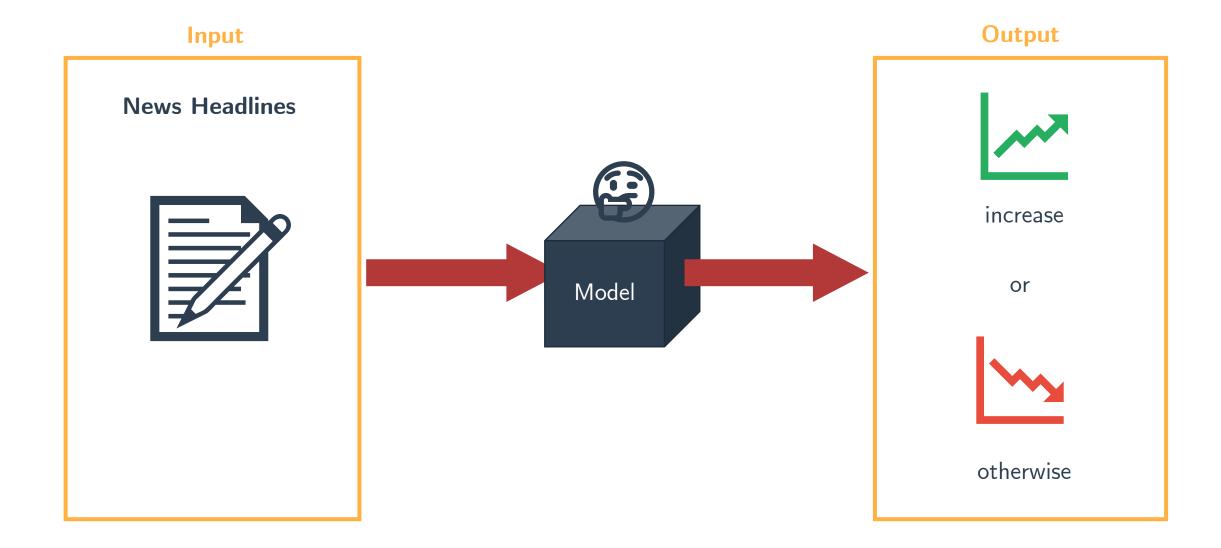
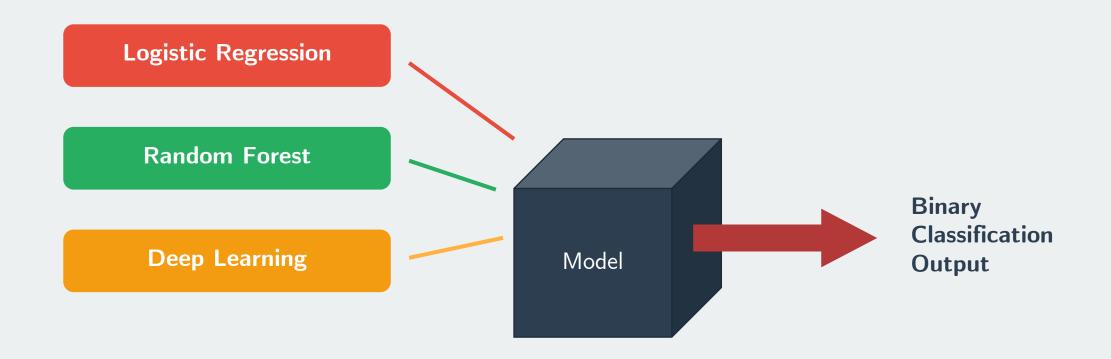
# Machine Learning Project Daily News for Stock Movement Prediction

