Social Bots Detection on Twitter

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2. Introduction

As the popularity of social networking and microblogging tools continues to grow, accounts controlled by automated programs, known as social bots (or bots for short), have risen from the horizon. Twitter, with the openness of its platform and API, has seen the rise of the machines. These bots automatically produce contents and even communicate with humans, trying to hide themselves among the mass population of Twitter. In this project, we are going to compare the performance of two different machine learning algorithms for detecting bot accounts.

1. Motivation

Although plenty of bots on Twitter are benign, there are still bad bots that mislead, exploit and manipulate online conversations with rumors, spam, phishing links and slander [1]. These bots randomly add other users as their friends and if the user follows back, the malicious contents will be displayed on his homepage [2]. They can also be alleged to help political candidates in the election [3], potentially undermining the democratic system. Therefore, we need to develop a method to detect these bots in social media, either to help Twitter manage the community [4], or to help human users identify who they are communicating with.

1. Related Work

[1] Emilio Ferrara, Onur Varol, Clayton Davis, Filippo Menczer, and Alessandro Flammini. 2015. The Rise of Social Bots. X, X, Article XX ( 201X), 11 pages.

[2] Z. Chu, S. Gianvecchio, H. Wang and S. Jajodia, "Detecting Automation of Twitter Accounts: Are You a Human, Bot, or Cyborg?," in IEEE Transactions on Dependable and Secure Computing, vol. 9, no. 6, pp. 811-824, Nov.-Dec. 2012.

doi: 10.1109/TDSC.2012.75

[3] J. P. Dickerson, V. Kagan and V. S. Subrahmanian, "Using sentiment to detect bots on Twitter: Are humans more opinionated than bots?," 2014 IEEE/ACM International Conference on Advances in Social Networks Analysis and Mining (ASONAM 2014), Beijing, 2014, pp. 620-627.

[4] N. Chavoshi, H. Hamooni and A. Mueen, "DeBot: Twitter Bot Detection via Warped Correlation," 2016 IEEE 16th International Conference on Data Mining (ICDM), Barcelona, 2016, pp. 817-822.

[5] Bot or Not: an end-to-end data analysis in Python, Erin Shellman. Available: http://www.erinshellman.com/bot-or-not/

1. Data.

Bots: PhotoLab Bot@Photolab\_Bot, Mobilised @mtm, LOUDBOT@LOUDBOT, AR Herald @ARherald, AugmentedAdvertsing @AugmentedAdvert, Debit Card@NeedADebitCard, Hurricane Sandy Help @SandyAid, Procrastibot@Procrastibot, It’s Spelt Caesar @itsspeltcaesar, Hundred Zeros@HundredZeros, Dear Assistant@DearAssistant, What the Fare@WhatTheFare, Museum Bot@MuseumBot, Pentametron@pentametron, Grammar Police@\_grammar\_,[Pixel Sorter](https://twitter.com/pixelsorter) [@pixelsorter](https://twitter.com/pixelsorter), Netfix Bot@netflix\_bot, Earthquake Robot@earthquakeBot, everyword@everyword, Thinkpiece Bot@thinkpiecebot,

Congress-edits@congressedits, Great Artist@greatartbot, olivia taters@oliviataters, moth generator@mothgenerator, MoMA Bot@MoMARobot, blah future@blahfuture, NYPD edits@NYPDedits, it’s always 4:20 @420worldclock, balh future@blahfuture, Two Headliens@TwoHeadlines, [poem.exe @poem\_exe](mailto:poem.exe@poem_exe), censusAmericans@censusAmericans, Holiday Bot@holidaybot4000, Countdown bot@lettergamebot, JustDiedBot @JustDiedBot, NotInventedHere@NotInventedHere, Perline Egg Craft@perlineggcraft, Botstructions@botstructions, TNY Poetry@tnypoetry, Rules for the Day @RulesForTheDay.

Doodlybot@Doodlybot, [phase↝chase](https://twitter.com/phasechase) [@phasechase](https://twitter.com/phasechase), NYT Anonymous@NYTAnon, RealDonaldContext@realDonaldCntxt, Good Music Recs@songchoicebot, Ominous Studies@OminousStudies, Kanye Jordan @kanyejordan, Sealth Mountain@StealthMountain, dronesweetie@dronesweetie, DSCOVR:EPIC @dscovr\_epic.

Not Bot: Donald J.Trump @realDonaldTrump,

1. Algorithms plan to use

Naïve Bayes, Random Forest [5]