

# On Twitter Bots Behaving Badly: Empirical Study of Code Patterns on GitHub



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# Internet Computing

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Cognitive Services and Intelligent Chatbots Theme Article Bots Acting Like Humans: Understanding and Preventing Harm Florian Daniel and Cinzia Cappiello University of New South Wales tecnico di Milano entities that act like humans in con via Twitter, on Facebook, in chats or Q&A sites. This paper studies how they may affect online conversations, provides a taxonomy of harms that may be caused, and discusses ow to prevent harm by studying when abuses occur MANY TECHNOLOGISTS CONSIDER chatbots one sites, online newspapers, emails, and the like of the hottest technologies in recent times, in the everywhere where there are human opinion fueled, for example, by Facebook's release of its Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegitimately. For example, Messenger API in 2016. In April 2017, mately or illegit cacebook reported 100 000 monthly active bots explicitly allows bots in its chats, while What shows the Messenger platform. In March 2017, Varol on the Messenger platform. In March 2017, Vario SApp states that it blocks phone numbers gener-et al.<sup>2</sup> estimated that between 9% and 15% of ating bot traffice.<sup>2</sup> Inspired by Bessi and Ferrara.<sup>6</sup> active Twitter accounts are bots (29–49 million accounts out of 328 million.<sup>3</sup> Gartner estimates active Twitter accounts are bots (29–49 million accounts out of 328 million.<sup>3</sup> Gartner requests will be the stud on the surface act like legitimate human handled by bots, while Inbent estimates 1.8 bli-users in online conversations. Induced by loss, while motivat estimates 1.5 or a loss in other conversations. I lon unique customer chabdou serves by 2021.<sup>1</sup> The technological advancements of chabdots undoubtedly produced a hype on its own, yet bots today are by far not limited to instant mea-terms in the serves and the conversations. Even though online bots are multiplying their most organizations and users still do not have the honowledge, skills, or understanding to craft a sucaging only. Bots permeate all kinds of online con-ersations in Twitter, Facebook, Instagram, Q&A

F. Daniel, C. Cappiello, B. Benatallah. Bots Acting Like Humans: Understanding and Preventing Harm. IEEE Internet Computing 23(2), 2019, Pages 40-49.

Date of publication 14 January 2019; date of current version Robot Ethics9 to address "urgent ethical quest

of *physical* robots, significant research and development efforts are being invested into *robotethics*.<sup>8</sup> The IEEE has a Technical Committee or

