

# Mache: No-Loss Trace Compaction

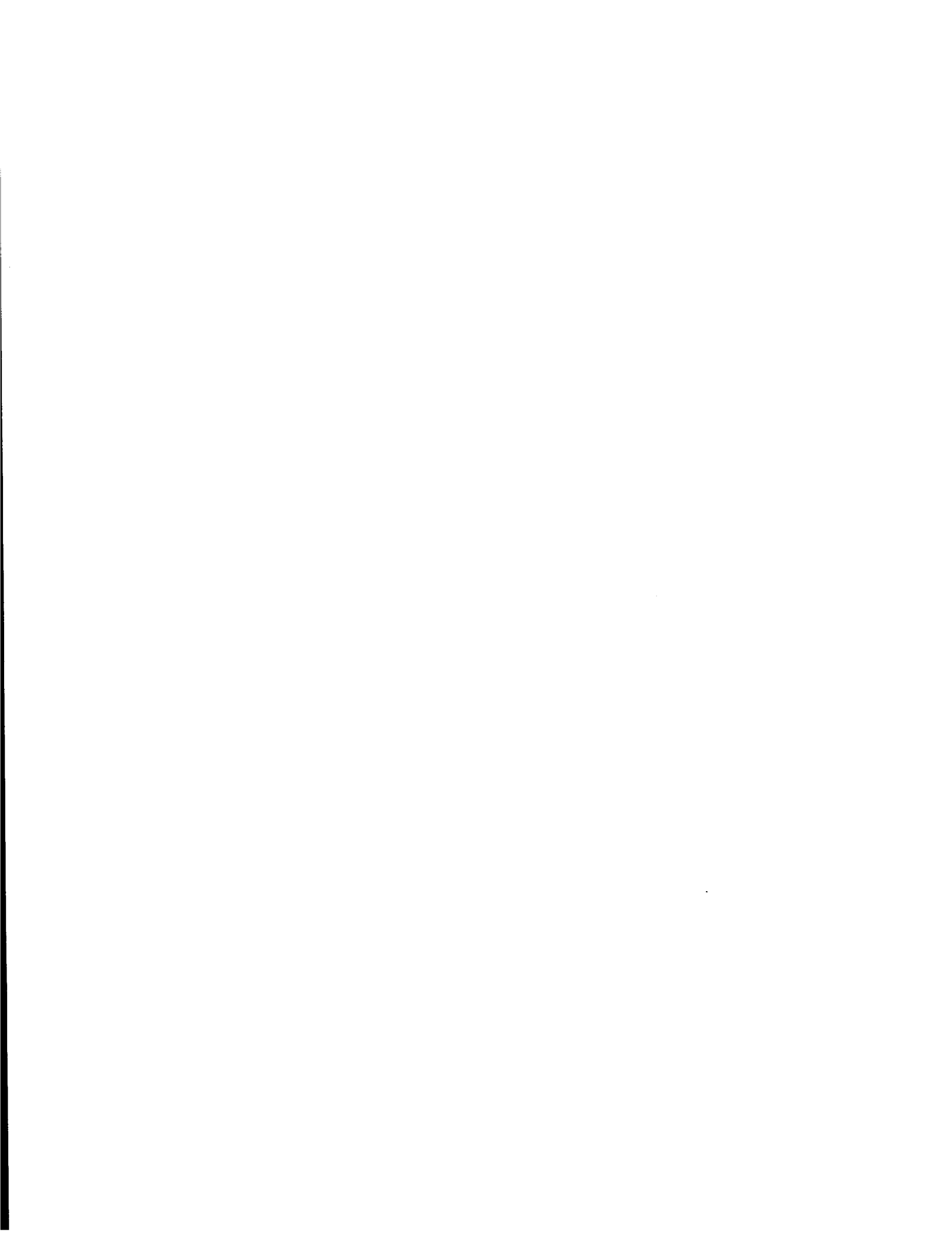
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September 15, 1988

## **Abstract**

Execution traces can be significantly compressed using their referencing locality. A simple observation leads to a technique capable of compressing execution traces by an order of magnitude; instruction-only traces are compressed by two orders of magnitude. This technique is unlike previously reported trace compression techniques in that it compresses without loss of information and, therefore, does not affect trace-driven simulation time or accuracy.



# Mache: No-Loss Trace Compaction\*

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## Abstract

Execution traces can be significantly compressed using their referencing locality. A simple observation leads to a technique capable of compressing execution traces by an order of magnitude; instruction-only traces are compressed by two orders of magnitude. This technique is unlike previously reported trace compression techniques in that it compresses without loss of information and, therefore, does not affect trace-driven simulation time or accuracy.

## 1 Introduction

Collecting and storing execution traces produces huge amounts of data that can tax the largest disk system (or, more precisely, other users of that disk system). Traces on the order of tens of millions of memory references requiring anywhere from forty to fifty megabytes are not uncommon.

In section 2, I discuss a basic method for reducing the space required to store an execution trace significantly. The technique has the following properties.

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\*This research was sponsored by the Defence Advanced Research Projects Agency (DoD), monitored by Space and Naval Warfare Systems Command under Contract N00039-88-C-0292.

<sup>†</sup>Supported in part by an AT&T Bell Laboratories Scholarship.

- No information is lost.
- Compaction is in direct proportion to the locality of the memory references in the trace.
- It is a one-pass algorithm with running time  $O(n)$ .

