

SOCIAL MEDIA TRENDS, 3<sup>RD</sup> OF OCTOBER

# Health of the Media Conversation: Using AI to Detect and Understand Harmful Speech

Eric Karstens

Funding Consultant

European Journalism Centre

Sophia Karakeva

Communications and Marketing Executive

Datascouting

Stavros Vologiannidis

Professor for the Technological Educational Institute  
of Central Macedonia and Founder

Datascouting

SOCIAL MEDIA TRENDS, 3<sup>RD</sup> OF OCTOBER

# Health of the Media Conversation: Using AI to Detect and Understand Harmful Speech

---

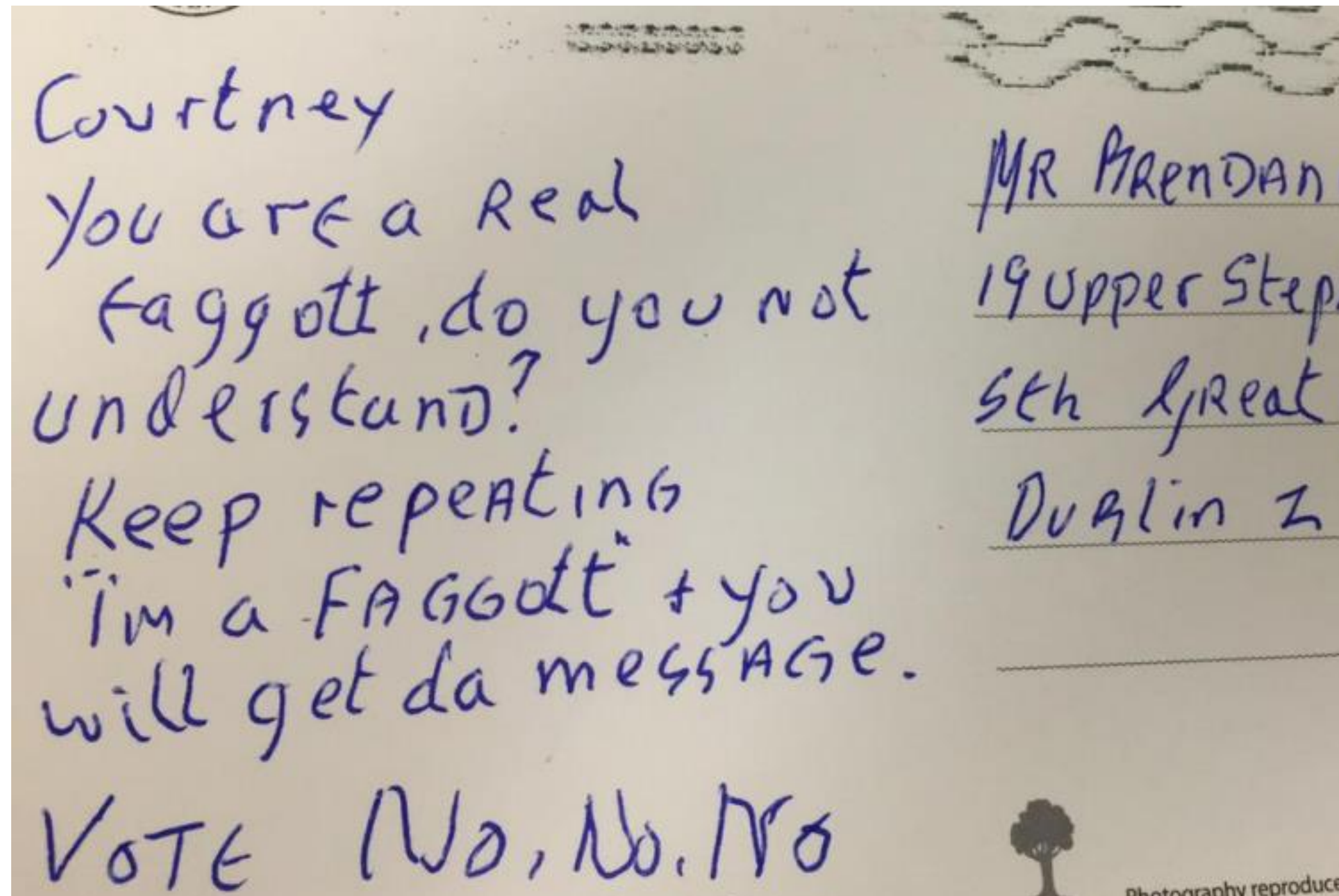
Journalism and hate speech

**Eric Karstens**

Funding Consultant

European Journalism Centre

# HATE MAIL THEN AND NOW



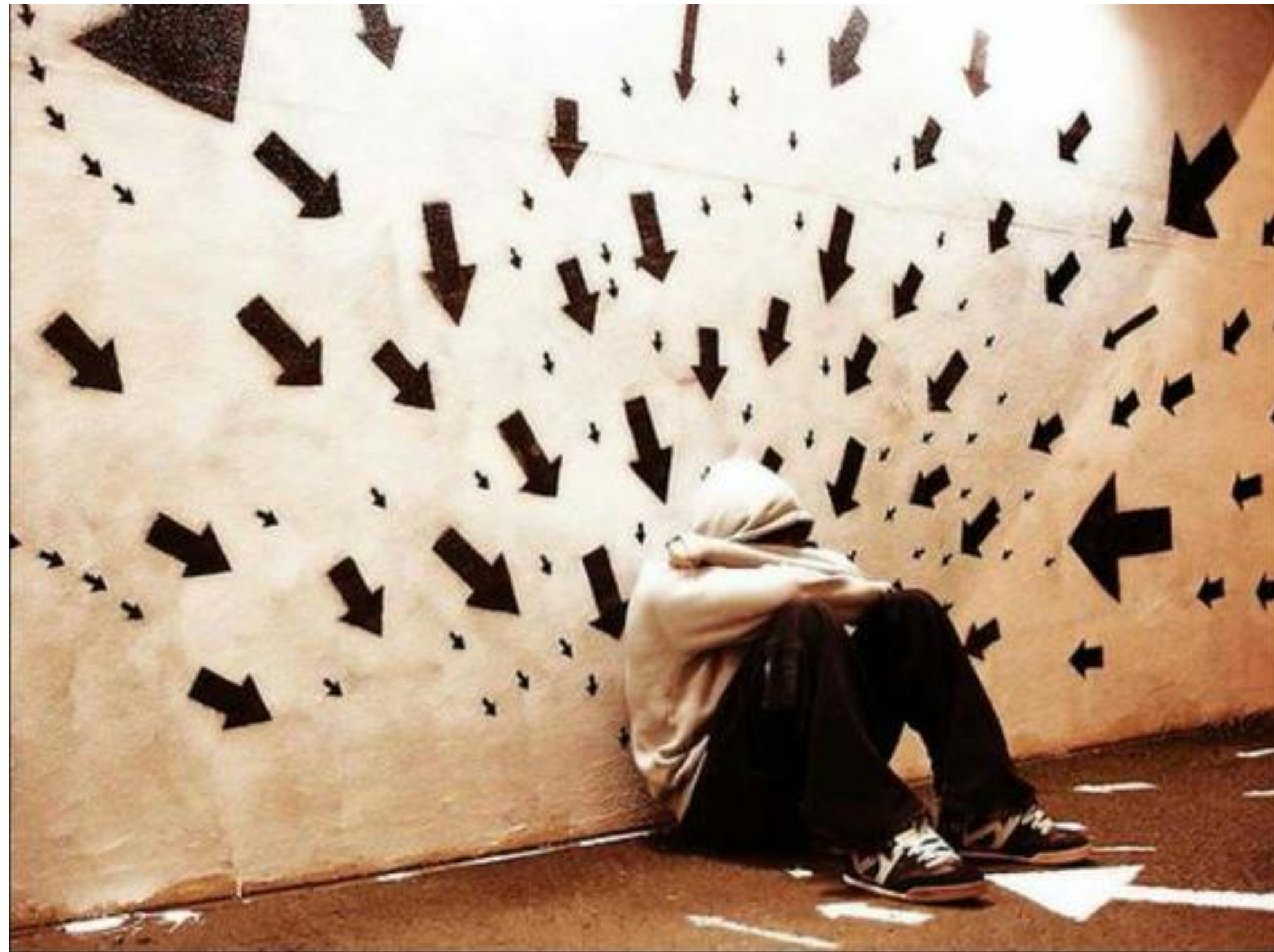
