LECTURE (October 21st)

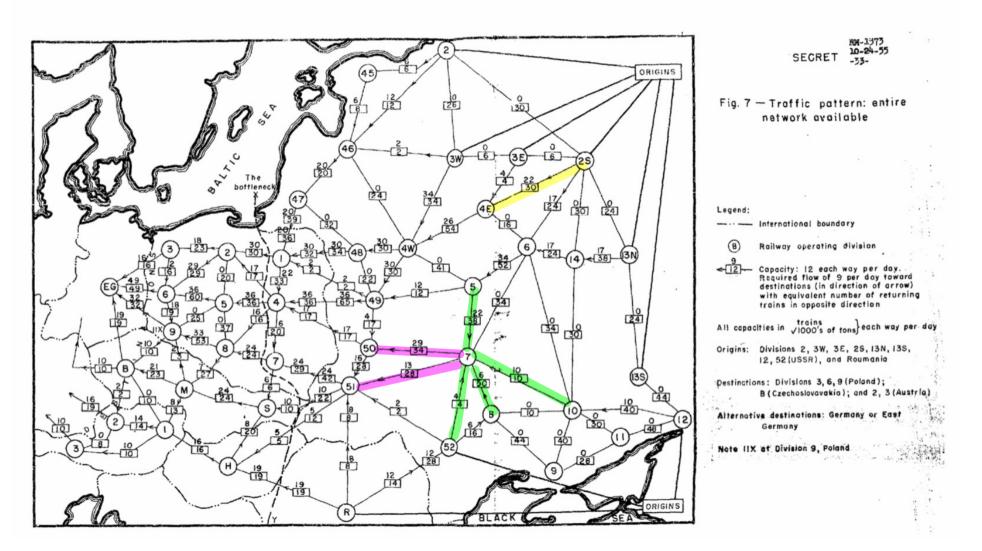
Maximum Flows & Minimum Cuts

Today we are going to look at flows, which might be a more familiar territory

- Given a graph with some data associated to it, we want to compute some structure within that graph
- You might have seen related examples in previous courses like CS225, where you saw minimum spanning trees & Dijkstra's algorithm, CS374 where you saw all pairs chartest path algorithm. Those structures that you are computing — shortest path, minimum spanning tree, etc. — are subgraphs of your input graph
- In particular, you want to pick out a subgraph of your graph that satisfy some optimality properties

Today, we are going to look at other kinds of optimal structure

Before we look at the definition, let us take a detour into history of the problem



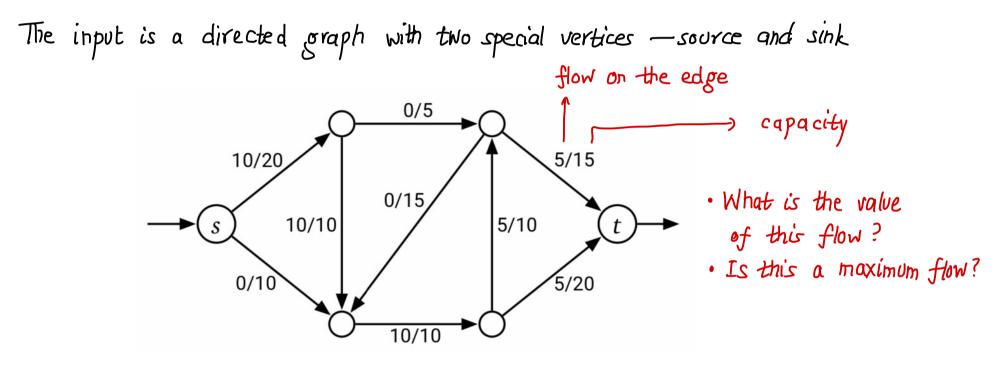
This is a map of Eastern Europe after the second world war when cold war was looming

The circles represent cities & the labels are just identifier The numbers on the box between two cities represent the number of trains that go between the citles daily, think of it as the apacity of the rail line The numbers outside the boxes and the arrow represents a schedule to use some sort of material which is mined at the box labeled origins

So, we want to ship material from Moscow to East Berlin Instead of using the entire capacity, we can send fewer trains on a single track

If you look at any city that is not Moscow or East Berlin, then you will notice that the amount of stuff going into the city is the same as the amount of stuff going out, since it does not make sense to have extra stuff flowing into intermediate city and by definition the material is only produced at the origin, so it does not make sense for more stuff to leave the city than what came in

