# **Matrix Multiplication**

CS 5220: Homework 1

Group 9

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## 1 Method

We refer to the tuning ideas on the note "Tuning on a single core".

## 1.1 Blocking

We can partition matrices into small blocks, and apply the block matrix multiplication. The advantage is that original matrices cannot fit into cache while the little blocks can. We compare different block sizes, including  $2^p$ ,  $1 \le p \le 10$ .

# 1.2 Loop Order

In the naive implementation, we loop over i, then j, then k.

An alternative loop order might be more cache friendly. For example, using the (i, j, k) order, we have to go across a row of A in the inner loop, which is a non-unit-stride access.

#### 1.2.1 Pure Loop Order

From the original implementation, we access the following:

```
A[k*M + i]
B[j*M + k]
C[j*M + i]
```

From a purely theoretical standpoint, the original loop order of i, j, k may not be the most efficient, as matrix A is being iterated through by indices dependent on k and i. This would imply that the loop should be reordered to have k and i be the innermost two loops, and j be the outermost one.

It should be noted that while this works in the initial implementation of matrix multiplication, if matrices are transposed, the loop ordering will need to be altered.

#### 1.2.2 Loop Order with Blocking

We compare different orders, including (i, k, j), (j, i, k), (j, k, i).

When we apply blocking, there are two layers of loop orders. For example, the algorithm for looping order  $(i, j, k) \times (j, i, k)$  is as follows.

```
for block i

for block j

for block k

for cell j

for cell i

for cell k

do calculation
```

Again, we compare different combinations of block size and two-layer looping order.

#### 1.3 Copy Optimization

We might run into conflict misses associated with cache. One way to solve the problem is to explicitly copy blocks into a contiguous block of local storage before multiplying them.

Another side benefit of copy optimization is that we can use it to gracefully deal with fringe blocks.

We add copy optimization to enhance blocking, and compare different block sizes, including  $2^p$ ,  $1 \le p \le 10$ .

### 1.4 Compiler Flag

It is worthwhile playing with the flags that control the compiler optimizations, since modern compilers do some types of optimizations much better than people do.

We compile the code with option -funroll-loops to unroll loops (basic loop unrolling is automatic with -O3).

# 2 Result

We run jobs on the totient cluster, and get the results depicted in several plots.

# 2.1 Blocking

See Figure 1 and 2 below for blocking with default loop order  $(i, j, k) \times (i, j, k)$ .

The plot shows that blocking performs the best with block size 64, 128, 256.

# 2.2 Loop Order

See Figure 3–7 below for loop order.

The plot shows that looping performs the best with loop order (j, k, i) for the basic imple-

mentation, as earlier hypothesized. Additionally, as expected, after matrix A is transposed, the loop order (j, k, i) is no longer the most efficient.

#### 2.3 Blocking and Loop Order

See Figure 8–13 below for blocking with loop order.

The plot shows that blocking and looping performs the best with block size 256, 512 and loop order  $(i, j, k) \times (j, k, i), (j, k, i) \times (j, k, i)$ .

### 2.4 Blocking and Copy Optimization

See Figure 14 and 15 below for blocking with default loop order  $(i, j, k) \times (i, j, k)$  and copy optimization.

The plot shows that blocking and copying performs the best with block size 8, 128, 512.

### 2.5 Blocking, Loop Order and Copy Optimization

See Figure 16–21 below for blocking with loop order and copy optimization.

The plot shows that blocking, looping and copying performs the best with block size 16,32 and loop order  $(i,j,k) \times (j,i,k)$ .

# 2.6 Compiler Flag

See Figure 22 below for unrolling loops in compiler flags.

The plot shows that unrolling loops in compiler flags alone does not make much difference.

# Reference

(1) David Bindel, Tuning on a single core, Applications of Parallel Computers (CS 5220), Fall 2011.