## HATE SPEECH AGAINST JOURNALISTS



# IMPACTS OF HATE SPEECH



## RECLAIMING AGENCY



SOCIAL MEDIA TRENDS, 3<sup>RD</sup> OF OCTOBER

# Health of the media conversation: using AI to detect and understand and harmful speech

---

Mapping and monitoring hate speech

Sophia Karakeva

Communications and Marketing Executive

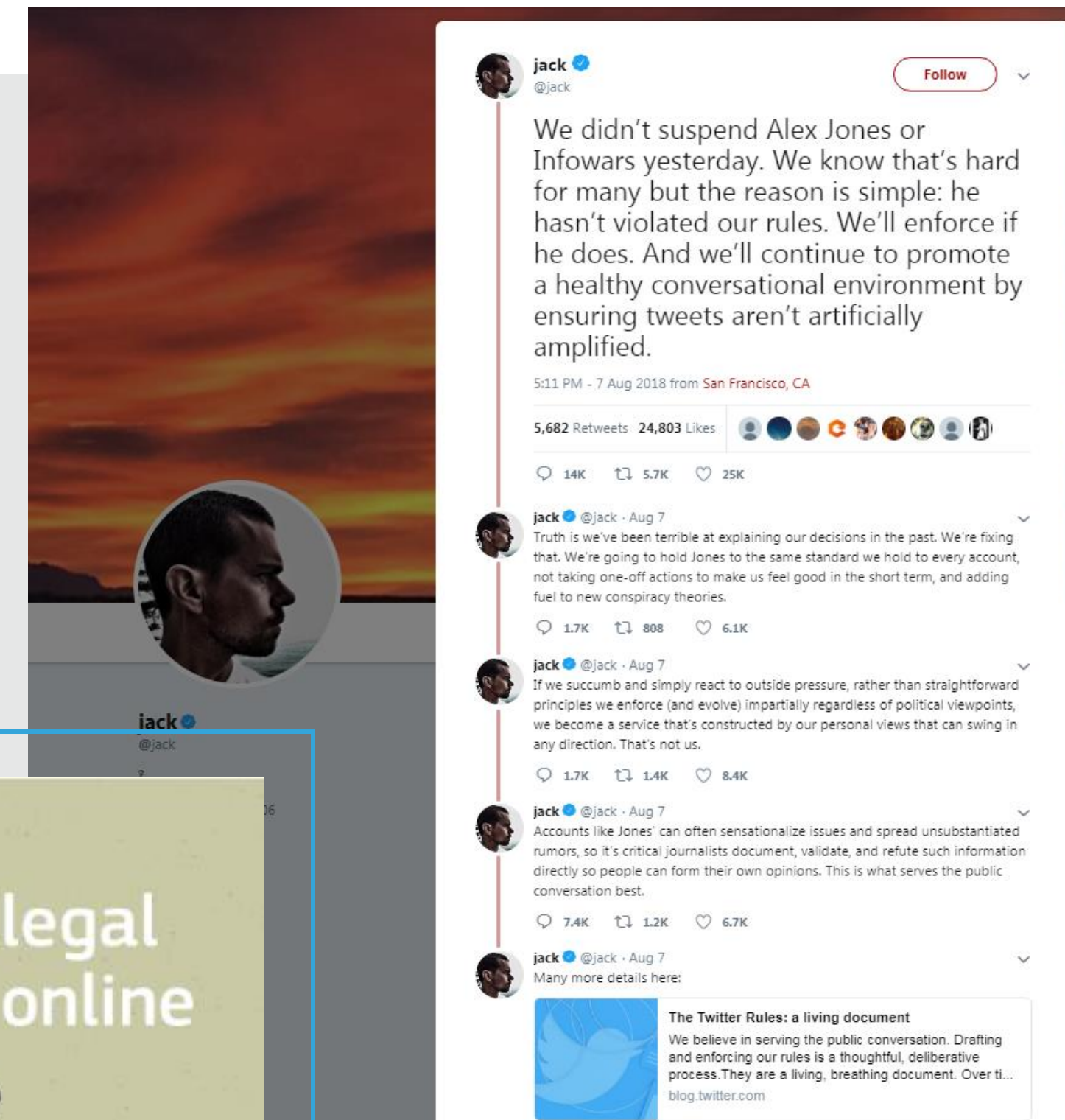
Datascouting

- 

“ *Truth is now considered #HateSpeech.  
Thanks progressives.* ”

