

# 数据结构与算法

Data Structure and Algorithm

## XVIII. 图论 II

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# Review

## 回顾



★ 数据结构与算法

★ 数学回顾

★ 数组

★ 数组列表

★ 搜索和排列

★ 递归与迭代

★ 二进制搜索

★ 分而治之

★ 链接列表

★ 散列表

★ 树

★ 堆

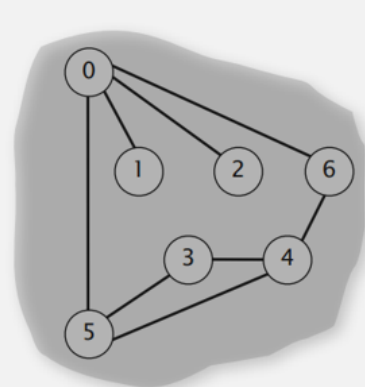
★ 图论-BFS,DFS

# 概述

- 连通分支
- 单路径最短路径
  - Dijkstra 算法
  - Bellman-Ford 算法
  - A\* 算法
- 生成树

# 连通图

- 在无向图中，若从顶点 $v$ 到顶点 $w$ 之间存在路径，则顶点 $v$ 和 $w$ 是连通的
- 在有向图中，连接顶点 $v$ 和顶点 $w$ 的路径中所有的边都必须同向。
- 如果图中任意两点都是连通的，那么图被称作连通图。



3 connected components

v	id[]
0	0
1	0
2	0
3	0
4	0
5	0
6	0
7	1
8	1
9	2
10	2
11	2
12	2

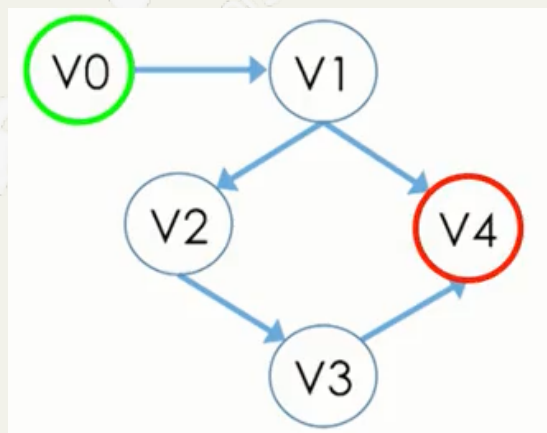
- Goal:** Partition vertices into connected components

## Connected components

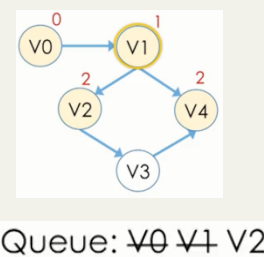
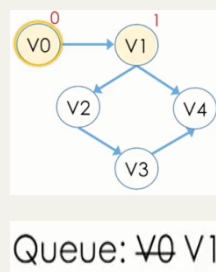
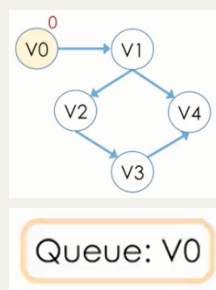
Initialize all vertices  $v$  as unmarked.

For each unmarked vertex  $v$ , run DFS to identify all vertices discovered as part of the same component.

