Code Compression

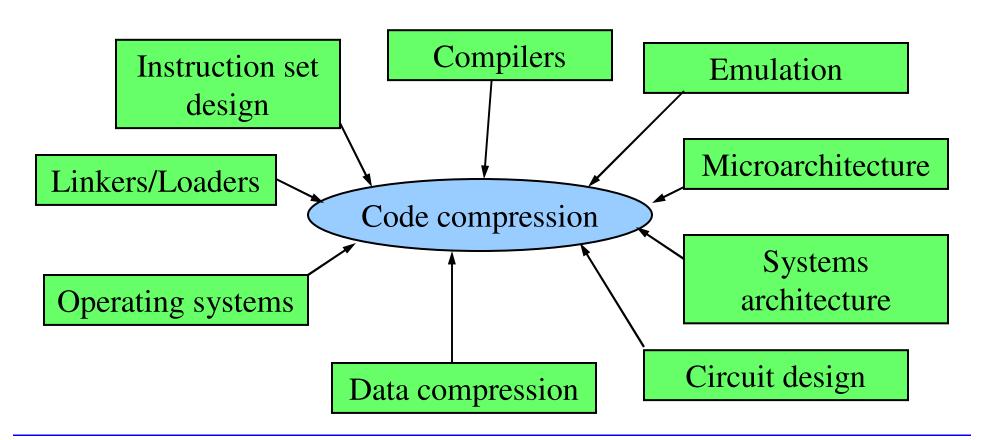
Charles Lefurgy

http://www.research.ibm.com/people/1/lefurgy

Austin Research Lab
IBM

Code Compression

- Compressing ordinary computer programs and executing the compressed form.
- Usually refers only to instruction (not data) memory



The problem

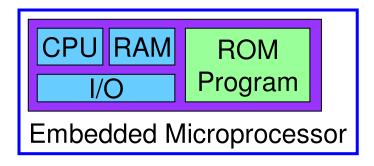
Microprocessor die cost

- Low cost is critical for high-volume, low-margin embedded systems
- Control cost by reducing area and increasing yield

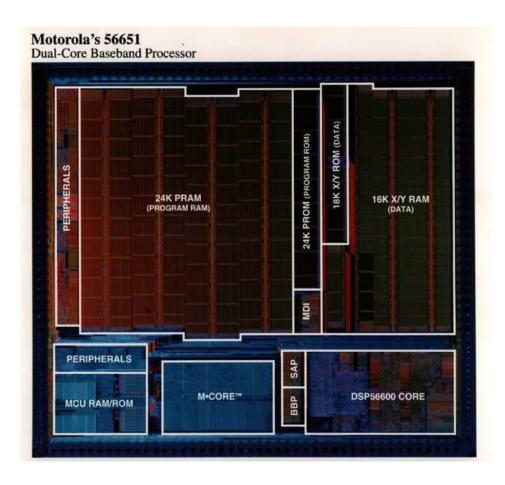
Increasing amount of on-chip memory

- Memory is 40-80% of die area [ARM, MCore]
- In control-oriented embedded systems, much of this is program memory

How can program memory be reduced?



System-on-chip

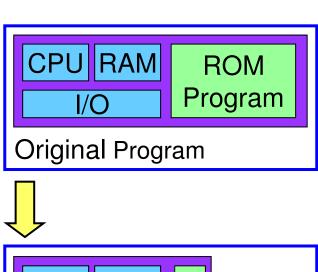




Solution

Code compression

- Reduce compiled code size
- Compress at compile-time
- Decompress at run-time









Outline

Compression methods

- Metrics
- Object code
- Gzip
- Static dictionary
- Adaptive dictionary
- Stream division

Implementations

- CCRP
- CodePack

Impact and issues

- Performance
- Energy
- Compiler optimizations

Alternatives to code compression

- Instruction set design
- Compiler optimizations
- Conclusion

Compression methods

Metrics
Object code
Gzip

Static dictionary
Adaptive dictionary (LZ)
Stream division

Metrics

Compression ratio

- Ranges from 0 to 1
- 1 is original code size

$$compression \ ratio = \frac{compressed \ size}{original \ size}$$

Execution time

Decoding efficiency

Energy

- Important for battery-operated system
- Compare to system without code compression

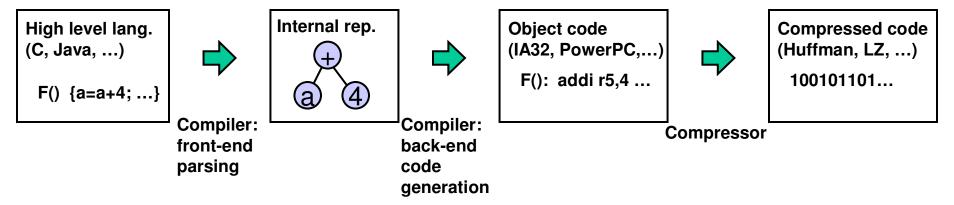
Power

- Especially for hardware implementations
- Chip cooling solution is constrained by maximum power dissipated

Code generation

Code representations:

- High level language
- Compiler internal format
- Object code



What to compress?

- This talk focuses on compressing object code.
- Compressing the high-level language and compiler formats has been proposed.

