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CS 265

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GitHub Link: https://github.com/shkailas/bril

(All optimization scripts can be found in bril/examples/local_optimizations)

Summary of Accomplishments:

In this task, I accomplished the following tasks:

- 1. Global (trivial) Dead Code Elimination
- 2. Local Dead Code Elimination
- 3. Local Value Numbering
 - a. Common Subexpression Elimination
 - b. Copy Propagation
 - c. Constant Propagation
 - d. Constant Folding

Algorithms and Design Decisions:

1. Global Dead Code Elimination

Global DCE was implemented almost identically to how it was demonstrated in class. There was very little variation or room for design decisions. To ensure layers of global dead code were removed, the global DCE algorithm was run until no new lines of code were removed. In terms of runtime optimality, a flag variable was used to mark whether there had been any changes in the program to determine if convergence was reached. Due to the use of python list comprehension, it was difficult to determine if instructions were removed (i.e. convergence was reached) without checking if the program before an iteration of global DCE was different than the program after. In hindsight, perhaps checking just the number of instructions instead of a list comparison may have been more efficient, but correctness remains the same.

Code Snippet:

2. Local Dead Code Elimination

Similar to global DCE, local DCE also very strictly adhered to the classroom implementation. The first step in this process was to create the maximal basic blocks. This was borrowed from the pre-built form_blocks function provided. As taught, local DCE was run on each block. Arguments in an instruction were handled before destinations so that unused instructions where a variable appeared in both the arguments and destination were handled appropriately. To ensure layers of local dead code were removed from each block, the local DCE algorithm was run until no new lines of code were removed per block. Some notable design decisions included indexing the instructions when iterating through a block. This made it easier to make sure that the correct instruction was deleted (and not another instruction that may have been identical but in a different location). Furthermore, to ensure this, instruction indices were stored until the end of a pass through a block, then sorted and removed in descending order so as to preserve the previous instruction indices. Finally, since the order of the blocks after local DCE was maintained, converting the modified blocks back to instruction format was as simple as combining all of the instructions in the order they appeared in the block, in block order.

Code Snippet:

- 3. Local Value Numbering
- a. Common Subexpression Elimination

The first step in this process was to create the maximal basic blocks. This was again borrowed from the pre-built form_blocks function provided. The common subexpression elimination code was written to adhere to the class pseudocode. The first major design decision was to write a line that essentially eliminated control flow statements and print statements from consideration. This was done under the assumption that instructions of this type typically do not involve non-singular expressions (where this algorithm would not really matter).

As expected, four maps were used to represent the LVN table, and a uniqueld variable was used whenever new entries were made involving numbering or the creation of temp variables to ensure uniqueness in the keys. Canonicalization was done relatively as expected, with values having the tuple form (<op>, <argument_1_num>, ...), and the creation of "symbolic" entries (for instructions where the arguments were variables that weren't defined before in the same block) was also handled with the value "symbolic<uniqueld>".

```
if "op" in block[i]:
    # if the instr has arguemnts (sum, id, mult, div)
    if "args" in block[i]:
        #args = [var2num[arg] for arg in block[i]["args"]]
        args = []
        for arg in block[i]["args"]:
        if not (arg in var2num or arg in temp_references): # symbolic case
            num2val[uniqueId] = "symbolic" + str(uniqueId)
            val2num["symbolic" + str(uniqueId)] = uniqueId
            var2num[arg] = uniqueId
            num2var[uniqueId] = arg
            uniqueId += 1

        if arg in temp_references:
            args.append( var2num[temp_references[arg]] )
        else:
            args.append( var2num[arg] )

        value = tuple([block[i]["op"]] + args)

elif block[i]["op"] == "const":
        value = tuple(("const"]+[block[i]["value"]])
```

Then the typical common subexpression elimination was conducted, using the val2num map to see if the subexpression was already computed. Note that only direct matches were optimized for (i.e. no commutativity, partial matches, etc.). If the subexpression was already calculated previously and the variable that it was stored in still had that value, then an id instruction replaced the subexpression instruction, reducing computation.

```
if value not in val2num:
    num = uniqueId
    val2num[value] = num
    num2val[num] = value
    uniqueId += 1

else:
    num = val2num[value]
    if num2var[num] in temp_references: # its clobbered
        pass
    else: # eliminate-able common subexpression
        block[i]["op"] = "id"
        block[i]["args"] = [num2var[num]]
```

