

# CS4234

## Optimiz(s)ation Algorithms

L8 – (Weighted) Max-Cardinality  
– (Bipartite) – Matching

<https://visualgo.net/en/matching>

This course material is now made available for public usage.  
Special acknowledgement to School of Computing, National University of Singapore  
for allowing Steven to prepare and distribute these teaching materials.

# CS3233/CS4234

## Dual Slides 😊



Dr. Steven Halim



# Roadmap (for CS4234)


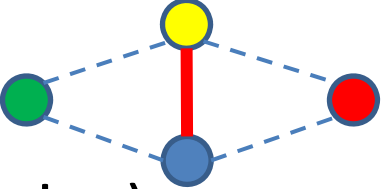
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## Graph Matching

- Overview
- Unweighted MCBM: Max Flow review, Augmenting Path, Hopcroft Karp's (nice *theoretical* result), Augmenting Path++
  - A few relevant applications will be discussed in tutorial ☺
- Weighted MCBM: (Min/Max) Cost (Max) Flow, Hungarian (**overview**)
- Unweighted MCM: Edmonds's Matching (**overview**)
- Weighted MCM: DP with Bitmask (**small graph only, review**)
  - Sorry, still unable to make it work for Christofides's 1.5-Approximation algorithm for M-R/NR-TSP as of year 2018

# Graph Matching

A matching (**marriage**) in a graph  $\mathbf{G} = (\mathbf{V}, \mathbf{E})$  (**real life**) is a subset  $\mathbf{M}$  of  $\mathbf{E}$  edges in  $\mathbf{G}$  (**special relationships**) such that no two of which meet at a common vertex (**no affair**)

Thus a.  and b.   
are matchings (red thick edge),

But trying to match both edges in c.   
is invalid since there is an overlapping vertex

# Max-Cardinality-Matching (MCM)

Usually, the problem asked in graph matching is the size (cardinality) of a maximum (not maximal) matching

A maximum matching is a matching that contains the largest possible number of edges

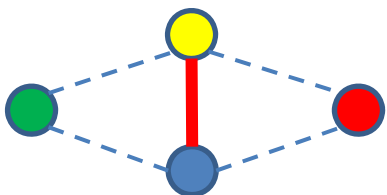
A possible variant: **Perfect** Matching

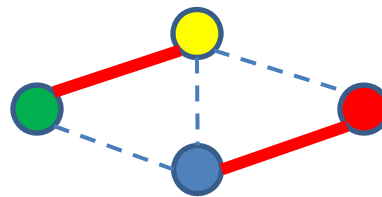
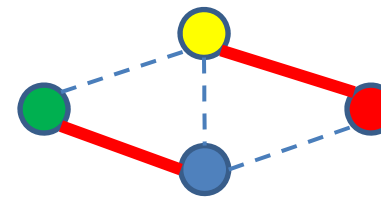
- MCM with no vertex left unmatched
  - PS: This is *impossible* on graph with odd number of vertices

# Examples

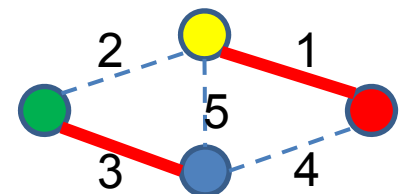
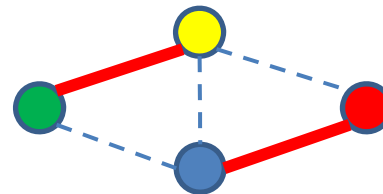
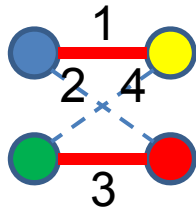
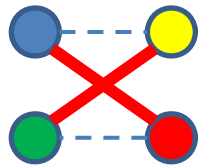
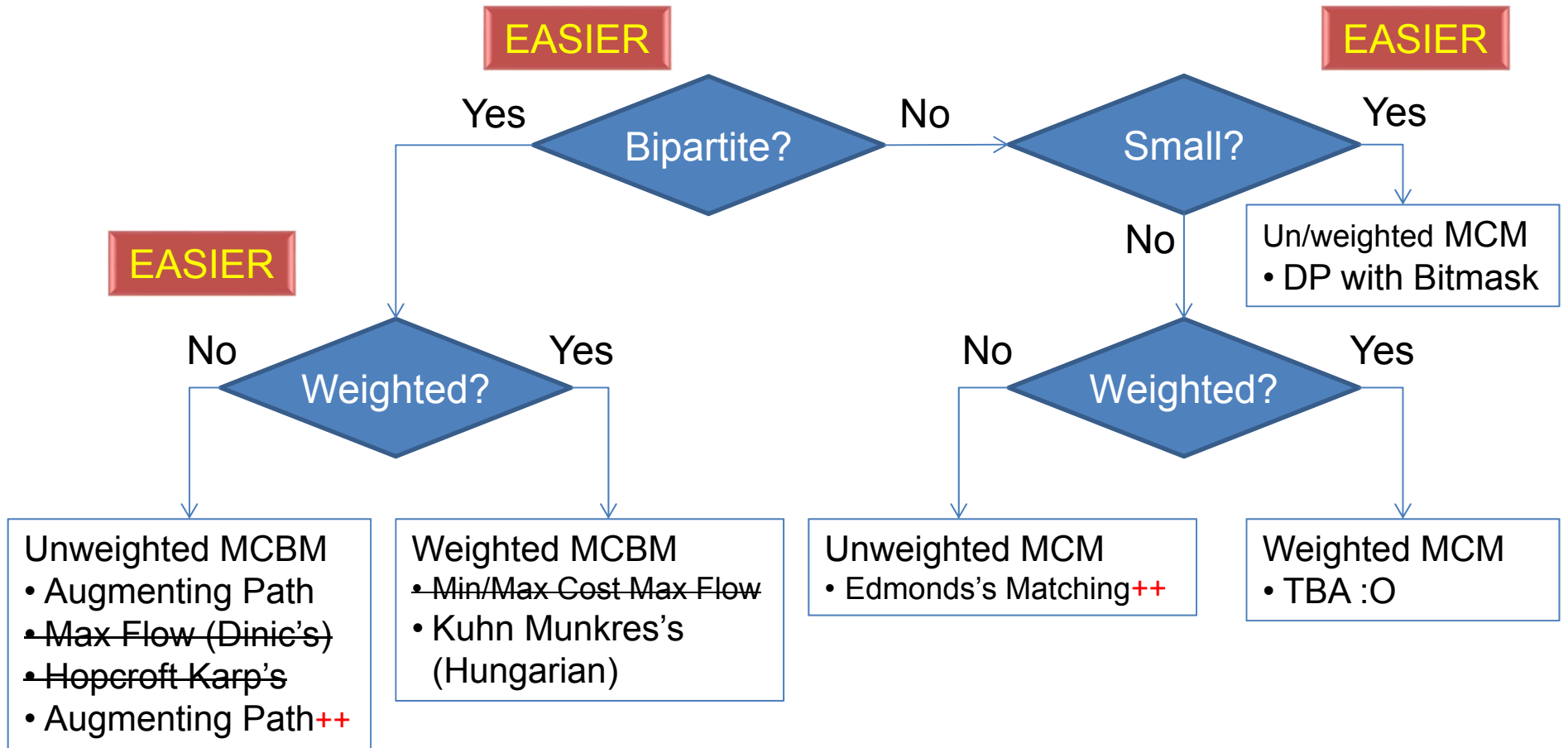
● is a max(imum) matching (0 matching)  
(no edge to be matched)

●—● is also a max(imum) matching (1 matching)  
(no other edge to be matched)

But  is not a max(imum) matching  
(it is a max(**imal**) matching btw)

as we can change it to  or   
(2 matchings)

# Types of Graph Matching



Solutions:

Max Flow (just use Dinic's, one of the best for bipartite matching graphs)

Augmenting Path Algorithm

Hopcroft Karp's Algorithm  $\approx$  Dinic's (for theoretical interest only)

Augmenting Path Algorithm++ (the ++ is very important :O)

# UNWEIGHTED MCBM

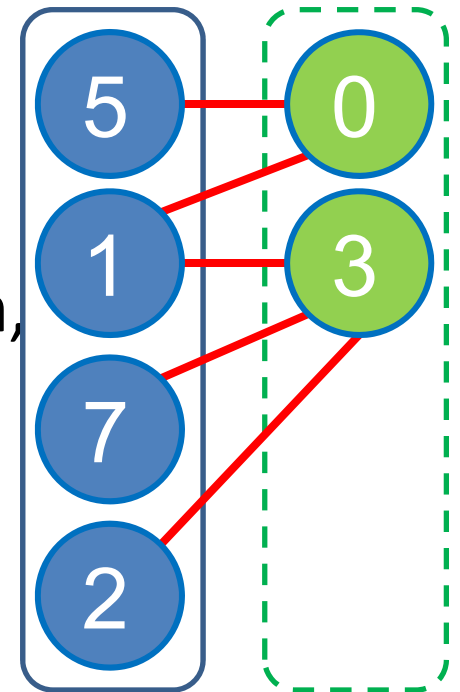


# Max-Cardinality-Bipartite-Matching (MCBM)

A **Bipartite** Graph is a graph whose vertices can be divided into two disjoint sets **X** and **Y** such that every edge can only connect a vertex in **X** to one in **Y**

Matching in this kind of graph is **a lot easier** than matching in general graph

So if we are given a graph matching problem, **our first check** is to see whether the given input graph is bipartite...



