

Approximation Algorithms

Admin

- Assignment 9 is due on Wednesday!
 - Questions on randomized algorithms and hashing
 - Last HW you will turn in
- Deferring lecture on randomized data structures: skip lists, cuckoo hash
- Will start approximation algorithm instead so you have some examples before you attempt HW 9 problems
- CS major advising: "drop in" hours this week (no reserved slots)
 - Can go to mine or any prof's hours and discuss plans
 - Important to fill out worksheet and submit transcript before the meeting

Approximate TSP

Traveling Salesman Problem

- Recall the traveling salesman problem: Given n cities labeled v_1, \dots, v_n , and distance function $d(i, j)$, the distance from city v_i to city v_j
- **TSP. (Decision Version)** Given target D , is there a tour that visits every city and returns to the starting city with total length at most D ?
- **NP complete problem.** Recall reduction from Hamiltonian cycle.
- Given directed graph $G = (V, E)$, define instance of TSP as:
 - City c_i for each node v_i
 - $d(c_i, c_j) = 1$ if $(v_i, v_j) \in E$
 - $d(c_i, c_j) = 2$ if $(v_i, v_j) \notin E$

Approximation Algorithms

- Cannot optimally solve NP hard problems like TSP efficiently
- **Goal of approximation algorithms:** design efficient algorithms that return solutions "close" to the optimal
- **(Optimization version).** Find the tour of every city with min total distance
- **(Approximation algorithm for TSP).**
 - Let OPT be the TSP tour of minimum total length, then a c -approximation algorithm finds a TSP tour of total length that is at most $c \cdot \text{OPT}$, for some constant $c > 1$
 - Example: a 2-approximation for TSP always returns a tour whose cost (total distance) is never more than twice that of the optimal solution

Bad News: Approx-TSP is hard

- **Claim.** There is no polynomial-time c -approximation algorithm for the general TSP problem, for any constant $c \geq 1$, unless $\mathbf{P} = \mathbf{NP}$.
- **Proof.** Suppose there is a poly-time c -approximation algorithm A that computes a TSP tour of total weight at most $c \cdot \text{OPT}$
- Show that A can be used to solve the Hamiltonian cycle problem
- Modified reduction from Hamiltonian cycle instance G to TSP instance:
 - $d(c_i, c_j) = 1$ if $(v_i, v_j) \in E$ and $d(c_i, c_j) = cn + 1$ if $(v_i, v_j) \notin E$
- If G has a Hamiltonian cycle: there is a tour of length exactly n
- If G does not have a Hamiltonian cycle, any tour has length at least $cn + 1$

Bad News: Approx-TSP is hard

- **Claim.** There is no polynomial-time c -approximation algorithm for the general TSP problem, for any constant $c \geq 1$, unless $P = NP$.
- **Proof. (Cont).** If G has a Hamiltonian cycle: there is a tour of length exactly n
- If G does not have a Hamiltonian cycle, any tour has length at least $cn + 1$
- A computes tour of length at most $cn \iff G$ has a Hamiltonian cycle:
 - A solves Hamiltonian cycle in polynomial time and $P = NP$
- **[More Bad news]**
For any function $f(n)$ that can be computed in polynomial time in n , there is no polynomial-time $f(n)$ -approximation for TSP on general weighted graphs, unless $P = NP$.

Good News: Metric TSP is Not

- While approximating TSP on general distances is NP hard, the common special case can be approximate easily
- **Metric TSP.** TSP problem on metric distances, that is, d satisfies:
 - $d(i, j) \leq d(i, k) + d(k, j)$ for any cities i, j, k [Triangle inequality]
 - $d(i, i) = 0$ and $d(i, j) \geq 0$ [Identity and Non-negative]
 - $d(i, j) = d(j, i)$ [Symmetric]
- Euclidean distances are an example of metric distances
- **Metric TSP** is still NP complete (reduction from undirected Ham cycle)
 - Setting $d(c_i, c_j) = 2$ when $(v_i, v_j) \notin E$ satisfies triangle inequality

Approximating Metric TSP

Approximating an NP Hard Problem

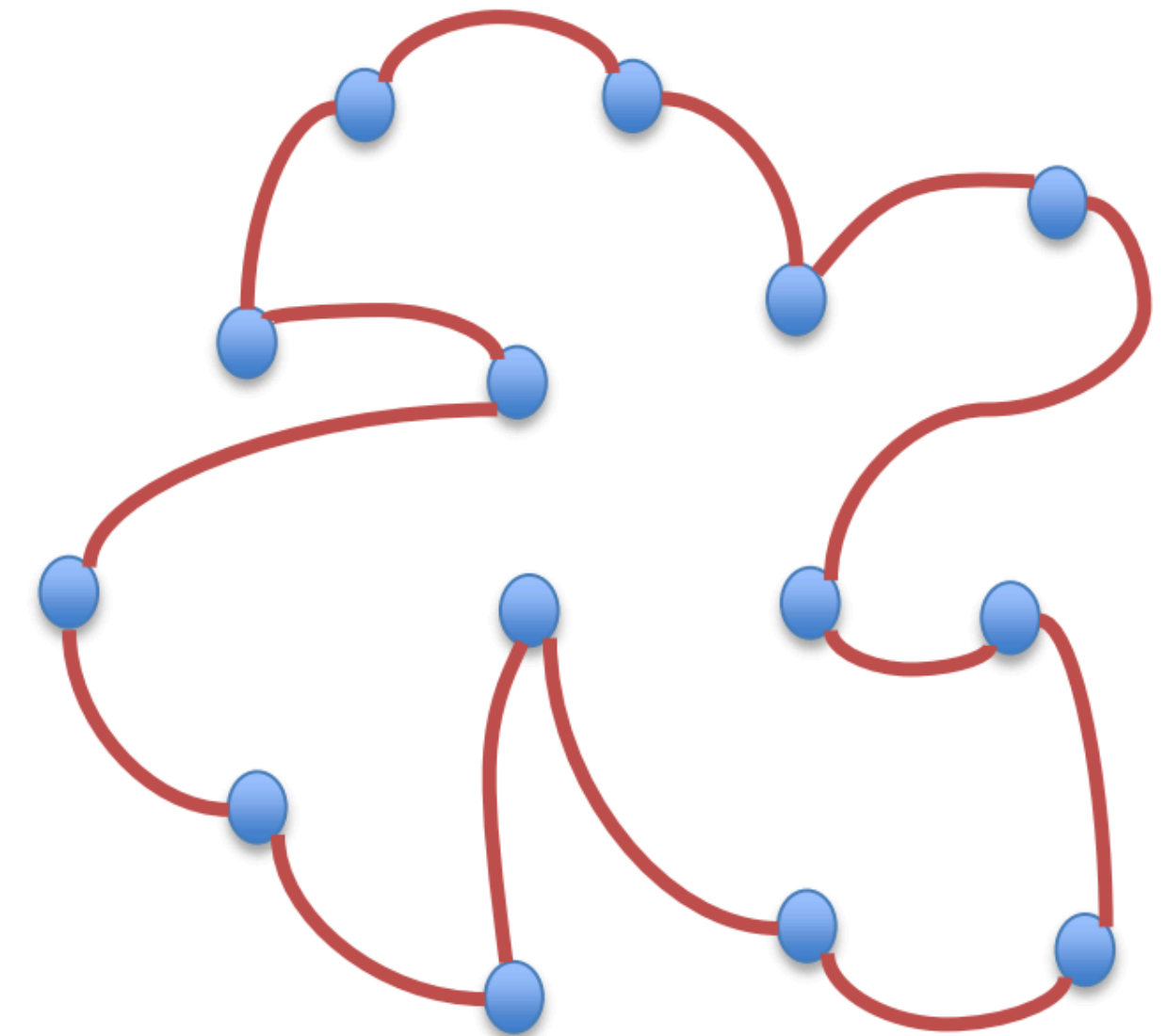
- (Metric TSP) Consider the **weighted complete graph** G where each vertex is a city, and each edge (i,j) for $i,j \in V$ has weight equal to the distance $d(i,j)$, where d satisfies the triangle inequality
- **(NP hard problem)**. Find the tour of every city with min total distance
- **(Goal: c -approximation)**. Design an algorithm that finds a TSP tour of total length that is at most $c \cdot \text{OPT}$, for some constant $c > 1$
- Remember, we don't actually know what the optimal algorithm or OPT actually is: we need to approximate without knowing that
- We do this by relating (upper and lower bounding) the cost of the approximation algorithm and OPT via a carefully chosen function

Approximating Metric TSP

- Consider the weighted complete graph G where each vertex is a city, and each edge (i,j) for $i,j \in V$ has weight equal to the distance $d(i,j)$, where d satisfies the triangle inequality
- **(Optimization version).** Find the tour of every city with min total distance
- Steps to follow when designing an approximation algorithm for a minimization problem
 - **Lower bound the optimal cost** by some function of input
 - Upper bound the cost of algorithm by **the same function**
- Will use MSTs to derive these upper and lower bounds for metric TSP
- We give a 2-approximation to metric-TSP using minimum spanning trees

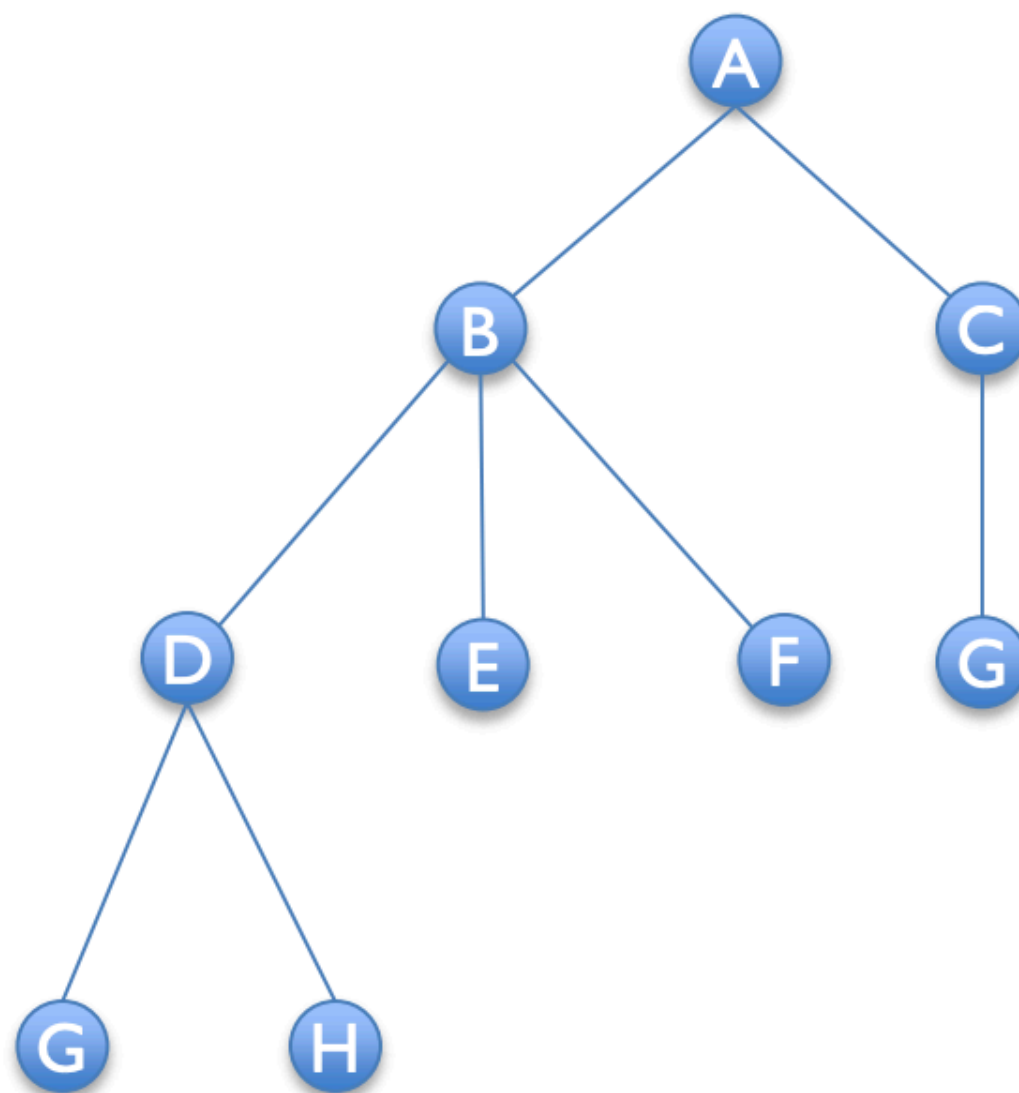
Lower Bound on OPT

- Note that an optimal tour must not visit a city more than once:
 - Optimal tour is a simple cycle that visits all nodes
- **Claim.** Let T be the minimum spanning tree of G then length of the optimal tour $\text{OPT} \geq w(T)$.
- **Proof.** Take an optimal tour of length OPT
- Drop an edge from it to obtain a spanning tree T'
- Distances/weights are non-negative, so $w(T') \leq \text{OPT}$
- $w(T) \leq w(T')$ (T is the MST)
- Thus $w(T) \leq \text{OPT}$ ■



