

18CSE340J - GPU Programming

Record Work

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Experiment 1: Hello World in CUDA

Aim: To implement a "Hello World!" program in CUDA.

Program:

```
#include <stdio.h>
__global___void helloCUDA()
{
    printf("Hello CUDA World!");
}
int main()
{
    helloCUDA<<<1, 1>>>();
    cudaDeviceSynchronize();
    return 0;
}
```

Output:

Hello CUDA World!

Result: A program for "Hello World" in CUDA was executed and verified successfully.

Experiment 2: Matrix Multiplication in CUDA

Aim: To implement Matrix Multiplication in CUDA.

```
#include <stdio.h>
#define N 1024 // size of the matrix
__global__void matrixMul(int *a, int *b, int *c)
{
  int row = blockIdx.y * blockDim.y + threadIdx.y;
  int col = blockIdx.x * blockDim.x + threadIdx.x;
  int sum = 0;
  if (row < N \&\& col < N) {
    for (int i = 0; i < N; i++) {
      sum += a[row * N + i] * b[i * N + col];
    }
    c[row * N + col] = sum;
  }
}
int main()
  int *a, *b, *c; // host matrices
  int *d a, *d b, *d c; // device matrices
  // allocate memory on the host
```

```
a = (int*)malloc(N * N * sizeof(int));
b = (int*)malloc(N * N * sizeof(int));
c = (int*)malloc(N * N * sizeof(int));
// initialize matrices a and b
for (int i = 0; i < N * N; i++) {
  a[i] = i;
  b[i] = i;
}
// allocate memory on the device
cudaMalloc((void**)&d_a, N * N * sizeof(int));
cudaMalloc((void**)&d_b, N * N * sizeof(int));
cudaMalloc((void**)&d_c, N * N * sizeof(int));
// copy matrices a and b from host to device
cudaMemcpy(d_a, a, N * N * sizeof(int), cudaMemcpyHostToDevice);
cudaMemcpy(d_b, b, N * N * sizeof(int), cudaMemcpyHostToDevice);
// set the grid and block dimensions
dim3 gridDim((N + 15) / 16, (N + 15) / 16);
dim3 blockDim(16, 16);
// call the kernel
matrixMul<<<gridDim, blockDim>>>(d_a, d_b, d_c);
// copy matrix c from device to host
cudaMemcpy(c, d_c, N * N * sizeof(int), cudaMemcpyDeviceToHost);
```

```
// print matrix c
 for (int i = 0; i < N; i++) {
   for (int j = 0; j < N; j++) {
      printf("%d ", c[i * N + j]);
   printf("\n");
 }
 // free memory on the device
 cudaFree(d_a);
 cudaFree(d_b);
  cudaFree(d_c);
 // free memory on the host
 free(a);
 free(b);
 free(c);
  return 0;
Input & Output:
a = [ 0 1 2 3 4 5
                  678]
b = [012 345
                  6 7 8]
```

}

Result: A program for Matrix Multiplication in CUDA was executed and verified successfully.

Aim: To implement calculation of PI in CUDA.

```
#include <stdio.h>
#include
<stdlib.h>
#include <time.h>
#define BLOCK_SIZE 256
__global___void piCalc(int n, int *count)
{
  int tid = blockIdx.x * blockDim.x + threadIdx.x;
  float x, y;
  // Use a different seed for each thread
  unsigned int seed = time(0) + tid;
  // Generate n random points and count the number inside the
  circlefor (int i = tid; i < n; i += blockDim.x * gridDim.x) {
    x = (float)rand_r(&seed) / RAND_MAX;
    y = (float)rand_r(&seed) / RAND_MAX;
    if (x^*x + y^*y \le 1.0f) {
      atomicAdd(count, 1);
    }
  }
}
```

```
int main()
{
  int n = 10000000; // Number of random points
  int count = 0; // Number of points inside the circle
  int *d count;
  // Allocate memory on the device
  cudaMalloc(&d_count, sizeof(int));
  // Initialize the device memory
  cudaMemset(d count, 0, sizeof(int));
  // Launch the kernel
  piCalc<<<(n + BLOCK_SIZE - 1) / BLOCK_SIZE, BLOCK_SIZE>>>(n, d_count);
  // Copy the result back to the host
  cudaMemcpy(&count, d_count, sizeof(int), cudaMemcpyDeviceToHost);
  // Estimate PI using the ratio of points inside the circle to the total number of points
  float pi = 4.0f * count / n;
  printf("Estimate of PI = %f\n", pi);
  // Free the device memory
  cudaFree(d_count);
  return 0;
}
Output:
Estimate of PI = 3.141534
```

