

# **Datathon and Machine Learning Competition on Antisemitism**



# Agenda & Objectives

## Agenda

1. Prerequisites & Registration
2. Challenge Description #1
3. Working with X's Advanced Search
4. Scraping with Bright Data
5. Pre-Processing the Data
6. Annotations Portal Walkthrough

## Objectives: What You'll Learn Today:

- How to register and get started
- How the Annotation Portal works
- How to approach the ML challenge
- Where to find tools and datasets
- How to succeed as a team

Download this Presentation: [Click the link](#)

# Prerequisites & Setup

Before You Start:



A computer with internet access



Twitter

A X account



A Gmail account



Access to Google Colab (<https://colab.research.google.com/>)

Stop & Do Now:



Register on the Annotation Portal: <https://annotate.osome.iu.edu>



Check Github: <https://github.com/AnnotationPortal/DatathonandHackathon.github.io/blob/main/README.md>



Read Challenge Description: [https://github.com/damieh1/datathon\\_2025/blob/main/Datathon\\_Challenge.pdf](https://github.com/damieh1/datathon_2025/blob/main/Datathon_Challenge.pdf)

# What is the Datathon?

Today's presentation will focus only on Part I of the Datathon competition.

## A Two-Part Challenge on Hate Speech Detection

### Part I: Annotation Competition (Workshop 13<sup>th</sup> & 20<sup>th</sup> July)

- Scrape Social Media Posts from X
- Apply a Framework to Classify Antisemitic Content
- Label and Annotate a Dataset

### Part II: Machine Learning Model (Workshop 27<sup>th</sup> July)

- Train a model using annotated data
- Submit predictions & Evaluation Metrics
- Error Analysis

# Instructions for Challenge #1

## Subtasks include:

1. Define your scraping focus (hashtags, user groups, topics) and document your rationale and potential biases.
2. Use the [Bright Data](#) interface to scrape at least 100 relevant user-generated posts from [X.com](#).
3. Annotate your data using a structured definition of antisemitism and hate speech.
4. Prepare a [X/Twitter dataset](#), and include a dataset report with label definitions, distribution information, and annotation rationale.

# Instructions for Challenge #1

## Deliverables for Challenge #1

- Adapting and implementing an existing definition of antisemitism
- Reporting how the data was scraped and which guidelines were used to classify and annotate the data in a standardized way

Gain **+10 bonus points** by evaluating the consistency of your team's annotations using an inter-annotator agreement (IAA) metric.

This means:

- Having at least two annotators label a shared subset of the data
- Calculating a formal agreement score, such as:
  - **Cohen's Kappa** (for binary or pairwise categorical annotation)
  - **Krippendorff's Alpha** (especially for multi-class or missing data)

Clearly report:

- Which subset was double-annotated
- Your score and a brief interpretation (e.g., "moderate agreement," "high agreement")

# Working With X's Advanced Search Function

Top of the pop-up menu

Advanced search

**Words**

All of these words  
Example: what's happening · contains both "what's" and "happening"

This exact phrase  
Example: happy hour · contains the exact phrase "happy hour"

Any of these words  
Example: cats dogs · contains either "cats" or "dogs" (or both)

None of these words  
Example: cats dogs · does not contain "cats" and does not contain "dogs"

These hashtags  
Example: #ThrowbackThursday · contains the hashtag #ThrowbackThursday

Search

Bottom of the pop-up menu

Advanced search

**Engagement**

Minimum replies  
Example: 280 · posts with at least 280 replies

Minimum Likes  
Example: 280 · posts with at least 280 Likes

Minimum reposts  
Example: 280 · posts with at least 280 reposts

**Dates**

From

Month Day Year

To

Month Day Year

Search

1. Go to: → X (Twitter) → <https://x.com/search-advanced>
2. Specify dates, e.g., May 8, 2024.
3. Click "Search."
4. Select posts with a minimum of 200 views.
5. Go to user profiles and copy URLs to a spreadsheet.
6. Goal: Find a wide range of users who engage in online discourse.

