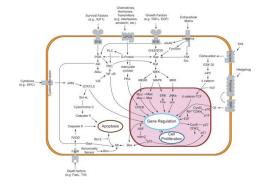
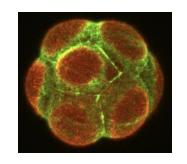




Julia Klein, Tatjana Petrov, Alberto d'Onofrio University of Konstanz
Centre for the Advanced Study of Collective Behaviour

## **Swarms (collectives)**





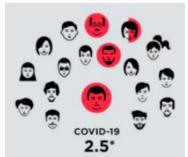












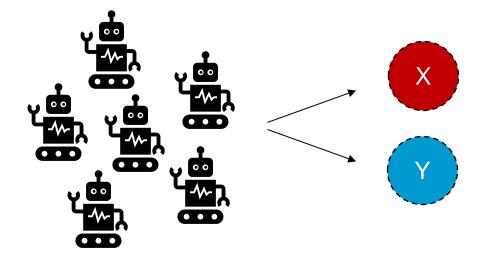


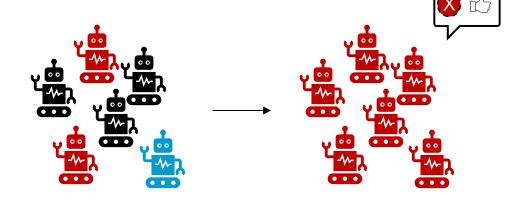


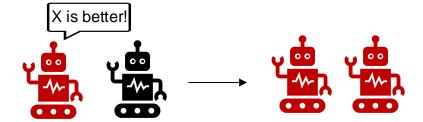


→ How do swarms agree on decisions?

## Collective decision making

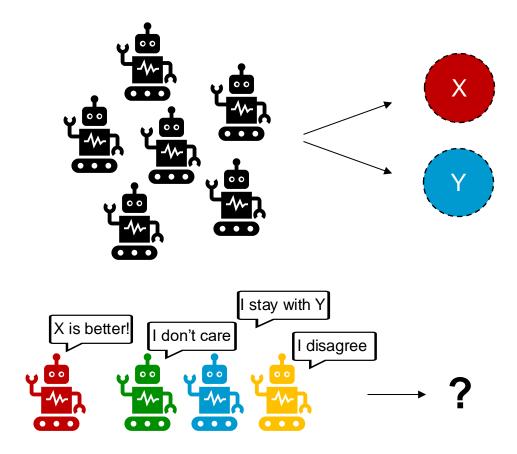


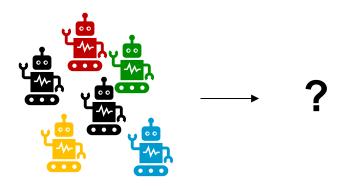




Ideally, consensus will be achieved with certain speed and accuracy

## Collective decision making



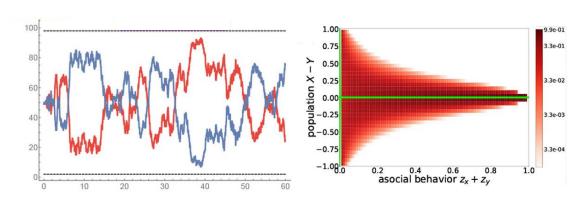


What happens in presence of disruptive (asocial) individuals?

# Collective decision making – disruptive individuals

#### Voter model with zealots

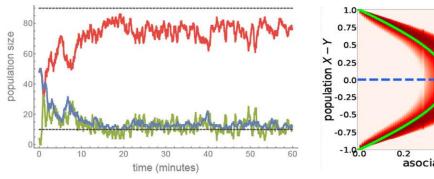
2 species, 2 reactions

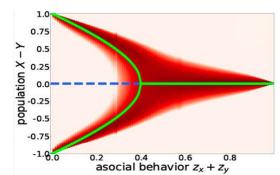


- Permanent indecision with already 4% zealots
- Swarm gets quickly locked into indecision state

#### **Cross-inhibition model with zealots**

3 species, 4 reactions



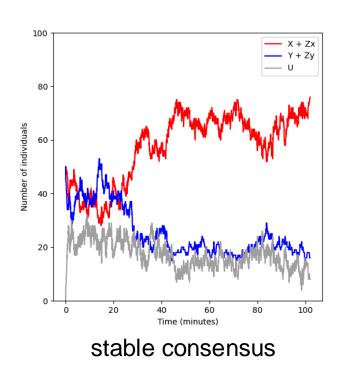


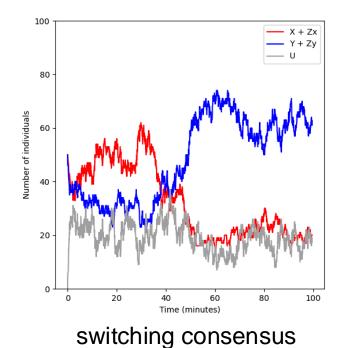
- Stable dynamics for 20% zealots!
- Swarm demonstrates resilience against relatively high levels of asocial behaviour

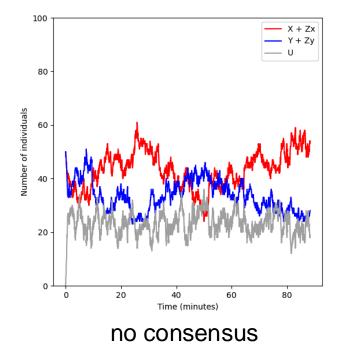
Reina, A., Zakir, R., De Masi, G., Ferrante, E.: Cross-inhibition leads to group consensus despite the presence of strongly opinionated minorities and asocial behaviour. Communications Physics 6(1), 236 (2023)

# **Motivation – group dynamics**

Stochastic **cross-inhibition model with 30% zealots** gives three qualitatively different scenarios!







