

CMPT 733 Final Project

Global Data Analysis for Social Unrest Prediction

Team ExaMine:

Raksha Harish, Tavleen Sahota, Harman Jit Singh

PROJECT DEVELOPMENT

PROBLEM STATEMENT

- 1. How can we predict social unrest events (protests, riots, etc.) across the globe based on factors like historical event patterns, poverty, hunger, crime rate, natural calamities, etc.?
- 2. What insights can be derived from the study of these factors?

APPROACHES

2

- 1. Use Temporal Burst Patterns in Historical Events data to predict impactful unrest events
- 2. Use Additional Data Sources that capture social, political and economic contexts such as-poverty, hunger(food prices), crime rates, etc. to predict social unrest.

CHALLENGES

- 1. Limited Domain Knowledge due to insufficient material for literature review
- 2. Data quota constraints when handling huge amounts of data
- 3. Integration of different datasets with inconsistent timelines and missing data for certain time periods

ACTION PLAN

- 1. Data Collection
- 2.EDA
- 3. Visualizing data patterns
- 4. Integration of data sources
- 5. Feature Selection and Engineering
- 6. Building Models for prediction
- 7. Build an interactive UI to access real-time social unrest predictions

FINAL PRODUCT

- fa So
- 1. An Interactive Dashboard using Dash by plotly facilitating Data Exploration, Social Unrest Prediction and Insights about possible unrest events based on twitter hashtags.
 - 2. Insights from Analysis

DATA SCIENCE PIPELINE

1 - Data Collection



Datasets used:

- GDELT Global Events dataset (using Google BigQuery)
- ACLED (Armed Conflict Data)
- World Food Program and UN Global Criminal Rates datasets
- LatestTwitter Data

2 - Data Exploration



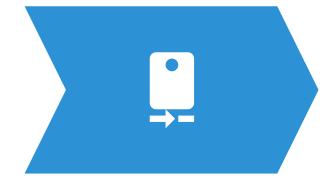
- Descriptive Statistics Distribution
 Visualization, Outlier
 Detection, Correlation
 Analysis, etc.
- Temporal Analysis
- Event sequence/pattern analysis

3 - Data Manipulation



- Data Filtering reduced 2.5TB to 1GB
- Data Aggregation
- ETL One hot encoding, categorization into classes, splitting
- Feature selection and engineering
- Feature importance analysis for modeling and prediction

4 - Modelling



- Social Unrest Scale
 Prediction (Goldstein Scale):
 - 1. Linear Reg + Lasso
 - 2. Random ForestRegressor Model3. H2O ai flow +Gradient Boosting
 - Machine
 4. H2O ai flow +
 - Isolation Forest Model
- Even sequence prediction:
 - Simulation using
 Markov chain and