# What is the role of social media when it comes to hate speech?

- ▶ Code of conduct
- ▶ Still not there
- ▶ Cooperation but diversity



# The legal and discursive characteristics of hate speech

- ▶ Legal liability
- ▶ Danger of over-regulation



## ARTICLE 19

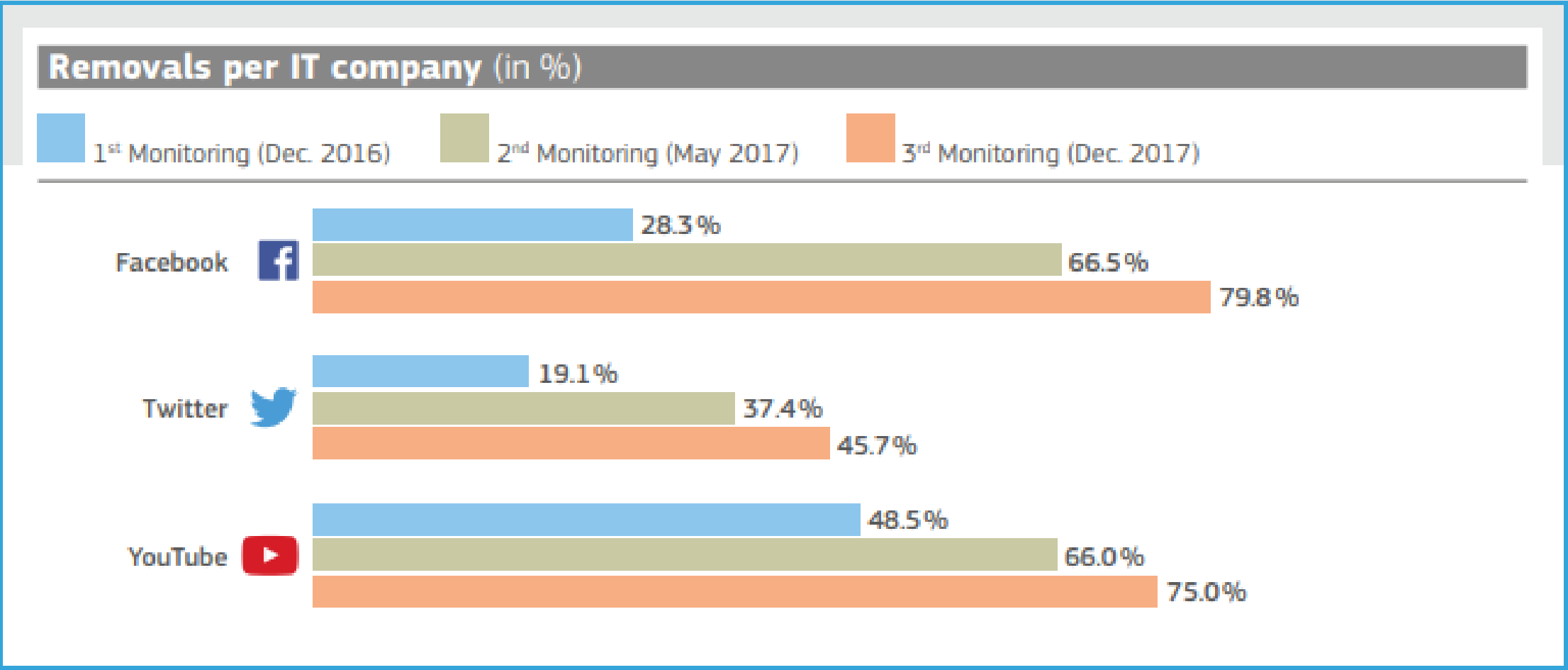
"Everyone has the right to freedom of opinion and expression; this right includes freedom to hold opinions without interference and to seek, receive and impart information and ideas through any media and regardless of frontiers."  
- Article 19; Universal Declaration of Human Rights

# Where does hate speech spread the most?



Source:  
The Flash Eurobarometer, September 2018  
The 3rd monitoring of EU's Code of Conduct, January 2018  
2017 Pew Research Center survey about online harassment

- ▶ Social networks the most common online daily activity
- ▶ Malta has the highest online hate speech in the EU
- ▶ Facebook has the largest amount of notifications
- ▶ 70% of hate speech content was removed in 2017



