



# BiPol

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**Classifying Political Bias from Tweets**

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# Team



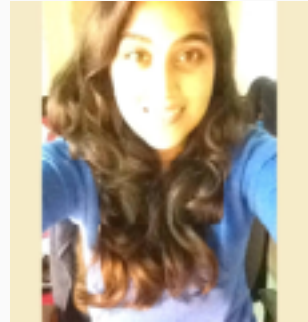
Saloni



Shivangi



Akbar



Anavi

# Motivation for Study

- Number of major celebrities using twitter
- These have large follower base
- Most used tool by politicians for political communication
- Important for voters to know - what is their behavior like?



# BiPol: Tool for Use by News Media

- Most news articles quote politician's Twitter handles
- Good tool for journalists/ researchers who want to study polarization on social media
- Pun on bye-polls ☺
- Goal: Detect Political Bias in Tweets



External affairs minister Sushma Swaraj on Sunday said Indians abroad facing problems can tweet to the respective embassy and tag her in the tweet, adding that she will personally monitor the missions' responses.

The minister, who has been known for reaching out to citizens through Twitter, said a 'new pattern' is being adopted.

"We are adopting a new pattern here. Pl (please) tweet your problem to the concerned Indian Embassy/authority and endorse the same to @sushmaswaraj.



## Detecting Political Bias: Major Challenges



Is this bias?

# Major Challenges

How to  
remove links?



Handling  
non-text  
media?

# Major Challenges

How to  
understand his  
'Mann ki baat'?





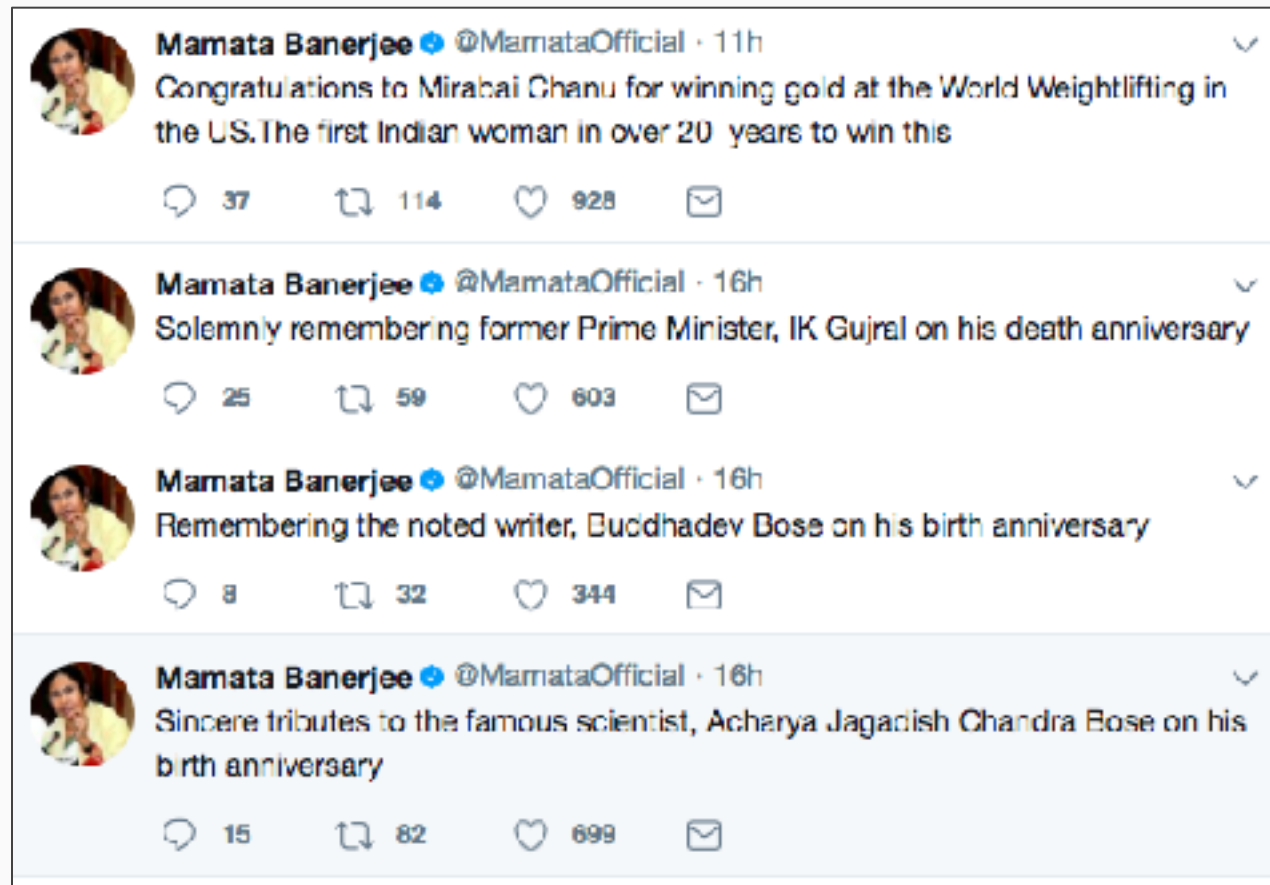
## Findings – Politician's Behavior on Twitter



