

# On the Load Balancing Problem

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joint work with

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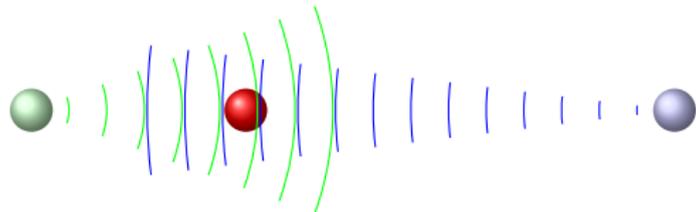
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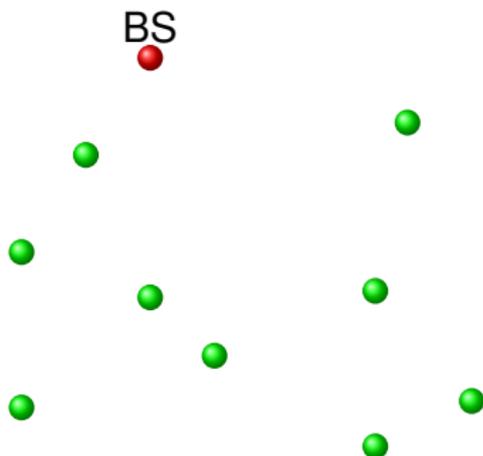
# Wireless sensor networks (WSN)

WSN is a special type of **ad-hoc wireless networks** such that its nodes are devices **with** embedded

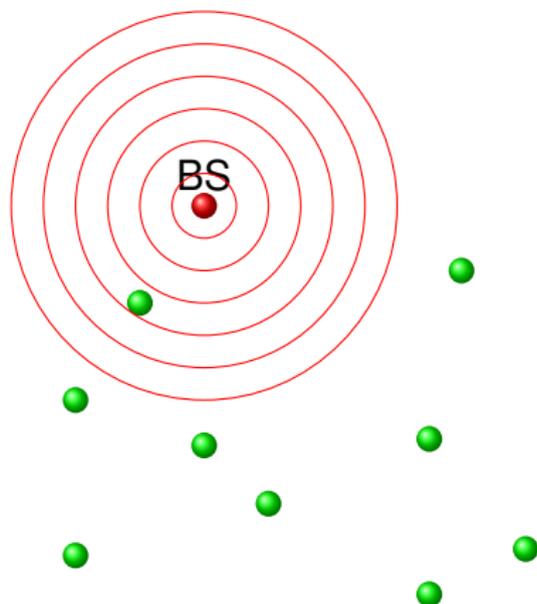
- microcontroller,
- sensor,
- FM radio and
- power source.



# Building WSN



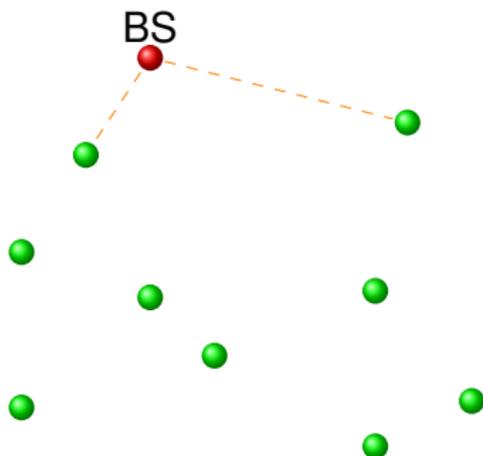
# Building WSN



- 1 Building reachability graph

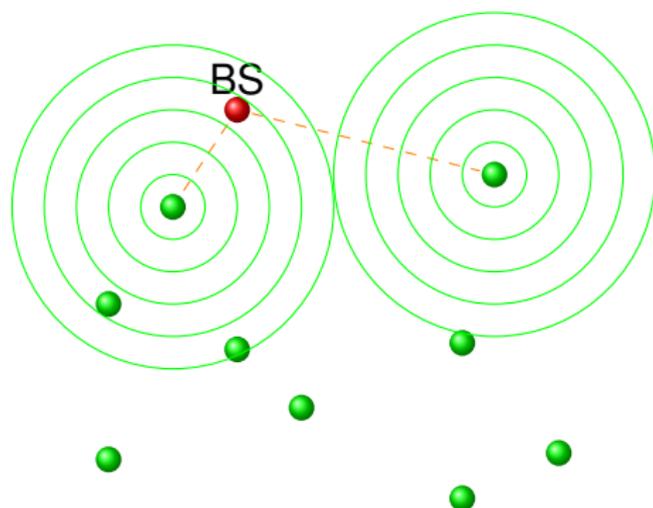
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- 1 Building reachability graph



