طراحی الگوریتم ها

مبحث شانزدهم: برش کمینه

سجاد شیرعلی شهرضا بهار 1402 سه شنبه، 19 اردیبهشت 1402

اطلاع رساني

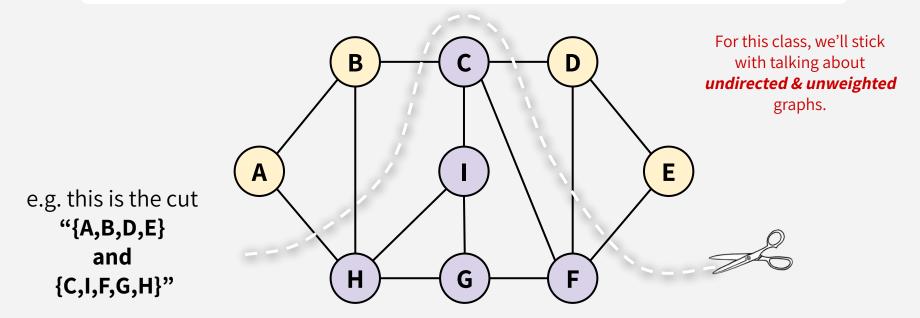
- بخش مرتبط کتاب برای این جلسه: 26
 مهلت ارسال تمرین سوم: صبح یکشنبه 24 اردیبهشت

برش کمینه

یک برش کمینه در گراف چیست؟

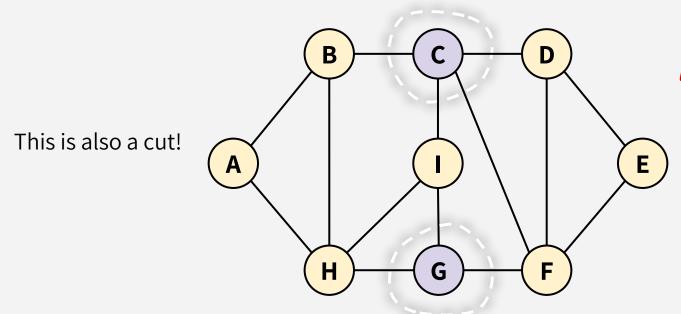
RECALL: CUTS IN GRAPHS

A **cut** is a partition of the vertices into two nonempty parts.



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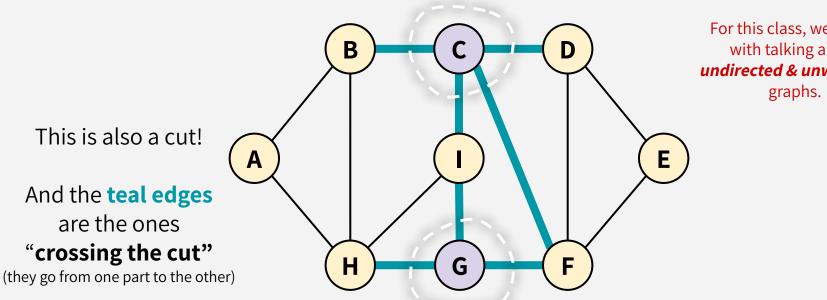
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For this class, we'll stick with talking about undirected & unweighted graphs.

EDGES THAT "CROSS" A CUT

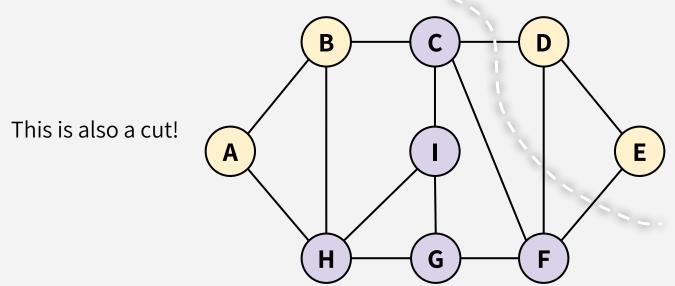
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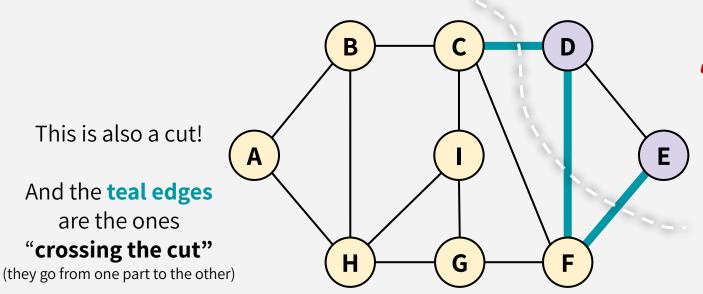
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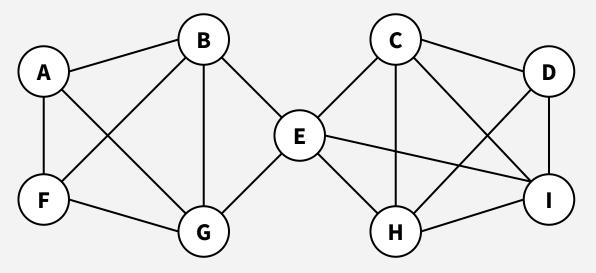
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For this class, we'll stick with talking about undirected & unweighted graphs.

A (global) **minimum cut** is a cut that has the fewest edges possible crossing it.

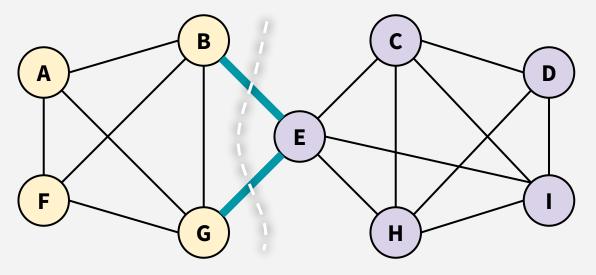
What is the (global) minimum cut in this graph?



Note: we emphasize "global" because later in class, we'll talk about *s*–*t minimum cuts*, which is a cut that separates specific nodes s and t (a slightly different concept).

A (global) **minimum cut** is a cut that has the fewest edges possible crossing it.

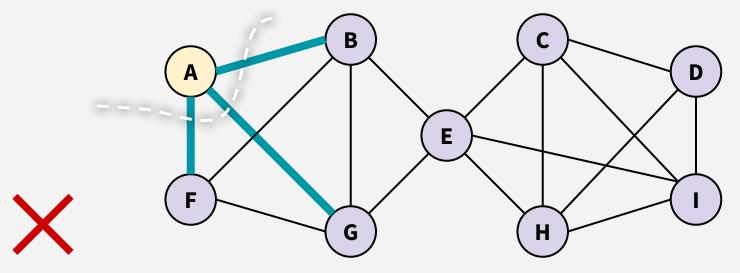
What is the (global) minimum cut in this graph?



This cut is the (global) minimum cut (it has only 2 edges crossing it)!

A (global) **minimum cut** is a cut that has the fewest edges possible crossing it.

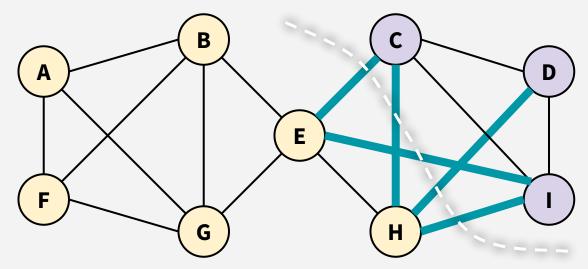
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What is the (global) minimum cut in this graph?





APPLICATIONS?

