

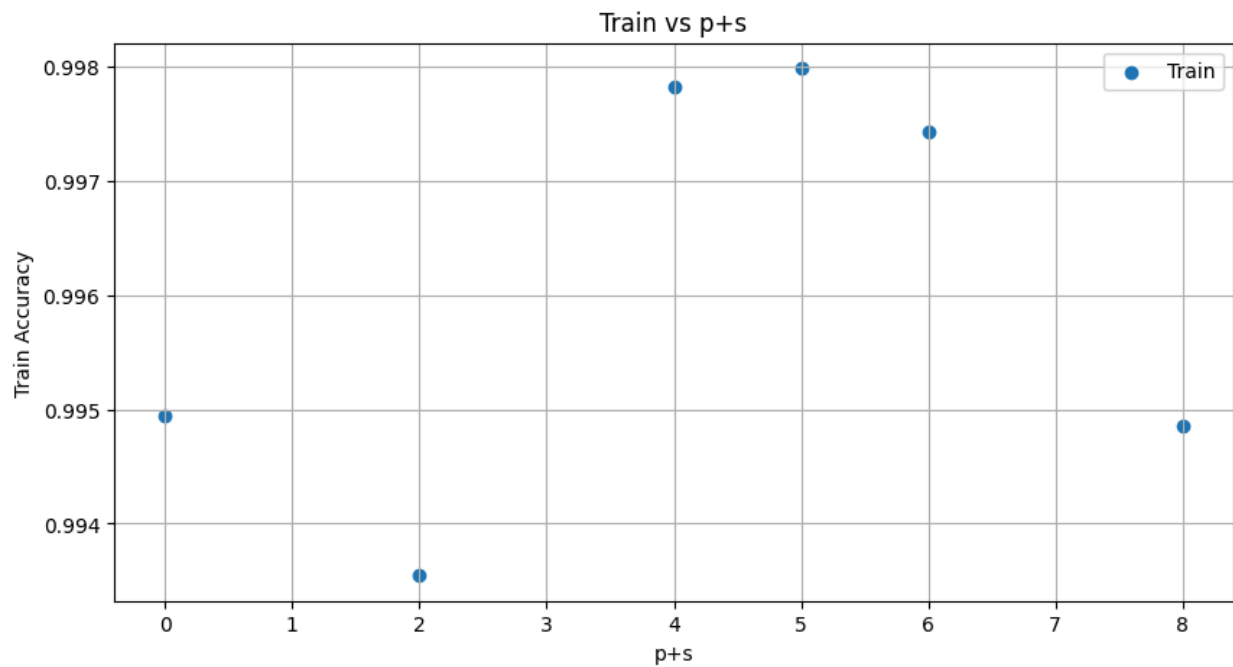
# FFNN

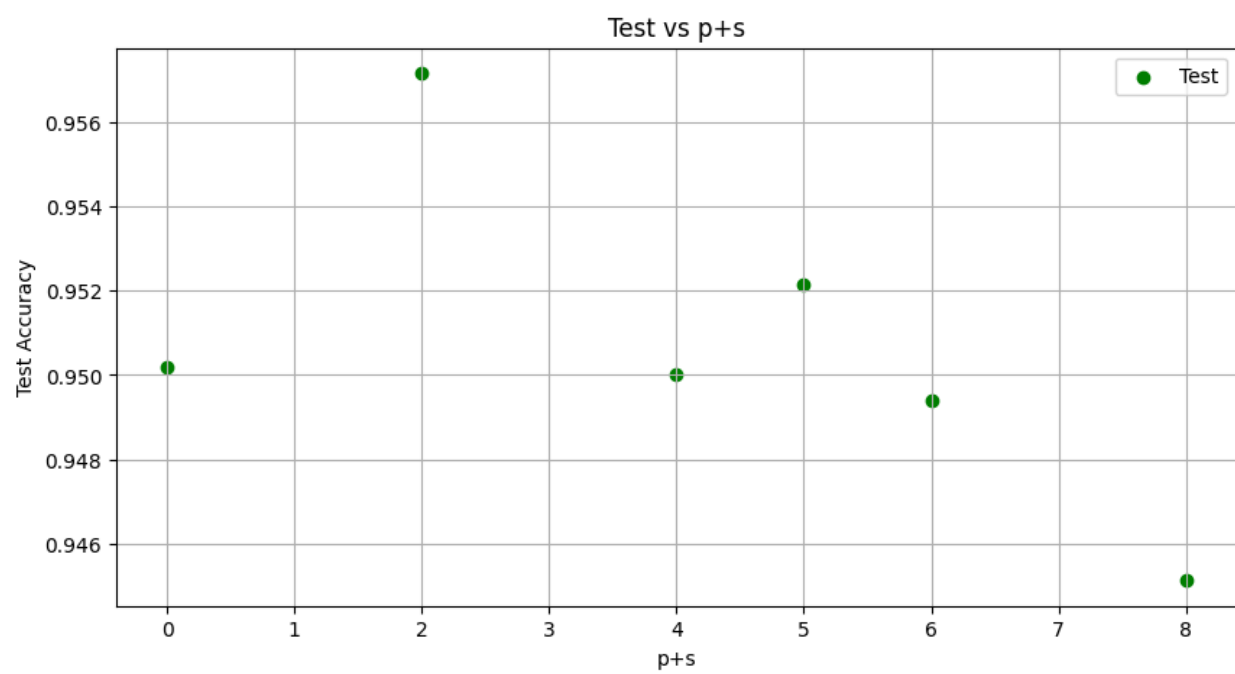
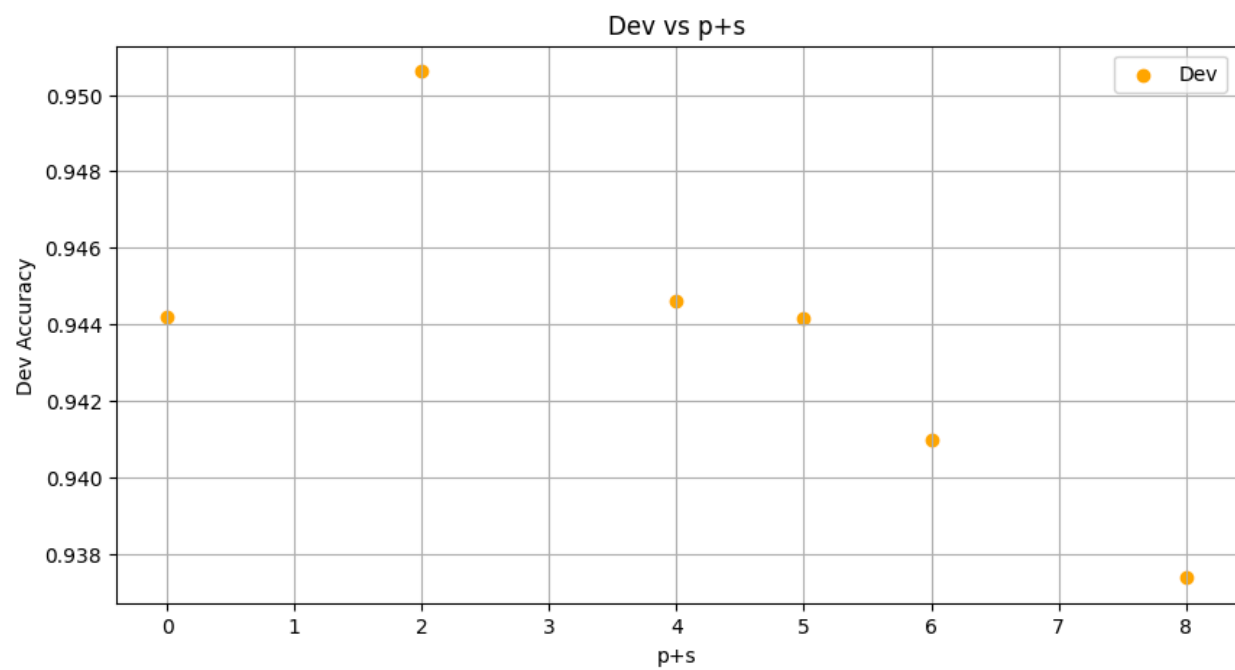
Epochs : 75

Chosen Model :  $p = s = 1$

{'PART': 0, 'OOV': 1, 'PRON': 2, 'PROPN': 3, 'NUM': 4, 'AUX': 5, 'VERB': 6, 'CCONJ': 7, 'ADJ': 8, 'NOUN': 9, 'ADV': 10, 'INTJ': 11, 'ADP': 12, 'DET': 13}

Index p+s	Train	Dev	Test
0	0.9949439934230809	0.9441990364348087	0.9501975683890578
2	0.9935463981091357	0.9506172839506173	0.9571428571428572
4	0.9978213955400267	0.9445950015055706	0.95
5	0.997985818518138	0.9441433303221921	0.9521276595744681
6	0.9974308909670127	0.9409816320385426	0.9493920972644377
8	0.9948617819340253	0.9373682625715146	0.9451367781155016





**p=0 s=0**

Training model...

Epoch 1/75, Loss: 0.8681635394124572  
Epoch 2/75, Loss: 0.35202915210193625  
Epoch 3/75, Loss: 0.2596902823499092  
Epoch 4/75, Loss: 0.20427245551747186  
Epoch 5/75, Loss: 0.16957314591487666  
Epoch 6/75, Loss: 0.14631193098899725  
Epoch 7/75, Loss: 0.1312463967565266  
Epoch 8/75, Loss: 0.1203126020260821  
Epoch 9/75, Loss: 0.11090921784008262  
Epoch 10/75, Loss: 0.10405567658258327  
Epoch 11/75, Loss: 0.0965028539062545  
Epoch 12/75, Loss: 0.0925034311991344  
Epoch 13/75, Loss: 0.08686759843637569  
Epoch 14/75, Loss: 0.08153747835024609  
Epoch 15/75, Loss: 0.07758819546454748  
Epoch 16/75, Loss: 0.07490421935380089  
Epoch 17/75, Loss: 0.07130005654339518  
Epoch 18/75, Loss: 0.06734352853627916  
Epoch 19/75, Loss: 0.06508594227702295  
Epoch 20/75, Loss: 0.06297782451222937  
Epoch 21/75, Loss: 0.06152307610673245  
Epoch 22/75, Loss: 0.056311103722366695  
Epoch 23/75, Loss: 0.05814633374795016  
Epoch 24/75, Loss: 0.05369755152768187  
Epoch 25/75, Loss: 0.05355855770064891  
Epoch 26/75, Loss: 0.053604977917400376  
Epoch 27/75, Loss: 0.05056259590245443  
Epoch 28/75, Loss: 0.051898523207232095  
Epoch 29/75, Loss: 0.04586906031170336  
Epoch 30/75, Loss: 0.05332299097200227  
Epoch 31/75, Loss: 0.042941992431107086  
Epoch 32/75, Loss: 0.04688231437844461  
Epoch 33/75, Loss: 0.042564824483232296  
Epoch 34/75, Loss: 0.042139033942730504  
Epoch 35/75, Loss: 0.04402194240849628  
Epoch 36/75, Loss: 0.04105270667875729  
Epoch 37/75, Loss: 0.04607790806931001  
Epoch 38/75, Loss: 0.037554445278483915  
Epoch 39/75, Loss: 0.03642117567689612  
Epoch 40/75, Loss: 0.040192398745865915  
Epoch 41/75, Loss: 0.035657210432148555  
Epoch 42/75, Loss: 0.03820701422792562

Epoch 43/75, Loss: 0.041372663328310494  
Epoch 44/75, Loss: 0.0318424846931743  
Epoch 45/75, Loss: 0.03427693372870227  
Epoch 46/75, Loss: 0.040103445995351  
Epoch 47/75, Loss: 0.03020267218690993  
Epoch 48/75, Loss: 0.032885601295572525  
Epoch 49/75, Loss: 0.029611401603815268  
Epoch 50/75, Loss: 0.03190327035162067  
Epoch 51/75, Loss: 0.028068236914333012  
Epoch 52/75, Loss: 0.03712659505356169  
Epoch 53/75, Loss: 0.026550498065685817  
Epoch 54/75, Loss: 0.03380538917438229  
Epoch 55/75, Loss: 0.025435179396556203  
Epoch 56/75, Loss: 0.042681078396663866  
Epoch 57/75, Loss: 0.029647464004336852  
Epoch 58/75, Loss: 0.030857799802173534  
Epoch 59/75, Loss: 0.024247071703986833  
Epoch 60/75, Loss: 0.033147514913023646  
Epoch 61/75, Loss: 0.02347862614205847  
Epoch 62/75, Loss: 0.023366906724897305  
Epoch 63/75, Loss: 0.026049193323242445  
Epoch 64/75, Loss: 0.023133217696116905  
Epoch 65/75, Loss: 0.03149266785512778  
Epoch 66/75, Loss: 0.027065620794480966  
Epoch 67/75, Loss: 0.023667975118439648  
Epoch 68/75, Loss: 0.022158168461175135  
Epoch 69/75, Loss: 0.025709881600115978  
Epoch 70/75, Loss: 0.02000249942470628  
Epoch 71/75, Loss: 0.02035523636468881  
Epoch 72/75, Loss: 0.02955523210419542  
Epoch 73/75, Loss: 0.019444085144663947  
Epoch 74/75, Loss: 0.019488982673491546  
Epoch 75/75, Loss: 0.028258233310696733  
Saving model...

Loading model...

**Metrics of train set:**

**sk\_accuracy\_train=0.9949439934230809**

classification\_rep\_train =

	precision	recall	f1-score	support
ADJ	1.00	0.98	0.99	1577
ADP	1.00	1.00	1.00	11098
ADV	0.93	0.95	0.94	371
AUX	1.00	0.99	1.00	1573

CCONJ	1.00	1.00	1.00	742
DET	1.00	1.00	1.00	4511
INTJ	1.00	0.99	0.99	335
NOUN	0.99	0.99	0.99	8786
NUM	0.89	0.96	0.93	785
PRON	1.00	1.00	1.00	2435
PROPN	1.00	1.00	1.00	11403
VERB	1.00	0.99	0.99	5039
<b>accuracy</b>			<b>0.99</b>	<b>48655</b>
<b>macro avg</b>	<b>0.98</b>	<b>0.99</b>	<b>0.99</b>	<b>48655</b>
<b>weighted avg</b>	<b>1.00</b>	<b>0.99</b>	<b>0.99</b>	<b>48655</b>

recall\_micro\_train=0.9949439934230809

recall\_macro\_train=0.9876571029441347

f1\_micro\_train=0.9949439934230809

f1\_macro\_train=0.985654847374831

confusion\_mat\_train =

```
[[ 1549      0      1      0      0      0      0      5      8      0     10
 4]
 [      0 11091      2      0      0      0      0      0      2      0      1
 2]
 [      0      0   351      0      0      0      0      2     11      0      2
 5]
 [      0      0      5  1565      0      0      0      2      1      0      0
 0]
 [      0      0      0      0   742      0      0      0      0      0      0
 0]
 [      0      0      0      0      0  4510      0      1      0      0      0
 0]
 [      0      0      0      0      0      0   331      0      1      0      0
 3]
 [      1      0      3      0      0      0      0  8741     23      0     12
 6]
 [      0      0      2      0      0      0      0      14    757      0      8
 4]
 [      0      0      0      0      0      0      0      5      0   2428      2
 0]
 [      1      1      4      0      0      0      0      9     25      0  11362
 1]
 [      0      0      9      0      0      0      0     21     20      0      7
 4982]]
```

**Metrics of dev set:**

**sk\_accuracy\_dev=0.9441990364348087**

```

classification_rep_dev =
      precision    recall  f1-score   support

   ADJ           1.00      0.94      0.97         223
   ADP           0.97      1.00      0.99        1456
   ADV           0.85      0.92      0.89          51
   AUX           0.98      0.77      0.86         296
  CCONJ           1.00      0.99      1.00         107
   DET           1.00      0.94      0.97         697
  INTJ           1.00      1.00      1.00          35
  NOUN           0.93      0.96      0.94        1130
   NUM           0.89      0.82      0.86         131
  PRON           0.95      1.00      0.98         314
  PROPN           0.98      0.96      0.97        1565
  VERB           0.89      0.99      0.94         637

 accuracy                0.96      6642
 macro avg      0.95      0.94      0.95      6642
 weighted avg    0.96      0.96      0.96      6642

