

Shortest Path Algorithm with Heaps

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1 Introduction

Shortest path problems are ones of the most fundamental combinatorial optimization problems with many applications. The efficient algorithm Dijkstra's algorithm is created by Edsger W. Dijkstra in 1956.

In this project we are going to compute the shortest paths using Dijkstra's algorithm. We'll implement the algorithm with two different heap structures, Binomial heap and Fibonacci heap. The goal of this project is to find the best data structure for the Dijkstra algorithm.

For testing, the USA road networks and the benchmark for the 9th DIMACS Implementation Challenge will be used. And more than 1000 queries will be used for evaluating the run times.

2 Data Structure/Algorithm

2.1 Dijkstra Algorithm

Algorithm 1 basic structure of dijkstra code

Input: $G(V, E)$

Output: $V.dist = shortestpath$

```
build the heap
source.distance  $\leftarrow 0$ 
insert(source)
temp  $\leftarrow extractMin()$ 
while temp do
    known[temp]  $\leftarrow 1$ 
    distance[temp]  $\leftarrow temp.distance$ 
    if temp = target then
        return distance[temp]
    end if
    for every nodes in the adjacent list of temp do
        node.distance = arc-length + distance[temp]
        insert(node)
    end for
    while known[temp] do
        temp = extractMin()
    end while
end while
```

2.2 Fiboacci Heap

A Fibonacci heap is composed of a collection of connected trees that result in a forest-like structure. Each tree within the Fibonacci heap follows a "heap-ordered" structure. The insertion and merge operation of this data structure just cost $O(1)$, and the extractMin operation costs $O(\log N)$

Fibonacci Heap maintains a pointer to the minimum values. All tree roots are connected using a circular doubly linked list, so all trees can be accessed using a single pointer.

Algorithm 2 Fibonacci heap-insert

```
addNode( node , root )
node->left = root->left
root->left->right = node
node->right = root
root->left = node
insert( node )
if keyNum=0 then
    min = node
else
    addNode(node,min)
    if node->key = min->key then
        min = node
    end if
end if
key++
```

The merge of two heaps:join root lists of Fibonacci heaps H1 and H2 and make a single Fibonacci heap H.Then compare the minimum elements of the two heaps. The pseudocodes are as follows:

Algorithm 3 Fibonacci heap-merge

```
catList(a,b)
tmp = a->right
a->right = b->right
b->right->left = a
b->right = tmp
tmp->left = b
combine( node )
if node==null then
    return
end if
if node->maxDegree > this->maxDegree then
    swap( node,this )
end if
if this->min==null then
    this->min = node->min
    this->keyNum = node->keyNum
    free(node->cons)
    delete node
else if node->min==null then
    free(node->cons)
    delete other
else
    catList( this->min,node->min )
    if this->key->node->key then
        this->key = node->key
    end if
    this->keyNum += node->keyNum
    free( node->cons)
    delete node
end if
```

Algorithm 4 Fibonacci heap-extractMin

removing min node means removing the tree it belongs to.

```
extractMin()
p = min
if p==p->right then
    min = null
else
    removeNNode(p)
    min = p->right
end if
p->left = p->right = p
return p
removeMin()
if min==null then
    return
end if
child = null
m = min
while m->child do
    child = m->child
    removeNode(child)
    if child->right = child then
        m->child = null
    else
        m->child = child->right
    end if
    addNode(child,min)
    child->parent = null
end while
removeNode(m)
if m->right= m then
    min = null
else
    min = m->right
    consolidate()
end if
keyNum-
delete m
```

2.3 Binominal Heap

A Binomial Heap is a set of Binomial Trees where each Binomial Tree follows the Min Heap property. And there can be at most one Binomial Tree of any degree.

