

Algorithm Theory, Tutorial 7

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Ex 1

Let us consider the following modified version of the contraction algorithm presented in the lecture. Instead of choosing an edge uniformly at random and merging its endpoints, in each step the modified algorithm chooses a pair of nodes in graph G uniformly at random and merges the two nodes into a single node.

- (a) Give an example graph of size at least n where the above algorithm does not work well, that is, where the probability of finding a minimum cut is exponentially small in n .
- (b) Prove that property for the example you gave in (a), i.e., show that the modified contraction algorithm has probability of finding a minimum cut at most c^n for some constant $c < 1$.

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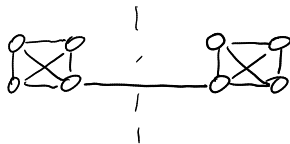
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- Draw example.

1a, Example pictures

Which pairs of nodes are bad for us (destroy our min-cut?)

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- We analyze the chance that this happens in the first $T = n/4$ contractions of the modified algorithm.
- Let n_1, n_2 be the number of remaining nodes in K_1, K_2 after T contractions. Since we contract at most $n/4$ pairs, we have $n_1, n_2 \geq n/4$.



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$$\mathbb{P}(\mathcal{E}_i) = 1 - \mathbb{P}(\overline{\mathcal{E}_i}) = 1 - \mathbb{P}(\text{"}u, v \text{ from different cliques"}) \leq 1 - 2 \cdot \frac{1}{4} \cdot \frac{1}{4} = \frac{7}{8}$$

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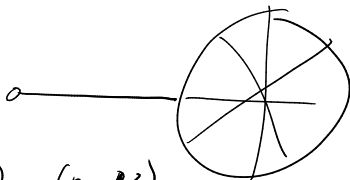
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- This is a necessary condition for finding the min-cut.
- Since our bounds $\mathbb{P}(\mathcal{E}_i) \leq 7/8$ for the probabilities of the events \mathcal{E}_i are independent from one another, we have

$$\mathbb{P}(\mathcal{E}) = \mathbb{P}\left(\bigcap_{i=1}^{n/4} \mathcal{E}_i\right) \leq (7/8)^{n/4} = \boxed{\left(\sqrt[4]{7/8}\right)^n} = c^n, \text{ where } c := \sqrt[4]{7/8} < 1.$$



$$\frac{n-1}{n} \cdot \frac{(n-2)}{n-1} \cdot \frac{(n-3)}{(n-2)} \cdot \dots \cdot \frac{2}{3}$$

$$= \frac{2}{n}$$

$$\leq C^n$$

$$C \leq 1$$

Metric TSP with Small Edge Weights

Consider the family of complete, weighted, undirected graphs $G = (V, E, w)$ in which all edges have weight either 1 or 2.

Remark: TSP is the Traveling Salesperson Problem. The goal is to find a tour, i.e., a permutation v_1, \dots, v_n of nodes, that minimizes the total weight of edges on that tour $w(v_1, v_n) + \sum_{i=1}^n w(v_i, v_{i+1})$.

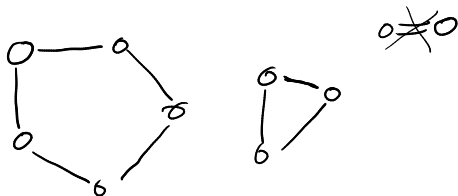
- Ⓐ Assume you have a subroutine that computes a *minimum* 2-matching for the above family of graphs in polynomial time. Describe an *efficient* algorithm that computes a $4/3$ -approximation for the TSP problem for graphs of this family.

Remark: A 2-matching is a subset $M \subseteq E$, so that every $v \in V$ is incident to exactly 2 edges in M .

- Ⓑ Prove that your algorithm computes a $4/3$ -approximation of TSP on the this family of graphs.

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- The minimum TSP is also a 2-matching.
- \Rightarrow Minimum T2-matching is lower bound for minimum TSP.
- Idea for Algorithm: Calculate minimum 2-matching and then merge the cycles to a ~~solution~~
approximation of the min tsp.

Algorithm

- Calculate the min 2-matching.

