

# 論文紹介: When are Distributed Algorithms Robust ?



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- Introduction
- Basic Framework (definition of controlled agents, network, failures, robustness, etc)
- Example (Average Consensus)
- How to make Algorithms Robust (byzantine general problems).



- Distributed Algorithm : Average-Consensus, rendezvous, sensor coverage, deployment.  
Etc
- Involving many agents which are cooperating to get a greater accuracy
- Huge numbers of Agents components will be cheap high failures rates
- Is the task still satisfactorily executed even when some agents fail to communicate or perform correctly ?



Definition I : A controlled agent

1.  $x(k)$   $X$  is the state
2.  $u(k)$   $U$  is the control input
3.  $x(0)$   $X(0)$  is the initial condition
4.  $f: X \times U \rightarrow X$  is a map defines dynamics  
ie.  $x(k+1) = f(x(k), u(k))$  or  
 $x(k+1) = A(k)x(k) + B(k)u(k)$



## Definition II : Network of controlled agents

1.  $I = \{1, \dots, N\}$  is the set of unique identifiers
2.  $A = \{A_i\}_{i \in I}$  is the set of controlled agents
3.  $G_{\text{comm}}$  is the set of allowed communication graph. Assume the undirected graphs (agent  $i$  is the neighbor of agent  $j$  if the two can communicate.)
4. Additional environmental variable:  $V$  (eg. Location of obstacles)



## Definition III : Cooperative task

$$C: \prod_i \{x_i(k)\}_{k=0}^{\infty} \times \prod_i \{u_i(k)\}_{k=0}^{\infty} \times \prod_i x_i(0) \times V \mapsto \mathfrak{R}^+$$

The aim of any algorithm that carries out the task is to **minimize** the cost function

## Definition IV : Cooperative Algorithm

is a choice of communication and control laws for every agent.



- Average Consensus: the task is to ensure each agent has the value  $m$  (arithmetic mean)
- Dynamics :  $x_i(k+1) = x_i(k) + u_i(k)$

- Cost Function :

$$C = \lim_{k \rightarrow \infty} \sum_{\text{all nodes } i} \left| x_i(k) - \frac{1}{N} \sum x_i(0) \right|^2$$

- Control input :

$$u_i(k) = -h \sum_{j: j \text{ is a neighbor of } i} |x_i(k) - x_j(k)|$$



- One way to characterize an algorithm is through the value of the task cost function  $C$  that it achieves.
- PC: Performance Cost by the algorithm can be a function of  $x(0)$  and  $V$ .
- Average Cost,  $PC_{\text{avg}}$  averaging PC as  $x(0)$  and  $V$  are chosen from a given set  $S$ .
- Worst Case,  $PC_{\text{wc}}$  compute the supremum of PC.
- PC also depends on the number of agents,  $PC_{\text{avg}}(N)$





When an agent fails, it alters the control law and the communication law that it follows. Failure modes can be defined as follows :

- Failure mode 1 : An agent may fail by simply **ceasing** to communicate with other agents.
- Failure mode 2 : An agent fails by setting its state value  $x_i(k)$  to a **constant** in the set  $X_i$ .
- Failure mode 3 : The agent alters the control input to set its state at every time step  $k$  to an **arbitrary** value in the set  $X_i$ .



## Definition V : Robustness of an Algorithm

An algorithm is said to be worst-case robust to a particular failure up to  $p$  agents if

$$PC_{wc}(N, p) = O(PC_{wc}(N - p))$$

Note :  $f(x) = O(g(x))$  iff there exists numbers  $x_0$  and  $M > 0$  s.t.  $|f(x)| \leq M|g(x)|$  For  $x > x_0$

- If  $PC_{wc}(N, p) = \Omega(PC_{wc}(N - p))$  but  $PC_{wc}(N, p) \neq \Theta(PC_{wc}(N - p))$

The algorithm is said to be worst-case non-robust



- **Worst case robustness** tells us if the algorithm will perform correctly for any set of initial condition.
- **Average case robustness** guarantees that the algorithm will perform correctly on an average.
- **Almost sure robustness** worst case robust except on a region with measure zero.