Section 4 examines why the technique works so well, and looks to see if any improvements are possible. While improvements are possible, the results are not sufficiently better to warrant the additional complexity.

## 2 Method

I assume that the memory traces of interest are similar to DineroIII input format [2]. Each memory reference is a pair (*label address*): the label identifies the type of reference and the address is an unsigned integer. In the memory traces used in this paper, there are three labels: a label for an instruction fetch, a label for data read, and a label for data write.

The technique described here, called *Maching* (rhymes with ‘caching’), loses no information and compresses general memory reference traces (instruction and data) by factors of ten to twenty times and instruction address traces by factors of thirty to two hundred times depending upon locality.

The basic technique is in Figure 1: when a (*label address*) pair is read, the difference  $\delta = \text{cache}_{\text{label}} - \text{address}$  is computed between this address and the last address for this label. If this difference is small enough, it is emitted in the low-order 14 bits of the codeword; otherwise the full address is emitted preceded by a ‘miss’ indicator.

The routines *emitByte*, *emit2bytes*, and *emit4bytes* pass their argument of the indicated size to a dynamic, one-pass compression scheme which takes advantage of repeating patterns in the sequence of bytes. I have used the Lempel-Ziv compression algorithm [9,10,11] implemented as the *compress* program un-

```

encode(label,address) {
     $\delta \leftarrow \text{cache}[\text{label}] - \text{address};$ 
    if ( $|\delta| \geq \text{threshold}$ ) {
        /* A 'miss'; emit the full address */
        codeByte  $\leftarrow (\text{missFlag} \ll 6) | \text{label};$ 
        emitByte(codeByte);
        emit4bytes(address);
    }
    else {
        /* A 'hit'; emit the shorter difference value */
        codeWord  $\leftarrow (\text{label} \ll 14) | \delta;$ 
        emit2bytes(codeWord);
    }
    cache[label]  $\leftarrow \text{address};$ 
}

```

10

Figure 1: The inner loop

der UNIX<sup>1</sup>. The Lempel-Ziv technique (hereafter referred to as LZ) is extraordinarily capable of detecting and exploiting the memory referencing patterns exposed by the differencing of the execution trace.

## 2.1 Traces

We have used several different execution traces of programs running to test the efficacy of *Mache* (see Table 1 for quantitative descriptions). COMP is a trace<sup>2</sup> of *compress* compressing a file of C source code (a concatenation of all the source code for a version of *Mache*). This file of source code was 90,872 bytes long and compressed to 38006 bytes. The *compress* program has a very tight inner loop for very high locality in its instruction references.

CC1 is a trace of the Gnu C compiler compiling the 303 line file *genemit.c*, which is one of the source files for the Gnu C compiler. The file contains 276

<sup>1</sup>UNIX is a trademark of AT&T Bell Laboratories.

<sup>2</sup>All traces reported in this paper were taken on a Sun workstation using the 68020.

name	# references	bytes	command
COMP	12,437,454	62,187,270	compress <sources.c >temp.Z
CC1	13,711,558	68,557,790	gcc genemit.c
VAX	13,898,315	69,491,575	vaxima <vax.in
TROFF	13,092,334	65,461,670	troff -t -Tip -rv1 paper.short
LATEX	27,221,840	136,109,200	latex todo
SCR-N	16,368,894	81,844,470	scrunch <scrunch.c
SCR-R	16,651,110	83,255,550	xscrunch <scrunch.c

Table 1: Summary of traces used

lines, is 4,664 bytes long sans comments, and has nine function definitions.

VAX is a trace of a version of Vaxima (a Macsyma-like symbolic processor) solving eight simultaneous equations in nine variables. The trace includes the reading of the input and the displaying of the answer. It was included since Vaxima is implemented in LISP, a language often charged with the sin of producing non-local memory referencing behavior.

TROFF is a trace of the troff typesetting system working on a short paper. The file is 7705 bytes long, and produces 9768 bytes of output. This trace is included because of the nature of the troff source code. Troff was more or less transliterated into C from an assembly language implementation that made heavy use of jump tables [8]. The resulting C code is peppered with *switch* statements and non-structured gotos, and has bad locality.

LATEX is a trace of the  $\text{\LaTeX}$  text processor constructing a “To Do List” form. The resulting typeset page of output has very little text and a lot of straight lines.  $\text{\LaTeX}$  is a macro package running on a Common  $\text{\TeX}$  version of Knuth’s text processing system [3].

SCR-N and SCR-R are traces of a Huffman encoder called *scrunch* compressing a 1529 line, 41,796 byte file (its own source) that is compressed to 27,345 bytes. SCR-N is the trace of a version of *scrunch* produced directly by the Gnu C compiler. SCR-R is the trace of a version of *scrunch* that has been modified to improve the locality of the code. See reference 6 for more information on this research. The important point is that these two traces show the effect of the