Sushma Swaraj replies actively to queries

## Findings - Politician's Behavior on Twitter

Mamata Banerjee typically tweets about anniversaries & birthday greetings



The image shows a screenshot of four tweets from Mamata Banerjee (@MamataOfficial). Each tweet includes her profile picture, name, handle, and time posted. The tweets are as follows:

- Tweet 1:** Posted 11h. Text: "Congratulations to Mirabai Chanu for winning gold at the World Weightlifting in the US. The first Indian woman in over 20 years to win this". Engagement: 37 replies, 114 retweets, 928 likes.
- Tweet 2:** Posted 16h. Text: "Solemnly remembering former Prime Minister, IK Gujral on his death anniversary". Engagement: 25 replies, 59 retweets, 603 likes.
- Tweet 3:** Posted 16h. Text: "Remembering the noted writer, Buddhadev Bose on his birth anniversary". Engagement: 8 replies, 32 retweets, 344 likes.
- Tweet 4:** Posted 16h. Text: "Sincere tributes to the famous scientist, Acharya Jagadish Chandra Bose on his birth anniversary". Engagement: 15 replies, 82 retweets, 699 likes.

# Implementation

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# Data Cleaning

## Data Sourcing



## Data Cleaning & Collating



# Attributes

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## Attributes Used

Attribute Name	Description	Example
Politician Handles (y/n)	Presence of the handle of a politician <a href="https://surendran.info/twitter-handles-politicians-india/">https://surendran.info/twitter-handles-politicians-india/</a>	Eg. “wonderful step by shri @narendramodi #DeMo” Presence → Bias
Controlled Vocab (y/n)	Presence of list of words (stop list) that could be indicators of political bias	Eg. “corruption”, “black money”, “communal”, “RSS” Presence → Bias
Party Mentions (y/n)	Presence of political party	Eg. ‘BJP’, ‘NDA’ Presence → Bias
Policy Mentions (y/n)	Presence of government policies	Eg. “Swachh Bharat”, “Jan Dhan” Presence → Bias

# Attributes Used

Attribute Name	Description	Example
PM Mention (y/n)	Presence of variations of 'PM'	Eg. "namo", "modi", "Prime Minister" Presence → Bias
Politician Names	Presence of just names (not handles) of politicians Scraped data of Lok Sabha and Rajya Sabha	Eg. "Rahul Gandhi" Presence → Bias
Greetings (y/n)	Presence of any kind of greeting (congratulatory, death, condolence)	Presence → Low (or no) bias
News Bias (y/n)	Presence of news handles, interpolated with their bias (0,1)	Eg. "@IndiaToday", "@TimesNow" Presence → Bias
Polarity Score (-1,1)	> 0 → Positive    > 0 → Negative 0 → Neutral	Positive / Negative → Bias
Subjectivity Score	0 → Objective    5 → 1 → Subjective	Objective → Bias

