

# Sentiment Analysis of the Speaker in Literary Works

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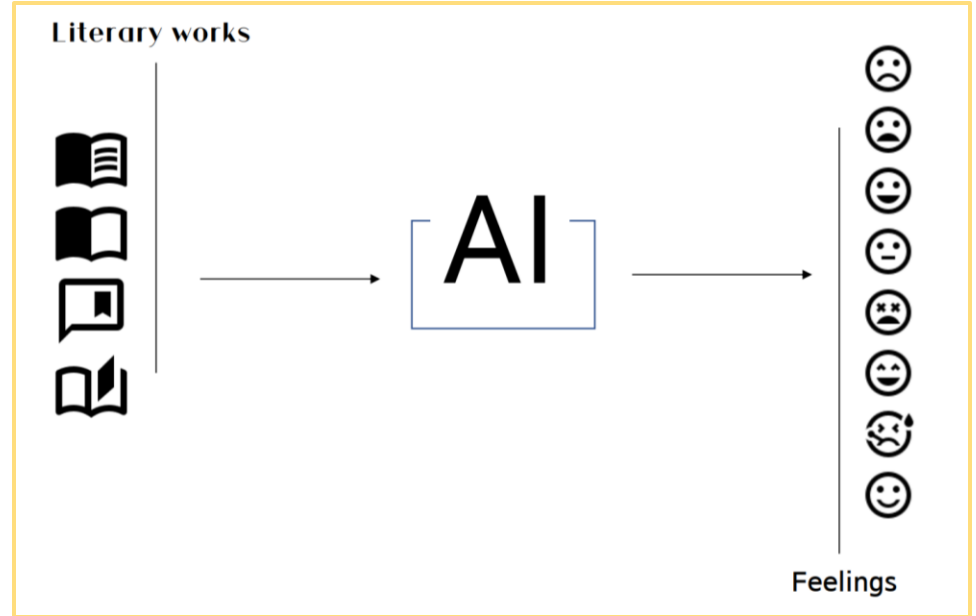
# Background

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# Topic

Sentiment classification  
of the speaker  
in literary works



# Classify into various emotions

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Pre-existing model



Our model



# Social Media Texts

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## Tweets



Reveal the emotions well  
Many researches on  
social media emotion analysis

**Training**

Emotion\_final.csv (2.26 MB)

Detail Compact Column

▲ Text	▲ Emotion
<b>21405</b> unique values	happy 33% sadness 29% Other (8165) 38%
i didnt feel humiliated	sadness
i can go from feeling so hopeless to so damned hopeful just from being around someone who cares and ...	sadness
im grabbing a minute to post i feel greedy wrong	anger
i am ever feeling nostalgic about the fireplace i will know that it is still on the property	love
i am feeling grouchy	anger
ive been feeling a little burdened lately wasnt sure why that was	sadness
ive been taking or	surprise