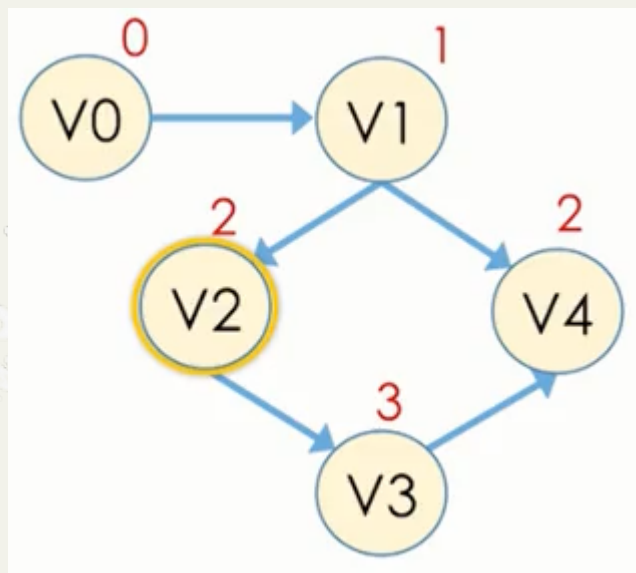
# 最短路径



- 从 V0 到 V4
- 让我们试试 BFS

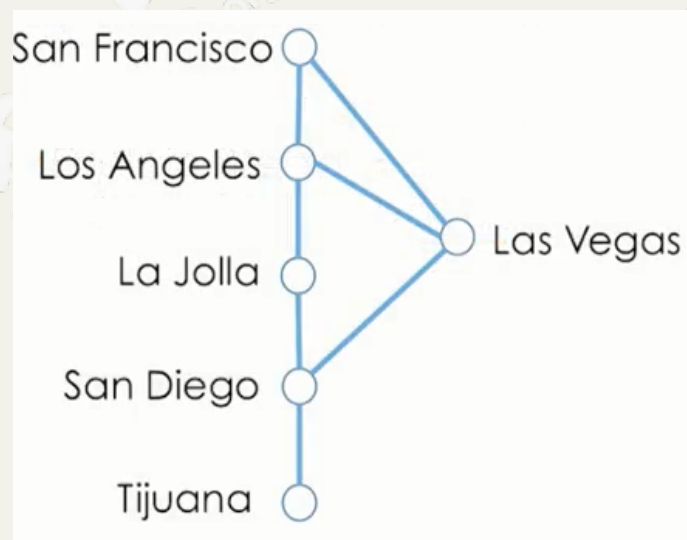


- 发现 V4
- 我们完成了吗?



- ⊙ BFS是否适用于所有图表?