Algorithm 5 Binomial heap-merge

```
treeMerge(p, c)
  parent[c]  $\leftarrow$  p
  sibling[c]  $\leftarrow$  child[p]
  child[p]  $\leftarrow$  c
  return p
heapMerge(H1, H2)
  node1  $\leftarrow$  head[H1]
  node2  $\leftarrow$  head[H2]
  while node1  $\neq$  NULL and node2  $\neq$  NULL do
    if degree[node1]  $\neq$  degree[node2] then
      H3  $\leftarrow$  node1
      node1  $\leftarrow$  sibling[node1]
    else
      H3  $\leftarrow$  node2
      node2  $\leftarrow$  sibling[node2]
    end if
    if pre = NULL then
      pre  $\leftarrow$  H3
      head[heap]  $\leftarrow$  H3
    else
      sibling[pre]  $\leftarrow$  H3
      pre  $\leftarrow$  H3
    end if
  end while
  if node1  $\neq$  NULL then
    sibling[H3]  $\leftarrow$  node1
  else
    sibling[H3]  $\leftarrow$  node2
  end if
  return heap
```

Algorithm 6 Binomial heap-union

```
union(H)
H ← heapMerge(H1, H2)
x ← head[H]
pre ← NULL
while sibling[x] != NULL do
  if degree[x] != degree[sibling[x]] then
    pre ← x
    x ← sibling[x]
  else if sibling[sibling[x]] != NULL then
    if degree[x] = degree[sibling[sibling[x]]] then
      pre ← x
      x ← sibling[x]
    else
      x ← treeMerge(x, sibling[x])
    end if
  if pre then
    sibling[pre] ← x
  else
    heap ← x
  end if
else
  x ← treeMerge(x, sibling[x])
  if pre != NULL then
    sibling[pre] ← x
  else
    heap ← x
  end if
end if
end while
return heap
```

Insert k to a Binomial Heap. This operation first creates a Binomial Heap with a single key k , then calls union on H and the new Binomial heap.

Extract the minimum element of the heap. This also uses a union. We first call getMin to find the minimum key Binomial Tree, then remove the node and create a new Binomial Heap by connecting all subtrees of the removed minimum node. This requires $O(\log N)$ time.

Algorithm 7 Binomial heap–insert and removeMin

```
insert(H, n)
  value[newHeap]  $\leftarrow$  n
  heap  $\leftarrow$  head[H]
  if heap = NULL then
    heap  $\leftarrow$  newHeap
  else
    heap  $\leftarrow$  union(heap, newHeap)
  end if
  return heap
removeMin(H)
  minnode  $\leftarrow$  heapMin(H)
  node  $\leftarrow$  head[H]
  while sibling[node] and sibling[node] do
    node  $\leftarrow$  sibling[node]
  end while
  sibling[node]  $\leftarrow$  sibling[sibling[node]]
  newnode  $\leftarrow$  child[minnode]
  nextnode  $\leftarrow$  null
  while newnode do
    head[newH]  $\leftarrow$  newnode
    parent[newnode]  $\leftarrow$  null
    tmpnode  $\leftarrow$  sibling[newnode]
    sibling[newnode]  $\leftarrow$  nextnode
    nextnode  $\leftarrow$  newnode
    newnode  $\leftarrow$  tmpnode
  end while
  return union(H, newH)
```

3 About Benchmark

3.1 What is Benchmark

The benchmark test platform is a research paradigm used to measure the performance of algorithms and compare the performance of different algorithms. The DIMACS Challenge website contains the shortest benchmark platform.

Specifically, we only use the short-circuit problem (no-nega-ARC) in which there

is No Negative side Single-Source Shortest Path Problem (NSSP) test benchmark.

3.2 Why use it

1

Reduce the impact of machine hardware with reference algorithms.

Since the program runtime is affected by many underlying architectures, in order to exclude machine influences on the performance of the comparison algorithms, DIMACS recommends using the time "relative to the standard algorithm" and comes with a standard NSSP solver.

2

A more precise timer.

The timer used by benchmark is `sys/time.h` from Linux, which provides a precision of 0.01ms.

3

Generate test data in batches.

The platform's '.ss' test files are not written to death, but are randomly generated by code, generating more than 400 individual test cases per graph.

Finally, the total running time is averaged to obtain the solving performance of the algorithm on this graph. Doing so avoids testpoint bias for the best or worst case.

