## Jupytermain\_show

## December 12, 2020

```
[1]: #from A1 import A1_extract_landmarks as A1_extract
    from A1 import A1_functions_all as A1_functions
    Using TensorFlow backend.
[2]: TaskA1_opt_models_dict, TaskA1_res_dict = A1_functions.get_A1_results()
    -----Task A1: Gird searching on Logistic
    regression-----
    Logistic Regression Grid search CV on Dataset A:
    Training scores:
    0.500 (+/-0.004) for {'C': 0.0001, 'penalty': 'l1', 'solver': 'saga'}
    0.834 (+/-0.005) for {'C': 0.0001, 'penalty': '12', 'solver': 'saga'}
    0.501 (+/-0.003) for {'C': 0.001, 'penalty': 'l1', 'solver': 'saga'}
    0.873 (+/-0.008) for {'C': 0.001, 'penalty': '12', 'solver': 'saga'}
   0.855 (+/-0.009) for {'C': 0.01, 'penalty': 'l1', 'solver': 'saga'}
    0.903 (+/-0.006) for {'C': 0.01, 'penalty': '12', 'solver': 'saga'}
    0.916 (+/-0.005) for {'C': 0.1, 'penalty': 'l1', 'solver': 'saga'}
    0.922 (+/-0.005) for {'C': 0.1, 'penalty': '12', 'solver': 'saga'}
    0.928 (+/-0.009) for {'C': 1, 'penalty': 'l1', 'solver': 'saga'}
    0.928 (+/-0.009) for {'C': 1, 'penalty': 'l2', 'solver': 'saga'}
    0.929 (+/-0.010) for {'C': 10, 'penalty': 'l1', 'solver': 'saga'}
    0.929 (+/-0.010) for {'C': 10, 'penalty': '12', 'solver': 'saga'}
    Validation scores:
    0.499 (+/-0.004) for {'C': 0.0001, 'penalty': 'l1', 'solver': 'saga'}
    0.834 (+/-0.029) for {'C': 0.0001, 'penalty': '12', 'solver': 'saga'}
    0.501 (+/-0.003) for {'C': 0.001, 'penalty': 'l1', 'solver': 'saga'}
    0.871 (+/-0.042) for {'C': 0.001, 'penalty': '12', 'solver': 'saga'}
    0.854 (+/-0.031) for {'C': 0.01, 'penalty': 'l1', 'solver': 'saga'}
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0.897 (+/-0.032) for {'C': 0.01, 'penalty': '12', 'solver': 'saga'}
0.910 (+/-0.016) for {'C': 0.1, 'penalty': '11', 'solver': 'saga'}
0.914 (+/-0.021) for {'C': 0.1, 'penalty': '12', 'solver': 'saga'}
0.917 (+/-0.018) for {'C': 1, 'penalty': '11', 'solver': 'saga'}
0.918 (+/-0.021) for {'C': 1, 'penalty': '12', 'solver': 'saga'}
0.918 (+/-0.018) for {'C': 10, 'penalty': '11', 'solver': 'saga'}
0.918 (+/-0.018) for {'C': 10, 'penalty': '12', 'solver': 'saga'}
```

Prediction on a pseudo test set (split from Dataset A):

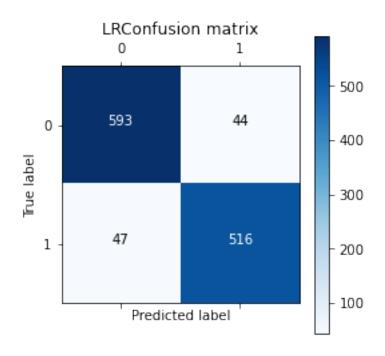
	precision	recall	II-score	support
0.0	0.93	0.93	0.93	637
1.0	0.92	0.92	0.92	563
accuracy			0.92	1200
macro avg	0.92	0.92	0.92	1200
weighted avg	0.92	0.92	0.92	1200

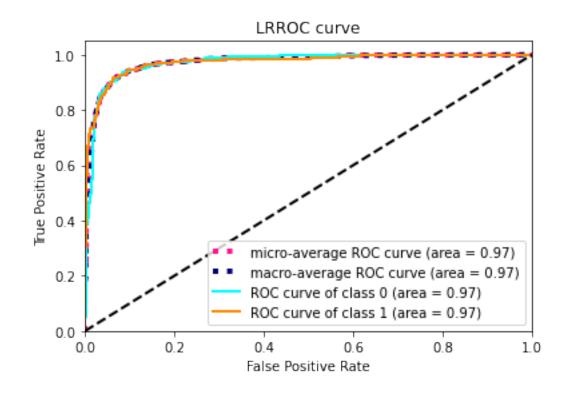
Accuracy: 0.924166666666667

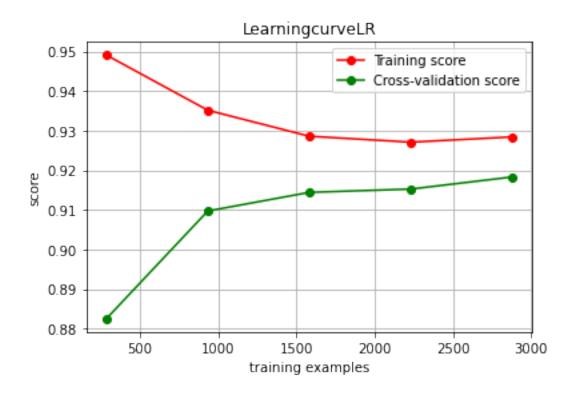
Best parameters found on Dataset A:

{'C': 1, 'penalty': '12', 'solver': 'saga'}

Average runtime per test instance: 6.55055046081543e-07







```
-----Task A1: Gird searching on
SVM Grid search CV on Dataset A:
Training scores:
0.502 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
0.891 (+/-0.010) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
0.502 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.502 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
0.891 (+/-0.010) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
0.502 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.502 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
0.891 (+/-0.010) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
0.502 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.502 (+/-0.000) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
0.891 (+/-0.010) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
0.661 (+/-0.041) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.502 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
0.917 (+/-0.005) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
0.502 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.502 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
0.917 (+/-0.005) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
0.502 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.801 (+/-0.008) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
0.917 (+/-0.005) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
0.822 (+/-0.007) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.502 (+/-0.000) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
0.917 (+/-0.005) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}
0.633 (+/-0.012) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.504 (+/-0.002) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.932 (+/-0.005) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
0.502 (+/-0.000) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.856 (+/-0.007) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
0.932 (+/-0.005) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
0.840 (+/-0.006) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.886 (+/-0.008) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
0.932 (+/-0.005) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
0.752 (+/-0.010) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.521 (+/-0.004) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
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0.932 (+/-0.005) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}

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0.602 (+/-0.032) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.860 (+/-0.005) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.939 (+/-0.005) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
0.841 (+/-0.006) for {'C': 1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.904 (+/-0.006) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.939 (+/-0.005) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
0.891 (+/-0.009) for {'C': 1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.950 (+/-0.004) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.939 (+/-0.005) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
0.695 (+/-0.013) for {'C': 1, 'gamma': 0.01, 'kernel': 'sigmoid'}
1.000 (+/-0.000) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.939 (+/-0.005) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.597 (+/-0.035) for {'C': 1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.901 (+/-0.009) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.942 (+/-0.003) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.889 (+/-0.007) for {'C': 10, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.937 (+/-0.004) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.942 (+/-0.003) for {'C': 10, 'gamma': 0.001, 'kernel': 'linear'}
0.881 (+/-0.018) for {'C': 10, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.992 (+/-0.002) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.942 (+/-0.003) for {'C': 10, 'gamma': 0.01, 'kernel': 'linear'}
0.689 (+/-0.011) for {'C': 10, 'gamma': 0.01, 'kernel': 'sigmoid'}
1.000 (+/-0.000) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.942 (+/-0.003) for {'C': 10, 'gamma': 0.1, 'kernel': 'linear'}
0.598 (+/-0.035) for {'C': 10, 'gamma': 0.1, 'kernel': 'sigmoid'}
Validation scores:
0.502 (+/-0.001) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
0.889 (+/-0.039) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
0.502 (+/-0.001) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.502 (+/-0.001) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
0.889 (+/-0.039) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
0.502 (+/-0.001) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.502 (+/-0.001) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
0.889 (+/-0.039) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
0.502 (+/-0.001) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.502 (+/-0.001) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
0.889 (+/-0.039) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
0.665 (+/-0.050) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.502 (+/-0.001) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
0.911 (+/-0.023) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
0.502 (+/-0.001) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.502 (+/-0.001) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
0.911 (+/-0.023) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
0.502 (+/-0.001) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.797 (+/-0.026) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
0.911 (+/-0.023) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
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0.821 (+/-0.021) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.502 (+/-0.001) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
0.911 (+/-0.023) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}
0.632 (+/-0.028) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.504 (+/-0.003) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.917 (+/-0.025) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
0.502 (+/-0.001) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.854 (+/-0.031) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
0.917 (+/-0.025) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
0.842 (+/-0.023) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.869 (+/-0.026) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
0.917 (+/-0.025) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
0.753 (+/-0.035) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.512 (+/-0.006) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
0.917 (+/-0.025) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}
0.593 (+/-0.043) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.860 (+/-0.026) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.916 (+/-0.021) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
0.842 (+/-0.021) for {'C': 1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.897 (+/-0.033) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.916 (+/-0.021) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
0.887 (+/-0.035) for {'C': 1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.901 (+/-0.021) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.916 (+/-0.021) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
0.696 (+/-0.043) for {'C': 1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.813 (+/-0.025) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.916 (+/-0.021) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.591 (+/-0.043) for {'C': 1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.897 (+/-0.033) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.914 (+/-0.011) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.890 (+/-0.032) for {'C': 10, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.914 (+/-0.030) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.914 (+/-0.011) for {'C': 10, 'gamma': 0.001, 'kernel': 'linear'}
0.874 (+/-0.043) for {'C': 10, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.903 (+/-0.018) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.914 (+/-0.011) for {'C': 10, 'gamma': 0.01, 'kernel': 'linear'}
0.688 (+/-0.045) for {'C': 10, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.818 (+/-0.029) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.914 (+/-0.011) for {'C': 10, 'gamma': 0.1, 'kernel': 'linear'}
0.591 (+/-0.045) for {'C': 10, 'gamma': 0.1, 'kernel': 'sigmoid'}
Prediction on a pseudo test set (split from Dataset A):
              precision
                           recall f1-score
                                              support
         0.0
                   0.93
                             0.93
                                       0.93
                                                  637
         1.0
                   0.92
                             0.92
                                       0.92
                                                  563
                                       0.93
                                                 1200
    accuracy
```

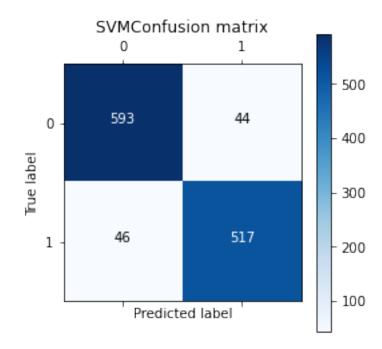
macro	avg	0.92	0.92	0.92	1200
weighted	avg	0.93	0.93	0.93	1200

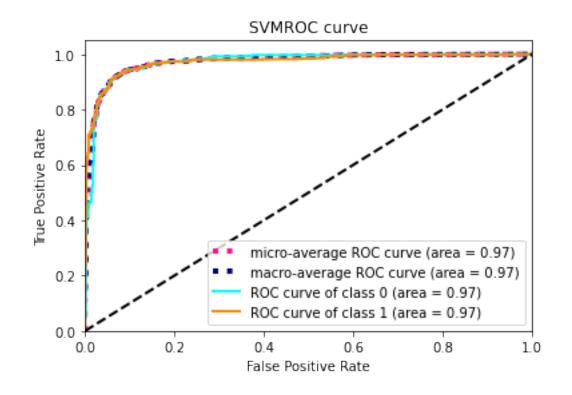
Accuracy: 0.925

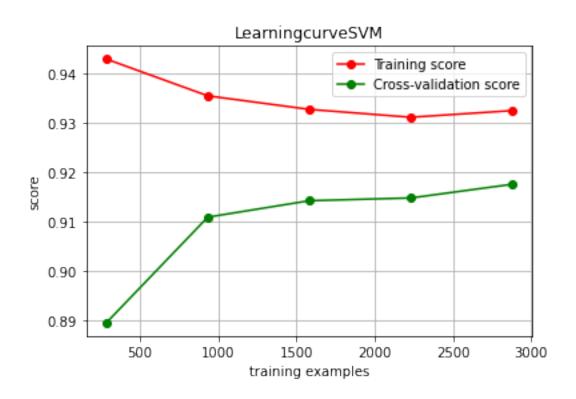
Best parameters found on Dataset A:

{'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}

Average runtime per test instance: 0.0001966351270675659







```
Forest-----
RF Grid search CV on Dataset A:
Training scores:
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 512}
1.000 (+/-0.000) for {'max depth': 128, 'n estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 512}
1.000 \ (+/-0.000) for {'max depth': 1024, 'n estimators': 1024}
Validation scores:
0.862 (+/-0.017) for {'max_depth': 64, 'n_estimators': 64}
0.864 (+/-0.026) for {'max_depth': 64, 'n_estimators': 128}
0.864 (+/-0.035) for {'max_depth': 64, 'n_estimators': 256}
0.864 (+/-0.024) for {'max_depth': 64, 'n_estimators': 512}
0.865 (+/-0.024) for {'max_depth': 64, 'n_estimators': 1024}
0.858 (+/-0.019) for {'max_depth': 128, 'n_estimators': 64}
0.861 (+/-0.032) for {'max_depth': 128, 'n_estimators': 128}
0.864 (+/-0.023) for {'max_depth': 128, 'n_estimators': 256}
0.866 (+/-0.034) for {'max_depth': 128, 'n_estimators': 512}
0.868 (+/-0.028) for {'max_depth': 128, 'n_estimators': 1024}
```

-----Task A1: Gird searching on Rando

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0.859 (+/-0.027) for {'max_depth': 256, 'n_estimators': 64}
0.865 (+/-0.025) for {'max_depth': 256, 'n_estimators': 128}
0.866 (+/-0.027) for {'max_depth': 256, 'n_estimators': 256}
0.867 (+/-0.024) for {'max_depth': 256, 'n_estimators': 512}
0.864 (+/-0.028) for {'max depth': 256, 'n estimators': 1024}
0.863 (+/-0.026) for {'max_depth': 512, 'n_estimators': 64}
0.861 \ (+/-0.019)  for {'max depth': 512, 'n estimators': 128}
0.865 (+/-0.032) for {'max_depth': 512, 'n_estimators': 256}
0.866 (+/-0.024) for {'max_depth': 512, 'n_estimators': 512}
0.867 \ (+/-0.027) \ for \ {'max_depth': 512, 'n_estimators': 1024}
0.862 (+/-0.019) for {'max_depth': 1024, 'n_estimators': 64}
0.863 (+/-0.028) for {'max_depth': 1024, 'n_estimators': 128}
0.864 (+/-0.023) for {'max_depth': 1024, 'n_estimators': 256}
0.863 \ (+/-0.029) \ for \ {\rm 'max\_depth': 1024, 'n\_estimators': 512}
0.867 \ (+/-0.024) \ for {'max_depth': 1024, 'n_estimators': 1024}
```

## Prediction on a pseudo test set (split from Dataset A): precision recall f1-score

support

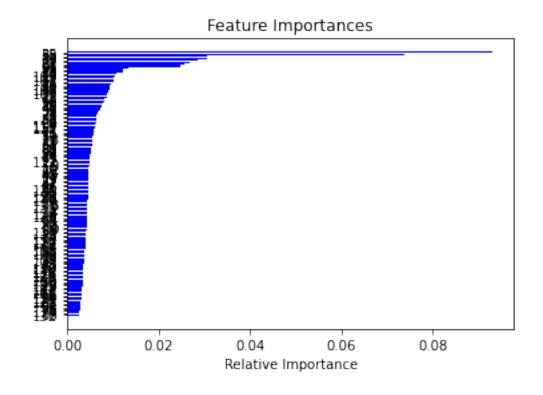
0.0	0.87	0.90	0.89	637
1.0	0.88	0.85	0.86	563
accuracy			0.88	1200
macro avg	0.88	0.87	0.88	1200
weighted avg	0.88	0.88	0.88	1200

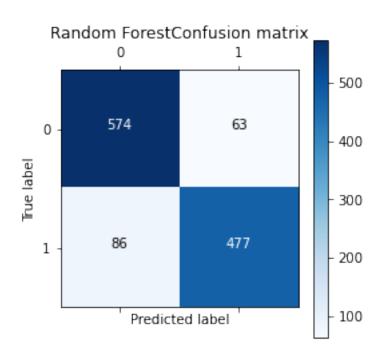
Accuracy: 0.8758333333333334

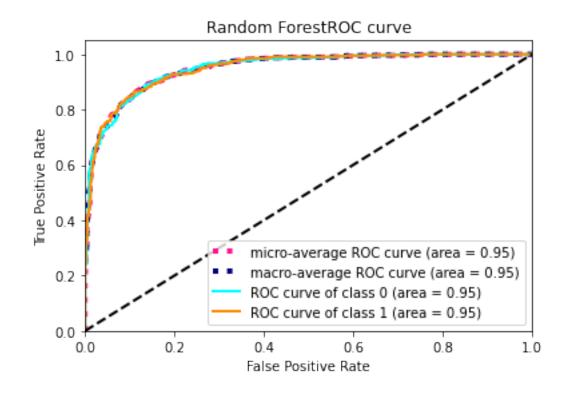
Best parameters found on Dataset A:

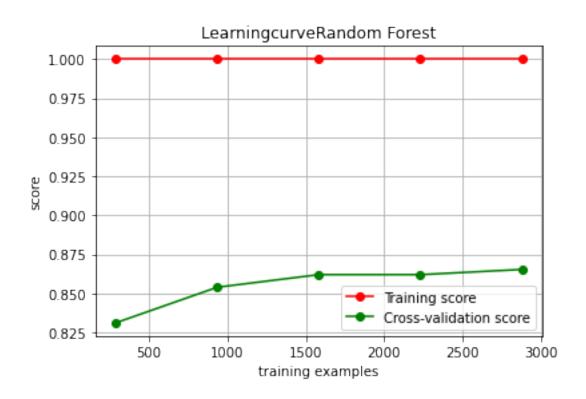
{'max\_depth': 128, 'n\_estimators': 1024}

Average runtime per test instance: 0.000469089945157369









-----Task A1: Gird searching on KNN-----\_\_\_\_\_\_ KNN Grid search CV on Dataset A: Training scores: 0.840 (+/-0.007) for {'n\_neighbors': 8} 0.836 (+/-0.009) for {'n\_neighbors': 16}  $0.830 (+/-0.009) for {'n_neighbors': 32}$ 0.820 (+/-0.007) for {'n\_neighbors': 64}  $0.815 \ (+/-0.005) \ for \ \{'n_neighbors': 128\}$ Validation scores: 0.799 (+/-0.021) for {'n\_neighbors': 8}  $0.808 (+/-0.010) for {'n_neighbors': 16}$  $0.812 (+/-0.022) for {'n_neighbors': 32}$ 0.812 (+/-0.019) for {'n\_neighbors': 64}  $0.809 (+/-0.009) for {'n_neighbors': 128}$ precision recall f1-score support

Prediction on a pseudo test set (split from Dataset A):

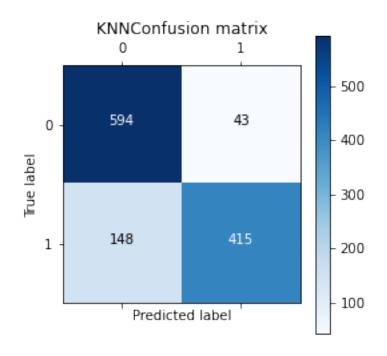
	F			
0.0	0.80	0.93	0.86	637
1.0	0.91	0.74	0.81	563
accuracy			0.84	1200
macro avg	0.85	0.83	0.84	1200
weighted avg	0.85	0.84	0.84	1200

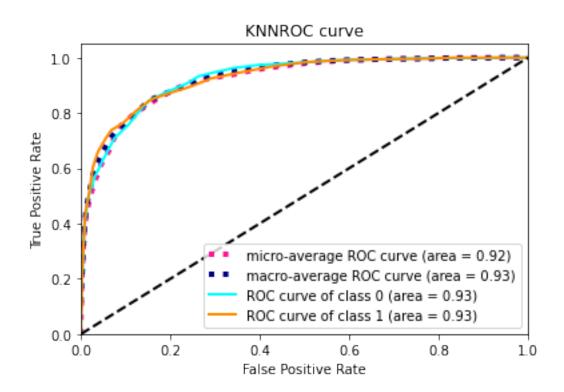
Accuracy: 0.8408333333333333

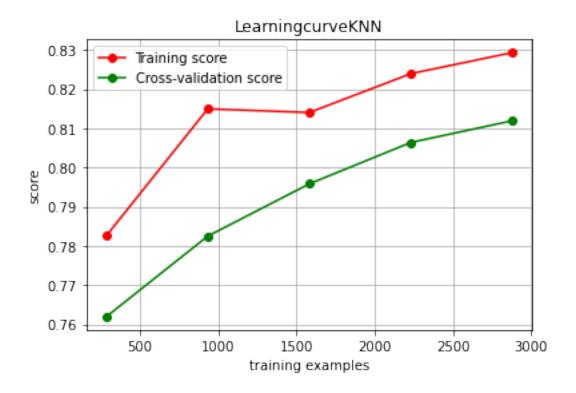
Best parameters found on Dataset A:

{'n\_neighbors': 32}

Average runtime per test instance: 0.0046141950289408365







Task	A1:	Gird	searching	on
MT.P				

## 

Squential with Kfold CV:

Training for fold 1  $\dots$ 

Adding layer 1:

Adding layer 2:

Adding layer 3:

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
dense_1 (Dense)	(None, 2176)	298112
dense_2 (Dense)	(None, 1088)	2368576
dropout_1 (Dropout)	(None, 1088)	0
dense_3 (Dense)	(None, 544)	592416

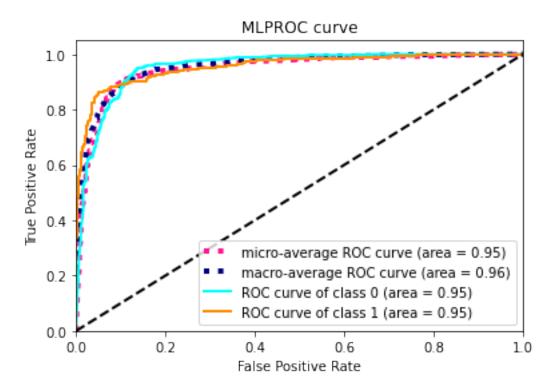
```
dropout_2 (Dropout) (None, 544)
-----
dense_4 (Dense)
                     (None, 272)
                                          148240
_____
dropout 3 (Dropout)
                 (None, 272)
                                          0
_____
dense_5 (Dense) (None, 2)
                                          546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
WARNING:tensorflow:From /Users/apple/opt/anaconda3/envs/AMLS/lib/python3.6/site-
packages/keras/backend/tensorflow_backend.py:422: The name tf.global_variables
is deprecated. Please use tf.compat.v1.global_variables instead.
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.5507 - val_loss: 0.5097 - val_accuracy: 0.7500
2880/2880 [============= ] - 2s 818us/step - loss: 0.4809 -
accuracy: 0.7830 - val_loss: 0.3649 - val_accuracy: 0.8389
Epoch 3/50
2880/2880 [============== ] - 2s 686us/step - loss: 0.3193 -
accuracy: 0.8653 - val_loss: 0.3237 - val_accuracy: 0.8528
Epoch 4/50
2880/2880 [============== ] - 3s 880us/step - loss: 0.3329 -
accuracy: 0.8618 - val_loss: 0.2598 - val_accuracy: 0.8972
Epoch 5/50
2880/2880 [============= ] - 3s 884us/step - loss: 0.2962 -
accuracy: 0.8806 - val_loss: 0.3196 - val_accuracy: 0.8611
Epoch 6/50
2880/2880 [============= ] - 2s 867us/step - loss: 0.2727 -
accuracy: 0.8899 - val loss: 0.2622 - val accuracy: 0.8792
Epoch 7/50
2880/2880 [============= ] - 3s 883us/step - loss: 0.2528 -
accuracy: 0.8993 - val_loss: 0.2832 - val_accuracy: 0.8861
Epoch 8/50
2880/2880 [============= ] - 3s 881us/step - loss: 0.2608 -
accuracy: 0.8899 - val_loss: 0.2523 - val_accuracy: 0.9111
Epoch 9/50
2880/2880 [============= ] - 3s 882us/step - loss: 0.2550 -
accuracy: 0.8941 - val_loss: 0.2370 - val_accuracy: 0.9111
Epoch 10/50
2880/2880 [============== ] - 3s 885us/step - loss: 0.2144 -
accuracy: 0.9142 - val_loss: 0.2486 - val_accuracy: 0.9069
```

Epoch 11/50

```
2880/2880 [============== ] - 3s 880us/step - loss: 0.2312 -
accuracy: 0.9080 - val_loss: 0.2402 - val_accuracy: 0.9000
Epoch 12/50
2880/2880 [============= - - 2s 854us/step - loss: 0.2066 -
accuracy: 0.9139 - val_loss: 0.2729 - val_accuracy: 0.8736
Epoch 13/50
2880/2880 [============ ] - 3s 887us/step - loss: 0.2057 -
accuracy: 0.9167 - val_loss: 0.4767 - val_accuracy: 0.8236
Epoch 14/50
2880/2880 [============= ] - 2s 868us/step - loss: 0.2402 -
accuracy: 0.9017 - val_loss: 0.2382 - val_accuracy: 0.9111
Epoch 15/50
2880/2880 [============= ] - 3s 873us/step - loss: 0.1966 -
accuracy: 0.9139 - val_loss: 0.2550 - val_accuracy: 0.9056
Epoch 16/50
2880/2880 [============ ] - 3s 879us/step - loss: 0.1745 -
accuracy: 0.9271 - val_loss: 0.3941 - val_accuracy: 0.8625
Epoch 17/50
2880/2880 [============= ] - 3s 872us/step - loss: 0.2009 -
accuracy: 0.9174 - val_loss: 0.2450 - val_accuracy: 0.9153
Epoch 18/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.1792 -
accuracy: 0.9271 - val_loss: 0.2545 - val_accuracy: 0.9014
Epoch 19/50
2880/2880 [============= ] - 3s 870us/step - loss: 0.1948 -
accuracy: 0.9243 - val_loss: 0.3225 - val_accuracy: 0.8861
Epoch 20/50
2880/2880 [============== ] - 2s 862us/step - loss: 0.1818 -
accuracy: 0.9243 - val_loss: 0.2759 - val_accuracy: 0.8972
Epoch 21/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.1562 -
accuracy: 0.9354 - val_loss: 0.2606 - val_accuracy: 0.9181
Epoch 22/50
2880/2880 [============= ] - 3s 876us/step - loss: 0.1750 -
accuracy: 0.9198 - val loss: 0.2838 - val accuracy: 0.8611
Epoch 23/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.1516 -
accuracy: 0.9354 - val_loss: 0.2892 - val_accuracy: 0.8917
Epoch 24/50
2880/2880 [============= ] - 3s 901us/step - loss: 0.1655 -
accuracy: 0.9236 - val_loss: 0.2235 - val_accuracy: 0.9250
Epoch 25/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.1354 -
accuracy: 0.9413 - val_loss: 0.2775 - val_accuracy: 0.9208
Epoch 26/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.1333 -
accuracy: 0.9413 - val_loss: 0.3202 - val_accuracy: 0.9056
Epoch 27/50
```

```
2880/2880 [============== ] - 3s 890us/step - loss: 0.1439 -
accuracy: 0.9413 - val_loss: 0.2804 - val_accuracy: 0.8806
Epoch 28/50
2880/2880 [============= ] - 3s 930us/step - loss: 0.1648 -
accuracy: 0.9285 - val_loss: 0.2332 - val_accuracy: 0.9181
Epoch 29/50
2880/2880 [============= ] - 3s 1ms/step - loss: 0.1330 -
accuracy: 0.9451 - val_loss: 0.2763 - val_accuracy: 0.9264
Epoch 30/50
accuracy: 0.9469 - val_loss: 0.2570 - val_accuracy: 0.9139
Epoch 31/50
2880/2880 [============= ] - 3s 900us/step - loss: 0.1188 -
accuracy: 0.9493 - val_loss: 0.3567 - val_accuracy: 0.9042
Epoch 32/50
2880/2880 [============ ] - 3s 896us/step - loss: 0.1458 -
accuracy: 0.9431 - val_loss: 0.2766 - val_accuracy: 0.8972
Epoch 33/50
2880/2880 [============= - 3s 918us/step - loss: 0.1171 -
accuracy: 0.9524 - val_loss: 0.3100 - val_accuracy: 0.9056
Epoch 34/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.1314 -
accuracy: 0.9479 - val_loss: 0.3310 - val_accuracy: 0.9028
Epoch 35/50
2880/2880 [============= ] - 3s 893us/step - loss: 0.1063 -
accuracy: 0.9552 - val_loss: 0.3036 - val_accuracy: 0.9181
Epoch 36/50
2880/2880 [============== ] - 3s 893us/step - loss: 0.1192 -
accuracy: 0.9538 - val_loss: 0.2530 - val_accuracy: 0.9319
Epoch 37/50
2880/2880 [============= ] - 3s 892us/step - loss: 0.1098 -
accuracy: 0.9576 - val_loss: 0.2738 - val_accuracy: 0.9222
Epoch 38/50
2880/2880 [============ ] - 3s 904us/step - loss: 0.0863 -
accuracy: 0.9646 - val loss: 0.3780 - val accuracy: 0.9056
Epoch 39/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.1110 -
accuracy: 0.9545 - val_loss: 0.3042 - val_accuracy: 0.9292
Epoch 40/50
2880/2880 [============= ] - 3s 894us/step - loss: 0.1281 -
accuracy: 0.9451 - val_loss: 0.2534 - val_accuracy: 0.9139
Epoch 41/50
2880/2880 [============= ] - 3s 922us/step - loss: 0.0776 -
accuracy: 0.9681 - val_loss: 0.3789 - val_accuracy: 0.9069
Epoch 42/50
2880/2880 [============= ] - 3s 910us/step - loss: 0.1437 -
accuracy: 0.9399 - val_loss: 0.3091 - val_accuracy: 0.9222
Epoch 43/50
```