Proposition I :

If an algorithm is non-robust for  $p$  failed agents to failure mode 2, it is non robust to  $p$  failed agents to mode 3.

Proof :

Let the control input used in the calculation of PC for failure mode 3(2) be given by  $\{u_i(k)\}_{3(2)}$  for agent  $i$  and messages sent be given by  $\{m_i(k)\}_{3(2)}$ . Consider the choice of the control inputs. The set in which the control inputs are allowed to vary for mode 3 also contains as a particular element  $\{u_i(k)\}_2$ .



Proof :

Since by definition, the cost in mode 3 is maximized by  $\{u_i(k)\}_3$ ; in particular, the cost achieved by using  $\{u_i(k)\}_2$  is not more than when  $\{u_i(k)\}_2$  is used. Thus,

$$PC_{wc} \left( N, p \right)_{\text{failure mode 3}} \geq PC_{wc} \left( N, p \right)_{\text{failure mode 2}}$$

If the algorithm is non-robust to failure mode 2, there exists a constant  $c$  s.t.

$$PC_{wc} \left( N, p \right)_{\text{failure mode 2}} \geq c PC_{wc} \left( N \right)$$



Proposition II :

If an algorithm is non-robust to failure of  $p$  agents in failure mode 3, it is also non-robust to failure of  $t$  agents in failure mode 3 where  $t \geq p$ .

Proof :

Consider the case when  $p$  agents fail and the choice of init. Conditions, control inputs and messages for failed agents that corresponds to the worst case of PC. Choose a set  $S$  of  $t-p$  func. Agents which control inputs and messages are  $\{u_i(k)\}$  and  $\{m_i(k)\}$ . Now, consider the case when  $t$  agents can fail.



Proof :

Choose the same init. Cond. as the previous case.

Let the  $t$  agents that fail be chosen s.t. they consist of the  $p$  agents that failed in the previous case and  $t-p$  agents in the set  $S$ . Also, let the  $p$  agents apply the same control inputs and transmit the same messages as the previous case. Thus, the evolution of the system will be identical to the case when only  $p$  agents failed. Hence,

$$PC_{wc} \{N, t\} \geq PC_{wc} \{N, p\}$$



# Example (Average Consensus)

- Since the algorithm requires connected graphs, we will assume that to be the case as long as no agents fail.
- Consider that the initial conditions to be chosen uniformly over the set  $[-1,1]$ .
- Assume that  $p$  agents that fail are allowed to be chosen so that the graph of the remaining  $N-p$  agents is disconnect. Then the algorithm is worst case non-robust to failure mode 1. If the graph remains connected, then the algorithm is worst-case, average case and a.s. robust to failure mode 1.





Proof :

Consider the case that we allow the graph of the remaining agents to be potentially disconnected. Let  $p=1$ ,  $N=2m+1$  and choose the graph of  $N$  agents as a line. Let the agent  $I$  fail s.t two distinct connected sub-groups of agents are formed, each with  $m$  agents. Also, the initial cond. Are chosen s.t. every agent in the first group has value 1 and in the other group has value  $-1$ . Thus,

$$PC_{wc} \{N, 1\} \geq \sum_{i=1}^m |1|^2 + \sum_{i=1}^m |-1|^2 = N - 1$$



Proof :

$PC_{wc}(N)=0$  as long as the graph was connected. Thus, the algorithm is worst case non-robust. If the graph remains connected,  $PC_{wc}(N,p)=PC_{wc}(N)=0$ . Hence, the algorithm is worst case robust.

- The algorithm is worst-case, average-case and a.s non-robust to failure mode 2.

Proof :

Consider the case when  $p=1$ . Let the initial cond. be s.t. the non-faulty  $N-1$  agents have values 0 while, the faulty agent has value 1.