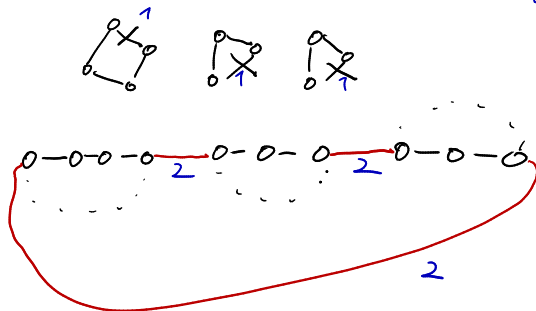
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worst case



$$\text{Approx} \leq 2 \text{ Two Match} + \frac{n}{3}$$

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- We compute

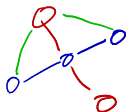
$$\frac{n}{3} \leq \frac{O}{3}$$

$$\boxed{S} \leq M + \frac{n}{3} \leq O + \frac{n}{3} \leq \boxed{\frac{4}{3}O} \Rightarrow \frac{4}{3} \text{ approx.}$$

Ex 3

Consider an undirected, unweighted Graph $G = (V, E)$ with n nodes. We are given a set \mathcal{P} of p *simple* paths in G , where each path has exactly ℓ nodes. We say a path $P \in \mathcal{P}$ is *covered* by a set Q of nodes if P has a node in Q . Then, the goal is to find a set $Q \subseteq V$ of nodes with minimum cardinality such that *every* path in \mathcal{P} is covered by Q .

Now consider the following simple greedy algorithm: It starts with $Q = \emptyset$. As long as there is a path not covered by Q , the node that covers the most uncovered paths is added to Q .



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- Since $s \leq p$ the claim follows.

$1 + \ln p$ approx.

Show that after selecting i nodes, there are at most $p(1 - \frac{\ell}{n})^i$ uncovered paths left.

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- There must be one node that covers at least $p\ell/n$ paths.
- The greedy algorithm selects such a node that covers most paths, we “loose” at least $p\ell/n$ many paths in that step, hence we have

$$u_1 \leq p - \frac{p\ell}{n} = p\left(1 - \frac{\ell}{n}\right)$$

many paths left.

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- Together we get

hypothesis

$$u_{i+1} \leq u_i - u_i \ell / n = u_i \cdot (1 - \ell / n) \leq p \left(1 - \frac{\ell}{n}\right)^i \left(1 - \frac{\ell}{n}\right) = p \left(1 - \frac{\ell}{n}\right)^{i+1}$$

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- This proves our claim

- Show that for $i > \frac{n}{\ell} \ln p$ all paths are covered. *Hint:* $1 - x < e^{-x}$ for all $x > 0$

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- \swarrow from 3b
- We show that $i > \frac{n}{\ell} \ln p$ implies $p(1-\frac{\ell}{n})^i < 1$ which means that all paths are covered by greedily selecting i nodes (realtime calculation)

$$\begin{aligned}
 i &> \frac{n}{\ell} \ln(p) \\
 \frac{\ell i}{n} &> \ln(p) \\
 e^{\frac{\ell i}{n}} &> p \\
 1 &> p \cdot e^{-\frac{\ell i}{n}} > p \cdot \left(1 - \frac{\ell}{n}\right)^i
 \end{aligned}$$

$\left(e^{-\frac{\ell}{n}}\right)^i$
 \swarrow + Hint

$$\begin{aligned}
& i > \frac{n}{\ell} \ln p \\
\iff & -\frac{i\ell}{n} < -\ln p \\
\iff & e^{-\frac{i\ell}{n}} < \frac{1}{p} \\
\iff & \left(e^{-\frac{\ell}{n}}\right)^i < \frac{1}{p} \\
\stackrel{\text{Hint}}{\implies} & \left(1 - \frac{\ell}{n}\right)^i < \frac{1}{p} \\
\iff & p\left(1 - \frac{\ell}{n}\right)^i < 1
\end{aligned}$$

Exercise 4

In the lecture, we showed that every (undirected) graph with edge connectivity λ has at most $n^{2\alpha}$ cuts of size at most $\alpha \cdot \lambda$. Use this fact to prove the following statement:

Let $G = (V, E)$ be an undirected graph with constant edge connectivity $\lambda \geq 1$ and let $p := \min \left\{ 1, \frac{c \ln n}{\lambda} \right\}$, where $c > 0$ is a constant. Assume that edge $e \in E$ is sampled independently with probability p . Let E_p be the set of sampled edges and let $G_p = (V, E_p)$ be the graph induced by the sampled edges. Show that if the constant c is chosen sufficiently large, the graph G_p is connected with high probability.