# 作为地图的地理地图

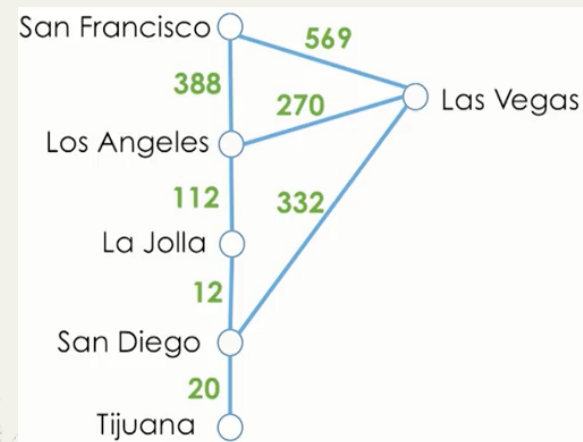


图中缺少了什么？

- 距离

还有什么？

- 速度限制
- 交通
- 登机交叉延误

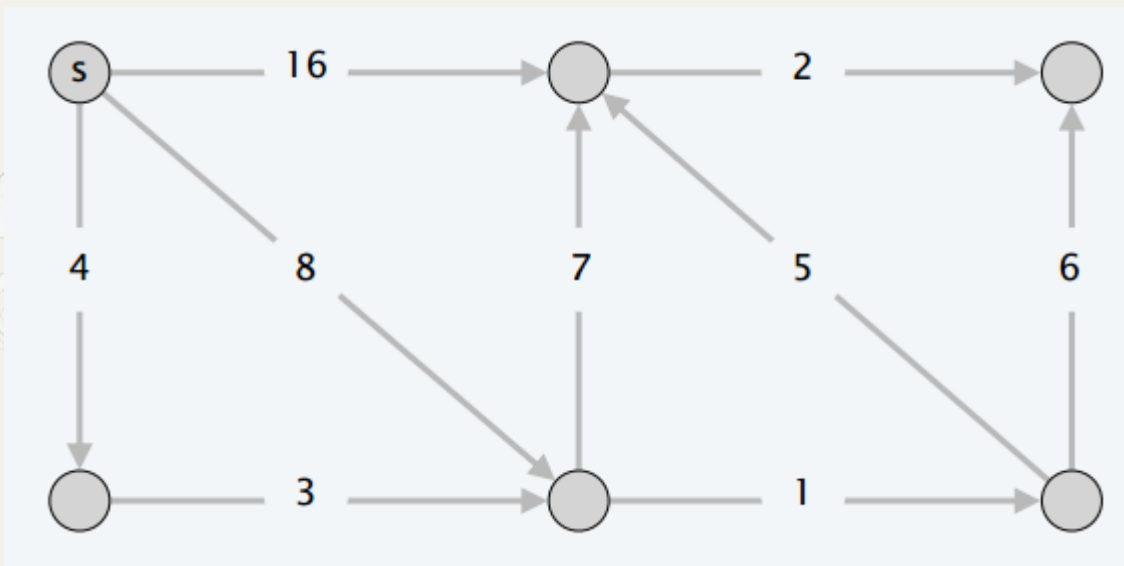


● BFS能够找到从圣地亚哥到旧金山的最短路径？

● BFS不考虑边缘权重，只考虑边缘数量

# 单源最短路径问题

- 给定一个图  $G = (V, E)$  和一个  $V$  中的“源”顶点  $u$ ， $V$  中的一个“目标”顶点  $v$ ，找到从  $u$  到  $v$  的最小代价路径。
- 给定图  $G = (V, E)$  和  $V$  中的“源”顶点  $s$ ，求出从  $s$  到  $V$  中每个顶点的最小代价路径。





# Dijkstra's 算法

## ◎ 回顾 BFS

- 如何跟踪下一步搜索的位置?
- 使用队列

## ◎ Dijkstra's 算法

- 贪婪
- 使用优先级队列
- 列表中添加元素{元素, 优先级}, 并从另一端删除最高优先级项
- 入队 → 添加一个{元素, 优先级}
- 队列 → 删除最高优先级的元素
- 优先级队列通常使用“堆”来实现, 并可以优先考虑低值 (Min-Heap) 或大值 (Max-Heap)



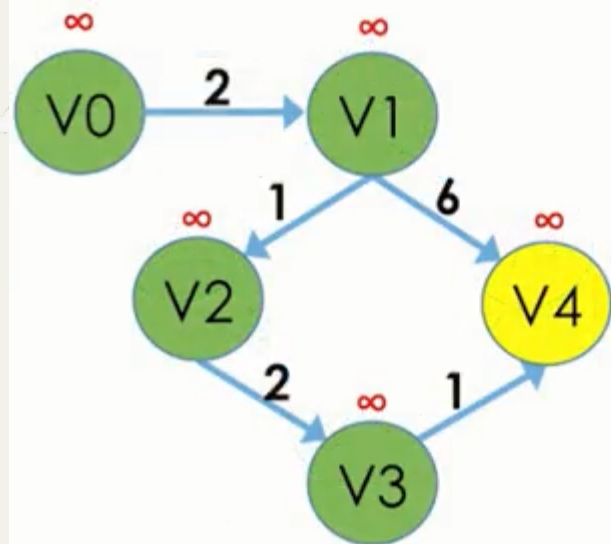
# 粗略的想法

- Maintain an estimate  $d[v]$  of the length  $\delta(s, v)$  of the shortest path for each vertex  $v$
- Always  $d[v] \geq \delta(s, v)$  and  $d[v]$  equals the length of a known path
  - ( $d[v] = \infty$  if we have no paths so far)
- Initially  $d[s] = 0$  and all the other  $d[v]$  values are set to  $\infty$ . The algorithm will then process the vertices one by one in some order.
  - The processed vertex's estimate will be validated as being real shortest distance, i.e.  $d[v] = \delta(s, v)$
- Here “processing a vertex  $u$ ” means finding new paths and updating  $d[v]$  for all  $v \in \text{Adj}[u]$  if necessary. The process by which an estimate is updated is called relaxation.
- When all vertices have been processed,
  - $d[v] = \delta(s, v)$  for all  $v$
- **Question 1:** How does the algorithm find new paths and do the relaxation?
- **Question 2:** In which order does the algorithm process the vertices one by one?