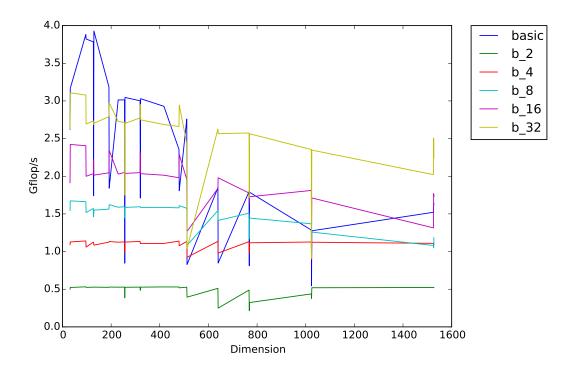


Figure 1: blocking with loop order  $(i,j,k) \times (i,j,k)$ 

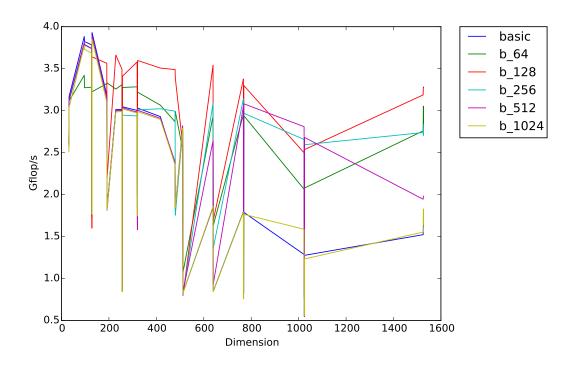


Figure 2: blocking with loop order  $(i, j, k) \times (i, j, k)$ 

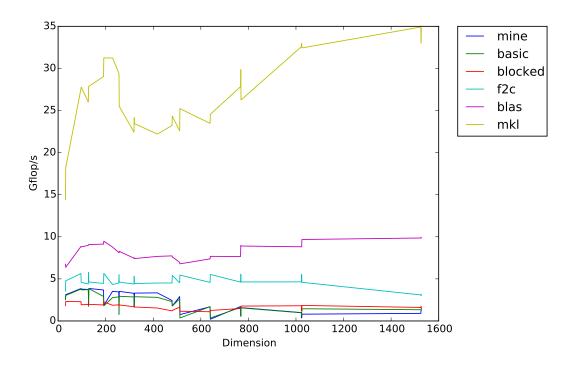


Figure 3: loop order (i, j, k)

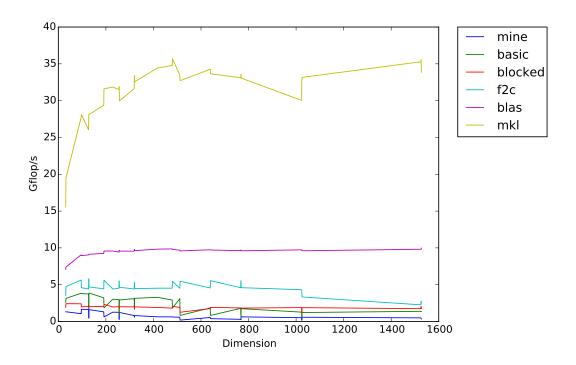


Figure 4: loop order (i, k, j)

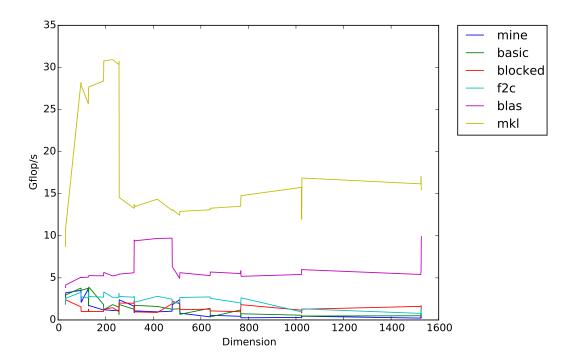


Figure 5: loop order (j, i, k)

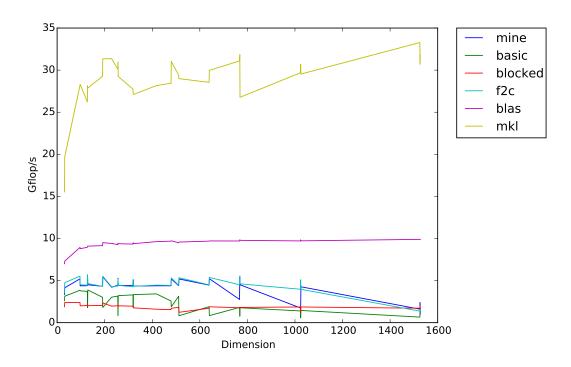


Figure 6: loop order (j, k, i)

