

# Fast Searches for Effective Optimization Phase Sequences

Prasad Kulkarni, Stephen Hines, Jason Hiser David Whalley, Jack Davidson, Douglas Jones

- Computer Science Department, Florida State University, Tallahassee, Florida
- Computer Science Department, University of Virginia, Charlottesville, Virginia
- Electrical and Computer Eng. Department, University of Illinois, Urbana, Illinois





### Phase Ordering Problem

- A single ordering of optimization phases will not always produce the best code
  - different applications
  - different compilers
  - different target machines
- Example
  - register allocation and instruction selection





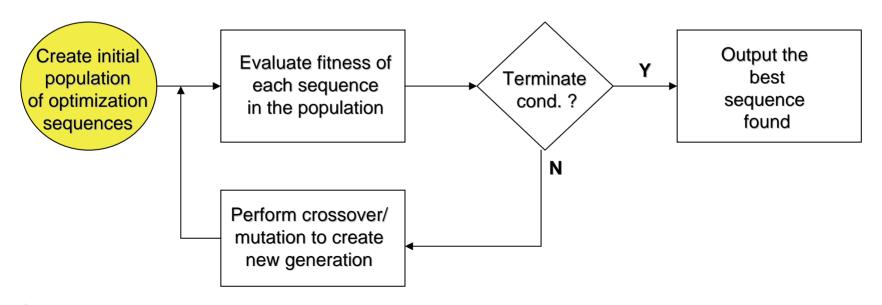
# Approaches to Addressing the Phase Ordering Problem

- Framework for formally specifying compiler optimizations.
- Single intermediate language representation
  - repeated applications of optimization phases
- Exhaustive search?
- Our approach
  - intelligent search of the optimization space using genetic algorithm





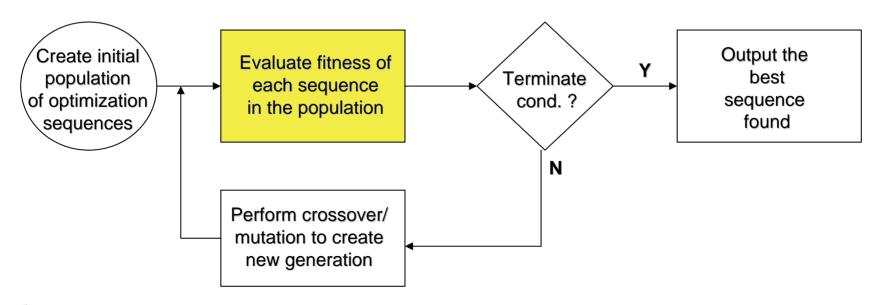
- A biased sampling search method
  - evolves solutions by merging parts of different solutions







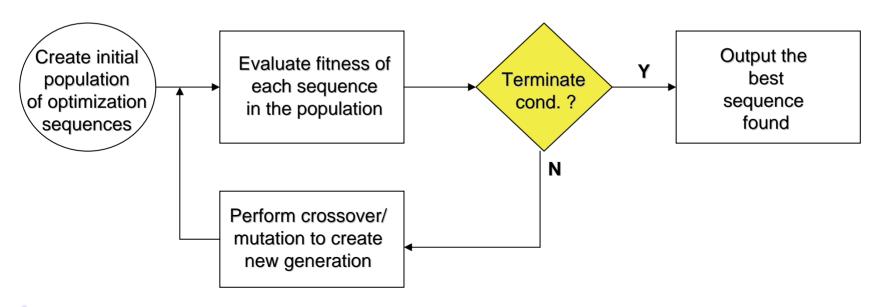
- A biased sampling search method
  - evolves solutions by merging parts of different solutions