See <https://visualgo.net/en/maxflow>, select modeling, Bipartite Matching, all ones

Time Complexity of such modeling:

Depends on the chosen Max Flow algorithm, **but much faster** than general case as the capacities are **all ones (unit weights)** and the graph **is bipartite**, e.g.  $O(m^2)$  for basic FF,  $O(\text{sqrt}(n)*m)$  for Dinic's

Finding MCBM by reducing this problem into

# MAX FLOW

## CP3.17B SECTION 4.6, 4.7 & 4.8.4

# CP3.17b Ex 4.8.4.1\*:

## Why the graph has to be directed?

- Try creating a counter example



# When To Use Max Flow Solution?

Only if we are solving Bipartite Matching **with Capacity** (the “Assignment Problem”)

- e.g. UVa 00259, CP3.17b Section 4.6.5

See <https://visualgo.net/en/maxflow>, select modeling, Bipartite Matching, random

- This variant will be **much slower** to solve if reduced to MCBM, imagine if the typical capacities of the edges are big, reducing the problem to MCBM will lead to a gigantic bipartite graph...

But if we are solving pure MCBM (capacities are all 1), just use the next algorithm (easier to code and faster)

Finding MCBM via

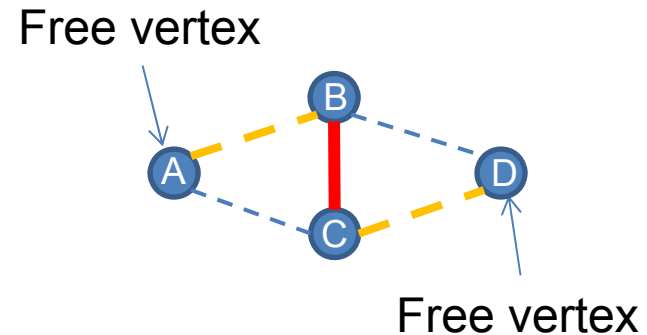
# **AUGMENTING PATH ALGORITHM**

## **CP3.17B SECTION 4.8.4++**

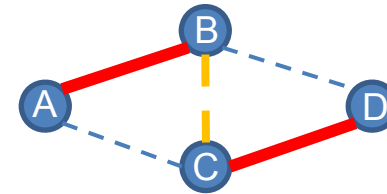
PS: Slightly different from similar term in Network Flow lecture

# Augmenting Path

In this graph, the path colored **orange(unmatched)-red(matched)-orange**: A-B-C-D is an augmenting path



We can flip the edge status to **red-orange-red**: A-B-C-D and the number of edges in the matching set increases by 1



# Augmenting Path Algorithm

Lemma (Claude Berge 1957):

Detailed proof omitted, you can search for external resources if you are interested

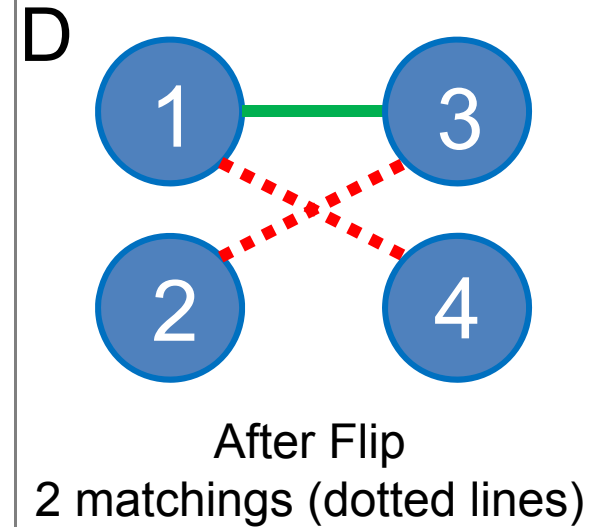
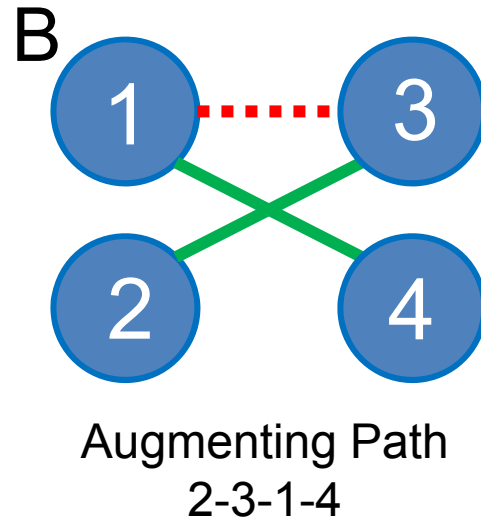
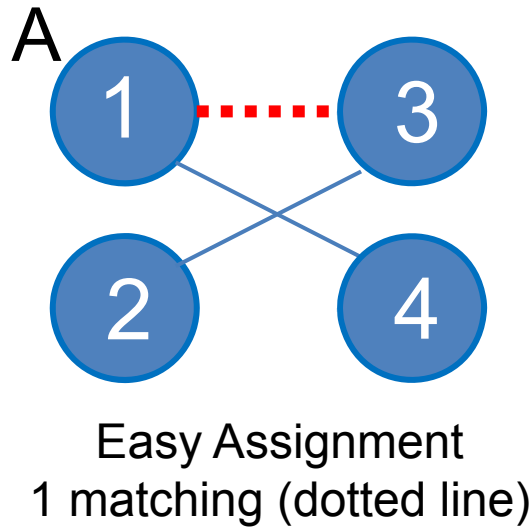
A matching  $M$  in  $G$  is maximum iff there is no more augmenting path in  $G$

$G$  does not necessarily be a Bipartite Graph in the lemma

Augmenting Path Algorithm is a simple  $O(V^*(V+E))$   
 $\approx O(VE)$  implementation of that lemma

- Find and then eliminate augmenting paths in  $G$ 
  - PS: Later we will see an  $O(kE)$  implementation of this thing,  $k \ll V$

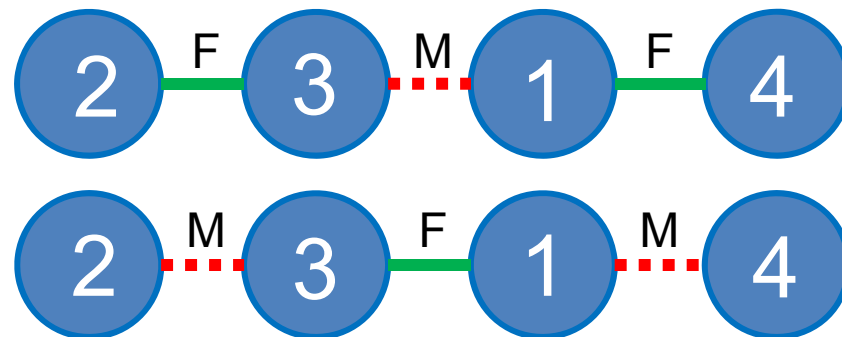
# Augmenting Path Algorithm



**C**

An augmenting path  
F=Free, M=Matched

Flip to increase matching  
from 1 to 2 matchings





Finding MCBM via

# **HOPCROFT KARP'S ALGORITHM (SECTION 9.13)**

<https://visualgo.net/en/matching>, load (almost) complete Bipartite Graph, and see how slow the standard Augmenting Path algorithm can be...

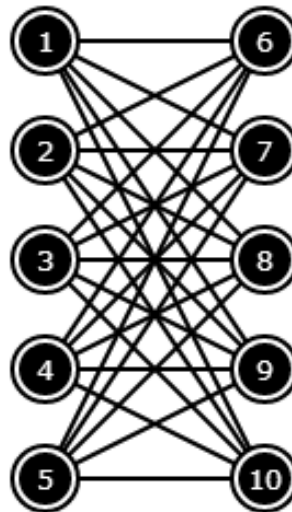
## An Extreme Test Case...