```
2880/2880 [============== ] - 3s 887us/step - loss: 0.0927 -
accuracy: 0.9646 - val_loss: 0.3102 - val_accuracy: 0.9194
Epoch 44/50
2880/2880 [============= ] - 3s 869us/step - loss: 0.1175 -
accuracy: 0.9597 - val_loss: 0.2746 - val_accuracy: 0.9222
Epoch 45/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.0745 -
accuracy: 0.9708 - val_loss: 0.4392 - val_accuracy: 0.8806
Epoch 46/50
                     =========] - 3s 881us/step - loss: 0.1065 -
2880/2880 [=======
accuracy: 0.9535 - val_loss: 0.3760 - val_accuracy: 0.9069
Epoch 47/50
2880/2880 [============= ] - 3s 882us/step - loss: 0.1001 -
accuracy: 0.9632 - val_loss: 0.3186 - val_accuracy: 0.9139
Epoch 48/50
2880/2880 [============= ] - 3s 875us/step - loss: 0.0964 -
accuracy: 0.9656 - val_loss: 0.2979 - val_accuracy: 0.9208
Epoch 49/50
2880/2880 [============= ] - 3s 883us/step - loss: 0.0875 -
accuracy: 0.9646 - val_loss: 0.2685 - val_accuracy: 0.9083
Epoch 50/50
2880/2880 [============== ] - 3s 894us/step - loss: 0.0611 -
accuracy: 0.9792 - val_loss: 0.5453 - val_accuracy: 0.9000
1200/1200 [=========== ] - 1s 562us/step
```

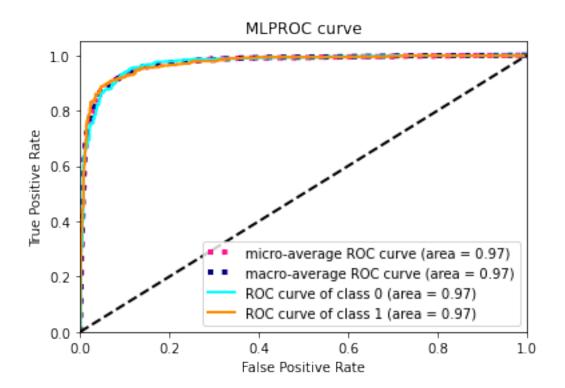


```
Squential with Kfold CV:
Training for fold 2 ...
Adding layer 1:
Adding layer 2:
Adding layer 3:
Model: "sequential 2"
._____
Layer (type)
                  Output Shape
                                     Param #
______
                   (None, 2176)
dense_6 (Dense)
                                      298112
                  (None, 1088)
dense_7 (Dense)
                                     2368576
 -----
dropout_4 (Dropout) (None, 1088)
dense_8 (Dense)
              (None, 544)
                                    592416
_____
dropout_5 (Dropout) (None, 544) 0
     _____
dense_9 (Dense)
                  (None, 272)
                                     148240
-----
               (None, 272)
dropout_6 (Dropout)
dense_10 (Dense) (None, 2) 546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============= ] - 8s 3ms/step - loss: 1.5425 -
accuracy: 0.5771 - val_loss: 0.4574 - val_accuracy: 0.7903
Epoch 2/50
2880/2880 [============= ] - 3s 883us/step - loss: 0.4389 -
accuracy: 0.8024 - val loss: 0.3372 - val accuracy: 0.8611
Epoch 3/50
2880/2880 [============== ] - 3s 881us/step - loss: 0.3690 -
accuracy: 0.8434 - val_loss: 0.3014 - val_accuracy: 0.8736
Epoch 4/50
2880/2880 [============= ] - 3s 895us/step - loss: 0.2717 -
accuracy: 0.8896 - val_loss: 0.2941 - val_accuracy: 0.8875
Epoch 5/50
2880/2880 [============= ] - 3s 883us/step - loss: 0.3056 -
accuracy: 0.8722 - val_loss: 0.3069 - val_accuracy: 0.8806
Epoch 6/50
2880/2880 [============= ] - 3s 892us/step - loss: 0.2445 -
```

```
accuracy: 0.8976 - val_loss: 0.3102 - val_accuracy: 0.8861
Epoch 7/50
2880/2880 [============= ] - 3s 886us/step - loss: 0.2244 -
accuracy: 0.9083 - val_loss: 0.3877 - val_accuracy: 0.8653
Epoch 8/50
2880/2880 [============= ] - 3s 879us/step - loss: 0.2834 -
accuracy: 0.8819 - val_loss: 0.3984 - val_accuracy: 0.8611
Epoch 9/50
2880/2880 [============= ] - 3s 883us/step - loss: 0.2256 -
accuracy: 0.9080 - val_loss: 0.2667 - val_accuracy: 0.8944
Epoch 10/50
2880/2880 [============== ] - 3s 888us/step - loss: 0.2251 -
accuracy: 0.9059 - val_loss: 0.3212 - val_accuracy: 0.8542
Epoch 11/50
2880/2880 [============= ] - 3s 876us/step - loss: 0.2231 -
accuracy: 0.9080 - val_loss: 0.2993 - val_accuracy: 0.9056
Epoch 12/50
2880/2880 [============= ] - 2s 864us/step - loss: 0.2031 -
accuracy: 0.9181 - val_loss: 0.3070 - val_accuracy: 0.8764
Epoch 13/50
2880/2880 [============= ] - 3s 896us/step - loss: 0.2015 -
accuracy: 0.9233 - val_loss: 0.3007 - val_accuracy: 0.8917
Epoch 14/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.2091 -
accuracy: 0.9149 - val_loss: 0.2467 - val_accuracy: 0.8958
Epoch 15/50
2880/2880 [============== ] - 3s 893us/step - loss: 0.1749 -
accuracy: 0.9323 - val_loss: 0.2856 - val_accuracy: 0.9014
2880/2880 [============= ] - 2s 865us/step - loss: 0.2114 -
accuracy: 0.9108 - val_loss: 0.3120 - val_accuracy: 0.8694
Epoch 17/50
2880/2880 [============= ] - 3s 886us/step - loss: 0.1698 -
accuracy: 0.9264 - val_loss: 0.2943 - val_accuracy: 0.8903
Epoch 18/50
2880/2880 [============= - 3s 878us/step - loss: 0.2003 -
accuracy: 0.9240 - val loss: 0.2649 - val accuracy: 0.9000
Epoch 19/50
2880/2880 [============= ] - 3s 884us/step - loss: 0.1537 -
accuracy: 0.9375 - val_loss: 0.2767 - val_accuracy: 0.8972
Epoch 20/50
2880/2880 [============= ] - 3s 898us/step - loss: 0.1651 -
accuracy: 0.9309 - val_loss: 0.3674 - val_accuracy: 0.8764
Epoch 21/50
2880/2880 [============ ] - 3s 881us/step - loss: 0.1532 -
accuracy: 0.9378 - val_loss: 0.3211 - val_accuracy: 0.9056
Epoch 22/50
2880/2880 [============= ] - 3s 875us/step - loss: 0.1690 -
```

```
accuracy: 0.9288 - val_loss: 0.3288 - val_accuracy: 0.8722
Epoch 23/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.1553 -
accuracy: 0.9347 - val_loss: 0.3175 - val_accuracy: 0.8819
Epoch 24/50
2880/2880 [============= ] - 3s 896us/step - loss: 0.1598 -
accuracy: 0.9347 - val_loss: 0.3071 - val_accuracy: 0.9083
Epoch 25/50
2880/2880 [============= ] - 3s 878us/step - loss: 0.1572 -
accuracy: 0.9333 - val_loss: 0.3806 - val_accuracy: 0.8931
Epoch 26/50
2880/2880 [============== ] - 3s 888us/step - loss: 0.1491 -
accuracy: 0.9344 - val_loss: 0.3273 - val_accuracy: 0.8931
Epoch 27/50
2880/2880 [============= ] - 3s 897us/step - loss: 0.1262 -
accuracy: 0.9444 - val_loss: 0.4130 - val_accuracy: 0.9153
Epoch 28/50
2880/2880 [============= ] - 2s 864us/step - loss: 0.1872 -
accuracy: 0.9278 - val_loss: 0.3045 - val_accuracy: 0.9000
Epoch 29/50
2880/2880 [============= ] - 3s 914us/step - loss: 0.1225 -
accuracy: 0.9497 - val_loss: 0.3550 - val_accuracy: 0.9042
Epoch 30/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.1507 -
accuracy: 0.9347 - val_loss: 0.3379 - val_accuracy: 0.9014
Epoch 31/50
2880/2880 [============== ] - 3s 885us/step - loss: 0.1158 -
accuracy: 0.9503 - val_loss: 0.3438 - val_accuracy: 0.8806
2880/2880 [============= ] - 3s 874us/step - loss: 0.1147 -
accuracy: 0.9514 - val_loss: 0.4197 - val_accuracy: 0.9042
Epoch 33/50
2880/2880 [============= ] - 3s 910us/step - loss: 0.1137 -
accuracy: 0.9542 - val_loss: 0.3728 - val_accuracy: 0.9000
Epoch 34/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.1274 -
accuracy: 0.9458 - val loss: 0.4012 - val accuracy: 0.8903
Epoch 35/50
2880/2880 [============== ] - 3s 899us/step - loss: 0.0949 -
accuracy: 0.9646 - val_loss: 0.4547 - val_accuracy: 0.9181
Epoch 36/50
2880/2880 [============= ] - 3s 900us/step - loss: 0.1530 -
accuracy: 0.9410 - val_loss: 0.3124 - val_accuracy: 0.9083
Epoch 37/50
2880/2880 [============ ] - 3s 891us/step - loss: 0.0786 -
accuracy: 0.9698 - val_loss: 0.4749 - val_accuracy: 0.8875
Epoch 38/50
2880/2880 [============= ] - 3s 877us/step - loss: 0.1538 -
```

```
accuracy: 0.9427 - val_loss: 0.2819 - val_accuracy: 0.9139
Epoch 39/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.1122 -
accuracy: 0.9611 - val_loss: 0.2752 - val_accuracy: 0.8958
Epoch 40/50
2880/2880 [============= ] - 3s 882us/step - loss: 0.1202 -
accuracy: 0.9552 - val_loss: 0.3065 - val_accuracy: 0.8875
Epoch 41/50
2880/2880 [============= ] - 3s 892us/step - loss: 0.0830 -
accuracy: 0.9684 - val_loss: 0.4704 - val_accuracy: 0.9069
Epoch 42/50
2880/2880 [============== ] - 3s 888us/step - loss: 0.1159 -
accuracy: 0.9542 - val_loss: 0.4802 - val_accuracy: 0.8736
Epoch 43/50
2880/2880 [============= ] - 3s 884us/step - loss: 0.1049 -
accuracy: 0.9601 - val_loss: 0.2950 - val_accuracy: 0.8958
Epoch 44/50
2880/2880 [============ ] - 3s 900us/step - loss: 0.0658 -
accuracy: 0.9771 - val_loss: 0.4492 - val_accuracy: 0.9097
Epoch 45/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.1034 -
accuracy: 0.9622 - val_loss: 0.5229 - val_accuracy: 0.8806
Epoch 46/50
2880/2880 [============= ] - 3s 875us/step - loss: 0.1358 -
accuracy: 0.9521 - val_loss: 0.3935 - val_accuracy: 0.9083
Epoch 47/50
2880/2880 [============= ] - 2s 867us/step - loss: 0.0606 -
accuracy: 0.9788 - val_loss: 0.4179 - val_accuracy: 0.9111
2880/2880 [============= ] - 3s 892us/step - loss: 0.1037 -
accuracy: 0.9635 - val_loss: 0.3792 - val_accuracy: 0.9056
2880/2880 [============= ] - 3s 877us/step - loss: 0.0645 -
accuracy: 0.9715 - val_loss: 0.4508 - val_accuracy: 0.9097
Epoch 50/50
2880/2880 [============= ] - 3s 895us/step - loss: 0.0587 -
accuracy: 0.9757 - val loss: 0.5628 - val accuracy: 0.8972
1200/1200 [============ ] - 1s 543us/step
```



\_\_\_\_\_\_

Squential with Kfold CV:

Training for fold 3  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

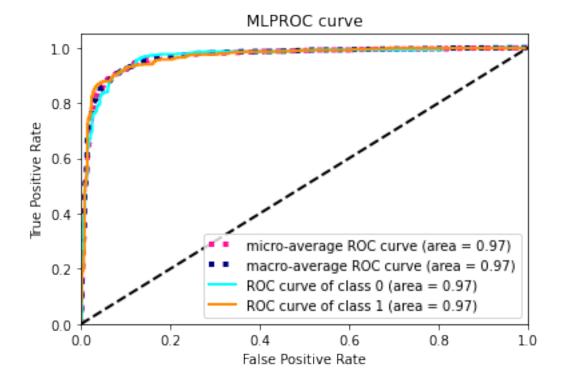
Model: "sequential\_3"

Layer (type)	Output Shape	Param #
dense_11 (Dense)	(None, 2176)	298112
dense_12 (Dense)	(None, 1088)	2368576
dropout_7 (Dropout)	(None, 1088)	0
dense_13 (Dense)	(None, 544)	592416
dropout_8 (Dropout)	(None, 544)	0
dense_14 (Dense)	(None, 272)	148240
dropout_9 (Dropout)	(None, 272)	0

```
dense_15 (Dense)
                        (None, 2)
                                              546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.5688 - val_loss: 0.4442 - val_accuracy: 0.8375
Epoch 2/50
2880/2880 [============== ] - 3s 878us/step - loss: 0.4584 -
accuracy: 0.7986 - val_loss: 0.3101 - val_accuracy: 0.8569
Epoch 3/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.3168 -
accuracy: 0.8701 - val_loss: 0.3495 - val_accuracy: 0.8403
Epoch 4/50
2880/2880 [============= ] - 3s 897us/step - loss: 0.2772 -
accuracy: 0.8851 - val_loss: 0.2641 - val_accuracy: 0.8875
Epoch 5/50
2880/2880 [============= ] - 3s 891us/step - loss: 0.3389 -
accuracy: 0.8597 - val_loss: 0.2826 - val_accuracy: 0.8875
Epoch 6/50
2880/2880 [============== ] - 3s 884us/step - loss: 0.2467 -
accuracy: 0.8990 - val_loss: 0.2603 - val_accuracy: 0.8944
Epoch 7/50
2880/2880 [============== ] - 3s 887us/step - loss: 0.2551 -
accuracy: 0.8924 - val_loss: 0.2459 - val_accuracy: 0.8931
2880/2880 [============= ] - 3s 886us/step - loss: 0.2334 -
accuracy: 0.9045 - val_loss: 0.3296 - val_accuracy: 0.8542
2880/2880 [============= ] - 2s 867us/step - loss: 0.2424 -
accuracy: 0.9010 - val_loss: 0.2624 - val_accuracy: 0.8792
Epoch 10/50
2880/2880 [============= - 3s 893us/step - loss: 0.2224 -
accuracy: 0.9066 - val_loss: 0.2507 - val_accuracy: 0.8931
Epoch 11/50
2880/2880 [============== ] - 2s 829us/step - loss: 0.2404 -
accuracy: 0.9073 - val_loss: 0.2797 - val_accuracy: 0.9014
Epoch 12/50
2880/2880 [============= ] - 2s 678us/step - loss: 0.2244 -
accuracy: 0.9167 - val_loss: 0.2389 - val_accuracy: 0.9097
Epoch 13/50
2880/2880 [============= ] - 2s 805us/step - loss: 0.2019 -
accuracy: 0.9135 - val_loss: 0.2751 - val_accuracy: 0.8903
Epoch 14/50
2880/2880 [============ ] - 3s 1ms/step - loss: 0.1949 -
```

```
accuracy: 0.9229 - val_loss: 0.2159 - val_accuracy: 0.9111
Epoch 15/50
2880/2880 [============== ] - 3s 1ms/step - loss: 0.1878 -
accuracy: 0.9243 - val_loss: 0.2228 - val_accuracy: 0.9194
Epoch 16/50
2880/2880 [============= ] - 3s 1ms/step - loss: 0.1813 -
accuracy: 0.9271 - val_loss: 0.2916 - val_accuracy: 0.9056
Epoch 17/50
2880/2880 [============= ] - 3s 1ms/step - loss: 0.2017 -
accuracy: 0.9240 - val_loss: 0.2434 - val_accuracy: 0.8958
Epoch 18/50
2880/2880 [============== ] - 3s 1ms/step - loss: 0.1751 -
accuracy: 0.9295 - val_loss: 0.2641 - val_accuracy: 0.8958
Epoch 19/50
2880/2880 [============= ] - 3s 874us/step - loss: 0.1734 -
accuracy: 0.9361 - val_loss: 0.2997 - val_accuracy: 0.8944
Epoch 20/50
2880/2880 [============ ] - 3s 891us/step - loss: 0.1643 -
accuracy: 0.9375 - val_loss: 0.3842 - val_accuracy: 0.8375
Epoch 21/50
2880/2880 [============= ] - 3s 873us/step - loss: 0.1769 -
accuracy: 0.9271 - val_loss: 0.2215 - val_accuracy: 0.9194
Epoch 22/50
2880/2880 [=========== ] - 3s 886us/step - loss: 0.1734 -
accuracy: 0.9347 - val_loss: 0.2614 - val_accuracy: 0.8986
Epoch 23/50
2880/2880 [============= ] - 3s 919us/step - loss: 0.1570 -
accuracy: 0.9399 - val_loss: 0.2498 - val_accuracy: 0.8917
2880/2880 [============= ] - 3s 878us/step - loss: 0.1181 -
accuracy: 0.9542 - val_loss: 0.2741 - val_accuracy: 0.9194
Epoch 25/50
2880/2880 [============= ] - 3s 937us/step - loss: 0.1978 -
accuracy: 0.9312 - val_loss: 0.3120 - val_accuracy: 0.8833
Epoch 26/50
2880/2880 [============= ] - 3s 875us/step - loss: 0.1508 -
accuracy: 0.9455 - val loss: 0.2504 - val accuracy: 0.9069
Epoch 27/50
2880/2880 [============== ] - 3s 885us/step - loss: 0.1338 -
accuracy: 0.9462 - val_loss: 0.2984 - val_accuracy: 0.9097
Epoch 28/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.1428 -
accuracy: 0.9431 - val_loss: 0.2729 - val_accuracy: 0.9139
Epoch 29/50
2880/2880 [============ ] - 3s 887us/step - loss: 0.1329 -
accuracy: 0.9493 - val_loss: 0.3580 - val_accuracy: 0.8958
Epoch 30/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.1279 -
```

```
accuracy: 0.9507 - val_loss: 0.3876 - val_accuracy: 0.8208
Epoch 31/50
2880/2880 [============= ] - 2s 867us/step - loss: 0.1612 -
accuracy: 0.9434 - val_loss: 0.2538 - val_accuracy: 0.8931
Epoch 32/50
2880/2880 [============= ] - 3s 898us/step - loss: 0.1047 -
accuracy: 0.9604 - val_loss: 0.3140 - val_accuracy: 0.9125
Epoch 33/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.1492 -
accuracy: 0.9417 - val_loss: 0.2438 - val_accuracy: 0.9222
Epoch 34/50
2880/2880 [============== ] - 3s 885us/step - loss: 0.1327 -
accuracy: 0.9524 - val_loss: 0.2782 - val_accuracy: 0.9097
Epoch 35/50
2880/2880 [============= ] - 3s 882us/step - loss: 0.1332 -
accuracy: 0.9472 - val_loss: 0.3154 - val_accuracy: 0.9111
Epoch 36/50
2880/2880 [============ ] - 3s 869us/step - loss: 0.0884 -
accuracy: 0.9694 - val_loss: 0.3910 - val_accuracy: 0.8847
Epoch 37/50
2880/2880 [============= ] - 3s 905us/step - loss: 0.1514 -
accuracy: 0.9417 - val_loss: 0.3238 - val_accuracy: 0.8847
Epoch 38/50
2880/2880 [============= ] - 3s 885us/step - loss: 0.1040 -
accuracy: 0.9583 - val_loss: 0.3120 - val_accuracy: 0.9014
Epoch 39/50
2880/2880 [============= ] - 3s 898us/step - loss: 0.0904 -
accuracy: 0.9635 - val_loss: 0.3731 - val_accuracy: 0.8931
2880/2880 [============= ] - 3s 907us/step - loss: 0.1342 -
accuracy: 0.9552 - val_loss: 0.3141 - val_accuracy: 0.9097
Epoch 41/50
2880/2880 [============= ] - 3s 875us/step - loss: 0.1032 -
accuracy: 0.9569 - val_loss: 0.3707 - val_accuracy: 0.9153
Epoch 42/50
2880/2880 [============= ] - 3s 874us/step - loss: 0.1036 -
accuracy: 0.9611 - val_loss: 0.3167 - val_accuracy: 0.9153
Epoch 43/50
2880/2880 [============= ] - 3s 873us/step - loss: 0.1005 -
accuracy: 0.9611 - val_loss: 0.3514 - val_accuracy: 0.9097
Epoch 44/50
2880/2880 [============= ] - 3s 873us/step - loss: 0.0804 -
accuracy: 0.9701 - val_loss: 0.3933 - val_accuracy: 0.9042
Epoch 45/50
2880/2880 [============] - 3s 884us/step - loss: 0.1022 -
accuracy: 0.9608 - val_loss: 0.3542 - val_accuracy: 0.8958
Epoch 46/50
2880/2880 [============= ] - 3s 882us/step - loss: 0.0712 -
```

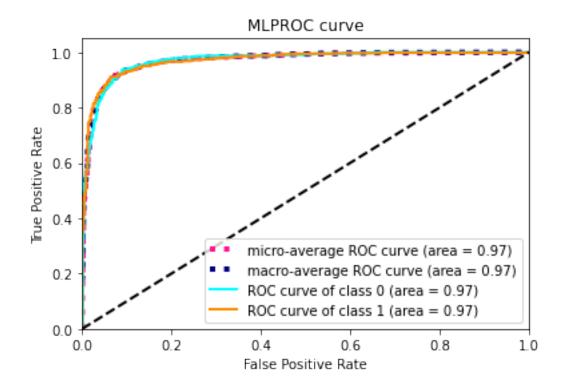


```
______
dense_16 (Dense)
                   (None, 2176)
                                      298112
dense_17 (Dense)
              (None, 1088)
                                     2368576
        ._____
dropout_10 (Dropout) (None, 1088)
_____
dense_18 (Dense) (None, 544)
                               592416
_____
dropout_11 (Dropout) (None, 544)
    -----
dense_19 (Dense) (None, 272)
                                     148240
-----
dropout_12 (Dropout) (None, 272)
_____
dense_20 (Dense) (None, 2)
                                     546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============= ] - 8s 3ms/step - loss: 1.5821 -
accuracy: 0.5767 - val_loss: 0.5356 - val_accuracy: 0.7319
Epoch 2/50
2880/2880 [============== ] - 3s 880us/step - loss: 0.4606 -
accuracy: 0.7913 - val_loss: 0.3222 - val_accuracy: 0.8556
2880/2880 [============= ] - 3s 891us/step - loss: 0.3276 -
accuracy: 0.8604 - val_loss: 0.3422 - val_accuracy: 0.8514
Epoch 4/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.3044 -
accuracy: 0.8729 - val_loss: 0.2594 - val_accuracy: 0.9000
Epoch 5/50
2880/2880 [============= ] - 3s 872us/step - loss: 0.2996 -
accuracy: 0.8736 - val loss: 0.3465 - val accuracy: 0.8403
Epoch 6/50
2880/2880 [============== ] - 3s 888us/step - loss: 0.2718 -
accuracy: 0.8889 - val_loss: 0.2598 - val_accuracy: 0.9042
Epoch 7/50
2880/2880 [============= ] - 3s 897us/step - loss: 0.2476 -
accuracy: 0.9021 - val_loss: 0.2503 - val_accuracy: 0.9056
Epoch 8/50
2880/2880 [============== ] - 3s 902us/step - loss: 0.2435 -
accuracy: 0.9035 - val_loss: 0.2437 - val_accuracy: 0.9069
Epoch 9/50
2880/2880 [============= ] - 3s 896us/step - loss: 0.2216 -
```

```
accuracy: 0.9080 - val_loss: 0.2328 - val_accuracy: 0.9139
Epoch 10/50
2880/2880 [============= ] - 3s 881us/step - loss: 0.2387 -
accuracy: 0.9118 - val_loss: 0.2679 - val_accuracy: 0.8958
Epoch 11/50
2880/2880 [============= ] - 3s 891us/step - loss: 0.2015 -
accuracy: 0.9198 - val_loss: 0.2734 - val_accuracy: 0.9000
Epoch 12/50
2880/2880 [============= ] - 3s 896us/step - loss: 0.2245 -
accuracy: 0.9056 - val_loss: 0.2635 - val_accuracy: 0.8986
Epoch 13/50
2880/2880 [=============== ] - 3s 884us/step - loss: 0.2187 -
accuracy: 0.9090 - val_loss: 0.2510 - val_accuracy: 0.9153
Epoch 14/50
2880/2880 [============= ] - 3s 899us/step - loss: 0.2099 -
accuracy: 0.9222 - val_loss: 0.2916 - val_accuracy: 0.8917
Epoch 15/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.1989 -
accuracy: 0.9219 - val_loss: 0.2533 - val_accuracy: 0.9028
Epoch 16/50
2880/2880 [============= ] - 3s 904us/step - loss: 0.1799 -
accuracy: 0.9278 - val_loss: 0.2699 - val_accuracy: 0.9069
Epoch 17/50
2880/2880 [============= ] - 3s 924us/step - loss: 0.1728 -
accuracy: 0.9292 - val_loss: 0.2685 - val_accuracy: 0.8833
Epoch 18/50
2880/2880 [============= ] - 3s 922us/step - loss: 0.1899 -
accuracy: 0.9163 - val_loss: 0.2814 - val_accuracy: 0.8986
2880/2880 [============= ] - 3s 897us/step - loss: 0.1681 -
accuracy: 0.9333 - val_loss: 0.2704 - val_accuracy: 0.9028
Epoch 20/50
2880/2880 [============= ] - 3s 912us/step - loss: 0.1711 -
accuracy: 0.9368 - val_loss: 0.3308 - val_accuracy: 0.9125
Epoch 21/50
2880/2880 [============= ] - 3s 878us/step - loss: 0.1978 -
accuracy: 0.9267 - val loss: 0.2581 - val accuracy: 0.9111
Epoch 22/50
2880/2880 [============== ] - 3s 893us/step - loss: 0.1586 -
accuracy: 0.9361 - val_loss: 0.2617 - val_accuracy: 0.8819
Epoch 23/50
2880/2880 [============= ] - 3s 894us/step - loss: 0.1517 -
accuracy: 0.9385 - val_loss: 0.3048 - val_accuracy: 0.9042
Epoch 24/50
2880/2880 [============ ] - 3s 882us/step - loss: 0.1537 -
accuracy: 0.9375 - val_loss: 0.2428 - val_accuracy: 0.9056
Epoch 25/50
2880/2880 [============= ] - 3s 896us/step - loss: 0.1267 -
```

```
accuracy: 0.9458 - val_loss: 0.2993 - val_accuracy: 0.8944
Epoch 26/50
2880/2880 [============ ] - 3s 872us/step - loss: 0.1638 -
accuracy: 0.9344 - val_loss: 0.2676 - val_accuracy: 0.9125
Epoch 27/50
2880/2880 [============= ] - 3s 891us/step - loss: 0.1437 -
accuracy: 0.9413 - val_loss: 0.3484 - val_accuracy: 0.8722
Epoch 28/50
2880/2880 [============= ] - 3s 903us/step - loss: 0.1404 -
accuracy: 0.9378 - val_loss: 0.3396 - val_accuracy: 0.8944
Epoch 29/50
2880/2880 [============== ] - 3s 880us/step - loss: 0.1479 -
accuracy: 0.9372 - val_loss: 0.2867 - val_accuracy: 0.8875
Epoch 30/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.1356 -
accuracy: 0.9420 - val_loss: 0.3211 - val_accuracy: 0.9167
Epoch 31/50
2880/2880 [============= ] - 3s 884us/step - loss: 0.1070 -
accuracy: 0.9542 - val_loss: 0.4485 - val_accuracy: 0.9014
Epoch 32/50
2880/2880 [============= ] - 3s 899us/step - loss: 0.1396 -
accuracy: 0.9465 - val_loss: 0.3373 - val_accuracy: 0.8958
Epoch 33/50
2880/2880 [============= ] - 3s 881us/step - loss: 0.1287 -
accuracy: 0.9476 - val_loss: 0.4457 - val_accuracy: 0.8778
Epoch 34/50
2880/2880 [============= ] - 3s 886us/step - loss: 0.1434 -
accuracy: 0.9358 - val_loss: 0.3041 - val_accuracy: 0.9139
2880/2880 [============= ] - 3s 909us/step - loss: 0.1118 -
accuracy: 0.9514 - val_loss: 0.3517 - val_accuracy: 0.8917
Epoch 36/50
2880/2880 [============= ] - 2s 863us/step - loss: 0.1013 -
accuracy: 0.9538 - val_loss: 0.4641 - val_accuracy: 0.8972
Epoch 37/50
2880/2880 [============= ] - 3s 892us/step - loss: 0.1361 -
accuracy: 0.9479 - val loss: 0.3464 - val accuracy: 0.8583
Epoch 38/50
2880/2880 [============== ] - 3s 890us/step - loss: 0.1328 -
accuracy: 0.9476 - val_loss: 0.3508 - val_accuracy: 0.8708
Epoch 39/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.1101 -
accuracy: 0.9542 - val_loss: 0.3373 - val_accuracy: 0.8944
Epoch 40/50
2880/2880 [============ ] - 3s 873us/step - loss: 0.1037 -
accuracy: 0.9517 - val_loss: 0.3233 - val_accuracy: 0.8819
Epoch 41/50
2880/2880 [============= ] - 3s 888us/step - loss: 0.1021 -
```