```

**recall\_micro\_dev=0.9591990364348087**

**recall\_macro\_dev=0.9405929340283952**

**f1\_micro\_dev=0.9591990364348087**

**f1\_macro\_dev=0.9458223402421592**

**confusion\_mat\_dev =**

```

[[ 209    0    6    0    0    0    0    4    1    0    3    0]
 [   0 1455    0    0    0    0    0    1    0    0    0    0]
 [   0    0   47    0    0    0    0    1    1    0    1    1]
 [   0    0    0  228    0    0    0   17    0    0    0   51]
 [   0    0    0    0  106    1    0    0    0    0    0    0]
 [   0   42    0    0    0   654    0    1    0    0    0    0]
 [   0    0    0    0    0    0   35    0    0    0    0    0]
 [   1    0    0    4    0    0    0 1086    3    0   17   19]
 [   0    0    1    0    0    0    0    0  108   15    5    2]
 [   0    0    0    0    0    0    0    0    0  313    1    0]
 [   0    0    1    0    0    0    0   57    6    0 1500    1]
 [   0    0    0    0    0    0    0    2    2    0    3  630]]

```

**Metrics of test set:**

**sk\_accuracy\_test=0.9501975683890578**

**classification\_rep\_test =**

```

      precision    recall  f1-score   support

   ADJ           0.90      0.94      0.92         218
   ADP           1.00      1.00      1.00        1491

```

ADV	0.86	0.64	0.74	76
AUX	0.97	0.91	0.94	239
CCONJ	1.00	0.98	0.99	109
DET	1.00	1.00	1.00	618
INTJ	1.00	1.00	1.00	36
NOUN	0.96	0.93	0.94	1240
NUM	0.83	0.80	0.81	128
PRON	0.95	1.00	0.97	286
PROPN	0.98	0.99	0.98	1516
VERB	0.89	0.96	0.93	623
<b>accuracy</b>			<b>0.97</b>	<b>6580</b>
<b>macro avg</b>	<b>0.95</b>	<b>0.93</b>	<b>0.94</b>	<b>6580</b>
<b>weighted avg</b>	<b>0.97</b>	<b>0.97</b>	<b>0.96</b>	<b>6580</b>

recall\_micro\_test=0.9651975683890578

recall\_macro\_test=0.9286033754720951

f1\_micro\_test=0.9651975683890578

f1\_macro\_test=0.9352826361549704

confusion\_mat\_test =

```
[[ 204    0    3    0    0    0    0    4    2    0    3    2]
 [   1 1489    0    0    0    0    0    0    0    0    0    1]
 [   20    0   49    0    0    0    0    1    1    0    4    1]
 [    0    0    0  218    0    0    0   21    0    0    0    0]
 [    0    0    0    0  107    2    0    0    0    0    0    0]
 [    0    0    0    0    0  616    0    2    0    0    0    0]
 [    0    0    0    0    0    0   36    0    0    0    0    0]
 [    0    0    0    6    0    0    0 1148    4    0   20   62]
 [    0    0    0    0    0    0    0    1  102   15    6    4]
 [    0    0    0    0    0    0    0    0    0  286    0    0]
 [    1    2    2    0    0    0    0    4   11    0 1495    1]
 [    0    0    3    0    0    0    0   16    3    0    0 601]]
```

**p=1 s=1**

Training model...

Epoch 1/75, Loss: 0.6653759327476034  
Epoch 2/75, Loss: 0.24612780980540638  
Epoch 3/75, Loss: 0.17956529387349995  
Epoch 4/75, Loss: 0.14623236873905504  
Epoch 5/75, Loss: 0.12734507840755793  
Epoch 6/75, Loss: 0.11387971372364404  
Epoch 7/75, Loss: 0.10279513666383021  
Epoch 8/75, Loss: 0.09631142775563624  
Epoch 9/75, Loss: 0.09051683592474218  
Epoch 10/75, Loss: 0.08663826996357145  
Epoch 11/75, Loss: 0.08059836413600248  
Epoch 12/75, Loss: 0.07890300723828014  
Epoch 13/75, Loss: 0.07569907029430442  
Epoch 14/75, Loss: 0.07206495610263404  
Epoch 15/75, Loss: 0.06965854867810013  
Epoch 16/75, Loss: 0.06654168398959887  
Epoch 17/75, Loss: 0.06776103299773441  
Epoch 18/75, Loss: 0.06090503466500975  
Epoch 19/75, Loss: 0.05945304968503259  
Epoch 20/75, Loss: 0.06136599870236543  
Epoch 21/75, Loss: 0.05752311471118298  
Epoch 22/75, Loss: 0.05470718825074573  
Epoch 23/75, Loss: 0.054227284087902596  
Epoch 24/75, Loss: 0.052194644925538686  
Epoch 25/75, Loss: 0.05035528555101992  
Epoch 26/75, Loss: 0.04776201048969097  
Epoch 27/75, Loss: 0.04898391138055789  
Epoch 28/75, Loss: 0.044625931757730726  
Epoch 29/75, Loss: 0.04492582787284277  
Epoch 30/75, Loss: 0.050213452582072396  
Epoch 31/75, Loss: 0.03999883194933826  
Epoch 32/75, Loss: 0.04095282671103168  
Epoch 33/75, Loss: 0.04284383164939349  
Epoch 34/75, Loss: 0.040742786326788495  
Epoch 35/75, Loss: 0.039360890332946416  
Epoch 36/75, Loss: 0.03997906423840919  
Epoch 37/75, Loss: 0.03551999892621987  
Epoch 38/75, Loss: 0.03639441412826951  
Epoch 39/75, Loss: 0.03632051697353788  
Epoch 40/75, Loss: 0.03595019084450405  
Epoch 41/75, Loss: 0.03698284534946648  
Epoch 42/75, Loss: 0.03526530713839465



Epoch 43/75, Loss: 0.030376939602445554  
Epoch 44/75, Loss: 0.030526801138143565  
Epoch 45/75, Loss: 0.03196187283669271  
Epoch 46/75, Loss: 0.029979200099296607  
Epoch 47/75, Loss: 0.03327043062509442  
Epoch 48/75, Loss: 0.02913917906094072  
Epoch 49/75, Loss: 0.0320390529950514  
Epoch 50/75, Loss: 0.027266101460495067  
Epoch 51/75, Loss: 0.031791719381291314  
Epoch 52/75, Loss: 0.02767385300836391  
Epoch 53/75, Loss: 0.02640724379135732  
Epoch 54/75, Loss: 0.026867136746667443  
Epoch 55/75, Loss: 0.02534136137651716  
Epoch 56/75, Loss: 0.02687800109666896  
Epoch 57/75, Loss: 0.027402799325904812  
Epoch 58/75, Loss: 0.026984180226998864  
Epoch 59/75, Loss: 0.025604903653692855  
Epoch 60/75, Loss: 0.026800537322496185  
Epoch 61/75, Loss: 0.027375409019695457  
Epoch 62/75, Loss: 0.020173560801264742  
Epoch 63/75, Loss: 0.024542906810541356  
Epoch 64/75, Loss: 0.02414080894774021  
Epoch 65/75, Loss: 0.024334698399322112  
Epoch 66/75, Loss: 0.026745803086403605  
Epoch 67/75, Loss: 0.018100517205771667  
Epoch 68/75, Loss: 0.019744566950996467  
Epoch 69/75, Loss: 0.026176343799496044  
Epoch 70/75, Loss: 0.02663622331811998  
Epoch 71/75, Loss: 0.017239858343085206  
Epoch 72/75, Loss: 0.025306915151201013  
Epoch 73/75, Loss: 0.01596333922314148  
Epoch 74/75, Loss: 0.02176945554171438  
Epoch 75/75, Loss: 0.029869008724830497  
Saving model...

Loading model...

**Metrics of train set:**

**sk\_accuracy\_train=0.9935463981091357**

classification\_rep\_train =

	precision	recall	f1-score	support
ADJ	0.97	0.98	0.98	1577
ADP	1.00	1.00	1.00	11098
ADV	0.95	0.93	0.94	371
AUX	1.00	1.00	1.00	1573

CCONJ	1.00	1.00	1.00	742
DET	1.00	1.00	1.00	4511
INTJ	1.00	0.98	0.99	335
NOUN	0.98	0.99	0.99	8786
NUM	0.95	0.98	0.96	785
PRON	1.00	1.00	1.00	2435
PROPN	1.00	1.00	1.00	11403
VERB	1.00	0.98	0.99	5039
<b>accuracy</b>			<b>0.99</b>	<b>48655</b>
<b>macro avg</b>	<b>0.99</b>	<b>0.99</b>	<b>0.99</b>	<b>48655</b>
<b>weighted avg</b>	<b>0.99</b>	<b>0.99</b>	<b>0.99</b>	<b>48655</b>

recall\_micro\_train=0.9935463981091357

recall\_macro\_train=0.9854504974831704

f1\_micro\_train=0.9935463981091357

f1\_macro\_train=0.9863229343901038

confusion\_mat\_train =

```
[[ 1544      2      0      0      0      0      0      15      6      0      10
  0]
 [      3 11090      2      0      0      0      0      1      1      0      1
  0]
 [      1      1   345      0      0      0      0      8      6      1      8
  1]
 [      3      0      0  1567      0      0      0      1      2      0      0
  0]
 [      0      0      0      0   742      0      0      0      0      0      0
  0]
 [      7      2      1      0      0  4490      0      8      0      0      2
  1]
 [      0      0      3      0      0      0   328      0      0      0      0
  4]
 [     17      1      0      0      0      0      0  8736     13      0     18
  1]
 [      2      0      1      0      0      0      0     10    767      0      4
  1]
 [      1      0      0      0      0      0      0      1      0  2433      0
  0]
 [      4      0      0      0      0      0      0     32      3      0 11364
  0]
 [      6      9     12      0      0      0      0     59      9      0      9
 4935]]
```

**Metrics of dev set:**

**sk\_accuracy\_dev=0.9506172839506173**

```

classification_rep_dev =
      precision    recall  f1-score   support

   ADJ           0.94      0.93      0.93        223
   ADP           0.97      1.00      0.98       1456
   ADV           0.77      0.80      0.79         51
   AUX           0.97      0.76      0.85        296
  CCONJ          1.00      0.99      1.00        107
   DET           1.00      0.93      0.96        697
  INTJ           1.00      0.97      0.99         35
  NOUN           0.91      0.95      0.93       1130
   NUM           0.90      0.85      0.87        131
  PRON           0.95      0.99      0.97        314
  PROPN          0.98      0.95      0.97       1565
  VERB           0.89      0.96      0.92        637

 accuracy                0.95      6642
 macro avg      0.94      0.92      0.93      6642
 weighted avg   0.95      0.95      0.95      6642

```

**recall\_micro\_dev=0.9506172839506173**

**recall\_macro\_dev=0.9238191977339062**

**f1\_micro\_dev=0.9506172839506173**

**f1\_macro\_dev=0.9298615697541335**

```

confusion_mat_dev =
[[ 207    1    7    0    0    0    0    5    1    0    1    1]
 [   1 1449    1    1    0    0    0    3    0    0    1    0]
 [   2    0   41    1    0    0    0    5    0    0    1    1]
 [   1    1    0  224    0    0    0   17    0    0    2   51]
 [   0    1    0    0  106    0    0    0    0    0    0    0]
 [   2   44    0    0    0  648    0    1    1    0    0    1]
 [   0    0    0    0    0    0   34    0    0    0    0    1]
 [   4    0    3    4    0    0    0 1074    3    0   22   20]
 [   1    0    0    0    0    0    0    2  111   15    2    0]
 [   0    0    0    0    0    0    0    1    0  312    1    0]
 [   2    1    0    0    0    0    0   61    5    0 1494    2]
 [   1    3    1    2    0    0    0   10    3    1    2  614]]

```

**Metrics of test set:**

**sk\_accuracy\_test=0.9571428571428572**

```

classification_rep_test =
      precision    recall  f1-score   support

   ADJ           0.85      0.94      0.89        218
   ADP           0.99      1.00      0.99       1491

```

ADV	0.85	0.59	0.70	76
AUX	0.96	0.91	0.93	239
CCONJ	0.99	0.97	0.98	109
DET	1.00	0.98	0.99	618
INTJ	0.97	1.00	0.99	36
NOUN	0.94	0.91	0.93	1240
NUM	0.87	0.80	0.83	128
PRON	0.95	0.99	0.97	286
PROPN	0.97	0.99	0.98	1516
VERB	0.90	0.94	0.92	623
<b>accuracy</b>			<b>0.96</b>	<b>6580</b>
<b>macro avg</b>	<b>0.94</b>	<b>0.92</b>	<b>0.93</b>	<b>6580</b>
<b>weighted avg</b>	<b>0.96</b>	<b>0.96</b>	<b>0.96</b>	<b>6580</b>

recall\_micro\_test=0.9571428571428572

recall\_macro\_test=0.9186666817258099

f1\_micro\_test=0.9571428571428572

f1\_macro\_test=0.9250386356217705

confusion\_mat\_test =

```
[[ 204    1    4    0    0    0    0    5    4    0    0    0]
 [   1 1484    0    1    0    1    0    1    0    0    1    2]
 [  21    0   45    1    0    0    0    4    1    0    4    0]
 [   0    0    0  218    0    0    0   21    0    0    0    0]
 [   0    0    0    0  106    0    0    1    0    0    2    0]
 [   2    0    0    1    0  607    0    5    0    0    3    0]
 [   0    0    0    0    0    0   36    0    0    0    0    0]
 [   4    1    0    7    1    0    0 1131    2    0   29   65]
 [   1    0    2    0    0    0    0    2  103   16    4    0]
 [   0    0    0    0    0    0    0    2    0  284    0    0]
 [   1    4    1    0    0    0    0    8    6    0 1496    0]
 [   5    2    1    0    0    2    1   18    3    0    7  584]]
```

**p=2 s=2**

Training model...

Epoch 1/75, Loss: 0.6669412596976083  
Epoch 2/75, Loss: 0.23950119581166532  
Epoch 3/75, Loss: 0.17561681083135514  
Epoch 4/75, Loss: 0.14639532106283576  
Epoch 5/75, Loss: 0.1245593965951838  
Epoch 6/75, Loss: 0.11139748842790771  
Epoch 7/75, Loss: 0.0986522575886247  
Epoch 8/75, Loss: 0.09120297909271567  
Epoch 9/75, Loss: 0.08413295384341025  
Epoch 10/75, Loss: 0.07834551974065705  
Epoch 11/75, Loss: 0.07611020553505854  
Epoch 12/75, Loss: 0.06790540271356663  
Epoch 13/75, Loss: 0.06560826117815187  
Epoch 14/75, Loss: 0.06278679532990375  
Epoch 15/75, Loss: 0.05956481083791942  
Epoch 16/75, Loss: 0.055731257978532046  
Epoch 17/75, Loss: 0.05453096036832907  
Epoch 18/75, Loss: 0.0532482283777859  
Epoch 19/75, Loss: 0.04892798243376577  
Epoch 20/75, Loss: 0.047698379530514874  
Epoch 21/75, Loss: 0.04262518438098062  
Epoch 22/75, Loss: 0.04187127529498071  
Epoch 23/75, Loss: 0.043483608516886624  
Epoch 24/75, Loss: 0.03786127808232445  
Epoch 25/75, Loss: 0.03751578032158504  
Epoch 26/75, Loss: 0.039012994108956034  
Epoch 27/75, Loss: 0.032980726797286995  
Epoch 28/75, Loss: 0.036319219190098805  
Epoch 29/75, Loss: 0.030915478442877762  
Epoch 30/75, Loss: 0.030771825520483825  
Epoch 31/75, Loss: 0.03091167194502119  
Epoch 32/75, Loss: 0.027442443065213575  
Epoch 33/75, Loss: 0.0267454694849118  
Epoch 34/75, Loss: 0.03180381721686755  
Epoch 35/75, Loss: 0.025704050450629526  
Epoch 36/75, Loss: 0.02538777613393421  
Epoch 37/75, Loss: 0.029070953044625574  
Epoch 38/75, Loss: 0.022571083300139683  
Epoch 39/75, Loss: 0.026267587507243402  
Epoch 40/75, Loss: 0.020667055550925115  
Epoch 41/75, Loss: 0.028094953834608468  
Epoch 42/75, Loss: 0.021779593421269075

Epoch 43/75, Loss: 0.021487854677080743  
Epoch 44/75, Loss: 0.02135819606964459  
Epoch 45/75, Loss: 0.023956931393654166  
Epoch 46/75, Loss: 0.02171284635889645  
Epoch 47/75, Loss: 0.021565944876335286  
Epoch 48/75, Loss: 0.021704151510148078  
Epoch 49/75, Loss: 0.01769023550826999  
Epoch 50/75, Loss: 0.015553645015377409  
Epoch 51/75, Loss: 0.02123445989517687  
Epoch 52/75, Loss: 0.018470431407131723  
Epoch 53/75, Loss: 0.023528578550442363  
Epoch 54/75, Loss: 0.017549729169839212  
Epoch 55/75, Loss: 0.020739572657214662  
Epoch 56/75, Loss: 0.019251615818965527  
Epoch 57/75, Loss: 0.014290918878349962  
Epoch 58/75, Loss: 0.02569019970423699  
Epoch 59/75, Loss: 0.013361575072304229  
Epoch 60/75, Loss: 0.017239952922943027  
Epoch 61/75, Loss: 0.015290537125674512  
Epoch 62/75, Loss: 0.018410970048088858  
Epoch 63/75, Loss: 0.01790050916941143  
Epoch 64/75, Loss: 0.010703432608934253  
Epoch 65/75, Loss: 0.021755999071883646  
Epoch 66/75, Loss: 0.01278015168899959  
Epoch 67/75, Loss: 0.021185490611762146  
Epoch 68/75, Loss: 0.010793651018968667  
Epoch 69/75, Loss: 0.014587920354463782  
Epoch 70/75, Loss: 0.015520555621741814  
Epoch 71/75, Loss: 0.016898193668220072  
Epoch 72/75, Loss: 0.012560404137468686  
Epoch 73/75, Loss: 0.013122266575809612  
Epoch 74/75, Loss: 0.016777434418168237  
Epoch 75/75, Loss: 0.013163483590862165  
Saving model...

