	STRINGSORT	QSORT
35	8192	4		STRINGSORT	QSORT	STRINGSORT	QSORT
36	8192	16		QUICKSORT	QSORT	QUICKSORT	QSORT
37	8192	64		QUICKERSORT	QSORT	QUICKSORT	QSORT

Fig. 1. Number of comparisons vs array size (random arrays) for *qsort*.

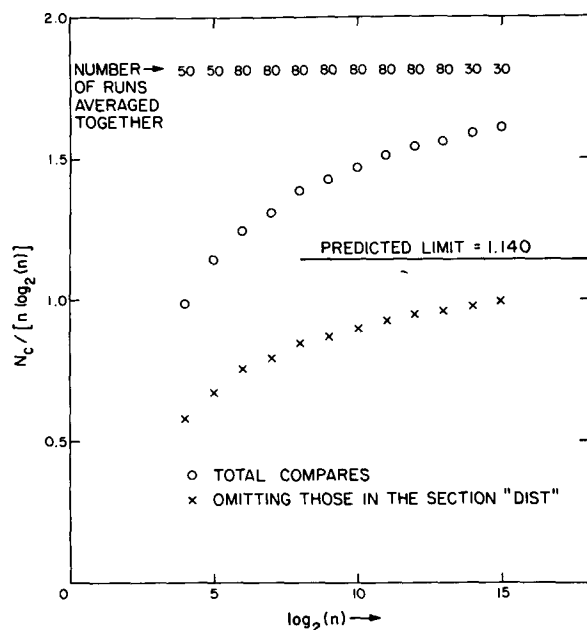


Table VIII.

Performance on Random Arrays (average of 10 runs)

	n	"quicksort"	"qsortx"	"qsort"
N_r	256	169	160	108
	1024	680	640	432
	4096	2722	2569	1726
N_c	256	2232	3226	2799
	1024	11676	17368	14966
	4096	59070	84702	75545

n = array size

N_r = number of calls to partitioning routine

N_c = number of comparisons

ported here can be used to select the most appropriate algorithm. If the choice is not clear, then the algorithms' performance, in the computing environment in question, should be tested explicitly. In any case, values of n , N_c , N_f , N_s , and N_r should be observed and monitored on an ongoing basis.

Acknowledgment. It is my pleasure to thank Dr. Fred Young for helping with this work, the Computer Center of the Smithsonian Astrophysical Observatory for providing a large amount of running time on its Control Data 6400 system, and the referees for helpful comments.

Appendix

Hoare states the following derivation of a theoretical minimum number of comparisons [7, p. 12]: "The theoretical minimum average number of comparisons required to sort n unequal randomly-ordered items may be estimated on information-theoretic considerations. As a result of a single binary comparison, the maximum entropy which may be destroyed is $-\log(2)$, while the original entropy of the randomly-ordered data is $-\log(n!)$; the final entropy of the sorted data is zero. The minimum number of comparisons required to achieve this reduction in entropy is $-\log(n!)/-\log(2) = \log_2(n!)$." Since values of this expression are not

readily obtainable for large n , Hoare also gives the approximation $\log_2(n!) \approx n \log_2(n)$.

This derivation has no particular regard for the detailed behavior of the published version of *quicksort*. The minimum possible number of comparisons for *quicksort* is greater than $\log_2(n!)$, because *quicksort* has been coded to require as many as $n + 2$ comparisons to split an array of length n . On the other hand, *quicksort* has been coded to make n comparisons to split an array of length n . Indeed, it may be possible to devise a scan requiring only $n - 1$ comparisons, since Y is one of the elements of the array to be partitioned and need not be compared with itself.

Optimum partitions, using the minimum number of comparisons, occur when segments are continually split into low and high subarrays of equal length. Consider the case $n = 11$ with a partitioning routine that requires $k = n + 2$ comparisons per scan and splits (rather than sorts) subarrays of length two (e.g., like *quicksort*). The first partition would require 13 comparisons; the result would be two segments of length five (and a middle part of length one, which needs no further consideration). Each segment of length five would require seven comparisons to be split into two segments, each of length two (and again with middle parts of length one). Arrays of length two are the smallest to be split, requiring four comparisons each. The total number of comparisons clearly is $13 + 7 + 7 + 4 + 4 + 4 + 4 + 4 = 43$. For $n = 11 + 11 + 1 = 23$, the first partition requires 25 comparisons and can produce two segments each of length 11. The minimum number of comparisons evidently is $43 + 43 + 25 = 111$. Similarly, with $n = 23 + 23 + 1 = 47$, the minimum number of comparisons is $111 + 111 + 49 = 271$.

If subarrays of length two are sorted rather than split (i.e.

Table A-1.