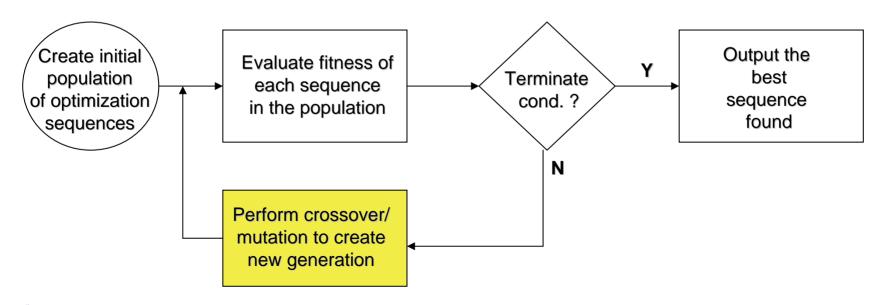
- A biased sampling search method
  - evolves solutions by merging parts of different solutions







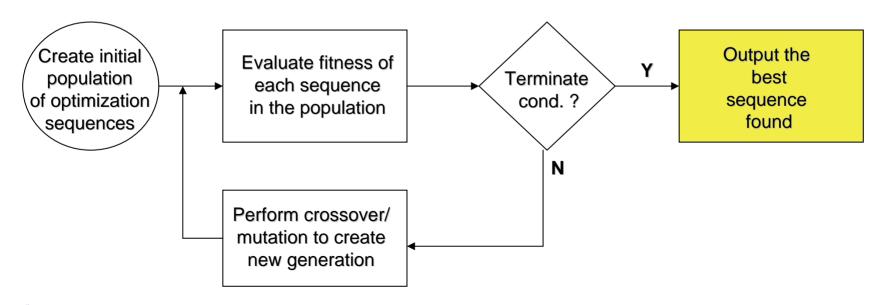
- A biased sampling search method
  - evolves solutions by merging parts of different solutions







- A biased sampling search method
  - evolves solutions by merging parts of different solutions

