CUDA Acceleration for Edge Detection

ECE 5720

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Motivation

- Edge detection is commonly implemented in various fields
  - Computer vision
  - Machine learning
- Applications
  - Depth perception
  - Object detection
    - Fingerprint recognition
    - License plate detection
- Computation speed is crucial for such implementation
  - Lower the high-resolution digital images for faster convolution operations.

Edge detection

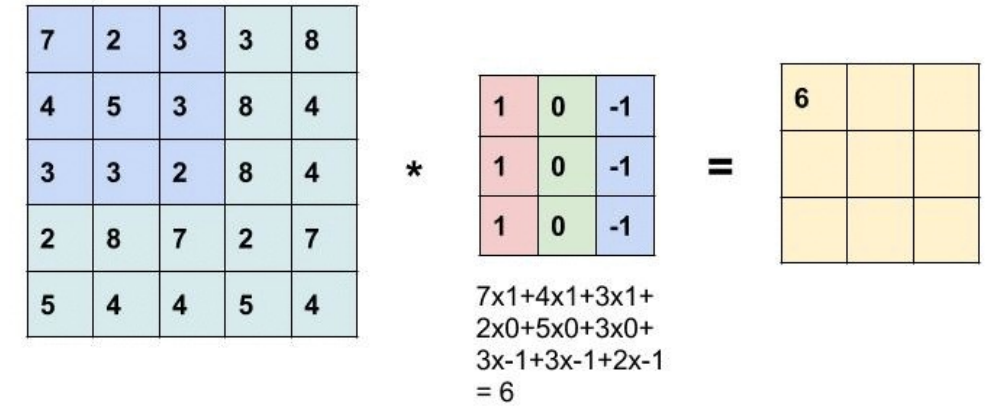
- What is edge detection?
  - Simplest definition: Sharp changes in the image brightness
    - The points where the image brightness varies sharply are called **the edges** of the image

Source: Wiki

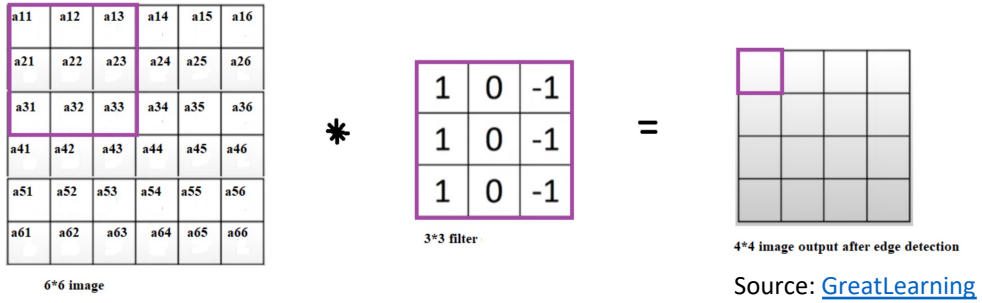
- Three major edge detection methods
  - Sobel, Canny, Fuzzy logic

Edge detection (cont.)

- How to process the high-resolution digital image?
  - Convolution comes to the rescue



Source: medium



Source: GreatLearning

- For the purple square:
  - $output =$

$$a_{11} * 1 + a_{12} * 0 + a_{13} * (-1) + a_{21} * 1 + a_{22} * 0 + a_{23} * (-1) + a_{31} * 1 + a_{32} * 0 + a_{33} * (-1)$$

Sobel Edge detection

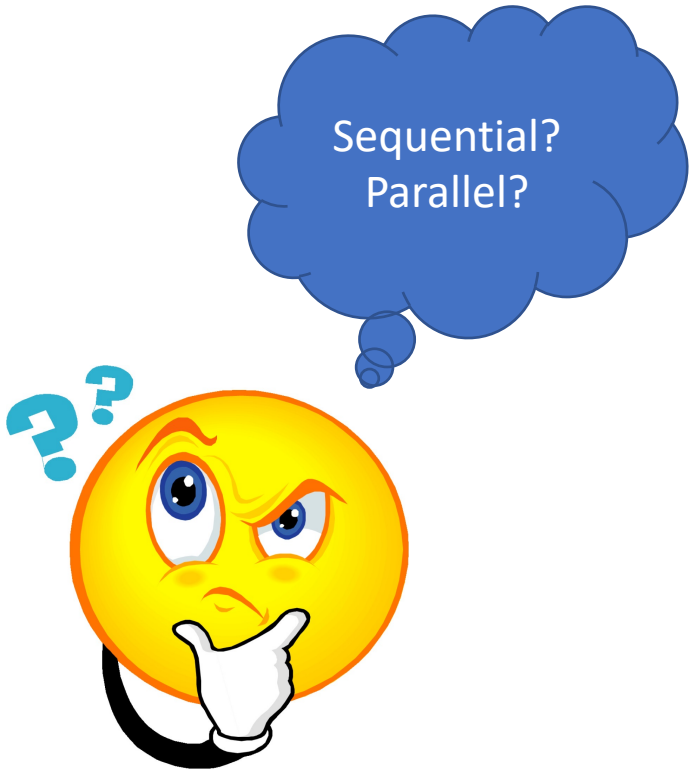
- Detect the horizontal and vertical edges in images
- Filters:
  - $G_x = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix} * A$ (vertical edge detection)
  - $G_y = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & 2 & -1 \end{bmatrix} * A$ (horizontal edge detection)

Sobel Edge detection (cont.)

