Minimum Cut, Bipartite Matching and Minimum Cost Maximum Flow with BGL

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November 16, 2016

Minimum Cut: Shopping Trip



Start from HB:

- Visit as many shops as possible.
- Return to HB after each shop.

Condition: Use each road on at most one trip.

Compute the bottleneck, i.e. the number of edge-disjoint paths. ⇒ Four shops.

Unrealistic condition!

(There are interesting streets in Zürich.)

Minimum Cut: Shopping Trip



Start from HB:

- Visit as many shops as possible.
- Return to HB after each shop.

Condition: Use beautiful roads more often.







Use Bahnhofstrasse up to three times.

Compute the weighted bottleneck, i.e. the minimum cut between s and t. \Rightarrow 6 shops.

Minimum Cut: Cuts and Flows

$$G = (V, E, s, t)$$
 a flow network. $S \subset V$ s.t. $s \in S, t \in V \setminus S$, e.g. $S = \{s, v_1, v_3\}$.

The value of the $(S, V \setminus S)$ -cut is

$$cap(S, V \setminus S) := outgoing capacity$$

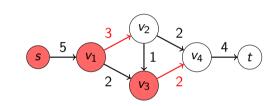
$$= \sum_{\substack{e=(u,v)\\u\in S,v\in V\setminus S}} \operatorname{cap}(e)$$
$$= 3 + 2 = 5.$$

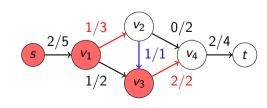
The value of the flow f from S to $V \setminus S$ is

$$f(S, V \setminus S) := \text{ outgoing flow } - \text{ incoming flow}$$

$$= \sum_{\substack{e = (u,v) \\ u \in S, v \in V \setminus S}} \text{ flow}(e) - \sum_{\substack{e = (v,u) \\ u \in S, v \in V \setminus S}} \text{ flow}(e)$$

$$= 1 + 2 - 1 = 2.$$





Minimum Cut: Maxflow-Mincut-Theorem

Theorem (Maxflow-Mincut-Theorem)

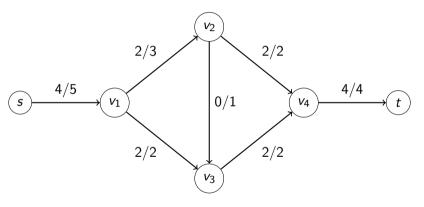
Let f be a s-t-flow in a graph G. Then f is a maximum flow if and only if

$$|f| = \min_{S: s \in S, t \notin S} cap(S, V \setminus S).$$

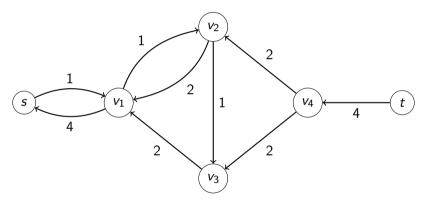
This allows us to easily find a minimum *s-t-*cut:

- Construct the residual graph $G_f := (V, E_f)$. For each edge $(u, v) \in G$ we have:
 - An edge $(u,v)\in G_f$ with capacity $\operatorname{cap}(e)-f(e),$ if $\operatorname{cap}(e)-f(e)>0.$
 - An edge $(v,u)\in G_f$ with capacity f(e), if f(e)>0.
- Since f is a maximum flow, there is no s-t path in the residual graph G_f .
- Take S to be all vertices in G_f reachable from s. $\Rightarrow (S, V \setminus S)$ is a minimum s-t-cut.

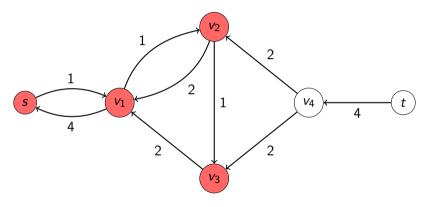
Graph G and a maximum flow f.



Residual graph G_f .

