

Fake News Data: ()

Assumptions:

1. Removed the rows that didn't have a label associated with it.
2. Used only the text to develop the classifier

Resources: MATLAB 2017b (Neural Network Toolbox and Parallel Computing Toolbox), NVIDIA K-40 GPU (for training)

Initial set up

```
clear

% Set the parameters below as true if you want to train the model.
doTrain = true;
doTest = true;
```

Loading the Dataset

```
% filename = "fake_or_real_news.xlsx";
% data = readtable(filename, 'TextType', 'string');
% save('fakeNewsData.mat')
load('fakeNewsData.mat');
```

Consider the rows where the dataset is labeled as FAKE or REAL.

Remove the rows where the text is blank

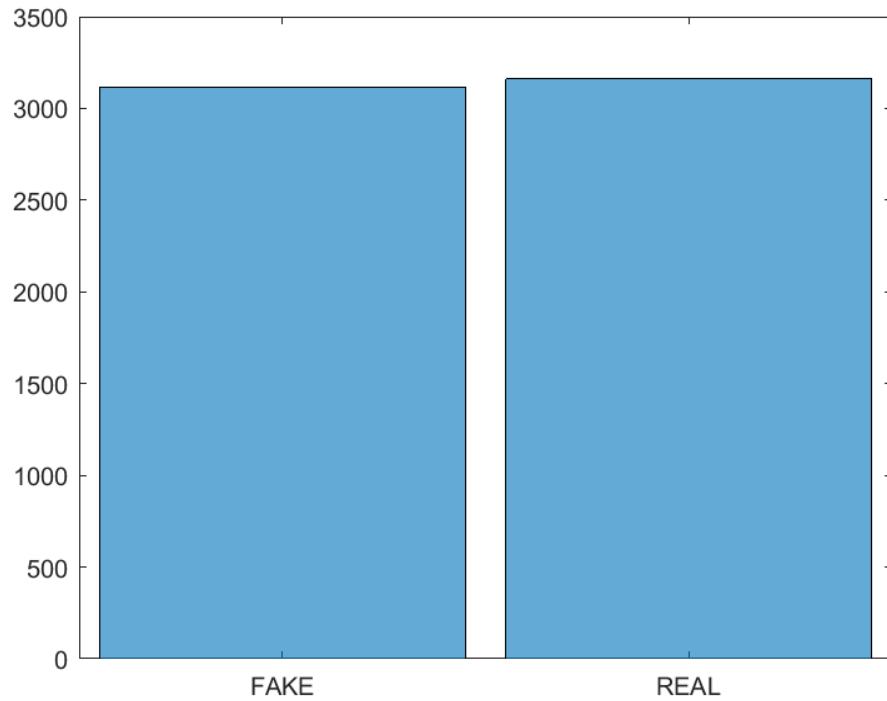
```
idxLabel = find(data.label == 'FAKE' & data.label == 'REAL');
data(idxLabel,:) = [];

idxEmpty = strlength(data.text) == 0;
data(idxEmpty,:) = [];

data.label = categorical(data.label);
```

visualizing the distribution of the classes. There appears to be non class imbalance

```
h = histogram(data.label);
```



Partitioning the dataset

```
cvp = cvpartition(data.label,'Holdout',0.1);
dataTrain = data(training(cvp),:);
dataTest = data(test(cvp),:);
```

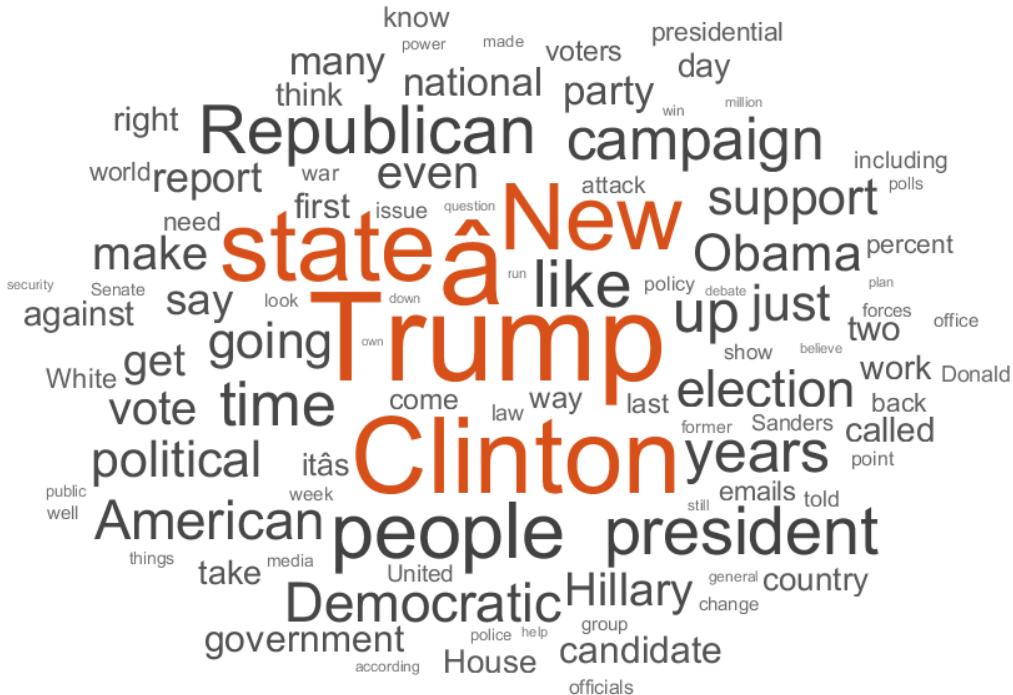
Partitioning into training and test sets

```
textDataTrain = dataTrain.text;
textDataTest = dataTest.text;
YTrain = dataTrain.label;
YTest = dataTest.label;
```

