







# Session 2: Toxicity Analysis with Communalytic

May 27, 2021

Socialmedialab.ca



The Social Media Lab is a

Multidisciplinary Research Laboratory at

Ryerson University in

Toronto, Canada

The lab studies how social media is changing the way people communicate, share information and form communities online, and how these changes impact society.



#### **Instructors**



Dr. Anatoliy Gruzd
Canada Research
Chair, Associate
Professor, Director of
Research at the
Ryerson University
Social Media Lab



Philip Mai M.A., J.D. Co-Director and Senior Researcher at the Ryerson University Social Media Lab

Socialmedialab.ca







# Video and slides from Session 1 is now available online at: communalytic.com



FAQ TUTORIALS PUBLICATIONS



APR **07** 2021 Social Media Lab's Computational Social Science of Q (CSS) Bootcamp – Summer 2021

By COMMUNALYTIC



What: CSS Bootcamp on Examining Online Discourse & Networks with Communalytic

When: 2<sup>nd</sup> and 4<sup>th</sup> Thursday of the month at 10 am -11:30 am (ET) between May and July, 2021.

Where: Zoom (see details below)
Free Registration via Zoom

3

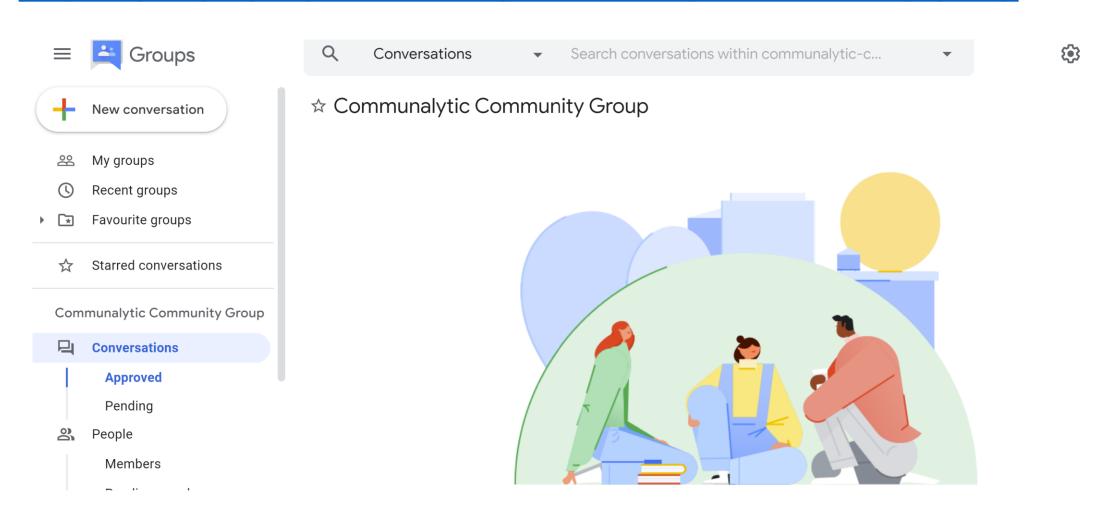
# CSS Bootcamp Schedule Summer 2021

Session #1	Getting Started with Communalytic: Data Collection from Reddit	May 13, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #2	Toxicity Analysis with Reddit Data using Perspective API	May 27, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #3	Getting Started with Communalytic: Data Collection from Twitter (Twitter Thread via API v2.0 and Twitter Academic Track)	June 10, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #4	Toxicity Analysis of Twitter Threads using Perspective API	June 24, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #5	Social Network Analysis of Signed Networks with Reddit and Twitter data	July 8, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #6	Getting Started with Communalytic: Data Collection from Facebook & Instagram (via CrowdTangle API) + Social Network Analysis of Two-mode Semantic Networks with CrowdTangle data	July 22, 2021, 10:00- 11:30am ( <u>EDT</u> )



# Join the Communalytic Community Group

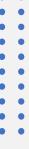
https://groups.google.com/u/1/g/communalytic-community-group



# We're Hiring a Postdoctoral Researcher to Study Dis/Mis-Information Campaigns at Scale



- Must have expertise in applying and evaluating various computational approaches for large-scale network visualization and analysis
- Ideal for candidates with a doctorate in Computational Social Science, Digital Sociology, Communication, Information Systems, Computer Science, Network Science, Complex Systems, Computer Engineering or a related field
- More Info: SocialMediaLab.ca



# Outline



About Communalytic and Anti-social Behaviour Research with Social Media Data



**Manual Content Analysis** 



Automated Dictionary-based Content Analysis



Machine Learningbased Content Analysis Perspective API
Toxicity Analysis with
Communalytic

# Communalytic is a research tool for studying online communities and online discourse.

Communalytic can collect and analyze public data from social media platforms. It uses advanced text and social network analysis techniques to automatically pinpoint toxic and anti-social interactions, identify influencers, map shared interests and the spread of misinformation, and detect signs of possible coordination among seemingly disparate actors.









collect

extract

analyze

visualize

How to choose between Communalytic Edu and Pro.	communalytic EDU	Communalytic PRO
Account Type	Free	\$349/6-mo. to support site infrastructure (server-side data collection, storage, processing, analysis and visualization)
Designed For	Students and is ideal for teaching and learning about social media analytics	Academic researchers and is ideal for large scale academic research projects
Account Caps	≤ 30K records shared across 3 datasets	≤ 10M records shared across 50 datasets
Reddit	Live-collection* of public posts from any public subreddit for ≤ 7 consecutive days (Limit: ≤ 30K posts)	Live* & historical collection of public posts from any public subreddit for ≤ 31 consecutive days (Limit: Account Caps)
Twitter Threads (API ver.2) req. Twitter developer's account	Public replies to any public tweet posted within the previous 7 days (Limit: ≤ 30K tweets)	Public replies to any public tweet posted within the previous 7 days (Limit: ≤500K tweets/month)
Twitter Academic Track (API ver.2) req. Application to Twitter	Not supported	Full-archive historical-search of tweets back to 2006 (Limit: ≤ 10M tweets/month)
CrowdTangle (FB/IG) URL Search req. a CrowdTangle account	Public Facebook or Instagram posts that shared the same URL (Limit: ≤ 30K posts)	Public Facebook or Instagram posts that shared the same URL (Limit: Account Caps)

<sup>\*</sup> Live-collection = the collection of posts/tweets <u>posted on or after</u> the date when you initiated the data collection.

How to choose between Communalytic Edu and Pro.	communalytic EDU	communalytic PRO
Exploratory Data Analysis (EDA)	<ul> <li>Emoji cloud (freq. used emojis)</li> <li>Word cloud (freq. used words)</li> <li>Time series (posts per day)</li> <li>Top posters (top 10)</li> </ul>	<ul> <li>Emoji cloud (freq. used emojis)</li> <li>Word cloud (freq. used words)</li> <li>Time series (posts per day)</li> <li>Top posters (top 10)</li> </ul>
Text Analysis	<u>Toxicity analysis</u> based on machine learning via <u>Google's Perspective API</u>	<u>Toxicity analysis</u> based on machine learning via <u>Google's Perspective API</u>
Social Network Analysis (SNA)	<ul> <li>Reply Network</li> <li>Reply Network with toxicity scores</li> <li>2-mode Semantic Network (for CrowdTangle FB and IG data only)</li> </ul>	<ul> <li>Reply Network</li> <li>Reply Network with toxicity scores</li> <li>2-mode Semantic Network (for CrowdTangle FB and IG data only)</li> </ul>
Additional features	<ul> <li>Simultaneously Run Multiple Data         Collectors: 1 Reddit, 1 Twitter &amp; 1         CrowdTangle</li> <li>Import existing datasets (CSV and         Gzip-ed CSV)</li> <li>Collaboration Friendly: Access to         various team collaboration features</li> </ul>	<ul> <li>Simultaneously Run Multiple Data         Collectors: 2 Reddit, 1 Twitter &amp; 1         CrowdTangle</li> <li>Import existing datasets (CSV and Gzip-ed CSV)</li> <li>Collaboration Friendly: Access to various team collaboration features</li> </ul>

## Notice:

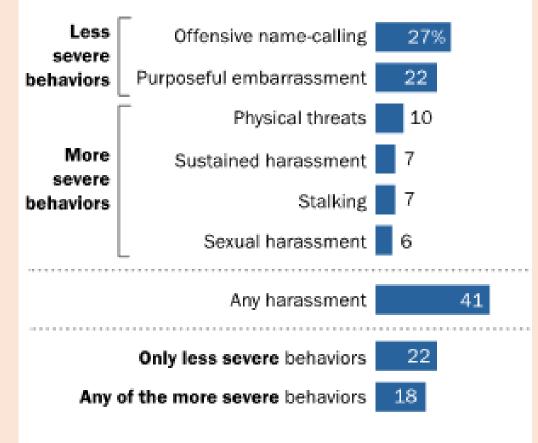
Sample posts included in this presentation are from real users. The content of some posts are offensive.