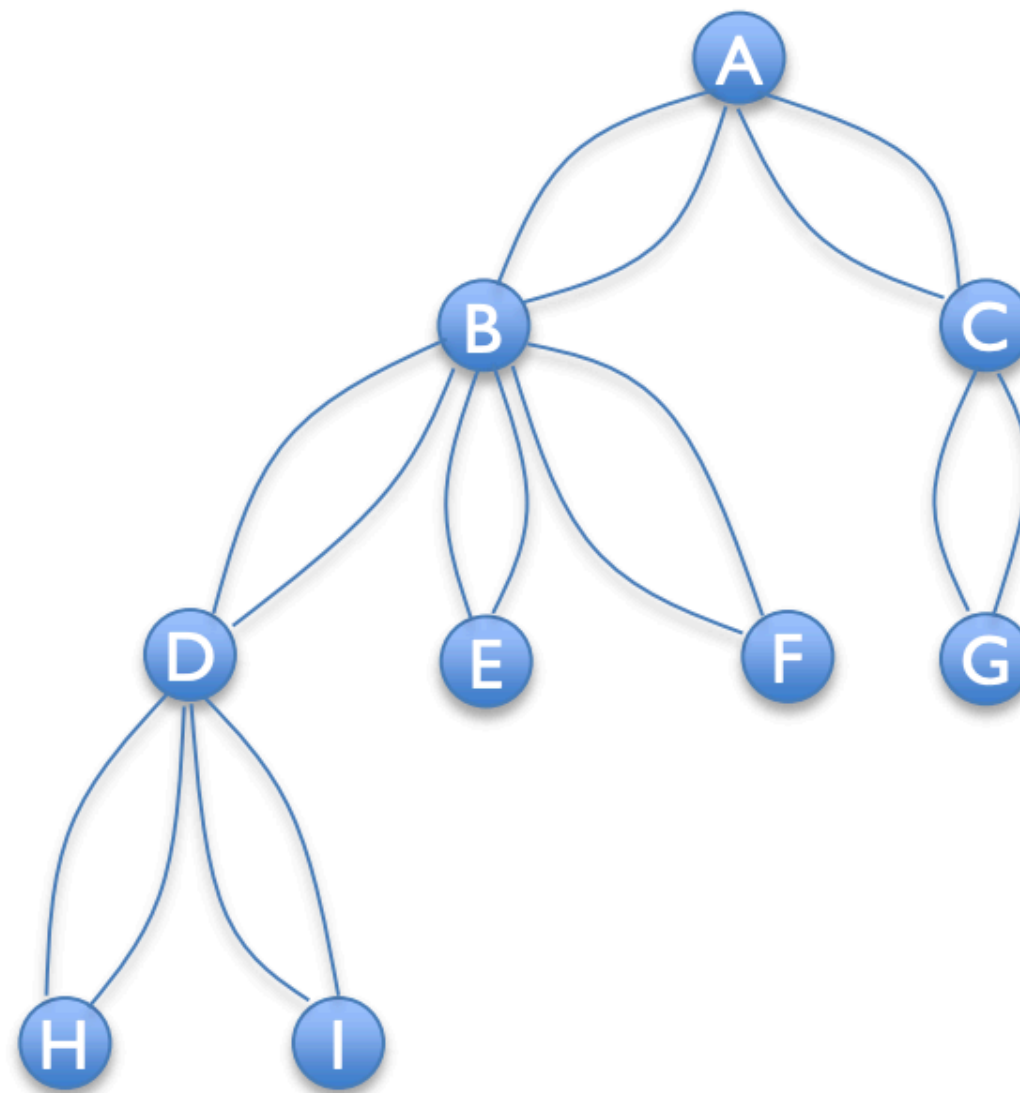
Double Tree Algorithm

- **Find a minimum spanning tree T**
- Duplicate every edge in T
- Find an Eulerian tour of resulting multi-graph
- Shortcut Euler tour to avoid repeated vertices



Double Tree Algorithm

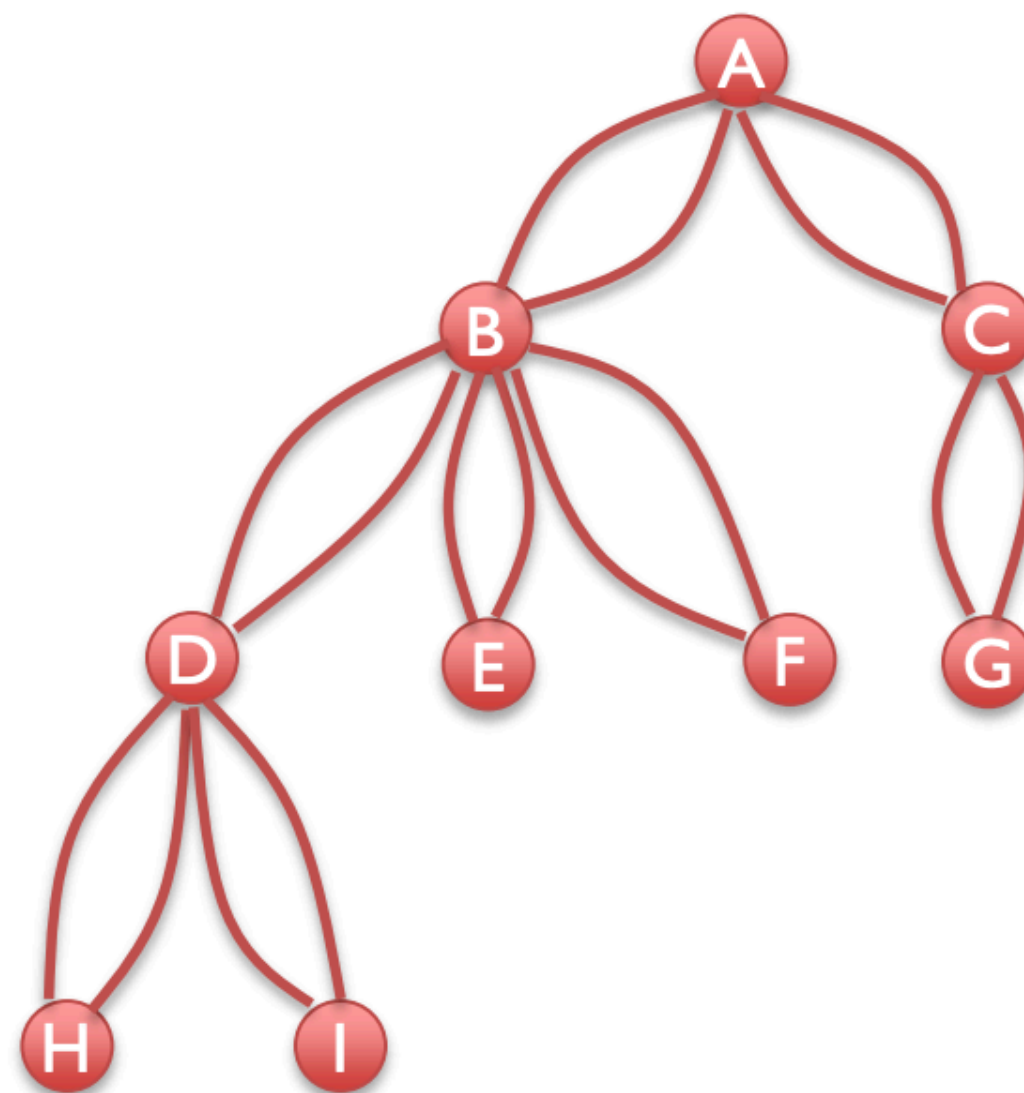
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Why must an Euler tour exist?

