Opinion dynamics



Learning outcomes

How things propagate across networks

Compare different models of diffusion

Choose the right model for the right job

Introduction

Started in the 1970s

Original models were from physics

Many disciplines, different uses

Introduction

Ideas, behaviours, rumours, information, opinions, fake news, etc.

Opinion Markets Insight

How herd behaviour drives action on r/WallStreetBets

Oxford research shows influence of social contagion on stocks such as Tesla and GameStop

IAN GOLDIN

+ Add to myFT



Simple models, complex models

Simple contagion

Agents can be "infected" directly

Complex contagion

Non-linear or repeated interactions for agents to be "infected"

Irreversible states

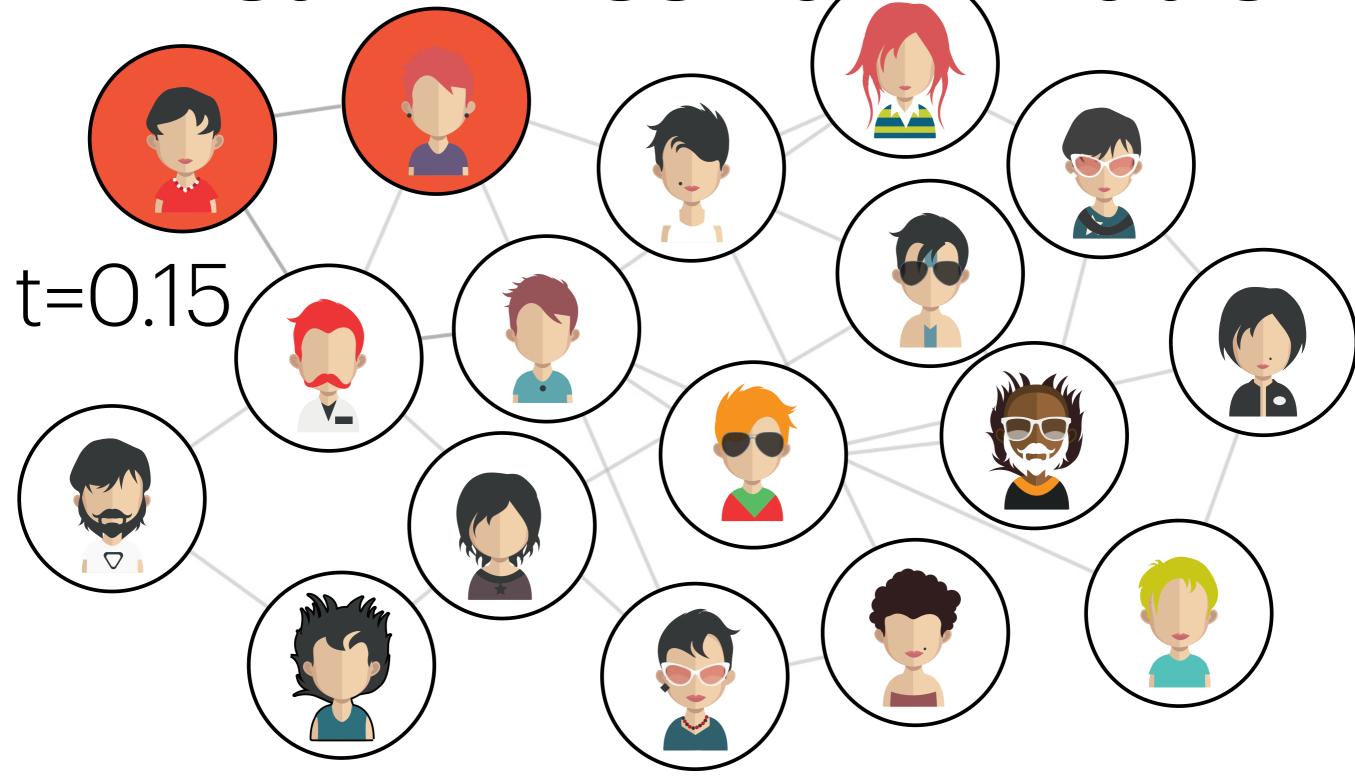
Easy to compute

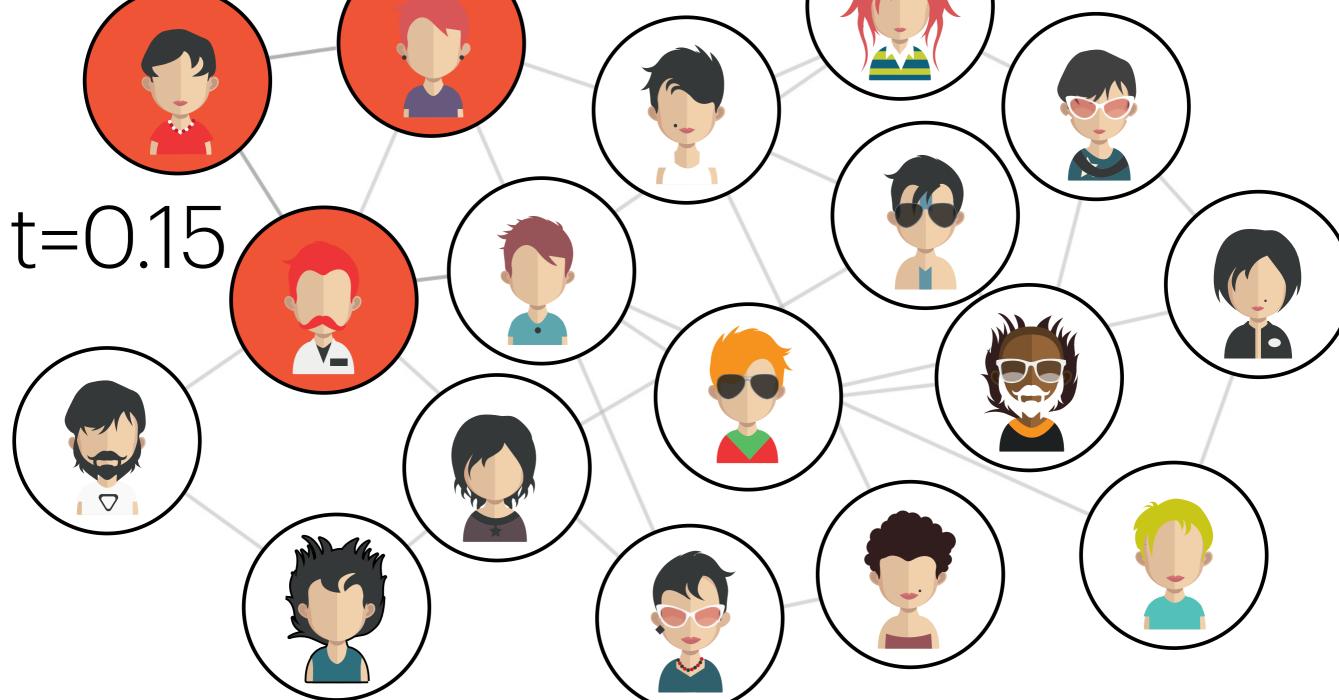
Easy to extend to influence maximisation