Minimum number of comparisons, for sorting arrays of length n with splitting routines requiring k comparisons per splitting scan, producing middle parts of length 1, and either splitting segments of length 2 (case a) or sorting segments of length 2 (case b). All values were divided by $n \log_2(n)$.

n	$\log_2(n!)$	k = n-1		k = n		k = n+1		k = n+2	
		a	b	a	b	a	b	a	b
11	0.664	0.578	0.578	0.762	0.657	0.946	0.736	1.130	0.815
23	0.716	0.634	0.634	0.779	0.702	0.923	0.769	1.067	0.836
47	0.756	0.682	0.682	0.801	0.739	0.919	0.797	1.038	0.854
95	0.789	0.721	0.721	0.822	0.771	0.923	0.820	1.024	0.870
191	0.813	0.753	0.753	0.841	0.797	0.929	0.840	1.016	0.884
383	0.834	0.780	0.780	0.857	0.818	0.935	0.857	1.012	0.895
767	0.850	0.801	0.801	0.871	0.836	0.940	0.871	1.010	0.905
1535	0.864	0.820	0.820	0.882	0.851	0.945	0.882	1.008	0.914
3071	0.876	0.835	0.835	0.892	0.864	0.950	0.892	1.007	0.921
6143	0.886	0.848	0.848	0.901	0.874	0.954	0.901	1.007	0.927
12287	0.894	0.859	0.859	0.908	0.883	0.957	0.908	1.006	0.932
24575	0.901	0.869	0.869	0.914	0.891	0.960	0.914	1.006	0.937
49151	0.907	0.877	0.877	0.919	0.898	0.962	0.920	1.005	0.941
98303	0.913	0.884	0.884	0.925	0.904	0.965	0.925	1.005	0.945
196607	0.918	0.891	0.891	0.929	0.910	0.967	0.929	1.005	0.948
393215	0.922	0.897	0.897	0.933	0.915	0.969	0.933	1.004	0.951
786431	0.926	0.902	0.902	0.936	0.919	0.970	0.936	1.004	0.953

Hillmore's modification), it would take $13 + 7 + 7 + 1 + 1 + 1 + 1 = 31$ comparisons to sort an array of length eleven. If, in addition, the splitting routine uses only $k = n$ comparisons per scan, then only $11 + 5 + 5 + 1 + 1 + 1 + 1 = 25$ comparisons would be needed to sort that same array.

The foregoing illustrates a recursive procedure that generates a certain sequence of array sizes n_i and the corresponding minimum number of comparisons c_i . The c_i depend on the number of comparisons per scan and on the handling of subarrays of length two; the n_i depend on the number of elements in the middle subarrays. (The values of c_i obtained in this way are probably local minima of the variation of c with n , since the n_i are deliberately chosen to make equal-length subarrays possible—no other values of n permit this.) Table A-1 shows values of c_i for middle subarrays of length one. There are two sets of values for each value of k : (1) for the case that subarrays of length two are partitioned, and (2) for the case that subarrays of length two are sorted. Values of $\log_2 (n_i!)$ are also shown. (These were obtained by summation, according to $\log_2 n! = \log_2 n + \log_2 [(n-1)!]$.) All numbers in Table A-1 have been divided by $n \log_2 (n)$, as described in Section 3.

Note first that for the array sizes shown, values of the approximation $n \log_2 (n)$ are considerably larger than $\log_2 (n!)$. Nevertheless, it happens that $n \log_2 (n)$ predicts the minimum number of comparisons quite well for a routine like *quicksort* ($k = n + 2$, case 1). However, its values, and those of $\log_2 (n!)$ as well, are too large for a routine like *quickersort* ($k = n$, case 2), and even more so for a routine with $k = n - 1$.

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L.D. Fosdick and

A.K. Cline, Editors

Algorithms

Submittal of an algorithm for consideration for publication in Communications of the ACM implies unrestricted use of the algorithm within a computer is permissible.

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Algorithm 475

Visible Surface Plotting Program [J6]

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National Center for Atmospheric Research is sponsored by the National Science Foundation

Key Words and Phrases: hidden line problem, computer graphics, contour surface

CR Categories: 3.65, 4.41, 8.2

Language: Fortran

[This program is not in ANSI Fortran. Nonstandard features are noted in the text. A demonstration driver is included to illustrate use of the subroutines. I/O unit 9 is used by this driver.—LDF.]

Description

This package of three routines produces a perspective picture of an arbitrary object or group of objects with the hidden parts not drawn. The objects are assumed to be stored in the format described below, a format which was chosen to facilitate the display of functions of three variables (Figure 1) or output from three-dimensional computer simulations (Figure 2). The basic method is to contour cuts through the array, starting with a cut nearest the observer. The algorithm leaves out the hidden parts of the contours by suppressing lines enclosed within lines produced while processing preceding cuts. The technique is described in detail in [2].

The object is defined in a three-dimensional array by setting words to one where the object is, and to zero where it is not. That is, the position in the array corresponds to a position in three-space, and the value of the array tells whether any object is present at that position or not. Because a large array is needed to define objects with good resolution, only a part of the array is passed to the package with each call.

There are three subroutines in the package. *INIT3D* is called

Fig. 1. Four contour surfaces of the wave function of a 3-P electron in a one electron atom: $50 \times 50 \times 50$ object cube, 100×100 screen model.