# One way to study online anti-social behaviour is by using a survey

#### Roughly four-in-ten Americans have personally experienced online harassment

% of U.S. adults who have experienced\_\_\_\_ online



Source: Survey conducted Jan. 9-23, 2017. "Online Harassment 2017"

PEW RESEARCH CENTER

Another way is to examine online content and interactions to look for "manifestations (acts) of anti-social behaviour"

Unlike a survey, when we use social media data to study anti-social behaviour, we are studying observed behaviour, not self-reported behaviour or perception.

# Example of "Manifestations (Acts) of Anti-Social Behaviour"

Hate speech (Southern & Harmer, 2019)

Impoliteness (Theocharis et al., 2016)

Rudeness (Su et al., 2018)

Incivility (Kenski, Coe, & Rains, 2017; Rossini, 2019)

Offensive comments (Kwon & Gruzd, 2017), and

Stereotyping (Southern & Harmer, 2019).

## When Studying Anti-Social Behaviour in Online Discourse

#### Things to keep in mind ...

- For some online groups, what is often referred to as 'anti-social' may be a communal norm and be practiced by group members to socialize;
- But we are interested in studying group dynamics where such behaviour may negatively affect the overall group cohesion and may have psychological and emotional consequences for individuals.
- There is also a concern that some forms of anti-social behaviour, such as hate speech, may galvanize xenophobic behaviour offline and lead to changing social norms at the societal level.
- We now know that what happens online doesn't always stay online.

PSMLabTO 15



# Outline



About Communalytic and Anti-social Behaviour Research with Social Media Data



**Manual Content Analysis** 



Automated Dictionary-based Content Analysis



Machine Learningbased Content Analysis Perspective API
Toxicity Analysis with
Communalytic



### RESEARCH QUESTIONS

RQ1: What is the prevalence of toxic/insulting messages targeting political candidates?



RQ2: Is there a difference in frequency of toxic/insulting messages directed at women versus men candidates on Twitter?

#### DATA COLLECTION



Compiled a comprehensive list of 2,144 #ELXN43 candidates



Identified 1,344 candidates with a public Twitter profile



Collected 363,706 public tweets in English directed at 1,116 candidates (Sept. 28 – Oct. 27, 2019)

#### **METHOD**

3 coders hand coded a **random sample** of 3,637 tweets (1% of 363,706) Only tweets
flagged by all 3
coders as either
toxic or insulting
were considered

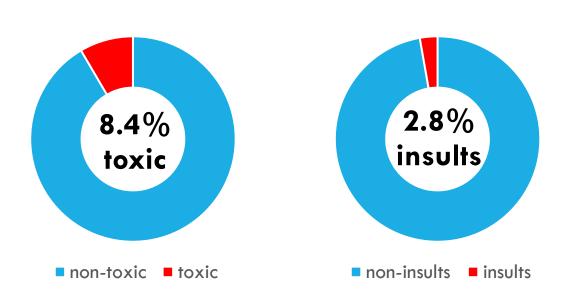


Tested a relationship between a **candidate's gender** & the likelihood of receiving toxic/insulting tweets

(chi-square test)

- A message is toxic when it is rude, disrespectful, or unreasonable
- A message is **insulting** when it is inflammatory/negative toward a particular person or a group of people

# RQ1: WHAT IS THE PREVALENCE OF TOXIC/INSULTING MESSAGES TARGETING POLITICAL CANDIDATES?

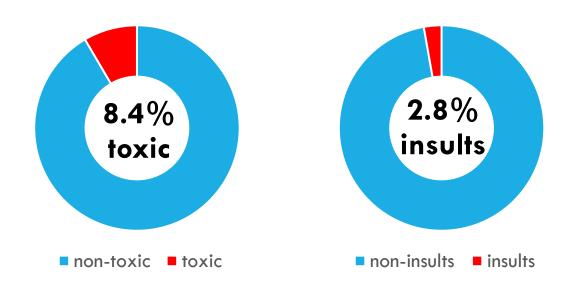


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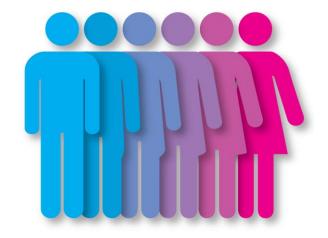


- A message is toxic: when it is rude, disrespectful, or unreasonable
- A message is insulting when it is inflammatory/negative toward a
  particular person or a group of people

#### Related work:

- Southern and Harmer (2019):9.8% of tweets targeting BritishMPs were uncivil
- Gorrel et al. (2019): less than 4% of tweets directed at British MPs were abusive
- Mead (2014); Subrahmanyam et al. (2006): swearing, dismissive insults, and abusive words to make up around 3% of online communications more broadly

# RQ2: IS THERE A DIFFERENCE IN FREQUENCY OF TOXIC/INSULTING MESSAGES DIRECTED AT WOMEN VERSUS MEN CANDIDATES ON TWITTER?



#### Chi-square test

Tested a relationship between a candidate's gender & the likelihood of receiving toxic/insulting tweets

#### Result:

No significant association between gender and receiving a toxic or insulting tweet

#### Related work:

- Gorrel et al. (2019): abuse on Twitter does not depend on gender (UK's MPs)
- Southern & Harmer (2019):
  women were more likely to
  receive certain types of uncivil
  tweets (UK's MPs)

### **IMPLICATIONS**

Out of 307 (8.4%) toxic and 101 (2.8%) insulting tweets flagged by our coders, the **majority of these posts** (255 toxic and 85 insulting tweets) **are still publicly available** as of January 2, 2020

While the overall percentage of toxic and insulting tweets was relatively low (<10%), it's not necessarily their quantity, but also their severity which may negatively impact one's well-being

Irrespective of one's gender, some candidates tend to experience more extreme cases of online violence and toxicity

Social media **platforms need** to take a more **proactive role** in preventing online harassment campaigns against their users

#### IMPLICATIONS — A WAY FORWARD



Coordinated and sustained online harassment and the use of toxic and insulting language by trolls and cyberbullies are ultimately about controlling who can be visible and have a voice in the public sphere.

#### POSSIBLE SOLUTIONS

- ☐ Boost referral-site filtering to prevent coordinated attacks from external site (e.g. 4Chan, Reddit, etc.)
- ☐ Streamline the abuse reporting process to make it more transparent and easier to track complaints
- ☐ Hire more staff to improve complaints response time

# Manual Content Analysis: Pros and Cons

Advantages

- Can detect implicit instances of anti-social acts;
- Can support an analysis within a domain, media format or language not yet/well supported by automation;
- Requires 2 or more independent coders;
- Difficult to get high intercoder reliability;
- Not scalable to analyze large datasets;

Disadvantages



#### **Manual Content Analysis**



Dictionary-based Content Analysis



Machine Learningbased Content Analysis

Perspective API
Toxicity Analysis
with
Communalytic

35

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Outline









Global Affairs Canada Affaires mondiales Canada



# Mapping out Violence Against Women (VAW) on Twitter: a Case of India

Priya Kumar, Anatoliy Gruzd, Philip Mai Social Media Lab

# Based on a published paper:

#### American Behavioral Scientist

# Mapping out Violence Against Women of Influence on Twitter Using the Cyber–Lifestyle Routine Activity Theory

Priya Kumar, Anatoliy Gruzd, Philip Mai



Article information ^







#### **Article Information**

Volume: 65 issue: 5, page(s): 689-711

Article first published online: January 29, 2021; Issue published: May 1, 2021

Priya Kumar<sup>1</sup>, Anatoliy Gruzd<sup>1</sup>, Philip Mai<sup>1</sup>

<sup>1</sup>Ryerson University, Toronto, Ontario, Canada

# Research Questions

 How is online violence against women manifested on Twitter in the Indian context?

 Do different Indian women of influence receive different types of online harassment on Twitter?

 Who are the posters of online harassment, abuse, and violence against women?

# Why India?

- With a population of 1.2 billion, India is commonly referred to as the largest democratic country in the world.
- A recent survey found 41% of women in India have experienced some form of harassment online (Bhargava, 2017).