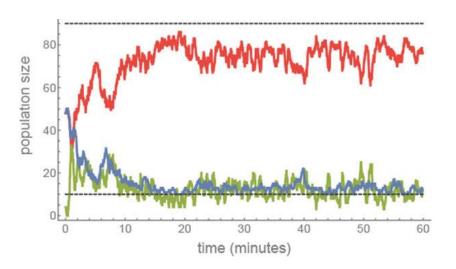
## **Cross-inhibition model**

$$(1) X + Y \stackrel{q_X}{\to} X + U$$

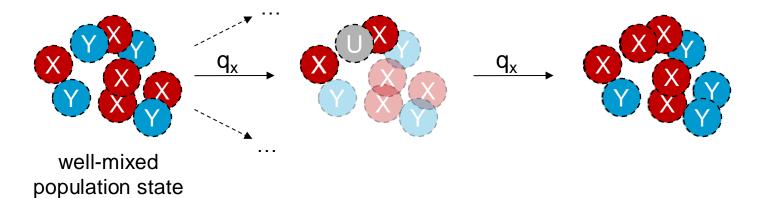
$$(2) X + Y \stackrel{q_y}{\to} U + Y$$

$$(3) X + U \stackrel{q_X}{\to} 2X$$

$$(4) Y + U \stackrel{q_y}{\to} 2Y$$



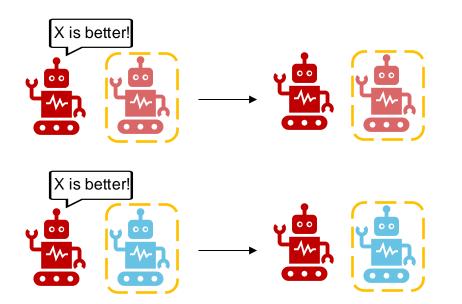
Swarm state evolves stochastically as a continuous-time Markov chain



## **Disruptive individuals**

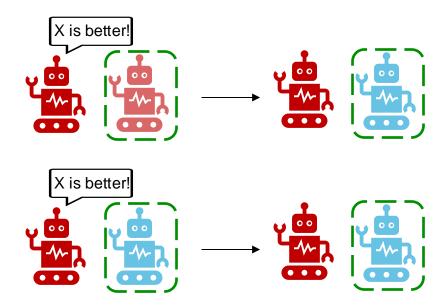


Stubborn individuals which never change their own opinion





Individuals which counter the opinion of the individual they interact with



## Cross-inhibition model with disruptive individuals

#### **Zealot dynamics**



$$(5) X + ZY \stackrel{q_y}{\to} U + Zy$$

$$(6) U + ZY \stackrel{q_y}{\to} Y + ZY$$

$$(7) Y + ZX \stackrel{q_X}{\to} U + ZX$$

$$(8) U + Z_X \stackrel{q_X}{\to} X + ZX$$

#### **Cross-inhibition model**

$$(1) X + Y \stackrel{q_{\chi}}{\to} X + U$$

$$(2) X + Y \stackrel{q_y}{\to} U + Y$$

$$(3) X + U \stackrel{q_{\chi}}{\to} 2X$$

$$(4) Y + U \stackrel{q_y}{\to} 2Y$$

#### **Contrarian dynamics**



$$(5) X + C_Y \stackrel{q_y}{\to} U + C_y$$

$$(6) U + C_Y \stackrel{q_Y}{\to} Y + C_Y$$

$$(7) X + C_X \stackrel{q_X}{\to} X + C_Y$$

$$(8) Y + C_X \stackrel{q_X}{\to} U + C_X$$

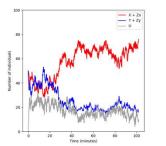
$$(9) U + CX \stackrel{q_X}{\to} X + CX$$

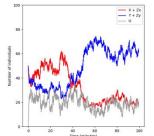
$$(10) Y + CY \stackrel{q_y}{\to} Y + CX$$

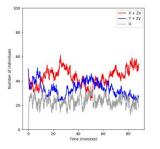
$$(11) CX + CX \stackrel{q_X}{\to} CY + CY$$

$$(12) CY + CY \stackrel{q_y}{\to} CX + CX$$

## **Research Questions**







## 1. Robustness analysis

How does the amount of disruptive individuals affect consensus reaching/switching?

#### 2. Combined effect

How does the **combination** of zealots and contrarians affect consensus reaching/switching?

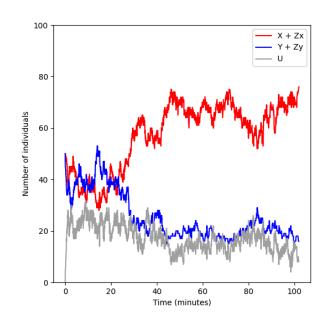
## 3. Group size effect

How does the **group size** affect consensus reaching/switching?

#### Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

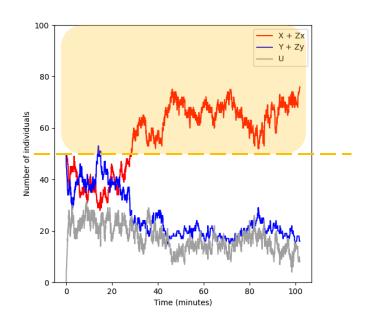
$$F_{\leq t}(G_{\leq h}(((x+Z_x+C_x\geq min_m) \land ((x+Z_x+C_x)-(y+Z_y+C_y)\geq d)) \lor ((y+Z_y+C_y\geq min_m) \land ((y+Z_y+C_y)-(x+Z_x+C_x)\geq d))))$$



#### Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

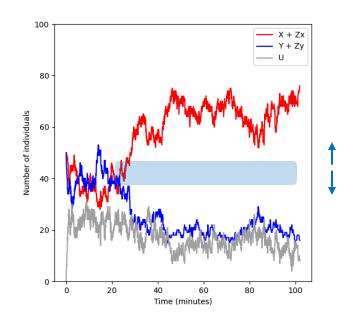
$$F_{\leq t}(G_{\leq h}(((x + Z_x + C_x \geq min_m) \land ((x + Z_x + C_x) - (y + Z_y + C_y) \geq d)) \lor ((y + Z_y + C_y \geq min_m) \land ((y + Z_y + C_y) - (x + Z_x + C_x) \geq d)))))$$



#### Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

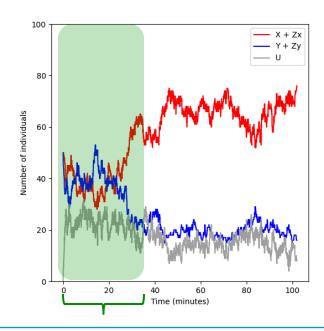
$$F_{\leq t}(G_{\leq h}(((x + Z_x + C_x \geq min_m) \land ((x + Z_x + C_x) - (y + Z_y + C_y) \geq d)) \lor ((y + Z_y + C_y \geq min_m) \land ((y + Z_y + C_y) - (x + Z_x + C_x) \geq d)))))$$



#### Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

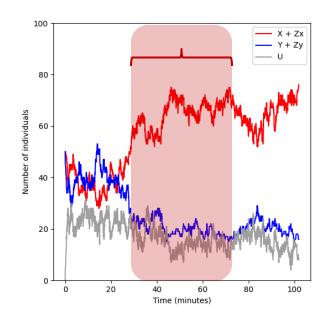
$$F_{\leq t} G_{\leq h}(((x + Z_x + C_x \geq min_m) \land ((x + Z_x + C_x) - (y + Z_y + C_y) \geq d)) \lor ((y + Z_y + C_y \geq min_m) \land ((y + Z_y + C_y) - (x + Z_x + C_x) \geq d)))))$$



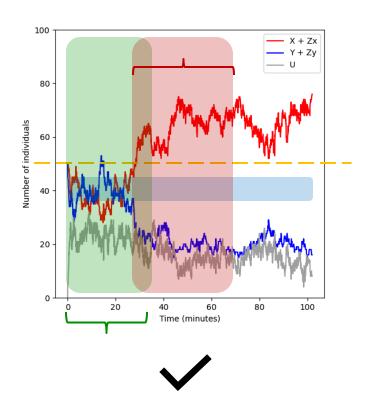
#### Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

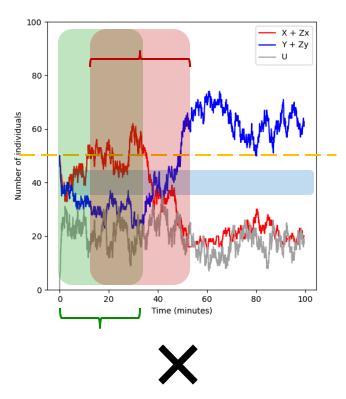
STEP 1: Formally describe stable consensus and switching consensus in BLTL

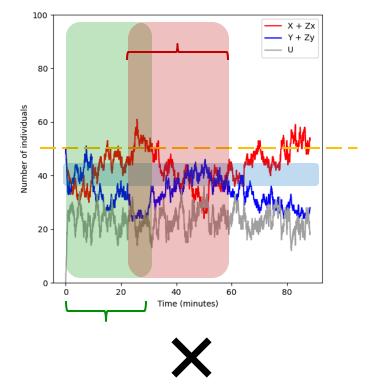
$$F_{\leq t}(G_{\leq h}((x + Z_x + C_x \geq min_m) \land ((x + Z_x + C_x) - (y + Z_y + C_y) \geq d)) \lor ((y + Z_y + C_y \geq min_m) \land ((y + Z_y + C_y) - (x + Z_x + C_x) \geq d)))))$$



#### Is this a stable consensus?



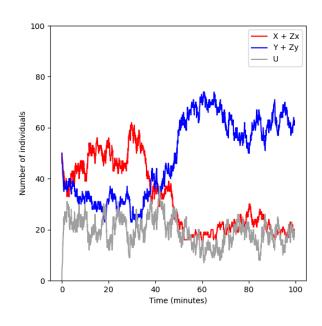




## Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

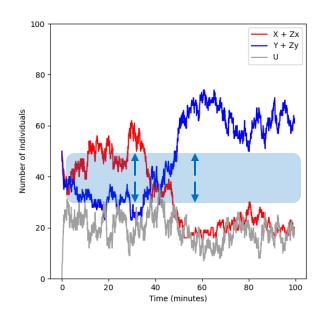
$$F_{\leq t}((((x + Z_x + C_x) - (y + Z_y + C_y) \geq d) \land (true \ U_{\leq s}((y + Z_y + C_y) - (x + Z_x + C_x) \geq d))) \lor (((y + Z_y + C_y) - (x + Z_x + C_x) \geq d) \land (true \ U_{\leq s}((x + Z_x + C_x) - (y + Z_y + C_y) \geq d))))$$



## Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

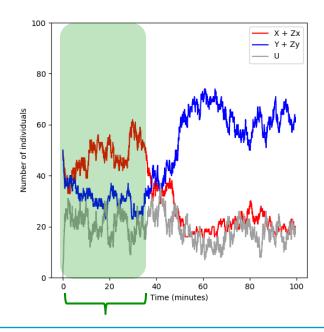
$$F_{\leq t}((((x + Z_x + C_x) - (y + Z_y + C_y) \geq d) \land (true \ U_{\leq s}((y + Z_y + C_y) - (x + Z_x + C_x) \geq d))) \lor (((y + Z_y + C_y) - (x + Z_x + C_x) \geq d) \land (true \ U_{\leq s}((x + Z_x + C_x) - (y + Z_y + C_y) \geq d))))$$



## Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

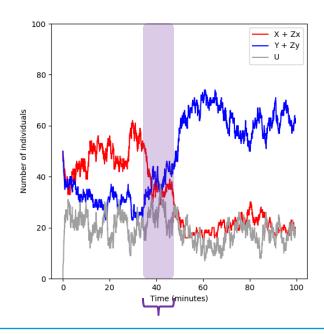
$$F_{\leq t}(((x + Z_x + C_x) - (y + Z_y + C_y) \geq d) \land (true \ U_{\leq s}((y + Z_y + C_y) - (x + Z_x + C_x) \geq d))) \lor (((y + Z_y + C_y) - (x + Z_x + C_x) \geq d) \land (true \ U_{\leq s}((x + Z_x + C_x) - (y + Z_y + C_y) \geq d))))$$



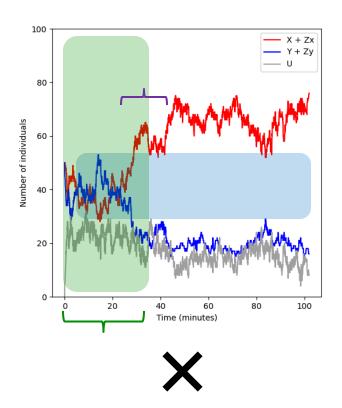
## Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

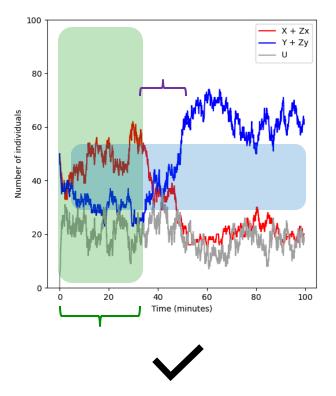
STEP 1: Formally describe stable consensus and switching consensus in BLTL