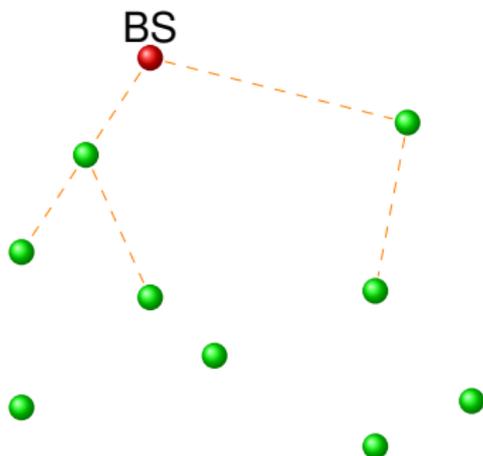
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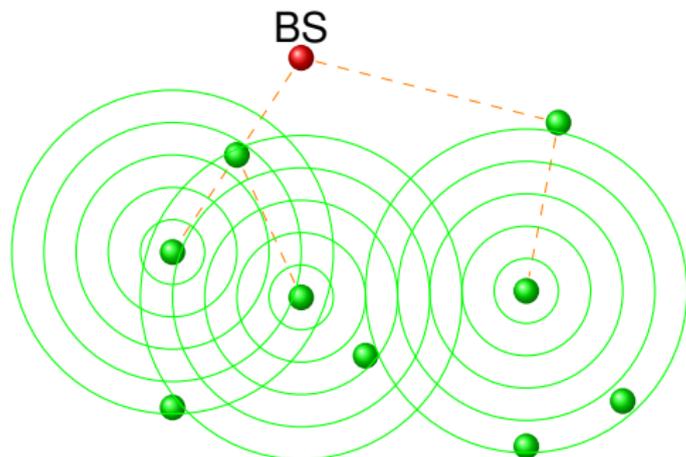
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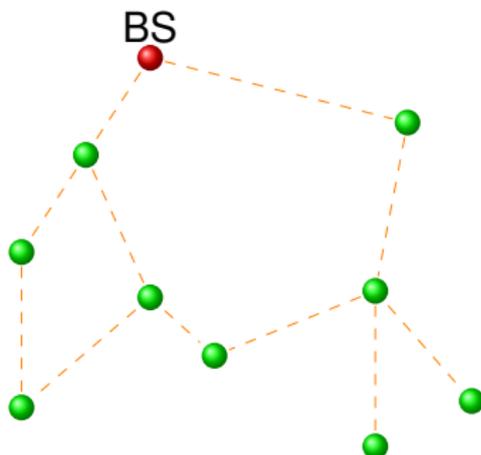
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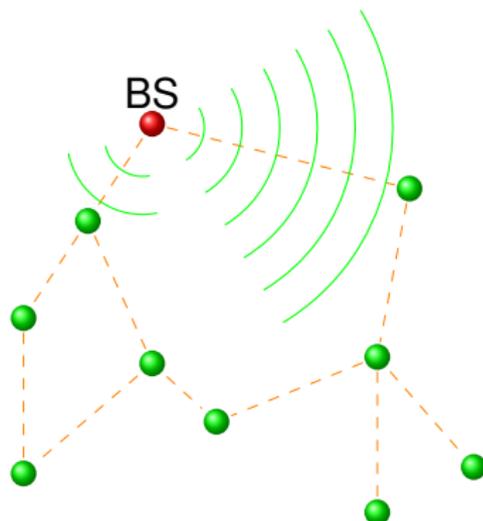


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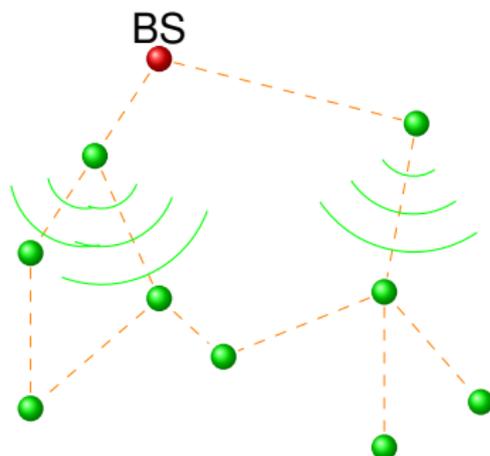


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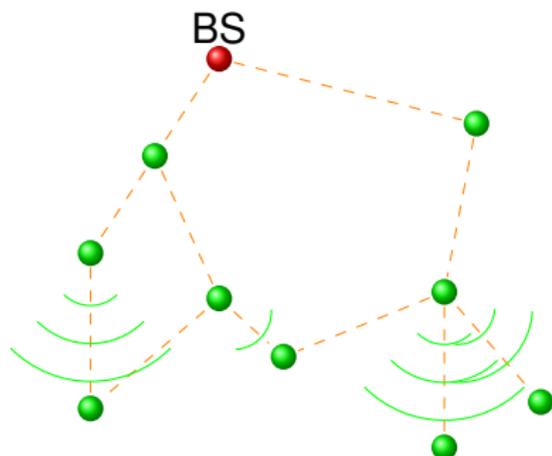
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- 2 Building BFS tree, establishing connections

# Building WSN



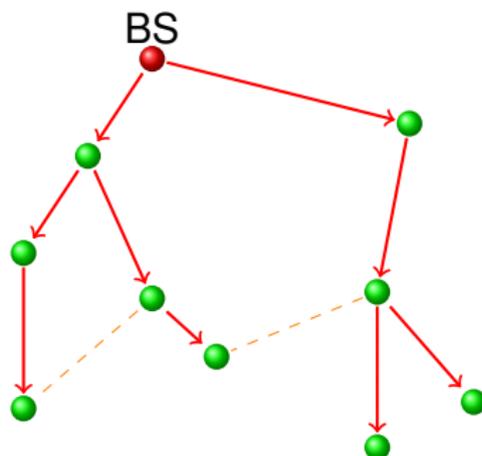
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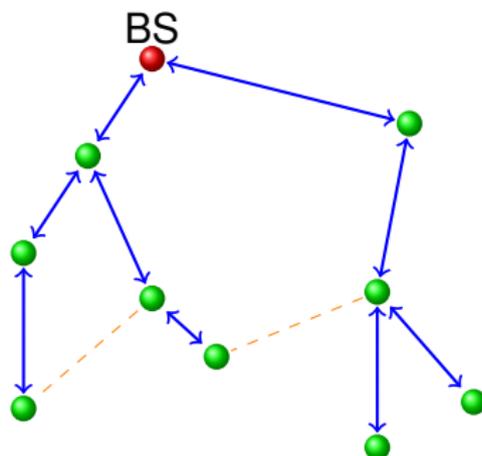
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# Building WSN



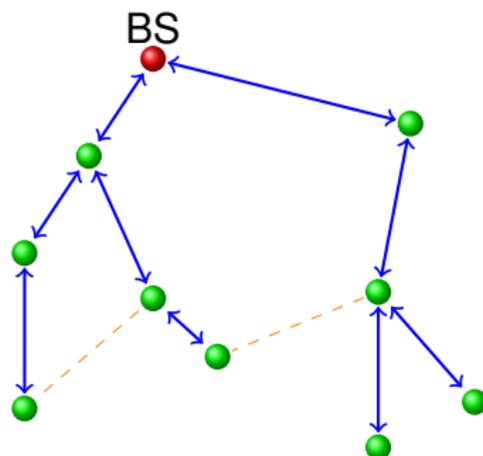
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# Building WSN