PSMLabTO 4:

ET Home > Slideshows > People > The 20 most influential global Indian women

#### The 20 most influential global Indian women

Jan 04, 2015, 02.01 PM IST



The 20 most influential global Indian women

#### 10 most powerful female politicians of India

New Delhi: Politics has always been male dominated like other fields in India. Participation of female in this complex world is not so desirable, but Females have come out to be the super power of

India TV News Desk [ Updated: June 12, 2015 18:34 IST ]



10 most powerful female politicians of india

# Study Sample: 101 Indian Women of Influence

#### **Politicians**

#### Mamata Banerjee

First Female Chief Minister of West Bengal @mamataofficial

- Named one of the 100 Most Influential People in the World (Time Magazine, 2012)
- 50<sup>th</sup> Most Influential in Finance (Bloomberg Markets, 2012)



#### **Celebrities**

#### **Deepika Padukone**

Actor

@deepikapadukone

- 24 million followers on Twitter
- Highest-paid actress in India (2018)



## Other Public Figures

#### **Barkha Dutt**

Journalist and News Anchor @BDUTT

- Columnist for Washington Post
- Awarded the Padma Shri (civilian honour) in 2008



## Method

#### **Content Analysis**

- Automated text analysis to detect online swearing (a potential sign of explicit harassment)
- Manual content analysis to validate the "swear word" dictionaries and explore the nature of online harassment

#### **Data Collection Tools**

- Netlytic (data collection, development of dictionaries)
- Excel and R (data cleaning, pre-processing)

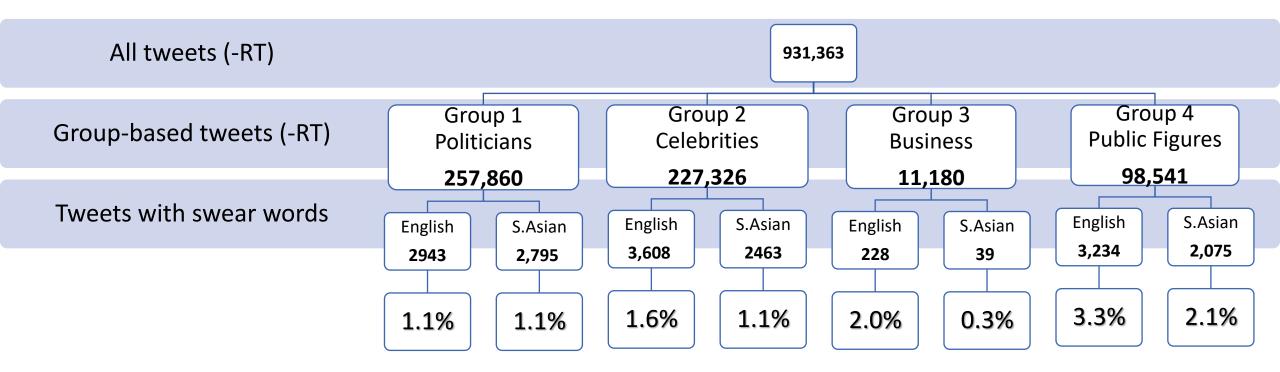
## **Swear Word Dictionaries**

- English (n = 584)
  - Based off

Kwon, K.H. & Gruzd, A. (2017). Is Offensive Commenting Contagious Online? Examining Public vs. Interpersonal Swearing in Response to Donald Trump's YouTube Campaign Videos. *Internet Research*. https://doi.org/10.1108/IntR-02-2017-0072

- South Asian (n = 759)
  - Original (based on an iterative review process)
  - Crowd-sourced (<u>www.youswear.com</u>; <u>www.hindilearner.com</u>)

# Data Collection



Swearing, dismissive insults, and abusive words characteristically make up under 3% of online communications (Mead, 2014; Subrahmanyam, Smahel, & Greenfield, 2006).

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# Results

# Different Accounts Mentioned

• Politicians, prominent Indian news outlets, media sources, and journalists often mentioned in the recorded tweets

# Different Categories of Perpetrators

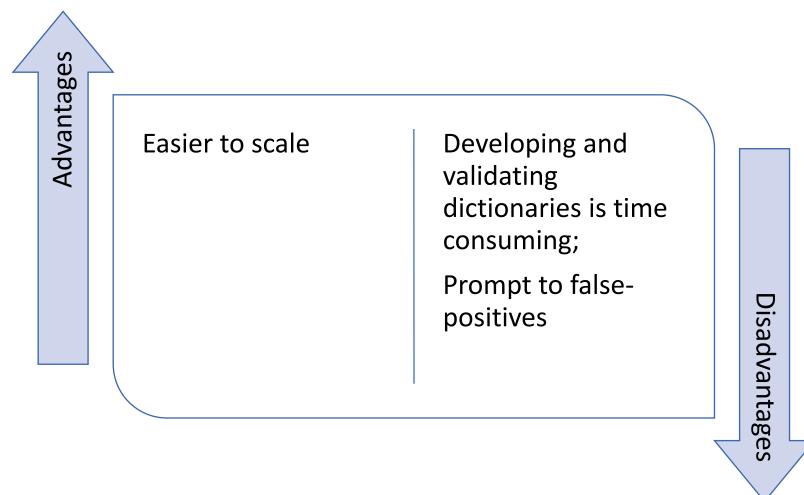
• 'News Junkies', 'Bollywood Fanatics', 'Lone-Wolves'

# Different Types of Abuse, Harassment and Violence

- Celebrities and journalists receive more sexualized and gendered attacks (body/slut-shaming)
- Dismissive and reactionary tweets to politicians and business CEOs based on professional decisions

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# Automated Dictionary-based Content Analysis: Pros and Cons



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### **Manual Content Analysis**



Automated Dictionary-based Content Analysis



Machine Learningbased Content Analysis

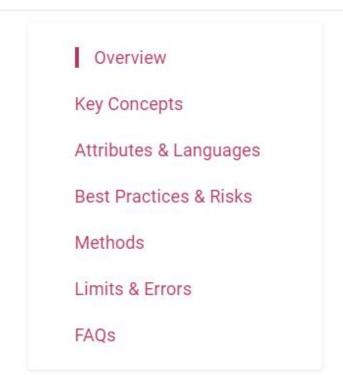
Perspective API
Toxicity Analysis
with
Communalytic

Outline

# Toxicity Analysis with Perspective API

## ♣ Perspective | Developers

About the API Docs Contact Us



Toxicity INPUT: TEXT "Shut up. You're Identity\_Attack Severe\_Toxicity an idiot!" Perspective Threat Insult API OUTPUT: SCORE Toxicity 0.99 Severe\_Toxicity 0.75 Insult 1.0 Likely\_To\_Reject Sexually\_Explicit Sexually\_Explicit 0.04 Profanity 0.93 Profanity Likely\_To\_Reject 0.99 Threat 0.15 Identity\_Attack 0.03

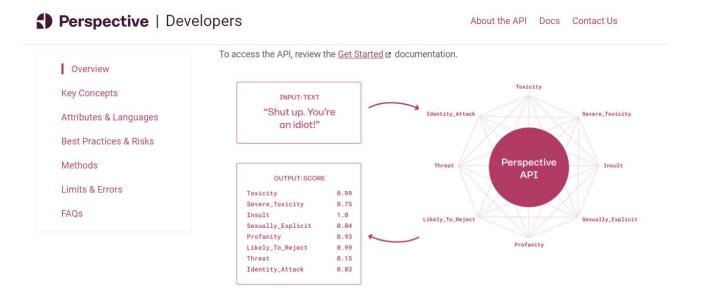
https://support.perspectiveapi.com/s/docs-get-started

To access the API, review the Get Started & documentation.

# Toxicity Analysis with Perspective API

- Training data: online forums including Wikipedia & New York Times comments, with crowdsourced labels (e.g., a comment is "toxic" or not)
- Machine Learning:

   Convolutional Neural
   Network (CNN) trained with
   GloVe word embeddings



https://support.perspectiveapi.com/s/docs-get-started

	Definition	Sample post
Toxicity	Rude, disrespectful, or unreasonable post	"This is one of stupidest things I've read but fuck me I laughed at the second line"
Severe toxicity		
Insult		
Identity attack		
Profanity		
Threat		

	Definition	Sample post
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Severe toxicity	Very hateful, aggressive, disrespectful post. This score is less sensitive to posts that include positive uses of curse words	"Fuck off pathetic loser, no one cares about your worthless opinion"
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Insult	Insulting, inflammatory, or negative post toward an individual or a group	"How fucking stupid is [Name]? That is pretty fucking stupid. What's next - a deep fake having him say racist things as a "social experiment"?
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Profanity		
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Profanity	Post with swear words or other obscene language	"Why vote for the [Political Party Name] when you know they won't win shit."
Threat		

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Profanity	Post with swear words or other obscene language	"Why vote for the [Political Party Name] when you know they won't win shit."
Threat	Post with an intention to inflict pain, injury, or violence against an individual or group	"Shoot all yellow vests! We have to kill all Nazis!"