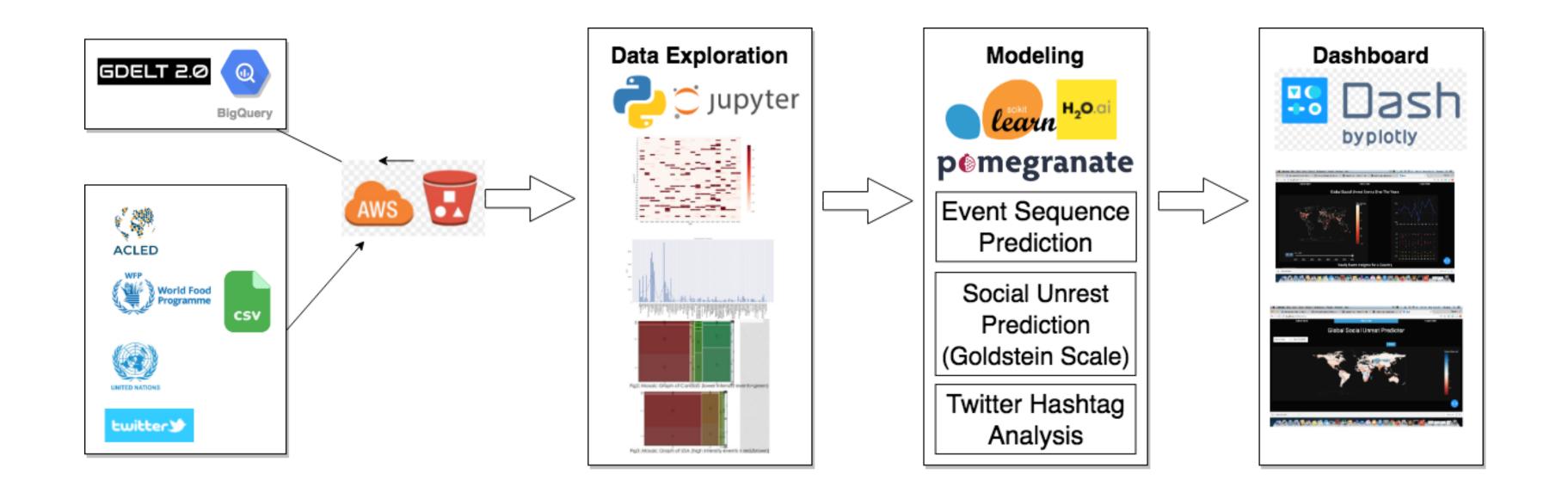
 Prediction using
 Hidden Markov Model

5 - Data Product



- Interactive Dashboard allows for Data Exploration and Social Unrest Prediction.
- Dynamic analysis of twitter hashtag streams indicating social unrest
- Insights from EDA

DATA AND TECHNOLOGIES



METHODOLOGY

- 1.EDA + Feature Engineering Exploring GDELT Dataset using Google BigQuery, and integrating with other datasets based on specific timeline to form an original bespoke dataset.
- 2. Modelling Using the bespoke dataset to predict social unrest by creating models that predict the GoldsteinScale (range of -10 to +10 -> no impact to violent event scale). This is used to simulate the sequence of events observed for a specific timeline that could possibly lead to future social unrest, using Hidden Markov Model.

Prediction of Goldstein Scale

Event Sequence Prediction

- 1. Model-1: Linear Regression with Lasso Accuracy of 59.01%
- 2. Model-2: Random Forest Regressor Accuracy of 78.17%
- 3. Model-3: H2O ai framework with Gradient Boosting Machine Accuracy of 86.31%
- 4. Model-4: H2O ai framework with Isolation Forest Model Accuracy of 64.03%

<u>CONCLUSION</u>: Best models to predict Goldstein Scale for Social Unrest are Model-2 and Model-3

- 1. Simulating the sequence of events for a specific timeline based on the 5 types of events that have occurred previously Markov Chain (MC) Simulation.
- 2. Using Hidden Markov Model (HMM) to predict the event sequence transition probability, firstly for MC simulated sequence, and then testing the model for a Random Event Sequence.

METHODOLOGY (Contd)

- 3. Twitter feed analysis to detect social unrest in real time Using tweepy and nltk packages with Twitter API for sentiment analysis to detect keywords (eg: protest, fight, violence, etc.) in tweets for trending twitter hashtags that indicate social unrest
- 4. Creating Dashboard The final product created is a dynamic, user-interactive dashboard with functionalities to query and visualize data, and make predictions