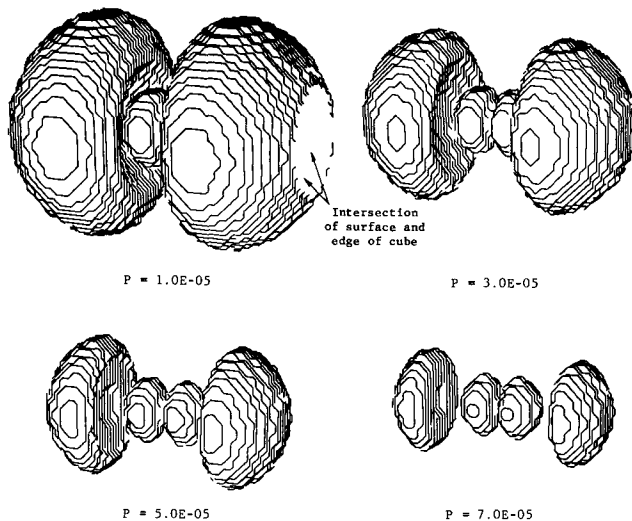


Fig. 2. Output from a three-dimensional cloud model: $100 \times 100 \times 60$ object cube, 200×200 screen model.

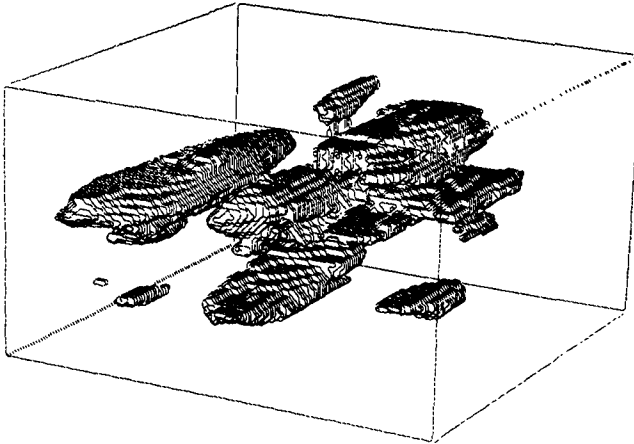
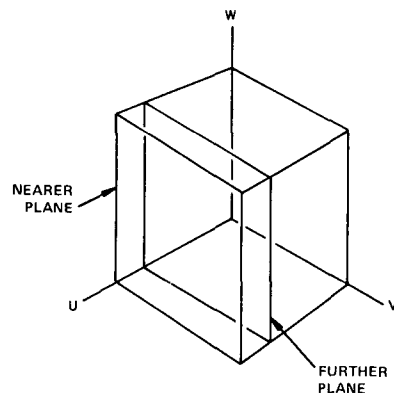


Fig. 3. Processing different parts of a three-dimensional array.



at the beginning of a picture. This call can be skipped sometimes if certain criteria are met and certain precautions are taken. See the comment lines for details. *SETORG* (which has an entry point *PERSPC*) does three-space to two-space perspective transformations. It is called by *INIT3D* and need not be called by the user. The mathematical method for the three-space to two-space transformation is due to Kubert, Szabo, and Giulieri [1]. *DANDR* (draw and remember) is called successively to process different parts of the three-dimensional array. For example, in Figure 3, the

nearer plane would be processed in the first call to *DANDR*, while the further plane would be processed in a subsequent call. A sample program is provided with the algorithm to illustrate this point.

Although this package was developed using NCAR's CDC machines with locally written systems and compilers, implementation on different machines or systems should not be too difficult regardless of the plotter. The algorithm has been tested on the Minnesota Fortran compiler (MNF), and when the following items are taken care of, should be portable.

There is a *PROGRAM* card in the demonstration program. There is an *ENTRY* statement in *SETORG*. *ENTRY* statements are nonstandard, but are generally portable. It could be eliminated, but the package would run longer. There are two machine-dependent variables used and described in *DANDR*. There is one system routine, *LINE*, called once and described in *DANDR*, which must be implemented or simulated to use this package. In three statements (which are marked) in *DANDR*, *.OR.* and *.AND.* are used for masking operations with integer variables. Some compilers may not produce the desired code, so references to machine language functions may have to be substituted. There is a nonstandard but common form of the *DATA* statement in *DANDR*. Functions which are assumed available are *SQRT*, *ACOS*, and *SIN*.

Figures 4 and 5 are referred to in the listing as the first picture and the second picture.