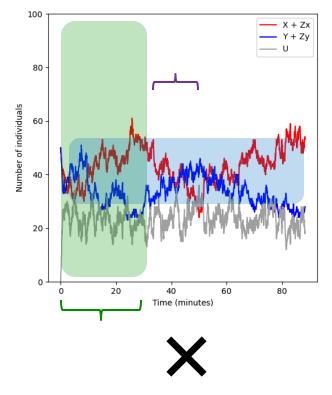
$$F_{\leq t}((((x + Z_x + C_x) - (y + Z_y + C_y) \geq d) \land (true \ U_{\leq s}((y + Z_y + C_y) - (x + Z_x + C_x) \geq d))) \lor (((y + Z_y + C_y) - (x + Z_x + C_x) \geq d) \land (true \ U_{\leq s}((x + Z_x + C_x) - (y + Z_y + C_y) \geq d))))$$



## Is this a switching consensus?







Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

STEP 1: Formally describe stable consensus and switching consensus in BLTL

- Five parameters: majority *m*, distance *d*, reaching time *t*, holding time *h*, switching time *s* 

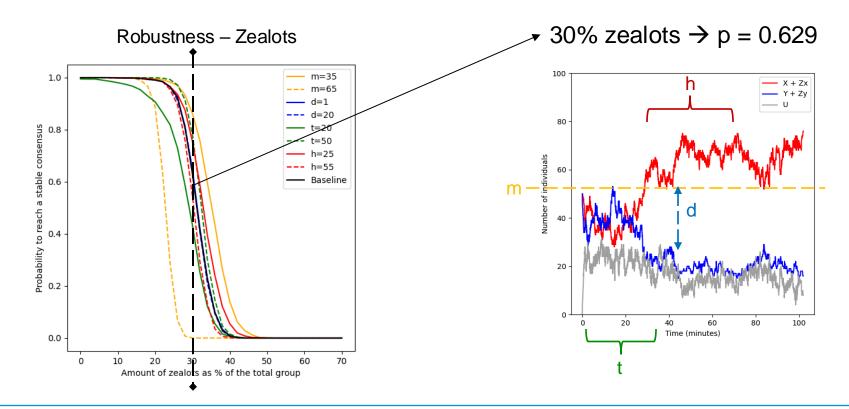
**STEP 2:** Apply model checking tools (*PRISM* and *PlasmaLab*) to explore the relevant scenarios:

- Varying number of zealots and contrarians to explore robustness
- Varying number of both to explore combined effect
- Varying total group size to explore group size effect
- Monte Carlo algorithm to estimate satisfaction probability
- Error margin ε=0.025, confidence bound Δ=0.01



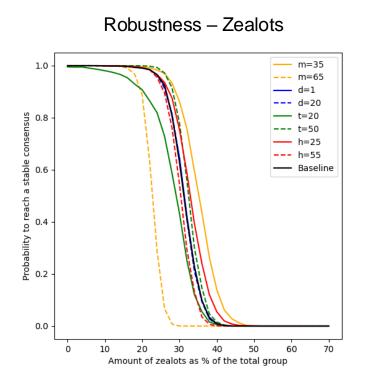
## Results – robustness of stable consensus

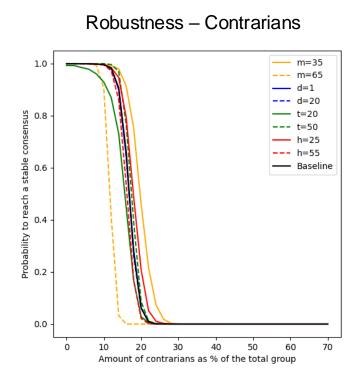
- Scenario: N = 100 robots, equivalent options X and Y ( $q_x = q_y$ ), initially #X=#Y, #U=0, #Zx=#Zy, #Cx=#Cy
- Baseline: m=50, d=10, t=35, h=40

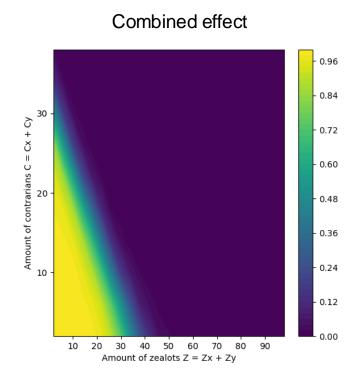


## Results – robustness of stable consensus

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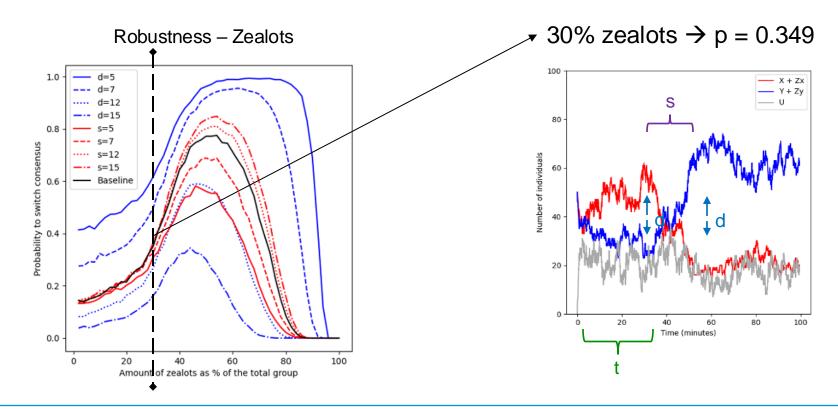




## Results – robustness of switching consensus

• Scenario: N = 100 robots, equivalent options X and Y  $(q_x = q_y)$ , initially #X = #Y, #U = 0, #Zx = #Zy, #Cx = #Cy