## Data

21,459

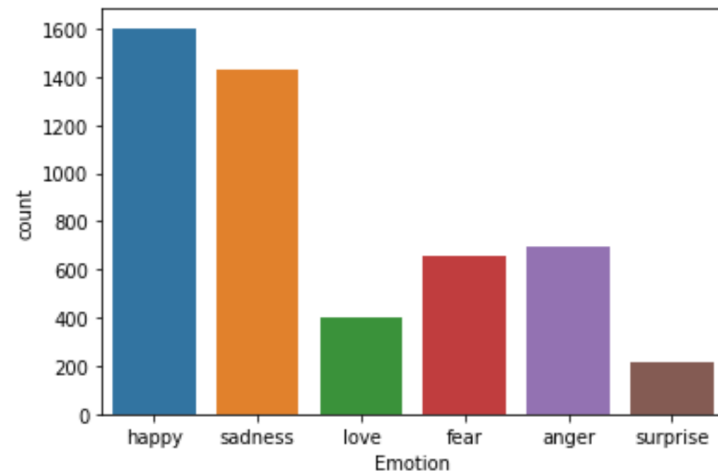
## Emotions

Happy, Fear, Anger,  
Sadness, Surprise, Love

## Split data

- Train: 65%
- Validation: 15%
- Test: 20%

# Dataset



The amount of data each emotion

# Baseline model - SVM

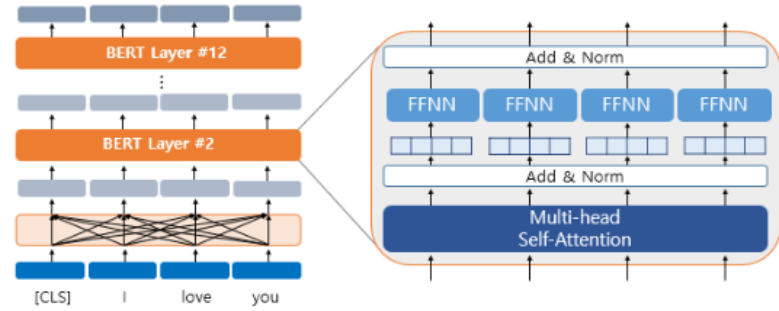
- Analyze data for classification and regression analysis.
- Kernel trick - efficiently perform a non-linear classification

## Result

Accuracy: 0.699  
F1-Score: 0.672

Accuracy: 0.699					
	precision	recall	f1-score	support	
0	0.891	0.410	0.562	139	
1	0.870	0.511	0.644	131	
2	0.605	0.947	0.738	320	
3	0.920	0.287	0.438	80	
4	0.757	0.847	0.799	287	
5	0.500	0.140	0.218	43	
accuracy			0.699	1000	
macro avg	0.757	0.524	0.567	1000	
weighted avg	0.744	0.699	0.672	1000	

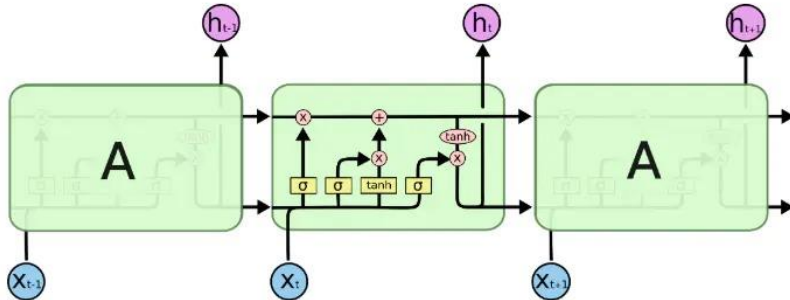




**LSTM**

Which model  
we should choose?

**BERT**



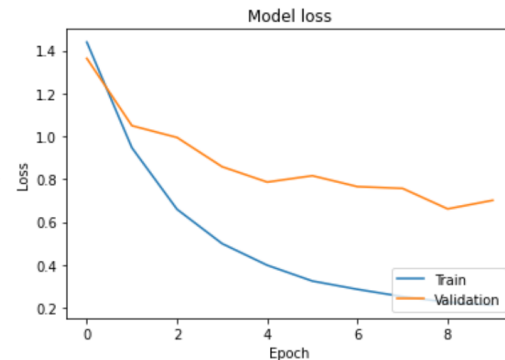
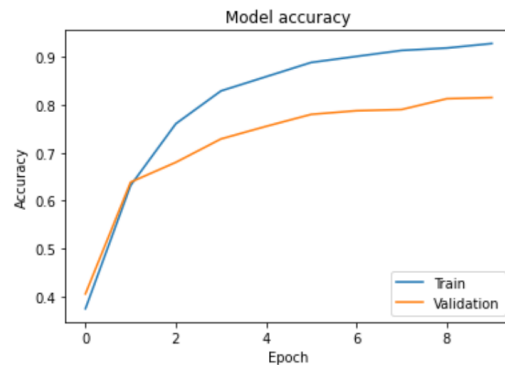
# LSTM

## Structure

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, None, 16)	32000
lstm (LSTM)	(None, 64)	20736
dense (Dense)	(None, 128)	8320
dense_1 (Dense)	(None, 64)	8256
dense_2 (Dense)	(None, 6)	390