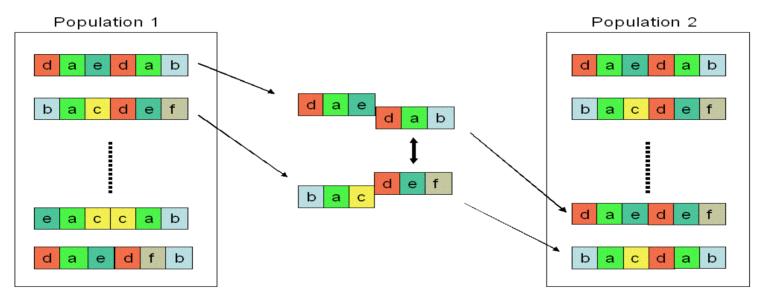






### Genetic Algorithm (cont...)

- Crossover
  - 20% sequences in each generation replaced

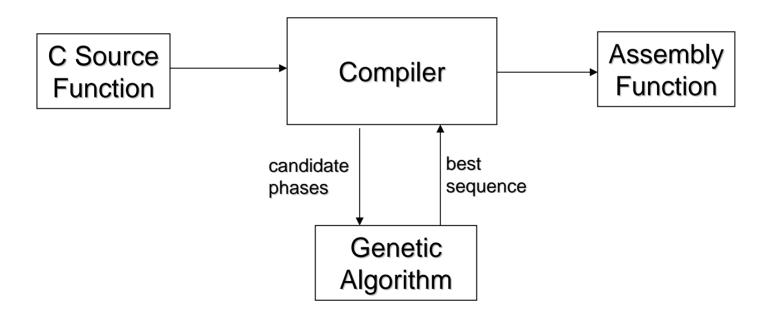


- Mutation
  - phases in each sequence replaced with a low probability





### Genetic Algorithm (cont...)







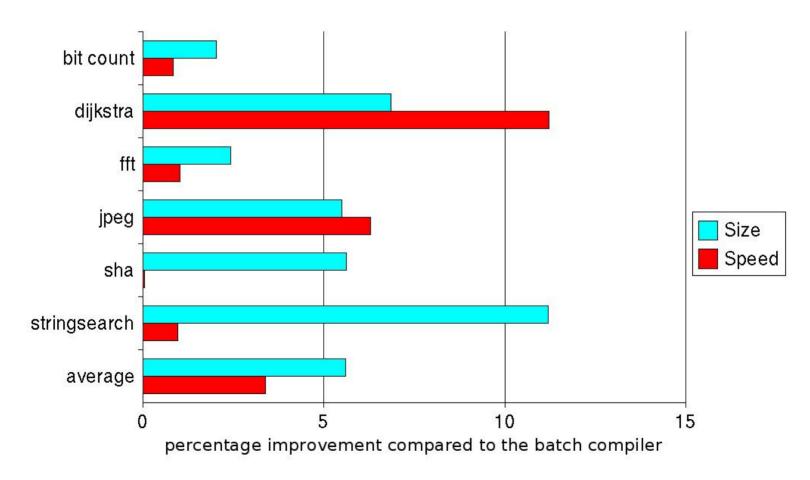
### Experiments

- Performed on six mibench benchmarks, which contained a total of 106 functions.
- Used 15 candidate optimization phases.
- Sequence length set to 1.25 times the number of successful batch phases.
- Population size set to 20.
- Performed 100 generations.
- Fitness value was 50% speed and 50% size.





### Genetic Algorithm — Results







### Our Earlier Work

- Published in LCTES '03
  - complete compiler framework
  - detailed description of the genetic algorithm
  - improvements given by the genetic algorithm for code-size, speed, and 50% of both factors
  - optimization sequences found by the genetic algorithm for each function
  - Finding Effective Optimization Phase Sequences –
     http://www.cs.fsu.edu/~whalley/papers/lctes03.ps





### Genetic Algorithm – Issues

- Very long search times
  - evaluating each sequence involves compiling, assembling, linking, execution and verification
  - simulation / execution on embedded processors is generally slower than general-purpose processors
- Reducing the search overhead
  - avoiding redundant executions of the application.
  - modifying the search to obtain comparable results in fewer generations.





### Methods for Avoiding Redundant Executions

- Detect sequences that have already been attempted.
- Detect sequences of phases that have been successfully applied.
- Check if an instance of this function has already been generated.
- Check if an equivalent function has already been generated.





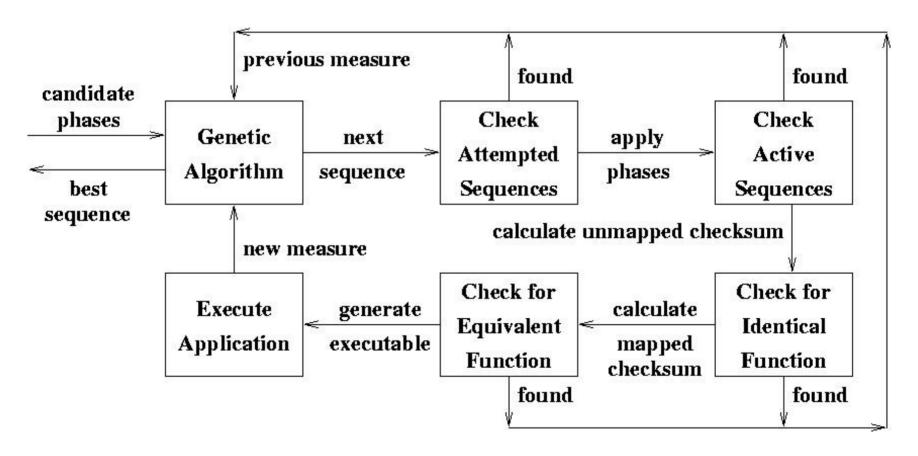
### Reducing the Search Overhead

- Avoiding redundant executions.
- Obtaining similar results in fewer generations.