# 粗略的答案

- **Question 1:** How does the algorithm find new paths and do the relaxation?
- **Answer: Finding new paths.** When processing a vertex  $u$ , the algorithm will examine all vertices  $v \in Adj[u]$ . For each vertex  $v \in Adj[u]$ , a new path from  $s$  to  $v$  is found (path from  $s$  to  $u$  + new edge).
  - We use **Greedy** algorithm. For each vertex  $v \in Adj[u]$ . The next vertex processed is always a vertex  $v \in Adj[u]$  for which  $d[u]$  is minimum
  - that is, we take the unprocessed vertex that is closest (by our estimate) to
- **Question 2:** In which order does the algorithm process the vertices one by one?
- **Answer: Relaxation.** If the length of the new path  $s$  to  $v$  is shorter than  $d[v]$ , then update  $d[v]$  to the length of this new path.

# Dijkstra's 算法



PQ:  
curr:  
visited:

## Dijkstra: Algorithm

Dijkstra(S, G):

Initialize: **Priority queue (PQ)**, visited HashSet,  
parent HashMap, and **distances to infinity**

Enqueue {S, 0} onto the **PQ**

while **PQ** is not empty:

    dequeue node curr from front of queue

**if(curr is not visited)**

**add curr to visited set**

**If curr == G return parent map**

        for each of curr's neighbors, n, not in visited set:

**add n to visited set**

**if path through curr to n is shorter**

**update n's distance**

**update curr as n's parent in parent map**

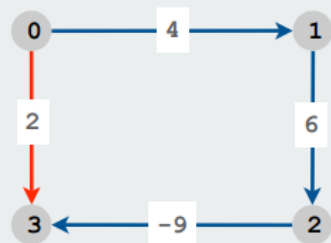
**enqueue {n, distance} into the PQ**

    // If we get here then there's no path

# Dijkstra's 挑战

- 负数会怎么样？

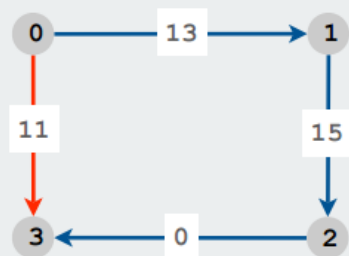
Dijkstra. Doesn't work with negative edge weights.



Dijkstra selects vertex 3 immediately after 0.  
But shortest path from 0 to 3 is 0→1→2→3.

- 我们试试

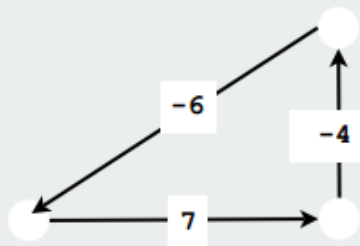
Re-weighting. Adding a constant to every edge weight also doesn't work.



Adding 9 to each edge changes the shortest path  
because it adds 9 to each segment, wrong thing to do  
for paths with many segments.

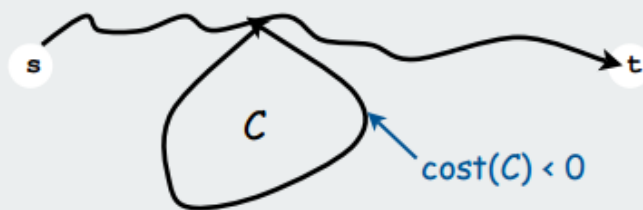
# 负循环

**Negative cycle.** Directed cycle whose sum of edge weights is negative.



**Observations.**

- If negative cycle  $C$  on path from  $s$  to  $t$ , then shortest path can be made **arbitrarily negative** by spinning around cycle
- There exists a shortest  $s$ - $t$  path that is simple.