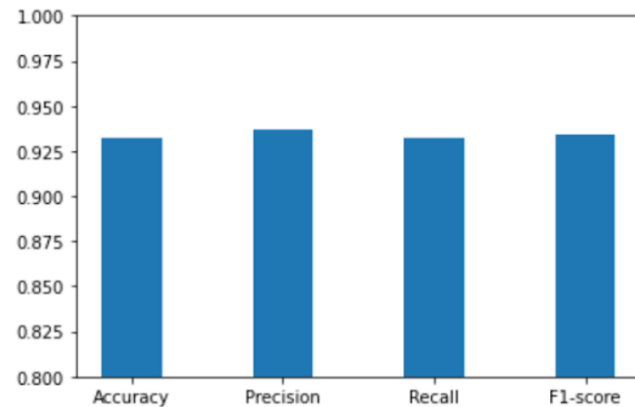
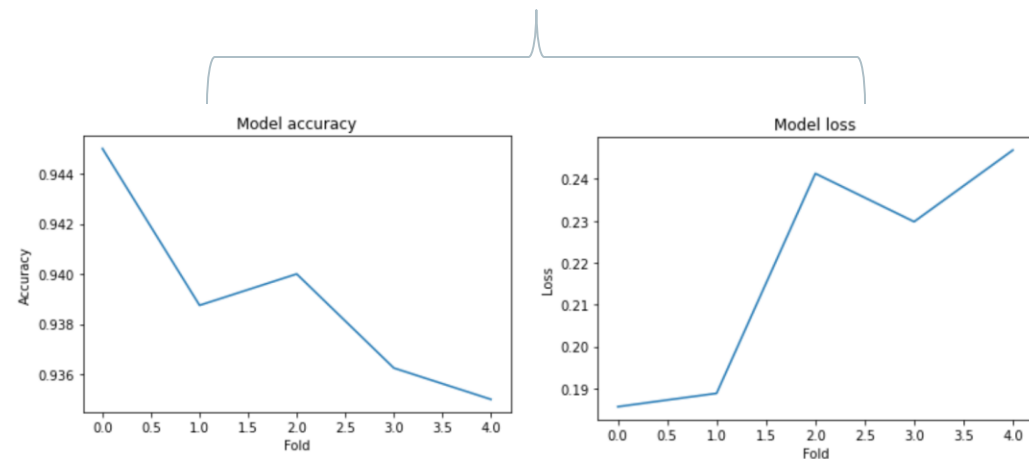
Total params: 69,702  
Trainable params: 69,702  
Non-trainable params: 0

## Training Result



# LSTM with K-fold cross validation

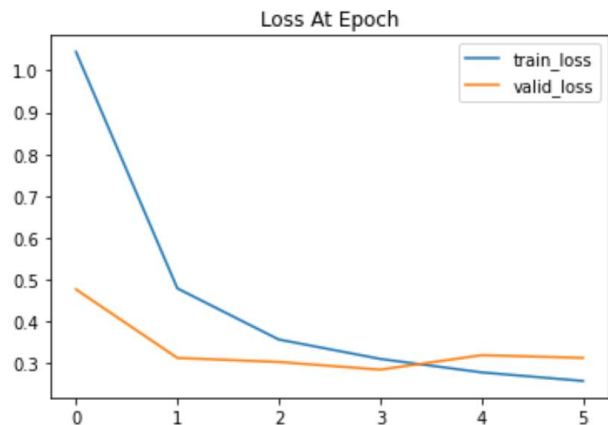
## The result of each fold



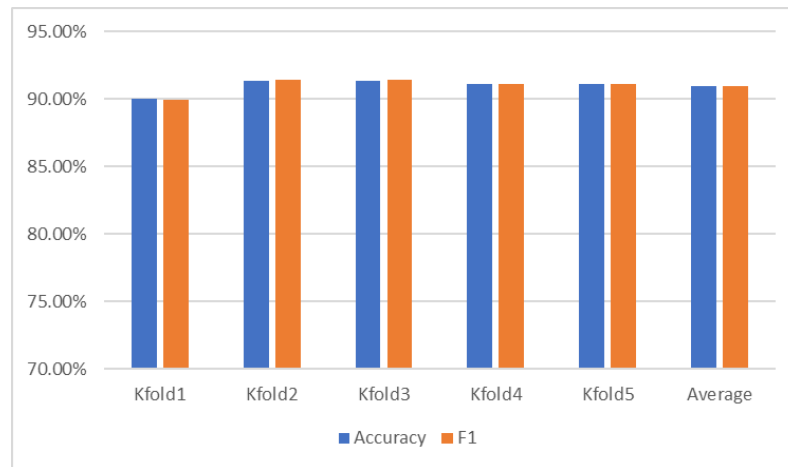
Accuracy	0.9325
Precision	0.9365
Recall	0.9325
F1-score	0.9337