A Complete Bipartite Graph  $K_{N,M}$ ,  $V=N+M$  &  $E = N*M$

Augmenting Path algorithm  $\rightarrow O((N+M)*(N*M))$

- If  $M = N$ , we have an  $O(N^3)$  solution, only OK for  $N \leq 200$

Example with  $N = M = 5$



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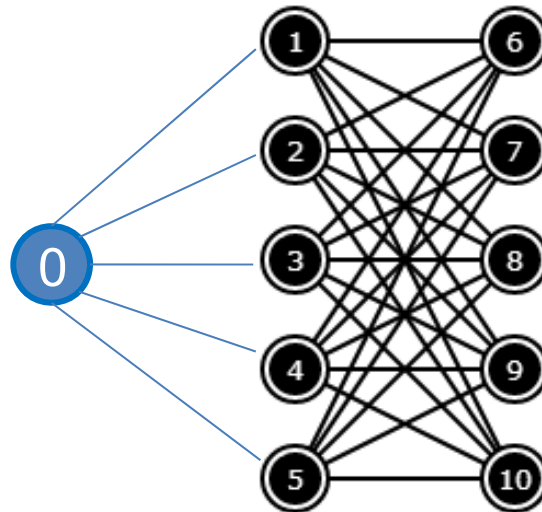
Steven Halim, SoC, NUS

<https://visualgo.net/en/matching>, reviving HK animation by Week 09 or 10??

# Hopcroft Karp's Algorithm (1973)

Key Idea (very similar to *Dinic's algorithm*\*):

- Use shortest augmenting paths from all free vertices (BFS)
- Run similar algorithm as the Augmenting Path Algorithm earlier (DFS), but now using this BFS information,
- Can have  $\geq 1$  matching(s) per iteration




# Hopcroft Karp's Algorithm (1973)

That is, up to  $\sqrt{V}$  iterations only

Hopcroft Karp's runs in  $O(\sqrt{V} * E)$ , proof omitted

- For the extreme test case in previous slide, this is  $O(\sqrt{N+M} * N * M)$
- With  $M = N$ , this is about  $O(N^{5/2})$ , OK for  $N \leq 600$ 
  - PS: Dinic's algorithm on Bipartite Matching graph also runs in  $O(N^{5/2})$

CP3.17b Ex 4.8.4.3\*?: Is this algorithm **must be learned**   
in order to do well in practice<sub>/programming contest</sub> ?

- CP Answer: NOT needed...
  - You can make the standard augmenting path algorithm avoids its worst case behavior by doing randomized greedy pre-processing (randomized to avoid adversary test case) 😊, try this in VisuAlgo: "Augmenting Path with Randomized Greedy Preprocessing",  $O(kE)$

# The ++ Code (1) 😊

```
vi match, vis; // global variables

int Aug(int L) { // return 1 if ∃ an augmenting path from L
    if (vis[L]) return 0; // return 0 otherwise
    vis[L] = 1;
    for (auto &R : AL[L])
        if (match[R] == -1 || Aug(match[R])) {
            match[R] = L;
            return 1; // found 1 matching
        }
    return 0; // no matching
}
```

# The ++ Code (2) 😊

```
int main() {
    ios::sync_with_stdio(false); cin.tie(NULL);

    // inside int main()
    // build unweighted bipartite graph
    // with directed edge left->right set

    unordered_set<int> freeV;
    for (int L = 0; L < Vleft; L++) // assume all vertices
        freeV.insert(L); // on left set are free initially
    match.assign(V, -1); // V is the number of vertices
    int MCBM = 0; // in bipartite graph
```

# The ++ Code (3) 😊

```
// Greedy pre-processing for trivial Augmenting Paths
for (int L = 0; L < Vleft; L++) { // O(V^2)
    vi candidates;
    for (auto &R : AL[L])
        if (match[R] == -1)
            candidates.push_back(R);
    if (candidates.size() > 0) {
        MCBM++;
        freeV.erase(L); // L is matched, no longer a free vertex
        int a = rand()%candidates.size(); // randomize this
        match[candidates[a]] = L; // greedy matching
    }
}
```

# The ++ Code (4) 😊

```
// for each of the k remaining free vertices
for (auto &f : freeV) {
    vis.assign(Vleft, 0); // reset before each recursion
    MCBM += Aug(f); // once f is matched,
} // f remains matched till end
cout << "Found " << MCBM << " matchings\n";

return 0;
}
```



# Live Solve

<https://open.kattis.com/problems/b>

From my Augmenting Path

Algorithm++ baseline code (C++)

Solution:

Min/Max Cost (Max) Flow (Overview Only)

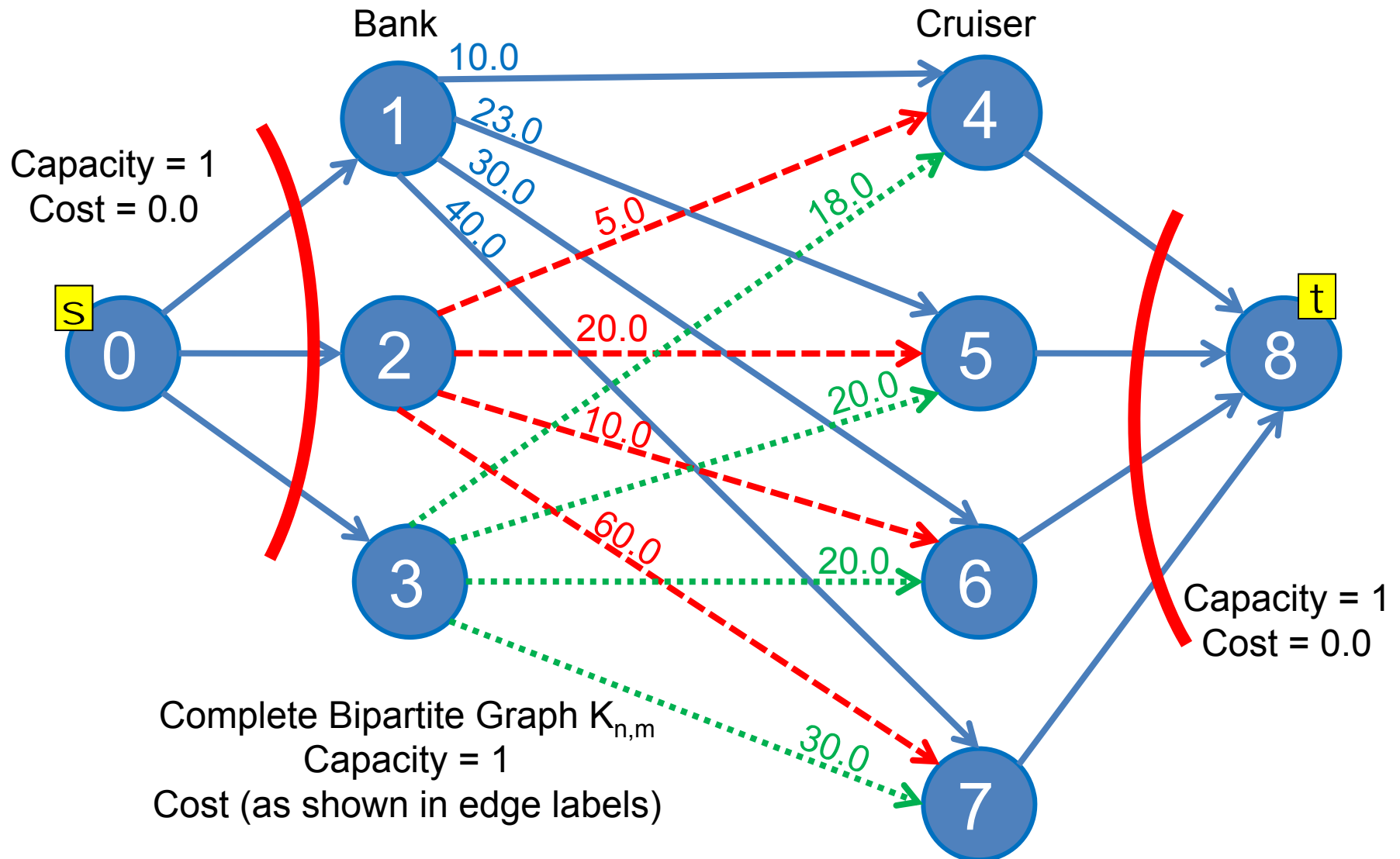
Kuhn Munkres's (Hungarian, Overview Only)

