## Fault-Tolerant Power Assignment and Backbone in Wireless Networks

by Paz Carmi, Matthew J. Katz, Michael Segal and Hanan Shpungin

## Outline

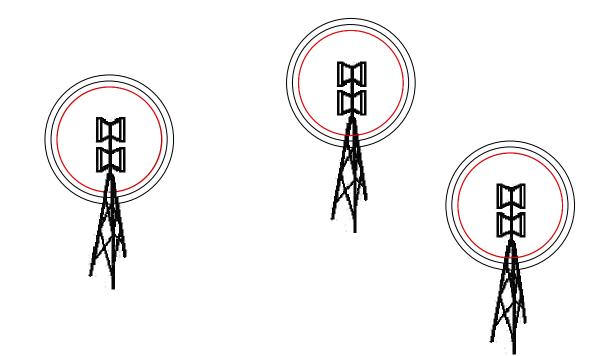
- Introduction
- Model & Problems
- Fault-Tolerant Power Assignment
- Connected Backbone Power Assignment
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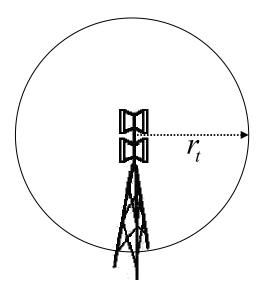
### • Wireless Ad-Hoc Network

Set of transceivers communicating by radio



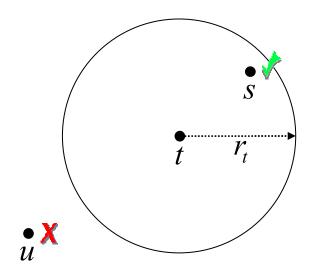
### • Wireless Ad-Hoc Network

• Each transceiver has a transmission power p(t)which results in a transmission range  $r_t$ 



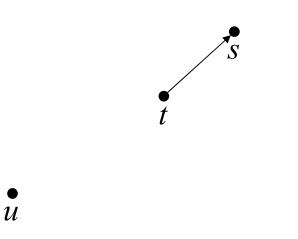
### • Wireless Ad-Hoc Network

• Transceiver *s* receives transmission from *t* only if  $d(t,s) \le r_t$ 



### • Wireless Ad-Hoc Network

As a result a directed communication graph is induced

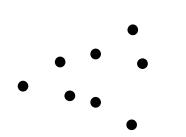


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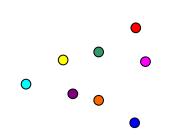
### Definition

• A set *T* of *n* transceivers  $t_1, t_2, \ldots, t_n$ 



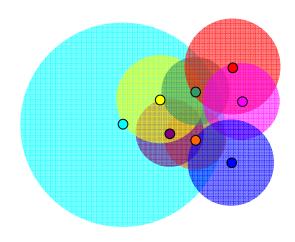
### Definition

- A set *T* of *n* transceivers  $t_1, t_2, \ldots, t_n$
- $A = A(T) = \{p(t) | t \in T\}$  is the power assignment



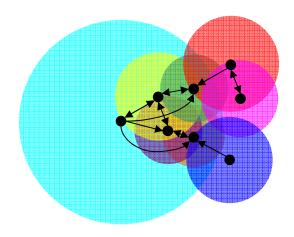
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### Definitions

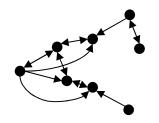
- A set *T* of *n* transceivers  $t_1, t_2, \ldots, t_n$
- $A = A(T) = \{p(t) | t \in T\}$  is the power assignment
- $H_A = (T, E_A)$  is the communication graph



### Definitions

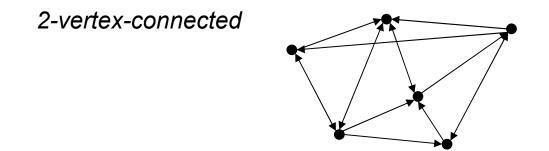
• A set *T* of *n* transceivers  $t_1, t_2, \ldots, t_n$ 