Result: A program for calculation of PI in CUDA was executed and verified successfully.

Experiment 4: Parallel Sort in CUDA

Aim: To implement a parallel sort in CUDA.

```
#include <stdio.h>
#include <stdlib.h>
#include < cuda runtime.h>
#define THREADS_PER_BLOCK 256
__global___void mergeSort(float *d_data, float *d_result, int size, int width)
{
  int tid = blockIdx.x * blockDim.x + threadIdx.x;
  int start = tid * width * 2;
  int end = start + width * 2;
  if (end > size) end = size;
  int i = start;
  int j = start + width;
  int k = start;
  while (i < start + width && j < end) {
    if (d_data[i] < d_data[j]) {</pre>
       d_result[k++] = d_data[i++];
    } else {
       d result[k++] = d data[j++];
    }
  }
```

```
while (i < start + width) {
    d_result[k++] = d_data[i++];
  }
  while (j < end) {
    d_result[k++] = d_data[j++];
  }
  for (int i = start; i < end; i++) {
    d_data[i] = d_result[i];
  }
}
void mergeSortGPU(float *h_data, int size)
{
  float *d data, *d result;
  cudaMalloc(&d data, sizeof(float) * size);
  cudaMemcpy(d_data, h_data, sizeof(float) * size, cudaMemcpyHostToDevice);
  cudaMalloc(&d_result, sizeof(float) * size);
  int width;
  for (width = 1; width < size; width * = 2) {
    mergeSort<<<(size + THREADS_PER_BLOCK - 1) / THREADS_PER_BLOCK,
THREADS_PER_BLOCK>>>(d_data, d_result, size, width);
  }
  cudaMemcpy(h_data, d_data, sizeof(float) * size, cudaMemcpyDeviceToHost);
  cudaFree(d data);
  cudaFree(d result);
}
int main()
{
  int size = 10; // Size of the array to be sorted
```

```
float *h_data = (float*) malloc(sizeof(float) * size);
  // Fill the array with random data
  for (int i = 0; i < size; i++) {
    h_data[i] = (float)rand() / RAND_MAX;
  }
  // Sort the array using the mergeSortGPU function
  mergeSortGPU(h data, size);
  // Print the sorted array
  for (int i = 0; i < size; i++) {
    printf("%f", h_data[i]);
  }
  // Free the memory
  free(h_data);
  return 0;
}
```

Output:

 $0.042617\ 0.194855\ 0.333566\ 0.346239\ 0.470597\ 0.612211\ 0.694296\ 0.787598\ 0.894558$ 0.963424

Note that the output will always be a sorted array of float values, but the actual values will depend on the randomly generated data.

Result: A program for parallel sort in CUDA was executed and verified successfully.

