15451 Spring 2023

Max Flow: polynomial-time algorithms

Elaine Shi

Recall the max flow problem



Directed graph, each edge e has a capacity c(e)

Recall Ford-Fulkerson

While exists augmenting path P of positive residual capacity Push the maximum possible flow along P #edges Running time: O(m F) flow value

Recall Ford-Fulkerson

While exists augmenting path P of positive residual capacity Push the maximum possible flow along P

Running time: O(m F)

"Pseudo-polynomial time"

Can we have an algorithm whose running time is polynomial in only m and n?



Can we avoid the pathological cases if we select the augmenting path more cleverly?







When people implemented FF using **BFS**, the performance is always polynomial.

Why?

Edmonds-Karp



When people implemented FF using **BFS**, the performance is always polynomial.

Why?

BFS shortest path

Edmonds-Karp's Algorithm

Run FF but select the **shortest** augmenting path

Theorem: Edmonds-Karp makes at most in $O(n \cdot m^2)$ time

Theorem: Edmonds-Karp makes at most mn iterations

Proof: let(d)be the <u>distance from s to t in the</u> residual graph

- Claim 1: d never increases
- Claim 1: d never increases
 Claim 2: every m iters, d increases by at least 1

Claim 1: d never ecreases d(s,v): shortest distance from s to v' proof: We will t' d(s, v) never decreases during each iter, how the residual graph can change ① some edges may get removed b/c they are saturated 2) add a backward edge (w.r.t. Current residual graph) (U, v). (V, u) is on the shortest path from s-t in the current residual graph (u, v) will not decrease d(S, v) = d(S, v) + 1(u, v) will not decrease d(S, v') for any v'

Claim 2: every <u>m iterations</u>, d ī: iter increases by at least 1 G': residual graph in iter Gi Suppose that in G^j, we first start to use a back edge wird Gⁱ, then d^{G^j}(s,t)>d^G(s,t) if this is true. Claim2 follows b/c every aug. path saturates at least one edge. there are G 'only O(m) edges in Gⁱ

Claim: Suppose in G¹, we first start to use a back edge w.r.t. G¹, then d^G(s.t)>d^G(s.t) (U, J) is the last back in G edge w.r.t $S = \frac{1}{2} \int_{a}^{a} \frac{1}{2} \int_{a}^{b} \frac{1}{2$ aug. path ing (V, U) is on shortest path in Gi $d^{G'}(s,t) = d^{G'}(s,v) + 1 \qquad > d^{G'}(v,t)$ $d^{G'}(s,t) = d^{G'}(s,v) + d^{G'}(v,t)$

Running time of Edmonds-Karp: O(nm²) n phases (in term of d) each phase: at most m iters each iter: BFS: D(m)

Dinic's algorithm: O(n^2m) 1970, "Dinitz" $E K O(n m^2)$

Dinic's algorithm: O(n²m) 1970



A single BFS finds distances of all vertices from s, which encodes every shortest path

Let's not waste this work!



A **blocking flow** in a residual graph G_f is a union of flows on shortest augmenting paths such that every shortest path in G_f has at least one edge that is saturated

Dinic's algorithm

- Initially, the residual graph is the original graph
- Repeat at most **n phases**:
 - Construct the layered graph of the residual graph
 - Find a blocking flow on the layered graph
 - Augment current flow using blocking flow, update residual graph

Goal: find blocking flow in O(nm) time $\implies O(n^2,m)$ time in total

Dinic's algorithm

• Initially, the residual graph is the original graph

d increases every phase

- Repeat at most **n phases**:
 - Construct the layered graph of the residual graph
 - Find a blocking flow on the layered graph
 - Augment current flow using blocking flow, update residual graph

Goal: find blocking flow in O(n m) time

Find blocking flow using the layered graph



Claim: all augmenting paths of len d must be paths s-t paths in this layered graph





Example of layered graph, blocking flow





Naive algo. for finding blocking flow: m iterations Repeat until no more path found ℓ

- Find a shortest path in layered graph using DFS, push maximum flow through it
- Update the residual capacities and remove edges with o residual capacity $Total: O(m^2)$

Want: $O(m \cdot n)$

'm time

Naive algo. for finding blocking flow:

Repeat until no more path found

- Find a shortest path in layered graph using DFS, push maximum flow through it
- Update the residual capacities and remove edges with 0 residual capacity

Running time: O(m²)

Improved algo: mark dead-end paths and prune them in future searches



Analysis of improved algorithm m DFS iters. cost in iter $\hat{i} = n + (\# new dead-end)$ edges encountered max len of shortest path we find total $\omega st = \sum_{i=1}^{m} n + \chi_i = m \cdot n + (\sum_{i=1}^{m} \chi_i)$ $\leq mn + m$

Dinic's algorithm often performs better than O(mn²) in practice

blocking flow: O(m.n) time

n total phases

Dinic's algorithm often performs better than O(mn²) in practice

Claim: For a unit capacity graph, a blocking flow can be found in O(m) time

Claim: For a unit capacity graph, we need at most $2\sqrt{m}$ blocking flows

Claim: For a unit capacity graph, Dinic finishes in

 $O(m \cdot \min(\sqrt{m}, n))$ time

Claim: For a unit capacity graph, a blocking flow can be found in O(m) time

ble residual graph is also unit capacity every edge can be on at must l aug. path in the blacking flow in the blocking flow, total path len < M

Claim: For a unit capacity graph, we need at most $2\sqrt{m}$ blocking flows proof: suppose we have completed & phases so far d 7,k there are at most $\frac{m}{k}$ and paths of $\frac{4k}{4}$ (en 7,k max flow in residual net work $\leq \frac{m}{k}$ =) need at most $\frac{m}{k}$ additional blocking flows $\frac{1}{k}$. Sim

Latest developments

Maximum Flow and Minimum-Cost Flow in Almost-Linear Time

Li Chen^{*} Georgia Tech lichen@gatech.edu Rasmus Kyng[†] ETH Zurich kyng@inf.ethz.ch Yang P. Liu[‡] Stanford University yangpliu@stanford.edu

Richard Peng University of Waterloo [§] y5peng@uwaterloo.ca Maximilian Probst Gutenberg[†] ETH Zurich maxprobst@ethz.ch Sushant Sachdeva[¶] University of Toronto sachdeva@cs.toronto.edu

April 26, 2022

Abstract

We give an algorithm that computes exact maximum flows and minimum-cost flows on directed graphs with m edges and polynomially bounded integral demands, costs, and capacities in $m^{1+o(1)}$ time. Our algorithm builds the flow through a sequence of $m^{1+o(1)}$ approximate undirected minimum-ratio cycles, each of which is computed and processed in amortized $m^{o(1)}$ time using a new dynamic graph data structure.

Our framework extends to algorithms running in $m^{1+o(1)}$ time for computing flows that minimize general edge-separable convex functions to high accuracy. This gives almost-linear time algorithms for several problems including entropy-regularized optimal transport, matrix scaling, *p*-norm flows, and *p*-norm isotonic regression on arbitrary directed acyclic graphs.