

New Approximation Results for Resource Replication Problems

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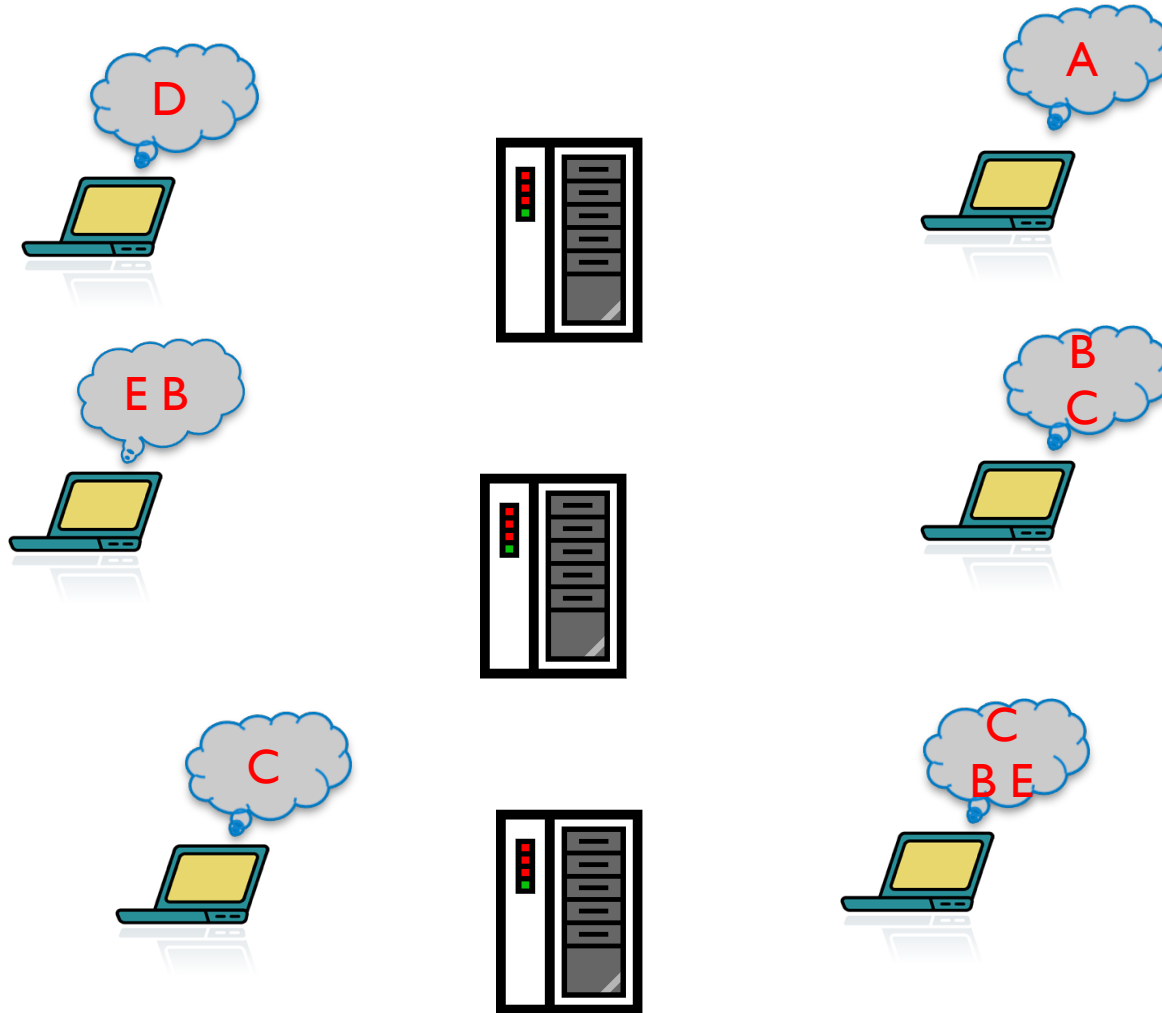
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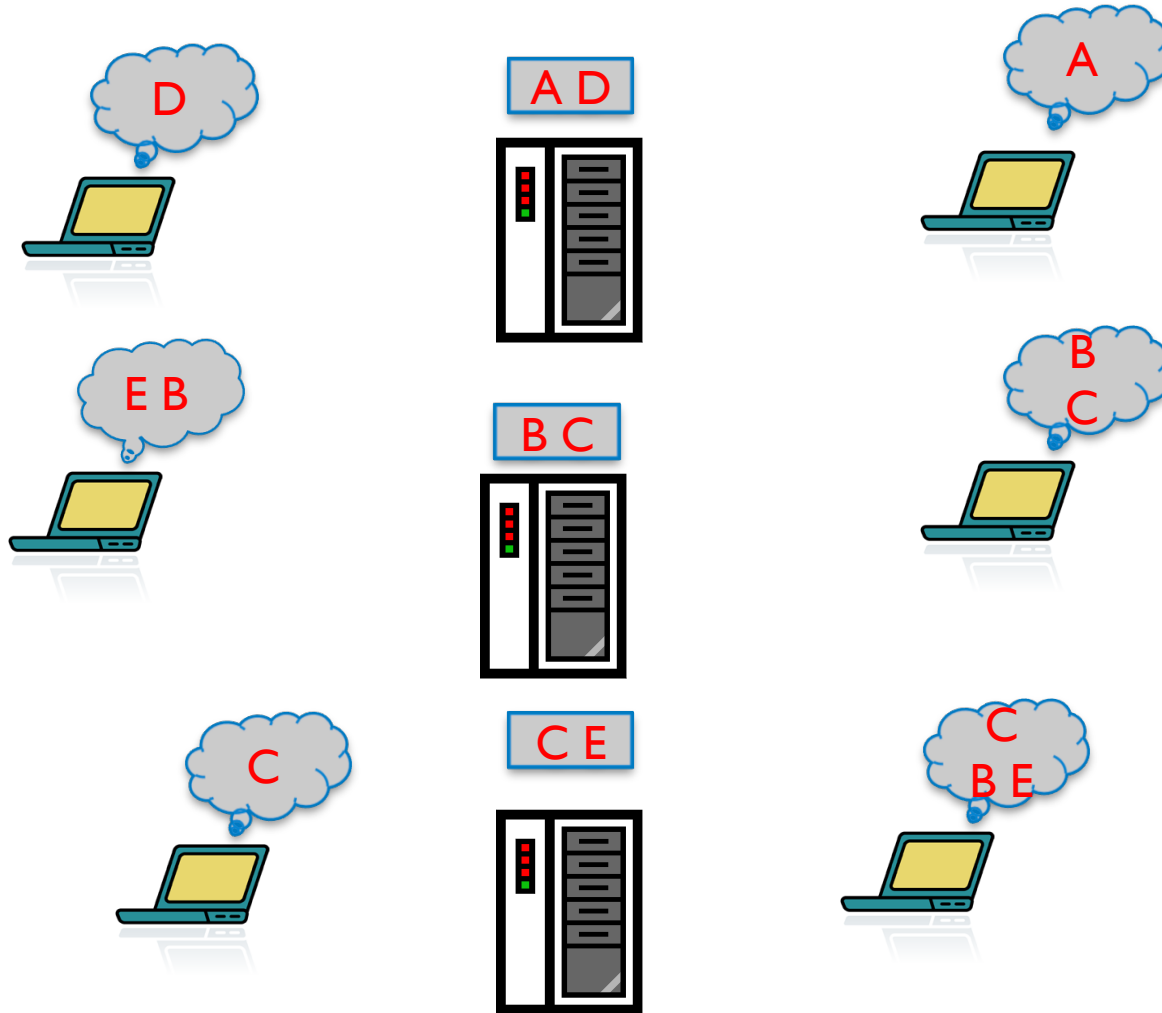