# BERT with K-fold cross validation

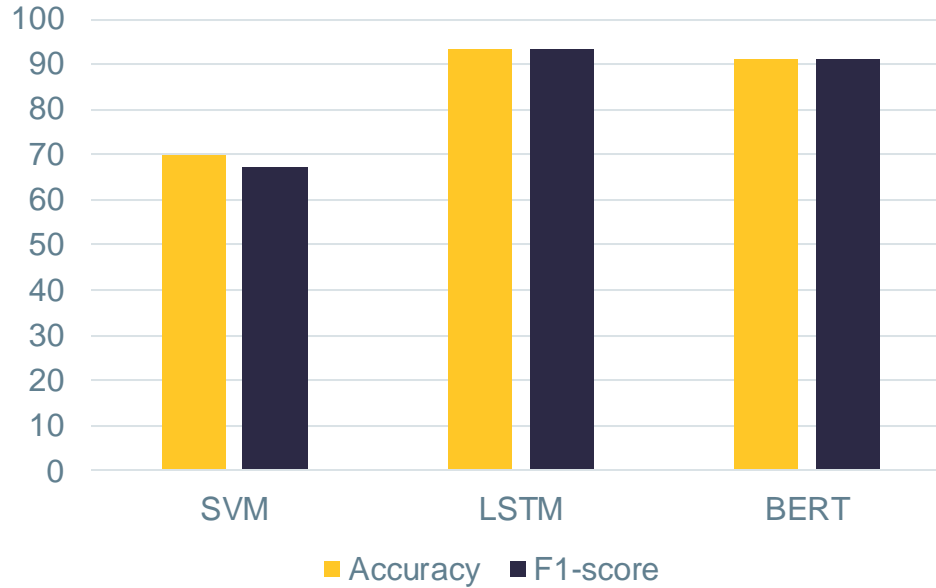
## Training Result



## The result of each fold



# Comparison of LSTM and BERT



# Create test dataset

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How to increase the association between two types of text?

# Create test dataset

- **Type**

Poem, Novel, Lyric, and YouTube comment

- **Amount of data**

Almost 30 data each type and each emotion

- **How to collect data**

Team members crawled the data

- **Criteria of collecting data**

Not appearing emotional words repeatedly

Filtering curses, and correcting spelling

Dataset	Test	Poem	Novel	Lyric	YouTube
anger	599	30	30	30	30
fear	530	30	30	30	30
happy	1406	30	30	30	30
love	328	31	31	30	30
sadness	1253	33	33	31	30
surprise	176	11	11	1	20
Total	4292	165	165	152	170

# Literary Work Test Data Set

## Poem / Love

Shy one, shy one,  
Shy one of my heart,  
She moves in the  
firelight  
Pensively apart.

And shy as a rabbit,  
Helpful and shy.  
To an isle in the water  
With her would I fly.

## Novel / Surprise

"My stars!" cried Aunt  
Polly.  
"Why, I can't believe it!"  
She stared at the fence.  
"You can work when  
you want to, Tom  
Sawyer!"

