

DEPARTMENT OF MATHEMATICS AND COMPUTING  
IIT(ISM) DHANBAD  
**GPU Computing Lab Manual**  
V -M.Tech (M&C)  
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Course Type	Course Code	Name of Course	L	T	P	Credit
DP	MCC302	GPU Computing Lab	0	0	2	2

### Course Objectives

- To understand the concepts of General-Purpose GPU Programming
- To understand GPU Architecture and Performance
- To learn parallel programming on Single and Multiple GPUs
- To understand programming in OpenCL and pyCUDA

### Learning Outcomes

Upon successful completion of this course, students will: Parallel programming skills on the GPU with CUDA

Sl. No.	Name of Experiment/Lab	Learning Outcomes
1.	Basic Programming and CUDA implementation	To write basic CUDA Programs
2.	Programs -Hello world, a Kernel Call and Passing Parameters	To write program to understand host, device and global functions
3.	Vector Sum and Dot Product	To write CUDA program for vector addition and Dot Product
4.	Matrix-Matrix Multiplication	To develop skill for writing Matrix-Matrix Multiplication
5.	Use of shared memory to reduce Global Memory Traffic	To write tiled Matrix-Matrix Multiplication and understand memory management

Sl. No.	Name of Experiment/Lab	Learning Outcomes
6.	Program on warp divergence issue	To understand warp divergence through CUDA program
7.	Implementation of Min/Max/Sum reduction algorithm	To write CUDA programme for reduction algorithm and warp divergence
8.	General purpose GPU computing with PyCUDA and PyOpenCL	To understand CUDA pyCUDA, OpenCL programming structure
9.	An overview of OpenCL, Important OpenCL concepts and Basic Program Structure, NumbaPro	To Write basic OpenCL program and use of NumbaPro
10.	GPU Computing Applications- A Case Study in Machine Learning	Case study using GPU programming

### Text Books

- Programming Massively Parallel Processors: A Hands-on Approach by David B. Kirk, Wen-mei W. Hwu, Elsevier, 2010[1].

### Reference Books

1. Shane Cook: CUDA Programming: A Developer's Guide to Parallel Computing with GPUs Applications of GPU computing series Morgan Kaufmann, Newnes, 2012[2].
2. Dr. Brian Tuomanen: Hands-On GPU Programming with Python and CUDA: Explore high-performance parallel computing with CUDA, Packt Publishing Ltd, 2018.

## 1 General Information and Instructions

### Requirements

HARDWARE Linux/Windows System with Graphics Card, SOFTWARE: CUDA, OpenCL, pyCUDA [3],[4]

### System Access and Other useful commands

1. Open terminal/putty/mobaxterm
2. Use Server IP (172.16.203.23) with ssh command (**ssh -Y userID@ServerIP**)
3. Write CUDA/OpenCL code using any editor (vi, vim, gedit etc), the extension of file may be .cu
4. At the end of file .bashrc write the followings(only first time):

```
export_PATH=/usr/local/cuda/bin:$PATH
export_LD_LIBRARY_PATH=/usr/local/cuda/lib:$LD_LIBRARY_PATH
```

5. Save the .bashrc and use command **source .bashrc**
6. Compile the CUDA code using NVCC (**\$ nvcc helloWorld.cu**) or (**\$ nvcc helloWorld.cu -o hello**)
7. Execute the program using **./a.out** or **./hello**

8. Use docker images for PyCUDA program as:

- `docker run --runtime=nvidia -v $HOME:$HOME -ti bryankp/pycuda:latest bash`
- Write PyCUDA using your favorite editor (like vim, Emacs etc.)<sup>[5]</sup>
- use command `python3` to execute the code.

9. You may use Colaboratory, or “Colab” for short, is a product from Google Research <https://colab.research.google.com> for pyCUDA and PyOpenCL practicals.

## Build Applications with Makefile<sup>[6]</sup>

Makefile for the following completion of a single file with output

```
$nvcc -g -G -Xcompiler -Wall main.cpp -o main.exe
```

is given as

```
#Makefile
NVCC=/usr/local/cuda/bin/nvcc
NVCC_FLAGS=-g -G -Xcompiler -Wall
main.exe: main.cpp
$$$$$(NVCC) $(NVCC_FLAGS) $< -o $@
```

For more information please visit <https://www.gnu.org/software/make>

## Components of Lab Manual

1. Sample Experiment
2. Aim
3. Program Logic/Steps and CUDA Code
4. Expected Outcomes

## Components of Lab Report

1. Cover page (Name of Student, Admission No. with Lab Date, and Submission Date)
2. Title
3. Objectives
4. Experiments
5. Algorithm/ Flowchart
6. CUDA Program with source code file
7. Input(if any)/Output
8. Discussion and Conclusion

## 2 Experiments

### Experiment 2.1 *Programs -Hello world, a Kernel Call and Passing Parameters*

**Objectives:** display Hello world on terminal from CPU & GPU threads  
**CUDA Sample Program**

```
1000 #include <stdio.h>
1001     __global__ void helloFromGPU ()
1002 {
1003     printf("Hello World from GPU!\n");
1004 }
1005
1006 int main(int argc, char **argv)
1007 {
1008     printf("Hello World from CPU!\n");
1009
1010     helloFromGPU<<<1, 5>>>();
1011     cudaDeviceReset ();
1012     return 0;
1013 }
```

helloWorld.cu

#### Output

```
[badam@isrohpc GPULAB]$ nvcc helloWorld.cu
[badam@isrohpc GPULAB]$ ./a.out
Hello World from CPU!
Hello World from GPU!
Hello World from GPU!
Hello World from GPU!
Hello World from GPU!
Hello World from GPU!
Hello World from GPU!
```

**Lab Exercise 2.1** *Display information from CPU and GPU as per the followings:*

1. Write a CUDA C program to display your 10-10 times name from CPU and GPU respectively.
2. Write a CUDA C program to display your 4 times Course Name, Name of Experiment and Date from CPU and GPU respectively.

### Experiment 2.2 *Check Device Information*

**Objectives:** Display information on the first CUDA device, including driver version, runtime version, compute capability, bytes of global memory  
**CUDA Sample Program**

```
1000 #include <cuda_runtime.h>
1001 #include <stdio.h>
```

```

1004 int main(int argc, char **argv)
1005 {
1006     printf("%s Starting...\n", argv[0]);
1007
1008     int deviceCount = 0;
1009     cudaGetDeviceCount(&deviceCount);
1010
1011     if (deviceCount == 0)
1012     {
1013         printf("There are no available device(s) that support CUDA\n");
1014     }
1015     else
1016     {
1017         printf("Detected %d CUDA Capable device(s)\n", deviceCount);
1018     }
1019     int dev = 0, driverVersion = 0, runtimeVersion = 0;
1020     cudaSetDevice(dev);
1021     cudaDeviceProp deviceProp;
1022     cudaGetDeviceProperties(&deviceProp, dev);
1023     printf("Device %d: \"%s\"\n", dev, deviceProp.name);
1024
1025     cudaDriverGetVersion(&driverVersion);
1026     cudaRuntimeGetVersion(&runtimeVersion);
1027
1028     printf("  CUDA Driver Version / Runtime Version  %d.%d / %d.%d\n", driverVersion / 1000,
1029           (driverVersion % 100) / 10, runtimeVersion / 1000, (runtimeVersion % 100) / 10);
1030     printf("  CUDA Capability Major/Minor version number: %d.%d\n", deviceProp.major,
1031           deviceProp.minor);
1032     printf("  Total amount of global memory:  %.2f GBytes (%llu " "bytes)\n", (float)
1033           deviceProp.totalGlobalMem / pow(1024.0, 3), (unsigned long long)deviceProp.totalGlobalMem
1034           );
1035     printf("  GPU Clock rate:  %.0f MHz (%0.2f " "GHz)\n", deviceProp.clockRate * 1e-3f,
1036           deviceProp.clockRate * 1e-6f);
1037     printf("  Memory Clock rate:  %.0f Mhz\n", deviceProp.memoryClockRate * 1e-3f);
1038     printf("  Memory Bus Width:  %d-bit\n", deviceProp.memoryBusWidth);
1039
1040     if (deviceProp.l2CacheSize)
1041     {
1042         printf("  L2 Cache Size:  %d bytes\n", deviceProp.l2CacheSize);
1043     }
1044
1045     printf("  Max Texture Dimension Size (x,y,z)  1D=(%d), "
1046           "2D=(%d,%d), 3D=(%d,%d,%d)\n", deviceProp.maxTexture1D,
1047           deviceProp.maxTexture2D[0], deviceProp.maxTexture2D[1],
1048           deviceProp.maxTexture3D[0], deviceProp.maxTexture3D[1],
1049           deviceProp.maxTexture3D[2]);
1050     printf("  Max Layered Texture Size (dim) x layers  1D=(%d) x %d, "
1051           "2D=(%d,%d) x %d\n", deviceProp.maxTexture1DLayered[0],
1052           deviceProp.maxTexture1DLayered[1], deviceProp.maxTexture2DLayered[0],
1053           deviceProp.maxTexture2DLayered[1], deviceProp.maxTexture2DLayered[2]);
1054     printf("  Total amount of constant memory:  %lu bytes\n",
1055           deviceProp.totalConstMem);
1056     printf("  Total amount of shared memory per block:  %lu bytes\n", deviceProp.
1057           sharedMemPerBlock);
1058     printf("  Total number of registers available per block:  %d\n", deviceProp.regsPerBlock)
1059     ;
1060     exit(EXIT_SUCCESS);
1061 }

```

DeviceInfo.cu

Expected output could be similar to the followings

```

[badam@isrohpc GPULAB]$ nvcc DeviceInfo.cu
[badam@isrohpc GPULAB]$ ./a.out
./a.out Starting...
Detected 2 CUDA Capable device(s)
Device 0: "GeForce GTX TITAN X"
  CUDA Driver Version / Runtime Version      10.0 / 9.1
  CUDA Capability Major/Minor version number: 5.2
  Total amount of global memory:             11.93 GBytes (12806062080 bytes)
  GPU Clock rate:                            1076 MHz (1.08 GHz)
  Memory Clock rate:                         3505 Mhz
  Memory Bus Width:                          384-bit
  L2 Cache Size:                             3145728 bytes
  Max Texture Dimension Size (x,y,z)         1D=(65536), 2D=(65536,65536), 3D=(4096,4096,4096)
  Max Layered Texture Size (dim) x layers    1D=(16384) x 2048, 2D=(16384,16384) x 2048
  Total amount of constant memory:           65536 bytes
  Total amount of shared memory per block:   49152 bytes
  Total number of registers available per block: 65536

```

**Lab Exercise 2.2** Write a CUDA program to display the following device information on the terminal:

1. Warp size:
2. Maximum number of threads per multiprocessor:
3. Maximum number of threads per block:
4. Maximum sizes of each dimension of a block:
5. Maximum sizes of each dimension of a grid:
6. Maximum memory pitch:

The the heading related to above points are: warpSize, maxThreadsPerMultiProcessor, maxThreadsPerBlock, maxThreadsDim[0],deviceProp.maxThreadsDim[1], deviceProp.maxThreadsDim[2], maxGridSize[0], , maxGridSize[1], maxGridSize[2], and memPitch