# Perspective API: Evaluation

True

Rate



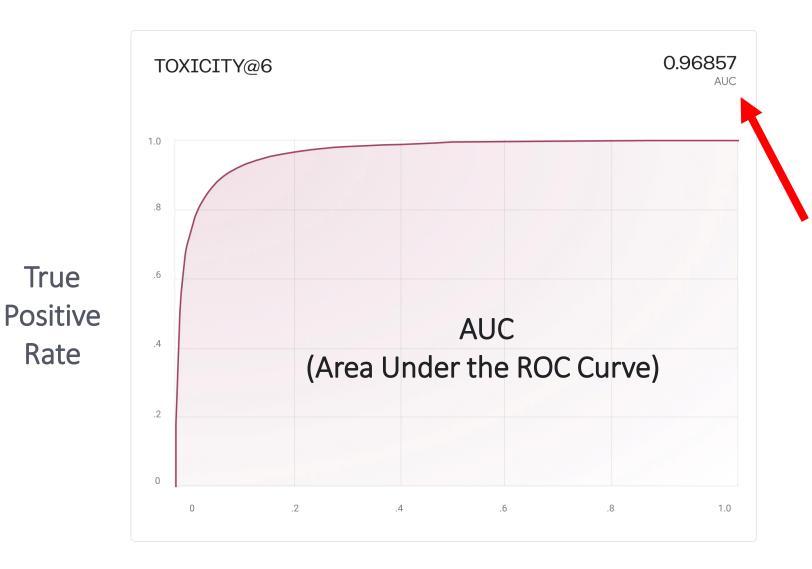
### **Receiver Operating Characteristic (ROC) Curve**

- a chart showing the performance of a classification model.

https://support.perspectiveapi.com/s /about-the-api-best-practices-risks

**False Positive Rate** 

# Perspective API: Evaluation



"AUC ranges in value from 0 to 1. A model whose predictions are 100% wrong has an AUC of 0.0; one whose predictions are 100% correct has an AUC of 1.0."

https://developers.google.com/machine-learning/crash-course/classification/roc-and-auc

### **AUC values**

[0.7-0.8) – acceptable [0.8 to 0.9) – excellent >=0.9 – outstanding (Mandrekar, 2015)

**False Positive Rate** 

# Unintended Bias in Machine Learning Models

### False "toxic" positives

A naively trained model will have some strong unintended biases illustrated by these false-positive examples...

Comment	Toxicity score
The Gay and Lesbian Film Festival starts today.	0.82
Being transgender is independent of sexual orientation.	0.52
A Muslim is someone who follows or practices Islam.	0.46

(Borkan, Dixon, Sorensen, Thain & Vasserman, 2019)

# Perspective API: Identity Subgroup Evaluation

Test dataset	Description
Subgroup AUC	Only examples that mention the specific identity subgroup.
	A low value in this metric => the model does a poor job of distinguishing between toxic and non-toxic comments that mention the identity.
BPSN (Background	Non-toxic examples that mention the identity & Toxic examples that do not.
Positive, Subgroup Negative) AUC	A low value in this metric => the model likely predicts higher toxicity scores than it should for non-toxic examples mentioning the identity.
BNSP (Background	Toxic examples that mention the identity & Non-toxic examples that do not.
Negative, Subgroup Positive) AUC	A low value in this metric => the model likely predicts lower toxicity scores than it should for toxic examples mentioning the identity.

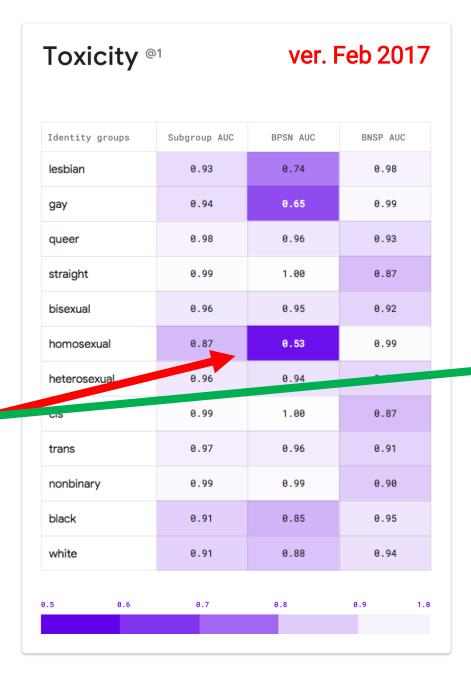
# Perspective API: Unitary Identity Subgroup Evaluation

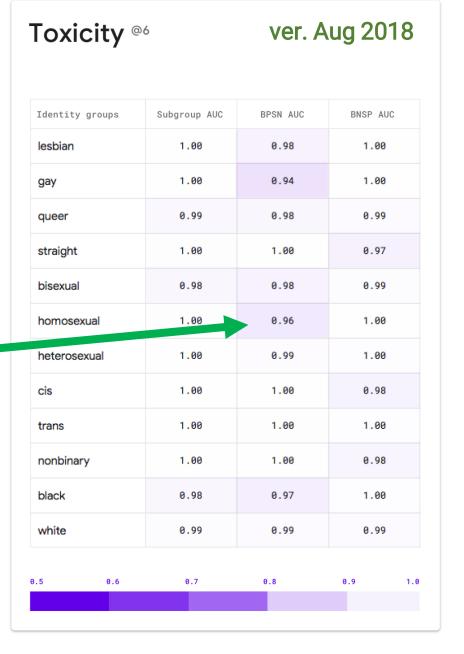
A low value in this metric

=> the model likely predicts
higher toxicity scores than
it should for non-toxic
examples mentioning the
identity.

### **AUC values**

[0.7-0.8) – acceptable [0.8 to 0.9) – excellent >=0.9 – outstanding (Mandrekar, 2015)





https://medium.com/jigsaw/increasing-transparency-in-machine-learning-models-311ee08ca58a

# Perspective API: Intersectional Identity Subgroup Evaluation

A low value in this metric

=> the model likely predicts
higher toxicity scores than
it should for non-toxic
examples mentioning the
identity.

### **AUC values**

[0.7-0.8) – acceptable [0.8 to 0.9) – excellent >=0.9 – outstanding (Mandrekar, 2015)



Toxicity ®	6 <b>V</b>	er. Aug	2018
Identity groups	Subgroup AUC	BPSN AUC	BNSP AUC
black gay	1.00	0.89	1.00
black queer	0.97	0.96	0.99
straight	0.99	0.99	0.98
black bisexual	0.95	0.93	0.99
black homosexual	1.00	0.92	1.00
black heterosexual	1.00	0.97	1.00
black cis	1.00	1.00	0.99
black trans	1.00	0.98	1.00
black nonbinary	1.00	1.00	0.99
white lesbian	1.00	0.98	1.00
white gay	1.00	0.95	1.00
white queer	1.00	0.99	0.99
white straight	1.00	1.00	0.98
white bisexual	1.00	0.98	0.99
white homosexual	1.00	0.97	1.00
white heterosexual	1.00	1.00	1.00
white cis	1.00	1.00	0.97
white trans	1.00	1.00	1.00
white nonbinary	1.00	1.00	0.98
5 θ.6	0.7	0.8	0.9 1.0

https://medium.com/jigsaw/increasing-transparency-in-machine-learning-models-311ee08ca58a

# How to access Perspective API

### via Python script

#### Python

Here is a sample request and response using the Python version of the Google API Client Libraries.