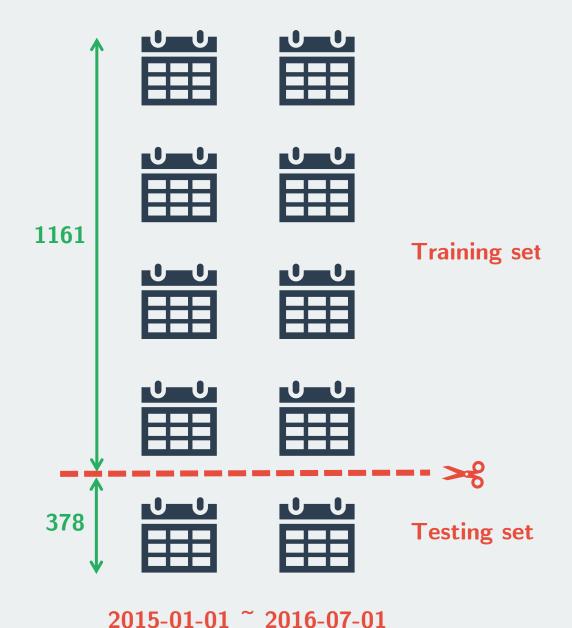
by

Zimin Luo 417124





#### 2008-08-08 ~ 2014-12-31

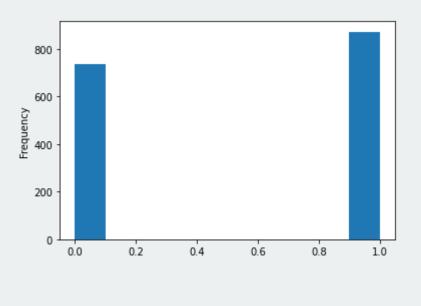


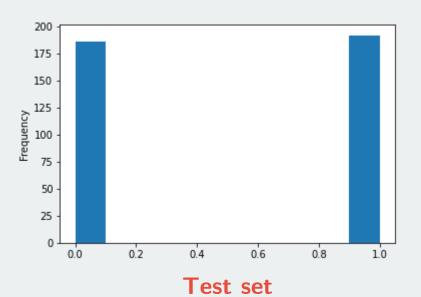
1. Combine all news headlines into one blob

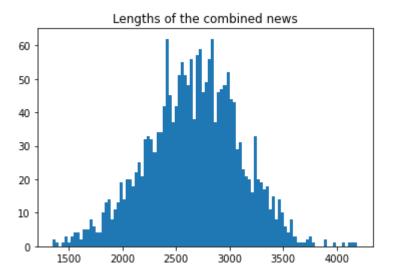
#### 2. Clean text:

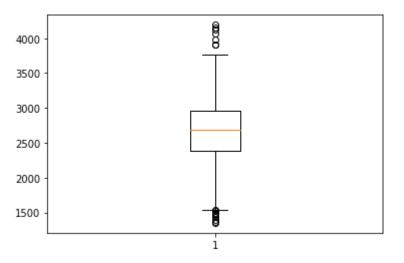
- normalize (convert to lower cases)
- remove HTML tags
- remove texts in brackets []
- remove punctuation
- remove digits
- fix concentration \*
- remove stop words \*
- stem words \*\*
- lemmatization \*\*