Hint: A graph is connected if and only if there is an edge across each of the $2^{n-1} - 2$ possible cuts. Analyze the probability that for a cut of a given size k in G , at least one edge is sampled in E_p . Then, use the upper bound from the lecture on the number of cuts of a given size and a union bound over all cuts of a given size in G . Finally, one can do a union bound over all possible cut sizes.

Ex 4 Solution

- We assume $p < 1$, otherwise $G_p = G$ is obviously connected.

$$\begin{aligned}(1-p)^k &= \left(1 - \frac{c \ln n}{\lambda}\right)^k \\ &\stackrel{(*)}{\leq} \exp\left(-\frac{kc \ln n}{\lambda}\right) \\ &= \exp\left(-\alpha c \ln n\right) = n^{-c\alpha}\end{aligned}\qquad (*) : 1 - x < e^{-x}$$

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- The graph G_p becomes disconnected if all edges of any cut are removed.
- We first consider one fixed cut of size $k := \alpha\lambda$
- The probability that we remove all edges of a cut of size $k := \alpha\lambda$ is $(1 - p)^k$. We obtain

$$\begin{aligned}(1 - p)^k &= \left(1 - \frac{c \ln n}{\lambda}\right)^k \\ &\stackrel{(*)}{\leq} \exp\left(-\frac{kc \ln n}{\lambda}\right) \\ &= \exp(-\alpha c \ln n) = \boxed{n^{-c\alpha}} \\ e^{-\alpha c \ln n} &= n^{-\alpha c}\end{aligned}$$

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fixed

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- Let us enumerate these cuts with C_1, \dots, C_ℓ with $\ell \leq n^{2\alpha}$
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- Let $\mathcal{E}^k := \bigcup_{i=1}^{\ell} \mathcal{E}_i^k$ be the event that all edges of at least one of the cuts C_i out of the cuts C_1, \dots, C_ℓ of size k are removed.

- Last slide: cut of size $\alpha\lambda$ has no edges with probability $< n^{-c\alpha}$
- In the lecture we saw that we have at most $n^{2\alpha}$ cuts of size $k = \alpha\lambda$.
- Let us enumerate these cuts with C_1, \dots, C_ℓ with $\ell \leq n^{2\alpha}$
- Let \mathcal{E}_i^k be the event that all edges of a specific cut C_i are removed.
- Let $\mathcal{E}^k := \bigcup_{i=1}^{\ell} \mathcal{E}_i^k$ be the event that all edges of at least one of the cuts C_i out of the cuts C_1, \dots, C_ℓ of size k are removed.
- We do a union bound and get

$$\begin{aligned}
 \mathbb{P}(\mathcal{E}^k) &= \mathbb{P}\left(\bigcup_{i=1}^{\ell} \mathcal{E}_i^k\right) \stackrel{\text{union bound}}{\leq} \sum_{i=1}^{\ell} \mathbb{P}(\mathcal{E}_i^k) \\
 &\leq \ell n^{-c\alpha} \leq n^{2\alpha} \cdot n^{-c\alpha} = n^{-(c-2)\alpha} = n^{-(c-2)k/\lambda}
 \end{aligned}$$

\downarrow
 $\ell \leq n^{2\alpha}$

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- Clearly we have $k \leq n$.
- Let $\mathcal{E} := \bigcup_{k=1}^n \mathcal{E}^k$. We obtain

$$\begin{aligned}
 \mathbb{P}(\mathcal{E}) &= \mathbb{P}\left(\bigcup_{k=1}^n \mathcal{E}^k\right) \leq \sum_{k=1}^n \mathbb{P}(\mathcal{E}^k) \\
 &\leq \sum_{k=1}^n n^{-(c-2)k/\lambda} \leq \sum_{k=1}^n n^{-(c-2)/\lambda} \quad \begin{array}{l} \swarrow k \text{ makes} \\ \swarrow \text{us smaller} \end{array} \\
 &= n \cdot n^{-(c-2)/\lambda} = n^{-(c-2-\lambda)/\lambda}
 \end{aligned}$$

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- This term is independent from the graph size n and thus indeed a constant as required.

