# Group final project - BAN400 fall 2020

Candidate: (...) 6/11/2020

## Contents

Project suggestion Group 30
Initial proposal:
Method:
Analysis:
Sources
Workflow:
1. get the data to run in R and tidyverse
2. Prepossesing and getting the data ready for modelling
3. Training the machine learning model
4. Test the model, and polish it (maybe go back to step 2 if need be)
5. make a shiny app, which will work as an interface to plug in new data and see if
the news stories is true or fake.

## Project suggestion Group 30

#### Initial proposal:

"Textual data analysis combined with regression analysis using data on fake/true news"

#### Method:

- We create regressors based on the words present in the fake and real news. So far we can create a sentiment factor, on both the body and the title of the article, and a keyword per word factor, also both on the title and the body.
- Then use a predictive model to see if the news are real or fake based on these factors. We can run a factor regression or perhaps a machine learning model, like an XGBoost model.
- Finally, we can create a "shiny ap" which allows up to paste news articles in, the ap will then preprocess the article and give us a score of "fake probability".

### **Analysis:**

- Tokenizing and prepossesing before doing any textual analysis, we need to preposess the data, which mean shaping it in order for the different models to read them.
- Topic modelling: Using the words in the dataset to define a concrete topic in each of the files. This could help us see which words to look out for, perhaps there is a topic which is "Hype up words", that might be more present in the fake news. We can create a "topic per word" score.
- Sentiment analysis to gather data on how negative /positive the fake news is compared with true news. Maybe there is a specific sentiment in the fake news that we can extract, we can add these numbers to our regression, to make a predictive model to see if the news are fake or not.
- Run a regression, were 1 is fake news and 0 is true news. It could be a linear regression, or we could experiment with some machine learning model. We keep whichever performs best.

#### Sources

For this we will use a data set from Kaggle: https://www.kaggle.com/clmentbisaillon/fake-and-real-news-dataset (Links to an external site.) with data from 2016 to 2017.

All graphs will be visualized using ggplot package.

- Use machine learning
- Apply it to live data

https://www.r-bloggers.com/2020/10/sentiment-analysis-in-r-with-custom-lexicon-dictionary-using-tidytext/"

 $https://rstudio-education.github.io/tidyverse-cookbook/import.html\ https://www.tidyverse.org/blog/2020/06/recipes-0-1-13/$ 

#### Workflow:

- 1. get the data to run in R and tidyverse
- 1.0 merge true and fake news datasets, adding a new dummy column if news is true or fake.
- 1.1 separate the data into training and test data.
- 2. Prepossesing and getting the data ready for modelling.
- 2.1 cleaning the words, and making them ready for the sensitivy analysis(making a frequency list)
- 2.1.1 Use recipes in order to make this cleaning process replicable for test data
- 2.2 Using the sensitivity analysis to make usable columns in a tibble for each news article.
- 2.3 gathering all the data into one tibble, which we'll use to train our machine learning model.
- 3. Training the machine learning model
- 4. Test the model, and polish it (maybe go back to step 2 if need be)
- 5. make a shiny app, which will work as an interface to plug in new data and see if the news stories is true or fake.
- 5.1 make the app more user friendly