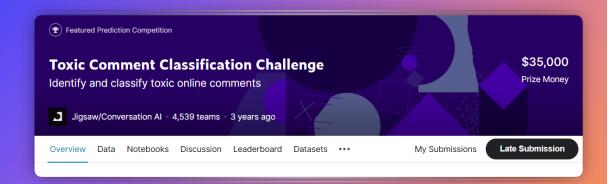
TOXIC COMMENT CLASSIFICATION

Simone Monti - 807994

Vittorio Maggio – 817034

University of Milano-Bicocca

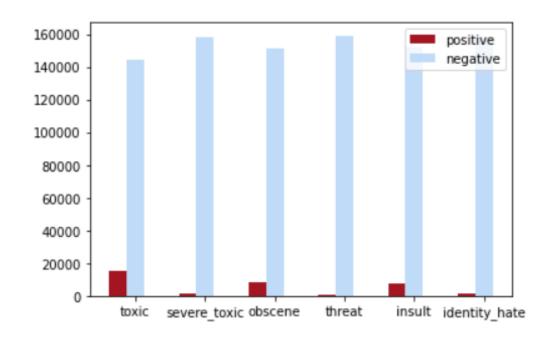


Deep learning approach

https://www.kaggle.com/c/jigsaw-toxic-comment-classification-challenge

DATASET EXPLORATION

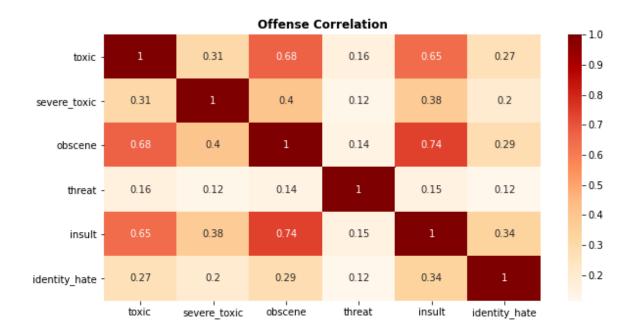




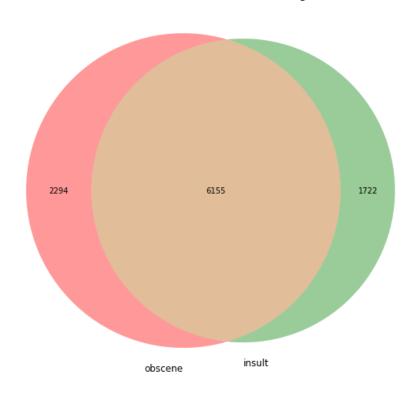
Total: 159571 comments

- Toxic
- Severe toxic
- Obscene
- Threat
- Insult
- Identity hate

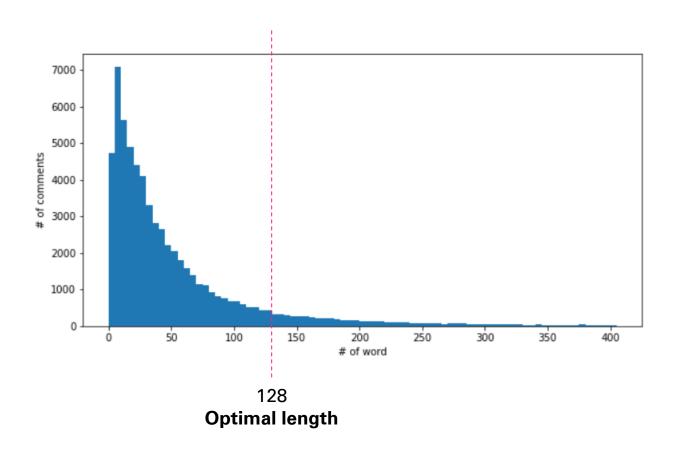
Correlation among categories



Obscene and insult comments Venn diagram



Comments length



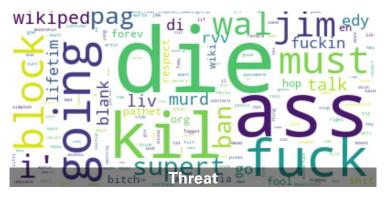
WordCloud







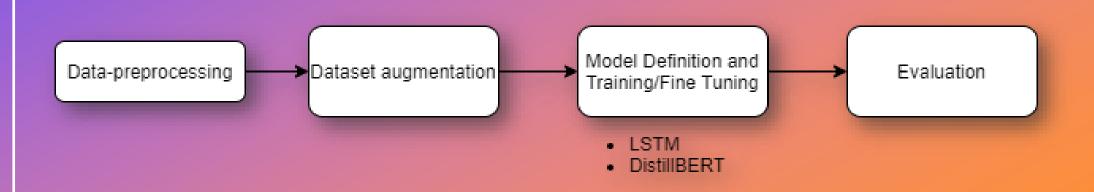






PIPELINE





DATASET PREPARATION



Pre-processing

- Lowercase
- Punctuation and spacing removal

Explanation\nWhy the edits made under my usern...