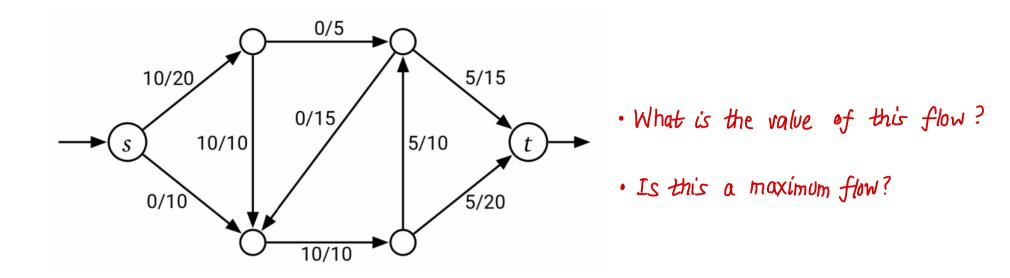
This is an example of the maximum flow problem



Each edge has a number associated to it called the capacity and you want to compute a second number for every edge called the flow value that satisfies the conservation constraint at all intermediate vertices, i.e., the flow coming in equals the flow coming out

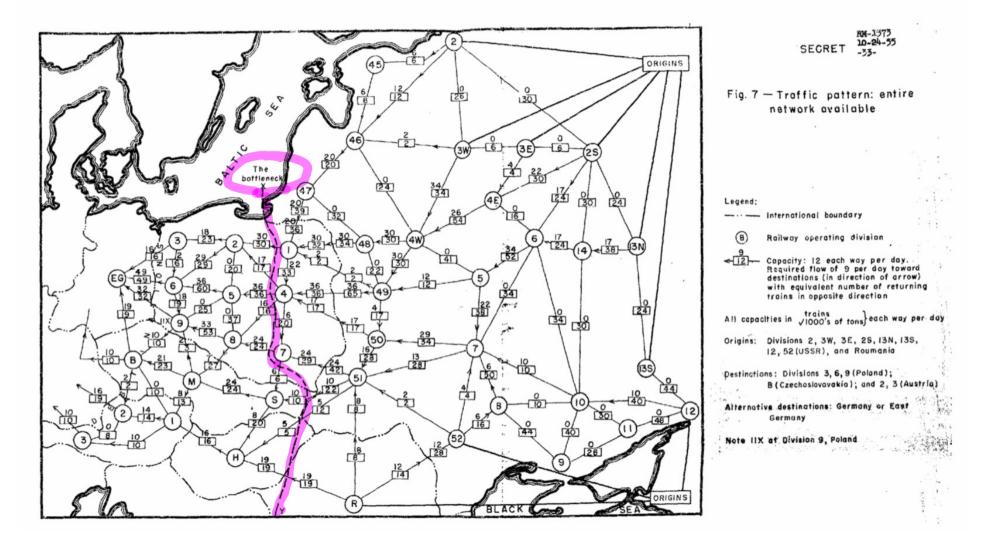
<u>Maximum Flow Problem</u> Given G = (V, E) directed capacity function $c : E \rightarrow IR_{30}$

One useful analogy is to think of the edges as pipes and the capacity as the capacity of the pipe before it explodes



We will see shortly why this is not a maximum flow & how to get to a maximum flow from here

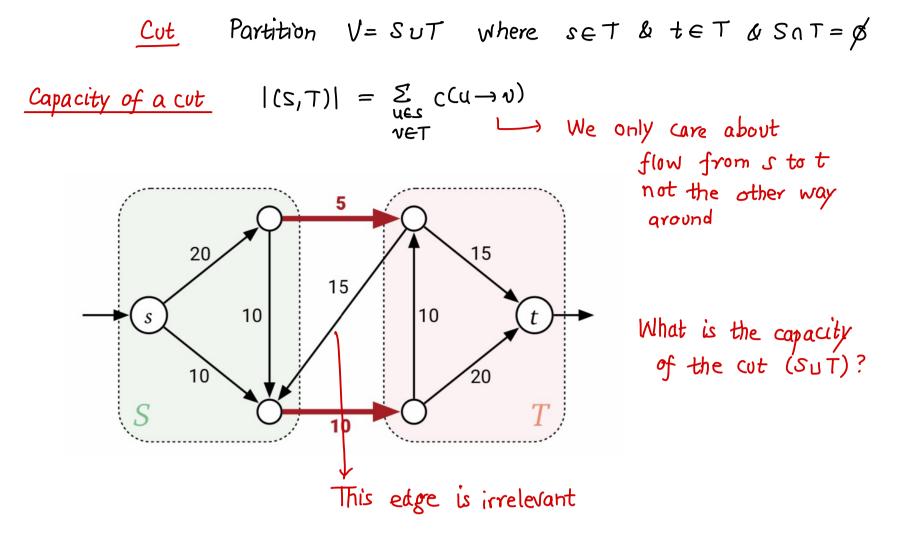
Before we do that, let's introduce the evil twin of the maximum flow problem called the minimum cut problem