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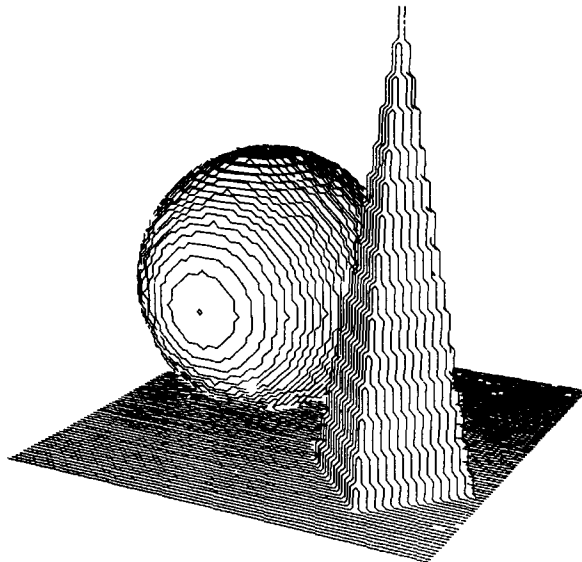
Algorithm

```

PROGRAM ACMTEST
C DEMONSTRATION PROGRAM
DIMENSION EYE(3), S(4), ST1(80,80,2), IS2(3,160)
DIMENSION IOBJ(80,80)
C USE WHOLE FRAME
S(1) = 0.
S(2) = 1.
S(3) = 0.
S(4) = 1.
C SET EYE POSITION
EYE(1) = 250.
EYE(2) = 150.
EYE(3) = 100.
C INITIALIZE PACKAGE
CALL INIT3D(EYE, 80, 80, 80, ST1, 3, 160, IS2, 9, S)
C CREATE AND PLOT TEST OBJECT
DO 50 I=1,80
  A = (I-50)**2
  DO 40 J=1,80
    C = (J-25)**2
    D = IABS(J-63) + IABS(I-25)
    DO 30 K=1,80
      C FL00R
        IF (K.EQ.1) GO TO 10
      C BALL
        IF (SQRT(A+C+(FL00R(K)-25)**2).LE.25.) GO TO 10
      C POINT
        IF (D.GT.FL00R(80-K)*.1875) GO TO 20
        IOBJ(J,K) = 1
        GO TO 30
        IOBJ(J,K) = 0
      20 CONTINUE
      30 CONTINUE
      40 CONTINUE
      CALL DANDR(80, 80, ST1, 3, 160, 160, IS2, 9, S, IOBJ,
        * 80)
      50 CONTINUE
C ADVANCE TO THE NEXT FRAME.
CALL FRAME
C A SECOND PICTURE WILL NOW BE CALLED USING THE SAME SIZE
C ARRAYS AND EYE POSITION. THIS MEANS THE CALL TO INIT3D.
C THE BIGGEST TIME CONSUMER, CAN BE SKIPPED IF THE FOLLOWING
C FOUR LINES ARE INCLUDED.
REWIND 9
DO 70 I=1,3
  DO 60 J=1,160
    IS2(I,J) = 0
  60 CONTINUE
  70 CONTINUE
C THIS PICTURE WILL BE THE T=4 CONTOUR SURFACE OF
C T=1/SQRT(U*U+V*V+W*W)+(.5-V)**2/SQRT(U*U+V*V).
DO 120 I=1,80
  U = (40.5-FL00R(I))/79.
  UU = U*U
  DO 110 J=1,80
    V = (FL00R(J)-40.5)/79.
    VV = V*V
    A = 1./SQRT(UU+VV)
    DO 100 K=1,80

```

Fig. 4. The first picture produced by the test program.