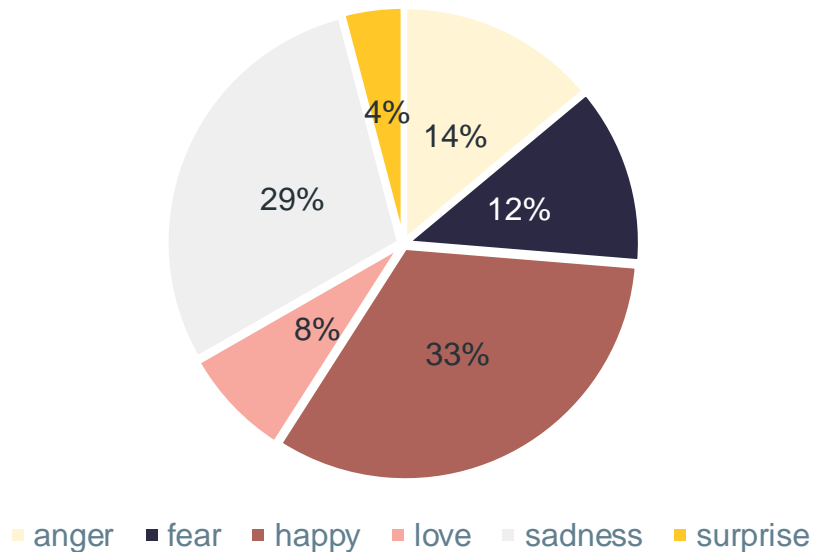
## YouTube / Anger

You guys really have  
the nerve to call him a  
"devoted father?" What  
a slap in the face to the  
kids!



# BERT Train Data set

Dataset	Train	Validation	Test	Total
anger	1,915	479	599	2,993
fear	1,698	424	530	2,652
happy	4,498	1,125	1,406	7,029
love	1,050	263	328	1,641
sadness	4,010	1,002	1,253	6,265
surprise	562	141	176	879
Total	13,733	3,434	4,292	21,459
Ratio	64%	16%	20%	100%