D'aww! He matches this background colour I'm s...

Hey man, I'm really not trying to edit war. It...

"\nMore\nI can't make any real suggestions on ...

You, sir, are my hero. Any chance you remember...



explanation why the edits made under my userna ...

daww he matches this background colour im seem...

hey man im really not trying to edit war its j...

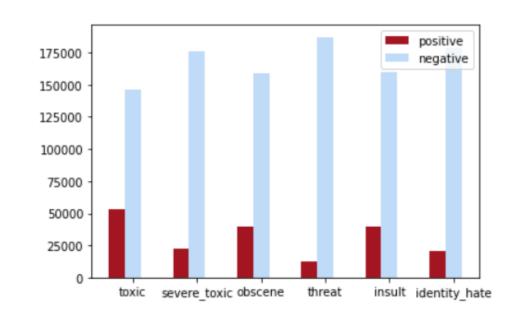
more i cant make any real suggestions on impro...

you sir are my hero any chance you remember wh...

Dataset augmentation: EDA

https://github.com/jasonwei20/eda_nlp

- Synonym Replacement (SR): Randomly choose n words from the sentence that are not stop words. Replace each of these words with one of its synonyms chosen at random.
- Random Insertion (RI): Find a random synonym of a random word in the sentence that is not a stop word. Insert that synonym into a random position in the sentence. Do this n times.
- Random Swap (RS): Randomly choose two words in the sentence and swap their positions. Do this n times.
- Random Deletion (RD): For each word in the sentence, randomly remove it with probability p.
- Used with threat(x20), severe_toxic(x10), identity_hate(x10)
- 199.131 elements

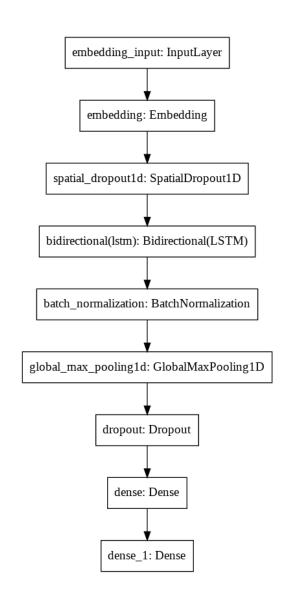


MODELS



LSTM – Architecture

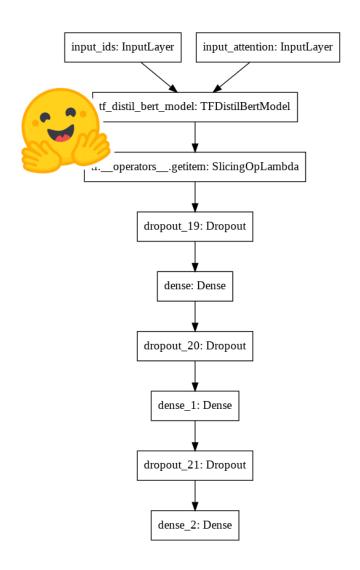
- Embedding layer (input_dim: 128, output_dim=128)
- SpatialDropout1D (rate: 0.3),
- Bi-directional LSTM layer with 42 neurons and with Tanh as activation function, and Sigmoid as recurrent activation,
- · BatchNormalization,
- GlobalMaxPool1d,
- Dropout (rate: 0.3),
- Dense layer with 18 neurons and Relu as activation function,
- Dense layer formed by 6 neurons with Sigmoid function as activation function (the output is composed by 6 different binary values)



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DistillBERT Fine-tuned – Architecture

- DistilBERT base uncased (6-layer, 768-hidden, 12-heads)
- Dropout (rate: 0.2)
- Dense layer with 256 neurons and Relu as activation function
- Dropout (rate: 0.2)
- Dense layer with 32 neurons and Relu as activation function
- Dopout (rate: 0.2)
- Dense layer formed by 6 neurons with Sigmoid function as activation function (the output is composed by 6 different binary values)



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Training

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LSTM

- HyperParametes:
 - Epochs: 50;
 - Batch size: 256;
 - Validation split: 0.2
 - Optimizer: Adam (learning rate: 0.0001);
 - Loss: Binary Cross Entropy;
- Early Stopping:
 - Monitor: validation loss;
 - Patience: 3,
 - restore_best_weights: True;

DistillBERT Fine-tuned

- HyperParametes:
 - Epochs: 6;
 - Batch size: 64;
 - Validation split: 0.2
 - Optimizer: Adam (learning rate: 0.00005);
 - Loss: Binary Cross Entropy;
- Early Stopping:
 - Monitor: validation loss;
 - Patience: 2,
 - restore_best_weights: True;