Loading model...

**Metrics of train set:**

**sk\_accuracy\_train=0.9978213955400267**

classification\_rep\_train =

	precision	recall	f1-score	support
ADJ	0.99	0.99	0.99	1577
ADP	1.00	1.00	1.00	11098
ADV	0.98	0.98	0.98	371
AUX	1.00	1.00	1.00	1573

CCONJ	1.00	1.00	1.00	742
DET	1.00	1.00	1.00	4511
INTJ	1.00	0.98	0.99	335
NOUN	1.00	0.99	1.00	8786
NUM	0.98	1.00	0.99	785
PRON	1.00	1.00	1.00	2435
PROPN	1.00	1.00	1.00	11403
VERB	1.00	1.00	1.00	5039
<b>accuracy</b>			<b>1.00</b>	<b>48655</b>
<b>macro avg</b>	<b>1.00</b>	<b>1.00</b>	<b>1.00</b>	<b>48655</b>
<b>weighted avg</b>	<b>1.00</b>	<b>1.00</b>	<b>1.00</b>	<b>48655</b>

recall\_micro\_train=0.9978213955400267

recall\_macro\_train=0.9952450385389895

f1\_micro\_train=0.9978213955400267

f1\_macro\_train=0.9952303008786357

confusion\_mat\_train =

```
[[ 1561      0      0      0      0      1      0      3      2      0      3
 7]
 [      0 11091      0      0      0      0      0      2      0      0      4
 1]
 [      1      0   365      0      0      0      0      0      1      0      4
 0]
 [      0      0      0  1573      0      0      0      0      0      0      0
 0]
 [      0      0      0      0   742      0      0      0      0      0      0
 0]
 [      0      0      0      0      0  4510      0      0      0      0      1
 0]
 [      0      0      6      0      0      0   329      0      0      0      0
 0]
 [    11      0      0      2      0      1      0  8737     10      0     23
 2]
 [      0      0      0      0      0      0      0      1   783      0      0
 1]
 [      2      0      0      0      0      0      0      1      0  2432      0
 0]
 [      0      0      0      0      0      0      0      3      2      0 11398
 0]
 [      1      0      0      0      0      1      0      1      3      0      5
 5028]]
```

Metrics of dev set:

sk\_accuracy\_dev=0.9445950015055706

```

classification_rep_dev =
      precision    recall  f1-score   support

   ADJ           0.94      0.88      0.91       223
   ADP           0.97      0.99      0.98      1456
   ADV           0.76      0.73      0.74        51
   AUX           0.96      0.75      0.84       296
  CCONJ          1.00      0.99      1.00       107
   DET           0.99      0.93      0.96       697
  INTJ           1.00      0.91      0.96        35
  NOUN           0.91      0.94      0.92      1130
   NUM           0.87      0.83      0.85       131
  PRON           0.95      0.99      0.97       314
  PROPN          0.97      0.95      0.96      1565
  VERB           0.87      0.97      0.92       637

 accuracy                0.94      6642
 macro avg      0.93      0.91      0.92      6642
 weighted avg   0.95      0.94      0.94      6642

```

**recall\_micro\_dev=0.9445950015055706**

**recall\_macro\_dev=0.9055198059501973**

**f1\_micro\_dev=0.9445950015055706**

**f1\_macro\_dev=0.917055619794931**

confusion\_mat\_dev =

```

[[ 196    1    3    0    0    1    0   12    5    0    2    3]
 [   0 1439    1    1    0    2    0    3    0    0    4    6]
 [   0    1   37    1    0    1    0    5    0    0    4    2]
 [   0    0    1  222    0    0    0   17    0    0    2  54]
 [   0    0    0    0  106    0    0    0    0    0    0    1]
 [   1   42    0    1    0  649    0    1    0    0    2    1]
 [   0    0    2    0    0    0   32    0    0    0    0    1]
 [   8    1    1    4    0    1    0  1063    4    0   26   22]
 [   0    0    0    0    0    0    0    3   109   15    4    0]
 [   0    0    0    1    0    0    0    0    0  312    1    0]
 [   1    1    1    0    0    1    0   62    4    0 1493    2]
 [   2    1    3    2    0    1    0    4    3    0    5  616]]

```

**Metrics of test set:**

**sk\_accuracy\_test=0.95**

classification\_rep\_test =

```

      precision    recall  f1-score   support

   ADJ           0.84      0.88      0.86       218
   ADP           1.00      0.99      0.99      1491

```



ADV	0.80	0.53	0.63	76
AUX	0.95	0.90	0.93	239
CCONJ	0.99	0.96	0.98	109
DET	0.99	0.99	0.99	618
INTJ	0.97	0.97	0.97	36
NOUN	0.94	0.90	0.92	1240
NUM	0.79	0.77	0.78	128
PRON	0.94	1.00	0.97	286
PROPN	0.96	0.98	0.97	1516
VERB	0.87	0.94	0.90	623
<b>accuracy</b>			<b>0.95</b>	<b>6580</b>
<b>macro avg</b>	<b>0.92</b>	<b>0.90</b>	<b>0.91</b>	<b>6580</b>
<b>weighted avg</b>	<b>0.95</b>	<b>0.95</b>	<b>0.95</b>	<b>6580</b>

recall\_micro\_test=0.95

recall\_macro\_test=0.9011578223882216

f1\_micro\_test=0.9500000000000001

f1\_macro\_test=0.9083186329604063

confusion\_mat\_test =

```
[[ 192    0    2    0    0    0    0    11    4    0    5    4]
 [   2 1476    2    2    1    2    0    2    1    0    2    1]
 [   19    0   40    2    0    0    1    3    0    0    5    6]
 [    0    0    0  216    0    0    0  21    0    0    2    0]
 [    1    0    0    0  105    1    0    0    0    0    2    0]
 [    1    0    1    0    0  609    0    3    0    0    3    1]
 [    0    0    0    0    0    0   35    0    0    0    0    1]
 [    5    0    1    7    0    0    0 1118    8    0   34   67]
 [    3    0    2    0    0    0    0    4   99   15    3    2]
 [    0    0    0    0    0    1    0    0    0  285    0    0]
 [    2    0    1    0    0    1    0    6    9    1 1493    3]
 [    3    3    1    1    0    2    0   17    4    1    8  583]]
```

**p=2 s=3**

Training model...

Epoch 1/75, Loss: 0.6066811263619992  
Epoch 2/75, Loss: 0.22933205328734882  
Epoch 3/75, Loss: 0.17128477451170007  
Epoch 4/75, Loss: 0.13629795124316185  
Epoch 5/75, Loss: 0.11856527653315965  
Epoch 6/75, Loss: 0.10581943224173201  
Epoch 7/75, Loss: 0.09537249560231148  
Epoch 8/75, Loss: 0.08646029437575171  
Epoch 9/75, Loss: 0.08122710557272059  
Epoch 10/75, Loss: 0.07334742390944742  
Epoch 11/75, Loss: 0.0698228381132357  
Epoch 12/75, Loss: 0.06510573317836504  
Epoch 13/75, Loss: 0.06014025303823983  
Epoch 14/75, Loss: 0.058310281188391225  
Epoch 15/75, Loss: 0.05704072710093289  
Epoch 16/75, Loss: 0.053776597283286764  
Epoch 17/75, Loss: 0.04991354832319242  
Epoch 18/75, Loss: 0.04713122321289517  
Epoch 19/75, Loss: 0.04218561193977689  
Epoch 20/75, Loss: 0.04510606291058356  
Epoch 21/75, Loss: 0.038483053487894214  
Epoch 22/75, Loss: 0.0379830002536333  
Epoch 23/75, Loss: 0.04013793948985319  
Epoch 24/75, Loss: 0.03216634290323837  
Epoch 25/75, Loss: 0.03753336011208  
Epoch 26/75, Loss: 0.03412311541538566  
Epoch 27/75, Loss: 0.02824934760778069  
Epoch 28/75, Loss: 0.03275061313618508  
Epoch 29/75, Loss: 0.033253448247863575  
Epoch 30/75, Loss: 0.02543197648778956  
Epoch 31/75, Loss: 0.025191040590637072  
Epoch 32/75, Loss: 0.03222002559424767  
Epoch 33/75, Loss: 0.030199145966179865  
Epoch 34/75, Loss: 0.02118405998950569  
Epoch 35/75, Loss: 0.022600579856360242  
Epoch 36/75, Loss: 0.02308721250918954  
Epoch 37/75, Loss: 0.020994994436135006  
Epoch 38/75, Loss: 0.026191695994982212  
Epoch 39/75, Loss: 0.02287128277579596  
Epoch 40/75, Loss: 0.02048208353646305  
Epoch 41/75, Loss: 0.02066043517036689  
Epoch 42/75, Loss: 0.01853551409434247

Epoch 43/75, Loss: 0.024111261559951896  
Epoch 44/75, Loss: 0.019578372295762408  
Epoch 45/75, Loss: 0.017679350775969934  
Epoch 46/75, Loss: 0.0215711614083332  
Epoch 47/75, Loss: 0.01713939163038143  
Epoch 48/75, Loss: 0.017011324452206728  
Epoch 49/75, Loss: 0.018371516570437785  
Epoch 50/75, Loss: 0.01911885707345314  
Epoch 51/75, Loss: 0.01736232925555088  
Epoch 52/75, Loss: 0.01692050680433803  
Epoch 53/75, Loss: 0.018151474712912  
Epoch 54/75, Loss: 0.011723280555726547  
Epoch 55/75, Loss: 0.01913941790003311  
Epoch 56/75, Loss: 0.019848266462664643  
Epoch 57/75, Loss: 0.012695891056716143  
Epoch 58/75, Loss: 0.008338540084068818  
Epoch 59/75, Loss: 0.019329817401498584  
Epoch 60/75, Loss: 0.022936138921599583  
Epoch 61/75, Loss: 0.01015816558886662  
Epoch 62/75, Loss: 0.0131897474774696  
Epoch 63/75, Loss: 0.02279871449152917  
Epoch 64/75, Loss: 0.011174469034143786  
Epoch 65/75, Loss: 0.017684894482143613  
Epoch 66/75, Loss: 0.01443913824708327  
Epoch 67/75, Loss: 0.014213146331462043  
Epoch 68/75, Loss: 0.01043119764159893  
Epoch 69/75, Loss: 0.01746341197240412  
Epoch 70/75, Loss: 0.011987612876596334  
Epoch 71/75, Loss: 0.01284054764689252  
Epoch 72/75, Loss: 0.01015789966021862  
Epoch 73/75, Loss: 0.018313447507931718  
Epoch 74/75, Loss: 0.014498993806539563  
Epoch 75/75, Loss: 0.007623427357451297  
Saving model...

Loading model...

**Metrics of train set:**

**sk\_accuracy\_train=0.997985818518138**

classification\_rep\_train =

	precision	recall	f1-score	support
ADJ	0.99	0.99	0.99	1577
ADP	1.00	1.00	1.00	11098
ADV	1.00	0.99	0.99	371
AUX	1.00	1.00	1.00	1573

CCONJ	1.00	1.00	1.00	742
DET	1.00	1.00	1.00	4511
INTJ	1.00	1.00	1.00	335
NOUN	1.00	1.00	1.00	8786
NUM	0.97	1.00	0.98	785
PRON	1.00	1.00	1.00	2435
PROPN	1.00	1.00	1.00	11403
VERB	1.00	1.00	1.00	5039
<b>accuracy</b>			<b>1.00</b>	<b>48655</b>
<b>macro avg</b>	<b>1.00</b>	<b>1.00</b>	<b>1.00</b>	<b>48655</b>
<b>weighted avg</b>	<b>1.00</b>	<b>1.00</b>	<b>1.00</b>	<b>48655</b>

**recall\_micro\_train=0.997985818518138**  
**recall\_macro\_train=0.9971744746310826**  
**f1\_micro\_train=0.997985818518138**  
**f1\_macro\_train=0.9967612133829397**

confusion\_mat\_train =

```

[[ 1568      0      0      0      0      0      0      5      3      0      0
 1]
 [      0 11093      0      0      0      0      0      2      0      0      0
 3]
 [      0      0   367      0      0      0      0      3      0      0      1
 0]
 [      0      0      0  1573      0      0      0      0      0      0      0
 0]
 [      0      0      0      0   742      0      0      0      0      0      0
 0]
 [      0      0      0      0      0  4510      0      1      0      0      0
 0]
 [      0      0      0      0      0      0   334      1      0      0      0
 0]
 [    10      0      0      1      0      0      0  8746     10      0    12
 7]
 [      0      0      0      0      0      0      0      2    782      0      0
 1]
 [      0      0      0      0      0      0      0      1      0  2434      0
 0]
 [      2      0      0      0      0      0      0      7      6      0 11387
 1]
 [      0      0      0      0      0      0      0     13      3      0      2
5021]]

```

**Metrics of dev set:**  
**sk\_accuracy\_dev=0.9441433303221921**

```

classification_rep_dev =
      precision    recall  f1-score   support

   ADJ           0.93      0.84      0.88        223
   ADP           0.97      0.99      0.98       1456
   ADV           0.75      0.75      0.75         51
   AUX           0.96      0.74      0.84        296
  CCONJ           0.97      0.99      0.98        107
   DET           0.99      0.93      0.96        697
  INTJ           1.00      1.00      1.00         35
  NOUN           0.90      0.95      0.92       1130
   NUM           0.91      0.85      0.88        131
  PRON           0.95      1.00      0.97        314
  PROPN           0.98      0.95      0.96       1565
  VERB           0.86      0.96      0.91        637

 accuracy
macro avg      0.93      0.91      0.92      6642
weighted avg   0.95      0.94      0.94      6642

```

```

recall_micro_dev=0.9441433303221921
recall_macro_dev=0.9120591242485095
f1_micro_dev=0.9441433303221921
f1_macro_dev=0.9195467230921373

```

```

confusion_mat_dev =
[[ 187    1    8    0    1    2    0   14    1    1    4    4]
 [   2 1443    1    0    1    0    0    4    0    0    0    5]
 [   1    0   38    2    0    0    0    5    0    0    3    2]
 [   1    1    0  220    0    2    0   18    0    0    1   53]
 [   0    0    0    0  106    0    0    0    0    0    0    1]
 [   2   41    0    1    0  649    0    1    0    0    0    3]
 [   0    0    0    0    0    0   35    0    0    0    0    0]
 [   6    0    0    5    0    0    0  1068    3    0   22   26]
 [   1    0    0    0    0    0    0    2   111   15    2    0]
 [   0    0    0    0    0    0    0    0    0  314    0    0]
 [   1    2    0    0    0    1    0   67    4    0 1487    3]
 [   0    0    4    0    1    0    0   10    3    1    5  613]]

```

#### Metrics of test set:

```

sk_accuracy_test=0.9521276595744681

```

```

classification_rep_test =
      precision    recall  f1-score   support

   ADJ           0.82      0.88      0.85        218
   ADP           0.99      0.99      0.99       1491

```

ADV	0.87	0.61	0.71	76
AUX	0.94	0.90	0.92	239
CCONJ	1.00	0.97	0.99	109
DET	0.99	0.98	0.99	618
INTJ	0.95	0.97	0.96	36
NOUN	0.94	0.91	0.92	1240
NUM	0.80	0.77	0.79	128
PRON	0.95	0.99	0.97	286
PROPN	0.97	0.98	0.98	1516
VERB	0.88	0.95	0.91	623

<b>accuracy</b>			<b>0.95</b>	<b>6580</b>
<b>macro avg</b>	<b>0.93</b>	<b>0.91</b>	<b>0.91</b>	<b>6580</b>
<b>weighted avg</b>	<b>0.95</b>	<b>0.95</b>	<b>0.95</b>	<b>6580</b>

recall\_micro\_test=0.9521276595744681

recall\_macro\_test=0.9084664736197056

f1\_micro\_test=0.9521276595744681

f1\_macro\_test=0.9146494786148335

confusion\_mat\_test =

```
[[ 191      2      2      0      0      1      0      11      5      0      3      3]
 [    4 1481      1      0      0      1      0      2      1      0      0      1]
 [   17      1   46      1      0      0      1      2      0      1      4      3]
 [    0      0      0  215      0      0      0     21      0      0      1      2]
 [    0      0      0      0  106      0      0      2      0      0      0      1]
 [    3      0      1      1      0   605      1      4      0      0      3      0]
 [    0      0      0      0      0      0   35      0      0      0      0      1]
 [    7      3      1     10      0      0      0  1128      5      0     22     64]
 [    4      0      0      1      0      0      0      4     99     15      3      2]
 [    0      0      0      0      0      2      0      0      0    284      0      0]
 [    4      1      2      0      0      1      0     13      9      0  1484      2]
 [    3      2      0      0      0      0      0     19      4      0      4   591]]
```

**p=3 s=3**

Training model...