tions prompted by and associated with robotic

Harm in human-bot interactions

# Psychological harm

Someone's psychological health or well-being get endangered or injured

# Legal harm

Someone becomes subject to law enforcement or prosecution

# Economic harm

Someone incurs a monetary cost or loses time

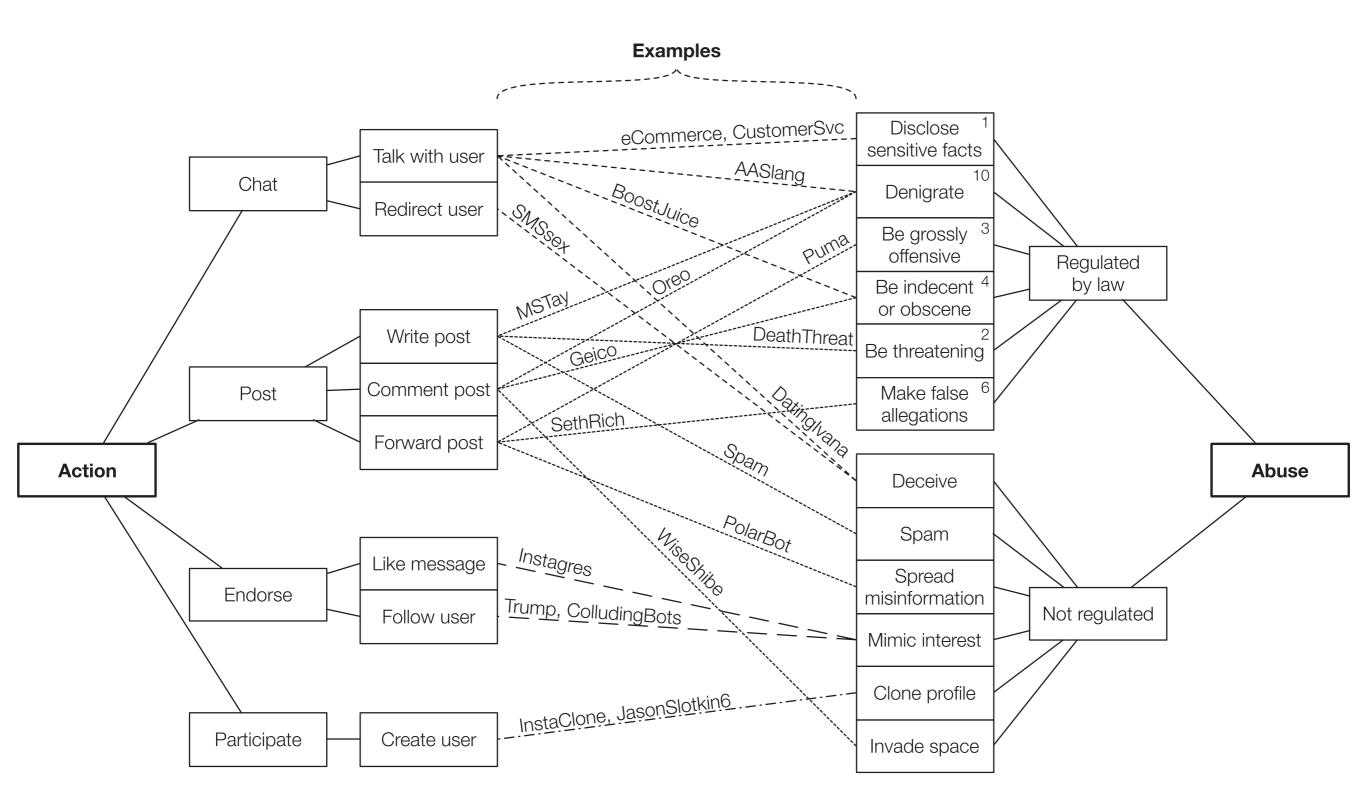
# Social harm

Someone's image or standing in a community gets affected negatively

# Democratic harm

Democratic rules and principles are undermined

#### Harm = consequence of an **action** and an **abuse**



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#### Literature on bots

- Bot development (frameworks, APIs, etc.)
- Bot detection from externally visible communications

**Problem** (goal of this paper)

Identify how harm is caused by bots

Understand likely underlying intentions

>> Abuse-oriented classification of bot **code repositories** published on GitHub Before going into the details...

Bots are not negative in general!

#### Platform policies and permissions



All platforms provide developers with **programmable interfaces** Typically allow programmatic access to **all functionalities** Users of the APIs must **authenticate** with the platforms Almost all platforms impose some kind of **limitation** 

#### Dataset

### Focus on **Twitter**

# Search by keywords Collection of code files + metadata

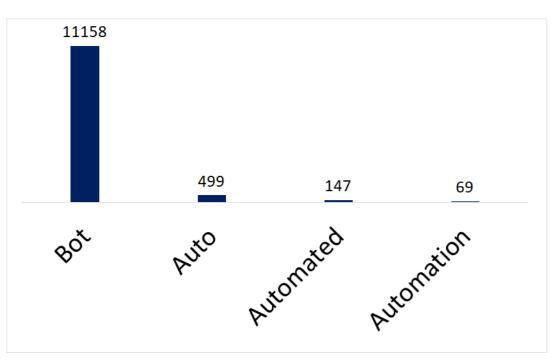


Fig. 1: Distribution of GitHub search results by searched keywords (includes all programming languages).

# Preliminary analysis

of actions implemented in repositories

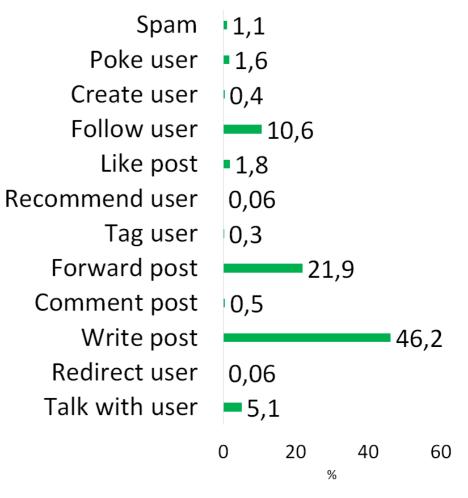


Fig. 2: Labels of repositories.