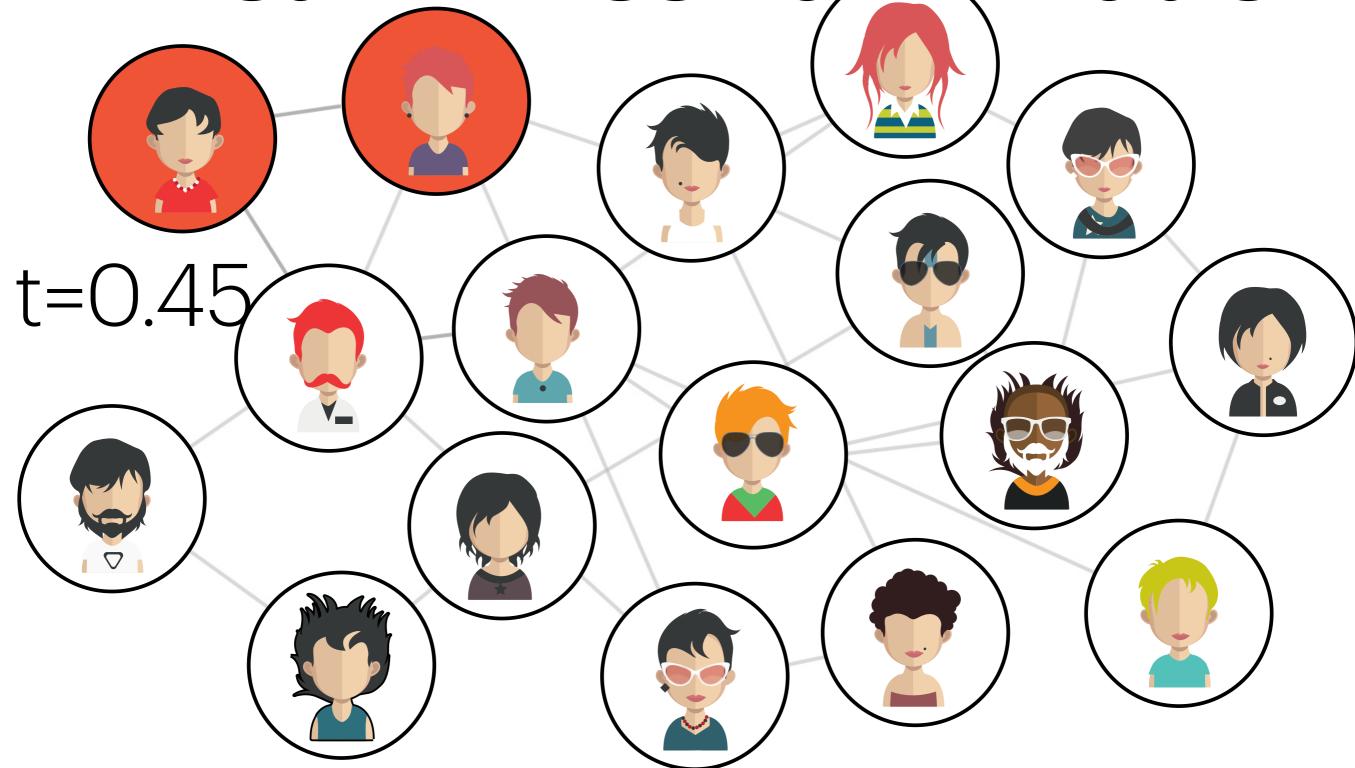
Good for:

information and rumor propagation

Irreversible states
Easy to compute
Toy model for complex contagion







Time to play!

www.ncase.me/crowds

Good for:

Group decisions, majority voting, etc.

Flipping states
Simple, analytical solution for equilibrium
Fails when opinions are fine-grained

Properties

"Always" reach consensus

Analytical formula for equilibrium

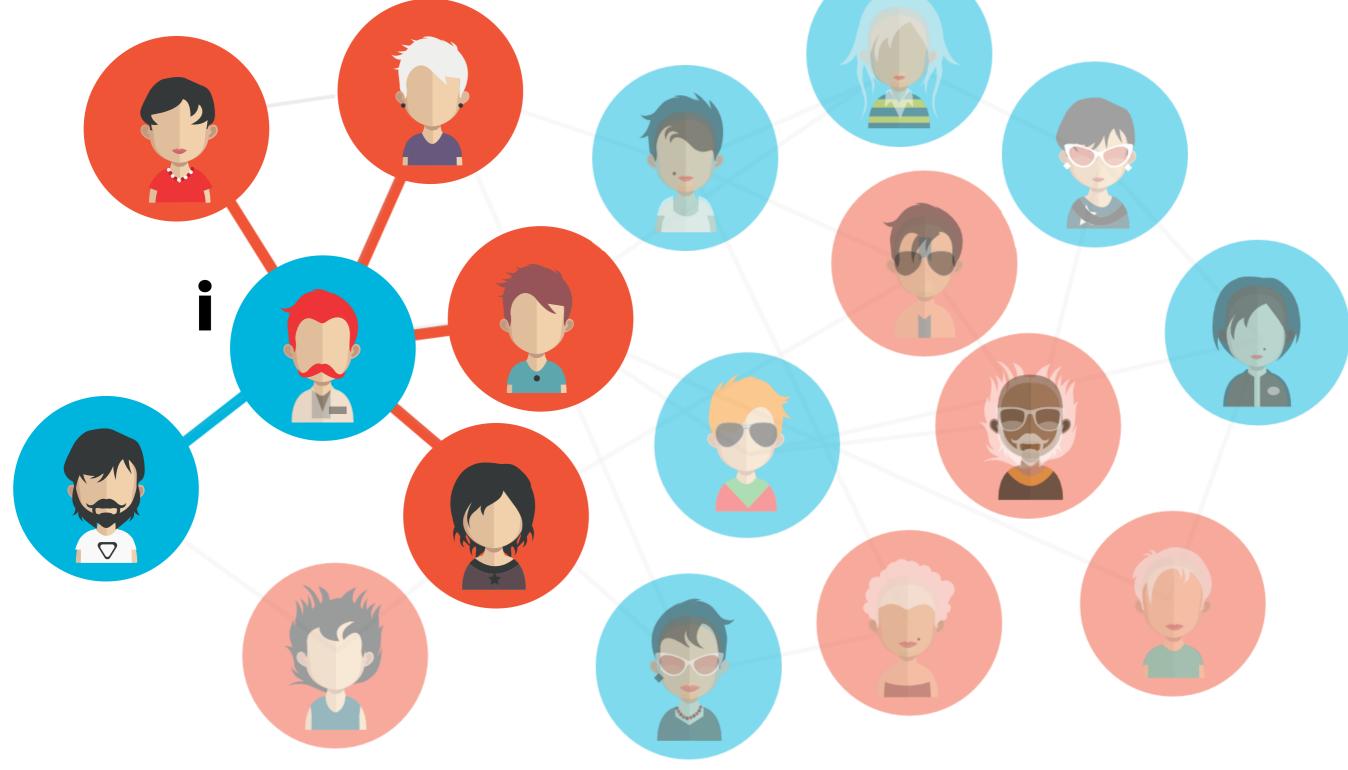
Many possible extensions

1) Pick a random node i2) pick a random neighbour j of node i

nodes

3) change opinion of i to opinion of j

Do this simultaneously for all the





Good for:

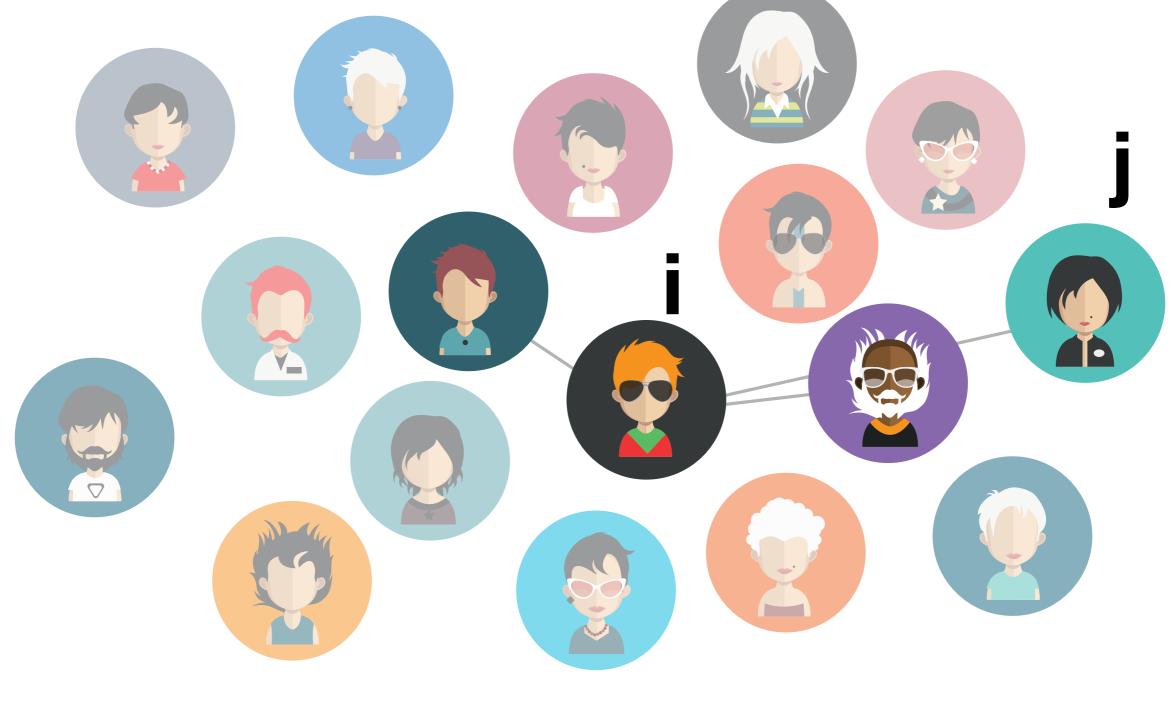
anything with binary states that can be reverted

Bounded confidence models

Continuous opinions
Usually between 0 and 1 but not necessarily

Easy to implement with agent-based models





Equations

$$\begin{cases} o_i(t+1) = o_i(t) + \mu(o_j(t) - o_i(t)) \\ o_j(t+1) = o_j(t) + \mu(o_i(t) - o_j(t)) \end{cases}$$

Opinion agent i

Equations

$$\begin{cases} o_i(t+1) = o_i(t) + \mu(o_j(t) - o_i(t)) \\ o_j(t+1) = o_j(t) + \mu(o_i(t) - o_j(t)) \end{cases}$$

Opinion agent i

Equations

Convergence parameter

$$\begin{cases} o_i(t+1) = o_i(t) + \mu(o_j(t) - o_i(t)) \\ o_j(t+1) = o_j(t) + \mu(o_i(t) - o_j(t)) \end{cases}$$