- 1. Install the Python client library 2.
- 2. Run the following commands:

```
Python
                                                                           Сору
Docs > Sample Requests > Python
1 from googleapiclient import discovery
2 import json
4 API_KEY = 'copy-your-api-key-here'
6 client = discovery.build(
    "commentanalyzer",
    "v1alpha1",
    developerKey=API KEY,
     discoveryServiceUrl="https://commentanalyzer.googleapis.com/$discovery/rest?vers
     static discovery=False,
14 analyze request = {
     'comment': { 'text': 'friendly greetings from python' },
     'requestedAttributes': {'TOXICITY': {}}
19 response = client.comments().analyze(body=analyze request).execute()
20 print(json.dumps(response, indent=2))
```

### via Web Interface in Communalytic



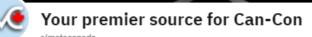
# Sample dataset

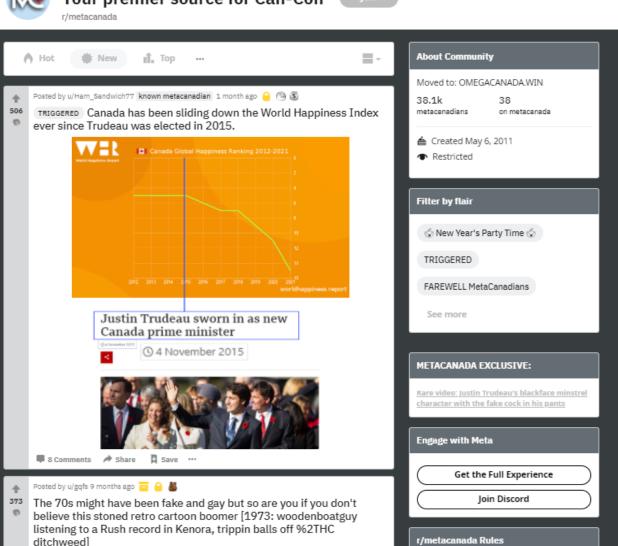
### r/metacanada

"Forum largely (but not exclusively) populated by conservatives"

Note: the group has now moved to another platform.







Always remember kids:

There're

1. No Doxxing

2. No brigading

3. Use NP for reddit links

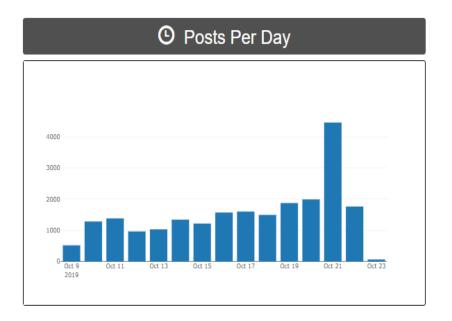
# Sample dataset

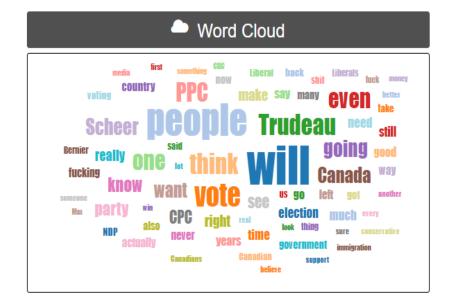
### r/metacanada

- Total records = 22k
- Oct 9- 23, 2019











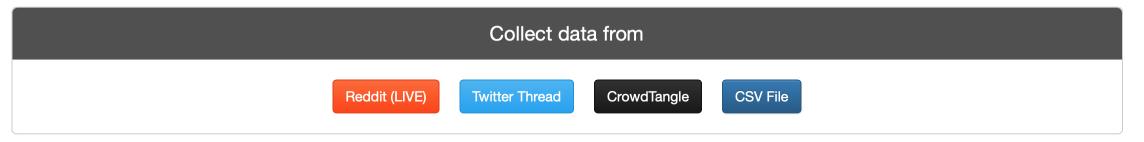


## My Datasets

FAQ

Tutorials

Server Time: May 11, 2021 15:04 UTC



Search dataset list...

### **API Keys**

Twitter Bearer Token:

Remove Key

CrowdTangle API:

Apply for academic/research access to Facebook's CrowdTangle here

**Enter Key** 

Perspective API:

-

**Enter Key** 

Perspective is a machine learning API by Jigsaw and Google designed to conduct a 'toxicity' analysis of online comments. To use this API within Communalytic, please follow these steps to generate an API key. Once generated, enter your API key in the text field above.

\*Note: We recommend using a personal Gmail account to request an API key, some institutional emails may block Google Cloud console / API key requests

## Perspective | Developers

Overview

Getting Started

Enable the API

Sample Requests

### Prerequisites

You must have a <u>Google account</u> , giving you access to the suite of Google products including Google Cloud.

You also must have a Google Cloud project to authenticate (but not necessarily host) your API requests. Go to the Google Cloud console and use an existing project or follow these steps to create a new one:

https://developers.perspectiveapi.com/s/docs-get-started

## Perspective | Developers

Overview

**Getting Started** 

Enable the API

Sample Requests

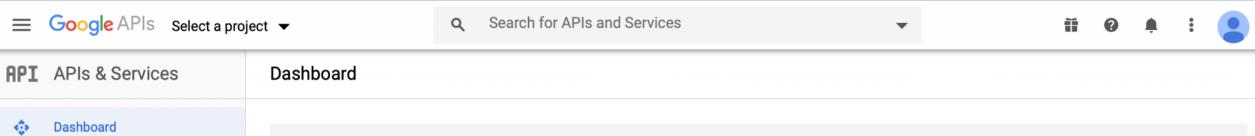
### Prerequisites

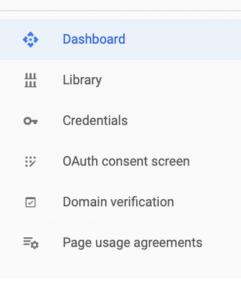
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https://developers.perspectiveapi.com/s/docs-get-started

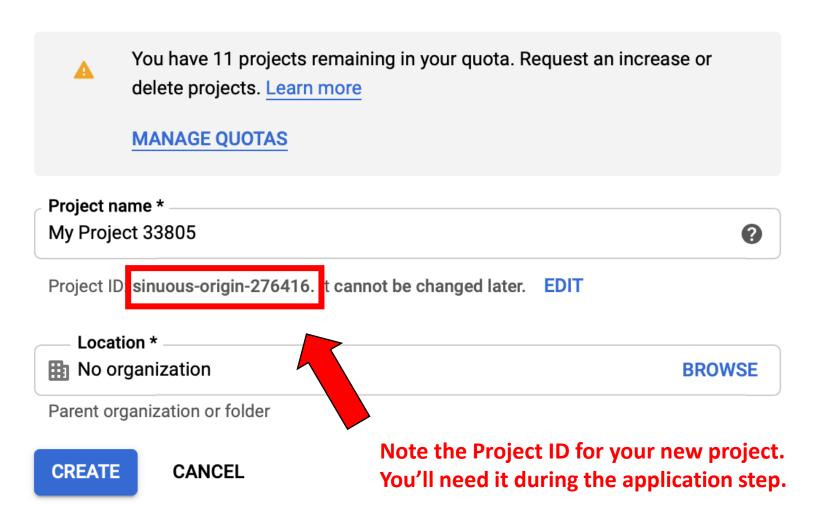








### **New Project**





### Get Access to Perspective API

In order to gain access to Perspective API, you will need a Google Cloud project (<a href="mailto:console.cloud.google.com">console.cloud.google.com</a>). Upon completion of this form, you will receive an email confirmation and be able to view and enable the API.