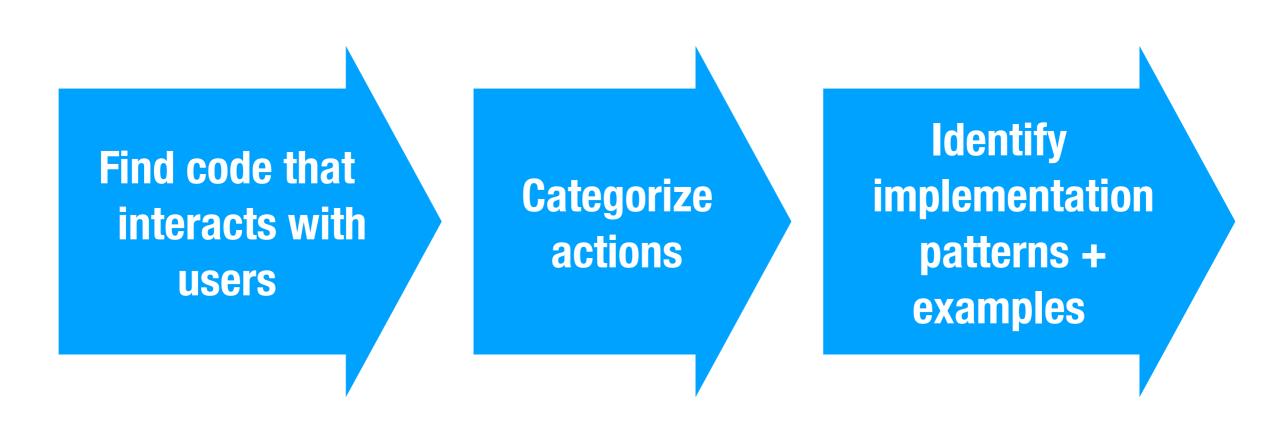
# **Starting point** Final dataset

# Selection criteria:

- Programming language = Python
- Exclusion of repositories that are out of scope
- 5 best repositories for each of the most used actions
- 5 random repositories from the rest
- 10 best repositories we could not classify
- 10 random repositories from the rest

# = 60 GitHub Twitter bot repositories programmed in Python average number of files per repository: 3 average number of lines of code: 192 average size of repository: 21.39 KBytes

#### Methodology



Systematic, manual code review!



Action	Description
Search	Search users or tweets using names, keywords, hashtags, ids or similar or by nav- igating social network relationships (e.g., friends of friends, followers of friends, friends of followers, followers of followers)
Follow	Follow users to establish social relationships
Like	Like tweets by other users to endorse them
Tweet	Post a new tweet to communicate content
Mention	Mention other users in tweets using @ to attract attention
Retweet	Re-post tweets by other users to endorse them
Talk to	Send direct messages to users to converse with them
Pause	Pause the conversation flow of the bot
Store	Store content retrieved from the social network for later use

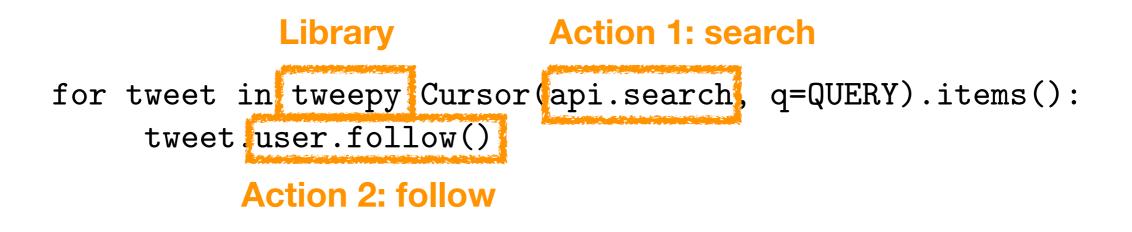
Synthesis of online communication actions implemented by Twitter bots