4 Test Result

Figure1 2 are the statistic of number of nodes and number of arcs when reading in different graphs. Figure3 is the running result on the benchmark. Figure 4 is the comparison result of different data structures.

Meanwhile, we also used the smart queue provided by the benchmark, and the result comes out that the heap structure greatly improved the performance of the algorithm.

From the picture we can see that the heaps have a superiority with comparison with the smart queue. The Fibonacci heap is quicker than the binomial heap, as the insertion and merge operation is quicker using the Fibonacci heap.

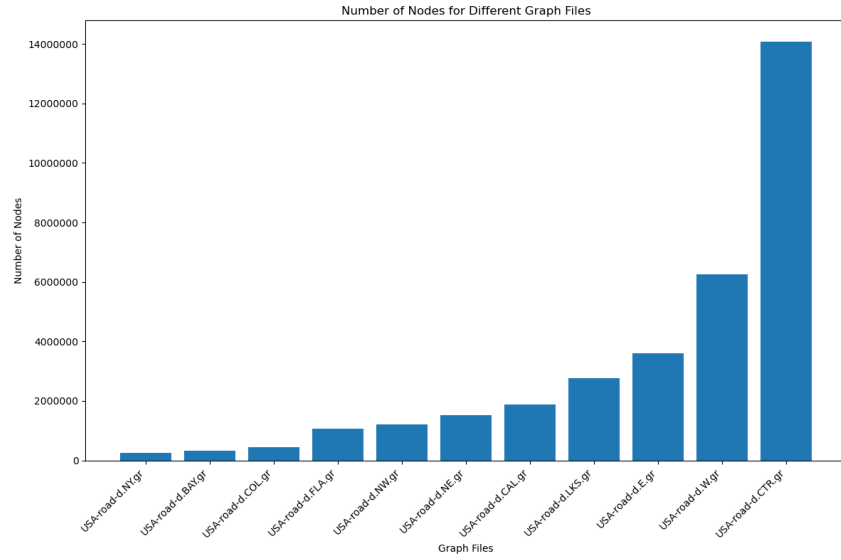


Figure 1: statistics of nodes number

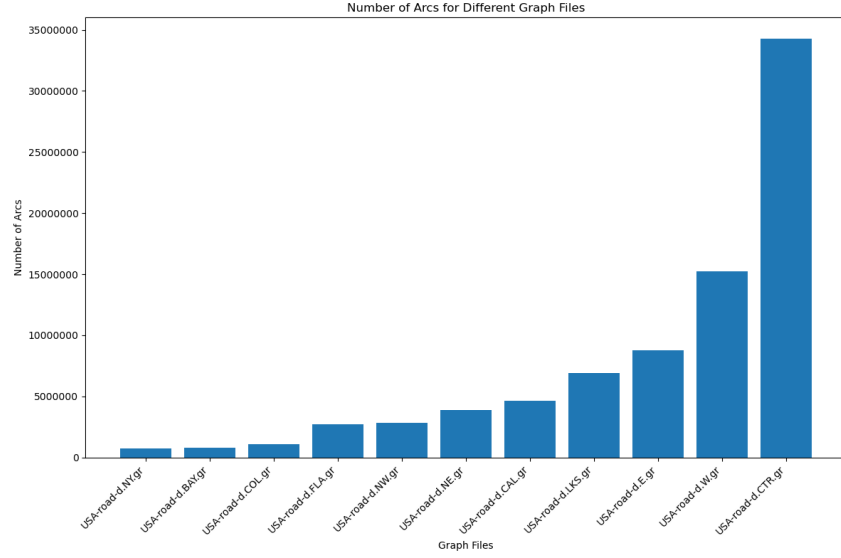


Figure 2: statistics of arcs number

```

* 9th DIMACS Implementation Challenge: Shortest Paths
* http://www.dis.uniroma1.it/~challenge9
* USA-road-t family ss core experiment

* Running ss solver on graph ../inputs/USA-road-t/USA-road-t.NY.gr
c -----
c SQ/SQP DIMACS Challenge version
c -----
c
c Nodes:                264346        Arcs:                733846
c MinArcLen:             2          MaxArcLen:            92366
c Trials:                 532
c Scans (ave):           264346.0    Improvements (ave):    0.0
c Time (ave, ms):        36.56