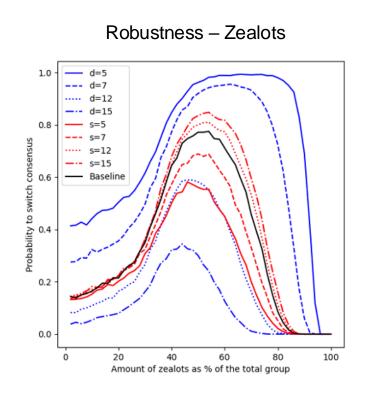
• Baseline: d=10, t=35, s=10

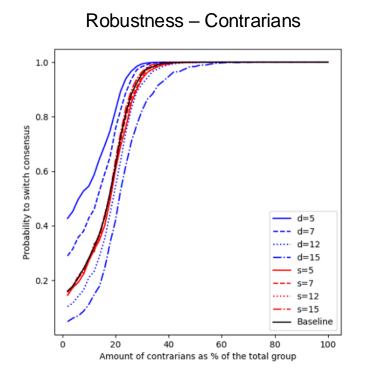


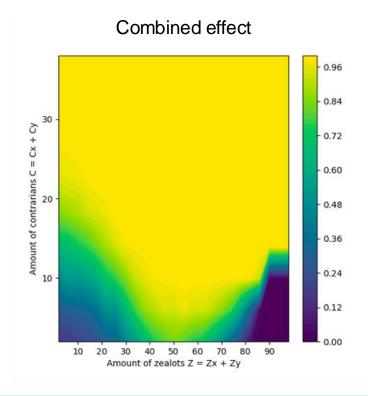
## Results – robustness of switching consensus

• Scenario: N = 100 robots, equivalent options X and Y  $(q_x=q_y)$ , initially #X=#Y, #U=0, #Zx=#Zy, #Cx=#Cy

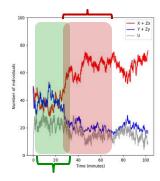
• Baseline: d=10, t=35, s=10





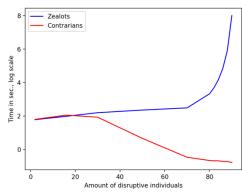


# Results – expected times



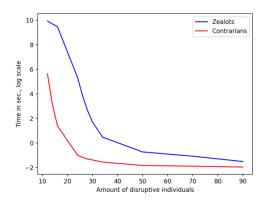
#### **Expected times to <u>reach</u> consensus**

#	2	16	30	50	70	80	82	84	86	88	90
Zealots	5.95	7.28	9.02	10.57	12.04	27.82	39.94	64.95	128.85	374.04	2975.68
Contrarians	6.07	7.81	6.89	1.95	0.63	0.52	0.51	0.51	0.49	0.49	0.46

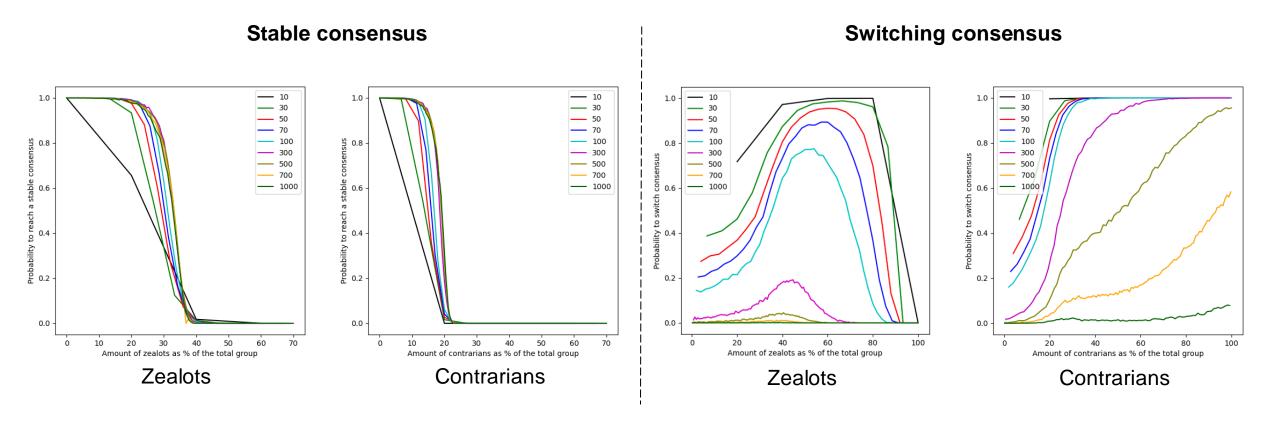


### **Expected times to hold consensus**

#	12	14	16	24	26	28	30	34	50	70	90
Zealots	20686.51	16368.28	13047.85	210.98	47.71	14.13	5.46	1.61	0.48	0.34	0.22
Contrarians	283.57	22.53	4.03	0.37	0.31	0.27	0.25	0.21	0.16	0.15	0.14



# Results – group size effect



... robust to group size scaling!

... **sensitive** to group size scaling!

## **Conclusion and outlook**

- > A small increase of disruptive individuals can drastically affect consensus dynamics
- > Our method with SMC allows to explore consensus beyond mean-field analysis or single simulation

#### > Stable consensus

- > Cross-inhibition model robust up to certain fraction of zealots/contrarians, then rapid phase transition
- Zealots are less harmful for reaching consensus than contrarians

#### Switching consensus

- > Range of zealots for which such trajectories occur with non-negligible probability, but very rare for high number of zealots
- Contrarians promote switching dynamics

#### > Future work

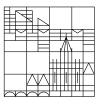
- > Group size effect: characterisation of a class of stochastic systems for which consensus reaching is robust to scaling
- Asymmetric model: what if only one decision is correct?
- Control theory: interventions over individuals for a global outcome







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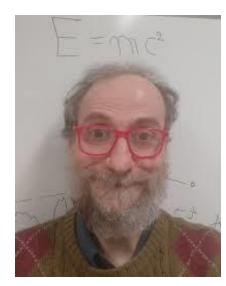
# Thank you very much!