Epoch 1/75, Loss: 0.6136231879543381  
Epoch 2/75, Loss: 0.23109073894840032  
Epoch 3/75, Loss: 0.17369523159205014  
Epoch 4/75, Loss: 0.14257437405733298  
Epoch 5/75, Loss: 0.12236861573867548  
Epoch 6/75, Loss: 0.10924358293169353  
Epoch 7/75, Loss: 0.09828827280250443  
Epoch 8/75, Loss: 0.089253970347351  
Epoch 9/75, Loss: 0.07971711097287905  
Epoch 10/75, Loss: 0.077430327652688  
Epoch 11/75, Loss: 0.06843705482887405  
Epoch 12/75, Loss: 0.06674183667781206  
Epoch 13/75, Loss: 0.06004759395429583  
Epoch 14/75, Loss: 0.05781155366379184  
Epoch 15/75, Loss: 0.05604326767724867  
Epoch 16/75, Loss: 0.04937365675293357  
Epoch 17/75, Loss: 0.0486428805256878  
Epoch 18/75, Loss: 0.042729378669892445  
Epoch 19/75, Loss: 0.042962512109836634  
Epoch 20/75, Loss: 0.039832713588680686  
Epoch 21/75, Loss: 0.03901295537335678  
Epoch 22/75, Loss: 0.034946161869283225  
Epoch 23/75, Loss: 0.0350869149795677  
Epoch 24/75, Loss: 0.02928802635264728  
Epoch 25/75, Loss: 0.03548459676486643  
Epoch 26/75, Loss: 0.02774214527897789  
Epoch 27/75, Loss: 0.02906649150464114  
Epoch 28/75, Loss: 0.032239798807433245  
Epoch 29/75, Loss: 0.025808440042114297  
Epoch 30/75, Loss: 0.02532763664652002  
Epoch 31/75, Loss: 0.02836155713044914  
Epoch 32/75, Loss: 0.025846273821412875  
Epoch 33/75, Loss: 0.02419761915481963  
Epoch 34/75, Loss: 0.023974995278144416  
Epoch 35/75, Loss: 0.020739304348079787  
Epoch 36/75, Loss: 0.02409619288076659  
Epoch 37/75, Loss: 0.023893752229001208  
Epoch 38/75, Loss: 0.017369652433574842  
Epoch 39/75, Loss: 0.021500226730614854  
Epoch 40/75, Loss: 0.025041418425960933  
Epoch 41/75, Loss: 0.01666209857973236  
Epoch 42/75, Loss: 0.017225050536275908

Epoch 43/75, Loss: 0.017854187815707342  
Epoch 44/75, Loss: 0.021187122239319043  
Epoch 45/75, Loss: 0.022711282643601925  
Epoch 46/75, Loss: 0.013168443667398966  
Epoch 47/75, Loss: 0.026872589642840984  
Epoch 48/75, Loss: 0.017059267377236842  
Epoch 49/75, Loss: 0.012087913963095736  
Epoch 50/75, Loss: 0.017103950323258305  
Epoch 51/75, Loss: 0.017553822486073296  
Epoch 52/75, Loss: 0.017402153708078984  
Epoch 53/75, Loss: 0.018220528052628183  
Epoch 54/75, Loss: 0.01992841178551755  
Epoch 55/75, Loss: 0.01091572596809779  
Epoch 56/75, Loss: 0.01529875124308467  
Epoch 57/75, Loss: 0.015908462319284142  
Epoch 58/75, Loss: 0.016718878429690068  
Epoch 59/75, Loss: 0.014323895355564581  
Epoch 60/75, Loss: 0.016232970950852632  
Epoch 61/75, Loss: 0.011443299988896864  
Epoch 62/75, Loss: 0.015495828033144306  
Epoch 63/75, Loss: 0.01215549170464137  
Epoch 64/75, Loss: 0.012961527832133708  
Epoch 65/75, Loss: 0.014078445021157926  
Epoch 66/75, Loss: 0.013271559616253006  
Epoch 67/75, Loss: 0.012557289755295778  
Epoch 68/75, Loss: 0.01706889579733555  
Epoch 69/75, Loss: 0.009507583748596928  
Epoch 70/75, Loss: 0.015332845086418323  
Epoch 71/75, Loss: 0.011060372136469168  
Epoch 72/75, Loss: 0.015503776062231647  
Epoch 73/75, Loss: 0.009370110531024465  
Epoch 74/75, Loss: 0.014439139431303268  
Epoch 75/75, Loss: 0.013925257731443338

Saving model...

Loading model...

**Metrics of train set:**

**sk\_accuracy\_train=0.9974308909670127**

classification\_rep\_train =

	precision	recall	f1-score	support
ADJ	0.99	0.99	0.99	1577
ADP	1.00	1.00	1.00	11098
ADV	0.99	0.99	0.99	371
AUX	1.00	1.00	1.00	1573



CCONJ	0.99	1.00	1.00	742
DET	1.00	1.00	1.00	4511
INTJ	0.99	1.00	0.99	335
NOUN	0.99	1.00	1.00	8786
NUM	0.99	0.99	0.99	785
PRON	1.00	1.00	1.00	2435
PROPN	1.00	1.00	1.00	11403
VERB	0.99	1.00	1.00	5039
<b>accuracy</b>			<b>1.00</b>	<b>48655</b>
<b>macro avg</b>	<b>1.00</b>	<b>1.00</b>	<b>1.00</b>	<b>48655</b>
<b>weighted avg</b>	<b>1.00</b>	<b>1.00</b>	<b>1.00</b>	<b>48655</b>

recall\_micro\_train=0.9974308909670127

recall\_macro\_train=0.9961855185173517

f1\_micro\_train=0.9974308909670127

f1\_macro\_train=0.9958864246054083

confusion\_mat\_train =

```
[[ 1563      0      0      0      2      0      0      9      1      0      1
1]
[      0 11094      0      0      0      1      0      0      0      0      1
2]
[      1      0   366      0      0      0      0      3      0      0      1
0]
[      0      0      0  1570      0      0      3      0      0      0      0
0]
[      0      0      0      0   742      0      0      0      0      0      0
0]
[      1      0      0      0      0  4502      0      4      0      0      2
2]
[      0      0      0      0      0      0   335      0      0      0      0
0]
[      9      0      1      1      1      2      1  8752      2      0      8
9]
[      0      0      0      0      1      0      0      4   777      0      0
3]
[      0      0      0      0      0      0      0      0      0  2435      0
0]
[      4      0      0      0      0      0      0      21      2      0 11363
13]
[      1      0      1      0      0      0      0      5      0      0      1
5031]]
```

Metrics of dev set:

sk\_accuracy\_dev=0.9409816320385426

```

classification_rep_dev =
      precision    recall  f1-score   support

   ADJ           0.89     0.86     0.88       223
   ADP           0.97     0.99     0.98      1456
   ADV           0.71     0.76     0.74        51
   AUX           0.95     0.73     0.83       296
  CCONJ           0.98     0.98     0.98       107
   DET           0.99     0.93     0.96       697
  INTJ           0.92     1.00     0.96        35
  NOUN           0.90     0.93     0.92      1130
   NUM           0.89     0.83     0.86       131
  PRON           0.95     0.99     0.97       314
  PROPN           0.98     0.95     0.96      1565
  VERB           0.86     0.97     0.91       637

 accuracy
macro avg      0.92    0.91    0.91    6642
weighted avg   0.94    0.94    0.94    6642

```

**recall\_micro\_dev=0.9409816320385426**

**recall\_macro\_dev=0.9107010930927868**

**f1\_micro\_dev=0.9409816320385426**

**f1\_macro\_dev=0.911280603976841**

confusion\_mat\_dev =

```

[[ 192    1    5    0    1    2    0   11    2    1    3    5]
 [   2 1440    2    0    0    1    0    4    0    1    2    4]
 [   0    1   39    0    0    0    0    4    1    1    5    0]
 [   0    1    2  216    0    0    3   17    0    0    2  55]
 [   0    0    0    0  105    0    0    1    0    0    0    1]
 [   2   43    1    1    0  648    0    1    0    0    1    0]
 [   0    0    0    0    0    0   35    0    0    0    0    0]
 [  15    1    3    8    0    2    0 1051    3    0   18  29]
 [   0    0    1    0    0    1    0    3  109   15    0    2]
 [   0    0    0    0    0    0    0    2    0  312    0    0]
 [   2    1    0    0    0    0    0   63    6    0 1487    6]
 [   2    2    2    2    1    0    0    5    1    0    6 616]]

```

**Metrics of test set:**

**sk\_accuracy\_test=0.9493920972644377**

classification\_rep\_test =

```

      precision    recall  f1-score   support

   ADJ           0.86     0.90     0.88       218
   ADP           0.99     0.99     0.99      1491

```

ADV	0.77	0.62	0.69	76
AUX	0.95	0.90	0.92	239
CCONJ	0.97	0.95	0.96	109
DET	0.99	0.98	0.99	618
INTJ	0.88	1.00	0.94	36
NOUN	0.93	0.90	0.91	1240
NUM	0.86	0.78	0.82	128
PRON	0.93	1.00	0.96	286
PROPN	0.97	0.98	0.97	1516
VERB	0.87	0.94	0.91	623

<b>accuracy</b>			<b>0.95</b>	<b>6580</b>
<b>macro avg</b>	<b>0.91</b>	<b>0.91</b>	<b>0.91</b>	<b>6580</b>
<b>weighted avg</b>	<b>0.95</b>	<b>0.95</b>	<b>0.95</b>	<b>6580</b>

**recall\_micro\_test=0.9493920972644377**

**recall\_macro\_test=0.9114722790171058**

**f1\_micro\_test=0.9493920972644377**

**f1\_macro\_test=0.9116387925449995**

confusion\_mat\_test =

```
[[ 196    0    2    1    2    0    1    9    1    0    1    5]
 [   3 1477    2    0    0    1    0    2    0    2    3    1]
 [  18    1   47    0    0    0    1    2    0    2    4    1]
 [   0    0    0  215    0    0    0   21    0    0    3    0]
 [   1    0    1    0  104    0    0    1    0    0    1    1]
 [   1    1    2    0    0  606    1    5    0    0    1    1]
 [   0    0    0    0    0    0   36    0    0    0    0    0]
 [   5    2    2   10    1    0    2 1112    7    0   28   71]
 [   0    1    1    1    0    0    0    6   100   15    2    2]
 [   0    0    0    0    0    1    0    0    0  285    0    0]
 [   3    1    1    0    0    0    0   20    5    0 1481    5]
 [   2    3    3    0    0    2    0   17    3    1    4  588]]
```

**p=4 s=4**

Training model...

Epoch 1/75, Loss: 0.5626572770080022  
Epoch 2/75, Loss: 0.2403129536594887  
Epoch 3/75, Loss: 0.1800828152990921  
Epoch 4/75, Loss: 0.14544266758121932  
Epoch 5/75, Loss: 0.1225804647723881  
Epoch 6/75, Loss: 0.1045258568120085  
Epoch 7/75, Loss: 0.09632886438257258  
Epoch 8/75, Loss: 0.08519852415781068  
Epoch 9/75, Loss: 0.0800808182352745  
Epoch 10/75, Loss: 0.06707518164077027  
Epoch 11/75, Loss: 0.0677103447174389  
Epoch 12/75, Loss: 0.05915198307322233  
Epoch 13/75, Loss: 0.05316585168582369  
Epoch 14/75, Loss: 0.056287070114094256  
Epoch 15/75, Loss: 0.04695392495950405  
Epoch 16/75, Loss: 0.042428462784315134  
Epoch 17/75, Loss: 0.043212575500961956  
Epoch 18/75, Loss: 0.039572288750267144  
Epoch 19/75, Loss: 0.03665015791859987  
Epoch 20/75, Loss: 0.033491335093213476  
Epoch 21/75, Loss: 0.032417380729817244  
Epoch 22/75, Loss: 0.03388855144511248  
Epoch 23/75, Loss: 0.029515575875814836  
Epoch 24/75, Loss: 0.03163067653659947  
Epoch 25/75, Loss: 0.025529922403853184  
Epoch 26/75, Loss: 0.02666165327787373  
Epoch 27/75, Loss: 0.026357483060923385  
Epoch 28/75, Loss: 0.027122807513322057  
Epoch 29/75, Loss: 0.0235691475522315  
Epoch 30/75, Loss: 0.022750605183666773  
Epoch 31/75, Loss: 0.025448588501646833  
Epoch 32/75, Loss: 0.019111906016084663  
Epoch 33/75, Loss: 0.018333248247331187  
Epoch 34/75, Loss: 0.02039696650104175  
Epoch 35/75, Loss: 0.019292128794703642  
Epoch 36/75, Loss: 0.021136131047686628  
Epoch 37/75, Loss: 0.018126870784858622  
Epoch 38/75, Loss: 0.022344241623556558  
Epoch 39/75, Loss: 0.01484474513560519  
Epoch 40/75, Loss: 0.021732423168113524  
Epoch 41/75, Loss: 0.014885076386801827  
Epoch 42/75, Loss: 0.019266160734490386

Epoch 43/75, Loss: 0.016178489500446716  
Epoch 44/75, Loss: 0.013748834753043995  
Epoch 45/75, Loss: 0.02307849664436166  
Epoch 46/75, Loss: 0.014427895025172385  
Epoch 47/75, Loss: 0.010091405794061372  
Epoch 48/75, Loss: 0.02134290929735452  
Epoch 49/75, Loss: 0.016990455804844054  
Epoch 50/75, Loss: 0.016353955074374597  
Epoch 51/75, Loss: 0.01394435920713485  
Epoch 52/75, Loss: 0.016922478951877314  
Epoch 53/75, Loss: 0.013722538957351973  
Epoch 54/75, Loss: 0.010712827724988862  
Epoch 55/75, Loss: 0.01313519796878119  
Epoch 56/75, Loss: 0.01596668890022546  
Epoch 57/75, Loss: 0.019857267397363623  
Epoch 58/75, Loss: 0.008543637250920883  
Epoch 59/75, Loss: 0.01624863701838281  
Epoch 60/75, Loss: 0.010221772269053558  
Epoch 61/75, Loss: 0.015544409213794949  
Epoch 62/75, Loss: 0.012159392875231524  
Epoch 63/75, Loss: 0.014255131743927167  
Epoch 64/75, Loss: 0.00638195187964159  
Epoch 65/75, Loss: 0.012012925659419195  
Epoch 66/75, Loss: 0.018569944874697288  
Epoch 67/75, Loss: 0.012050978502651303  
Epoch 68/75, Loss: 0.012550602032640315  
Epoch 69/75, Loss: 0.008885750291539964  
Epoch 70/75, Loss: 0.015780557215214427  
Epoch 71/75, Loss: 0.011134464023758204  
Epoch 72/75, Loss: 0.0070732269813792315  
Epoch 73/75, Loss: 0.015522285835747664  
Epoch 74/75, Loss: 0.01502310316802834  
Epoch 75/75, Loss: 0.009570330699243  
Saving model...