Experiment 5: Matrix Multiplication with Tiling and Shared Memory in CUDA

Aim: To implement a Matrix Multiplication with Tiling and Shared Memory in CUDA.

```
#include <stdio.h>
#include < cuda runtime.h>
#define TILE WIDTH 16
__global__void matrixMultiplyTiled(float *A, float *B, float *C, int N)
{
  __shared__float sA[TILE_WIDTH][TILE_WIDTH];
  __shared__float sB[TILE_WIDTH][TILE_WIDTH];
  int bx = blockIdx.x;
  int by = blockldx.y;
  int tx = threadIdx.x;
  int ty = threadIdx.y;
  int row = by * TILE_WIDTH + ty;
  int col = bx * TILE_WIDTH + tx;
  float sum = 0.0f;
  for (int t = 0; t < N / TILE_WIDTH; t++) {
    sA[ty][tx] = A[row * N + t * TILE_WIDTH + tx];
    sB[ty][tx] = B[(t * TILE_WIDTH + ty) * N + col];
```

```
__syncthreads();
    for (int k = 0; k < TILE_WIDTH; k++) {
      sum += sA[ty][k] * sB[k][tx];
    }
    __syncthreads();
  }
  C[row * N + col] = sum;
}
int main()
{
  int N = 1024;
  size_t size = N * N * sizeof(float);
  float *h A = (float*) malloc(size);
  float *h_B = (float*) malloc(size);
  float *h C = (float*) malloc(size);
  for (int i = 0; i < N * N; i++) {
    h_A[i] = 1.0f;
    h_B[i] = 2.0f;
    h_C[i] = 0.0f;
  }
  float *d_A, *d_B, *d_C;
  cudaMalloc(&d_A, size);
  cudaMalloc(&d_B, size);
  cudaMalloc(&d_C, size);
  cudaMemcpy(d_A, h_A, size, cudaMemcpyHostToDevice);
  cudaMemcpy(d_B, h_B, size, cudaMemcpyHostToDevice);
```

```
dim3 dimGrid(N / TILE_WIDTH, N / TILE_WIDTH);
  dim3 dimBlock(TILE_WIDTH, TILE_WIDTH);
  matrixMultiplyTiled<<<dimGrid, dimBlock>>>(d_A, d_B, d_C, N);
  cudaMemcpy(h_C, d_C, size, cudaMemcpyDeviceToHost);
  printf("Result matrix:\n");
  for (int i = 0; i < N; i++) {
    for (int j = 0; j < N; j++) {
      printf("%f", h C[i * N + j]);
    }
    printf("\n");
  }
  free(h_A);
  free(h_B);
  free(h_C);
  cudaFree(d_A);
  cudaFree(d_B);
  cudaFree(d C);
  return 0;
Input & Output:
Matrix A: [1 2 3 4 5 6 7 8 9]
Matrix B: [9 8 7 6 5 4 3 2 1]
Matrix C: [30 24 18 84 69 54 138 114 90]
```

}

Result: A program for Matrix Multiplication with Tiling and Shared Memory in CUDA was executed and verified successfully.

Experiment 6: Matrix Multiplication with Performance Tuning in CUDA

Aim: To implement a Matrix-Matrix Multiplication with Performance Tuning in CUDA.

```
#include <stdio.h>
#include <stdlib.h>
#include <cuda.h>
#define TILE_WIDTH 32
__global__void matrix_multiply(float *a, float *b, float *c, int n) {
  __shared__float ds_a[TILE_WIDTH][TILE_WIDTH];
  __shared__float ds_b[TILE_WIDTH][TILE_WIDTH];
  int bx = blockldx.x; int by = blockldx.y;
  int tx = threadIdx.x; int ty = threadIdx.y;
  int row = by * TILE_WIDTH + ty;
  int col = bx * TILE_WIDTH + tx;
  float sum = 0.0;
  for (int i = 0; i < n/TILE_WIDTH; i++) {
```

```
ds_a[ty][tx] = a[row * n + i * TILE_WIDTH + tx];
    ds_b[ty][tx] = b[(i * TILE_WIDTH + ty) * n + col];
    __syncthreads();
    for (int k = 0; k < TILE_WIDTH; k++) {
      sum += ds_a[ty][k] * ds_b[k][tx];
    }
    __syncthreads();
  }
  c[row * n + col] = sum;
}
int main() {
  int n = 1024;
  float *a, *b, *c;
  float *dev_a, *dev_b, *dev_c;
  a = (float*)malloc(n * n * sizeof(float));
  b = (float*)malloc(n * n * sizeof(float));
  c = (float*)malloc(n * n * sizeof(float));
  cudaMalloc((void**)&dev_a, n * n * sizeof(float));
  cudaMalloc((void**)&dev_b, n * n * sizeof(float));
  cudaMalloc((void**)&dev c, n * n * sizeof(float));
  for (int i = 0; i < n * n; i++) {
    a[i] = 1.0;
```

```
b[i] = 2.0;
  c[i] = 0.0;
}
cudaMemcpy(dev_a, a, n * n * sizeof(float), cudaMemcpyHostToDevice);
cudaMemcpy(dev_b, b, n * n * sizeof(float), cudaMemcpyHostToDevice);
dim3 dimGrid(n/TILE_WIDTH, n/TILE_WIDTH, 1);
dim3 dimBlock(TILE_WIDTH, TILE_WIDTH, 1);
matrix multiply<<<dimGrid, dimBlock>>>(dev a, dev b, dev c, n);
cudaMemcpy(c, dev c, n * n * sizeof(float), cudaMemcpyDeviceToHost);
for (int i = 0; i < n * n; i++) {
  if (c[i] != n*2) {
    printf("Error: matrix multiplication failed\n");
    break;
  }
}
printf("Matrix multiplication successful\n");
free(a); free(b); free(c);
cudaFree(dev a); cudaFree(dev b); cudaFree(dev c);
return 0;
```