To handle clobbered variables, an aggressive approach was taken to introduce temporary variables. This was done partly because I wanted to try and implement it, but mostly to save precomputed entries in the value-number tables so that the computation could be reused (instead of being deleted when the clobbered variable was deleted). Temp variables were created using the naming format "<original_name>_temp<uniqueld>" so that their names would be unique. These temp variable names were also linked to their original names using a temp_references dictionary so that the most recent temp variable could always be utilized when converting from var names to num (see earlier code snippet).

```
if block[i]["dest"] in var2num:
    temp_name = block[i]["dest"]+"_temp"+str(uniqueId)
    uniqueId +=1

    temp_references[block[i]["dest"]] = temp_name
    block[i]["dest"] = temp_name

    var2num[temp_name] = num
    num2var[num] = temp_name

else:
    var2num[ block[i]["dest"] ] = num
    num2var[num] = block[i]["dest"]
```

Lastly, all of the blocks' instructions were merged back together in a similar manner as was done with local DCE.

b. Copy Propagation

This copy propagation also began with creating maximal blocks. This implementation of copy propagation was written to replace variables who share the same value (at that instruction in the same block) with the variable that appears earliest in the code. This was done in a separate pass from the LVN pass so that its results could be more usefully utilized by the LVN pass. Equivalence between variables was only determined when an "id" instruction was done.

```
for block in blocks:
    equivalences = {}
    for i in range(0, len(block)): # for instruction block[i]
    if "op" in block[i] and block[i]["op"] == "id":
        if block[i]["args"][0] in equivalences:
```

```
equiv = equivalences[block[i]["args"][0]]
block[i]["args"] = [ equiv ]
equivalences[block[i]["dest"]] = equiv
else:
    equivalences[block[i]["dest"]] = block[i]["args"][0]
```

Code was also written to 1. propagate these equivalences through argument variables and 2. to remove equivalence relationships after a variable has been modified.

```
elif "op" in block[i] and block[i]["op"] in operations_that_modify_dest:
    if "args" in block[i]:
        block[i]["args"] = [equivalences[arg] if (arg in equivalences) else arg for arg in
block[i]["args"]]
    equivalences.pop(block[i]["dest"], None)# clobbered
```

Lastly, all of the blocks' instructions were merged back together in a similar manner as was done with local DCE. This implementation of copy propagation was done so that more local dead code would be revealed, since not all the variable assignments are really necessary in the program where previously defined variables would have sufficed.

c. Constant Propagation & Folding

Constant Propagation and Folding were done together and added to the LVN pass. This was done under the belief that constant-related optimizations would not help with subexpression optimizations as much and vice versa, since constant propagation would essentially remove the need for a subexpression anyways. The process boiled down to identifying instructions that could potentially benefit from constant propagation or folding. I specifically avoided division by 0 examples since this I did not want to implement what the optimizer should do if this case was reached. If it was determined that every argument in the instruction could be traced to a constant value, then the instruction was rewritten to be a "const" instruction with the appropriate calculated value using a self-written evaluate_operation function (not shown here).

After the instruction was rewritten, the four state tables needed to be modified to reflect these changes and impacts of clobbering (as was done with common subexpression elimination)

```
# update the state tables
num = uniqueId
value = tuple([block[i]["op"]] + [folding_result])
val2num[value] = num
num2val[num] = value
if block[i]["dest"] in var2num:
        temp_name = block[i]["dest"]+"_temp"+str(uniqueId)
        temp_references[block[i]["dest"]] = temp_name
        block[i]["dest"] = temp_name
        var2num[temp_name] = num
        num2var[num] = temp_name
else:
        var2num[ block[i]["dest"]] = num
        num2var[num] = block[i]["dest"]
```

Lastly, all of the blocks' instructions were merged back together in a similar manner as was done with local DCE.

In general, optimizations were ordered in the following format: Global DCE, local DCE, copy propagation, constant propagation, constant folding, common subexpression elimination, global DCE, local DCE. The DCEs surround the LVN optimizations since they reduce the work of the LVN optimizations and the LVN optimizations may leave more dead code that can be cleaned up. Copy propagation was chosen next because it reduces the number of variables to deal with and may make common subexpression elimination easier. Constant propagation and constant folding were done together due to the similar nature of pushing constants through the program. Their place in the pipeline also came before common subexpression since there was an inherent preference to have constants in the program over variable expressions, but otherwise these could be flipped.