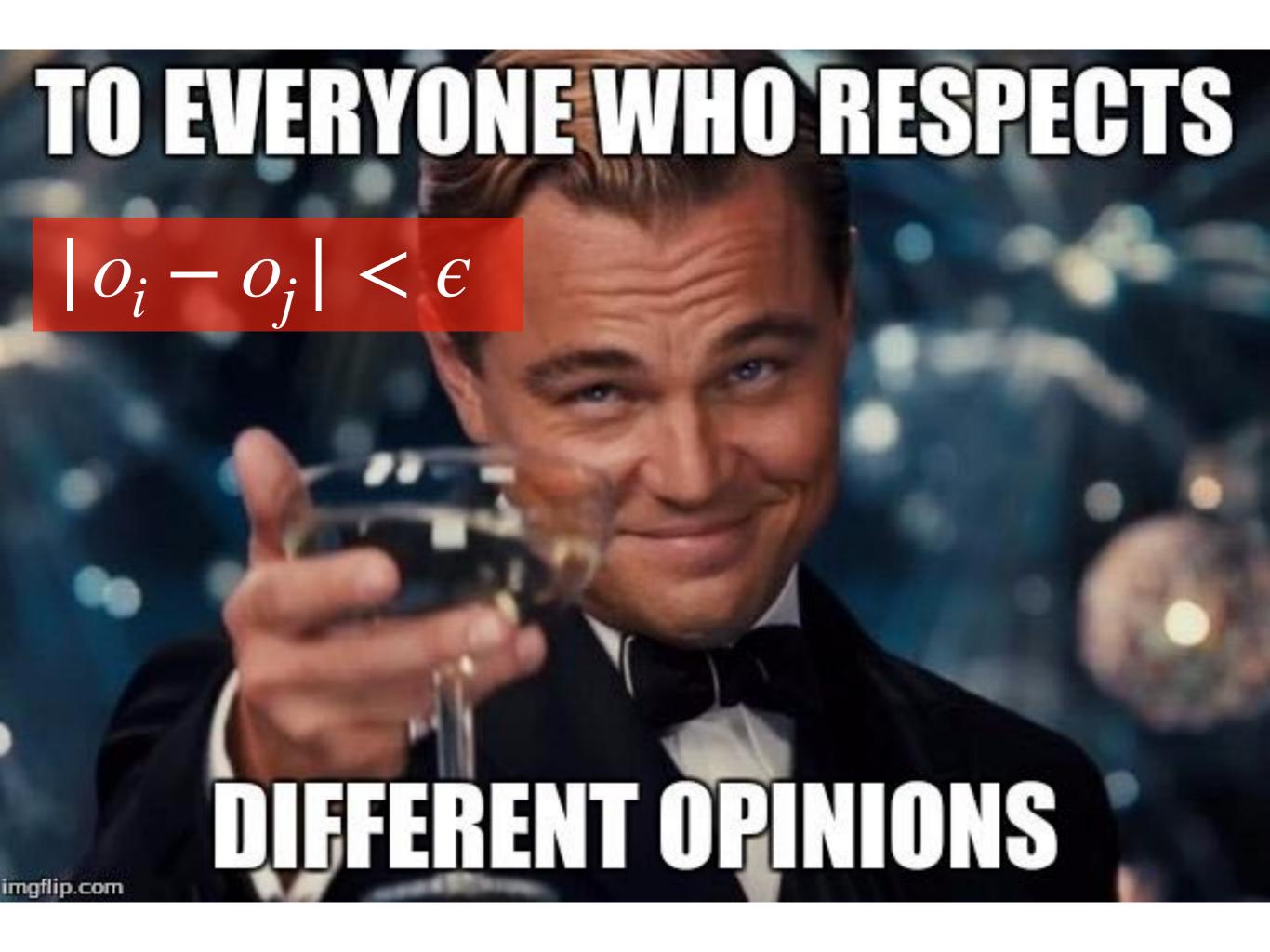
Opinion agent i

Equations

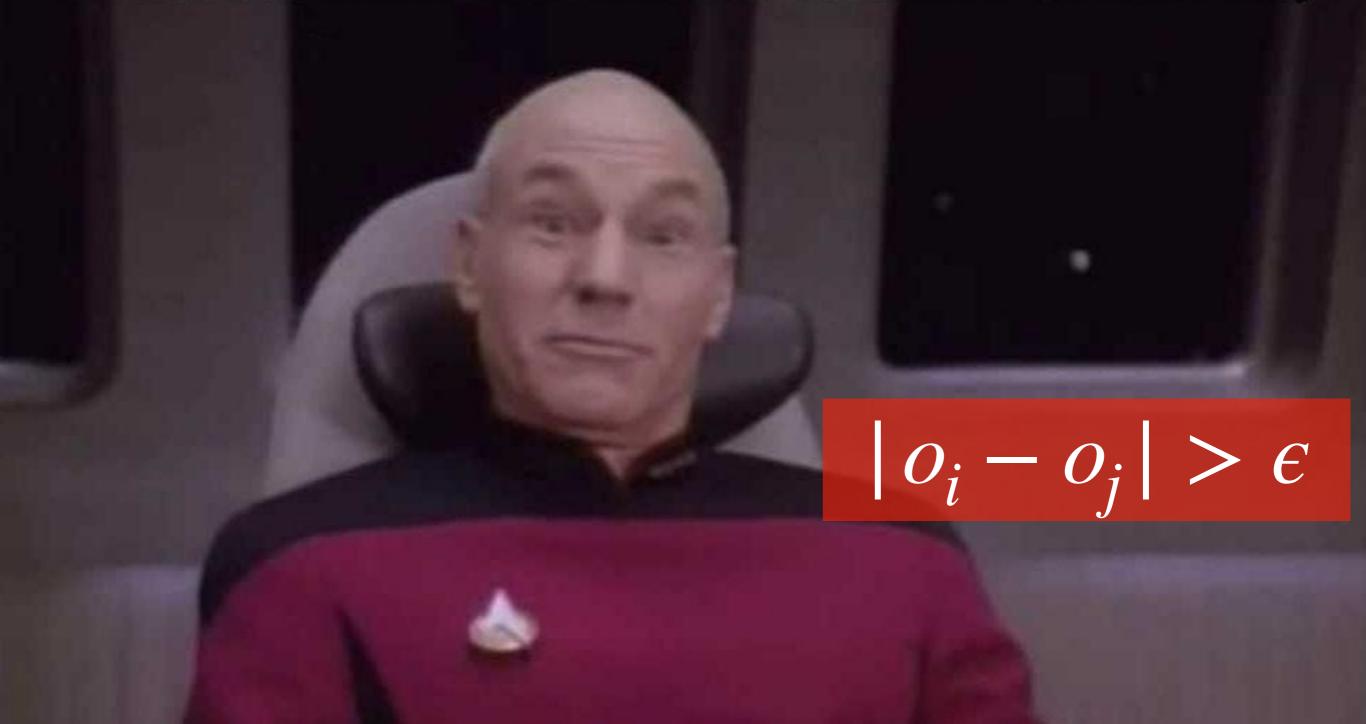
Convergence parameter

$$\begin{cases} o_i(t+1) = o_i(t) + \mu(o_j(t) - o_i(t)) \\ o_j(t+1) = o_j(t) + \mu(o_i(t) - o_j(t)) \end{cases}$$