**Expected output could be similar to the followings:**

```

Warp size: 32
Maximum number of threads per multiprocessor: 2048
Maximum number of threads per block: 1024
Maximum sizes of each dimension of a block: 1024 x 1024 x 64
Maximum sizes of each dimension of a grid: 2147483647 x 65535 x 65535
Maximum memory pitch: 2147483647 bytes

```

**Experiment 2.3** *Display the dimensions of grid and a thread block*

**Objectives:** Display the dimensions number of threads in block and number of block in the grid

**CUDA Sample Program**

```

1000 #include <cuda_runtime.h>
1001 #include <stdio.h>
1002
1003 __global__ void checkIndex(void)
1004 {
1005     printf("threadIdx:(%d, %d, %d)\n", threadIdx.x, threadIdx.y, threadIdx.z);
1006     printf("blockIdx:(%d, %d, %d)\n", blockIdx.x, blockIdx.y, blockIdx.z);
1007
1008     printf("blockDim:(%d, %d, %d)\n", blockDim.x, blockDim.y, blockDim.z);
1009     printf("gridDim:(%d, %d, %d)\n", gridDim.x, gridDim.y, gridDim.z);
1010 }
1011
1012 int main(int argc, char **argv)
1013 {
1014     // define total data element
1015     int nElem = 3;
1016
1017     // define grid and block structure
1018     dim3 block(3);
1019     dim3 grid((nElem + block.x - 1) / block.x);
1020
1021     // check grid and block dimension from host side
1022     printf("grid.x %d grid.y %d grid.z %d\n", grid.x, grid.y, grid.z);
1023     printf("block.x %d block.y %d block.z %d\n", block.x, block.y, block.z);
1024
1025     // check grid and block dimension from device side
1026     checkIndex<<<grid, block>>>();
1027
1028     // reset device before you leave
1029     cudaDeviceReset();
1030
1031     return(0);
1032 }

```

DimensionsGridBlock.cu

Expected output could be similar to the followings

```

[badam@isrohpc GPULAB]$ nvcc DimensionsGridBlock.cu
[badam@isrohpc GPULAB]$ ./a.out
grid.x 1 grid.y 1 grid.z 1
block.x 3 block.y 1 block.z 1
threadIdx:(0, 0, 0)
threadIdx:(1, 0, 0)
threadIdx:(2, 0, 0)
blockIdx:(0, 0, 0)
blockIdx:(0, 0, 0)
blockIdx:(0, 0, 0)
blockDim:(3, 1, 1)
blockDim:(3, 1, 1)
blockDim:(3, 1, 1)
gridDim:(1, 1, 1)
gridDim:(1, 1, 1)
gridDim:(1, 1, 1)

```

## Experiment 2.4 *Define grid and Blocks*

**Objectives:** Display grid and block structure  
**CUDA Sample Program**

```
1000 #include <cuda_runtime.h>
1001 #include <stdio.h>
1002 int main(int argc, char **argv)
1003 {
1004     // define total data element
1005     int nElem = 1024;
1006
1007     // define grid and block structure
1008     dim3 block (1024);
1009     dim3 grid ((nElem + block.x - 1) / block.x);
1010     printf("grid.x %d block.x %d \n", grid.x, block.x);
1011
1012     // reset block
1013     block.x = 512;
1014     grid.x = (nElem + block.x - 1) / block.x;
1015     printf("grid.x %d block.x %d \n", grid.x, block.x);
1016
1017     // reset block
1018     block.x = 256;
1019     grid.x = (nElem + block.x - 1) / block.x;
1020     printf("grid.x %d block.x %d \n", grid.x, block.x);
1021
1022     // reset block
1023     block.x = 128;
1024     grid.x = (nElem + block.x - 1) / block.x;
1025     printf("grid.x %d block.x %d \n", grid.x, block.x);
1026
1027     // reset device before you leave
1028     cudaDeviceReset();
1029
1030     return(0);
1031 }
```

GridBlock.cu

**Expected output could be similar to the followings**

```
[badam@isrohpc GPULAB]$ nvcc GridBlock.cu
[badam@isrohpc GPULAB]$ ./a.out
grid.x 1 block.x 1024
grid.x 2 block.x 512
grid.x 4 block.x 256
grid.x 8 block.x 128
```

## Experiment 2.5 *Vector Addition on GPU*

**Objectives:** Element wise sum of vector  
**CUDA Sample Program**



```

1000 #include <cuda_runtime.h>
1001 #include <stdio.h>
1002 #define N 10
1003 __global__ void VecAddGPU(int *a, int *b, int *c)
1004 {
1005     int i=threadIdx.x;
1006     if (i<N)
1007     {
1008         c[i]=a[i]+b[i];
1009     }
1010 }
1011
1012 int main(int argc, char **argv)
1013 {
1014     int a[N], b[N], c[N];
1015     int *dev_a, *dev_b, *dev_c;
1016     // allocate the memory on device
1017     cudaMalloc((void**)&dev_a, N*sizeof(int));
1018     cudaMalloc((void**)&dev_b, N*sizeof(int));
1019     cudaMalloc((void**)&dev_c, N*sizeof(int));
1020     for (int i=0; i<N; i++){
1021         a[i] = -i;
1022         b[i] = i * i;
1023     }
1024     //Copy data from host to device
1025     //cudaError_t=
1026     cudaError_t err=cudaMemcpy(&dev_a, a, N*sizeof(int), cudaMemcpyHostToDevice);
1027     if (err!=cudaSuccess)
1028     {
1029         printf("%s in %s at line %d\n", cudaGetErrorString(err), __FILE__, __LINE__);
1030         exit(1);
1031     }
1032     cudaMemcpy(dev_b, b, N*sizeof(int), cudaMemcpyHostToDevice);
1033     //kernel launch
1034     VecAddGPU<<<1,N>>>(dev_a, dev_b, dev_c);
1035     //Copy result from device to host
1036     cudaMemcpy(c, dev_c, N*sizeof(int), cudaMemcpyDeviceToHost);
1037
1038     for (int i=0; i<N; i++)
1039     {
1040         printf("%d+%d=%d\n", a[i], b[i], c[i]);
1041     }
1042     cudaFree(dev_a);
1043     cudaFree(dev_b);
1044     cudaFree(dev_c);
1045     return 0;
1046 }

```

VecAddGPU.cu

Expected output could be similar to the followings

```

[badam@isrohpc GPULAB]$ nvcc VecAddGPU.cu
[badam@isrohpc GPULAB]$ ./a.out
0+0=0
-1+1=0
-2+4=2
-3+9=6
-4+16=12

```

-5+25=20  
-6+36=30  
-7+49=42  
-8+64=56  
-9+81=72

**Lab Exercise 2.3** Write a CUDA program to display

1. Display grid, block and thread details for a block of size (256,3,1):

Expected output could be similar to that of in 2.4

**Lab Exercise 2.4** Write a CUDA program to display

1. Distance between two vectors  $x$  and  $y$  where  $x = \{i^2\}_{i=1}^n$ ,  $y = \{(2i + 1)\}_{i=1}^n$  and  $n = 1024$ . Also find the Euclidean norms of  $x$  and  $y$  respectively.
2. Find the standard deviation of  $y = \{(2i + 1)\}_{i=1}^n$  and  $n = 1024$ .

Experiment 2.6 *Matrix-Matrix sum on GPU*

**Objectives: Sum of two matrices**  
**CUDA Sample Program**

```
1000 #include <cuda_runtime.h>
#include <stdio.h>
1002
void initialData(float *ip, const int size)
1004 {
    int i;
1006     for(i = 0; i < size; i++)
1008     {
        ip[i] = i;
1010     }
1012     return;
}
1014 void displayMatrix(float *A, int nx,int ny)
{
1016 int idx;
    for (int i = 0; i < nx; i++)
1018     {
        for (int j = 0; j < ny; j++)
1020         {
            idx = i*ny + j;
1022             printf(" %f ",A[idx]);
        }
1024         printf("\n");
    }
1026     return;
}
```

```

1028 }
1030 // grid 1D block 1D
1032 __global__ void sumMatrixOnGPU(float *MatA, float *MatB, float *MatC, int nx,
                                int ny)
1034 {
1036     unsigned int ix = threadIdx.x + blockIdx.x * blockDim.x;
1038     if (ix < nx)
1040         for (int iy = 0; iy < ny; iy++)
1042             {
1044                 int idx = iy * nx + ix;
1046                 MatC[idx] = MatA[idx] + MatB[idx];
1048             }
1050 }
1052 int main(int argc, char **argv)
1054 {
1056     // set up data size of matrix
1058     int nx = 4;
1060     int ny = 5;
1062
1064     int nxy = nx * ny;
1066     int nBytes = nxy * sizeof(float);
1068     printf("Matrix size: nx %d ny %d\n", nx, ny);
1070
1072     // malloc host memory
1074     float *h_A, *h_B, *h_C;
1076     h_A = (float *)malloc(nBytes);
1078     h_B = (float *)malloc(nBytes);
1080     h_C = (float *)malloc(nBytes);
1082
1084     // initialize data at host side
1086     initialData(h_A, nxy);
1088     initialData(h_B, nxy);
1090
1092     // malloc device global memory
1094     float *d_MatA, *d_MatB, *d_MatC;
1096     cudaMalloc((void **)&d_MatA, nBytes);
1098     cudaMalloc((void **)&d_MatB, nBytes);
1100     cudaMalloc((void **)&d_MatC, nBytes);
1102
1104     // transfer data from host to device
1106     cudaMemcpy(d_MatA, h_A, nBytes, cudaMemcpyHostToDevice);
1108     cudaMemcpy(d_MatB, h_B, nBytes, cudaMemcpyHostToDevice);
1110
1112     // invoke kernel at host side
1114     int dimx = 32;
1116     dim3 block(dimx, 1);
1118     dim3 grid((nx + block.x - 1) / block.x, 1);
1120
1122     sumMatrixOnGPU<<<grid, block>>>(d_MatA, d_MatB, d_MatC, nx, ny);
1124     cudaDeviceSynchronize();
1126
1128     // copy kernel result back to host side
1130     cudaMemcpy(h_C, d_MatC, nBytes, cudaMemcpyDeviceToHost);
1132     displayMatrix(h_C, nx, ny);
1134
1136     // free device global memory

```

```

1092   cudaFree(d_MatA);
1094   cudaFree(d_MatB);
1094   cudaFree(d_MatC);

1096   // free host memory
1098   free(h_A);
1098   free(h_B);
1098   free(h_C);

1100   // reset device
1102   cudaDeviceReset();

1104   return (0);
}

```

sumMatrix1D.cu

**Expected output could be similar to the followings**

```

[bsk@gr02 ]$ nvcc sumMatrix1D.cu
[bsk@gr02 ]$ ./a.out
Matrix size: nx 4 ny 5
0.000000  2.000000  4.000000  6.000000  8.000000
10.000000 12.000000 14.000000 16.000000 18.000000
20.000000 22.000000 24.000000 26.000000 28.000000
30.000000 32.000000 34.000000 36.000000 38.000000