\* Required

#### Contact Information

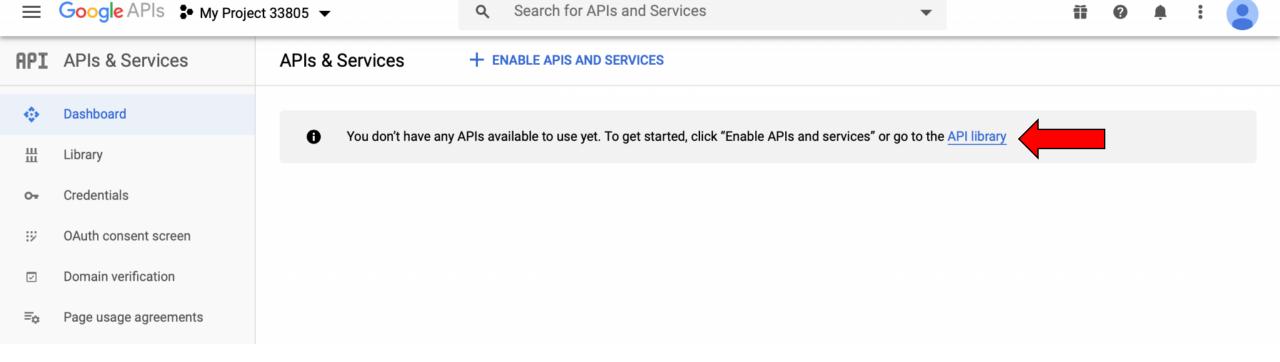
Full Name \*

Your answer

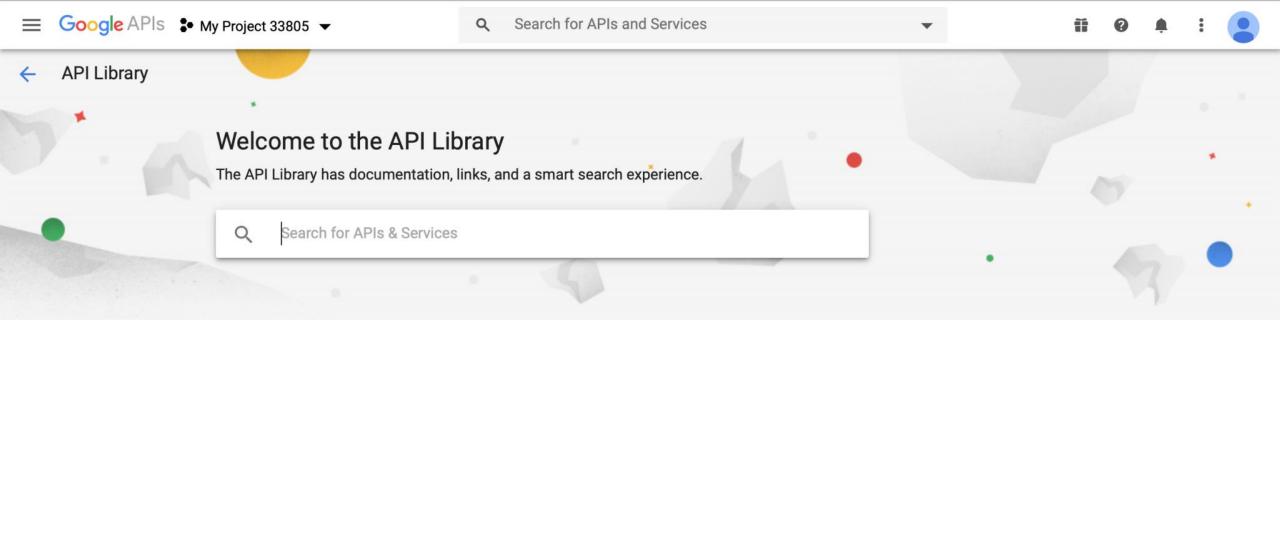
Email Address \*

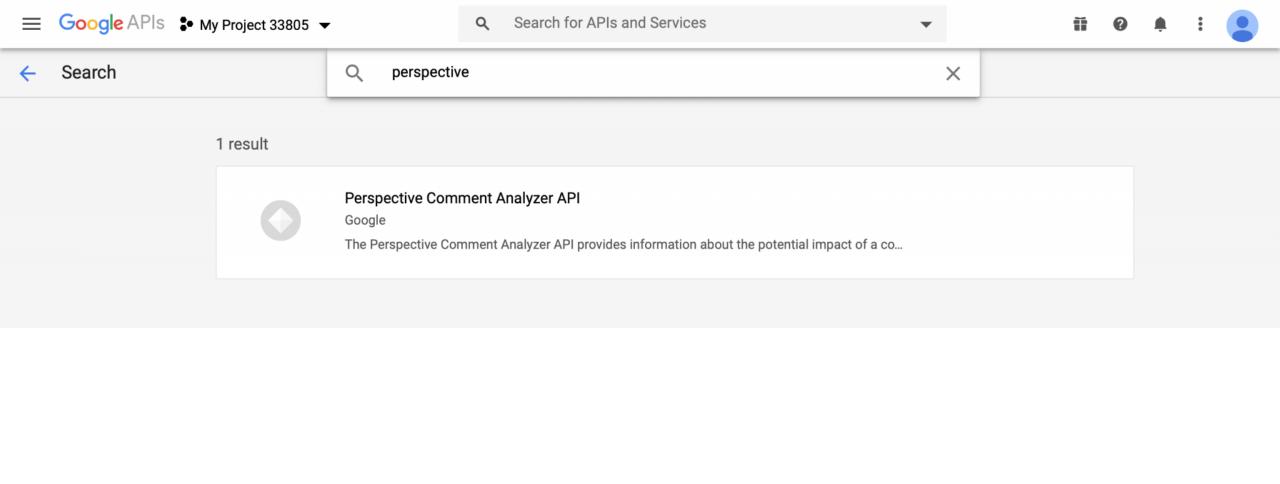
Please provide the email address you used to access the Google Cloud console.

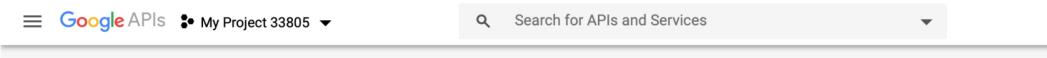
Your answer



https://console.developers.google.com/apis/library/commentanalyzer.googleapis.com













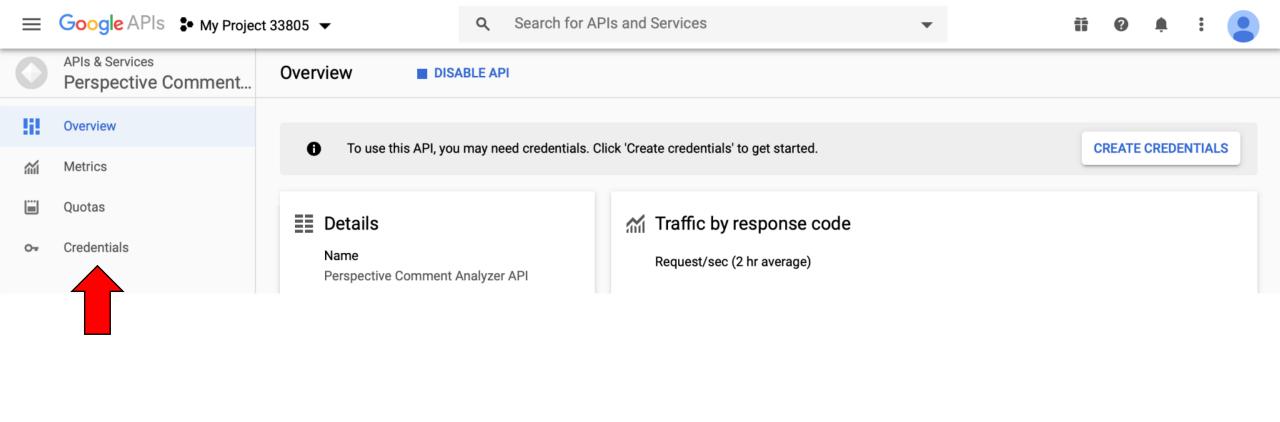
### Perspective Comment Analyzer API

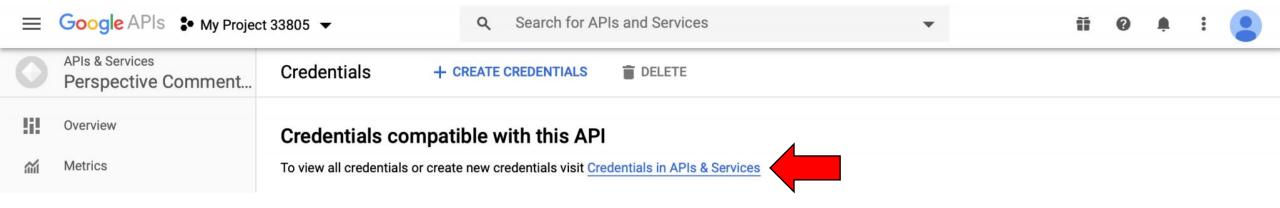
#### Google

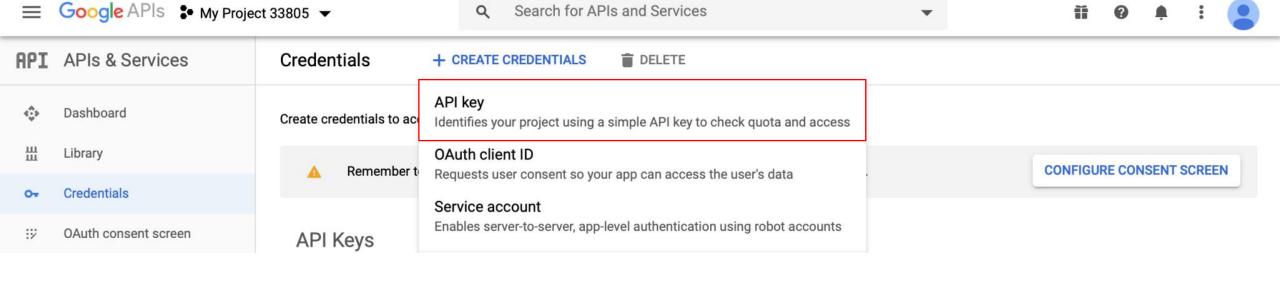
The Perspective Comment Analyzer API provides information about the potential impact of a comment...