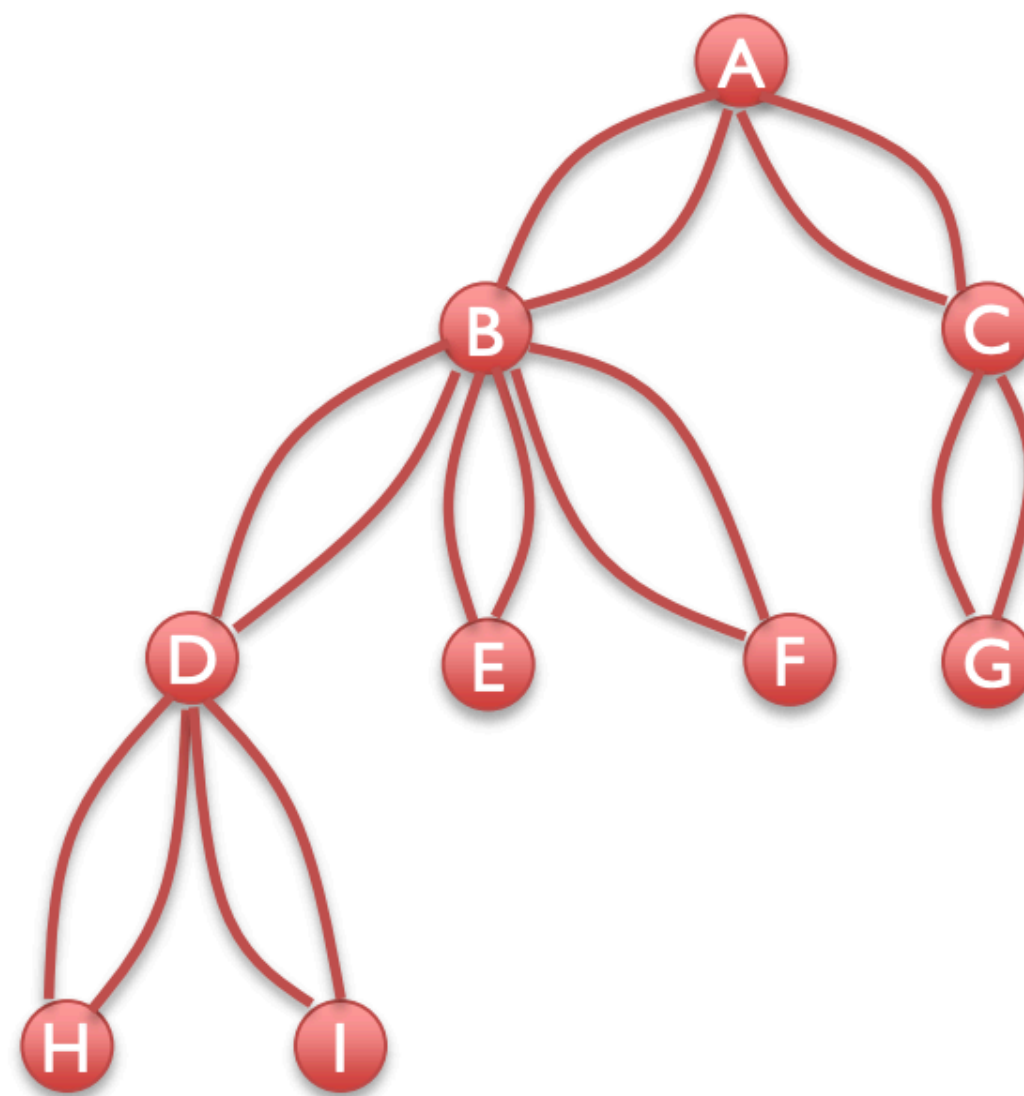


A,B,D,H,D,I,D,B,E,B,F,B,A,C,G,C,A

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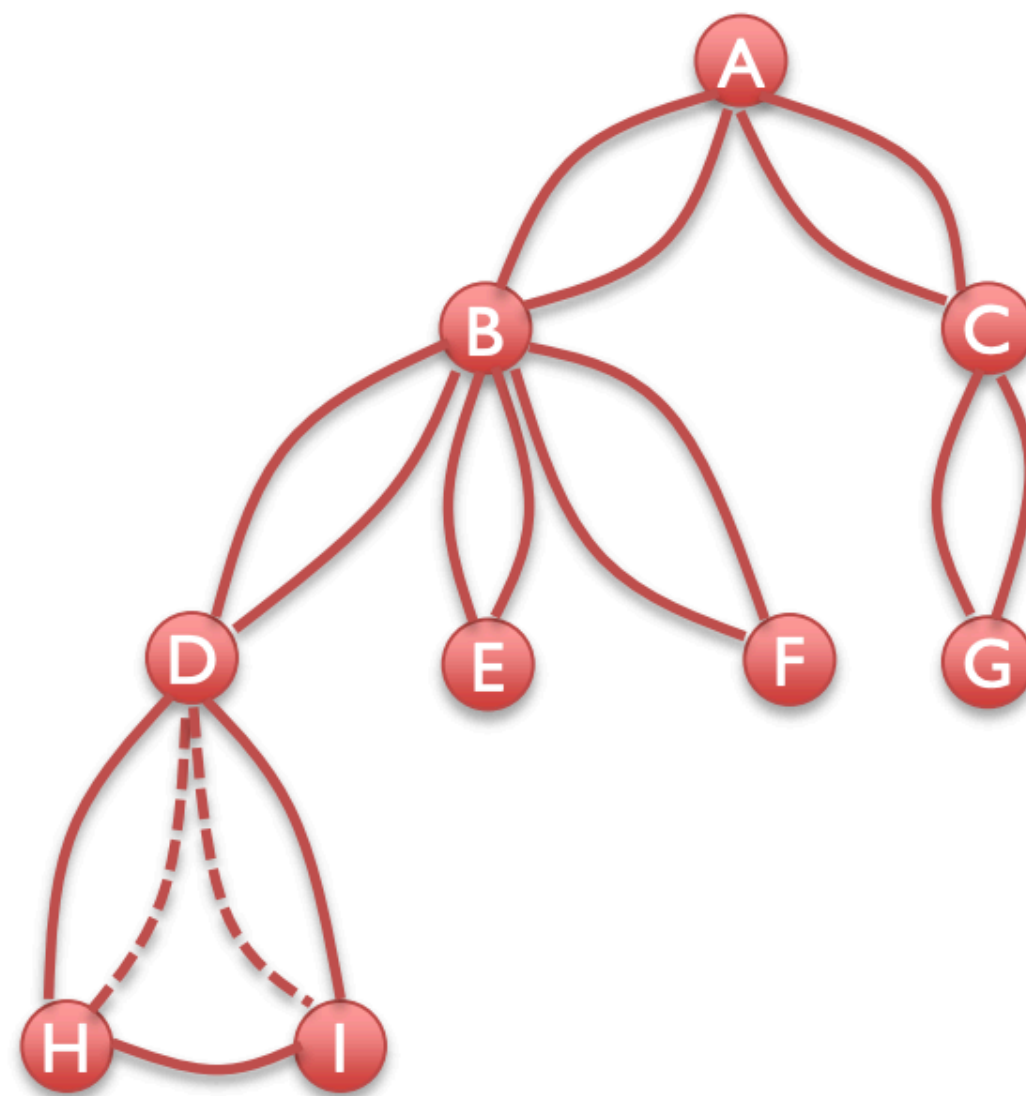
A graph has an Euler tour iff all nodes have even degree