- $A = A(T) = \{p(t) | t \in T\}$  is the power assignment
- $H_A = (T, E_A)$  is the communication graph
- $C_A = \sum_{t \in T} p(t) = \sum_{t \in T} r_t^{\alpha}$  is the cost of the assignment



### Definitions

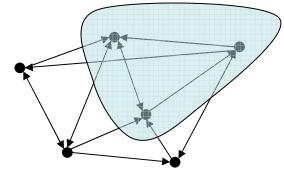
 A graph G = (V, E) is k-vertex-connected if for any two nodes u, v ∈ V there exist k-vertex-disjoint paths connecting u to v



### Definitions

For graph G = (V, E), a subset D ⊆ V is a *connected backbone* if G restricted to D is strongly connected and for each t ∈ T \ D there exists u ∈ D so that e = (u,t) ∈ E





- **Problem 1** (*k*-vertex-connectivity)
  - Input:A set T of transceivers, and a parameter k > 1Output:A power assignment A(T) with minimal<br/>possible cost  $C_A$ , where  $H_A$  is k-vertex<br/>connected

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#### O(k)-approximation algorithm

- **Problem 2** (connected backbone)
  - *Input:* A set *T* of transceivers
  - **Output:** A subset *D* of *T* and a power assignment A(D)with minimal possible cost  $C_A$ , where  $H_A$ (restricted to *D*) is *strongly connected*, and for each  $t \in T \setminus D$ , there exists  $u \in D$ , such that  $d(u,t) \leq r_u$

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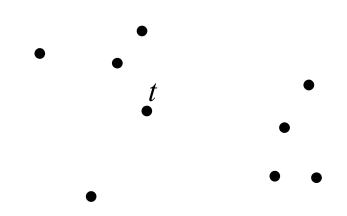
**Constant-factor approximation algorithm in**  $O(n \log n)$ 

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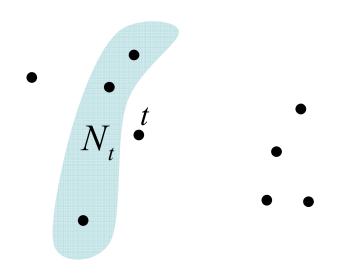
### Definitions

• For each  $t \in T$ , let  $N_t \subseteq T$  be a set of k closest nodes to t



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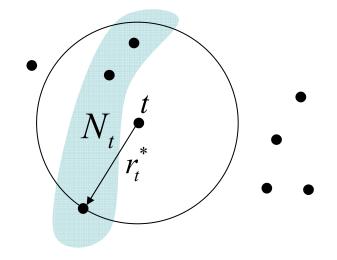
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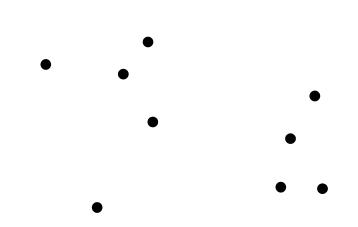
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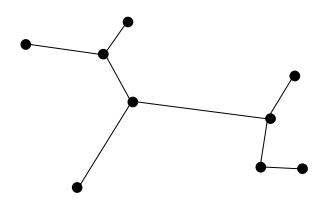
• Let 
$$r_t^* = \max_{t \in N_t} d(t, t')$$



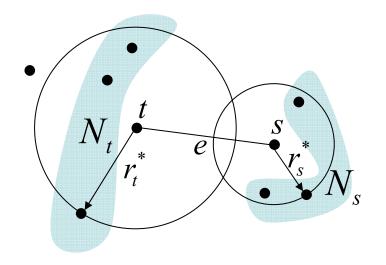
- The algorithm
  - Assign each  $t \in T$  the range  $r_t^*$  (denote  $A_k$ )
  - Compute an MST of *T*



