

Our first step to optimizing the program was to jump directly into SSE by using the example script already written in Bitbucket. The hope was that the SSE code was a fast enough kernel that we can build a good multiplier around it; indeed, it an immediate speed-up was noticed! It approximately doubled our naive multiplication. See figure 1.

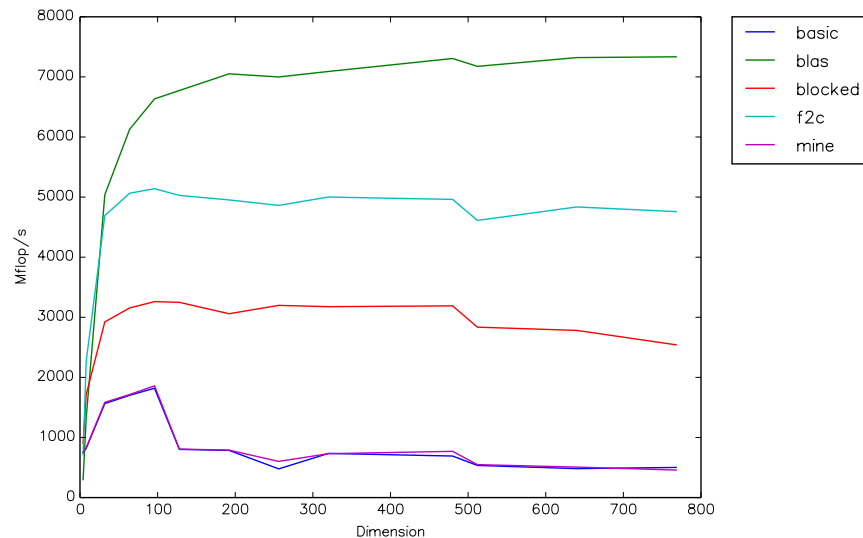


Figure 1: Our initial set-up using the SSE code provided. Note that due to laziness, the “blocked” line is actually our matrix multiply.

What the code does is fairly basic:

1. Initialize arrays At, Bt.
2. Takes in matrix A, B and copies them into At, Bt such that we have memory layout of 2 by P blocks in A and P by 2 blocks in B.
3. Loop over the blocks in A and B and multiply, adding the 2 by 2 matrix from the output into the C matrix.

We originally compiled this SSE code via the Intel compiler, but weirdly we found that compiling it with the GCC compiler results in a roughly 33% speedup over the Intel compiler! We’re actually not sure why, but various flags of the ICC doesn’t seem to coax the same performance out of the code.

This was all tested on even sizes initially, as there was no need to pad the matrices with anything. The odd cases are handled in the natural way, which was to pad a row and column and take the slight hit to the performance. When implemented, our performance was still pretty decent. See figure 2.

The next step to optimizing this was to play with the flags and use tiny code tricks.

Tricks

One trick that we played with was to do some loop reordering. In our main block, we have the following

```
for (bi = 0; bi < n_blocks; ++bi) {
    const int i = bi * BLOCK_SIZE;
    for (bj = 0; bj < n_blocks; ++bj) {
        const int j = bj * BLOCK_SIZE;
        // trans is our ``transitional`` matrix which contains both At and Bt
        kdgemm2P2(n, tempC, &trans[bi*n*2], &trans[temp + bj*n*2]);
    }
}
```

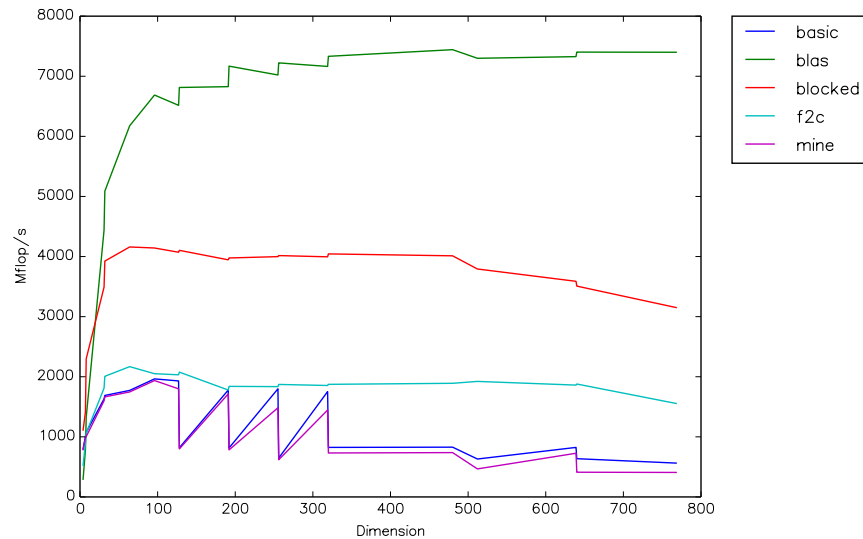


Figure 2: After adding the odd cases in (note the “blocked” line is ours).