```
accuracy: 0.9587 - val_loss: 0.5991 - val_accuracy: 0.8653
Epoch 42/50
2880/2880 [============= ] - 3s 879us/step - loss: 0.1273 -
accuracy: 0.9566 - val_loss: 0.3580 - val_accuracy: 0.9028
Epoch 43/50
2880/2880 [============ ] - 3s 901us/step - loss: 0.1138 -
accuracy: 0.9531 - val_loss: 0.3530 - val_accuracy: 0.9069
Epoch 44/50
2880/2880 [============ ] - 3s 897us/step - loss: 0.1067 -
accuracy: 0.9628 - val_loss: 0.3820 - val_accuracy: 0.8986
Epoch 45/50
2880/2880 [============== ] - 3s 893us/step - loss: 0.0709 -
accuracy: 0.9688 - val_loss: 0.5421 - val_accuracy: 0.8639
Epoch 46/50
2880/2880 [============= ] - 3s 893us/step - loss: 0.1240 -
accuracy: 0.9476 - val_loss: 0.3126 - val_accuracy: 0.9028
Epoch 47/50
2880/2880 [============= ] - 3s 903us/step - loss: 0.0712 -
accuracy: 0.9740 - val_loss: 0.4670 - val_accuracy: 0.9111
Epoch 48/50
2880/2880 [============= ] - 3s 873us/step - loss: 0.1220 -
accuracy: 0.9524 - val_loss: 0.3788 - val_accuracy: 0.9083
Epoch 49/50
2880/2880 [============ ] - 3s 908us/step - loss: 0.1202 -
accuracy: 0.9569 - val_loss: 0.2964 - val_accuracy: 0.8958
Epoch 50/50
2880/2880 [============= ] - 3s 875us/step - loss: 0.0704 -
accuracy: 0.9719 - val_loss: 0.3623 - val_accuracy: 0.9125
```



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Squential with Kfold CV:

Training for fold 5  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

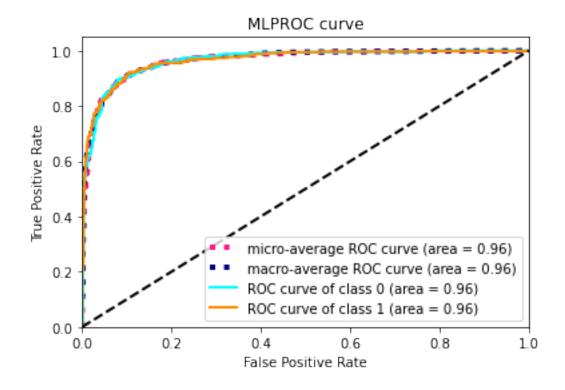
Model: "sequential\_5"

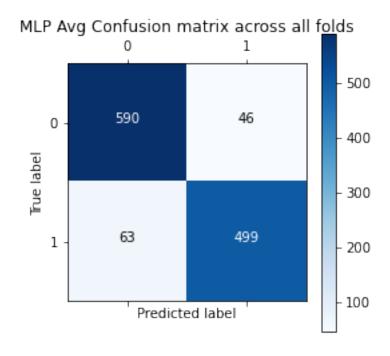
Layer (type)	Output Shape	Param #
dense_21 (Dense)	(None, 2176)	298112
dense_22 (Dense)	(None, 1088)	2368576
dropout_13 (Dropout)	(None, 1088)	0
dense_23 (Dense)	(None, 544)	592416
dropout_14 (Dropout)	(None, 544)	0
dense_24 (Dense)	(None, 272)	148240
dropout_15 (Dropout)	(None, 272)	0

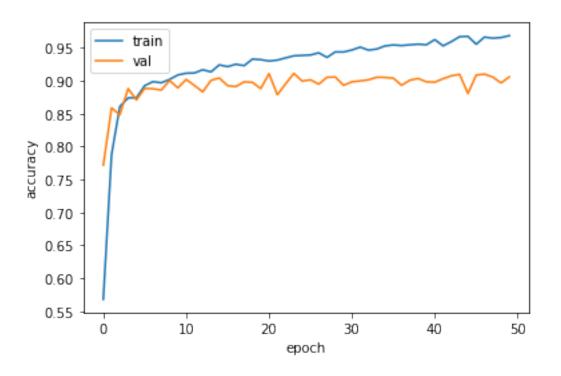
```
dense_25 (Dense)
                        (None, 2)
                                               546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.5670 - val_loss: 0.5837 - val_accuracy: 0.7500
Epoch 2/50
2880/2880 [============== ] - 2s 865us/step - loss: 0.5222 -
accuracy: 0.7646 - val_loss: 0.2951 - val_accuracy: 0.8819
Epoch 3/50
2880/2880 [============= ] - 3s 897us/step - loss: 0.3301 -
accuracy: 0.8660 - val_loss: 0.3842 - val_accuracy: 0.8250
Epoch 4/50
2880/2880 [============= ] - 3s 899us/step - loss: 0.3271 -
accuracy: 0.8615 - val_loss: 0.2828 - val_accuracy: 0.8694
Epoch 5/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.2929 -
accuracy: 0.8861 - val_loss: 0.2721 - val_accuracy: 0.8861
Epoch 6/50
2880/2880 [============== ] - 3s 901us/step - loss: 0.2757 -
accuracy: 0.8885 - val_loss: 0.3153 - val_accuracy: 0.8792
Epoch 7/50
2880/2880 [============== ] - 3s 892us/step - loss: 0.2644 -
accuracy: 0.8924 - val_loss: 0.2587 - val_accuracy: 0.8903
2880/2880 [============= ] - 3s 896us/step - loss: 0.2451 -
accuracy: 0.9059 - val_loss: 0.2595 - val_accuracy: 0.8972
2880/2880 [============= ] - 3s 912us/step - loss: 0.2587 -
accuracy: 0.9000 - val_loss: 0.2544 - val_accuracy: 0.9042
Epoch 10/50
2880/2880 [============= ] - 3s 902us/step - loss: 0.2349 -
accuracy: 0.9066 - val loss: 0.2349 - val accuracy: 0.8972
Epoch 11/50
2880/2880 [============== ] - 3s 885us/step - loss: 0.2151 -
accuracy: 0.9153 - val_loss: 0.2845 - val_accuracy: 0.9042
Epoch 12/50
2880/2880 [============= ] - 3s 883us/step - loss: 0.2436 -
accuracy: 0.9059 - val_loss: 0.2430 - val_accuracy: 0.9069
Epoch 13/50
2880/2880 [============ ] - 3s 879us/step - loss: 0.1900 -
accuracy: 0.9222 - val_loss: 0.3116 - val_accuracy: 0.8958
Epoch 14/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.2246 -
```

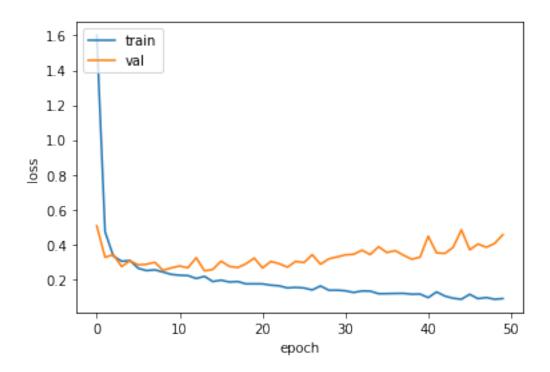
```
accuracy: 0.9066 - val_loss: 0.2443 - val_accuracy: 0.8944
Epoch 15/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.1721 -
accuracy: 0.9295 - val_loss: 0.2610 - val_accuracy: 0.8931
Epoch 16/50
2880/2880 [============= ] - 3s 884us/step - loss: 0.2222 -
accuracy: 0.9156 - val_loss: 0.2489 - val_accuracy: 0.9194
Epoch 17/50
2880/2880 [============ ] - 3s 873us/step - loss: 0.1715 -
accuracy: 0.9299 - val_loss: 0.3150 - val_accuracy: 0.8722
Epoch 18/50
2880/2880 [============== ] - 2s 848us/step - loss: 0.1864 -
accuracy: 0.9191 - val_loss: 0.2662 - val_accuracy: 0.8972
Epoch 19/50
2880/2880 [============= ] - 2s 754us/step - loss: 0.1734 -
accuracy: 0.9347 - val_loss: 0.2711 - val_accuracy: 0.9083
Epoch 20/50
2880/2880 [============= ] - 2s 823us/step - loss: 0.1822 -
accuracy: 0.9319 - val_loss: 0.2474 - val_accuracy: 0.9181
Epoch 21/50
2880/2880 [============= ] - 3s 879us/step - loss: 0.1787 -
accuracy: 0.9240 - val_loss: 0.2594 - val_accuracy: 0.9014
Epoch 22/50
2880/2880 [============= ] - 3s 875us/step - loss: 0.1500 -
accuracy: 0.9385 - val_loss: 0.3720 - val_accuracy: 0.8806
Epoch 23/50
2880/2880 [============== ] - 3s 869us/step - loss: 0.1916 -
accuracy: 0.9260 - val_loss: 0.2837 - val_accuracy: 0.9083
2880/2880 [============= ] - 3s 889us/step - loss: 0.1499 -
accuracy: 0.9406 - val_loss: 0.2953 - val_accuracy: 0.8986
Epoch 25/50
2880/2880 [============= ] - 3s 880us/step - loss: 0.1451 -
accuracy: 0.9420 - val_loss: 0.2410 - val_accuracy: 0.9069
Epoch 26/50
2880/2880 [============= - 3s 901us/step - loss: 0.1481 -
accuracy: 0.9413 - val loss: 0.3099 - val accuracy: 0.8903
Epoch 27/50
2880/2880 [============= ] - 3s 874us/step - loss: 0.1420 -
accuracy: 0.9403 - val_loss: 0.3623 - val_accuracy: 0.8972
Epoch 28/50
2880/2880 [============= ] - 3s 893us/step - loss: 0.1680 -
accuracy: 0.9413 - val_loss: 0.2777 - val_accuracy: 0.9014
Epoch 29/50
2880/2880 [============ ] - 3s 869us/step - loss: 0.1455 -
accuracy: 0.9392 - val_loss: 0.3060 - val_accuracy: 0.9153
Epoch 30/50
2880/2880 [============= ] - 3s 907us/step - loss: 0.1297 -
```

```
accuracy: 0.9455 - val_loss: 0.3341 - val_accuracy: 0.9139
Epoch 31/50
2880/2880 [============= ] - 2s 865us/step - loss: 0.1613 -
accuracy: 0.9375 - val_loss: 0.2934 - val_accuracy: 0.9139
Epoch 32/50
2880/2880 [============= ] - 3s 896us/step - loss: 0.1116 -
accuracy: 0.9552 - val_loss: 0.3662 - val_accuracy: 0.8903
Epoch 33/50
2880/2880 [============= ] - 3s 871us/step - loss: 0.1508 -
accuracy: 0.9382 - val_loss: 0.4586 - val_accuracy: 0.9028
Epoch 34/50
2880/2880 [============== ] - 3s 882us/step - loss: 0.1199 -
accuracy: 0.9601 - val_loss: 0.3912 - val_accuracy: 0.9111
Epoch 35/50
2880/2880 [============= ] - 3s 895us/step - loss: 0.1353 -
accuracy: 0.9476 - val_loss: 0.5052 - val_accuracy: 0.8875
Epoch 36/50
2880/2880 [============= ] - 3s 885us/step - loss: 0.1207 -
accuracy: 0.9552 - val_loss: 0.3438 - val_accuracy: 0.8986
Epoch 37/50
2880/2880 [============= ] - 3s 896us/step - loss: 0.1120 -
accuracy: 0.9517 - val_loss: 0.3998 - val_accuracy: 0.9139
Epoch 38/50
2880/2880 [=========== ] - 3s 870us/step - loss: 0.1134 -
accuracy: 0.9611 - val_loss: 0.3662 - val_accuracy: 0.9139
Epoch 39/50
2880/2880 [============= ] - 3s 945us/step - loss: 0.1432 -
accuracy: 0.9458 - val_loss: 0.2814 - val_accuracy: 0.9056
2880/2880 [============= ] - 3s 880us/step - loss: 0.0851 -
accuracy: 0.9674 - val_loss: 0.4307 - val_accuracy: 0.9000
Epoch 41/50
2880/2880 [============= ] - 3s 884us/step - loss: 0.1015 -
accuracy: 0.9618 - val_loss: 0.4137 - val_accuracy: 0.8958
Epoch 42/50
2880/2880 [============= ] - 3s 907us/step - loss: 0.1426 -
accuracy: 0.9545 - val loss: 0.2893 - val accuracy: 0.9028
Epoch 43/50
2880/2880 [============= ] - 3s 910us/step - loss: 0.1012 -
accuracy: 0.9594 - val_loss: 0.4294 - val_accuracy: 0.9069
Epoch 44/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.0799 -
accuracy: 0.9663 - val_loss: 0.4129 - val_accuracy: 0.9153
Epoch 45/50
2880/2880 [============ ] - 3s 880us/step - loss: 0.0705 -
accuracy: 0.9760 - val_loss: 0.5586 - val_accuracy: 0.8833
Epoch 46/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.1258 -
```









Average scores for pesudo test set across all folds:

> Accuracy: 0.9080000042915344 (+- 0.008259694488400967)

> Loss: 0.43945329840729636

> Avg runtime per test instance: 0.0015766371488571168

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## 

## 

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Squential with Kfold CV:

Training for fold 1  $\dots$ 

Adding layer 1:

Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_6"

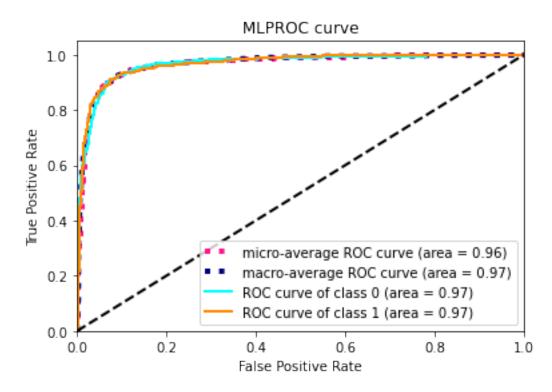
Layer (type)	Output Shape	Param #
dense_26 (Dense)	(None, 4352)	596224
dense_27 (Dense)	(None, 2176)	9472128
dropout_16 (Dropout)	(None, 2176)	0

```
dense_28 (Dense) (None, 1088)
                                  2368576
_____
dropout_17 (Dropout)
                (None, 1088)
_____
dense 29 (Dense) (None, 544)
                                 592416
_____
dropout_18 (Dropout) (None, 544)
_____
            (None, 272)
dense 30 (Dense)
                                 148240
dropout_19 (Dropout) (None, 272)
dense_31 (Dense) (None, 2) 546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============= ] - 15s 5ms/step - loss: 3.0720 -
accuracy: 0.5059 - val_loss: 0.6910 - val_accuracy: 0.5431
Epoch 2/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.7191 -
accuracy: 0.5146 - val_loss: 0.6581 - val_accuracy: 0.7667
Epoch 3/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.7065 -
accuracy: 0.5719 - val_loss: 0.5130 - val_accuracy: 0.7806
accuracy: 0.6983 - val_loss: 0.4873 - val_accuracy: 0.8361
2880/2880 [============ ] - 8s 3ms/step - loss: 0.4322 -
accuracy: 0.8434 - val_loss: 0.3770 - val_accuracy: 0.8625
Epoch 6/50
accuracy: 0.8694 - val loss: 0.4953 - val accuracy: 0.8347
Epoch 7/50
accuracy: 0.8788 - val_loss: 0.3334 - val_accuracy: 0.8958
Epoch 8/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.3341 -
accuracy: 0.8806 - val_loss: 0.2946 - val_accuracy: 0.9028
Epoch 9/50
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.3258 -
accuracy: 0.8712 - val_loss: 0.2786 - val_accuracy: 0.8931
Epoch 10/50
```

```
accuracy: 0.8934 - val_loss: 0.3713 - val_accuracy: 0.8264
Epoch 11/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2707 -
accuracy: 0.8938 - val_loss: 0.2490 - val_accuracy: 0.9097
Epoch 12/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2627 -
accuracy: 0.8951 - val_loss: 0.2985 - val_accuracy: 0.8944
Epoch 13/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2321 -
accuracy: 0.9097 - val_loss: 0.2747 - val_accuracy: 0.9153
Epoch 14/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2680 -
accuracy: 0.8885 - val_loss: 0.2635 - val_accuracy: 0.8972
Epoch 15/50
accuracy: 0.9153 - val_loss: 0.2405 - val_accuracy: 0.9083
Epoch 16/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2586 -
accuracy: 0.8951 - val_loss: 0.2673 - val_accuracy: 0.8944
Epoch 17/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2075 -
accuracy: 0.9167 - val_loss: 0.2351 - val_accuracy: 0.9139
Epoch 18/50
accuracy: 0.9052 - val_loss: 0.2534 - val_accuracy: 0.9042
Epoch 19/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1939 -
accuracy: 0.9212 - val_loss: 0.2568 - val_accuracy: 0.9069
accuracy: 0.9090 - val_loss: 0.2550 - val_accuracy: 0.9153
Epoch 21/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1938 -
accuracy: 0.9260 - val_loss: 0.2389 - val_accuracy: 0.9042
Epoch 22/50
accuracy: 0.9087 - val loss: 0.2753 - val accuracy: 0.9069
Epoch 23/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2032 -
accuracy: 0.9219 - val_loss: 0.2401 - val_accuracy: 0.9125
Epoch 24/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1703 -
accuracy: 0.9292 - val_loss: 0.2701 - val_accuracy: 0.8819
Epoch 25/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2415 -
accuracy: 0.9066 - val_loss: 0.2336 - val_accuracy: 0.9139
Epoch 26/50
2880/2880 [============ ] - 7s 3ms/step - loss: 0.1889 -
```

```
accuracy: 0.9229 - val_loss: 0.2186 - val_accuracy: 0.9181
Epoch 27/50
2880/2880 [============ ] - 9s 3ms/step - loss: 0.1679 -
accuracy: 0.9253 - val_loss: 0.2568 - val_accuracy: 0.9111
Epoch 28/50
accuracy: 0.9108 - val_loss: 0.2126 - val_accuracy: 0.9222
Epoch 29/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1565 -
accuracy: 0.9413 - val_loss: 0.2748 - val_accuracy: 0.9042
Epoch 30/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1879 -
accuracy: 0.9208 - val_loss: 0.2939 - val_accuracy: 0.9042
Epoch 31/50
accuracy: 0.9302 - val_loss: 0.2714 - val_accuracy: 0.8972
Epoch 32/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1670 -
accuracy: 0.9340 - val_loss: 0.2679 - val_accuracy: 0.9153
Epoch 33/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1866 -
accuracy: 0.9292 - val_loss: 0.2551 - val_accuracy: 0.9042
Epoch 34/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1625 -
accuracy: 0.9330 - val_loss: 0.2901 - val_accuracy: 0.9167
Epoch 35/50
2880/2880 [=============== ] - 8s 3ms/step - loss: 0.1659 -
accuracy: 0.9392 - val_loss: 0.2756 - val_accuracy: 0.9014
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1717 -
accuracy: 0.9333 - val_loss: 0.2793 - val_accuracy: 0.9236
Epoch 37/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1626 -
accuracy: 0.9306 - val_loss: 0.2230 - val_accuracy: 0.9194
Epoch 38/50
accuracy: 0.9323 - val loss: 0.2481 - val accuracy: 0.9167
Epoch 39/50
accuracy: 0.9410 - val_loss: 0.2714 - val_accuracy: 0.8986
Epoch 40/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1859 -
accuracy: 0.9240 - val_loss: 0.3051 - val_accuracy: 0.9208
Epoch 41/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1548 -
accuracy: 0.9375 - val_loss: 0.3298 - val_accuracy: 0.9194
Epoch 42/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1685 -
```

```
accuracy: 0.9316 - val_loss: 0.2908 - val_accuracy: 0.8667
Epoch 43/50
accuracy: 0.9375 - val_loss: 0.2691 - val_accuracy: 0.9292
Epoch 44/50
accuracy: 0.9385 - val_loss: 0.3034 - val_accuracy: 0.8875
Epoch 45/50
accuracy: 0.9285 - val_loss: 0.2985 - val_accuracy: 0.9056
Epoch 46/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1542 -
accuracy: 0.9420 - val_loss: 0.2363 - val_accuracy: 0.9167
Epoch 47/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1315 -
accuracy: 0.9434 - val_loss: 0.2403 - val_accuracy: 0.9208
Epoch 48/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1418 -
accuracy: 0.9451 - val_loss: 0.2638 - val_accuracy: 0.9056
Epoch 49/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1523 -
accuracy: 0.9392 - val_loss: 0.2371 - val_accuracy: 0.9264
Epoch 50/50
2880/2880 [=====
                 accuracy: 0.9476 - val_loss: 0.2290 - val_accuracy: 0.9347
1200/1200 [========== ] - 2s 1ms/step
```

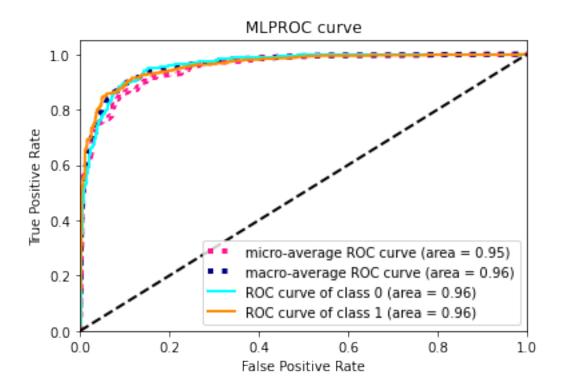


```
Squential with Kfold CV:
Training for fold 2 ...
Adding layer 1:
Adding layer 2:
Adding layer 3:
Adding layer 4:
Model: "sequential_7"
                            Param #
Layer (type) Output Shape
______
dense_32 (Dense)
                 (None, 4352)
                                  596224
             (None, 2176)
dense_33 (Dense)
                                 9472128
dropout_20 (Dropout) (None, 2176)
_____
dense_34 (Dense) (None, 1088)
                                  2368576
______
dropout_21 (Dropout) (None, 1088)
dense_35 (Dense) (None, 544)
                                  592416
_____
dropout_22 (Dropout) (None, 544)
_____
dense_36 (Dense)
                 (None, 272)
                                  148240
dropout_23 (Dropout) (None, 272)
dense_37 (Dense) (None, 2) 546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============== ] - 14s 5ms/step - loss: 2.8035 -
accuracy: 0.5139 - val_loss: 0.6995 - val_accuracy: 0.4819
Epoch 2/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.7215 -
accuracy: 0.5080 - val_loss: 0.6919 - val_accuracy: 0.4819
Epoch 3/50
accuracy: 0.5267 - val_loss: 0.5771 - val_accuracy: 0.8000
```

```
Epoch 4/50
accuracy: 0.6524 - val_loss: 0.4606 - val_accuracy: 0.7958
2880/2880 [============= ] - 8s 3ms/step - loss: 0.4368 -
accuracy: 0.8347 - val_loss: 0.3443 - val_accuracy: 0.8639
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.3805 -
accuracy: 0.8566 - val_loss: 0.3386 - val_accuracy: 0.8708
Epoch 7/50
accuracy: 0.8552 - val_loss: 0.3256 - val_accuracy: 0.8736
Epoch 8/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2600 -
accuracy: 0.8979 - val_loss: 0.3141 - val_accuracy: 0.8792
Epoch 9/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.3207 -
accuracy: 0.8792 - val_loss: 0.3049 - val_accuracy: 0.8833
Epoch 10/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2406 -
accuracy: 0.9042 - val_loss: 0.2975 - val_accuracy: 0.8944
Epoch 11/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.3191 -
accuracy: 0.8795 - val_loss: 0.2883 - val_accuracy: 0.8917
Epoch 12/50
2880/2880 [============== ] - 6s 2ms/step - loss: 0.2290 -
accuracy: 0.9097 - val_loss: 0.2885 - val_accuracy: 0.8833
Epoch 13/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2518 -
accuracy: 0.9062 - val_loss: 0.3758 - val_accuracy: 0.8569
Epoch 14/50
accuracy: 0.9187 - val_loss: 0.3444 - val_accuracy: 0.8569
Epoch 15/50
accuracy: 0.9028 - val_loss: 0.2828 - val_accuracy: 0.9000
Epoch 16/50
accuracy: 0.9201 - val_loss: 0.3488 - val_accuracy: 0.9014
Epoch 17/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.2364 -
accuracy: 0.9045 - val_loss: 0.3299 - val_accuracy: 0.8986
Epoch 18/50
accuracy: 0.9264 - val_loss: 0.2716 - val_accuracy: 0.9056
Epoch 19/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2519 -
accuracy: 0.9021 - val_loss: 0.3182 - val_accuracy: 0.8750
```

```
Epoch 20/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1891 -
accuracy: 0.9281 - val_loss: 0.2825 - val_accuracy: 0.8944
Epoch 21/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2088 -
accuracy: 0.9201 - val_loss: 0.3359 - val_accuracy: 0.8917
2880/2880 [=========== ] - 5s 2ms/step - loss: 0.1670 -
accuracy: 0.9312 - val_loss: 0.3306 - val_accuracy: 0.8944
Epoch 23/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2195 -
accuracy: 0.9059 - val_loss: 0.3332 - val_accuracy: 0.8542
Epoch 24/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1858 -
accuracy: 0.9250 - val_loss: 0.3470 - val_accuracy: 0.8889
Epoch 25/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1957 -
accuracy: 0.9240 - val_loss: 0.3159 - val_accuracy: 0.8667
Epoch 26/50
accuracy: 0.9274 - val_loss: 0.3747 - val_accuracy: 0.8611
Epoch 27/50
accuracy: 0.9167 - val_loss: 0.3360 - val_accuracy: 0.9042
Epoch 28/50
accuracy: 0.9392 - val_loss: 0.3798 - val_accuracy: 0.8917
Epoch 29/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1836 -
accuracy: 0.9250 - val_loss: 0.3560 - val_accuracy: 0.8931
Epoch 30/50
accuracy: 0.9399 - val_loss: 0.4579 - val_accuracy: 0.8194
Epoch 31/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1817 -
accuracy: 0.9229 - val_loss: 0.3380 - val_accuracy: 0.9083
Epoch 32/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1714 -
accuracy: 0.9271 - val_loss: 0.4290 - val_accuracy: 0.8778
Epoch 33/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1401 -
accuracy: 0.9427 - val_loss: 0.3813 - val_accuracy: 0.9069
Epoch 34/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1711 -
accuracy: 0.9372 - val_loss: 0.3591 - val_accuracy: 0.8778
Epoch 35/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1453 -
accuracy: 0.9434 - val_loss: 0.2892 - val_accuracy: 0.9014
```

```
Epoch 36/50
accuracy: 0.9417 - val_loss: 0.3803 - val_accuracy: 0.8944
Epoch 37/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.1433 -
accuracy: 0.9451 - val_loss: 0.3592 - val_accuracy: 0.8931
2880/2880 [=========== ] - 6s 2ms/step - loss: 0.1588 -
accuracy: 0.9323 - val_loss: 0.5134 - val_accuracy: 0.8750
Epoch 39/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1375 -
accuracy: 0.9434 - val_loss: 0.3619 - val_accuracy: 0.8944
Epoch 40/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.1637 -
accuracy: 0.9323 - val_loss: 0.3242 - val_accuracy: 0.8875
Epoch 41/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1772 -
accuracy: 0.9278 - val_loss: 0.3959 - val_accuracy: 0.8958
Epoch 42/50
accuracy: 0.9573 - val_loss: 0.4685 - val_accuracy: 0.8806
Epoch 43/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1368 -
accuracy: 0.9420 - val_loss: 0.3459 - val_accuracy: 0.9042
Epoch 44/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1749 -
accuracy: 0.9306 - val_loss: 0.3688 - val_accuracy: 0.8653
Epoch 45/50
accuracy: 0.9462 - val_loss: 0.3683 - val_accuracy: 0.8889
Epoch 46/50
accuracy: 0.9493 - val_loss: 0.3405 - val_accuracy: 0.8792
Epoch 47/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1303 -
accuracy: 0.9451 - val_loss: 0.5053 - val_accuracy: 0.8764
Epoch 48/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1419 -
accuracy: 0.9448 - val_loss: 0.4397 - val_accuracy: 0.8750
Epoch 49/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1140 -
accuracy: 0.9566 - val_loss: 0.4230 - val_accuracy: 0.8903
Epoch 50/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1405 -
accuracy: 0.9441 - val_loss: 0.3749 - val_accuracy: 0.8569
1200/1200 [=========== ] - 1s 524us/step
```



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Squential with Kfold CV:

Training for fold 3  $\dots$ 

Adding layer 1:

Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_8"

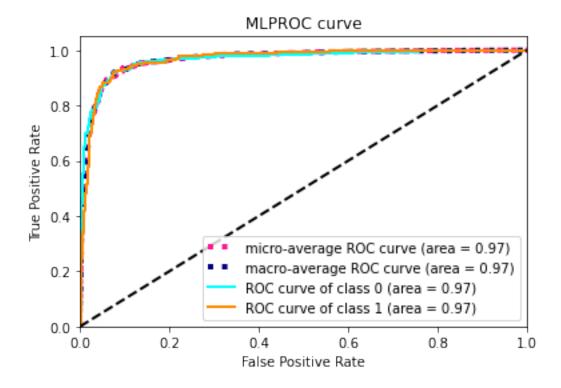
Layer (type)	Output Shape	Param #
dense_38 (Dense)	(None, 4352)	596224
dense_39 (Dense)	(None, 2176)	9472128
dropout_24 (Dropout)	(None, 2176)	0
dense_40 (Dense)	(None, 1088)	2368576
dropout_25 (Dropout)	(None, 1088)	0
dense_41 (Dense)	(None, 544)	592416
dropout_26 (Dropout)	(None, 544)	0

```
(None, 272)
dense_42 (Dense)
                                 148240
dropout_27 (Dropout) (None, 272)
                                 0
_____
dense_43 (Dense) (None, 2)
                                 546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.4948 - val_loss: 0.6870 - val_accuracy: 0.4931
Epoch 2/50
accuracy: 0.5170 - val_loss: 0.6461 - val_accuracy: 0.6972
Epoch 3/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.6925 -
accuracy: 0.6174 - val_loss: 0.4841 - val_accuracy: 0.7958
Epoch 4/50
accuracy: 0.8035 - val_loss: 0.4358 - val_accuracy: 0.8458
Epoch 5/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.3487 -
accuracy: 0.8708 - val_loss: 0.3361 - val_accuracy: 0.8750
Epoch 6/50
accuracy: 0.8465 - val_loss: 0.3491 - val_accuracy: 0.8708
Epoch 7/50
accuracy: 0.8872 - val_loss: 0.2678 - val_accuracy: 0.8778
Epoch 8/50
accuracy: 0.8826 - val_loss: 0.3797 - val_accuracy: 0.8583
Epoch 9/50
accuracy: 0.8962 - val_loss: 0.2401 - val_accuracy: 0.8944
Epoch 10/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2732 -
accuracy: 0.8944 - val_loss: 0.2348 - val_accuracy: 0.9097
accuracy: 0.9003 - val_loss: 0.2526 - val_accuracy: 0.9083
Epoch 12/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2817 -
accuracy: 0.8941 - val_loss: 0.2239 - val_accuracy: 0.9125
```