- 1 Building reachability graph
- 2 Building BFS tree, establishing connections
- 3 Establishing secure connections

# Building WSN

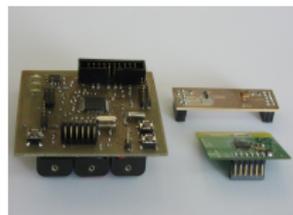
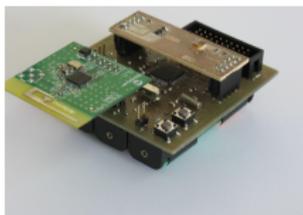


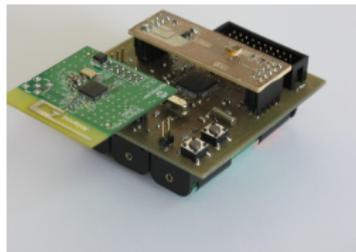
- 1 Building reachability graph
- 2 Building BFS tree, establishing connections
- 3 Establishing secure connections
- 4 Secure communication

# WSN based on CDMA technology

## WSN based on CDMA technology

- consists of node that **are able to communicate each other with respect to their physical limitations and mutual distance**,
- the sink of the network (**base station**) **has relatively large computational capabilities and energy sources**,
- the number of communication channels available at the sensors is **limited** (say 16).





# $k$ -Path Vertex Cover Problem (hereditary)

# Generalised Scheduling Problem (induced hereditary reductions)

# Graph theoretical approach

The limited number of sensor channels leads to the problem of finding BSF-tree with bounded degree.

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## Bad news

The problem of finding spanning tree with bounded degree is NP-complete.

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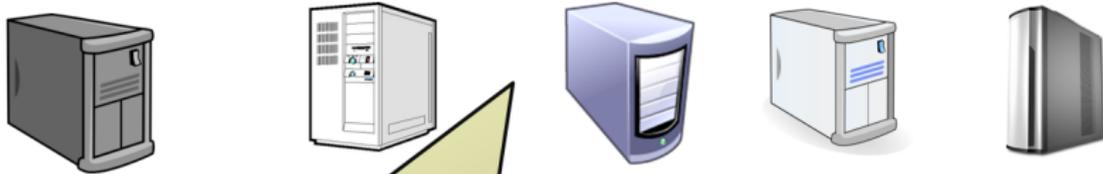
## Bad news

The problem of finding spanning tree with bounded degree is NP-complete.

## Good news

We can restrict our consideration to a bipartite graph that is formed by two layers in BSF-tree.

# Load balancing problem

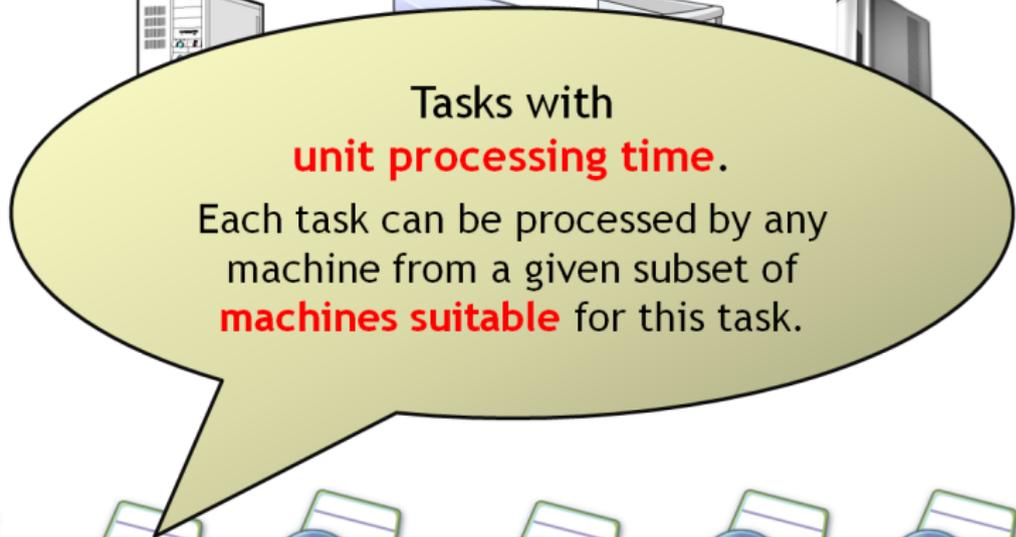
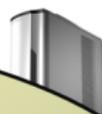


**Set of machines**  
differ in computational  
resources, data  
resources, ...

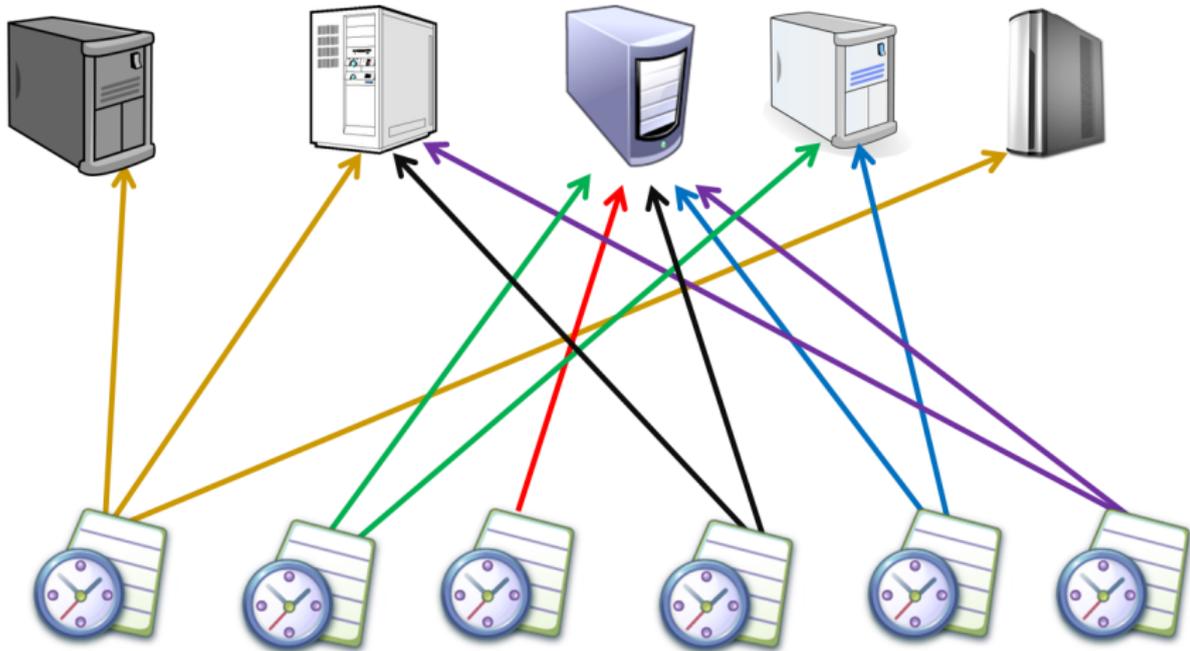
**Set of tasks**  
to be processed by  
machines