visualize training data using a word cloud

```
figure
wordcloud(textDataTrain);
title("Training Data")
```

Training Data



Little Preprocessing

```
textDataTrain = erasePunctuation(textDataTrain);
textDataTrain = lower(textDataTrain);
documentsTrain = tokenizedDocument(textDataTrain);
```

Word Embedding

```
tic;
embeddingDimension = 100;
embeddingEpochs = 50;

if doTrain
    emb = trainWordEmbedding(documentsTrain, ...
        'Dimension',embeddingDimension, ...
        'NumEpochs',embeddingEpochs, ...
        'Verbose',1)
else
    load('emb.mat');
end
```

Training: 100% Loss: 1.64922 Remaining time: 0 hours 0 minutes.

emb =

wordEmbedding with properties:

Dimension: 100

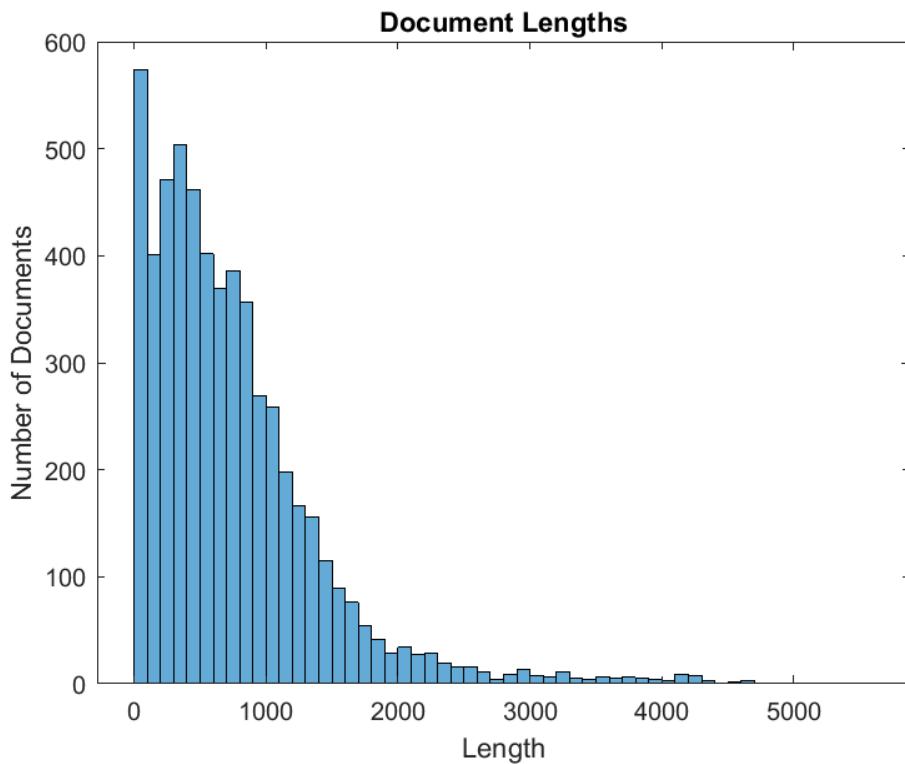
```
Vocabulary: [1x28987 string]
```

```
toc
```

```
Elapsed time is 732.518802 seconds.
```

length of docs

```
documentLengths = doclength(documentsTrain);  
figure  
histogram(documentLengths)  
title("Document Lengths")  
xlabel("Length")  
ylabel("Number of Documents")
```



Since most of the docs have tokens of size 1200, padding with seq length of 1200

```
sequenceLength = 1200;  
documentsTruncatedTrain = docfun(@(words) words(1:min(sequenceLength,end)),documentsTrain);  
save('emb_1200')
```

Converting the doc to sequences to be fed into the LSTM

```
XTrain = doc2sequence(embs,documentsTruncatedTrain);
```

Starting parallel pool (parpool) using the 'local' profile ...
connected to 12 workers.