```
Epoch 13/50
accuracy: 0.8986 - val_loss: 0.4471 - val_accuracy: 0.7917
Epoch 14/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2497 -
accuracy: 0.9035 - val_loss: 0.2345 - val_accuracy: 0.9125
Epoch 15/50
2880/2880 [=========== ] - 5s 2ms/step - loss: 0.2399 -
accuracy: 0.9010 - val_loss: 0.2720 - val_accuracy: 0.9014
Epoch 16/50
accuracy: 0.9160 - val_loss: 0.2211 - val_accuracy: 0.9153
Epoch 17/50
accuracy: 0.9097 - val_loss: 0.2363 - val_accuracy: 0.9236
Epoch 18/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2011 -
accuracy: 0.9198 - val_loss: 0.5242 - val_accuracy: 0.8278
Epoch 19/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2252 -
accuracy: 0.9167 - val_loss: 0.2188 - val_accuracy: 0.9194
Epoch 20/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1977 -
accuracy: 0.9267 - val_loss: 0.2406 - val_accuracy: 0.9069
Epoch 21/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2464 -
accuracy: 0.9035 - val_loss: 0.2226 - val_accuracy: 0.9097
Epoch 22/50
accuracy: 0.9316 - val_loss: 0.2522 - val_accuracy: 0.8972
Epoch 23/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1717 -
accuracy: 0.9333 - val_loss: 0.2637 - val_accuracy: 0.9167
Epoch 24/50
accuracy: 0.9240 - val_loss: 0.3051 - val_accuracy: 0.8958
Epoch 25/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.2220 -
accuracy: 0.9142 - val_loss: 0.2362 - val_accuracy: 0.9139
Epoch 26/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1903 -
accuracy: 0.9302 - val_loss: 0.2350 - val_accuracy: 0.9306
Epoch 27/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1913 -
accuracy: 0.9264 - val_loss: 0.2235 - val_accuracy: 0.9181
Epoch 28/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2000 -
accuracy: 0.9156 - val_loss: 0.2161 - val_accuracy: 0.9139
```

```
Epoch 29/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1742 -
accuracy: 0.9309 - val_loss: 0.2483 - val_accuracy: 0.8986
Epoch 30/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1777 -
accuracy: 0.9281 - val_loss: 0.2568 - val_accuracy: 0.9111
Epoch 31/50
2880/2880 [=========== ] - 5s 2ms/step - loss: 0.1582 -
accuracy: 0.9347 - val_loss: 0.2389 - val_accuracy: 0.9153
Epoch 32/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.2105 -
accuracy: 0.9170 - val_loss: 0.2350 - val_accuracy: 0.9069
Epoch 33/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.1622 -
accuracy: 0.9295 - val_loss: 0.2759 - val_accuracy: 0.9194
Epoch 34/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1570 -
accuracy: 0.9354 - val_loss: 0.3078 - val_accuracy: 0.8958
Epoch 35/50
2880/2880 [=============== ] - 5s 2ms/step - loss: 0.1916 -
accuracy: 0.9285 - val_loss: 0.2619 - val_accuracy: 0.9153
Epoch 36/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1568 -
accuracy: 0.9382 - val_loss: 0.2523 - val_accuracy: 0.9042
Epoch 37/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1826 -
accuracy: 0.9240 - val_loss: 0.2555 - val_accuracy: 0.9139
Epoch 38/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1424 -
accuracy: 0.9431 - val_loss: 0.3166 - val_accuracy: 0.9278
Epoch 39/50
accuracy: 0.9365 - val_loss: 0.2354 - val_accuracy: 0.9125
Epoch 40/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1667 -
accuracy: 0.9378 - val_loss: 0.2361 - val_accuracy: 0.9264
Epoch 41/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1652 -
accuracy: 0.9378 - val_loss: 0.2895 - val_accuracy: 0.8917
Epoch 42/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1457 -
accuracy: 0.9465 - val_loss: 0.2838 - val_accuracy: 0.9153
Epoch 43/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1525 -
accuracy: 0.9403 - val_loss: 0.2299 - val_accuracy: 0.9236
Epoch 44/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1648 -
accuracy: 0.9417 - val_loss: 0.2242 - val_accuracy: 0.9333
```

```
Epoch 45/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1446 -
accuracy: 0.9427 - val_loss: 0.2398 - val_accuracy: 0.9222
Epoch 46/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1172 -
accuracy: 0.9556 - val_loss: 0.2465 - val_accuracy: 0.9319
Epoch 47/50
accuracy: 0.9378 - val_loss: 0.2821 - val_accuracy: 0.9222
Epoch 48/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1747 -
accuracy: 0.9354 - val_loss: 0.2583 - val_accuracy: 0.9194
Epoch 49/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1388 -
accuracy: 0.9486 - val_loss: 0.2353 - val_accuracy: 0.9167
Epoch 50/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1405 -
accuracy: 0.9444 - val_loss: 0.2524 - val_accuracy: 0.9194
1200/1200 [=========== ] - 1s 498us/step
```



\_\_\_\_\_\_

Squential with Kfold CV: Training for fold 4 ... Adding layer 1: Adding layer 2: Adding layer 3: Adding layer 4: Model: "sequential\_9"

-----Layer (type) Output Shape \_\_\_\_\_\_ (None, 4352) dense 44 (Dense) 596224 ----dense\_45 (Dense) (None, 2176) 9472128 dropout\_28 (Dropout) (None, 2176) 0 ----dense\_46 (Dense) (None, 1088) 2368576 ----dropout\_29 (Dropout) (None, 1088) ----dense\_47 (Dense) (None, 544) 592416 ----dropout\_30 (Dropout) (None, 544) \_\_\_\_\_ dense\_48 (Dense) (None, 272) 148240 \_\_\_\_\_\_ dropout\_31 (Dropout) (None, 272) \_\_\_\_\_ 546 (None, 2) dense\_49 (Dense) \_\_\_\_\_\_

Total params: 13,178,130 Trainable params: 13,178,130 Non-trainable params: 0

\_\_\_\_\_\_

Train on 2880 samples, validate on 720 samples Epoch 1/50

2880/2880 [============= ] - 8s 3ms/step - loss: 2.8803 -

accuracy: 0.5299 - val\_loss: 0.7284 - val\_accuracy: 0.5125

Epoch 2/50

2880/2880 [=========== ] - 6s 2ms/step - loss: 0.7414 -

accuracy: 0.4938 - val\_loss: 0.6927 - val\_accuracy: 0.4875

Epoch 3/50

2880/2880 [=========== ] - 7s 2ms/step - loss: 0.7232 -

accuracy: 0.5108 - val\_loss: 0.8221 - val\_accuracy: 0.4875

Epoch 4/50

2880/2880 [============= ] - 5s 2ms/step - loss: 0.7454 -

accuracy: 0.6094 - val\_loss: 0.3991 - val\_accuracy: 0.8556

Epoch 5/50

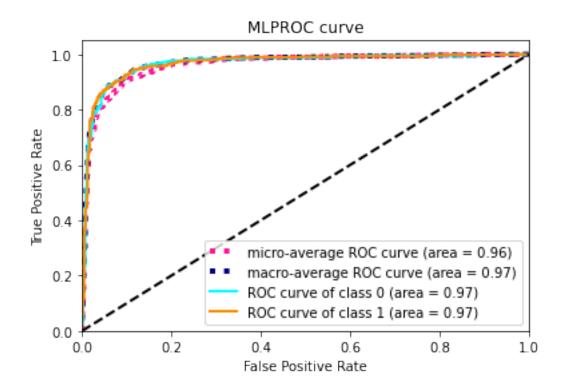
accuracy: 0.8330 - val\_loss: 0.3029 - val\_accuracy: 0.8764

Epoch 6/50

```
accuracy: 0.8760 - val_loss: 0.3790 - val_accuracy: 0.8347
Epoch 7/50
accuracy: 0.8674 - val_loss: 0.2750 - val_accuracy: 0.8792
Epoch 8/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.3613 -
accuracy: 0.8552 - val_loss: 0.2665 - val_accuracy: 0.8889
Epoch 9/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2670 -
accuracy: 0.9031 - val_loss: 0.2914 - val_accuracy: 0.8875
Epoch 10/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.3275 -
accuracy: 0.8701 - val_loss: 0.2721 - val_accuracy: 0.9000
Epoch 11/50
accuracy: 0.8944 - val_loss: 0.2361 - val_accuracy: 0.9097
Epoch 12/50
accuracy: 0.9146 - val_loss: 0.3855 - val_accuracy: 0.8431
Epoch 13/50
accuracy: 0.8736 - val_loss: 0.2446 - val_accuracy: 0.9125
Epoch 14/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2174 -
accuracy: 0.9181 - val_loss: 0.2697 - val_accuracy: 0.8958
Epoch 15/50
accuracy: 0.8920 - val_loss: 0.2492 - val_accuracy: 0.8944
Epoch 16/50
2880/2880 [============= ] - 7s 2ms/step - loss: 0.2018 -
accuracy: 0.9240 - val_loss: 0.2294 - val_accuracy: 0.9083
Epoch 17/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2334 -
accuracy: 0.9087 - val loss: 0.3403 - val accuracy: 0.8597
Epoch 18/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2194 -
accuracy: 0.9135 - val_loss: 0.2487 - val_accuracy: 0.9236
Epoch 19/50
2880/2880 [=========== ] - 5s 2ms/step - loss: 0.2024 -
accuracy: 0.9177 - val_loss: 0.3175 - val_accuracy: 0.8944
Epoch 20/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2496 -
accuracy: 0.9017 - val_loss: 0.2198 - val_accuracy: 0.9153
Epoch 21/50
accuracy: 0.9253 - val_loss: 0.2600 - val_accuracy: 0.8931
Epoch 22/50
```

```
accuracy: 0.9212 - val_loss: 0.2701 - val_accuracy: 0.9097
Epoch 23/50
accuracy: 0.9177 - val_loss: 0.2555 - val_accuracy: 0.9069
Epoch 24/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1884 -
accuracy: 0.9281 - val_loss: 0.3606 - val_accuracy: 0.9056
Epoch 25/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1757 -
accuracy: 0.9323 - val_loss: 0.2506 - val_accuracy: 0.9042
Epoch 26/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2147 -
accuracy: 0.9038 - val_loss: 0.2752 - val_accuracy: 0.9097
Epoch 27/50
accuracy: 0.9351 - val_loss: 0.2382 - val_accuracy: 0.9181
Epoch 28/50
accuracy: 0.9240 - val_loss: 0.3213 - val_accuracy: 0.8861
Epoch 29/50
accuracy: 0.9271 - val_loss: 0.2737 - val_accuracy: 0.9194
Epoch 30/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1603 -
accuracy: 0.9333 - val_loss: 0.3226 - val_accuracy: 0.9097
Epoch 31/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1960 -
accuracy: 0.9222 - val_loss: 0.2813 - val_accuracy: 0.9111
Epoch 32/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2082 -
accuracy: 0.9226 - val_loss: 0.2332 - val_accuracy: 0.9153
Epoch 33/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1522 -
accuracy: 0.9396 - val loss: 0.2986 - val accuracy: 0.9056
Epoch 34/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1808 -
accuracy: 0.9233 - val_loss: 0.2648 - val_accuracy: 0.8917
Epoch 35/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1813 -
accuracy: 0.9319 - val_loss: 0.2622 - val_accuracy: 0.9181
Epoch 36/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1515 -
accuracy: 0.9410 - val_loss: 0.2820 - val_accuracy: 0.9125
Epoch 37/50
accuracy: 0.9448 - val_loss: 0.2997 - val_accuracy: 0.9083
Epoch 38/50
```

```
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1690 -
accuracy: 0.9385 - val_loss: 0.3219 - val_accuracy: 0.9111
Epoch 39/50
accuracy: 0.9427 - val loss: 0.3304 - val accuracy: 0.8972
Epoch 40/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1716 -
accuracy: 0.9365 - val_loss: 0.2649 - val_accuracy: 0.9153
Epoch 41/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1499 -
accuracy: 0.9424 - val_loss: 0.3913 - val_accuracy: 0.8944
Epoch 42/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1635 -
accuracy: 0.9368 - val_loss: 0.2505 - val_accuracy: 0.9153
Epoch 43/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.1178 -
accuracy: 0.9573 - val_loss: 0.3642 - val_accuracy: 0.9194
Epoch 44/50
accuracy: 0.9323 - val_loss: 0.3341 - val_accuracy: 0.9069
Epoch 45/50
accuracy: 0.9503 - val_loss: 0.3841 - val_accuracy: 0.9139
Epoch 46/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1882 -
accuracy: 0.9299 - val_loss: 0.3265 - val_accuracy: 0.8875
Epoch 47/50
2880/2880 [=============== ] - 5s 2ms/step - loss: 0.1396 -
accuracy: 0.9451 - val_loss: 0.4016 - val_accuracy: 0.8889
Epoch 48/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1529 -
accuracy: 0.9378 - val_loss: 0.3519 - val_accuracy: 0.9056
Epoch 49/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1391 -
accuracy: 0.9438 - val loss: 0.4898 - val accuracy: 0.8556
Epoch 50/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1666 -
accuracy: 0.9354 - val_loss: 0.2672 - val_accuracy: 0.8958
1200/1200 [=========== ] - 1s 457us/step
```



\_\_\_\_\_\_

Squential with Kfold CV:

Training for fold 5  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3: Adding layer 4:

Model: "sequential\_10"

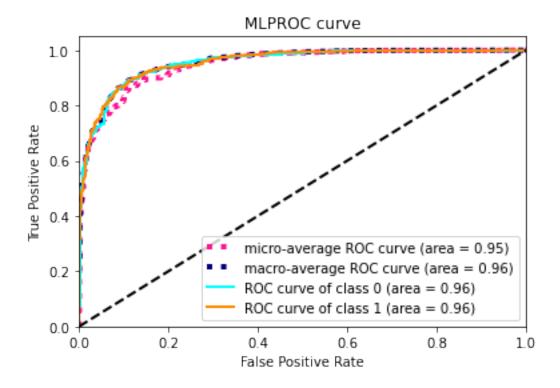
Layer (type)	Output Shape	Param #
dense_50 (Dense)	(None, 4352)	596224
dense_51 (Dense)	(None, 2176)	9472128
dropout_32 (Dropout)	(None, 2176)	0
dense_52 (Dense)	(None, 1088)	2368576
dropout_33 (Dropout)	(None, 1088)	0
dense_53 (Dense)	(None, 544)	592416
dropout_34 (Dropout)	(None, 544)	0

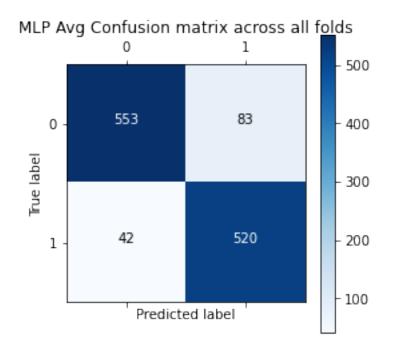
```
(None, 272)
dense_54 (Dense)
                                    148240
dropout_35 (Dropout) (None, 272)
_____
dense_55 (Dense) (None, 2)
                                    546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============== ] - 7s 2ms/step - loss: 2.1300 -
accuracy: 0.5198 - val_loss: 0.6921 - val_accuracy: 0.5264
Epoch 2/50
accuracy: 0.5142 - val_loss: 0.6847 - val_accuracy: 0.5625
Epoch 3/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.7475 -
accuracy: 0.5295 - val_loss: 0.6567 - val_accuracy: 0.6389
Epoch 4/50
accuracy: 0.7010 - val_loss: 0.4485 - val_accuracy: 0.8444
Epoch 5/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.4005 -
accuracy: 0.8538 - val_loss: 0.5102 - val_accuracy: 0.8403
Epoch 6/50
accuracy: 0.8465 - val_loss: 0.3151 - val_accuracy: 0.8833
Epoch 7/50
accuracy: 0.8847 - val_loss: 0.3092 - val_accuracy: 0.8889
Epoch 8/50
accuracy: 0.8750 - val_loss: 0.3001 - val_accuracy: 0.8972
Epoch 9/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.2861 -
accuracy: 0.8997 - val_loss: 0.2657 - val_accuracy: 0.8931
Epoch 10/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.2588 -
accuracy: 0.8931 - val_loss: 0.2634 - val_accuracy: 0.8986
accuracy: 0.8844 - val_loss: 0.2599 - val_accuracy: 0.8917
Epoch 12/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.2221 -
accuracy: 0.9167 - val_loss: 0.3947 - val_accuracy: 0.8708
```

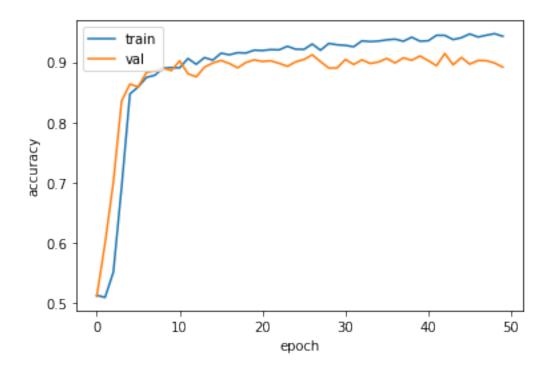
```
Epoch 13/50
accuracy: 0.8941 - val_loss: 0.2548 - val_accuracy: 0.9014
Epoch 14/50
accuracy: 0.9090 - val_loss: 0.2824 - val_accuracy: 0.8958
Epoch 15/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.2513 -
accuracy: 0.9056 - val_loss: 0.3036 - val_accuracy: 0.8889
Epoch 16/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.2351 -
accuracy: 0.9191 - val_loss: 0.2676 - val_accuracy: 0.8944
Epoch 17/50
accuracy: 0.9215 - val_loss: 0.2540 - val_accuracy: 0.8931
Epoch 18/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2313 -
accuracy: 0.9135 - val_loss: 0.3376 - val_accuracy: 0.8917
Epoch 19/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2303 -
accuracy: 0.9181 - val_loss: 0.2426 - val_accuracy: 0.9014
Epoch 20/50
accuracy: 0.9337 - val_loss: 0.3096 - val_accuracy: 0.8875
Epoch 21/50
accuracy: 0.9215 - val_loss: 0.2753 - val_accuracy: 0.9069
Epoch 22/50
accuracy: 0.9115 - val_loss: 0.3017 - val_accuracy: 0.9028
Epoch 23/50
accuracy: 0.9233 - val_loss: 0.2901 - val_accuracy: 0.9000
Epoch 24/50
accuracy: 0.9253 - val_loss: 0.3471 - val_accuracy: 0.8931
Epoch 25/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1814 -
accuracy: 0.9306 - val_loss: 0.2631 - val_accuracy: 0.9042
Epoch 26/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2095 -
accuracy: 0.9215 - val_loss: 0.3057 - val_accuracy: 0.9028
Epoch 27/50
accuracy: 0.9469 - val_loss: 0.2765 - val_accuracy: 0.9097
Epoch 28/50
2880/2880 [============= ] - 7s 2ms/step - loss: 0.2141 -
accuracy: 0.9094 - val_loss: 0.3254 - val_accuracy: 0.8903
```

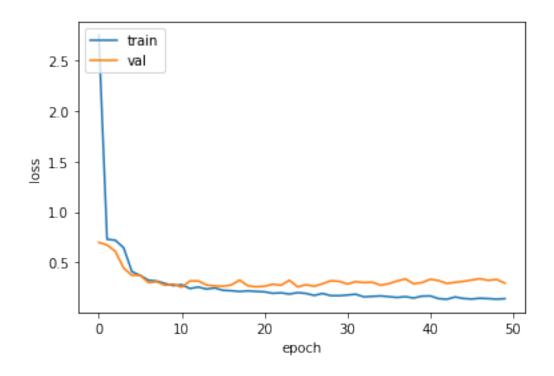
```
Epoch 29/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1768 -
accuracy: 0.9312 - val_loss: 0.4522 - val_accuracy: 0.8361
Epoch 30/50
accuracy: 0.9219 - val_loss: 0.2449 - val_accuracy: 0.9069
Epoch 31/50
2880/2880 [=========== ] - 6s 2ms/step - loss: 0.1764 -
accuracy: 0.9292 - val_loss: 0.3152 - val_accuracy: 0.8903
Epoch 32/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.1792 -
accuracy: 0.9271 - val_loss: 0.3937 - val_accuracy: 0.8653
Epoch 33/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1672 -
accuracy: 0.9358 - val_loss: 0.3032 - val_accuracy: 0.8833
Epoch 34/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1600 -
accuracy: 0.9420 - val_loss: 0.3113 - val_accuracy: 0.9069
Epoch 35/50
accuracy: 0.9319 - val_loss: 0.2950 - val_accuracy: 0.8653
Epoch 36/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1874 -
accuracy: 0.9309 - val_loss: 0.2645 - val_accuracy: 0.8958
Epoch 37/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1435 -
accuracy: 0.9465 - val_loss: 0.4558 - val_accuracy: 0.8583
Epoch 38/50
2880/2880 [============== ] - 6s 2ms/step - loss: 0.1823 -
accuracy: 0.9274 - val_loss: 0.2927 - val_accuracy: 0.9056
Epoch 39/50
2880/2880 [============ ] - 6s 2ms/step - loss: 0.1431 -
accuracy: 0.9434 - val_loss: 0.2538 - val_accuracy: 0.9125
Epoch 40/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.1550 -
accuracy: 0.9434 - val_loss: 0.3835 - val_accuracy: 0.9014
Epoch 41/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2076 -
accuracy: 0.9319 - val_loss: 0.2715 - val_accuracy: 0.9111
Epoch 42/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1330 -
accuracy: 0.9507 - val_loss: 0.3197 - val_accuracy: 0.8931
Epoch 43/50
accuracy: 0.9451 - val_loss: 0.2613 - val_accuracy: 0.8958
Epoch 44/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1494 -
accuracy: 0.9444 - val_loss: 0.2935 - val_accuracy: 0.8847
```

Epoch 45/50 2880/2880 [============ ] - 5s 2ms/step - loss: 0.1551 accuracy: 0.9351 - val\_loss: 0.2830 - val\_accuracy: 0.9083 Epoch 46/50 2880/2880 [============ ] - 5s 2ms/step - loss: 0.1184 accuracy: 0.9576 - val\_loss: 0.4911 - val\_accuracy: 0.8681 Epoch 47/50 accuracy: 0.9372 - val\_loss: 0.2769 - val\_accuracy: 0.9056 Epoch 48/50 2880/2880 [============ ] - 5s 2ms/step - loss: 0.1100 accuracy: 0.9601 - val\_loss: 0.3091 - val\_accuracy: 0.9069 Epoch 49/50 2880/2880 [============== ] - 6s 2ms/step - loss: 0.1492 accuracy: 0.9483 - val\_loss: 0.2839 - val\_accuracy: 0.9042 Epoch 50/50 2880/2880 [============ ] - 5s 2ms/step - loss: 0.1401 accuracy: 0.9434 - val\_loss: 0.3615 - val\_accuracy: 0.8514 1200/1200 [=========== ] - 1s 672us/step









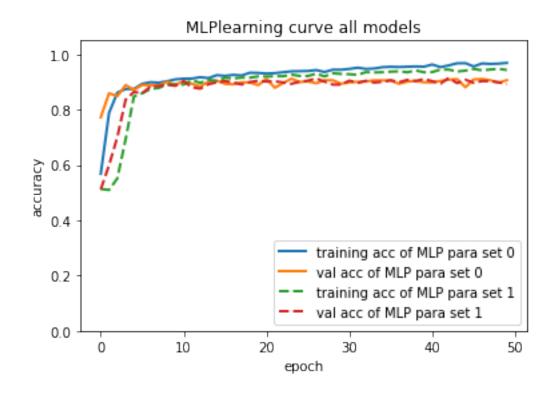
Average scores for pesudo test set across all folds:

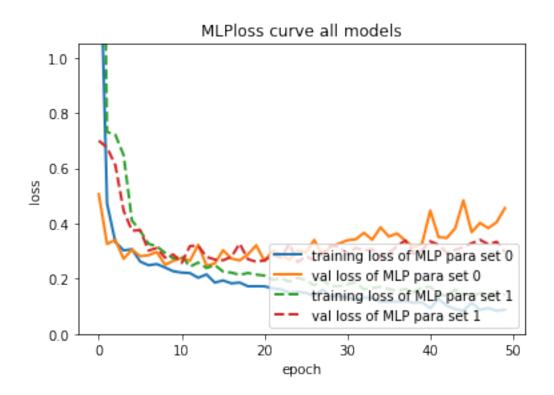
> Accuracy: 0.8950000047683716 (+- 0.018461679153953428)

> Loss: 0.29445346718033155

> Avg runtime per test instance: 0.0018388656377792355

-----





```
Opt model parameter found on the pseudo test set:
    {'num_hidden_layer': 3, 'hidden_layer_activation': ['relu', 'tanh', 'relu'],
    'dropout': [0.5, 0.25, 0.125], 'last_activation': 'softmax'}
    Best average pseudo test set accuracy score with the opt model:
    0.9080000042915344
    Average runtime per test instance: 0.0015766371488571168
[5]: from A2 import A2 functions all as A2 functions
[7]: TaskA2 opt models dict, TaskA2 res dict = A2 functions.get A2 results()
    -----Reading Task A2 dataset-----
    -----Task A2: Gird searching on Logistic
    regression-----
    Logistic Regression Grid search CV on Dataset A:
    Training scores:
    0.514 (+/-0.000) for {'C': 0.0001, 'penalty': 'l1', 'solver': 'saga'}
    0.881 (+/-0.008) for {'C': 0.0001, 'penalty': '12', 'solver': 'saga'}
    0.545 (+/-0.076) for {'C': 0.001, 'penalty': 'l1', 'solver': 'saga'}
    0.889 (+/-0.005) for {'C': 0.001, 'penalty': '12', 'solver': 'saga'}
    0.878 (+/-0.005) for {'C': 0.01, 'penalty': 'l1', 'solver': 'saga'}
    0.898 (+/-0.002) for {'C': 0.01, 'penalty': '12', 'solver': 'saga'}
    0.899 (+/-0.005) for {'C': 0.1, 'penalty': 'l1', 'solver': 'saga'}
    0.903 (+/-0.004) for {'C': 0.1, 'penalty': '12', 'solver': 'saga'}
    0.905 (+/-0.004) for {'C': 1, 'penalty': 'l1', 'solver': 'saga'}
    0.906 (+/-0.005) for {'C': 1, 'penalty': '12', 'solver': 'saga'}
    0.907 (+/-0.008) for {'C': 10, 'penalty': '11', 'solver': 'saga'}
    0.907 (+/-0.008) for {'C': 10, 'penalty': '12', 'solver': 'saga'}
    Validation scores:
    0.514 (+/-0.001) for {'C': 0.0001, 'penalty': '11', 'solver': 'saga'}
    0.880 (+/-0.018) for {'C': 0.0001, 'penalty': '12', 'solver': 'saga'}
    0.545 (+/-0.072) for {'C': 0.001, 'penalty': 'l1', 'solver': 'saga'}
    0.886 (+/-0.012) for {'C': 0.001, 'penalty': '12', 'solver': 'saga'}
    0.876 (+/-0.013) for {'C': 0.01, 'penalty': 'l1', 'solver': 'saga'}
    0.898 (+/-0.009) for {'C': 0.01, 'penalty': '12', 'solver': 'saga'}
    0.895 (+/-0.012) for {'C': 0.1, 'penalty': 'l1', 'solver': 'saga'}
    0.896 (+/-0.009) for {'C': 0.1, 'penalty': '12', 'solver': 'saga'}
```

```
0.895 (+/-0.015) for {'C': 1, 'penalty': 'l1', 'solver': 'saga'}
0.892 (+/-0.020) for {'C': 1, 'penalty': 'l2', 'solver': 'saga'}
0.890 (+/-0.023) for {'C': 10, 'penalty': 'l1', 'solver': 'saga'}
0.891 (+/-0.022) for {'C': 10, 'penalty': 'l2', 'solver': 'saga'}
```

Prediction on a pseudo test set (split from Dataset A):

precision recall f1-score support

	procession	TOOUTT	II DOOLO	Duppor
0.0	0.87	0.90	0.88	596
1.0	0.90	0.87	0.88	604
accuracy			0.88	1200
macro avg	0.88	0.88	0.88	1200
weighted avg	0.88	0.88	0.88	1200

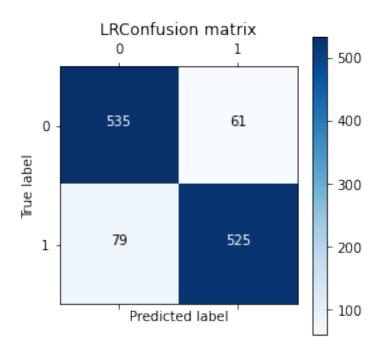
Accuracy: 0.8833333333333333

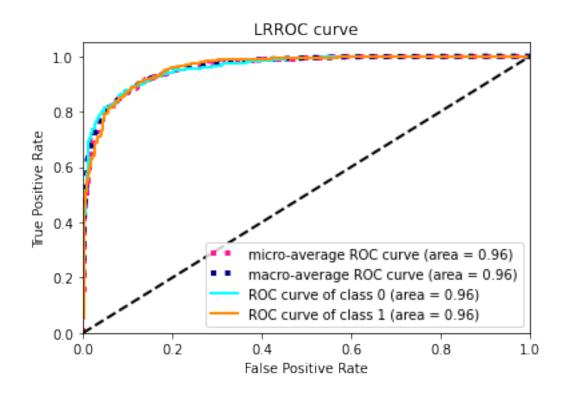
Best parameters found on Dataset A:

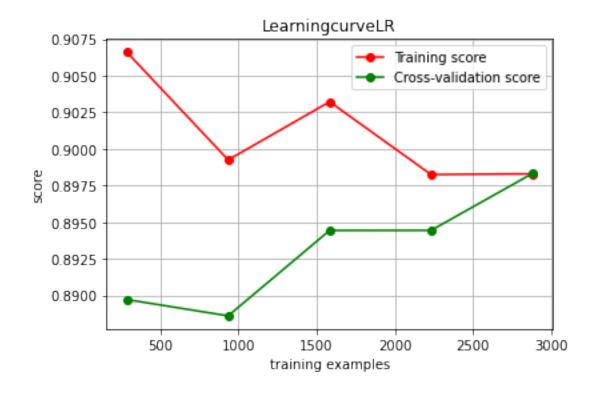
{'C': 0.01, 'penalty': '12', 'solver': 'saga'}

Average runtime per test instance: 1.5525023142496745e-06

-----