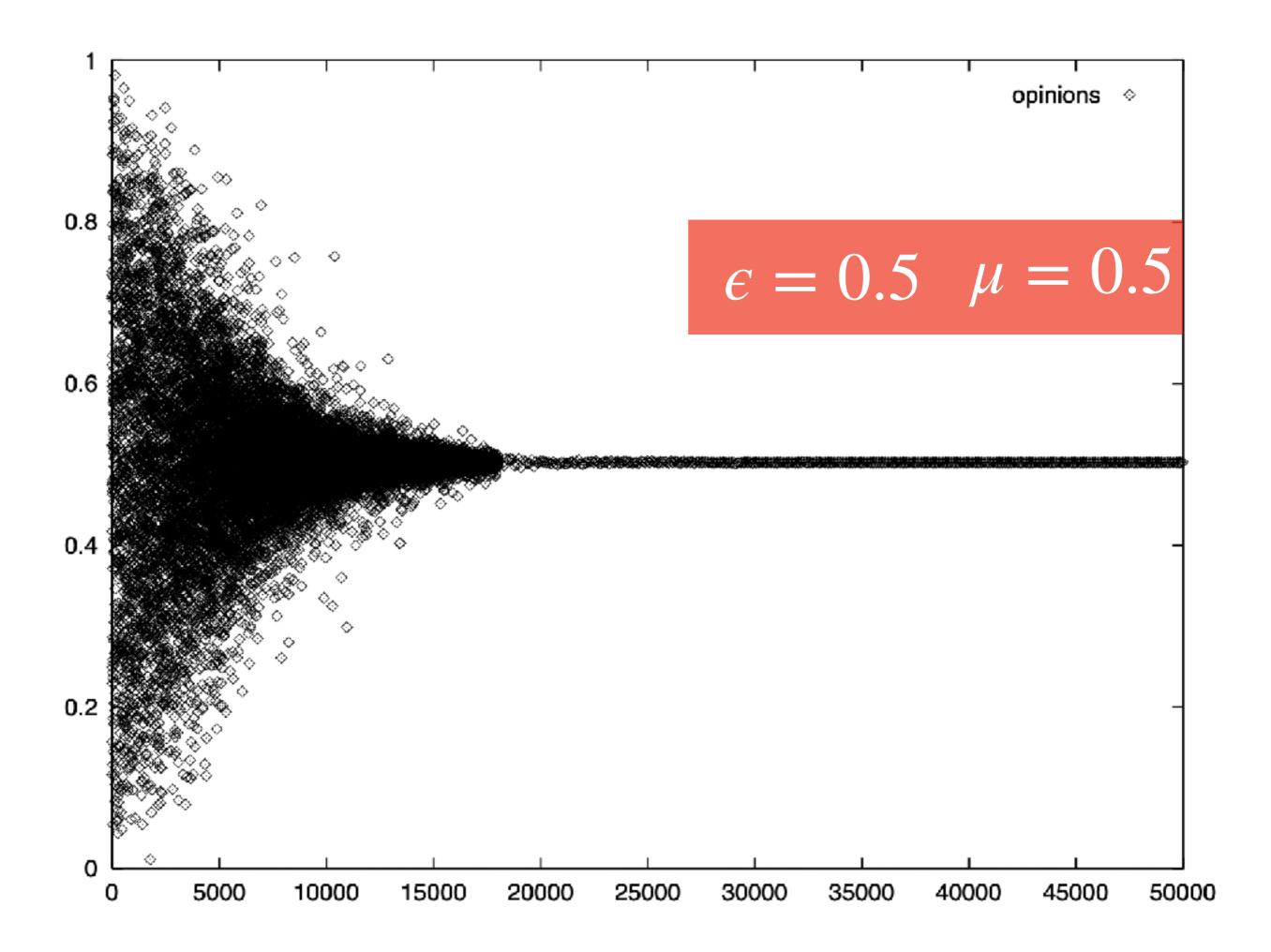
Only If $|o_i(t) - o_j(t)| < \epsilon$

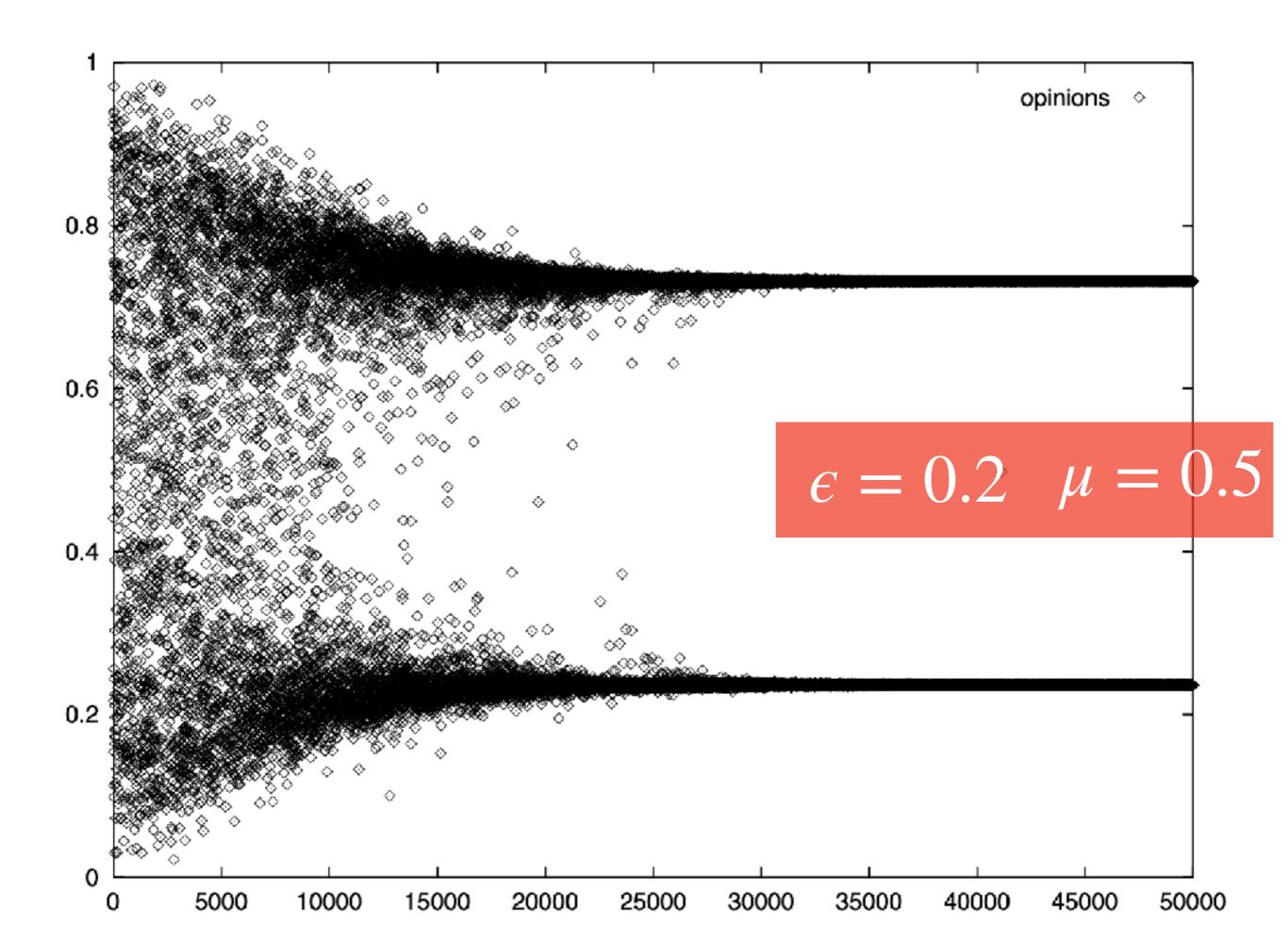


I DON'T ALWAYS LISTEN TO OTHER'S OPINIONS,



BUT WHEN I DO, I PUT ON MY 'I DON'T GIVE A SHIT' FACE.





Deffuant model

Good for:

when you need continuous opinions

Case study



Case study

An opinion dynamics model for prediction markets

Prediction markets



Prediction markets

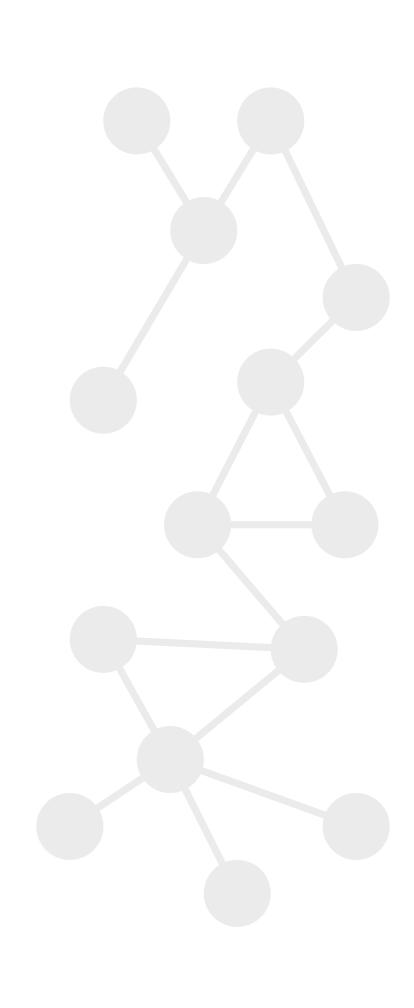
Markets on events that have a **definite** end-point

(e.g. political elections, football match, price of stock at time t)

Contract pays 1 if correct 0 otherwise

Price reflect **perceived probability** of that outcome to occur

Agents start with random opinions (uniform, between 0 and 1)



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They "live" on a barabasi-albert network

Agents start with random opinions (uniform, between 0 and 1)

They "live" on a **barabasi-albert** network

They communicate with random neighbour

Agents start with random opinions (uniform, between 0 and 1)