### **Results** Code patterns of action implementations

		. Taxonomy of code pa	itterns used for the implementation of actions.
	Action	Pattern	Description
Coorob	Search	User search	Search user account by name, keyword, id or similar
Search		Tweet search	Search tweets by keyword or hashtag
		Trend search	Search trending topics or hashtags by location
Follow	Follow	$Indiscriminate\ follow$	Follow users without checking suitability of users, user- names or content shared
		Whitelist-based follow	Follow only users whose attributes or tweets match some element of a given whitelist
		Blacklist-based follow	Don't follow users whose attributes or tweets satisfy one or more criteria specified in a blacklist
		Phantom follow	Follow users and unfollow them as soon as a given con- dition is satisfied, e.g, a limit of friends reached or being followed back
l iko	Like	Indiscriminate like	Like tweets without checking suitability of content, user or username
		User searchSearTweet searchSearTrend searchSearIndiscriminate followFollorMhitelist-based followFollorBlacklist-based followDon' or mPhantom followFollorIndiscriminate likeLike or usWhitelist-based likeLike matorBlacklist-based likeLike matorBlacklist-based likeDon' or meMass likeAggrFixed-content tweetThe collerAI-generated tweetThe be comeTrusted source tweetThe be comeTweet with opt-inTweet weet the herIndiscriminate mention of aBlacklist-based mentionMem of aBlacklist-based retweetSear 	Like only tweets by users whose attributes or content match some element of a whitelist
	SearchUse TweetFollowIndeFollowIndeWhenBlackPhotoIndeLikeIndeMaMaTweetFiaAI-TraTweetFiaAI-TraTweetIndeImage: Search of the search of	Blacklist-based like	Don't like tweets whose attributes or users match an element of a blacklist
		Mass like	Aggressively like tweets of given users
	Tweet	Fixed-content tweet	The content of the tweet is taken from a fixed, static collection of predefined messages
		AI-generated tweet	The text of the tweet is automatically generated using AI/NLP tools
		Trusted source tweet	The content of the tweet is taken from a source that can be considered trusted
		Tweet with opt-in	Tweets are sent only to people who ask to interact with the bot, sending it a message or mentioning it in a tweet
Mention	Mention	Indiscriminate mention	Mention other users without checking suitability of user- name or content shared
	Follow Fo	Targeted mention	Classify users on the basis of their tweets and mention them in targeted messages
		Whitelist-based mention	Mention only users whose attributes match some element of a whitelist
		Blacklist-based mention	Don't mention users whose attributes match elements of a blacklist
Rativoat	Retweet	Indiscriminate retweet	Retweet tweets without checking content or username for suitability
		Whitelist-based retweet	Retweet content only from users whose attributes match some element of a whitelist
		Blacklist-based retweet	Don't retweet tweets whose attributes or users satisfy some condition expressed in a blacklist
		Mass retweet	Aggressively retweet multiple tweets by selected users
Talk to	Talk to	Indiscriminate talk	Send direct, instant messages to users without checking their suitability
		Talk with opt-in	Reply only to messages sent to the bot (passive behavior)
		AI-generated talk	Generate messages using AI/NLP tools
		Fixed-content talk	Take message from a fixed list of predefined phrases
			Classify users based on their tweets or attributes and target message accordingly
Pause	Pause		Use pauses in instant messages to deliver human-like conversation experience to other humans
Ctoro		Satisfy API constraints	Use as short as possible pauses just to avoid being blocked by API usage limitations
SUIE	Store	Store persistently	Store retrieved content or user information for later use

Table 2: Taxonomy of code patterns used for the implementation of actions.

Action	Pattern	Description						
Search	User search	Search user account by name, keyword, id or similar						
	Tweet search	Search tweets by keyword or hashtag						
	Trend search	Search trending topics or hashtags by location						
Follow	Indiscriminate follow	Follow users without checking suitability of users, user- names or content shared						
	Whitelist-based follow	Follow only users whose attributes or tweets match some element of a given whitelist						
	Blacklist-based follow	Don't follow users whose attributes or tweets satisfy one or more criteria specified in a blacklist						
	Phantom follow	Follow users and unfollow them as soon as a given con- dition is satisfied, e.g, a limit of friends reached or being followed back						
Like	Indiscriminate like	Like tweets without checking suitability of content, user or username						
	Whitelist-based like	Like only tweets by users whose attributes or content match some element of a whitelist						
	Blacklist-based like	Don't like tweets whose attributes or users match an el- ement of a blacklist						
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	Tweet with opt-in	Tweets are sent only to people who ask to interact with the bot, sending it a message or mentioning it in a tweet						