# Load balancing problem



# Load balancing problem



# Load balancing problem



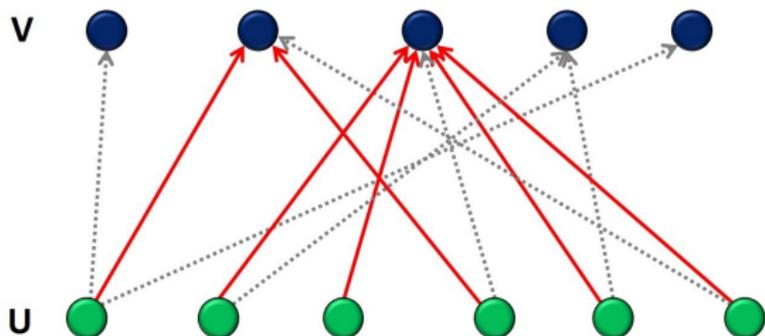
**Input:** a unweighted bipartite graph  
**Goal:** assign each task to a suitable machine



# Semi-matchings

**Semi-matching** in a bipartite graph  $G = (U, V, E)$ :

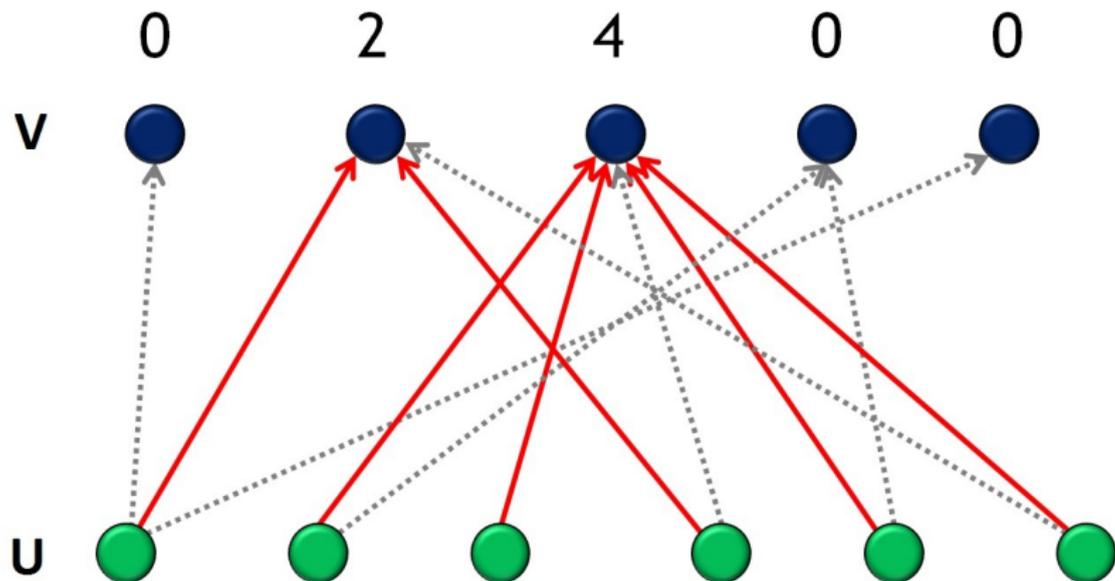
- any subset  $M \subseteq E$  such that  $\deg_M(u) \leq 1$  for all  $u \in U$
- each task is assigned to at most one machine



**Maximum semi-matching** - maximizes the number of assigned tasks; if there is *no other restriction* then

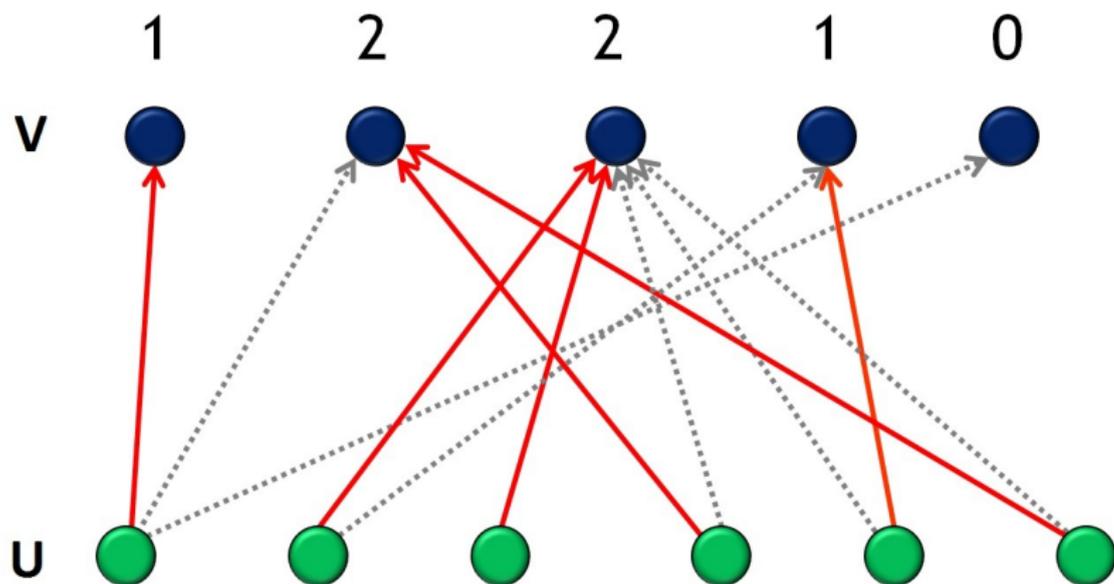
- any subset  $M \subseteq E$  such that  $\deg_M(u) = 1$  for all  $u \in U$
- always exists, **many** maximum semi-matchings