EVALUATION



Evaluated on the test dataset provided by the competition

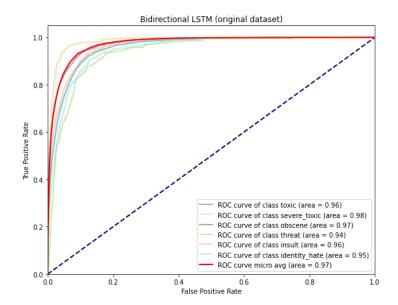
LSTM

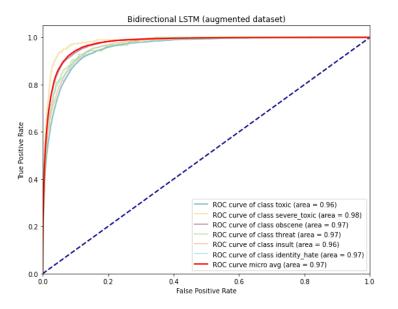
ORIGINAL DATASET

AUGMENTED DATASET

	Precision	Recall	F1-score	support
Toxic	0.61	0.75	0.67	6090
Severe toxic	0.35	0.26	0.30	367
$\overline{\mathrm{Obscene}}$	0.66	0.73	0.69	3691
${f Threat}$	0.00	0.00	0.00	211
Insult	0.62	0.63	0.63	3427
Identity_hate	0.60	0.00	0.01	712
Micro avg	0.62	0.66	0.64	14498

	Precision	Recall	F1-score	support
Toxic	0.52	0.85	0.65	6090
Severe_toxic	0.42	0.13	0.20	367
Obscene	0.62	0.77	0.69	3691
Threat	0.33	0.02	0.04	211
${\bf Insult}$	0.61	0.67	0.64	3427
Identity_hate	0.78	0.14	0.24	712
Micro avg	0.57	0.72	0.63	14498





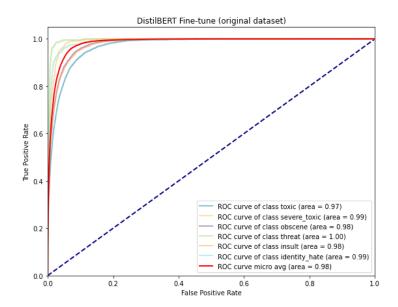
DistilBERT Fine-tuned

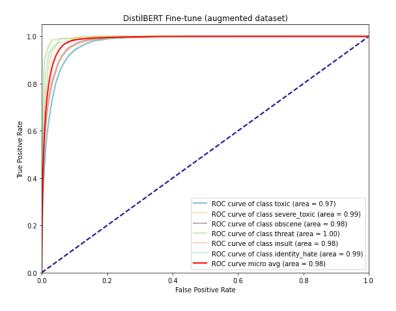
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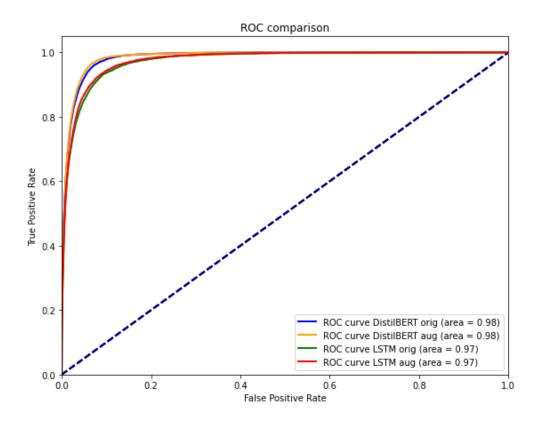
	Precision	Recall	F1-score	support
Toxic	0.50	0.92	0.65	6090
Severe toxic	0.27	0.75	0.39	367
$\overline{\mathrm{Obscene}}$	0.58	0.83	0.68	3691
Threat	0.44	0.62	0.51	211
${\bf Insult}$	0.61	0.77	0.68	3427
Identity_hate	0.56	0.61	0.58	712
Micro avg	0.53	0.84	0.65	14498

	Precision	Recall	F1-score	support
Toxic	0.50	0.94	0.65	6090
Severe toxic	0.26	0.72	0.39	367
Obscene	0.52	0.89	0.65	3691
Threat	0.46	0.57	0.51	211
Insult	0.51	0.87	0.65	3427
Identity_hate	0.51	0.97	0.58	712
Micro avg	0.50	0.89	0.64	14498





Comparison



	Dataset originale		Dataset aumentato		
	LSTM	DistilBERT Fine tune	LSTM	DistilBERT Fine tune	
AUC	0.97	0.98	0.97	0.98	
$\mathbf{F1}$	0.64	0.65	0.63	0.64	

THANK YOU FOR YOUR ATTENTION

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