## Overview of Avoiding Redundant Executions







# Finding Redundant Attempted Sequences

- Same optimization phase sequence may be reattempted
  - Crossover operation producing a previously attempted sequence
  - Mutation not occurring on any of the phases in the sequence
  - Mutation changing phases, but producing a previously attempted sequence





# Finding Redundant Attempted Sequences (cont...)

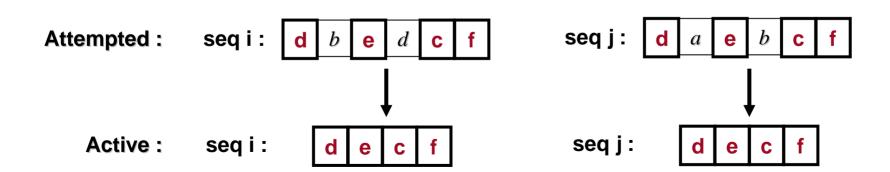
# Before mutation After mutation seq i: d a e d c f seq j: f a c b c d seq j: f a c a c d seq k: f a c b c d





### Finding Redundant Active Sequences

- An active optimization phase is one that is able to complete one or more transformations.
- Dormant phases do not affect the compilation.
- Compiler must indicate if phase was active.







### Detecting Identical Code

- Sometimes identical code for a function can be generated from different active sequences.
- Some phases are essentially independent
  - branch chaining and register allocation
- Sometimes more than one way to produce the same code.





### Detecting Identical Code (cont...)

Example:

```
r[2] = 1; r[3] = r[4] + r[2]; r[3] = r[4] + r[2]; r[3] = r[4] + r[2]; r[3] = r[4] + 1; r[3] = r[4] + 1;
```

Used CRC checksums to compare function instances.





### Detecting Equivalent Code

- Code generated by different optimization sequences may be equivalent, but not identical.
- Some optimization phases consume registers.
- Different ordering of such phases may result in equivalent instructions, but different registers being used.





### Detecting Equivalent Code (cont...)

```
sum = 0;
for (i = 0; i < 1000; i++ )
    sum += a [ i ];</pre>
```

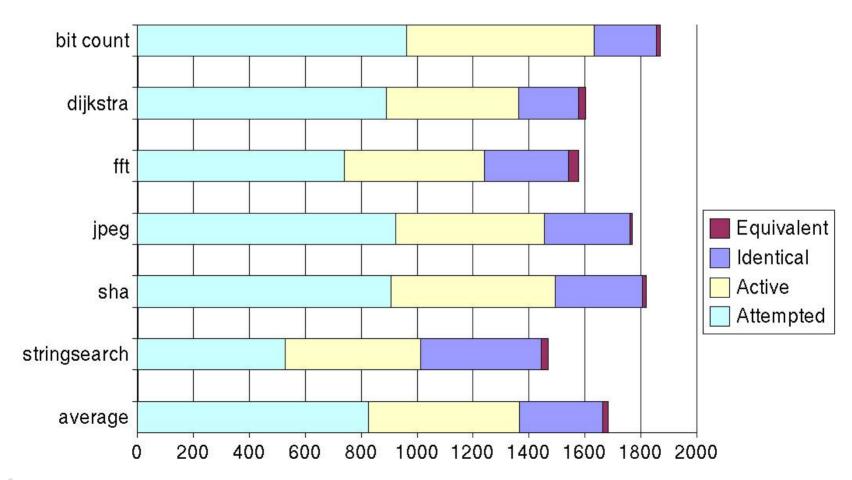
#### Source Code

```
r[10]=0;
                         r[11]=0;
  r[12]=HI[a];
                         r[10]=HI[a];
  r[12]=r[12]+LO[a];
                         r[10]=r[10]+LO[a];
  r[1]=r[12];
                         r[1]=r[10];
  r[9]=4000+r[12];
                         r[9]=4000+r[10];
L3
                       L3
  r[8]=M[r[1]];
                         r[8]=M[r[1]];
  r[10]=r[10]+r[8];
                         r[11]=r[11]+r[8];
  r[1]=r[1]+4;
                         r[1]=r[1]+4;
  IC=r[1]?r[9];
                         IC=r[1]?r[9];
  PC=IC<0,L3;
                         PC=IC<0,L3;
 Register Allocation
                         Code Motion before
 before Code Motion
                         Register Allocation
```

```
r[32]=0;
  r[33]=HI[a];
  r[33]=r[33]+LO[a];
  r[34]=r[33];
  r[35]=4000+r[33];
L3
  r[36]=M[r[34]];
  r[32]=r[32]+r[36];
  r[34]=r[34]+4;
  IC=r[34]?r[35];
  PC=IC<0,L3;
   After Mapping
     Registers
```