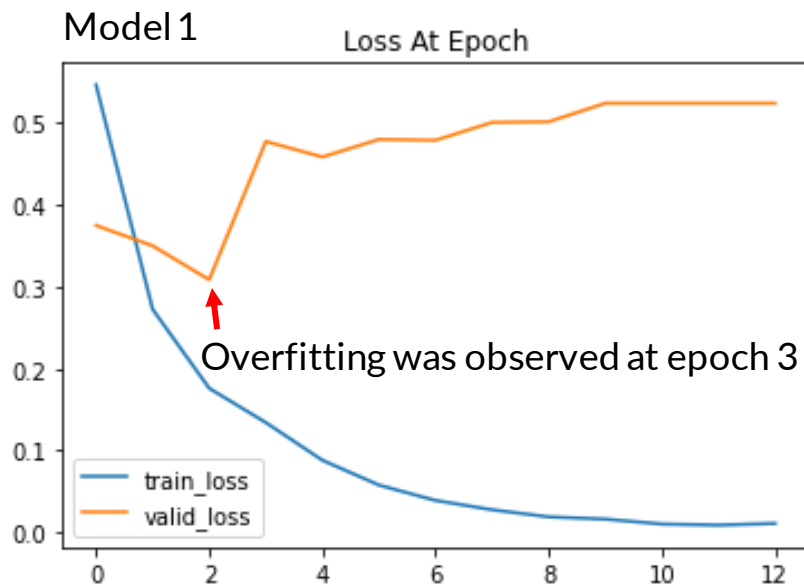
# Our model - BERT

## Train Bert with full data

### Hyper Parameters

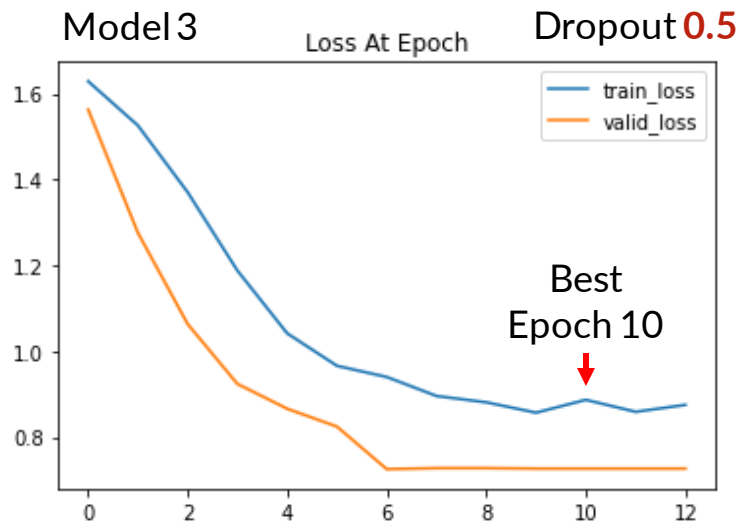
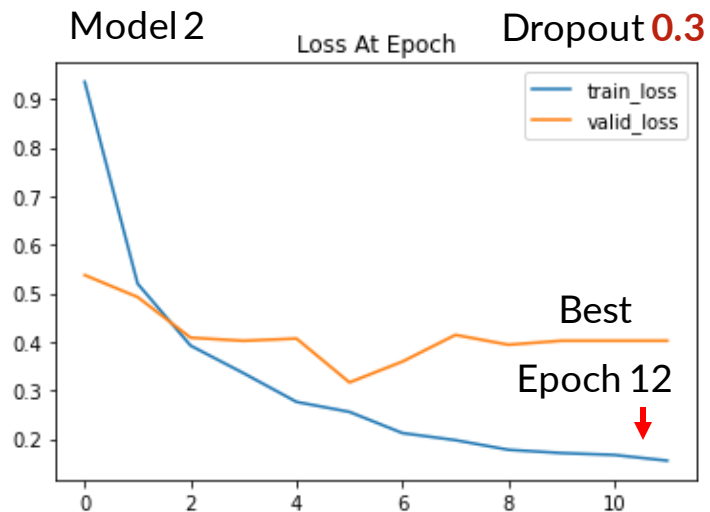
Learning rate =  $1e-5$ , eps =  $1e-8$

dropout = 0.1, epoch = 13, batch-size = 3



# Our model - BERT

Change Hyper Parameters  
Try to avoid overfitting



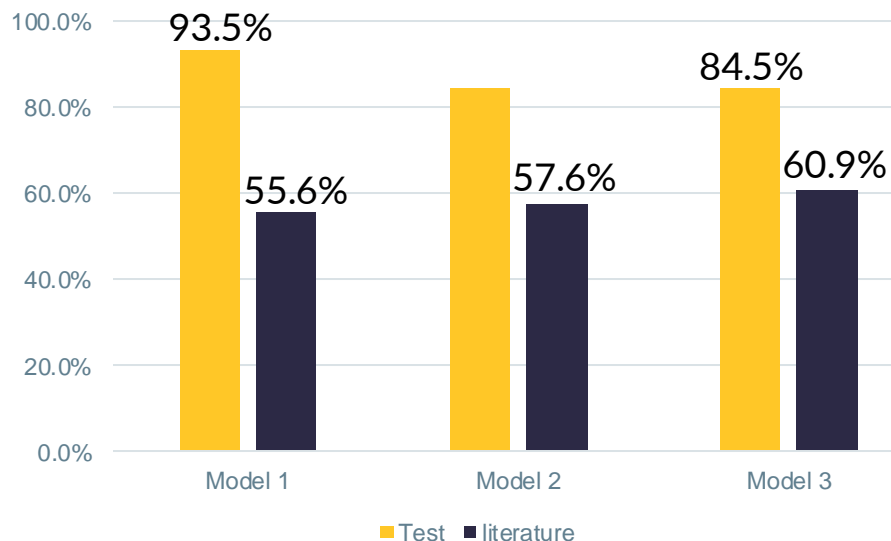
Train loss and validation loss show decreasing trend and converge to a certain value.  
In model2 and model3 it seems that **overfitting is highly unlikely**.