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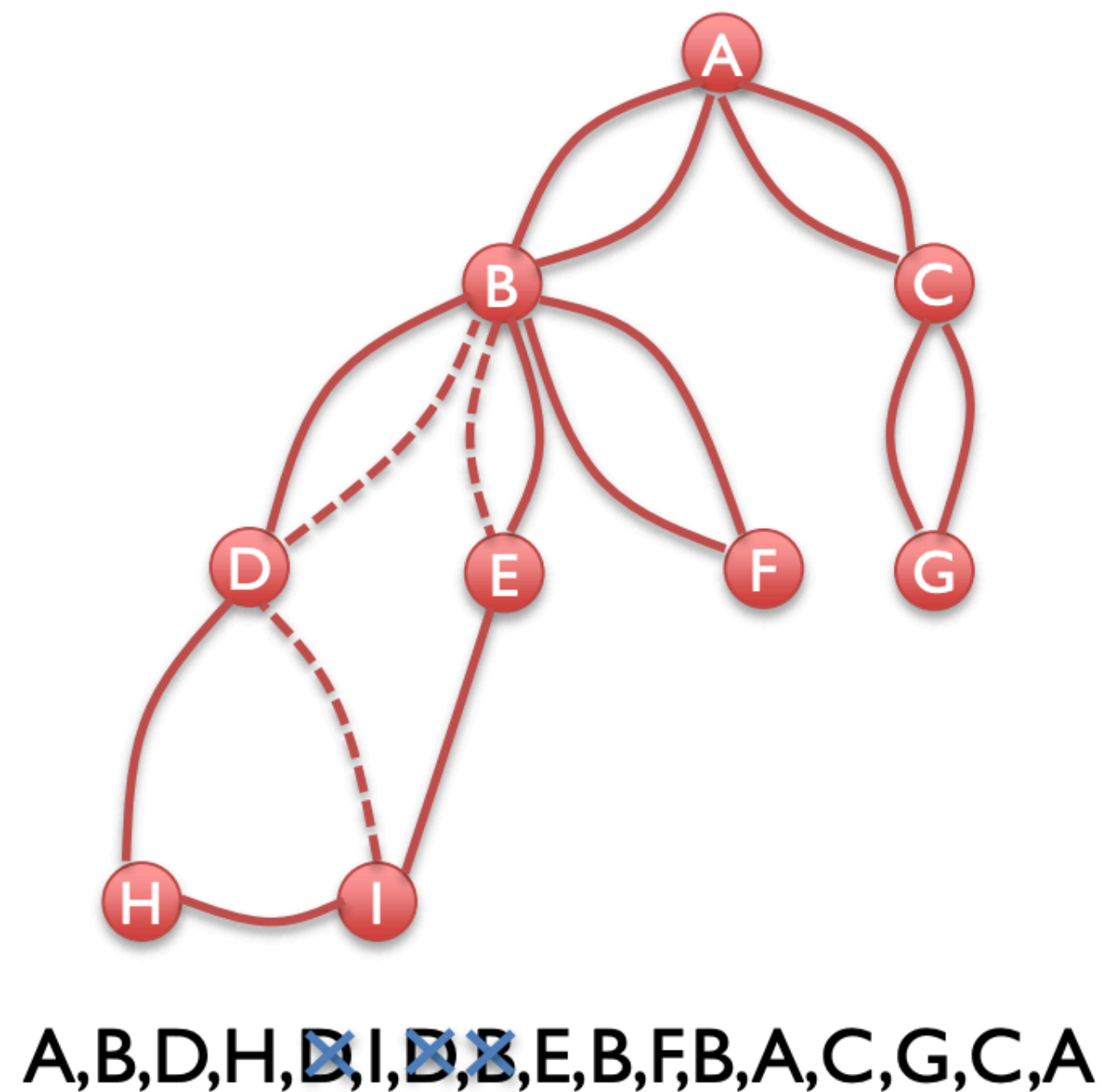
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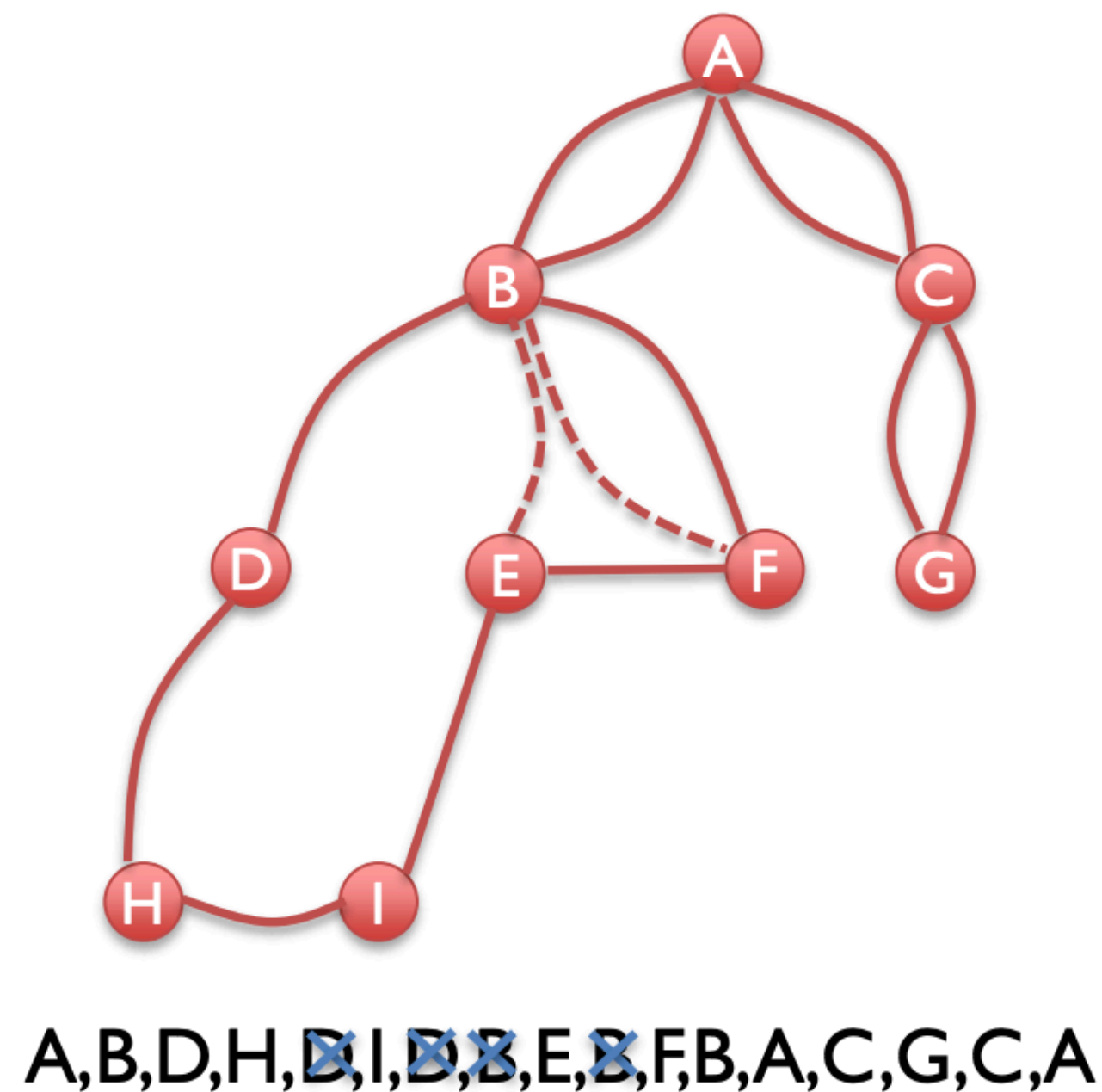
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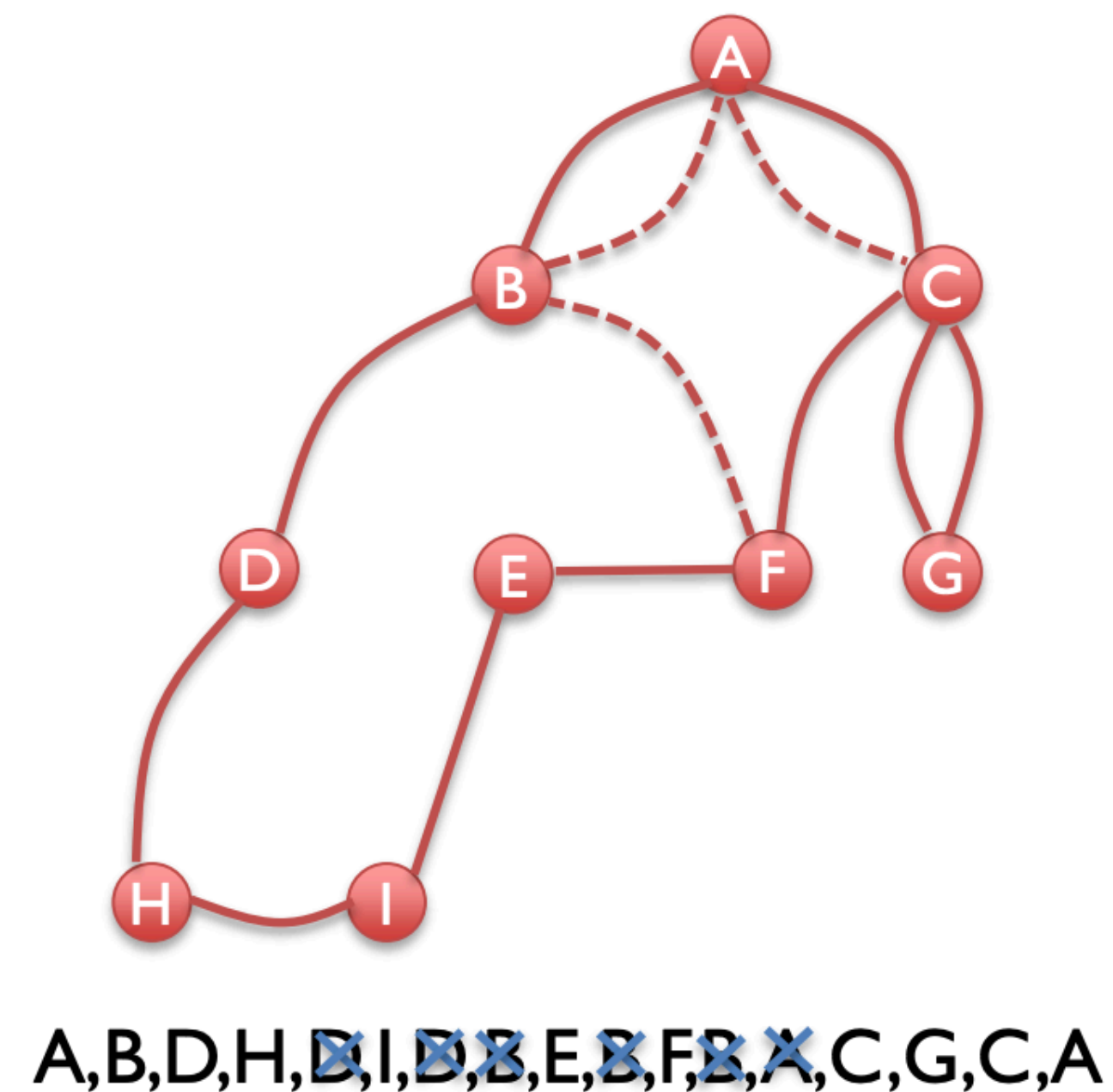
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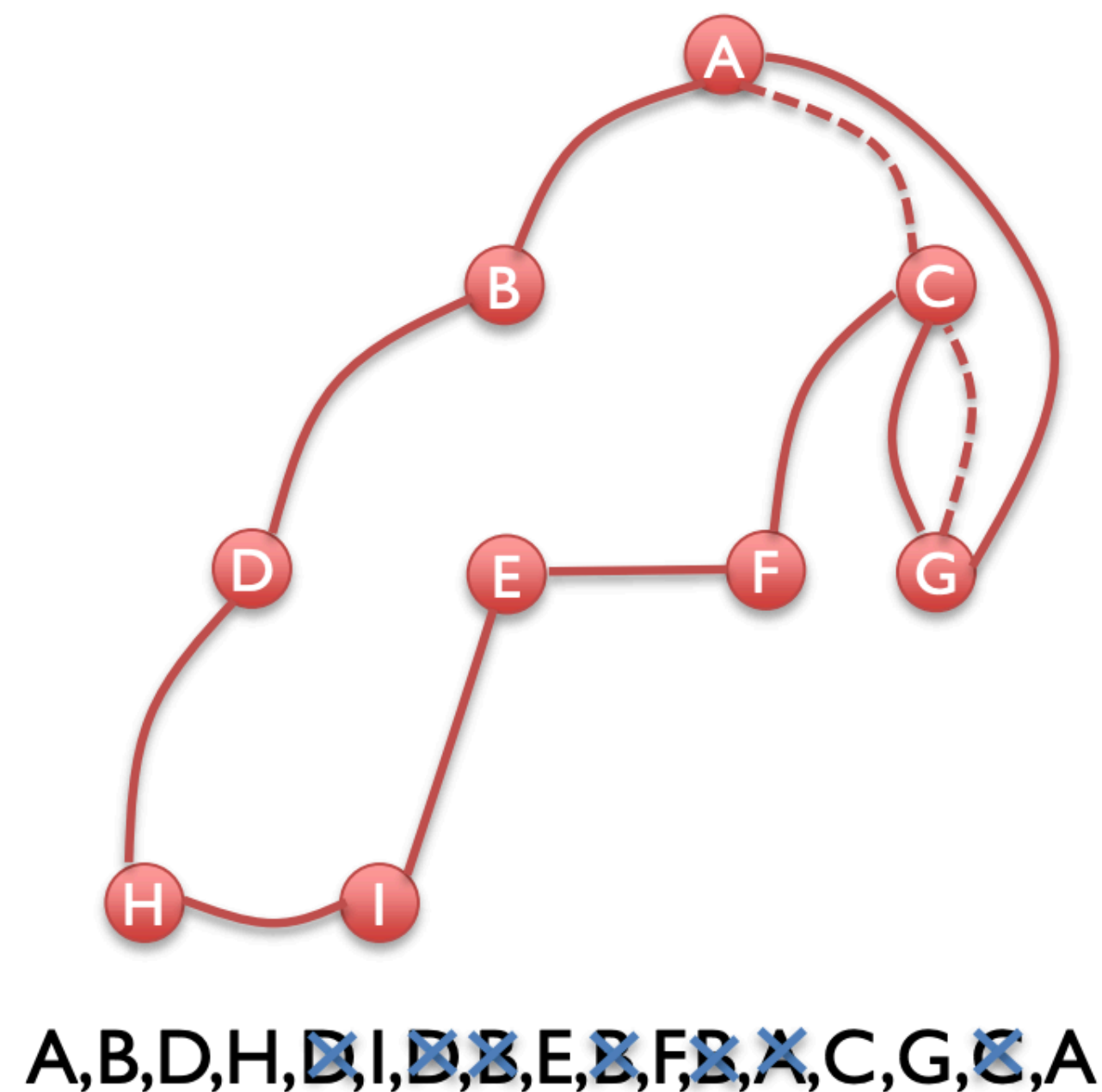
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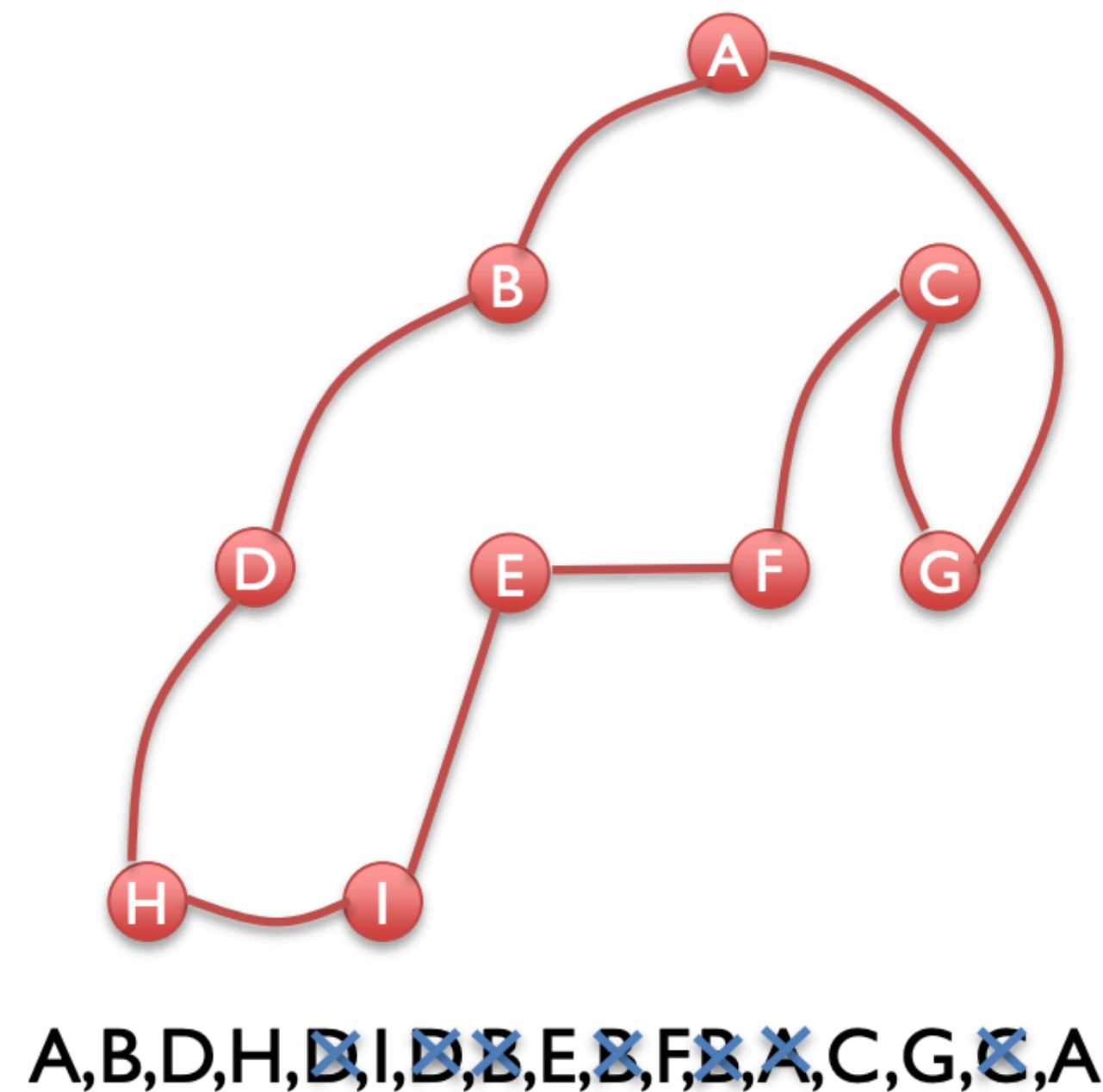
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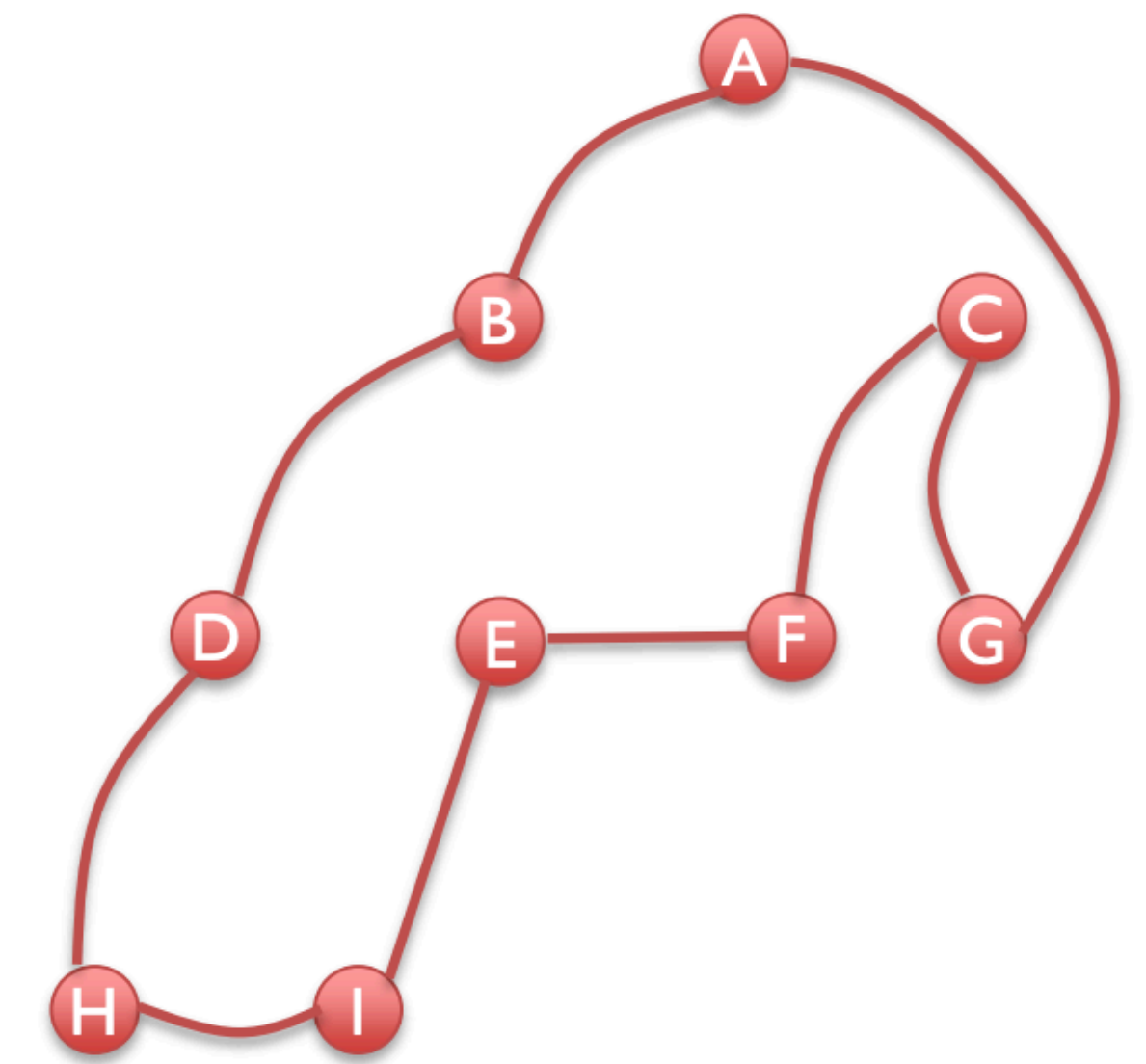
Double Tree Algorithm