#### **Training set**









## Word cloud of positive news ["Label" = 1]

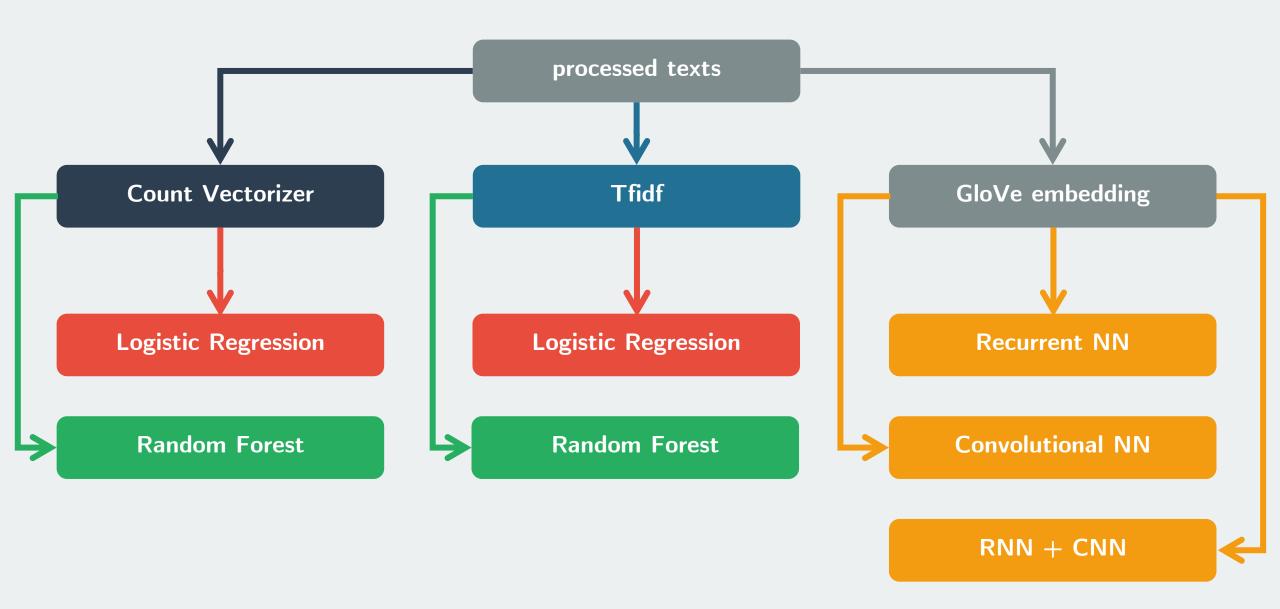


### Word cloud of negative news ["Label" = 0]



#### **Topic Modeling**

```
Topic: 0
0.001*"protest" + 0.001*"egypt" + 0.001*"iran" + 0.000*"amp" + 0.000*"russia" + 0.000*"gaza" + 0.000*"russian" + 0.000*"israel" + 0.000*"canada" + 0.000*"egyptian"
Topic: 1
0.001*"israel" + 0.001*"gaza" + 0.000*"isra" + 0.000*"news" + 0.000*"korea" + 0.000*"china" + 0.000*"palestinian" + 0.000*"north" + 0.000*"nuclear" + 0.000*"bank"
Topic: 2
Words: 0.001*"gaza" + 0.001*"isra" + 0.001*"ukrain" + 0.001*"israel" + 0.000*"un" + 0.000*"amp" + 0.000*"fire" + 0.000*"nuclear" + 0.000*"protest" + 0.000*"russia"
Topic: 3
0.001*"korea" + 0.001*"north" + 0.0001*"gaza" + 0.000*"palestinian" + 0.000*"iran" + 0.000*"riot" + 0.000*"isra" + 0.000*"russia" + 0.000*"amp" + 0.000*"human"
Topic: 4
0.001*"korea" + 0.000*"protest" + 0.000*"rape" + 0.000*"syria" + 0.000*"south" + 0.000*"israel" + 0.000*"iran" + 0.000*"north" + 0.000*"gaza" + 0.000*"palestinian"
Topic: 5
0.001* "wikileak" + 0.001* "isra" + 0.000* "russia" + 0.000* "israel" + 0.000* "gaza" + 0.000* "war" + 0.000* "militari" + 0.000* "drug" + 0.000* "presid" + 0.000* "nuclear"
Topic: 6
0.001*"amp" + 0.001*"korea" + 0.001*"syria" + 0.001*"north" + 0.001*"isra" + 0.000*"iran" + 0.000*"wikileak" + 0.000*"protest" + 0.000*"russia" + 0.000*"nuclear"
Topic: 7
0.001*"syria" + 0.001*"ukrain" + 0.000*"gaza" + 0.000*"oil" + 0.000*"right" + 0.000*"citi" + 0.000*"russian" + 0.000*"snowden" + 0.000*"forc" + 0.000*"amp"
Topic: 8
0.001*"gaza" + 0.001*"protest" + 0.001*"israel" + 0.000*"isra" + 0.000*"snowden" + 0.000*"isi" + 0.000*"iran" + 0.000*"russia" + 0.000*"wikileak" + 0.000*"drug"
Topic: 9
0.001*"ukrain" + 0.001*"gaza" + 0.001*"palestinian" + 0.001*"russia" + 0.001*"wikileak" + 0.001*"israel" + 0.000*"war" + 0.000*"protest" + 0.000*"korea" + 0.000*"isra"
```



#### Count Vectorizer

- uni-gram
- bi-gram
- tri-gram

#### Tfidf

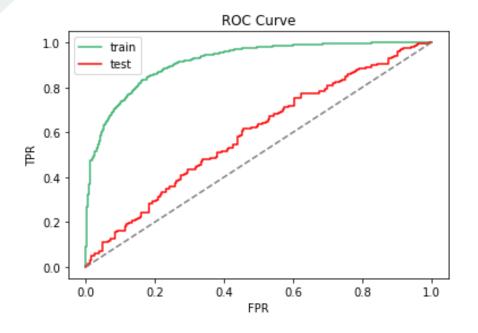
- uni-gram
- bi-gram
- tri-gram
- bi-gram + custom settings •



**Logistic Regression** 

The AUC Score is: 0.566 The F1 score is: 0.629

The co	onfusion	matrix is:			
		precision	recall	f1-score	support
	0	0.59	0.41	0.49	186
	1	0.56	0.72	0.63	192
a	ccuracy			0.57	378
mad	cro avg	0.57	0.57	0.56	378
weight	ted avg	0.57	0.57	0.56	378



