

CS512: Advanced Machine Learning.
Assignment 3: Adversarial Training on Sequence Classification

Garima Gupta: ggupta22@uic.edu

Sai Teja Karnati: skarna3@uic.edu

Shubham Singh: ssing57@uic.edu

Wangfei Wang: wwang75@uic.edu

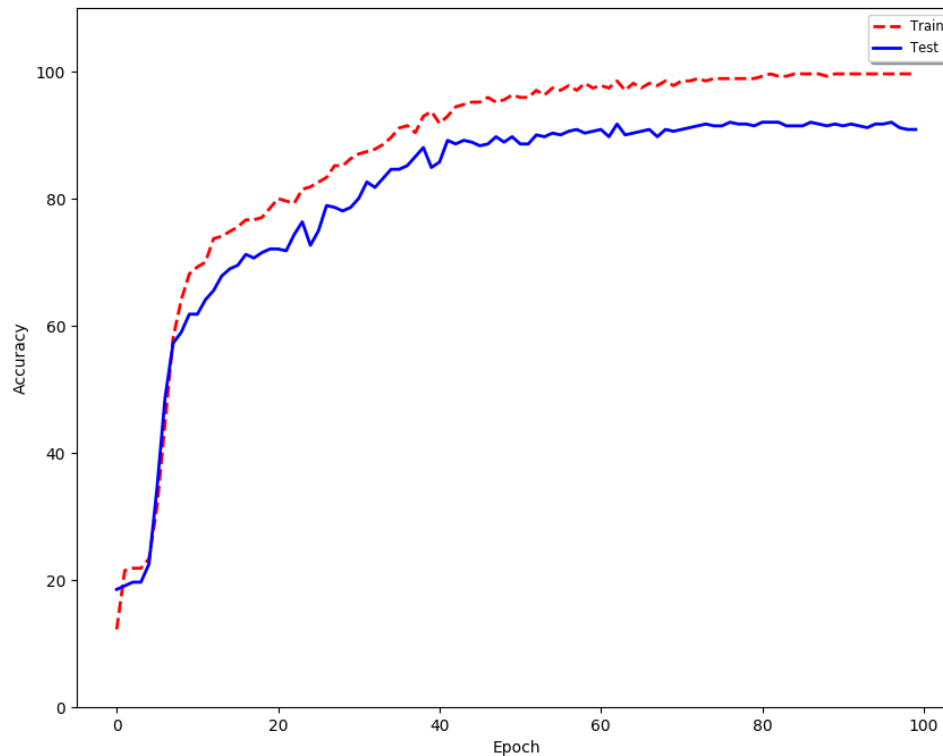
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1 Introduction

2 (15 points) Training the Basic Model

Hyperparameters values:

```
batch_size = 27, hidden_size = 10, basic_epoch = 100, out_channels = 64, kernel_size = 10, stride = 3, lr = 1e-3 (learning rate), weight_decay = 1e-3.
```



3 (10 points) Save and Load Pretrained Model

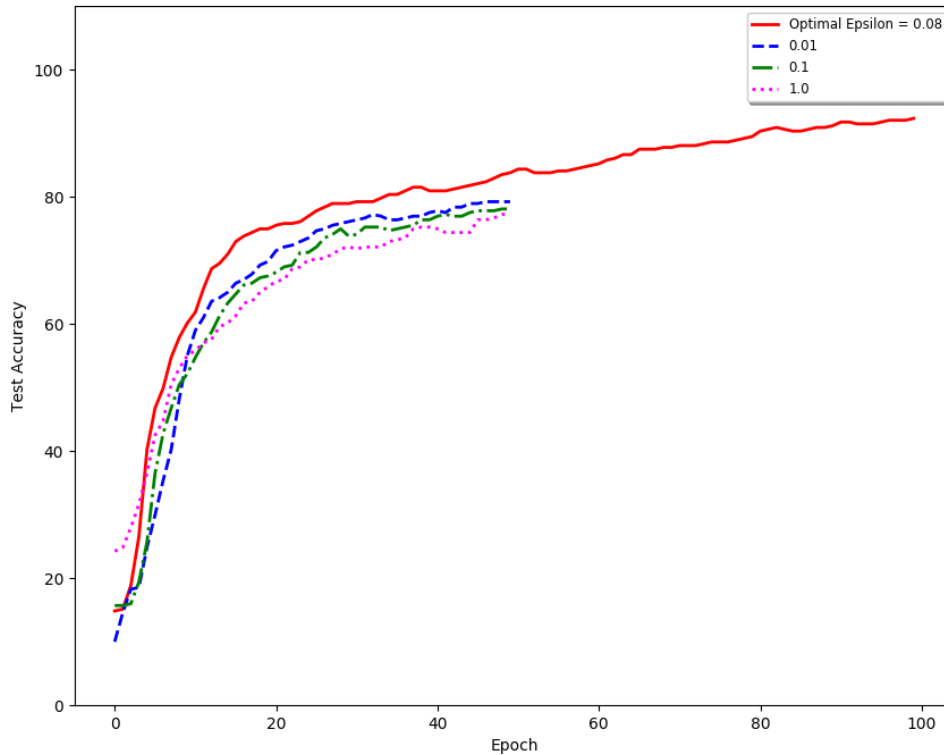
See code in `training.py` under the comment Part 3, Save and Load model.

4 (25 points) Adversarial Training as Regularization

- a (10 points) See the `compute_perturbation` function in `training.py`.
- b (5 points) See the branch `mode = 'AdvLSTM'` in `LSTMClassifier` in `Classifier.py`.
- c (10 points)

Among the ϵ 's we have tried ($\epsilon = [0, 2, 4, 6, 8, 10, 0.001, 0.01, 0.08, 0.1, 1, 10, 100, 1000]$), $\epsilon = 0.08$ gives the optimal performance at the end of 100 epochs. The other hyperparameters were set the same as those in the basic model. The adversarial training improved the test accuracy although this improvement is not significant. The basic model test accuracy is about 92.02%, while with adversarial training, the test accuracy reached 92.3%. The output of basic model test accuracy and test accuracy with adversarial training were stored in `BasicModel_test.txt`, and first 100 rows of `AdvModel_acc.txt` in folder `Figures`.

As shown in the figure, the performance of the model changes slightly with the change of ϵ between $[0.01, 0.1, 1]$, meaning our model is pretty robust to disturbance. At the end of epoch 50, $\epsilon = 0.01$ seems to give the best test accuracy among $\epsilon = [0.01, 0.1, 1]$.



5 (40 points) Adversarial Training as Proximal Mapping

- a (30 points) We have implemented the ProxLSTMCCell in `ProxLSTM.py`. We have implemented `forward` pass, but we are having issues with `backward` pass.
- b (10 points) We have written a branch in `LSTMClassifier` that can handle `mode = 'ProxLSTM'`. Because we haven't made `backward` pass work, we cannot perform the experiment with different $\epsilon = \lambda^{-1}\sigma^2 = [0.1, 1.0, 5.0]$. But we can imagine that by using the similar code in `plot.py`, we can generate curves similar to that in the previous section. We can also imagine that

by adding adversarial training as proximal mapping would further increase the test accuracy, especially when compared with basic model.

6 (10 points) Dropout and Batch Normalization

- a **(5 points)** We have implemented a dropout layer in `Classifier.py`.
- b **(5 points)** We have implemented a batch normalization layer in `Classifier.py`.