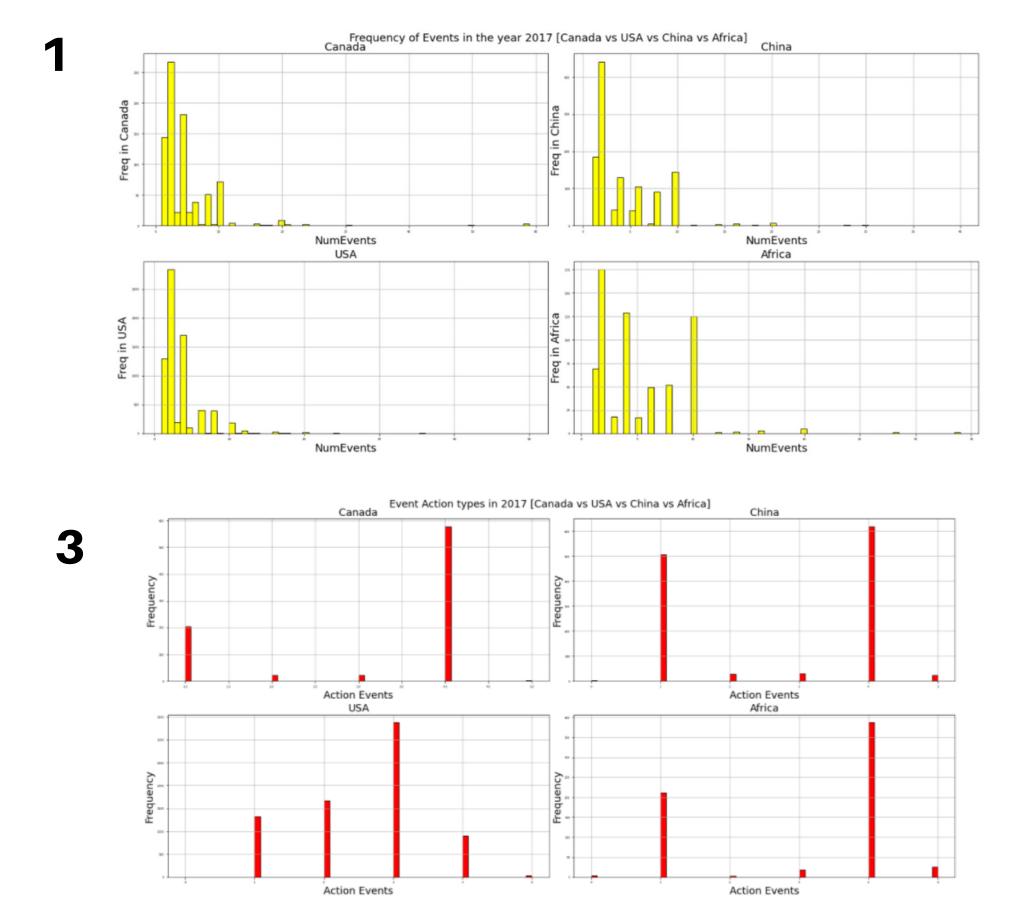
Twitter feed analysis to detect social unrest in real time

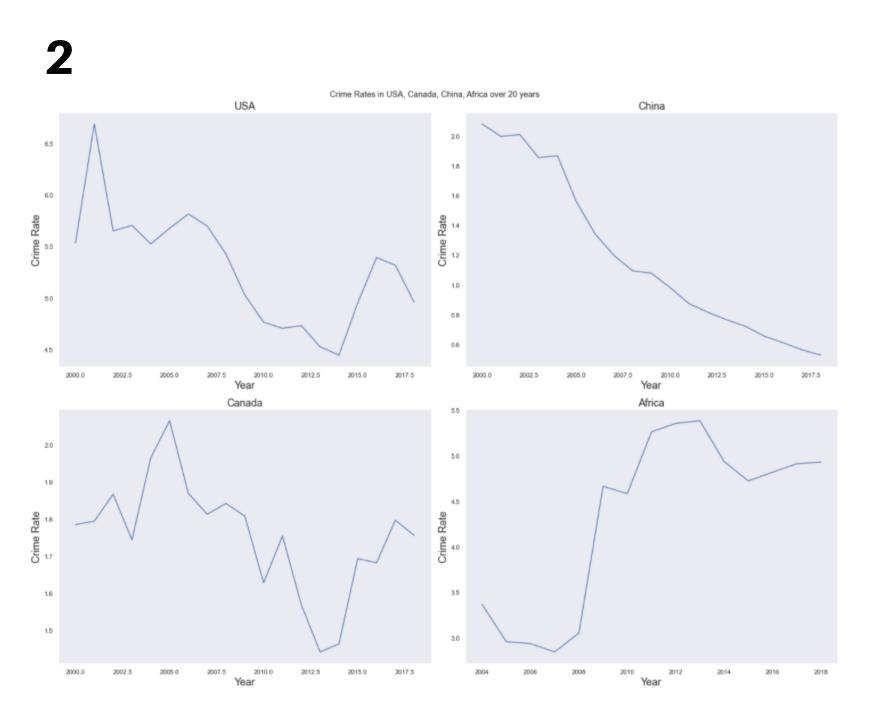
- 1. For the countries with Goldstein Scale values above threshold, top 50 trending hashtags are identified and plotted based on the tweet volume.
- 2. Hashtags and tweets analysis:
 - For each hashtag, Tweets are searched, cleaned and analysed for unrest related keywords.
 - Hashtag with most unrest related keywords identified.
 - Related hashtags found for identified hashtag and plotted.