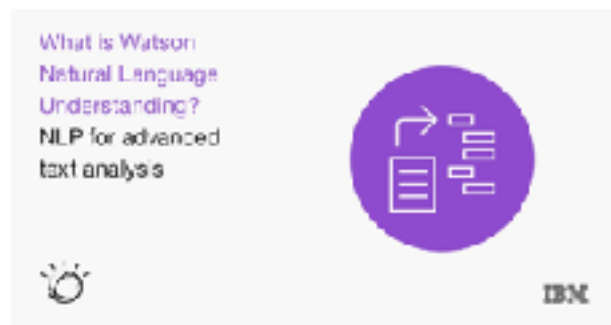
# Data Scraping Sources– For Attribute Creation

1. List of twitter handles for politicians
2. List of all politician names
3. List of all policy names (bigrams & trigrams used)
4. Names of all Political Parties, Election Commission Website.





# Tools/APIs



langdetect

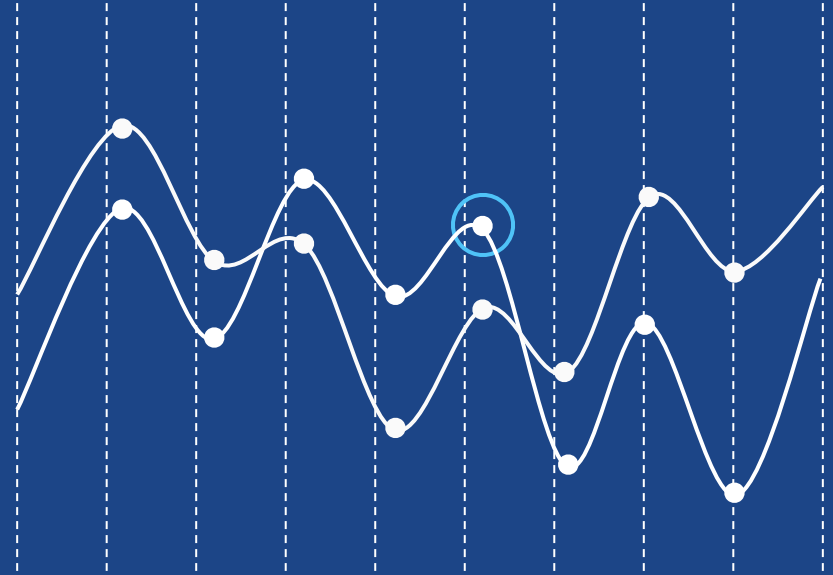
build passing

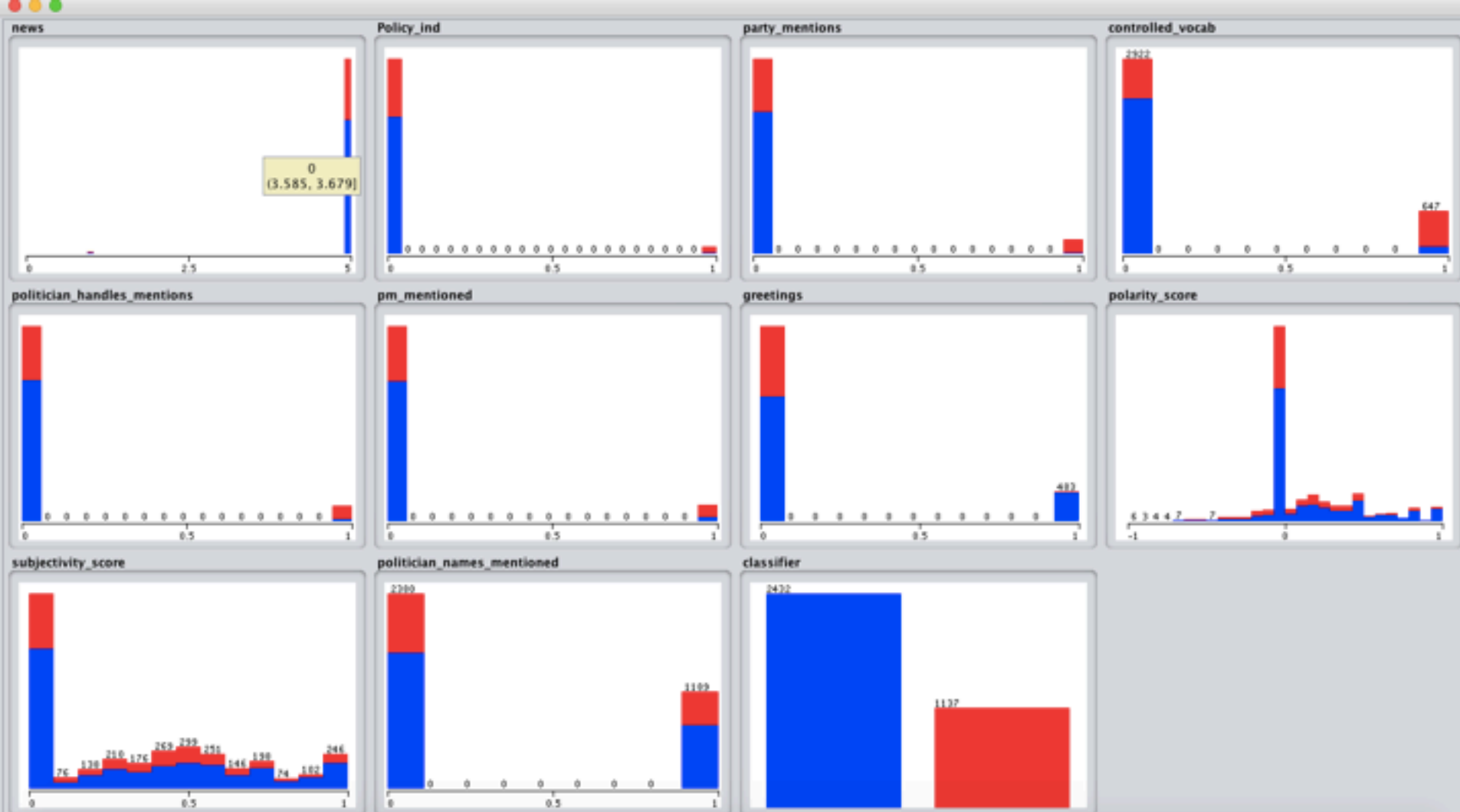
Part of Google's [language detection](#) library (version from 03/03/2014) to Python.

# Dataset

[illegible]

# Results & Conclusion





Preprocess Classify Cluster Associate Select attributes Visualize

**Classifier**

Choose NaiveBayes

**Test options**

☐ Use training set  
☐ Supplied test set  
☒ Cross-validation Folds 10  
☐ Percentage split % 65  
 More options...

(Wom) classifier

Start Stop

**Result list (right-click for options)**

06:16:31 - bayes.NaiveBayes

**Classifier output**

weight sum 2432 1137  
precision 1 1

greetings  
mean 0.1391 0.0202  
std. dev. 0.3916 0.1667  
weight sum 2432 1137  
precision 1 1

Time taken to build model: 0.04 seconds

=== Stratified cross-validation ===  
=== Summary ===

Correctly Classified Instances 3103 89.745 %  
Incorrectly Classified Instances 360 10.255 %  
Kappa statistic 0.7681  
Mean absolute error 0.1441  
Root mean squared error 0.3817  
Relative absolute error 37.7683 %  
Root relative squared error 64.7441 %  
Total Number of Instances 3169

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.938	0.190	0.914	0.938	0.926	0.761	0.895	0.931	no
	0.010	0.062	0.000	0.010	0.034	0.761	0.005	0.047	yes
Weighted Avg.	0.897	0.149	0.896	0.897	0.897	0.761	0.895	0.904	

=== Confusion Matrix ===

a \ b	classified as	
	a = no	b = yes
2182 250   a = no	2182	250
216 521   b = yes	216	521



## BiPol

Tweet:

@nitickumar hihr govtrife with lies and corruption

SUBMIT

Attributes	Values
Policy mentioned	0
Party mentioned	0
Controlled vocabulary	1
Political handles count	1
PM mentioned	0
Hashtag present	0
Greeting	0
Bias	The text is politically biased

Characters remaining: 227