Worse news: need a different **problem**

# Bellman-Ford 算法

Initialize  $\text{distTo}[s] = 0$  and  $\text{distTo}[v] = \infty$  for all other vertices.

Repeat  $V$  times:

- Relax each edge.

```
for (int i = 0; i < G.V(); i++)  
    for (int v = 0; v < G.V(); v++)  
        for (DirectedEdge e : G.adj(v))  
            relax(e);
```

← pass i (relax each edge)

- 时间复杂度?
- 暴力解法
- $O(EV)$

# Bellman-Ford 改进

**Observation.** If  $\text{distTo}[v]$  does not change during pass  $i$ ,  
no need to relax any edge pointing from  $v$  in pass  $i+1$ .

**FIFO implementation.** Maintain **queue** of vertices whose  $\text{distTo}[]$  changed.



be careful to keep at most one copy  
of each vertex on queue (why?)

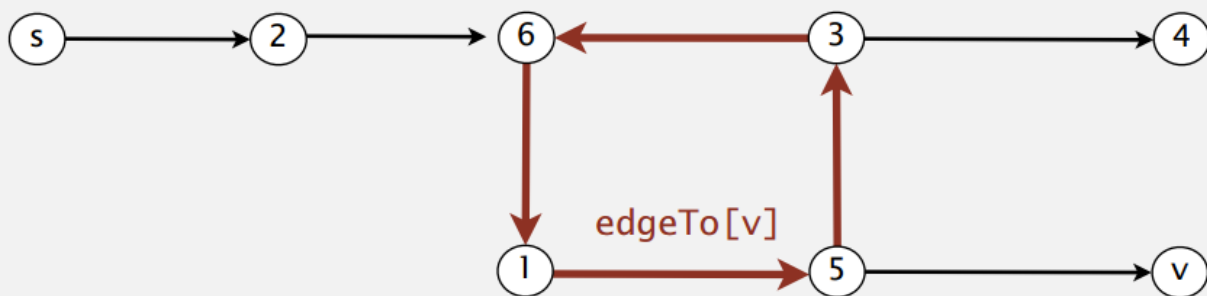
**Overall effect.**

- The running time is still proportional to  $E \times V$  in worst case.
- But much faster than that in practice.



# 寻找负循环

**Observation.** If there is a negative cycle, Bellman-Ford gets stuck in loop, updating `distTo[]` and `edgeTo[]` entries of vertices in the cycle.



**Proposition.** If any vertex `v` is updated in pass `v`, there exists a negative cycle (and can trace back `edgeTo[v]` entries to find it).