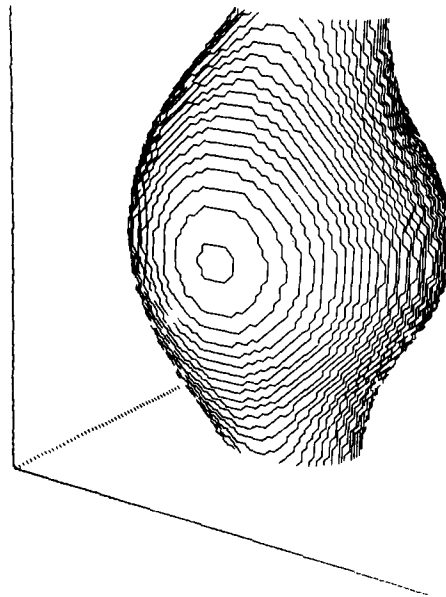
```

C THE FOLLOWING CARD ADDS AXES.
  IF (I*J.EQ.1 .OR. I*K.EQ.1 .OR. J*K.EQ.1) GO TO 80
  W = (FLOAT(K)-40.5)/79.
  IF (1./SQRT(UU+VV+W*W))+(.5-V)**2*A.LE.4.) GO TO 90
  I0BJ(J,K) = 1
  GO TO 100
  I0BJ(J,K) = 0
  90 CONTINUE
  100 CONTINUE
  110 CALL DANDK(80, 80, ST1, 3, 160, 160, IS2, 9, S, I0BJ,
    * 80)
  120 CONTINUE
C FLUSH PLOT BUFFER
  CALL FRAME
  STOP
  END

SUBROUTINE INIT3D(EYE, NU, NV, NW, ST1, LX, NY, IS2, IU,
  * S)
  DIMENSION EYE(3), ST1(NV,NW,2), IS2(LX,NY), S(4)
C BY THOMAS WRIGHT
C COMPUTING FACILITY
C THE NATIONAL CENTER FOR ATMOSPHERIC RESEARCH
C BOULDER, COLORADO 80302
C NCAR IS SPONSORED BY THE NATIONAL SCIENCE FOUNDATION.
C THE METHOD IS DESCRIBED IN DETAIL IN - A ONE-PASS HIDDEN-
C LINE REMOVER FOR COMPUTER DRAWN THREE-SPACE OBJECTS. PROC
C 1972 SUMMER COMPUTER SIMULATION CONFERENCE, 261-267, 1972.
C THIS VERSION IS FOR USE ON CDC 6000 OR 7000 COMPUTERS.
C THIS PACKAGE OF ROUTINES PLOTS 3-DIMENSIONAL OBJECTS WITH
C HIDDEN PARTS NOT SHOWN. OBJECTS ARE STORED IN AN ARRAY,
C WITH THE POSITION IN THE ARRAY CORRESPONDING TO A LOCATION
C IN 3-SPACE AND THE VALUE OF THE ARRAY ELEMENT TELLING IF
C ANY OBJECT IS PRESENT AT THE LOCATION.
C INIT3D IS AN INITIALIZATION ROUTINE FOR THIS PACKAGE. IT
C IS CALLED, THEN A SEQUENCE OF CALLS ARE MADE TO DANDK TO
C PRODUCE A PICTURE.
C EYE AN ARRAY 3 LONG CONTAINING THE U, V, AND W COORDI-
C NATES OF THE EYE POSITION. OBJECTS ARE CONSIDERED
C TO BE IN A BOX WITH 2 EXTREME CORNERS AT (1,1,1) AND
C (NU,NV,NW). THE EYE POSITION MUST HAVE POSITIVE
C COORDINATES AWAY FROM THE COORDINATE PLANES U=0,
C V=0, AND W=0. WHILE GAINING EXPERIENCE WITH THE
C PACKAGE, USE EYE(1)=5*NU, EYE(2)=4*NW, EYE(3)=3*NU.
C NU U DIRECTION LENGTH OF THE BOX CONTAINING THE OBJECTS
C NV V DIRECTION LENGTH OF THE BOX CONTAINING THE OBJECTS
C NW W DIRECTION LENGTH OF THE BOX CONTAINING THE OBJECTS
C ST1 A SCRATCH ARRAY AT LEAST NV*NW*2 WORDS LONG.
C LX FIRST DIMENSION OF A SCRATCH ARRAY, IS2, USED BY THE
C PACKAGE FOR REMEMBERING WHERE IT SHOULD NOT DRAW.
C NY SECOND DIMENSION OF IS2. SEE DANDK COMMENTS.
C IS2 A SCRATCH ARRAY AT LEAST LX*NY WORDS LONG.
C IU UNIT NUMBER OF SCRATCH FILE FOR THE PACKAGE. ST1
C WILL BE WRITTEN NU TIMES ON THIS FILE.
C S AN ARRAY 4 LONG WHICH CONTAINS THE COORDINATES OF
C THE AREA WHERE THE PICTURE IS TO BE DRAWN. THAT IS,
C ALL PLOTTING COORDINATES GENERATED WILL BE BOUNDED
C AS FOLLOWS-- X COORDINATES WILL BE BETWEEN S(1) AND
C S(2), Y COORDINATES WILL BE BETWEEN S(3) AND S(4).
C TO PREVENT DISTORTION, HAVE S(2)-S(1)=S(4)-S(3).
C IF SEVERAL PICTURES ARE TO BE DRAWN WITH THE SAME SIZE
C ARRAYS AND EYE POSITION AND THE USER REWINDS IU AND FILLS
C IS2 WITH ZEROS, INIT3D NEED NOT BE CALLED FOR OTHER THAN
C THE FIRST PICTURE.
C SET UP TRANSFORMATION ROUTINE FOR THIS LINE OF SIGHT.
  U = NU
  V = NV
  W = NW
  CALL SETORG(U*.5, V*.5, W*.5, EYE(1), EYE(2), EYE(3))