There are several immediate use cases for finding global minimum cuts!

Clustering

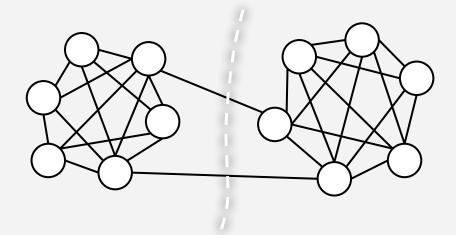
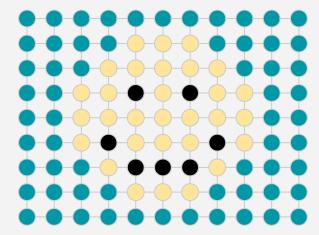


Image Segmentation*



*Although we're only working with unweighted graphs for now, we could represent an image using large edge weights between similar pixels





یک الگوریتم تصادفی برای یافتن برش کمینه سراسری در گراف بدون جهت

LAS VEGAS vs. MONTE CARLO

LAS VEGAS ALGORITHMS

Guarantees correctness!

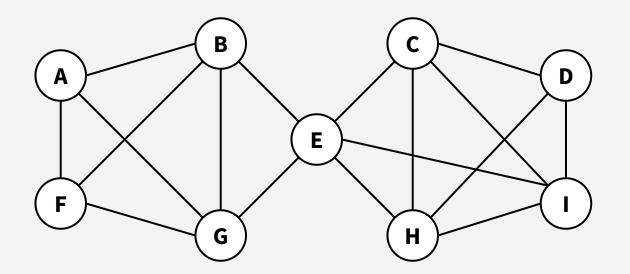
But the runtime is a random variable. (i.e. there's a chance the runtime could take awhile)

MONTE CARLO ALGORITHMS

Correctness is a random variable. (i.e. there's a chance the output is wrong)

But the runtime is guaranteed!

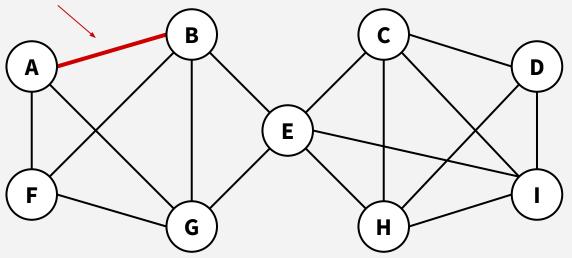
Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



As always, let's see an example of the algorithm in action!

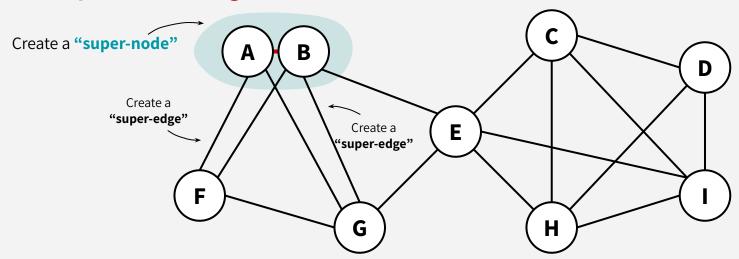
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Pick a random edge!



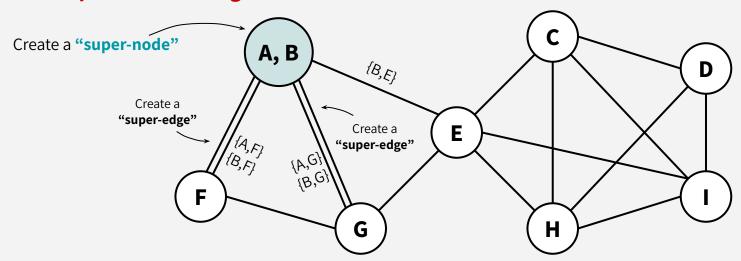
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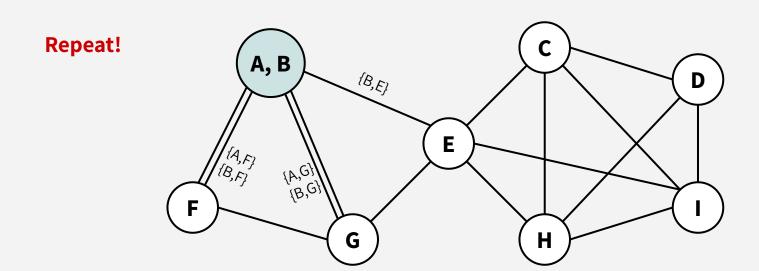
Now perform an "edge contraction"!

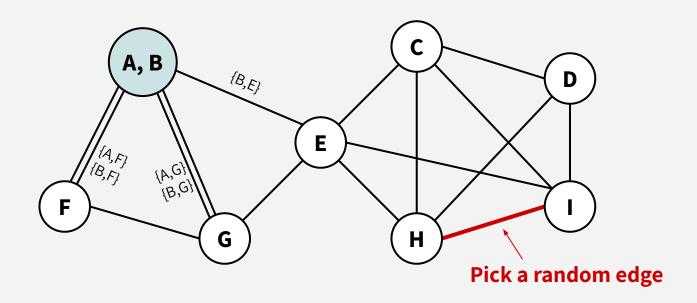


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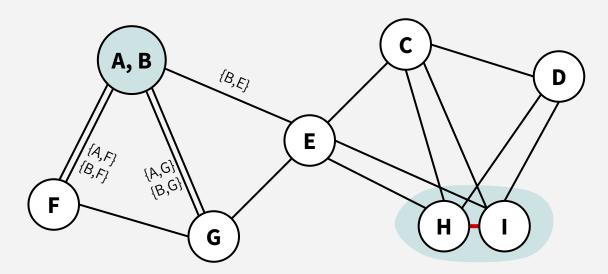
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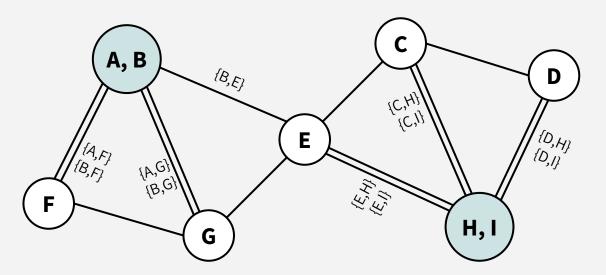


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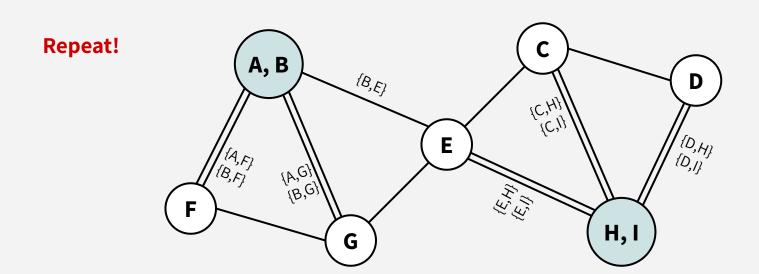