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University of Konstanz

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Tatjana Petrov



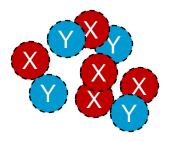
Alberto d'Onofrio

$$(1) X + Y \stackrel{q_X}{\to} X + X$$

$$(2) X + Y \xrightarrow{q_y} Y + Y$$

Swarm state evolves as a continuous-time Markov chain

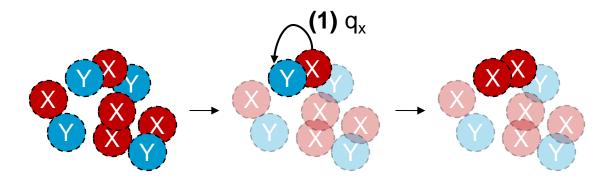
→ classical population model



$$(1) X + Y \stackrel{q_X}{\rightarrow} X + X$$

$$(2) X + Y \stackrel{q_y}{\to} Y + Y$$

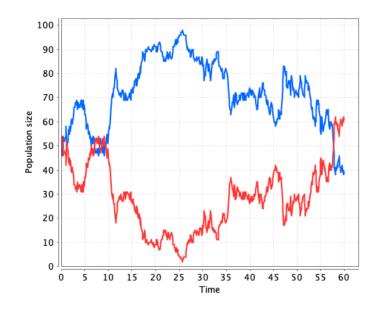
Swarm state evolves as a continuous-time Markov chain
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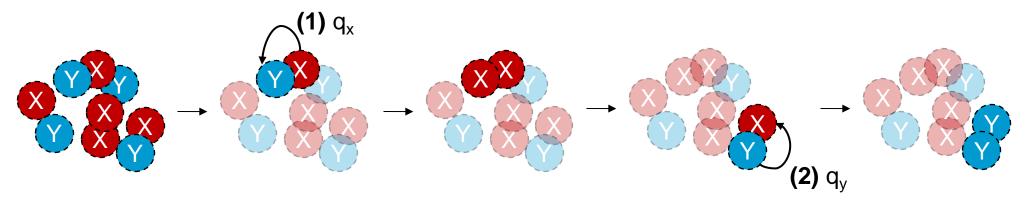


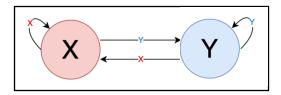
$$(1) X + Y \stackrel{q_X}{\to} X + X$$

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Swarm state evolves as a continuous-time Markov chain
→ classical population model





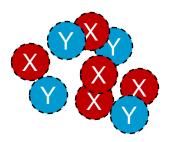


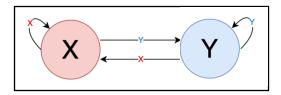
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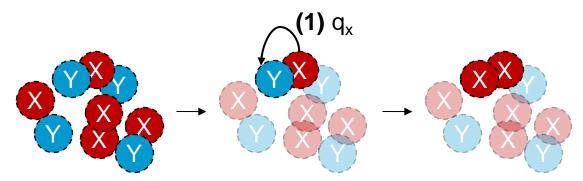


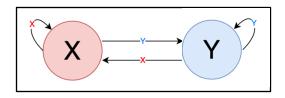
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Swarm state evolves as a continuous-time Markov chain

→ classical population model

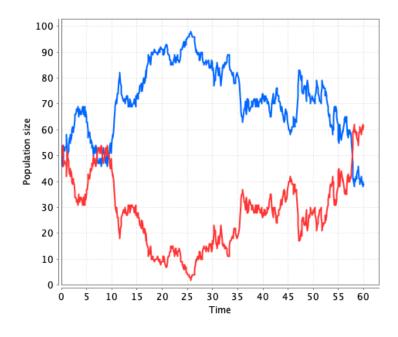


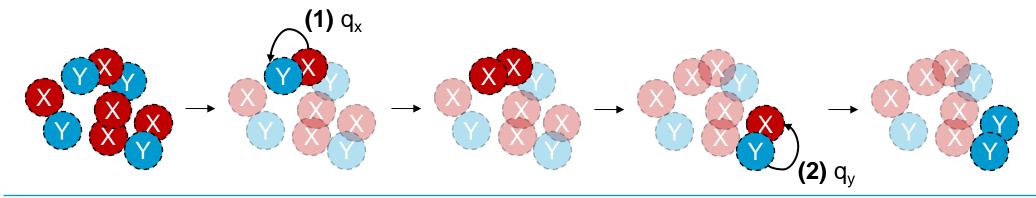


$$(1) X + Y \stackrel{q_X}{\to} X + X$$

$$(2) X + Y \stackrel{q_y}{\rightarrow} Y + Y$$

Swarm state evolves as a continuous-time Markov chain
→ classical population model

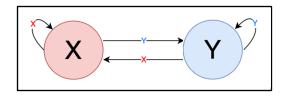




## **Decision making: Voter model**

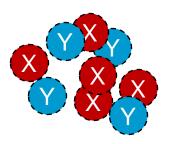
$$(1) X + Y \stackrel{q_X}{\to} X + X$$

$$(2) X + Y \stackrel{q_y}{\to} Y + Y$$



Swarm state evolves as a continuous-time Markov chain

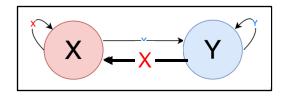
→ classical population model



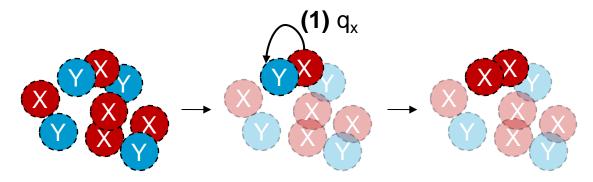
## **Decision making: Voter model**

$$(1) X + Y \stackrel{q_X}{\rightarrow} X + X$$

$$(2) X + Y \xrightarrow{q_y} Y + Y$$



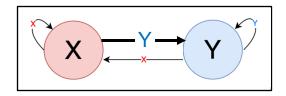
Swarm state evolves as a continuous-time Markov chain
→ classical population model



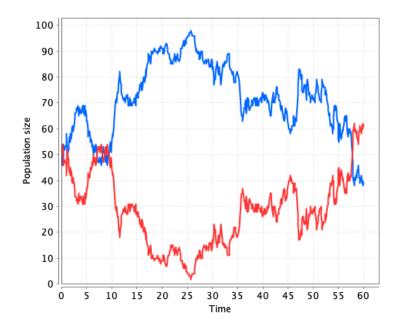
## **Decision making: Voter model**

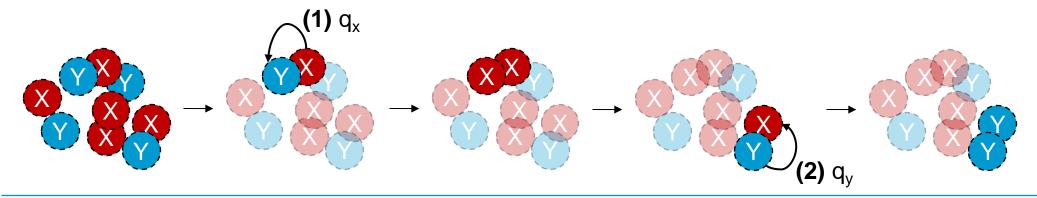
$$(1) X + Y \stackrel{q_X}{\to} X + X$$

$$(2) X + Y \stackrel{q_y}{\rightarrow} Y + Y$$



Swarm state evolves as a continuous-time Markov chain
→ classical population model