- The Sobel kernels can compute the gradient with smoothing
- $G_x = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} * ([+1 \;\; 0 \;\; -1] * A) \quad \Longrightarrow$ increasing the right direction
- $G_y = \begin{bmatrix} +1 \\ 0 \\ -1 \end{bmatrix} * ([1 \;\; 2 \;\; 1] * A) \quad \Longrightarrow$ increasing the down direction
- For each point in the image, the gradient approximation can be combined as the gradient magnitude:
  - $G = \sqrt{G_x^2 + G_y^2}$

Implementation

```
1  function sobel_edge_detector(A)   % A as the image
2    Gx = [-1 0 1; -2 0 2; -1 0 1]
3    Gy = [-1 -2 -1; 0 0 0; 1 2 1]
4
5    image_row = size(A, 1)
6    image_col = size(A, 2)
7    mag = zeros(A)  % Magnitude
8
9    for i=1:image_row-2
10     for j=1:image_col-2
11       S1 = sum(sum(Gx.*A(i:i+2, j:j+2)))
12       S2 = sum(sum(Gy.*A(i:i+2, j:j+2)))
13       mag(i+1, j+1) = sqrt(S1^2 + S2^2)
14     end for
15   end for
16
17   threshold = 70
18   output_image = max(mag, threshold)
19   output_image(output_image == round(threshold)) = 0;
20   return output_image
21 end function
```



Parallel approaches

- OpenMP

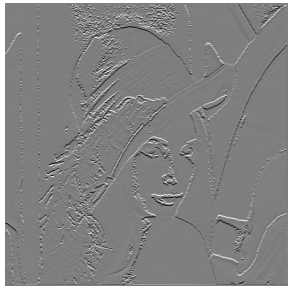
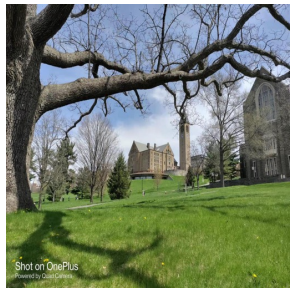
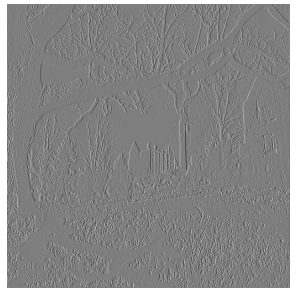
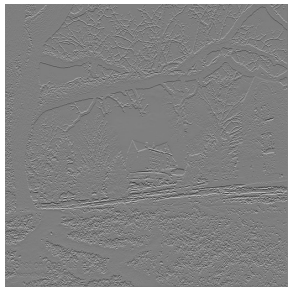
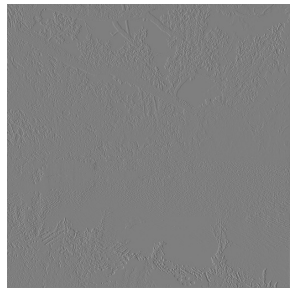
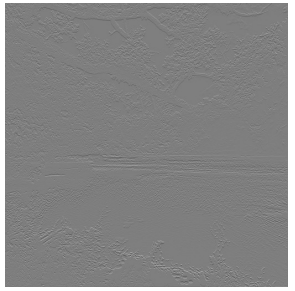
```
1  void sobel_edge_filter_omp(image, out_image){
2  int gx, gy, i, j;
3  #pragma omp parallel for default(shared) private( i, j, gx, gy)
4      for( i = 1; i < image->height - 2; i++ ) {
5          for( j = 1; j < image->width - 2; j++ ) {
6              gx = convolution(i,j,sobel_filter_x);
7              gy = convolution(i,j,sobel_filter_y);
8              out_image->imageData[i][j] = sqrt(gx*gx + gy*gy);
9              out_image->gx[i][j] = gx;
10             out_image->gy[i][j] = gy;
11         }
12     }
13 }
```