Loading model...

**Metrics of train set:**

**sk\_accuracy\_train=0.9948617819340253**

classification\_rep\_train =

	precision	recall	f1-score	support
ADJ	1.00	0.94	0.97	1577
ADP	1.00	1.00	1.00	11098
ADV	0.98	0.93	0.95	371
AUX	0.99	1.00	1.00	1573

CCONJ	1.00	1.00	1.00	742
DET	1.00	1.00	1.00	4511
INTJ	0.97	1.00	0.99	335
NOUN	0.99	1.00	0.99	8786
NUM	0.97	0.98	0.98	785
PRON	1.00	1.00	1.00	2435
PROPN	1.00	1.00	1.00	11403
VERB	0.99	1.00	0.99	5039
<b>accuracy</b>			<b>0.99</b>	<b>48655</b>
<b>macro avg</b>	<b>0.99</b>	<b>0.99</b>	<b>0.99</b>	<b>48655</b>
<b>weighted avg</b>	<b>0.99</b>	<b>0.99</b>	<b>0.99</b>	<b>48655</b>

**recall\_micro\_train=0.9948617819340253**  
**recall\_macro\_train=0.986216971029593**  
**f1\_micro\_train=0.9948617819340253**  
**f1\_macro\_train=0.9882420468176026**  
confusion\_mat\_train =

[[ 1482	3	7	1	0	0	1	58	8	0	4
13]										
[ 0 11092	0	1	0	0	1	3	0	0	0	0
1]										
[ 0 0 344	0	0	0	8	11	0	0	0	1	
7]										
[ 0 0 0 1572	0	0	0	0	0	0	0	0	0	0
1]										
[ 0 0 0 0 739	0	0	0	0	0	0	0	0	0	0
3]										
[ 1 8 0 0 0 4500	0	0	0	0	0	0	0	0	0	0
2]										
[ 0 0 0 0 0 0 335	0	0	0	0	0	0	0	0	0	0
0]										
[ 2 0 0 3 0 0 0 8748	6	0	12							
15]										
[ 0 0 1 0 0 0 0 5 773	0	0	0							
6]										
[ 0 0 0 0 0 0 0 0 0 2435	0									
0]										
[ 0 2 0 4 0 2 0 28 7 1 11347										
12]										
[ 0 1 0 0 0 0 0 0 0 0 0 0										
5038]]										

**Metrics of dev set:**  
**sk\_accuracy\_dev=0.9373682625715146**

```

classification_rep_dev =
      precision    recall  f1-score   support

   ADJ           0.97      0.79      0.87        223
   ADP           0.97      0.99      0.98       1456
   ADV           0.79      0.65      0.71         51
   AUX           0.95      0.74      0.83        296
  CCONJ          0.99      0.99      0.99        107
   DET           0.99      0.93      0.96        697
  INTJ           0.92      1.00      0.96         35
  NOUN           0.88      0.94      0.91       1130
   NUM           0.86      0.76      0.80        131
  PRON           0.95      0.99      0.97        314
  PROPJ          0.98      0.94      0.96       1565
  VERB           0.84      0.97      0.90        637

 accuracy
macro avg      0.92      0.89      0.90      6642
weighted avg   0.94      0.94      0.94      6642

```

**recall\_micro\_dev=0.9373682625715146**

**recall\_macro\_dev=0.8913191734410507**

**f1\_micro\_dev=0.9373682625715146**

**f1\_macro\_dev=0.9038903907478538**

confusion\_mat\_dev =

```

[[ 177    0    3    0    0    1    0   26    4    0    4    8]
 [   1 1445    0    1    0    2    0    1    0    0    0    6]
 [   0    0   33    1    0    0    2    4    2    0    4    5]
 [   0    1    1  220    0    0    0   19    0    0    1  54]
 [   0    0    0    0  106    0    0    0    0    0    0    1]
 [   1   45    0    1    0  645    0    3    0    1    1    0]
 [   0    0    0    0    0    0   35    0    0    0    0    0]
 [   2    2    2    6    0    2    0 1063    6    0   20   27]
 [   0    0    0    0    0    1    0   10   99   15    0    6]
 [   0    0    0    0    0    0    0    1    0  312    0    1]
 [   0    1    3    1    1    1    0   74    3    0 1471   10]
 [   2    1    0    2    0    0    1    7    1    0    3  620]]

```

**Metrics of test set:**

**sk\_accuracy\_test=0.9451367781155016**

classification\_rep\_test =

```

      precision    recall  f1-score   support

   ADJ           0.89      0.79      0.84        218
   ADP           0.99      0.99      0.99       1491

```

ADV	0.85	0.54	0.66	76
AUX	0.93	0.90	0.91	239
CCONJ	1.00	0.94	0.97	109
DET	0.99	0.98	0.98	618
INTJ	0.88	0.97	0.92	36
NOUN	0.92	0.90	0.91	1240
NUM	0.76	0.71	0.73	128
PRON	0.95	0.99	0.97	286
PROPN	0.97	0.98	0.97	1516
VERB	0.86	0.96	0.91	623

<b>accuracy</b>			<b>0.95</b>	<b>6580</b>
<b>macro avg</b>	<b>0.91</b>	<b>0.89</b>	<b>0.90</b>	<b>6580</b>
<b>weighted avg</b>	<b>0.95</b>	<b>0.95</b>	<b>0.94</b>	<b>6580</b>

recall\_micro\_test=0.9451367781155016

recall\_macro\_test=0.8878807760439787

f1\_micro\_test=0.9451367781155016

f1\_macro\_test=0.8975692535220482

confusion\_mat\_test =

```
[[ 173    1    4    0    0    1    0   17    7    0    5   10]
 [    0 1481    0    2    0    1    0    1    2    0    2    2]
 [   16    0   41    1    0    0    4    6    3    1    2    2]
 [    1    2    0  214    0    0    0   21    0    0    1    0]
 [    0    0    0    0  103    0    0    1    0    0    1    4]
 [    0    4    1    1    0  603    1    7    1    0    0    0]
 [    0    0    0    0    0    0   35    0    0    0    0    1]
 [    2    2    2   10    0    0    0  1115    9    0   29   71]
 [    2    0    0    0    0    0    0   15   91   14    2    4]
 [    0    0    0    0    0    1    0    1    0  284    0    0]
 [    0    2    0    3    0    1    0   17    6    0 1481    6]
 [    0    3    0    0    0    2    0   16    1    0    3  598]]
```



### LSTM

"red"="Train Accuracy"

"blue"="Train Loss"

"green"="Validation Loss"

"yellow"="Validation Accuracy"

Chosen Best Model : lstm\_model\_1.pt

Model : lstm\_model\_1.pt

Parameters:

```
model = LSTM(64, 64, 1, len(trainData.vocab), len(trainData.tagVocab))
```

```
optimizer = torch.optim.Adam(model.parameters(), lr=0.01)
```

```
train_loss_list, valid_loss_list, train_accuracy_list,
```

```
valid_accuracy_list = train(model, trainData, optimizer, criterion,  
valData, 1, "Test1")
```

Training model...

Validation loss: 0.0377047061920166

Epoch 1 loss: 0.21563559322993256

Validation loss: 0.027747202664613724

Epoch 2 loss: 0.024322354617014305

Validation loss: 0.02634860761463642

Epoch 3 loss: 0.019135417632147003

Validation loss: 0.025797950103878975

Epoch 4 loss: 0.017105293588768413

Validation loss: 0.025858696550130844

Validation loss increased

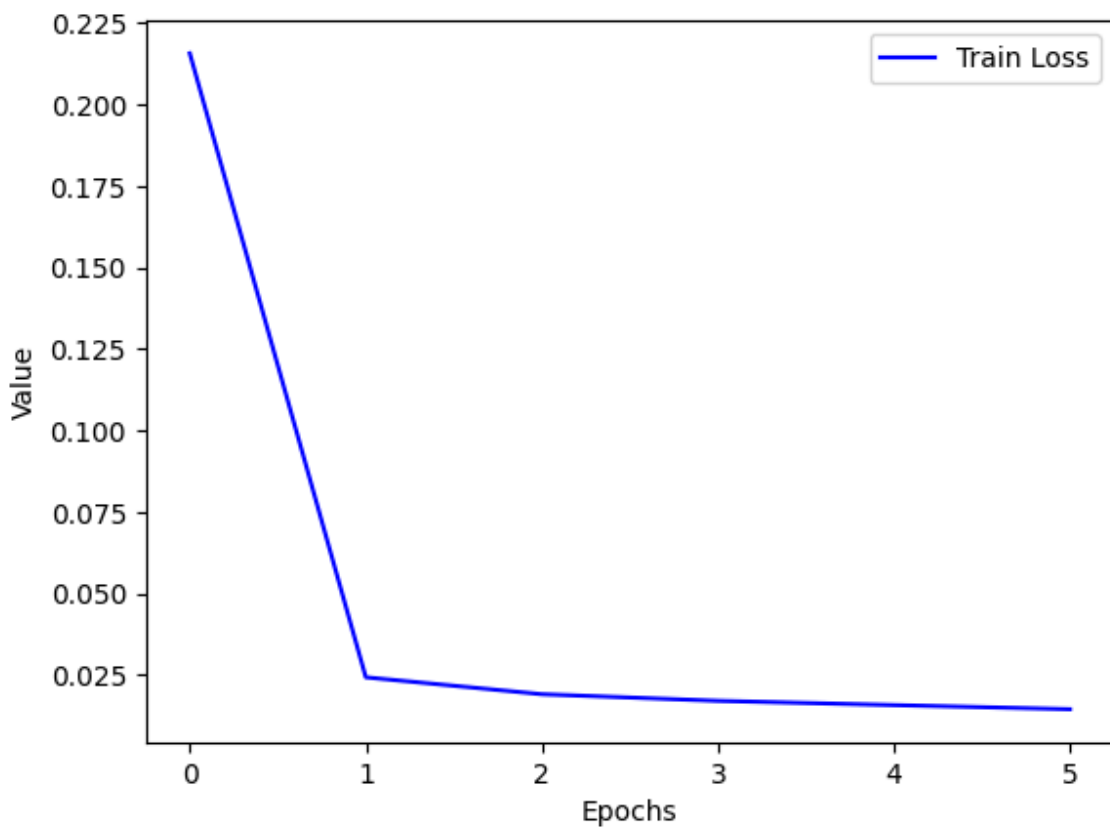
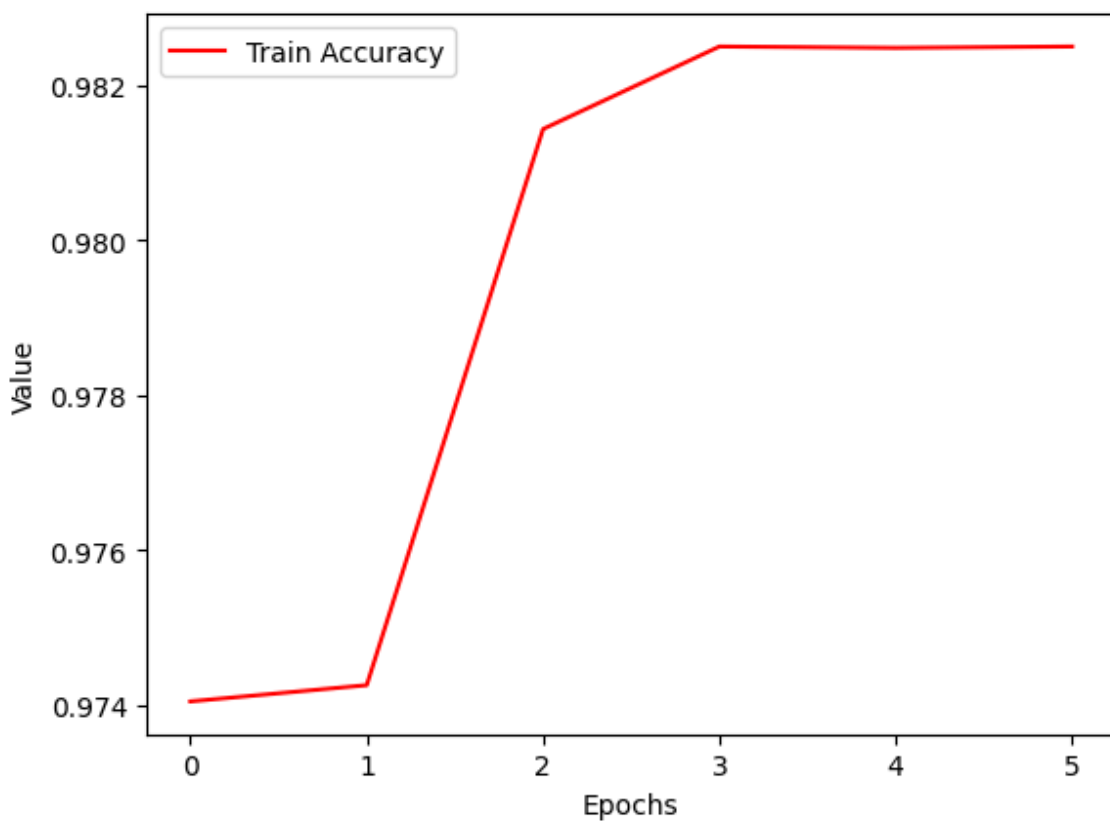
Epoch 5 loss: 0.015801586294129713

