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# Kaggle Competetion Toxic Comment Classification Challenge



• <a href="https://www.kaggle.com/c/jigsaw-toxic-comment-classification-challenge/overview">https://www.kaggle.com/c/jigsaw-toxic-comment-classification-challenge/overview</a>

# Kaggle Competetion Overview

• Discussing things you care about can be difficult. The threat of abuse and harassment online means that many people stop expressing themselves and give up on seeking different opinions. Platforms struggle to effectively facilitate conversations, leading many communities to limit or completely shut down user comments.

### Kaggle Competetion Overview

• The <u>Conversation AI</u> team, a research initiative founded by <u>Jigsaw</u> and Google (both a part of Alphabet) are working on tools to help improve online conversation. One area of focus is the study of negative online behaviors, like toxic comments (i.e. comments that are rude, disrespectful or otherwise likely to make someone leave a discussion). So far they've built a range of publicly available models served through the <u>Perspective API</u>, including toxicity. But the current models still make errors, and they don't allow users to select which types of toxicity they're interested in finding (e.g. some platforms may be fine with profanity, but not with other types of toxic content).

## Kaggle Competetion Overview

- In this competition, you're challenged to build a multi-headed model that's capable of detecting different types of of toxicity like threats, obscenity, insults, and identity-based hate better than Perspective's <u>current models</u>. You'll be using a dataset of comments from Wikipedia's talk page edits. Improvements to the current model will hopefully help online discussion become more productive and respectful.
- Disclaimer: the dataset for this competition contains text that may be considered profane, vulgar, or offensive.

Implementing Kaggle Competetion with Scikit-Learn

Importing libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

#### Implementing Kaggle Competetion with Scikit-Learn

#### Data Preprocessing

- ✓ Dividing the data into attributes and labels
- ✓ dividing the data into training and testing sets
- ✓ replace(r'^\s\*\$', np.NaN, regex=True)
- $\checkmark$  dropna(axis = 0, how ='any')
- ✓ reset\_index(drop=True)

#### Implementing Kaggle Competetion with Scikit-Learn

#### Exploratory Data Analysis

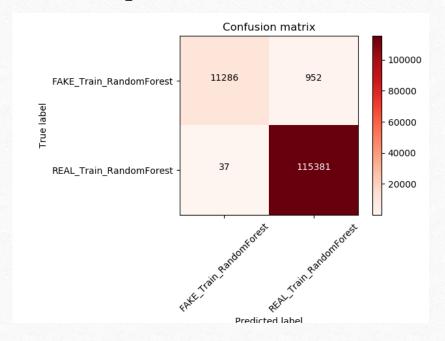
- ✓ To get a feel of how our dataset actually looks, execute the following command: bankdata.head()
- ✓ You can see that all of the attributes in the dataset are numeric. The label is also numeric i.e. 0 and 1.

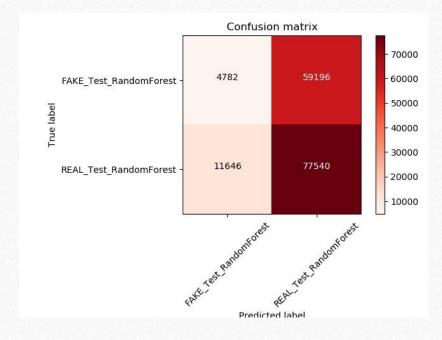
#### Implementing Kaggle Competetion with Scikit-Learn

- Evaluating the Algorithm
  - ✓ Confusion matrix, precision, recall, and F1 measures are the most commonly used metrics for classification tasks.
    - ✓ from sklearn.metrics import classification\_report, confusion\_matrix
    - ✓ print(confusion\_matrix(y\_test,y\_pred))
    - ✓ print(classification\_report(y\_test,y\_pred))

## Kaggle Competetion Random Forest

• The output of the **Random Forest** looks like this:





## Kaggle Competetion Random Forest

• The output of the **Random Forest** looks like this:

classification\_report for RandomForestClassifier\_train Model

precision recall f1-score support

-1 0.99 1.00 1.00 115418
1 1.00 0.92 0.96 12238

avg / total 0.99 0.99 0.99 127656

Accuracy\_RandomForestClassifier\_train: 0.992

Cohens kappa\_train\_RandomForestClassifier: 0.953763

[[115381

952 11286]]

 classification\_report for RandomForestClassifier\_test Model

 precision
 recall
 f1-score
 support

 -1
 0.57
 0.87
 0.69
 89186

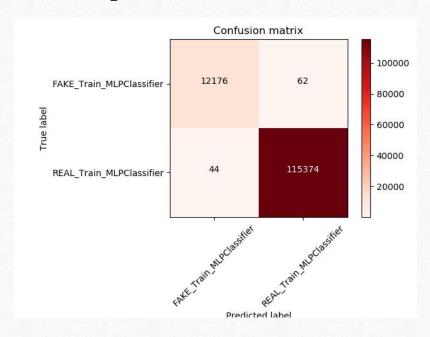
 1
 0.29
 0.07
 0.12
 63978

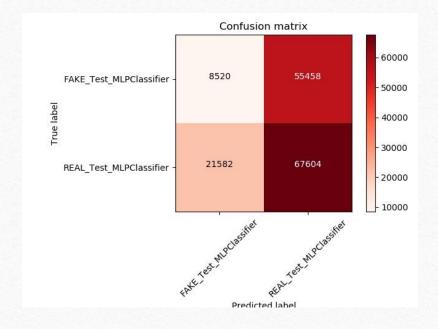
 avg / total
 0.45
 0.54
 0.45
 153164

 Accuracy\_RandomForestClassifier\_test:
 0.537

# Kaggle Competetion MLP

• The output of the **MLP** looks like this:





# Kaggle Competetion MLP

• The output of the **MLP** looks like this:

classification_report	for	MLPClassifier_	_train	Model
-----------------------	-----	----------------	--------	-------

	precision	recall	f1-score	support
-1 1	1.00 1.00	1.00 0.99	1.00 1.00	115418 12238
avg / total	1.00	1.00	1.00	127656

Accuracy\_MLPClassifier\_train: 0.999

Cohens kappa\_train\_MLPClassifier: 0.995207

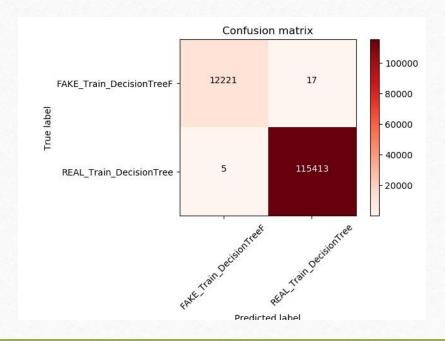
[[115374 44] [ 62 12176]] classification\_report for MLPClassifier\_test Model

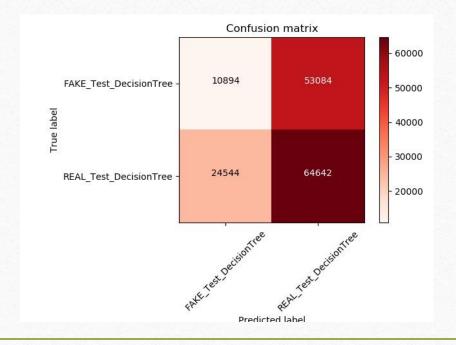
	precision	recall	f1-score	support
-1 1	0.55 0.28	0.76 0.13	0.64 0.18	89186 63978
avg / total	0.44	0.50	0.45	153164

Accuracy\_MLPClassifier\_test: 0.497

## Kaggle Competetion DecisionTree

• The output of the **DecisionTree** looks like this:





## Kaggle Competetion DecisionTree

• The output of the **DecisionTree** looks like this:

classificati	ion_report for	Decisio	nTree_train	Model
	precision	recall	f1-score	support
-1 1	1.00 1.00	1.00 1.00	1.00 1.00	115418 12238
avg / total	1.00	1.00	1.00	127656

Accuracy\_DecisionTree\_train: 1.000

classification\_report for DecisionTree\_test Model

	precision	recall	f1-score	support
-1	0.55	0.72	0.62	89186
1	0.31	0.17	0.22	63978
avg / total	0.45	0.49	0.46	153164

Accuracy\_DecisionTree\_test: 0.493