Object code

• Example: PowerPC code from ijpeg benchmark in SPEC95

| Offset Bytes Assembly code | | | | |
|----------------------------|-------------|----------|---|--|
| 51ec0 | 34 e7 ff ff | addic. | r7,r7,-1 | |
| 51ec4 | 81 83 00 18 | lwz | r12,24(r3) | |
| 51ec8 | 80 63 00 20 | lwz | r3,32(r3) | |
| 51ecc | 4d 80 00 20 | bltlr | | |
| 51ed0 | 81 04 00 00 | lwz | r8,0(r4) | |
| 51ed4 | 38 84 00 04 | addi | r4,r4,4 | |
| 51ed8 | 39 40 00 00 | li r10,0 | | |
| 51edc | 54 c9 10 3a | rlwinm | r9,r6,2,0,29 | |
| 51ee0 | 7c 8a 60 40 | cmplw | cr1, r10, r12 | |
| 51ee4 | 81 65 00 00 | lwz | r11,0(r5) | |
| 51ee8 | 38 c6 00 01 | addi | r6,r6,1 | |
| 51eec | 7d 29 58 2e | lwzx | r9, r9, r11 | |
| 51ef0 | 40 84 00 1c | bge | <pre>cr1,00051f0c <grayscale_convert+4c></grayscale_convert+4c></pre> | |
| 51ef4 | 88 08 00 00 | lbz | r0,0(r8) | |
| 51ef8 | 7c 09 51 ae | stbx | r0,r9,r10 | |
| 51efc | 39 4a 00 01 | addi | r10,r10,1 | |
| 51f00 | 7c 8a 60 40 | cmplw | cr1, r10, r12 | |
| 51f04 | 7d 08 1a 14 | add | r8, r8, r3 | |
| 51f08 | 41 84 ff ec | blt | <pre>cr1,00051ef4 <grayscale_convert+34></grayscale_convert+34></pre> | |
| 51f0c | 34 e7 ff ff | addic. | r7,r7,-1 | |
| 51f10 | 40 80 ff c0 | bge | 00051ed0 <grayscale_convert+10></grayscale_convert+10> | |
| 51f14 | 4e 80 00 20 | blr | | |
| | | | | |



Data to be compressed.

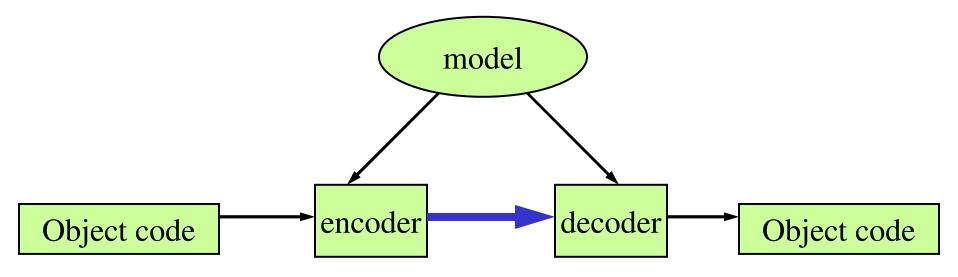
Data compression

Model

- What are the symbols in the input? (instructions, fields, bytes, etc.)
- What are their frequencies? (Fixed or varying?)

Encoder/Decoder

- How to encode a single symbol?
 - Most common symbols have the shortest codes
- Example: Huffman



Transmit compressed object code

Why not just use gzip?

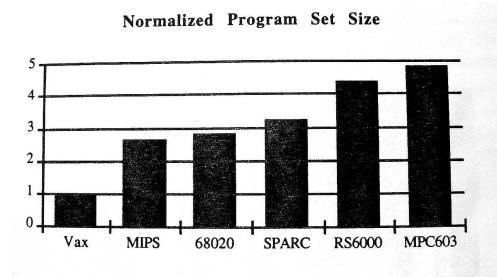
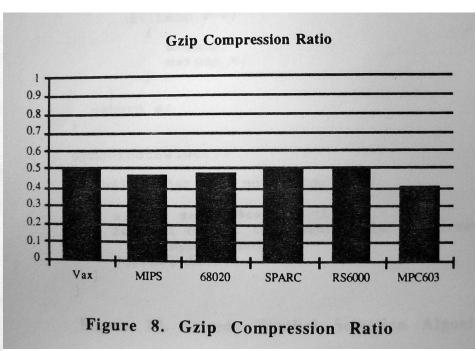


Figure 4. Sum of Program Sizes for Each Machine (Normalized to the VAX 11/750)



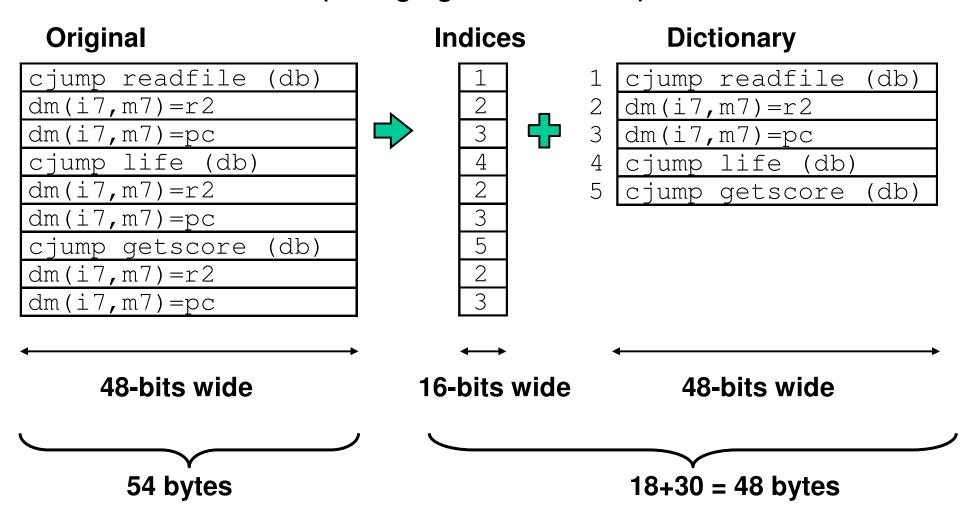
[Kozuch & Wolfe, Int. Conf. on Computer Design, 1994]