They "live" on a barabasi-albert network

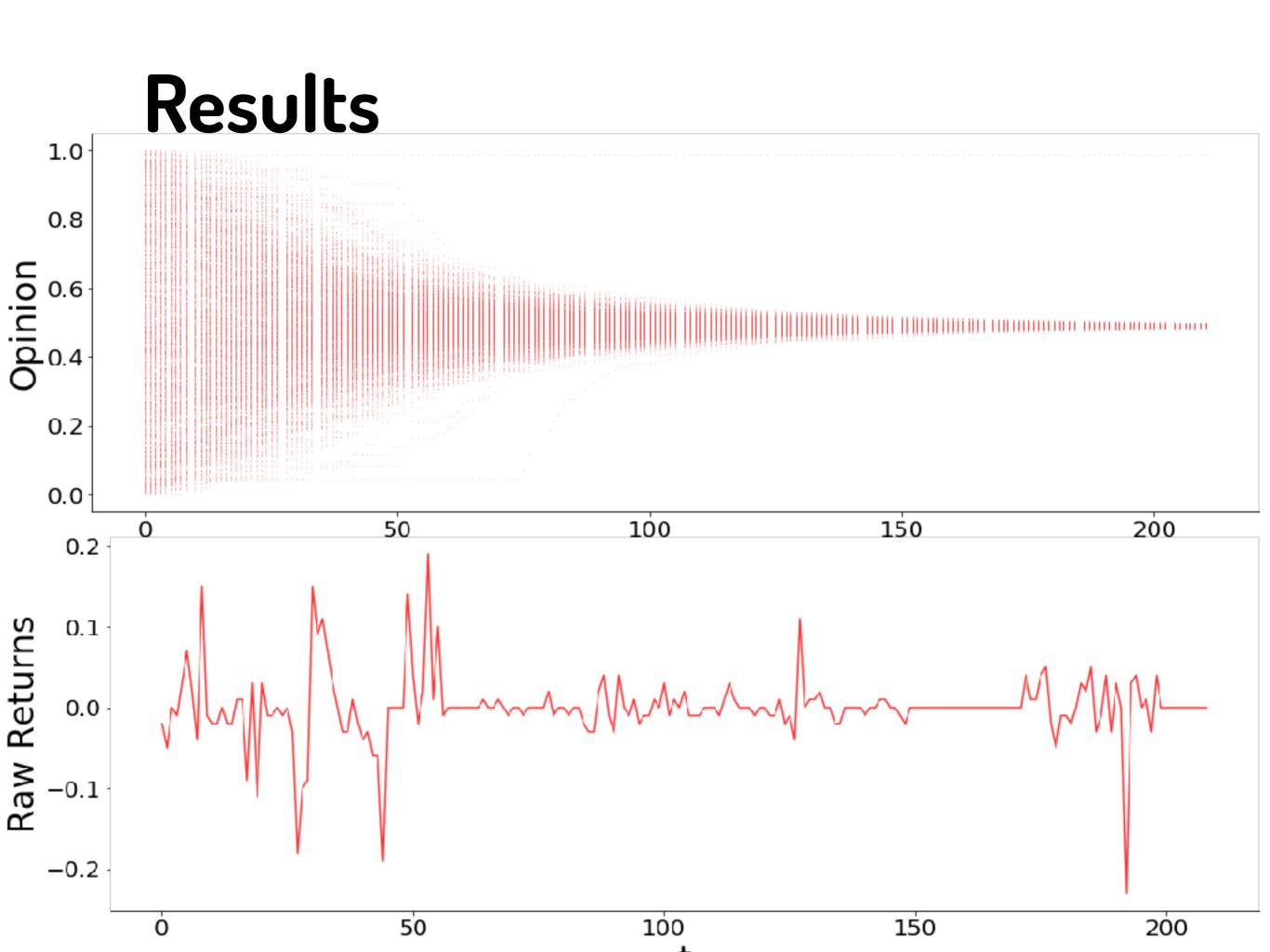
They communicate with random neighbour

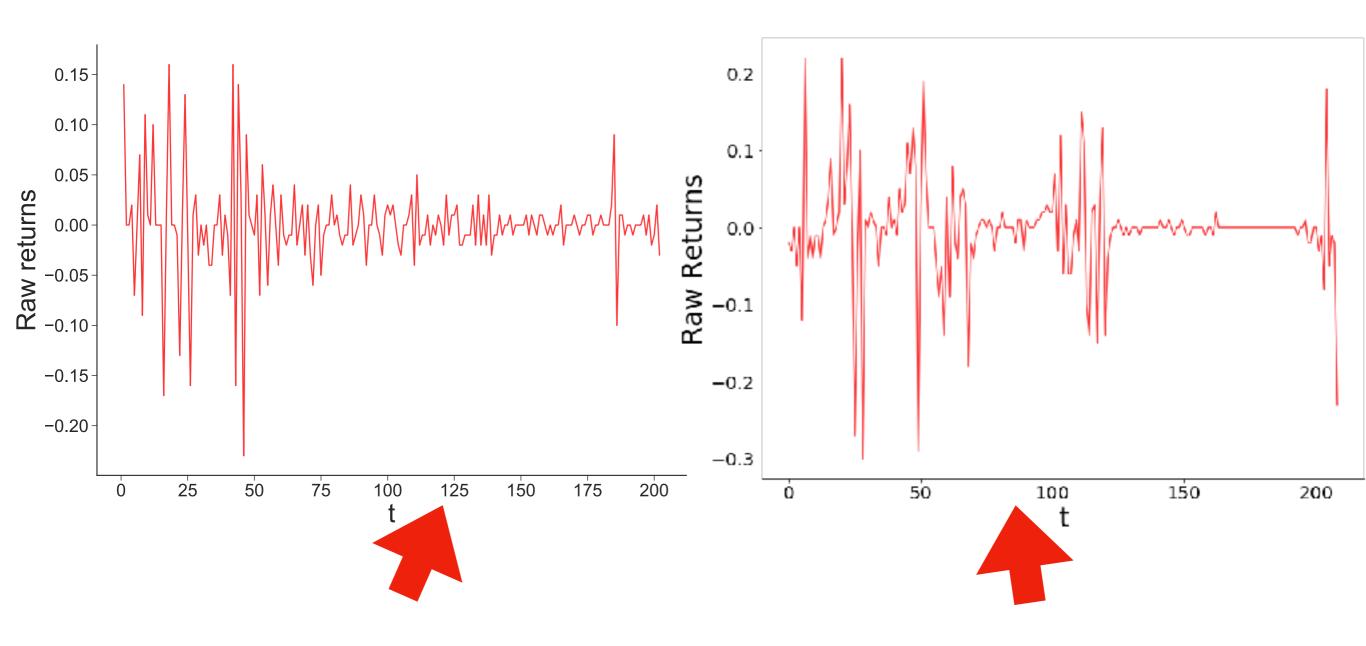
Some of them participate in the market. They buy if price is less than their opinion, and sell otherwise

Simulations

Generate price data with simulations

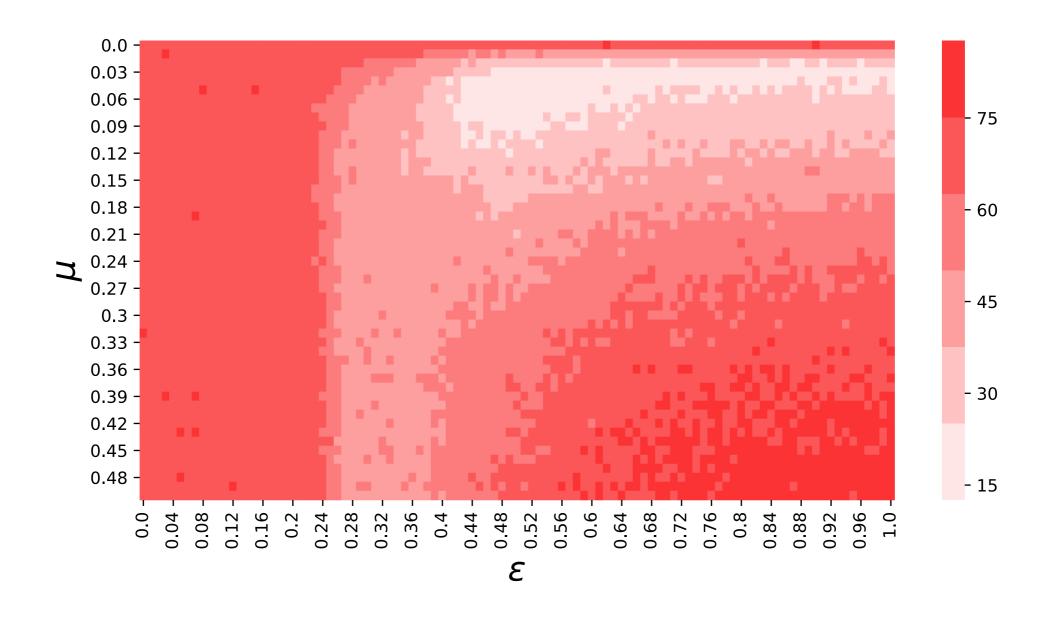
Compare with **empirical data** (3385 markets)

