

# 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

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**Algorithm 1** basic structure of dijkstra code

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**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  $O(\log N)$

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**Algorithm 2** structure of Fibonacci heap

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$j_1$

---

## 2.3 Binominal Heap

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**Algorithm 3** 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
```

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**Algorithm 4** Binomial heap-union

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

---

---

**Algorithm 5** Binomial heap–insert 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

### pro1

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.

### pro2

A more precise timer.

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

### pro3

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

## 5 Analysis

### 5.1 Time Complexity

The construct of the adjacent list will cost  $O(V + E)$ . With per insertion costing  $O(1)$ , the total cost of insertion will be  $O(E)$ . Extracting minimum elements will cost  $O(V \log E)$ .

The time complexity will be  $O(V \log E)$  for most sparse matrix. Which is 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.*