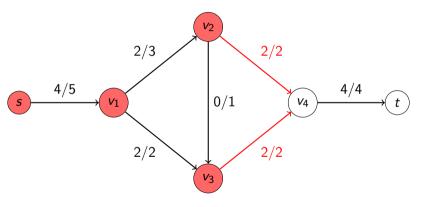


Residual graph G_f .



$$|S| = \{s, v_1, v_2, v_3\}.$$

Residual graph G_f .



$$|S| = \{s, v_1, v_2, v_3\}. \Rightarrow cap(S, V \setminus S) = |f| = 4.$$

Minimum Cut: Code

Example code: BFS on the residual graph G_f . \rightarrow tut9_bgl_residual_bfs.cpp

```
93 // BFS to find vertex set S
94 vector <int> vis(N, false); // visited flags
95 std::queue<int> Q; // BFS queue (from std:: not boost::)
96 vis[src] = true; // Mark the source as visited
97 Q. push (src);
98 while (!Q.empty()) {
       const int u = Q.front();
99
      Q.pop();
100
      OutEdgelt ebeg. eend:
101
      for (tie(ebeg, eend) = out_edges(u, G); ebeg != eend; ++ebeg) {
102
           const int v = target(*ebeg, G);
103
           // Only follow edges with spare capacity
104
           if (rescapacitymap[*ebeg] == 0 || vis[v]) continue;
105
           vis[v] = true;
106
         Q. push (v);
107
108
109 }
```

Minimum Cut: Proof of the Maxflow-Mincut-Theorem

Theorem (Maxflow-Mincut-Theorem)

Let f be a s-t-flow in a graph G. Then the following are equivalent:

- 2 f is a maxflow.
- \blacksquare There is no s-t path in the residual graph G_f .

Proof.

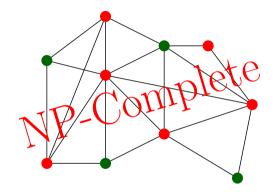
- 1) \implies 2) f cannot be bigger than $\min_{S} \operatorname{cap}(S, V \setminus S)$.
- 2) \Longrightarrow 3) If there was s-t path in G_f , f could be extended.
- 3) \Longrightarrow 1) Take S to be all vertices in G_f reachable from S. Then all edges from S to $V \setminus S$ must be fully saturated by the flow! But then $|f| = f(S, V \setminus S) = \text{cap}(S, V \setminus S)$.

Minimum cut: Algorithm

- **I** Compute maximum flow f and the residual graph G_f .
- **2** Compute the set of vertices *S*:
 - S is reachable from the source s in G_f .
 - BFS on edges with residual capacity > 0.
- 3 Output (depending on the task):
 - All vertices in S.
 - All edges going from S to $V \setminus S$.

Bipartite Matchings: Vertex Cover and Independent Set I

- Maximum independent set Largest $T \subseteq V$, such that $\nexists u, v \in T : (u, v) \in E$.
- Minimum vertex cover Smallest $S \subseteq V$, such that $\forall (u, v) \in E : u \in S \lor v \in S$.
- Complementary problems!



Bipartite Matchings: Vertex Cover and Independent Set II

Theorem (König: MinVC is simpler on bipartite graphs!)

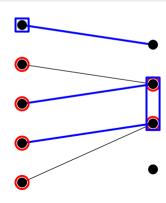
In a bipartite graph, the number of edges in a maximum matching is equal to the number of vertices in a minimum vertex cover.

Algorithm:

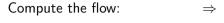
- Maximum matching M, $V = L \cup R$. Find all unmatched vertices in L, label them as visited.
- 2 Starting at visited vertices search (BFS) left to right along edges from $E \setminus M$ and right to left along edges from M. Label each found vertex as visited.
- Minimum vertex cover all unvisited in L and all visited in R.

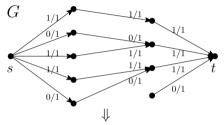
Careful! Step 2 can take several rounds.

Easy Implementation?

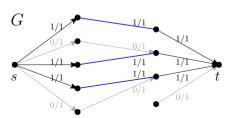


Bipartite Matchings: Modeling via Flow!

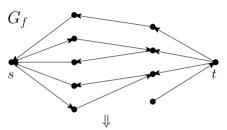




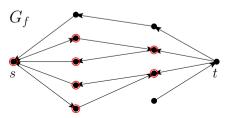
Find the matching:



Compute the residual graph G_f :



Find visited vertices with BFS from s:



MaxFlowMinCut and Bipartite Matching with BGL: Summary

What you should remember:

Edge Cut

- Theorem: maximum amount of s-t-flow = minimum capacity of any s-t-cut
- Finding the cut: BFS on residual graph starting from s.

Vertex Cover

- Minimum vertex cover and maximum independent set are hard problems.
- Bipartite graphs allow fast minimum vertex cover based on maximum matching.
- Finding the minimum vertex cover: BFS on residual graph from s.

Minimum Cost for a Bipartite Matching

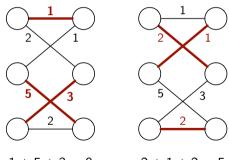
What if the edges in a matching also have an associated costs?