Data compression assumptions

| | For generic data | For computer programs | | |
|---------------------------|---------------------------------------|--|--|--|
| Туре | Lossless or lossy. | Lossless. | | |
| Data length | Possibly infinite. | Finite. | | |
| Number of passes | Single. | No restrictions. | | |
| Input context | Long. | Short (< 1000 bytes). | | |
| Decompression entry point | From beginning only. | From any instruction or function boundary. | | |
| Code alignment | Bit-aligned. | Probably word-aligned for fast decoding. | | |
| Compression speed | Important for real-time applications. | Not important. Done at compile time. | | |
| Data content | Use original data. | May apply code optimizations that result in better compression. (e.g. register allocation) | | |
| Example: | Gzip | CodePack | | |

Example of dictionary compression

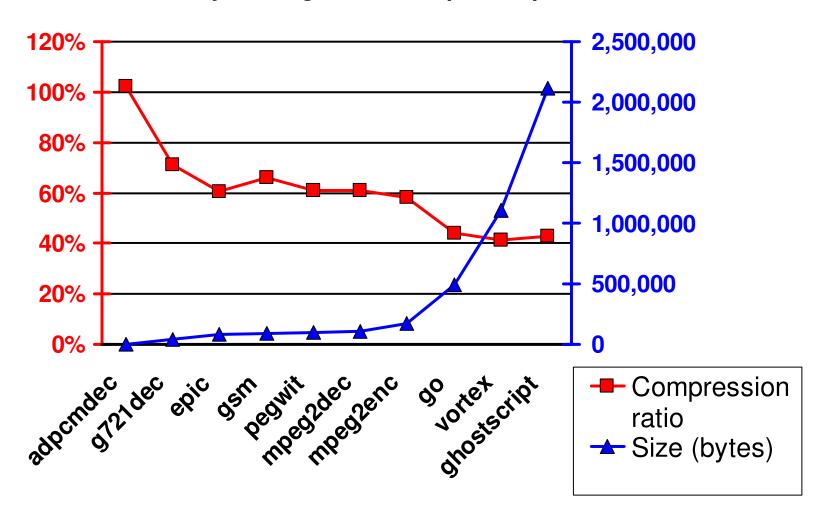
ADI SHARC DSP code. (from go:g2.c in SPEC 95)



Compression ratio = 48/54 = .89

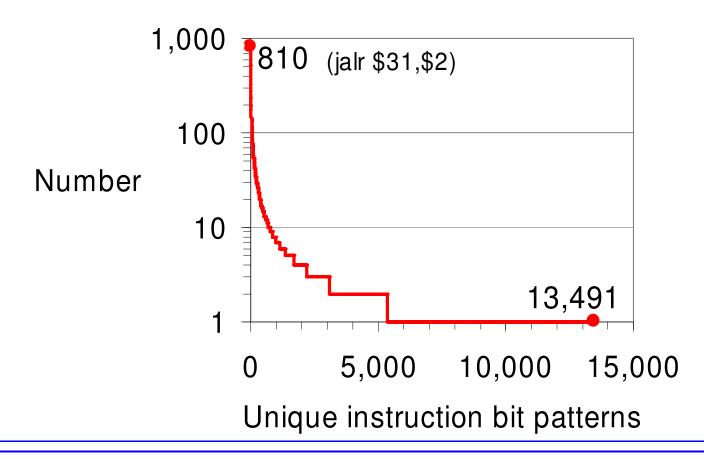
SHARC Experiments

- Dictionary compression applied to SHARC DSP programs
- Instructions are 6 bytes long. Contain up to 3 operations.



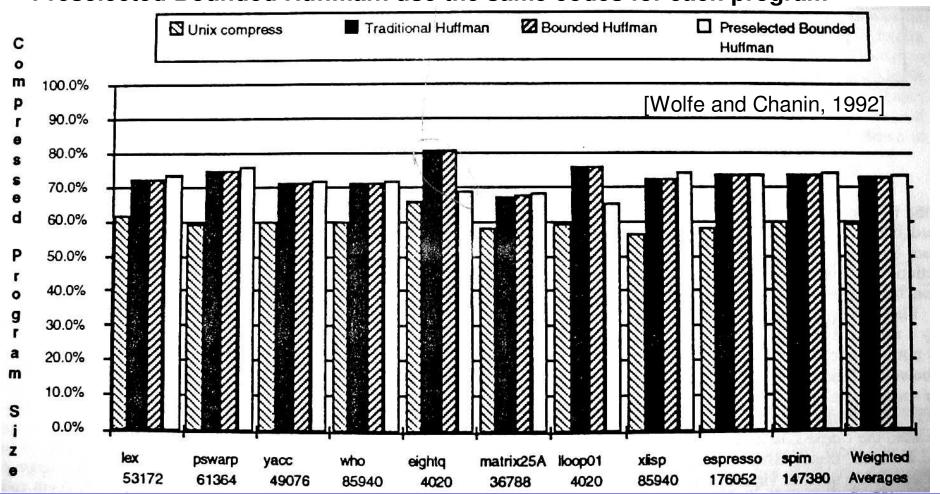
Instruction-based dictionary compression