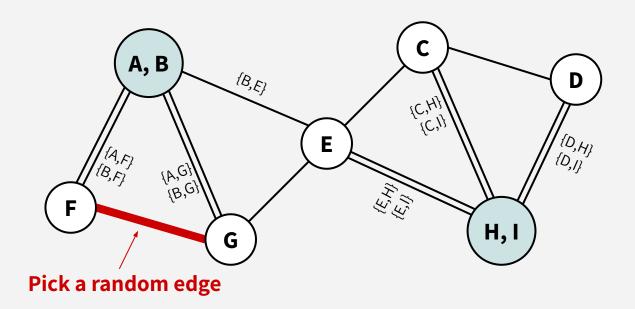
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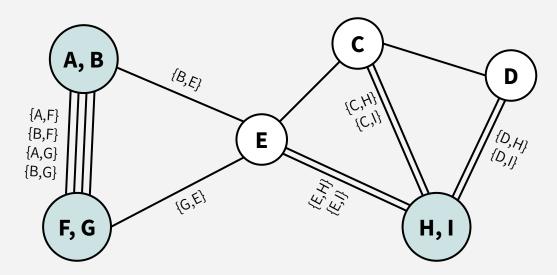


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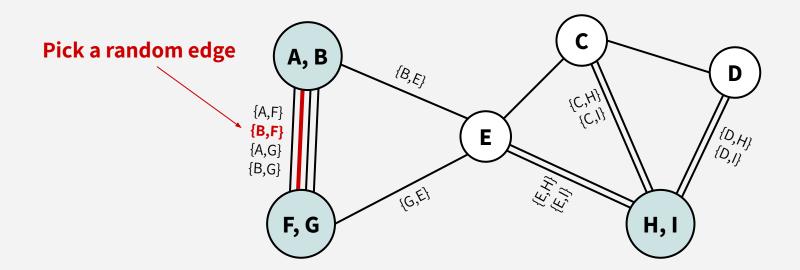


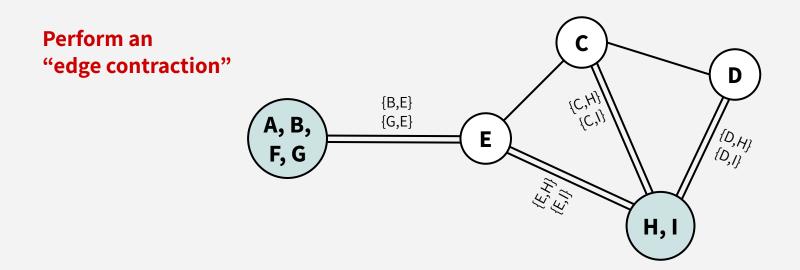


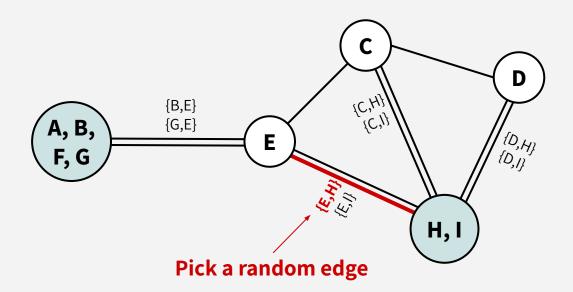
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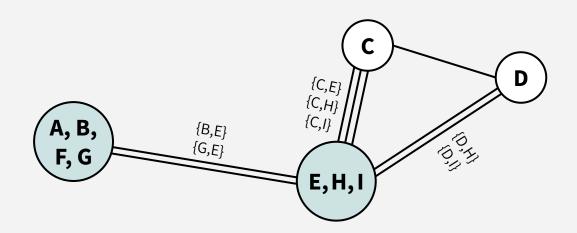
Perform an "edge contraction"



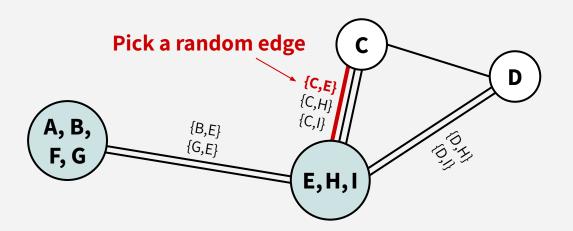




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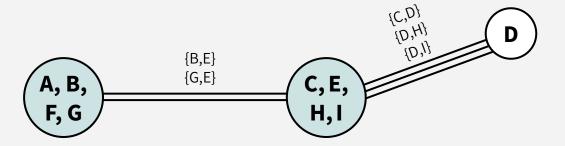


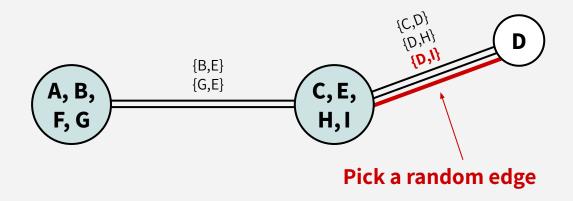
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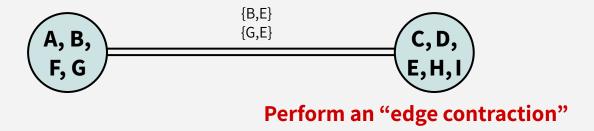


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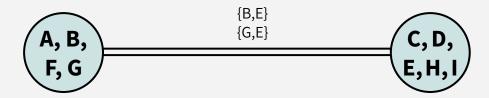






Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

We return the cut given by the remaining 2 super-vertices!

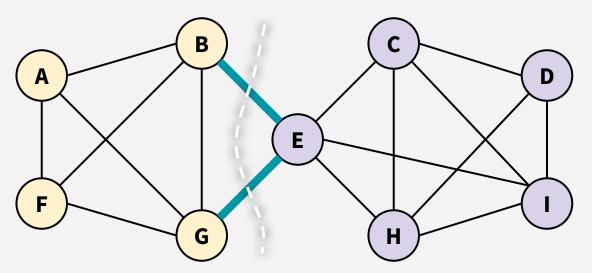


KARGER'S ALGORITHM

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

We return the cut given by the remaining 2 super-vertices!

(Just bringing back our original graph to see what this corresponds to)





زمان اجرا و درستی الگوریتم کارگر

KARGER'S ALGORITHM

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KARGER(G=(V,E)):
   vertices of G' = {supernode(v) for v in V}
   E'_{u,v} = \{(u,v)\} \text{ for } (u,v) \text{ in } E
   E'_{u,v} = \{\} for (u,v) not in E
   F = a copy of E
   while |G'| > 2:
       \{(u,v)\}\ = \ uniformly \ random \ edge \ in \ F
       MERGE_SUPERNODES(u, v)
       F = F \setminus E_{u,v}
   return the cut given by the remaining two supernodes!
MERGE_SUPERNODES(u,v):
   x' = supernode(nodes in u' U nodes in v')
   for each w' in G' \setminus \{u', v'\}:
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KARGER'S ALGORITHM: RUNTIME

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(we can be more clever with a union-find data structure but don't worry about that for today!)

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The while loop executes n – 2 times...

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Runtime (w/o fancy data structures): $O(n^2)$

KARGER'S ALGORITHM

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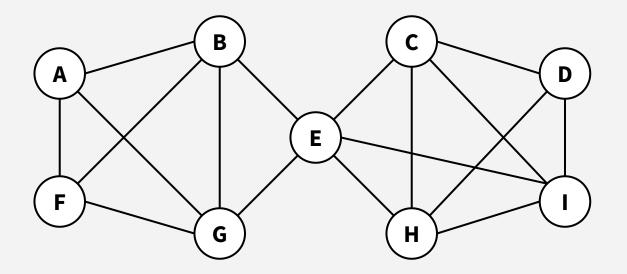
(Just bringing bac our original grap to see what this corresponds to)

How fast is this? Does it always work?