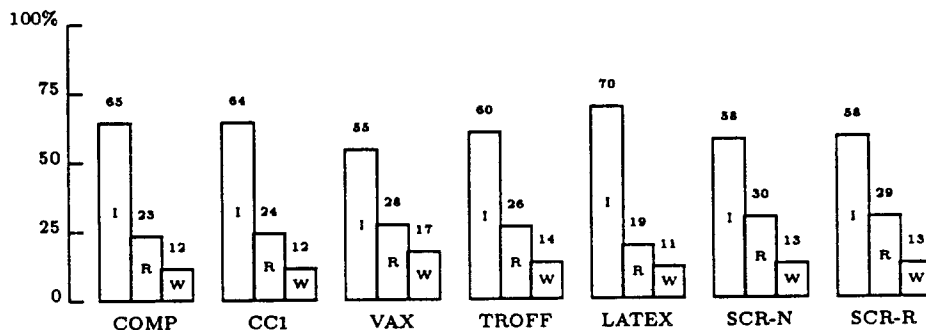


Figure 2: Breakdown of memory references in each trace

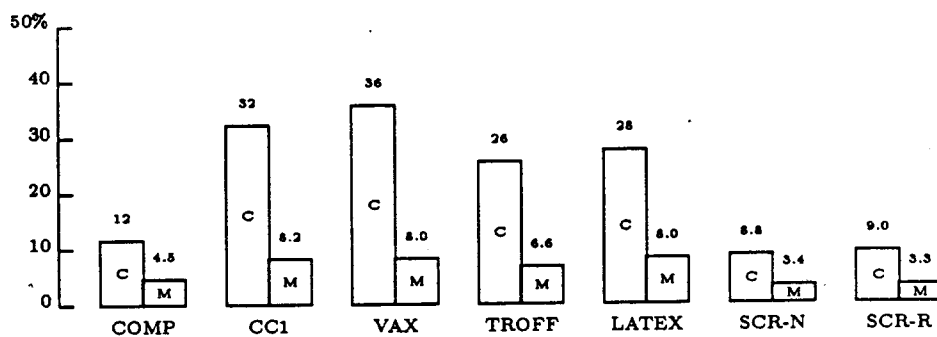


Figure 3: compress vs. Mache, percent of original trace

locality of an execution trace on *Mache*'s performance.

In Figure 2 we see that the traces are similar in their mix: 50-60% instruction references, 20-30% data reads and 15-25% data writes. The numbers for any one trace may not add to 100 due to rounding errors. In particular, SCR-N has slightly less than 30% reads, and slightly more than 58% instruction fetches, while SCR-R has more instruction fetches than SCR-N, but the same number of reads and writes.

### 3 Results

Figure 3 shows the results of compressing each trace by *compress* (C) and *Mache* (M). From these traces we see that *Mache* almost always creates a file at least

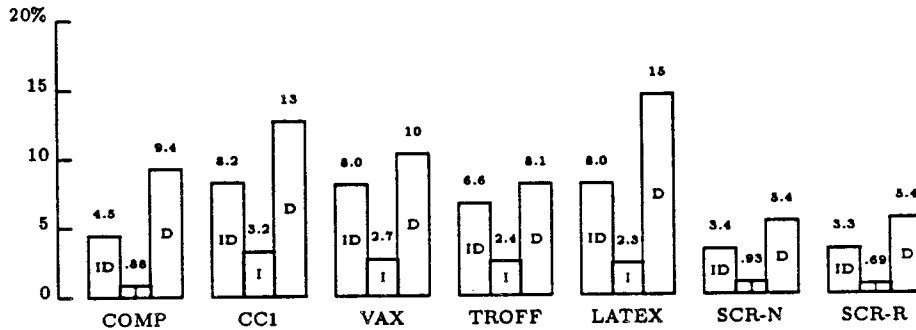


Figure 4: Compression of instruction-only and data-only traces

ten times smaller than the original trace file and over three times smaller than that produced by *compress*.

This scheme works exceptionally well for instruction-only traces. Figure 4 compares the compaction achieved on the full trace (ID), with the compaction achieved on only instruction references (I) and the data references (D). The percentages in the graph are relative to the sizes of the traces with all memory references not of interest thrown out. For example, the full trace COMP has 12,437,454 references, occupying 62,187,270 bytes in raw form. Of those references, 8,065,326 are instruction references which take 40,326,630 bytes in raw form, but only 0.88% of that (355,143 bytes) in compressed form. It is apparent that the bulk of the space savings for a complete trace is derived from the regularity in the instruction stream.

Traces SCR-N and SCR-R are of particular note since they show the sensitivity of the compression scheme to the locality of the trace. Both are traces of the same program working on the same data. SCR-N is a trace of the program as it is produced by the Gnu C compiler, and SCR-R is a trace of the program after its basic blocks and procedures were reorganized to improve the program's behavior in an instruction cache. In all other respects, the two versions of the program were the same. The difference in compression between the two traces is attributable solely to the differences in the locality of the memory references be-



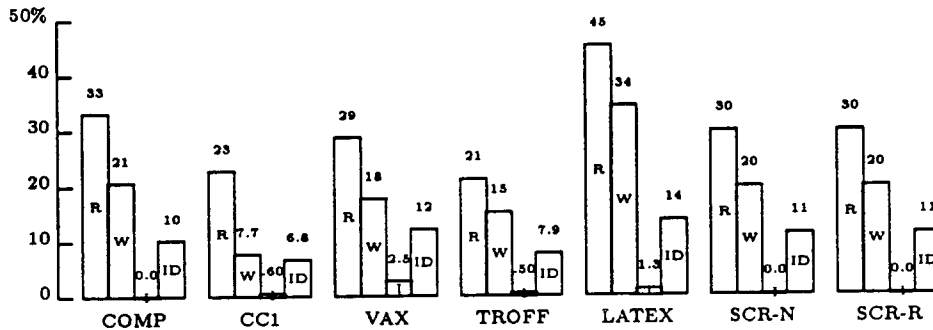


Figure 5: Miss rates

tween the two versions of the program. It is interesting to note that even though trace SCR-R contains more instructions fetched and executed than SCR-N (see Table 2), the reorganized program nevertheless executes faster and the trace compresses better due to the improved locality of the code.

