

# Agreement with oblivious robots

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# The agreement invariant

- to share the same (an approximate) location
  - gathering (convergence)
- to not share the same location
  - scattering
- to acknowledge the same chief
  - leader election
- to form and maintain the same pattern
  - flocking, pattern formation, connectivity

# Agreement QoS

- Agreement time
- Faults resilience
- Scheduler freedom

# Agreement feasibility

- Without **additional assumptions** exact agreement is impossible to achieve

# Tools to bypass the impossibility results

- increased knowledge (multiplicity, common axes/origin/orientation)
- constrained scheduler (bounded, fair)
- constrained number of faulty robots
- random choices

# The GATHERING case

- Acknowledgements

- Ando, Oasa, Suzuki, Yamashita [Trans. on Robotics and Automation 1999]
- Flocchini, Prencipe, Santoro, Widmayer [TCS 2005]
- Agmon and Peleg [SIAM J. of Computing 2006]
- Cohen and Peleg [STACS 2006]
- Dieudonné and Petit [FUN2007]
- Defago, Gradinariu, Messika, Raipan-Parvédy [DISC 2006 + new results]
- Clémant, Defago, Gradinariu, Izumi [IPL 2010]

# Fault-free environment

Scheduler	multiplicity			
	deterministic		probabilistic	
	$n = 2$	$n \geq 3$	$n = 2$	$n \geq 3$
unfair	NO(L.1)	NO(L.10)	<b>OK(L.8)</b>	<b>OK(L.18)</b>
unfair centr.	<b>OK(L.9)</b>	NO(L.10)	OK(L.8)	<b>OK(L.18)</b>
fair	<b>NO(L.1)</b>	<b>OK(L.2)</b>	OK(L.8)	OK(L.2)
fair centr.	OK(L.9)	OK(L.2)	OK(L.8)	OK(L.2)
fair bounded		OK(L.2)	OK(L.8)	OK(L.2)
fair $k$ -bounded		OK(L.2)	OK(L.8)	OK(L.2)
fair bounded reg.		OK(L.2)	OK(L.8)	OK(L.2)
fair centr. bounded reg.	OK(L.9)	OK(L.2)	OK(L.8)	OK(L.2)

Scheduler	without multiplicity			
	deterministic		probabilistic	
	$n = 2$	$n \geq 3$	$n = 2$	$n \geq 3$
unfair	NO(L.3)	NO(L.3)	<b>OK(L.8)</b>	NO(L.12)
unfair centr.	<b>OK(L.9)</b>	NO(L.7)	OK(L.8)	NO(L.12)
fair	<b>NO(L.3)</b>	<b>NO(L.3)</b>	OK(L.8)	NO(L.12)
fair centr.	OK(L.9)	NO(L.7)	OK(L.8)	<b>NO(L.12)</b>
fair bounded		NO(L.7)	OK(L.8)	<b>OK(L.13)</b>
fair $k$ -bounded		NO(L.7)	OK(L.8)	OK(L.13)
fair bounded reg.		NO(L.7)	OK(L.8)	OK(L.13)
fair centr. bounded reg.	OK(L.9)	<b>NO(L.7)</b>	OK(L.8)	OK(L.13)

# Probabilistic gathering

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## Functions:

*observe\_neighbors* :: returns the set of robots within visibility range of robot  $p$  (the set of  $p$ 's neighbors). Note that, in a system with unlimited visibility, *observe\_neighbors* returns all the robots in the network.

## Actions:

$\mathcal{A}_1 :: true \rightarrow$

$\mathcal{N}_p = \text{observe\_neighbors}();$   
with probability  $\alpha = \frac{1}{|\mathcal{N}_p \cup \{p\}|}$  do  
select a robot  $q \in \mathcal{N}_p \cup \{p\};$   
move towards  $q;$

*Remark: with probability  $1 - \alpha$ , the position remains unchanged;*

- convergence in  $O(n^2)$  rounds when strong multiplicity knowledge is used
- **no multiplicity** knowledge: exponential convergence time



# Probabilistic gathering

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**Functions:** *observe\_neighbors* :: returns the set of robots in the system  
*maximal\_multiplicity* :: returns the set of points with maximal multiplicity  
*local\_position* :: returns the local position of the robot  $p$   
(i.e., the locations shared by the largest number of robots as returned by the observe function);

