

Fake News Detection

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References:

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2. Ankesh Anand, Tanmoy Chakraborty, Noseong Park (2016), We used Neural Networks to Detect Clickbaits: You won't believe what happened Next!
3. Martin Potthast, Johannes Kiesel, Kevin Reinartz, Janek Bevendorff, and Benno Stein (2017), A stylometric inquiry into hyperpartisan and fake news.
4. Srivastava, N., Hinton, G.E., Krizhevsky, A., Sutskever, I., Salakhutdinov, R.: Dropout: a simple way to prevent neural networks from overfitting. Journal of Machine Learning Research 15(1) (2014) 1929–1958
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TASK AND EXPERIENCE:

Task: Designing and training a neural network architecture to classify news articles as fake or real based on linguistic features.

Three types of fake news ^[1]:

- Serious Fabrication
- Hoaxes
- Satire

Features to be focused on:

- Sentiment
- Grammar
- Punctuations
- Click baits

Dataset: Dataset from previous research:

- Clickbait detection: 15,000 ^[2]
- Buzzed dataset: 1627 news articles including 299 fake news articles ^[3]
- Word List: 800 profanities

MODEL AND ARCHITECTURE:

Model: Gated Recurrent Unit

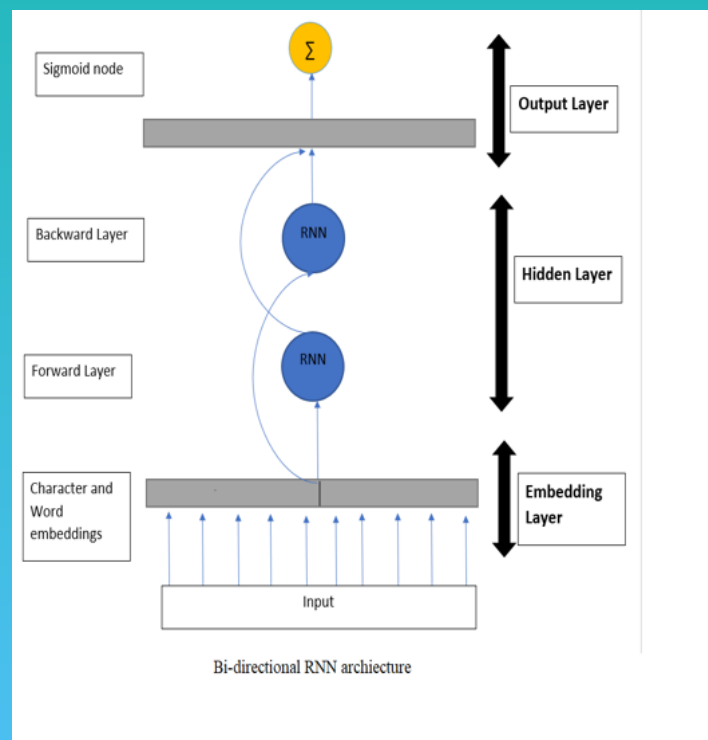
Training: Mini-batch gradient descent

Cost Function: Binary Cross Entropy Loss

Optimization: ADAM optimizer for parameter updates

Regularization to avoid

overfitting: Dropout technique ^[4]



PERFORMANCE MEASURE:

10-fold cross validation using:

- Accuracy
- Precision
- Recall
- F1 measure

Comparison baseline (previous results):

- Bag of words approach: 0.52 ^[3]
- Stylometric approach: 0.55 ^[3]
- NLP approach using ngrams, punctuations, syntax, readability etc: 75% ^[5]

Expected Result/ Aim:

- Accuracy $\geq 75\%$