#### **Count Vectorizer**

• bi-gram

#### Tfidf

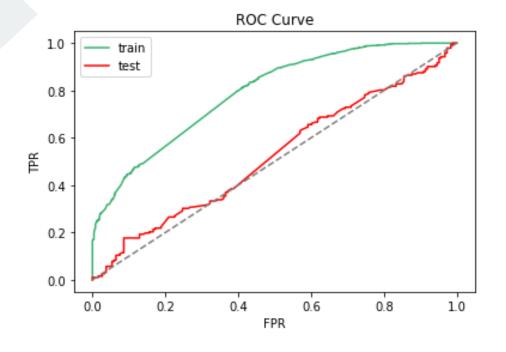
• bi-gram + custom settings

#### **Random Forest**

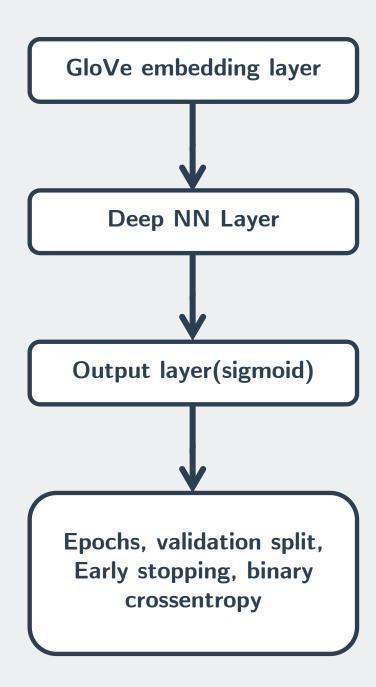
- AdaBoost
- **XGBoost**

The AUC Score is: 0.529 The F1 score is: 0.599

The confusi	on	matrix is:			
		precision	recall	f1-score	support
	0	0.53	0.37	0.44	186
	_				
	1	0.53	0.69	0.60	192
				0.53	270
accurac	y			0.53	378
macro av	g	0.53	0.53	0.52	378
weighted av	g	0.53	0.53	0.52	378