# Working With Bright Data

Initiate Dictionary Logs

Add inputs ⓘ

url ⓘ Required max\_number\_of\_posts

https://x.com/RealCandaceO	250
https://x.com/TuckerCarlson	250
https://x.com/infowars	250
https://x.com/jacksonhinkle	250

Count: 4

+ Add input

☒ Include errors report with the results

☐ Define delivery settings

☐ Select a custom output schema

Start collecting

My scrapers / X (formerly Twitter) - Profiles

**X (formerly Twitter) - Profiles**

Collect profiles by profile URL

x.com Public web data 1.6K

Average response time per input: 2s

Starts at: \$0.0015 per record

Initiate Dictionary Logs

Do you need pre-collected data as well? You can find it on our dataset marketplace. Find Data Now

Download logs

s\_mcc3ipzn1hsxcqm

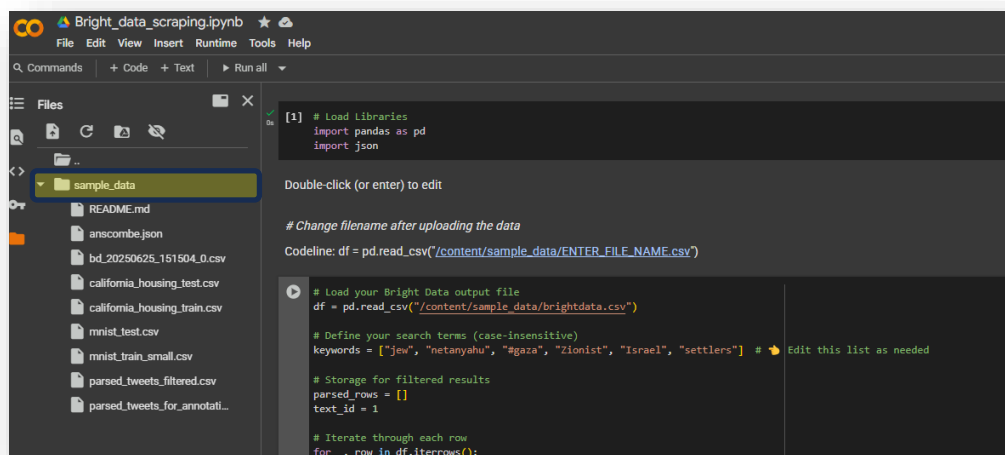
Snapshot ID	Status	Scraper name	Timestamp	Collection time	Success rate	Number of records	Number of records
s_mcc3ipzn1hsxcqm	Ready	X (formerly Twitter) - Profiles - collect by URL	2025-06-25 15:15:04	21s	100.00%	4	4

Download

1. Go to: Bright Data → Web Scrapes → X (Twitter) → Posts → Discover by URL → [https://brightdata.com/cp/scrapers/no\\_code](https://brightdata.com/cp/scrapers/no_code)
2. Click: Add Inputs → <https://x.com/RandomXUser>
3. Specify Number of Posts → max. 250 per User
4. Start Collecting → Runs the Query
5. Download Output as .CSV



# Working With Google Colab



```
[1] # Load libraries
import pandas as pd
import json

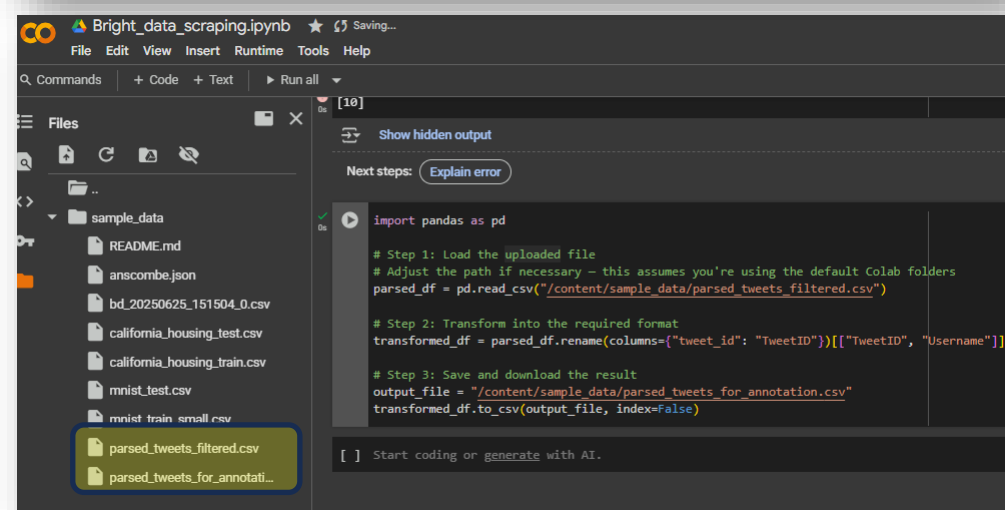
# Change filename after uploading the data
Codeline: df = pd.read_csv("/content/sample_data/ENTER_FILE_NAME.csv")

# Load your Bright Data output file
df = pd.read_csv("/content/sample_data/brightdata.csv")

# Define your search terms (case-insensitive)
keywords = ["jew", "netanyahu", "agaza", "zionist", "israel", "settlers"] # Edit this list as needed

# Storage for filtered results
parsed_rows = []
text_id = 1

# Iterate through each row
for _, row in df.iterrows():
```



```
[10]

Next steps: Explain error

import pandas as pd

# Step 1: Load the uploaded file
# Adjust the path if necessary - this assumes you're using the default Colab folders
parsed_df = pd.read_csv("/content/sample_data/parsed_tweets_filtered.csv")

# Step 2: Transform into the required format
transformed_df = parsed_df.rename(columns={"tweet_id": "TweetID"})[["TweetID", "Username"]]

# Step 3: Save and download the result
output_file = "/content/sample_data/parsed_tweets_for_annotation.csv"
transformed_df.to_csv(output_file, index=False)

[ ] Start coding or generate with AI.
```