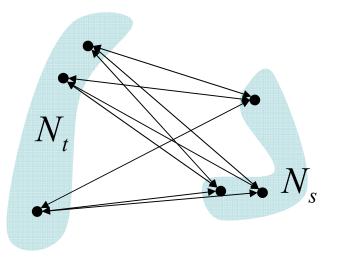
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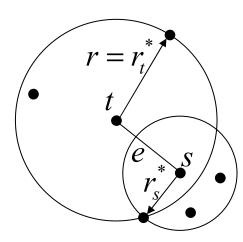
- The algorithm
  - For each edge e = (t, s) of MST increase the range of the nodes in  $N_t \cup N_s$  such that each node  $t' \in N_t$ can reach all nodes in  $N_s$ , and vice versa (denote  $A_k$ )



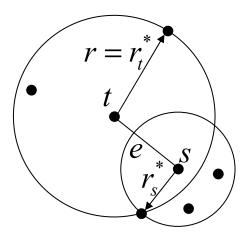
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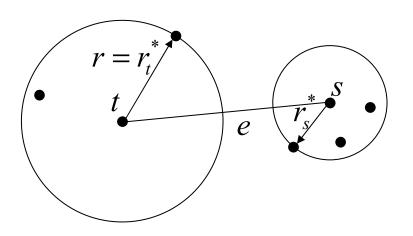
- Proof sketch
  - Let  $r = \max\{r_t^*, r_s^*\}$
  - In  $A_k$  each  $t' \in N_t \cup N_s$  is assigned at most |e|+2r
  - Case 1:  $|e| \leq r$



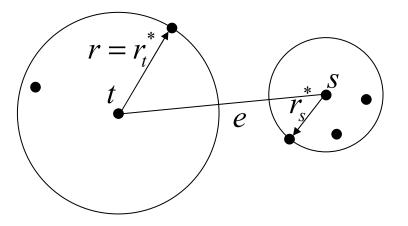
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  - Case 1:  $|e| \le r \implies \max_{\substack{t \in N_t, s \in N_s}} d(t', s') \le 3r$



- Proof sketch
  - Let  $r = \max\{r_t^*, r_s^*\}$
  - In  $A_k$  each  $t' \in N_t \cup N_s$  is assigned at most |e|+2r
  - Case 2: |*e*|>*r*



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  - In  $A_k$  each  $t' \in N_t \cup N_s$  is assigned at most |e| + 2r
  - Case 2:  $|e| > r \implies \max_{\substack{t \in N_t, s \in N_s}} d(t', s') \le 3 |e|$



- Proof sketch
  - Let  $r = \max\{r_t^*, r_s^*\}$
  - In  $A_k$  each  $t' \in N_t \cup N_s$  is assigned at most |e|+2r
  - Easy to see  $C_{A_k} \leq C_{A_k^*}$

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  - Kirousis et al. proved  $C_{\text{MST}} \leq C_{A_1^*}$

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  - Let  $r = \max\{r_t^*, r_s^*\}$
  - In  $A_k$  each  $t' \in N_t \cup N_s$  is assigned at most |e|+2r
  - Easy to see  $C_{A_{k}} \leq C_{A_{k}^{*}}$
  - Kirousis et al. proved  $C_{\text{MST}} \leq C_{A_1^*}$
  - As a result

$$C_{A_k} \leq O(k) \cdot (C_{A_k^*} + C_{\text{MST}}) \leq O(k) \cdot C_{A_k^*}$$

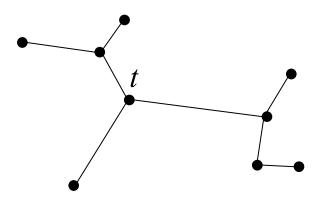
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# **Connected Backbone Power Assignment**