F

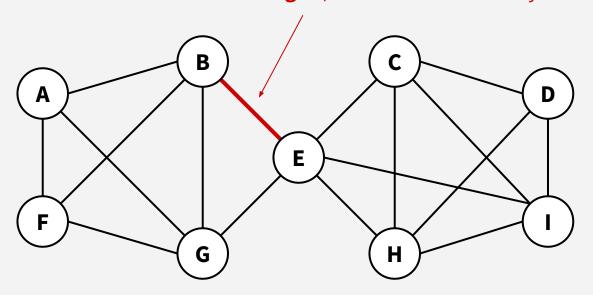
G

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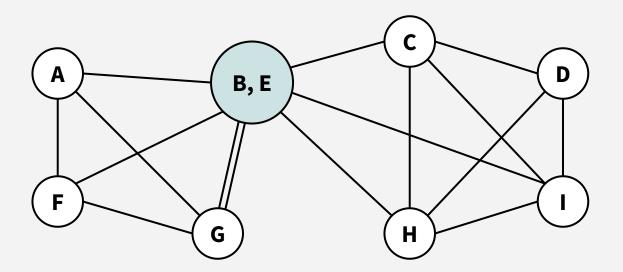


Let's see what happens if we're not so lucky all the time...

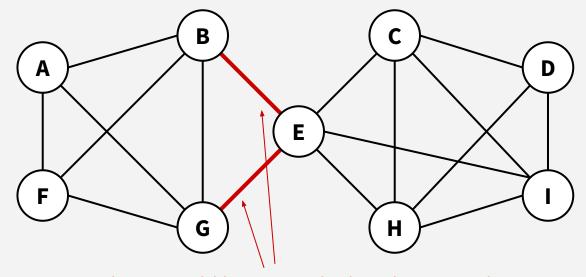
Pick a random edge! (this is a more "unlucky" choice)



After we do the "edge contraction", there's no way to return a cut that separates B and E.



In fact, if Karger's algorithm ever randomly selects **edges in the min-cut**, then it won't return that min-cut.



These would be very unlucky edges to pick...

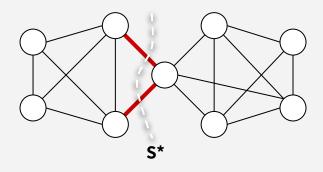
CLAIM:

The probability that Karger's algorithm returns a minimum cut is $\geq 1/\binom{n}{2}$

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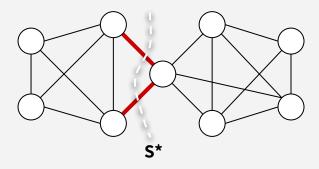


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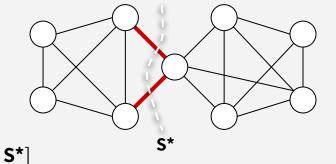
= Pr[\mathbf{e_1} doesn't cross S^*]

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...

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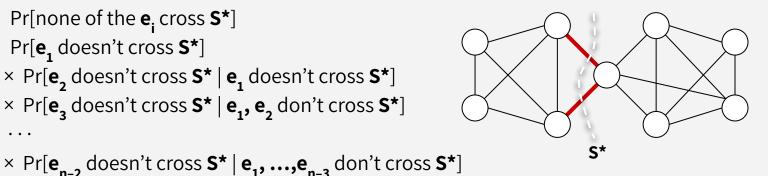


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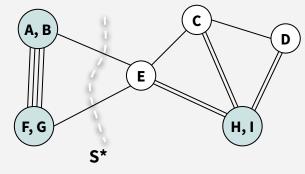
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Pr[return S*] = Pr[none of the e; cross S*]
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                  × Pr[e, doesn't cross S* | e, doesn't cross S*]
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```



$$\geq 1/\binom{n}{2}$$

We need to show this! We can find an expression for: $Pr[e_i doesn't cross S^* | e_1, ..., e_{i-1} don't cross S^*]$

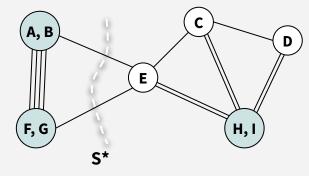
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This is equivalent to answering this:

Suppose after j-1 edge choices, Karger hasn't messed up yet! What's the probability that our j th edge, e_i , also doesn't cross S*?

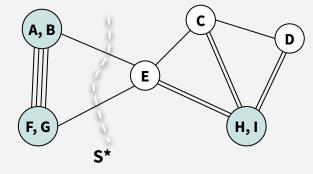


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Say there are k edges that cross S*.



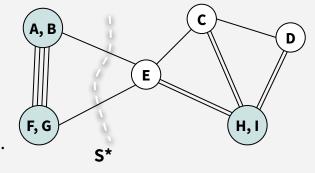
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All nodes must currently have at least k (original) edges coming out.



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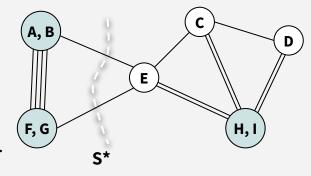
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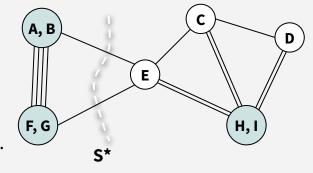
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Thus, there are at least k(n-j+1)/2 remaining edges to choose from!



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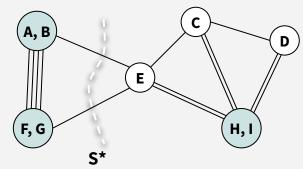
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Since the number of nodes decreases by 1 each time we choose an edge, there are n - (j - 1) = n - j + 1 nodes left, each with at least k edges.



PROOF (cont'd): What is $Pr[e_i doesn't cross S^* | e_1, ..., e_{i-1} don't cross S^*]$?

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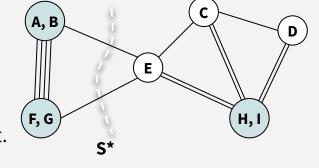
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So, the probability that our jth edge is also *safe* is:



We basically computed

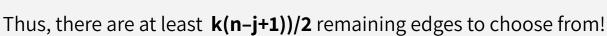
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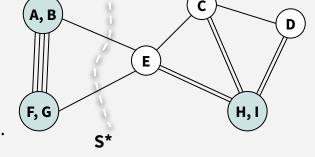


Since the number of nodes decreases by 1 each time we choose an edge, there

1 - Pr[lanure]

=1 - (k / total # remaining edges)

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So, the probability that our ith edge is also safe is:

$$\leq 1 - rac{k}{rac{k(n-j+1)}{2}}$$

$$1 - \frac{2}{n-j+1} = \frac{n-j-1}{n-j+1}$$

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PROOF (cont'd): Suppose S^* is a min-cut, and suppose we select edges $e_1, e_2, ..., e_{n-2}$. Then:

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From previous slide:

 $\Pr[\mathsf{e_{j}} \ \mathsf{doesn't} \ \mathsf{cross} \ \mathsf{S^{\star}} \ | \ \mathsf{e_{1},...,e_{j-1}} \ \mathsf{don't} \ \mathsf{cross} \ \mathsf{S^{\star}}]$

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• • •

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≥ \left(\frac{2}{n(n-1)}\right) = 1/\binom{n}{2}
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From previous slide:

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We proved the claim!

But 1/(n choose 2) doesn't seem like a very high probability.

For example, with our example of n = 9, Karger would return the $\{A,B,C,D\} - \{E,F,G,H,I\}$ min cut with probability 1/(9 choose 2) = 0.028...

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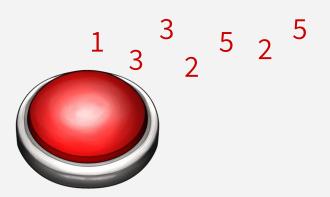
To boost our chances of success, let's employ some repetition!

$$\geq \left(\frac{2}{n(n-1)}\right) = 1/\binom{n}{2}$$

Suppose we have a magic button that produces one of 5 numbers {a,b,c,d,e}, uniformly at random when we push it. (We don't know what {a,b,c,d,e} are).