```

Fig. 5. The second picture produced by the test program.



```

C FIND EXTREMES IN TRANSFORMED SPACE.
  CALL PERSPC(1., 1., W, D, YI, D)
  CALL PERSPC(U, V, 1., D, YB, D)
  CALL PERSPC(U, 1., 1., XL, D, D)
  CALL PERSPC(1., V, 1., XR, D, D)
C ADJUST EXTREMES TO PREVENT DISTORTION WHEN GOING FROM
C TRANSFORMED SPACE TO PLOTTER SPACE.
  DIF = (XR-XL-YI+YB)*.5
  IF (DIF) 10, 30, 20
  10 XL = XL + DIF
  XR = XR - DIF
  GO TO 30
  20 YB = YB - DIF
  YI = YI + DIF
  30 REWIND IU
C FIND THE PLOTTER COORDINATES OF THE 3-SPACE LATTICE POINTS
  C1 = .9*(S(2)-S(1))/(XR-XL)
  C2 = .05*(S(2)-S(1)) + S(1)
  C3 = .9*(S(4)-S(3))/(YI-YB)
  C4 = .05*(S(4)-S(3)) + S(3)
  DO 60 I=1,NU
    U = NU + I - 1
    DO 50 J=1,NV
      V = J
      DO 40 K=1,NW
        CALL PERSPC(U, V, FLOAT(K), X, Y, D)
        ST1(J,K,1) = C1*(X-XL) + C2
        ST1(J,K,2) = C3*(Y-YB) + C4
      40 CONTINUE
    50 CONTINUE
  60 WRITE THEM ON UNIT IU.
    WRITE (IU) ST1
  60 CONTINUE
  REWIND IU
C ZERO OUT ARRAY WHERE VISIBILITY IS REMEMBERED.
  DO 80 J=1,NY
    DO 70 I=1,LX
      IS2(I,J) = 0
    70 CONTINUE
  80 CONTINUE
  RETURN
  END

SUBROUTINE SETORG(X, Y, Z, XT, YT, ZT)
C THIS ROUTINE IMPLEMENTS THE 3-SPACE TO 2-SPACE TRANSFOR-
C MATION BY KUBER, SZABO AND GIULIEMI, THE PERSPECTIVE
C REPRESENTATION OF FUNCTIONS OF TWO VARIABLES. J. ACM 15,
C 2, 193-204, 1968.
C SETORG ARGUMENTS
C X,Y,Z ARE THE 3-SPACE COORDINATES OF THE INTERSECTION
C OF THE LINE OF SIGHT AND THE IMAGE PLANE. THIS
C POINT CAN BE THOUGHT OF AS THE POINT LOOKED AT.
C XT,YT,ZT ARE THE 3-SPACE COORDINATES OF THE EYE POSITION.
C PERSPC ARGUMENTS
C X,Y,Z ARE THE 3-SPACE COORDINATES OF A POINT TO BE
C TRANSFORMED.
C XT,YT THE RESULTS OF THE 3-SPACE TO 2-SPACE TRANSFOR-
C MATION.
C ZT NOT USED.
C STORE THE PARAMETERS OF THE SETORG CALL FOR USE WHEN
C PERSPC IS CALLED.
  AX = X
  AY = Y
  AZ = Z
  EX = XT
  EY = YT
  EZ = ZT
C AS MUCH COMPUTATION AS POSSIBLE IS DONE DURING EXECUTION
C OF SETORG SINCE PERSPC IS CALLED THOUSANDS OF TIMES FOR

```


Remark on Algorithm 179 [S14]

Incomplete Beta Ratio

[Oliver G. Ludwig, *Comm. ACM* 6 (June 1963), 314]

Nancy E. Bosten and E.L. Battiste [Recd. 1 Sept. 1972 and 15 Mar. 1973]

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Description

Algorithm 179 (modified to include the remark by M.C. Pike and I.D. Hill [1]) computes the Incomplete Beta Ratio using this equation

$$I_x(p, q) = \frac{INFSUM \cdot x^p \cdot \Gamma(PS+p)}{\Gamma(PS) \cdot \Gamma(p+1)} + \frac{x^p \cdot (1-x)^q \cdot \Gamma(p+q) \cdot FINSUM}{\Gamma(p) \cdot \Gamma(q+1)}$$

INFSUM and *FINSUM* represent two series summations defined as follows:

$$INFSUM = \sum_{i=0}^{\infty} \frac{(1-PS)_i \cdot p}{p+i} \frac{x^i}{i!}, \text{ where}$$
$$(1-PS)_i = 1 \quad [i=0]$$
$$= (1-PS) \cdot (2-PS) \cdots (i-PS) = \frac{\Gamma(1+i-PS)}{\Gamma(1-PS)} \quad [i>0]$$

$$\text{and } FINSUM = \sum_{i=1}^{[q]} \frac{q \cdot (q-1) \cdots (q-i+1)}{(p+q-1)(p+q-2) \cdots (p+q-i)} \frac{1}{(1-x)^i},$$

where $[q]$ is equal to the largest integer less than q . If $[q] = 0$, then *FINSUM* = 0. *PS* is defined as

PS = 1, if q is an integer; otherwise
 $= q - [q]$.

By rearranging Algorithm 179 so that scaling can be introduced, the argument range of p and q can be extended and accuracy can be improved.

Since $I_x(p, q)$ is a probability and, therefore, bounded $[0, 1]$, and *INFSUM* and *FINSUM* are series having only positive terms, we see that $I_x(p, q)$ is a collection of terms all of which are positive and bounded in the range $[0, 1]$ if: (1) each term of *INFSUM* is multiplied by $(x^p \cdot \Gamma(PS+p))/(\Gamma(PS) \cdot \Gamma(p+1))$; and (2) each term of *FINSUM* is multiplied by $(x^p \cdot (1-x)^q \cdot \Gamma(p+q))/(\Gamma(p) \cdot \Gamma(q+1))$.

Knowing this fact, we can apply a scaling procedure to the algorithm. *INFSUM* is a decreasing series. If the product of the first term of *INFSUM* and its multiplicative factor would underflow, then the sum of this series could be set to zero and all calculations involving underflow could be avoided. This is handled in the modification of the algorithm given below. However, since *INFSUM* is a decreasing series, underflows may occur later in the calculations. No attempt has been made to handle them here.