- **ljpeg benchmark** (MIPS gcc 2.7.2.3 -O2)
 - 49,566 static instructions
 - 13,491 unique instructions
 - 1% of unique instructions cover 29% of static instructions



Byte-oriented Huffman compression

- Symbols are 8-bit bytes
- Bounded Huffman: limit codes to 16-bits max
 - Use escape code to encode original byte if code is longer than 16 bits.
- Preselected Bounded Huffman: use the same codes for each program



From bytes to fields

MIPS instruction format

- 32-bit fixed-length instructions
- 3 types of instructions
- Fields do not align to byte boundaries
 - Poor for 1-byte Huffman encoding

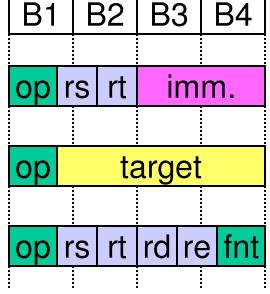
I-Type

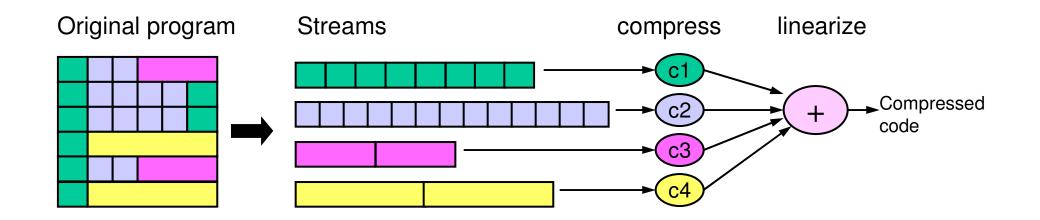
R-Type

J-Type op

Stream compression

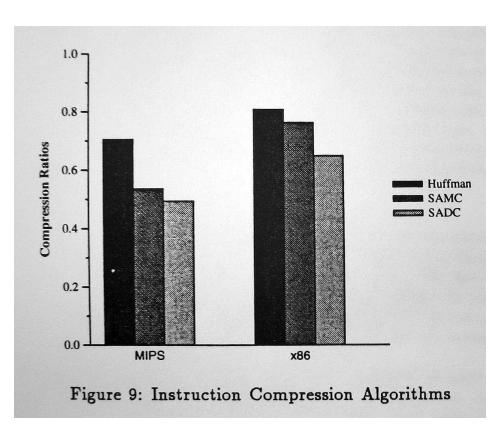
- Compress each field type separately
- Improve similarity between symbols





Semiadaptive Dictionary Compression

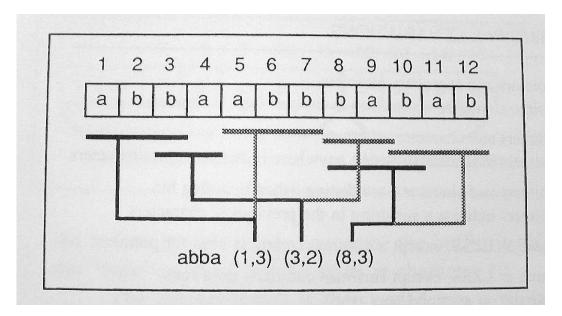
- Example of a higher-order model for code compression
- SADC achieves 50% compression ratios
 - Divide MIPS instruction into streams for each instruction field.
 - Opcode
 - Register
 - Immediate
 - Long immediate
 - Markov model for next-bit probabilities.
 - Use arithmetic coding on each stream.
 - Opcode dictionary to encode frequently used sequences of opcodes.
 - Semi-adaptive: probabilities and dictionary are different for each program.



[Lekatsas & Wolf, 1998]

Lempel-Ziv: adaptive dictionaries

- Encode several symbols at a time
 - Create dictionary of recently seen strings of symbols
- Use sliding window of recent input to find matching strings
 - Assumes that next symbols will look similar to ones recently seen.
 - Automatically adapts as symbol frequencies change
 - Larger window (context) yields better compression
- Basis for popular compression programs: pkzip, gzip, etc.



Dictionary

| abb | (1,3) |
|-----|-------|
| ba | (3,2) |
| bab | (8,3) |
| ••• | |

[Bell, Cleary, and Witten, *Text Compression*]

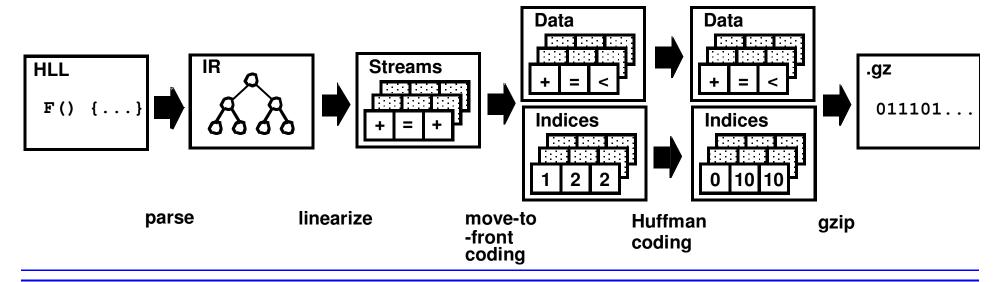
"Wire code"

Overview

- [Ernst et al., 1997]
- Wire codes
- Compress compiler representation

Results

- 1/5 size of SPARC program
- Good for sending code over a network
- Must decompress and compile using just-in-time compiler



Implementations

General issues

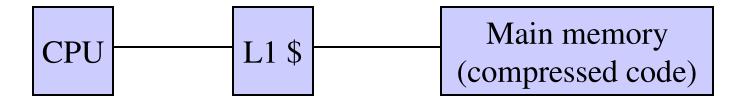
CCRP: widely studied method

CodePack: a commercial solution

Issues

Where to decompress?