In Figure 5 the 'miss rate' for each kind of reference for each trace is plotted. The miss rate is the percentage of references that require a full address to be emitted because the difference between the last and current address is too large to be encoded in a short code word. Each trace has four miss rates shown, one for each kind of memory reference (read, write, and instruction fetch) and the overall miss rate.

#### 4 Why it works

There are four reasons why this scheme works: (1) encoding differences between 32-bit addresses often requires fewer bits than encoding the addresses, (2) the differences tend to cluster much more than the original addresses, (3) the stream of differences is much more regular than the stream of addresses, and (4) the LZ compression technique works solely by discovering and encoding common sequences of bytes in the source. These attributes can be seen in Figure 6. The two columns on the left are the first few memory references from the LATEX

original trace		difference trace		comments
2	2020	3,2	2020	first instruction ref
0	efffd64	3,0	efffd64	first data read, from stack
0	efffd66	0	2	second data read
2	2022	2	2	
2	2024	2	2	
2	2026	2	2	
2	2028	2	2	
2	202a	2	2	
2	202c	2	2	
2	202e	2	2	
2	2030	2	2	
2	2032	2	2	
1	40000	3,1	40000	first data write
1	40002	1	2	
2	2034	2	2	
1	efffd60	3,1	efbfd5e	a data write 'miss'
1	efffd62	1	2	
2	2036	2	2	
1	efffd5c	1	-6	
1	efffd5e	1	2	
2	2038	2	2	
1	efffd58	1	-6	
1	efffd5a	1	2	
2	203a	2	2	
2	203c	2	2	
2	203e	2	2	
1	efffd54	1	-6	
1	efffd56	1	2	
2	2066	2	28	
2	2096	2	30	
0	efffd54	0	-12	
0	efffd52	0	-2	
2	2040	2	-56	
2	2042	2	2	
2	2044	2	2	
1	efffd54	1	-2	
1	efffd56	1	2	
2	2098	2	54	
2	209a	2	2	
2	209c	2	2	
2	2d898	3,2	2b7fc	an instruction fetch 'miss'
2	2d89a	2	2	

Figure 6: Sample trace from LATEX

trace, with the resulting stream of differences in the middle column. (The notation "3,2" means that a reference with label 2 missed.) Differencing results in more redundancy, the numbers required to represent the differences are much smaller, and there are patterns in the difference stream that are not apparent in the original. (All numbers are in hexadecimal).

Finally, the distribution of byte-values in the trace file has little to do with the final compression. For example, the number of bytes in the raw trace file CC1 is 68,557,790, and the number of bytes after differencing is 27,830,296. The entropies of the two traces are 5.40 and 3.55 respectively, implying that a Huffman encoding of the raw file would be about 46.28 megabytes, and of the difference file about 12.35 megabytes. However, these are nowhere near the 5,649,247 bytes produced by the combination of differencing and compressing.

The best compression is not achieved by looking at individual byte values and their distributions, but by the discovery and encoding of common sub-sequences exposed by differencing. 'Bit twiddling' variations based on this basic method result in negligible differences.

## 5 Refinements and Variations

I have tried several variations of the scheme presented so far to compress trace files even further. In actual fact, most of these 'variations' were tried first, and led to the conclusion that the simpler scheme presented above was the better one. I'll recapitulate the ontogeny of the technique, and then report on the results of the variations tried along the way.

The original scheme came from thinking about caches (and hence the name *Mache*). The first attempt simply coded three caches, one for instruction fetches, one for data reads, and another for data writes. Then, whenever a (*label address*) pair was read, *cache<sub>label</sub>* was examined for a slot *i* which contained the address. If such a slot were found (a 'hit'), a 16-bit word was created which encoded

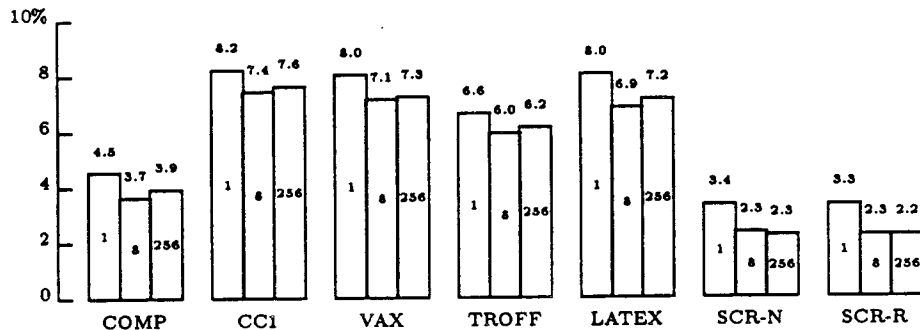


Figure 7: Matching with 1/8/256-slot caches

the label (and hence the cache), the slot number, and the offset within the slot. If it were not found (a 'miss') a header byte was emitted that encoded the miss for this label, followed by the emission of the four byte address; this stream of encoded cache-hit records is then compressed with an LZ algorithm. I tried several of the traditional replacement algorithms for caches and, not surprisingly, LRU worked best.