APPROX 2012

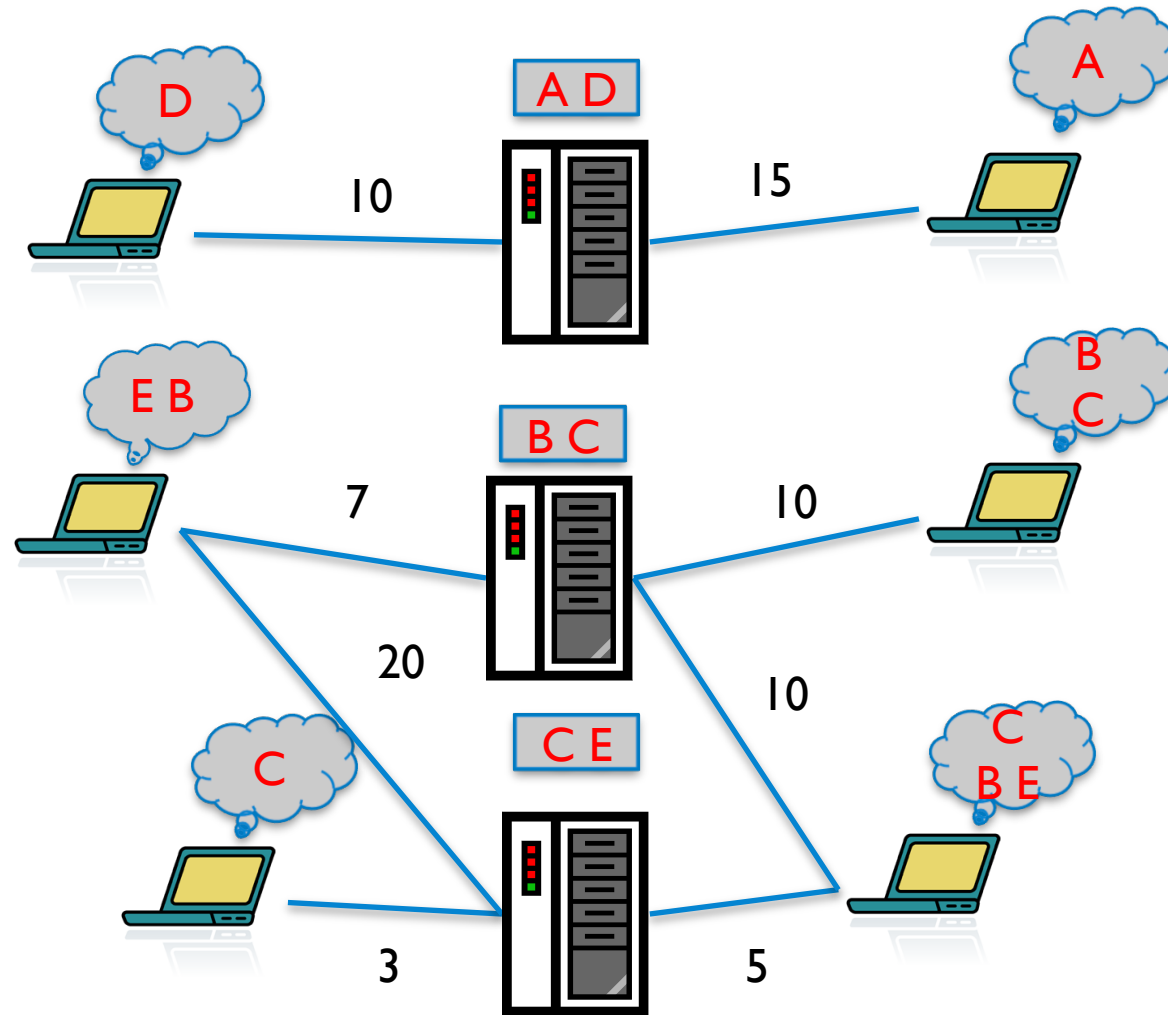
Resource Replication Problem



Resource Replication Problem

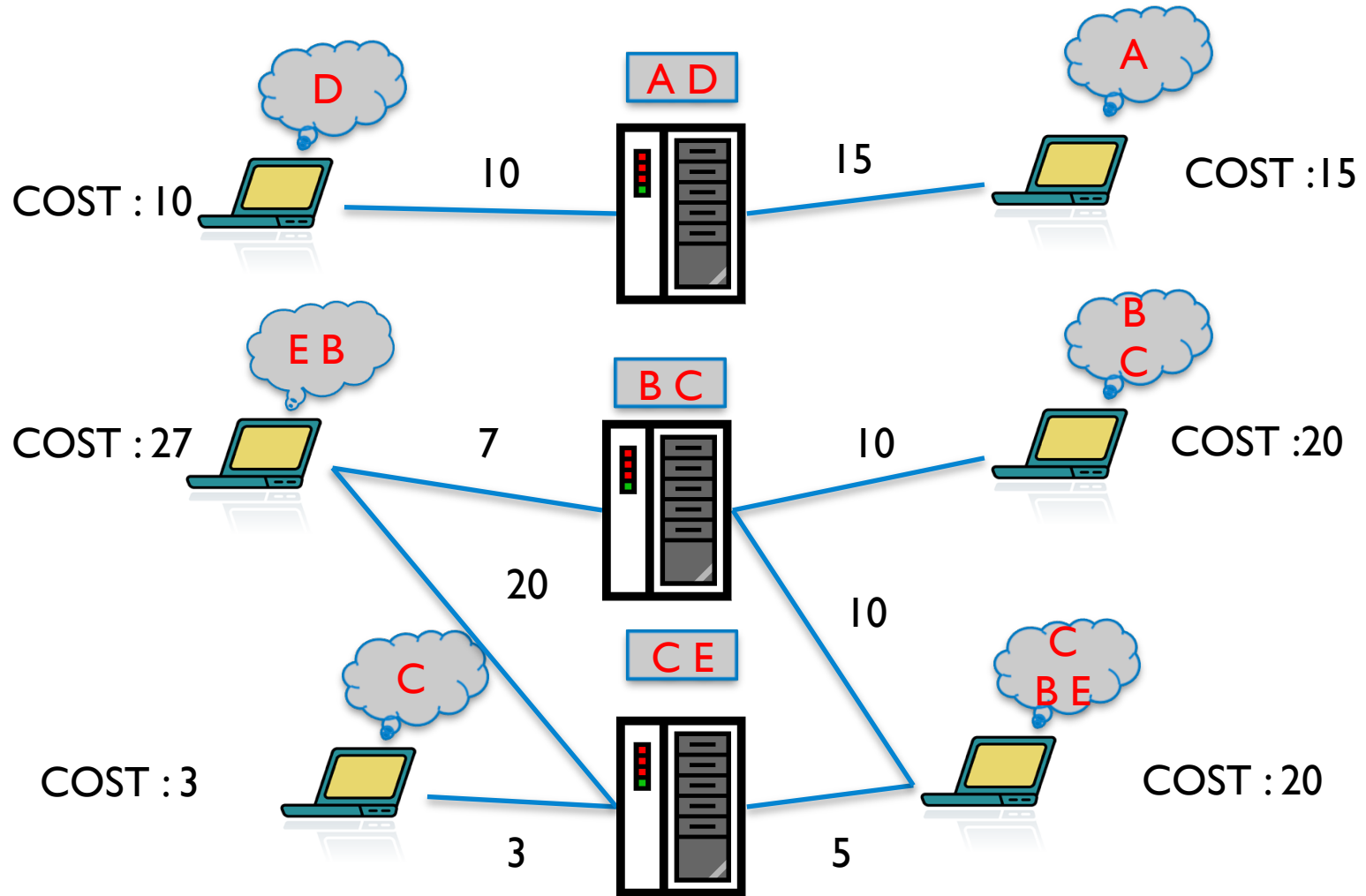


Resource Replication Problem



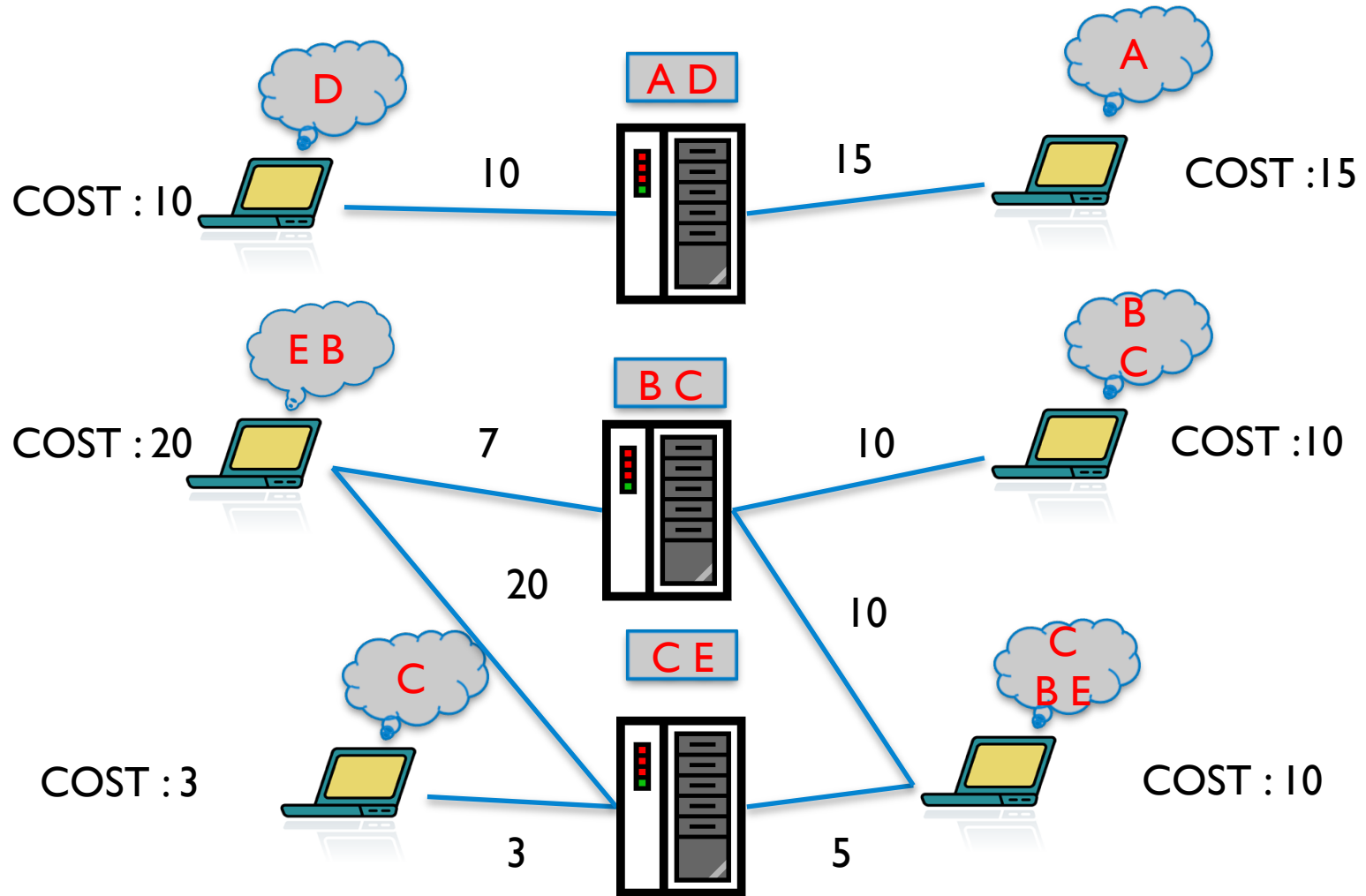
Min Sum Objective

MIN SUM
COST: 95



Min Max Objective

MIN MAX
COST: 20





Objectives

- **Min Sum**
 - An LP based IO approximation by Baev, Rajaraman and Swamy[SODA 2001, SIAM J. Compt. 2008].
 - Not covered in this talk
- **Min Max distance**
 - Studied by Ko and Rubenstein[2003,2004].
 - This is the subject of current talk.

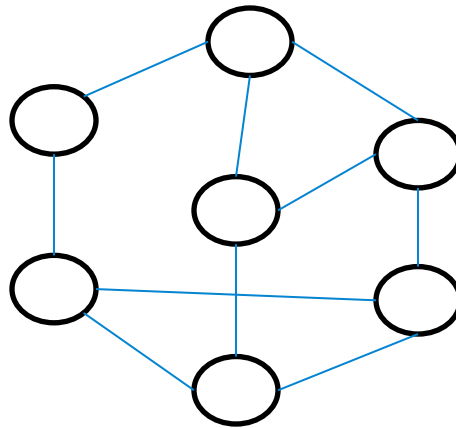


Motivation: Resource Replication

- Need of data replication arises in many contexts
 - File replica distribution
 - Reduce retrieval time: improves data locality
 - Fault tolerance
 - Distributed computation
 - Distribution of loads: improves latency
 - Channel allocation in wireless networks
 - Channel capacity can improve if nodes assigned to the same channel are far apart: reduce interference

Resource Replication

- Basic Problem:
 - Given K resources, N servers to host resources. Each client needs every resource.