1. Go to: Google Colab → <https://colab.research.google.com>
2. Upload: Bright Data Output .CSV
3. Parse Data  
→ Run Code on Colab
4. Prepare Data for Annotation Portal  
→ Run Code on Colab
5. Download compatible .CSV Output

Click this →

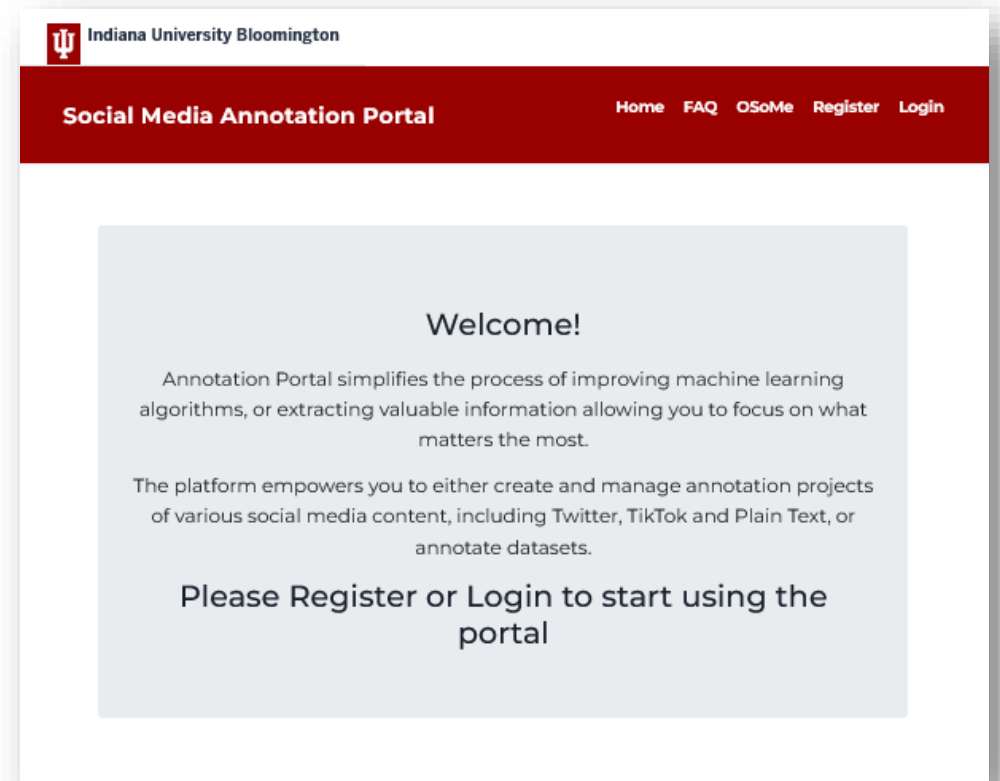


# Annotation Portal Walkthrough

## How to Use the Portal:

1. Register and & Login
2. Appoint a Manager to Your Team
3. Create New Project
4. Create a Sample
5. Create Annotation Scheme and Questions
6. Important! Do not start annotating before the schema has been fully created.
7. Export when done