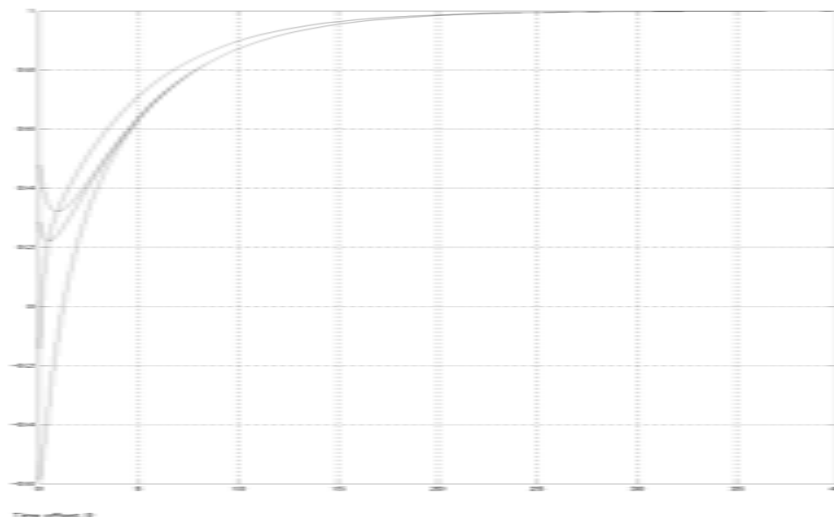
# Example (Average Consensus)

Proof :

Thus, the algorithm will converge with each agent achieving the value 1, as against converging to the correct mean for N-1 agents, which is 0. Thus,

$$PC_{wc} \left[ \sum_{i=1}^N |1 - 0| \right] = N$$

Since  $PC_{wc}(N) \neq 0$ , the algorithm is non-robust.





## Byzantine Generals problem :

- A general needs to transmit a value  $v$  to  $N$  commanders s.t. when the algorithm terminates,
  1. All the functional(loyal) commanders make the same decision about the value. The final value of the non-loyal commanders is not concerned.
  2. If the General is functional, all functional commanders receive the correct value.
- Assume that the General is Functional.



- Consider the Cost

$$C = \lim_{k \rightarrow \infty} \sum_{i=1}^N \|x_i(k) - v\|^2$$

- $x_i$  is the final decision of the  $i$ -th loyal commander and  $N$  is the number of loyal commanders.
- Let's study the robustness properties of three algorithms that solve the problem.
- In the First one, assumed the general to be node 1 whose state remains at  $v$ . Every other agent updates its state as



$$x_i(k+1) = x_i(k) - h \sum_{j \neq i} |x_i(k) - x_j(k)|$$

- The algorithm solves for all agents are functional.
- When  $p$  agents fail according to failure mode 2, as long as a node has a path from the failed agent that does not include the general, it does not converge to the value  $v$ . Thus the algorithm is worst-case and average-case non-robust for any  $p$ .
- It is also non-robust for failure mode 3.



- In the second algorithm, assume that less than one-third agents fail. For simplicity, the comm.graph is assumed to be fully connected.
- The algorithm proceeds as :
  1. At time step 1, the general transmits its value  $v$  to all the commanders.
  2. At time step 2, every commander transmits its estimation to every other commander.
  3. At time step 3, every commander calculates a **majority** of what it has heard and outputs its estimate of the decision.



- The algorithm is both worst-case and average-case robust to failure mode 3.
- The third algorithm involves including a fault detection step in algorithm 1.
- For node  $i$ , let  $N_i$  be the neighbor set of  $i$ . When agent  $i$  communicates with agent  $j$  at time  $k$ , it transmits **4 quantities**:  $x_i(k)$  (denoted by  $a_i(k)$ ),  $x_i(k-1)$  (denoted by  $b_i(k)$ ),  $\sum_{l \in N_i} x_l(k-1)$  (denoted by  $d_i(k)$ ) and  $N_i$ .





- Given these quantities, each node carries out the following checks :
  1. It checks if  $a_i(k-1)=b_i(k)$ .
  2. It checks if  $a_i(k)=(1-hN_i)b_i(k)+hN_i d_i(k)$ .
- If both checks are successful, it carries out the same step as average consensus algorithm, otherwise it identifies the node  $i$  as faulty agents and disregards it from that time on.



- Distributedness in algorithms **does not inherently** lead to robustness.
- To make algorithms robust, in general, we need to ensure the agents receive **enough information** from their neighbors to be able to **detect and isolate** faulty agents.