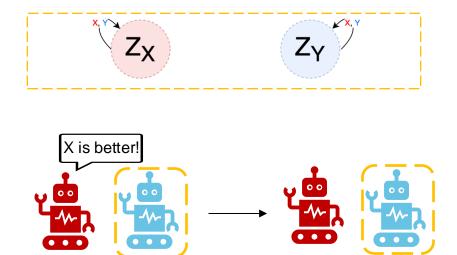




#### **Disruptive individuals**

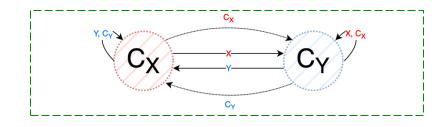
#### **Zealots**

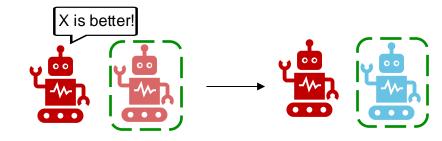
Stubborn individuals which never change their own opinion



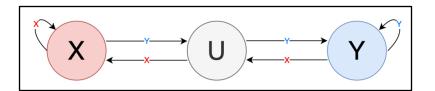
#### **Contrarians**

 Individuals which counter the opinion of the individual they interact with





## **Decision making: cross-inhibition model**

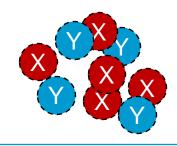


$$(1) X + Y \stackrel{q_X}{\to} X + U$$

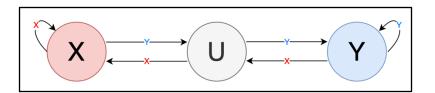
$$(2) X + Y \stackrel{q_y}{\to} U + Y$$

$$(3) X + U \stackrel{q_X}{\to} 2X$$

$$(4) Y + U \stackrel{q_y}{\to} 2Y$$



## **Decision making: cross-inhibition model**

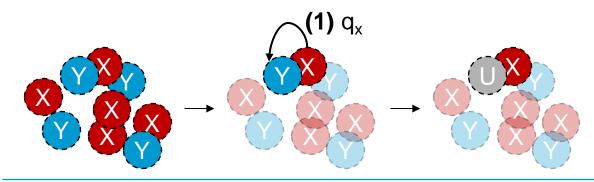


$$(1) X + Y \stackrel{q_X}{\rightarrow} X + U$$

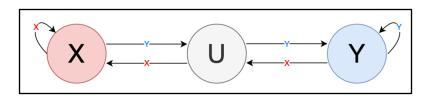
$$(2) X + Y \stackrel{q_y}{\to} U + Y$$

$$(3) X + U \stackrel{q_X}{\to} 2X$$

$$(4) Y + U \stackrel{q_y}{\to} 2Y$$



#### Decision making: cross-inhibition model

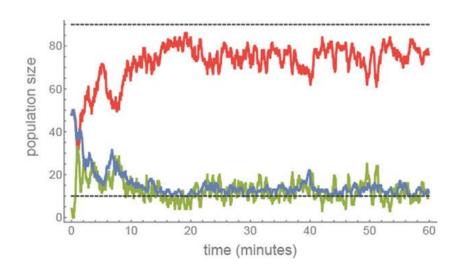


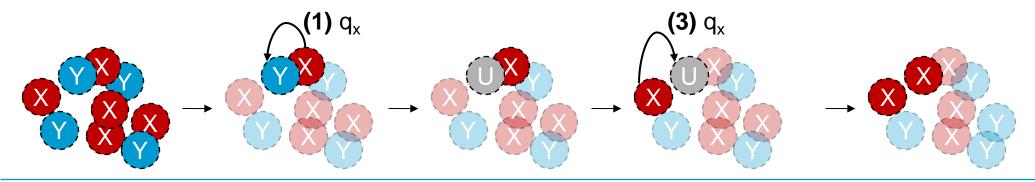
$$(1) X + Y \stackrel{q_X}{\to} X + U$$

$$(2) X + Y \stackrel{q_y}{\to} U + Y$$

$$(3) X + U \stackrel{q_X}{\rightarrow} 2X$$

$$(4) Y + U \stackrel{q_y}{\to} 2Y$$

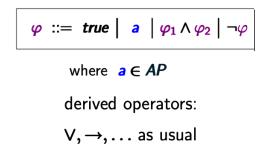




## **Approach**

#### Statistical Model Checking of properties in Bounded Linear Temporal Logic (BLTL)

based on Linear Temporal Logic (LTL)



 BLTL: (F, G, X, U, W) are bounded by temporal bound

$$F_{\leq t} \phi$$

$$G_{\leq t} \phi$$

$$X_{\leq t} \phi$$

$$U_{\leq t} \phi$$

$$W_{\leq t} \phi$$

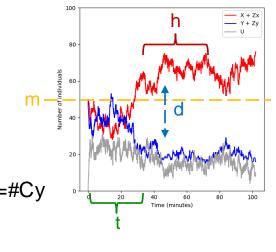
Textual	Symbolic	Explanation	Diagram				
Unary op	perators:						
Χφ	$\bigcirc \varphi$	ne <b>X</b> t: $\phi$ has to hold at the next state.	•—	φ	····>• —	→•	>
<b>F</b> φ	$\Diamond \varphi$	Finally: $\phi$ eventually has to hold (somewhere on the subsequent path).	•—	→•······	φ	→•	>
Gφ	$\Box \varphi$	<b>G</b> lobally: $\phi$ has to hold on the entire subsequent path.	φ	φ	φ	φ	φ
Binary o	perators:						
ψ <b>U</b> φ	$\psi\mathcal{U}arphi$	<b>U</b> ntil: $\psi$ has to hold <i>at least</i> until $\phi$ becomes true, which must hold at the current or a future position.	Ψ	Ψ	Ψ	φ	>
ψ <b>R</b> φ	$\psi  \mathcal{R}  arphi$	Release: $\phi$ has to be true until and including the point where $\psi$ first becomes true; if $\psi$ never becomes true, $\phi$ must remain true forever.	φ •— φ	φ ••••••••••••••••••••••••••••••••••••	φ φ	φ,ψ ••••••••••••••••••••••••••••••••••••	φ>
ψ <b>W</b> φ	$\psi \mathcal{W} arphi$	<b>W</b> eak until: $\psi$ has to hold <i>at least</i> until $\phi$ ; if $\phi$ never becomes true, $\psi$ must remain true forever.	Ψ	Ψ ••••••••••••••••••••••••••••••••••••	Ψ Ψ •••••	φ ••••••••••••••••••••••••••••••••••••	Ψ>