Parallel approaches

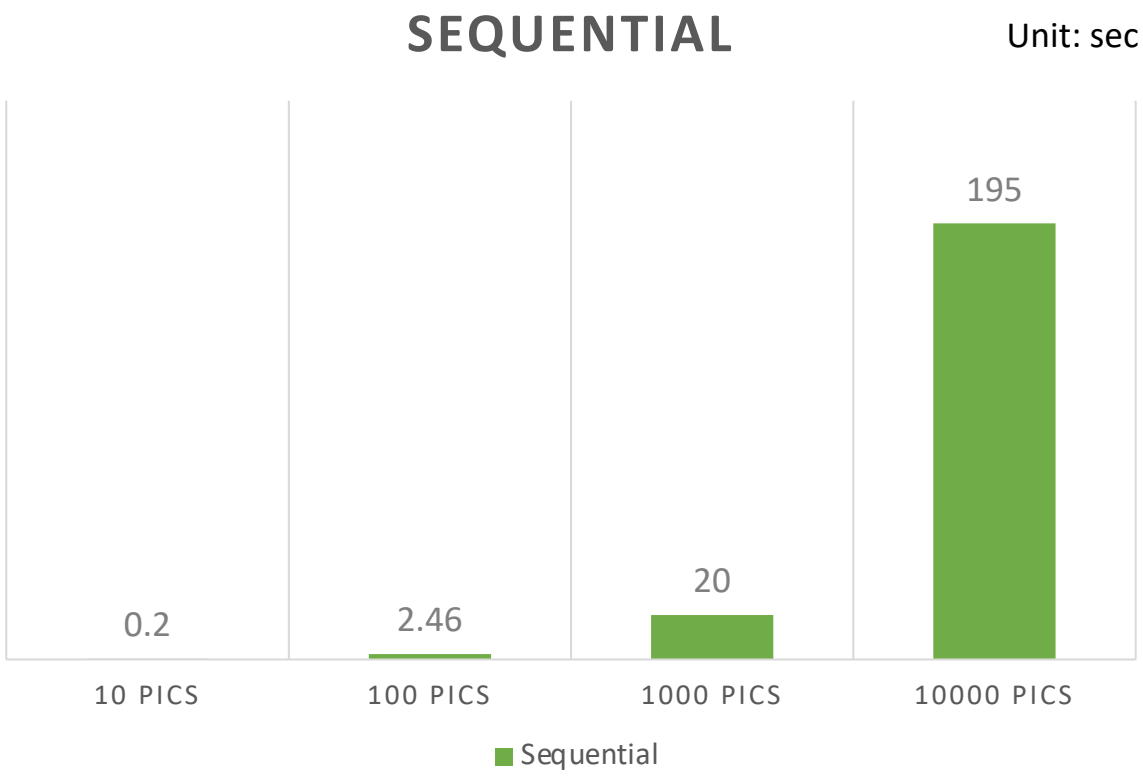
- Cuda

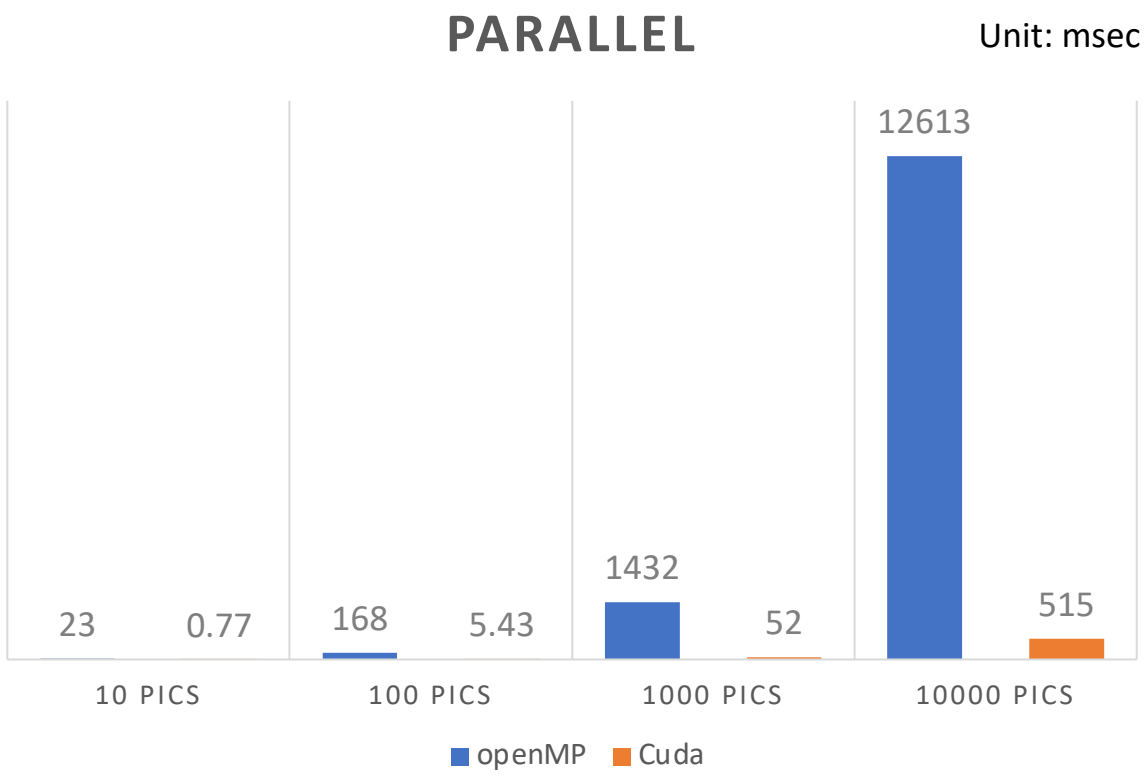
```
__global__ void sobel_edge_filter_cuda(int* dataIn, int* dataOut, int imgHeight,  int imgWidth) {
    int x_index = threadIdx.x + blockIdx.x * blockDim.x;
    int y_index = threadIdx.y + blockIdx.y * blockDim.y;
    int index = y_index * imgWidth + x_index;
    int Gx = 0;
    int Gy = 0;
    if (x_index > 0 && x_index < imgWidth - 2 && y_index > 0 && y_index < imgHeight - 2) {
        Gx = convolution(x_index,y_index,sobel_filter_x);
        Gy = convolution(x_index,y_index,sobel_filter_y);
        dataOut[index] = sqrt(Gx * Gx + Gy * Gy);
    }
}
```