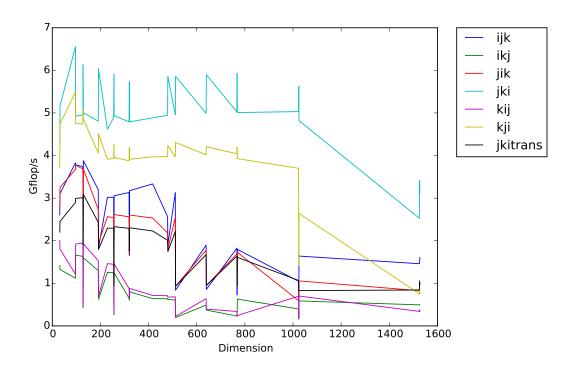


Figure 7: loop orders with transpose

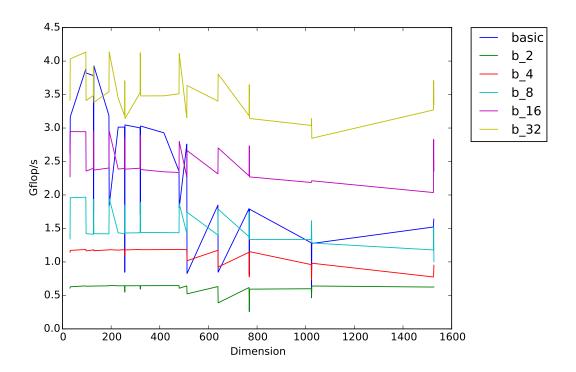


Figure 8: blocking with loop order  $(i,j,k)\times(j,k,i)$ 

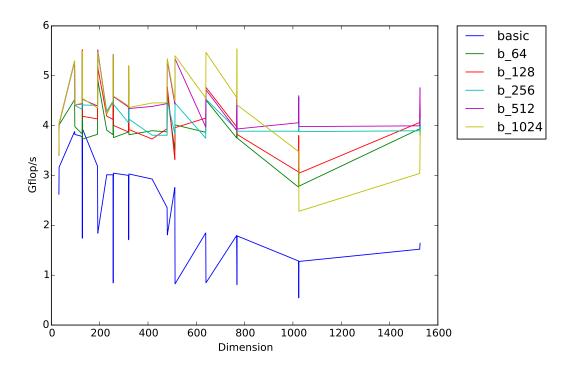


Figure 9: blocking with loop order  $(i,j,k) \times (j,k,i)$ 10

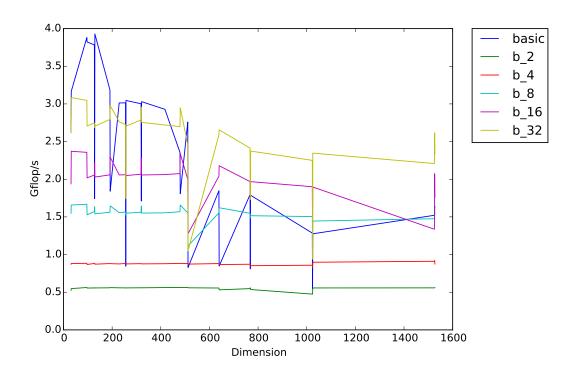


Figure 10: blocking with loop order  $(j,k,i)\times(i,j,k)$ 

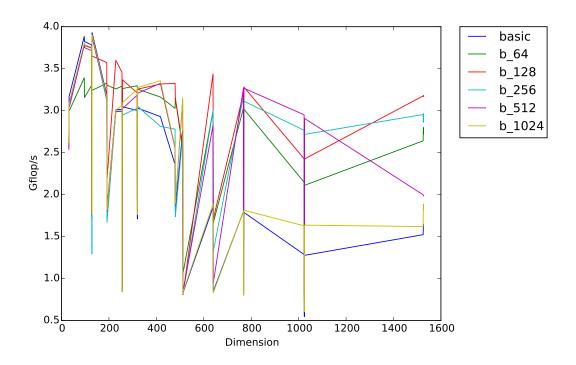


Figure 11: blocking with loop order  $(j,k,i)\times(i,j,k)$ 

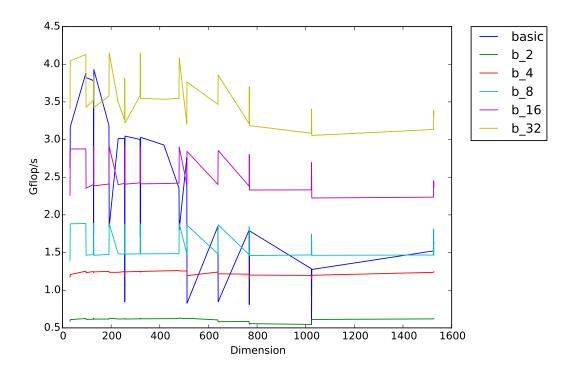


Figure 12: blocking with loop order  $(j,k,i)\times(j,k,i)$ 