```

**Lab Exercise 2.5** Write a CUDA program to demonstrate the followings

1. Allocate Device Memory
2. Transfer Data(Matrices A and B) from host to device
3. Sum two matrices using 2D grid
4. Transfer result(Matrix C) from device to host
5. Print the result in matrix format

**Expected output could be similar to that of in 2.6**

**Lab Exercise 2.6** Write a CUDA program to demonstrate

1. Allocate Device Memory
2. Transfer Data(Matrices A and B) from host to device
3. Sum two matrices using 2D grid with different block sizes
4. Transfer result(Matrix C) from device to host
5. Print the result in matrix format

6. Show the effect of block size and grid size in terms of total run time.

## Experiment 2.7 Matrix-Matrix Multiplication on GPU

**Objectives:** Multiply of two matrices  
**CUDA Sample Program**

```
1000 #include <cuda_runtime.h>
1001 #include <stdio.h>
1002 #define N 3
1003 __global__ void MatrixMulKernel(float* MatA, float* MatB, float* MatC,
1004 int Width) {
1005     // Calculate the row index of the P element and M
1006     int Row = blockIdx.y*blockDim.y+threadIdx.y;
1007     // Calculate the column index of P and N
1008     int Col = blockIdx.x*blockDim.x+threadIdx.x;
1009     if ((Row < Width) && (Col < Width)) {
1010         float Pvalue = 0;
1011         // each thread computes one element of the block sub-matrix
1012         for (int k = 0; k < Width; ++k) {
1013             Pvalue += MatA[Row*Width+k]*MatB[k*Width+Col];
1014         }
1015         MatC[Row*Width+Col] = Pvalue;
1016     }
1017 }
1018
1019 void initialData(float *ip, const int size)
1020 {
1021     int i;
1022     for(i = 0; i < size; i++)
1023     {
1024         ip[i] = ((float)rand())/(float)(RAND_MAX);
1025     }
1026 }
1027
1028 return;
1029 }
1030 void displayMatrix(float *A, int nx,int ny)
1031 {
1032     int idx;
1033     for (int i = 0; i < nx; i++)
1034     {
1035         for (int j = 0; j < ny; j++)
1036         {
1037             idx = i*ny + j;
1038             printf(" %f ",A[idx]);
1039         }
1040         printf("\n");
1041     }
1042     return;
1043 }
1044
1045 int main(int argc, char **argv)
1046 {
1047
1048     // set up data size of matrix
1049     int Width=N;
1050     int nx = Width;
1051     int ny = Width;
1052
1053     int nxy = nx * ny;
```

```

1054     int nBytes = nxy * sizeof(float);
1055     printf("Matrix size: nx %d ny %d\n", nx, ny);
1056
1057     // malloc host memory
1058     float *h_A, *h_B,*h_C;
1059     h_A = (float *)malloc(nBytes);
1060     h_B = (float *)malloc(nBytes);
1061     h_C = (float *)malloc(nBytes);
1062
1063     // initialize data at host side
1064     initialData(h_A, nxy);
1065     initialData(h_B, nxy);
1066
1067     // malloc device global memory
1068     float *d_MatA, *d_MatB, *d_MatC;
1069     cudaMalloc((void **)&d_MatA, nBytes);
1070     cudaMalloc((void **)&d_MatB, nBytes);
1071     cudaMalloc((void **)&d_MatC, nBytes);
1072
1073     // transfer data from host to device
1074     cudaMemcpy(d_MatA, h_A, nBytes, cudaMemcpyHostToDevice);
1075     cudaMemcpy(d_MatB, h_B, nBytes, cudaMemcpyHostToDevice);
1076
1077     // invoke kernel at host side
1078     int bdimx = 16;
1079     int bdimy=16;
1080     dim3 block(bdimx, bdimy);
1081     dim3 grid((nx + block.x - 1) / block.x, (ny + block.y - 1) / block.y,1);
1082
1083     MatrixMulKernel<<<grid, block>>>(d_MatA, d_MatB, d_MatC,Width);
1084     cudaDeviceSynchronize();
1085
1086     // copy kernel result back to host side
1087     cudaMemcpy(h_C, d_MatC, nBytes, cudaMemcpyDeviceToHost);
1088     printf("Matrix A is=\n");
1089     displayMatrix(h_A, nx, ny);
1090     printf("Matrix B is=\n");
1091     displayMatrix(h_B, nx, ny);
1092     printf("The product of Matrix A and Matrix B is=\n");
1093     displayMatrix(h_C, nx, ny);
1094
1095     // free device global memory
1096     cudaFree(d_MatA);
1097     cudaFree(d_MatB);
1098     cudaFree(d_MatC);
1099
1100     // free host memory
1101     free(h_A);
1102     free(h_B);
1103     free(h_C);
1104
1105     // reset device
1106     cudaDeviceReset();
1107
1108     return (0);
1109 }

```

matrix-matrix-multGPU.cu

Expected output could be similar to the followings

```

[badam@isrohpc GPULAB]$ nvcc matrix-matrix-multGPU.cu
[badam@isrohpc GPULAB]$ ./a.out
Matrix size: nx 3 ny 3
Matrix A is=
0.000000  1.000000  2.000000
3.000000  4.000000  5.000000
6.000000  7.000000  8.000000
Matrix B is=
0.000000  1.000000  2.000000
3.000000  4.000000  5.000000
6.000000  7.000000  8.000000
The product of Matrix A and Matrix B is=
15.000000  18.000000  21.000000
42.000000  54.000000  66.000000
69.000000  90.000000  111.000000

```

**Lab Exercise 2.7** Write a CUDA program to demonstrate the followings

1. Allocate Device Memory
2. Transfer Data(Matrices A, B and C) from host to device
3. Find the Product of three matrices  $A*B*C$  using 2D grid
4. Transfer result from device to host
5. Print the result in matrix format

**Expected output could be similar to that of in 2.7**

**Lab Exercise 2.8** Write a CUDA program to demonstrate

1. Allocate Device Memory
2. Transfer Data(Matrices A and B) from host to device
3. Find the transpose (TA and TB) of matrices A and B in parallel on GPU
4. Find the product of A and B and TA and TB
5. Transfer results from device to host
6. Print the result matrices and their differences
7. Show the effect of block size and grid size in terms of total run time.

**Experiment 2.8 Use of Makefile with Main program, Distance Kernel and Header Kernel**

## Objectives: Use of Makefile CUDA Sample Program

```
1000 #include "DistKernel.h"
1001 #include <stdlib.h>
1002 #define N 16
1003 float scale(int i, int n)
1004 {
1005     return ((float)i)/(n - 1);
1006 }
1007 int main()
1008 {
1009     const float ref = 0.5f;
1010
1011     float *in = (float*)calloc(N, sizeof(float));
1012
1013     float *out = (float*)calloc(N, sizeof(float));
1014
1015     // Compute scaled input values
1016
1017     for (int i = 0; i < N; ++i)
1018     {
1019         in[i] = scale(i, N);
1020     }
1021
1022     // Compute distances for the entire array
1023
1024     distanceArray(out, in, ref, N);
1025     free(in);
1026     free(out);
1027     return 0;
1028 }
```

distanceMain.cpp

```
1000 #include "DistKernel.h"
1001 #include <stdio.h>
1002 #define TPB 16
1003 __device__ float distance(float x1, float x2)
1004 {
1005     return sqrt((x2 - x1)*(x2 - x1));
1006 }
1007
1008 __global__ void distanceKernel(float *d_out, float *d_in, float ref)
1009 {
1010     const int i = blockIdx.x*blockDim.x + threadIdx.x;
1011     const float x = d_in[i];
1012     d_out[i] = distance(x, ref);
1013     printf("i = %2d: dist from %f to %f is %f.\n", i, ref, x, d_out[i]);
1014 }
1015 void distanceArray(float *out, float *in, float ref, int len)
1016 {
1017     // Declare pointers to device arrays
1018     float *d_in = 0;
1019     float *d_out = 0;
1020     // Allocate memory for device arrays
1021     cudaMalloc(&d_in, len*sizeof(float));
1022     cudaMalloc(&d_out, len*sizeof(float));
1023     // Copy input data from host to device
1024     cudaMemcpy(d_in, in, len*sizeof(float), cudaMemcpyHostToDevice);
1025     // Launch kernel to compute and store distance values
1026     distanceKernel<<<len/TPB, TPB>>>(d_out, d_in, ref);
1027     // Copy results from device to host
```



```

1028 cudaMemcpy(out, d_out, len*sizeof(float), cudaMemcpyDeviceToHost);
    // Free the memory allocated for device arrays
1030 cudaFree(d_in);
    cudaFree(d_out);
1032 }

```

DistKernel.cu

```

1000 #ifndef KERNEL_H
    #define KERNEL_H
1002 // Kernel wrapper for computing array of distance values
    void distanceArray(float *out, float *in, float ref, int len);
1004 #endif

```

DistKernel.h

```

1000 NVCC = /usr/local/cuda/bin/nvcc
    NVCC_FLAGS = -g -G -Xcompiler -Wall
1002
    all: distanceMain.exe
1004 distanceMain.exe: distanceMain.o DistKernel.o
    $(NVCC) $^ -o $@
1006
    distanceMain.o: distanceMain.cpp DistKernel.h
1008 $(NVCC) $(NVCC_FLAGS) -c $< -o $@
1010
    DistKernel.o: DistKernel.cu DistKernel.h
    $(NVCC) $(NVCC_FLAGS) -c $< -o $@

```

Makefile

**Use make to compile the files: Output will be similar to the following**

```

[badam@isrohpc GPULAB]$ make
/usr/local/cuda/bin/nvcc -g -G -Xcompiler -Wall -c distanceMain.cpp -o distanceMain.o
/usr/local/cuda/bin/nvcc -g -G -Xcompiler -Wall -c DistKernel.cu -o DistKernel.o
/usr/local/cuda/bin/nvcc distanceMain.o DistKernel.o -o distanceMain.exe

```

**Run distanceMain.exe the output will be similar to the following**

```

[badam@isrohpc GPULAB]$ ./distanceMain.exe
i = 0: dist from 0.500000 to 0.000000 is 0.500000.
i = 1: dist from 0.500000 to 0.066667 is 0.433333.
i = 2: dist from 0.500000 to 0.133333 is 0.366667.
i = 3: dist from 0.500000 to 0.200000 is 0.300000.
i = 4: dist from 0.500000 to 0.266667 is 0.233333.
i = 5: dist from 0.500000 to 0.333333 is 0.166667.
i = 6: dist from 0.500000 to 0.400000 is 0.100000.
i = 7: dist from 0.500000 to 0.466667 is 0.033333.
i = 8: dist from 0.500000 to 0.533333 is 0.033333.
i = 9: dist from 0.500000 to 0.600000 is 0.100000.
i = 10: dist from 0.500000 to 0.666667 is 0.166667.
i = 11: dist from 0.500000 to 0.733333 is 0.233333.

```

```

i = 12: dist from 0.500000 to 0.800000 is 0.300000.
i = 13: dist from 0.500000 to 0.866667 is 0.366667.
i = 14: dist from 0.500000 to 0.933333 is 0.433333.
i = 15: dist from 0.500000 to 1.000000 is 0.500000.