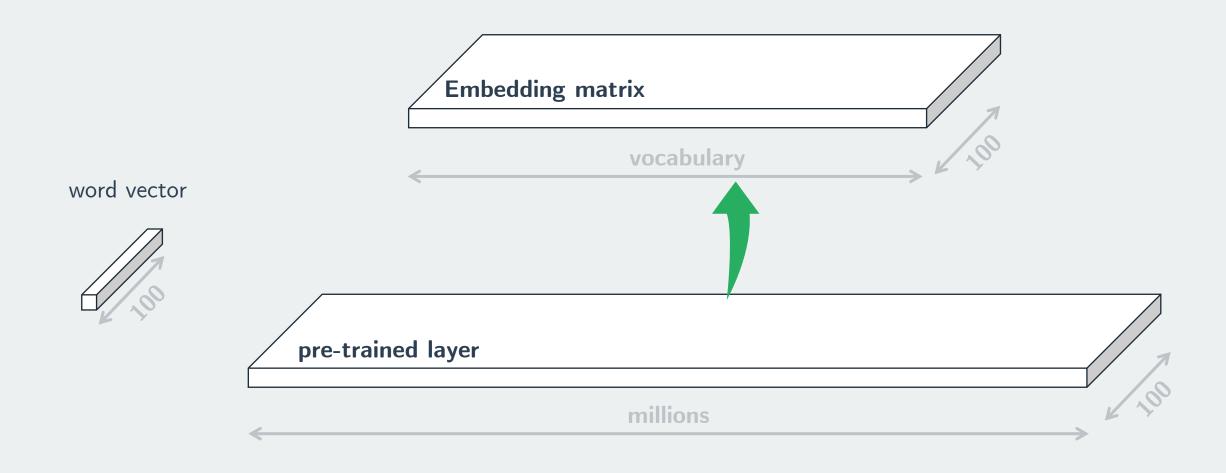


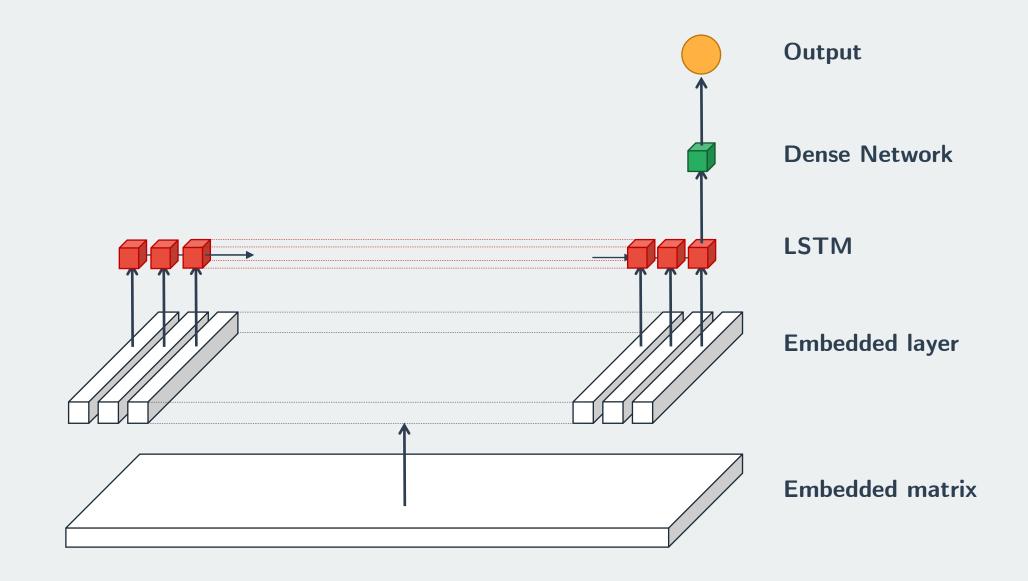
#### **Deep Learning**



#### GloVe embedding layer

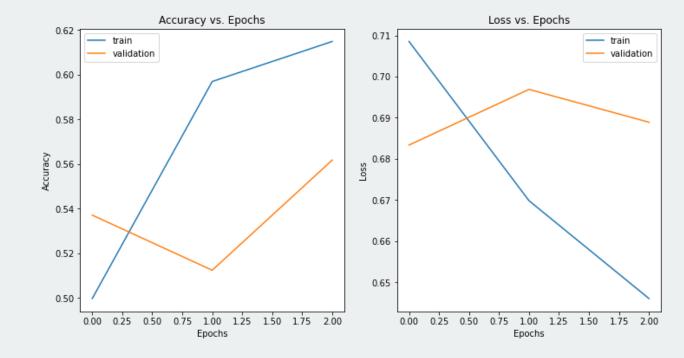
- Maximum length of each combined headline: 2500
- Dimension of word vector: 100