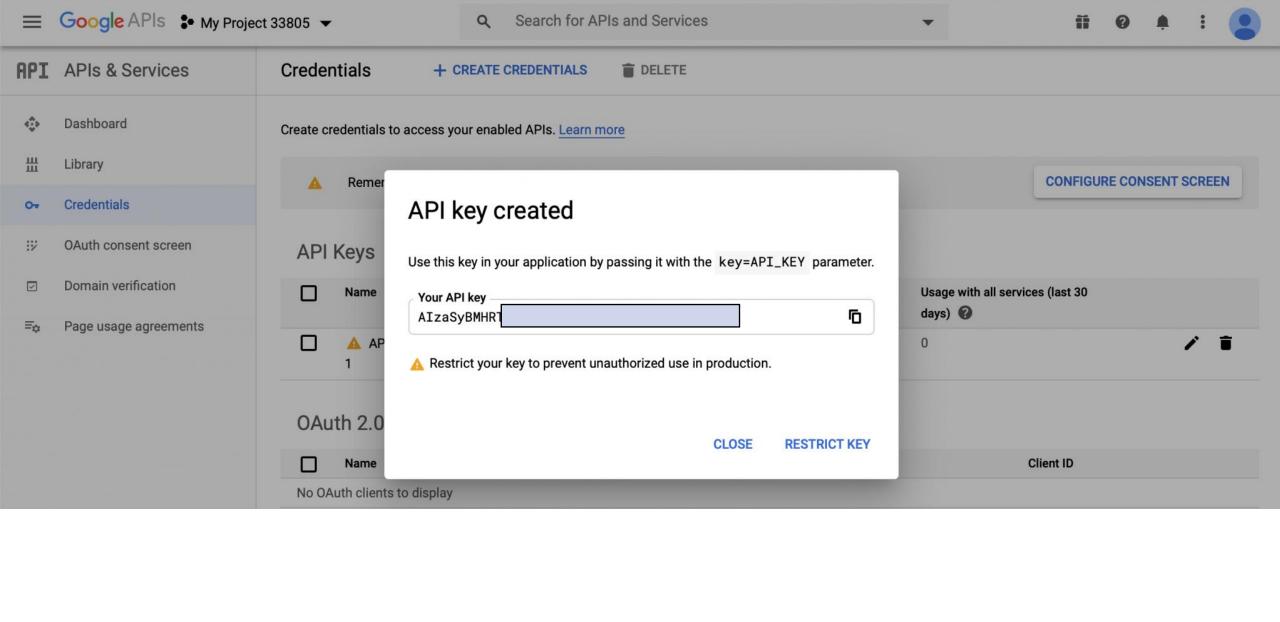


https://console.developers.google.com/apis/library/commentanalyzer.googleapis.com









## **API Keys**

Twitter Bearer Token:

Remove Key

CrowdTangle API:

Apply for academic/research access to Facebook's CrowdTangle here

**Enter Key** 

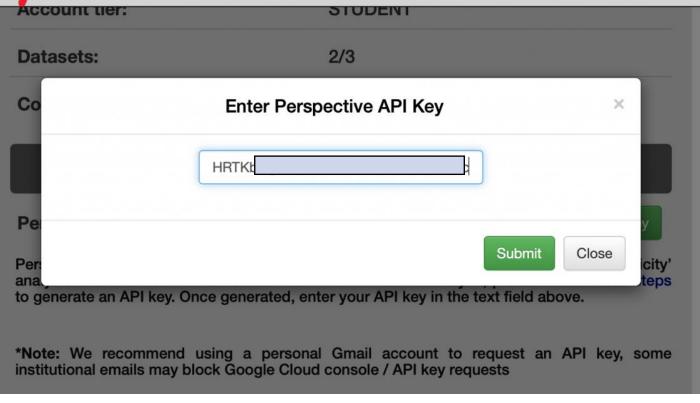
Perspective API:

-

**Enter Key** 

Perspective is a machine learning API by Jigsaw and Google designed to conduct a 'toxicity' analysis of online comments. To use this API within Communalytic, please follow these steps to generate an API key. Once generated, enter your API key in the text field above.

\*Note: We recommend using a personal Gmail account to request an API key, some institutional emails may block Google Cloud console / API key requests



### **API Keys Twitter Bearer** AAAAAAAA Remove Key Token: CrowdTangle Apply for academic/research access to Facebook's **Enter Key** API: CrowdTangle here **Perspective AlzaSyBMH** Remove Key API:

Perspective is a machine learning API by Jigsaw and Google designed to conduct a 'toxicity' analysis of online comments. To use this API within Communalytic, please follow these steps to generate an API key. Once generated, enter your API key in the text field above.

\*Note: We recommend using a personal Gmail account to request an API key, some institutional emails may block Google Cloud console / API key requests

FAQ

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**Publications** 

■ My Datasets

1 -

Reddit (LIVE)

**Twitter Thread** 

CrowdTangle

**CSV File** 

×

Search dataset list...

My Datasets

**Shared With Me** 

Total records in the account: 186 / 30000

Dataset Name	Number of Records	Toxicity Analysis	Network Analysis	Download Dataset	Collaborators	Delete
metacanada till Feb 1 Q reddit : metacanada	6,675 🗠	+	*	<b>⊥</b>	1	
metacanada 4 book chapter <b>Q</b> reddit:-	22,560 ? Check for missed submissions	4	*	<u>.</u>	1	
metacanada Q reddit : metacanada	6,703 🛂	4	*	<u>+</u>	1	

FAQ

Tutorials

**Publications** 

■ My Datasets



#### ← Back to My Datasets

#### ■ Overview

Dataset Name: metacanada 4 book

Name: chapter

Platform: reddit

Subreddit:

Collection started:

2020-02-11 02:11

Collection before:

2020-02-11 02:11

Records: 22,560

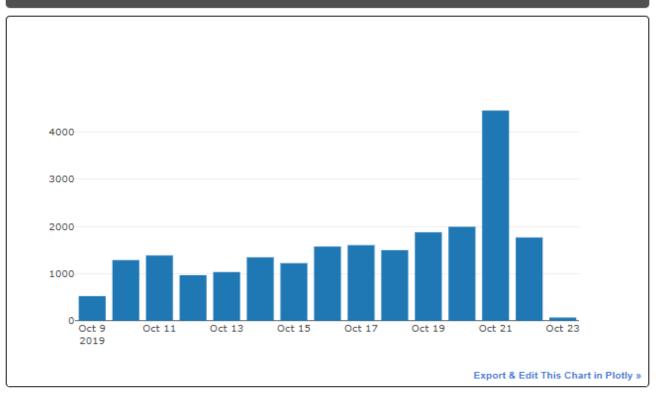
#### Posts Per Day

Word Cloud

11 Top Ten Posters

- Toxicity Analysis
- \* Network Analysis
- Download Dataset

## Posts Per Day





\_**◀**≡

#### **■** Overview

Dataset Name: metacanada 4 book

**Communalytic** 

chapter

Platform:

reddit

Subreddit: -

Collection started:

2020-02-11 02:11

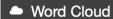
Collection before:

2020-02-11 02:11

Records:

22,560

#### O Posts Per Day



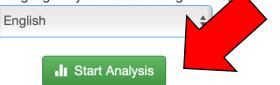
↓¹ Top Ten Posters

- Toxicity Analysis
- \* Network Analysis

## **Toxicity Analysis**

The Toxicity analysis can only run with one language mode at a time.

Please select the primary language of your dataset using the drop own menu below.



- Toxicity Analysis is based on a machine learning API called Perspective API by Google. If you already have an API key for Perspective, it can be added to your account under My Profile; otherwise, see this guide on how to generate a new Perspective API.
- Due to the post length restriction imposed by the API, Communalytic can only analyze the first 3000 characters of each post. Posts longer than 3000 characters will be automatically truncated.
- Link's and URLs will be removed before text is sent to API for analysis.
- The resulting toxicity scores will be added to the export files and available for download via the "Export Posts" and "Export Network" options.
- The API currently supports the following languages: English, French, German, Italian, Portuguese, Spanish.