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Double Tree Analysis

- **Claim.** Double-tree algorithm is a 2-approximation to TSP.
- **Proof.** The Euler tour visits every edge of MST T exactly twice, thus the length of tour $= 2 \cdot w(T)$
- Due to triangle inequality, shortcutting the tour does not increase length
 - (Here is where metric distances play a role!)
- Thus length of short-circuited tour $\leq 2 \cdot w(T)$
- Since $w(T) \leq \text{OPT}$, we get that our tour length is $\leq 2 \cdot \text{OPT}$ ■



A,B,D,H,~~I~~,~~D~~,~~B~~,E,~~F~~,~~A~~,C,G,~~C~~,A

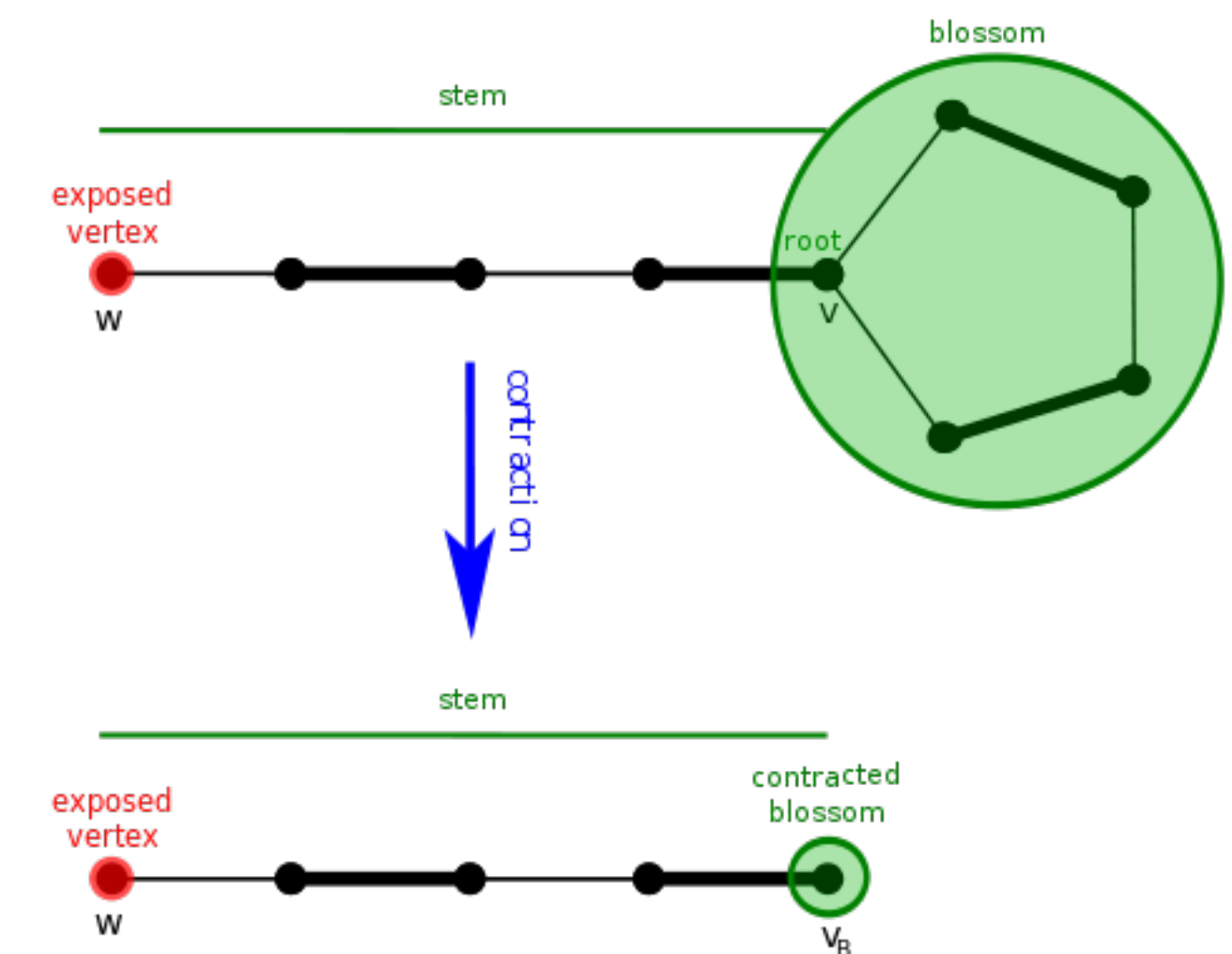
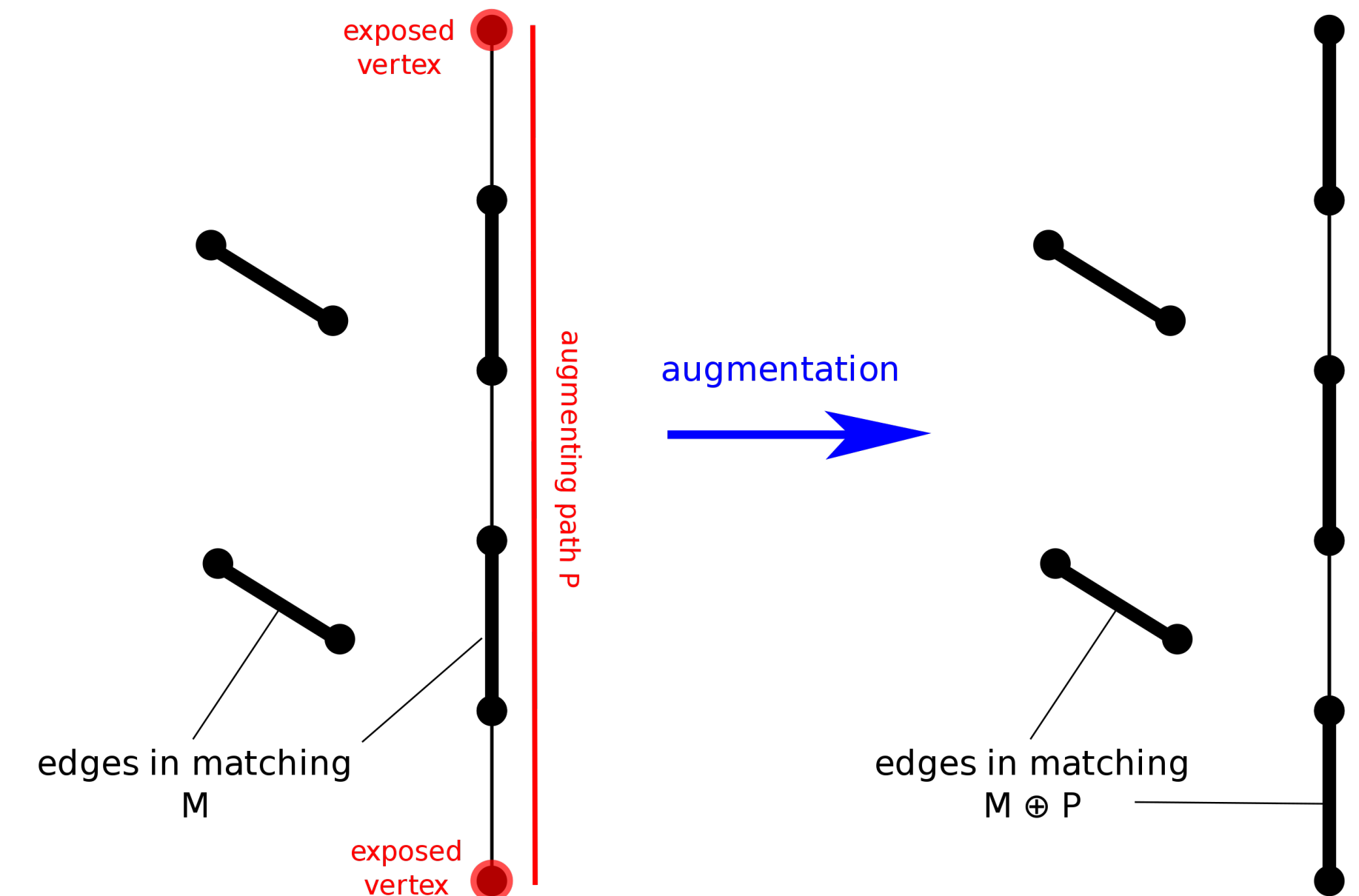
Christofides-Serdyukov Algorithm

Christofides Algorithm [Christofides 76][Serdyukov 76]

- Doubling the edges of MST is one way to achieve even degree nodes, but is there a cheaper way to augment the tree to obtain an Eulerian tour?
- What is the parity of odd degree vertices in an undirected graph?
 - Even number of odd degree vertices!
- **Christofides algorithm.** Starts with an MST, but fixes the parity of odd degree vertices by augmenting it with a matching
- **Matching.** A set of edges such that no two are adjacent
- **Perfect matching.** Every vertex is incident to exactly one edge in the matching
- **Fact We'll Use.** Minimum cost “perfect” matchings of any graph can be computed in polynomial time.

Minimum Cost Matching

- Won't see in this class, unfortunately
- Edmond's "blossom" algorithm
- $O(|E| |V|^2)$ (slow, but much better than exponential)
- Somewhat similar to Ford-Fulkerson:
 - Use special structure to prove that we just need to find augmenting paths
 - Use data structures so that we can find augmenting paths quickly
- Tricky part: "augmenting paths" are more complicated when finding a matching

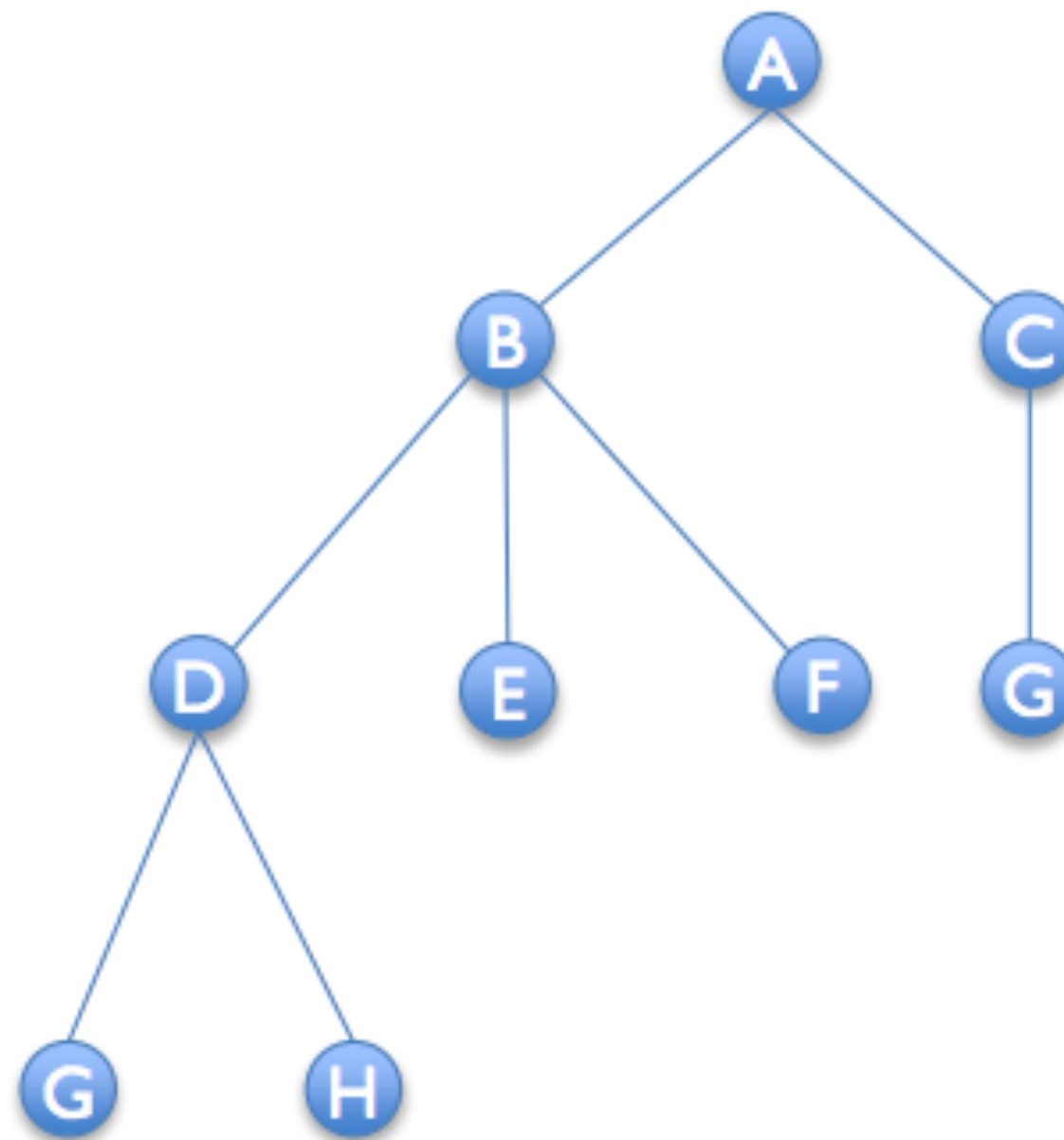


Christofides Algorithm

- Find the minimum spanning tree T
- Compute O : the set of odd degree vertices in T
- Find the min-cost perfect matching M of subgraph induced by O
- Return shortcut of Euler tour of $T \cup M$