```

**Lab Exercise 2.9** Write a CUDA program to demonstrate the followings

1. Write a header file for declaring add and multiply functions
2. Write a device functions to add the two Matrices in GPU
3. Then find the Square of resultant Matrix using global function
4. Transfer result from device to host
5. Print the result

**Expected output could be a single matrix**

**Lab Exercise 2.10** Write a CUDA program to demonstrate the followings

1. Write a header file for declaring functions(device and global)
2. Write a device functions to transpose of matrix A in GPU
3. Then find the product of A and  $A^T$  using global function
4. Transfer result from device to host
5. Print the result

**Expected output could be a single matrix**

### Experiment 2.9 *Tiled Matrix-Matrix Multiplication*

**Objectives:** Use tiled algorithm for Matrix-Matrix multiplication  
**CUDA Sample Program**

```

1000 #include <cuda_runtime.h>
1001 #include <stdio.h>
1002 #include "funcDef.h"
1003 #define N 8
1004 #define TILE_WIDTH 2
1005 __global__ void MatrixMulKernel(float* MatA, float* MatB, float* MatC,
1006 int Width) {
1007     //Shared memory allocation
1008     __shared__ float Mds[TILE_WIDTH][TILE_WIDTH];
1009     __shared__ float Nds[TILE_WIDTH][TILE_WIDTH];
1010     int bx = blockIdx.x; int by = blockIdx.y;

```

```

1012 int tx = threadIdx.x; int ty = threadIdx.y;
1013 int Row = by * TILE_WIDTH + ty;
1014 int Col = bx * TILE_WIDTH + tx;
1015
1016 float Pvalue = 0;
1017 for (int ph = 0; ph < Width/TILE_WIDTH; ++ph) {
1018 // Collaborative loading of A and B tiles into shared memory
1019 Mds[ty][tx] = MatA[Row*Width + ph*TILE_WIDTH + tx];
1020 Nds[ty][tx] = MatB[(ph*TILE_WIDTH + ty)*Width + Col];
1021 __syncthreads();
1022 //dot product using shared memory
1023 for (int k = 0; k < TILE_WIDTH; ++k) {
1024     Pvalue += Mds[ty][k] * Nds[k][tx];
1025 }
1026 __syncthreads();
1027 MatC[Row*Width+Col] = Pvalue;
1028 }
1029
1030 void displayMatrix(float *A, int nx,int ny)
1031 {
1032 int idx;
1033 for (int i = 0; i < nx; i++)
1034 {
1035     for (int j = 0; j < ny; j++)
1036     {
1037         idx = i*ny + j;
1038         printf(" %f ",A[idx]);
1039     }
1040     printf("\n");
1041 }
1042 return;
1043 }
1044
1045 int main(int argc, char **argv)
1046 {
1047
1048 // set up data size of matrix
1049 int Width=N;
1050 int nx = Width;
1051 int ny = Width;
1052
1053 int nxy = nx * ny;
1054 int nBytes = nxy * sizeof(float);
1055 printf("Matrix size: %d by %d\n", nx, ny);
1056 printf("Tile size: %d by %d\n", TILE_WIDTH, TILE_WIDTH);
1057
1058 // malloc host memory
1059 float *h_A, *h_B,*h_C;
1060 h_A = (float *)malloc(nBytes);
1061 h_B = (float *)malloc(nBytes);
1062 h_C = (float *)malloc(nBytes);
1063
1064 // initialize data at host side
1065 initialData(h_A, nxy);
1066 initialData(h_B, nxy);
1067
1068 // malloc device global memory
1069 float *d_MatA, *d_MatB, *d_MatC;
1070 cudaMalloc((void **)&d_MatA, nBytes);
1071 cudaMalloc((void **)&d_MatB, nBytes);
1072 cudaMalloc((void **)&d_MatC, nBytes);
1073
1074 // transfer data from host to device

```

```

1076   cudaMemcpy(d_MatA, h_A, nBytes, cudaMemcpyHostToDevice);
1077   cudaMemcpy(d_MatB, h_B, nBytes, cudaMemcpyHostToDevice);
1078
1079   // invoke kernel at host side
1080   int bdimx = TILE_WIDTH;
1081   int bdimy=TILE_WIDTH;
1082   dim3 block(bdimx, bdimy);
1083   dim3 grid((nx + block.x - 1) / block.x, (ny + block.y - 1) / block.y,1);
1084
1085   MatrixMulKernel<<<grid, block>>>(d_MatA, d_MatB, d_MatC,Width);
1086   cudaDeviceSynchronize();
1087
1088   // copy kernel result back to host side
1089   cudaMemcpy(h_C, d_MatC, nBytes, cudaMemcpyDeviceToHost);
1090   printf("Matrix A is=\n");
1091   displayMatrix(h_A, nx, ny);
1092   printf("Matrix B is=\n");
1093   displayMatrix(h_B, nx, ny);
1094   printf("The product of Matrix A and Matrix B is=\n");
1095   displayMatrix(h_C, nx, ny);
1096
1097   // free device global memory
1098   cudaFree(d_MatA);
1099   cudaFree(d_MatB);
1100   cudaFree(d_MatC);
1101
1102   // free host memory
1103   free(h_A);
1104   free(h_B);
1105   free(h_C);
1106
1107   // reset device
1108   cudaDeviceReset();
1109
1110   return (0);
1111 }

```

Tiled-Mat-multGPU.cu

```

1000 #include "funcDef.h"
1001 void initialData(float *ip, const int size)
1002 {
1003     int i;
1004
1005     for(i = 0; i < size; i++)
1006     {
1007         ip[i] = ((float)rand()/(float)(RAND_MAX));
1008     }
1009
1010     return;
1011 }

```

initialDataMatAB.cu

```

1000 #ifndef FUNCDEF_H
1001 #define FUNCDEF_H
1002 void initialData(float *ip, const int size);
1003 #endif

```

funcDef.h

```

1000 NVCC = /usr/local/cuda/bin/nvcc
      #NVCC_FLAGS = -g -G -Xcompiler -Wall
1002
      #all: distanceMain.exe
1004 Tiled-Mat-multGPU.exe: Tiled-Mat-multGPU.o initialDataMatAB.o funcDef.h
      $(NVCC) Tiled-Mat-multGPU.cu initialDataMatAB.o -o Tiled-Mat-multGPU.exe
1006 Tiled-Mat-multGPU.o: Tiled-Mat-multGPU.cu funcDef.h initialDataMatAB.o
      $(NVCC) -c Tiled-Mat-multGPU.cu initialDataMatAB.o
1008 initialDataMatAB.o: funcDef.h initialDataMatAB.cu
      $(NVCC) -c initialDataMatAB.cu
1010 clean:
      rm -f Tiled-Mat-multGPU.exe

```

Makefile2

The output will be similar to the followings

```

[badam@isrohpc GPULAB]$ make -f Makefile2
/usr/local/cuda/bin/nvcc -c Tiled-Mat-multGPU.cu initialDataMatAB.o
/usr/local/cuda/bin/nvcc Tiled-Mat-multGPU.cu initialDataMatAB.o -o Tiled-Mat-multGPU.exe
[badam@isrohpc GPULAB]$ ./Tiled-Mat-multGPU.exe
Matrix size: 8 by 8
Tile size: 2 by 2
Matrix A is=
 0.840188 0.394383 0.783099 0.798440 0.911647 0.197551 0.335223 0.768230
 0.277775 0.553970 0.477397 0.628871 0.364784 0.513401 0.952230 0.916195
 0.635712 0.717297 0.141603 0.606969 0.016301 0.242887 0.137232 0.804177
 0.156679 0.400944 0.129790 0.108809 0.998924 0.218257 0.512932 0.839112
 0.612640 0.296032 0.637552 0.524287 0.493583 0.972775 0.292517 0.771358
 0.526745 0.769914 0.400229 0.891529 0.283315 0.352458 0.807725 0.919026
 0.069755 0.949327 0.525995 0.086056 0.192214 0.663227 0.890233 0.348893
 0.064171 0.020023 0.457702 0.063096 0.238280 0.970634 0.902208 0.850920
Matrix B is=
 0.266666 0.539760 0.375207 0.760249 0.512535 0.667724 0.531606 0.039280
 0.437638 0.931835 0.930810 0.720952 0.284293 0.738534 0.639979 0.354049
 0.687861 0.165974 0.440105 0.880075 0.829201 0.330337 0.228968 0.893372
 0.350360 0.686670 0.956468 0.588640 0.657304 0.858676 0.439560 0.923970
 0.398437 0.814767 0.684219 0.910972 0.482491 0.215825 0.950252 0.920128
 0.147660 0.881062 0.641081 0.431953 0.619596 0.281059 0.786002 0.307458
 0.447034 0.226107 0.187533 0.276235 0.556444 0.416501 0.169607 0.906804
 0.103171 0.126075 0.495444 0.760475 0.984752 0.935004 0.684445 0.383188
The product of Matrix A and Matrix B is=
 1.836571 2.588725 2.984560 3.674901 3.222222 2.906765 2.833551 3.107897
 1.606581 2.257570 2.642685 2.914744 3.035270 2.768578 2.426692 2.922653
 0.980174 1.811519 2.140080 2.251573 2.072760 2.403044 1.876315 1.488523
 1.090756 1.782398 1.928511 2.370534 2.102371 1.812182 2.199538 2.137506
 1.465810 2.494980 2.685837 3.086129 3.034677 2.511495 2.702769 2.496949
 1.685890 2.520235 2.969749 3.164906 3.116278 3.173994 2.568542 2.928237
 1.434502 2.054889 2.120121 2.223845 2.140972 1.920228 1.896632 2.210005
 1.092188 1.533195 1.680568 2.035834 2.515924 1.758628 1.904237 2.138672

```

**Lab Exercise 2.11** Write a CUDA program to demonstrate the followings

1. Write a header file for declaring Matrix-Matrix Multiplication
2. Write a device functions to multiply two matrices using tiled algorithm and without tiled algorithm in GPU
3. Transfer result from device to host
4. Print the result

Expected output could be two matrices

### Experiment 2.10 *Sum reduction*

**Objectives:** Write a CUDA program for sum reduction  
**CUDA Sample Program**

```

1000 #include <cuda_runtime.h>
1001 #include <stdio.h>
1002 #define N 1000
1003 #define BD 256
1004 #define CHECK(call)
1005 {
1006     const cudaError_t error = call;
1007     if (error != cudaSuccess)
1008     {
1009         fprintf(stderr, "Error: %s:%d, ", __FILE__, __LINE__);
1010         fprintf(stderr, "code: %d, reason: %s\n", error,
1011             cudaGetErrorString(error));
1012         exit(1);
1013     }
1014 }
1015 __global__ void sumReduce(float *dev_a, float *dev_b)
1016 {
1017     //unsigned int i=blockIdx.x*blockDim.x+threadIdx.x;
1018     __shared__ float partialSum[BD];
1019     //for (unsigned int ph = 0; ph < N/BD; ++ph)
1020     {
1021         // Collaborative loading
1022         //if(ph==blockIdx.x)
1023         //{
1024         partialSum[threadIdx.x] = dev_a[blockIdx.x*blockDim.x+threadIdx.x];
1025         //}
1026         //__syncthreads();
1027         unsigned int t = threadIdx.x;
1028         for (unsigned int stride = 1; stride < blockDim.x; stride *= 2)
1029         {
1030             __syncthreads();
1031             if (t % (2*stride) == 0)
1032             {
1033                 partialSum[t] += partialSum[t+stride];
1034             }
1035         }
1036         dev_b[0] = partialSum[0];
1037     }
1038 }