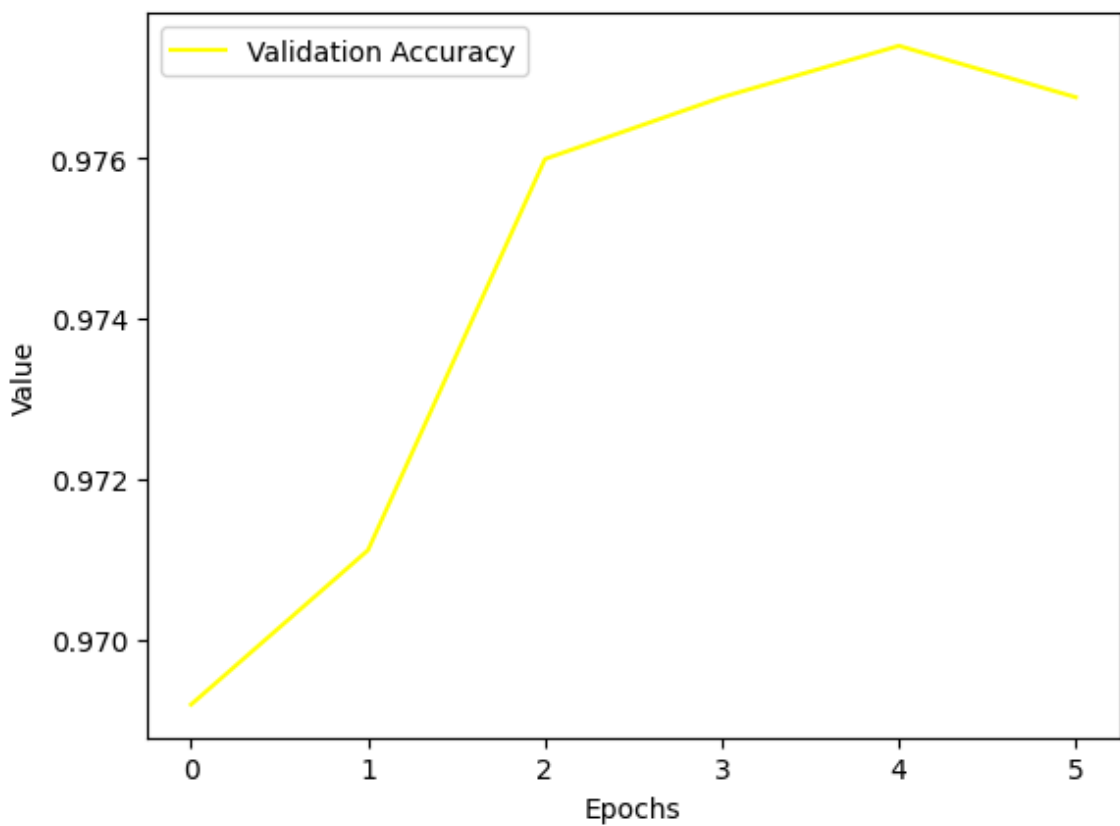
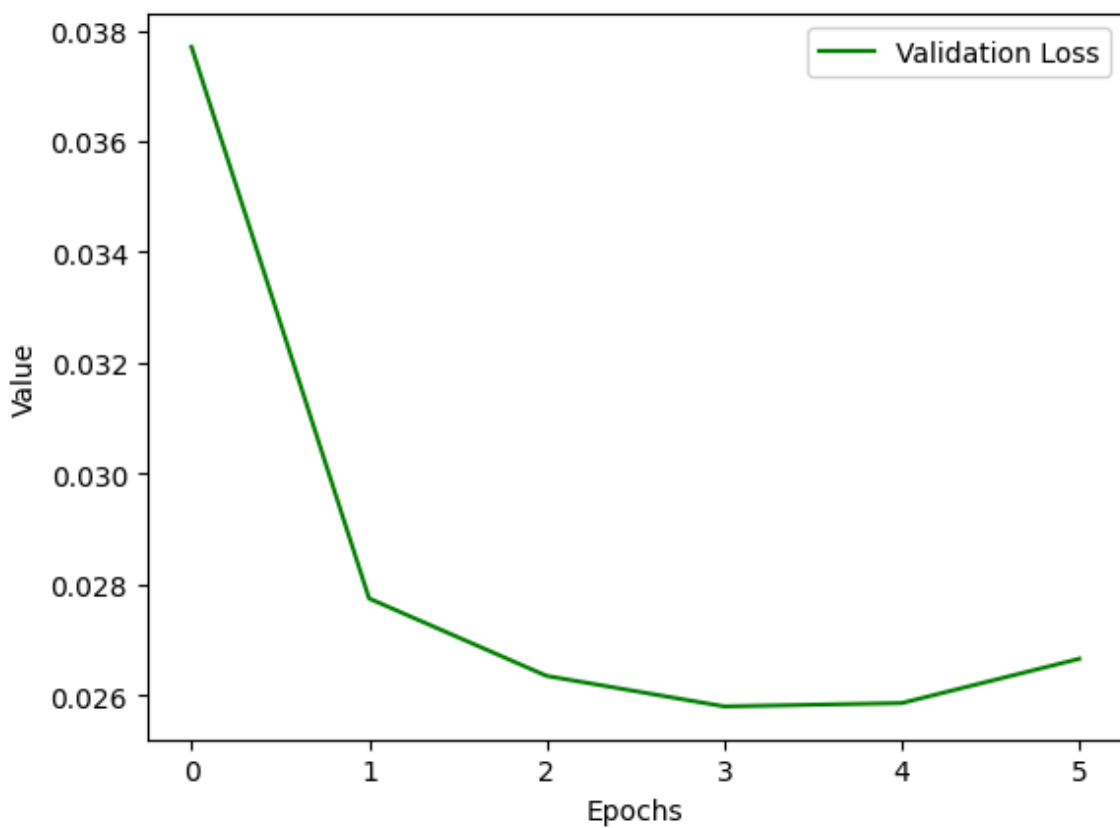
Validation loss: 0.026658661663532257

Validation loss increased

Early stopping

Epoch 6 loss: 0.014542925633740291





Loading model...

**Training Accuracy: 0.9825021413463387**

**Metrics of val set:**

**valData\_sk\_accuracy=0.9767591165896251**

valData\_classification\_rep =

	precision	recall	f1-score	support
<EOT>	1.00	1.00	1.00	573
<SOT>	1.00	1.00	1.00	573
adj	0.96	0.89	0.93	227
adp	0.99	0.99	0.99	1415
adv	0.81	0.81	0.81	59
aux	0.97	0.97	0.97	266
cconj	1.00	1.00	1.00	107
det	0.99	0.86	0.92	568
intj	1.00	1.00	1.00	35
noun	0.98	0.99	0.98	1143
num	0.94	0.99	0.96	131
part	0.90	0.89	0.90	73
pron	0.86	0.99	0.92	414
propn	0.99	0.99	0.99	1551
verb	0.98	0.98	0.98	653
<b>accuracy</b>			<b>0.98</b>	<b>7788</b>
<b>macro avg</b>	<b>0.96</b>	<b>0.96</b>	<b>0.96</b>	<b>7788</b>
<b>weighted avg</b>	<b>0.98</b>	<b>0.98</b>	<b>0.98</b>	<b>7788</b>

**valData\_recall\_micro=0.9767591165896251**

**valData\_recall\_macro=0.9577204100124747**

**valData\_f1\_micro=0.9767591165896251**

**valData\_f1\_macro=0.9573463444351754**

valData\_confusion\_mat =

```
[[ 573    0    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0  573    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0    0  203    0    7    0    0    0    0    11    3    0    0
2
    1]
[   0    0    0 1407    0    0    0    0    0    1    0    6    1
0
    0]
```

```

[ 0 0 5 0 48 1 0 0 0 2 1 0 0
0
2]
[ 0 0 0 0 0 258 0 0 0 0 0 0 0
0
8]
[ 0 0 0 0 0 0 107 0 0 0 0 0 0
0
0]
[ 0 0 0 10 0 0 0 491 0 0 1 0 66
0
0]
[ 0 0 0 0 0 0 0 0 35 0 0 0 0
0
0]
[ 0 0 3 0 0 0 0 0 0 1130 1 0 0
9
0]
[ 0 0 0 1 0 0 0 0 0 0 130 0 0
0
0]
[ 0 0 0 8 0 0 0 0 0 0 0 65 0
0
0]
[ 0 0 0 0 0 0 0 4 0 1 0 1 408
0
0]
[ 0 0 0 0 1 0 0 0 0 7 2 0 0
1539
2]
[ 0 0 0 0 3 6 0 0 0 2 1 0 0
1
640]]

```

# Metrics of test set:

testData\_sk\_accuracy=0.9770381836945304

testData\_classification\_rep =

	precision	recall	f1-score	support
<EOT>	1.00	1.00	1.00	586
<SOT>	1.00	1.00	1.00	586
adj	0.91	0.95	0.93	220
adp	1.00	1.00	1.00	1434
adv	0.82	0.67	0.74	76
aux	0.95	1.00	0.97	256

cconj	1.00	1.00	1.00	109
det	0.99	0.87	0.92	512
intj	1.00	1.00	1.00	36
noun	0.99	0.99	0.99	1166
num	0.88	0.96	0.92	127
part	0.98	0.96	0.97	56
pron	0.85	0.98	0.91	392
propn	0.99	0.99	0.99	1567
verb	0.99	0.96	0.97	629
<b>accuracy</b>			<b>0.98</b>	<b>7752</b>
<b>macro avg</b>	<b>0.96</b>	<b>0.96</b>	<b>0.95</b>	<b>7752</b>
<b>weighted avg</b>	<b>0.98</b>	<b>0.98</b>	<b>0.98</b>	<b>7752</b>

testData\_recall\_micro=0.9770381836945304

testData\_recall\_macro=0.9553220963694653

testData\_f1\_micro=0.9770381836945304

testData\_f1\_macro=0.954684369349297

testData\_confusion\_mat =

```
[[ 586    0    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0  586    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0    0  210    0    3    0    0    0    0    0    1    0    0    0
4
    2]
[   0    0    1 1429    0    0    0    0    0    0    1    0    1    2
0
    0]
[   0    0   15    2   51    1    0    0    0    0    1    2    0    0
3
    1]
[   0    0    0    0    0   255    0    0    0    0    0    0    0    0
0
    1]
[   0    0    0    0    0    0   109    0    0    0    0    0    0    0
0
    0]
[   0    0    0    1    1    0    0  443    0    1    0    0    0    65
1
    0]
[   0    0    0    0    0    0    0    0    0   36    0    0    0    0
0
```

[illegible]

Model : lstm\_model\_2.pt

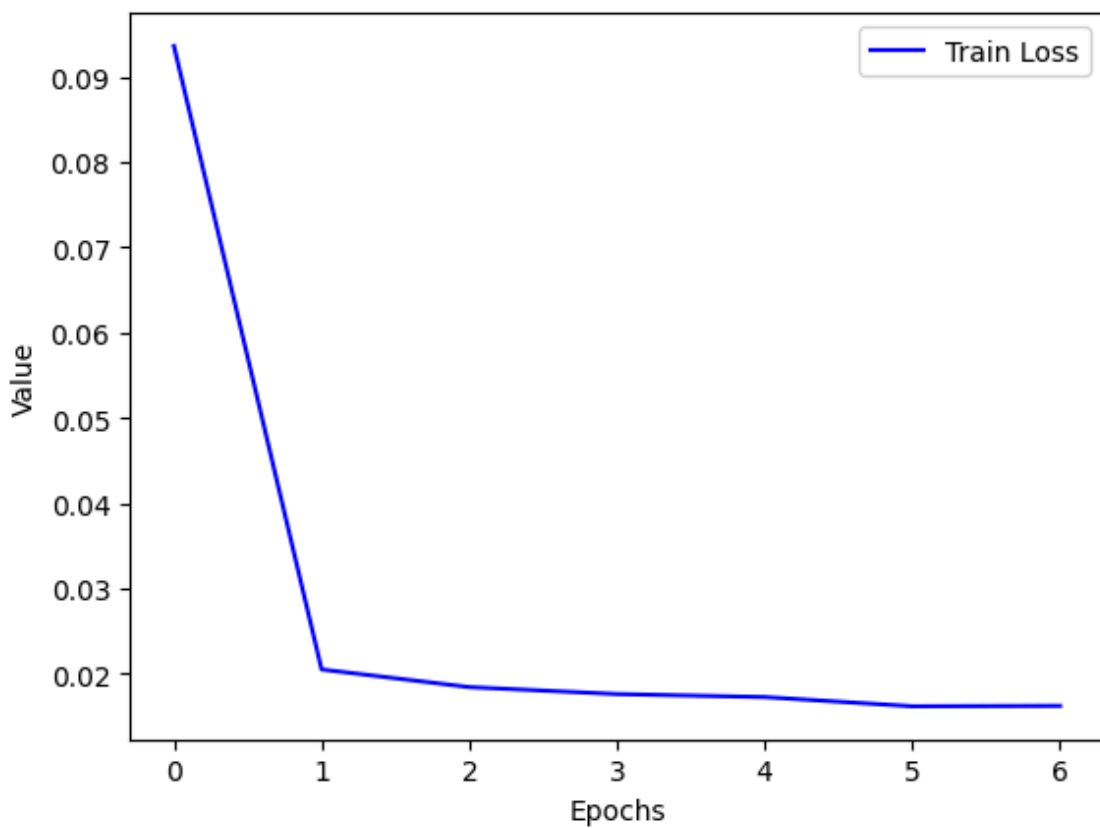
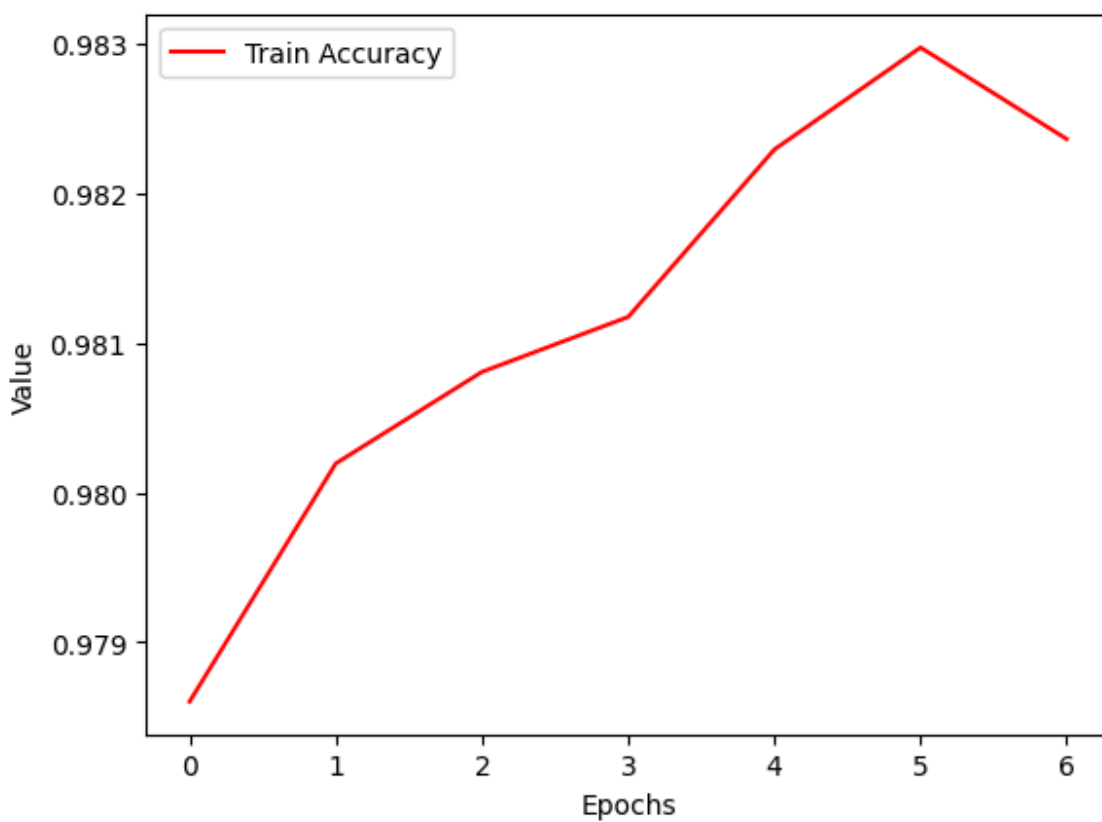
Parameters:

