General Matrix Multiplication Optimization

Apple Zhang

College of Computer Science and Software Engineering, Shenzhen University

December 5, 2022





Table of Contents

Optimization Methods

2 Experiments





Table of Contents

Optimization Methods

2 Experiments





Trivial Matrix Multiplication

$$C_{ij} = \sum_{r=1}^{n} A_{ir} B_{rj}. \tag{1}$$

```
1 void gemm_trivial(int m, int n, int k, double *a, int lda,
                                 double *b, int ldb,
2
                                 double *c, int ldc)
3
     // TRIVIAL MATRIX MULTIPLICATION
     for (int j = 0; j < n; j++) {
        for (int i = 0; i < m; i++) {
           for (int r = 0; r < k; r++) {
              C(i, j) += A(i, r) * B(r, j);
10
11
12
13 }
```





Simple cache optimization

Change the iteration order: $jir \rightarrow jri$.

```
void gemm_cache(int m, int n, int k, double *a, int lda,
                               double *b, int ldb,
2
                               double *c, int ldc)
3
     // OPTIMIZE THE ORDER OF MATRIX MULTIPLICATION
5
     for (int j = 0; j < n; j++) {
        for (int r = 0; r < k; r++) {
           register double brj = B(r, j);
           for (int i = 0; i < m; i++) {
               C(i, i) += A(i, r) * bri;
10
11
12
13
14 }
```





Simple cache optimization

Scan the matrix in column order, and reduce the cache miss.

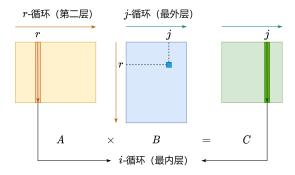


Figure: Simple cache opitmization



Compute 1x4 vector

Avoid reading the same elements from matrix A repeatedly.

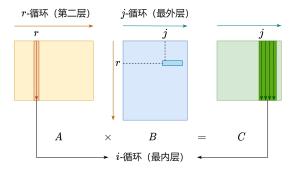


Figure: Compute four elements in each iteration.





December 5, 2022

Compute 4x4 block

Scan four rows of A and four columns of B.

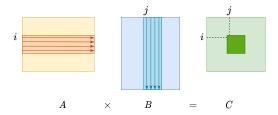


Figure: Compute a 4x4 matrix in each iteration.



Compute 4x4 block with AVX2

Consider SIMD, use AVX2 instructions for acceleration.

```
for (int r = 0; r < k; r++) {
1
             ar reg.v = mm256 load pd(&A(0, r));
2
            b0_reg.v = _mm256_broadcast_sd(br0_ptr++);
            b1 reg.v = mm256 broadcast sd(br1 ptr++);
5
            b2 \text{ reg.v} = \text{mm256 broadcast sd(br2 ptr++)};
6
            b3 req.v = _{mm256\_broadcast\_sd(br3\_ptr++)};
7
            c0 \text{ reg.v} = mm256 \text{ add pd(c0 reg.v,}
9
                      mm256 mul pd(ar reg.v. b0 reg.v));
10
            c1 \text{ reg.v} = \text{mm256} \text{ add pd(c1 reg.v,}
11
                      _mm256_mul_pd(ar_req.v, b1_req.v));
12
            c2\_reg.v = \_mm256\_add\_pd(c2\_reg.v,
13
                      _mm256_mul_pd(ar_reg.v, b2_reg.v));
14
            c3_{req.v} = _{mm256_{add_pd}(c3_{req.v},
15
                      mm256 mul pd(ar reg.v, b3 reg.v));
16
17
```





Block matrices optimization

Since for matrix multiplication, we have

$$C = AB = \begin{bmatrix} A_1 & A_2 & \cdots & A_K \end{bmatrix} \begin{bmatrix} B_1 \\ B_2 \\ \vdots \\ B_K \end{bmatrix}$$
 (2)

we can partition the matrices for better spatial locality.





Block matrices optimization

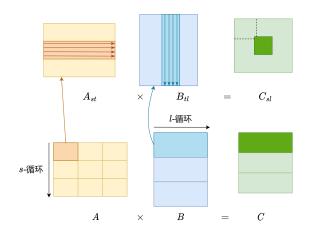


Figure: Block matrix multiplication



Reorganize the matrices

Rearrange the elements of the matrices at beginning for better spatial locality.

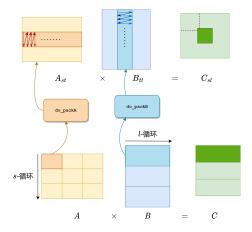


Figure: Rearrange the elements of matrices



Multi-thread with OpenMP

Use OpenMP to implement multi-thread.

```
1 void gemm pack memory(int m, int n, int k, double *a, int lda,
                                double *b, int ldb,
2
                                double *c, int ldc)
3
4
     for (int 1 = 0: 1 < k: 1 += ikk) {
5
        int lb = min(k - l. ikk):
  #ifdef OPENMP // MULTI-THREAD VIA OPENMP
        #pragma omp parallel for schedule(dynamic)
8
  #endif
        for (int s = 0; s < m; s += ikm) {
10
           int sb = min(m - s, ikm);
11
           inner_kernel_packAB(sb, n, lb, &A(s, 1), lda, &B(l, 0),
12
                ldb, &C(s, 0), ldc);
13
14
15 }
```

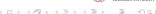


Table of Contents

Optimization Methods

2 Experiments





Evaluation

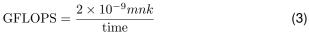
- OS: Linux (WSL)
- CPU: Intel i5-11400H

```
(base) apple@LAPTOP-INDIFA6M:~/codes/csexp1-gemm$ ./gemm test 1024 6
 #thread -> 6/12
 Trivial:
                3.006579 sec
 Cache:
              0.328501 sec
 1x4:
              0.229344 sec
 4x4:
             0.510782 sec
 4x4-avx: 0.327548 sec
          0.120619 sec
 Block:
 packOpenMP: 0.032284 sec
 max elem-wise error = 1.63425e-13
 CORRECT
 final speedup: 93.129073x
(base) apple@LAPTOP-INDIFA6M:~/codes/csexp1-gemm$
```

Figure: The results of the program.



Evaluation



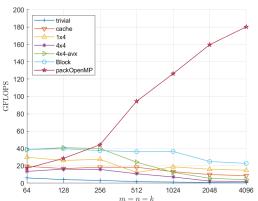


Figure: GFLOPS of each optimization.



Evaluation

But compared with Matlab (MKL)...

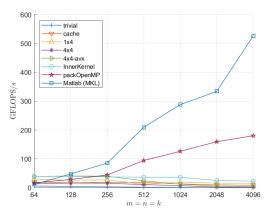


Figure: GFLOPS of each optimization.



Thanks!

My full codes can be viewed in

https://github.com/Apple-Zhang/optimize-gemm