- Between memory and L1 cache. (Focus of this talk)
 - Part of the memory system. Invisible to core processor.
 - Code in the cache can execute without more decompression.
 - Improves CPU performance.
- Between L1 cache and instruction execution.
 - Part of core processor. Decoder must be modified.
 - Instructions must be decompessed each time they are executed.
 - · Fewer off-chip accesses.
 - Improves memory performance.



More issues

Blocking: unit of decompression

- Large blocks allow for more context and better compression
- Large blocks slow execution
 - Jumps into middle of block: must decompress first instructions
 - May jump out of block before reaching the end. Decompress unused instructions.

Should dictionary be different for each program?

- Smaller compressed size
- Adapts better to each compiler and program.
- Could be a barrier to wide adoption. Must re-load decoder to decompress the next program.

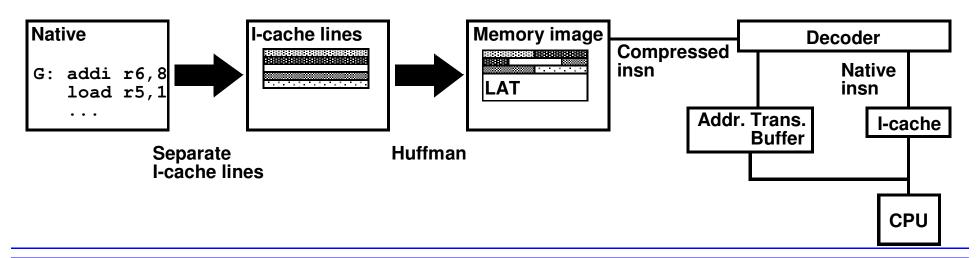
CCRP: compressed code RISC processor

Overview

- [Kozuch and Wolfe, 1994]; [Benes et al. 1998]
- Compressed Code RISC Processor (CCPR)
- Huffman encode cache lines
- Address translation for random access to cache lines.
 - LAT: line address table
- Programs run from -10% to +30% faster than conventional system.
 - Faster when memory is slow or instruction cache miss ratio is high.

Results

- 73% compression ratio for MIPS
- 0.8μ CMOS, 0.75 mm², decompression output 163 MB/s



CCRP address translation

LAT: line address table

- Input: a program address
- Output: the corresponding compressed code address

LAT entry (8 bytes)

- Encodes 8 cache lines. 24-bit base address, 5 bit offsets.
- Base address: first address of the compressed block
- L1...Lk: offset to compressed cache line
- Address of nth cache line = base + L1 + ... + Ln
- LAT overhead is 3% of compressed code.

| Base Addr. | L ₁ | L ₂ | L ₃ | | L ₈ |
|------------|----------------|----------------|----------------|--|----------------|
| LAT Entry | | | | | |

| k | 1 | k | 2 | b | 3 | k | 4 |
|---|-----|-----|----------|----|----|----|----------|
| | Li | n e | 1 | | | Li | - |
| n | e 2 | 2 | | Ä | ne | 3 | } |
| | | | Lir | пe | 4 | | |
| | | | | | | | |

Byte-aligned compressed cache lines

CodePack

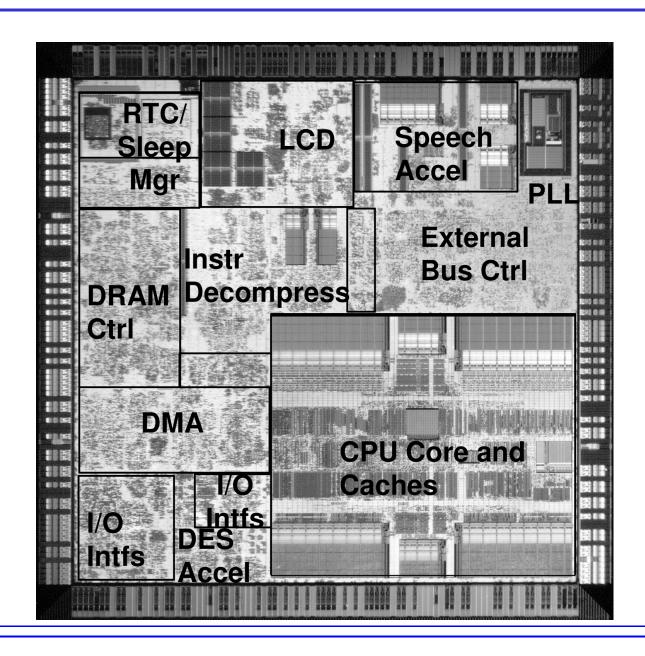
Overview

- The only widely-deployed code compression method
- IBM
- PowerPC instruction set
- 60% compression ratio, ±10% performance [IBM]
 - · performance gain due to prefetching