$|O|$ must be even

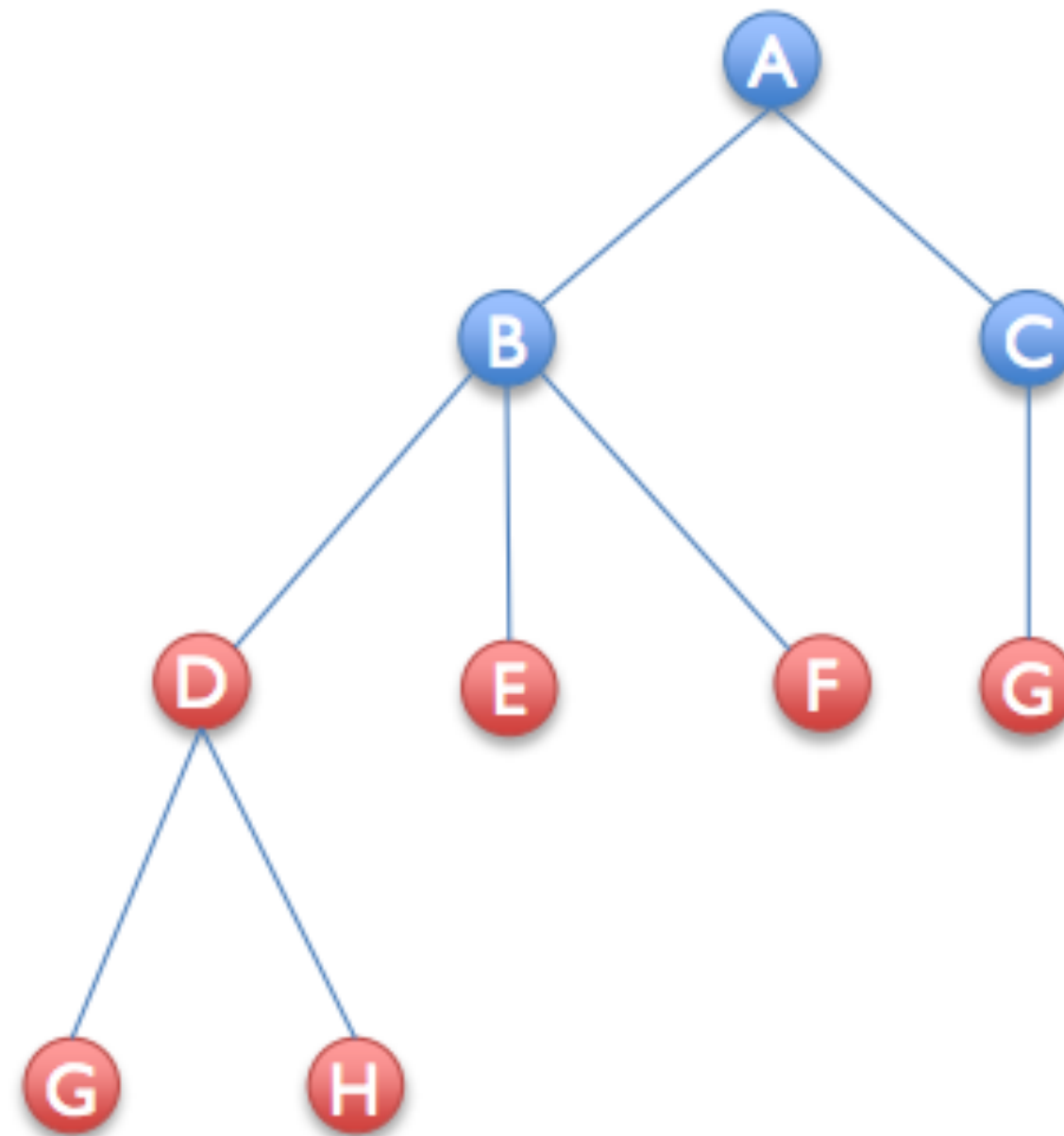
All odd-degree vertices
and any edges
connecting them



Christofides Algorithm

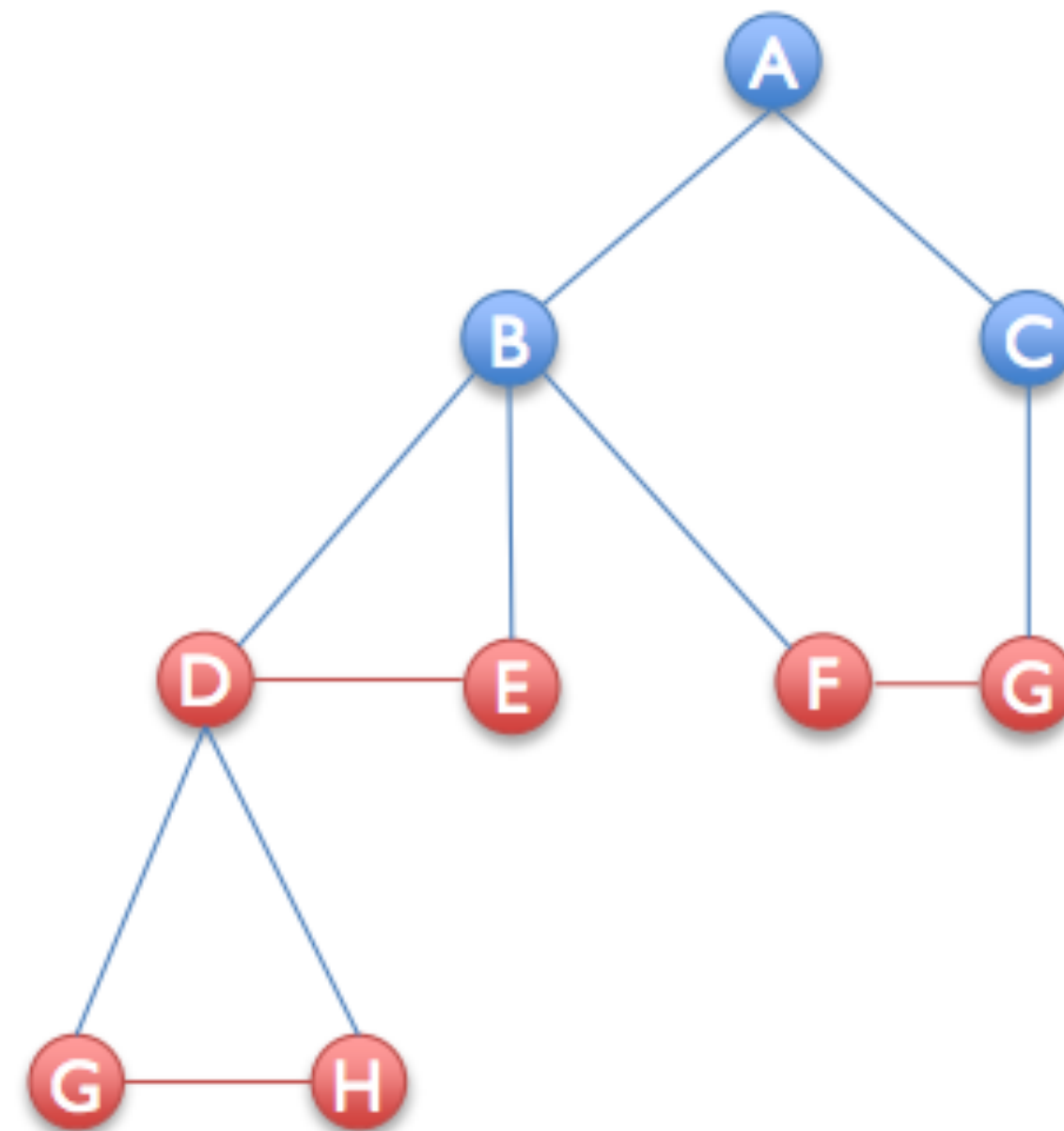
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What does adding M do to O ?