# **WEIGHTED MCBM**



# UVa 10746 (1)

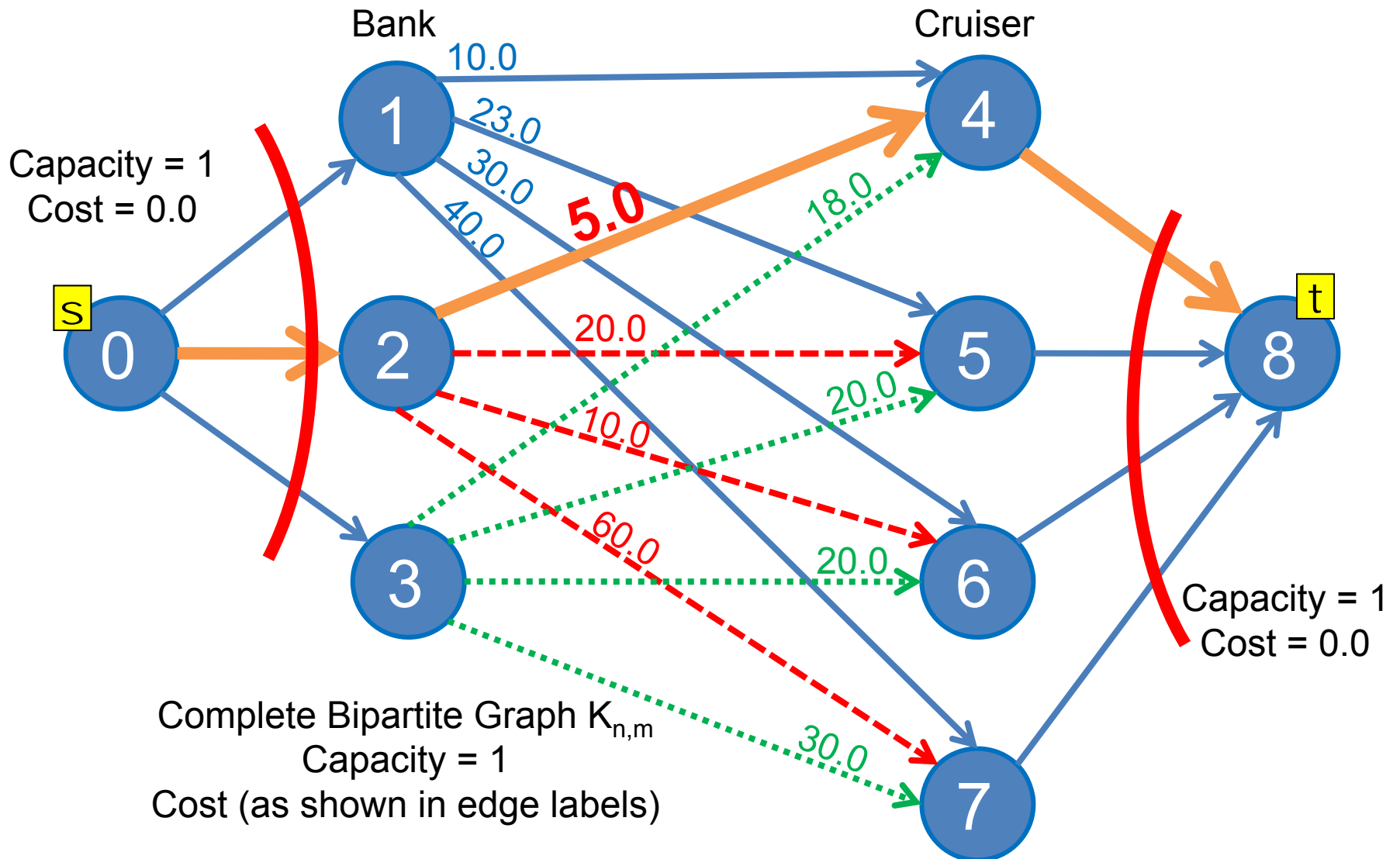
Find Max Flow (=3 here)  
of Minimum Cost (still in P)  
That is, Min Cost Bipartite Matching...