This map was made by RAND corporation in the 50s — a lot of algorithm research originated there in the 1950s

This map was classified until 1999, when an optimization researcher Lex Schriver wrote to the US government to declassify it

The bottleneck is the smallest cost of destroying all the rail lines to disconnect Moscow from East Berlin — cost equals the capacity of the rail line This is called the minimum cut problem. The input is exactly the same with a source s & target t but now we are trying to separate the source & target, i.e., divide the vertices in two parts where one contains s & the other contains t. Such a partition is called a cut.



We want the cut with the smallest capacity, i.e., the smallest cost to disconnect s from t

The max-flow min-cut theorem, whose proof we will see in today's lecture says that in a priven proph

Max-flow Min-cut Theorem

The proof of this theorem will also give us an algorithm to compute
both the maximum flow & the minimum Cut
Let's see the easy direction of the proof first max
$$|f| \le \min |(S,T)|$$

Pick any feasible flow f and any cut (S,T)
Value of f is $|f| = \sum_{u}^{n} f(s \rightarrow w) - \sum_{u}^{n} f(u \rightarrow s)$



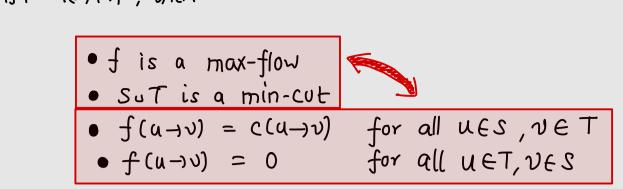
Since all vertices except s,t have the same of flow going in as coming out, we have

 \Rightarrow max $|f| \leq \min |(S,T)|$

We have used all the properties of the flow - value of the flow - conservation - non-pegotivity

non-negativity
feasibility

But one of the thing this implies that if we find a flow & a cut that have the same value, then the flow must be a maximum flow & cut must be the minimum cut and all the inequalities must be tight If If I= I(S,T) , then



Let's see the proof of the other direction

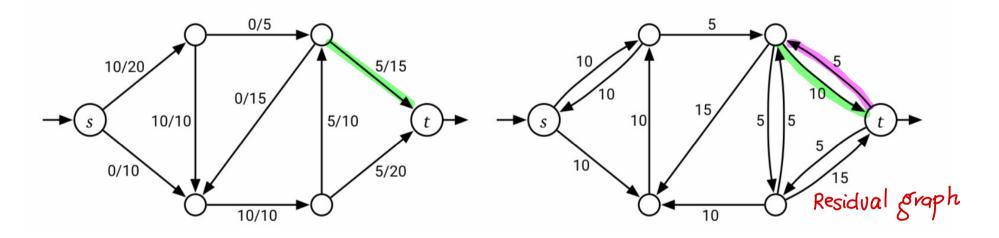
$$\max(f) > \min(S,T)$$

The proof will be easier if we assume there is at most one edge between any two vertices — this is easy to handle [Why?]

Pick your favorite flow f, we define a new flow problem using residual capacities & residual graphs

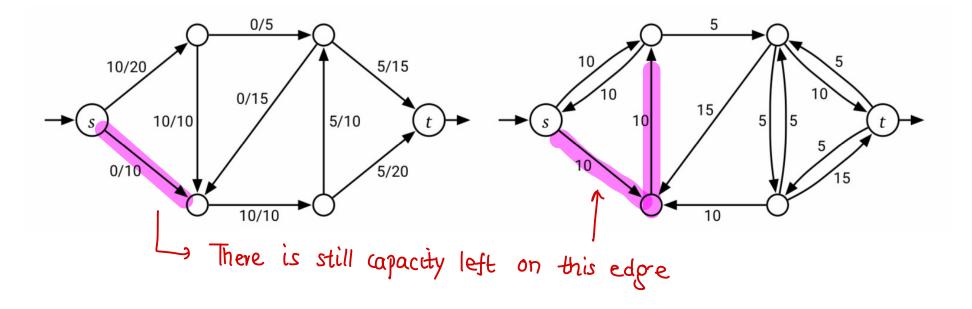
These capture how much of the capacity is not being used