```

```

1040 int main(int argc, char **argv)
    {
1042     float a[N], b[N];
        float *dev_a, *dev_b;
1044     int bdimx = BD;
        float elapsedTime;
1046     dim3 block(bdimx);
        dim3 grid((N + block.x - 1) / block.x, 1, 1);
1048     cudaEvent_t start, stop;
        CHECK(cudaEventCreate(&start));
1050     CHECK(cudaEventCreate(&stop));
        printf("Array Size is=%d\n", N);
1052     // allocate the memory on device
        CHECK(cudaMalloc((void**)&dev_a, N*sizeof(float)));
1054     CHECK(cudaMalloc((void**)&dev_b, N*sizeof(float)));
        for(int i=0; i<N; i++){
1056         a[i] = ((float)rand()/(float)(RAND_MAX));
        }
1058     //Cuda events for time measure
        CHECK(cudaEventRecord(start, 0));
1060     cudaMemcpy(dev_a, a, N*sizeof(float), cudaMemcpyHostToDevice);
        CHECK(cudaEventRecord(stop, 0));
1062     CHECK(cudaEventSynchronize(stop));
        cudaEventElapsedTime(&elapsedTime, start, stop);
1064     printf("Time to do memory transfer of array a, from host to device is %8.6f ms\n",
            elapsedTime);
        CHECK(cudaEventRecord(start, 0));
1066     sumReduce<<<<grid, block>>>(dev_a, dev_b);
        //Copy result from device to host
1068     CHECK(cudaMemcpy(b, dev_b, N*sizeof(float), cudaMemcpyDeviceToHost));
        CHECK(cudaEventRecord(stop, 0));
1070     CHECK(cudaEventSynchronize(stop));
        cudaEventElapsedTime(&elapsedTime, start, stop);
1072     printf("Time to do sum reduction is %8.6f ms\n", elapsedTime);
        printf("Sum=%f\n", b[0]);
1074     cudaEventDestroy(start);
        cudaEventDestroy(stop);
1076     cudaFree(dev_a);
        cudaFree(dev_b);
1078     return 0;
    }

```

SumReduction.cu

The output will be similar to the followings

```

[badam@isrohpc GPULAB]$ nvcc SumReduction.cu
[badam@isrohpc GPULAB]$ ./a.out
Array Size is=1000
Time to do memory transfer of array a, from host to device is 0.0 ms
Time to do sum reduction is 0.1 ms
Sum=112.796783

```

**Lab Exercise 2.12** Write a CUDA program to demonstrate the followings

1. Write a header file for declaring Error function
2. Write a device functions to do the sum reduction with less warp divergence

3. Print the execution time of the kernel and compare with classical sum reduction as given in 2.11
4. Print the result

Expected output could be similar to 2.11

Experiment 2.11 *Numerical accuracy of fusing a multiply-add*

**Objectives:** Write a CUDA program for to illustrates the effect on numerical accuracy of fusing and multiply-add into a single MAD instruction.

**CUDA Sample Program**

```

1000 #include <stdio.h>
1001 #include <stdlib.h>
1002 #define CHECK(call)
1003 {
1004     const cudaError_t error = call;
1005     if (error != cudaSuccess)
1006     {
1007         fprintf(stderr, "Error: %s:%d, ", __FILE__, __LINE__);
1008         fprintf(stderr, "code: %d, reason: %s\n", error,
1009                 cudaGetErrorString(error));
1010         exit(1);
1011     }
1012 }
1013 /*A fused multiplyadd (FMA or fmad https://en.wikipedia.org/wiki/Multiply%E2%80%93accumulate_operation) is a floating-point multiplyadd (MAD) operation performed in one
1014    step, with a single rounding. That is, where an unfused multiplyadd would compute
1015    the product b * c, round it to N significant bits, add the result to a, and round back
1016    to N significant bits, a fused multiplyadd would compute the entire expression a + (b
1017    * c) to its full precision before rounding the final result down to N significant bits
1018    . */
1019
1020 __global__ void fmad_kernel(double x, double y, double *out)
1021 {
1022     int tid = blockIdx.x * blockDim.x + threadIdx.x;
1023
1024     if (tid == 0)
1025     {
1026         *out = x * x + y;
1027     }
1028 }
1029
1030 double host_fmad_kernel(double x, double y)
1031 {
1032     return x * x + y;
1033 }
1034
1035 int main(int argc, char **argv)
1036 {
1037     double *d_out, h_out;
1038     double x = 2.891903;
1039     double y = -3.980364;
1040
1041     double host_value = host_fmad_kernel(x, y);
1042     CHECK(cudaMalloc((void **)&d_out, sizeof(double)));
1043     fmad_kernel<<<1, 32>>>(x, y, d_out);

```



```

1040 CHECK(cudaMemcpy(&h_out, d_out, sizeof(double),
                  cudaMemcpyDeviceToHost));
1042
1043 if (host_value == h_out)
1044 {
1045     printf("The device output the same value as the host.\n");
1046     printf("The device output is %.20f and the host output is=%.20f\n",h_out, host_value);
1047 }
1048 else
1049 {
1050     printf("The device output and host values are different, (host-device) is =%e.\n",
1051           fabs(host_value - h_out));
1052     printf("The device output is %.20f and the host output is=%.20f\n",h_out, host_value);
1053 }
1054 return 0;
}

```

fp-mad.cu

The output will be similar to the followings

```

[badam@isrohpc GPULAB]$ nvcc fp-mad.cu
[badam@isrohpc GPULAB]$ ./a.out
The device output and host values are different, (host-device) is =8.881784e-16.
The device output is 4.38273896140900109941 and the host output is=4.38273896140900021123

```

### Experiment 2.12 *Floating-point's inability*

**Objectives:** Write a CUDA program for to illustrates the effect of floating-point's inability due to single and double precision.

**CUDA Sample Program**

```

1000 #include <stdio.h>
1001 #include <stdlib.h>
1002 #define CHECK(call)
1003 {
1004     const cudaError_t error = call;
1005     if (error != cudaSuccess)
1006     {
1007         fprintf(stderr, "Error: %s:%d, ", __FILE__, __LINE__);
1008         fprintf(stderr, "code: %d, reason: %s\n", error,
1009               cudaGetErrorString(error));
1010         exit(1);
1011     }
1012 }
1013 __global__ void kernel(float *F, double *D)
1014 {
1015     int tid = blockIdx.x * blockDim.x + threadIdx.x;
1016
1017     if (tid == 0)
1018     {
1019         *F = 128.1;
1020         *D = 128.1;

```

```

1022 }
1024 int main(int argc, char **argv)
1026 {
1028     float *deviceF;
1028     float h_deviceF;
1028     double *deviceD;
1028     double h_deviceD;
1030
1030     float hostF = 128.1;
1032     double hostD = 128.1;
1034
1034     CHECK(cudaMalloc((void **)&deviceF, sizeof(float)));
1034     CHECK(cudaMalloc((void **)&deviceD, sizeof(double)));
1036     kernel <<<1, 32>>>(deviceF, deviceD);
1036     CHECK(cudaMemcpy(&h_deviceF, deviceF, sizeof(float),
1038                     cudaMemcpyDeviceToHost));
1038     CHECK(cudaMemcpy(&h_deviceD, deviceD, sizeof(double),
1040                     cudaMemcpyDeviceToHost));
1042
1042     printf("Host single-precision representation of 128.1 = %.20f\n", hostF);
1042     printf("Host double-precision representation of 128.1 = %.20f\n", hostD);
1044     printf("Device single-precision representation of 128.1 = %.20f\n", h_deviceF);
1044     printf("Device double-precision representation of 128.1 = %.20f\n", h_deviceD);
1046     printf("Device and host single-precision representation equal? %s\n",
1046           hostF == h_deviceF ? "yes" : "no");
1048     printf("Device and host double-precision representation equal? %s\n",
1048           hostD == h_deviceD ? "yes" : "no");
1050
1050     return 0;
1052 }

```

fpaccuracy.cu

The output will be similar to the followings

```

[badam@isrohpc GPULAB]$ ./a.out
Host single-precision representation of 128.1 = 128.10000610351562500000
Host double-precision representation of 128.1 = 128.09999999999999431566
Device single-precision representation of 128.1 = 128.10000610351562500000
Device double-precision representation of 128.1 = 128.09999999999999431566
Device and host single-precision representation equal? yes
Device and host double-precision representation equal? yes

```

**Lab Exercise 2.13** Write a CUDA program to demonstrate the followings:

1. Kahan summation algorithm
2. Write a header file for declaring Error function
3. Write a device functions to do sum of all the element of an arrays with and without sorting.
4. Write a program do demonstrate Atomic Sum and Atomic Subtraction of two numbers after the one increment of their values by two threads.
5. Print the execution time of the kernel and compare the accuracy of results 2.12.

---

**Algorithm 1** Kahan summation algorithm

---

**Require:** *input array*,  $n = \text{input array length}$ **Ensure:** *sum*

```
1: var sum  $\leftarrow$  0.0
2: var c  $\leftarrow$  0.0
3: i  $\leftarrow$  1
4: while i  $\leq$  n do
5:   var y  $\leftarrow$  input[i] - c
6:   var t  $\leftarrow$  sum + y
7:   c  $\leftarrow$  (t - sum) - y
8:   sum  $\leftarrow$  t
9:   i  $\leftarrow$  i + 1
10: end while
```

---

Expected output could be similar to 2.11

**Experiment 2.13** *Single Stream*

**Objectives:** Write a CUDA program to create a single stream with the use of paged-locked memory and asynchronous data transfer.