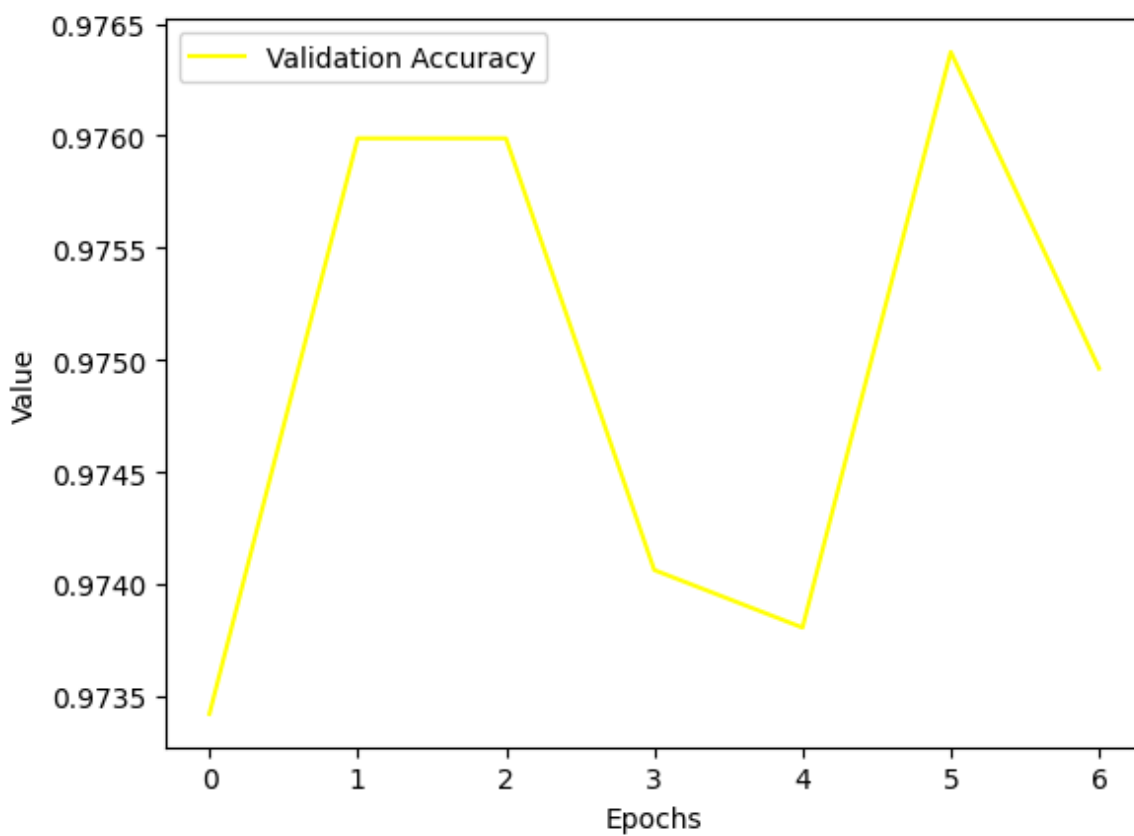
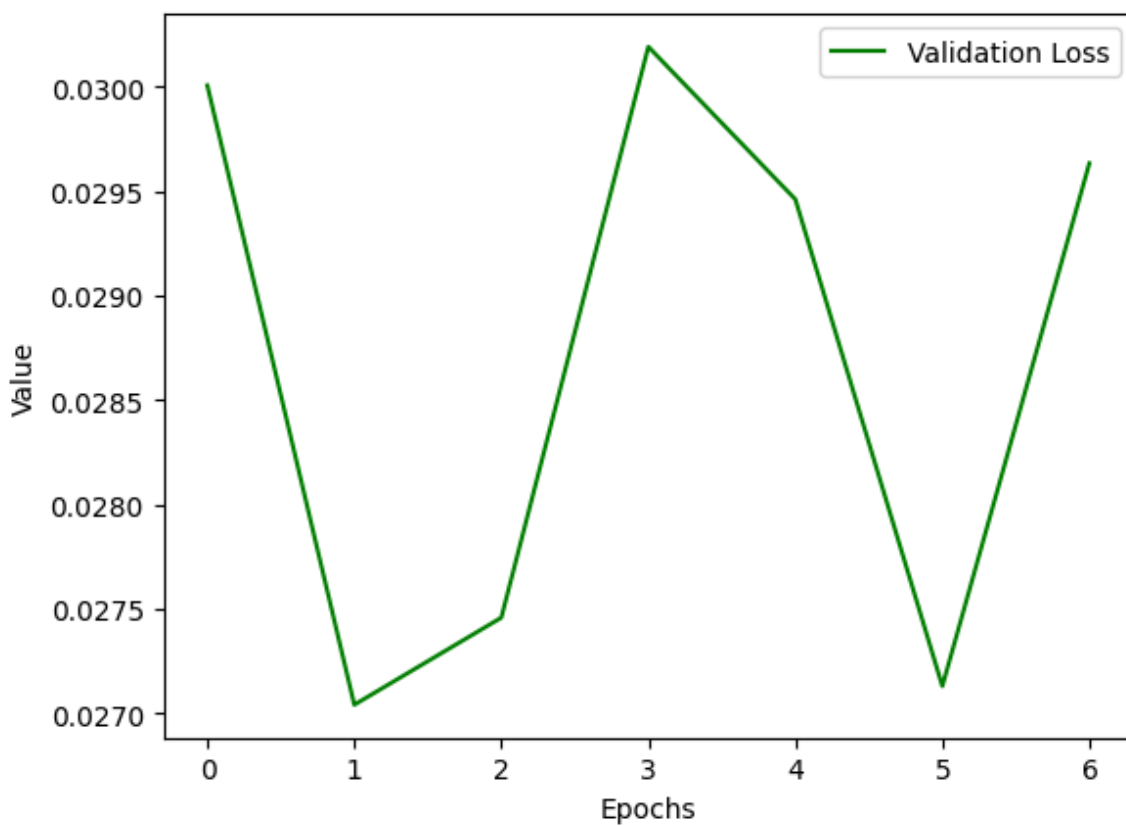
```
model = LSTM(128, 128, 1, len(trainData.vocab),
len(trainData.tagVocab))
optimizer = torch.optim.Adam(model.parameters(), lr=0.025)
train_loss_list, valid_loss_list, train_accuracy_list,
valid_accuracy_list = train(model, trainData, optimizer, criterion,
valData, 4, "Test2")
```

Training model...

```
Validation loss: 0.03000626340508461
Epoch 1 loss: 0.09359400103619295
Validation loss: 0.027040695771574974
Epoch 2 loss: 0.020520129703235492
Validation loss: 0.027457769960165024
Validation loss increased
Epoch 3 loss: 0.01844894391164851
Validation loss: 0.030191484838724136
Validation loss increased
Epoch 4 loss: 0.017633233306262253
Validation loss: 0.029461102560162544
Epoch 5 loss: 0.017281378668024025
Validation loss: 0.0271297600120306
Epoch 6 loss: 0.016205696170148786
Validation loss: 0.02963349223136902
Validation loss increased
Epoch 7 loss: 0.016253406140230485
```







Loading model...

**Training Accuracy: 0.9823622983201357**

**Metrics of val set:**

**valData\_sk\_accuracy=0.9749614791987673**

valData\_classification\_rep =

	precision	recall	f1-score	support
<EOT>	1.00	1.00	1.00	573
<SOT>	1.00	1.00	1.00	573
adj	0.98	0.88	0.93	227
adp	0.98	1.00	0.99	1415
adv	0.82	0.85	0.83	59
aux	0.97	0.96	0.96	266
cconj	1.00	1.00	1.00	107
det	1.00	0.85	0.92	568
intj	1.00	1.00	1.00	35
noun	0.98	0.98	0.98	1143
num	0.96	0.99	0.97	131
part	0.92	0.79	0.85	73
pron	0.86	0.99	0.92	414
propn	0.99	0.99	0.99	1551
verb	0.96	0.98	0.97	653
<b>accuracy</b>			<b>0.97</b>	<b>7788</b>
<b>macro avg</b>	<b>0.96</b>	<b>0.95</b>	<b>0.96</b>	<b>7788</b>
<b>weighted avg</b>	<b>0.98</b>	<b>0.97</b>	<b>0.97</b>	<b>7788</b>

**valData\_recall\_micro=0.9749614791987673**

**valData\_recall\_macro=0.9514734869262078**

**valData\_f1\_micro=0.9749614791987673**

**valData\_f1\_macro=0.9550053318960793**

valData\_confusion\_mat =

```
[[ 573    0    0    0    0    0    0    0    0    0    0    0    0
 0
    0]
 [   0  573    0    0    0    0    0    0    0    0    0    0    0
 0
    0]
 [   0    0  199    0   10    0    0    0    0    13    1    0    0
 3
    1]
 [   0    0    1 1409    0    0    0    0    0    0    0    5    0
 0
    0]
```

```

[ 0 0 3 0 50 1 0 0 0 2 1 0 0
0
2]
[ 0 0 0 0 0 255 0 0 0 0 0 0 0
0
11]
[ 0 0 0 0 0 0 107 0 0 0 0 0 0
0
0]
[ 0 0 0 15 0 0 0 485 0 1 0 0 67
0
0]
[ 0 0 0 0 0 0 0 0 35 0 0 0 0
0
0]
[ 0 0 0 0 1 1 0 0 0 1125 2 0 0
7
7]
[ 0 0 0 0 0 0 0 0 0 0 130 0 0
1
0]
[ 0 0 0 14 0 1 0 0 0 0 0 58 0
0
0]
[ 0 0 0 1 0 0 0 1 0 0 1 0 411
0
0]
[ 0 0 0 0 0 0 0 0 0 5 0 0 0
1542
4]
[ 0 0 0 3 0 5 0 0 0 3 1 0 0
0
641]]

```

# Metrics of test set:

testData\_sk\_accuracy=0.9788441692466461

testData\_classification\_rep =

	precision	recall	f1-score	support
<EOT>	1.00	1.00	1.00	586
<SOT>	1.00	1.00	1.00	586
adj	0.93	0.95	0.94	220
adp	0.99	1.00	1.00	1434
adv	0.87	0.70	0.77	76
aux	0.97	1.00	0.98	256

cconj	1.00	0.99	1.00	109
det	0.99	0.86	0.92	512
intj	1.00	1.00	1.00	36
noun	0.99	0.99	0.99	1166
num	0.94	0.98	0.96	127
part	0.98	0.95	0.96	56
pron	0.85	0.98	0.91	392
propn	0.99	0.99	0.99	1567
verb	0.99	0.97	0.98	629
<b>accuracy</b>			<b>0.98</b>	<b>7752</b>
<b>macro avg</b>	<b>0.97</b>	<b>0.96</b>	<b>0.96</b>	<b>7752</b>
<b>weighted avg</b>	<b>0.98</b>	<b>0.98</b>	<b>0.98</b>	<b>7752</b>

testData\_recall\_micro=0.9788441692466461

testData\_recall\_macro=0.9565420087002782

testData\_f1\_micro=0.9788441692466461

testData\_f1\_macro=0.9595992246174666

testData\_confusion\_mat =

```
[[ 586    0    0    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0  586    0    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0    0  208    0    4    0    0    0    0    0    3    1    0    0
3
    1]
[   0    0    0 1430    0    0    0    0    0    0    0    0    1    2
0
    1]
[   0    0   14    1   53    1    0    0    0    2    0    0    0    0
5
    0]
[   0    0    0    0    0  255    0    0    0    0    0    0    0    0
0
    1]
[   0    0    0    0    0    0  108    1    0    0    0    0    0    0
0
    0]
[   0    0    0    1    0    0    0  441    0    0    0    0    0    68
1
    1]
[   0    0    0    0    0    0    0    0    0  36    0    0    0    0
0
```

[illegible]

**Model : lstm\_model\_3.pt**

**Parameters:**

```
model = LSTM(32, 32, 1, len(trainData.vocab), len(trainData.tagVocab))  
optimizer = torch.optim.Adam(model.parameters(), lr=0.001)  
train_loss_list, valid_loss_list, train_accuracy_list,  
valid_accuracy_list = train(model, trainData, optimizer, criterion,  
valData, 4, "Test3")
```

Training model...

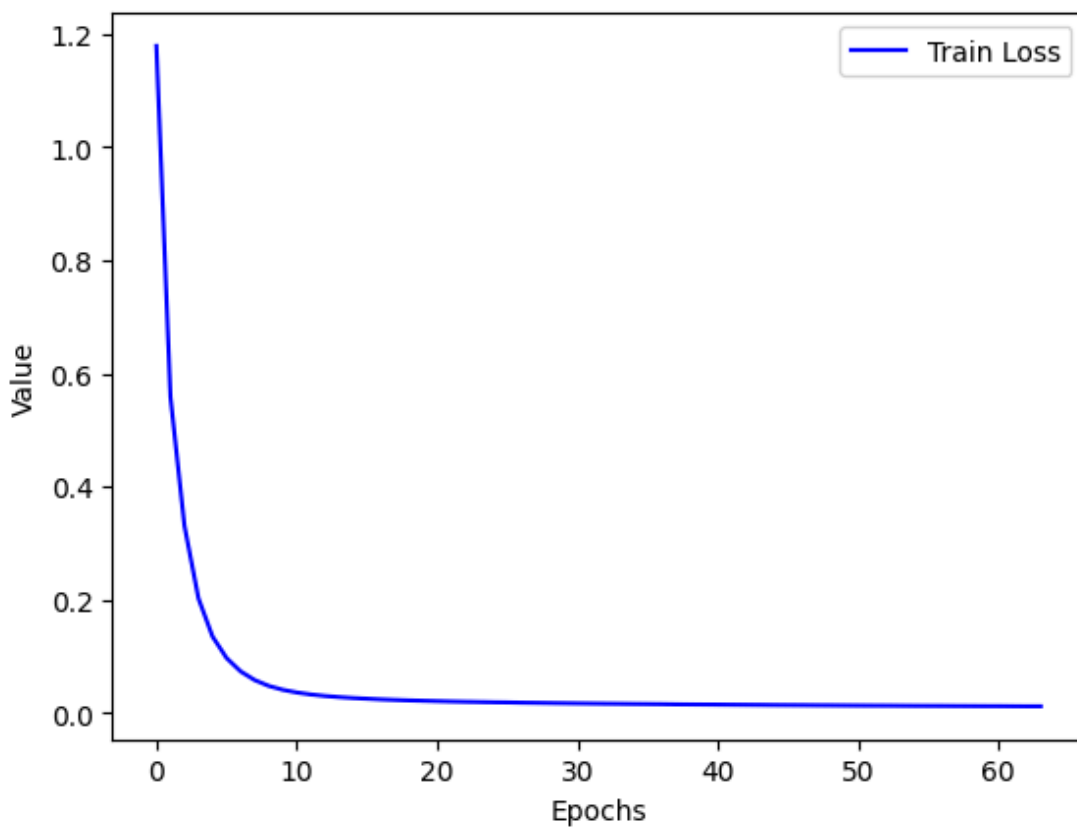
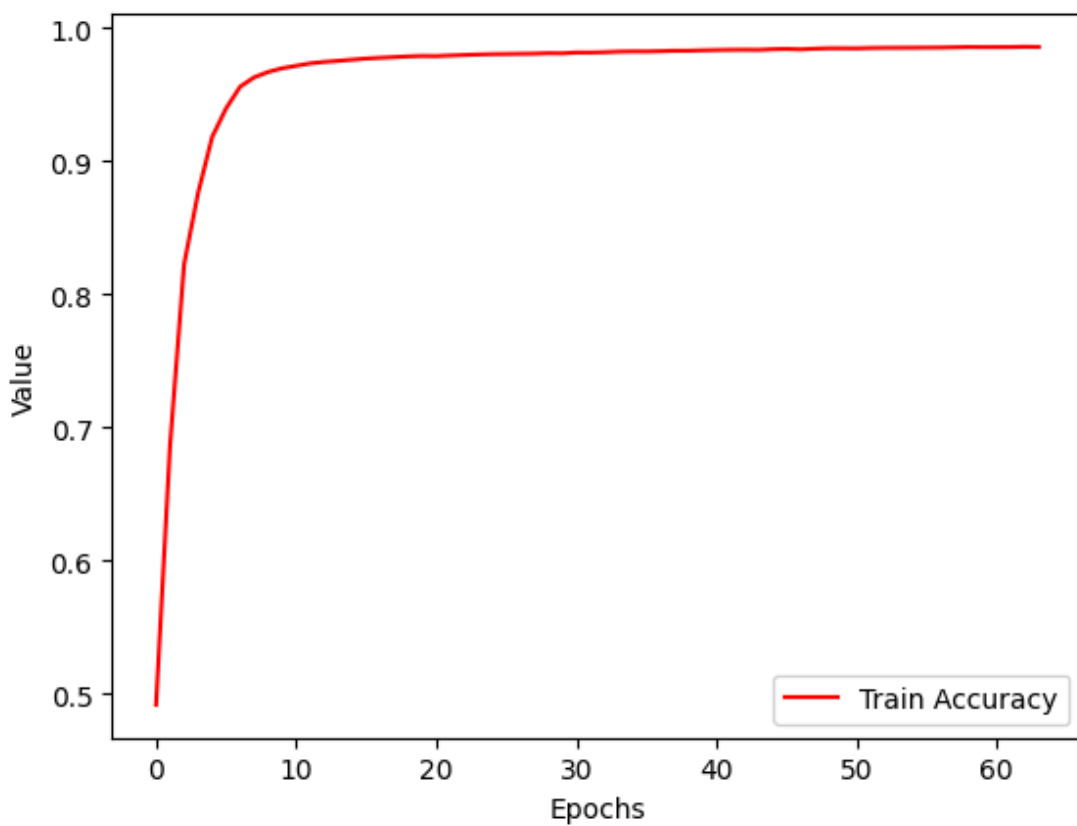
```
Validation loss: 0.8267819881439209  
Epoch 1 loss: 1.1788430547536308  
Validation loss: 0.4886995553970337  
Epoch 2 loss: 0.5591191226866707  
Validation loss: 0.295136034488678  
Epoch 3 loss: 0.33042034795924796  
Validation loss: 0.18995629251003265  
Epoch 4 loss: 0.20208776341890222  
Validation loss: 0.13233111798763275  
Epoch 5 loss: 0.1347428599733915  
Validation loss: 0.09854885190725327  
Epoch 6 loss: 0.09672956324335355  
Validation loss: 0.0768512487411499  
Epoch 7 loss: 0.07314069457908175  
Validation loss: 0.06284667551517487  
Epoch 8 loss: 0.057820455560377286  
Validation loss: 0.05355506390333176  
Epoch 9 loss: 0.047502332408703975  
Validation loss: 0.0473376139998436  
Epoch 10 loss: 0.04061129728137557  
Validation loss: 0.042440060526132584  
Epoch 11 loss: 0.03583047377751834  
Validation loss: 0.0392436683177948  
Epoch 12 loss: 0.032324917211350224  
Validation loss: 0.03660435602068901  
Epoch 13 loss: 0.029683746890957232  
Validation loss: 0.03498818352818489  
Epoch 14 loss: 0.027686630625889372  
Validation loss: 0.033405497670173645  
Epoch 15 loss: 0.02606684780582341  
Validation loss: 0.031818464398384094  
Epoch 16 loss: 0.02479759951942225  
Validation loss: 0.03159680590033531  
Epoch 17 loss: 0.023608263442987828  
Validation loss: 0.030357791110873222  
Epoch 18 loss: 0.022679021283150164
```