With a little thought came the realization that if the cache were made into a LIFO stack with LRU replacement, - i.e. when a slot was hit, it became slot zero and all the slots in between were moved down - then almost all hits would be to slot zero. This would increase the amount of redundancy in the resulting cache-hit stream. This modification produced significant improvements, and was the method of choice for some time. Note that the method presented in section 2 is this method using a cache with one slot.

But what is the optimal number of slots to produce the best compression? Figure 7 shows each of the traces *Mached* with different sizes of caches: the original 1-slot cache, an 8-slot cache, and a 256-slot cache. Apparently, it makes little difference.

To test this further, I tried compressing a single trace using different numbers of cache slots. Figure 8 is the result of *Maching* the CC1 trace using one, two, four, six, eight, ten, sixteen, thirty-two and 256 slots in the caches.

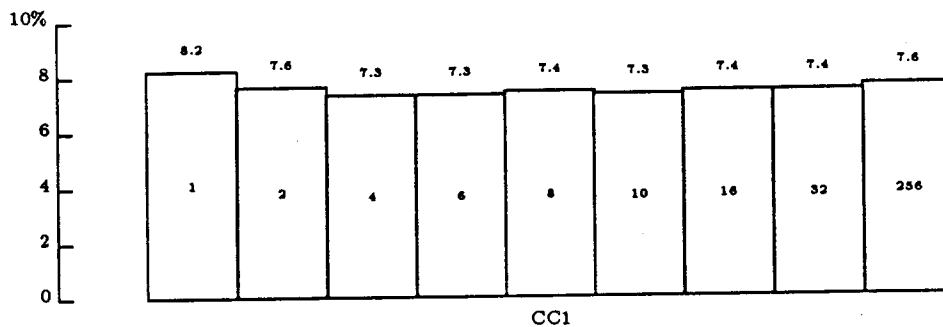


Figure 8: Matching CC1 with 1/2/4/6/8/10/16/32/256-slot caches

Even though there appears to be a minimum somewhere between six and ten slots<sup>3</sup> there is apparently not enough improvement over the 1-slot scheme (which is exactly equivalent to the ‘differencing’ method presented in section 2) to warrant the additional complexity.

There were a few other variations tried that “seemed like a good idea at the time” but which produced either worse results or negligibly different results. I will not report details on them here, but just give the bottom line. For example, the thought occurred to me that if each instruction word in memory could be given its own data cache, the memory reference patterns associated with each instruction would readily become apparent. Since not every word can be given such a cache, we approximated it by hashing the address of an instruction word into a set of 256 caches. However, either the number of caches was insufficient, or the patterns were not uncovered, because the results were surprisingly disappointing.

Another scheme used an eight-bit byte to encode hits and misses rather than the 16 bit words used in the experiments reported elsewhere in this paper. The results were negligibly different, and not always better. When using an eight-bit encoding, any differences larger than about 40 require a 32-bit word to record

<sup>3</sup>The ‘bump’ at eight slots in Figure 7 is probably attributable to vagaries of the LZ compression scheme; in particular, when the hash table in LZ fills up and the compression ratio begins to fall, LZ clears everything and starts over afresh.

the 'miss'. Such misses are now much more frequent, and account for a larger percentage of the resulting compression file.

And, finally, I split the trace out into three separate instruction, data-read, and data-write streams. This required the existence of a fourth stream to record the order of the reads, writes, and instruction fetches. The reasoning behind this said that if instruction streams could be reduced two orders of magnitude when taken by themselves, perhaps that compression combined with increased compression of the data-read and data-write streams would result in a better compression overall. Alas and alack, the fourth stream (the order stream) turns out to be pretty random already, and did not compress sufficiently to make everything come out smaller on the whole. While it was true that the sum of the sizes of the compressed instruction, data-read, and data-write streams was smaller than the size of the original compressed trace file, the size of the compressed order file was too large, and made the total file larger. Besides which, this scheme was far too complicated to recommend in good conscience.

One variation I *haven't* tried is run-length encoding: that is, after differencing, encoding the number of times the same cache-hit value occurs sequentially. This run-length encoded stream can then be passed to the LZ compression scheme. While this *may* produce impressive results for instruction-only trace files, I suspect that it will do nothing for data-only traces, and will have marginal benefit for full traces. The results presented in the next section support this conjecture.

## 5.1 Other Applications

This scheme should not be relegated to compressing only trace data. Any stream of data that contains first-order difference patterns can benefit from differencing followed by LZ encoding. For example, image bit maps, particularly those using multiple bits per pixel (e.g. color images), and analog-to-digital data that changes relatively slowly over time would be very susceptible to compression by

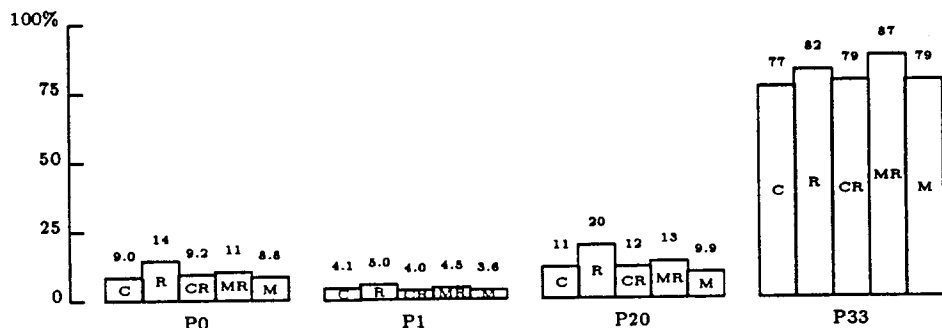


Figure 9: Mache vs. compress vs. run length encoding; percent of original file

this technique.