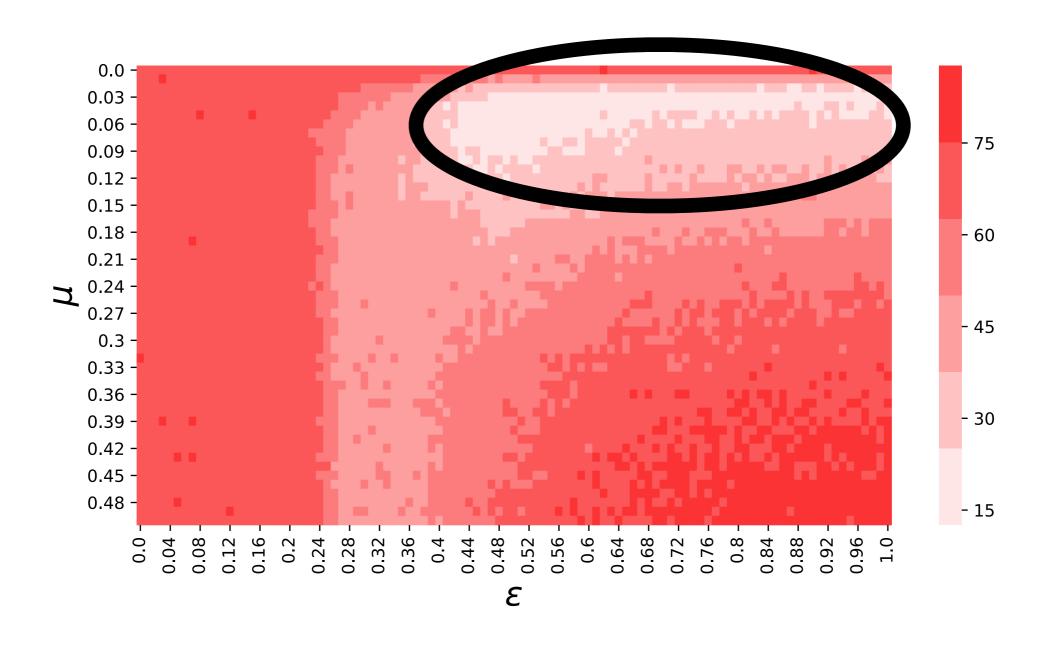


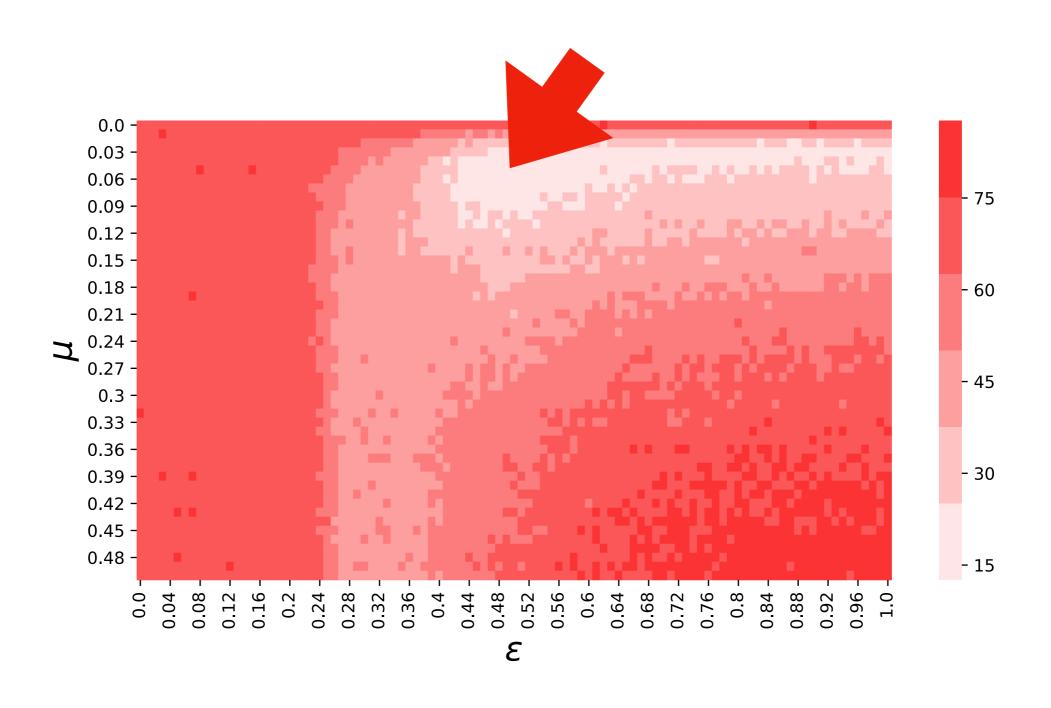


Historical (Brexit)