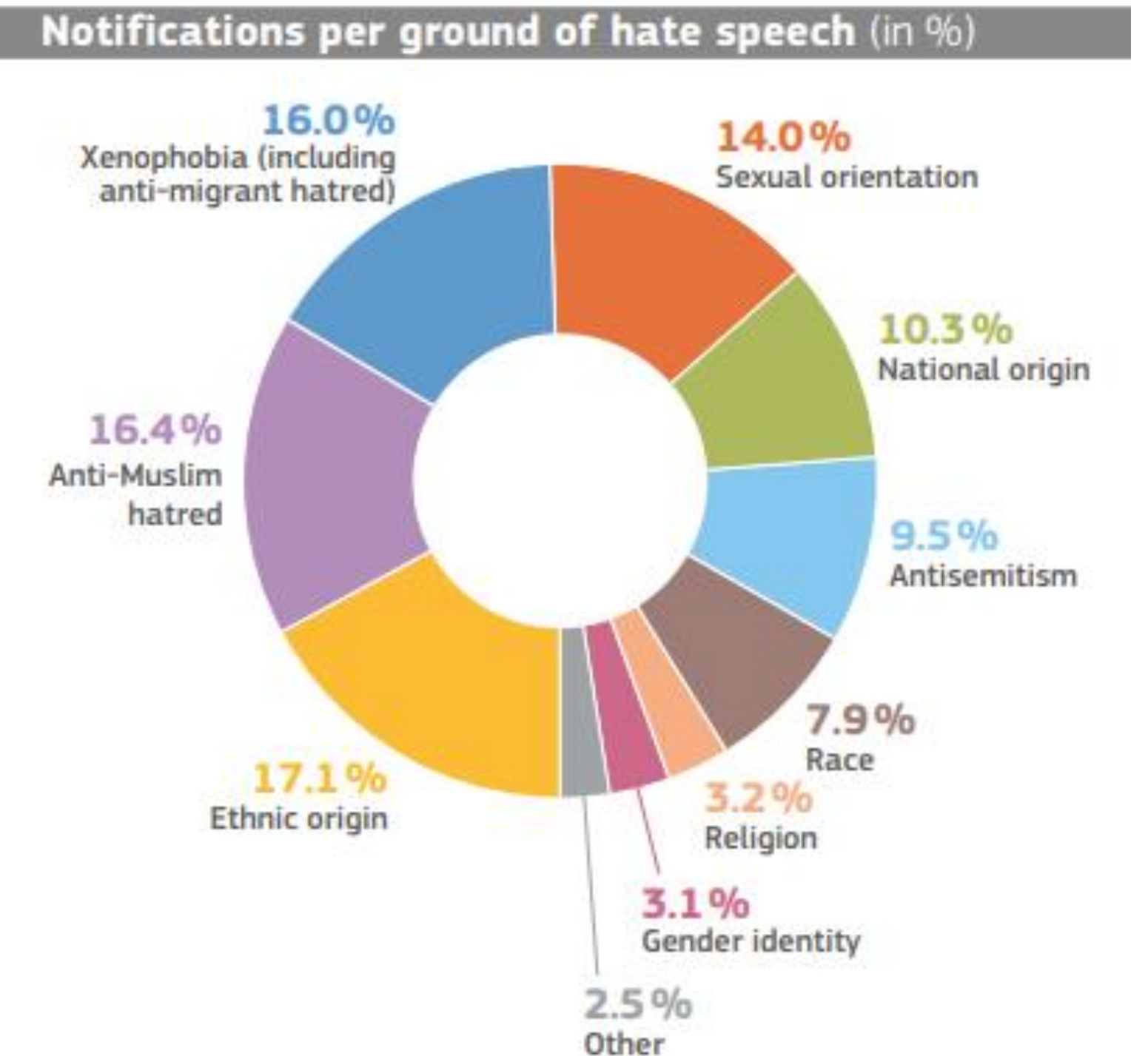
# Who is most targeted?