Results and Analysis:

To measure the value of these optimizations and their ability to work together, I used the TCDE benchmark suite and the LVN benchmark suite. Below are the different pipelines of optimizations used on each benchmark:

- 1. TCDE benchmarks
- a. Baseline
- b. Global Dead Code Elimination (global_dce)
- c. Local Dead Code Elimination (global_local_dce)

Test Case	Optimization Level	Result
reassign	baseline	3
reassign	global_dce	3
reassign	global_local_dce	2
combo	baseline	6
combo	global_dce	5
combo	global_local_dce	4
skipped	baseline	4
skipped	global_dce	4
skipped	global_local_dce	4
double	baseline	6
double	global_dce	4
double	global_local_dce	4
simple	baseline	5
simple	global_dce	4
simple	global_local_dce	4
double-pass	baseline	6
double-pass	global_dce	4
double-pass	global_local_dce	4
reassign-dkp	baseline	3
reassign-dkp	global_dce	3

reassign-dkp	global_local_dce	2
diamond	baseline	6
diamond	global_dce	6
diamond	global_local_dce	6

- 2. LVN Benchmarks (of those that would run)
- a. Global/Local Dead Code Elimination (dce)
- b. Global/Local Dead Code Elimination + Common Subexpression Elimination (dce_cse)
- c. Global/Local Dead Code Elimination + Common Subexpression Elimination + Global/Local Dead Code Elimination (dce_cse_dce)
- d. Global/Local Dead Code Elimination + Common Subexpression Elimination + Global/Local Dead Code Elimination (dce_cse_dce)
- e. Global/Local Dead Code Elimination + Copy Propagation + Common Subexpression Elimination + Global/Local Dead Code Elimination (dce_copy_cse_dce)
- f. Global/Local Dead Code Elimination + Copy Propagation + Constant Propagation + Constant Folding + Common Subexpression Elimination + Global/Local Dead Code Elimination (dce_copy_const_cse_dce)

Note: logical-operators, divide-by-zero, fold-comparisons tests would not run on the brench tool and were run by hand.

logical-operators	baseline	missing
logical-operators	dce	0
logical-operators	dce_cse	0
logical-operators	dce_cse_dce	0
logical-operators	dce_copy_cse_dce	0
logical-operators	dce_copy_const_cse_dce	0
reassign	baseline	3
reassign	dce	2
reassign	dce_cse	2
reassign	dce_cse_dce	2
reassign	dce_copy_cse_dce	2
reassign	dce_copy_const_cse_dce	2
idchain	baseline	5
idchain	dce	5
idchain	dce_cse	5
idchain	dce_cse_dce	5
idchain	dce_copy_cse_dce	5
idchain	dce_copy_const_cse_dce	2
redundant-dce	baseline	6
redundant-dce	dce	6
redundant-dce	dce_cse	6
redundant-dce	dce_cse_dce	6
redundant-dce	dce_copy_cse_dce	6
redundant-dce	dce_copy_const_cse_dce	2
nonlocal-clobber	baseline	4

nonlocal-clobber	dce	4
nonlocal-clobber	dce_cse	4
nonlocal-clobber	dce_cse_dce	3
nonlocal-clobber	dce_cse_dce	3
nonlocal-clobber	dce_copy_cse_dce dce_copy_const_cse_dce	3
divide-by-zero	baseline	5
	dce	5
divide by zero		5
divide by zero	dce_cse	
divide-by-zero	dce_cse_dce	5
divide-by-zero	dce_copy_cse_dce	5
divide-by-zero	dce_copy_const_cse_dce	5
clobber-arg	baseline	3
clobber-arg	dce	3
clobber-arg	dce_cse	3
clobber-arg	dce_cse_dce	0
clobber-arg	dce_copy_cse_dce	0
clobber-arg	dce_copy_const_cse_dce	0
idchain-nonlocal	baseline	6
idchain-nonlocal	dce	6
idchain-nonlocal	dce_cse	6
idchain-nonlocal	dce_cse_dce	6
idchain-nonlocal	dce_copy_cse_dce	6
idchain-nonlocal	dce_copy_const_cse_dce	6
nonlocal	baseline	7
nonlocal	dce	7
nonlocal	dce_cse	7
nonlocal	dce_cse_dce	7
nonlocal	dce_copy_cse_dce	7
nonlocal	dce_copy_const_cse_dce	5
commute	baseline	6
commute	dce	6
commute	dce_cse	6
commute	dce_cse_dce	6
commute	dce_copy_cse_dce	6
commute	dce_copy_const_cse_dce	2
rename-fold	baseline	7
rename-fold	dce	7
rename-fold	dce_cse	7
rename-fold	dce_cse_dce	6
rename-fold	dce_copy_cse_dce	6
rename-fold	dce_copy_const_cse_dce	4
redundant	baseline	6
redundant	dce	6
redundant	dce_cse	6
redundant	dce_cse_dce	6
redundant	dce_copy_cse_dce	6
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redundant	dce_copy_const_cse_dce	2
clobber	baseline	10
clobber	dce	5
clobber	dce_cse	5
clobber	dce_cse_dce	5
clobber	dce_copy_cse_dce	5
clobber	dce_copy_const_cse_dce	2
idchain-prop	baseline	5
idchain-prop	dce	5
idchain-prop	dce_cse	5
idchain-prop	dce_cse_dce	5
idchain-prop	dce_copy_cse_dce	5
idchain-prop	dce_copy_const_cse_dce	5
clobber-fold	baseline	10
clobber-fold	dce	5
clobber-fold	dce_cse	5
clobber-fold	dce_cse_dce	5
clobber-fold	dce_copy_cse_dce	5
clobber-fold	dce_copy_const_cse_dce	2
fold-comparisons	baseline	15
fold-comparisons	dce	0
fold-comparisons	dce_cse	0
fold-comparisons	dce_cse_dce	0
fold-comparisons	dce_copy_cse_dce	0
fold-comparisons	dce_copy_const_cse_dce	0