To show this, we have compressed four color bit-images<sup>4</sup>: P0 is a picture of the Mandelbrot set [4], and P1, P20, and P33 are magnified details of this set. The size of each image is 722,532 bytes: eight bits of color information per pixel for an 850 by 850 image. In terms of visual complexity, P1 is the least complex, and P33 is the most complex, with P20 falling somewhere in between. This is reflected in the pictures' respective compressibility. Figure 9 shows that, with the exception of P33, *Mache* (M) obtains the better compression over *compress* (C), a run length encoding (R), and a *compress* of the run length encoding (CR). *Maching* the original picture also does better than *Maching* the run length encoded version of the picture (MR).

## 5.2 Performance

It should be fairly obvious from Figure 1 that the overhead required to do the differencing is minimal: keep the last address and pass to the LZ compression routine the difference between the last and the current address. The fact that even with this overhead *Mache* is often faster than *compress* may be surprising. Since differencing produces a file that is approximately 40% of the size of the original, the LZ routine has fewer bytes as input. Given that a good portion

<sup>4</sup>Black and white versions of the pictures appear in an appendix to this paper.

of the time in LZ compression is spent hashing, shifting, *anding*, and *oring* variable length bit strings, the overhead of a subtraction and a compare or two is more than compensated by the work saved in the LZ compression routine. However, LZ works much faster when decompressing a file, so the overhead in *Mache* of caching and adding addresses represents a greater proportion of the execution time when it is decompressing. Nevertheless, the performance figures are very close. For example, on a MIPS processor, it took 408 CPU seconds to *compress* the COMP trace, but only 315 seconds to *Mache* it. To *uncompress* takes 215 CPU seconds, but to *unMache* it takes 229 seconds. (For comparison, it takes the *cmp* program over 170 seconds to compare the uncompressed 65Mb COMP trace with a copy of itself.) *Mache* compressed a smaller execution trace from 691,460 bytes to 67,583 bytes in 3.3 CPU seconds, while *compress* took it down to 235,971 bytes in 4.7 seconds. *Mache* took 2.5 seconds to recover the trace, while *compress* required 3.3 seconds.

Bottom line: *Mache* is fast.

## 6 Related Work

Most of the related work in trace compaction has concentrated on decreasing the amount of time required to do a simulation (of caches, virtual memory, etc.) as well as the amount of space required to store the cache. These compression schemes have often thrown away information that is relatively unimportant to the intended simulations. The technique reported in this paper has concentrated solely on space savings. If decreasing simulation times are important, then one of the following techniques may prove useful.

Alan Smith [7] reports on two methods for reducing the size of address traces. The first method, stack deletion, deletes the top  $k$  levels of the LRU stack. Information loss introduces an error rate in paging simulations on the order of less than 1% while reducing the trace size from 25-95%. The second



method, snapshot, reduces trace sizes by a factor of 5 to 100 with an introduced error rate similar to that for stack deletion. However, these error rates appear to be valid only for paging or cache studies using full associative placement, and do not apply to set-associative caches.

Thomas Puzak [5] reported a method he called *tape stripping* in which a direct-mapped cache with a fixed block size is simulated. Only the misses and run-length of the hits are recorded from the simulation, resulting in compaction on the order of 90-95%. He proves that there is no error introduced if the stripped trace is used to simulate caches with more sets and the same block size. He also shows that if different block sizes are used, the introduced relative error is small ( $\leq 1\%$ ) and becomes negligible as the block size becomes large. He suggests using different traces that have been stripped using the block size of interest. Of course, this requires keeping the original raw trace around, in which case *Mache* will be useful.

Anant Agarwal [1] studied cache behavior in multi-processor systems with hierarchical caches. His compaction technique can produce reductions of one to two orders of magnitude, and introduces simulation errors on the order of 10-15% in measured miss ratios due to information loss.

## 7 Conclusions

*Mache* is a simple and surprisingly effective technique for compacting execution traces. It uses differencing of memory references to expose patterns in program execution that are not immediately available in the raw trace. Using a compaction scheme that looks for and encodes long common sequences of bytes results in compressed trace files less than 10% of original size for complete traces, and on the order of 0.5% the size of the original trace for instruction-only traces.

**Acknowledgements:** The final form of this paper been improved considerably by constructive input from several people. Thanks to Paul Hilfinger,

David Wood, Charlie Farnum and especially Alan Smith for their comments.  
Any problems that remain are mine, and are probably the result of my not  
paying closer attention to their suggestions.

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The complex-plane coordinates of the lower left-hand corner of each picture can be found in the box in the upper right. 'Size' refers to the number of pixels to a side of the image on the original color display.