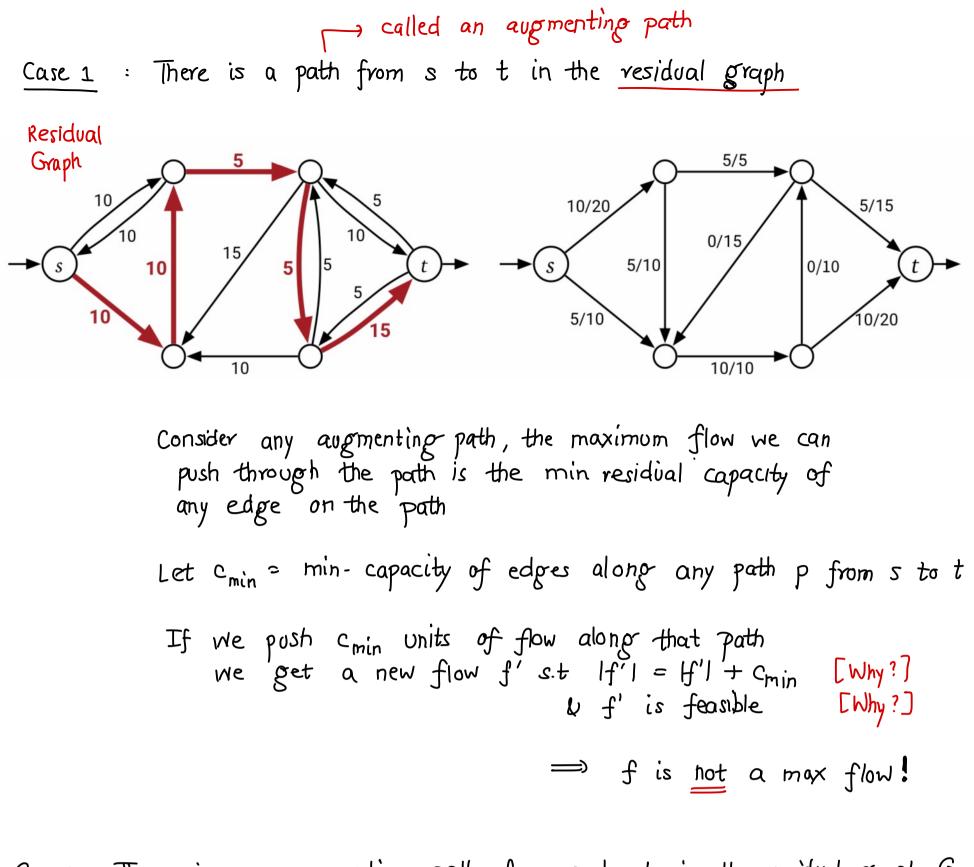
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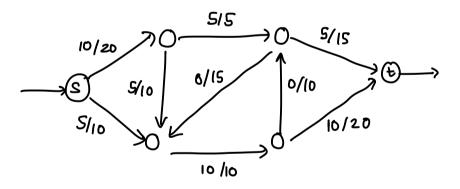
Define residual capacity for flow f by $C_{f}(u \rightarrow v) = \begin{cases} c(u \rightarrow v) - f(u \rightarrow v) & \text{if } u \rightarrow v \in E \\ f(v \rightarrow u) & \text{if } v \rightarrow u \in E \\ 0 & 0/\omega \end{cases}$

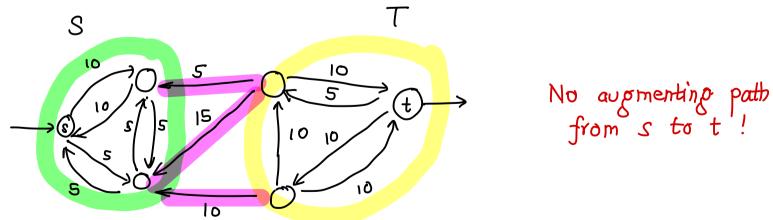
[Why?] Residual capacities are always non-negrative





There is no augmenting path from s to t in the residual proph Gr Case 2





Let
$$S = all$$
 vertices reachable from s in G_f
 $T = V \setminus S$

For every vertex
$$u \in S$$
, $v \in T$
if $u \rightarrow v \in E$, then $f(u \rightarrow v) = C(u \rightarrow v)$
if $v \rightarrow u \in E$, then $f(u \rightarrow v) = 0$

Recall, what we saw earlier

If If I = I(S,T) I, then • f is a max-flow • SuT is a min-cut • f(u+v) = c(u+v) for all u \in S, v \in T • f(u+v) = 0 for all u \in T, v \in S Thus, f is a max-flow & (S,T) is a min-cut This algorithm was discovered by Ford-Fulkerson in 1952 Augmenting Paths Algorithm Initialize $f \leftarrow 0$ $G_{f} \leftarrow G$ While there is a path p from s to t in G_{f} push flow along p rebuild G_{f} Return f

Runtime of this algorithm? NEXT LECTURE

