Estratégias Algorítmicas 2020/21 Week 11 – Graph Algorithms (SCC and MaxFlow)



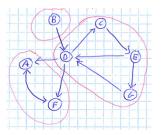
Universidade de Coimbra

Outline

- 1. Strongly Connected Components
- 2. Max-flow Algorithms
 - 2.1 Ford-Fulkerson Algorithm
 - 2.2 Edmond-Karp Algorithm
 - 2.3 Applications

Introduction

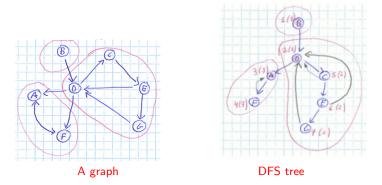
Given a directed graph G = (V, A), a subgraph G' is a strongly connected component if there exists a path between a vertex and every other vertex in G' and this subgraph has maximal size.



3 strongly connected components

Strongly Connected Components

A vertex v is the root of a connected component if low[v] = dfs[v]



The vertices in each strongly connected component under the root are stored in a stack. It is possible to solve it with Tarjan Algorithm in O(|V| + |A|).

Strongly Connected Components

```
Function Tarjan(v)
  low[v] = dfs[v] = t
  t = t + 1
  push(S, v)
  for each arc (v, w) \in A do
     if dfs[w] has no value then
        Tarjan(w)
        low[v] = min(low[v], low[w])
     else if w \in S then
        low[v] = min(low[v], dfs[w])
  if low[v] = dfs[v] then
     C = \emptyset
     repeat
        w = pop(S)
        push(C, w)
     until w = v
     push(Scc, C)
```

Scc collects all stacks of strongly connected components. This algorithm must be called for every unvisited vertex in the graph.

Introduction

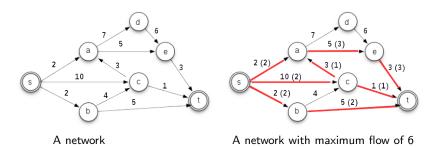
Given a directed graph G = (V, A), where each arc (u, v) has a capacity c(u, v) > 0 (a *network*), and two vertices *source s* and *sink t*, the maximum flow problem consists of maximizing the total amount of flow from s to t subject to two constraints:

- 1. Flow on arc (u, v) does not exceed c(u, v)
- 2. For every vertex $v \neq s,t$, incoming flow is equal to outgoing flow.

This problem arises in many real-life situations:

- Routing as many packets as possible on a network, given a bandwidth capacity
- Sending as many trucks as possible, where roads have limits on the number of trucks per unit time

Introduction

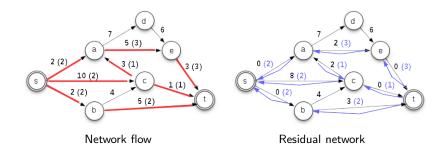


How to compute the maximum flow?

A residual network $G_R = (V, A_R)$ has the same vertices as the original network G and two different types of arcs:

- 1. For each arc (u, v) with capacity c(u, v) and flow f(u, v) < c(u, v), there exists an arc (u, v) in the residual network with capacity c(u, v) f(u, v)
- 2. For each arc (u, v) with capacity c(u, v) and flow f(u, v) > 0, there exists an arc (v, u) in the residual network with capacity f(u, v)

An augmenting path is an path from source to sink in the residual network G^* that should send as much flow as possible.



```
Function FF(G_R)

mflow = 0

while there is a s-t path P in G_R do

f_p = \min\{c(u,v) \mid (u,v) \in P\}

for each arc (u,v) in P do

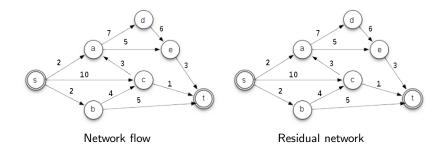
c(u,v) = c(u,v) - f_p

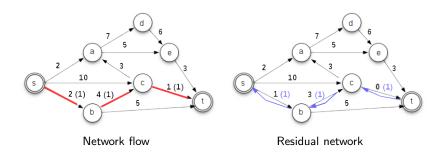
c(v,u) = c(v,u) + f_p

mflow = mflow + f_p

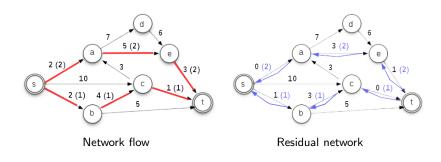
return mflow
```

For each arc (u, v) in G, G_R contains an arc (u, v) with capacity c(u, v) and an arc (v, u) with capacity 0.