Left Padding

```
for i = 1:numel(XTrain)
    XTrain{i} = leftPad(XTrain{i},sequenceLength);
end
XTrain(1:5');
```

Train Network

```
inputSize = embeddingDimension;
outputSize = 100;
numClasses = numel(categories(YTrain));

layers = [ ...
    sequenceInputLayer(inputSize)
    lstmLayer(outputSize,'OutputMode','last')
    fullyConnectedLayer(numClasses)
    softmaxLayer
    classificationLayer]
```

```
layers =
5x1 Layer array with layers:

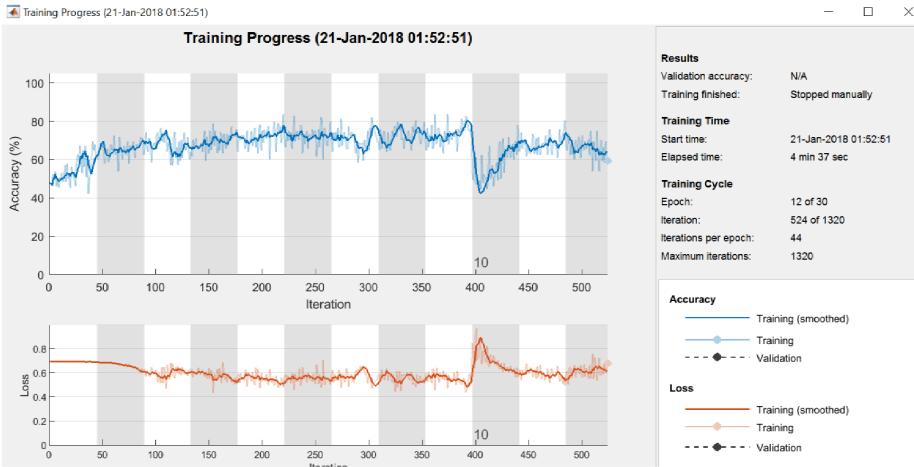
1  ''  Sequence Input      Sequence input with 100 dimensions
2  ''  LSTM                 LSTM with 100 hidden units
3  ''  Fully Connected     2 fully connected layer
4  ''  Softmax              softmax
5  ''  Classification Output crossentropyex
```

```
options = trainingOptions('sgdm', ...
    'InitialLearnRate',0.01, ...
    'Plots','training-progress', ...
    'Verbose',1)
```

Train Network (training with default options)

```
if doTrain
    net = trainNetwork(XTrain,YTrain,layers,options);
    save('firstPassNetwork.mat')
else
    load('lstmFake_firstPass.mat');
    imshow('firstPassNetwork.PNG');
```

end



Running the network on default hyperparameters yielded low accuracy with drops in training progress. This suggests that the initial learning rate be adjusted, and possibly, reduced after every nth epoch so that the objective func reaches a global minima and not diverge. Hence, adding a drop factor of 0.2 after every 2 epochs so that the steps for the learning rate become smaller with time

```
options = trainingOptions('sgdm', ...
    'LearnRateSchedule','piecewise',...
    'InitialLearnRate',0.05, ...
    'LearnRateDropFactor',0.2,... ...
    'LearnRateDropPeriod',2,... ...
    'Shuffle','never',...
    'Plots','training-progress', ...
    'Verbose',1)
```

```
options =
TrainingOptionsSGDM with properties:

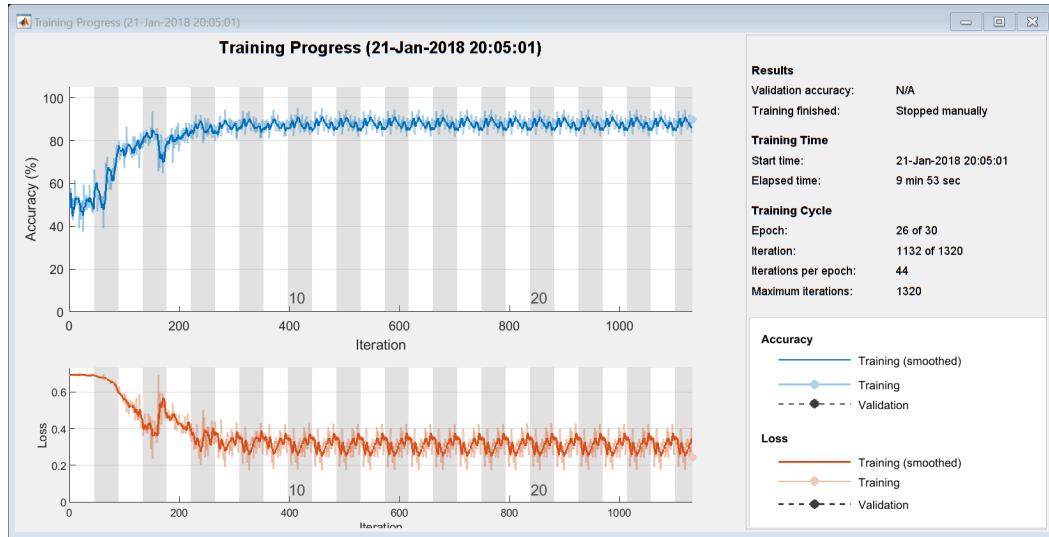
    Momentum: 0.9000
    InitialLearnRate: 0.0500
    LearnRateScheduleSettings: [1x1 struct]
    L2Regularization: 1.0000e-04
    GradientThresholdMethod: 'l2norm'
    GradientThreshold: Inf
    MaxEpochs: 30
    MiniBatchSize: 128
    Verbose: 1
    VerboseFrequency: 50
    ValidationData: []
    ValidationFrequency: 50
    ValidationPatience: 5
    Shuffle: 'never'
    CheckpointPath: ''
ExecutionEnvironment: 'auto'
    WorkerLoad: []
    OutputFcn: []
    Plots: 'training-progress'
SequenceLength: 'longest'
```