Q1: How many times do we need to push the button, in expectation, to see the minimum value?

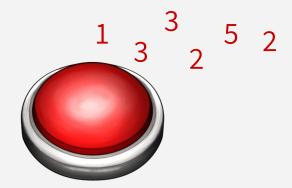
Q2: What is the probability that you have to push it more than 5 times? 10 times?



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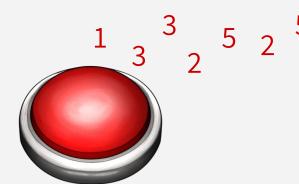


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 - E[# times we push until min] = 1/(0.2) = 5

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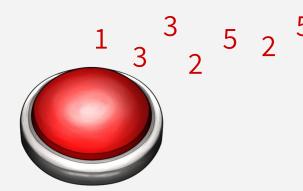
2. We've done this less frequently:

 $Pr[we don't see min after T times] = (1 - 0.2)^T$

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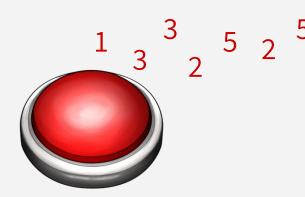
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Pr[we don't see min after T times] = $(1 - 0.2)^T$ Pr[we don't see min after 5 times] = $(1 - 0.2)^5 \approx 0.33$

Suppose we have a magic button that produces one of 5 numbers {a,b,c,d,e}, uniformly at random when we push it. (We don't know what {a,b,c,d,e} are).

Q1: How many times do we need to push the button, in expectation, to see the minimum value?

Q2: What is the probability that you have to push it more than 5 times? 10 times?



1. We've done this same calculation a bunch of times:

E[# times we push until min] = 1/(0.2) = 5

2. We've done this less frequently:

Pr[we don't see min after T times] = $(1 - 0.2)^T$ Pr[we don't see min after 5 times] = $(1 - 0.2)^5 \approx 0.33$ Pr[we don't see min after 10 times] = $(1 - 0.2)^{10} \approx 0.1$

To boost our chances of success, let's repeat Karger multiple times and return the smallest cut found!

- •Run Karger's! The cut size is 6!
- •Run Karger's! The cut size is 3!
- •Run Karger's! The cut size is 3!
- •Run Karger's! The cut size is 2!
- •Run Karger's! The cut size is 5!

 $(1 - 0.2)^T$

is the probability that our repetition fails, i.e. that the minimum value out of the T tries is *not* the true min cut value!

THE QUESTION:

How many times should we run Karger's to succeed with probability ≥ 1-δ?

Here, δ represents an upper bound on the probability of *failure* we're willing to tolerate (it's a parameter we can choose, e.g. it can be some small constant like 0.01)

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- First, we'll make a ~clever~ observation that 1-p ≤ e^{-p} (if you plot the lines you'll see why!)
- Thus, we have $(1-p)^T \le e^{-pT}$. We want $e^{-pT} \le \delta \implies$ we want $-pT \le \ln \delta \implies$ we want $T \ge (1/p)(-\ln \delta)$

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• Choose T to be at least $(1/n)\ln(1/\delta)$ i.e. run Karger's at least

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KARGER'S: PUTTING IT ALL TOGETHER

Runtime of Karger's (without fancy data structures): **O(n²)**

To successfully find a minimum cut with probability $\geq 1-\delta$, repeat Karger's at least $\binom{n}{2} \ln(\frac{1}{\delta})$ times.

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Treating δ as a constant (e.g. 0.01), the total runtime to succeed w/ prob. $\geq 1-\delta =$

$$O\left(n^2 \cdot \binom{n}{2} \ln\left(\frac{1}{\delta}\right)\right) = O(n^4)$$

If you want different success probabilities, this overall runtime would be adjusted as well (by adjusting the number of trials needed). Keep in mind that this can be improved by using a union-find data structure as well.

KARGER'S: PUTTING IT ALL TOGETHER

Runtime of Karger's (without fancy data structures): **O(n²)**

This is pretty good!

In practice, it's often good enough to use Monte-Carlo algorithms that'll succeed with probability 0.99. Repetition was the key!!!

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GENERAL TAKEAWAY

- If we have a Monte Carlo algorithm with a small success probability
- And we can check how good a solution is (e.g. cost of cut)
- Then we can **boost** our success probability via repetition! Repeat it a bunch of times and take the best solution

GENERAL TAKEAWAY

- If we have a Monte Carlo algorithm with a small success probability
- And we can check how good a solution is (e.g. cost of cut)
- Then we can **boost** our success probability via repetition! Repeat it a bunch of times and take the best solution

Unfortunately, repeating an O(n²) time algorithm many times can still be quite expensive...

CAN WE DO BETTER?

Absolutely!!!

The **Karger-Stein** algorithm can achieve a better runtime of **O(n²log²n)**.

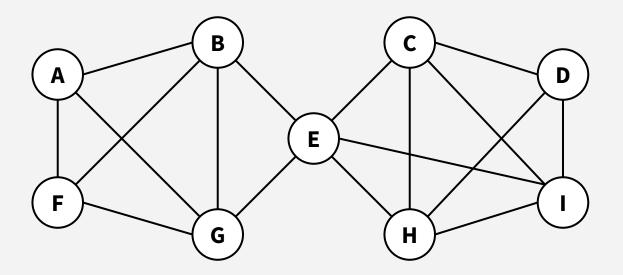
THE KEY: We'll do the repetitions in a clever way so that we save on runtime! Note: this is a trickier algorithm, so we'll just sketch the high-level approach here.



الگوريتم كارگر-استين

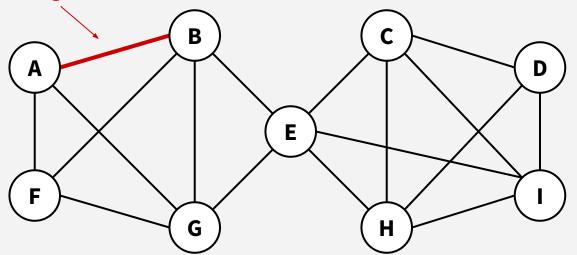
راه بهتری برای تکرار الگوریتم کارگِر

Let's brainstorm how we might save on repetitions...



Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

Pick a random edge!

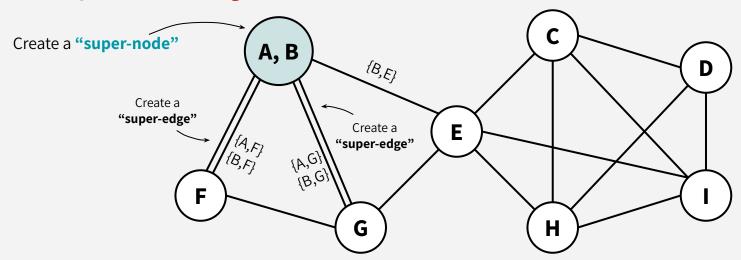


There are 17 edges, and 15 are good to contract

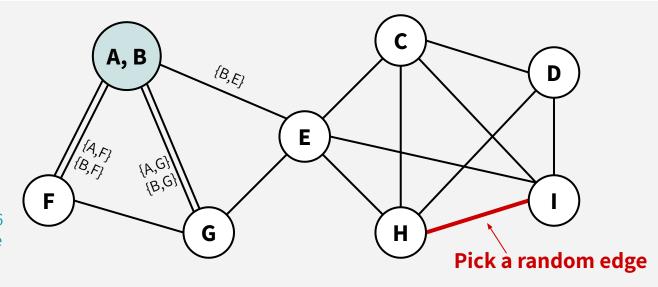
Probability that we didn't mess up = 15/17

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

Now perform an "edge contraction"!