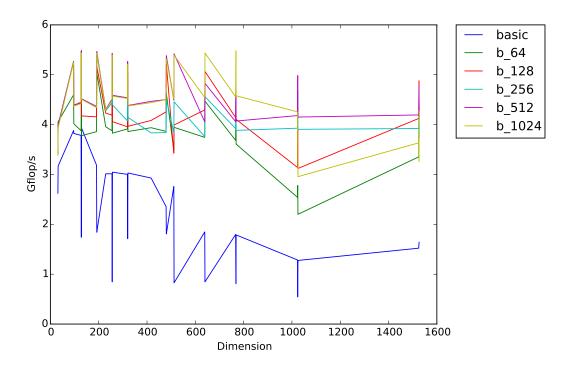


Figure 13: blocking with loop order  $(j, k, i) \times (j, k, i)$ 

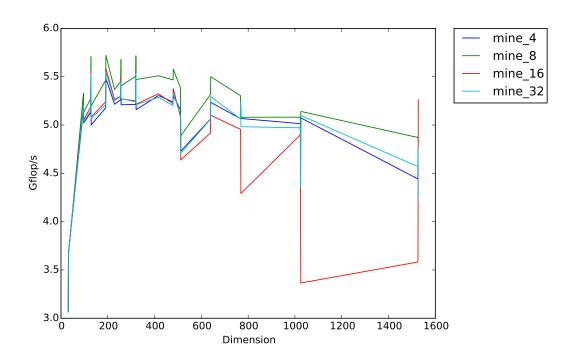


Figure 14: blocking with loop order  $(i,j,k)\times(i,j,k)$  and copy optimization

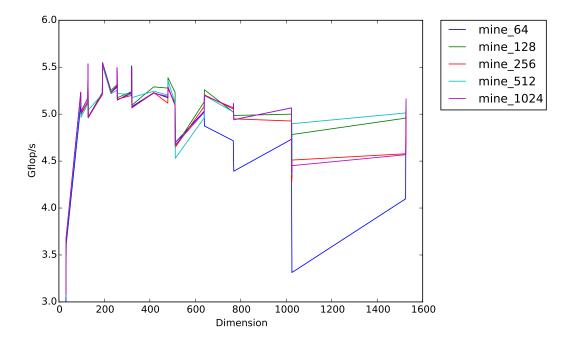


Figure 15: blocking with loop order  $(i,j,k) \times (i,j,k)$  and copy optimization

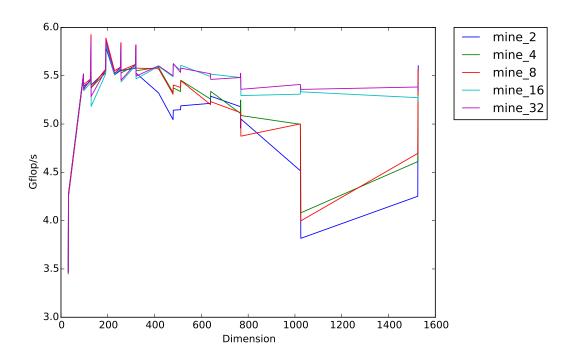


Figure 16: blocking with loop order  $(i,j,k) \times (j,i,k)$  and copy optimization

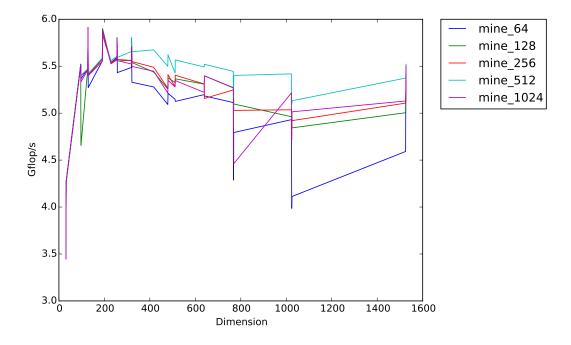


Figure 17: blocking with loop order  $(i,j,k) \times (j,i,k)$  and copy optimization