**CUDA Sample Program**

```
1000 #include <stdio.h>
1001 #include <stdlib.h>
1002 #define N (1024*1024)
1003 #define FULL_DATA_SIZE (N*20)
1004 #define CHECK(call)
1005 {
1006     const cudaError_t error = call;
1007     if (error != cudaSuccess)
1008     {
1009         fprintf(stderr, "Error: %s:%d, ", __FILE__, __LINE__);
1010         fprintf(stderr, "code: %d, reason: %s\n", error,
1011             cudaGetErrorString(error));
1012         exit(1);
1013     }
1014 }
1015
1016 __global__ void kernel( int *a, int *b, int *c ) {
1017     int idx = threadIdx.x + blockIdx.x * blockDim.x;
1018     if (idx < N) {
1019         c[idx] = (a[idx] + b[idx]) / 2.0;
1020     }
1021 }
1022
1023 int main( void ) {
1024     cudaDeviceProp prop;
1025     int whichDevice;
1026     CHECK( cudaGetDevice( &whichDevice ) );
1027     CHECK( cudaGetDeviceProperties( &prop, whichDevice ) );
1028     if (!prop.deviceOverlap) {
1029         printf( "Device will not handle overlaps, so no speed up from streams\n" );
1030         return 0;
1031     }
```

```

1032     }
1034     cudaEvent_t      start , stop;
1035     float            elapsedTime;
1036
1037     cudaStream_t      stream;
1038     int *host_a, *host_b, *host_c;
1039     int *dev_a, *dev_b, *dev_c;
1040
1041     // start the timers
1042     CHECK( cudaEventCreate( &start ) );
1043     CHECK( cudaEventCreate( &stop ) );
1044
1045     // initialize the stream
1046     CHECK( cudaStreamCreate( &stream ) );
1047
1048     // allocate the memory on the GPU
1049     CHECK( cudaMalloc( (void**)&dev_a,
1050                      N * sizeof(int) ) );
1051     CHECK( cudaMalloc( (void**)&dev_b,
1052                      N * sizeof(int) ) );
1053     CHECK( cudaMalloc( (void**)&dev_c,
1054                      N * sizeof(int) ) );
1055
1056     // allocate host locked memory, used to stream
1057     CHECK( cudaHostAlloc( (void**)&host_a,
1058                          FULL_DATA_SIZE * sizeof(int),
1059                          cudaHostAllocDefault ) );
1060     CHECK( cudaHostAlloc( (void**)&host_b,
1061                          FULL_DATA_SIZE * sizeof(int),
1062                          cudaHostAllocDefault ) );
1063     CHECK( cudaHostAlloc( (void**)&host_c,
1064                          FULL_DATA_SIZE * sizeof(int),
1065                          cudaHostAllocDefault ) );
1066
1067     for (int i=0; i<FULL_DATA_SIZE; i++) {
1068         host_a[i] = rand();
1069         host_b[i] = rand();
1070     }
1071
1072     CHECK( cudaEventRecord( start , 0 ) );
1073     // now loop over full data, in bite-sized chunks
1074     for (int i=0; i<FULL_DATA_SIZE; i+= N) {
1075         // copy the locked memory to the device, async
1076         CHECK( cudaMemcpyAsync( dev_a, host_a+i,
1077                               N * sizeof(int),
1078                               cudaMemcpyHostToDevice,
1079                               stream ) );
1080         CHECK( cudaMemcpyAsync( dev_b, host_b+i,
1081                               N * sizeof(int),
1082                               cudaMemcpyHostToDevice,
1083                               stream ) );
1084
1085         kernel<<<N/256,256,0,stream>>>( dev_a, dev_b, dev_c );
1086
1087         // copy the data from device to locked memory
1088         CHECK( cudaMemcpyAsync( host_c+i, dev_c,
1089                               N * sizeof(int),
1090                               cudaMemcpyDeviceToHost,
1091                               stream ) );
1092     }
1093     // copy result chunk from locked to full buffer
1094     CHECK( cudaStreamSynchronize( stream ) );
1095
1096

```

```

1098 CHECK( cudaEventRecord( stop, 0 ) );
1100 CHECK( cudaEventSynchronize( stop ) );
1100 CHECK( cudaEventElapsedTime( &elapsedTime,
1102                               start, stop ) );
1102 printf( "The single stream with ID %d was created and the total time taken for (
1102 data transfer, computation) is %8.6f ms\n", stream, elapsedTime );
1104 // cleanup the streams and memory
1104 CHECK( cudaFreeHost( host_a ) );
1106 CHECK( cudaFreeHost( host_b ) );
1106 CHECK( cudaFreeHost( host_c ) );
1108 CHECK( cudaFree( dev_a ) );
1108 CHECK( cudaFree( dev_b ) );
1110 CHECK( cudaFree( dev_c ) );
1110 CHECK( cudaStreamDestroy( stream ) );
1112
1112 return 0;
1114 }

```

SingleStream.cu

The output will be similar to the followings

```

[badam@isrohpc GPULAB]$ ./a.out
The single stream with ID 27547856 was created and
the total time taken for (data transfer, computation) is 21.894527 ms

```

**Lab Exercise 2.14** Write a CUDA program to demonstrate the followings:

1. Write a header file for declaring Error function
2. Write a CUDA program to perform the sum of arrays and to find the maximum of array using double-streams with paged-locked memory and asynchronous data transfer.
3. Print the execution time of the memory transfers and computations

Expected output could be similar to 2.13

**Experiment 2.14** Device information through a PyCUDA Program

**Objectives:** Write a PyCUDA Program for displaying GPU Device information  
**PyCUDA Sample Program**

```

1000 import pycuda.driver as drv
1000 drv.init()
1002 print ("%d device(s) found." % drv.Device.count())
1002 for ordinal in range(drv.Device.count()):
1004     dev = drv.Device(ordinal)
1004     print ("Device #%d: %s" % (ordinal, dev.name()))
1006     print (" Compute Capability: %d.%d" % dev.compute_capability())
1006     print (" Total Memory: %s KB" % (dev.total_memory()//(1024)))

```

pycudaDevInfo.py

The output will be similar to the followings

```
[badam@isrohpc GPULAB]$ python3 pycudaDevInfo.py
2 device(s) found.
Device #0: GeForce GTX TITAN X
  Compute Capability: 5.2
  Total Memory: 12505920 KB
Device #1: GeForce GTX 680
  Compute Capability: 3.0
  Total Memory: 2047552 KB
[badam@isrohpc GPULAB]$
```

### Experiment 2.15 *Simple PyCUDA Program*

**Objectives:**Demonstrate workflow with PyCUDA program for computing 2 times of all the elements of a Matrix

**PyCUDA Sample Program**

```
1000 import pycuda.driver as cuda
1001 import pycuda.autoinit
1002 from pycuda.compiler import SourceModule
1003 import numpy
1004 a=numpy.random.randn(5,5)
1005 a=a.astype(numpy.float32)
1006
1007 a_gpu=cuda.mem_alloc(a.nbytes)
1008 cuda.memcpy_htod(a_gpu,a)
1009
1010 mod=SourceModule("""
1011     __global__ void doubleMatrix(float *a)
1012     {
1013         int idx=threadIdx.x+threadIdx.y*4;
1014         a[idx]*=2;
1015     }
1016 """)
1017 func=mod.get_function("doubleMatrix")
1018 func(a_gpu,block=(5,5,1))
1019 a_doubled=numpy.empty_like(a)
1020 cuda.memcpy_dtoh(a_doubled,a_gpu)
1021 print("Original Matrix")
1022 print(a)
1023 print("Double Matrix After PyCUDA Execution")
1024 print(a_doubled)
```

SimpleProg.py

The output will be similar to the followings

```
[badam@isrohpc GPULAB]$ python3 SimpleProg.py
Original Matrix
[[ 0.03760774  1.604201  -0.39076883  0.30589864 -1.1251544 ]
 [-1.1846496  -0.29308596 -1.1174204  -0.24432212  2.4030788 ]
 [ 0.10457493 -0.5216858  -0.30098775  1.8247517  0.22829506]
```

```

[-0.34854513  1.0348203  -0.2788698  -0.69622207 -1.5858915 ]
[-0.03784035  0.7266018   0.36599657 -0.7192867  -1.5846182 ]]
Double Matrix After PyCUDA Execution
[[ 0.07521549  3.208402  -0.78153765  0.6117973  -2.2503088 ]
 [-2.3692992  -0.5861719  -2.2348409  -0.48864424  4.8061576 ]
 [ 0.20914985 -1.0433716  -0.6019755  3.6495035  0.45659012]
 [-0.69709027  2.0696406  -0.5577396  -1.3924441  -3.171783  ]
 [-0.0756807   0.7266018   0.36599657 -0.7192867  -1.5846182 ]]

```

Lab Exercise 2.15 Write a PyCUDA program to demonstrate the followings:

1. Allocate host and device memories for three matrices A, B, C
2. Transfer data of matrices A, B from host to device
3. Performance Matrix and Matrix multiplication

Expected output could be similar to 2.13

Experiment 2.16 PyCUDA with GPUArray and NumbaPro

Objectives: (1) PyCUDA GPUArray program for computing 2 times of all the elements of a Matrix

Objectives: (2) PyCUDA NumbaPro program for Matrix-Matrix Multiplication  
PyCUDA Sample Program

```

1000 import pycuda.gpuarray as gpuarray
1001 import pycuda.driver as cuda
1002 import pycuda.autoinit
1003 from numba import guvectorize
1004 import numpy as np
1005 a_gpu = gpuarray.to_gpu(np.random.randn(5,5).astype(np.float32))
1006 a_doubled = (2*a_gpu).get()
1007 print ("ORIGINAL MATRIX")
1008 print (a_gpu)
1009 print ("DOUBLED MATRIX AFTER PyCUDA EXECUTION USING GPUARRAY CALL")
1010 print (a_doubled)
1011
1012
1013
1014 @guvectorize(['void(int64[:, :], int64[:, :], int64[:, :])'],
1015             '(m,n),(n,p)->(m,p)')
1016 def matmul(A, B, C):
1017     m, n = A.shape
1018     n, p = B.shape
1019     for i in range(m):
1020         for j in range(p):
1021             C[i, j] = 0
1022             for k in range(n):
1023                 C[i, j] += A[i, k] * B[k, j]
1024
1025 dim = 10
1026 A = np.random.randint(dim, size=(dim, dim))
1027 B = np.random.randint(dim, size=(dim, dim))

```

```

1028
1030 C = matmul(A, B)
1032 print("INPUT MATRIX A")
1034 print(":\n%s" % A)
1036 print("INPUT MATRIX B")
print(":\n%s" % B)
print("RESULT MATRIX C = A*B")
print(":\n%s" % C)

```

GPUArrayNumbaPro.py

The output will be similar to the followings

```

[badam@isrohpc GPULAB]$ python3 GPUArrayNumbaPro.py
ORIGINAL MATRIX
[[-0.9645335  0.68201447 -0.1265066  -0.4648311  0.04720533]
 [-0.16570863  1.1681297  0.4138771  1.9788967  0.8305167 ]
 [-0.46075127  1.0986797  -0.08944886  1.4339496  1.0959916 ]
 [-1.1752167  1.1341211  -1.0027288  0.38856977 -0.49455154]
 [-0.51226526  0.3633518  -0.44432122  0.33385575  0.3319665 ]]
DOUBLED MATRIX AFTER PyCUDA EXECUTION USING GPUARRAY CALL
[[-1.929067  1.3640289 -0.2530132 -0.9296622  0.09441066]
 [-0.33141726  2.3362594  0.8277542  3.9577935  1.6610334 ]
 [-0.92150253  2.1973593 -0.17889772  2.8678992  2.1919832 ]
 [-2.3504333  2.2682421 -2.0054576  0.77713954 -0.9891031 ]
 [-1.0245305  0.7267036 -0.88864243  0.6677115  0.663933  ]]
INPUT MATRIX A
:
[[1 4 9 4 0 4 6 8 2 0]
 [6 6 1 6 7 9 0 6 5 3]
 [1 6 3 0 6 4 8 7 0 4]
 [2 4 8 2 3 8 6 7 4 9]
 [7 0 2 4 2 6 3 9 9 6]
 [3 2 7 4 2 8 3 4 9 7]
 [9 6 1 0 2 5 4 9 5 4]
 [4 5 6 2 8 7 9 2 1 4]
 [8 8 0 8 2 7 0 2 9 4]
 [2 1 9 5 7 7 5 5 2 0]]
INPUT MATRIX B
:
[[4 1 0 2 3 0 2 0 9 6]
 [2 8 9 3 1 0 1 0 2 7]
 [5 7 3 9 2 1 4 0 9 0]
 [0 1 7 9 4 5 4 8 6 5]
 [8 4 5 2 7 3 6 9 2 0]
 [4 1 9 1 1 8 8 9 4 4]
 [3 3 6 1 6 1 0 6 5 7]
 [7 9 3 2 1 2 6 3 6 4]
 [7 7 7 9 4 2 4 0 7 4]
 [4 7 2 6 3 9 9 4 2 7]]
RESULT MATRIX C = A*B