Creation of Dynamic, User-Interactive Dashboard

- 1. Used Dash by plotly to make interactive dashboard allowing users to visualize key insights from historical data, predict future unrest and investigate Twitter feed related to unrest.
- 2. Functionalities include:
 - Time-Series Visualization (global & country-specific)
 - Comparative Analysis of key features
 - Real-time Prediction of Likely Unrest Events and Goldstein
 Scales across the world for a specific date
 - Visualization of latest twitter hashtags associated with unrest

RESULTS

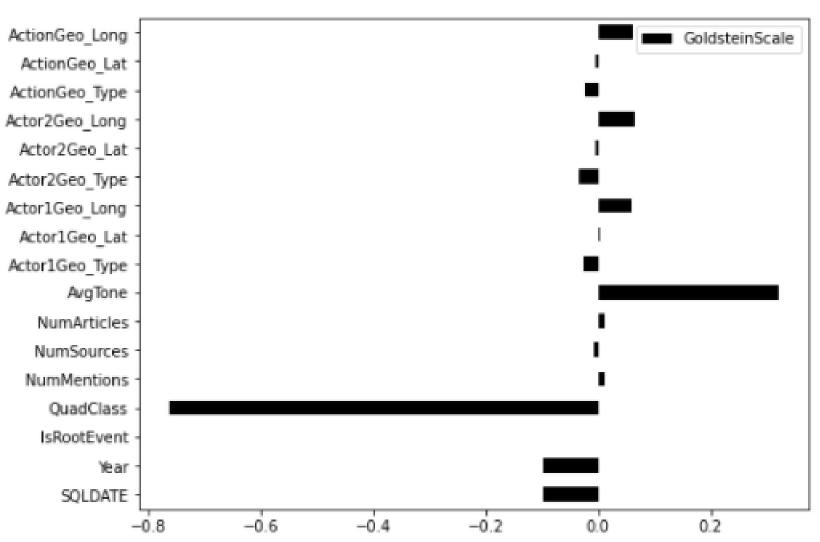




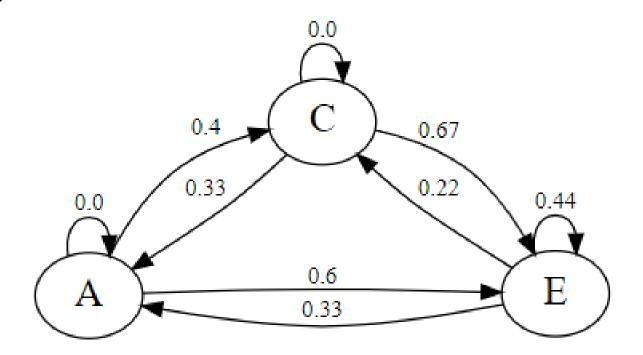
Figures: Initial EDA for Frequency of regular events (1), Crime Rates (2), and Frequency of occurrence of Action event types (3)

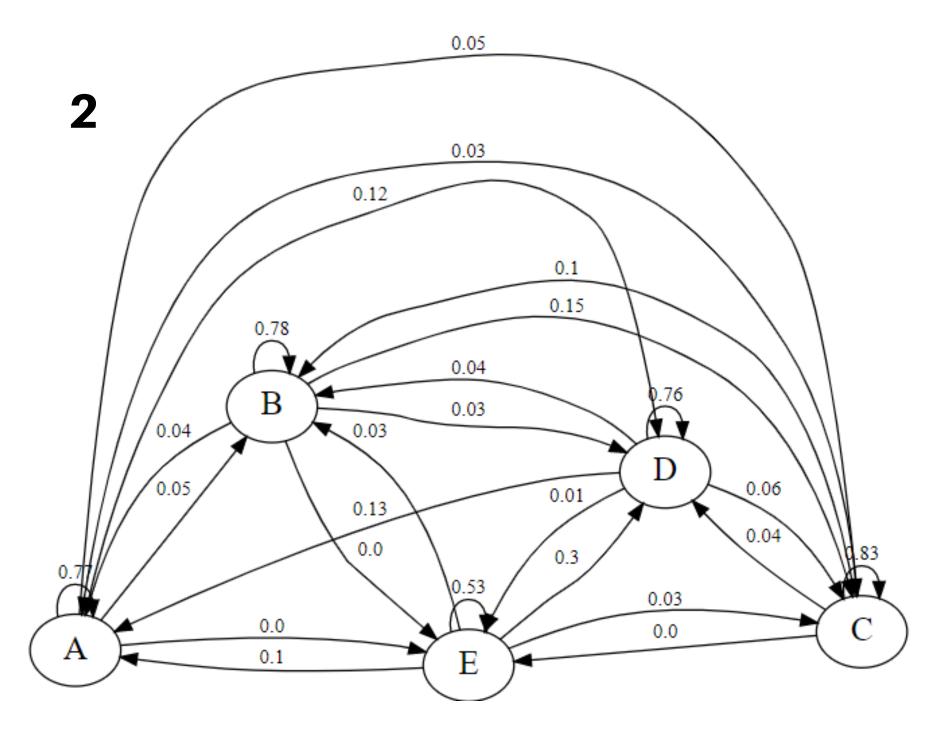
RESULTS (Contd)