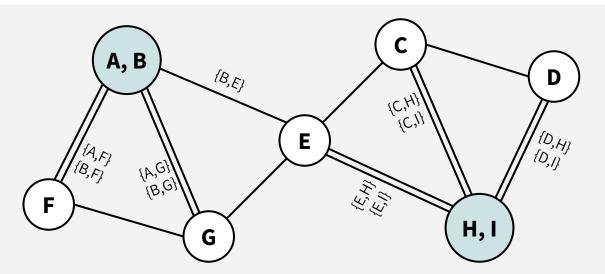
Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



Now there are only 16 edges, since the edge between A and B is gone.

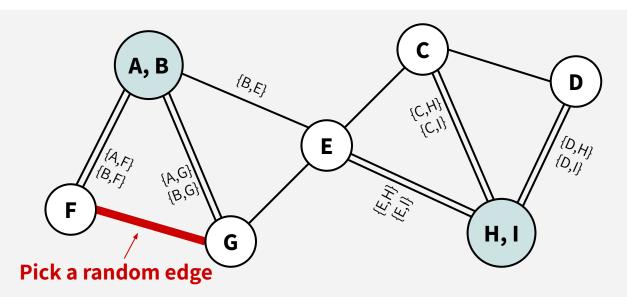
Probability that we didn't mess up = 14/16

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



Now perform an "edge contraction"!

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

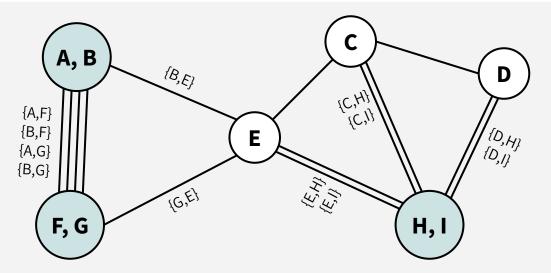


There are 15 edges, since the edge between H and I is gone now.

Again, only 2 edges are bad to pick still.

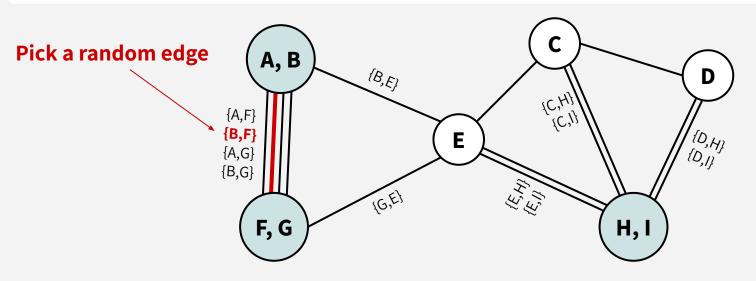
Probability that we didn't mess up = 13/15

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



Perform an "edge contraction"

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

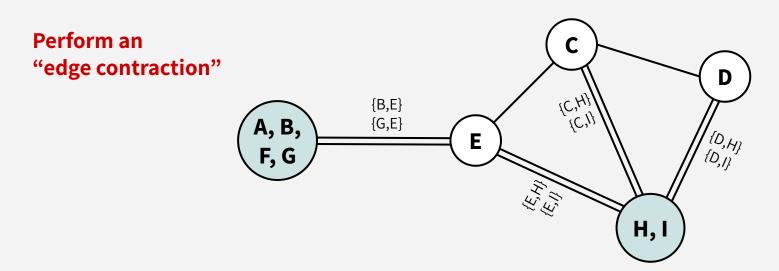


There are now 14 edges, since the edge between F and G is gone now.

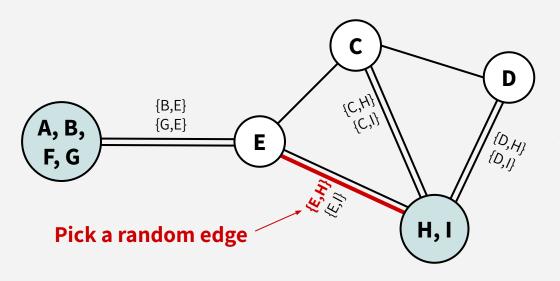
Again, only 2 edges are bad to pick still.

Probability that we didn't mess up = 12/14

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

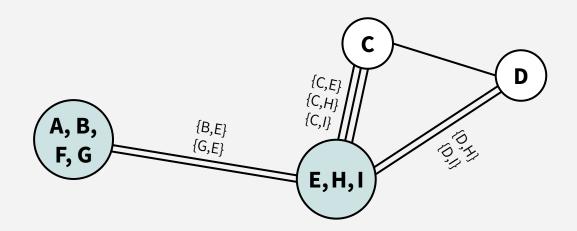


There are now only 10 edges, since the last contraction got rid of 4 edges!

Again, only 2 edges are bad to pick still.

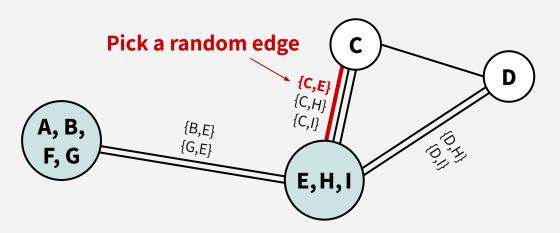
Probability that we didn't mess up = 8/10

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



Perform an "edge contraction"

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



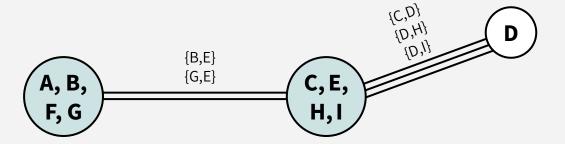
There are now only 8 edges, since the last contraction got rid of 2 edges!

Again, only 2 edges are bad to pick still.

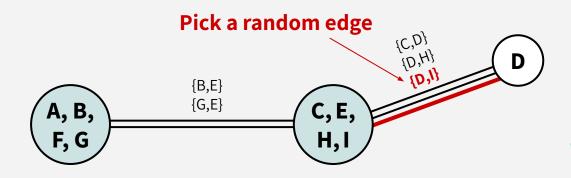
Probability that we didn't mess up = 6/8

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

Perform an "edge contraction"



Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



There are now only 5 edges, since the last contraction got rid of 3 edges!

Again, only 2 edges are bad to pick still.

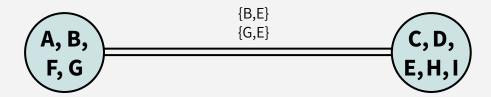
Probability that we didn't mess up = 3/5

Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.



Pick a random edge, **contract** it, and repeat until you only have 2 vertices left.

Here we stop because we only have 2 nodes left!



MORE LIKELY TO MESS UP OVER TIME!

At the beginning of a run of Karger, it's pretty likely our random choice is fine.

The probability that we mess up gets worse and worse over time.



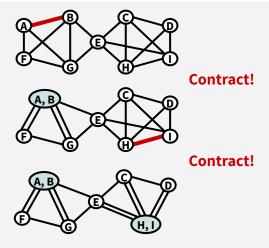
Moral of the walkthrough:

Repeating the beginning of the algorithm is wasteful!!! We weren't really that likely to mess up, so why bother protecting those choices with redundancy? (i.e. why bother using repetition for our earlier choices?)