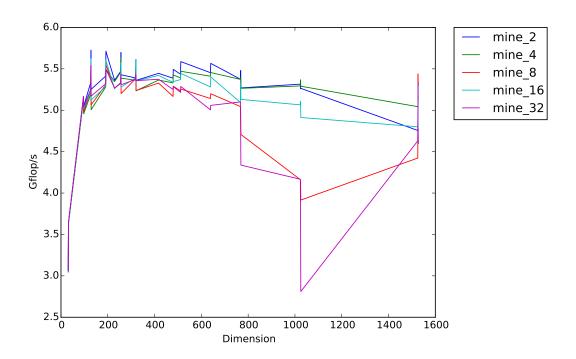


Figure 18: blocking with loop order  $(j,i,k)\times(j,i,k)$  and copy optimization

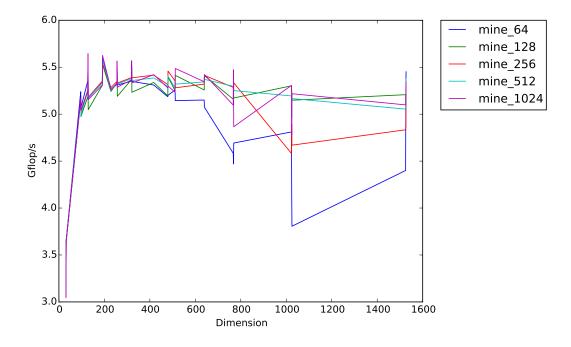


Figure 19: blocking with loop order  $(j,i,k)\times(j,i,k)$  and copy optimization

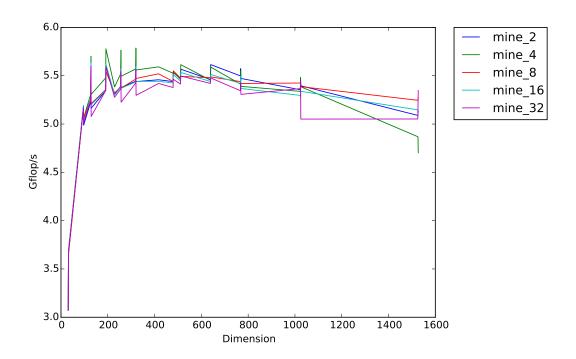


Figure 20: blocking with loop order  $(j,k,i)\times(j,i,k)$  and copy optimization

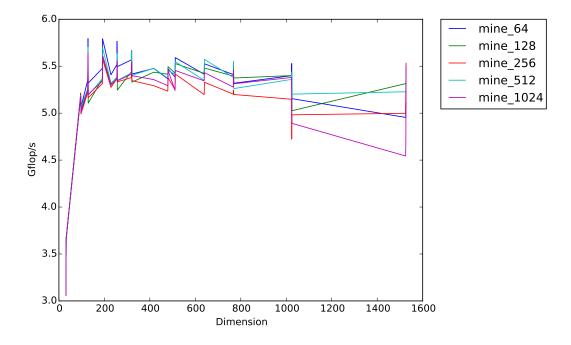


Figure 21: blocking with loop order  $(j,k,i)\times(j,i,k)$  and copy optimization

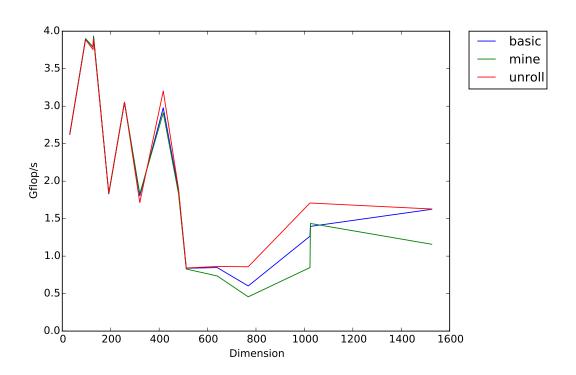


Figure 22: unrolling loops in compiler flags