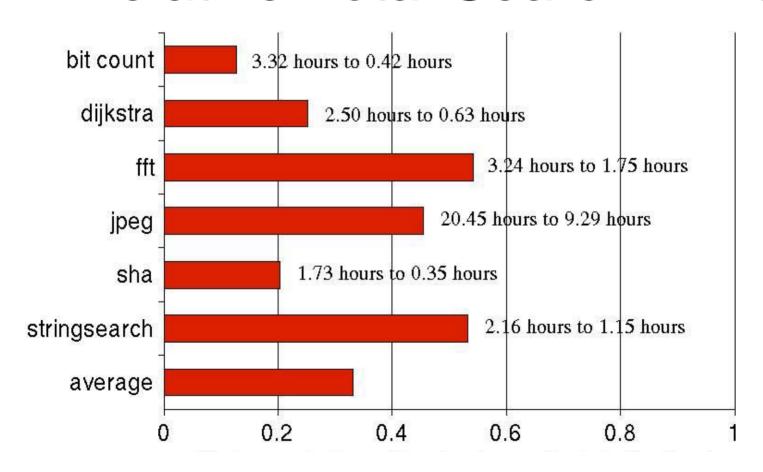
### Number of Avoided Executions







### Relative Total Search Time







### Reducing the Search Overhead

- Avoiding redundant executions.
- Obtaining similar results in fewer generations.





### Producing Similar Results in Fewer Generations

- Can reduce search time by running the genetic algorithm for fewer generations.
- Can obtain better results in the same number of generations.
- We evaluate four methods for reducing the number of required generations to find the best sequence in the search.





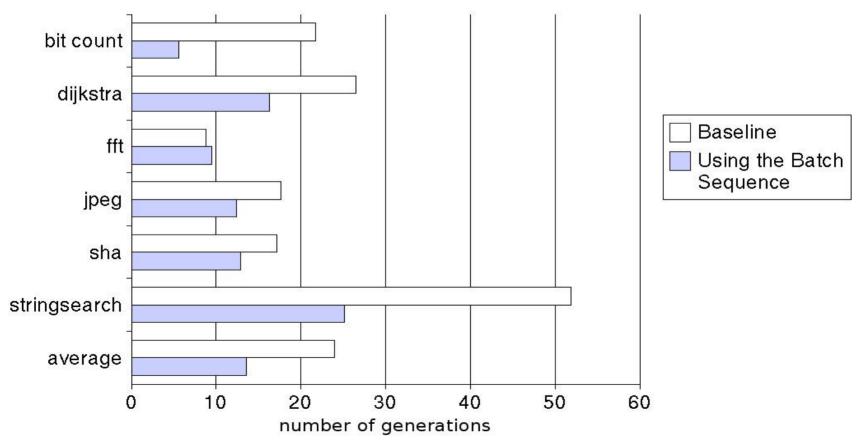
### Using the Batch Sequence

- Capture the active sequence of phases applied by the batch compiler.
- Place this sequence in the initial population.
- May allow the genetic algorithm to converge faster to the best sequence it can find.