# Result

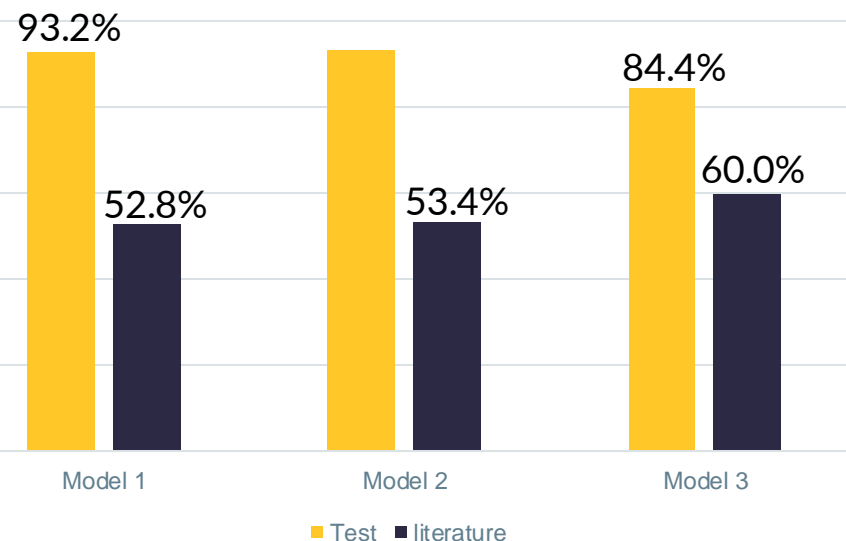
Model		test	lyric	poem	Youtube	novel
Model1	Accuracy	93.2%	40.13%	58.79%	67.65%	54.36%
	F1 Score	93.18%	37.22%	55.32%	65.66%	52.93%
Model2	Accuracy	<u>93.5%</u>	<u>48.68%</u>	56.36%	68.82%	55.90%
	F1 Score	93.53%	37.22%	55.79%	66.16%	54.38%
Model3	Accuracy	84.5%	44.74%	<u>69.70%</u>	<u>71.76%</u>	<u>56.92%</u>
	F1 Score	84.38%	43.84%	69.34%	71.09%	56.03%