# Which semi-matching is better?



Workload distribution (sorted loads): **4, 2, 0, 0, 0**

# Which semi-matching is better?



Workload distribution (sorted loads): **2, 2, 1, 1, 0**

# Optimal semi-matchings

**Cost** of a semi-matching  $M$  (the total completion time):

$$\text{cost}(M) = \sum_{v \in V} \frac{\text{deg}_M(v) \cdot (\text{deg}_M(v) + 1)}{2}$$

## Optimal semi-matching

- a maximum semi-matching  $M$  such that  $\text{cost}(M)$  is minimal
- a maximum semi-matching  $M$  such that its degree (workload) distribution is **lexicographically minimal**
  - shown by Bokal et al. to be equivalent with  $\text{cost}$ -minimal semi-matching (and also other cost measures)
  - in the previous example:  $(4, 2, 0, 0, 0)$  vs.  $(2, 2, 1, 1, 0)$

Our **optimality criterion**: **lexicographical minimality**

## Previous work

Algorithms for computing **an optimal semi-matchings**:

- $O(n^3)$  by Horn (1973) and Bruno et al. (1974)
- $O(n \cdot m)$  by Lovász et al. (2006, JAlgor)
- $O(\min\{n^{3/2}, m \cdot n\} \cdot m)$  by Lovász et al. (2006, JAlgor)
- $O(n \cdot m)$  by Bokal et al. (2009) for generalized setting
- $O(\sqrt{n} \cdot m \cdot \log n)$  by Fakcharoenphol et al. (2010, ICALP)

Algorithms are based on finding (cost-reducing) **alternating paths** with some properties.

**Maximum matchings** in bipartite graphs:

- $O(\sqrt{n} \cdot m)$  by Micali and Vazirani (1980)
- $O(n^\omega)$  by Mucha and Sankowski (2004)
  - $\omega$  is the exponent of the best known **matrix multiplication** algorithm
  - randomized algorithm, better for **dense graphs**

# Our work

Can we construct an algorithm for computing an optimal semi-matching that breaks through  $O(n^{2.5})$  barrier for dense graphs?

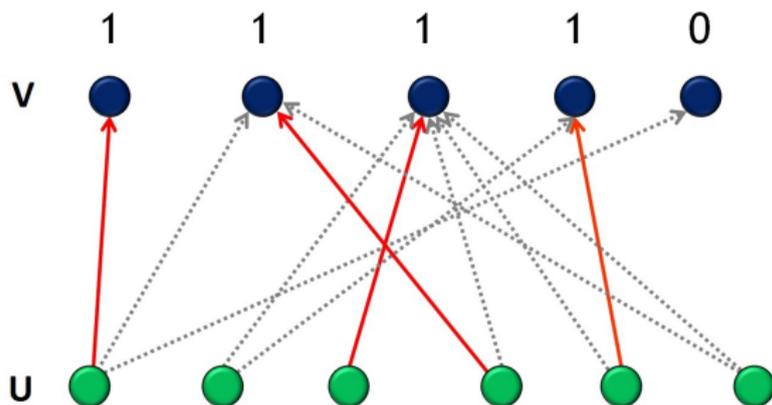
**Answer:** YES, we can

And moreover (side results):

- **new approach** for computing an optimal semi-matching: **divide and conquer** strategy instead of cost-reducing alternating paths
  - divide and conquer = more suitable for **parallel computation**
- **reduction** to a variant of *maximum bounded-degree semi-matching*
  - can be solved by different algorithms and approaches (e.g. maximum matchings, reduction to matrix multiplication)

## Limited workload for $V$ -vertices

**Restriction:** a machine can process only limited number of tasks, e.g. 1 task:

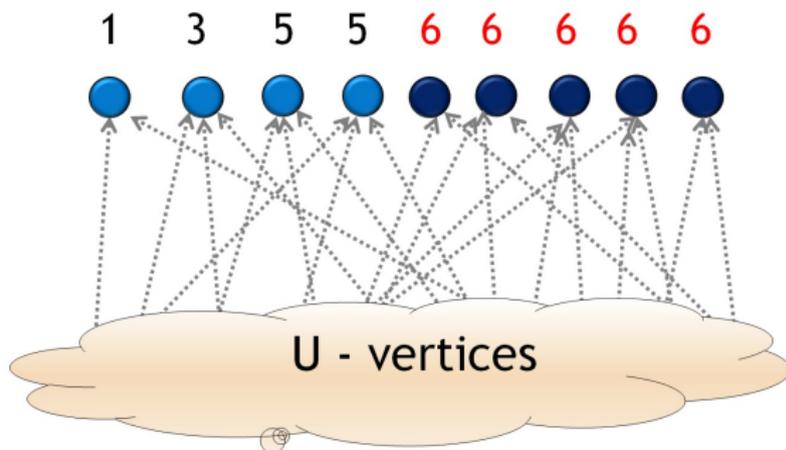


**Intuition:**

- there can be unassigned tasks
  - $U$ -vertices not incident to a matching edge
- larger workload limit for machines = more assigned tasks