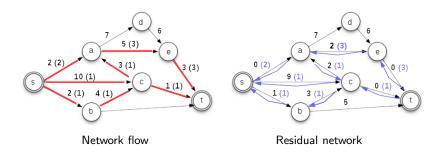




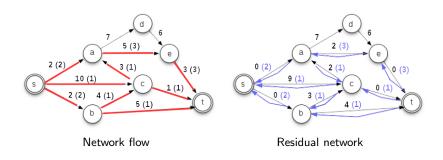
Send 1 unit of flow through (s,b,c,t) in the residual network



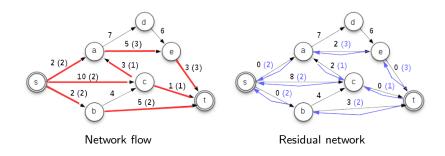
Send 2 units of flow through (s, a, e, t) in the residual network



Send 1 unit of flow through (s, c, a, e, t) in the residual network



Send 1 unit of flow through (s, b, t) in the residual network

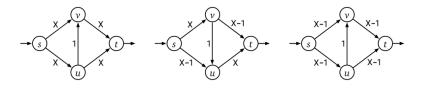


Send 1 unit of flow through (s, c, b, t) in the residual network, which cancels the flow from c to b in the network

Edmond-Karp Algorithm

Ford-Fulkerson works under any choice of a *s-t* path. However, an arbitrary choice may give a bad worst case time complexity.

Consider that the s-t path passes always through vertices u and v:

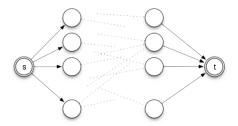


The number of iterations until it stops is twice the value of the maximum flow.

But if the *s-t* path with the least number of arcs is chosen (with BFS), the time complexity reduces to $O(|V| \cdot |E|^2)$ for any network (Edmond-Karp Algorithm).

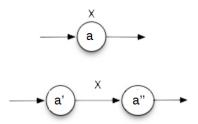
Variants

Many sources and/or many sinks: Add a source vertex that conects to all sources and/or add a sink vertex to which all sinks are connected. The new arcs must have an "infinite" capacity.



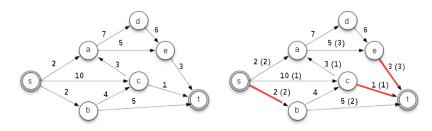
Variants

A vertex a with capacity constraint x: Split the vertex into two, a' and a'', and connect them by an arc with capacity x.



Minimum Cut Problem

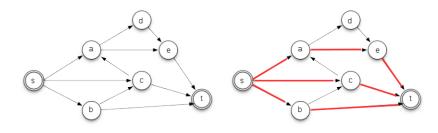
Given a directed graph G = (V, A), where each arc (u, v) has a weight w(u, v), and two vertices s and t, the minimum cut problem consists of finding the set of arcs to be removed (a cut) with the least total weight such that s and t become unreachable.



Solve a maximum flow problem where the arc weight is the arc capacity. The least total weight is equal to the maximum flow (max-flow min-cut theorem).

Arc-Disjoint Paths Problem

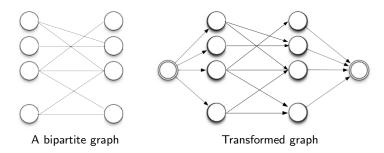
Given a directed graph G = (V, A), and two vertices s and t, the arc-disjoint paths problem consists of finding the maximum number of s-t paths that do not have arcs in common.



Add capacity one to each arc and solve the corresponding maximum flow problem. The number of arc-disjoint paths is equal to the maximum flow.

Maximum Bipartite Matching

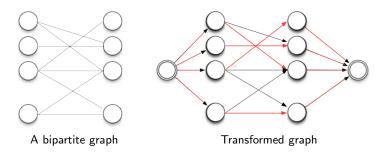
Given a directed graph G = (U, V, E), where U and V are two disjoint sets of vertices, and E is a set of edges, each of which connects an element of U to an element of V, the maximum bipartite matching problem consists of finding the maximum number of edges such that no two edges share a vertex (a matching).



Transformed graph: add a source that connects to all vertices in U and a sink to which all vertices in V are connected. Arc capacity is equal to 1.

Maximum Bipartite Matching

Given a directed graph G = (U, V, E), where U and V are two disjoint sets of vertices, and E is a set of edges, each of which connects an element of U to an element of V, the maximum bipartite matching problem consists of finding the maximum number of edges such that no two edges share a vertex (a matching).



The number of edges in a maximal bipartite matching is equal to the maximum flow in the transformed graph