SequencePaddingValue: 0

```

if doTrain
    net = trainNetwork(XTrain,YTrain,layers,options);
    save('thirdPassNetwork.mat')
else
    load('lstmFake_secondPass_88.mat');
    imshow([('secondPassNetwork.PNG')]);
end

```



Training on single GPU.

Epoch	Iteration	Time Elapsed (hh:mm:ss)	Mini-batch Accuracy	Mini-batch Loss	Base Learning Rate
1	1	00:00:01	52.34%	0.6931	0.0500
2	50	00:00:26	57.03%	0.6825	0.0500
3	100	00:00:52	67.19%	0.5932	0.0100
4	150	00:01:18	82.81%	0.4485	0.0100
5	200	00:01:44	78.13%	0.4621	0.0020
6	250	00:02:11	85.16%	0.3944	0.0020
7	300	00:02:38	89.06%	0.3042	0.0004
8	350	00:03:05	87.50%	0.3628	0.0004
10	400	00:03:32	86.72%	0.3637	8.0000e-05
11	450	00:03:59	89.84%	0.2624	1.6000e-05
12	500	00:04:25	90.63%	0.2145	1.6000e-05
13	550	00:04:51	85.94%	0.3001	3.2000e-06
14	600	00:05:17	86.72%	0.3004	3.2000e-06
15	650	00:05:44	88.28%	0.3031	6.4000e-07
16	700	00:06:10	82.81%	0.4268	6.4000e-07
18	750	00:06:36	84.38%	0.3277	1.2800e-07
19	800	00:07:02	87.50%	0.3284	2.5600e-08
20	850	00:07:29	89.84%	0.2725	2.5600e-08
21	900	00:07:55	89.84%	0.2792	5.1200e-09
22	950	00:08:21	86.72%	0.3289	5.1200e-09
23	1000	00:08:47	89.84%	0.2418	1.0240e-09
24	1050	00:09:14	85.94%	0.3474	1.0240e-09
25	1100	00:09:37	85.16%	0.3342	2.0480e-10
26	1132	00:09:53	89.84%	0.2418	2.0480e-10

Warning: Variable 'XTrain' was not saved. For variables larger than 2GB use MAT-file version 7.3 or later.

testing

```
textDataTest = erasePunctuation(textDataTest);
textDataTest = lower(textDataTest);
documentsTest = tokenizedDocument(textDataTest);
```

Convert the docs to seq

```
documentsTruncatedTest = docfun(@(words) words(1:min(sequenceLength,end)),documentsTest);
XTest = doc2sequence(emb,documentsTruncatedTest);
```

Starting parallel pool (parpool) using the 'local' profile ...
connected to 12 workers.

```
for i=1:numel(XTest)
    XTest{i} = leftPad(XTest{i},sequenceLength);
end
XTest(1:5)
```

ans = 1x5 cell array
{100x1200 single} {100x1200 single} {100x1200 single} {100x1200 single} {100x1200 single}

Inference

```
YPred = classify(net,XTest);
```

accuracy

```
accuracy = sum(YPred == YTest)/numel(YPred)
```

accuracy = 0.8947

Plot confusion

```
plotconfusion(YTest,YPred);
```

Confusion Matrix			
Output Class	Target Class		
	FAKE	REAL	
FAKE	281 44.8%	36 5.7%	88.6% 11.4%
REAL	30 4.8%	280 44.7%	90.3% 9.7%
	90.4% 9.6%	88.6% 11.4%	89.5% 10.5%