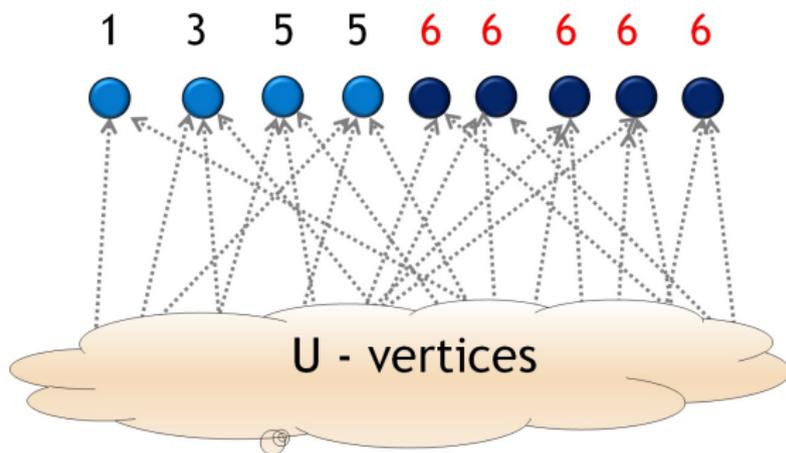
# Limited workload for $V$ -vertices

Maximum semi-matching with workload limit 6  
(max. 6 tasks per machine):



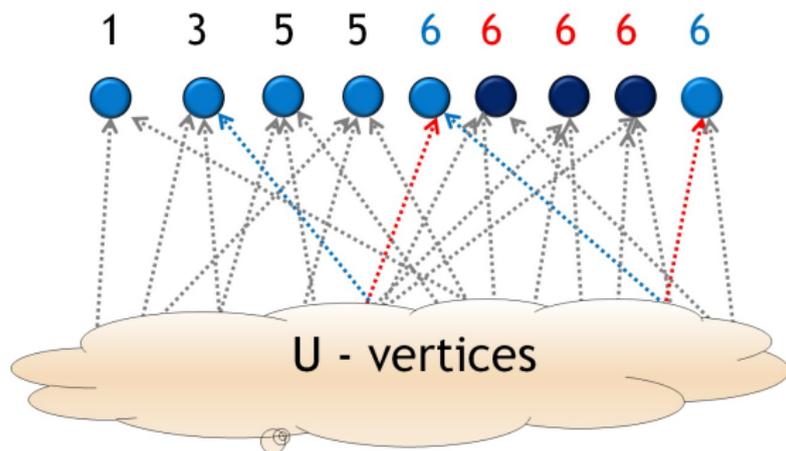
Is it necessary to increase workload limit for all  $V$ -vertices (machines)  
in order to match all  $U$ -vertices?

# Intuition related to limited workload



- **no sense** to increase the workload limit for vertices (machines) that are **not fully loaded** in a given maximum semi-matching

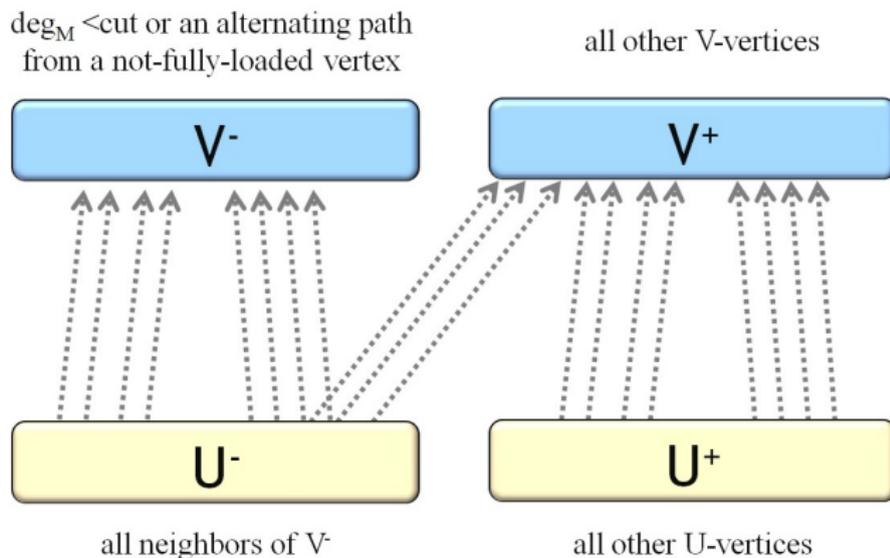
# Are all fully-loaded vertices good candidates?



- **no sense** to increase the workload limit for fully loaded vertices (machines) that are **endpoints of an alternating path** starting in a non-fully loaded vertex

# Intuition: How to divide the problem

Maximum semi-matching  $M$  respecting a workload limit  $cut$ :



Find an optimal semi-matching

- in  $G^- = (U^-, V^-, E^-)$  by "decreasing" workload limits
- in  $G^+ = (U^+, V^+, E^+)$  by "increasing" workload limits

# (Sub)problem instances

$LSM(G)$  - a set of all optimal semi-matchings for  $G$

**Input/problem instances:**  $(G, down, up, M_f)$

- an input bipartite graph  $G = (U, V, E)$  such that
  - $\forall M \in LSM(G), \forall v \in V : down \leq deg_M(v) \leq up$
- a semi-matching  $M_f$  in  $G$  such that
  - $\forall v \in V : deg_{M_f}(v) \geq down$

**Goal:** if  $(G, down, up, M_f)$  is an input, compute an optimal semi-matching for  $G$

**Starting point:**  $(G, 0, \infty, \emptyset)$

- $G$  is a graph, in which we want to find an optimal semi-matching
- all preconditions are satisfied

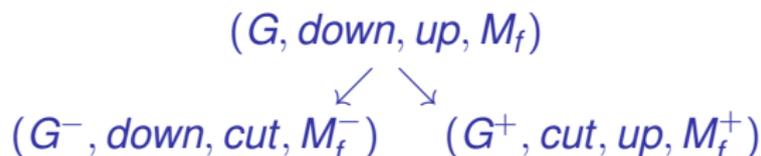
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- a semi-matching  $M_f$  in  $G$  such that
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**Divide phase** for  $cut$  ( $down \leq cut \leq up$ ):