1. TCDE benchmarks Analysis

The TCDE data reveals that our local and global analysis reduced the number of instructions in almost every benchmark. This was to be expected, since many of these programs were simple and had some obvious dead code in them. Furthermore, running both local and global dead code analyses on these benchmarks by hand revealed that using our classroom environments, these algorithms were working as expected (assuming I did it correctly by hand).

2. LVN benchmarks Analysis

The LVN data also reveals the potency of compounding each optimization as it was developed. Some of the biggest jumps in instruction count across the board were from 1. no optimizations to local/global DCE 2. local/global DCE to local/global DCE+common subexpression elimination + local/global DCE, and 3. Global/Local Dead Code Elimination + Copy Propagation + Common Subexpression Elimination + Global/Local Dead Code Elimination to Global/Local Dead Code Elimination + Copy Propagation + Constant

Propagation + Constant Folding + Common Subexpression Elimination + Global/Local Dead Code Elimination.

The first was expected, considering the previous result. The second major jump was likely due to how the common subexpression elimination was converting many instructions to id instructions while also rendering those id instructions as dead code since the same result was stored in another variable (also the dataset happened to have lots of these examples). The last major jump was likely from the constant propagation/folding addition, since this step removed many instructions and simplified them to const instructions. Taken together with the fact that there weren't many basic blocks in these benchmarks, there were likely lots of instructions that became dead code generated by these optimizations (since there were lots of instructions whose arguments were const vars in the same block), which got cleaned up in the final sweeps of DCE. It was cool to see how the instructions drastically dropped; however, the style of benchmarks contributed to that heavily. That being said, there are definitely times when my copy propagation does not work as intended. Some of the

Beyond a theoretical understanding, these benchmarks also provide some evidence of the correctness of our algorithm in not touching the observable features of the code. In this case, observable refers to the printed output of the program, and every optimization pipeline had the same results as the baseline.

Optimized Programs:

Here is a program that my optimizations do well on. This example features elements of common subexpression elimination, copy propagation, constant propagation and folding, and cleanup by DCE sweeps. It also effectively handles instances of variable clobbering:

```
@main {
  a: int = const 4;
  b: int = const 2;
  c: int = const 2;
  sum1: int = add a b;
  sum2: int = add a b;
  prod1: int = mul sum1 sum2;
  copy1: int = id prod1;
  copy2: int = id copy1;
  copy3: int = id copy2;
  # Clobber both sums.
  sum1: int = const 0;
  sum2: int = const 0;
  sum3: int = add a b;
  prod2: int = mul sum3 sum3;
  print prod2;
```

```
.testLabel:
    f: int = const 100;
    f: int = const 42;
    print f;
    f: int = const 4;
    b: int = const 2;
    c: int = const 1;
    d: int = add f b;
    e: int = add c d;
    print d;
}
```

Optimized version:

```
@main {
  prod2: int = const 36;
  print prod2;
.testLabel:
  f: int = const 42;
  print f;
  d: int = const 6;
  print d;
}
```

Here is a script that is not optimized well. The optimization suffers due to lack of handling commutativity (add a b vs. add b a), lack of optimizations around dataflow (i.e. skipping useless mylabel, sharing state across blocks with only one entry point), and lack of obvious constant folding scenarios that can be hard coded (multiplying by 0). In fact, the "optimized" version does not make any changes except some constant folding:

```
@main {
    a: int = const 36;
    b: int = const 47;
    c: int = add a b;
    print c;
    print d;
    jmp termination;

.dumblabel:
    f: int = add a b;
    d: int = add b a;
    print f;
    g: int = const 0;
    h: int = mul f g;
.termination:
```

```
print a;
}
```

Optimized version:

```
@main {
    a: int = const 36;
    b: int = const 47;
    c: int = const 83;
    print c;
    print d;
    jmp termination;
.dumblabel:
    f: int = add a b;
    d: int = add b a;
    print f;
.termination:
    print a;
}
```