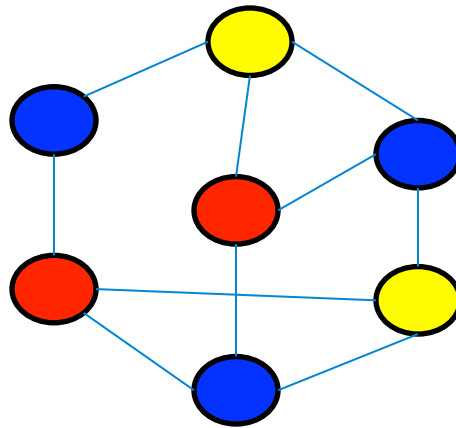


Edge denotes distance 1.
Shortest Path Metric.

- Goal: *minimize the maximum distance* each client needs to travel to get all the required resources.

Resource Replication Problem

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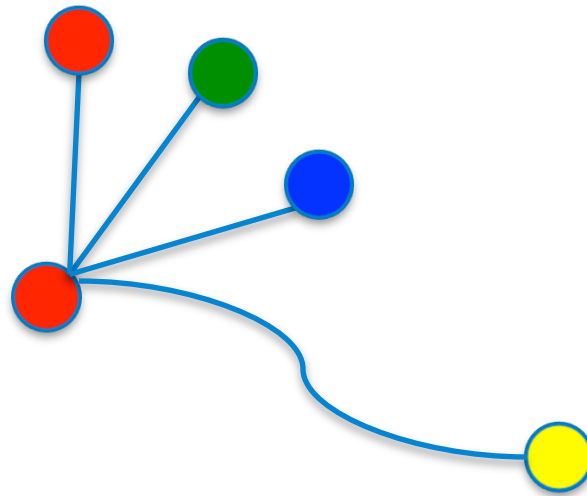


Basic Resource Replication Problem

- Studied by Ko and Rubenstein[ICNP 2003]
 - Distributed 3 approximation using local search
 - Local search may not converge in polynomial time
- We give a very simple 3 approximation which can be implemented in a distributed scenario as well.
 - Idea: Extending threshold graph technique for k center type algorithms

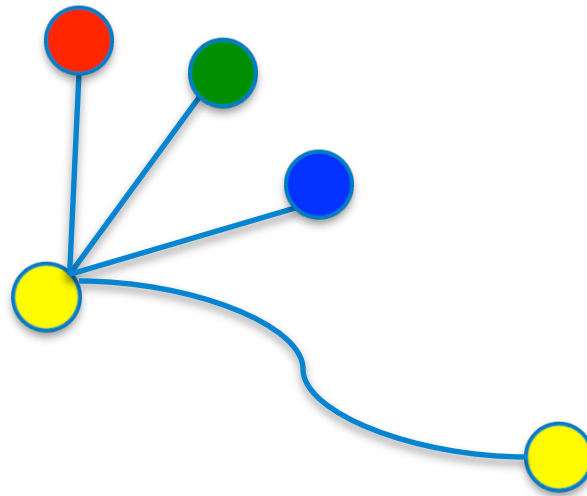
Ko and Rubenstein Algorithm

- Local Search Step-



Ko and Rubenstein Algorithm

- Local Search Step-



- Guaranteed to give a 3 approximation.
Not polynomial time.

Our Results.

Problem Variant	Approx. Achieved	Hardness of Approximation	Description
Basic Resource Replication (BRR) Problem	3 approx.	No $2 - \epsilon$ approx. possible	Every node requires every resource. Each node can store exactly one resource type.
Subset Resource Replication (SRR) Problem	3 approx.	No $3 - \epsilon$ approx. possible	Nodes require arbitrary subset of resource types. Each node has designated capacity.
BRR with Outliers	3 approx.	No $2 - \epsilon$ approx. possible	Same as BRR. Need to satisfy at least M nodes.

Our Results (Contd.)

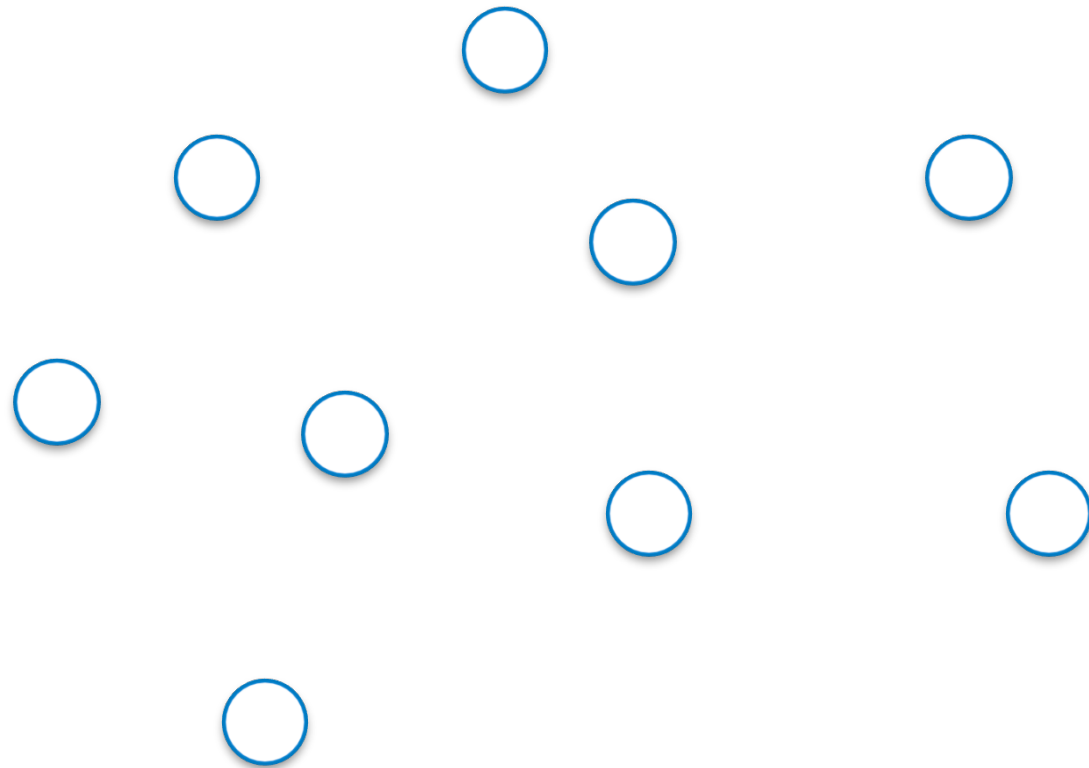
Problem Variant	Approx. Achieved	Hardness of Approximation	Description
SRR with Outliers	None	NP-Hard to approximate within any non-trivial approx..	Same as SRR. Need to satisfy at least M nodes.
BRR with Load Balancing	4 approx. with a violation of load capacity of factor 2.	No $2 - \epsilon$ approx.	Same as BRR. Each server can serve at most k clients.



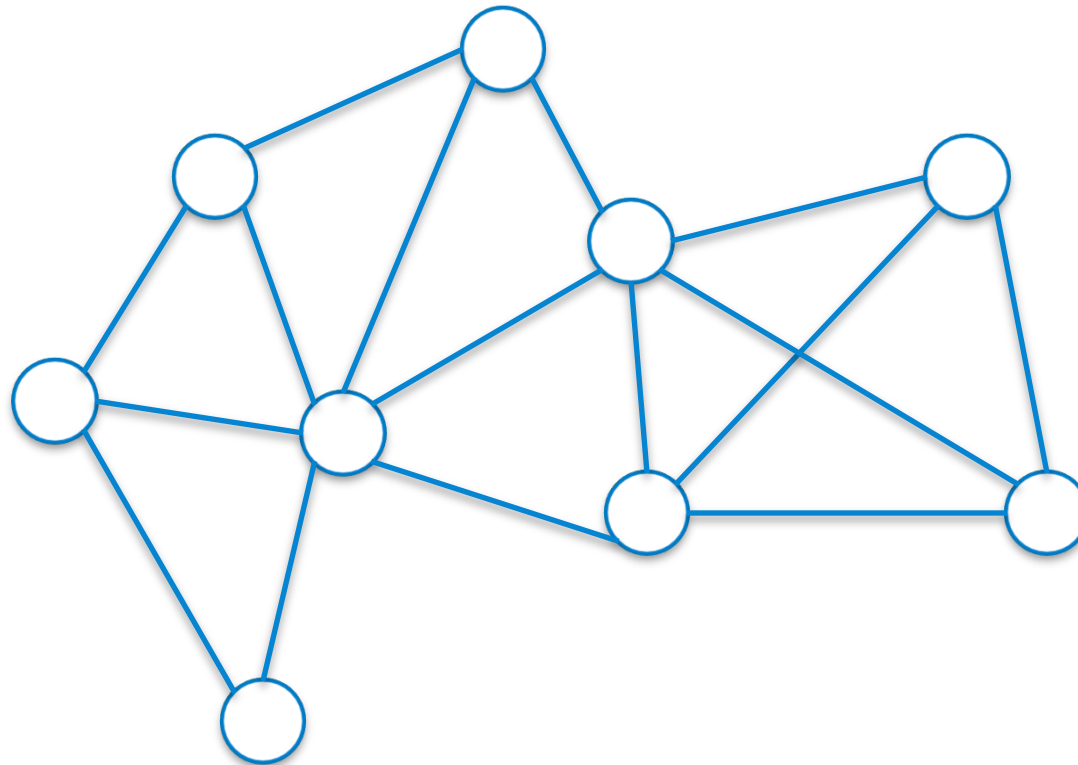
Our Algorithm

- Based on the threshold graph technique
 - Similar to the k-center.
- 3 approx. in polynomial time.
- Can be implemented in a distributed fashion.