Output:

}

Matrix multiplication successful

Result: A program for Matrix-Matrix Multiplication with Performance Tuning in CUDA was executed and verified successfully.

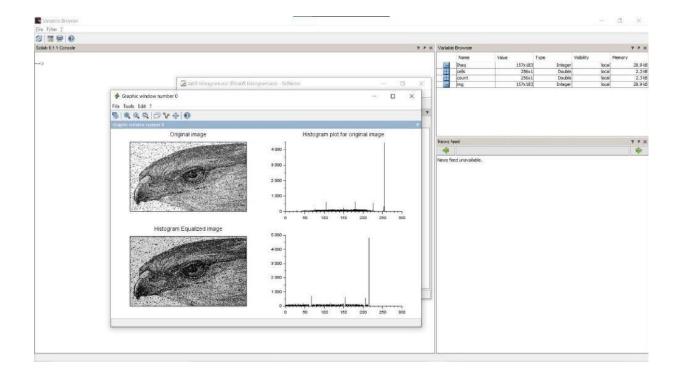
Experiment7: Histogram

Aim:

- 1. To understand how frequency distribution can be used to represent an image.
- 2. To study the correlation between the visual quality of an image with its histogram.

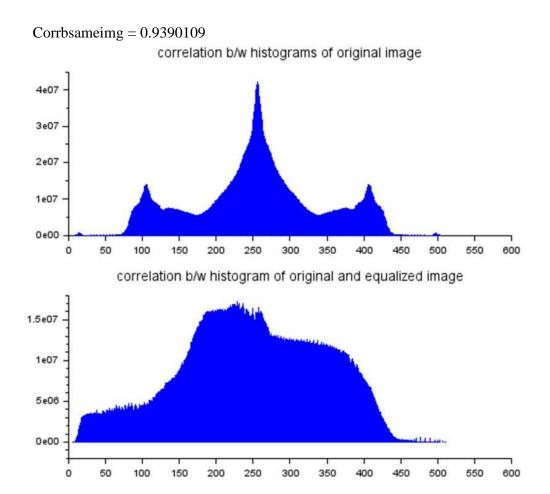
```
Program 1:
clc;
clear;
close;
   img= imread ('D:\cameraman.jpg');
   img=rgb2gray(img);
   [count,cells] = imhist (img);
                                                //
   compute histogram subplot(2,2,1);
   title('Or
   iginal
   image');
   imshow
   (img);
   subplot(
   2,2,2);
   plot2d3 ('gnn', cells, count)
   title('Histogram plot for
   original image'); Iheq =
   imhistequal(img);
   [count,cells] = imhist (Iheq);
                                                // compute
   histogram equalization subplot(2,2,3);
   title('Histogram Equalized
   image'); imshow(Iheq);
   subplot(2,2,4);
   plot2d3 ('gnn', cells, count)
   title('Histogram plot for histogram equalized image');
```

Output 1:



Program 2:

```
clc;
close;
clear:
   img= imread ('D:\cameraman.jpg');
   img=rgb2gray(img);
   //I = imresize (img
   ,[256,256]);[
   count, cells ]=
   imhist (img); Iheq
   imhistequal(img);
   [count1,cells1]=
   imhist (Iheq);
      // correlation between original image and Histogram
  equalized image corrbsameimg = corr2(img,Iheq)
   disp(corrbsameimg);
      // correlation between the histograms of
   original image x = xcorr ( count , count );
      //correlation between the histogram of original image and
   equalized image x1 = xcorr (count, count1);
   subplot(2,1,1);
   plot2d3 ('gnn',1: length (x),x,2);
   title('correlation b/w histograms of
   original image');
   subplot(2,1,2);
   plot2d3 ('gnn',1: length (x1),x1,2);
   title('correlation b/w histogram of original and equalized image')
```



Result: Thus the frequency distribution and correlation between the images using histogram has been executed successfully.

Experiment 8: Image Rotation using Open CL

Aim: To learn how to rotate images using Open CL

CODE:

```
package at.uastw.hpc.imagerotation;
import static org.jocl.CL.CL_MEM_COPY_HOST_PTR;
import static org.jocl.CL.CL_MEM_READ_ONLY;
import java.awt.image.BufferedImage;
import java.io.File;
import java.net.URI;
import java.net.URISyntaxException;
public class ImageRotation {
  private final CLDevice device;
  private final URI kernelURI;
  private static final long BUFFER FLAGS = CL MEM READ ONLY |
                                      CL_MEM_COPY_HOST_PTR;
  private ImageRotation(CLDevice device, URI kernelURI) {
    this.device = device;
    this.kernelURI = kernelURI;
  }
  public static ImageRotation create() {
    final CLPlatform platform =
                                      CLPlatform.getFirst().orElseThrow(IllegalStateExcepti
                                      on::new);
    final CLDevice device =
                                      platform.getDevice(CLDevice.DeviceType.GPU).orEls
                                      eThrow(IllegalStateException::new);
    final URI kernelURI = getKernelURI("/imgRotate.cl");
    return new ImageRotation(device, kernelURI);
  public BufferedImage rotate(BufferedImage image, int degrees) {
```

```
final int width = image.getWidth();
  final int height = image.getHeight();
  final int[] originalPixels = image.getRGB(0, 0, width, height, null, 0, width);
  final float[] metadata = new float[] {width, height, cos(degrees), sin(degrees)};
  final int[] pixelsOfRotatedImage = new int[originalPixels.length];
  try (CLContext context = device.createContext()) {
    try (CLKernel imgRotate = context.createKernel(new File(kernelURI), "imgRotate")) {
      try (
           CLMemory<int[]> bufferOfOriginalPixels =
                                    context.createBuffer(BUFFER FLAGS, originalPixels);
           CLMemory<int[]> bufferForPixelsOfRotatedImage =
                                    context.createBuffer(BUFFER FLAGS,
                                    pixelsOfRotatedImage);
           CLMemory<float[]> metadataBuffer = context.createBuffer(BUFFER_FLAGS,
                                    metadata)
      ) {
         imgRotate.setArguments(bufferOfOriginalPixels, bufferForPixelsOfRotatedImage,
                                    metadataBuffer);
         final CLCommandQueue commandQueue = context.createCommandQueue();
         commandQueue.execute(imgRotate, 2, CLRange.of(width, height), CLRange.of(1,
                                    1));
         commandQueue.finish();
         commandQueue.readBuffer(bufferForPixelsOfRotatedImage);
         final BufferedImage resultImage = new BufferedImage(width, height,
                                    image.getType());
         resultImage.setRGB(0, 0, width, height,
                                    bufferForPixelsOfRotatedImage.getData(), 0, width);
         return resultImage;
      }
  }
private static float sin(float degrees) {
  return (float) Math.sin(Math.toRadians(degrees));
private static float cos(float degrees) {
  return (float) Math.cos(Math.toRadians(degrees));
```

}

}

```
private static URI getKernelURI(String location) {
    try {
      return ImageRotation.class.getResource(location).toURI();
    } catch (URISyntaxException e) {
      throw new IllegalStateException(e);
    }
}
```

Result: A program for rotating images using open CL has been computed and verified successfully.