```



```

:
[[161 208 201 175 97 87 146 128 230 152]
 [222 214 274 191 143 173 243 222 238 209]
 [184 213 206 109 128 111 166 175 166 176]
 [243 284 264 230 149 200 269 208 266 231]
 [237 234 218 216 142 169 242 173 277 221]
 [227 245 256 257 142 189 251 180 270 206]
 [215 233 206 145 119 115 190 130 243 228]
 [212 207 255 164 172 147 201 227 228 200]
 [185 204 278 232 135 160 208 167 251 244]
 [203 187 228 187 144 130 197 211 242 135]]

```

### Experiment 2.17 *PyCUDA with GPUArray and NumbaPro*

Objectives: Device information using pyOpenCL  
PyOpenCL Sample Program

```

1000 import pyopencl as cl
1002
1004 def print_device_info() :
1004     print('\n' + '=' * 60 + '\nOpenCL Platforms and Devices')
1006     for platform in cl.get_platforms():
1006         print('=' * 60)
1008         print('Platform - Name: ' + platform.name)
1008         print('Platform - Vendor: ' + platform.vendor)
1010         print('Platform - Version: ' + platform.version)
1010         print('Platform - Profile: ' + platform.profile)
1012
1012         for device in platform.get_devices():
1014             print(' ' + '-' * 56)
1014             print(' Device - Name: ' \
1016                 + device.name)
1016             print(' Device - Type: ' \
1018                 + cl.device_type.to_string(device.type))
1018             print(' Device - Max Clock Speed: {0} Mhz'\
1020                 .format(device.max_clock_frequency))
1020             print(' Device - Compute Units: {0}'\
1022                 .format(device.max_compute_units))
1022             print(' Device - Local Memory: {0:.0f} KB'\
1024                 .format(device.local_mem_size/1024.0))
1024             print(' Device - Constant Memory: {0:.0f} KB'\
1026                 .format(device.max_constant_buffer_size/1024.0))
1026             print(' Device - Global Memory: {0:.0f} GB'\
1028                 .format(device.global_mem_size/1073741824.0))
1028             print(' Device - Max Buffer/Image Size: {0:.0f} MB'\
1030                 .format(device.max_mem_alloc_size/1048576.0))
1030             print(' Device - Max Work Group Size: {0:.0f}'\
1032                 .format(device.max_work_group_size))
1032
1032     print('\n')
1034 if __name__ == "__main__":
1034     print_device_info()

```

DeviceInfoPyCL.py

The output will be similar to the followings

```
[badam@isrohpc GPULAB]$ python3 DeviceInfoPyCL.py
```

```
=====
OpenCL Platforms and Devices
=====
```

```
Platform - Name:  NVIDIA CUDA
Platform - Vendor: NVIDIA Corporation
Platform - Version: OpenCL 1.2 CUDA 10.0.292
Platform - Profile: FULL_PROFILE
```

```
-----
Device - Name:  GeForce GTX TITAN X
Device - Type:  ALL | GPU
Device - Max Clock Speed:  1076 Mhz
Device - Compute Units:  24
Device - Local Memory:  48 KB
Device - Constant Memory: 64 KB
Device - Global Memory: 12 GB
Device - Max Buffer/Image Size: 3053 MB
Device - Max Work Group Size: 1024
-----
```

```
-----
Device - Name:  GeForce GTX 680
Device - Type:  ALL | GPU
Device - Max Clock Speed:  1058 Mhz
Device - Compute Units:  8
Device - Local Memory:  48 KB
Device - Constant Memory: 64 KB
Device - Global Memory: 2 GB
Device - Max Buffer/Image Size: 500 MB
Device - Max Work Group Size: 1024
-----
```

Lab Exercise 2.16 *Write a PyOpen program to demonstrate the followings:*

1. *Allocate host and device memories for three matrices A, B, C*
2. *Transfer data of matrices A, B from host to device*
3. *Performance addition of two Matrices*

Expected output could be similar to 2.17

Experiment 2.18 *NBody simulation*

Objectives: To implement a very simple two-stage NBody simulation  
CUDA Sample Program

```
1000 #include "common.h"
      #include <sys/time.h>
1002 #include <stdio.h>
```

```

#include <stdlib.h>
1004 #include <omp.h>
#include SINGLE_PREC
1006 #ifndef DOUBLE_PREC
#define SINGLE_PREC
1008 #endif
#endif

1010
#ifdef SINGLE_PREC
1012
typedef float real;
1014 #define MAX_DIST 200.0f
#define MAX_SPEED 100.0f
1016 #define MASS 2.0f
#define DT 0.00001f
1018 #define LIMIT_DIST 0.000001f
#define POW(x,y) powf(x,y)
1020 #define Sqrt(x) sqrtf(x)

1022 #else // SINGLE_PREC

1024 typedef double real;
#define MAX_DIST 200.0
1026 #define MAX_SPEED 100.0
#define MASS 2.0
1028 #define DT 0.00001
#define LIMIT_DIST 0.000001
1030 #define POW(x,y) pow(x,y)
#define Sqrt(x) sqrt(x)
1032

#endif // SINGLE_PREC

1034
#ifdef VALIDATE
1036
/**
1038 * Host implementation of the NBody simulation.
**/
1040 static void h_nbody_update_velocity(real *px, real *py,
real *vx, real *vy,
1042 real *ax, real *ay,
int N, int *exceeded_speed, int id)
1044 {
real total_ax = 0.0f;
1046 real total_ay = 0.0f;

real my_x = px[id];
real my_y = py[id];
1050
int i = (id + 1) % N;
1052
while (i != id)
1054 {
real other_x = px[i];
1056 real other_y = py[i];

real rx = other_x - my_x;
real ry = other_y - my_y;
1058

real dist2 = rx * rx + ry * ry;
1060

if (dist2 < LIMIT_DIST)
1062 {
dist2 = LIMIT_DIST;
1064 }
1066
}
}

```

```

1068     real dist6 = dist2 * dist2 * dist2;
1070     real s = MASS * (1.0f / Sqrt(dist6));
1072     total_ax += rx * s;
1074     total_ay += ry * s;
1076     i = (i + 1) % N;
1078 }
1080 ax[id] = total_ax;
1082 ay[id] = total_ay;
1084 vx[id] = vx[id] + ax[id];
1086 vy[id] = vy[id] + ay[id];
1088 real v = Sqrt(POW(vx[id], 2.0) + POW(vy[id], 2.0));
1090 if (v > MAX_SPEED)
1092 {
1094     *exceeded_speed = *exceeded_speed + 1;
1096 }
1098 }
1099 static void h_nbody_update_position(real *px, real *py,
1101                                     real *vx, real *vy,
1103                                     int N, int *beyond_bounds, int id)
1105 {
1107     px[id] += (vx[id] * DT);
1109     py[id] += (vy[id] * DT);
1111     real dist = Sqrt(POW(px[id], 2.0) + POW(py[id], 2.0));
1113     if (dist > MAX_DIST)
1115     {
1117         *beyond_bounds = 1;
1119     }
1121 }
1123 #endif // VALIDATE
1125 /**
1127  * CUDA implementation of simple NBody.
1129  */
1131 __global__ void d_nbody_update_velocity(real *px, real *py,
1133                                         real *vx, real *vy,
1135                                         real *ax, real *ay,
1137                                         int N, int *exceeded_speed)
1139 {
1141     int tid = blockIdx.x * blockDim.x + threadIdx.x;
1143     real total_ax = 0.0f;
1145     real total_ay = 0.0f;
1147     if (tid >= N) return;
1149     real my_x = px[tid];
1151     real my_y = py[tid];
1153     int i = (tid + 1) % N;
1155     while (i != tid)
1157     {
1159         real other_x = px[i];
1161         real other_y = py[i];
1163         real rx = other_x - my_x;
1165         real ry = other_y - my_y;

```

```

1134     real dist2 = rx * rx + ry * ry;
1136     if (dist2 < LIMIT_DIST)
1138     {
1140         dist2 = LIMIT_DIST;
1142     }
1144     real dist6 = dist2 * dist2 * dist2;
1146     real s = MASS * (1.0f / SQRT(dist6));
1148     total_ax += rx * s;
1150     total_ay += ry * s;
1152     i = (i + 1) % N;
1154 }
1156 ax[tid] = total_ax;
1158 ay[tid] = total_ay;
1160 vx[tid] = vx[tid] + ax[tid];
1162 vy[tid] = vy[tid] + ay[tid];
1164 real v = SQRT(POW(vx[tid], 2.0) + POW(vy[tid], 2.0));
1166 if (v > MAX_SPEED)
1168 {
1170     atomicAdd(exceeded_speed, 1);
1172 }
1174 }
1176
1178 __global__ void d_nbody_update_position(real *px, real *py,
1180                                         real *vx, real *vy,
1182                                         int N, int *beyond_bounds)
1184 {
1186     int tid = blockIdx.x * blockDim.x + threadIdx.x;
1188     if (tid >= N) return;
1190     px[tid] += (vx[tid] * DT);
1192     py[tid] += (vy[tid] * DT);
1194     real dist = SQRT(POW(px[tid], 2.0) + POW(py[tid], 2.0));
1196     if (dist > MAX_DIST)
1198     {
1200         *beyond_bounds = 1;
1202     }
1204 }
1206
1208 static void print_points(real *x, real *y, int N)
1210 {
1212     int i;
1214     for (i = 0; i < N; i++)
1216     {
1218         printf("%.20e %.20e\n", x[i], y[i]);
1220     }
1222 }
1224
1226 int main(int argc, char **argv)
1228 {
1230     int i;
1232     int N = 30720;
1234     int block = 256;
1236     int iter, niters = 50;