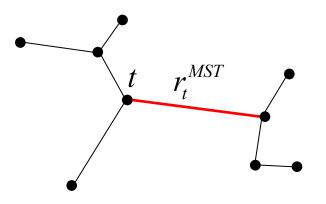
### Definitions

• Given the MST of *T*, for any node  $t \in T$ , let  $r_t^{MST}$  be the size of the longest edge adjacent to *t* 

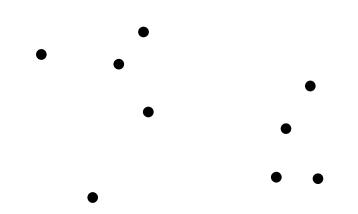


#### Definitions

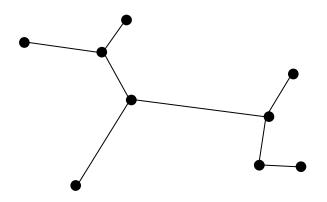
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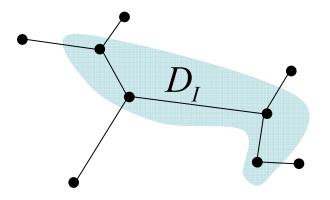
- The algorithm
  - Compute an MST of *T*



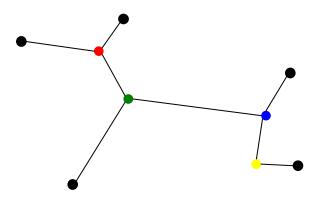
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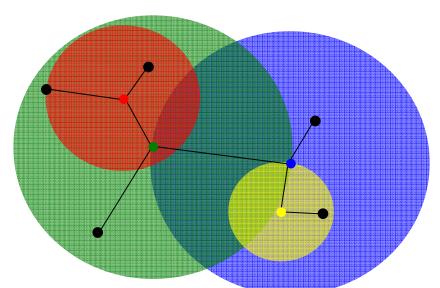
- The algorithm
  - Compute an MST of *T*
  - Let  $D_I$  be the set of all internal nodes of MST



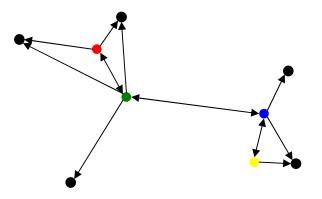
- The algorithm
  - Compute an MST of *T*
  - Let  $D_I$  be the set of all internal nodes of MST
  - Assign each  $u \in D_I$  with  $r_u^{MST}$  (denote A)



- The algorithm
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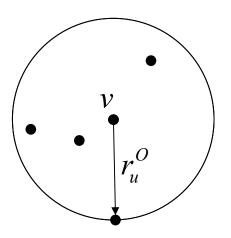


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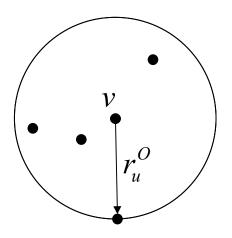


- Proof sketch
  - Construct a power assignment *B* for which it holds  $C_B \leq c_1 \cdot C_{\text{OPT}}$  and  $C_A \leq c_2 \cdot C_B$ , as a result obtaining  $C_A \leq c \cdot C_{\text{OPT}}$
  - B is derived from OPT

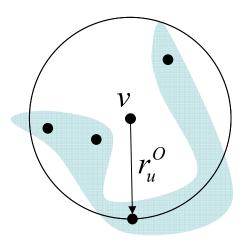
- Proof sketch
  - Let  $D_{\text{OPT}}$  be the *connected backbone* in OPT
  - For each node  $v \in D_{OPT}$  let  $r_v^O$  be the transmission range of v in OPT



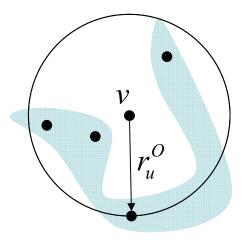
- Proof sketch
  - For each node  $v \in D_{OPT}$  let  $T_v$  be all the nodes within distance  $r_v^O$  from v



