Use machine learning to predict Twitter user's education level based on their tweets



Sarbottam Piya, PhD
Sarbottam.piya@gmail.com

Github.com/spiya

### Introduction



Education is the most important indicators of life outcomes



Education impacts an individual's perception about social and political changes



Can the opinion of an individual be used as a predictor to identify their education level?







Social media platforms such as Facebook and Twitter are a source of large amounts of publicly available user-generated data



Understanding the demographics of social media users is beneficial for targeted public messaging campaigns Develop a machine learning model that predicts the education level of Twitter users based on the opinions they expressed in Twitter









- Business/companies
  - Targeted advertisement
- Policymakers
  - Understand opinion of people

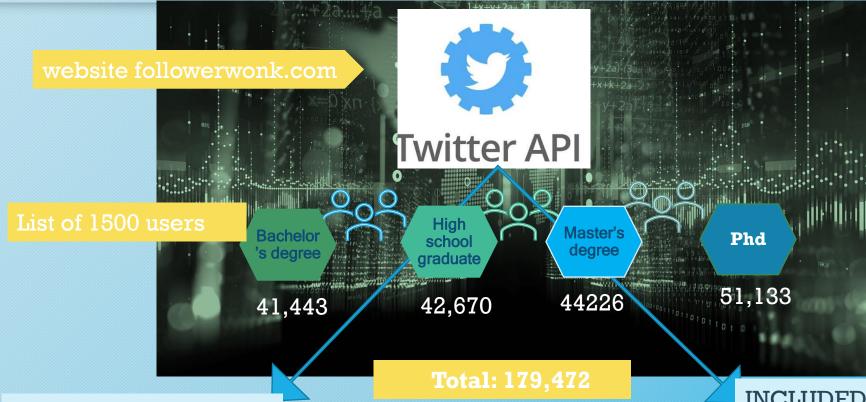






#### Dataset





#### **EXCLUDED**:

Users that have less than 10 tweets in English

#### INCLUDED:

50 tweets excluding re-tweets for each user Other tweet and twitter user features

# Data wrangling



**Total** 40,387 tweets





#### **Deleted** tweets

- <5 words
- Non-English
- **Duplicates**



high school graduate

degree

Master's degree

₹9,716



10,294

Phd

4075 Twitter users (≤10 tweets per user)

### Text normalization



- Delete the words that were used for tagging other users
- Extend the shortened words
- Characters were converted to lowercase
- Numerical characters, punctuation marks, single characters, whitespace, accented characters were deleted
- Emojis were replaced with words
- Weblinks were replaced with their domain name only.
- Delete the stop words
- Text lemmatization





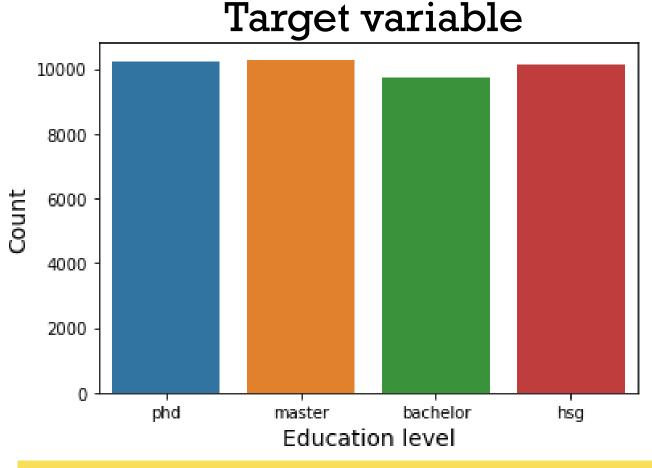
Delete the tweets that contain words 'phd', 'master', 'bachelor', 'high school', or 'degree'

Delete the entire tweets if the tweet contains word that has higher than 0.5 similarity with the above mentioned five words.



# Exploratory data analysis

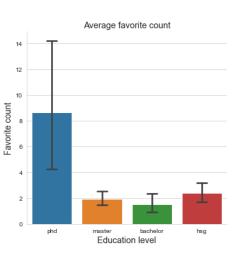


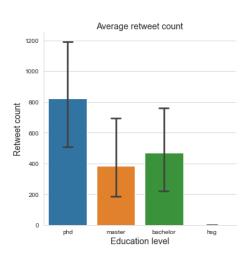


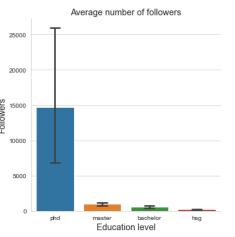
Total number of tweets for each class in the dataset

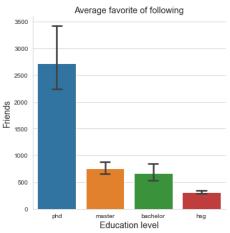
#### Twitter users with PhD are more active on Twitter