**Actions:**  $\mathcal{N}_p = \text{observe\_neighbors}()$ ;  
**if**  $|\text{maximal\_multiplicity}(\mathcal{N}_p)| > 1$  **then**  
    with probability  $\frac{1}{2n}$   
        let  $q$  be a point chosen arbitrarily from  $\text{maximal\_multiplicity}(\mathcal{N}_p) \setminus lp()$ ;  
        move towards  $q$ ;  
    with probability  $1 - \frac{1}{2n}$   
        do not move;  
**if**  $|\text{maximal\_multiplicity}(\mathcal{N}_p)| = 1$  **then**  
    let  $q$  be the point given by  $\text{maximal\_multiplicity}(\mathcal{N}_p)$ ;  
    move towards  $q$ ;

- convergence in  $O(l)$  rounds
- **strong multiplicity** knowledge

# Probabilistic gathering

**Actions:**

```
 $\mathcal{N}_p = \text{observe\_neighbors}();$   
if  $\mathcal{N}_p$  includes two or more multiplicity points then  
    Execute Algorithm 6.1;  
else if  $\mathcal{N}_p$  includes a single multiple point  $q$  then  
    Move towards  $q$ ;  
else  
    let  $q$  be a point chosen arbitrarily from  $\mathcal{N}_p$ ;  
    With probability  $\frac{1}{2n}$   
        move towards  $q$ ;  
    With probability  $1 - \frac{1}{2n}$   
        do not move;
```

- convergence in  $O(\log n \log(\log n))$  rounds
- **weak multiplicity** knowledge
- use a simple scattering scheme

# Two scattering schemes

Weak multiplicity knowledge  $O(\log n \log \log(n))$

```
Compute the Voronoi diagram;  
 $Cell_i :=$  Voronoi cell where  $r_i$  is located;  
 $Current\_Pos :=$  position where  $r_i$  is located;  
if  $Random() = 0$  then  
    Move toward an arbitrary position in  $Cell_i$ , different from  $Current\_Pos$ ;  
else Do not move;
```

Strong multiplicity knowledge  $O(1)$

```
Compute the Voronoi diagram;  
 $Cell_i :=$  Voronoi cell where  $r_i$  is located;  
 $Current\_Pos :=$  position where  $r_i$  is located;  
Let  $Pos$  be a set of  $2n^2$  positions in  $Cell_i$ ;  
    Move toward a position in  $Pos$  chosen uniformly at random,
```

# Two lessons

- refining the multiplicity knowledge may improve the complexity
- refining the scheduler may change de history

# Crash-prone environment

Scheduler	multiplicity		without multiplicity	
	deterministic	probabilistic	deterministic	probabilistic
$f = 1$				
unfair		<b>OK(L.18)</b>	NO(L.3)	NO(L.12)
unfair centr.		OK(L.18)	NO(L.7)	NO(L.12)
fair	<b>!!!OK(L.4)!!!</b>	OK(L.4)	NO(L.3)	NO(L.12)
fair centr.	OK(L.4)	OK(L.4)	NO(L.7)	NO(L.12)
fair bounded	OK(L.4)	OK(L.4)	NO(L.7)	OK(L.15)
fair $k$ -bounded	OK(L.4)	OK(L.4)	NO(L.7)	OK(L.15)
fair bounded reg.	OK(L.4)	OK(L.4)	NO(L.7)	OK(L.15)
fair centr. bounded reg.	OK(L.4)	OK(L.4)	NO(L.7)	OK(L.15)
$f \geq 2$				
unfair		<b>OK(L.18)</b>	NO(L.16)	NO(L.16)
unfair centr.		OK(L.18)	NO(L.16)	NO(L.16)
fair		OK(L.18)	NO(L.16)	NO(L.16)
fair centr.	<b>OK(L.17)</b>	OK(L.17)	NO(L.16)	NO(L.16)
fair bounded		OK(L.18)	NO(L.16)	NO(L.16)
fair $k$ -bounded		OK(L.18)	NO(L.16)	NO(L.16)
fair bounded reg.		OK(L.18)	NO(L.16)	NO(L.16)
fair centr. bounded reg.	OK(L.17)	OK(L.17)	<b>NO(L.16)</b>	<b>NO(L.16)</b>

