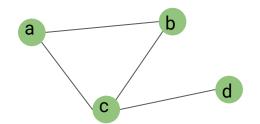
Project 2: Graph Algorithms

Iskandar, Andrel, Yong Qiang, Wayne, Jin Han

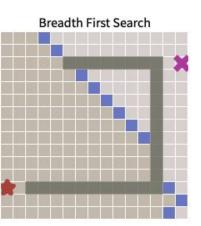
Overview



Graph Algorithms



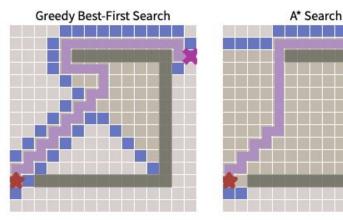
Undirected	Granh



X	а	b	С	d
а	0	1	1	0
b	1	0	1	0
С	1	1	0	1
d	0	0	1	0

Adjacency matrix

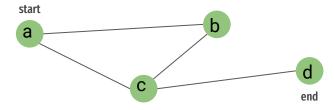
Used for data processing: Conversion from edge list



A* Search

Pseudocode

- f(n) = g(n) cost + h(n) heuristic
- openList : Nodes
- closedList



```
openList = [ nodes ], closedList = []
g(start) = 0, h(start) = heuristic(start,end)
f(start) = g(start) + h(start)
loop openList when Not Null
         M = top node of openList
         If m == end
                  End function
         Add M to closeList
         Remove M at openList
         Loop childnodes as n of M:
                  If n exist in closeList:
                           Continue
                  Cost = g(M) + distance(M,n)
                  If n in closedList AND cost < g(n):
                           Remove n from openList for faster path
                  If n in openList AND cost < g(n):
                           Remove n from closedList
                  If n not in openList AND n not in closedList:
                           Add n to openList
                           g(n) = cost
                           h(n) = heuristic(n,end)
```

f(n) = g(n) + h(n)

Analysis of A* Search

- Combines BFS (efficiency) & Greedy (speed)
- Estimate shortest path to nearest hospital
- Time Complexity
 - Best Case: O(E) --- target node beside start node
 - Worse Case: O(|V|+|E|) --- no path found / node cannot be reached
 - \circ Average Case: O(1/e + (V+E)/V)

Empirical Study

- Dire scenarios and disasters taken into consideration.
- Performance and Speed of algorithms
- Theoretic Analysis and Evaluation:
 - Breadth First Search
 - Depth First Search
 - Greedy Search

Analysis Overview of Algorithms

Algorithm	Random Graph Generated	Road Network Graph in Pennsylvania	Road Network Graph in Texas
A* Search	Fast, Optimal	Fast, Optimal	Fast, Optimal
Breadth First Search	Slow, Optimal	Slow, Optimal	Slow, Optimal
Last First Search	Average speed, Not-optimal	Average speed, Not-optimal	Average speed, Not-optimal
Greedy Search	Fast, Less-Optimal	Fast, Less-Optimal	Fast, Less-Optimal



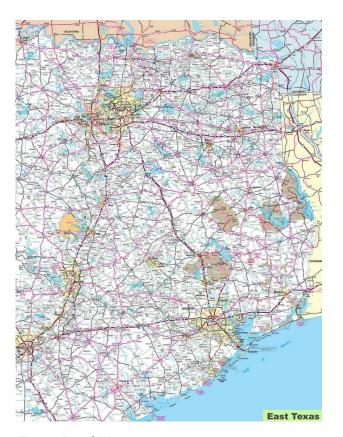
Real road network

- Huge data
- Process in to graph / adjacency matrix



Pennsylvania Road Map

• 1,088,092 Nodes + 1,541,898 Edges



Texas Road Map

• 1,379,917 Nodes + 1,921,660 Edges

Which to use?