```
-----Task A2: Gird searching on
SVM Grid search CV on Dataset A:
Training scores:
0.514 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
0.894 (+/-0.003) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
0.514 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.514 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
0.894 (+/-0.003) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
0.514 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.514 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
0.894 (+/-0.003) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
0.678 (+/-0.008) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.514 (+/-0.000) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
0.894 (+/-0.003) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
0.862 (+/-0.008) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.514 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
0.904 (+/-0.003) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
0.515 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.877 (+/-0.004) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
0.904 (+/-0.003) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
0.795 (+/-0.009) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.865 (+/-0.006) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
0.904 (+/-0.003) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
0.880 (+/-0.005) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.514 (+/-0.000) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
0.904 (+/-0.003) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}
0.771 (+/-0.009) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.875 (+/-0.003) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.907 (+/-0.004) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
0.796 (+/-0.008) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.887 (+/-0.005) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
0.907 (+/-0.004) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
0.881 (+/-0.004) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.897 (+/-0.003) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
0.907 (+/-0.004) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
0.834 (+/-0.002) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.540 (+/-0.007) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
0.907 (+/-0.004) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}
0.725 (+/-0.005) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.886 (+/-0.004) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.913 (+/-0.004) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
```

```
0.881 (+/-0.004) for {'C': 1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.902 (+/-0.003) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.913 (+/-0.004) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
0.893 (+/-0.004) for {'C': 1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.928 (+/-0.003) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.913 (+/-0.004) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
0.795 (+/-0.005) for {'C': 1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.998 (+/-0.001) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.913 (+/-0.004) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.723 (+/-0.006) for {'C': 1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.900 (+/-0.005) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.913 (+/-0.007) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.894 (+/-0.003) for {'C': 10, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.915 (+/-0.005) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.913 (+/-0.007) for {'C': 10, 'gamma': 0.001, 'kernel': 'linear'}
0.875 (+/-0.021) for {'C': 10, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.977 (+/-0.002) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.913 (+/-0.007) for {'C': 10, 'gamma': 0.01, 'kernel': 'linear'}
0.791 (+/-0.006) for {'C': 10, 'gamma': 0.01, 'kernel': 'sigmoid'}
1.000 (+/-0.000) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.913 (+/-0.007) for {'C': 10, 'gamma': 0.1, 'kernel': 'linear'}
0.723 (+/-0.006) for {'C': 10, 'gamma': 0.1, 'kernel': 'sigmoid'}
Validation scores:
0.514 (+/-0.001) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
0.892 (+/-0.014) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
0.514 (+/-0.001) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.514 (+/-0.001) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
0.892 (+/-0.014) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
0.514 (+/-0.001) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.514 (+/-0.001) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
0.892 (+/-0.014) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
0.678 (+/-0.024) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.514 (+/-0.001) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
0.892 (+/-0.014) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
0.859 (+/-0.022) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.514 (+/-0.001) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
0.897 (+/-0.006) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
0.515 (+/-0.001) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.877 (+/-0.021) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
0.897 (+/-0.006) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
0.793 (+/-0.019) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.863 (+/-0.025) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
0.897 (+/-0.006) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
0.880 (+/-0.014) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.514 (+/-0.001) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
```

0.897 (+/-0.006) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}

```
0.770 (+/-0.020) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.876 (+/-0.017) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.894 (+/-0.017) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
0.794 (+/-0.018) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.887 (+/-0.012) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
0.894 (+/-0.017) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
0.881 (+/-0.016) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.888 (+/-0.020) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
0.894 (+/-0.017) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
0.829 (+/-0.008) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.529 (+/-0.007) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
0.894 (+/-0.017) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}
0.727 (+/-0.016) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.885 (+/-0.012) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.890 (+/-0.020) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
0.881 (+/-0.015) for {'C': 1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.899 (+/-0.017) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.890 (+/-0.020) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
0.894 (+/-0.012) for {'C': 1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.890 (+/-0.017) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.890 (+/-0.020) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
0.787 (+/-0.016) for {'C': 1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.865 (+/-0.029) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.890 (+/-0.020) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.728 (+/-0.011) for {'C': 1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.897 (+/-0.012) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.889 (+/-0.021) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.892 (+/-0.014) for {'C': 10, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.900 (+/-0.019) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.889 (+/-0.021) for {'C': 10, 'gamma': 0.001, 'kernel': 'linear'}
0.876 (+/-0.023) for {'C': 10, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.878 (+/-0.019) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.889 (+/-0.021) for {'C': 10, 'gamma': 0.01, 'kernel': 'linear'}
0.786 (+/-0.019) for {'C': 10, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.861 (+/-0.024) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.889 (+/-0.021) for {'C': 10, 'gamma': 0.1, 'kernel': 'linear'}
0.727 (+/-0.011) for {'C': 10, 'gamma': 0.1, 'kernel': 'sigmoid'}
Prediction on a pseudo test set (split from Dataset A):
                           recall f1-score
              precision
                                              support
         0.0
                   0.89
                             0.89
                                       0.89
                                                  596
         1.0
                   0.89
                             0.89
                                       0.89
                                                  604
                                                  1200
    accuracy
                                       0.89
   macro avg
                   0.89
                             0.89
                                       0.89
                                                  1200
weighted avg
                   0.89
                             0.89
                                       0.89
                                                  1200
```

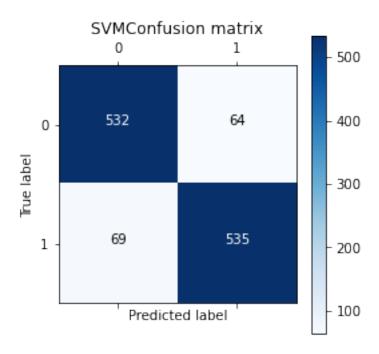
Accuracy: 0.889166666666667

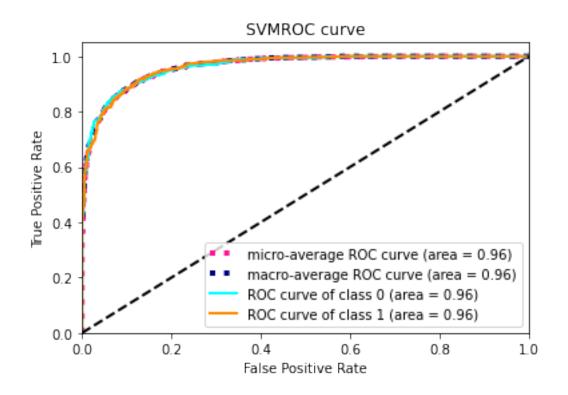
Best parameters found on Dataset A:

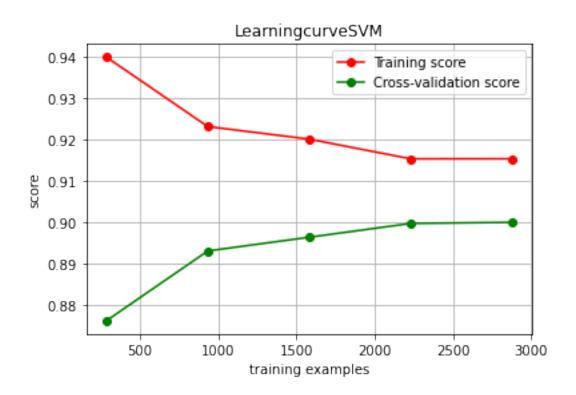
{'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}

Average runtime per test instance: 0.00027517338593800863

\_\_\_\_\_







```
Forest-----
RF Grid search CV on Dataset A:
Training scores:
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 512}
1.000 (+/-0.000) for {'max depth': 128, 'n estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 512}
1.000 (+/-0.000) for {'max depth': 1024, 'n estimators': 1024}
Validation scores:
0.889 (+/-0.008) for {'max_depth': 64, 'n_estimators': 64}
0.892 (+/-0.019) for {'max_depth': 64, 'n_estimators': 128}
0.890 (+/-0.013) for {'max_depth': 64, 'n_estimators': 256}
0.892 (+/-0.015) for {'max_depth': 64, 'n_estimators': 512}
0.893 (+/-0.009) for {'max_depth': 64, 'n_estimators': 1024}
0.889 (+/-0.012) for {'max_depth': 128, 'n_estimators': 64}
0.887 (+/-0.018) for {'max_depth': 128, 'n_estimators': 128}
0.892 (+/-0.013) for {'max_depth': 128, 'n_estimators': 256}
0.890 (+/-0.009) for {'max_depth': 128, 'n_estimators': 512}
0.892 (+/-0.016) for {'max_depth': 128, 'n_estimators': 1024}
```

-----Task A2: Gird searching on Rando

```
0.894 (+/-0.023) for {'max_depth': 256, 'n_estimators': 64}
0.894 (+/-0.011) for {'max_depth': 256, 'n_estimators': 128}
0.892 (+/-0.015) for {'max_depth': 256, 'n_estimators': 256}
0.893 (+/-0.008) for {'max_depth': 256, 'n_estimators': 512}
0.892 (+/-0.014) for {'max depth': 256, 'n estimators': 1024}
0.887 (+/-0.020) for {'max_depth': 512, 'n_estimators': 64}
0.887 \ (+/-0.011) for {'max depth': 512, 'n estimators': 128}
0.891 (+/-0.011) for {'max_depth': 512, 'n_estimators': 256}
0.891 (+/-0.012) for {'max_depth': 512, 'n_estimators': 512}
0.893 (+/-0.011) for {'max_depth': 512, 'n_estimators': 1024}
0.887 (+/-0.007) for {'max_depth': 1024, 'n_estimators': 64}
0.885 (+/-0.016) for {'max_depth': 1024, 'n_estimators': 128}
0.892 (+/-0.009) for {'max_depth': 1024, 'n_estimators': 256}
0.892 \ (+/-0.015) \ for \ {\rm 'max\_depth': 1024, 'n\_estimators': 512}
0.891 (+/-0.013) for {'max_depth': 1024, 'n_estimators': 1024}
```

Prediction on a pseudo test set (split from Dataset A): precision recall f1-score

support

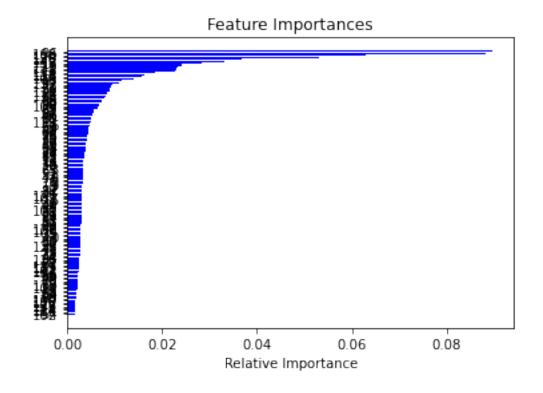
	-			
0.0	0.86	0.90	0.88	596
1.0	0.90	0.86	0.88	604
accuracy			0.88	1200
macro avg	0.88	0.88	0.88	1200
weighted avg	0.88	0.88	0.88	1200

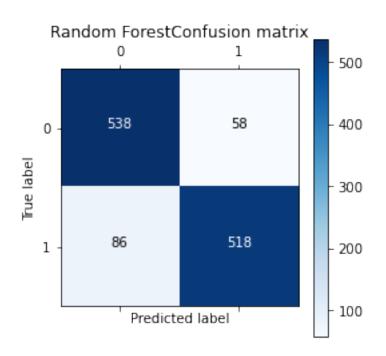
Accuracy: 0.88

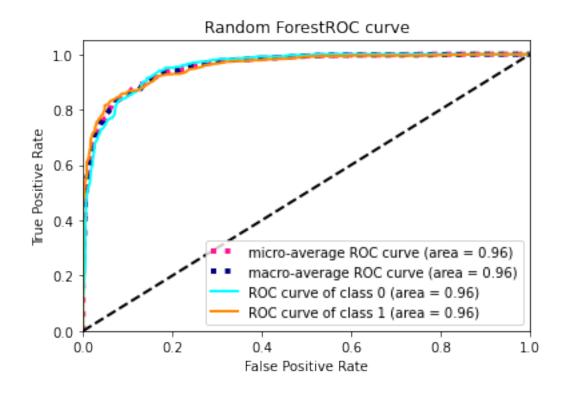
Best parameters found on Dataset A:

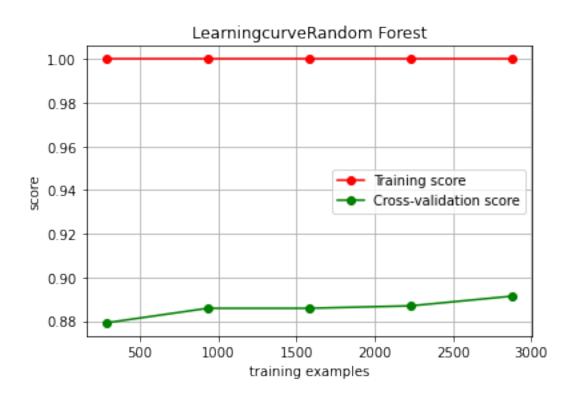
{'max\_depth': 256, 'n\_estimators': 128}

Average runtime per test instance: 5.51903247833252e-05









-----Task A2: Gird searching on KNN-----\_\_\_\_\_\_ KNN Grid search CV on Dataset A: Training scores:

0.894 (+/-0.009) for {'n\_neighbors': 8} 0.885 (+/-0.007) for {'n\_neighbors': 16}  $0.884 \ (+/-0.005)$  for  $\{'n_neighbors': 32\}$  $0.884 \ (+/-0.006)$  for  $\{'n_neighbors': 64\}$ 0.885 (+/-0.005) for {'n\_neighbors': 128}

### Validation scores:

0.870 (+/-0.016) for {'n\_neighbors': 8}  $0.876 \ (+/-0.022) \ for \{'n_neighbors': 16\}$  $0.877 \ (+/-0.020) \ for \{'n_neighbors': 32\}$  $0.881 (+/-0.015) for {'n_neighbors': 64}$  $0.882 (+/-0.014) for {'n_neighbors': 128}$ 

Prediction on a pseudo test set (split from Dataset A): precision recall f1-score support

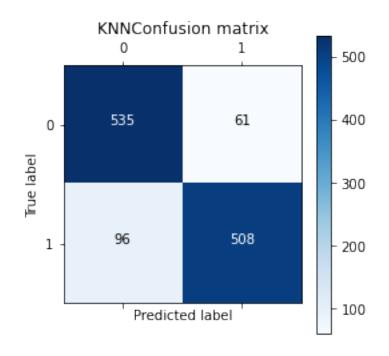
	F			FF
0.0	0.85	0.90	0.87	596
1.0	0.89	0.84	0.87	604
accuracy			0.87	1200
macro avg	0.87	0.87	0.87	1200
weighted avg	0.87	0.87	0.87	1200

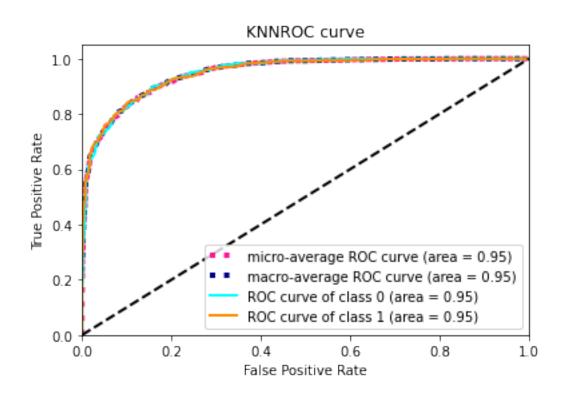
Accuracy: 0.869166666666666

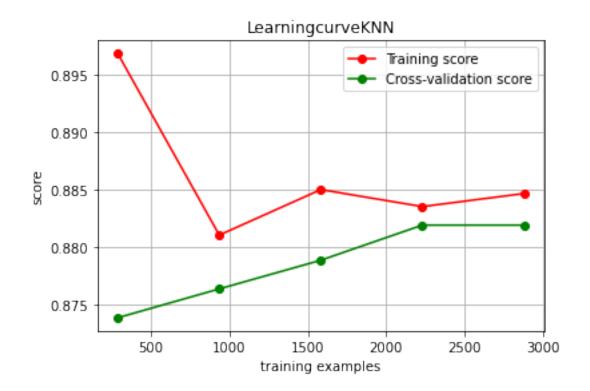
Best parameters found on Dataset A:

{'n\_neighbors': 128}

Average runtime per test instance: 0.001910872459411621







Task	A2:	Gird	searching	on
MI P				

# 

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Squential with Kfold CV:

Training for fold 1  $\dots$ 

Adding layer 1:

Adding layer 2:

Adding layer 3:

Model: "sequential\_11"

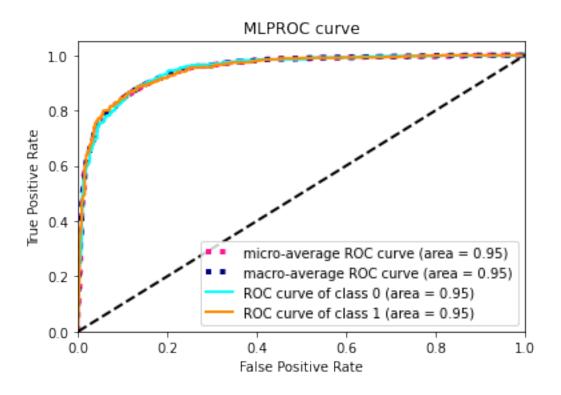
Layer (type)	Output Shape	Param #
dense_56 (Dense)	(None, 2176)	298112
dense_57 (Dense)	(None, 1088)	2368576
dropout_36 (Dropout)	(None, 1088)	0
dense_58 (Dense)	(None, 544)	592416

```
dropout_37 (Dropout) (None, 544)
.....
dense_59 (Dense)
                     (None, 272)
                                         148240
_____
dropout 38 (Dropout) (None, 272)
                                         0
_____
dense_60 (Dense) (None, 2)
                                         546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.6167 - val_loss: 0.3165 - val_accuracy: 0.8653
Epoch 2/50
2880/2880 [============= ] - 2s 672us/step - loss: 0.2700 -
accuracy: 0.8903 - val_loss: 0.3231 - val_accuracy: 0.8694
Epoch 3/50
2880/2880 [============= - - 2s 702us/step - loss: 0.2882 -
accuracy: 0.8792 - val_loss: 0.2612 - val_accuracy: 0.8792
Epoch 4/50
2880/2880 [============== ] - 2s 674us/step - loss: 0.2582 -
accuracy: 0.8913 - val_loss: 0.2821 - val_accuracy: 0.8736
Epoch 5/50
2880/2880 [============= ] - 2s 672us/step - loss: 0.2543 -
accuracy: 0.8885 - val_loss: 0.2681 - val_accuracy: 0.8806
2880/2880 [============= ] - 2s 691us/step - loss: 0.2585 -
accuracy: 0.8962 - val_loss: 0.3080 - val_accuracy: 0.8847
Epoch 7/50
2880/2880 [============= ] - 2s 689us/step - loss: 0.2407 -
accuracy: 0.9000 - val_loss: 0.2841 - val_accuracy: 0.8792
Epoch 8/50
2880/2880 [============== ] - 2s 666us/step - loss: 0.2281 -
accuracy: 0.9073 - val loss: 0.2870 - val accuracy: 0.8806
Epoch 9/50
2880/2880 [============== ] - 2s 663us/step - loss: 0.2517 -
accuracy: 0.9010 - val_loss: 0.2802 - val_accuracy: 0.8833
Epoch 10/50
2880/2880 [============= ] - 2s 666us/step - loss: 0.2047 -
accuracy: 0.9122 - val_loss: 0.3538 - val_accuracy: 0.8792
Epoch 11/50
2880/2880 [============= ] - 2s 671us/step - loss: 0.2580 -
accuracy: 0.9000 - val_loss: 0.2873 - val_accuracy: 0.8917
Epoch 12/50
2880/2880 [============= ] - 2s 674us/step - loss: 0.2213 -
```

```
accuracy: 0.9104 - val_loss: 0.2687 - val_accuracy: 0.8861
Epoch 13/50
2880/2880 [============= ] - 2s 686us/step - loss: 0.2123 -
accuracy: 0.9090 - val_loss: 0.2995 - val_accuracy: 0.8889
Epoch 14/50
2880/2880 [============= - - 2s 689us/step - loss: 0.2207 -
accuracy: 0.9083 - val_loss: 0.3242 - val_accuracy: 0.8847
Epoch 15/50
2880/2880 [============= ] - 2s 685us/step - loss: 0.1927 -
accuracy: 0.9198 - val_loss: 0.3176 - val_accuracy: 0.8986
Epoch 16/50
accuracy: 0.9135 - val_loss: 0.2907 - val_accuracy: 0.8847
Epoch 17/50
2880/2880 [============= ] - 2s 677us/step - loss: 0.1743 -
accuracy: 0.9253 - val_loss: 0.3448 - val_accuracy: 0.8944
Epoch 18/50
2880/2880 [============= ] - 2s 681us/step - loss: 0.1808 -
accuracy: 0.9215 - val_loss: 0.3586 - val_accuracy: 0.8931
Epoch 19/50
2880/2880 [============= ] - 2s 669us/step - loss: 0.1972 -
accuracy: 0.9198 - val_loss: 0.3965 - val_accuracy: 0.8736
Epoch 20/50
2880/2880 [============= ] - 2s 683us/step - loss: 0.1609 -
accuracy: 0.9295 - val_loss: 0.3853 - val_accuracy: 0.8778
Epoch 21/50
2880/2880 [============== ] - 2s 678us/step - loss: 0.1725 -
accuracy: 0.9253 - val_loss: 0.3580 - val_accuracy: 0.8764
2880/2880 [============== ] - 2s 665us/step - loss: 0.1564 -
accuracy: 0.9312 - val_loss: 0.3916 - val_accuracy: 0.8847
2880/2880 [============= ] - 2s 684us/step - loss: 0.1840 -
accuracy: 0.9212 - val_loss: 0.4411 - val_accuracy: 0.8764
Epoch 24/50
2880/2880 [============= ] - 2s 677us/step - loss: 0.1474 -
accuracy: 0.9385 - val loss: 0.4424 - val accuracy: 0.8514
Epoch 25/50
2880/2880 [============== ] - 2s 685us/step - loss: 0.1583 -
accuracy: 0.9323 - val_loss: 0.4163 - val_accuracy: 0.8847
Epoch 26/50
2880/2880 [============= ] - 2s 661us/step - loss: 0.1509 -
accuracy: 0.9382 - val_loss: 0.5442 - val_accuracy: 0.8653
Epoch 27/50
2880/2880 [============= ] - 2s 676us/step - loss: 0.1348 -
accuracy: 0.9458 - val_loss: 0.4648 - val_accuracy: 0.8819
Epoch 28/50
2880/2880 [============= ] - 2s 676us/step - loss: 0.1443 -
```

```
accuracy: 0.9385 - val_loss: 0.5526 - val_accuracy: 0.8736
Epoch 29/50
2880/2880 [============= ] - 2s 671us/step - loss: 0.1136 -
accuracy: 0.9514 - val_loss: 0.6664 - val_accuracy: 0.8861
Epoch 30/50
2880/2880 [============= - - 2s 669us/step - loss: 0.1539 -
accuracy: 0.9444 - val_loss: 0.4499 - val_accuracy: 0.8847
Epoch 31/50
2880/2880 [============== ] - 2s 688us/step - loss: 0.1130 -
accuracy: 0.9514 - val_loss: 0.5451 - val_accuracy: 0.8736
Epoch 32/50
2880/2880 [=============== ] - 2s 681us/step - loss: 0.1221 -
accuracy: 0.9497 - val_loss: 0.5304 - val_accuracy: 0.8847
Epoch 33/50
2880/2880 [============= ] - 2s 677us/step - loss: 0.1298 -
accuracy: 0.9563 - val_loss: 0.4849 - val_accuracy: 0.8861
Epoch 34/50
2880/2880 [============= ] - 2s 667us/step - loss: 0.1030 -
accuracy: 0.9604 - val_loss: 0.9952 - val_accuracy: 0.8653
Epoch 35/50
2880/2880 [============= ] - 2s 681us/step - loss: 0.1211 -
accuracy: 0.9580 - val_loss: 0.5980 - val_accuracy: 0.8778
Epoch 36/50
2880/2880 [============== ] - 2s 678us/step - loss: 0.1242 -
accuracy: 0.9580 - val_loss: 0.4105 - val_accuracy: 0.8667
Epoch 37/50
2880/2880 [============== ] - 2s 666us/step - loss: 0.0998 -
accuracy: 0.9622 - val_loss: 0.5587 - val_accuracy: 0.8792
2880/2880 [============= ] - 2s 680us/step - loss: 0.0841 -
accuracy: 0.9694 - val_loss: 0.5210 - val_accuracy: 0.8681
Epoch 39/50
2880/2880 [============= ] - 2s 665us/step - loss: 0.0963 -
accuracy: 0.9625 - val_loss: 0.4570 - val_accuracy: 0.8417
Epoch 40/50
2880/2880 [============= ] - 2s 674us/step - loss: 0.0779 -
accuracy: 0.9681 - val loss: 0.6022 - val accuracy: 0.8806
Epoch 41/50
2880/2880 [============== ] - 2s 676us/step - loss: 0.1728 -
accuracy: 0.9451 - val_loss: 0.4350 - val_accuracy: 0.8806
Epoch 42/50
2880/2880 [============= ] - 2s 667us/step - loss: 0.0524 -
accuracy: 0.9847 - val_loss: 0.5793 - val_accuracy: 0.8819
Epoch 43/50
2880/2880 [============ ] - 2s 661us/step - loss: 0.0817 -
accuracy: 0.9743 - val_loss: 0.5997 - val_accuracy: 0.8278
Epoch 44/50
2880/2880 [============= ] - 2s 658us/step - loss: 0.0727 -
```

```
accuracy: 0.9722 - val_loss: 0.6618 - val_accuracy: 0.8806
Epoch 45/50
2880/2880 [============== ] - 2s 665us/step - loss: 0.1183 -
accuracy: 0.9625 - val_loss: 0.5416 - val_accuracy: 0.8681
Epoch 46/50
2880/2880 [============== ] - 2s 656us/step - loss: 0.0485 -
accuracy: 0.9830 - val_loss: 0.8821 - val_accuracy: 0.8806
Epoch 47/50
2880/2880 [============= ] - 2s 649us/step - loss: 0.0933 -
accuracy: 0.9622 - val_loss: 0.6086 - val_accuracy: 0.8792
Epoch 48/50
2880/2880 [============== ] - 2s 665us/step - loss: 0.0271 -
accuracy: 0.9885 - val_loss: 0.9270 - val_accuracy: 0.8722
Epoch 49/50
2880/2880 [============= ] - 2s 656us/step - loss: 0.0687 -
accuracy: 0.9743 - val_loss: 0.7302 - val_accuracy: 0.8528
Epoch 50/50
2880/2880 [============= ] - 2s 676us/step - loss: 0.0610 -
accuracy: 0.9729 - val_loss: 0.7779 - val_accuracy: 0.8819
                         ======== ] - Os 255us/step
1200/1200 [=====
```



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Squential with Kfold CV: Training for fold 2 ...

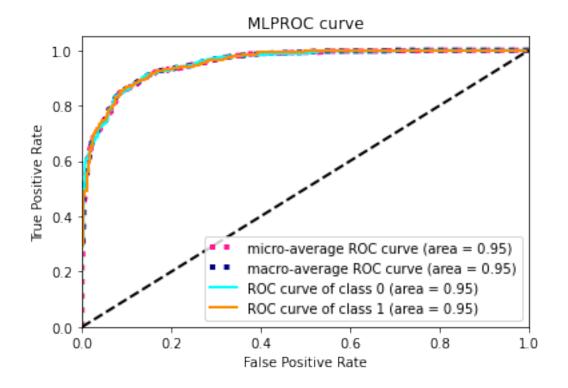
```
Adding layer 2:
Adding layer 3:
Model: "sequential_12"
 .-----
Layer (type)
                Output Shape
______
                 (None, 2176)
dense 61 (Dense)
                                  298112
-----
dense_62 (Dense) (None, 1088)
                                 2368576
dropout_39 (Dropout) (None, 1088)
                           0
-----
dense_63 (Dense) (None, 544)
                                 592416
-----
dropout_40 (Dropout) (None, 544)
-----
dense_64 (Dense)
                 (None, 272)
                                 148240
-----
dropout_41 (Dropout) (None, 272)
_____
dense_65 (Dense) (None, 2) 546
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.6757 - val_loss: 0.2952 - val_accuracy: 0.8736
2880/2880 [============== ] - 2s 665us/step - loss: 0.2618 -
accuracy: 0.8882 - val_loss: 0.2830 - val_accuracy: 0.8861
Epoch 3/50
2880/2880 [============= - - 2s 660us/step - loss: 0.2525 -
accuracy: 0.8913 - val loss: 0.2799 - val accuracy: 0.8875
Epoch 4/50
2880/2880 [============== ] - 2s 664us/step - loss: 0.2599 -
accuracy: 0.8892 - val_loss: 0.3768 - val_accuracy: 0.8306
Epoch 5/50
2880/2880 [=============== ] - 2s 670us/step - loss: 0.2570 -
accuracy: 0.8941 - val_loss: 0.3314 - val_accuracy: 0.8722
Epoch 6/50
2880/2880 [============== ] - 2s 665us/step - loss: 0.2506 -
accuracy: 0.8948 - val_loss: 0.2947 - val_accuracy: 0.8875
Epoch 7/50
2880/2880 [============= ] - 2s 657us/step - loss: 0.2468 -
```

Adding layer 1:

```
accuracy: 0.8969 - val_loss: 0.2823 - val_accuracy: 0.8792
Epoch 8/50
2880/2880 [============= ] - 2s 667us/step - loss: 0.2686 -
accuracy: 0.8847 - val_loss: 0.2747 - val_accuracy: 0.8847
Epoch 9/50
2880/2880 [============= - - 2s 682us/step - loss: 0.2421 -
accuracy: 0.8979 - val_loss: 0.2894 - val_accuracy: 0.8944
Epoch 10/50
2880/2880 [============== ] - 2s 665us/step - loss: 0.2132 -
accuracy: 0.9087 - val_loss: 0.3696 - val_accuracy: 0.8597
Epoch 11/50
2880/2880 [=============== ] - 2s 664us/step - loss: 0.2355 -
accuracy: 0.8983 - val_loss: 0.2920 - val_accuracy: 0.8944
Epoch 12/50
2880/2880 [============= ] - 2s 668us/step - loss: 0.2141 -
accuracy: 0.9101 - val_loss: 0.2996 - val_accuracy: 0.8722
Epoch 13/50
2880/2880 [============= ] - 2s 664us/step - loss: 0.2117 -
accuracy: 0.9090 - val_loss: 0.2941 - val_accuracy: 0.8722
Epoch 14/50
2880/2880 [============= ] - 2s 707us/step - loss: 0.2062 -
accuracy: 0.9104 - val_loss: 0.3328 - val_accuracy: 0.8875
Epoch 15/50
2880/2880 [============== ] - 2s 678us/step - loss: 0.1993 -
accuracy: 0.9083 - val_loss: 0.3467 - val_accuracy: 0.8875
Epoch 16/50
2880/2880 [============== ] - 2s 683us/step - loss: 0.2052 -
accuracy: 0.9094 - val_loss: 0.3267 - val_accuracy: 0.8819
Epoch 17/50
2880/2880 [============= ] - 2s 653us/step - loss: 0.1883 -
accuracy: 0.9125 - val_loss: 0.3544 - val_accuracy: 0.8500
Epoch 18/50
2880/2880 [============= ] - 2s 678us/step - loss: 0.1966 -
accuracy: 0.9094 - val_loss: 0.3522 - val_accuracy: 0.8917
Epoch 19/50
2880/2880 [============= ] - 2s 674us/step - loss: 0.1664 -
accuracy: 0.9250 - val loss: 0.3418 - val accuracy: 0.8778
Epoch 20/50
2880/2880 [============== ] - 2s 686us/step - loss: 0.1750 -
accuracy: 0.9243 - val_loss: 0.3963 - val_accuracy: 0.8764
Epoch 21/50
2880/2880 [============= ] - 2s 671us/step - loss: 0.1693 -
accuracy: 0.9201 - val_loss: 0.3540 - val_accuracy: 0.8611
Epoch 22/50
2880/2880 [============= ] - 2s 664us/step - loss: 0.1826 -
accuracy: 0.9184 - val_loss: 0.3446 - val_accuracy: 0.8653
Epoch 23/50
2880/2880 [============= ] - 2s 662us/step - loss: 0.1399 -
```