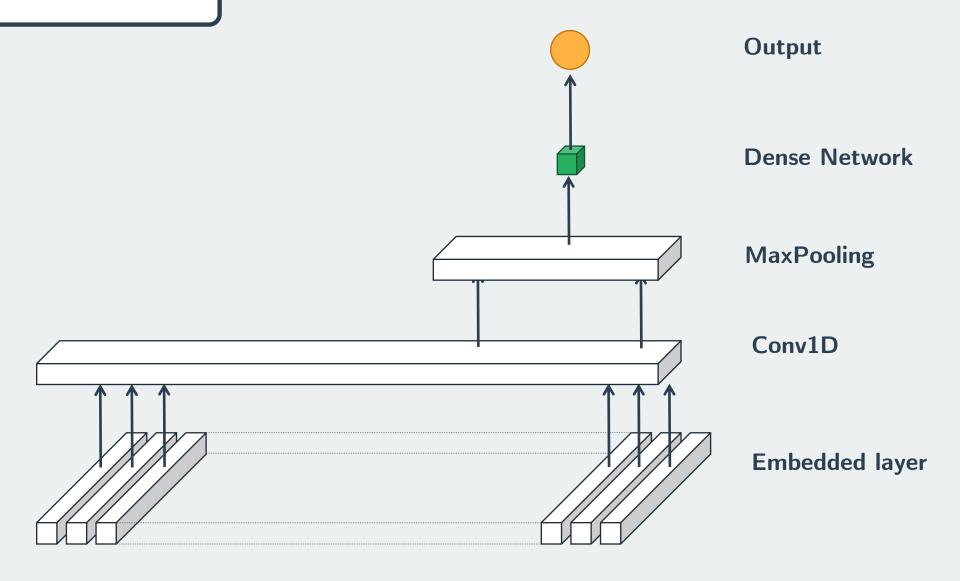




#### RNN

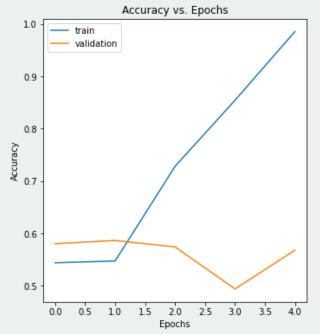
Model	Variants	Accuracy score
LSTM (RNN)	Default	0.5159
LSTM (RNN)	learning rate = 0.005	0.5238
LSTM (RNN)	dropout = 0.2	0.5132
LSTM (RNN)	dropout = 0.2, reccurent dropout = 0.2	0.4815
LSTM (RNN)	2 LSTM Layers	0.5132
LSTM (RNN)	optimizer = Nadam	0.4894
GRU (RNN)	Default	0.5026
GRU (RNN)	optimizer = Nadam	0.4894

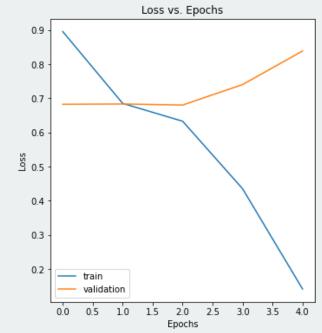




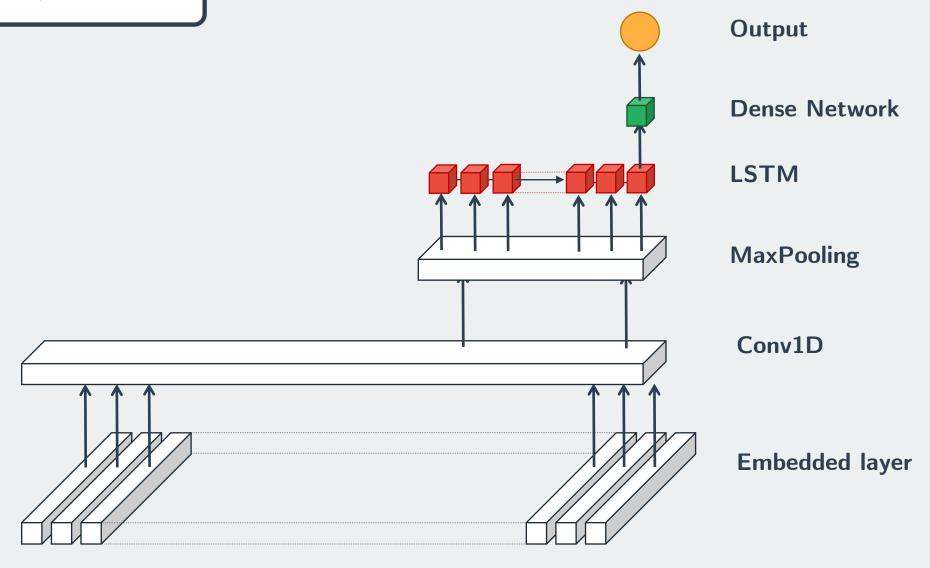
#### CNN

Model	Variants	Accuracy score
CNN	filters=32, kernel_size=5	0.5079
CNN	filters=64, kernel_size=5	0.4603
CNN	filters=100, kernel_size=5	0.5
CNN	filters=32, kernel_size=3	0.5026
CNN	filters=32, kernel_size=2	0.4841





CNN + RNN



#### CNN + RNN

Model	Variants	Accuracy score	0.625 - 0.600 -		train validation
CNN + LSTM	filters=32, kernel_size=5	0.5079	0.575 -		
CNN + GRU	filters=32, kernel_size=5	0.5106	0.550 -	0.690	
CNN + GRU	filters=32, kernel_size=5, learning_rate = 0.005	0.5	- 0.500 -	0.685	
CNN + GRU	filters=10, kernel_size=5, learning_rate = 0.005	0.5265	0.475 -		
CNN + GRU	filters=10, kernel_size=5, learning_rate = 0.007	0.4868	0.450 - 0.425 -	0.670 -	
				0.0 0.5 1.0 1.5 2.0 2.5 3.0 3.5 4.0 0.0 0.5 1.0 1.5 2.0 2 Epochs	2.5 3.0 3.5 4.0

Accuracy vs. Epochs

Loss vs. Epochs

## thank you