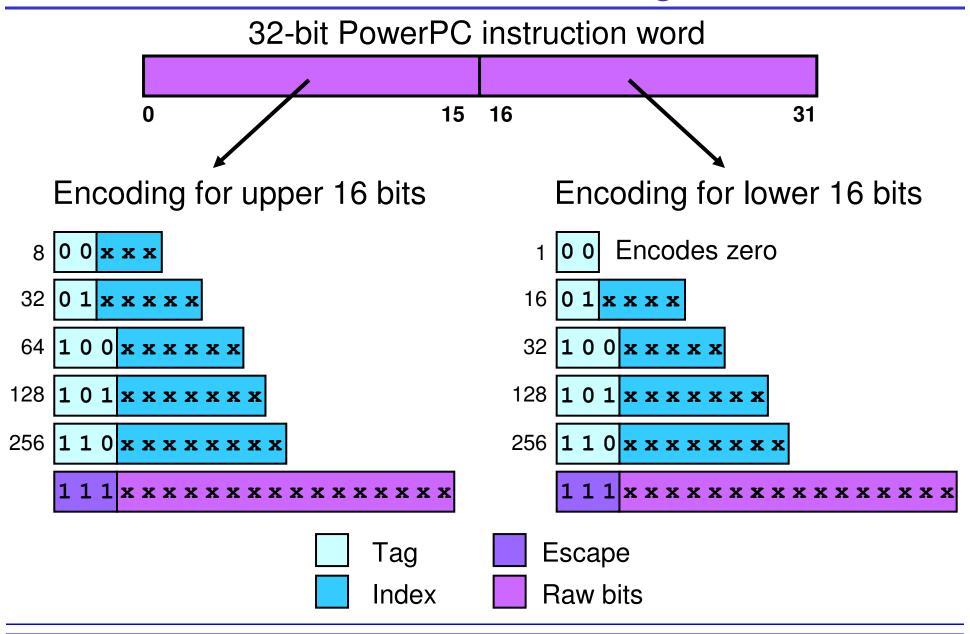
Implementation

- Binary executables are compressed after compilation
- Compression dictionaries tuned to application
- Decompression occurs on L1 cache miss
 - L1 caches hold decompressed data
 - Decompress 2 cache lines at a time (16 insns)
- PowerPC core is unaware of compression