Graph G and resources {R, G, B}

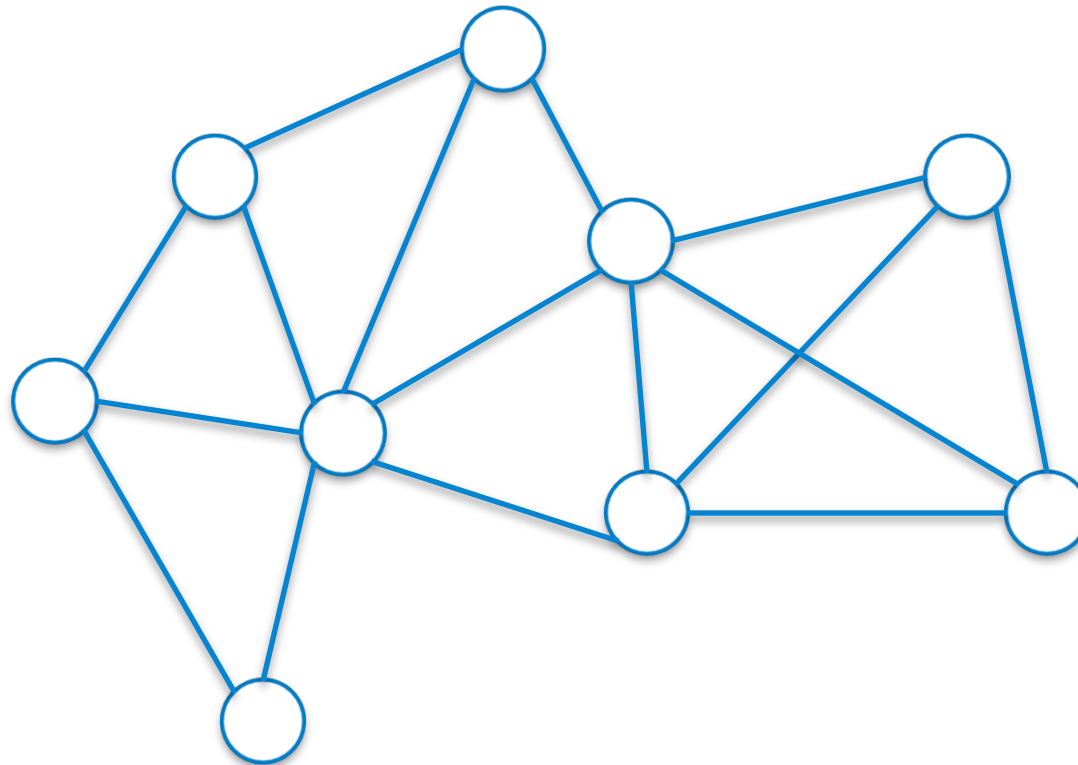


Guess the optimal distance δ and
Construct **threshold graph**.

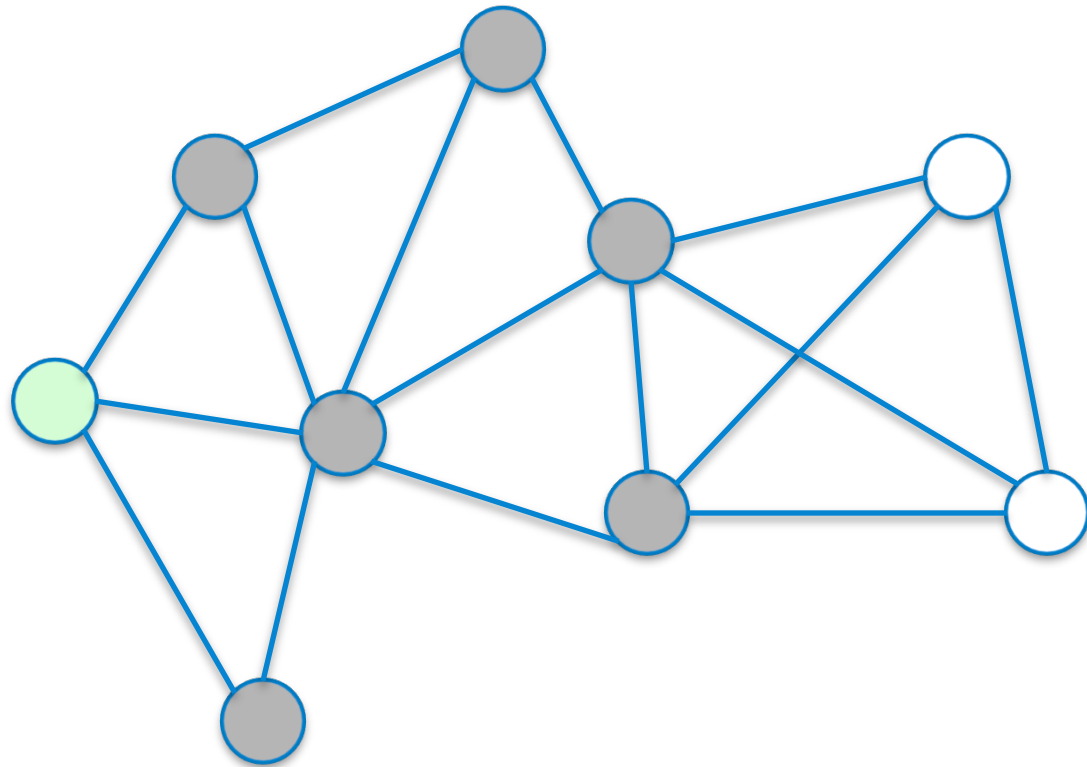


————— : Distance $\leq \delta$

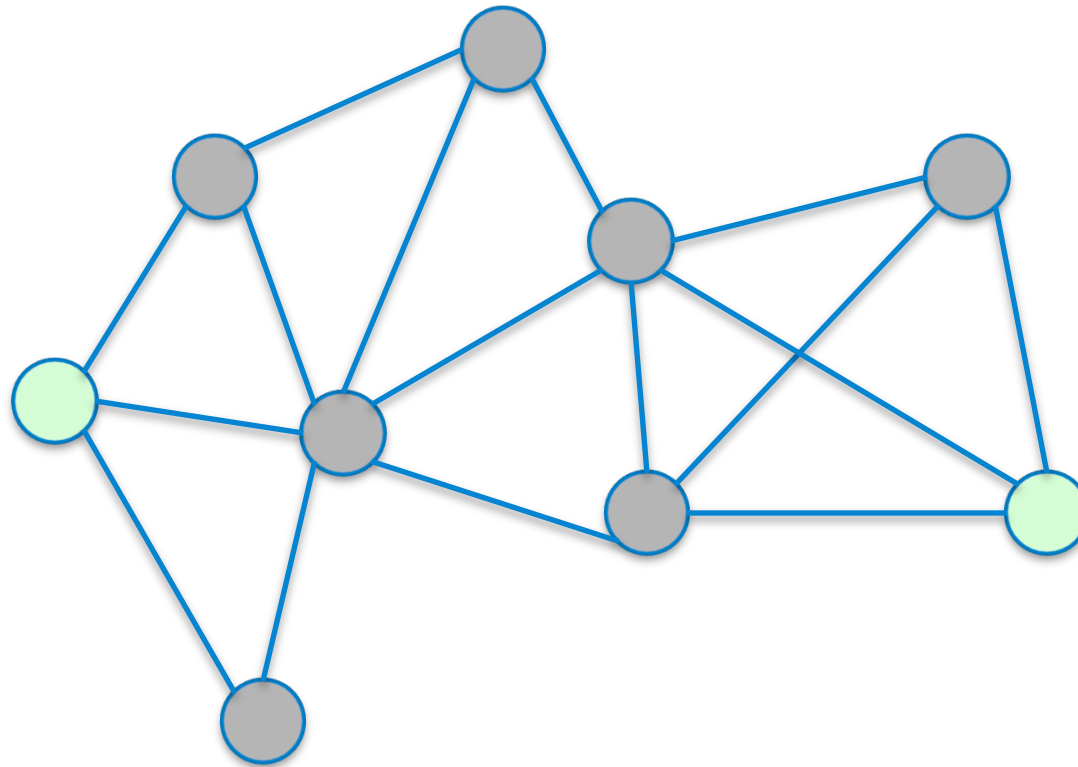
- Keep picking nodes and delete nodes within two hops.



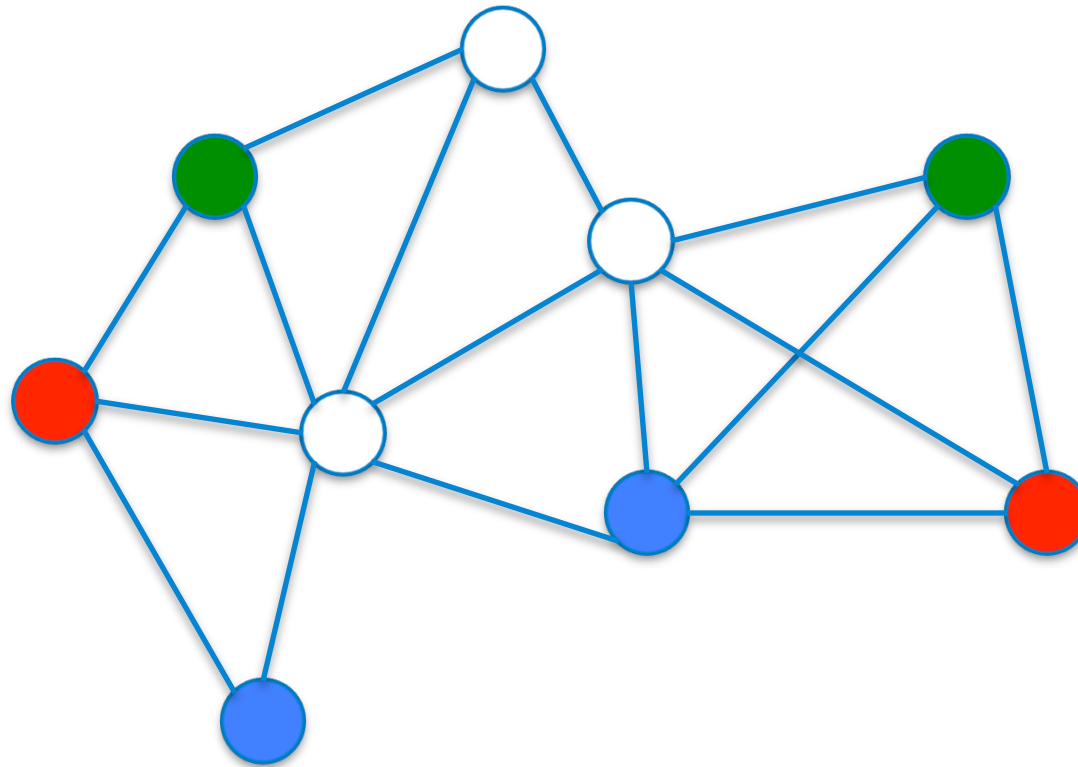
- Keep picking nodes and delete nodes within 2 hops. Call this set MIS.