# Byzantine-prone environment

Scheduler	multiplicity; deterministic					
	$f = 1$			$f \geq 2$		
	$n = 3$	$n \geq 4$ (even)	$n \geq 4$ (odd)	$n = 3$	$n \geq 4$ (even)	$n \geq 4$ (odd)
unfair	NO(L.5)	NO(L.5)	NO(L.5)	NO(L.5)	NO(L.5)	NO(L.5)
unfair centr.		NO(L.21)		NO(L.22)	NO(L.21)	NO(L.22)
fair	<b>NO(L.5)</b>	NO(L.5)	NO(L.5)	NO(L.5)	NO(L.5)	NO(L.5)
fair centr.		NO(L.21)	<b>OK(N.2)</b>	NO(L.22)	NO(L.21)	NO(L.22)
fair bounded		NO(L.21)		NO(L.22)	NO(L.21)	NO(L.22)
$(k \geq n - 1)$ -bounded		NO(L.21)		NO(L.22)	NO(L.21)	NO(L.22)
$(\Gamma(n, f) \leq k \leq n - 2)$ -bounded		<b>OK(C.4)</b>		NO(L.22)	NO(L.21)	NO(L.22)
$(2 \leq k < \Gamma(n, f))$ -bounded		<b>OK(C.4)</b>				
fair bounded reg.		OK(C.4)				
centr. $(k \geq n - 1)$ -bound.		<b>NO(L.21)</b>	OK(N.2)	NO(L.22)	NO(L.21)	NO(L.22)
centr. $(\Gamma(n, f) \leq k \leq n - 2)$ -bound.		OK(C.4)	OK(N.2)	<b>NO(L.22)</b>	<b>NO(L.21)</b>	<b>NO(L.22)</b>
centr. $(2 \leq k < \Gamma(n, f))$ -bound.		OK(C.4)	OK(N.2)			
fair centr. bounded reg.		OK(C.4)	OK(N.2)			
fully synchronized		<b>OK(L.6)</b>	<b>OK(L.6)</b>		<b>OK(L.6)</b> if $n \geq 3f + 1$	

$$\Gamma(n, f) = \left\lceil \frac{n-f}{f} \right\rceil \text{ if } n \text{ even ; } \Gamma(n, f) = \left\lceil \frac{n-f}{f-1} \right\rceil \text{ if } n \text{ odd.}$$

# Byzantine-prone environment

## Functions:

*observe\_neighbors* :: returns the set of robots within the vision range of robot  $p$  (the set of  $p$ 's neighbors);

*maximal\_multiplicity* :: returns the set of robots with the maximal multiplicity;

## Actions:

```
 $\mathcal{A}_1 :: true \rightarrow$   
   $\mathcal{N}_p = \text{observe\_neighbors}();$   
  if  $p \in \text{maximal\_multiplicity}(\mathcal{N}_p) \wedge |\text{maximal\_multiplicity}(\mathcal{N}_p)| > 1$  then  
    with probability  $\frac{1}{|\text{maximal\_multiplicity}(\mathcal{N}_p)|}$  do  
      select a robot  $q \in \text{maximal\_multiplicity}(\mathcal{N}_p);$   
      move towards  $q;$   
  else  
    select a robot  $q \in \text{maximal\_multiplicity}(\mathcal{N}_p);$   
    move towards  $q;$ 
```

- strong multiplicity knowledge
- probabilistic scheduler
- exponential convergence time !!!!!

# Byzantine resilient CONVERGENCE

Computation Model	Scheduler	Bounds
ATOM	fully synchronous	$n > 3f$
ATOM	fully synchronous	$n > 2f$
ATOM	$k$ -bounded	$n > 3f$
CORDA	$k$ -bounded	$n > 4f$
CORDA	$k$ -bounded	$n > 3f$
CORDA	fully asynchronous	$n > 5f$

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  - Cohen and Peleg
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# Open problems

- Byzantine resilient gathering
  - investigate the power of probabilistic algorithms
- Byzantine resilient convergence
  - multi-dimensional spaces
- connections with classical distributed computing area
  - convergence ~ approximate consensus
  - scattering ??? renaming
  - gathering ??? consensus / k-set agreement
  - robot models ??? distributed models