PowerPC 405 LP



CodePack encoding



CodePack system

CodePack is part of the memory system

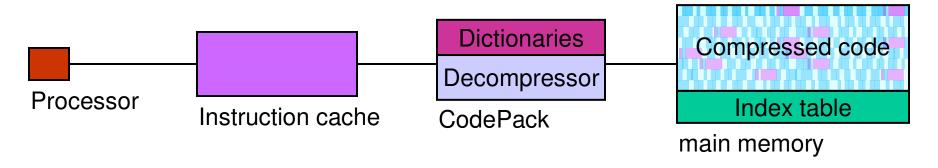
After L1 instruction cache

Dictionaries

Contain 16-bit upper and lower halves of instructions

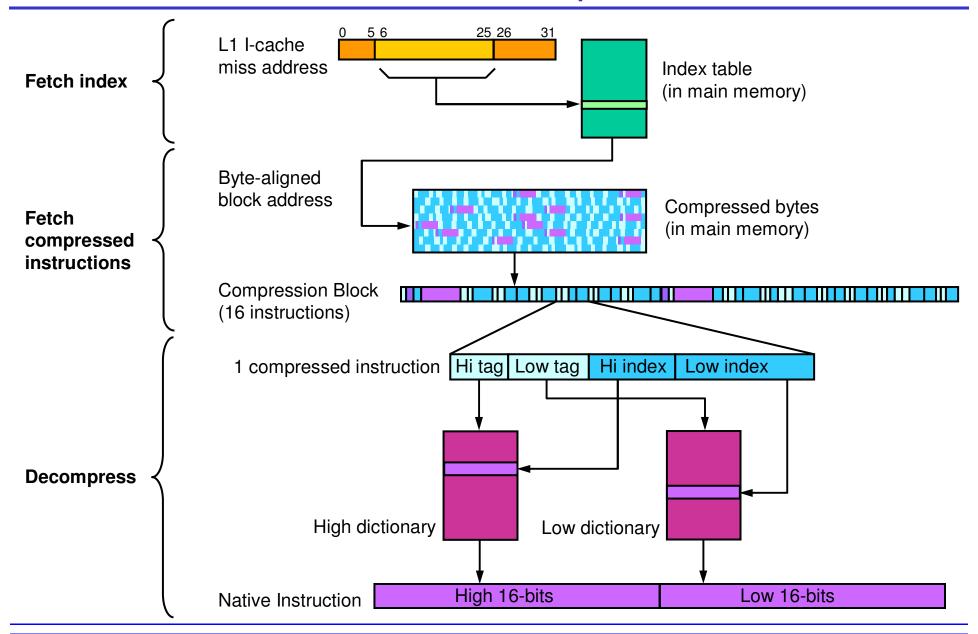
Index table

Maps instruction address to compressed code address



Instruction memory hierarchy

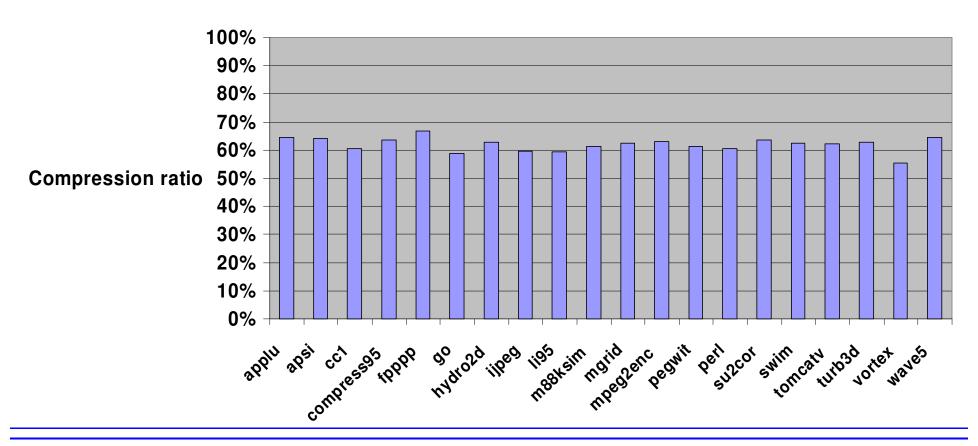
CodePack decompression



Compression ratio

•
$$compression\ ratio = \frac{compressed\ size}{original\ size}$$

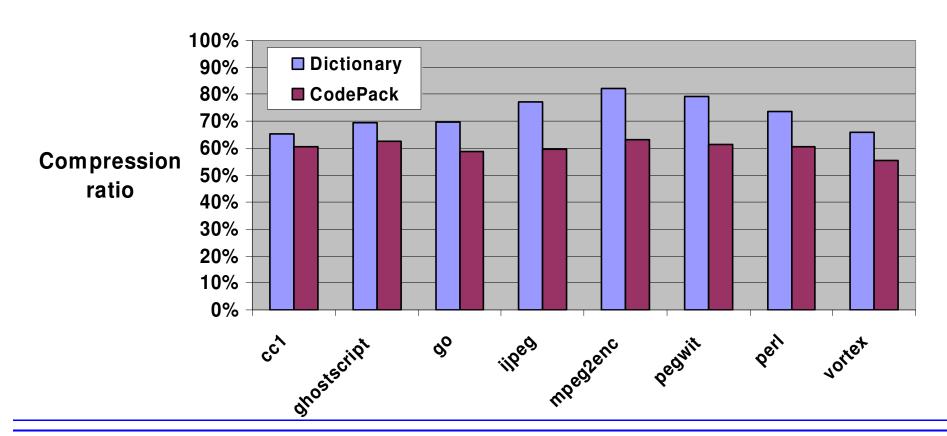
Average: 62%



Compression ratio

• $compression\ ratio = \frac{compressed\ size}{original\ size}$

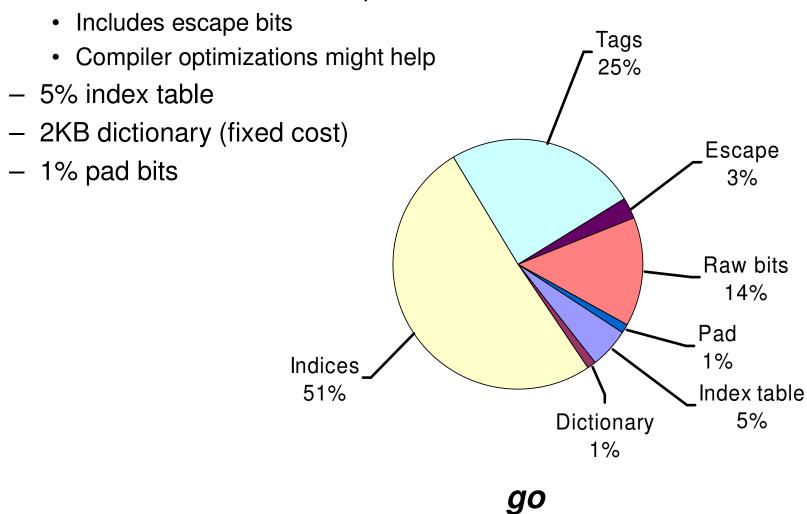
CodePack: 55% - 63%Dictionary: 65% - 82%



CodePack programs

Compressed executable

— 17%-25% raw bits: not compressed!



Impact and Issues

Performance
Energy
Compiler optimization

Can code compression improve performance?

Evidence both ways.

Yes:

- Fewer main memory accesses required.
- Less swapping, less use of overlays, etc.
- Loading compressed code from disk and compiling it can be faster than loading native code.
- If compressed instructions can be stored in cache, then caches are effectively bigger.

No:

- Decode time can increase latency of executing instruction
 - Compressed instruction in L1 cache must be decoded each time they are executed.
- Increased cache miss latency (CodePack and CCRP)

Can code compression save power?

Many studies, but no definitive answers.

Results are simulated, not measured on real hardware.

Yes:

- Less data is transmitted over memory bus: less bit flips.
- Less memory is required.
- Less memory accesses.
- Narrower memory bus can be used.
- If code runs faster, power-down modes can be used more often.

No:

- Slowdown causes CPU and peripherals to stay in power-up mode longer.
- Time for program to complete has a first-order impact on energy used.
 - CPU energy cost overwhelms any gain in memory/bus energy.