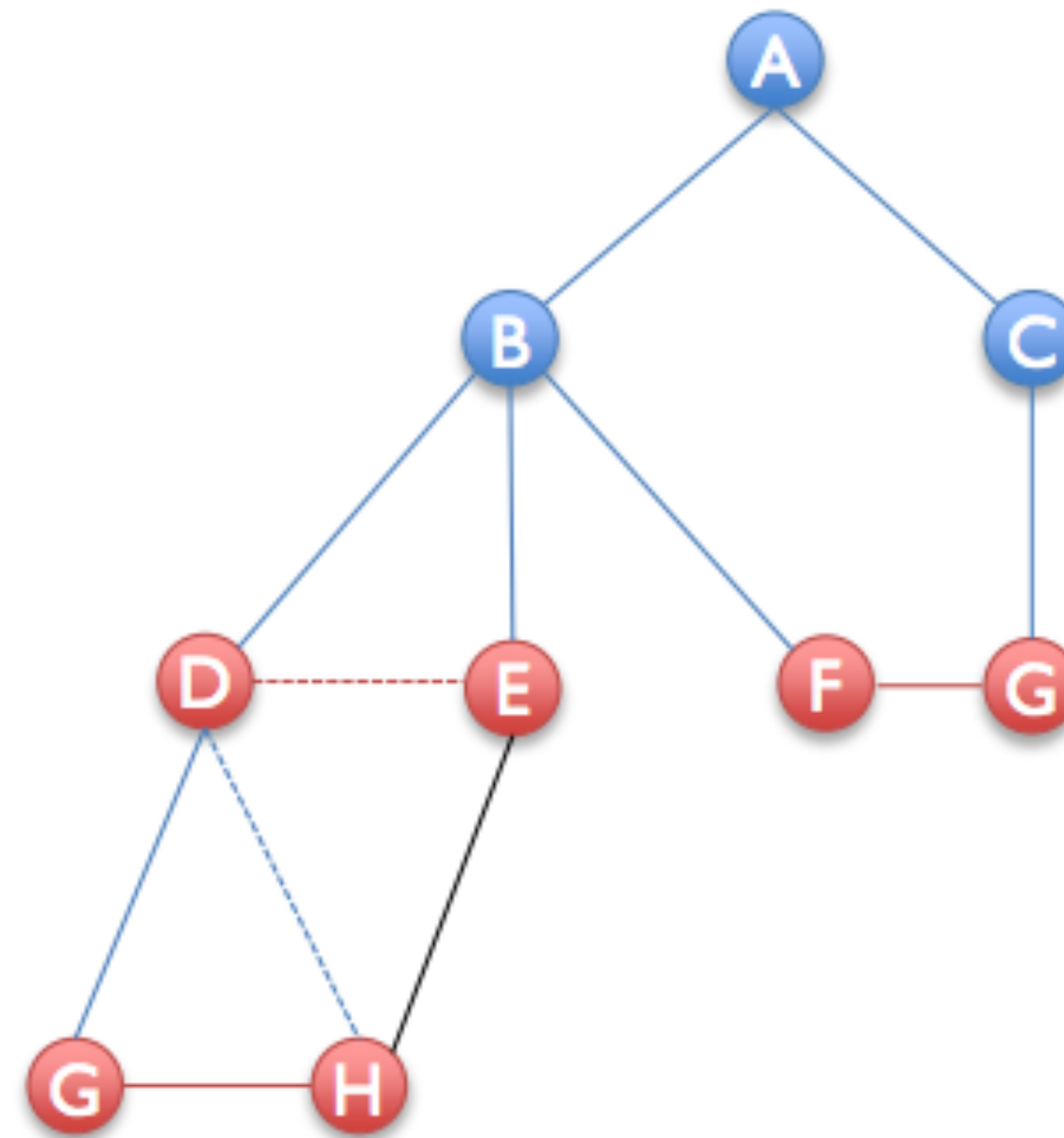
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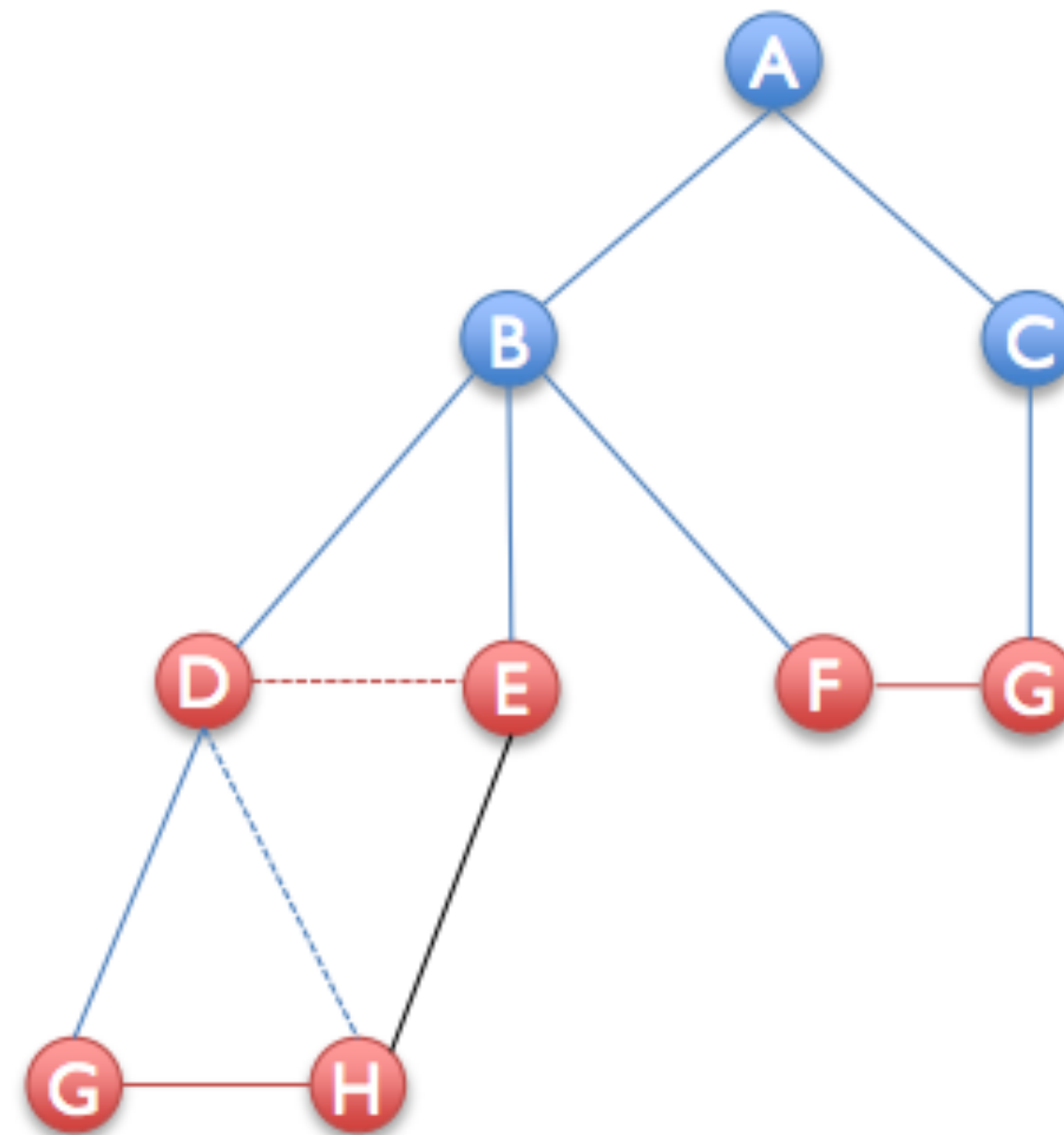
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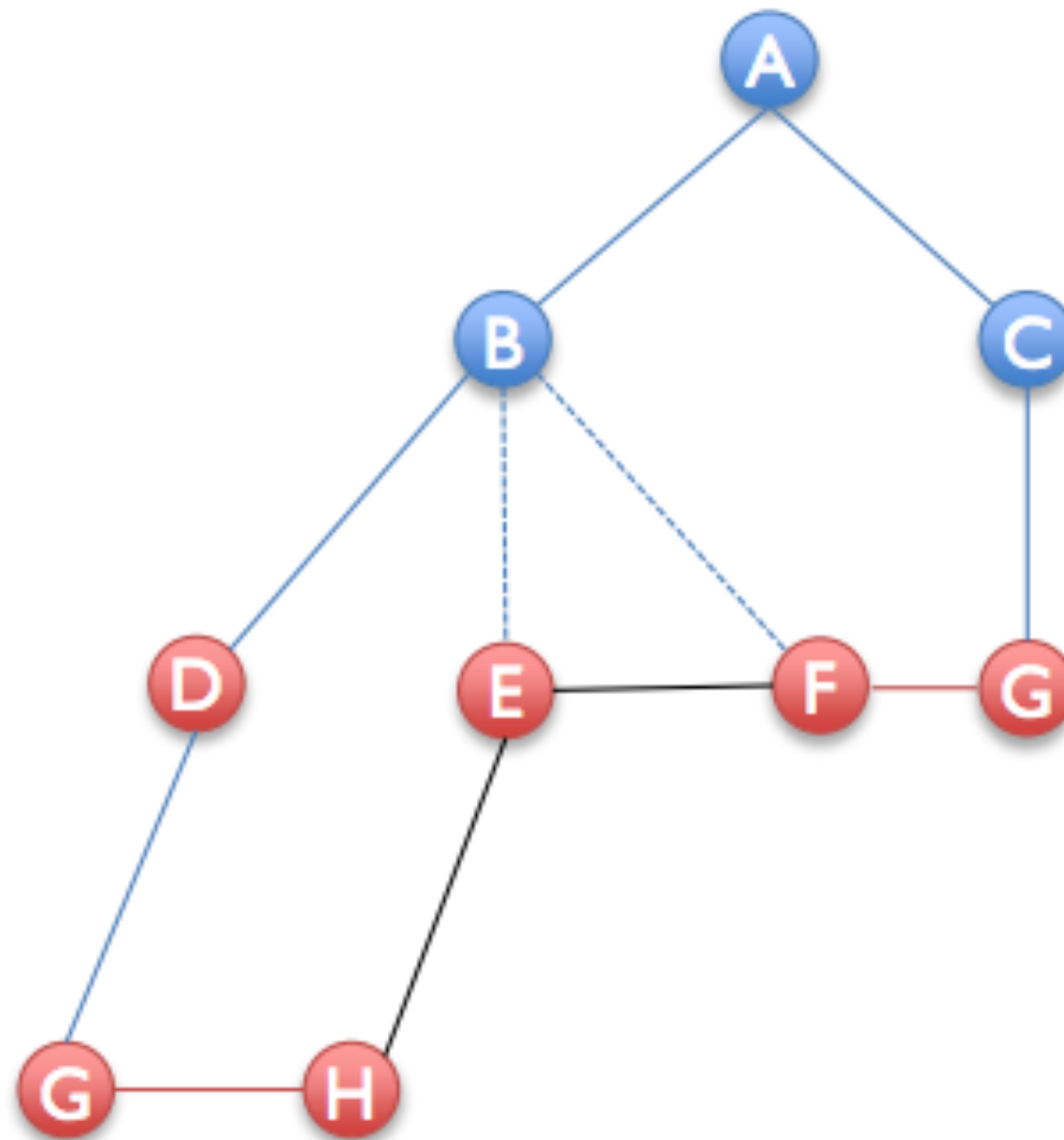
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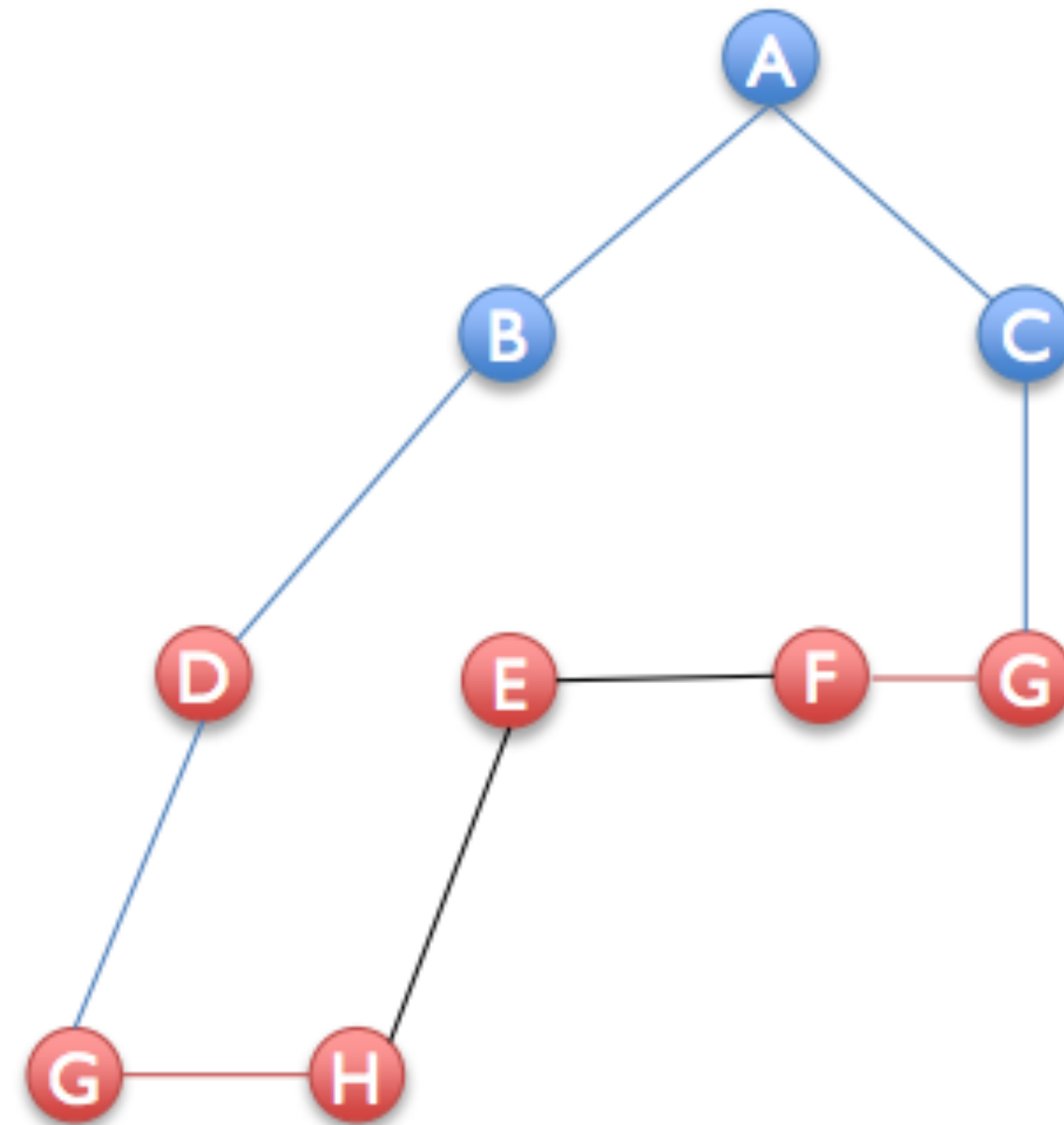
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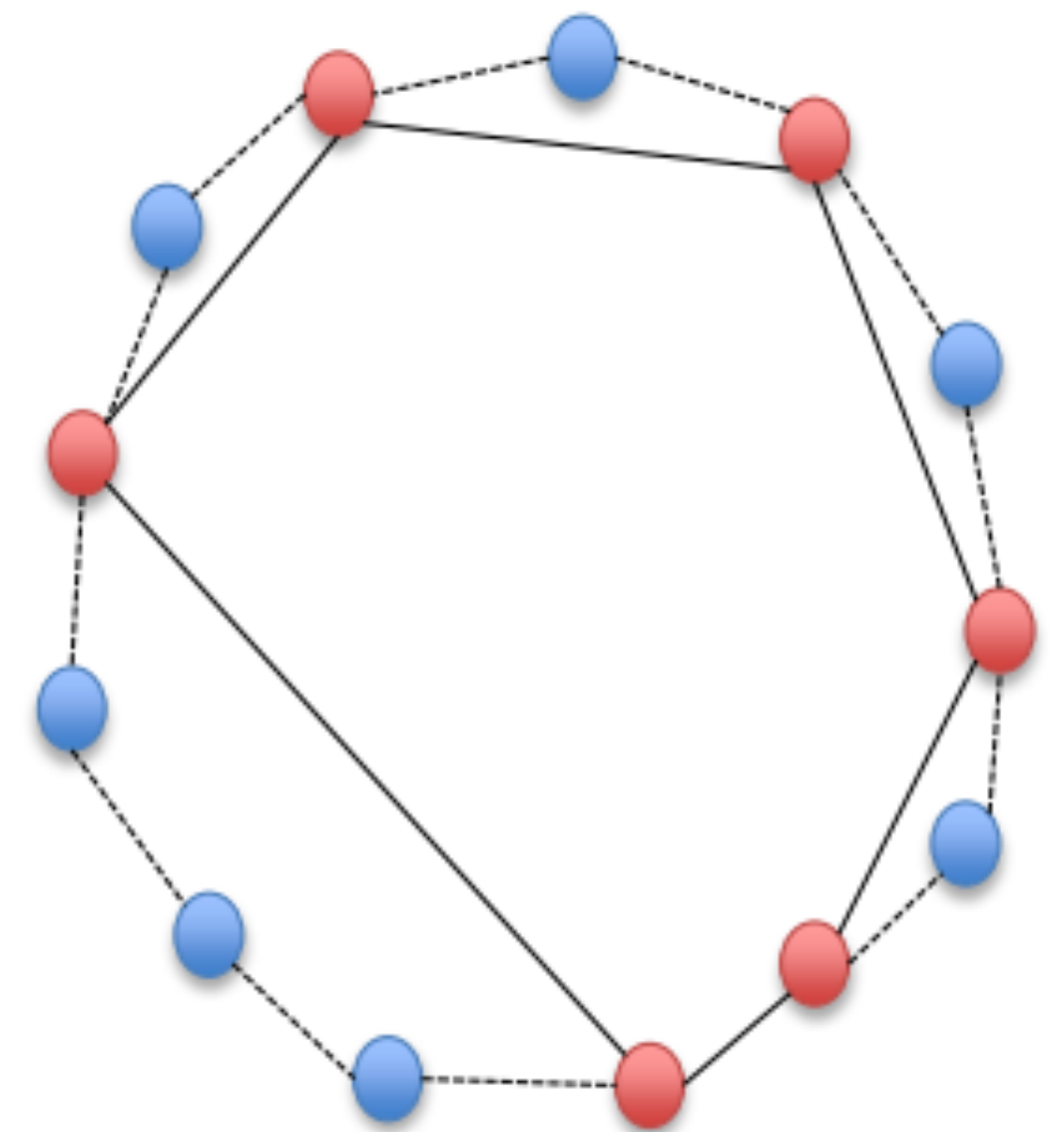
Christofides Analysis