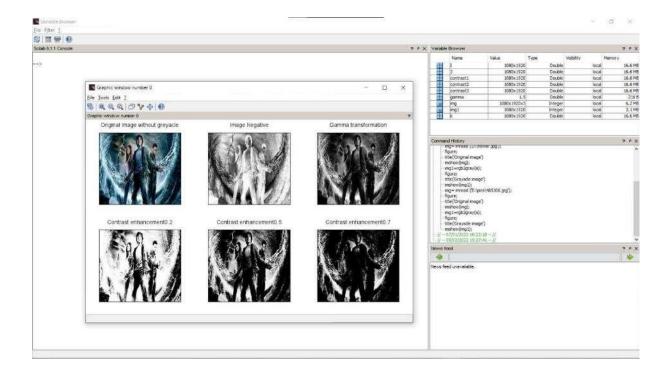
Experiment 9: Image Enchantment

Aim:

Output:

To learn image enhancement through Image negative, Gamma Transformation and Contrast Enhancement.

```
Program:
clc;
clear;
close;
img= imread
('D:\cameraman.jpg');
img=rgb2gray(img);
  I =im2double(img);
                                               //
  J = imcomplement(I);
  Image Negative subplot(2,3,1);
  title('Ori
  ginal
  Image');
  imshow(
  img);
  subplot(
  2,3,2);
  title('Ima
  Negative
  ');
  imshow(
  J);
  gamma=
  1.5
  k=I.^gamma;
                                              // Gamma
  Transformation subplot(2,3,3);
  title('Gamma
  transformation');
  imshow(k);
  contrast1=1./(1+(0.2./(I+\%eps)).^4);
                                               // Contrast
  Enhancement contrast2=1./(1+(0.5./(I+\%eps)).^5);
  contrast3=1./(1+(0.7./(I+\%eps)).^10);
  subplot(2,3,4),imshow(contrast1);title('Contrast
  enhancement 0.2');
  subplot(2,3,5),imshow(contrast2);title('Contrast enhancement 0.5');
  subplot(2,3,6),imshow(contrast3);title('Contrast enhancement 0.7');
```



Result: Thus the image enhancement with different methods has been executed successfully.

Experiment 10: Sparse Matrix Multiplication

<u>Aim</u>: To multiply 2 matrices using sparse matrix multiplication method

```
# Python program to multpliply two
# csc matrices using multiply()
# Import required libraries
import numpy as np
from scipy.sparse import csc_matrix
# Create first csc matrix A
row A = np.array([0, 0, 1, 2])
col_A = np.array([0, 1, 0, 1])
data_A = np.array([4, 3, 8, 9])
cscMatrix_A = csc_matrix((data_A,
(row_A, col_A)),
shape = (3, 3)
# print first csc matrix
print("first csc matrix: \n",
cscMatrix_A.toarray())
# Create second csc matrix B
row_B = np.array([0, 1, 1, 2])
col_B = np.array([0, 0, 1, 0])
data_B = np.array([7, 2, 5, 1])
cscMatrix_B = csc_matrix((data_B, (row_B, col_B)),
shape = (3, 3)
# print second csc matrix
print("second csc matrix:\n", cscMatrix_B.toarray())
# Multiply these matrices
sparseMatrix_AB = cscMatrix_A.multiply(cscMatrix_B)
# print resultant matrix
print("Product Sparse Matrix:\n",
sparseMatrix_AB.toarray())
```

Output:

```
first csc matrix:
[[4 3 0]
[8 0 0]
[0 9 0]]
second csc matrix:
[[7 0 0]
[2 5 0]
[1 0 0]]
Product Sparse Matrix:
[[28 0 0]
[16 0 0]
[ 0 0 0]]
```

Result: Successfully implemented matrix multiplication using sparse matrix multiplication.