Pict = p000  
Real = -2.0000000000000000  
Imag = -1.2500000000000000  
Side = 2.5000000000000000  
Size = 659

Current ColorMap = bw128

Canvas Real = -2.0000000000000000  
Canvas Imag = -1.2500000000000000  
Canvas Side = 2.5000000000000000

Done

Lead	Bisp	Inv.	Note	Quit
P000	P10	P20	P1	P2
P1	P24	P12	P31	P30
P25	P11	P34	P18	P33
P29	P32	P15	P16	P17
P16				

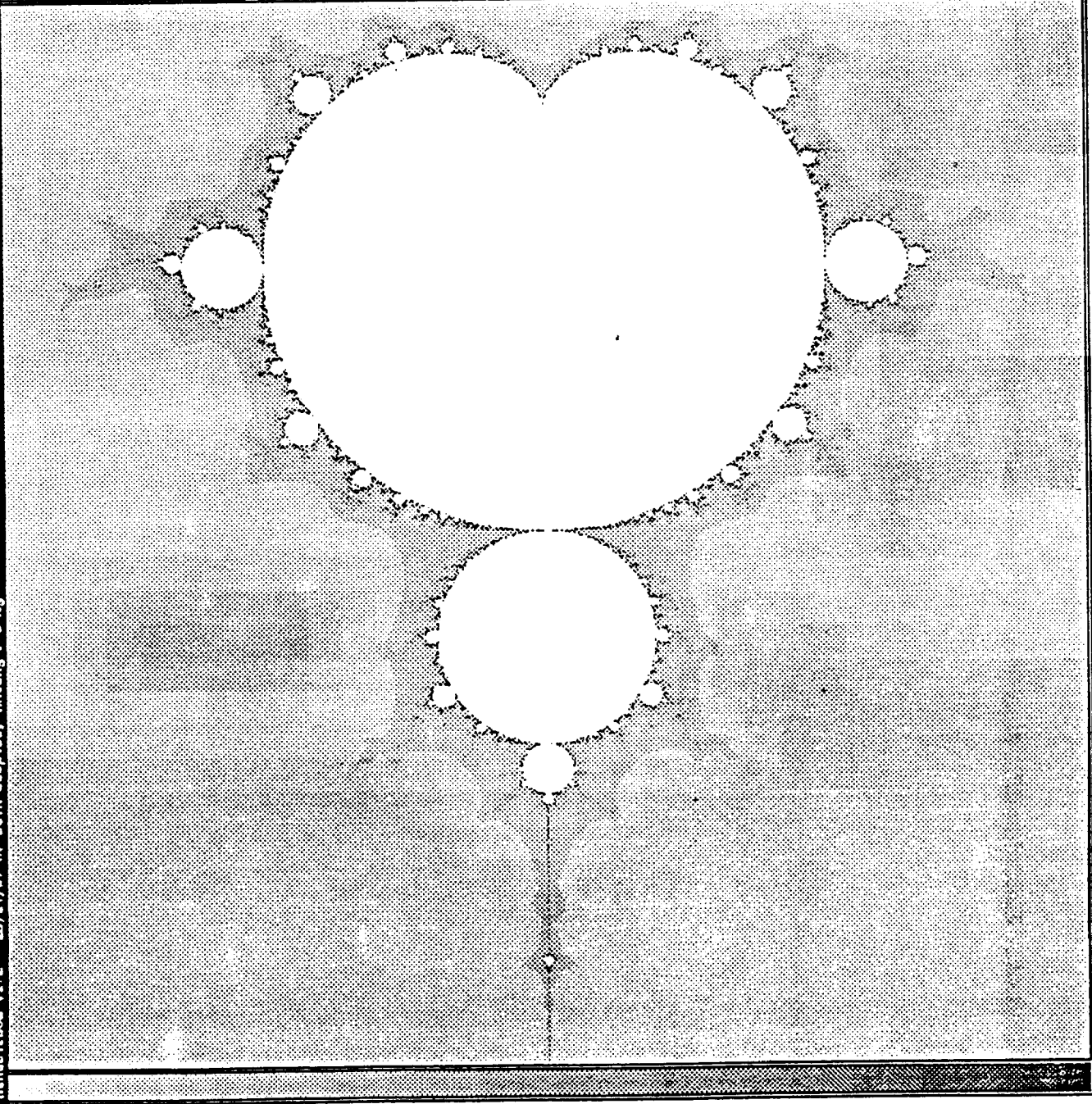
Red Green Blue

Rotate Speed:

- Default
- 1/5 Rotate
- Full Rotate
- Invert
- Map Rotate

ColorMaps

- bw000
- bw001
- bw002
- bw004
- bw016
- bw032
- bw128
- cs001
- cs002
- cs004
- cs016
- lin01
- lin02
- lin03
- lin16
- slp01
- slp02
- slp04
- slp16



Pict = p1

Real = -0.760000000000000  
Imag = 0.010000000000000  
Side = 0.020000000000000  
Size = 450

Current ColorMap = bw128

Canvas Real = -0.760000000000000  
Canvas Imag = 0.010000000000000  
Canvas Side = 0.020000000000000

Bone

Lead	Blsp	Inv.	Note	Quit
p000	p10	p20	p1	p2
p4	p28	p12	p31	p35
p25	p11	p34	p13	p33
p29	p32	p15	p14	p17
p16				

Red: GREEN: BLUE:

Rotate Speed:

1/5 Rotate

Full Rotate

Invert

Map Rotate

ColorMaps

bw000	bw001	bw002
bw004	bw016	bw032
bw128	cos01	cos02
cos04	cos16	lin01
lin02	lin03	lin16
stp01	stp02	stp04
stp16		

Pict = p20  
Real = -0.63235290000000  
Imag = -0.70000000000000  
Side = 0.65882353000000  
Size = 850

Current ColorMap = hv128  
Canvas Real = -0.63235290000000  
Canvas Imag = -0.70000000000000  
Canvas Side = 0.65882353000000