Compiler optimizations for code compression

Example: Instruction selection

- Repetition improves compression
- Choose PC-relative or absolute branches for similarity
- Improves compression ratio by over 10% for Spec95
- Reduces dictionary size by 50% for some benchmarks
- Removes many "singleton" instructions

PC-relative branches to same target cause different instruction words

80d4: e59f0010 ldr r0,&"hello" 80d8: eb000237 bl 89bc <printf> 80dc: e59f000c ldr r0,&"goodbye" 80e0: eb000235 bl 89bc <printf>



Using absolute addressing makes instruction words the same and compressible

```
80d4: e59f0010 ldr r0,&"hello"
80d8: eb0089bc bla 89bc <printf>
80dc: e59f000c ldr r0,&"goodbye"
80e0: eb0089bc bla 89bc <printf>
```

Alternatives to code compression

New instruction sets Compiler optimizations

Alternatives to code compression

New instruction set

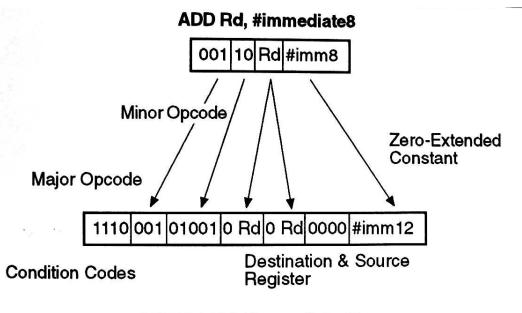
- ARM → Thumb
- Smaller, but could still compressed more.

Compiler optimization for small code-size

- Limited effect on code size. 10% is typical.
- Procedure abstraction

Thumb

- Thumb instruction set is based on ARM.
 - Processor can switch between ARM and Thumb instruction sets.
- 16-bit instructions (ARM is 32-bit)
- 8 32-bit general registers (ARM has 16)
- Destructive (2 register) instructions
- Load/store architecture
- Removed instructions
 - Multiply-accumulate
 - Atomic memory operations
 - Reverse subtract
 - Co-processor operations
 - Conditional Execution
 - In-line shifts



ADD Rd, Rd, #immediate12

[Microprocessor Report, 1995]

Figure 1. The Thumb instruction decompression logic expands opcodes and register references into their 32-bit ARM equivalents.

Thumb performance

- Compression ratio = 0.7
- Runs faster narrow busses.
 - Instructions can be read with fewer memory accesses
- Runs slower on wide busses.
 - 15-20% more dynamic instructions are executed.

Hybrid programs

- Use Thumb for infrequently used functions. (Most of the program.)
- Use ARM for the few performance-critical functions.
- Best compilers help you decide how to trade-off code size and performance.

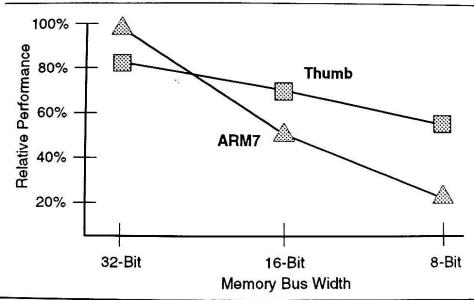


Figure 2. Comparing ARM7 cores with and without Thumb shows the effects of memory bus width on performance. The 16-bit Thumb instruction set surpasses ARM in low-cost systems. (Source: ARM).

© 1995 MicroDesign Resources

[Microprocessor Report, 1995]

Compiler optimizations for small code

Procedure abstraction [Standish, 1976]

- Use function call mechanism to abstract common code
- Apply to source code, compiler IR, or object code

```
// Count 2 lists
                                      // Count 2 lists
                                      int G(p) {
F() {
 total = 0;
                                        total = 0;
  while (a_ptr) {
                                        while (p) {
    total++;
                                          total++;
    a_ptr = a_ptr->next;
                                          p = p->next;
  a = total;
                                        return(total);
  total = 0;
  while (b_ptr) {
                                      F() {
    total++;
                                        a = G(a ptr);
    b_ptr = b_ptr->next;
                                        b = G(b ptr);
  b = total;
```

Conclusions

Does code compression help size?

Yes. 30-50% reduction for object code. 80% reduction for compiler IR + JIT.

Does code compression help performance?

- Possibly, in the right situations. (slow memory, narrow bus)
- Often decompression step causes systems to run slower.
- Hybrid programs (compressed and native code) can reduce performance impact.

Does code compression help energy consumption?

- Helps memory and bus power (fewer accesses)
- May not help full system power. Remains to be demonstrated.

Future?

- No new industrial solutions in last few years.
 - But still new ISAs. Thumb-2 can mix 16-bit and 32-bit instructions freely.
 - Larger register sets (IA-64, MMX, Vectors)
- Sensors networks. An ideal application?
- Cell phones. Ever smaller with more features.
- DRAM scaling is slowing: 4x/3years $\rightarrow 2x/3$ years (driven by PC market)
- Main memory compression (compress/decompress in memory controller)

References

- A. Beszédes et al., "Survey of Code-Size Reduction Methods", ACM Computing Surveys, Vol. 35, No. 3, September, 2003, pp. 223-267.
 - Code compaction and compression.
- "Software Techniques for Program Compaction", Guest editors B. De Sutter and K. De Bosschere, *Communications of the ACM*, Vol. 46, No. 8, August, 2003, pp. 33-34.
- The Code Compression Bibliography http://www.iro.umontreal.ca/~latendre/codeCompression/
 - 150+ citations.
- T. C. Bell, I. H. Witten, and J. G. Cleary, *Text Compression*, Prentice Hall, 1990.
 - Good reference on non-lossy data compression.

End