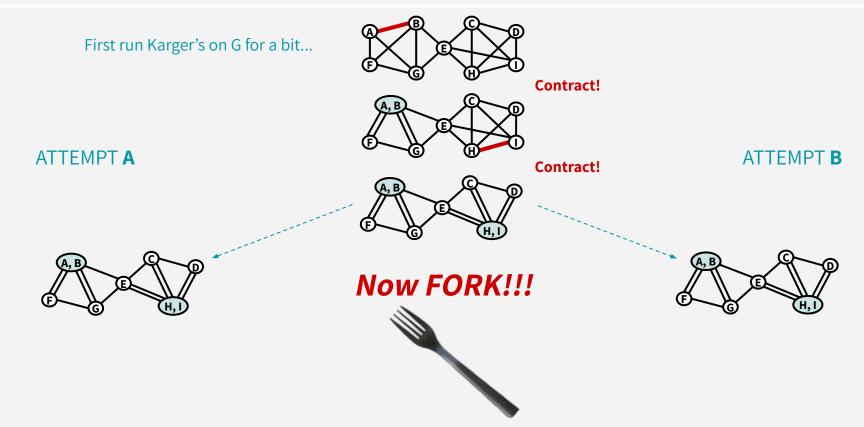


OUR GAMEPLAN (IN PICTURES)

First run Karger's on G for a bit...

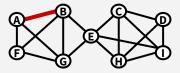


OUR GAMEPLAN (IN PICTURES)

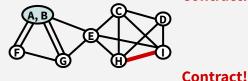


OUR GAMEPLAN (IN PICTURES)

First run Karger's on G for a bit...



Contract!

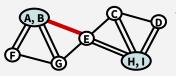


ATTEMPT B

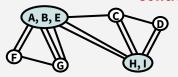
But it's okay since branch B made a good choice 😁

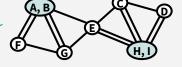
ATTEMPT A

This branch made a bad choice 😢



Contract!



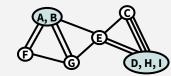


Now FORK!!!

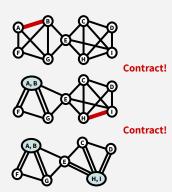




Contract!



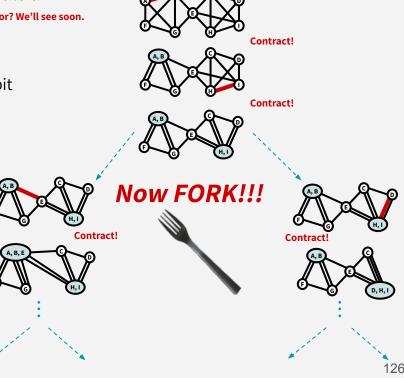
First run Karger's on G for a bit until there are $\frac{n}{\sqrt{2}}$ supernodes left. Why the $\sqrt{2}$ factor? We'll see soon.



First run Karger's on G for a bit until there are $\frac{n}{\sqrt{2}}$ supernodes left. Why the $\sqrt{2}$ factor? We'll see soon. Then split/fork into two independent copies, G₁ and G₂ Contract! Independently resume Karger's algorithm on G₁ and G₂ for a bit Contract! (until you reduce # of supernodes by another factor of $\sqrt{2}$) Now FORK!!! Contract!

- First run Karger's on G for a bit until there are $\frac{n}{\sqrt{2}}$ supernodes left.

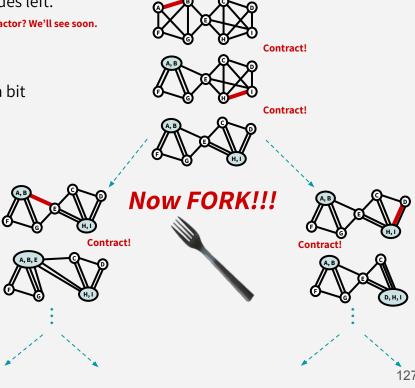
 Why the $\sqrt{2}$ factor? We'll see soon.
- Then split/fork into two independent copies, G₁ and G₂
- Independently resume Karger's algorithm on G_1 and G_2 for a bit (until you reduce # of supernodes by another factor of $\sqrt{2}$)
- Then split each of those into two independent copies....
- Etc....



- First run Karger's on G for a bit until there are $\frac{n}{\sqrt{2}}$ supernodes left. Why the $\sqrt{2}$ factor? We'll see soon.
- Then split/fork into two independent copies, G₁ and G₂
- Independently resume Karger's algorithm on G₁ and G₂ for a bit (until you reduce # of supernodes by another factor of $\sqrt{2}$)
- Then split each of those into two independent copies....
- Etc....

Basically:

Run Kargers for a bit, split into two copies, recurse on each, then return whichever copy's answer was better (i.e. the smaller cut)



KARGER-STEIN PSEUDOCODE

There are some details left out, but here's the skeleton:

KARGER-STEIN(G=(V,E)):

- n ← |V|
- if n < 4: find a min-cut by brute force ← O(1) time
- ullet Run Karger's algorithm on G until $\left\lfloor \frac{n}{\sqrt{2}} \right\rfloor$ supernodes remain
- $G_1 \& G_2 \leftarrow copies of what's left in G$
- $S_1 = KARGER-STEIN(G_1)$
- S₂ = KARGER-STEIN(G₂)
- return whichever of S₁, S₂ is the smaller cut

KARGER-STEIN (AS RECURSION TREE)

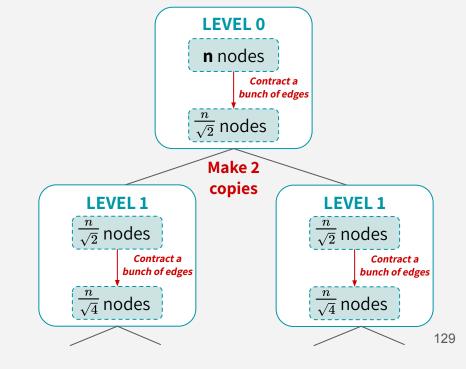
SHRINKING FACTOR: $\sqrt{2}$

DEPTH:
$$\log_{\sqrt{2}}(n) = \frac{\log n}{\log \sqrt{2}} = 2\log n$$

OF LEAVES: $2^{2\log n} = n^2$

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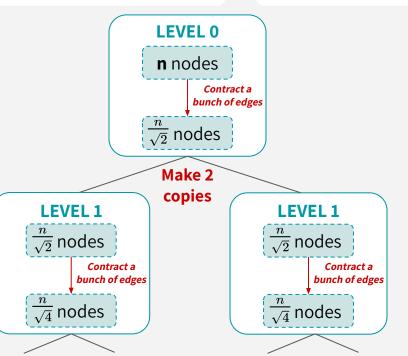


KARGER-STEIN RUNTIME

SHRINKING FACTOR: $\sqrt{2}$

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OF LEAVES: $2^{2\log n}=n^2$



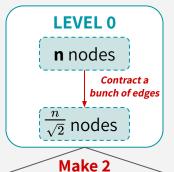
• Amount of work per level = amount of work needed to reduce the number of nodes by a factor of $\sqrt{2}$

KARGER-STEIN RUNTIME

SHRINKING FACTOR: $\sqrt{2}$

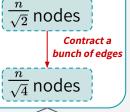
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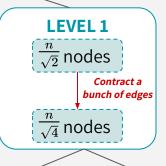
OF LEAVES: $2^{2\log n} = n^2$



That's at most O(n²) -- that's the time it takes to run
Karger's algorithm once in full!