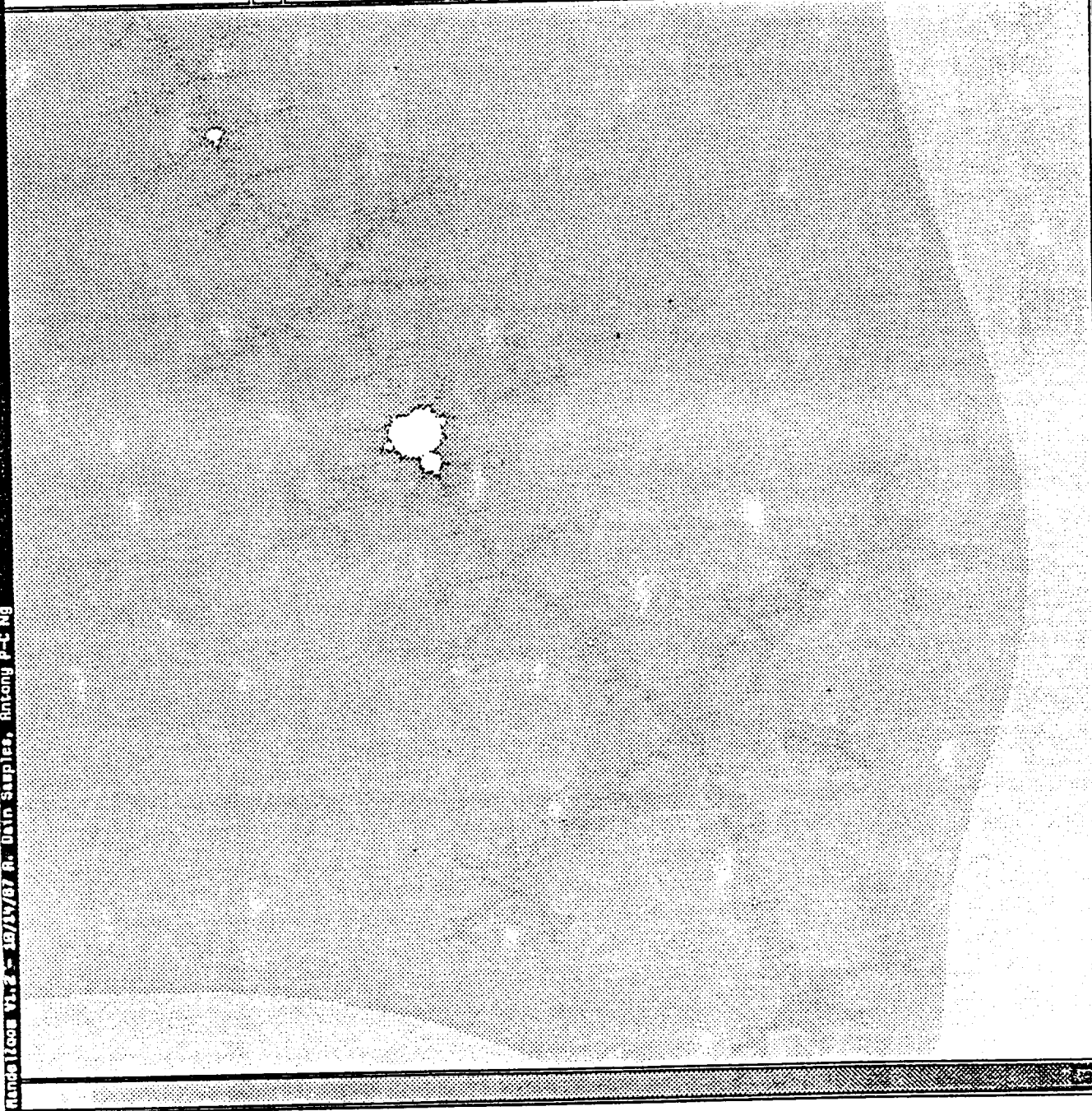
Color	Disp	Inv.	Note	Quit
p000	p10	p20	p1	p2
p4	p24	p12	p31	p30
p25	p11	p34	p13	p33
p29	p32	p15	p14	p17
p16				

Rotate Speed:  GRID  BLUE

- Default
- 3/5 Rotate
- Full Rotate
- Invert
- Map Rotate

ColorMaps

bv000	bv001	bv002
bv004	bv016	bv032
bv128	cos01	cos02
cos04	cos16	lin01
lin02	lin03	lin16
stp01	stp02	stp04
stp16		



Fact - p33  
Real - 1.255425605364  
Imag - 0.0271253702422  
Side - 0.00020905190311  
Size - 850

Current CalcMap - bw128

Canvas Real - 1.255425605364  
Canvas Imag - 0.0271253702422  
Canvas Side - 0.00020905190311

Date

Calc	Disp	Inv	Note	Quit
p00	p10	p20	p1	p2
p1	p20	p12	p31	p30
p25	p11	p34	p13	p14
p29	p32	p15	p14	p17
p16				

Rotate Speed:

Rotate Speed:

- Default
- 1/5 Rotate
- Full Rotate
- Invert
- Map Rotate

CalcMap

- bw000
- bw001
- bw002
- bw004
- bw016
- bw032
- bw128
- cos01
- cos02
- cos16
- lin01
- lin02
- lin03
- lin16
- slp01
- slp02
- slp04
- slp16