The second summation is more complicated. The series is decreasing if $q/(q+p-1)(1-x)$ is less than 1. If an individual term becomes less than $1.E-6$ times the previous sum, calculation can be legitimately terminated since no additivity is apparent. If a term of the decreasing series is less than an arbitrarily small constant (*EPS2*), calculation is also terminated. This is done to prevent underflows in the later terms.

If the series is increasing, the first terms may underflow. In this case a power of ϵ_1 (machine precision $\sim 1.E-78$ on the IBM 360/370) may be factored from each term in *FINSUM* (times its multiplier). These terms cannot be added to the sum since they are less than machine precision; however, they are useful in retaining the accuracy of the initial terms, which are then used recursively. By the nature of the problem, we know that any term in *FINSUM*, times its multiplier, must be less than or equal to 1, but we have factored out powers of ϵ_1 . Therefore, if a term of *FINSUM* becomes greater than 1, we know that rescaling, by multiplying the term by ϵ_1 , is in order.

Testing on the IBM 360/195 has shown that, by rearranging the calculations of the original Algorithm 179, and thus including

scaling, the input range of the algorithm can be greatly extended with a high degree of accuracy.

MDBETA requires a double precision function *DLGAMA* which computes the log of the gamma function. ACM Algorithm 291 may be used. *MDBETA* was tested against the SSP routine *BDTR* given in the manual *System/360 Scientific Subroutine Package (360A-CM-03X) Version III Programmer's Manual*, H20-0205. *MDBETA* ran 3.5 times faster than *BDTR* with greater accuracy. For example, in the case $x = .5$, $p = 2000$ and $q = 2000$, *MDBETA* gave the correct result, .5, while *BDTR* gave an answer of .497026. The IMSL subroutine, *MDBIN*, was used for an additional comparison when p and q are integers. *MDBIN* maintains IBM 370/360 single precision accuracy (approximately six significant digits). Over the tests performed the maximum difference occurred in the fifth significant digit when p and q were less than 200. Three to four significant digits of accuracy can be expected with p and q as large as 2000.

Acknowledgments. The above ideas are the application of ideas learned from the late Hirono Kuki. Routine *MDBETA* originated from a code which resides in IMSL Library 1. We thank Wayne Fullerton, from the University of California, Los Alamos Scientific Laboratory, for refereeing the paper.

Algorithm

```
SUBROUTINE MDBETA(X, P, Q, PROB, IER)
C FUNCTION - INCOMPLETE BETA PROBABILITY
C DISTRIBUTION FUNCTION
C USAGE - CALL MDBETA (X,P,Q,PROB,IER)
C PARAMETERS
C X - VALUE TO WHICH FUNCTION IS TO BE INTEGRATED. X
C MUST BE IN THE RANGE (0,1) INCLUSIVE.
C P - INPUT (1ST) PARAMETER (MUST BE GREATER THAN 0)
C Q - INPUT (2ND) PARAMETER (MUST BE GREATER THAN 0)
C PROB - OUTPUT PROBABILITY THAT A RANDOM VARIABLE FROM A
C BETA DISTRIBUTION HAVING PARAMETERS P AND Q
C WILL BE LESS THAN OR EQUAL TO X.
C IER - ERROR PARAMETER.
C IER = 0 INDICATES A NORMAL EXIT
C IER = 1 INDICATES THAT X IS NOT IN THE RANGE
C (0,1) INCLUSIVE.
C IER = 2 INDICATES THAT P AND/OR Q IS LESS THAN
C OR EQUAL TO 0.
C DOUBLE PRECISION PS, PX, Y, PI, DP, INFSUM, CNT, VH, XB,
C * DQ, C, EPS, EPS1, ALEPS, FINSUM, PQ, D4, EPS2, DLGAMA
C DOUBLE PRECISION FUNCTION DLGAMA
C MACHINE PRECISION
C DATA EPS/1.D-6/
C SMALLEST POSITIVE NUMBER REPRESENTABLE
C DATA EPS1/1.D-78/
C NATURAL LOG OF EPS1
C DATA ALEPS/-179.6016D0/
C ARBITRARILY SMALL NUMBER
C DATA EPS2/1.D-50/
C CHECK RANGES OF THE ARGUMENTS
Y = X
IF ((X.LE.1.0) .AND. (X.GE.0.0)) GO TO 10
IER = 1
GO TO 140
10 IF ((P.GT.0.0) .AND. (Q.GT.0.0)) GO TO 20
IER = 2
GO TO 140
20 IER = 0
IF (X.GT.0.5) GO TO 30
INT = 0
GO TO 40
C SWITCH ARGUMENTS FOR MORE EFFICIENT USE OF THE POWER
C SERIES
30 INT = 1
TEMP = P
P = Q
Q = TEMP
Y = 1.D0 - Y
40 IF (X.NE.0. .AND. X.NE.1.) GO TO 60
C SPECIAL CASE - X IS 0. OR 1.
50 PROB = 0.
GO TO 130
60 IB = Q
TEMP = IB
PS = Q - FLOAT(IB)
IF (Q.EQ.TEMP) PS = 1.D0
DP = P
DQ = Q
PX = DP*DLOG(Y)
PQ = DLGAMA(DP+DQ)
PI = DLGAMA(DP)
C = DLGAMA(DQ)
D4 = DLOG(DP)
IF (Y.GT.EPS) GO TO 70
C SPECIAL CASE - X IS CLOSE TO 0. OR 1.
XB = PX + PQ - D4 - PI - C
IF (XB.LE.ALEPS) GO TO 50
PROB = DEXP(XB)
GO TO 130
C DLGAMA IS A FUNCTION WHICH CALCULATES THE DOUBLE
C PRECISION LOG GAMMA FUNCTION
70 XB = PX + DLGAMA(PS+DP) - DLGAMA(PS) - D4 - PI
C SCALING
IB = XB/ALEPS
INFSUM = 0.D0
```

```

C FIRST TERM OF A DECREASING SERIES WILL UNDERFLOW
  IF (IB.NE.0) GO TO 90
  INFSUM = DEXP(XB)
  CNT = INFSUM*DP
C CNT WILL EQUAL DEXP(TEMP)*(1.D0-PS)*P*Y**I/FACTORIAL(I)
  VH = 0.0D0
  80 VH = VH + 1.D0
  CNT = CNT*(VH-PS)*Y/VH
  XB = CNT/(DP*VH)
  INFSUM = INFSUM + XB
  IF (XB/EPS.GT.INFSUM) GO TO 80
C DLGAMA IS A FUNCTION WHICH CALCULATES THE DOUBLE
C PRECISION LOG GAMMA FUNCTION
  90 FINSUM = 0.D0
  IF (DQ.LE.1.D0) GO TO 120
  XB = PX + DQ*DLOG(1.D0-Y) + PQ - PI - DLOG(DQ) - C
C SCALING
  IB = XB/ALEPS
  IF (IB.LT.0) IB = 0
  C = 1.D0/(1.D0-Y)
  CNT = DEXP(XB-FLOAT(IB)*ALEPS)
  PS = DQ
  VH = DQ
  PI = (PS*C)/(DP*VH-1.D0)
  XB = PI*CNT
  IF (XB.LE.EPS2 .AND. PI.LE.1.D0) GO TO 120
  100 VH = VH - 1.D0
  IF (VH.LE.0.0D0) GO TO 120
  IF (PI.LE.1.D0 .AND. CNT/PS.LE.FINSUM) GO TO 120
  CNT = (PS*C*CNT)/(DP*VH)
  IF (CNT.LE.1.D0) GO TO 110
C RESCALE
  IB = IB - 1
  CNT = CNT*EPS1
  110 PS = VH
  IF (IB.EQ.0) FINSUM = FINSUM + CNT
  GO TO 100
  120 PROB = FINSUM + INFSUM
  130 IF (INT.EQ.0) GO TO 140
  PROB = 1.0 - PROB
  TEMP = P
  P = Q
  Q = TEMP
  140 RETURN
  END