# Number of Generations When Using the Batch Sequence







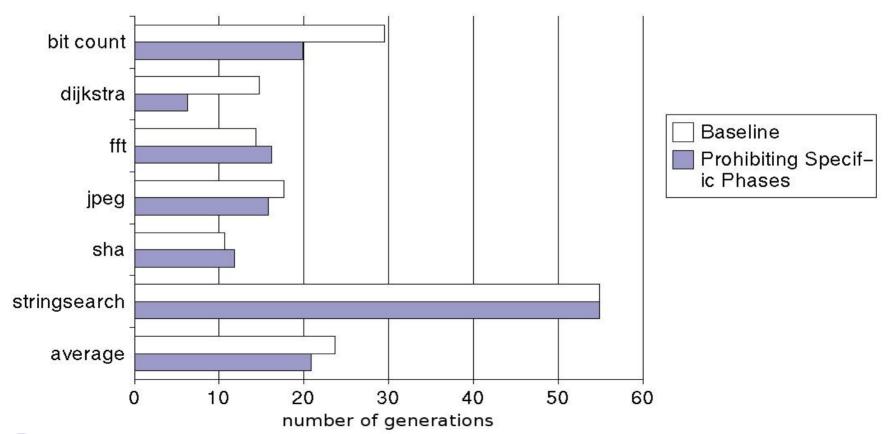
### Prohibiting Specific Phases

- Perform static analysis on the function.
  - No loops, then no loop optimizations.
  - No scalar variables, then no register allocation.
  - Only one basic block, then no unreachable code elimination and no branch optimizations.
  - Etc.
- Such phases are prohibited from being attempted for the entire search for that function.





### Number of Generations When Prohibiting Specific Phases







### Prohibiting Prior Dormant Phases

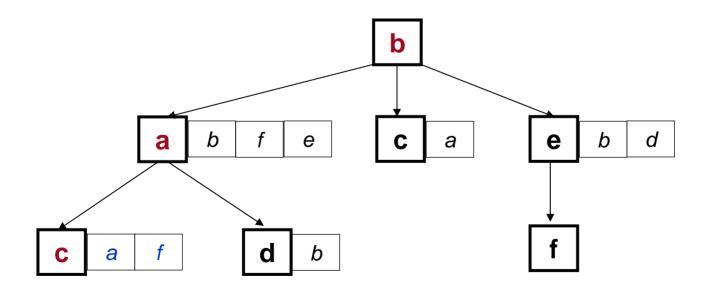
- Some phases will be found to be dormant given a specific prefix of active phases.
- If encounter the same prefix, then do not allow these prior dormant phases to be reattempted.
- Keep a tree of active prefixes and store the dormant phases with each node in the tree.
- Changed the genetic algorithm by forcing a prior dormant phase to mutate until finding a phase that has been active or not yet attempted with the prefix.





# Prohibiting Prior Dormant Phases (cont...)

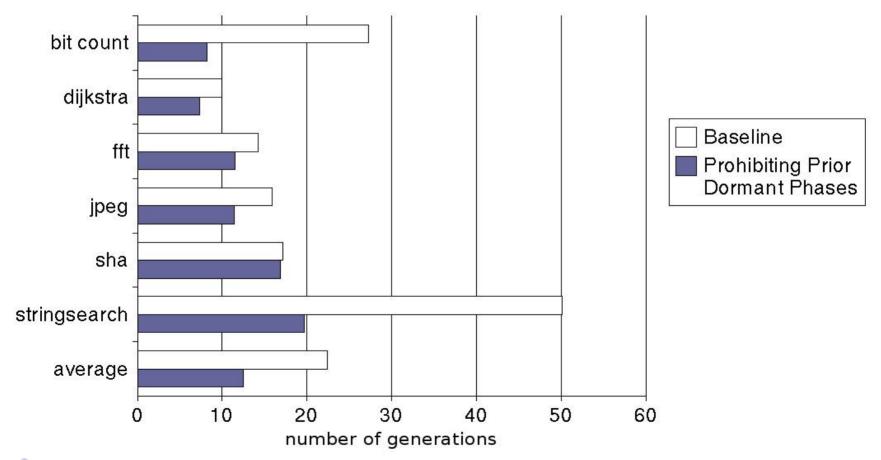
a and f are dormant phases given the active prefix of bac in the tree.