```
accuracy: 0.9326 - val_loss: 0.4212 - val_accuracy: 0.8750
Epoch 24/50
2880/2880 [============= ] - 2s 673us/step - loss: 0.1711 -
accuracy: 0.9267 - val_loss: 0.3707 - val_accuracy: 0.8722
Epoch 25/50
2880/2880 [============= - - 2s 681us/step - loss: 0.1653 -
accuracy: 0.9278 - val_loss: 0.3757 - val_accuracy: 0.8736
Epoch 26/50
2880/2880 [============= ] - 2s 681us/step - loss: 0.1633 -
accuracy: 0.9274 - val_loss: 0.3681 - val_accuracy: 0.8972
Epoch 27/50
2880/2880 [=============== ] - 2s 648us/step - loss: 0.1287 -
accuracy: 0.9420 - val_loss: 0.4437 - val_accuracy: 0.8875
Epoch 28/50
2880/2880 [============= ] - 2s 677us/step - loss: 0.1706 -
accuracy: 0.9233 - val_loss: 0.3896 - val_accuracy: 0.8931
Epoch 29/50
2880/2880 [============== ] - 2s 664us/step - loss: 0.1069 -
accuracy: 0.9535 - val_loss: 0.4315 - val_accuracy: 0.8833
Epoch 30/50
2880/2880 [============= ] - 2s 661us/step - loss: 0.1467 -
accuracy: 0.9368 - val_loss: 0.4741 - val_accuracy: 0.8806
Epoch 31/50
accuracy: 0.9507 - val_loss: 0.4199 - val_accuracy: 0.8764
Epoch 32/50
2880/2880 [============== ] - 2s 666us/step - loss: 0.0935 -
accuracy: 0.9604 - val_loss: 0.6921 - val_accuracy: 0.8611
2880/2880 [============= ] - 2s 654us/step - loss: 0.1166 -
accuracy: 0.9528 - val_loss: 0.5261 - val_accuracy: 0.8694
Epoch 34/50
2880/2880 [============= ] - 2s 682us/step - loss: 0.1366 -
accuracy: 0.9406 - val_loss: 0.4879 - val_accuracy: 0.8903
Epoch 35/50
2880/2880 [============== - - 2s 689us/step - loss: 0.1162 -
accuracy: 0.9510 - val loss: 0.4976 - val accuracy: 0.8792
Epoch 36/50
2880/2880 [============== ] - 2s 680us/step - loss: 0.0747 -
accuracy: 0.9691 - val_loss: 0.5380 - val_accuracy: 0.8542
Epoch 37/50
2880/2880 [============== ] - 2s 655us/step - loss: 0.0894 -
accuracy: 0.9604 - val_loss: 0.6493 - val_accuracy: 0.8708
Epoch 38/50
2880/2880 [============== ] - 2s 648us/step - loss: 0.1238 -
accuracy: 0.9507 - val_loss: 0.5079 - val_accuracy: 0.8944
Epoch 39/50
2880/2880 [============= ] - 2s 676us/step - loss: 0.0946 -
```

```
accuracy: 0.9646 - val_loss: 0.5390 - val_accuracy: 0.8778
Epoch 40/50
2880/2880 [============= ] - 2s 671us/step - loss: 0.0936 -
accuracy: 0.9601 - val_loss: 0.7018 - val_accuracy: 0.8694
Epoch 41/50
2880/2880 [============ ] - 2s 679us/step - loss: 0.1003 -
accuracy: 0.9646 - val_loss: 0.4850 - val_accuracy: 0.8861
Epoch 42/50
2880/2880 [============= ] - 2s 658us/step - loss: 0.0664 -
accuracy: 0.9767 - val_loss: 0.5643 - val_accuracy: 0.8833
Epoch 43/50
accuracy: 0.9712 - val_loss: 0.6564 - val_accuracy: 0.8722
Epoch 44/50
2880/2880 [============== ] - 2s 867us/step - loss: 0.0853 -
accuracy: 0.9698 - val_loss: 0.6263 - val_accuracy: 0.8792
Epoch 45/50
2880/2880 [============ ] - 3s 907us/step - loss: 0.0662 -
accuracy: 0.9767 - val_loss: 0.5801 - val_accuracy: 0.8764
Epoch 46/50
2880/2880 [============= ] - 3s 888us/step - loss: 0.0966 -
accuracy: 0.9628 - val_loss: 0.5319 - val_accuracy: 0.8792
Epoch 47/50
2880/2880 [============= ] - 3s 914us/step - loss: 0.0420 -
accuracy: 0.9826 - val_loss: 0.6731 - val_accuracy: 0.8750
Epoch 48/50
2880/2880 [============== ] - 3s 895us/step - loss: 0.0404 -
accuracy: 0.9865 - val_loss: 0.9514 - val_accuracy: 0.8750
2880/2880 [============= ] - 3s 912us/step - loss: 0.0662 -
accuracy: 0.9788 - val_loss: 0.6868 - val_accuracy: 0.8833
Epoch 50/50
2880/2880 [============= ] - 3s 908us/step - loss: 0.1418 -
accuracy: 0.9590 - val_loss: 0.3868 - val_accuracy: 0.8833
1200/1200 [========= ] - 1s 522us/step
```



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Squential with Kfold CV:

Training for fold 3  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

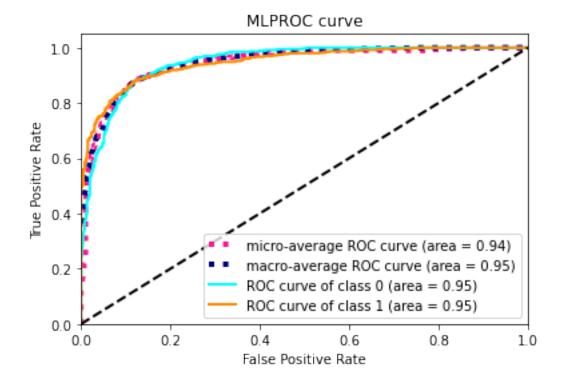
Model: "sequential\_13"

Layer (type)	Output Shape	Param #
dense_66 (Dense)	(None, 2176)	298112
dense_67 (Dense)	(None, 1088)	2368576
dropout_42 (Dropout)	(None, 1088)	0
dense_68 (Dense)	(None, 544)	592416
dropout_43 (Dropout)	(None, 544)	0
dense_69 (Dense)	(None, 272)	148240
dropout_44 (Dropout)	(None, 272)	0

```
dense_70 (Dense)
                        (None, 2)
                                              546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.6687 - val_loss: 0.2822 - val_accuracy: 0.8819
Epoch 2/50
2880/2880 [============== ] - 3s 878us/step - loss: 0.2601 -
accuracy: 0.8951 - val_loss: 0.2621 - val_accuracy: 0.8764
Epoch 3/50
2880/2880 [============= ] - 3s 894us/step - loss: 0.2513 -
accuracy: 0.8951 - val_loss: 0.2804 - val_accuracy: 0.8889
Epoch 4/50
2880/2880 [============= ] - 3s 876us/step - loss: 0.2802 -
accuracy: 0.8875 - val_loss: 0.2570 - val_accuracy: 0.8861
Epoch 5/50
2880/2880 [============= ] - 3s 925us/step - loss: 0.2441 -
accuracy: 0.9000 - val_loss: 0.2606 - val_accuracy: 0.8875
Epoch 6/50
2880/2880 [============= ] - 3s 924us/step - loss: 0.2276 -
accuracy: 0.9031 - val_loss: 0.2806 - val_accuracy: 0.8903
Epoch 7/50
2880/2880 [============= ] - 3s 915us/step - loss: 0.2865 -
accuracy: 0.8774 - val_loss: 0.2724 - val_accuracy: 0.8833
2880/2880 [============= ] - 3s 904us/step - loss: 0.2381 -
accuracy: 0.9028 - val_loss: 0.2676 - val_accuracy: 0.8819
2880/2880 [============= ] - 3s 908us/step - loss: 0.2231 -
accuracy: 0.9014 - val_loss: 0.3284 - val_accuracy: 0.8722
Epoch 10/50
2880/2880 [============= - 3s 910us/step - loss: 0.2361 -
accuracy: 0.9031 - val loss: 0.3256 - val accuracy: 0.8847
Epoch 11/50
2880/2880 [============= ] - 3s 917us/step - loss: 0.2430 -
accuracy: 0.9042 - val_loss: 0.2893 - val_accuracy: 0.8889
Epoch 12/50
2880/2880 [============= ] - 3s 987us/step - loss: 0.2147 -
accuracy: 0.9073 - val_loss: 0.3055 - val_accuracy: 0.8875
Epoch 13/50
2880/2880 [============ ] - 3s 928us/step - loss: 0.2168 -
accuracy: 0.9083 - val_loss: 0.3062 - val_accuracy: 0.8889
Epoch 14/50
2880/2880 [============= ] - 3s 907us/step - loss: 0.2059 -
```

```
accuracy: 0.9174 - val_loss: 0.3462 - val_accuracy: 0.8736
Epoch 15/50
2880/2880 [============ ] - 3s 900us/step - loss: 0.1875 -
accuracy: 0.9194 - val_loss: 0.2985 - val_accuracy: 0.8875
Epoch 16/50
2880/2880 [============= ] - 2s 854us/step - loss: 0.2043 -
accuracy: 0.9167 - val_loss: 0.2989 - val_accuracy: 0.8833
Epoch 17/50
2880/2880 [============= ] - 2s 679us/step - loss: 0.1843 -
accuracy: 0.9229 - val_loss: 0.3003 - val_accuracy: 0.8917
Epoch 18/50
2880/2880 [============== ] - 3s 938us/step - loss: 0.1808 -
accuracy: 0.9240 - val_loss: 0.2795 - val_accuracy: 0.8889
Epoch 19/50
2880/2880 [============= ] - 3s 894us/step - loss: 0.1701 -
accuracy: 0.9267 - val_loss: 0.3941 - val_accuracy: 0.8875
Epoch 20/50
2880/2880 [============= ] - 3s 904us/step - loss: 0.1989 -
accuracy: 0.9139 - val_loss: 0.3566 - val_accuracy: 0.8861
Epoch 21/50
2880/2880 [============= ] - 3s 869us/step - loss: 0.1734 -
accuracy: 0.9253 - val_loss: 0.3401 - val_accuracy: 0.8889
Epoch 22/50
2880/2880 [============= ] - 3s 892us/step - loss: 0.2281 -
accuracy: 0.9146 - val_loss: 0.2824 - val_accuracy: 0.8931
Epoch 23/50
2880/2880 [============== ] - 3s 893us/step - loss: 0.1418 -
accuracy: 0.9424 - val_loss: 0.4321 - val_accuracy: 0.8889
2880/2880 [============= ] - 3s 916us/step - loss: 0.1578 -
accuracy: 0.9330 - val_loss: 0.3554 - val_accuracy: 0.8847
Epoch 25/50
2880/2880 [============= ] - 3s 888us/step - loss: 0.1542 -
accuracy: 0.9368 - val_loss: 0.4161 - val_accuracy: 0.8986
Epoch 26/50
2880/2880 [============= ] - 3s 883us/step - loss: 0.1265 -
accuracy: 0.9413 - val loss: 0.4374 - val accuracy: 0.8639
Epoch 27/50
2880/2880 [============== ] - 3s 920us/step - loss: 0.1697 -
accuracy: 0.9288 - val_loss: 0.3318 - val_accuracy: 0.8889
Epoch 28/50
2880/2880 [============= ] - 3s 893us/step - loss: 0.1299 -
accuracy: 0.9438 - val_loss: 0.3369 - val_accuracy: 0.8861
Epoch 29/50
2880/2880 [============ ] - 3s 891us/step - loss: 0.1138 -
accuracy: 0.9545 - val_loss: 0.6935 - val_accuracy: 0.8639
Epoch 30/50
2880/2880 [============= ] - 3s 904us/step - loss: 0.1377 -
```

```
accuracy: 0.9451 - val_loss: 0.5088 - val_accuracy: 0.8667
Epoch 31/50
2880/2880 [============= ] - 3s 947us/step - loss: 0.1475 -
accuracy: 0.9396 - val_loss: 0.4445 - val_accuracy: 0.8958
Epoch 32/50
2880/2880 [============= ] - 3s 902us/step - loss: 0.1512 -
accuracy: 0.9441 - val_loss: 0.3933 - val_accuracy: 0.8833
Epoch 33/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.0969 -
accuracy: 0.9625 - val_loss: 0.5315 - val_accuracy: 0.8681
Epoch 34/50
2880/2880 [============== ] - 3s 889us/step - loss: 0.1269 -
accuracy: 0.9451 - val_loss: 0.4198 - val_accuracy: 0.8958
Epoch 35/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.0660 -
accuracy: 0.9743 - val_loss: 0.6070 - val_accuracy: 0.8833
Epoch 36/50
2880/2880 [============= ] - 3s 912us/step - loss: 0.0995 -
accuracy: 0.9556 - val_loss: 0.5572 - val_accuracy: 0.8861
Epoch 37/50
2880/2880 [============= ] - 3s 903us/step - loss: 0.1067 -
accuracy: 0.9632 - val_loss: 0.5880 - val_accuracy: 0.8153
Epoch 38/50
2880/2880 [============= ] - 3s 895us/step - loss: 0.1219 -
accuracy: 0.9465 - val_loss: 0.4858 - val_accuracy: 0.8778
Epoch 39/50
2880/2880 [============= ] - 3s 897us/step - loss: 0.0843 -
accuracy: 0.9694 - val_loss: 0.5749 - val_accuracy: 0.8972
2880/2880 [============= ] - 3s 896us/step - loss: 0.1245 -
accuracy: 0.9542 - val_loss: 0.4278 - val_accuracy: 0.8458
Epoch 41/50
2880/2880 [============= ] - 3s 893us/step - loss: 0.0839 -
accuracy: 0.9674 - val_loss: 0.4948 - val_accuracy: 0.8819
Epoch 42/50
2880/2880 [============= ] - 3s 991us/step - loss: 0.0744 -
accuracy: 0.9722 - val loss: 0.5031 - val accuracy: 0.8847
Epoch 43/50
2880/2880 [============= ] - 3s 931us/step - loss: 0.0997 -
accuracy: 0.9622 - val_loss: 0.4847 - val_accuracy: 0.8639
Epoch 44/50
2880/2880 [============= ] - 3s 903us/step - loss: 0.0859 -
accuracy: 0.9705 - val_loss: 0.6577 - val_accuracy: 0.8778
Epoch 45/50
2880/2880 [============ ] - 3s 878us/step - loss: 0.0651 -
accuracy: 0.9736 - val_loss: 0.5535 - val_accuracy: 0.8708
Epoch 46/50
2880/2880 [============= ] - 3s 899us/step - loss: 0.0388 -
```



```
Squential with Kfold CV:

Training for fold 4 ...

Adding layer 1:

Adding layer 2:

Adding layer 3:

Model: "sequential_14"

Layer (type)

Output Shape

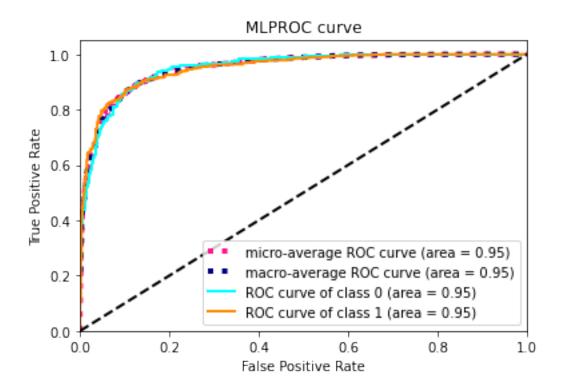
Param #
```

```
______
dense_71 (Dense)
                   (None, 2176)
                                      298112
dense_72 (Dense)
              (None, 1088)
                                     2368576
        ._____
dropout_45 (Dropout) (None, 1088)
_____
dense_73 (Dense) (None, 544)
                               592416
_____
dropout_46 (Dropout) (None, 544)
    -----
dense_74 (Dense) (None, 272)
                                     148240
-----
dropout_47 (Dropout) (None, 272)
_____
dense_75 (Dense) (None, 2)
                                     546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============= ] - 8s 3ms/step - loss: 1.4471 -
accuracy: 0.7215 - val_loss: 0.2626 - val_accuracy: 0.9014
Epoch 2/50
2880/2880 [============== ] - 3s 889us/step - loss: 0.2779 -
accuracy: 0.8816 - val_loss: 0.2465 - val_accuracy: 0.8944
2880/2880 [============= ] - 3s 943us/step - loss: 0.2619 -
accuracy: 0.8872 - val_loss: 0.2581 - val_accuracy: 0.8931
Epoch 4/50
2880/2880 [============= ] - 3s 902us/step - loss: 0.3084 -
accuracy: 0.8715 - val_loss: 0.2345 - val_accuracy: 0.9097
Epoch 5/50
2880/2880 [============= ] - 3s 906us/step - loss: 0.2573 -
accuracy: 0.8910 - val loss: 0.2325 - val accuracy: 0.9069
Epoch 6/50
2880/2880 [============== ] - 3s 897us/step - loss: 0.2558 -
accuracy: 0.8910 - val_loss: 0.3424 - val_accuracy: 0.8444
Epoch 7/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.2860 -
accuracy: 0.8830 - val_loss: 0.2642 - val_accuracy: 0.8903
Epoch 8/50
2880/2880 [============= ] - 3s 911us/step - loss: 0.2487 -
accuracy: 0.8948 - val_loss: 0.2324 - val_accuracy: 0.8986
Epoch 9/50
2880/2880 [============= ] - 3s 909us/step - loss: 0.2318 -
```

```
accuracy: 0.9031 - val_loss: 0.2556 - val_accuracy: 0.9056
Epoch 10/50
2880/2880 [============= ] - 3s 900us/step - loss: 0.2403 -
accuracy: 0.9003 - val_loss: 0.2421 - val_accuracy: 0.9125
Epoch 11/50
2880/2880 [============= ] - 3s 906us/step - loss: 0.2327 -
accuracy: 0.8962 - val_loss: 0.2378 - val_accuracy: 0.8986
Epoch 12/50
2880/2880 [============= ] - 3s 902us/step - loss: 0.2399 -
accuracy: 0.8990 - val_loss: 0.2657 - val_accuracy: 0.9042
Epoch 13/50
2880/2880 [============== ] - 3s 909us/step - loss: 0.2026 -
accuracy: 0.9104 - val_loss: 0.2469 - val_accuracy: 0.8972
Epoch 14/50
2880/2880 [============= ] - 3s 900us/step - loss: 0.2384 -
accuracy: 0.9090 - val_loss: 0.2553 - val_accuracy: 0.8958
Epoch 15/50
2880/2880 [============= ] - 3s 890us/step - loss: 0.1885 -
accuracy: 0.9229 - val_loss: 0.3107 - val_accuracy: 0.8778
Epoch 16/50
2880/2880 [============= ] - 3s 906us/step - loss: 0.2093 -
accuracy: 0.9104 - val_loss: 0.2585 - val_accuracy: 0.9014
Epoch 17/50
2880/2880 [============= ] - 3s 910us/step - loss: 0.1929 -
accuracy: 0.9191 - val_loss: 0.2882 - val_accuracy: 0.8861
Epoch 18/50
2880/2880 [============= ] - 3s 900us/step - loss: 0.1856 -
accuracy: 0.9201 - val_loss: 0.3660 - val_accuracy: 0.8875
2880/2880 [============= ] - 3s 895us/step - loss: 0.1997 -
accuracy: 0.9181 - val_loss: 0.2769 - val_accuracy: 0.9000
Epoch 20/50
2880/2880 [============= ] - 3s 895us/step - loss: 0.1742 -
accuracy: 0.9253 - val_loss: 0.3008 - val_accuracy: 0.8819
Epoch 21/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.2014 -
accuracy: 0.9170 - val_loss: 0.3057 - val_accuracy: 0.8722
Epoch 22/50
2880/2880 [============= ] - 3s 914us/step - loss: 0.1767 -
accuracy: 0.9278 - val_loss: 0.3052 - val_accuracy: 0.8889
Epoch 23/50
2880/2880 [============= ] - 3s 891us/step - loss: 0.1490 -
accuracy: 0.9326 - val_loss: 0.4810 - val_accuracy: 0.8500
Epoch 24/50
2880/2880 [============ ] - 3s 872us/step - loss: 0.1648 -
accuracy: 0.9302 - val_loss: 0.3284 - val_accuracy: 0.8972
Epoch 25/50
2880/2880 [============= ] - 2s 707us/step - loss: 0.1650 -
```

```
accuracy: 0.9285 - val_loss: 0.3273 - val_accuracy: 0.8569
Epoch 26/50
2880/2880 [============ ] - 3s 900us/step - loss: 0.1588 -
accuracy: 0.9295 - val_loss: 0.2960 - val_accuracy: 0.8889
Epoch 27/50
2880/2880 [============= ] - 3s 900us/step - loss: 0.1352 -
accuracy: 0.9417 - val_loss: 0.3957 - val_accuracy: 0.8708
Epoch 28/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.1513 -
accuracy: 0.9337 - val_loss: 0.3651 - val_accuracy: 0.8889
Epoch 29/50
2880/2880 [============== ] - 3s 889us/step - loss: 0.1405 -
accuracy: 0.9476 - val_loss: 0.3597 - val_accuracy: 0.8972
Epoch 30/50
2880/2880 [============= ] - 3s 929us/step - loss: 0.1290 -
accuracy: 0.9441 - val_loss: 0.3216 - val_accuracy: 0.8736
Epoch 31/50
2880/2880 [============= ] - 3s 903us/step - loss: 0.1325 -
accuracy: 0.9448 - val_loss: 0.3807 - val_accuracy: 0.8875
Epoch 32/50
2880/2880 [============= ] - 3s 894us/step - loss: 0.1140 -
accuracy: 0.9521 - val_loss: 0.4041 - val_accuracy: 0.8917
Epoch 33/50
2880/2880 [============ ] - 3s 908us/step - loss: 0.1260 -
accuracy: 0.9483 - val_loss: 0.5162 - val_accuracy: 0.8847
Epoch 34/50
2880/2880 [============== ] - 3s 891us/step - loss: 0.1284 -
accuracy: 0.9524 - val_loss: 0.4059 - val_accuracy: 0.8917
2880/2880 [============= ] - 3s 907us/step - loss: 0.0944 -
accuracy: 0.9649 - val_loss: 0.4876 - val_accuracy: 0.8875
Epoch 36/50
2880/2880 [============= ] - 3s 894us/step - loss: 0.1464 -
accuracy: 0.9538 - val_loss: 0.2882 - val_accuracy: 0.8847
Epoch 37/50
2880/2880 [============= - 3s 905us/step - loss: 0.1421 -
accuracy: 0.9521 - val loss: 0.3105 - val accuracy: 0.8931
Epoch 38/50
2880/2880 [============= ] - 3s 903us/step - loss: 0.0831 -
accuracy: 0.9663 - val_loss: 0.4540 - val_accuracy: 0.8653
Epoch 39/50
2880/2880 [============= ] - 3s 878us/step - loss: 0.0733 -
accuracy: 0.9760 - val_loss: 0.6443 - val_accuracy: 0.8611
Epoch 40/50
2880/2880 [============ ] - 3s 921us/step - loss: 0.1108 -
accuracy: 0.9566 - val_loss: 0.4913 - val_accuracy: 0.8778
Epoch 41/50
2880/2880 [============= ] - 3s 898us/step - loss: 0.0812 -
```

```
accuracy: 0.9653 - val_loss: 0.5369 - val_accuracy: 0.8819
Epoch 42/50
2880/2880 [============= ] - 3s 897us/step - loss: 0.0576 -
accuracy: 0.9778 - val_loss: 0.5726 - val_accuracy: 0.8736
Epoch 43/50
2880/2880 [============= ] - 3s 925us/step - loss: 0.1176 -
accuracy: 0.9524 - val_loss: 0.5216 - val_accuracy: 0.8667
Epoch 44/50
2880/2880 [============= ] - 3s 897us/step - loss: 0.0825 -
accuracy: 0.9674 - val_loss: 0.6639 - val_accuracy: 0.8778
Epoch 45/50
2880/2880 [============== ] - 3s 908us/step - loss: 0.0820 -
accuracy: 0.9701 - val_loss: 0.5551 - val_accuracy: 0.8931
Epoch 46/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.1083 -
accuracy: 0.9615 - val_loss: 0.4892 - val_accuracy: 0.8972
Epoch 47/50
2880/2880 [============= ] - 3s 908us/step - loss: 0.0931 -
accuracy: 0.9670 - val_loss: 0.4419 - val_accuracy: 0.8958
Epoch 48/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.0497 -
accuracy: 0.9823 - val_loss: 0.6675 - val_accuracy: 0.8861
Epoch 49/50
2880/2880 [============ ] - 3s 940us/step - loss: 0.0463 -
accuracy: 0.9854 - val_loss: 0.7629 - val_accuracy: 0.8847
Epoch 50/50
2880/2880 [============= ] - 3s 903us/step - loss: 0.1593 -
accuracy: 0.9503 - val_loss: 0.3756 - val_accuracy: 0.9014
1200/1200 [=========== ] - 1s 647us/step
```



\_\_\_\_\_\_

Squential with Kfold CV:

Training for fold 5  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

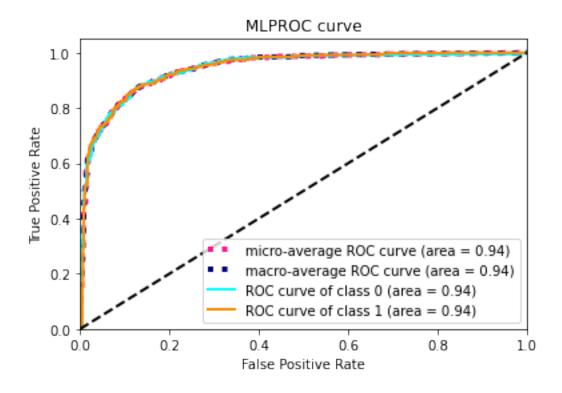
Model: "sequential\_15"

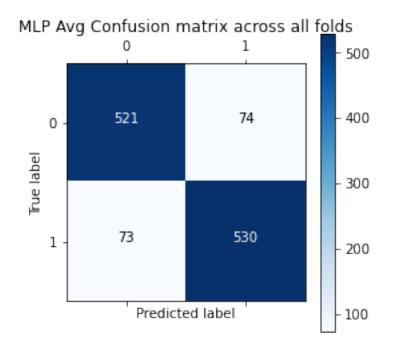
Layer (type)	Output Shape	Param #
dense_76 (Dense)	(None, 2176)	298112
dense_77 (Dense)	(None, 1088)	2368576
dropout_48 (Dropout)	(None, 1088)	0
dense_78 (Dense)	(None, 544)	592416
dropout_49 (Dropout)	(None, 544)	0
dense_79 (Dense)	(None, 272)	148240
dropout_50 (Dropout)	(None, 272)	0

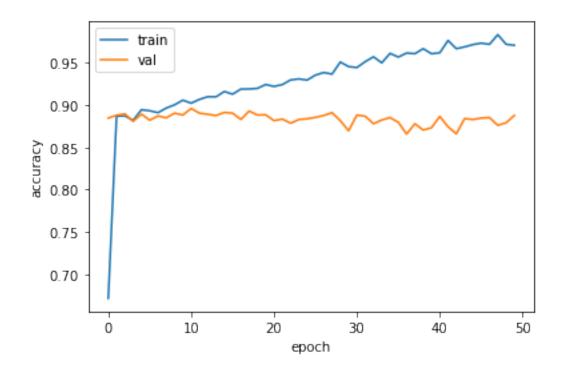
```
dense_80 (Dense)
                        (None, 2)
                                              546
______
Total params: 3,407,890
Trainable params: 3,407,890
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.6747 - val_loss: 0.2449 - val_accuracy: 0.9014
Epoch 2/50
2880/2880 [============== ] - 3s 879us/step - loss: 0.2779 -
accuracy: 0.8799 - val_loss: 0.2211 - val_accuracy: 0.9139
Epoch 3/50
2880/2880 [============= ] - 3s 915us/step - loss: 0.2607 -
accuracy: 0.8840 - val_loss: 0.2233 - val_accuracy: 0.9000
Epoch 4/50
2880/2880 [============= ] - 3s 888us/step - loss: 0.3169 -
accuracy: 0.8705 - val_loss: 0.2268 - val_accuracy: 0.9042
Epoch 5/50
2880/2880 [============= ] - 3s 902us/step - loss: 0.2500 -
accuracy: 0.8993 - val_loss: 0.2195 - val_accuracy: 0.9000
Epoch 6/50
2880/2880 [============= ] - 3s 895us/step - loss: 0.2785 -
accuracy: 0.8819 - val_loss: 0.2205 - val_accuracy: 0.9042
Epoch 7/50
2880/2880 [============= ] - 3s 884us/step - loss: 0.2461 -
accuracy: 0.8983 - val_loss: 0.2717 - val_accuracy: 0.9042
2880/2880 [============= ] - 3s 908us/step - loss: 0.2613 -
accuracy: 0.8934 - val_loss: 0.2845 - val_accuracy: 0.8792
Epoch 9/50
2880/2880 [============= ] - 2s 861us/step - loss: 0.2490 -
accuracy: 0.8986 - val_loss: 0.2208 - val_accuracy: 0.8972
Epoch 10/50
2880/2880 [============= - 3s 893us/step - loss: 0.2397 -
accuracy: 0.9052 - val loss: 0.2510 - val accuracy: 0.9056
Epoch 11/50
2880/2880 [============= ] - 3s 900us/step - loss: 0.2241 -
accuracy: 0.9132 - val_loss: 0.2349 - val_accuracy: 0.9069
Epoch 12/50
2880/2880 [============= ] - 3s 916us/step - loss: 0.2235 -
accuracy: 0.9073 - val_loss: 0.2274 - val_accuracy: 0.9028
Epoch 13/50
2880/2880 [============ ] - 3s 907us/step - loss: 0.2124 -
accuracy: 0.9132 - val_loss: 0.2677 - val_accuracy: 0.8986
Epoch 14/50
2880/2880 [============ ] - 3s 901us/step - loss: 0.2202 -
```

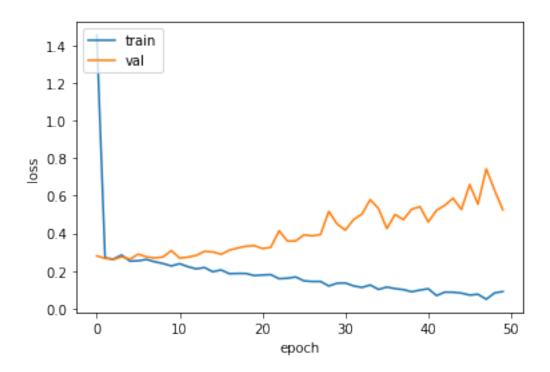
```
accuracy: 0.9042 - val_loss: 0.2622 - val_accuracy: 0.8972
Epoch 15/50
2880/2880 [============= ] - 3s 910us/step - loss: 0.2112 -
accuracy: 0.9104 - val_loss: 0.2339 - val_accuracy: 0.9056
Epoch 16/50
2880/2880 [============= ] - 3s 906us/step - loss: 0.2079 -
accuracy: 0.9149 - val_loss: 0.2675 - val_accuracy: 0.9014
Epoch 17/50
2880/2880 [============= ] - 3s 902us/step - loss: 0.1876 -
accuracy: 0.9153 - val_loss: 0.2652 - val_accuracy: 0.8931
Epoch 18/50
2880/2880 [============== ] - 3s 884us/step - loss: 0.1905 -
accuracy: 0.9208 - val_loss: 0.2535 - val_accuracy: 0.9042
Epoch 19/50
2880/2880 [============= ] - 3s 892us/step - loss: 0.1995 -
accuracy: 0.9097 - val_loss: 0.2474 - val_accuracy: 0.9028
Epoch 20/50
2880/2880 [============= ] - 3s 899us/step - loss: 0.1697 -
accuracy: 0.9292 - val_loss: 0.2343 - val_accuracy: 0.9208
Epoch 21/50
2880/2880 [============= ] - 3s 905us/step - loss: 0.1759 -
accuracy: 0.9229 - val_loss: 0.2388 - val_accuracy: 0.9097
Epoch 22/50
2880/2880 [============= ] - 3s 891us/step - loss: 0.1592 -
accuracy: 0.9295 - val_loss: 0.3032 - val_accuracy: 0.8861
Epoch 23/50
2880/2880 [============= ] - 3s 911us/step - loss: 0.1786 -
accuracy: 0.9205 - val_loss: 0.2947 - val_accuracy: 0.9028
2880/2880 [============= ] - 3s 887us/step - loss: 0.1661 -
accuracy: 0.9267 - val_loss: 0.2978 - val_accuracy: 0.9097
Epoch 25/50
2880/2880 [============= ] - 3s 908us/step - loss: 0.1977 -
accuracy: 0.9233 - val_loss: 0.2632 - val_accuracy: 0.9056
Epoch 26/50
2880/2880 [============== ] - 3s 906us/step - loss: 0.1386 -
accuracy: 0.9413 - val loss: 0.3143 - val accuracy: 0.9125
Epoch 27/50
2880/2880 [============= ] - 3s 891us/step - loss: 0.1523 -
accuracy: 0.9347 - val_loss: 0.2954 - val_accuracy: 0.9097
Epoch 28/50
2880/2880 [============= ] - 3s 1ms/step - loss: 0.1258 -
accuracy: 0.9444 - val_loss: 0.3254 - val_accuracy: 0.9139
Epoch 29/50
2880/2880 [============ ] - 3s 904us/step - loss: 0.1259 -
accuracy: 0.9483 - val_loss: 0.4333 - val_accuracy: 0.8778
Epoch 30/50
2880/2880 [============= ] - 3s 887us/step - loss: 0.1073 -
```