Min Cost so far =  
 $0 + 5.0 + 0 = 5.0$

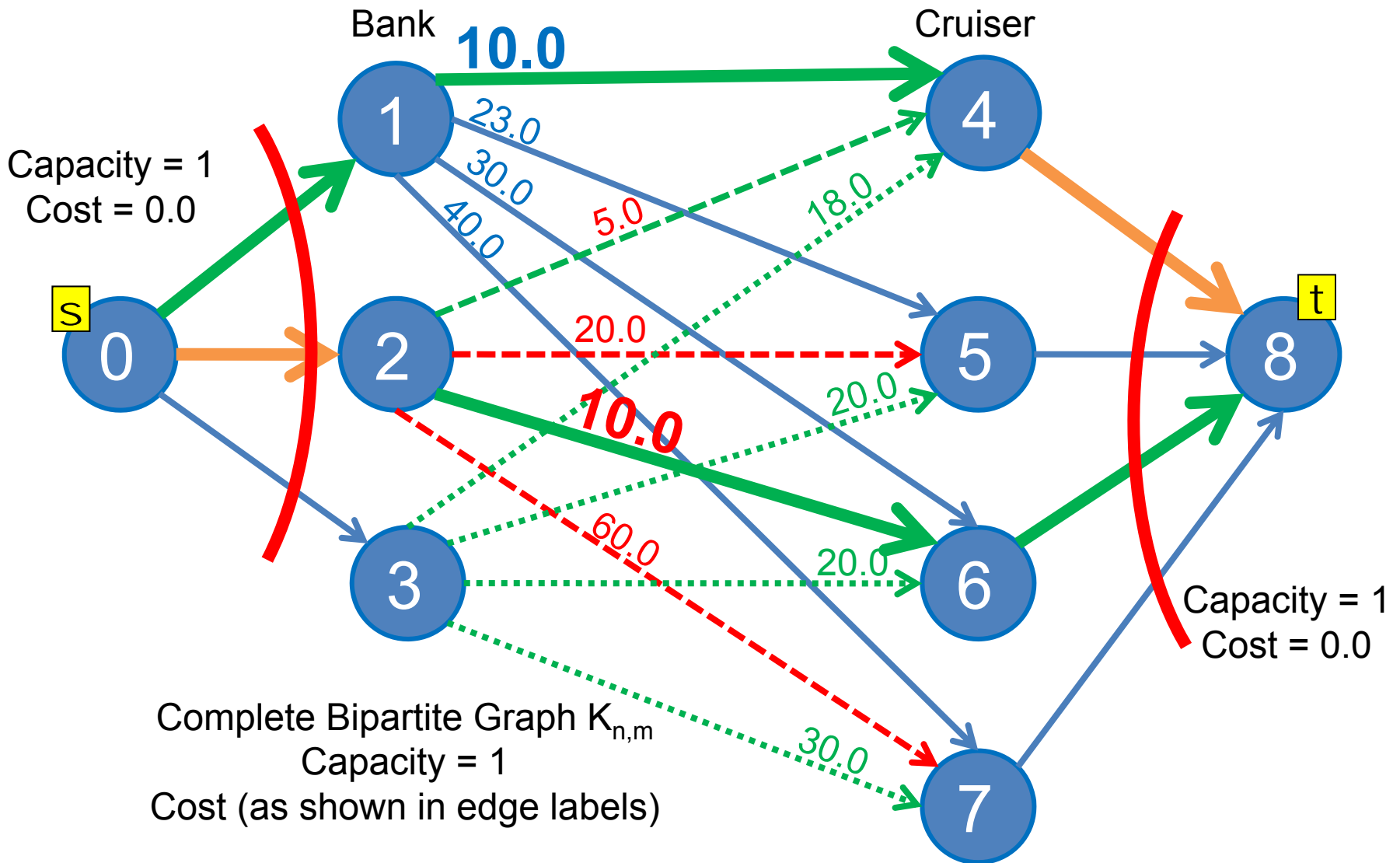


# UVa 10746 (2)



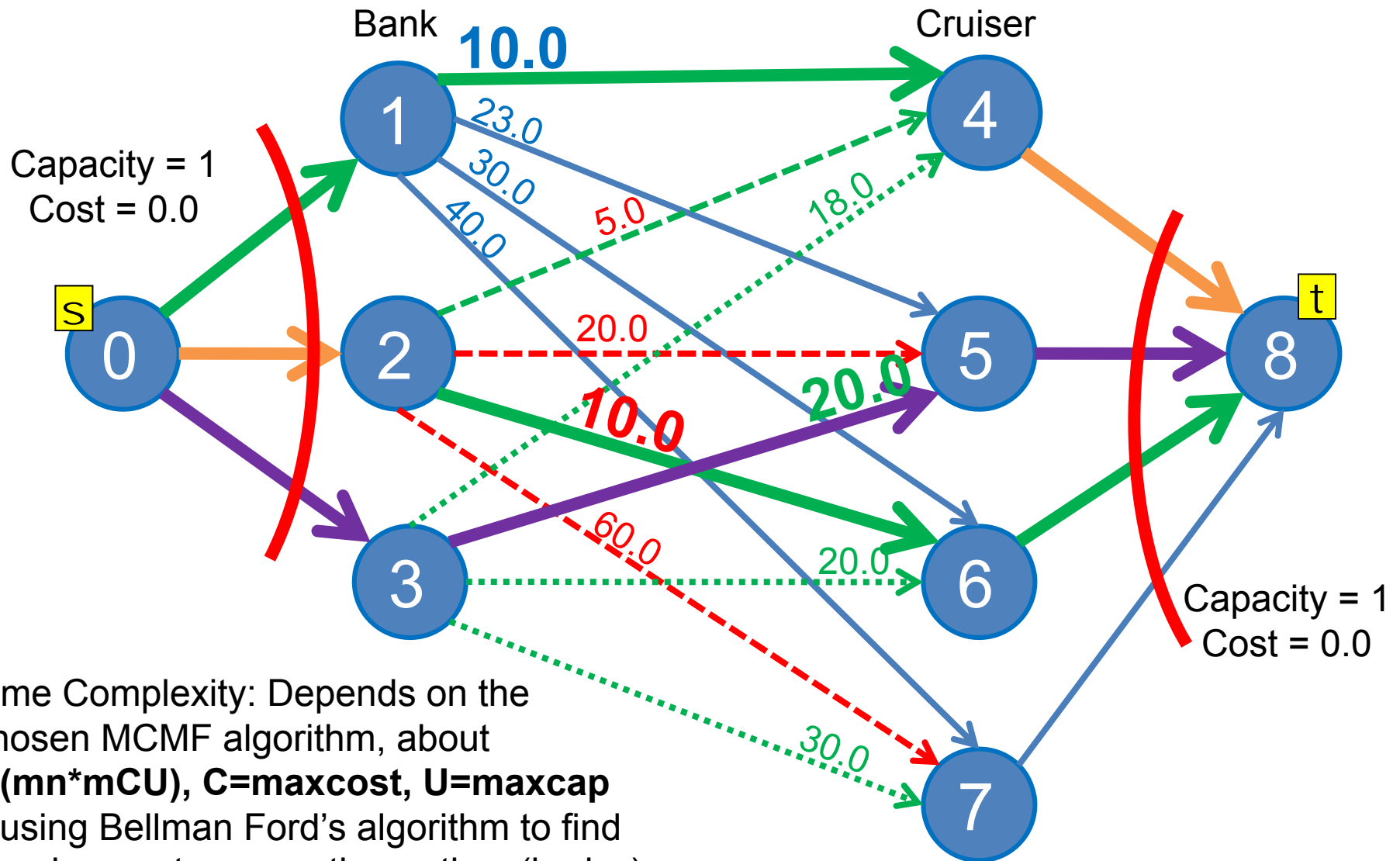
# UVa 10746 (3)

Min Cost so far =  
 $0 + 5.0 + 0 +$   
 $0 + 10.0 - 5.0 + 10.0 + 0 = 20.0$



# UVa 10746 (4)

Min Cost so far =  
 $0 + 5.0 + 0 +$   
 $0 + 10.0 - 5.0 + 10.0 + 0 +$   
 $0 + 20.0 + 0 = 40.0$



Time Complexity: Depends on the chosen MCMF algorithm, about  $O(mn \cdot mCU)$ ,  $C = \text{maxcost}$ ,  $U = \text{maxcap}$  if using Bellman Ford's algorithm to find the cheapest augmenting path... (bad...)

# Kuhn Munkres's (Hungarian) Algorithm

Harold Kuhn and James Munkres name their algorithm based on the work of two other Hungarian mathematicians (Denes Konig and Jano Egervary)

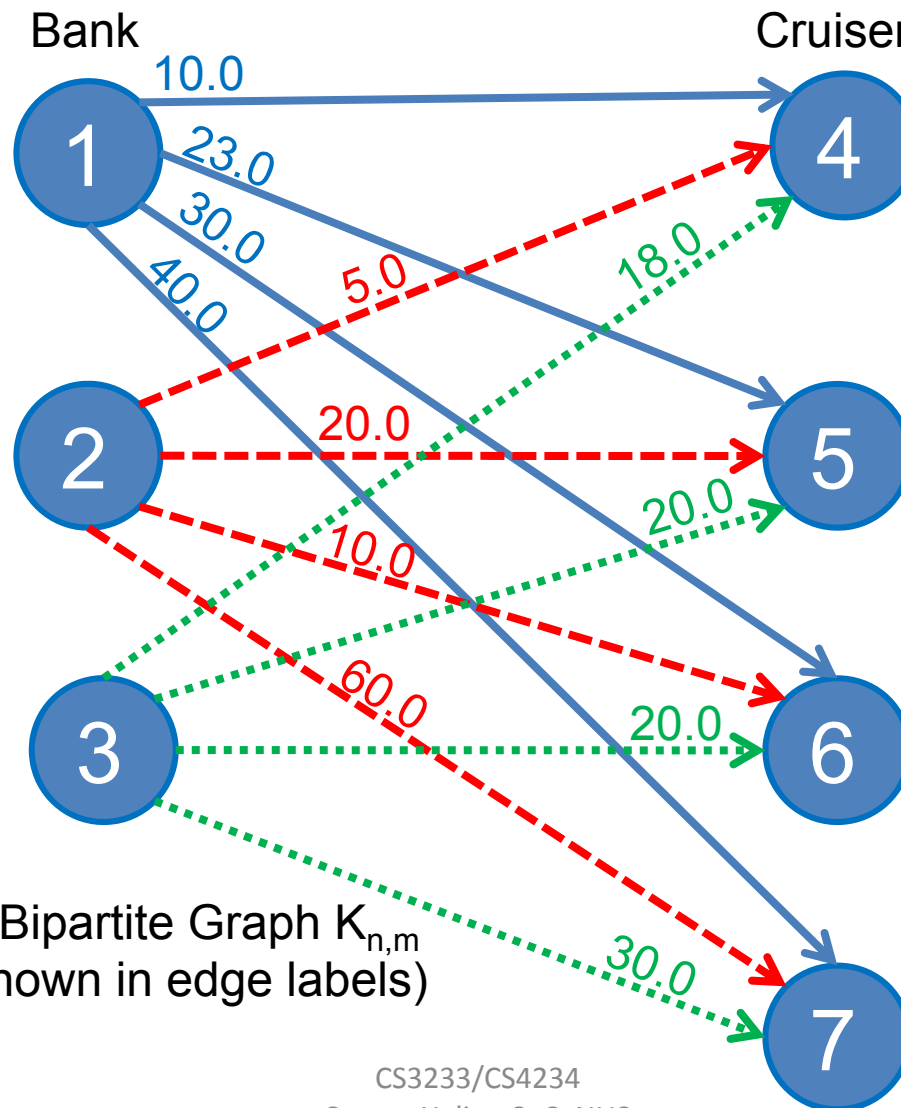
A rough explanation using this \*other\* animation tool:

[https://www-m9.ma.tum.de/graph-algorithms/matchings-hungarian-method/index\\_en.html](https://www-m9.ma.tum.de/graph-algorithms/matchings-hungarian-method/index_en.html)