# Number of Generations When Prohibiting Prior Dormant Phases

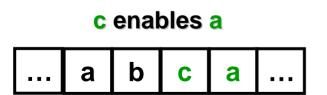


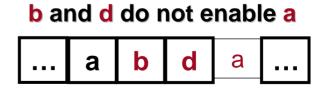




### Prohibiting Un-enabled Phases

- Most optimization phases when performed cannot be applied again until enabled.
  - ex: Register allocation will not be enabled by most branch optimizations



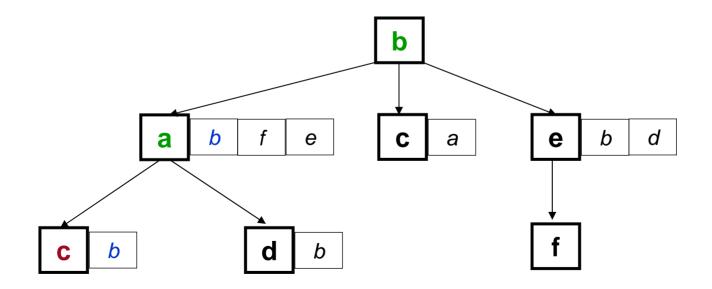






### Prohibiting Unenabled Phases (cont.)

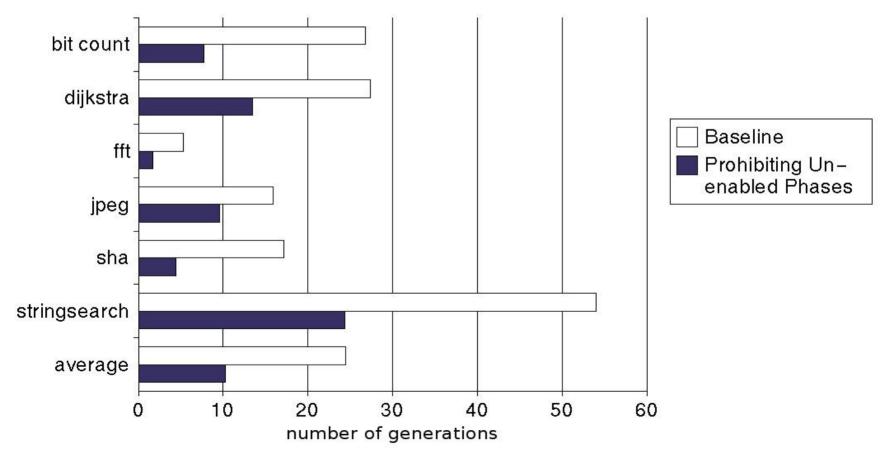
Assume b can be enabled by a, but cannot be enabled by c. Given the prefix bac, then b cannot be active at this point.







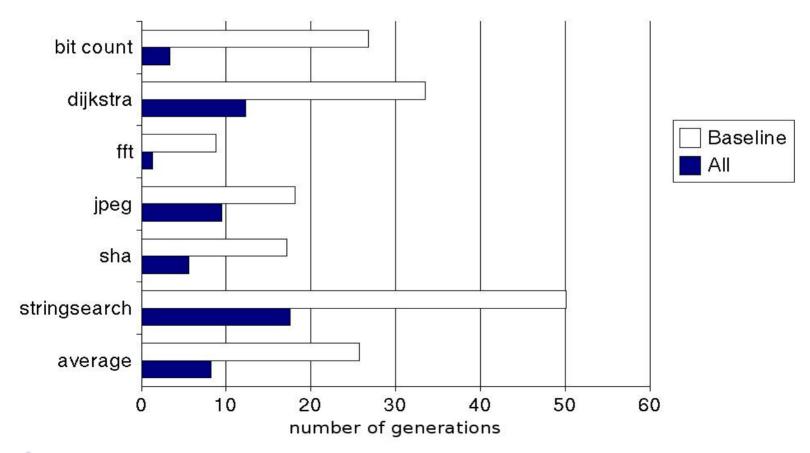
# Number of Generations When Prohibiting Unenabled Phases