- cardinality of the matching is no longer the only objective
- second priority: minimize the total cost

We search for the cheapest among all maximum matchings.

This does not fit into our model of network flow problems anymore.

two maximum matchings of different cost



$$1 + 5 + 3 = 9$$

$$2+1+2=5$$

More General: Minimum Cost Maximum Flow

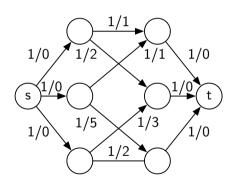
We extend the network flow problem by allowing edge costs that occur per unit of flow.

Input: A flow network consisting of

- \blacksquare directed graph G = (V, E)
- lacksquare source and sink $s,t\in V$
- edge capacity $cap : E \to \mathbb{N}$
- **edge cost** $cost: E \to \mathbb{Z}$.

Output: A flow function f with minimal $cost(f) = \sum_{e \in E} f(e) \cdot cost(e)$ among all flows with maximal |f|.

This can model much more than just minimum cost bipartite matching.



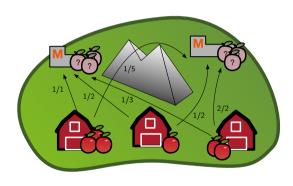
Legend: capacity/cost

Example: Fruit Delivery

Migros wants to schedule fruit deliveries from their farmers to their shops.

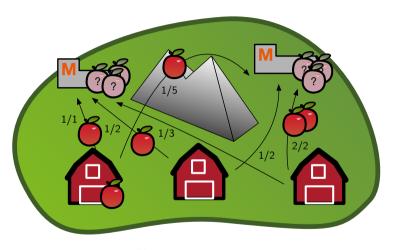
They know all important parameters;

- production per farm [in kg]
- demand per shop [in kg]
- transportation capacity [in kg] and transportation cost [in Fr. pro kg] for every farm-shop pair



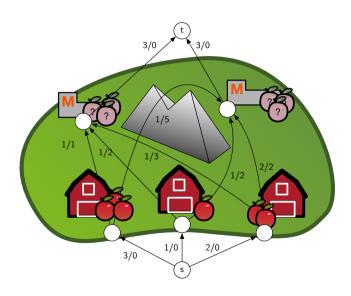
Note: This is not just a bipartite matching, even though the graph is bipartite. One farm might deliver to multiple shops (and vice versa).

Example: Fruit Delivery



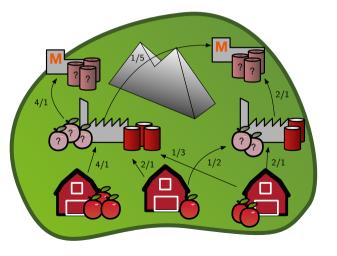
Flow: 2 + 1 + 2 = 5Cost: $1 \cdot 1 + 1 \cdot 5 + 1 \cdot 2 + 2 \cdot 2 = 12$

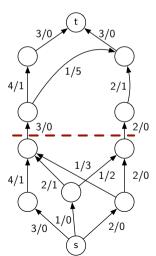
Example: Fruit Delivery



Extended Example: Canned Fruit Delivery

Extension: Canned fruit requires transportation to and from a canning factory.





There are two algorithms available in BGL (available in BGL v1.55+):

- cycle_canceling()
 - slow, but can handle negative costs
 - needs a maximum flow to start with (call e.g. push_relabel_max_flow before)
 - runtime $\mathcal{O}(C \cdot (nm))$ where C is the cost of the initial flow
 - [BGL documentation], [BGL example].
- successive_shortest_path_nonnegative_weights()
 - faster, but works only for non-negative costs
 - sum up all residual capacities at the source to get the flow value
 - runtime $\mathcal{O}(|f| \cdot (m + n \log n))$
 - [BGL documentation], [BGL example].

Useful things to know:

- costs implemented as edge_weight_t property
- a call find_flow_cost() to compute the cost of the flow

Weights and capacities, just one more nesting level in the typedefs:

Code file: → tut9 bgl mincostmaxflow.cpp

```
27 // Graph Type with nested interior edge properties for Cost Flow Algorithms
28 typedef adjacency list traits < vecS, vecS, directedS > Traits;
29 typedef adjacency_list < vecS, vecS, directedS, no_property,</pre>
       property < edge_capacity_t, long,</pre>
30
           property < edge residual capacity t, long,
31
                property<edge reverse t, Traits::edge descriptor,
32
                    property \langle edge weight t, long \rangle \rangle \rangle \rangle Graph; // new!
33
34 // Interior Property Maps
35 typedef property map < Graph, edge capacity t >::type EdgeCapacityMap;
36 typedef property_map<Graph, edge_weight_t >::type EdgeWeightMap; // new!
37 typedef property map<Graph, edge residual capacity t >::type ResCapacityMap;
38 typedef property map<Graph, edge reverse t >::type ReverseEdgeMap;
39 typedef graph traits < Graph > :: vertex descriptor Vertex;
40 typedef graph_traits < Graph > :: edge_descriptor Edge;
41 typedef graph_traits < Graph > :: out_edge_iterator OutEdgelt;
```

Extending the edge adder:

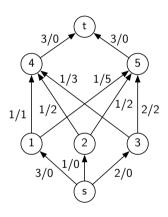
```
46 struct EdgeAdder {
47
      EdgeAdder(Graph & G. EdgeCapacityMap &capacitymap.
48
          EdgeWeightMap &weightmap, ReverseEdgeMap &revedgemap)
49
           : G(G), capacitymap(capacitymap), weightmap(weightmap),
50
             revedgemap(revedgemap) {}
51
      void addEdge(int u, int v, long c, long w) {
52
          Edge e, reverseE;
53
           tie(e, tuples::ignore) = add edge(u, v, G);
54
           tie(reverseE, tuples::ignore) = add edge(v. u. G):
55
          capacitymap[e] = c;
56
          weightmap[e] = w; // new!
57
          capacitymap[reverseE] = 0;
58
          weightmap[reverseE] = -w; // new!
59
          revedgemap[e] = reverseE;
60
          revedgemap[reverseE] = e;
61
```

Building the graph

```
70 const int N=7:
71 \text{ const int } V \text{ source} = 0:
72 const int v farm1 = 1:
73 const int v farm2 = 2;
74 const int v farm3 = 3:
75 const int v shop1 = 4;
76 const int v shop2 = 5;
77 const int v target = 6:
78
                                                         1/0
79 // Create Graph and Maps
                                                     3/0
  Graph G(N):
81 EdgeCapacityMap capacity = get(edge_capacity, G);
82 EdgeWeightMap weight = get(edge_weight, G);
 ReverseEdgeMap rev_edge = get(edge_reverse, G);
  ResCapacityMap res_capacity = get(edge_residual_capacity, G);
  EdgeAdder eaG(G, capacity, weight, rev_edge);
```

Add the edges:

```
eaG.addEdge(v source, v farm1,3,0);
  eaG.addEdge(v source.v farm2.1.0):
  eaG.addEdge(v source.v farm3.2.0):
91
  eaG.addEdge(v farm1.v shop1.1.1);
  eaG.addEdge(v farm1, v shop2,1,5);
  eaG.addEdge(v farm2, v shop1,1,2);
  eaG.addEdge(v_farm2,v_shop2,1,2);
  eaG.addEdge(v farm3, v shop1,1,3);
  eaG.addEdge(v farm3.v shop2.2.2);
98
  eaG.addEdge(v shop1.v target.3.0);
 eaG.addEdge(v_shop2, v_target, 3, 0);
```



Running the algorithm:

```
104 // Option 1:
105 int flow = push_relabel_max_flow(G, v_source, v_target);
106 cycle_canceling(G);
107 int cost = find_flow_cost(G);
108 cout << flow << endl; // 5
109 cout << cost << endl: // 12</pre>
```

Running the algorithm:

```
113 // Option 2: (only for non-negative weights!)
114 successive_shortest_path_nonnegative_weights(G, v_source, v_target);
115 int cost = find_flow_cost(G);
116 int flow = 0;
117 // Iterate over all edges leaving the source to sum up the flow values.
118 OutEdgelt e, eend;
119 for(tie(e, eend) = out_edges(vertex(v_source,G), G); e != eend; ++e) {
120    flow += capacity[*e] - res_capacity[*e];
121 }
122 cout << flow << endl; // 5
123 cout << cost << endl; // 12</pre>
```

Min Cost Max Flow with BGL: Summary

What you should remember:

- Minimum cost maximum flow is a powerful and versatile modeling tool.
- With BGL, it is not much harder to use than the regular network flow algorithms.
- Reformulating the problem as a graph without negative costs allows us to use a faster algorithm in BGL.
- Sample implementations are available on the Judge and in the BGL Docs.

If you are interested in the theory behind these algorithms: (not needed for this course)

■ Stanford CS 261 by Prof. Tim Roughgarden, Spring 2016, full lectures on Youtube