YouTube spam detection Artificial Intelligence for CyberSecurity

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Dataset and goal

Total of 1956 YouTube comments from 5 different (famous) musical videos¹ with the following features:

- Comment ID
- Author
- Date
- Content
- Class

INTRODUCTION

Total of 1956 YouTube comments from 5 different (famous) musical videos¹ with the following features:

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Goal: recognize and differentiate between legitimate (ham) comments and spam comments!

¹https://archive.ics.uci.edu/dataset/380/youtube+spam+collection

Example entries

	COMMENT_ID	AUTHOR	DATE	CONTENT	CLASS
0	LZQPQhLyRh80UYxNuaDWhIGQYNQ96luCg-AYWqNPjpU		2013-11-07T06:20:48	Huh, anyway check out this you[tube] channel:	1
1	LZQPQhLyRh_C2cTtd9MvFRJedxydaVW-2sNg5Diuo4A				1
2	LZQPQhLyRh9MSZYnf8djyk0gEF9BHDPYrrK-qCczIY8				1
3	z13jhp0bxqncu512g22wvzkasxmvvzjaz04				1
4	z13fwbwp1oujthgqj04chlngpvzmtt3r3dw	GsMega		watch?v=vtaRGgvGtWQ Check this out .	1
1951	_2viQ_Qnc6-bMSjqyL1NKj57ROicCSJV5SwTrw-RFFA	Katie Mettam	2013-07-13T13:27:39.441000		0
1952	_2viQ_Qnc6-pY-1yR6K2FhmC5i48-WuNx5CumlHLDAI				0
1953	_2viQ_Qnc6_k_n_Bse9zVhJP8tJReZpo8uM2uZfnzDs		2013-07-13T12:09:31.188000		0
1954	_2viQ_Qnc6_yBt8UGMWyg3vh0PulTqcqyQtdE7d4Fl0			Shakira u are so wiredo	0
1955	_2viQ_Qnc685RPw1aSa1tfrluHXRvAQ2rPT9R06KTqA			Shakira is the best dancer	0

Class equal to one indicates spam!

Word cloud



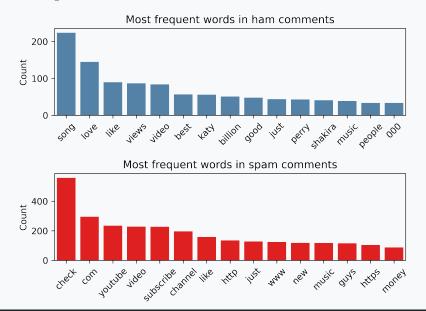


subscribe_channel

Ham word cloud

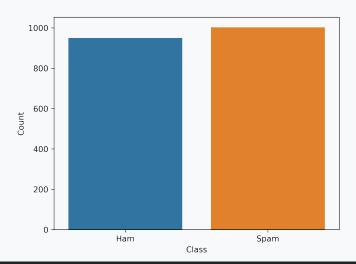
Spam word cloud

Most frequent words



Balanced dataset

Classes are (basically) balanced!



Cleaning

- Checked for null values (only a few in date)
- Removed useless features (comment ID, author and date)
- Removed duplicates (only 3 duplicates, which affected the balance positively)
- Replaced HTML tags and entities in comments (e.g. replaced
 with \n)

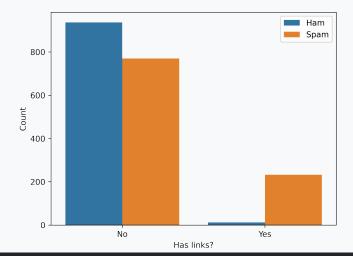
Adding useful features

Tried to extract possible features that may indicate a spamming behaviour:

- Links
- YouTube links (spamming one's channel)
- Use of non-ASCII characters (e.g. emojis)
- Number of characters, words, and sentences
- Number of uppercase letters

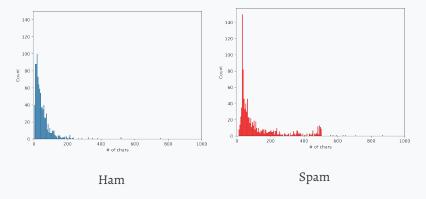
Adding useful features - Links

Presence of links may indicate spam:



Adding useful features - Characters

Spam comments have a much higher peak, a longer tail, and a second smaller peak at about 500 characters.



Adding useful features - Heatmap



Approach

Classification has been performed with 4 classifiers:

- Support Vector Machine
- Multinomial naïve Bayes
- Decision tree
- Random tree

and with 3 different preprocessings:

- Stemming with Porter stemmer
- Stemming with Snowball stemmer
- Lemmatization

Performance evaluation

Results obtain from K-fold (10 folds):

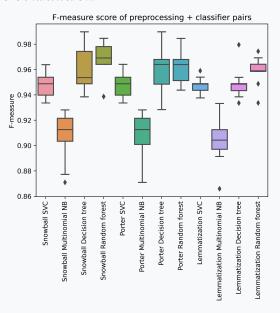
SVM		Multinomial NB	Decision tree	Random forest
Snowball	0.948	0.908	0.957	0.964
Porter	0.949	0.907	0.957	0.962
Lemmatization	0.947	0.903	0.949	0.958

F-measure

	SVM	Multinomial NB	Decision tree	Random forest
Snowball	0.948	0.908	0.958	0.964
Porter	0.949	0.907	0.96	0.964
Lemmatization	0.947	0.904	0.949	0.961

Accuracy

Performance evaluation



Testing the null hypothesis

The random forest classifier has performed the best, whereas the preprocessing has a much smaller impact on final results. Wilcoxon test can be used to determine if there is a statistical difference between preprocessing methods (with a fixed classifier).

Preprocessing pair	P-value
Snowball - Porter	0.4962
Snowball - Lemmatization	0.1934
Porter - Lemmatization	0.1055

Random forest with different preprocessing

Using the conventional acceptance of statistical significance at 0.05 (5%), we confirm the null hypothesis: the difference between the three proprocessing methods is not statistically significant!

Performance evaluation - Other results

In general, Wilcoxon test allows determining that in this dataset, for the tested classifiers and preprocessing algorithms:

- The use of a different preprocessing is usually not significant, but it is for the decision tree classifier
- The use of a different classifier is almost always significant

Conclusions and future work

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- The best classifier turned out to be the Random forest model. As the classifier has very good performances, the initial goal can be considered achieved
- The use of K-fold and of the Wilcoxon test ensure that results are statistically significant



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Improvements/future work:

- Trying other algorithms and/or preprocessing methods, which may lead to even higher performances
- The dataset is not large at all. To completely ensure that results can
 be trusted, it would be needed to use a much larger dataset (possibly
 with at least tens of thousand of comments). Unfortunately, no such
 dataset was found online