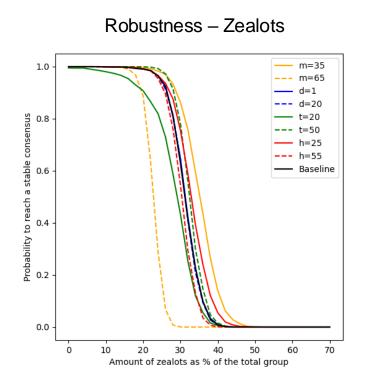
Boyer, B., Corre, K., Legay, A., Sedwards, S.: Plasma-lab: A flexible, distributable statistical model checking library. In: International Conference on Quantitative Evaluation of Systems. pp. 160–164. Springer (2013)

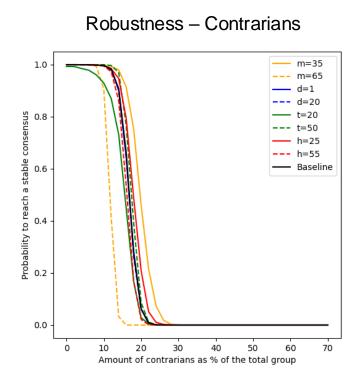
# Results – robustness of stable consensus

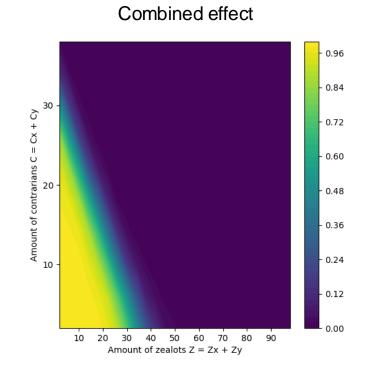
• Scenario: N = 100 robots, equivalent options X and Y  $(q_x=q_y)$ , initially #X=#Y, #U=0, #Zx=#Zy, #Cx=#Cy

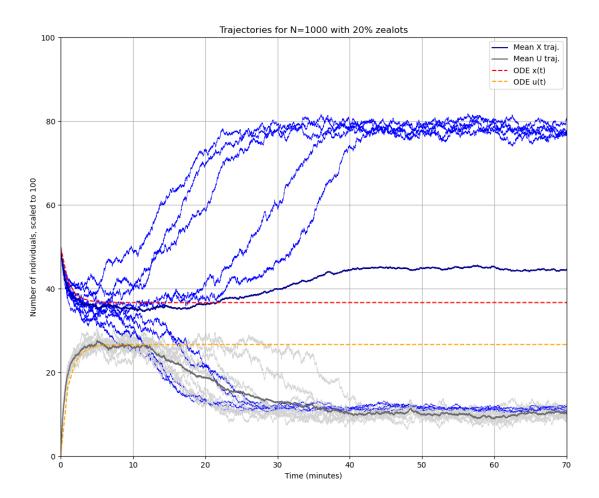
• Baseline: m=50, d=10, t=35, h=40

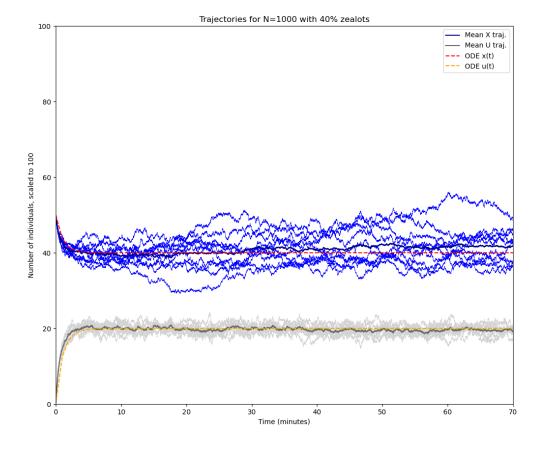






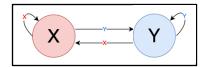


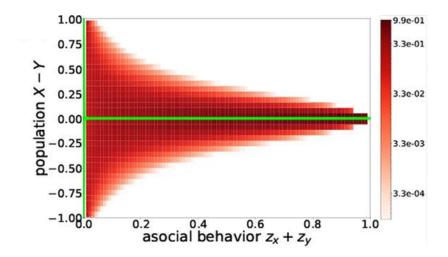




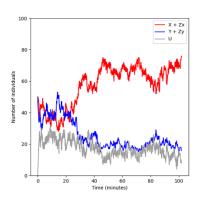
## Studied Model of Decision-Making

**Voter Model** 



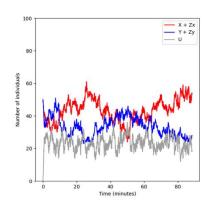


→ in presence of asocial individuals, the swarm gets quickly locked into an indecision state



No zealots

→ quick, stable consensus



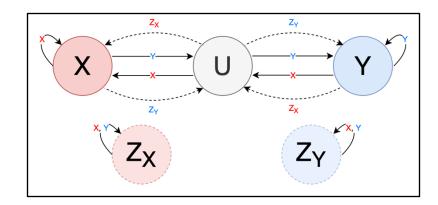
2% zealots

→ permanent indecision

Reina, A., Zakir, R., De Masi, G., Ferrante, E.: Cross-inhibition leads to group consensus despite the presence of strongly opinionated minorities and asocial behaviour. Communications Physics 6(1), 236 (2023)

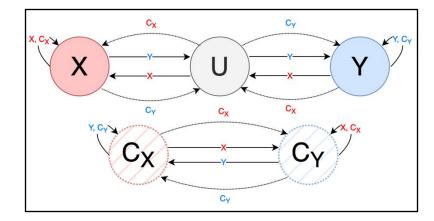
#### **Studied Model with Disruptive Individuals**

#### **Cross-Inhibition model with Zealots**



- Zealots: stubborn individuals which never change their own opinion
- Four additional reactions, where 'pure' agents interact with zealots & adjust their own states

#### **Cross-Inhibition model with Contrarians**



- Contrarians: individuals which counter the opinion of the individual they interact with
- Eight additional reactions, where contrarians influence 'pure' individuals & are influenced by others with the same opinion