



# CIS 530 Computational Linguistics: Sentiment Analysis on Movie Reviews

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# Introduction

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- Online reviews for products and comments on social media -> Sentiment Analysis
- Application/Social Impact:
  - improve recommendation system of social apps
  - quickly gather attitudes towards products
- Goal: text information -> sentiment/attitudes
  - eg: predict sentiment score from reviews

# Introduction

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- Models:
  - simple baseline: logistic regression
  - strong baseline: vanilla Bert
  - extension 1: fine tuning Bert
  - extension 2: BiLSTM-CNN

# Data

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- Available Data: Amazon Product Data, Tweets, IMDB movie reviews, ect.
- Our choice: Movie Reviews from Rotten Tomatoes
  - sentiment score from 0 - 4

# Evaluation Metric

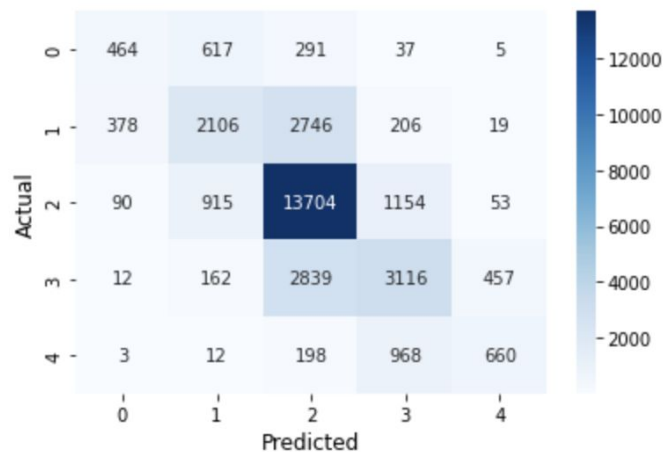
- F1 score:

$$F_1 = \frac{2}{\text{recall}^{-1} + \text{precision}^{-1}} = 2 \frac{\text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}} = \frac{2\text{tp}}{2\text{tp} + \text{fp} + \text{fn}}$$

- Macro F1 score:  $\text{sum}(\text{F1 scores}) / \# \text{ classes}$

# Simple Baseline

- Bag of Words + Logistic Regression
- Results:

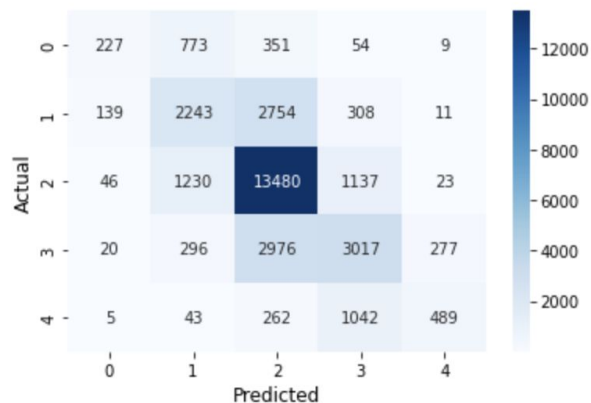


Classification Metrics

	precision	recall	f1-score	support
0	0.49	0.33	0.39	1414
1	0.55	0.39	0.45	5455
2	0.69	0.86	0.77	15916
3	0.57	0.47	0.52	6586
4	0.55	0.36	0.43	1841
accuracy			0.64	31212
macro avg	0.57	0.48	0.51	31212
weighted avg	0.62	0.64	0.62	31212

# Strong Baseline

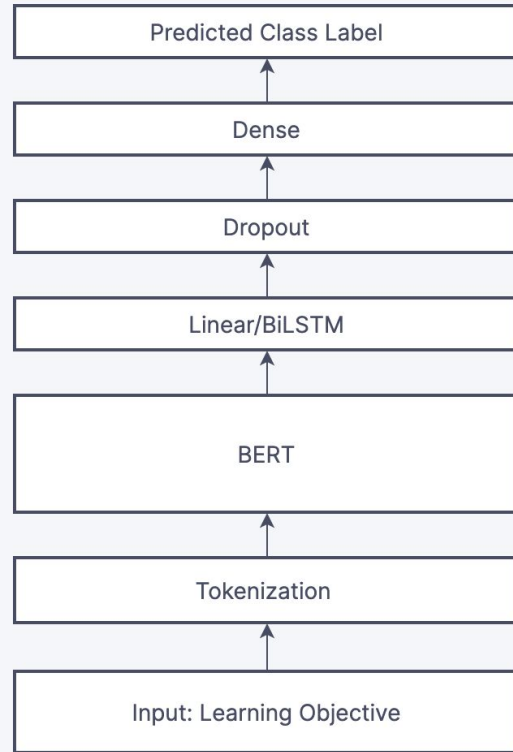
- BERT: context information
- BERT encoder + Logistic Regression
- Results:



Classification Metrics

	precision	recall	f1-score	support
0	0.52	0.16	0.25	1414
1	0.49	0.41	0.45	5455
2	0.68	0.85	0.75	15916
3	0.54	0.46	0.50	6586
4	0.60	0.27	0.37	1841
accuracy			0.62	31212
macro avg	0.57	0.43	0.46	31212
weighted avg	0.61	0.62	0.60	31212

# Model Extension I: Fine-Tuning BERT





# Model Extension I: Training Details

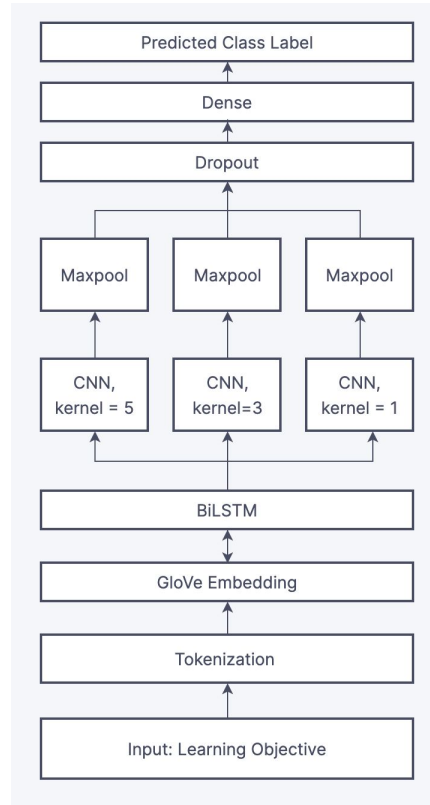
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- Epoch: 31 epochs  
1st epoch: train on weights of Bert and Fine-tuning  
Remaining epochs: only train on weight of Fine-tuning
- 2 Loss functions:
  - Cross Entropy Loss
  - Weighted Cross Entropy Loss

# Model Extension I: Evaluation

Section	Model	Accuracy	Macro F1	Weighted F1
Simple Baseline	Bag-of-words+Logistic	0.64	0.51	0.62
Published Baseline	BERT+Logistic	0.62	0.46	0.60
Extension1	BERT+Linear+Unweighted Loss	<b>0.69</b>	0.61	0.69
Extension1	BERT+Linear+Weighted Loss	0.68	0.61	0.68
Extension1	BERT+BiLSTM+Unweighted Loss	0.68	0.61	0.68
Extension1	BERT+BiLSTM+Weighted Loss	0.68	<b>0.62</b>	<b>0.69</b>
Extension2	GloVe+BiLSTM+CNN	0.67	0.59	0.67

# Model Extension 2: BiLSTM-CNN



# Model Extension 2: BiLSTM-CNN Detail

Section	Model	Accuracy	Macro F1	Weighted F1
Simple Baseline	Bag-of-words+Logistic	0.64	0.51	0.62
Published Baseline	BERT+Logistic	0.62	0.46	0.60
Extension1	BERT+Linear+Unweighted Loss	<b>0.69</b>	0.61	0.69
Extension1	BERT+Linear+Weighted Loss	0.68	0.61	0.68
Extension1	BERT+BiLSTM+Unweighted Loss	0.68	0.61	0.68
Extension1	BERT+BiLSTM+Weighted Loss	0.68	<b>0.62</b>	<b>0.69</b>
Extension2	GloVe+BiLSTM+CNN	0.67	0.59	0.67

# Error Analysis

- Does well in predicting extremely negative (class 0) and extremely positive (class 4)
- As feeling becomes neutral, gets more incorrect predictions - class 2 worst
- Mostly incorrectly predicted as the nearby class(es)

True	Predict 0	Predict 1	Predict 2	Predict 3	Predict 4
0	30541	632	30	9	0
1	611	29466	994	141	3
2	125	2318	26687	2018	64
3	7	201	1366	28717	921
4	0	10	18	625	30559

# Conclusion

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- Simple baseline: bag of words + logistic regression
- Strong baseline: BERT + logistic regression
- Extension 1: BERT + neural network (MLP & BiLSTM) + weighted/unweighted loss function
- Extension 2: GloVe.6B.300d + BiLSTM-CNN
- Best performance achieved by BERT + BiLSTM + weighted cross entropy loss - **accuracy 0.68, F1-score 0.69**

# Thank You

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- Thank you for listening!