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- Place k resources in each vertex of MIS's delta neighborhood.





Why does it work ?

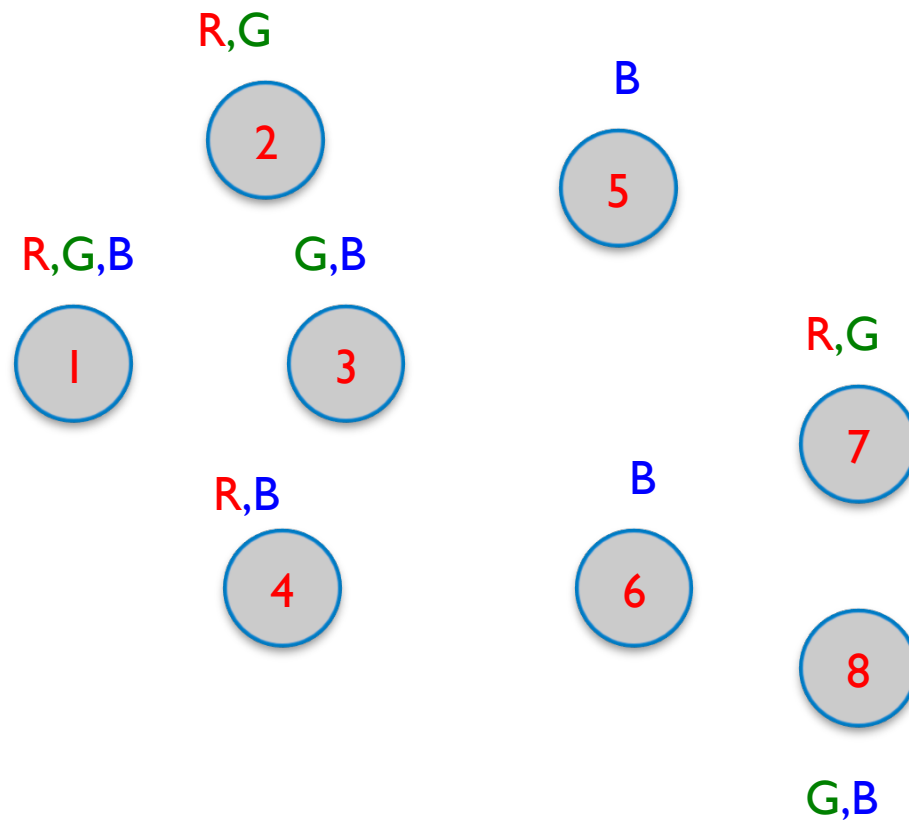
- Every vertex is within 2 hops of MIS.
- Each vertex in MIS has degree at least $k-1$.
- Placing colors on MIS and neighbors makes sure that every vertex has every color within 3 hops.
- Hence, we obtain a 3 approximation.



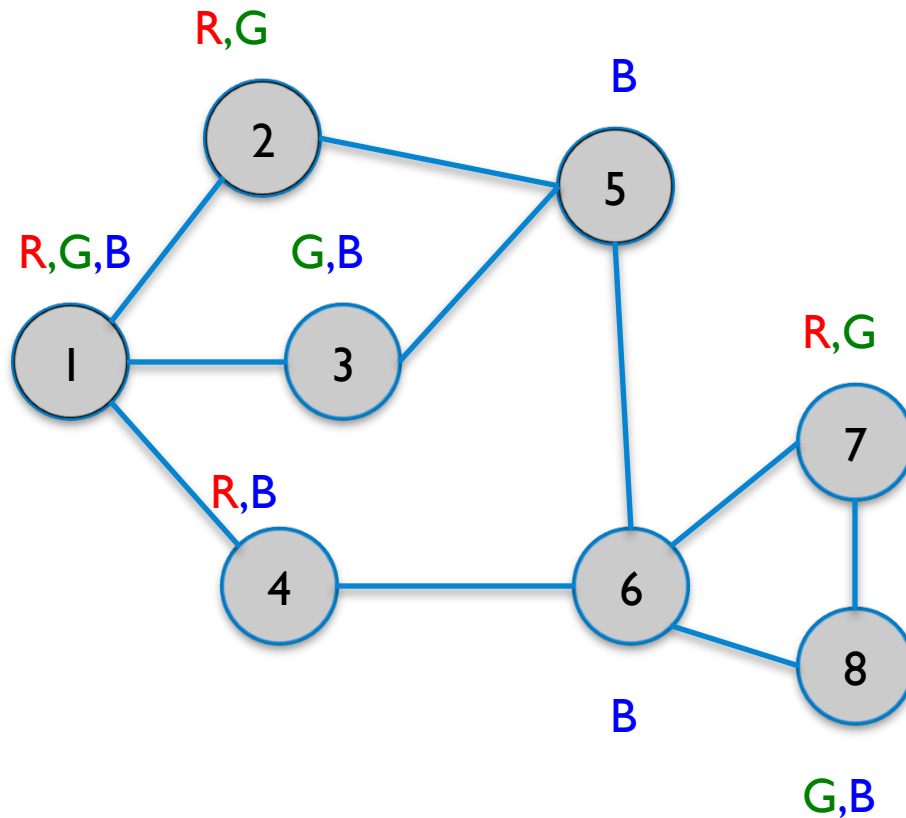
Subset Resource Replication Problem

- Want ~~all~~ **SUBSET** of colors.
- Can store ~~one~~ s_v colors per server
- Same Min Max objective.
- Also studied by Ko and Rubenstein[2004]
 - their heuristic does not provide any approximation guarantee.

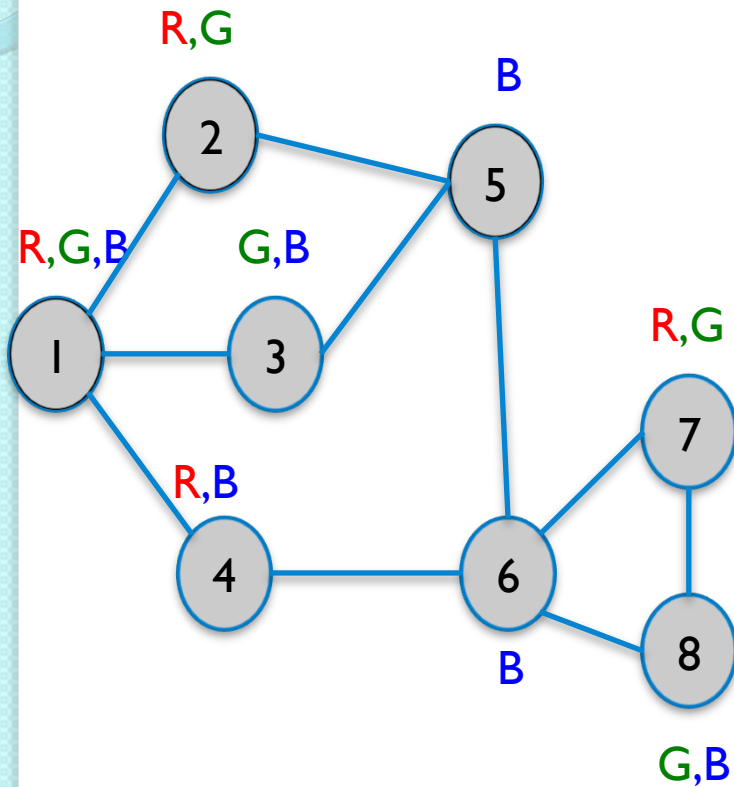
Given G and resources {R,G,B}



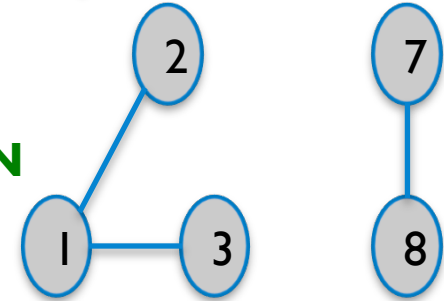
Guess $\delta \rightarrow$ threshold graph



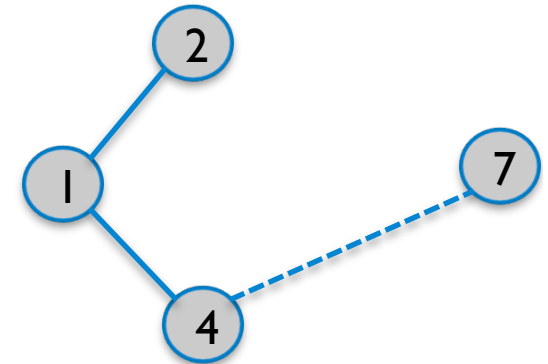
Decompose into sub-graphs.