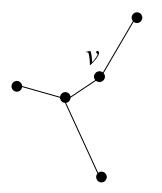
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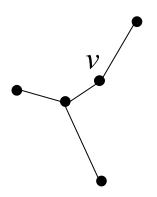
- Proof sketch
  - For each node  $v \in D_{OPT}$  let  $T_v$  be all the nodes within distance  $r_v^O$  from v
  - For each node  $v \in D_{OPT}$  compute  $MST_v$  of  $T_v \cup \{v\}$



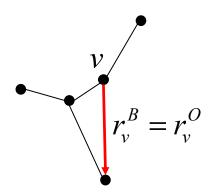
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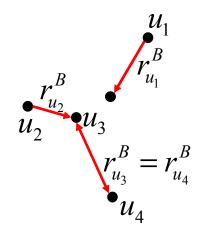
- Proof sketch
  - In *B*:
    - Each node  $v \in D_{OPT}$  is assigned  $r_v^B = r_v^O$



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- Proof sketch
  - In *B*:
    - Each node  $v \in D_{OPT}$  is assigned  $r_v^B = r_v^O$
    - Each node  $u \in T \setminus D_{OPT}$  is assigned  $r_u^B = r_u^{MST_v}$



Proof sketch

• 
$$C_B \leq c_1 \cdot C_{\text{OPT}}$$

• Carmi et al. showed that

$$\sum_{e \in MST_v} area(D_e) \leq 5area(\bigcup_{e \in MST_v} D_e)$$



Proof sketch

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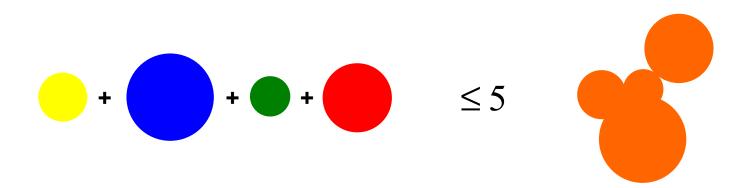
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• Using this fact we obtain

$$C_B = O(C_{\text{OPT}})$$

- Proof sketch
  - $\bullet C_A \leq c_2 \cdot C_B$ 
    - Kirousis et al. proved that given an MST assigning each node  $v \in T$  with  $r_v^{MST}$  yields a 2-factor approximation for *strong-connectivity* (denote  $A_{SC}$ )

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    - Using this fact we obtain

$$C_A \leq C_{A_{SC}} \leq 2C_{A_1^*} \leq 2C_B$$

- Proof sketch
  - Therefore

$$C_A = O(C_{\text{OPT}})$$

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#### Summary

- **Problem 1** (*k*-vertex-connectivity)
  - Input:A set T of transceivers, and a parameter k > 1Output:A power assignment A(T) with minimal<br/>possible cost  $C_A$ , where  $H_A$  is k-vertex<br/>connected

#### O(k)-approximation algorithm

#### Summary

#### • **Problem 2** (connected backbone)

*Input:* A set *T* of transceivers

**Output:** A subset *D* of *T* and a power assignment A(D)with minimal possible cost  $C_A$ , where  $H_A$ (restricted to *D*) is *strongly connected*, and for each  $t \in T \setminus D$ , there exists  $u \in D$ , such that  $d(u,t) \leq r_u$ 

**Constant-factor approximation algorithm in**  $O(n \log n)$ 

#### Summary

#### • **Problem 3** (*k*-connected backbone)

*Input:* A set *T* of transceivers, and a parameter k > 1

*Output:* A subset *D* of *T* and a power assignment A(D)with minimal possible cost  $C_A$ , where  $H_A$ (restricted to *D*) is *k*-vertex connected, and for each  $t \in T \setminus D$ , there exists  $u_1, u_2, \ldots, u_k \in D$ , such that  $d(u_i, t) \leq r_{u_i}$ 

 $O(k^3)$ -approximation algorithm