**Key property:**

- $\forall M^- \in LSM(G^-), \forall M^+ \in LSM(G^+): M^- \cup M^+ \in LSM(G)$

## Trivial case (or why is $M_f$ required)

**Input:**  $(G, down, up, M_f)$ , where  $up - down \leq 1$

**Problem:** How to compute  $M \in LSM(G)$ ?

**First idea:**

- compute a maximum semi-matching  $M$  for load limit  $up$
- it can happen that  $M \notin LSM(G)$ :
  - $(3, 2, 2, 2, 2, 2) \in LSM(G)$  vs.  $(3, 3, 3, 3, 1, 0) \notin LSM(G)$

**Solution:**

- utilizing  $M_f$  with  $deg_{M_f}(v) \geq down$  for all  $v \in V$ , **transform** semi-matching  $M$  to a semi-matching  $M_B$  such that
  - $|M| = |M_B|$
  - $down \leq deg_{M_B}(v) \leq up$  for all  $v \in V$
- it can be shown that  $M_B \in LSM(G)$
- transformation can be realized in the linear time

# Dividing subroutine - idea

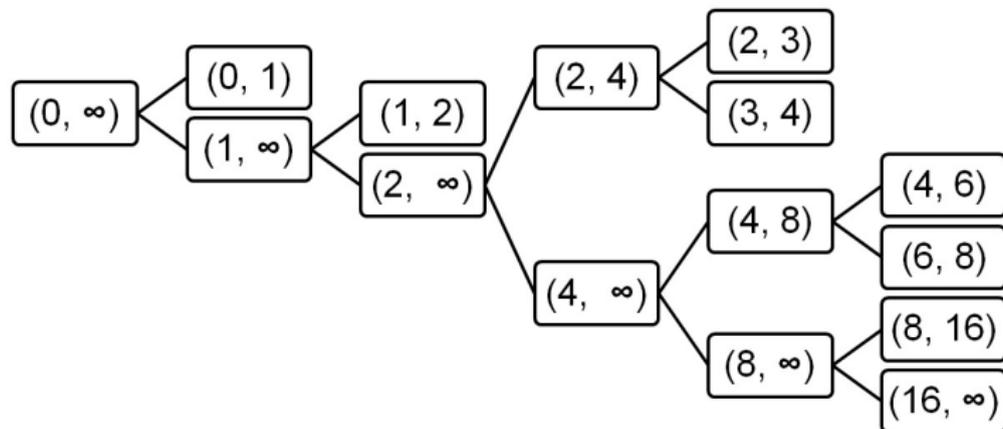
**Input instance:**  $(G, \text{down}, \text{up}, M_f)$

**Computation:**

- 1 compute a maximum semi-matching  $M$  for workload limit  $\text{cut}$
- 2 compute  $M_B$  by rebalancing  $M$  with respect to  $M_f$
- 3 compute  $V^-$ ,  $V^+$ ,  $U^-$ , and  $U^+$  considering workload of  $V$ -vertices
- 4 compute induced subgraphs  $G^- = (U^-, V^-, E^-)$  and  $G^+ = (U^+, V^+, E^+)$
- 5 compute  $M_f^- = M_B \cap E^-$  and  $M_f^+ = M_B \cap E^+$
- 6 return  $(G^-, \text{down}, \text{cut}, M_f^-)$  and  $(G^+, \text{cut}, \text{up}, M_f^+)$

# Main algorithm - Divide and conquer

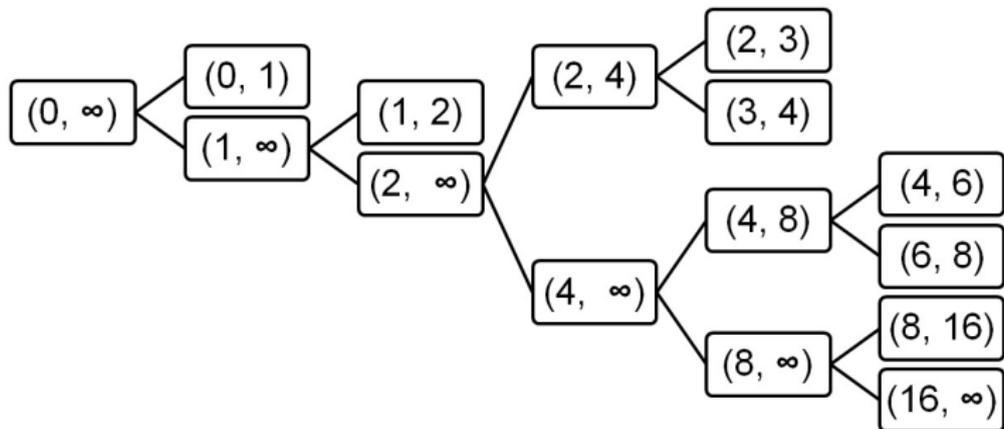
Computational tree starting with  $(G, 0, \infty, \emptyset)$ :



- **Divide and conquer:**  $(down, up)$  is always divided into 2 subintervals (of almost equal size)
- **Doubling:**  $(down, \infty)$  is divided to  $(down, 2 \cdot down)$  and  $(2 \cdot down, \infty)$

# Main algorithm - Computation

Computational tree starting with  $(G, 0, \infty, \emptyset)$ :



- after  $O(\log n)$  levels, graphs of subproblems are empty
  - there is no subgraph of  $G$  for which a semi-matching with load of a  $V$ -vertex at least  $n + 1$  exists

# Maximum semi-matching with workload limits?

- in each step of the algorithm, we need a maximum semi-matching that respects the workload limits

## Problem (Bounded-degree semi-matching )

**Instance:** A bipartite graph  $G = (U, V, E)$  with  $n = |U| + |V|$  vertices and  $m = |E|$  edges; a capacity mapping  $c : V \rightarrow \mathbb{N}$  satisfying  $\sum_{v \in V} c(v) \leq 2 \cdot n$ .