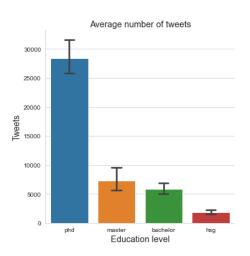






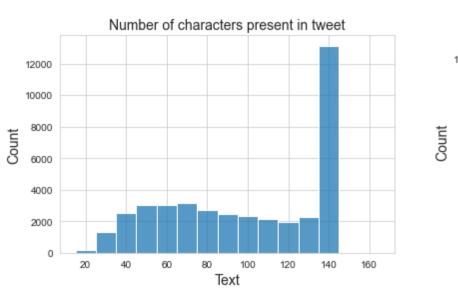


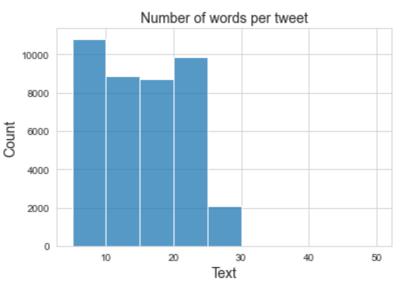


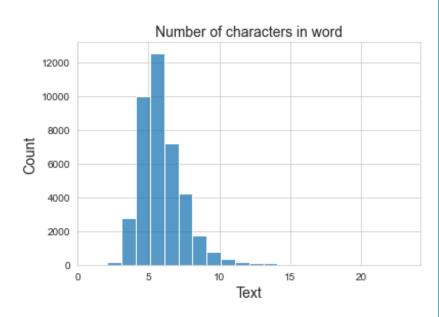


### Tweet features





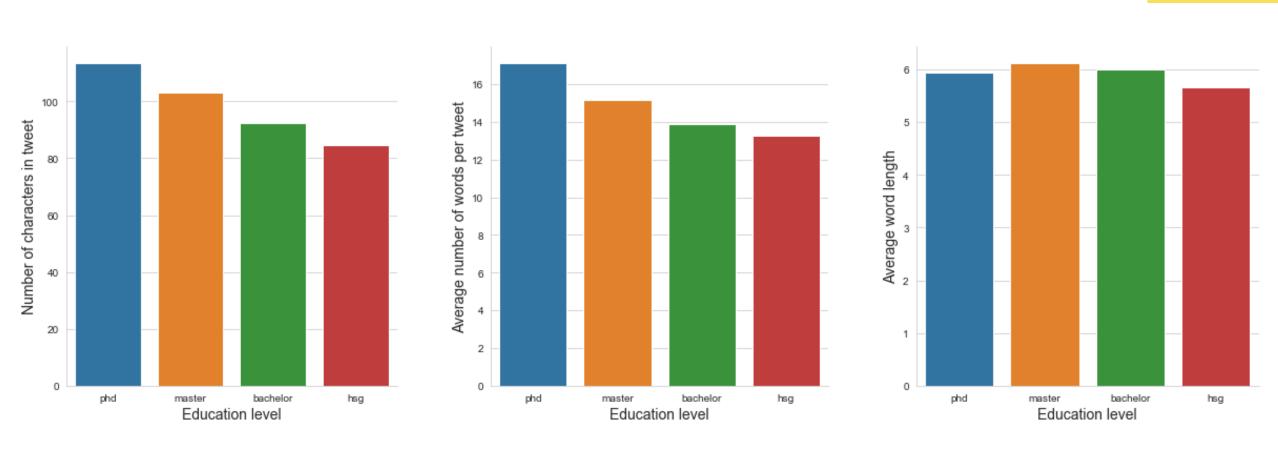




Most of the tweets were 140 characters in length with 5 to 25 words and average word length between 4 to 7 characters

### Tweet features based on education level

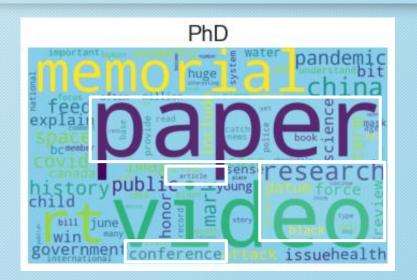


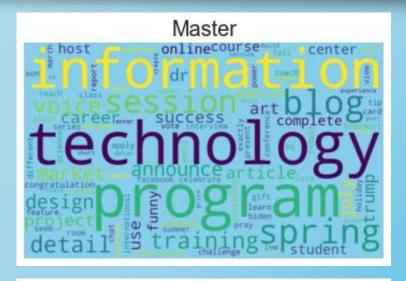


Average number of characters in a tweet and average number of words per tweet significantly increased as the education level increased