```
accuracy: 0.9576 - val_loss: 0.4942 - val_accuracy: 0.8431
Epoch 31/50
2880/2880 [============= ] - 3s 908us/step - loss: 0.1582 -
accuracy: 0.9361 - val_loss: 0.2955 - val_accuracy: 0.9083
Epoch 32/50
2880/2880 [============= - - 2s 828us/step - loss: 0.1215 -
accuracy: 0.9510 - val_loss: 0.3477 - val_accuracy: 0.9139
Epoch 33/50
2880/2880 [============= ] - 2s 702us/step - loss: 0.0927 -
accuracy: 0.9663 - val_loss: 0.4544 - val_accuracy: 0.8819
Epoch 34/50
2880/2880 [============== ] - 3s 868us/step - loss: 0.1322 -
accuracy: 0.9517 - val_loss: 0.5877 - val_accuracy: 0.8694
Epoch 35/50
2880/2880 [============= ] - 3s 910us/step - loss: 0.1116 -
accuracy: 0.9580 - val_loss: 0.4683 - val_accuracy: 0.8986
Epoch 36/50
2880/2880 [============= ] - 2s 860us/step - loss: 0.1280 -
accuracy: 0.9483 - val_loss: 0.3336 - val_accuracy: 0.9069
Epoch 37/50
2880/2880 [============= ] - 3s 878us/step - loss: 0.0922 -
accuracy: 0.9705 - val_loss: 0.3938 - val_accuracy: 0.8708
Epoch 38/50
2880/2880 [============= ] - 3s 881us/step - loss: 0.0907 -
accuracy: 0.9715 - val_loss: 0.3950 - val_accuracy: 0.8847
Epoch 39/50
2880/2880 [============= ] - 3s 892us/step - loss: 0.0991 -
accuracy: 0.9615 - val_loss: 0.4263 - val_accuracy: 0.8750
2880/2880 [============= ] - 3s 950us/step - loss: 0.0844 -
accuracy: 0.9653 - val_loss: 0.4864 - val_accuracy: 0.8931
Epoch 41/50
2880/2880 [============= ] - 3s 976us/step - loss: 0.0899 -
accuracy: 0.9674 - val_loss: 0.3499 - val_accuracy: 0.9028
Epoch 42/50
2880/2880 [============= ] - 3s 889us/step - loss: 0.0948 -
accuracy: 0.9712 - val loss: 0.3929 - val accuracy: 0.8486
Epoch 43/50
2880/2880 [============== ] - 3s 890us/step - loss: 0.0683 -
accuracy: 0.9740 - val_loss: 0.4843 - val_accuracy: 0.9000
Epoch 44/50
2880/2880 [============= ] - 3s 893us/step - loss: 0.1088 -
accuracy: 0.9649 - val_loss: 0.3260 - val_accuracy: 0.9056
Epoch 45/50
2880/2880 [============ ] - 3s 880us/step - loss: 0.0805 -
accuracy: 0.9753 - val_loss: 0.4009 - val_accuracy: 0.9069
Epoch 46/50
2880/2880 [============= ] - 3s 886us/step - loss: 0.0665 -
```









Average scores for pesudo test set across all folds:

> Accuracy: 0.8764999985694886 (+- 0.002758821595188)

> Loss: 0.5773166338602701

> Avg runtime per test instance: 0.0019980326493581136

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## 

# 

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Squential with Kfold CV:

Training for fold 1  $\dots$ 

Adding layer 1:

Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_16"

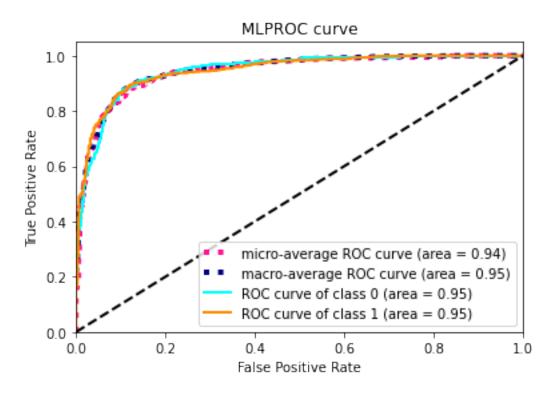
Layer (type)	Output Shape	Param #
dense_81 (Dense)	(None, 4352)	596224
dense_82 (Dense)	(None, 2176)	9472128
dropout_51 (Dropout)	(None, 2176)	0

```
dense_83 (Dense) (None, 1088)
                                  2368576
_____
                (None, 1088)
dropout_52 (Dropout)
_____
dense 84 (Dense) (None, 544)
                                 592416
_____
dropout_53 (Dropout) (None, 544)
_____
           (None, 272)
dense_85 (Dense)
                                 148240
dropout_54 (Dropout) (None, 272)
dense_86 (Dense) (None, 2) 546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============= ] - 16s 6ms/step - loss: 2.7476 -
accuracy: 0.5122 - val_loss: 0.6232 - val_accuracy: 0.5347
Epoch 2/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.5528 -
accuracy: 0.7566 - val_loss: 0.3717 - val_accuracy: 0.8653
Epoch 3/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.3031 -
accuracy: 0.8885 - val_loss: 0.2753 - val_accuracy: 0.8889
accuracy: 0.8802 - val_loss: 0.3042 - val_accuracy: 0.8861
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2788 -
accuracy: 0.8941 - val_loss: 0.3620 - val_accuracy: 0.8778
Epoch 6/50
accuracy: 0.8736 - val loss: 0.2890 - val accuracy: 0.8833
Epoch 7/50
accuracy: 0.8889 - val_loss: 0.2756 - val_accuracy: 0.8778
Epoch 8/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2689 -
accuracy: 0.8990 - val_loss: 0.2691 - val_accuracy: 0.8917
Epoch 9/50
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.2750 -
accuracy: 0.8979 - val_loss: 0.3755 - val_accuracy: 0.8569
Epoch 10/50
```

```
accuracy: 0.8875 - val_loss: 0.2740 - val_accuracy: 0.8861
Epoch 11/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2699 -
accuracy: 0.8941 - val_loss: 0.3855 - val_accuracy: 0.8625
Epoch 12/50
accuracy: 0.8983 - val_loss: 0.3113 - val_accuracy: 0.8667
Epoch 13/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2871 -
accuracy: 0.8816 - val_loss: 0.2865 - val_accuracy: 0.8722
Epoch 14/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2538 -
accuracy: 0.8972 - val_loss: 0.4874 - val_accuracy: 0.8042
Epoch 15/50
accuracy: 0.8997 - val_loss: 0.2886 - val_accuracy: 0.8819
Epoch 16/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2660 -
accuracy: 0.8920 - val_loss: 0.3222 - val_accuracy: 0.8819
Epoch 17/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2268 -
accuracy: 0.9062 - val_loss: 0.2834 - val_accuracy: 0.8778
Epoch 18/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2110 -
accuracy: 0.9163 - val_loss: 0.3194 - val_accuracy: 0.8764
Epoch 19/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2316 -
accuracy: 0.9052 - val_loss: 0.3430 - val_accuracy: 0.8681
accuracy: 0.9083 - val_loss: 0.3365 - val_accuracy: 0.8597
Epoch 21/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2412 -
accuracy: 0.9062 - val_loss: 0.2995 - val_accuracy: 0.8819
Epoch 22/50
2880/2880 [============= ] - 10s 3ms/step - loss: 0.2058 -
accuracy: 0.9208 - val loss: 0.3080 - val accuracy: 0.8778
Epoch 23/50
2880/2880 [============== ] - 10s 4ms/step - loss: 0.2289 -
accuracy: 0.9080 - val_loss: 0.3333 - val_accuracy: 0.8847
Epoch 24/50
2880/2880 [============= ] - 9s 3ms/step - loss: 0.2043 -
accuracy: 0.9205 - val_loss: 0.3003 - val_accuracy: 0.8847
Epoch 25/50
2880/2880 [============= ] - 10s 3ms/step - loss: 0.1998 -
accuracy: 0.9167 - val_loss: 0.3426 - val_accuracy: 0.8667
Epoch 26/50
2880/2880 [============= ] - 9s 3ms/step - loss: 0.2030 -
```

```
accuracy: 0.9229 - val_loss: 0.3541 - val_accuracy: 0.8542
Epoch 27/50
2880/2880 [============== ] - 9s 3ms/step - loss: 0.1856 -
accuracy: 0.9285 - val_loss: 0.3719 - val_accuracy: 0.8778
Epoch 28/50
accuracy: 0.9215 - val_loss: 0.3363 - val_accuracy: 0.8708
Epoch 29/50
2880/2880 [============= ] - 9s 3ms/step - loss: 0.2078 -
accuracy: 0.9128 - val_loss: 0.3816 - val_accuracy: 0.8792
Epoch 30/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1789 -
accuracy: 0.9299 - val_loss: 0.4561 - val_accuracy: 0.8722
Epoch 31/50
accuracy: 0.9302 - val_loss: 0.4405 - val_accuracy: 0.8875
Epoch 32/50
2880/2880 [============ ] - 9s 3ms/step - loss: 0.1894 -
accuracy: 0.9330 - val_loss: 0.3894 - val_accuracy: 0.8806
Epoch 33/50
2880/2880 [============ ] - 9s 3ms/step - loss: 0.1804 -
accuracy: 0.9302 - val_loss: 0.4127 - val_accuracy: 0.8722
Epoch 34/50
2880/2880 [============= ] - 9s 3ms/step - loss: 0.1769 -
accuracy: 0.9292 - val_loss: 0.4147 - val_accuracy: 0.8681
Epoch 35/50
2880/2880 [============== ] - 9s 3ms/step - loss: 0.2012 -
accuracy: 0.9233 - val_loss: 0.3287 - val_accuracy: 0.8931
2880/2880 [============= ] - 9s 3ms/step - loss: 0.1642 -
accuracy: 0.9351 - val_loss: 0.3742 - val_accuracy: 0.8681
Epoch 37/50
2880/2880 [============= ] - 9s 3ms/step - loss: 0.1727 -
accuracy: 0.9312 - val_loss: 0.3257 - val_accuracy: 0.8819
Epoch 38/50
accuracy: 0.9316 - val loss: 0.4294 - val accuracy: 0.8625
Epoch 39/50
accuracy: 0.9306 - val_loss: 0.5126 - val_accuracy: 0.8569
Epoch 40/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1692 -
accuracy: 0.9306 - val_loss: 0.5037 - val_accuracy: 0.8819
Epoch 41/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1592 -
accuracy: 0.9361 - val_loss: 0.4393 - val_accuracy: 0.8750
Epoch 42/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1482 -
```

```
accuracy: 0.9424 - val_loss: 0.4765 - val_accuracy: 0.8514
Epoch 43/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1831 -
accuracy: 0.9288 - val_loss: 0.4523 - val_accuracy: 0.8722
Epoch 44/50
accuracy: 0.9389 - val_loss: 0.4537 - val_accuracy: 0.8583
Epoch 45/50
accuracy: 0.9431 - val_loss: 0.5700 - val_accuracy: 0.8694
Epoch 46/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1595 -
accuracy: 0.9361 - val_loss: 0.4501 - val_accuracy: 0.8569
Epoch 47/50
accuracy: 0.9483 - val_loss: 0.4488 - val_accuracy: 0.8750
Epoch 48/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2016 -
accuracy: 0.9323 - val_loss: 0.3818 - val_accuracy: 0.8694
Epoch 49/50
accuracy: 0.9556 - val_loss: 0.4375 - val_accuracy: 0.8500
Epoch 50/50
2880/2880 [=====
                  ========] - 8s 3ms/step - loss: 0.1609 -
accuracy: 0.9347 - val_loss: 0.3806 - val_accuracy: 0.8681
1200/1200 [========== ] - 2s 1ms/step
```

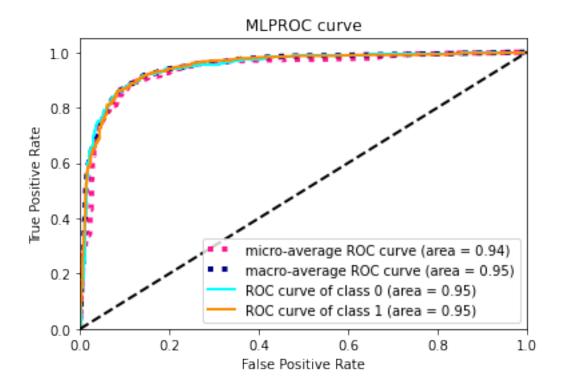


```
Squential with Kfold CV:
Training for fold 2 ...
Adding layer 1:
Adding layer 2:
Adding layer 3:
Adding layer 4:
Model: "sequential_17"
Layer (type) Output Shape
______
dense_87 (Dense)
                (None, 4352)
                                596224
            (None, 2176)
dense_88 (Dense)
                                9472128
dropout_55 (Dropout) (None, 2176)
_____
dense_89 (Dense) (None, 1088)
                                2368576
______
dropout_56 (Dropout) (None, 1088)
dense_90 (Dense) (None, 544)
                                592416
_____
dropout_57 (Dropout) (None, 544)
_____
dense_91 (Dense) (None, 272)
                                148240
dropout_58 (Dropout) (None, 272)
dense_92 (Dense) (None, 2) 546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
2880/2880 [============= ] - 15s 5ms/step - loss: 2.6515 -
accuracy: 0.5677 - val_loss: 0.6666 - val_accuracy: 0.8736
Epoch 2/50
accuracy: 0.6174 - val_loss: 0.4178 - val_accuracy: 0.8694
Epoch 3/50
accuracy: 0.8826 - val_loss: 0.4894 - val_accuracy: 0.8403
```

```
Epoch 4/50
accuracy: 0.8774 - val_loss: 0.3923 - val_accuracy: 0.8264
2880/2880 [============= ] - 8s 3ms/step - loss: 0.3257 -
accuracy: 0.8795 - val_loss: 0.2811 - val_accuracy: 0.8778
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.3412 -
accuracy: 0.8740 - val_loss: 0.3365 - val_accuracy: 0.8833
Epoch 7/50
accuracy: 0.8924 - val_loss: 0.3362 - val_accuracy: 0.8611
Epoch 8/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2899 -
accuracy: 0.8840 - val_loss: 0.3514 - val_accuracy: 0.8389
Epoch 9/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2716 -
accuracy: 0.8927 - val_loss: 0.3698 - val_accuracy: 0.8514
Epoch 10/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2842 -
accuracy: 0.8885 - val_loss: 0.4094 - val_accuracy: 0.8194
Epoch 11/50
accuracy: 0.8865 - val_loss: 0.3067 - val_accuracy: 0.8667
Epoch 12/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2619 -
accuracy: 0.8938 - val_loss: 0.3090 - val_accuracy: 0.8944
Epoch 13/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2733 -
accuracy: 0.9010 - val_loss: 0.2937 - val_accuracy: 0.8931
Epoch 14/50
accuracy: 0.8965 - val_loss: 0.3299 - val_accuracy: 0.8625
Epoch 15/50
accuracy: 0.8962 - val_loss: 0.2934 - val_accuracy: 0.8694
Epoch 16/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2497 -
accuracy: 0.8993 - val_loss: 0.3532 - val_accuracy: 0.8458
Epoch 17/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2256 -
accuracy: 0.9069 - val_loss: 0.3549 - val_accuracy: 0.8417
Epoch 18/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2421 -
accuracy: 0.9038 - val_loss: 0.2988 - val_accuracy: 0.8944
Epoch 19/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2433 -
accuracy: 0.9038 - val_loss: 0.2897 - val_accuracy: 0.8944
```

```
Epoch 20/50
accuracy: 0.9128 - val_loss: 0.3339 - val_accuracy: 0.8889
Epoch 21/50
accuracy: 0.8958 - val_loss: 0.3227 - val_accuracy: 0.8875
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.2044 -
accuracy: 0.9181 - val_loss: 0.3126 - val_accuracy: 0.8944
Epoch 23/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2251 -
accuracy: 0.9118 - val_loss: 0.3064 - val_accuracy: 0.8722
Epoch 24/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2124 -
accuracy: 0.9194 - val_loss: 0.3312 - val_accuracy: 0.8778
Epoch 25/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1951 -
accuracy: 0.9201 - val_loss: 0.3669 - val_accuracy: 0.8597
Epoch 26/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2149 -
accuracy: 0.9042 - val_loss: 0.3057 - val_accuracy: 0.8875
Epoch 27/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1959 -
accuracy: 0.9240 - val_loss: 0.3091 - val_accuracy: 0.8944
Epoch 28/50
accuracy: 0.9260 - val_loss: 0.3564 - val_accuracy: 0.9014
Epoch 29/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1986 -
accuracy: 0.9181 - val_loss: 0.3189 - val_accuracy: 0.8708
Epoch 30/50
accuracy: 0.9240 - val_loss: 0.3822 - val_accuracy: 0.8722
Epoch 31/50
accuracy: 0.9215 - val_loss: 0.3323 - val_accuracy: 0.8903
Epoch 32/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2072 -
accuracy: 0.9108 - val_loss: 0.3678 - val_accuracy: 0.8583
Epoch 33/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1831 -
accuracy: 0.9236 - val_loss: 0.3412 - val_accuracy: 0.8875
Epoch 34/50
2880/2880 [============= - - 8s 3ms/step - loss: 0.2014 -
accuracy: 0.9087 - val_loss: 0.3635 - val_accuracy: 0.8431
Epoch 35/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1665 -
accuracy: 0.9392 - val_loss: 0.3283 - val_accuracy: 0.8875
```

```
Epoch 36/50
accuracy: 0.9372 - val_loss: 0.3684 - val_accuracy: 0.8889
Epoch 37/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1764 -
accuracy: 0.9267 - val_loss: 0.3349 - val_accuracy: 0.8917
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.1567 -
accuracy: 0.9372 - val_loss: 0.3436 - val_accuracy: 0.8861
Epoch 39/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1715 -
accuracy: 0.9274 - val_loss: 0.3944 - val_accuracy: 0.8958
Epoch 40/50
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.1642 -
accuracy: 0.9312 - val_loss: 0.3418 - val_accuracy: 0.8764
Epoch 41/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1691 -
accuracy: 0.9340 - val_loss: 0.4182 - val_accuracy: 0.8875
Epoch 42/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1608 -
accuracy: 0.9309 - val_loss: 0.4583 - val_accuracy: 0.8153
Epoch 43/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1581 -
accuracy: 0.9333 - val_loss: 0.4011 - val_accuracy: 0.8944
Epoch 44/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1648 -
accuracy: 0.9389 - val_loss: 0.3639 - val_accuracy: 0.9000
Epoch 45/50
accuracy: 0.9399 - val_loss: 0.3668 - val_accuracy: 0.8778
Epoch 46/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1460 -
accuracy: 0.9448 - val_loss: 0.4534 - val_accuracy: 0.8833
Epoch 47/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1960 -
accuracy: 0.9306 - val_loss: 0.3789 - val_accuracy: 0.8944
Epoch 48/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1391 -
accuracy: 0.9455 - val_loss: 0.4058 - val_accuracy: 0.8806
Epoch 49/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1577 -
accuracy: 0.9337 - val_loss: 0.4215 - val_accuracy: 0.8736
Epoch 50/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1304 -
accuracy: 0.9490 - val_loss: 0.4478 - val_accuracy: 0.8917
1200/1200 [========== ] - 2s 1ms/step
```



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Squential with Kfold CV:

Training for fold 3  $\dots$ 

Adding layer 1: Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_18"

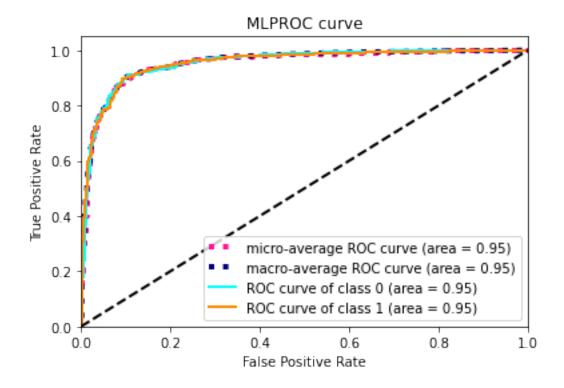
Layer (type)	Output Shape	Param #
dense_93 (Dense)	(None, 4352)	596224
dense_94 (Dense)	(None, 2176)	9472128
dropout_59 (Dropout)	(None, 2176)	0
dense_95 (Dense)	(None, 1088)	2368576
dropout_60 (Dropout)	(None, 1088)	0
dense_96 (Dense)	(None, 544)	592416
dropout_61 (Dropout)	(None, 544)	0

```
(None, 272)
dense_97 (Dense)
                                    148240
dropout_62 (Dropout) (None, 272)
                                    Ο
_____
dense_98 (Dense) (None, 2)
                                    546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.5552 - val_loss: 0.6539 - val_accuracy: 0.4958
Epoch 2/50
accuracy: 0.6003 - val_loss: 0.3858 - val_accuracy: 0.8667
Epoch 3/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.3764 -
accuracy: 0.8573 - val_loss: 0.5026 - val_accuracy: 0.8306
Epoch 4/50
accuracy: 0.8753 - val_loss: 0.2632 - val_accuracy: 0.9083
Epoch 5/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2937 -
accuracy: 0.8858 - val_loss: 0.2447 - val_accuracy: 0.9125
Epoch 6/50
accuracy: 0.8826 - val_loss: 0.2691 - val_accuracy: 0.8833
Epoch 7/50
accuracy: 0.8771 - val_loss: 0.3314 - val_accuracy: 0.8597
Epoch 8/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2823 -
accuracy: 0.8944 - val_loss: 0.4300 - val_accuracy: 0.8083
Epoch 9/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.3191 -
accuracy: 0.8809 - val_loss: 0.2538 - val_accuracy: 0.9139
Epoch 10/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2640 -
accuracy: 0.8951 - val_loss: 0.2674 - val_accuracy: 0.8972
accuracy: 0.8913 - val_loss: 0.3390 - val_accuracy: 0.8653
Epoch 12/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2965 -
accuracy: 0.8830 - val_loss: 0.2970 - val_accuracy: 0.8653
```

```
Epoch 13/50
accuracy: 0.8965 - val_loss: 0.2612 - val_accuracy: 0.8903
Epoch 14/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2701 -
accuracy: 0.8955 - val_loss: 0.2513 - val_accuracy: 0.9042
Epoch 15/50
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.2564 -
accuracy: 0.9031 - val_loss: 0.4532 - val_accuracy: 0.8208
Epoch 16/50
accuracy: 0.8861 - val_loss: 0.2459 - val_accuracy: 0.9042
Epoch 17/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2531 -
accuracy: 0.8969 - val_loss: 0.2816 - val_accuracy: 0.8931
Epoch 18/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2278 -
accuracy: 0.9045 - val_loss: 0.3271 - val_accuracy: 0.8833
Epoch 19/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2791 -
accuracy: 0.8889 - val_loss: 0.2617 - val_accuracy: 0.8972
Epoch 20/50
accuracy: 0.9087 - val_loss: 0.2941 - val_accuracy: 0.8764
Epoch 21/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2234 -
accuracy: 0.9035 - val_loss: 0.2915 - val_accuracy: 0.8917
Epoch 22/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2319 -
accuracy: 0.9056 - val_loss: 0.2543 - val_accuracy: 0.9042
Epoch 23/50
accuracy: 0.9198 - val_loss: 0.2955 - val_accuracy: 0.8931
Epoch 24/50
accuracy: 0.9045 - val_loss: 0.3086 - val_accuracy: 0.8986
Epoch 25/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2048 -
accuracy: 0.9208 - val_loss: 0.2886 - val_accuracy: 0.8972
Epoch 26/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2300 -
accuracy: 0.9125 - val_loss: 0.2779 - val_accuracy: 0.9000
Epoch 27/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2113 -
accuracy: 0.9128 - val_loss: 0.3034 - val_accuracy: 0.8833
Epoch 28/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2039 -
accuracy: 0.9108 - val_loss: 0.2723 - val_accuracy: 0.9000
```

```
Epoch 29/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2217 -
accuracy: 0.9118 - val_loss: 0.2880 - val_accuracy: 0.8847
Epoch 30/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1772 -
accuracy: 0.9271 - val_loss: 0.2990 - val_accuracy: 0.8944
Epoch 31/50
2880/2880 [=========== ] - 8s 3ms/step - loss: 0.2091 -
accuracy: 0.9215 - val_loss: 0.2933 - val_accuracy: 0.9028
Epoch 32/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1935 -
accuracy: 0.9198 - val_loss: 0.2933 - val_accuracy: 0.8972
Epoch 33/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.2169 -
accuracy: 0.9160 - val_loss: 0.3538 - val_accuracy: 0.8611
Epoch 34/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.2035 -
accuracy: 0.9142 - val_loss: 0.2850 - val_accuracy: 0.8667
Epoch 35/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1674 -
accuracy: 0.9330 - val_loss: 0.2922 - val_accuracy: 0.8986
Epoch 36/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1751 -
accuracy: 0.9302 - val_loss: 0.3174 - val_accuracy: 0.9028
Epoch 37/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1756 -
accuracy: 0.9295 - val_loss: 0.4567 - val_accuracy: 0.8917
Epoch 38/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1875 -
accuracy: 0.9257 - val_loss: 0.3012 - val_accuracy: 0.8889
Epoch 39/50
accuracy: 0.9444 - val_loss: 0.4646 - val_accuracy: 0.8903
Epoch 40/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1713 -
accuracy: 0.9292 - val_loss: 0.5024 - val_accuracy: 0.8819
Epoch 41/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1805 -
accuracy: 0.9260 - val_loss: 0.3365 - val_accuracy: 0.8917
Epoch 42/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1611 -
accuracy: 0.9354 - val_loss: 0.3087 - val_accuracy: 0.8903
Epoch 43/50
2880/2880 [============== ] - 8s 3ms/step - loss: 0.1467 -
accuracy: 0.9403 - val_loss: 0.4310 - val_accuracy: 0.8944
Epoch 44/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1888 -
accuracy: 0.9365 - val_loss: 0.3109 - val_accuracy: 0.8889
```