Results

RGB	Gray	G_x	G_y	Sobel
				
				
				

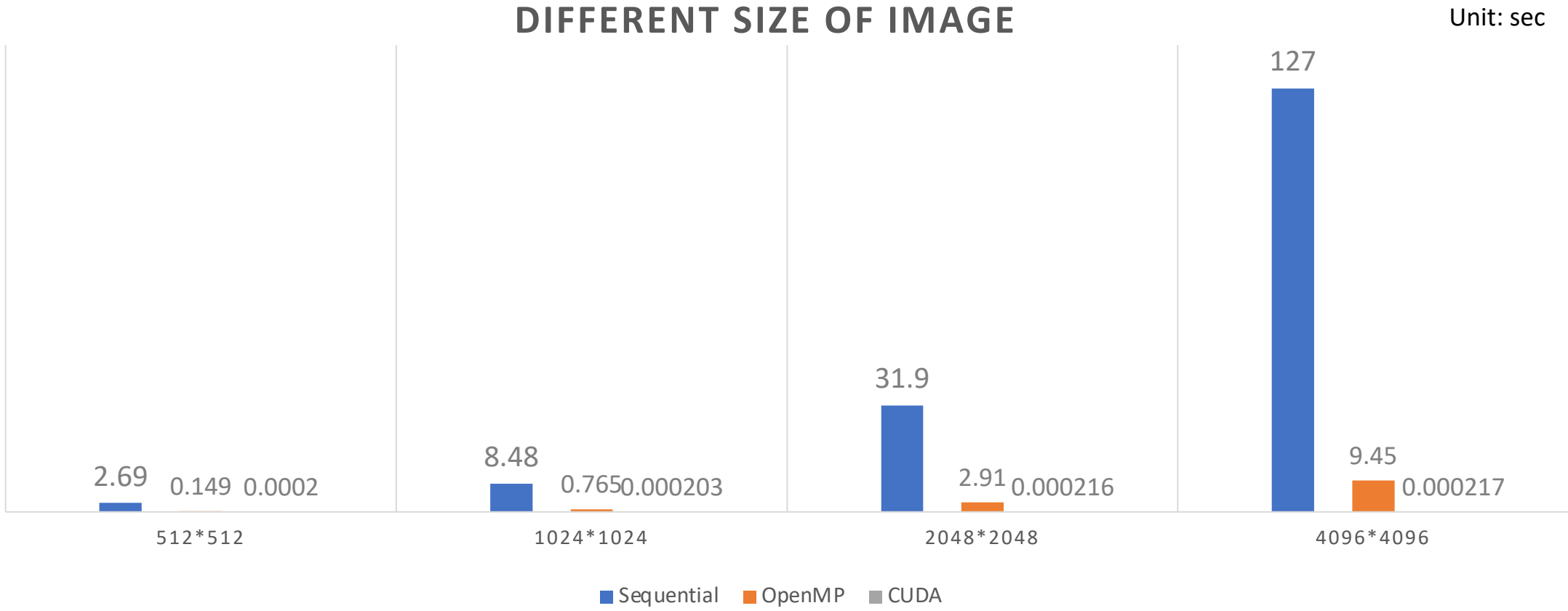
Comparisons





Comparisons

images: 100
threads in openMP: 10
Cuda size: grid(30, 30), block(300, 300)



Comparisons (plus loading image time)