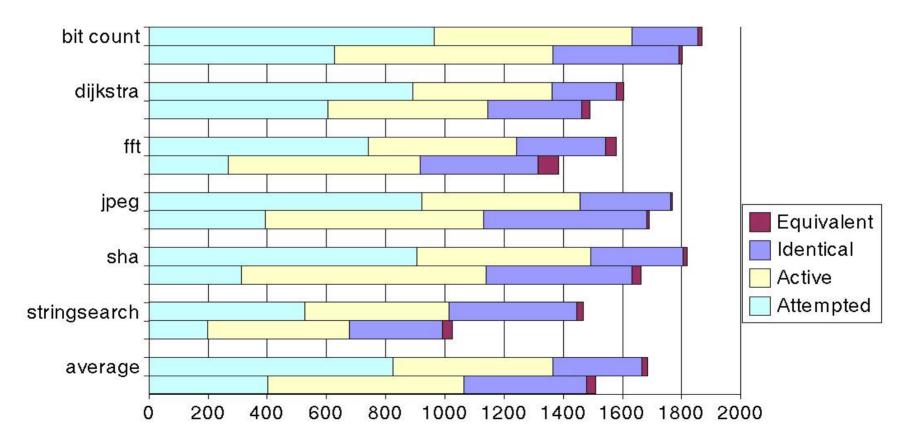
### Number of Generations When Applying All Techniques







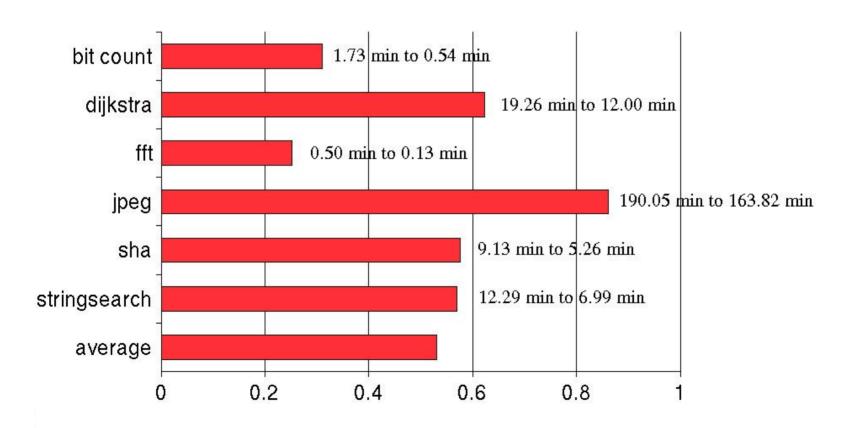
# Number of Avoided Executions When Reducing the Number of Generations







# Relative Search Time before Finding the Best Sequence







### Related Work

- Superoptimizers
  - instruction selection: Massalin
  - branch elimination: Granlund, Kenner
- Iterative compilation techniques using performance feedback information.
  - loop unrolling, software pipelining, blocking
- Using genetic algorithms to improve compiler optimizations
  - Parallelizing loop nests: Nisbet
  - Improving compiler heuristics: Stephenson et al.
  - Optimization sequences: Cooper et al.





### **Future Work**

- Detecting likely active phases given active phases that precede it.
- Varying the characteristics of the search.
- Parallelize the genetic algorithm.





### Conclusions

- Avoiding executions:
  - Important for genetic algorithm to know if attempted phases were active or dormant to avoid redundant active sequences.
  - Same code is often generated by different active sequences.
- Reducing the number of generations required to find the best sequence in the search:
  - Inserting the batch compilation active sequence is simple and effective.
  - Can use static analysis and empirical data to often detect when phases cannot be active.