## **Toxicity Analysis**

### Analysis in Progress ...

You may close this window and visit it later.

Progress: 59 / 4451

Check progress in 43 sec

Estimated Time Left: 1h 20m

Cancel Analysis

### **Toxicity Analysis**

There were 22560 comments analyzed in English

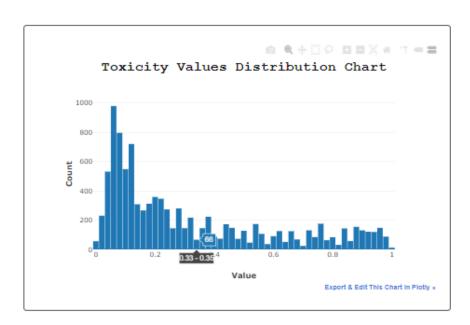
Reset Analysis

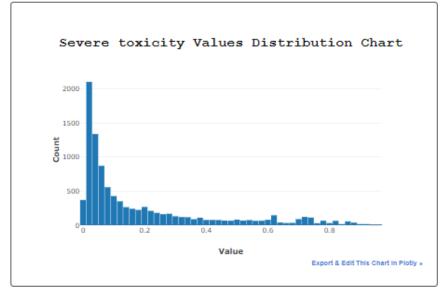
▲ The Toxicity analysis can only run with one language mode at a time. To change the primary language selection for this dataset, click on the Reset Analysis button and rerun the analysis as needed.

Click on the highest/lowest values to see the top posts for each category

Download the Toxicity analysis results as a CSV file.

	Average for dataset	Highest valu	owest value	
Toxicity 6	0.32	1.00	0.00	
Severe toxicity 😉	0.19	0.95	0.00	
Identity attack 😉	0.25	0.98	0.00	
Insuit 😉	0.28	0.99	0.00	
Profanity 6	0.26	0.99	0.00	
Threat 😉	0.22	0.99	0.01	
Sexually Explicit 😉	0.17	1.00	0.00	
Flirtation 6	0.30	0.98	0.02	



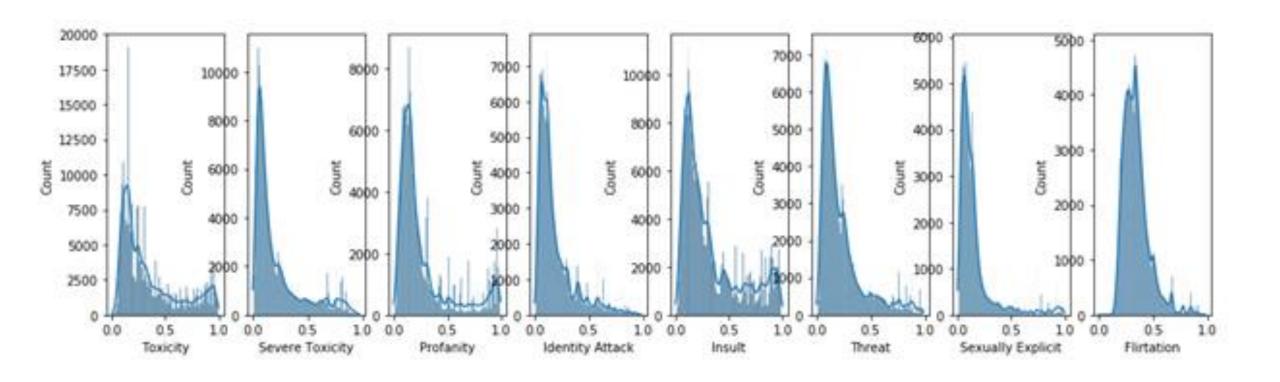


### Ten Highest Scored Posts

### With Highest toxicity Scores in r/-

S.No	Post	Score	Туре	Author	Posted On
1	Everyone who voted ppc	1.00	Submission	monkeygoneape	2019/10/21 10:10:17
2	Fuck off you whiny bitch. " my coalition "?? Wtf you talking about you shut in lonely loser	0.99	Reply	buckiliketofuck	2019/10/20 09:10:41
3	fuck off ya big fat bitch	0.99	Reply	IFIFIFIFIFOKIEDOKIE	2019/10/19 02:10:34
4	Fuck trolls and fuck you.	0.99	Reply	ourtomato	2019/10/22 10:10:01
5	Your such a fucking tool. PPC did cost the Cons anything. Scheer cost us because he is a spineless idiot.	0.99	Comment	trump997964	2019/10/21 11:10:19
6	Fuck Iran	0.99	Comment	Mew16	2019/10/18 04:10:12
7	Fuck Iran	0.99	Reply	None	2019/10/18 10:10:31
8	SUCK IT GOODALE YOU FUCKING CUCK.	0.99	Comment	DontFallForHillary	2019/10/21 11:10:13
9	LMAO. Serves you right, trying to vote and shit. You stupid fucking loser.	0.99	Comment	chimpchimp7	2019/10/13 08:10:25
10	Because you're a dick, asshole	0.99	Reply	barosa	2019/10/16 02:10:07

## Distribution of toxicity scores



# Choosing an appropriate threshold

	Number and Percentage of Posts with the Scores					
Threshold	>=0.7		>=0.8		>=0.9	
Toxicity	3376	15.0%	2287	10.1%	1198	5.3%
Severe toxicity	1401	6.2%	497	2.2%	54	0.2%
Insult	2658	11.8%	1515	6.7%	709	3.1%
Profanity	3358	14.9%	2671	11.8%	1595	7.1%
Identity attack	1114	4.9%	538	2.4%	99	0.4%
Threat	386	1.7%	241	1.1%	52	0.2%

## Perspective API: Evaluation

True

Rate



## **Receiver Operating Characteristic (ROC) Curve**

- a chart showing the performance of a classification model.

https://support.perspectiveapi.com/s /about-the-api-best-practices-risks

**False Positive Rate** 

# Correlation across toxicity scores

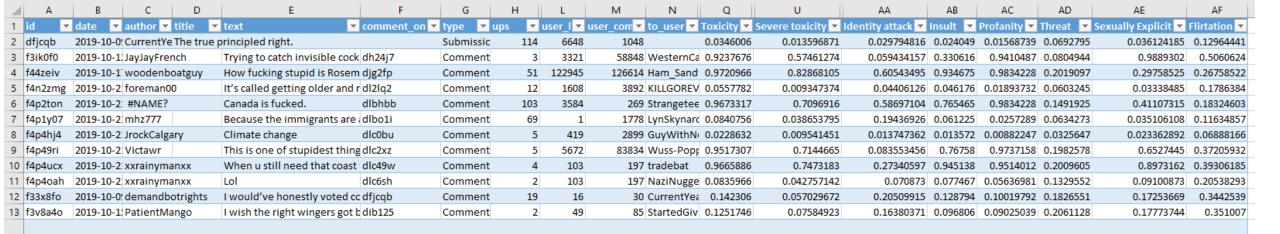
	Toxicity	Severe toxicity	Insult	Profanity	Identity attack	Threat
Toxicity	1	0.948	0.962	0.96	0.688	0.475
Severe toxicity	0.948	1	0.908	0.942	0.668	0.517
Insult	0.962	0.908	1	0.917	0.728	0.457
Profanity	0.96	0.942	0.917	1	0.578	0.402
Identity attack	0.688	0.668	0.728	0.578	1	0.503
Threat	0.475	0.517	0.457	0.402	0.503	1

Note: All correlation values are significant at the 0.01 level (2-tailed)



#### Toxicity scores





# CSS Bootcamp Schedule Summer 2021

Session #1	Getting Started with Communalytic: Data Collection from Reddit	May 13, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #2	Toxicity Analysis with Reddit Data using Perspective API	May 27, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #3	Getting Started with Communalytic: Data Collection from Twitter (Twitter Thread via API v2.0 and Twitter Academic Track)	June 10, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #4	Toxicity Analysis of Twitter Threads using Perspective API	June 24, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #5	Social Network Analysis of Signed Networks with Reddit and Twitter data	July 8, 2021, 10:00- 11:30am ( <u>EDT</u> )
Session #6	Getting Started with Communalytic: Data Collection from Facebook & Instagram (via CrowdTangle API) + Social Network Analysis of Two-mode Semantic Networks with CrowdTangle data	July 22, 2021, 10:00- 11:30am ( <u>EDT</u> )



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