- Cost of TSP tour returned is at most $w(T) + w(M)$
 - T is the MST of the graph, and M is the minimum cost perfect matching on the subgraph induced by O (odd degree nodes in T)
- We know $\text{OPT} \geq w(T)$
- To relate the costs, we lower bound the OPT in terms of the cost of M
- **Claim.** Let OPT be the length of the optimal tour and let M be a minimum-cost perfect matching on the complete subgraph induced by O , the odd degree nodes in MST T , then

$$w(M) = \sum_{e \in M} w_e \leq \frac{1}{2} \cdot \text{OPT}$$

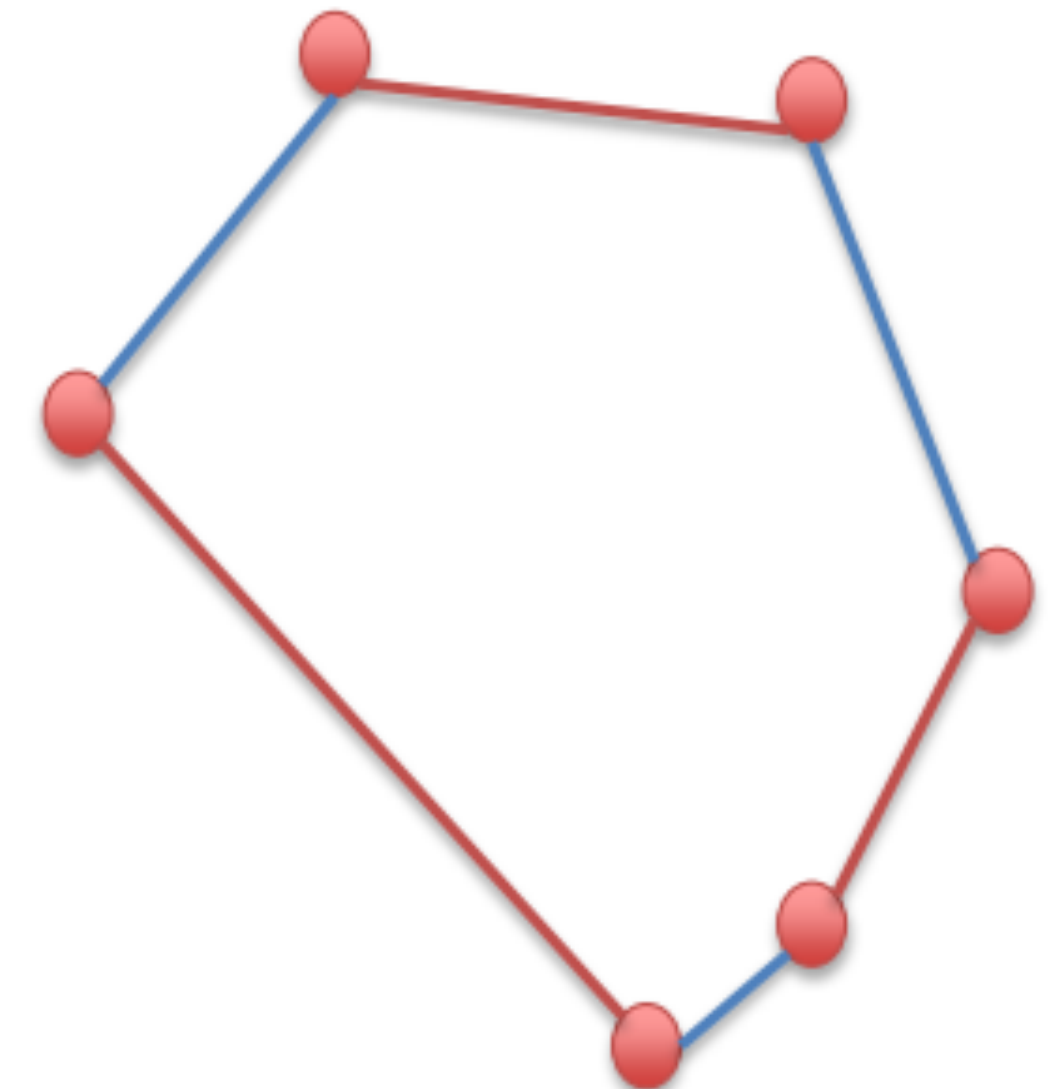
Christofides Analysis

- **Proof of claim.** Consider an optimal tour with cost OPT and consider vertices in O , the odd-degree vertices in T
- Shortcut optimal tour to obtain tour of vertices in O
- By triangle inequality the cost of tour can only decrease



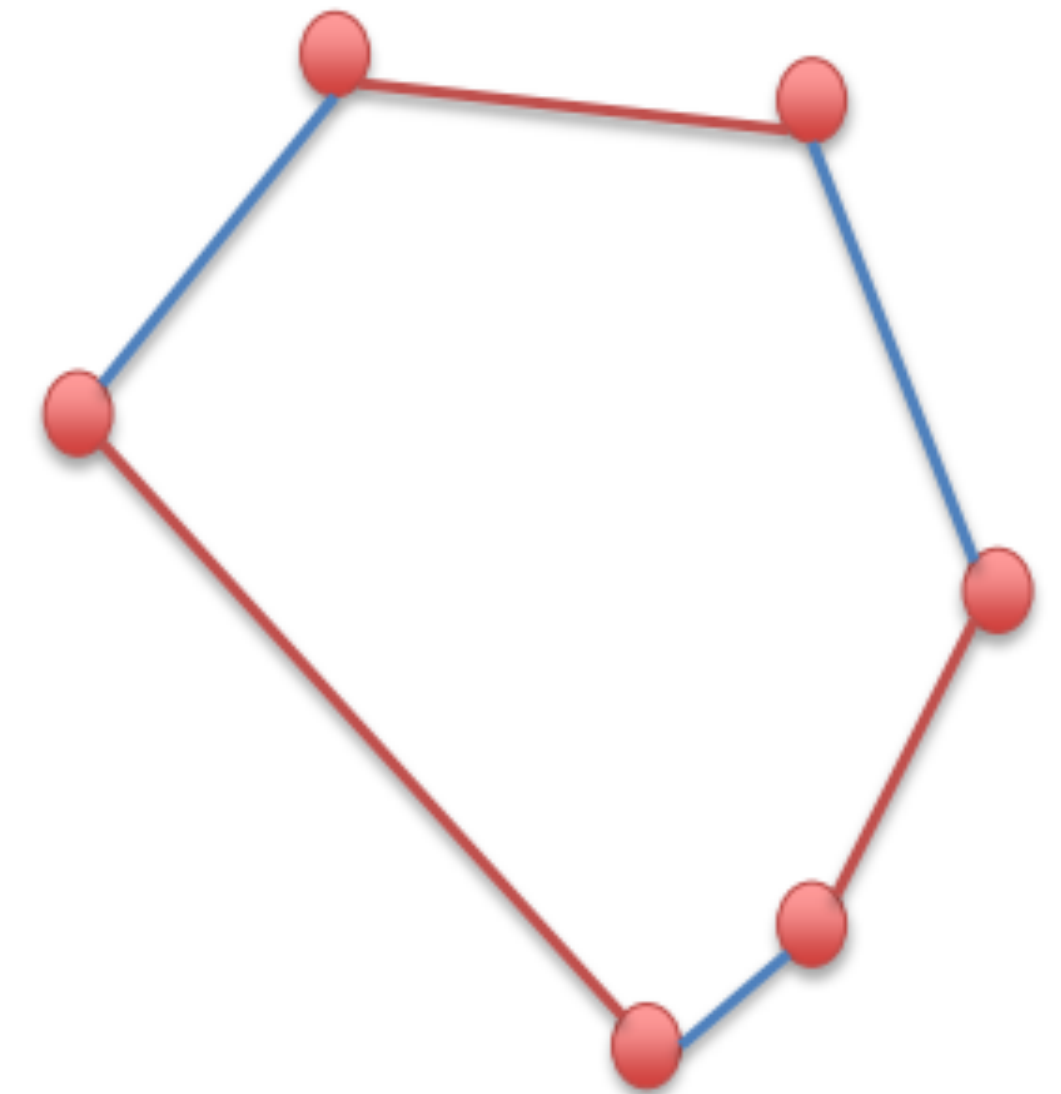
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- Shortcut optimal tour to obtain tour of vertices in O
- By triangle inequality the cost of tour can only decrease
- Consider matchings M_1, M_2 created by alternating edges on this tour
- $w(M_1) + w(M_2) \leq \text{OPT}$
- Then, $\min\{w(M_1), w(M_2)\} \leq \text{OPT}/2$
- $w(M) \leq \min\{w(M_1), w(M_2)\}$, where M : min-cost perfect matching on subgraph induced by O
- Thus, $w(M) \leq \text{OPT}/2$



Wrapping Up

- Cost of TSP tour returned by Christofides $\leq w(T) + w(M)$
- We showed that OPT (optimal cost) $\geq w(T)$ and $\text{OPT} \geq 2 \cdot w(M)$
- Thus, cost of TSP tour returned by Christofides
$$\leq \text{OPT} + \text{OPT}/2$$
$$\leq 1.5 \text{ OPT}$$
- Christofides is a 1.5 approximation to TSP



TSP: Summary

- Held & Karp [1970s] developed a heuristic for calculating a lower bound on a TSP tour (coincides with a linear program known as Held-Karp relaxation)
 - Conjectured to give a $4/3$ -approximation
- [Papadimitriou & Vempala, 2000's] NP-hard to approximate metric TSP within $220/219 \sim 1.0004$
 - Simplified and slightly improved by Lampis'12
- “Four decades after its discovery, Christofides’ algorithm was the best approximation algorithm known for metric TSP”

No PTAS

Last Summer

- This past summer [Karlin, Klein, Shayan] (unpublished):
 - 1.499 approximation
- “Euclidean TSP” does have a PTAS! [Aurora 98] [Mitchell 99]
- Understanding the approximability of TSP is a major open problem in TCS

Christofide's isn't optimal!

A (Slightly) Improved Approximation Algorithm for Metric TSP

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University of Washington

September 1, 2020

Abstract

For some $\epsilon > 10^{-36}$ we give a $3/2 - \epsilon$ approximation algorithm for metric TSP.

Acknowledgments

- Some of the material in these slides are taken from
 - Kleinberg Tardos Slides by Kevin Wayne (<https://www.cs.princeton.edu/~wayne/kleinberg-tardos/pdf/04GreedyAlgorithmsI.pdf>)
 - Jeff Erickson's Algorithms Book (<http://jeffe.cs.illinois.edu/teaching/algorithms/book/Algorithms-JeffE.pdf>)
 - Lecture slides: <https://web.stanford.edu/class/archive/cs/cs161/cs161.1138/>