# Comparison the result

## Accuracy



## F1 Score



The best hyper-parameter about Literary works

**Model 3** dropout = 0.5, epoch = 10

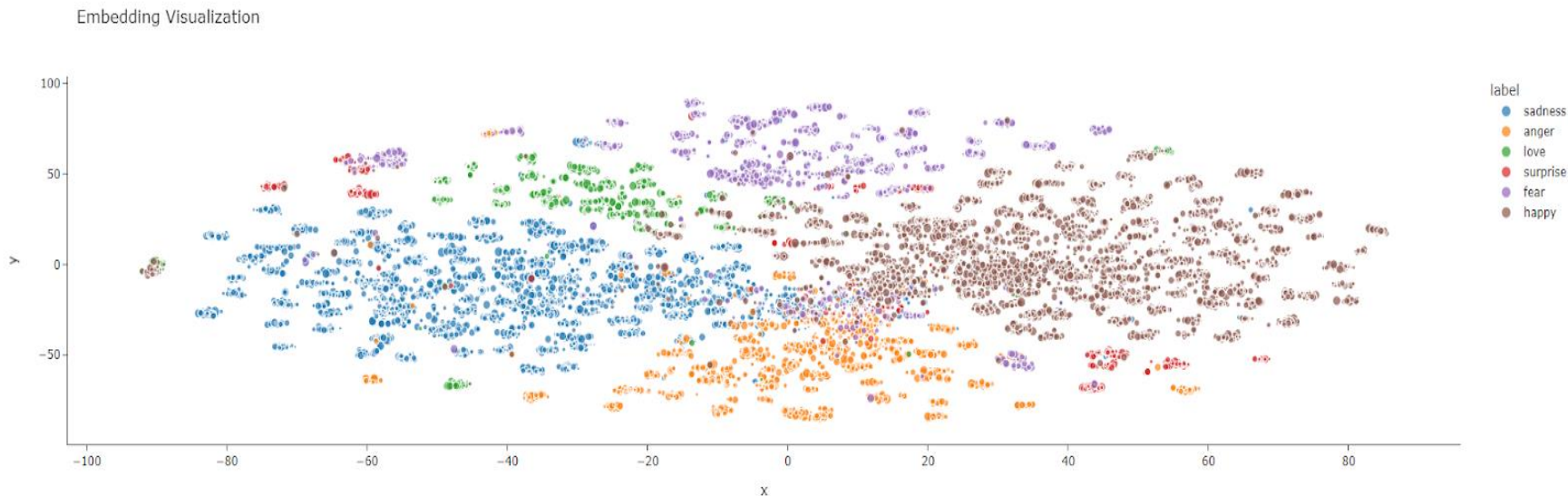
# Our model – BERT Detail Table

	test	lyric	poem	Youtube	novel
anger	93.2%	43.3%	70.0%	70.0%	63.2%
fear	89.6%	23.3%	60.0%	73.3%	72.4%
happy	93.8%	76.7%	83.3%	100.0%	53.1%
love	89.0%	13.3%	12.9%	20.0%	37.5%
sadness	96.4%	41.9%	78.8%	80.0%	80.6%
surprise	83.0%	100.0%	27.3%	60.0%	21.2%
Total	93.2%	40.1%	58.8%	67.6%	54.4%
F1 Score	93.18%	37.22%	55.32%	65.66%	52.93%

	test	lyric	poem	Youtube	novel
anger	82.6%	26.7%	66.7%	63.3%	63.2%
fear	79.4%	23.3%	53.3%	66.7%	51.7%
happy	87.1%	76.7%	76.7%	93.3%	59.4%
love	57.6%	43.3%	58.1%	43.3%	46.9%
sadness	93.1%	51.6%	90.9%	93.3%	90.3%
surprise	75.0%	100.0%	72.7%	70.0%	30.3%
Total	84.5%	44.7%	69.7%	71.8%	56.9%
F1 Score	84.38%	43.84%	69.34%	71.09%	56.03%

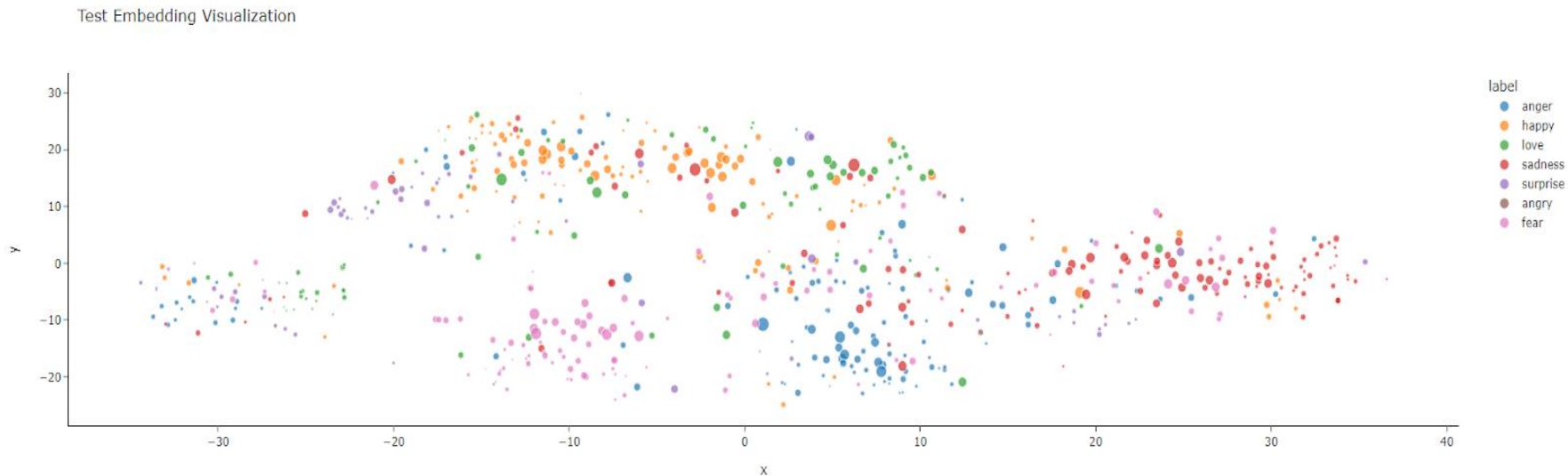
# Result – Visualization

Scatter plot with train and validation data label clusters



# Result – Visualization

## Scatter plot with test data label clusters





**THANKS**