```

```

1198     real *d_px, *d_py;
1199     real *d_vx, *d_vy;
1200     real *d_ax, *d_ay;
1201     real *h_px, *h_py;
1202     int *d_exceeded_speed, *d_beyond_bounds;
1203     int exceeded_speed, beyond_bounds;
1204 #ifndef VALIDATE
1205     int id;
1206     real *host_px, *host_py;
1207     real *host_vx, *host_vy;
1208     real *host_ax, *host_ay;
1209     int host_exceeded_speed, host_beyond_bounds;
1210 #endif // VALIDATE
1211
1212 #ifdef SINGLE_PREC
1213     printf("Using single-precision floating-point values\n");
1214 #else // SINGLE_PREC
1215     printf("Using double-precision floating-point values\n");
1216 #endif // SINGLE_PREC
1217
1218 #ifdef VALIDATE
1219     printf("Running host simulation. WARNING, this might take a while.\n");
1220 #endif // VALIDATE
1221
1222     h_px = (real *) malloc(N * sizeof(real));
1223     h_py = (real *) malloc(N * sizeof(real));
1224
1225 #ifdef VALIDATE
1226     host_px = (real *) malloc(N * sizeof(real));
1227     host_py = (real *) malloc(N * sizeof(real));
1228     host_vx = (real *) malloc(N * sizeof(real));
1229     host_vy = (real *) malloc(N * sizeof(real));
1230     host_ax = (real *) malloc(N * sizeof(real));
1231     host_ay = (real *) malloc(N * sizeof(real));
1232 #endif // VALIDATE
1233
1234     for (i = 0; i < N; i++)
1235     {
1236         real x = (rand() % 200) - 100;
1237         real y = (rand() % 200) - 100;
1238
1239         h_px[i] = x;
1240         h_py[i] = y;
1241 #ifdef VALIDATE
1242         host_px[i] = x;
1243         host_py[i] = y;
1244 #endif // VALIDATE
1245     }
1246
1247     CHECK(cudaMalloc((void **)&d_px, N * sizeof(real)));
1248     CHECK(cudaMalloc((void **)&d_py, N * sizeof(real)));
1249
1250     CHECK(cudaMalloc((void **)&d_vx, N * sizeof(real)));
1251     CHECK(cudaMalloc((void **)&d_vy, N * sizeof(real)));
1252
1253     CHECK(cudaMalloc((void **)&d_ax, N * sizeof(real)));
1254     CHECK(cudaMalloc((void **)&d_ay, N * sizeof(real)));
1255
1256     CHECK(cudaMalloc((void **)&d_exceeded_speed, sizeof(int)));
1257     CHECK(cudaMalloc((void **)&d_beyond_bounds, sizeof(int)));
1258
1259     CHECK(cudaMemcpy(d_px, h_px, N * sizeof(real), cudaMemcpyHostToDevice));
1260     CHECK(cudaMemcpy(d_py, h_py, N * sizeof(real), cudaMemcpyHostToDevice));
1261
1262     CHECK(cudaMemset(d_vx, 0x00, N * sizeof(real)));

```

```

1264     CHECK(cudaMemset(d_vy, 0x00, N * sizeof(real)));
1265 #ifndef VALIDATE
1266     memset(host_vx, 0x00, N * sizeof(real));
1267     memset(host_vy, 0x00, N * sizeof(real));
1268 #endif // VALIDATE
1269
1270     CHECK(cudaMemset(d_ax, 0x00, N * sizeof(real)));
1271     CHECK(cudaMemset(d_ay, 0x00, N * sizeof(real)));
1272 #ifndef VALIDATE
1273     memset(host_ax, 0x00, N * sizeof(real));
1274     memset(host_ay, 0x00, N * sizeof(real));
1275 #endif // VALIDATE
1276
1277     double start = seconds();
1278
1279     for (iter = 0; iter < niters; iter++)
1280     {
1281         CHECK(cudaMemset(d_exceeded_speed, 0x00, sizeof(int)));
1282         CHECK(cudaMemset(d_beyond_bounds, 0x00, sizeof(int)));
1283
1284         d_nbody_update_velocity<<<(N / block, block)>>>(d_px, d_py, d_vx, d_vy,
1285             d_ax, d_ay, N, d_exceeded_speed);
1286         d_nbody_update_position<<<(N / block, block)>>>(d_px, d_py, d_vx, d_vy,
1287             N, d_beyond_bounds);
1288     }
1289
1290     CHECK(cudaDeviceSynchronize());
1291     double exec_time = seconds() - start;
1292
1293 #ifndef VALIDATE
1294     for (iter = 0; iter < niters; iter++)
1295     {
1296         printf("iter=%d\n", iter);
1297         host_exceeded_speed = 0;
1298         host_beyond_bounds = 0;
1299
1300         #pragma omp parallel for
1301         for (id = 0; id < N; id++)
1302         {
1303             h_nbody_update_velocity(host_px, host_py, host_vx, host_vy,
1304                 host_ax, host_ay, N, &host_exceeded_speed,
1305                 id);
1306         }
1307
1308         #pragma omp parallel for
1309         for (id = 0; id < N; id++)
1310         {
1311             h_nbody_update_position(host_px, host_py, host_vx, host_vy,
1312                 N, &host_beyond_bounds, id);
1313         }
1314     }
1315 #endif // VALIDATE
1316
1317     CHECK(cudaMemcpy(&exceeded_speed, d_exceeded_speed, sizeof(int),
1318         cudaMemcpyDeviceToHost));
1319     CHECK(cudaMemcpy(&beyond_bounds, d_beyond_bounds, sizeof(int),
1320         cudaMemcpyDeviceToHost));
1321     CHECK(cudaMemcpy(h_px, d_px, N * sizeof(real), cudaMemcpyDeviceToHost));
1322     CHECK(cudaMemcpy(h_py, d_py, N * sizeof(real), cudaMemcpyDeviceToHost));
1323
1324     print_points(h_px, h_py, 10);
1325     printf("Any points beyond bounds? %s, # points exceeded velocity %d/%d\n",

```

```

1328         beyond_bounds > 0 ? "true" : "false", exceeded_speed,
1329         N);
1330     printf("Total execution time %f s\n", exec_time);
1331
1332 #ifndef VALIDATE
1333     double error = 0.0;
1334
1335     for (i = 0; i < N; i++)
1336     {
1337         double dist = sqrt(pow(h_px[i] - host_px[i], 2.0) +
1338                             pow(h_py[i] - host_py[i], 2.0));
1339         error += dist;
1340     }
1341
1342     error /= N;
1343     printf("Error = %.20e\n", error);
1344 #endif // VALIDATE
1345
1346     return 0;
1347 }

```

nbodyGPU.cu

```

1000 #include <sys/time.h>
1001
1002 #ifndef COMMON_H
1003 #define COMMON_H
1004
1005 #define CHECK(call)
1006 {
1007     \
1008     const cudaError_t error = call;
1009     \
1010     if (error != cudaSuccess)
1011     {
1012         \
1013         fprintf(stderr, "Error: %s:%d, ", __FILE__, __LINE__);
1014         \
1015         fprintf(stderr, "code: %d, reason: %s\n", error,
1016                 cudaGetErrorString(error));
1017         \
1018         exit(1);
1019     }
1020 }
1021
1022 #define CHECK_CUBLAS(call)
1023 {
1024     \
1025     cublasStatus_t err;
1026     \
1027     if ((err = (call)) != CUBLAS_STATUS_SUCCESS)
1028     {
1029         \
1030         fprintf(stderr, "Got CUBLAS error %d at %s:%d\n", err, __FILE__,
1031                 __LINE__);
1032         \
1033     }
1034 }

```



```

1024     exit(1);
1025 }
1026 }
1027 #define CHECK_CURAND(call)
1028 {
1029     \
1030     curandStatus_t err;
1031     \
1032     if ((err = (call)) != CURAND_STATUS_SUCCESS)
1033     {
1034         \
1035         fprintf(stderr, "Got CURAND error %d at %s:%d\n", err, __FILE__,
1036             \
1037             __LINE__);
1038         \
1039         exit(1);
1040     }
1041 }
1042 #define CHECK_CUFFT(call)
1043 {
1044     \
1045     cufftResult err;
1046     \
1047     if ((err = (call)) != CUFFT_SUCCESS)
1048     {
1049         \
1050         fprintf(stderr, "Got CUFFT error %d at %s:%d\n", err, __FILE__,
1051             \
1052             __LINE__);
1053         \
1054         exit(1);
1055     }
1056 }
1057 #define CHECK_CUSPARSE(call)
1058 {
1059     \
1060     cusparseStatus_t err;
1061     \
1062     if ((err = (call)) != CUSPARSE_STATUS_SUCCESS)
1063     {
1064         \
1065         fprintf(stderr, "Got error %d at %s:%d\n", err, __FILE__, __LINE__);
1066         \
1067         cudaError_t cuda_err = cudaGetLastError();
1068         \
1069         if (cuda_err != cudaSuccess)
1070         {
1071             \
1072             fprintf(stderr, "  CUDA error \"%s\" also detected\n",

```

```

1060     \
1061     \         cudaGetErrorString(cuda_err));
1062     \     }
1063     \     exit(1);
1064     \ }
1065 }
1066 inline double seconds()
1067 {
1068     struct timeval tp;
1069     struct timezone tzp;
1070     int i = gettimeofday(&tp, &tzp);
1071     return ((double)tp.tv_sec + (double)tp.tv_usec * 1.e-6);
1072 }
1073
1074 #endif // _COMMON_H

```

common.h

The output will be similar to the followings

```

[badam@isrohpc GPULAB]$ nvcc nbodyGPU.cu
[badam@isrohpc GPULAB]$ ./a.out
Using single-precision floating-point values
8.29334487915039062500e+01 -1.40215482711791992188e+01
7.68599090576171875000e+01 1.49849863052368164062e+01
9.29465179443359375000e+01 3.49770202636718750000e+01
8.58866806030273437500e+01 -8.03452301025390625000e+00
-5.09801368713378906250e+01 -7.88040084838867187500e+01
6.19454917907714843750e+01 -7.29803695678710937500e+01
-9.94156074523925781250e+00 -4.09241943359375000000e+01
6.29818801879882812500e+01 2.60619792938232421875e+01
3.99804077148437500000e+01 -7.39264450073242187500e+01
7.19351501464843750000e+01 3.59459609985351562500e+01
Any points beyond bounds? true, # points exceeded velocity 28529/30720
Total execution time 1.251625 s

```

Lab Exercise 2.17 Write a CUDA program to demonstrate the followings:

1. Allocate host and device memories
2. Transfer data from host to device
3. Perform N-body problem for large number of bodies( such a Solar System with thousands of Asteroids)

Expected output could be similar to 2.18

## References

- [1] Kirk, D. & Hwu, W. *Programming Massively Parallel Processors: A Hands-on Approach* (Elsevier Science, 2016).
- [2] Sanders, J. & Kandrot, E. *CUDA by Example: An Introduction to General-Purpose GPU Programming, Portable Documents* (Pearson Education, 2010).
- [3] Stroustrup, B. *The C++ Programming Language: 4th Edition*. Always learning (Addison-Wesley, 2013).
- [4] Chapman, S. *FORTRAN FOR SCIENTISTS & ENGINEERS* (McGraw-Hill Education, 2017).
- [5] Zaccane, G. *Python Parallel Programming Cookbook* (Packt Publishing, 2015).
- [6] Stevens, W. & Rago, S. *Advanced Programming in the UNIX Environment*. Addison-Wesley professional computing series (Addison-Wesley, 2008).