**Question:** Find a semi-matching  $M$  in  $G$  with maximum number of edges such that  $\deg_M(v) \leq c(v)$  for all  $v \in V$ .

**Time complexity notation:**  $T_{BDSM}(n, m)$  for a graph  $n$  vertices and  $m$  edges.

**Total time for computing an optimal semi-matching:**

$$O((n + m + T_{BDSM}(n, m)) \cdot \log n)$$

# Bounded-degree semi-matching

## Reduction to maximum matching:

- make  $c(v)$  **copies** of each  $V$ -vertex  $v$
- new graph has at most  $3 \cdot n$  vertices
- apply algorithm for maximum matching in  $O(n^\omega)$  by Mucha and Sankowski

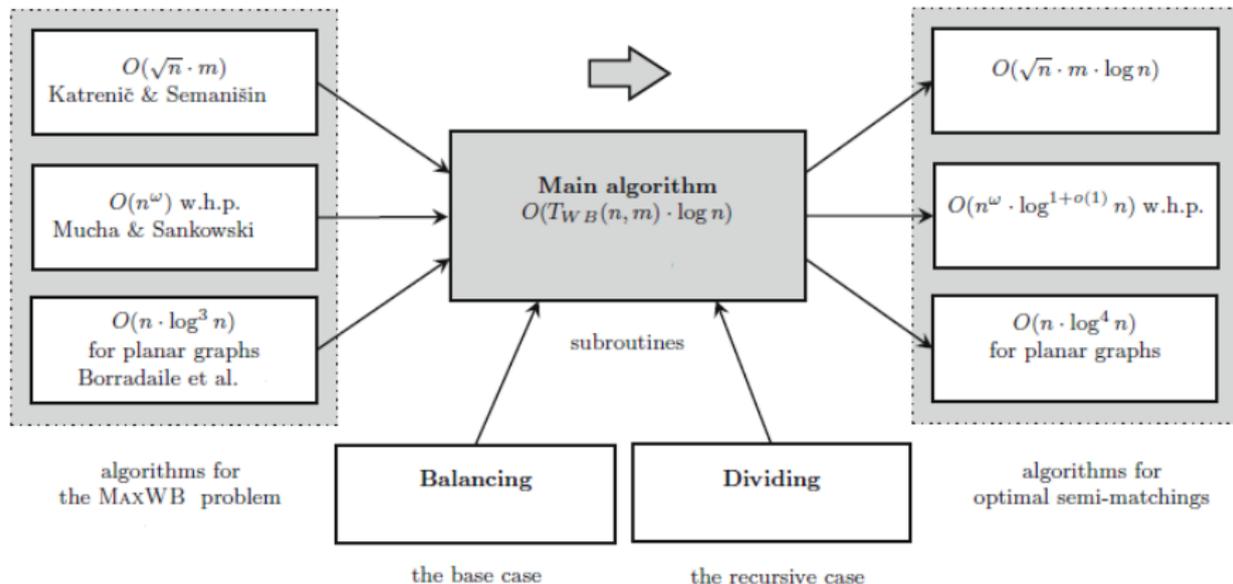
$$O(n^\omega \cdot \log n)$$

## Reduction to $(1, c)$ -semi-matchings:

- $(1, c)$ -semi-matching is bounded-degree semi-matching **without** condition  $\sum_{v \in V} c(v) \leq 2 \cdot n$
- due to algorithm by Katrenič and G.S.,  $(1, c)$ -semi-matching can be computed in time  $O(\sqrt{n} \cdot m)$

$$O(\sqrt{n} \cdot m \cdot \log n)$$

# Conclusion



## Related works

### Semi-matching problem

- algorithm for optimal **weighted semi-matching** with time complexity  $O(n^2 \cdot m)$  (Harada et al. 2007)
- **distributed deterministic 2-approximation** algorithms with time complexity  $O(\Delta^5)$  (Czygrinow et al. 2012, 2016)
- a **deterministic one-pass streaming** algorithm that for any  $0 \leq \epsilon \leq 1$  uses computes an  $O(n^{(1-\epsilon)/2})$  and with  $O(\log n)$  passes computes an  $O(\log n)$  approximation (Konrad et al. 2016)

### Distributed Backup Placement problem

- a **distributed** algorithm which finds placement in polylogarithmic time and approximation ratio  $O\left(\frac{\log n}{\log \log n}\right)$  (Halldórsson et al. 2015)

## Future work

A **generalisation** introduced by Bokal et al.

An  $(f, g)$ -**quasi-matching** in a bipartite graph  $G = (U \cup V, E)$ : any set of edges  $M \subseteq E$  such that

- each vertex  $u \in U$  is incident with at most  $f(u)$  edges of  $M$
- each vertex  $v \in V$  is incident with at least  $g(v)$  edges of  $M$ .

### Algorithm complexity

- Bokal et al. provided algorithm with complexity  $O(m.g(B))$ , where  $g(B) = \sum_{v \in B} g(v)$
- our reduction lemma should allow an **improvement of this result by factor  $\log n$  and factor  $\sqrt{g(B)}$** .

# References

-  D. Bokal, B. Brešar, J. Jerebic:  
A generalization of Hungarian method and Hall's theorem with applications in wireless sensor networks  
Discret. Appl. Math. 160(4–5), 460–470 (2012)
-  J. Fakcharoenphol, B. Laekhanukit, D. Nanongkai  
Faster algorithms for semi-matching problems  
ACM Trans. Algorithms 10(3), 14:1–14:23 (2014)
-  Katrenič, J., Semanišin, G.  
Maximum semi-matching problem in bipartite graphs  
Discuss. Math. Graph Theory 33, 559–569 (2013)
-  F. Galčík, J. Katrenič, G. Semanišin  
On computing an optimal semi-matching  
(to appear in Algorithmica, already published electronically)



Thank you for your attention

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