Size	MFlops (flipped)	MFlops (as shown)
767	2660.82	3516.27
768	2348.95	3586.11
769	2229.79	3503.37

Table 1: Table of loop ordering

```
}
}
```

We noted that flipping b_i and b_j 's loop order hurts the performance a lot. For example, at the far end of the spectrum of the sizes the different is quite noticable. See table 1, so we determined that Loop reordering was also applied to the copy commands:

```
if (odd == 0) {
  for (int j = 0; j < n; j++) {
    for (int i = 0; i < m; i++) {
      At[i*n*2 + 2*j] = A[j*n + i*2];
      At[i*n*2 + 2*j + 1] = A[j*n + i*2 + 1];
    }
  }
}
```

while it wasn't obvious which order should be better as A_t increments via j via A increments via i , it turns out j, i ordering is *slightly* faster.

Another small speed increase laid with the serial performance of the copying. Initially, we actually declared to different arrays, one for A and one for B . We find that if we just create one giant array of size $2n^2$, and copy A, B into that giant array, performance increase slightly.

Compiler Flags (Sep)

Memory Usage

One thing we tried was to break up the P by 2 and 2 by P matrices into smaller block. Theoretically, since the size of a double is 8 bytes, we need to fit $2(2P) * 8 + 4 = 32P + 4$ bytes into the cache. While the level 1 cache of the educational nodes are quite low, it seems to not matter from the plots (our maximum size that we can fit in the L1 cache is $P = 128$). If we look at the level 2 cache, our array size can go up to 1024 theoretically, which (theoretically) shouldn't impede us.

If we look at the cachegrind output for our program, we see the following:

```
[sj423@en-cluster02 matmul]$ valgrind --tool=cachegrind ./matmul-blocked
==18730== Cachegrind, a cache and branch-prediction profiler
==18730== Copyright (C) 2002-2013, and GNU GPL'd, by Nicholas Nethercote et al.
==18730== Using Valgrind-3.9.0 and LibVEX; rerun with -h for copyright info
==18730== Command: ./matmul-blocked
==18730==
--18730-- warning: L3 cache found, using its data for the LL simulation.
--18730-- warning: pretending that LL cache has associativity 24 instead of actual 16
Compiler: gcc
Options: -O3 -funroll-loops -msse4.2 -ffast-math -mtune=generic -march=corei7
Description: Simple blocked dgemm.

==18730==
==18730== I refs: 62,147,347,847
==18730== I1 misses: 1,539
==18730== L1i misses: 1,518
==18730== I1 miss rate: 0.00%
==18730== L1i miss rate: 0.00%
==18730==
==18730== D refs: 32,109,636,396 (27,962,135,435 rd + 4,147,500,961 wr)
==18730== D1 misses: 1,976,702,658 ( 1,966,158,637 rd + 10,544,021 wr)
==18730== L1d misses: 7,256,578 ( 5,016,816 rd + 2,239,762 wr)
==18730== D1 miss rate: 6.1% ( 7.0% + 0.2% )
==18730== L1d miss rate: 0.0% ( 0.0% + 0.0% )
==18730==
==18730== LL refs: 1,976,704,197 ( 1,966,160,176 rd + 10,544,021 wr)
==18730== LL misses: 7,258,096 ( 5,018,334 rd + 2,239,762 wr)
==18730== LL miss rate: 0.0% ( 0.0% + 0.0% )
```

which actually mimics the performance of the BLAS cache stats:

```
[sj423@en-cluster02 matmul]$ valgrind --tool=cachegrind ./matmul-blas
==1449== Cachegrind, a cache and branch-prediction profiler
==1449== Copyright (C) 2002-2013, and GNU GPL'd, by Nicholas Nethercote et al.
==1449== Using Valgrind-3.9.0 and LibVEX; rerun with -h for copyright info
==1449== Command: ./matmul-blas
==1449==
--1449-- warning: L3 cache found, using its data for the LL simulation.
--1449-- warning: pretending that LL cache has associativity 24 instead of actual 16
Compiler: gcc
Options: -O3 -funroll-loops -msse4.2 -ffast-math -mtune=generic -march=corei7
Description: System CBLAS dgemm.
```

```
==1449==
==1449== I refs:  59,157,389,231
==1449== I1 misses:      2,399
==1449== LLi misses:     2,297
==1449== I1 miss rate:    0.00%
==1449== LLi miss rate:   0.00%
==1449==
==1449== D refs:  30,847,547,276 (26,700,497,225 rd + 4,147,050,051 wr)
==1449== D1 misses: 1,670,642,244 ( 1,665,753,557 rd + 4,888,687 wr)
==1449== LLd misses: 3,023,479 ( 2,584,556 rd +   438,923 wr)
==1449== D1 miss rate:  5.4% (      6.2%  +      0.1% )
==1449== LLd miss rate:  0.0% (      0.0%  +      0.0% )
==1449==
==1449== LL refs:  1,670,644,643 ( 1,665,755,956 rd + 4,888,687 wr)
==1449== LL misses:  3,025,776 ( 2,586,853 rd +   438,923 wr)
==1449== LL miss rate:  0.0% (      0.0%  +      0.0% )
```

Note that it's a similar miss rate in terms of percentages, but our algorithm had to use much more read/writes than the BLAS algorithm, which contributes to inefficiencies.

SSE (Nick)