1



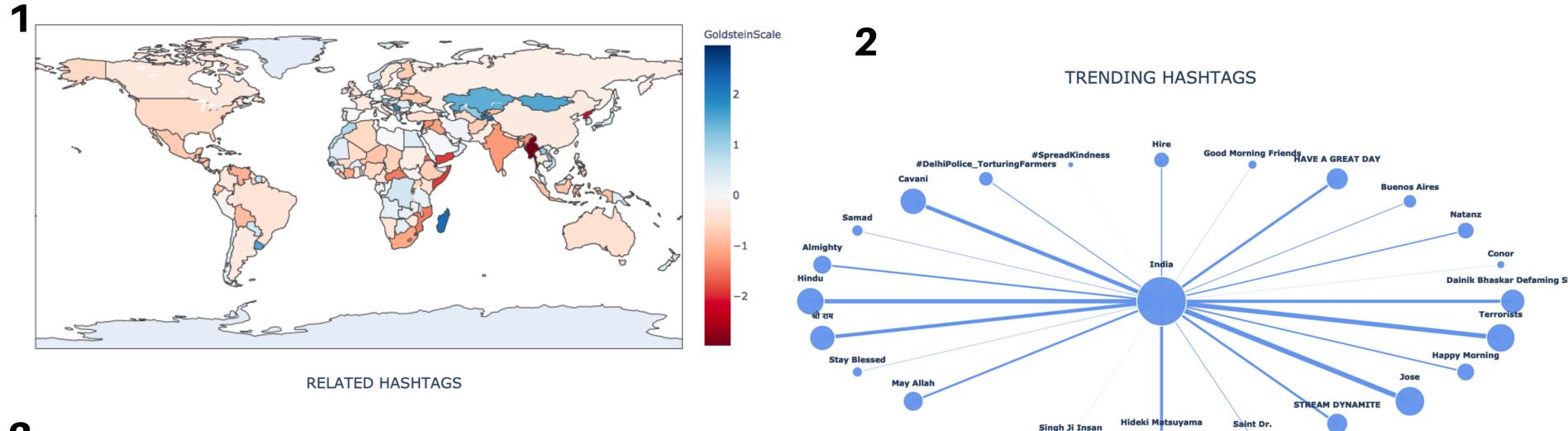
3 Graph for Random Sequence

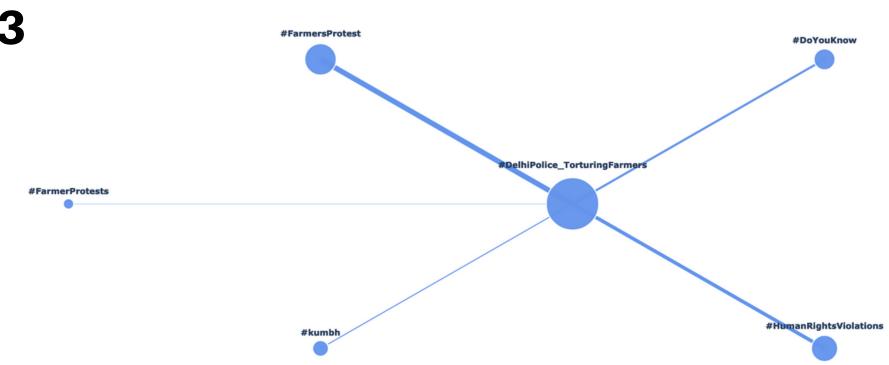




Figures: Pearson Correlation for most relevant features (1), Initial MC simulation for event sequence (2), HMM prediction for random input event sequence (3)