# UVa 10746 (5)

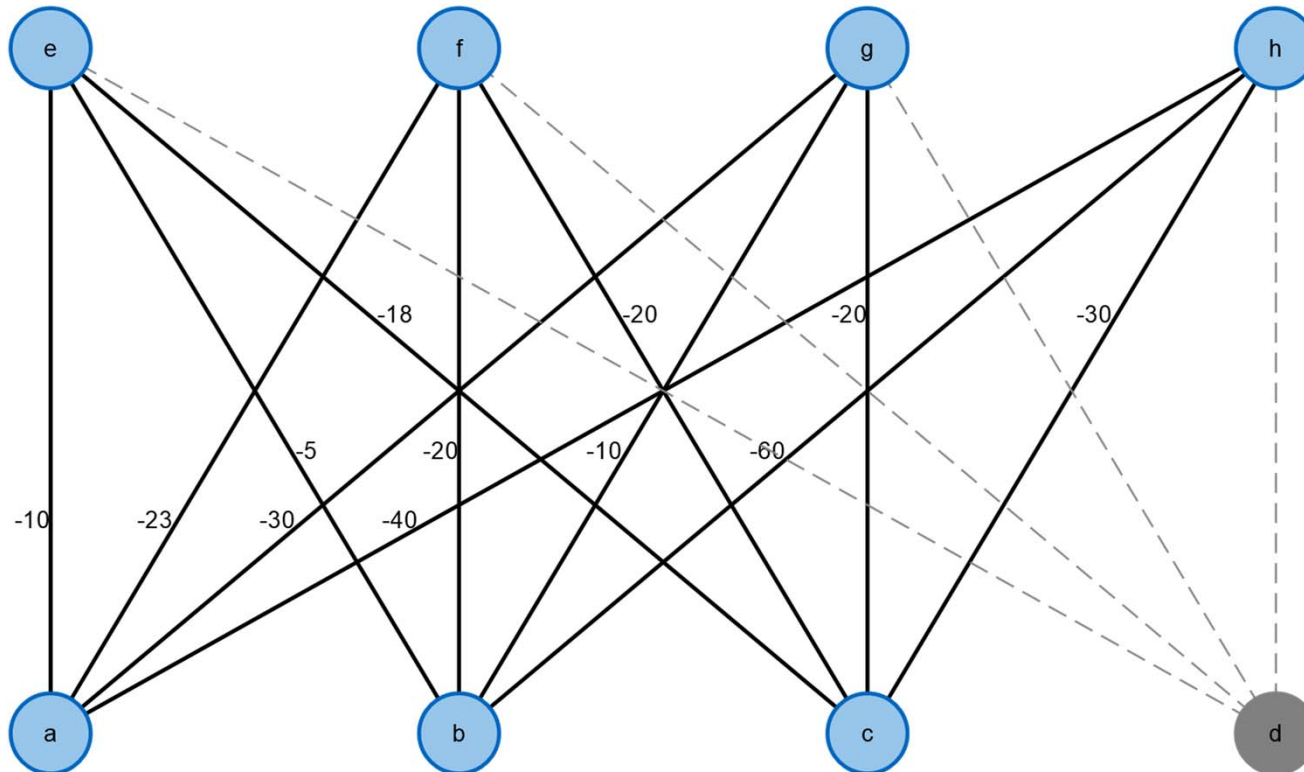
Find MCBM (=3 here)  
of Minimum Cost (still in P)  
That is, Min Cost Bipartite Matching...



Complete Bipartite Graph  $K_{n,m}$   
Cost (as shown in edge labels)

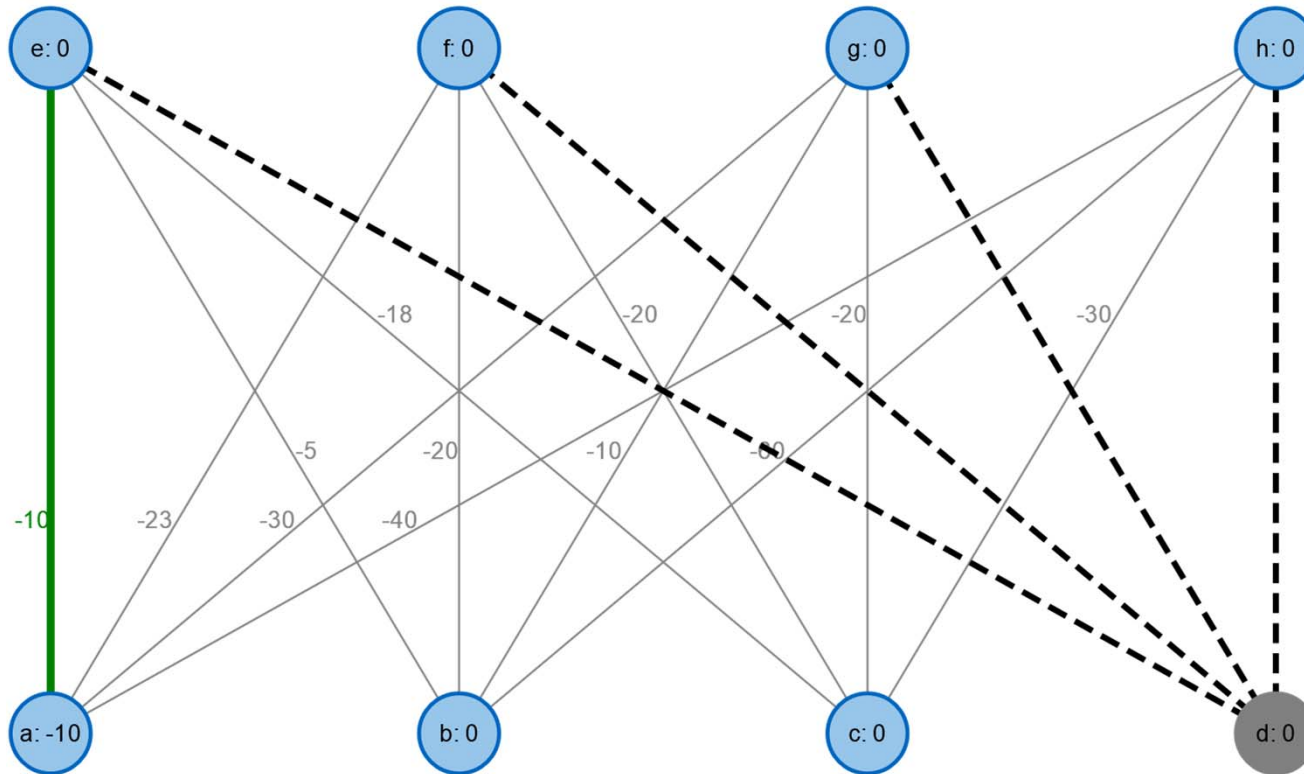


# UVa 10746 (6)



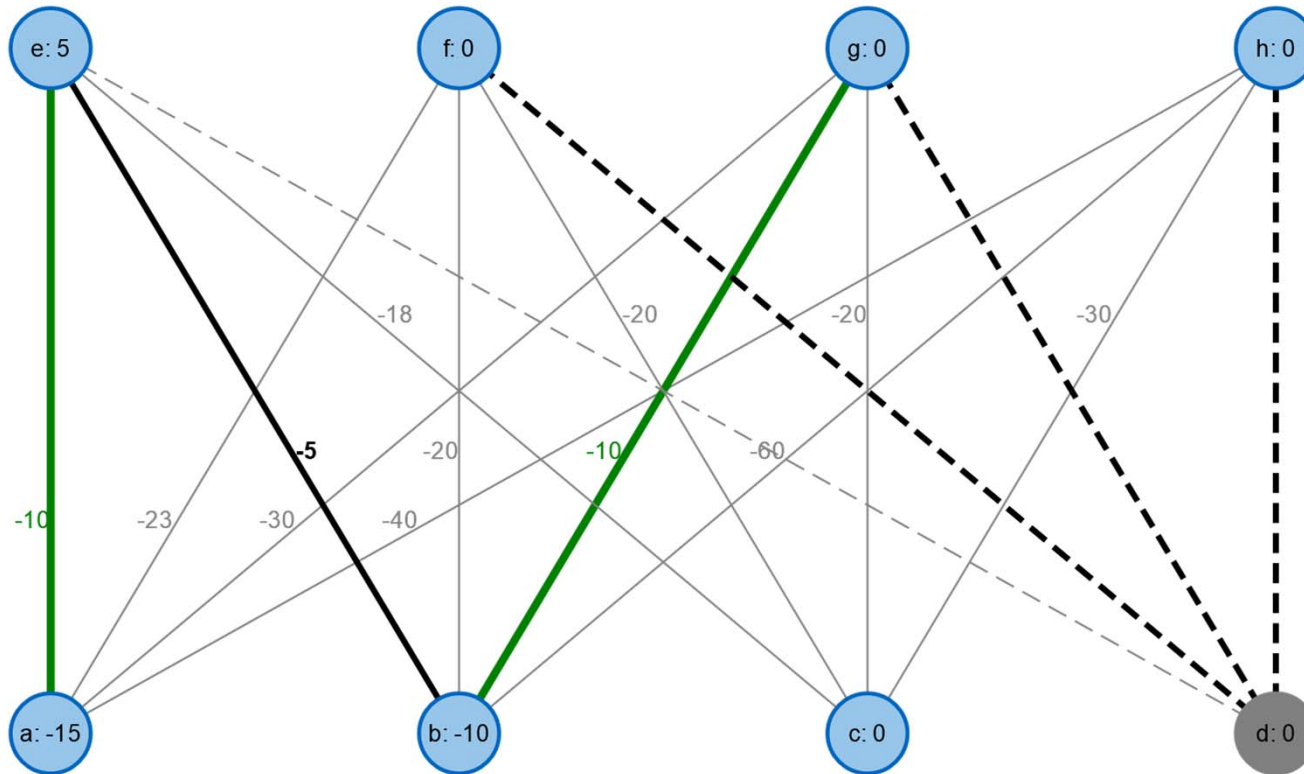
This tool [https://www-m9.ma.tum.de/graph-algorithms/matchings-hungarian-method/index\\_en.html](https://www-m9.ma.tum.de/graph-algorithms/matchings-hungarian-method/index_en.html) is on maximization problem on complete bipartite graph, so I have to convert the weights to negative and add dummy vertex/edges to make  $K_{4,4}$ . The layout is top row = right set and bottom row = left set, vertex 'd' is 'virtual'

