Late submission policy: Submission can be at the most 2 days late. There will be a 10% penalty for each day after the due date (cumulative).

In class, we described the structure of the optimized MMM code produced by ATLAS. The “computational heart” of this code is the mini-kernel that multiplies an NBxNB block of matrix A with an NBxNB block of matrix B into an NBxNB block of matrix C, where NB is chosen so that the working set of this computation fits in the cache. The mini-kernel itself is performed by repeatedly calling a micro-kernel that multiplies an MUx1 column vector of matrix A with a 1xNU row vector of matrix B into an MUXNU block of matrix C. The values of MU and NU are chosen so that the micro-kernel can be performed out of the registers. Pseudocode for the mini-kernel is shown below (note that this code assumes that NB is a multiple of MU and NU).

```c
//mini-kernel
for (int j = 0; j < NB; j += MU)
  for (int i = 0; i < NB; i += NU)
    load C[i..i+MU-1, j..j+NU-1] into registers
    for (int k = 0; k < NB; k++)
      //micro-kernel
      load A[i..i+MU-1,k] into registers
      load B[k,j..j+NU-1] into registers
      multiply A’s and B’s and add to C’s
      store C[i..i+MU-1, j..j+NU-1]
```

The data type in the matrix should be doubles.

For each optimization below:

- Compile your code in ICC with flags `-O3 -fp-model precise`.
- Submit your run to the job scheduler on Stampede at TACC - use the ‘serial’ queue. Since the values you obtain will depend a lot on the machine you use, you must use Stampede for the numbers you report.
• Report the performance of your code in GFLOPS. The number of floating point operations in matrix multiplication is \(2 \times N^3\), where \(N\times N\) is the size of each matrix and so, FLOPS is given by \(2 \times N^3/time\) (1 GFLOP = \(10^9\) FLOPS).

• Run the code to multiply matrices of various sizes (at least 5) and plot a graph of GFLOPS vs. matrix size. Explain your results briefly.

Note: Each optimization is cumulative, i.e., you implement one optimization on top of another.
As a reference, for 4096x4096 size matrices, the performance of \(i j k\) matrix multiplication version on Stampede is around 0.29 GFLOPS, the performance of matrix multiplication using BLAS is around 22 GFLOPS.

**Register-blocking**

To measure the impact of register-blocking without cache-blocking, implement register-blocking by writing a function for performing MMM, using the mini-kernel code with \(NB = N\) (you should verify that this implements MMM correctly). You can use the results in the Yotov et al. study of the ATLAS system to determine good values for \(MU\) and \(NU\), or you can experiment with different values to find good values.

Note: You will have to write some clean-up code to handle leftover pieces of the matrices when the value of \(N\) is not a multiple of your values for \(MU\) and \(NU\).

**Cache-blocking**

Modify the above code to implement both register-blocking and cache-blocking. You will have to wrap three loops around the mini-kernel to get a full MMM. Use any method you wish to determine a good value for \(NB\).

Note: You will have to write some clean-up code to handle leftover pieces of the matrices when the value of \(N\) is not a multiple of your value for \(NB\).

**Data copying**

You can improve the performance of your kernel by copying blocks of data into contiguous storage as explained in the Yotov et al. paper. Modify the above code to implement data copying (along with register-blocking and cache-blocking).

**Vectorization**

Modify the above code (along with register-blocking, cache-blocking, and data copying) to use vector registers and vector intrinsics instead of scalar registers.
**Fastest**

20 points

We will have a competition for the best performing matrix multiplication code. Your code should be correct for any square matrices; the performance will be measured only for 4096x4096 size matrices. You may use any of the techniques described above or from the literature but your code should not call any external library and you are not allowed to copy source code from anywhere. You will be graded on performance. You will not get any points if we are not able to reproduce your results.

**Implementation notes**

- Refer PHiPAC coding guidelines for writing “portable assembly language programs” in the C language.

- Stampede on TACC: Use the login node only for development - do not run or debug any executable on it. Run and debug your applications using the job scheduler. Read this to learn how to submit jobs.

- To understand the performance of your code, use PAPI to measure performance counters like L1 cache misses. To program with PAPI on stampede, run:
  
  ```
  module load papi
  module help papi
  module avail
  ```
  
  For help on using the module, run:
  
  ```
  papi
  ```
  
  Read the PAPI manual for more information, including example code.

**Deliverables**

Submit (to canvas) an archive (preferably, `.tar.gz/.tgz`) of your (one) fastest code containing all optimizations (cumulative) and a report containing all plots and analysis. Briefly describe a way to verify correctness of your code. You will not be given any points if we are not able to verify correctness for randomly chosen square matrices.