RESULTS (Contd)



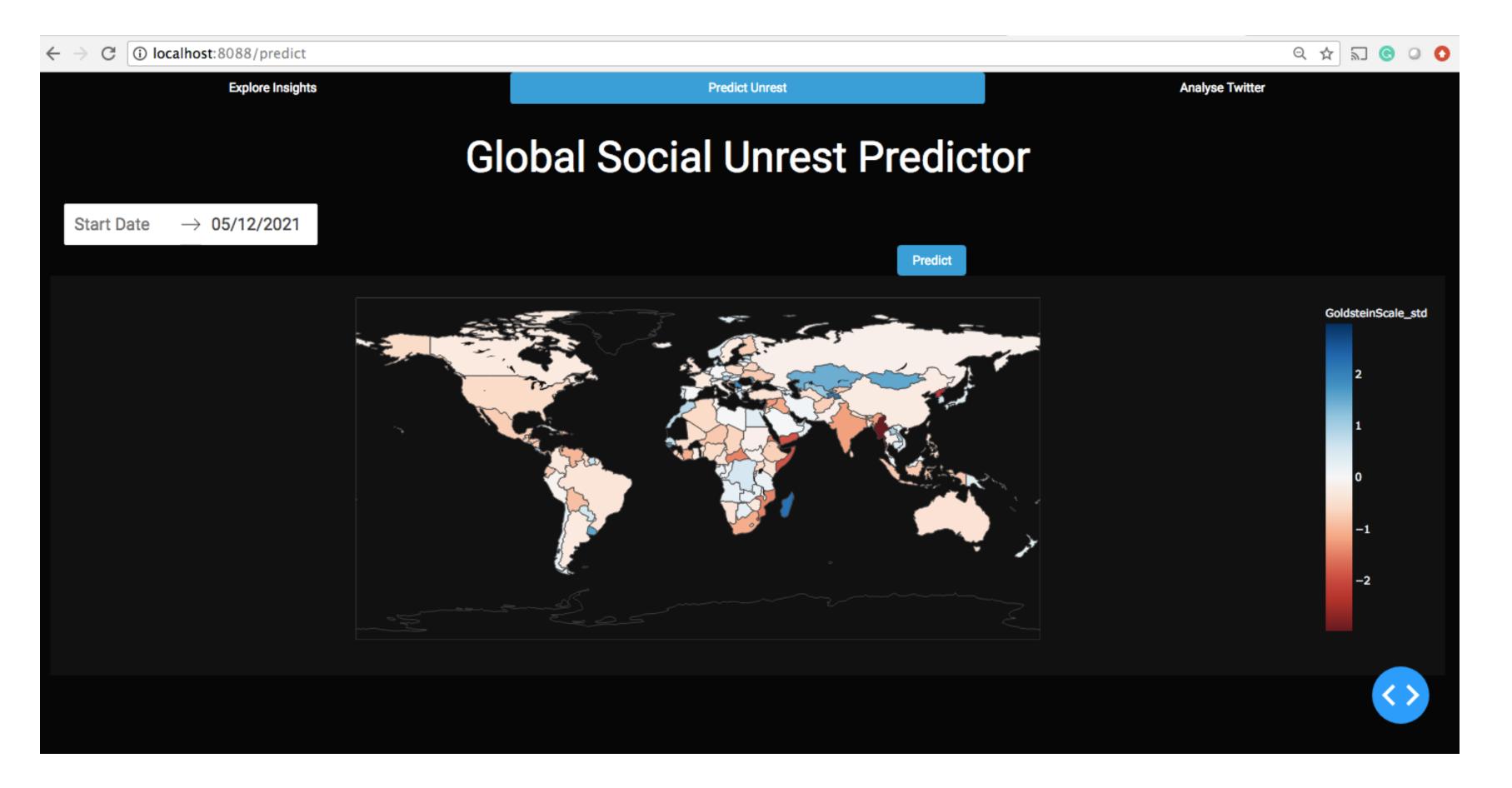


Figures: Social Unrest around the world (Mar-Apr, 2021) (1), Trending Hashtags for India (2), Related hashtags to identified unrest related hashtag (3)

DASHBOARD UI-1



DASHBOARD UI-2



THANK YOU