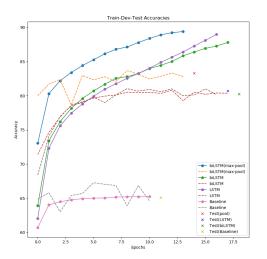
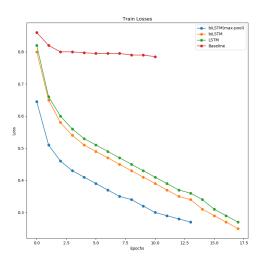
Assignment 1: InferSent

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1 Training process





(a) Train-Dev-Test Accuracy

(b) Train loss

Figure 1: Test accuracy achieved by all the models: **Baseline** = 65.07%, **LSTM** = 80.73%, **biLSTM** = 80.26%, **biLSTM**(max-pool) = 83.32%

2 Results and Analysis

Model	MR	CR	SUBJ	MPQA	SST	TREC	MRPC	SICK-R	SICK-E	STS-14
Baseline	77.19	78.14	91.1	87.85	80.29	83.0	72.87/81.38	0.80	78.49	0.54/0.55
LSTM	70.15	74.38	82.25	85.39	74.03	58.8	72.0/81.66	0.83	81.96	0.49/0.48
biLSTM	74.4	78.6	89.28	87.88	78.14	83.6	73.16/81.98	0.87	84.6	0.55/0.57
biLSTM(max-pool)	78.08	81.35	92.24	88.85	82	88.4	74.38/81.44	0.88	85.26	0.64/0.65

Table 1: SentEval scores on 10 downstream tasks of the 4 models.

Model	NLI		Transfer	
	dev	test	macro	micro
Baseline	64.6	65.07	81.1	82.73
LSTM	80.73	80.0	75.87	75.65
biLSTM	80.26	79.83	81.21	82.65
biLSTM(max-pool)	83.32	82.83	83.82	85.18

Table 2: Macro- and Micro- Accuracies on SentEval scores ('transfer task') of the 4 models along with training performance measures ('NLI task').

PERFORMANCE OVER SENTENCE LENGTHS

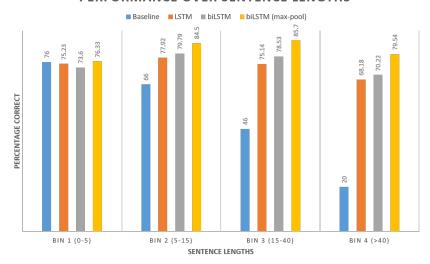


Figure 2: Performance of the models over different ranges of the sentence lengths. We can observe that both biLSTM models perform much better than others as the sentence length increases. For shorter sentences, performance of all the models is comparable.

A Appendix

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Premise: Excellent
Hypothesis: Pathetic
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Base = [9.6518, 2.8749, -12.3981] (Entailment) LSTM = [-1.7245, 0.5521, 1.1724] (Contradiction) biLSTM = [-4.9323, 1.2784, 3.6504] (Contradiction) biLSTM(max-pool) = [-15.0683, 1.5829, 13.4708] (Contradiction)

Premise: This burger is very good
Hypothesis: This burger is very bad

Base = [2.3129, 2.1854, -4.4383] (Entailment)
LSTM = [-0.0933, 0.3281, -0.2348] (Neutral)
biLSTM = [-2.6843, -0.1019, 2.7698] (Contradiction)
biLSTM(max-pool) = [-3.0693, 0.2361, 2.8359] (Contradiction)

Premise: I am a boy
Hypothesis: I am not a boy

Base = [2.8127, 0.7847, -3.5601] (Entailment) LSTM = [0.4299, -0.1724, -0.2575]] (Entailment) biLSTM = [0.4546, -0.3356, -0.1364] (Entailment) biLSTM(max-pool) = [-0.8157, -1.1039, 1.9276] (Contradiction)

Premise: The man is riding a bike wearing a blue helmet

Hypothesis: The woman is riding a bike wearing a blue helmet Base = [2.0862, -0.4986, -1.5574] (Entailment)
LSTM = [-0.5291, -0.5784, 1.1075] (Contradiction)
biLSTM = [-1.6187, -1.0998, 2.6845] (Contradiction)

biLSTM(max-pool) = [-7.2812, -1.5112, 8.7917]] (Contradiction)

Premise: A lady is in the park Hypothesis: A lady is in the house

Base = [0.9784, 0.1472, -1.0930] (Entailment)
LSTM = [-1.2440, -0.5133, 1.7574] (Contradiction)
biLSTM = [-3.6487, -1.1938, 4.8015] (Contradiction)

biLSTM(max-pool) = [-4.7834, -2.0677, 6.8549] (Contradiction)

Premise: I tried to make this a really long sentence but I am failing so bad that I have to write something that does not even make sense Hypothesis: This is a short sentence

Base = [1.9736, 1.4264, -3.3548] (Entailment) LSTM = [0.3392, 0.2548, -0.5939] (Entailment) biLSTM = [-4.6421, 1.4537, 3.1921] (Contradiction) biLSTM(max-pool) = [1.4613, 0.5478, -2.0038] (Contradiction)

Premise: I can tell you that I do not like fast cars and also I can say for sure that I do not like bikes

Hypothesis: I can tell you that I like fast cars and also I can say for sure that I do not like bikes

Base = [2.2253, 0.8985, -3.0780] (Entailment) LSTM = [1.6140, 0.0471, -1.6611] (Entailment) biLSTM = [0.9712, 0.2159, -1.1899] (Entailment) biLSTM(max-pool) = [0.5556, -0.0174, -0.5308] (Entailment)

Premise: I can tell you that I do not like fast cars and also I can say for sure that I do not like bikes

Hypothesis: I can tell you that I like fast cars and also I can say for sure that I like bikes

Base = [2.2253, 0.8985, -3.0780] (Entailment) LSTM = [1.1728, 0.0794, -1.2522] (Entailment) biLSTM = [-0.1934, 0.0876, 0.0975] (Contradiction) biLSTM(max-pool) = [-0.7368, -0.1455, 0.8880] (Contradiction)