= search users pattern + indiscriminate follow pattern

# Function definition def mentions count, max\_seconds\_ago, id blacklist) : Mention search return [mention for mention in api.mentions\_timeline(count=count) if not mention.id in id\_blacklist ]

**Blacklist inclusion check** 

= blacklist-based mention pattern



Patterns may

Enable an abuse

Logic that by design performs an abuse

Prevent an abuse

#### Logic that prevents the bot from performing an abuse

Be vulnerable to content abuse

Interactions with users and/or content that may be inappropriate

Be vulnerable to trust abuse

Forward, store or analyze content retrieved from users

tion	Pattern	A Contraction	0000 0	Digrate Wile Pack		E hacen lange	Un.	She of the	Colio Magali	(Ja)	Mi.		
ollow	Indiscriminate follow	<b>—</b>									GO		
	Whitelist-based follow					—		—	—		GO		_
	Blacklist-based follow								—		GO		_
	Phantom follow										GO		-
Like	Indiscriminate like			<u></u>							GO		_
	Whitelist-based like										GO		_
	Blacklist-based like										GO		_
	Mass like			$\delta$	<u></u>	—			GO		GO		—
Tweet	Fixed-content tweet	8							GO				_
	Al-generated tweet		Δ			∕		—	GO		—	GO	_
	Trusted source tweet	8	0						GO			_	-
Mention	Indiscriminate mention		GO	GO	GO				GO		GO		_
Mention	Opt-in mention		GO	GO	GO	—	—	—				—	—
	Targeted mention		$\Delta$			—		—	—		—	—	—
	Whitelist-based mention	_							—		GO		_
	Blacklist-based mention		۲	۲	۲	—					GO		-
Retweet	Indiscriminate retweet										GO		_
	Whitelist-based retweet										GO	—	_
	Blacklist-based retweet								—		GO	—	_
	Mass retweet	-		٨	٨			à	GO	à	GO		-
Talk to	Indiscriminate talk	_				—			GO		GO		_
	Fixed-content talk							—	GO		GO	—	_
	AI-generated talk							—	GO		GO	GO	_
	Talk with opt-in			—				—			۲		
	Targeted talk	-					—		GO		GO	—	-
Pause	Mimic human	_						GO					_
	Satisfy API contraints	-					—	-	GO			—	-
Store	Store persistently	A										$\mathbf{A}$	_

Results

## Pattern-effect matrix =

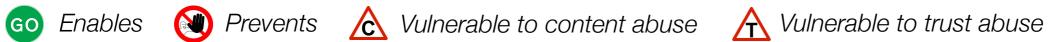
potential effects of patterns

#### Zoom into Follow patterns

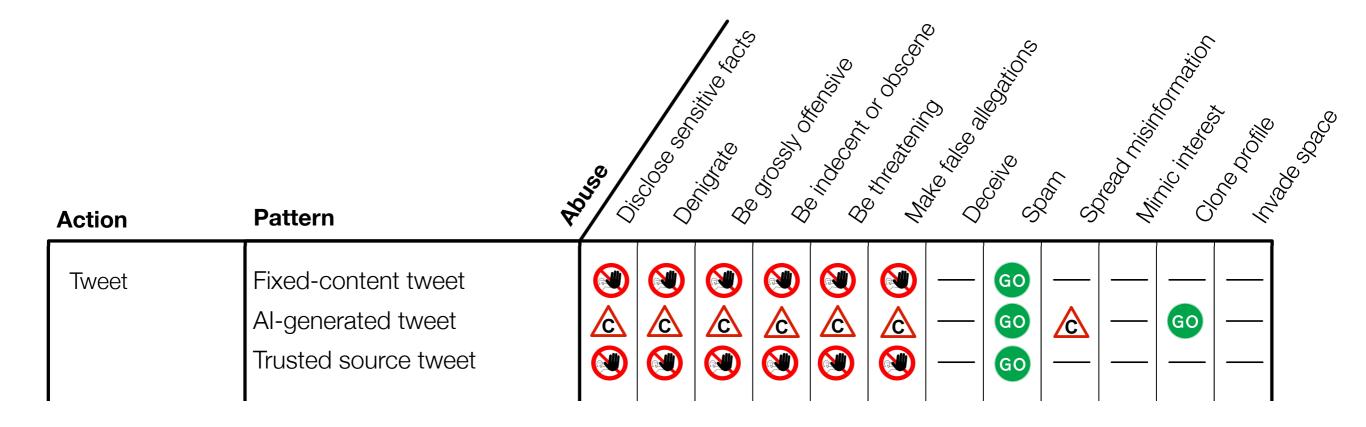
Action	Pattern	<b>Loss</b> Open de la participa de	хо <sub>со</sub>
Follow	Indiscriminate follow Whitelist-based follow Blacklist-based follow Phantom follow	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	