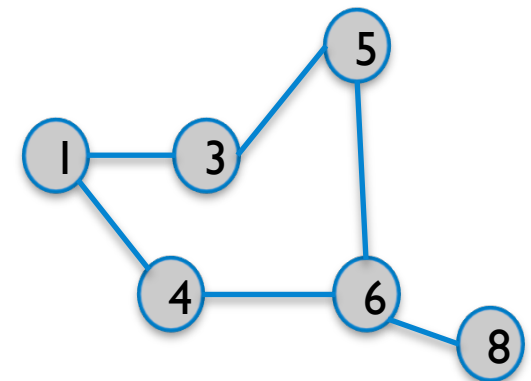
GREEN



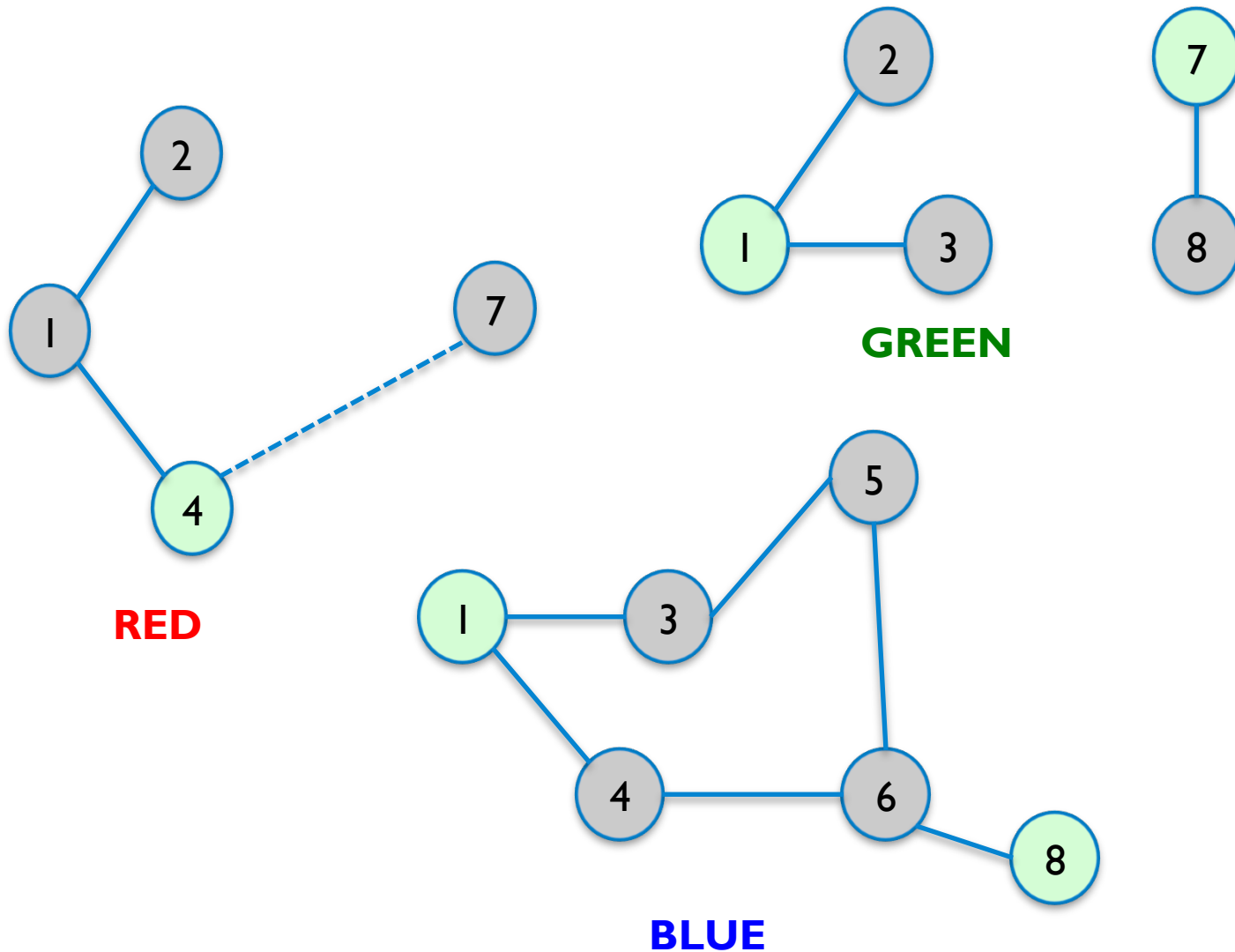
RED



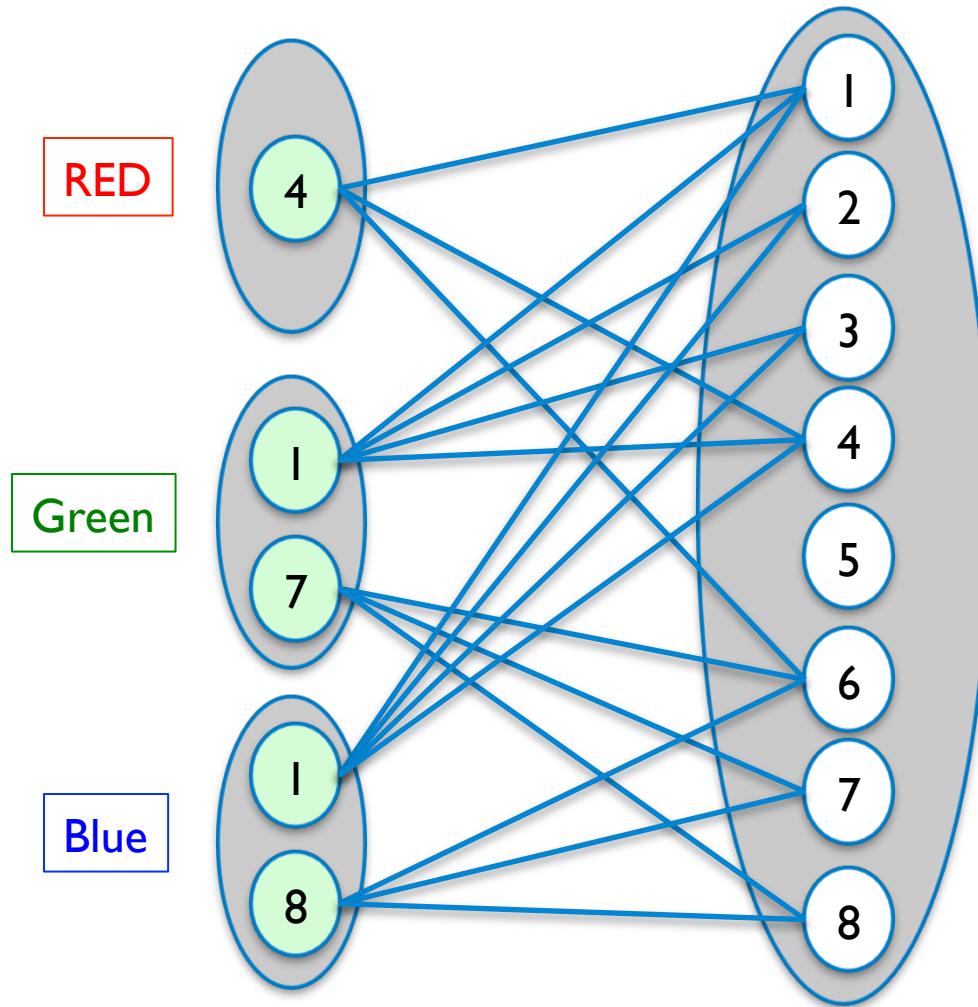
BLUE



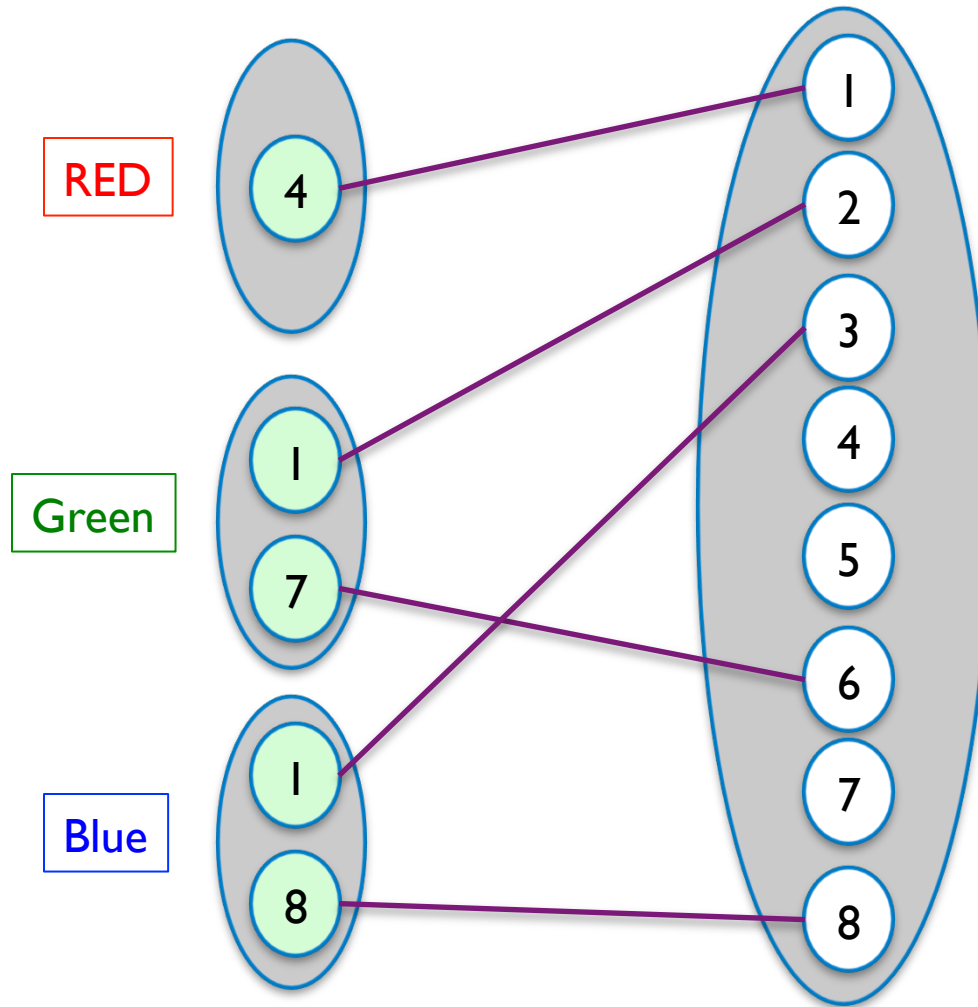
Compute **MIS** for each color graph



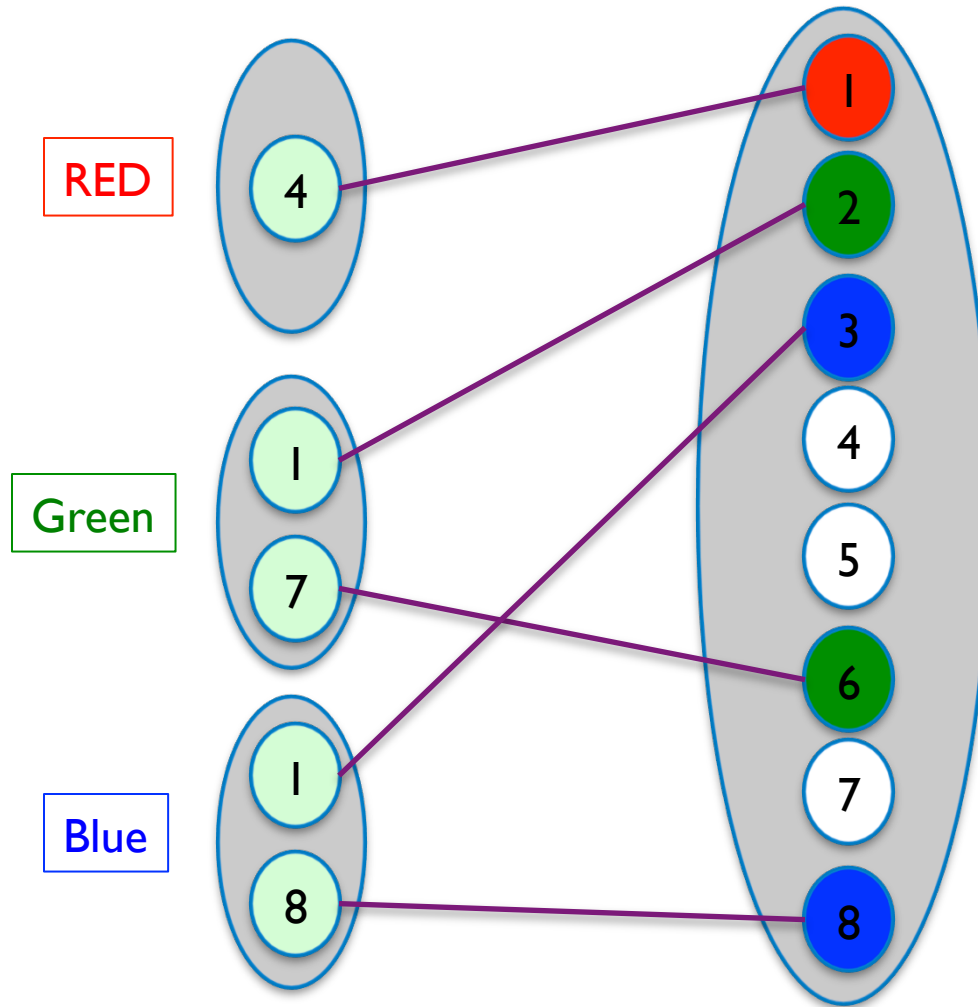
Compute a matching



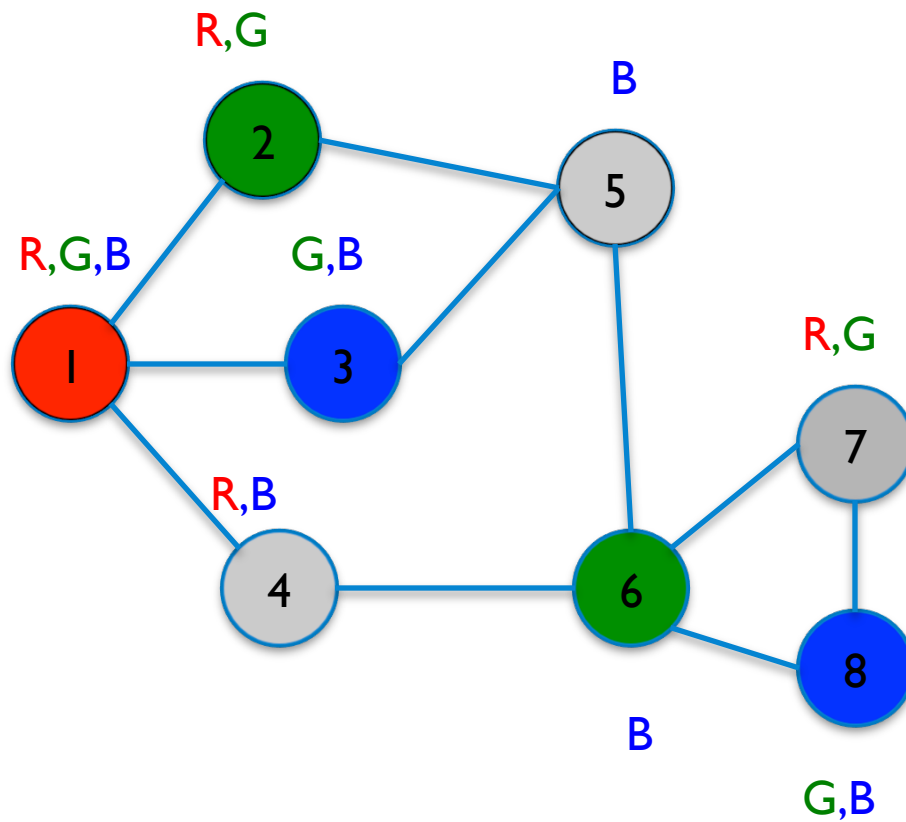
Compute a matching



Compute a matching



Assigning the colors.





Subset Resource Replication Problem

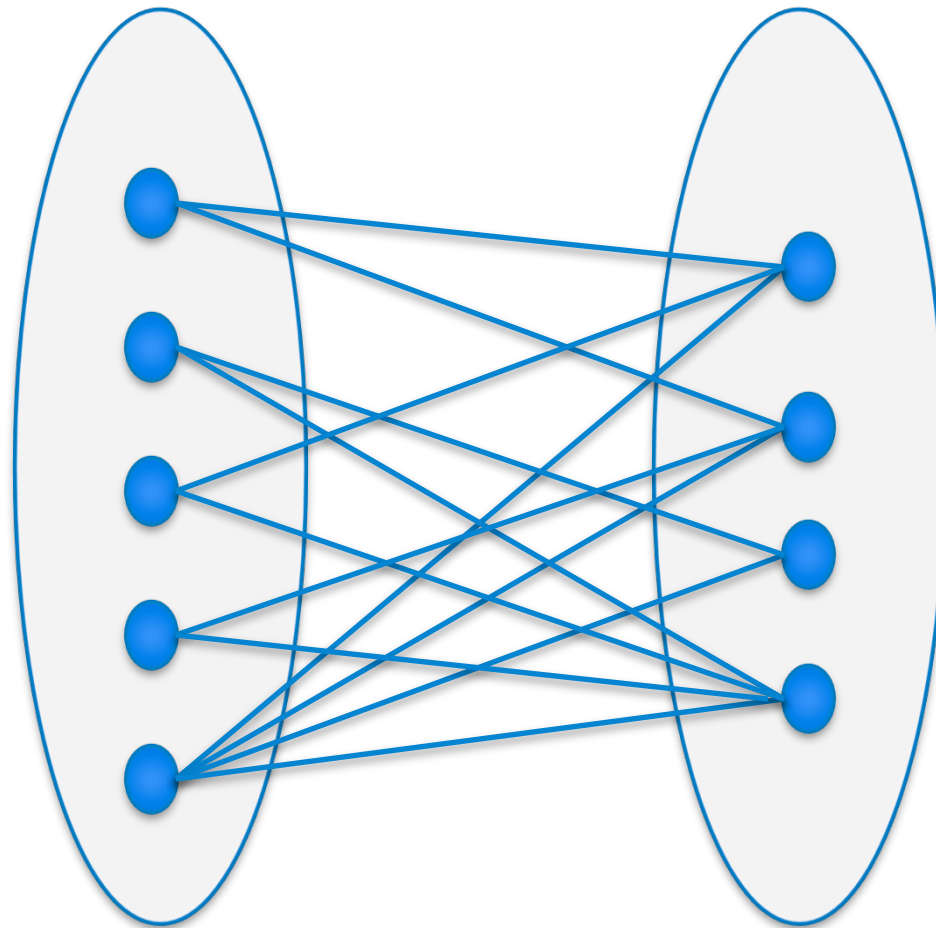