- ▶ Ethnic origin, anti-Muslim hatred, & xenophobia
- ▶ Gender matters to social media
- ▶ The developing world

“

**Facebook has been a useful instrument for those seeking to spread hate, in a context where for most users Facebook is the internet**

United Nations,  
Human Rights Council, Report of the Independent International Fact-Finding Mission on Myanmar, August 2018



Source: 3<sup>rd</sup> monitoring exercise of the European Union's Code of Conduct, released in January 2018

# Hate me, hate me not

***“It’s like public lynching... It has made be frightened for my physical safety when I am out in the streets.”***

Turkish journalist Amberin Zaman

- ▶ Hate me not: not allowed on our platform
- ▶ Hate me: often the most engaged content
- ▶ Hate me not: strong measures to take it down
- ▶ Hate me: not everything has the same value
- ▶ Hate me: highly reactive, high error rate
- ▶ Hate me not: automation and AI tools
- ▶ Hate me: “new normal”
- ▶ Hate me not: education, campaigns, regulation

The image shows a screenshot of a Twitter thread and a Facebook post. The Twitter thread consists of two tweets. The first tweet, from a user with a blurred profile picture, reads: "This morning I woke up to a rape and death threat directed at my 5 year old daughter. That this is part of my work life is unacceptable." It is dated 1:04 PM - 27 Jul 2016 and has 1,438 retweets and 1,348 likes. The second tweet is a reply to @maggieNYT, reading: "Fuck you fuck you fuck you fuck you. Maggie haberman is a lying bitch and an enabler of racism, misogyny, fascism and treason." The Facebook post is from a user named Amberin Zaman and reads: "American Trumpanzee idiots setting their Nikes on fire while Pakistani Islamist idiots are setting Dutch products on fire. Can someone tell them they're more alike than identical twins? #GlobalDeplorables". A warning box at the top of the Facebook post states: "This post goes against our Community Standards. Only you can see this post because it goes against our standards on hate speech."

- ## Hate Speech

- 
- ```

graph TD
    START([START]) --> Q1{IS THERE A MENTION OF A PC?}
    Q1 -- Yes --> Q2{IS THE PC MENTIONED ALONG WITH OTHER DETAILS?}
    Q1 -- No --> Q3{IS THERE A MENTION OF A QUASI PC? (MIGRANTS)}
    Q2 -- Yes --> Q4{ARE ALL THE DETAILS ALSO PROTECTED?}
    Q2 -- No --> Q5{IS THE PC/SUBSET BEING ATTACKED BY?}
    Q4 -- Yes --> Q5
    Q4 -- No --> Q6[CONSIDER OTHER POLICIES]
    Q5 --> Q7[IS THE QPC/SUBSET BEING ATTACKED BY?]
    Q7 --> Q8[CONSIDER OTHER POLICIES]
    Q8 --> Q9[DELETE]
    Q3 -- Yes --> Q10[CONSIDER OTHER POLICIES]
    Q3 -- No --> Q11[DELETE]
  
```
- Legend:**  
 → Yes (Green arrow)  
 → No (Orange arrow)
- Flowchart Steps:**
- START**
  - IS THERE A MENTION OF A PC?**
    - Yes:** Proceed to Step 3.
    - No:** Proceed to Step 5.
  - IS THE PC MENTIONED ALONG WITH OTHER DETAILS?**
    - Yes:** Proceed to Step 4.
    - No:** Proceed to Step 6.
  - ARE ALL THE DETAILS ALSO PROTECTED?**
    - Yes:** Proceed to Step 6.
    - No:** Proceed to Step 7.
  - IS THE PC/SUBSET BEING ATTACKED BY?**
    - Yes:** Proceed to Step 8.
    - No:** Proceed to Step 9.
  - IS THE QPC/SUBSET BEING ATTACKED BY?**
    - Yes:** Proceed to Step 8.
    - No:** Proceed to Step 9.
  - CONSIDER OTHER POLICIES**
  - DELETE**
- Attack Types (Step 6):**
- ENCOURAGING VIOLENCE.
  - COMPARING THEM TO ANIMALS, DISEASE OR FILTH.
  - CONSIDER THEM SUBHUMAN.
  - CONSIDER THEM: SEX OFFENDERS, MURDERERS OR TERRORISTS.
- Attack Types (Step 7):**
- ENCOURAGING VIOLENCE.
  - CALLING FOR SEGREGATION.
  - CALLING FOR EXCLUSION.
  - DEGRADING GENERALIZATIONS.
  - DISMISSING THE PC.
  - CURSING THE PC.
  - USING SLURS.