# Most predictive words

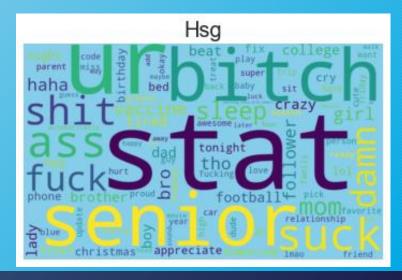














### **Text Vectorization**



CountVectorizer

TfidfVectorizer







Evaluated Naive Bayes, Logistic regression, Random Forest model and Linear Support Vector classification

Hyperparameter tuning was done with grid search with 5-fold cross-validation

Accuracy was used as the indicator of evaluation metrics.





Model	Accuracy		
	CountVectorizer	TfidfVectorizer	
Naïve Bayes	0.49	0.48	
Logistic regression	0.49	0.49	
Random Forest	0.46	0.47	
Linear SVM	0.49	0.49	

# Accuracy of models with text and tweets and user information as predictors



	Accuracy		
Model	CountVectorizer	TfidfVectorizer	
Naïve Bayes	0.5	0.5	
Logistic regression	0.54	0.56	
Random Forest	0.67	0.67	
Linear SVM	0.51	0.54	





#### Random Forest model (CountVectorizer)

	precision	recall	fl-score	support
HSG	0.58	0.86	0.69	2546
Bachelor	0.65	0.46	0.54	2383
Master	0.66	0.38	0.48	2405
PhD	0.8	0.97	0.88	2548
accuracy			0.67	9882
macro avg	0.67	0.67	0.65	9882
weighted avg	0.67	0.67	0.65	9882

## Targeting specific class for business cases



- Sometimes it may be of interest for some businesses to predict the education level of a specific class
- For example, an university want to advertise about their graduate program to twitter users with bachelor degree
- Multiclass target variable could be converted to binary variable
- Bachelor vs others

## Selection of the best model



Model	Accuracy		
	CountVectorizer	TfidfVectorizer	
Naïve Bayes	0.65	0.67	
Logistic regression	0.78	0.78	
Random Forest	0.80	0.79	
Linear SVM	0.78	0.78	



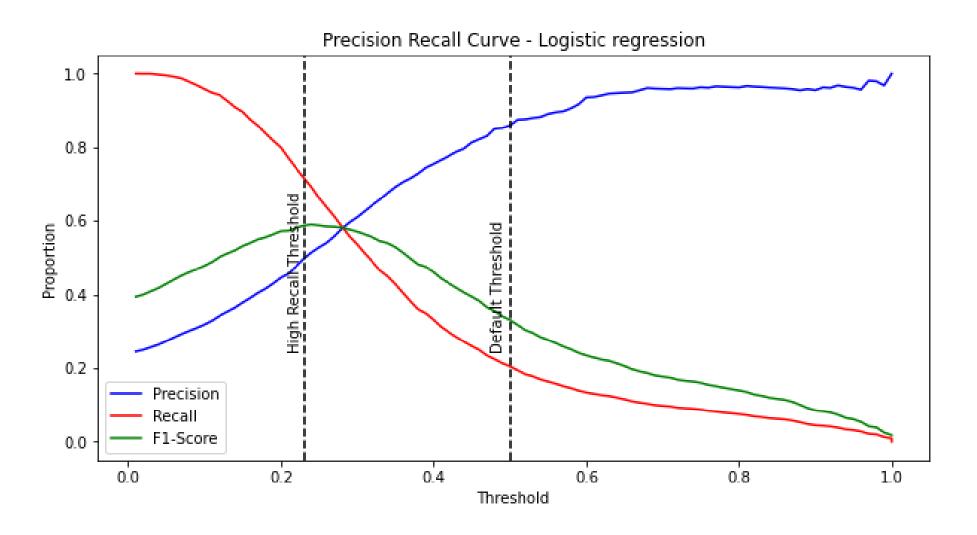
## Classification report of the best performing model

#### Random Forest model (CountVectorizer), Threshold = 0.231

	precision	recall	fl-score	support
Others	0.89	0.79	0.84	7499
bachelor	0.51	0.69	0.59	2383
accuracy			0.77	9882
macro avg	0.70	0.74	0.71	9882
weighted avg	0.80	0.77	0.78	9882

# Plot demonstrating F1 score, precision and recall at different thresholds





#### Conclusions

- Model performs better when information about the Twitter user and tweet were included as predictors along with text data
- Random Forest model performed the best with 67% accuracy
- Converting multiclass target variable to binary variable could improve the performance of the model to predict the specific class
- Excluding users that are posting several tweets advertising some products and additions of tweets from more users in the dataset could further improve the performance of the model



# Thank You

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