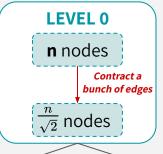
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KARGER-STEIN RUNTIME

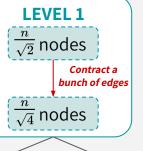
SHRINKING FACTOR: $\sqrt{2}$

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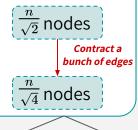
OF LEAVES: $2^{2\log n} = n^2$



Make 2 copies



LEVEL 1



- Amount of work per level = amount of work needed to reduce the number of nodes by a factor of $\sqrt{2}$
- That's at most O(n²) -- that's the time it takes to run
 Karger's algorithm once in full!
- Our recurrence relation is

$$T(n)=2Tig(rac{n}{\sqrt{2}}ig)+O(n^2)$$

The Master Theorem says:

$$T(n) = O(n^2 \log n)$$

Suppose we contract $n-\frac{n}{\sqrt{2}}$ edges. CLAIM:

Pr[None of those $n - \frac{n}{\sqrt{2}}$ edges cross S*] $\approx 1/2$ (when n is large)

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Pr[None of those n-t edges cross **S***]

- = Pr[**e**₁ doesn't cross **S***]
 - × Pr[e, doesn't cross S* | e, doesn't cross S*]
 - \times Pr[$\mathbf{e_3}$ doesn't cross $\mathbf{S^*} \mid \mathbf{e_1}, \mathbf{e_2}$ don't cross $\mathbf{S^*}$]

• •

 \times Pr[$\mathbf{e}_{\mathbf{n-t}}$ doesn't cross $\mathbf{S}^* \mid \mathbf{e}_{\mathbf{1}}, ..., \mathbf{e}_{\mathbf{n-t-1}}$ don't cross \mathbf{S}^*]

From earlier:

 $Pr[e_i doesn't cross S^* | e_1,...,e_{i-1} don't cross S^*]$

$$= \frac{n-j-1}{n-j+1}$$

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 \times Pr[$\mathbf{e}_{\mathbf{n}-\mathbf{t}}$ doesn't cross $\mathbf{S}^* \mid \mathbf{e}_{\mathbf{1}}, ..., \mathbf{e}_{\mathbf{n}-\mathbf{t}-\mathbf{1}}$ don't cross \mathbf{S}^*]

$$\geq \left(\frac{n-2}{n}\right) \left(\frac{n-3}{n-1}\right) \left(\frac{n-4}{n-2}\right) \left(\frac{n-5}{n-3}\right) \dots \left(\frac{t+1}{t+3}\right) \left(\frac{t}{t+2}\right) \left(\frac{t-1}{t+1}\right)$$

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$$= \frac{t \cdot (t-1)}{n \cdot (n-1)}$$

Suppose we contract $n-\frac{n}{\sqrt{2}}$ edges. CLAIM:

Pr[None of those $n - \frac{n}{\sqrt{2}}$ edges cross S*] $\approx 1/2$ (when n is large)

Suppose S^* is a min-cut, and suppose we select edges $e_1, e_2, ..., e_{n-t}$. Then:

Pr[None of those n-t edges cross **S***]

$$\geq \left(\frac{n-2}{n}\right) \left(\frac{n-3}{n-1}\right) \left(\frac{n-4}{n-2}\right) \left(\frac{n-5}{n-3}\right) \dots \left(\frac{t+1}{t+3}\right) \left(\frac{t}{t+2}\right) \left(\frac{t-1}{t+1}\right)$$

$$= \frac{t \cdot (t-1)}{n \cdot (n-1)} \qquad \text{(now choose } t = \frac{n}{\sqrt{2}} \text{)}$$

$$= \frac{\frac{n}{\sqrt{2}} \cdot \left(\frac{n}{\sqrt{2}} - 1\right)}{n \cdot (n-1)} \qquad \approx \qquad \frac{1}{2} \qquad \text{(when n is large)}$$

Suppose we contract $n-\frac{n}{\sqrt{2}}$ edges. CLAIM:

Pr[None of those $n - \frac{n}{\sqrt{2}}$ edges cross S*] $\approx 1/2$ (when n is large)

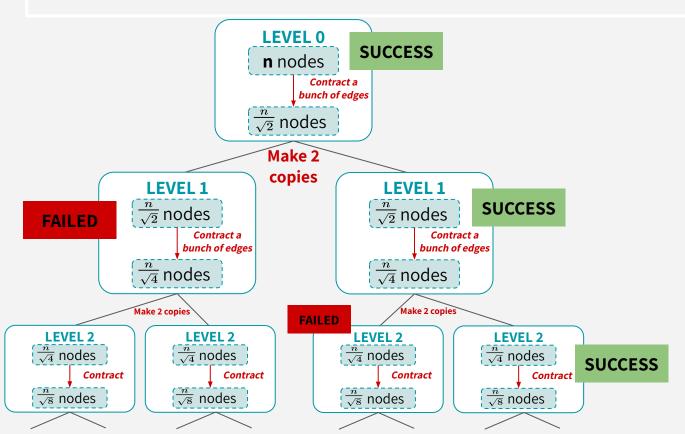
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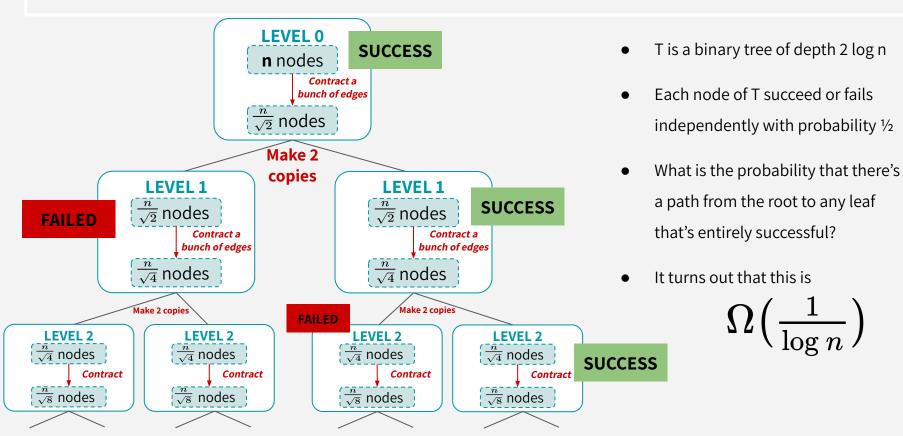
This means that in every "blob" of running Karger for a bit, the probability of success is ½!

$$= \frac{\frac{n}{\sqrt{2}} \cdot (\frac{n}{\sqrt{2}} - 1)}{n \cdot (n - 1)} \approx \frac{1}{2}$$
 (when n is large)

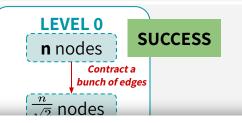
KARGER-STEIN PROB. OF SUCCESS



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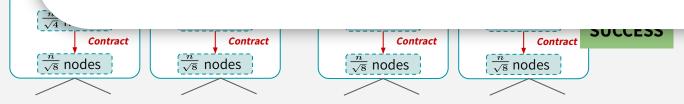
KARGER-STEIN PROB. OF SUCCESS



- T is a binary tree of depth 2 log n
- Each node of T succeed or fails independently with probability 1/2

The probability of success for one run of KARGER-STEIN is

$$\Omega\left(\frac{1}{\log n}\right)$$



ALTOGETHER NOW

- ullet The probability of success for one run of KARGER-STEIN is $\ \Omega\Big(rac{1}{\log n}\Big)$
- We can amplify the success probability with repetition
 - \circ Run Karger-Stein $O((\log n) \cdot (\log \frac{1}{\delta}))$ times to achieve success probability 1 δ
- Each iteration takes time O(n² log n)
 - o That's what we proved before.
- Choosing delta = 0.01 → the total runtime is

$$O(n^2 \log n \cdot \log n) = O(n^2 \log^2 n)$$