50 FIBEP  
WORLD MEDIA  
INTELLIGENCE  
CONGRESS

COPENHAGEN  
OCTOBER 1-3 2018  
MARRIOTT HOTEL



**DATAScoutING**  
Actionable Information



# What's being done to fight it?

- ▶ Expanded hate speech rules
- ▶ More human annotation
- ▶ Academic & scientific research
- ▶ Sophisticated technology



SOCIAL MEDIA TRENDS, 3RD OF OCTOBER

HEALTH OF THE MEDIA CONVERSATION:  
USING AI TO DETECT AND UNDERSTAND AND  
HARMFUL SPEECH

---

AUTOMATICALLY TAGGING HATE SPEECH

Stavros Vologiannidis

ASSISTANT PROFESSOR  
FOUNDER OF DATASCOUTING

# HATE SPEECH AND TECHNOLOGY

- ▶ Hate speech is nothing new
- ▶ But the problem is much more evident in the age of **social media**
- ▶ Several attempts have been made to automatically identify hate speech...
- ▶ **Challenge: separating hate speech from offensive speech**
  - ▶ Hate speech changes forms, uses different expressions
  - ▶ Keyword based identification techniques do not work well

# MACHINE LEARNING APPROACH

- ▶ Use **Natural Language Processing** and **Machine Learning** techniques in order to identify hate speech
- ▶ **Main stages** in Text Analytics with Machine Learning using **Neural Networks**
  - ▶ Transform a **document** (social media article) to **numbers** (features)
  - ▶ Create a **Machine Learning model** based on your architecture of your choice and **train** it as follows:
    - ▶ Model **looks** at documents and calculates and answer
    - ▶ Model **compares** answer against the correct one (annotated by human experts)
    - ▶ Model **tweaks** itself so that if it finds the same document again it would be more likely to calculate the correct answer
  - ▶ **Repeat**

# DOCUMENT -> NUMBERS

- ▶ **How to transform a document to numbers?**
  - ▶ A) Convert a **word to numbers** (vector of a few hundreds of numbers)
  - ▶ B) Convert a **document** = lots of words **to math**
- ▶ **A) Word -> numbers: Word embedding algorithms**
- ▶ We need to retain some **semantic properties** of the word such as similarity
  - ▶ Words that share common contexts are close as mathematical objects
  - ▶ King - Man + Woman = Queen
- ▶ We have several options of such algorithms and **pretrained** models (Wikipedia,...)
  - ▶ Word2vec (Google), FastText (Facebook), .....
- ▶ **B) Document -> numbers: math operations on words, can include NLP**

# MODEL ARCHITECTURE

- ▶ **Model architecture: Convolutional Neural Networks on top of character level word embeddings**
- ▶ **Input** : Document
- ▶ **Output**: Hate or not (or additional classes such as personal attack, etc)
- ▶ **Challenge**: we need **lots of annotated (labeled) documents to train** (and test) algorithms

# ACTIVE LEARNING APPROACH

- ▶ Active Learning is a methodology that can greatly **reduce the amount of annotated data** required to train a machine learning model.
- ▶ It does this by **prioritizing** the labeling work for the experts .
  - ▶ The model **looks** at unlabeled data (cheap)
  - ▶ **Identifies which data** it is **most confused about** and **requests labels** from experts for just those (human is involved - expensive).
  - ▶ It **trains** on that small amount of labeled data

## ACTIVE LEARNING - BETTER DATA > MORE DATA

- ▶ **Minimize costs/time** by reducing expert resources
  - ▶ Get the same or better classifier accuracy usually with **3 to 10 times less effort**
- ▶ **Better data is more useful than more data...**
- ▶ The best active learning based model is the one with **great accuracy** and the **minimum number of requests towards experts**

## CURRENT STATE OF HATE SPEECH DETECTION

- ▶ **Social media companies** work extensively on hate speech detection
- ▶ Only **small datasets** are available in the research literature, even for the english language
- ▶ Different **hate speech definitions**, intensify the above problem
- ▶ Several researchers and research groups work on the subject, with a current **success rate of approximately 80%**

# OUR VISION

- ▶ **Hate speech**
  - ▶ Create a scalable active learning based system for identifying hate speech with direct feedback from journalists
  - ▶ Support multiple languages (English, French, German, Greek, Spanish)
- ▶ **DataScouting Media Intelligence Products**
  - ▶ Include technology from the project to DataScouting's media intelligence solutions

## Thank you

**Eric Karstens**

Funding Consultant

European journalism centre

karstens@ejc.net

**Sophia Karakeva**

Marketing and Communications Executive

Datascouting

soka@datascouting.com

**Stavros Vologiannidis**

Professor for the Technological Educational Institute  
of Central Macedonia and Founder

Datascouting

svol@datascouting.com