#### Zoom into **Tweet patterns**

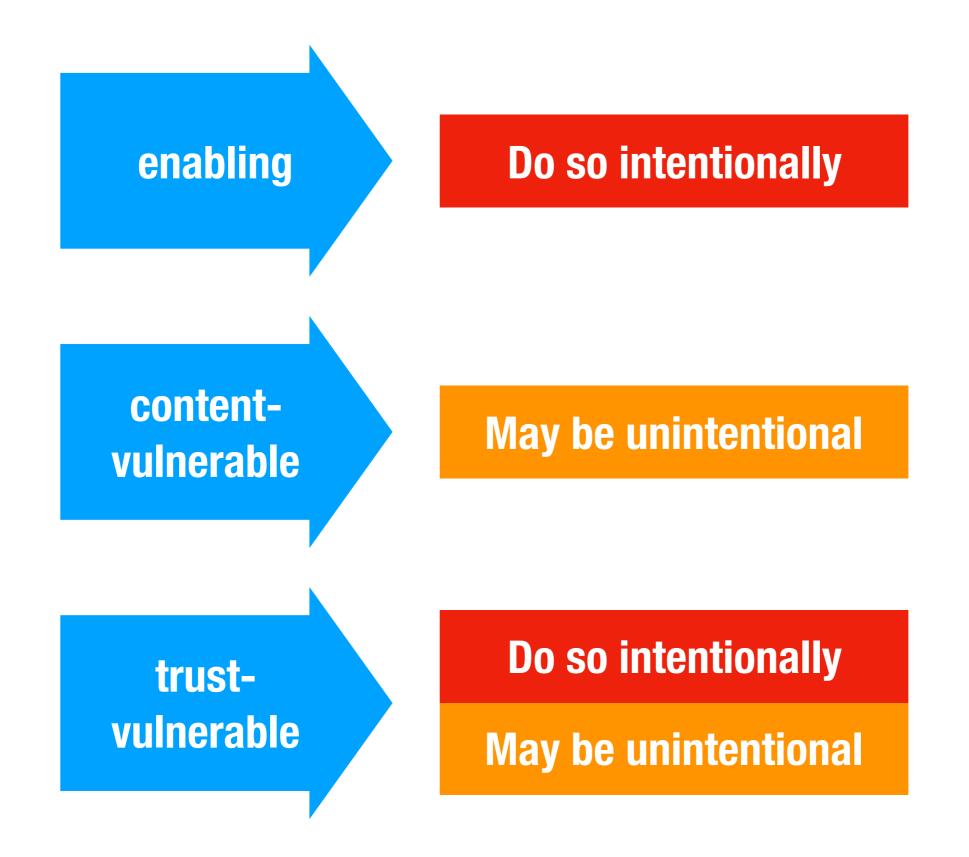




Prevents

 $\land$  Vulnerable to content abuse  $\land$  Vulnerable to trust abuse

**Results** Coming back to the "**why**" question... and using some technical considerations on the nature of patterns...



Summing up

**Original** perspective on bots for online communication: code

Contributions to state of the art:

- Identified **31 patterns** and 9 actions from 60 repositories (~ 80 hours of manual code review + x of discussion)
- 2. Discussed **effects** of patterns and mapped patterns to potential abuses
- 3. Technical interpretation of **intentionality** underlying bot implementations

**Next**: formal language for action patterns + patterns search engine for automated pattern retrieval from all collected repositories