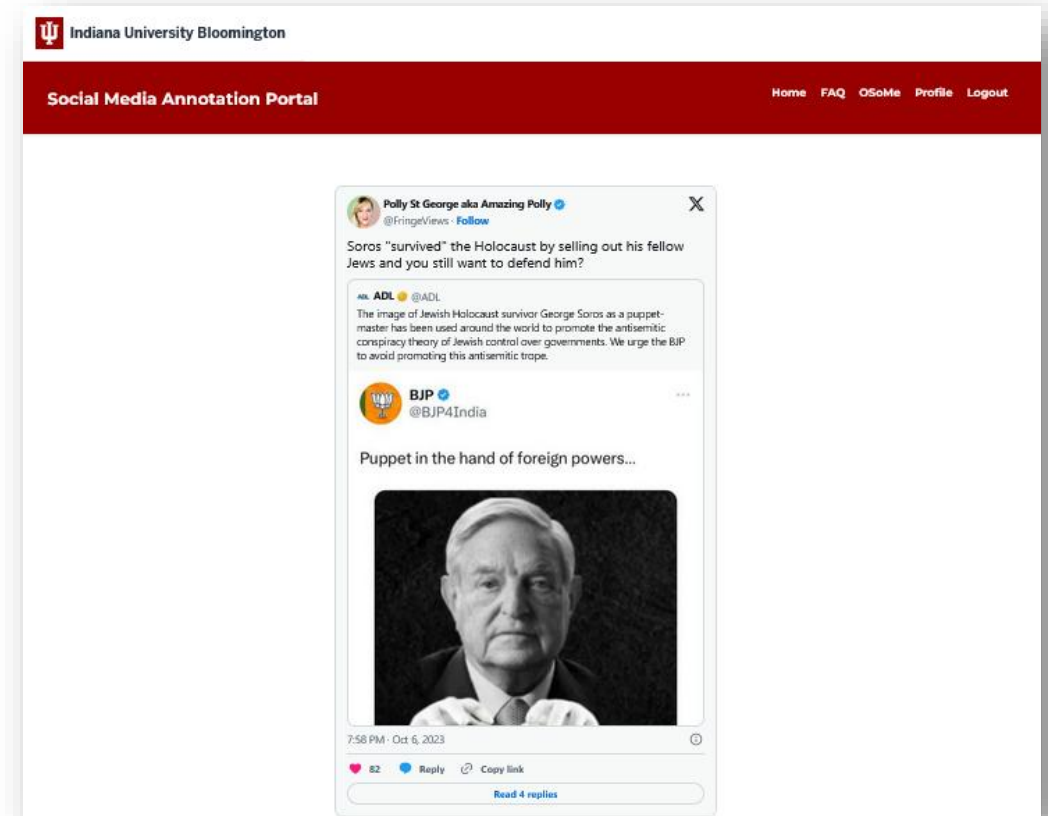
 Demo Pause: Let's annotate 1 example together!



<https://annotate.osome.iu.edu/>

# Annotation Best Practices

- ✓ **Tips for Quality Annotation:**
- Read carefully – context matters
  - Be consistent with labels
  - Use the comment field if unsure
  - Check your export for errors





# Summary & Final Tips

Okay, before you jump in, let's make sure you crossed everything on the checklist.

## Recap Checklist:

- ☐ Registered on the Portal, X
- ☐ Setup Annotation Scheme
- ☐ Registered Accessed GitHub and Colab
- ☐ Explored ML challenge resources
- ☐ Read Challenge Descriptions 1 and 2 thoroughly

-  Questions? Reach out to your instructor or organizers.
-  Remember: Collaboration, Curiosity & Critical Thinking!

# Instructions for Challenge #2

## Pre-Annotated Datasets:

- Use our pre-annotated gold standard dataset to build and evaluate a hate speech detection system.
- Select and combine the following curated datasets:
  - [Antisemitism on Twitter: A Dataset for Machine Learning and Text Analytics](#)
  - [Antisemitism on X: Trends in Counter-Speech and Israel-Related Discourse Before and After October 7](#)
- Datasets are classified as either biased or non-biased

## Subtasks include:

1. Download the (Goldstandard/GroundTruth) datasets listed above
2. Use state-of-the-art transformer models to train and fine-tune a system to detect antisemitic content.
3. Evaluate your model:
  - Report precision, recall, F1-score, and display a confusion matrix.
  - List the hyperparameters used for training.
  - Conduct error analysis and provide qualitative examples, especially false positives.


# ML Challenge & Resources

## **Phase 2: Machine Learning Challenge**

- Download gold-standard dataset
- Use Colab or local notebook to train

## **GitHub Resources Include:**

- NLP Tools and Code Examples
- Sample Datasets and Jupyter Notebooks

 Participation in the third workshop on July 27 provides further information.

# Instructions for Challenge #2

## Deliverables for Challenge #2

- Build and evaluate a transformer-based system for detecting antisemitic content using our pre-annotated gold standard datasets.
- Fine-tune a transformer model using the provided annotated datasets.

## Code Submission

- Clearly document the model used and include a summary of your training setup (train/test split, random seed, training strategy).
- Upload or link your training script(s), configuration files, and any preprocessing pipeline.

Earn up to **+10 bonus points** by testing your model on **unseen data**:

- Collect a small new sample of tweets by using Bright Data' scraping services
- Manually annotate 20–30 examples using the same label scheme
- Apply your trained model to this new set
- Report performance and reflect on how well the model generalizes

# Thanks for your attention

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