- *Theorem 1: There exists a polynomial time 3-approximation algorithm for the Subset Resource Replication problem.*



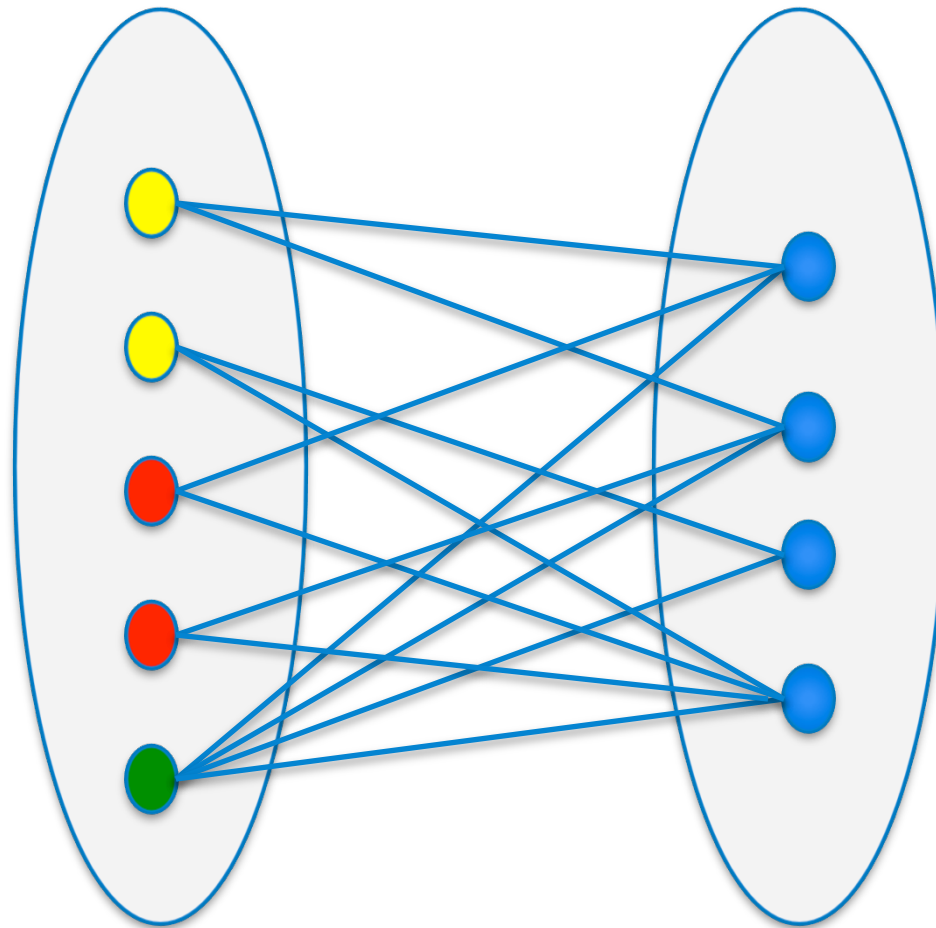
(3- ϵ) Hardness of Approximation

- Reduction from One Sided Domatic Number Problem (Feige, Halldorsson, Kortsarz and Srinivasan).
- Similar to the dominating set based reduction for the k-center problem.
- A similar reduction from Domatic Number Problem gives a (2- ϵ) hardness of approximation for Basic Resource Replication Problem.