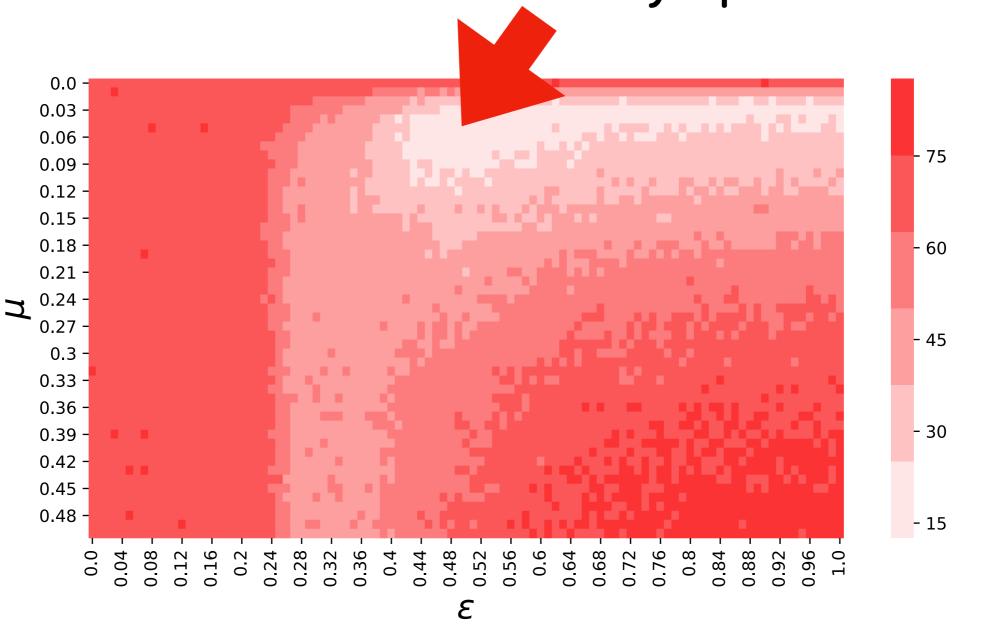
Simulated

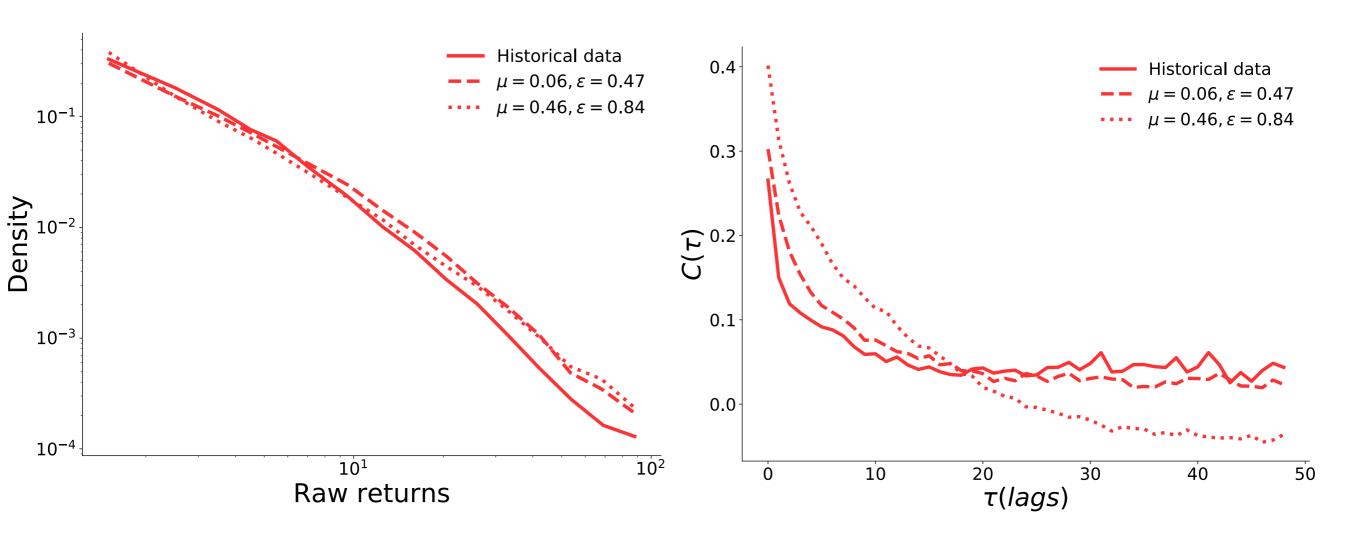






People exchange opinions with many, but barely update theirs

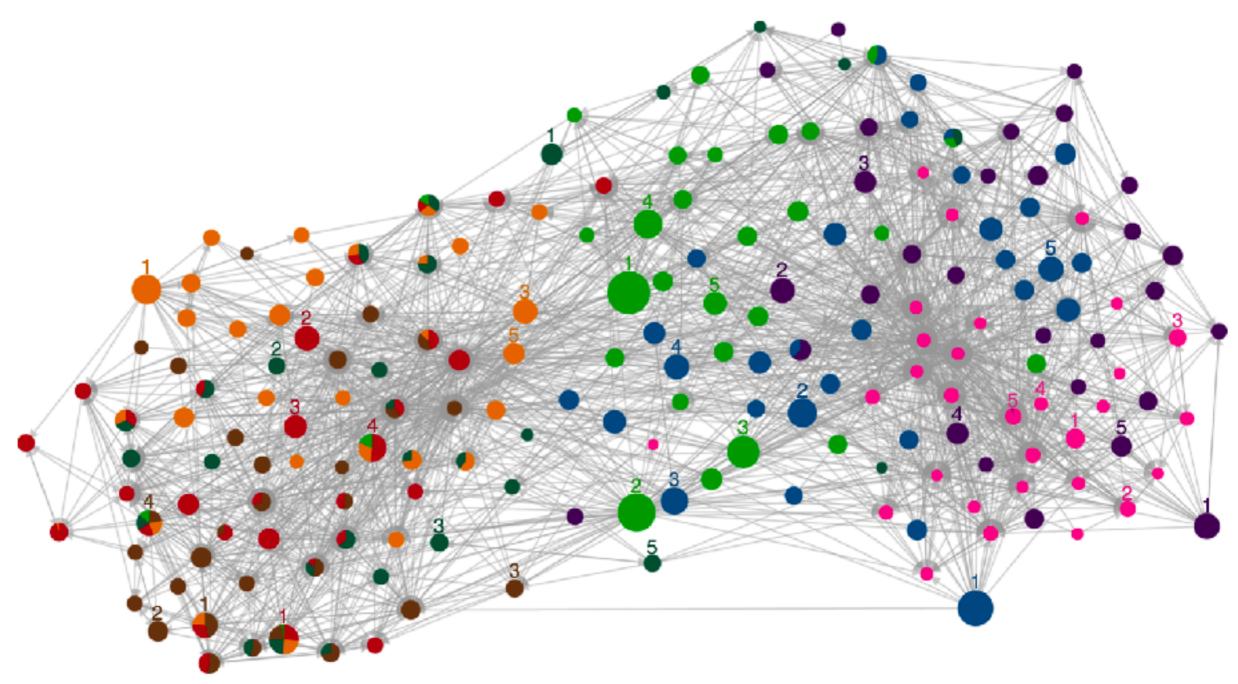




Case study II

Fake news and empirical Opinion dynamics on twitter

Empirical opinion dynamics



Fake news

- 1 @PrisonPlanet
- 2 @RealAlexJones
- 3 @ zerohedge
- 4 @DRUDGE_REPORT46
- 5 @realDonaldTrump^{1,2,4}

Extreme bias (right)

- 1 @realDonaldTrump^{2,4,5}
- 2 @ DailyCaller
- 3 @ BreitbartNews.
- 4 @ wikileaks
- 5 @ DRUDGE_REPORT^{4,4}

Right

- 1 @FoxNews
- 2 @realDonaldTrump^{1,4,5}
- 3 @dcexaminer
- 4 @DRUDGE_REPORT^{4,5}
- 5 @nypost

Right leaning 1 @WSJ

- 2 @WashTimes
- 3 @RT com 4 @realDonaldTrump^{1,2,5}
- 5 @RT_America

Center 1 @CNN

- 2 @thehill
- 3 @politico

5 @Reuters

- 4 @CNNPolitics 4 @NBCNews
- Left leaning 1 @ nytimes

5 @ Slate

- 2 @washingtonpost 2 @TIME
- 3 @ ABC

Left

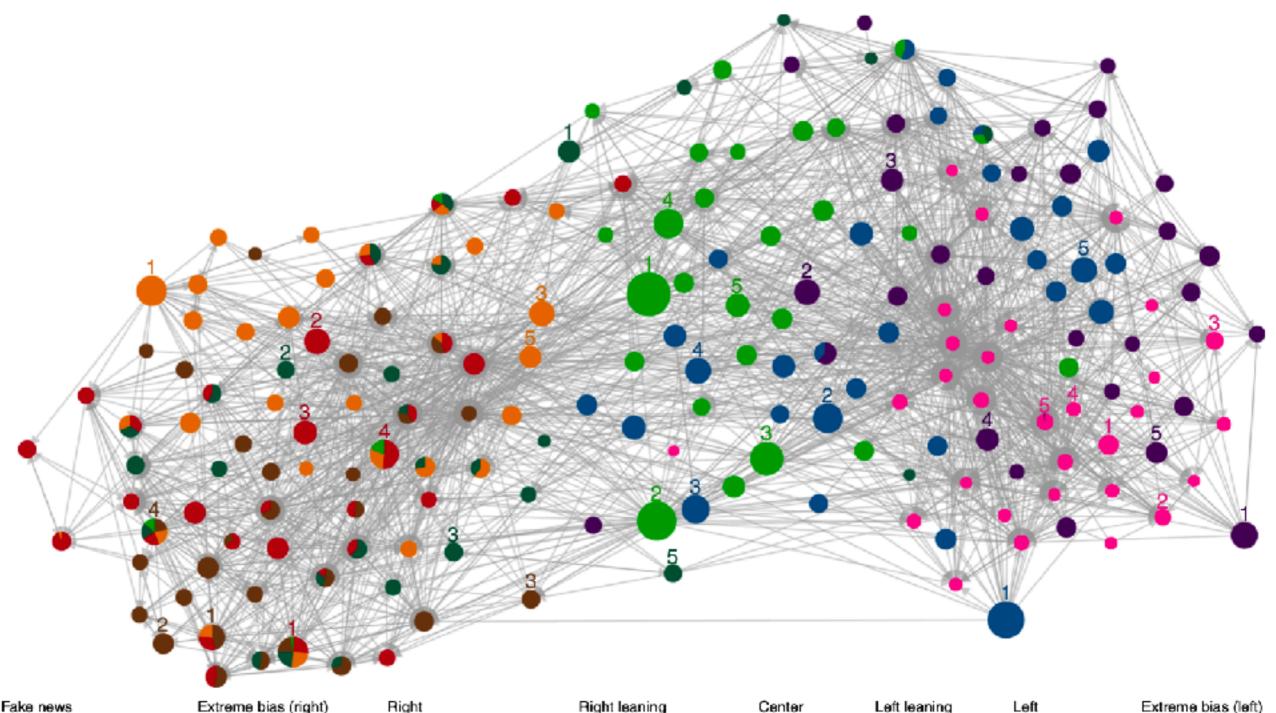
1 @ HuffPost

- 2 @ PalmerReport 3 @thedailybeast 3 @peterdaou
- 4 @ RawStory 4 @crooksandliars
- 5 @ HuffPostPol 5 @ BoldBlueWave

Extreme bias (left)

1 @ Bipartisanism

Empirical opinion dynamics



Fake news

1 @PrisonPlanet

2 @RealAlexJones

- 3 @zerohedge
- 4 @DRUDGE REPORT**
- 5 @realDonaldTrump1,2,4

Extreme bias (right)

- 1 @realDonaldTrump
- 2 @DailyCaller
- 3 @BreitbartNews
- 4 @wikileaks
- 5 @DRUDGE_REPORT4.4

Right

- 1 @ FoxNews
- 2 @realDonaldTrump 1,4,5
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- 1 @WSJ
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Which model would you use?

Video becoming viral
Marketing
Opinions on stock markets

Summary

Several models of opinion dynamics Practical applications Case studies