```

Remark on Algorithm 419 [C2]

Zeros of a Complex Polynomial [M.A. Jenkins and J.F. Traub, *Comm. ACM* 15 (Feb. 1972), 97-99]

David H. Withers [Rec. 9 Oct. 1972 and 14 May 1973]
IBM, Essex Junction, VT 04352

The published algorithm has performed satisfactorily for all except one (degenerate) case. When removing zeros at the origin, the algorithm does not stop if all roots have been located. An error will occur if the polynomials, $X^N = 0$ or $a_N = 0$ are given to the algorithm. The difficulty may be avoided by inserting after statement 40 the statement

IF (NN.EQ. 1) RETURN

The referee pointed out the second type of degenerate case above and two typographical errors:

1. In the initialization of constants section *COSR* should be initialized by *COSR* = -.069756474.
2. In the *FUNCTIONS SCALE* and *CMOD*, the declaration of *DSQRT* as *DOUBLE PRECISION* was accidentally typed as *DSQURT*.

Remark on Algorithm 431 [H]

A Computer Routine for Quadratic and Linear Programming Problems [H] [Arunachalam Ravindran, *Comm. ACM* 15 (Sept., 1972), 818]

Arunachalam Ravindran [Recd. 12 Mar. 1973]
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A small error has been brought to my notice in this algorithm. The error is in defining the matrix M. It should read as

$$M = \begin{pmatrix} Q+Q' & -A' \\ A & 0 \end{pmatrix}.$$

Graphics and
Image Processing

W. Newman
Editor

Scan Conversion Algorithms for a Cell Organized Raster Display

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and
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Raster scan computer graphics with "real time" character generators have previously been limited to alphanumeric characters. A display has been described which extends the capabilities of this organization to include general graphics.

Two fundamentally different scan conversion algorithms which have been developed to support this display are presented. One is most suitable to non-interactive applications and the other to interactive applications. The algorithms were implemented in Fortran on the CDC6400 computer. Results obtained from the implementations show that the noninteractive algorithms can significantly reduce display file storage requirements at little cost in execution time over that of a conventional raster display. The interactive algorithm can improve response time and reduce storage requirements.

Key Words and Phrases: graphics, scan conversion, raster display, line drawing, discrete image, dot generation, matrix displays

CR Categories: 4.41, 6.35, 8.2

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