One Sided Domatic Number.

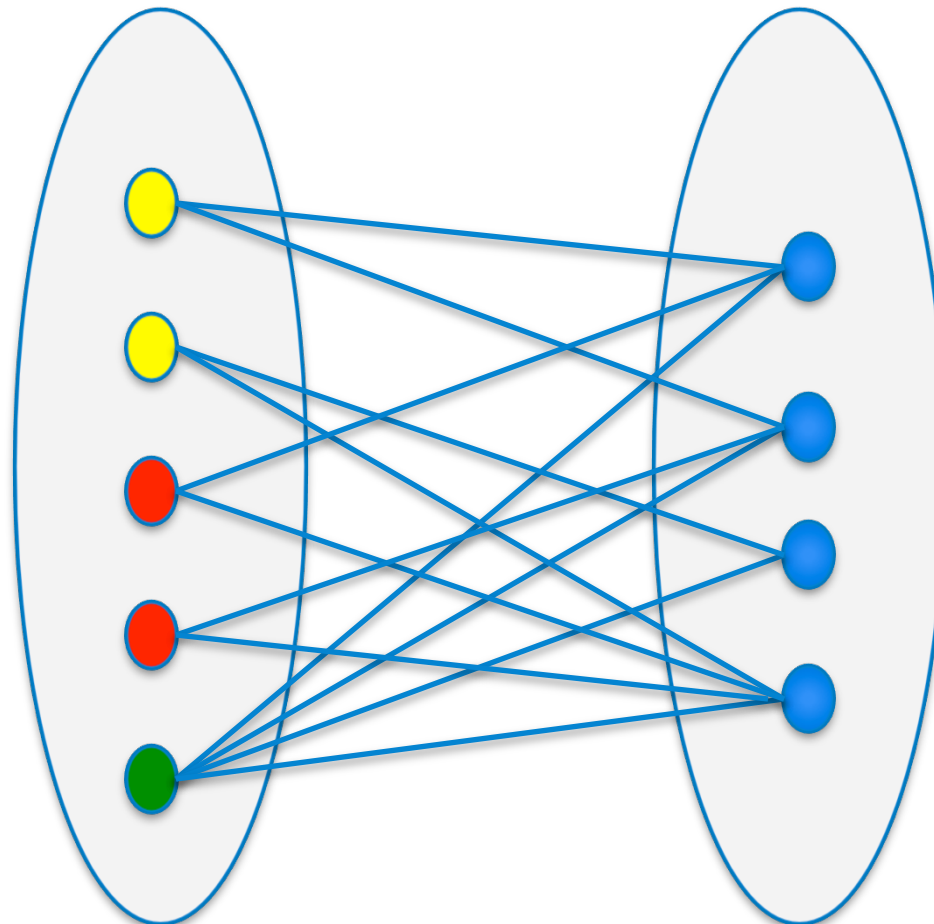


One Sided Domatic Number.



Domatic Number = 3

Reduction:



Blue Nodes:
Require all colors.
0 Capacity

Distance Metric:
Each edge – 1
Anti edge – 3

Other Vertices:
Unit capacity.
No resource
required



Outlier Version

- Basic Resource Replication problem
 - 3-approximation
- Basic Resource Replication problem with bounds on number of replicas of each resource
 - 5-approximation
- Subset Resource Replication problem
 - No nontrivial approximation factor unless $P=NP$

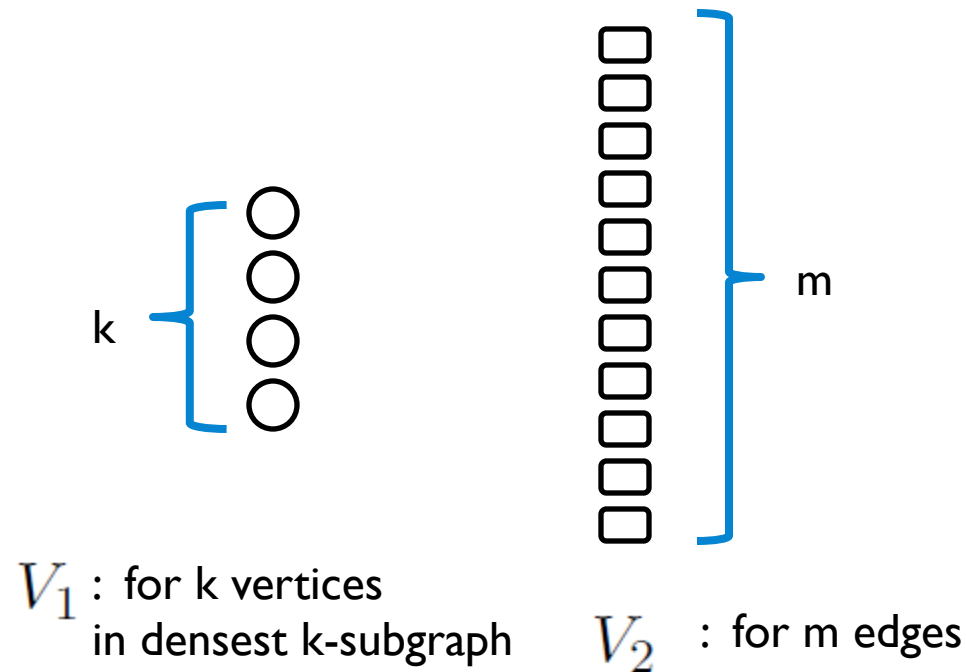


Outlier Version: Robust Subset Resource Replication Problem (RSRR)

- *Theorem : Assuming $P \neq NP$, there is no polynomial time algorithm which gives a positive approximation ratio for Robust Resource Replication Problem.*
- A polynomial time reduction of the densest k-subgraph problem to the problem of deciding the “feasibility” of RSRR.

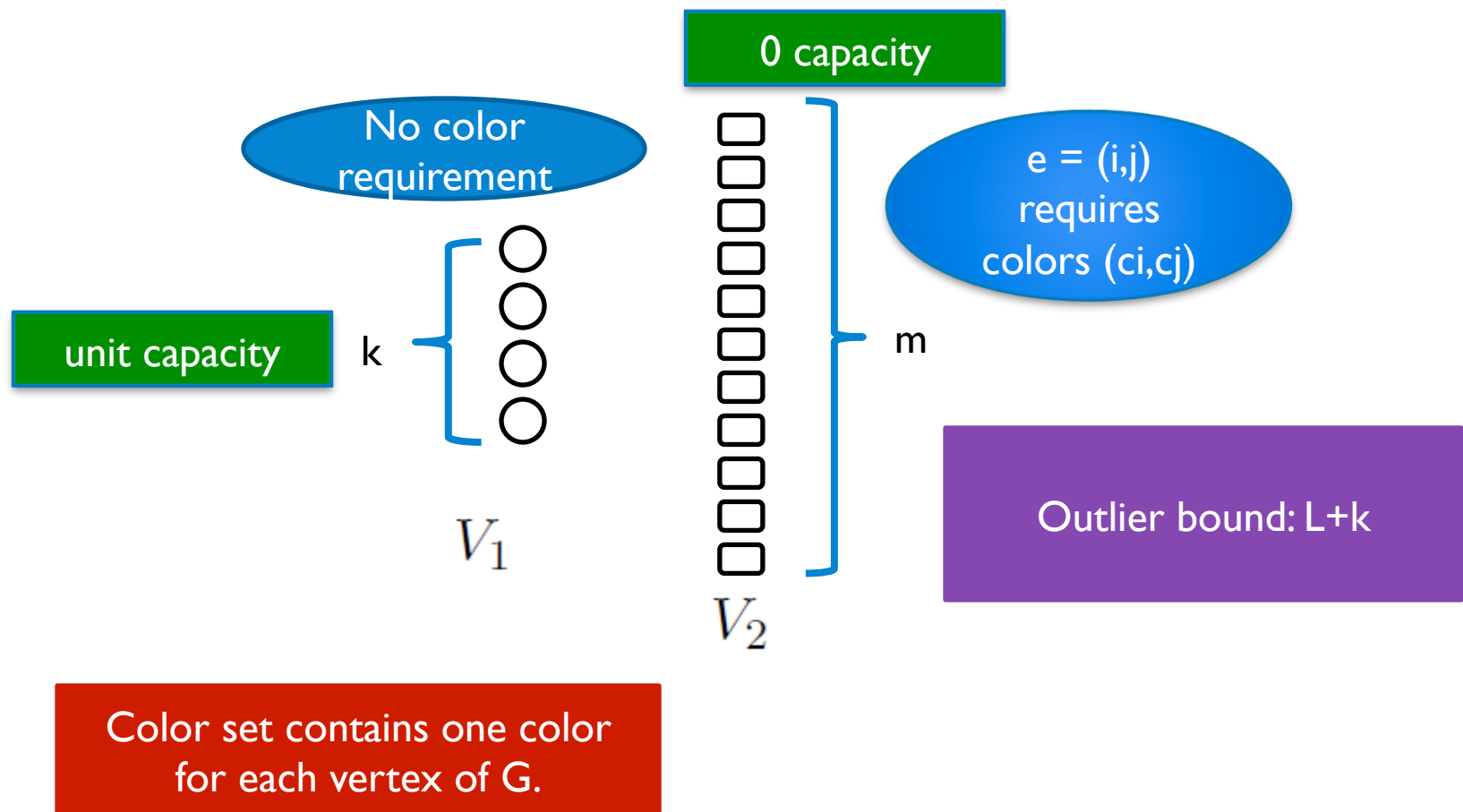
Robust Subset Resource Replication Problem (RSRR)

- Decision version of densest k -sub-graph problem:
 - Instance $\mathcal{I} = (G = (V, E), k, L)$, $|V| = n$, $|E| = m$, decide if in G , there exists a sub-graph of size k containing L edges.
- Construct an instance of RSRR as follows:



Robust Subset Resource Replication Problem (RSRR)

- Construct an instance of RSRR as follows:





Future Direction

- Studying other objectives such as Min Sum with bound on number of replicas/ cost for replication
- Extending Load constraint to Subset Resource Replication problem/ Matching lower and upper bound for Basic Resource Replication Problem etc.

Questions ?