# UVa 10746 (7)



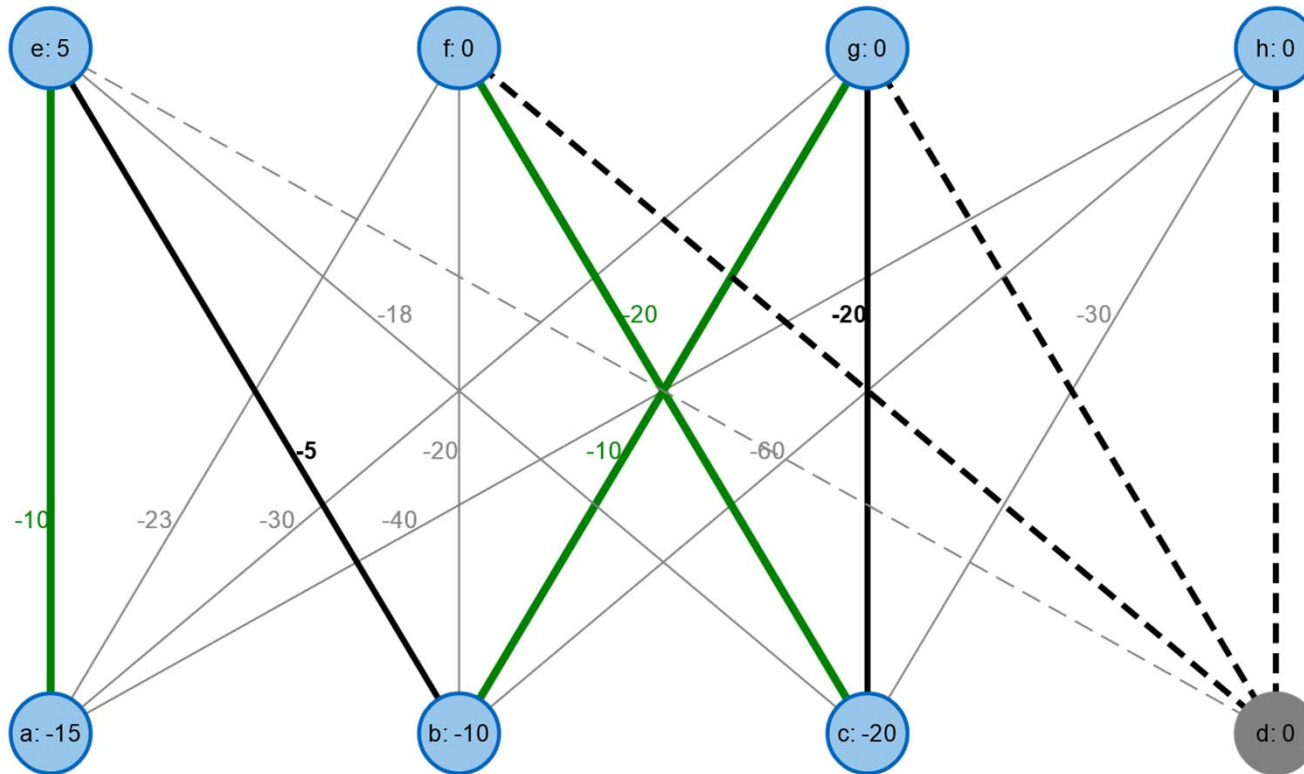
First matching with cost 10 (-10)

# UVa 10746 (8)



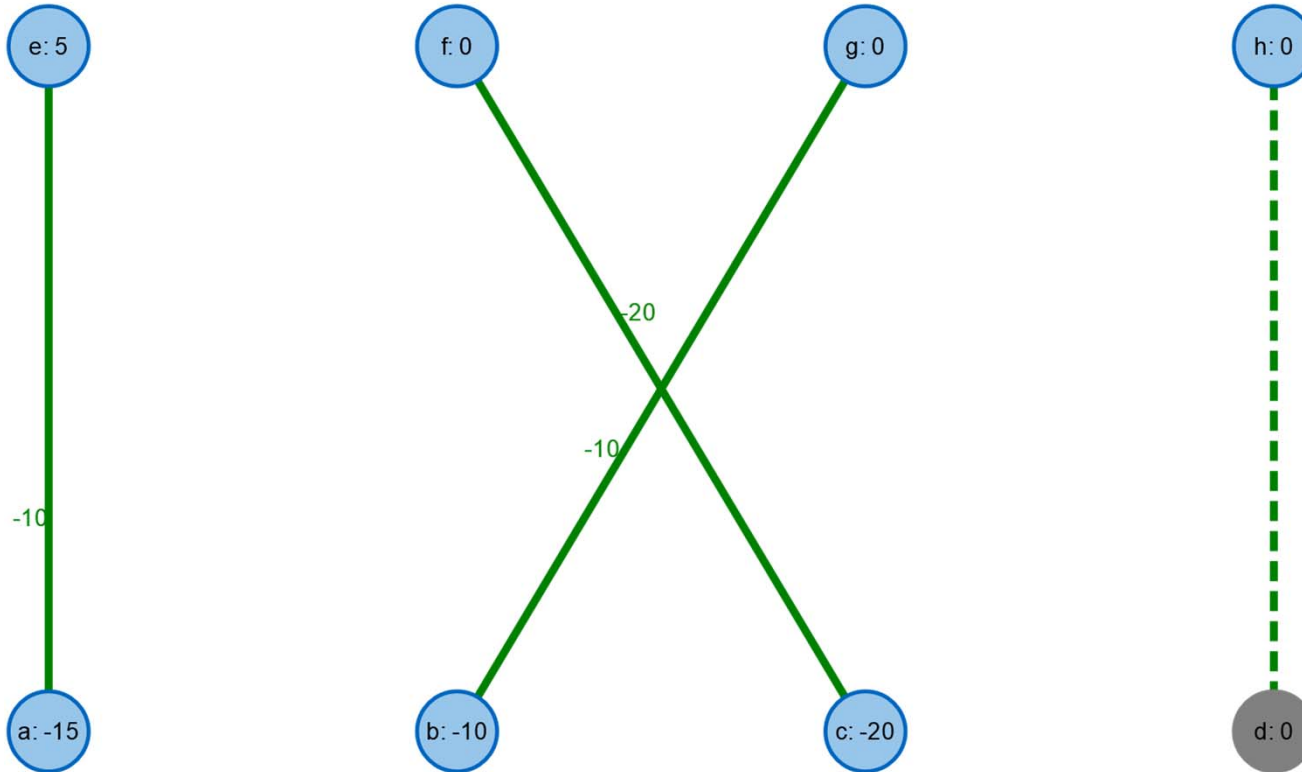
Second matching with cost 10 (-10) too (total 10+10 = 20) (-20)

# UVa 10746 (9)



Second matching with cost 20 (-20) too (total  $10+10+20 = 40$ ) (-40)

# UVa 10746 (10)



The last matching is just a dummy

# Kuhn Munkres's (Hungarian) Algorithm

A good Hungarian algorithm implementation runs in  $O(V^3)$ , thus it is a much better algorithm for **Weighted MCBM** problem

- You are allowed to quote this info verbatim in exam
  - Focus on the modeling of the *complete* weighted bipartite graph
  - And on whether it is a maximization or a minimization problem
- References:
  - <https://github.com/jaehyunp/stanfordacm/blob/master/code/MinCostMatching.cc>
  - [http://e-maxx.ru/algo/assignment\\_hungary](http://e-maxx.ru/algo/assignment_hungary)
  - <https://brilliant.org/wiki/hungarian-matching/>

Solution:

Edmonds's Matching Algorithm (Overview only)

DP with Bitmask (Small Graph, discussed later)

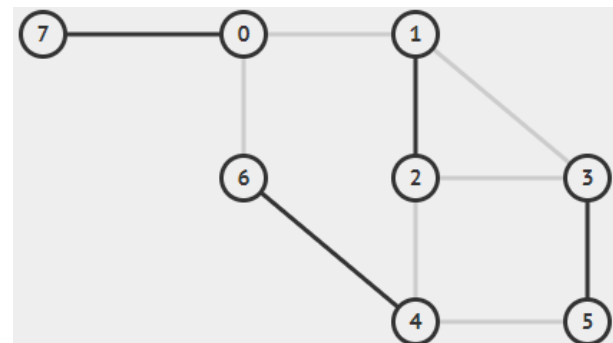
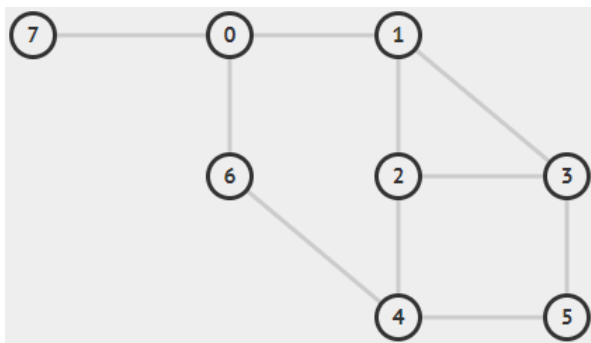
# **UNWEIGHTED MCM**

<https://visualgo.net/en/matching>, select “Unweighted General” tab

# Non-Bipartite Graph and Blossom

A graph is not bipartite if it has at least one odd-length cycle

What is the MCM of this non-bipartite graph?