```

Figure 3: a shot of running result

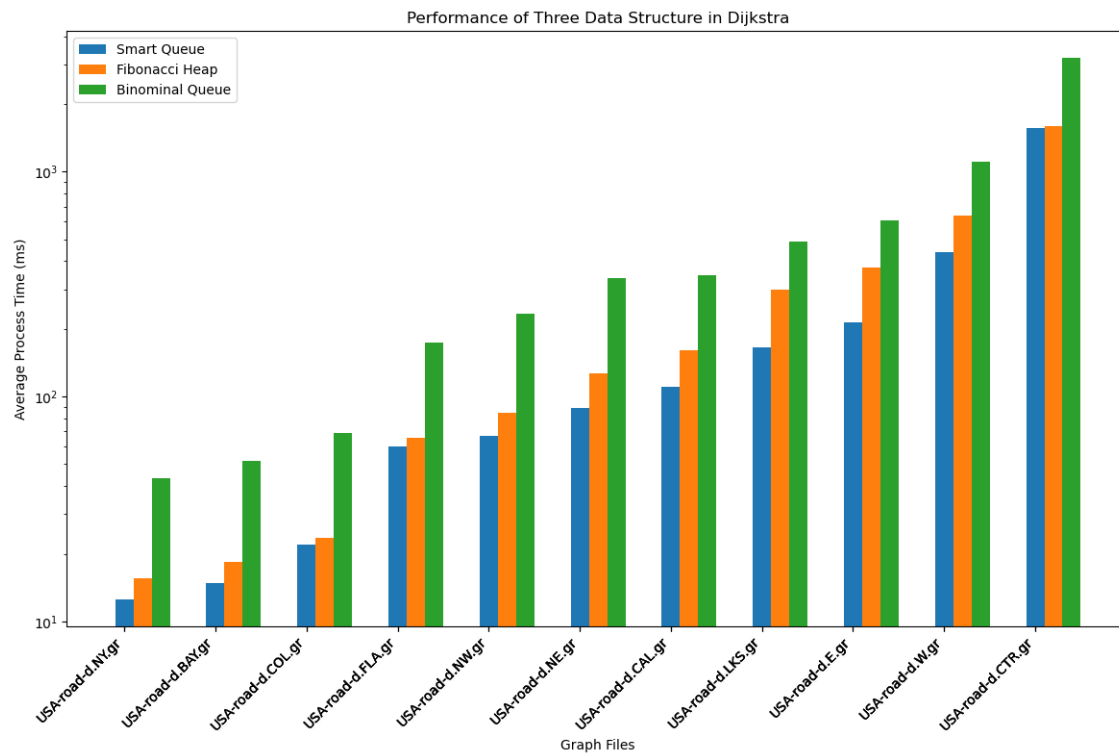


Figure 4: compare the average run times of three data structure

5 Analysis

5.1 Time Complexity

Adjacent List

The construct of the adjacent list will cost $O(V + E)$.

Insertion

When using the Fibonacci heap, per insertion will cost $O(1)$, so the total cost of insertion will be $O(E)$. And when using a binomial heap each insertion costs $O(\log N)$, so the operation in total costs $O(E \log E)$, which make for the reason why the binomial heap costs more time than the Fibonacci heap.

ExtractMin

Both extracting minimum elements operation will cost $O(V \log E)$ totally.

The time complexity will be $O(V \log E)$ for most sparse matrix. And for dense matrix, Fibonacci heap costs $O(V \log E)$ and binomial heap costs $O(E \log E)$, slower than the Fibonacci heap, both more quickly than the $O(V^2)$ complexity when using a table.

5.2 Space Complexity

With the adjacent list costing $O(V + E)$ and the heap costing $O(E)$, the space complexity is $O(V + E)$.

Declaration

We hereby declare that all the work done in this project titled "Roll Your Own Mini Search Engine" is of our independent effort as a group.