**In practice.** Check for negative cycles more frequently.

# 负循环应用

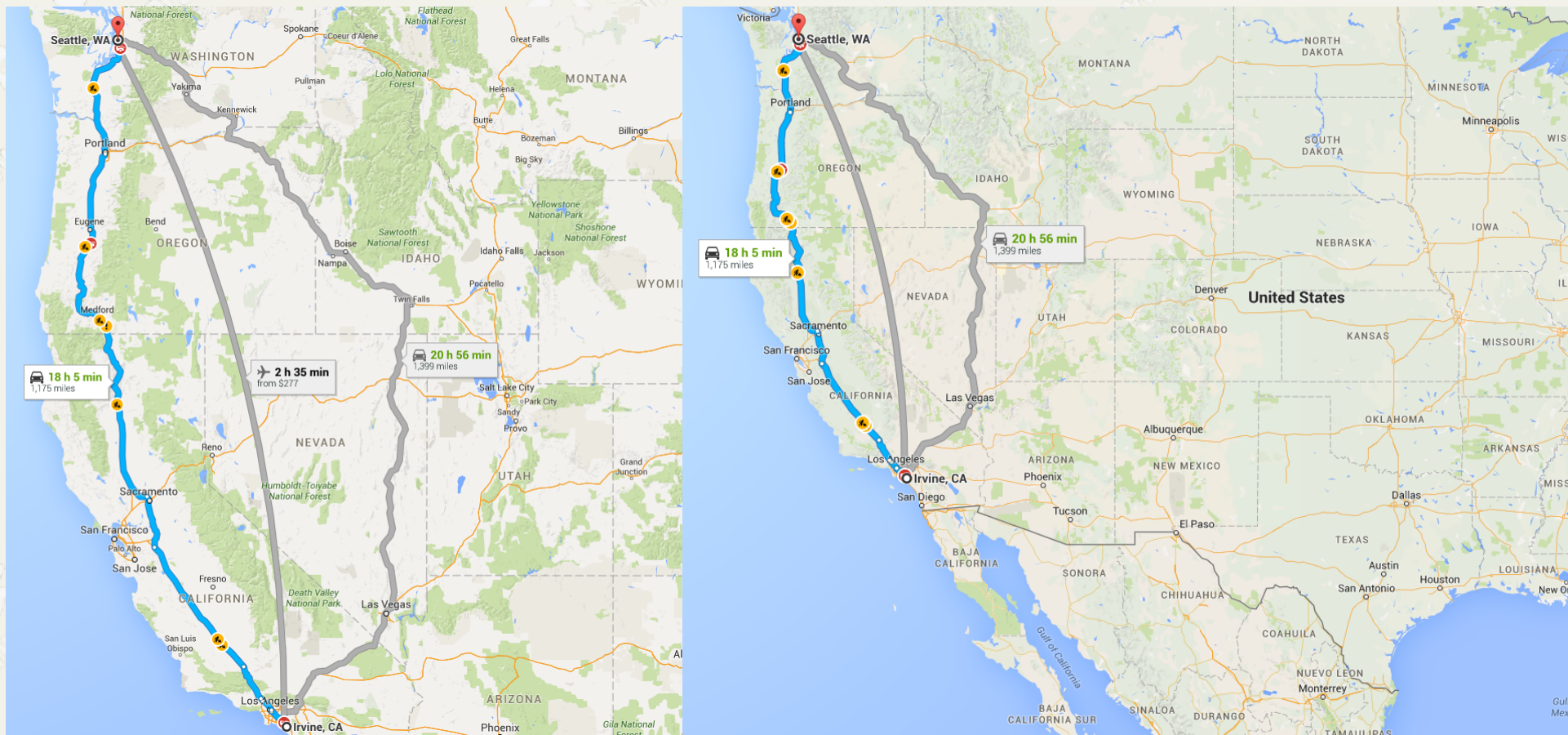
**Problem.** Given table of exchange rates, is there an arbitrage opportunity?

	USD	EUR	GBP	CHF	CAD
USD	1	0.741	0.657	1.061	1.011
EUR	1.350	1	0.888	1.433	1.366
GBP	1.521	1.126	1	1.614	1.538
CHF	0.943	0.698	0.620	1	0.953
CAD	0.995	0.732	0.650	1.049	1

**Ex.** \$1,000  $\Rightarrow$  741 Euros  $\Rightarrow$  1,012.206 Canadian dollars  $\Rightarrow$  \$1,007.14497.

$$1000 \times 0.741 \times 1.366 \times 0.995 = 1007.14497$$

# Dijkstra 局限性



- Dijkstra is BFS
- 并可能到达凤凰城，达拉斯，丹佛和盐湖城
- Dijkstra只考虑距离信号源的距离

# A\* 算法

- Dijkstra's 算法
  - 是基于优先级队列排序
  - $G(n)$ : 从起始顶点到顶点  $n$  的距离
- 我们也应该考虑距离目标
- A\* 算法
  - $G(n)$ : 从起始顶点到顶点  $n$  的距离
  - $H(n)$ : 从顶点  $n$  到目标顶点的启发式估计成本
  - $F(n) = g(n) + h(n)$
  - Dijkstra 可以看作  $h(n) = 0$  的特例
  - 保证找到最短路径如果估计永远不是高估
  - 在前面的例子中,
  - 低估: 使用直线距离
  - 这很容易计算, 我们有经度和纬度
  - 只是改变优先功能

# 数据结构与算法

Data Structure and Algorithm

## XVIII. 图论II 结束

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