It is harder to find augmenting path (Berge’s Lemma) in such graph due to alternating cycles called **blossoms**



<https://visualgo.net/en/matching>, select “Unweighted General” tab

# Blossom Shrinking/Expansion

Shrinking these blossoms (recursively)  
will make this problem “easy” again



# When To Use Edmonds's Matching?

This algorithm is a bit hard to implement...

$O(V^3)$  library code is preferred

- Only used for unweighted MCM with  $V \in [22..200^*]$ ; complex code
  - If  $V \leq [19..21]$ , see [this](#)
- Randomized greedy pre-processing is ALSO APPLICABLE here!!
  - Again, you can quote this verbatim for final assessment
  - Focus on the unweighted non-bipartite graph modeling

For code, just use external source

- <http://codeforces.com/blog/entry/49402> (C++)
- [https://sites.google.com/site/indy256/algo/edmonds\\_matching](https://sites.google.com/site/indy256/algo/edmonds_matching) (Java)

Solution(s):

DP with Bitmask (only for small graph)

Edmonds' Matching (future work...)

# **WEIGHTED MCM**



# No? Choice... (only for $V \leq [19..21]$ )

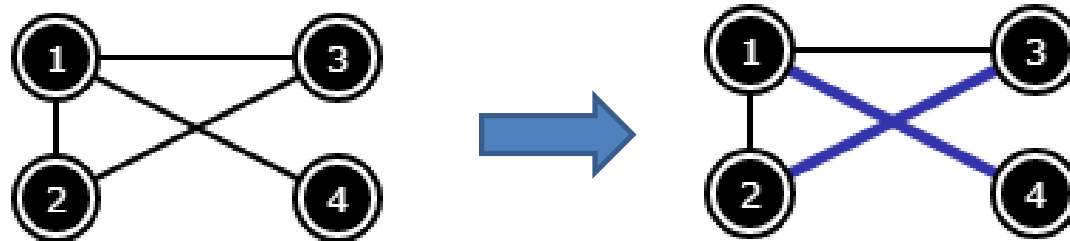
```
ii wMCM(int mask) {
    if (mask == (1<<N)-1) return ii(0, 0);
    if (memo[mask] != ii(-1, -1)) return memo[mask];
    int p1, p2;
    for (p1 = 0; p1 < N; p1++) if (!(mask & (1<<p1))) break;
    ii ans = wMCM(mask | (1<<p1)); // p1 unmatched
    for (p2 = p1+1; p2 < N; p2++)
        if (!(mask & (1<<p2)) && cost[p1][p2]) {
            ii nxt = wMCM(mask | (1<<p1) | (1<<p2)); // match p1-p2
            nxt.first += 2; nxt.second += cost[p1][p2];
            if ((nxt.first > ans.first) || // equal # matching
                ((nxt.first == ans.first) &&
                 (nxt.second < ans.second))) // or smaller cost
                ans = nxt;
        }
    return memo[mask] = ans;
}
```

<https://visualgo.net/en/recursion>, select the “Matching” example...

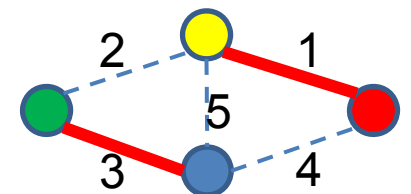
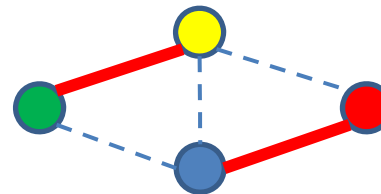
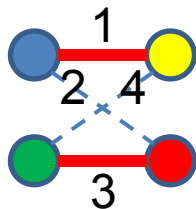
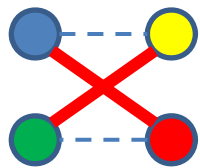
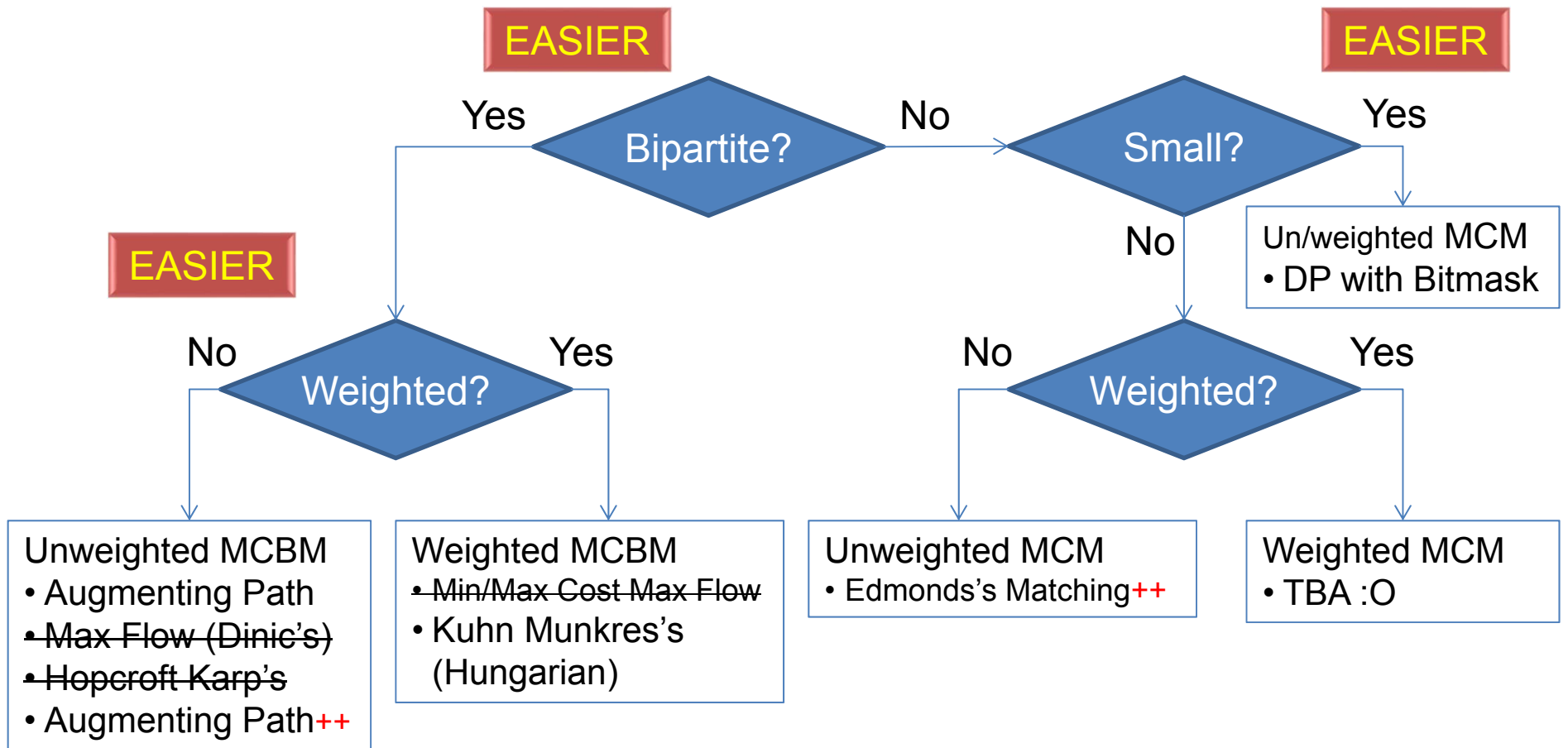


# Also for Small Unweighted MCM Too

Just set all weights = 1 in the previous code



# Summary (so far...)



# References

- Mostly CP3.17b, Section 4.7, 4.8.4, 9.9 (still empty, being updated), 9.13, 9.17 (still empty, being updated), and 9.21 😊
- TopCoder PrimePairs, RookAttack solution
- <http://www.comp.nus.edu.sg/~cs6234/2009/Lectures/Match-sl-PC.pdf> (Prof LeongHW's/P Karras slides)
- More matching exercises/applications during T07