CPU	GPU
(instruction)	(data processing)
 Serial processing Random branching instruction stream 	 Parallel processing Data stream Vast amount of data

GPU Processing of road map

- Data stream eliminate nested loops
- Convert codes to Cuda for GPU to read and run
- Dual GPU for faster processing





GPU Data Processing

Pseudocode

```
Roadmap = ReadRoadMap("roadmap.txt")
CudaArray = Cuda.Transfer(Roadmap)
Use GPU 0:
           CudaArray.SortAscending
          totalNodes = CudaArray.CountUniqueNodes
Use GPU 1:
           startNode, endNode = CudaArray.HorizontalSplit
Use GPU 0,1: (Matrix)
           currentRow = 0
           matrixRow = Cuda.ZeroValRows(totalNodes)
           Loop totalNodes:
                      if(current != startNode):
                                 AppendTextFile(matrixRow)
                                 Current = startNode
                                 matrixRow = Cuda.ZeroValRows(totalNodes)
                      matrixRow[endNode] = 1
           Else:
                      AppendTextFile(matrixRow)
Use GPU 0,1: Adjacency List
           adjList = [], currentNode = 0
           Loop i in totalNodes:
                     if(currentNode == U):
                                U = startNode[i]
                                 V = endNode[i]
                      Else:
                                 AppendTextFile(adjList)
                                 adjList = ∏
                      adjList[U].append(V)
```

Data Process

```
CudaProcessing.py
                      adiList.txt ×
Project 2 > Graph > 🖹 adjList.txt
           [1, 6309, 6353]
           [3, 4, 7]
           [2, 309]
           [2, 273, 274, 388]
           [6, 8, 9]
           [5, 305, 309, 310, 1060307]
           [2, 8, 16, 3998]
           [5, 7]
           [5, 12, 10464]
           [11, 22, 77]
           [10, 44]
           [9, 13, 14]
           [12, 15, 20]
          [14, 310, 312]
```

```
λ python CudaProcessing.py
Edge List to Adj List Completed 14.256047500000001
```

```
def ConvertToAdjList(arr1):
    #Device 1 is Asus Strix RTX 3090
    #Cp.Cuda.Device(1).use()
    cpSortedArr = arr1[arr1[:,0].argsort()]
    start,end = cp.hsplit(cpSortedArr, 2)
    node = cp.unique(cpSortedArr).size

#Device 0 is MSI Gaming X Trio RTX 3090
#Cp.Cuda.Device(0).use()
adjList = [[] for k in range(node)]

#Line 52 ~ 58 are executed in parallel
#Line 78 wait for both of my GPU to finish calculating for continuing to the next line
#cp.cuda.Device({0,1}).synchronize()

#Using both GPU to process at the same time
#cp.cuda.Device({0,1}).use()
for i in range(node):
    u = int(start[i].get())
    v = int(smt[i].get())
    adjList[u].append(v)

np.savetxt("adjList.txt", adjList, delimiter=" ", fmt="%s")
```

Data Process

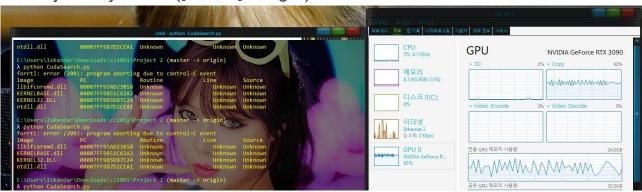
```
CudaProcessing.pv
                 matrixList.txt ×
Project 2 > Graph > a matrixList.txt
       [0 1 0 ... 0 0 0]
    2 [100...000]
    3 [000 ... 000]
    4 [001...000]
    5 [001...000]
    6 [000...000]
    7 [000...000]
    8 [001...000]
    9 [000...000]
   10 [000...000]
   11 [000 ... 000]
   12 [000...000]
       [0 0 0 ... 0 0 0]
```

```
λ python CudaProcessing.py
Edge List to Matrix List Completed 16.4808332
```

```
def CompressedAdjMatrix(arr1):
   size = cp.unique(arr1).size
   cpSortedArr = arr1[arr1[:,0].argsort()]
   arr = cp.zeros(size, dtype="int32")
   cur = 0
   with open("matrixList.txt", 'w') as f:
       for row, col in cpSortedArr:
           if(cur != row):
               f.write(cp.array str(arr) + "\n")
               cur = row
               arr = cp.zeros(size, dtype="int32")
           arr[col.get()] = 1
           f.write(cp.array str(arr) + "\n")
           f.close()
```

Analysis of GPU Processing

- Utilize CUDA Cores
- Parallel Processing
- Converts Edge List to Adjacency Matrix and List
- Time Complexity
 - C = Cuda Cores
 - Adjacency Matrix: O([V²/C] + log C)
 - Adjacency List: O([V+E/C] + log C)



Conclusion

- A* Search (fast & optimal)
- Analysis of GPU and CPU processing
- GPU vs CPU for huge data processing