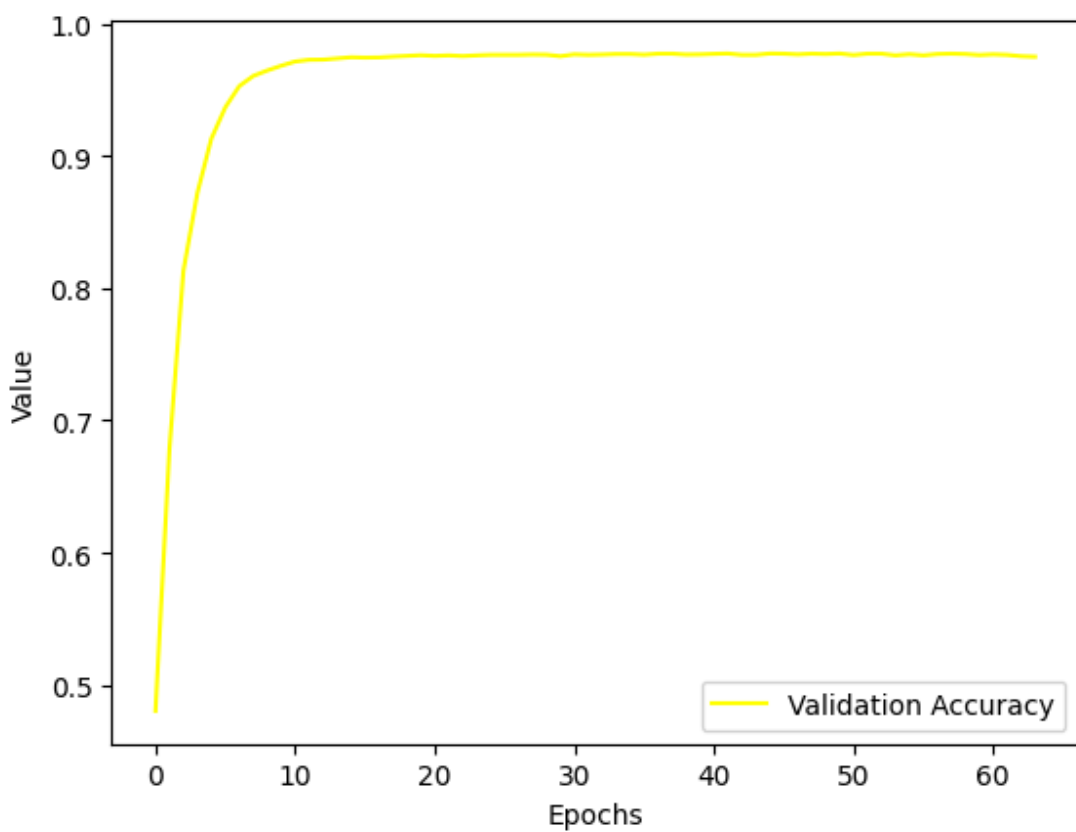
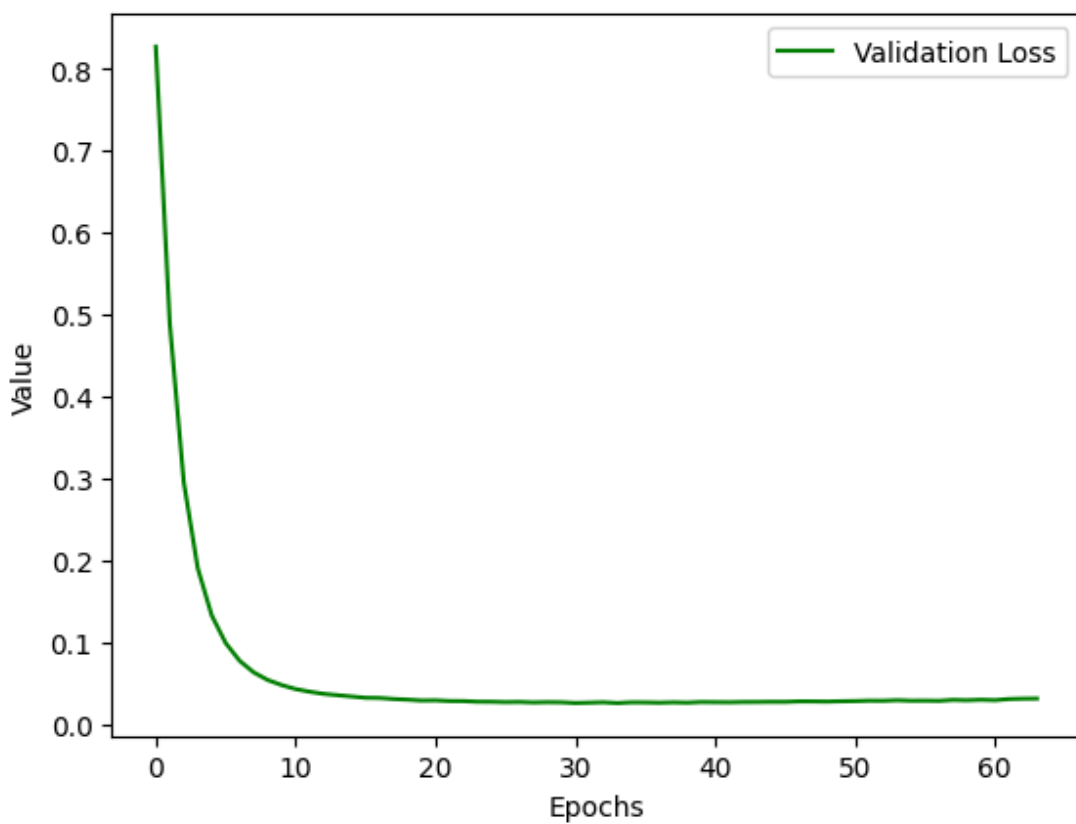
Validation loss: 0.029563358053565025  
Epoch 19 loss: 0.021862877296534047  
Validation loss: 0.028543829917907715  
Epoch 20 loss: 0.021176391802450168  
Validation loss: 0.02871806174516678  
Validation loss increased  
Epoch 21 loss: 0.020534275016233103  
Validation loss: 0.02789631485939026  
Epoch 22 loss: 0.01991789512781065  
Validation loss: 0.027775801718235016  
Epoch 23 loss: 0.01943285186747228  
Validation loss: 0.02702191285789013  
Epoch 24 loss: 0.01901908071517055  
Validation loss: 0.026976900175213814  
Epoch 25 loss: 0.018581606315643484  
Validation loss: 0.026539985090494156  
Epoch 26 loss: 0.018295775654155817  
Validation loss: 0.026751255616545677  
Validation loss increased  
Epoch 27 loss: 0.01782489790747971  
Validation loss: 0.02613043412566185  
Epoch 28 loss: 0.01753297629557661  
Validation loss: 0.02650010585784912  
Validation loss increased  
Epoch 29 loss: 0.017196768502107086  
Validation loss: 0.02629624679684639  
Epoch 30 loss: 0.016911642952585845  
Validation loss: 0.025424018502235413  
Epoch 31 loss: 0.016631542743225373  
Validation loss: 0.025847498327493668  
Validation loss increased  
Epoch 32 loss: 0.016386105230689715  
Validation loss: 0.026316218078136444  
Validation loss increased  
Epoch 33 loss: 0.016109455529767187  
Validation loss: 0.025360288098454475  
Epoch 34 loss: 0.015895636696646464  
Validation loss: 0.026251591742038727  
Validation loss increased  
Epoch 35 loss: 0.015636823199161175  
Validation loss: 0.026158278807997704  
Epoch 36 loss: 0.015469984599708844  
Validation loss: 0.025815628468990326  
Epoch 37 loss: 0.015253929355279056  
Validation loss: 0.026259060949087143



Validation loss increased  
Epoch 38 loss: 0.014901455112997054  
Validation loss: 0.025864901021122932  
Epoch 39 loss: 0.014717084185373205  
Validation loss: 0.02665800228714943  
Validation loss increased  
Epoch 40 loss: 0.014591307945744092  
Validation loss: 0.02648916095495224  
Epoch 41 loss: 0.014425085544752986  
Validation loss: 0.02635684609413147  
Epoch 42 loss: 0.014228231284016771  
Validation loss: 0.02668827772140503  
Validation loss increased  
Epoch 43 loss: 0.014047954471281437  
Validation loss: 0.026729028671979904  
Validation loss increased  
Epoch 44 loss: 0.013908561180109408  
Validation loss: 0.02691001445055008  
Validation loss increased  
Epoch 45 loss: 0.013688542278928321  
Validation loss: 0.026850424706935883  
Epoch 46 loss: 0.013564047125626855  
Validation loss: 0.027451658621430397  
Validation loss increased  
Epoch 47 loss: 0.01344645422980634  
Validation loss: 0.02739724889397621  
Epoch 48 loss: 0.013257850755575989  
Validation loss: 0.02720111981034279  
Epoch 49 loss: 0.013149918404532902  
Validation loss: 0.02758977562189102  
Validation loss increased  
Epoch 50 loss: 0.012963550099269  
Validation loss: 0.027819911018013954  
Validation loss increased  
Epoch 51 loss: 0.012824700655304451  
Validation loss: 0.028300490230321884  
Validation loss increased  
Epoch 52 loss: 0.012696115143680528  
Validation loss: 0.028203044086694717  
Epoch 53 loss: 0.012611193947299426  
Validation loss: 0.0288703516125679  
Validation loss increased  
Epoch 54 loss: 0.012467601381714887  
Validation loss: 0.028275655582547188  
Epoch 55 loss: 0.012309835083893876

Validation loss: 0.028335047885775566  
Validation loss increased  
Epoch 56 loss: 0.012215626040306776  
Validation loss: 0.028130531311035156  
Epoch 57 loss: 0.012145123837182104  
Validation loss: 0.02924933098256588  
Validation loss increased  
Epoch 58 loss: 0.012005117603702776  
Validation loss: 0.028833895921707153  
Epoch 59 loss: 0.011917717761791019  
Validation loss: 0.029429219663143158  
Validation loss increased  
Epoch 60 loss: 0.011805665036385407  
Validation loss: 0.02893529273569584  
Epoch 61 loss: 0.011763814669936451  
Validation loss: 0.030299536883831024  
Validation loss increased  
Epoch 62 loss: 0.011619369621013306  
Validation loss: 0.030742354691028595  
Validation loss increased  
Epoch 63 loss: 0.01147879188908125  
Validation loss: 0.030897025018930435  
Validation loss increased  
Epoch 64 loss: 0.011484069381354016





Loading model...

**Training Accuracy: 0.985910815110039**

**Metrics of val set:**

**valData\_sk\_accuracy=0.9748330765279918**

valData\_classification\_rep =

	precision	recall	f1-score	support
<EOT>	1.00	1.00	1.00	573
<SOT>	1.00	1.00	1.00	573
adj	0.98	0.89	0.93	227
adp	0.98	0.99	0.99	1415
adv	0.87	0.81	0.84	59
aux	0.99	0.94	0.96	266
cconj	1.00	1.00	1.00	107
det	0.99	0.87	0.93	568
intj	1.00	0.97	0.99	35
noun	0.98	0.99	0.98	1143
num	0.95	0.96	0.96	131
part	0.92	0.90	0.91	73
pron	0.86	0.99	0.92	414
propn	0.99	0.99	0.99	1551
verb	0.95	0.98	0.97	653
<b>accuracy</b>			<b>0.97</b>	<b>7788</b>
<b>macro avg</b>	<b>0.96</b>	<b>0.95</b>	<b>0.96</b>	<b>7788</b>
<b>weighted avg</b>	<b>0.98</b>	<b>0.97</b>	<b>0.97</b>	<b>7788</b>

**valData\_recall\_micro=0.9748330765279918**

**valData\_recall\_macro=0.9523248095715787**

**valData\_f1\_micro=0.9748330765279918**

**valData\_f1\_macro=0.9574733859194965**

valData\_confusion\_mat =

```
[[ 573    0    0    0    0    0    0    0    0    0    0    0    0
0
    0]
 [   0  573    0    0    0    0    0    0    0    0    0    0    0
0
    0]
 [   0    0  202    0    6    0    0    0    0    14    0    0    0
3
    2]
 [   0    0    0 1404    0    0    0    1    0    0    0    6    3
0
    1]
```

```

[ 0 0 4 1 48 0 0 0 0 1 1 0 1
0
3]
[ 0 0 0 0 0 249 0 0 0 0 0 0 0
0
17]
[ 0 0 0 0 0 0 107 0 0 0 0 0 0
0
0]
[ 0 0 0 13 0 0 0 494 0 0 1 0 60
0
0]
[ 0 0 0 0 0 0 0 0 34 0 0 0 0
0
1]
[ 0 0 0 0 1 0 0 0 0 1127 2 0 0
9
4]
[ 0 0 0 0 0 0 0 0 0 2 126 0 0
2
1]
[ 0 0 0 7 0 0 0 0 0 0 0 66 0
0
0]
[ 0 0 0 1 0 0 0 4 0 1 0 0 408
0
0]
[ 0 0 1 0 0 0 0 0 0 6 1 0 0
1540
3]
[ 0 0 0 3 0 3 0 0 0 3 1 0 0
2
641]]

```

# Metrics of test set:

testData\_sk\_accuracy=0.9780701754385965

testData\_classification\_rep =

	precision	recall	f1-score	support
<EOT>	1.00	1.00	1.00	586
<SOT>	1.00	1.00	1.00	586
adj	0.92	0.95	0.93	220
adp	1.00	1.00	1.00	1434
adv	0.87	0.70	0.77	76
aux	0.99	0.97	0.98	256

cconj	1.00	1.00	1.00	109
det	0.98	0.87	0.92	512
intj	1.00	1.00	1.00	36
noun	0.99	0.99	0.99	1166
num	0.97	0.93	0.95	127
part	0.98	0.98	0.98	56
pron	0.85	0.97	0.91	392
propn	0.99	0.99	0.99	1567
verb	0.96	0.99	0.98	629
<b>accuracy</b>			<b>0.98</b>	<b>7752</b>
<b>macro avg</b>	<b>0.97</b>	<b>0.96</b>	<b>0.96</b>	<b>7752</b>
<b>weighted avg</b>	<b>0.98</b>	<b>0.98</b>	<b>0.98</b>	<b>7752</b>

testData\_recall\_micro=0.9780701754385965

testData\_recall\_macro=0.9557698626049818

testData\_f1\_micro=0.9780701754385965

testData\_f1\_macro=0.9600307956029778

testData\_confusion\_mat =

```
[[ 586    0    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0  586    0    0    0    0    0    0    0    0    0    0    0
0
    0]
[   0    0  209    0    3    0    0    0    0    4    0    0    0
2
    2]
[   0    0    0 1430    0    0    0    0    0    0    0    1    2
0
    1]
[   0    0   15    1   53    1    0    0    0    2    0    0    0
3
    1]
[   0    0    0    0    0  249    0    0    0    0    0    0    0
0
    7]
[   0    0    0    0    0    0  109    0    0    0    0    0    0
0
    0]
[   0    0    0    2    0    0    0  445    0    0    0    0    63
1
    1]
[   0    0    0    0    0    0    0    0    36    0    0    0    0
0
```

[illegible]