```
Epoch 45/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1517 -
accuracy: 0.9431 - val_loss: 0.4029 - val_accuracy: 0.8833
Epoch 46/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1633 -
accuracy: 0.9340 - val_loss: 0.3644 - val_accuracy: 0.8972
Epoch 47/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1365 -
accuracy: 0.9469 - val_loss: 0.3793 - val_accuracy: 0.8875
Epoch 48/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.1652 -
accuracy: 0.9316 - val_loss: 0.4218 - val_accuracy: 0.8944
Epoch 49/50
accuracy: 0.9375 - val_loss: 0.3030 - val_accuracy: 0.8764
Epoch 50/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.1337 -
accuracy: 0.9497 - val_loss: 0.3443 - val_accuracy: 0.9042
1200/1200 [=========== ] - 2s 1ms/step
```



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Squential with Kfold CV: Training for fold 4 ... Adding layer 1: Adding layer 2: Adding layer 3: Adding layer 4:

Epoch 6/50

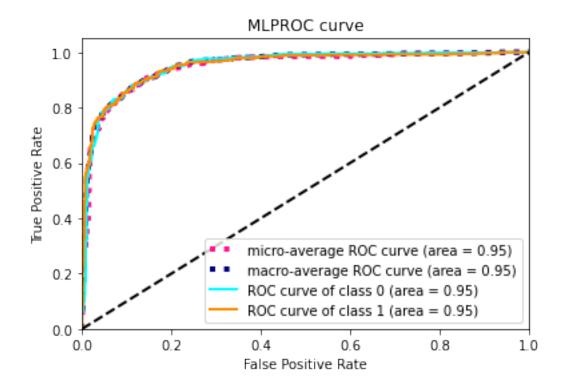
Model: "sequential\_19"

_			
Layer (type)	_	•	Param #
		4352)	596224
dense_100 (Dense)	(None,	2176)	9472128
dropout_63 (Dropout)		2176)	0
dense_101 (Dense)	(None,	1088)	2368576
dropout_64 (Dropout)	(None,	1088)	0
dense_102 (Dense)	(None,	544)	592416
dropout_65 (Dropout)	(None,	544)	0
dense_103 (Dense)	(None,	272)	148240
dropout_66 (Dropout)	(None,	272)	0
dense_104 (Dense)			546
Total params: 13,178,130 Trainable params: 13,178,130 Non-trainable params: 0			
Train on 2880 samples, valid		-	
2880/2880 [===================================			•
2880/2880 [===================================			-
Epoch 3/50 2880/2880 [===================================			-
Epoch 4/50 2880/2880 [===================================			-
Epoch 5/50 2880/2880 [===================================			•

```
accuracy: 0.8667 - val_loss: 0.3093 - val_accuracy: 0.8833
Epoch 7/50
accuracy: 0.8948 - val_loss: 0.3598 - val_accuracy: 0.8778
Epoch 8/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.3011 -
accuracy: 0.8785 - val_loss: 0.3298 - val_accuracy: 0.8792
Epoch 9/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2592 -
accuracy: 0.9042 - val_loss: 0.4309 - val_accuracy: 0.8375
Epoch 10/50
2880/2880 [============ ] - 8s 3ms/step - loss: 0.3165 -
accuracy: 0.8646 - val_loss: 0.3092 - val_accuracy: 0.8806
Epoch 11/50
accuracy: 0.9010 - val_loss: 0.3136 - val_accuracy: 0.8750
Epoch 12/50
accuracy: 0.9007 - val_loss: 0.3186 - val_accuracy: 0.8722
Epoch 13/50
accuracy: 0.9052 - val_loss: 0.3294 - val_accuracy: 0.8736
Epoch 14/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2541 -
accuracy: 0.9028 - val_loss: 0.3281 - val_accuracy: 0.8722
Epoch 15/50
accuracy: 0.9042 - val_loss: 0.3946 - val_accuracy: 0.8569
Epoch 16/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2378 -
accuracy: 0.9028 - val_loss: 0.3473 - val_accuracy: 0.8792
Epoch 17/50
accuracy: 0.9132 - val loss: 0.3093 - val accuracy: 0.8694
Epoch 18/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2713 -
accuracy: 0.8962 - val_loss: 0.3820 - val_accuracy: 0.8806
Epoch 19/50
2880/2880 [============= ] - 8s 3ms/step - loss: 0.2215 -
accuracy: 0.9003 - val_loss: 0.3886 - val_accuracy: 0.8819
Epoch 20/50
accuracy: 0.9208 - val_loss: 0.3782 - val_accuracy: 0.8653
Epoch 21/50
accuracy: 0.9056 - val_loss: 0.4440 - val_accuracy: 0.8708
Epoch 22/50
```

```
accuracy: 0.9087 - val_loss: 0.3714 - val_accuracy: 0.8514
Epoch 23/50
accuracy: 0.9146 - val_loss: 0.4341 - val_accuracy: 0.8778
Epoch 24/50
2880/2880 [============= ] - 6s 2ms/step - loss: 0.2124 -
accuracy: 0.9187 - val_loss: 0.3761 - val_accuracy: 0.8819
Epoch 25/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1913 -
accuracy: 0.9215 - val_loss: 0.4140 - val_accuracy: 0.8458
Epoch 26/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2113 -
accuracy: 0.9184 - val_loss: 0.3789 - val_accuracy: 0.8750
Epoch 27/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2296 -
accuracy: 0.9056 - val_loss: 0.3365 - val_accuracy: 0.8833
Epoch 28/50
accuracy: 0.9253 - val_loss: 0.3952 - val_accuracy: 0.8875
Epoch 29/50
accuracy: 0.9146 - val_loss: 0.3621 - val_accuracy: 0.8653
Epoch 30/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1890 -
accuracy: 0.9267 - val_loss: 0.6145 - val_accuracy: 0.8750
Epoch 31/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1983 -
accuracy: 0.9309 - val_loss: 0.3968 - val_accuracy: 0.8681
Epoch 32/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1963 -
accuracy: 0.9219 - val_loss: 0.5336 - val_accuracy: 0.8764
Epoch 33/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1911 -
accuracy: 0.9226 - val loss: 0.4552 - val accuracy: 0.8806
Epoch 34/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1858 -
accuracy: 0.9271 - val_loss: 0.4206 - val_accuracy: 0.8764
Epoch 35/50
2880/2880 [=========== ] - 5s 2ms/step - loss: 0.1704 -
accuracy: 0.9368 - val_loss: 0.4951 - val_accuracy: 0.8778
Epoch 36/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1877 -
accuracy: 0.9271 - val_loss: 0.4396 - val_accuracy: 0.8778
Epoch 37/50
accuracy: 0.9292 - val_loss: 0.4557 - val_accuracy: 0.8528
Epoch 38/50
```

```
accuracy: 0.9236 - val_loss: 0.3664 - val_accuracy: 0.8764
Epoch 39/50
accuracy: 0.9385 - val_loss: 0.4372 - val_accuracy: 0.8694
Epoch 40/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1718 -
accuracy: 0.9278 - val_loss: 0.4725 - val_accuracy: 0.8694
Epoch 41/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1530 -
accuracy: 0.9441 - val_loss: 0.4034 - val_accuracy: 0.8861
Epoch 42/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1848 -
accuracy: 0.9316 - val_loss: 0.4045 - val_accuracy: 0.8750
Epoch 43/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1508 -
accuracy: 0.9417 - val_loss: 0.4452 - val_accuracy: 0.8694
Epoch 44/50
accuracy: 0.9375 - val_loss: 0.4787 - val_accuracy: 0.8597
Epoch 45/50
accuracy: 0.9392 - val_loss: 0.4517 - val_accuracy: 0.8736
Epoch 46/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1625 -
accuracy: 0.9378 - val_loss: 0.4192 - val_accuracy: 0.8764
Epoch 47/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1596 -
accuracy: 0.9420 - val_loss: 0.4411 - val_accuracy: 0.8764
Epoch 48/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1619 -
accuracy: 0.9333 - val_loss: 0.4205 - val_accuracy: 0.8833
Epoch 49/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1128 -
accuracy: 0.9510 - val loss: 0.5024 - val accuracy: 0.8653
Epoch 50/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1835 -
accuracy: 0.9299 - val_loss: 0.4067 - val_accuracy: 0.8806
1200/1200 [=========== ] - 1s 458us/step
```



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Squential with Kfold CV:

Training for fold 5  $\dots$ 

Adding layer 1: Adding layer 2:

Adding layer 3: Adding layer 4:

Model: "sequential\_20"

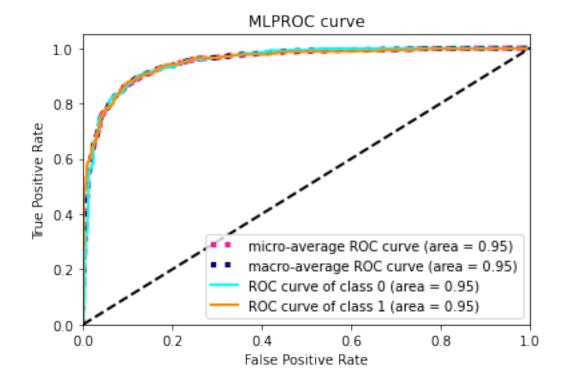
Layer (type)	Output Shape	Param #
dense_105 (Dense)	(None, 4352)	596224
dense_106 (Dense)	(None, 2176)	9472128
dropout_67 (Dropout)	(None, 2176)	0
dense_107 (Dense)	(None, 1088)	2368576
dropout_68 (Dropout)	(None, 1088)	0
dense_108 (Dense)	(None, 544)	592416
dropout_69 (Dropout)	(None, 544)	0

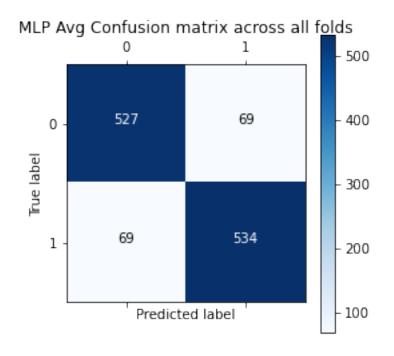
```
(None, 272)
dense_109 (Dense)
                                  148240
dropout_70 (Dropout) (None, 272)
_____
dense_110 (Dense) (None, 2)
                                   546
______
Total params: 13,178,130
Trainable params: 13,178,130
Non-trainable params: 0
Train on 2880 samples, validate on 720 samples
Epoch 1/50
accuracy: 0.5274 - val_loss: 0.6440 - val_accuracy: 0.5889
Epoch 2/50
accuracy: 0.6722 - val_loss: 0.3602 - val_accuracy: 0.8569
Epoch 3/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.4916 -
accuracy: 0.8059 - val_loss: 0.3003 - val_accuracy: 0.8944
Epoch 4/50
accuracy: 0.8816 - val_loss: 0.2838 - val_accuracy: 0.8944
Epoch 5/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.3430 -
accuracy: 0.8781 - val_loss: 0.2938 - val_accuracy: 0.8958
Epoch 6/50
accuracy: 0.8868 - val_loss: 0.2922 - val_accuracy: 0.8847
Epoch 7/50
accuracy: 0.8833 - val_loss: 0.2659 - val_accuracy: 0.8681
Epoch 8/50
accuracy: 0.8938 - val_loss: 0.2761 - val_accuracy: 0.8861
Epoch 9/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.3238 -
accuracy: 0.8701 - val_loss: 0.3141 - val_accuracy: 0.8375
Epoch 10/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2829 -
accuracy: 0.8913 - val_loss: 0.3341 - val_accuracy: 0.8778
accuracy: 0.8896 - val_loss: 0.2849 - val_accuracy: 0.8986
Epoch 12/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2901 -
accuracy: 0.8913 - val_loss: 0.2611 - val_accuracy: 0.8875
```

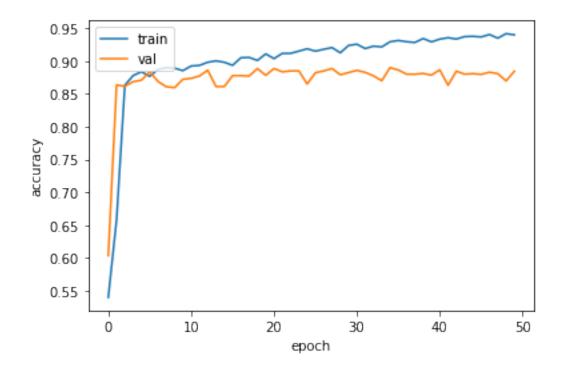
```
Epoch 13/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2565 -
accuracy: 0.9066 - val_loss: 0.2884 - val_accuracy: 0.9014
Epoch 14/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2376 -
accuracy: 0.9094 - val_loss: 0.3454 - val_accuracy: 0.8625
Epoch 15/50
2880/2880 [=========== ] - 5s 2ms/step - loss: 0.3029 -
accuracy: 0.8878 - val_loss: 0.2665 - val_accuracy: 0.8764
Epoch 16/50
accuracy: 0.8875 - val_loss: 0.3015 - val_accuracy: 0.8778
Epoch 17/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2443 -
accuracy: 0.9021 - val_loss: 0.2689 - val_accuracy: 0.9069
Epoch 18/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.2390 -
accuracy: 0.9069 - val_loss: 0.3660 - val_accuracy: 0.8514
Epoch 19/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2347 -
accuracy: 0.9066 - val_loss: 0.2570 - val_accuracy: 0.9014
Epoch 20/50
accuracy: 0.9042 - val_loss: 0.2692 - val_accuracy: 0.9014
Epoch 21/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2373 -
accuracy: 0.9076 - val_loss: 0.2672 - val_accuracy: 0.9111
Epoch 22/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2397 -
accuracy: 0.9056 - val_loss: 0.2952 - val_accuracy: 0.8903
Epoch 23/50
accuracy: 0.9045 - val_loss: 0.2812 - val_accuracy: 0.8972
Epoch 24/50
accuracy: 0.9128 - val_loss: 0.2857 - val_accuracy: 0.8819
Epoch 25/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2227 -
accuracy: 0.9146 - val_loss: 0.4127 - val_accuracy: 0.8556
Epoch 26/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2146 -
accuracy: 0.9177 - val_loss: 0.2874 - val_accuracy: 0.8944
Epoch 27/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.2152 -
accuracy: 0.9191 - val_loss: 0.3216 - val_accuracy: 0.8861
Epoch 28/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1993 -
accuracy: 0.9191 - val_loss: 0.3126 - val_accuracy: 0.8833
```

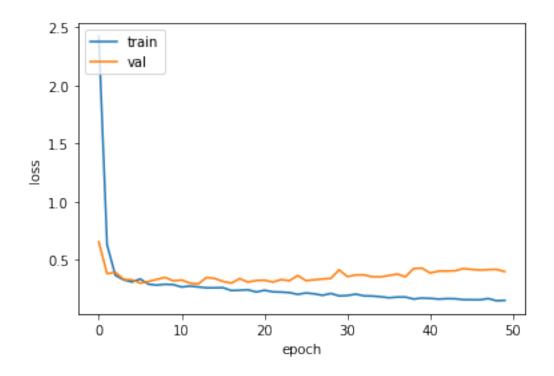
```
Epoch 29/50
accuracy: 0.9059 - val_loss: 0.3495 - val_accuracy: 0.8958
Epoch 30/50
accuracy: 0.9118 - val_loss: 0.3170 - val_accuracy: 0.8986
Epoch 31/50
2880/2880 [=========== ] - 5s 2ms/step - loss: 0.1807 -
accuracy: 0.9243 - val_loss: 0.3150 - val_accuracy: 0.8806
Epoch 32/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.2370 -
accuracy: 0.9101 - val_loss: 0.2622 - val_accuracy: 0.9014
Epoch 33/50
accuracy: 0.9215 - val_loss: 0.2865 - val_accuracy: 0.8861
Epoch 34/50
2880/2880 [============ ] - 5s 2ms/step - loss: 0.1751 -
accuracy: 0.9288 - val_loss: 0.2881 - val_accuracy: 0.8958
Epoch 35/50
accuracy: 0.9149 - val_loss: 0.3228 - val_accuracy: 0.8931
Epoch 36/50
accuracy: 0.9274 - val_loss: 0.3224 - val_accuracy: 0.8944
Epoch 37/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1819 -
accuracy: 0.9309 - val_loss: 0.3140 - val_accuracy: 0.8819
Epoch 38/50
accuracy: 0.9240 - val_loss: 0.3301 - val_accuracy: 0.8861
Epoch 39/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1659 -
accuracy: 0.9309 - val_loss: 0.3174 - val_accuracy: 0.8931
Epoch 40/50
accuracy: 0.9274 - val_loss: 0.3214 - val_accuracy: 0.8833
Epoch 41/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1801 -
accuracy: 0.9271 - val_loss: 0.3351 - val_accuracy: 0.8931
Epoch 42/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1550 -
accuracy: 0.9378 - val_loss: 0.3658 - val_accuracy: 0.8833
Epoch 43/50
2880/2880 [============== ] - 5s 2ms/step - loss: 0.1909 -
accuracy: 0.9240 - val_loss: 0.2885 - val_accuracy: 0.8931
Epoch 44/50
2880/2880 [============= ] - 5s 2ms/step - loss: 0.1657 -
accuracy: 0.9347 - val_loss: 0.4194 - val_accuracy: 0.8931
```

Epoch 45/50 2880/2880 [============= ] - 5s 2ms/step - loss: 0.1953 accuracy: 0.9233 - val\_loss: 0.3339 - val\_accuracy: 0.9000 2880/2880 [============ ] - 5s 2ms/step - loss: 0.1529 accuracy: 0.9312 - val\_loss: 0.3964 - val\_accuracy: 0.8847 accuracy: 0.9347 - val\_loss: 0.4108 - val\_accuracy: 0.8819 Epoch 48/50 2880/2880 [============ ] - 5s 2ms/step - loss: 0.1650 accuracy: 0.9319 - val\_loss: 0.4480 - val\_accuracy: 0.8764 Epoch 49/50 2880/2880 [============= ] - 5s 2ms/step - loss: 0.1649 accuracy: 0.9312 - val\_loss: 0.4241 - val\_accuracy: 0.8833 Epoch 50/50 2880/2880 [============ ] - 5s 2ms/step - loss: 0.1462 accuracy: 0.9368 - val\_loss: 0.4159 - val\_accuracy: 0.8778 1200/1200 [=========== ] - 1s 476us/step









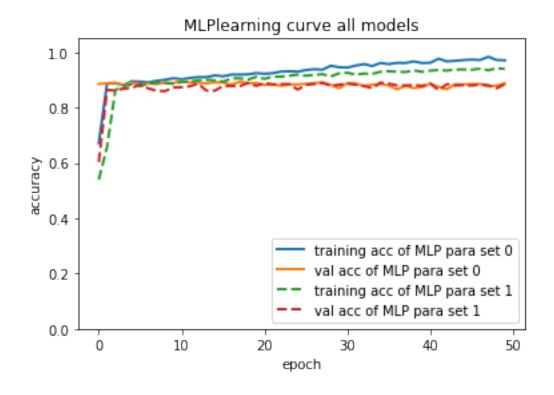
Average scores for pesudo test set across all folds:

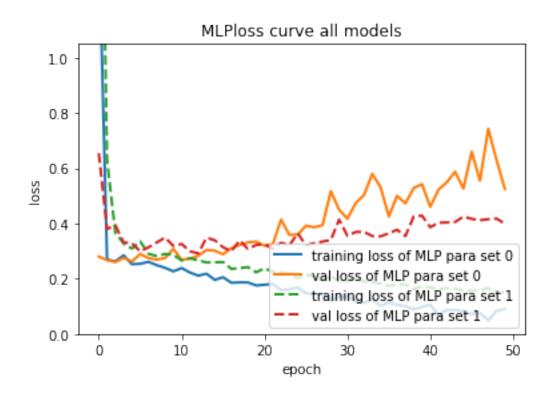
> Accuracy: 0.8846666574478149 (+- 0.006227174625387633)

> Loss: 0.40233149101336796

> Avg runtime per test instance: 0.003231040994326274

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```
Opt model parameter found on the pseudo test set:
    {'num_hidden_layer': 4, 'hidden_layer_activation': ['relu', 'tanh', 'relu',
    'tanh'], 'dropout': [0.5, 0.25, 0.125, 0.0625], 'last_activation': 'softmax'}
    Best average pseudo test set accuracy score with the opt model:
    0.8846666574478149
    Average runtime per test instance: 0.003231040994326274
[8]: from B1 import B1_functions_all as B1_functions
[9]: TaskB1_opt_models_dict, TaskB1_res_dict = B1_functions.get_B1_results()
              ------Reading Task B1 dataset------Reading Task B1 dataset-----
    -----Task B1: Gird searching on
    SVM Grid search CV on Dataset A:
    Training scores:
    0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
    0.592 (+/-0.012) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
    0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
    0.592 (+/-0.012) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
    0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
    0.592 (+/-0.012) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
    0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
    0.592 (+/-0.012) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
    0.216 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
    0.738 (+/-0.007) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
    0.216 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
    0.738 (+/-0.007) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
    0.267 (+/-0.016) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
    0.738 (+/-0.007) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
    0.216 (+/-0.000) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
    0.738 (+/-0.007) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}
    0.215 (+/-0.001) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
    0.770 (+/-0.008) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
    0.493 (+/-0.009) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
    0.770 (+/-0.008) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
    0.605 (+/-0.008) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
    0.770 (+/-0.008) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
    0.216 (+/-0.000) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
```

```
0.770 (+/-0.008) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}
0.494 (+/-0.015) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.775 (+/-0.010) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
0.675 (+/-0.009) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.775 (+/-0.010) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
0.859 (+/-0.005) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.775 (+/-0.010) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
1.000 (+/-0.000) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.775 (+/-0.010) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.661 (+/-0.005) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.777 (+/-0.007) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.815 (+/-0.007) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.777 (+/-0.007) for {'C': 10, 'gamma': 0.001, 'kernel': 'linear'}
0.985 (+/-0.003) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.777 (+/-0.007) for {'C': 10, 'gamma': 0.01, 'kernel': 'linear'}
1.000 (+/-0.000) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.777 (+/-0.007) for {'C': 10, 'gamma': 0.1, 'kernel': 'linear'}
Validation scores:
0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
0.583 (+/-0.035) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
0.583 (+/-0.035) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
0.583 (+/-0.035) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
0.216 (+/-0.000) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
0.583 (+/-0.035) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
0.216 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
0.724 (+/-0.029) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
0.216 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
0.724 (+/-0.029) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
0.265 (+/-0.009) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
0.724 (+/-0.029) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
0.216 (+/-0.000) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
0.724 (+/-0.029) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}
0.216 (+/-0.002) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.741 (+/-0.027) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
0.485 (+/-0.007) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
0.741 (+/-0.027) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
0.554 (+/-0.034) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
0.741 (+/-0.027) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
0.216 (+/-0.000) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
0.741 (+/-0.027) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}
0.487 (+/-0.018) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.721 (+/-0.022) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
0.651 (+/-0.038) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.721 (+/-0.022) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
```

```
0.714 (+/-0.045) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.721 (+/-0.022) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
0.520 (+/-0.042) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.721 (+/-0.022) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.647 (+/-0.027) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.707 (+/-0.019) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.760 (+/-0.044) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.707 (+/-0.019) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.707 (+/-0.040) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.707 (+/-0.019) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.535 (+/-0.043) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.707 (+/-0.019) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
```

Prediction on a pseudo test set (split from Dataset A):

	precision	recall	f1-score	support
0	0.81	0.77	0.79	389
1	0.79	0.72	0.75	400
2	0.70	0.81	0.75	392
3	0.76	0.72	0.74	390
4	0.81	0.83	0.82	383
accuracy			0.77	1954
macro avg	0.77	0.77	0.77	1954
weighted avg	0.77	0.77	0.77	1954

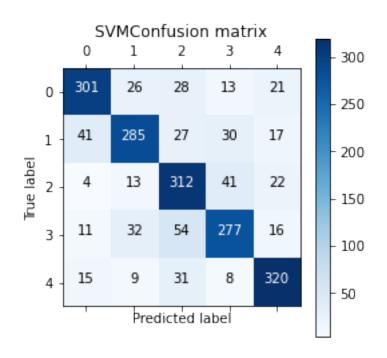
Accuracy: 0.77021494370522

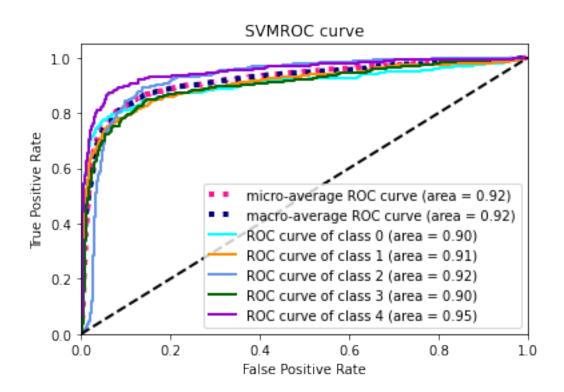
Best parameters found on Dataset A:

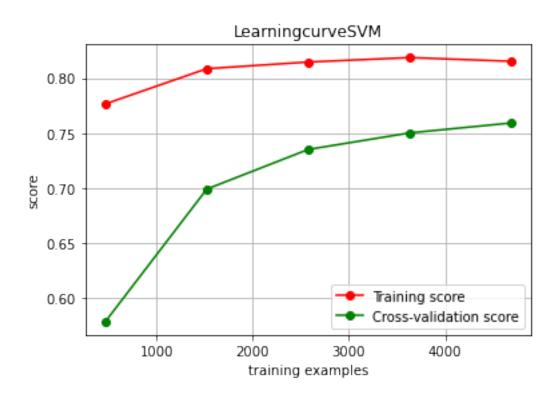
{'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}

Average runtime per test instance: 0.0013042773780705618

-----







-----Task B1: Gird searching on Rando

\_\_\_\_\_

RF Grid search CV on Dataset A:

## Training scores:

```
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 64, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 128, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 256}
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 512}
```

```
1.000 (+/-0.000) for {'max_depth': 256, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 64}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 128}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 256}
1.000 (+/-0.000) for {'max depth': 512, 'n estimators': 512}
1.000 (+/-0.000) for {'max_depth': 512, 'n_estimators': 1024}
1.000 (+/-0.000) for {'max depth': 1024, 'n estimators': 64}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 128}
1.000 (+/-0.000) for {'max depth': 1024, 'n estimators': 256}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 512}
1.000 (+/-0.000) for {'max_depth': 1024, 'n_estimators': 1024}
Validation scores:
0.703 (+/-0.018) for {'max_depth': 64, 'n_estimators': 64}
0.714 (+/-0.021) for {'max_depth': 64, 'n_estimators': 128}
0.715 (+/-0.020) for {'max_depth': 64, 'n_estimators': 256}
0.718 (+/-0.021) for {'max_depth': 64, 'n_estimators': 512}
0.718 (+/-0.029) for {'max_depth': 64, 'n_estimators': 1024}
0.697 (+/-0.034) for {'max depth': 128, 'n estimators': 64}
0.713 (+/-0.020) for {'max_depth': 128, 'n_estimators': 128}
0.716 (+/-0.020) for {'max_depth': 128, 'n_estimators': 256}
0.717 (+/-0.031) for {'max_depth': 128, 'n_estimators': 512}
0.721 (+/-0.025) for {'max_depth': 128, 'n_estimators': 1024}
0.705 (+/-0.021) for {'max_depth': 256, 'n_estimators': 64}
0.712 (+/-0.018) for {'max_depth': 256, 'n_estimators': 128}
0.717 (+/-0.023) for {'max_depth': 256, 'n_estimators': 256}
0.720 (+/-0.022) for {'max_depth': 256, 'n_estimators': 512}
0.720 (+/-0.024) for {'max_depth': 256, 'n_estimators': 1024}
0.709 (+/-0.012) for {'max_depth': 512, 'n_estimators': 64}
0.707 (+/-0.028) for {'max_depth': 512, 'n_estimators': 128}
0.720 (+/-0.018) for {'max_depth': 512, 'n_estimators': 256}
0.717 (+/-0.017) for {'max_depth': 512, 'n_estimators': 512}
0.718 \ (+/-0.024)  for {\max_{depth'}: 512, 'n_{estimators'}: 1024}
0.703 \ (+/-0.022)  for {'max depth': 1024, 'n estimators': 64}
0.712 (+/-0.021) for {'max_depth': 1024, 'n_estimators': 128}
0.717 \ (+/-0.023)  for {'max depth': 1024, 'n estimators': 256}
0.718 (+/-0.021) for {'max_depth': 1024, 'n_estimators': 512}
0.718 \ (+/-0.020) \ for \ {\rm 'max\_depth': 1024, 'n\_estimators': 1024}
Prediction on a pseudo test set (split from Dataset A):
              precision
                           recall f1-score
                                              support
```

-				
0	0.80	0.66	0.72	389
1	0.71	0.68	0.70	400
2	0.69	0.82	0.75	392
3	0.71	0.70	0.71	390
4	0.79	0.84	0.82	383

accuracy			0.74	1954
macro avg	0.74	0.74	0.74	1954
weighted avg	0.74	0.74	0.74	1954

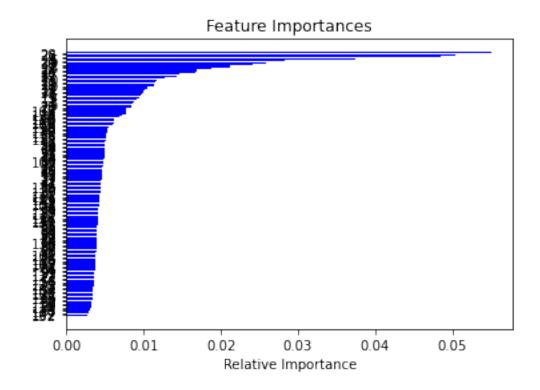
Accuracy: 0.7389969293756398

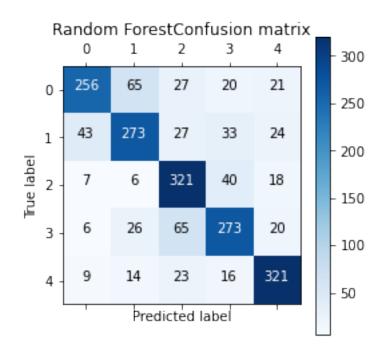
Best parameters found on Dataset A:

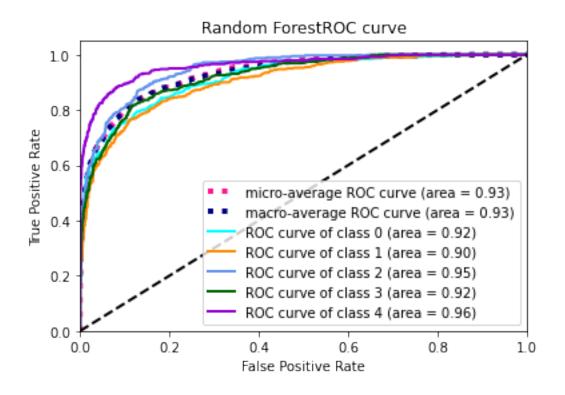
{'max\_depth': 128, 'n\_estimators': 1024}

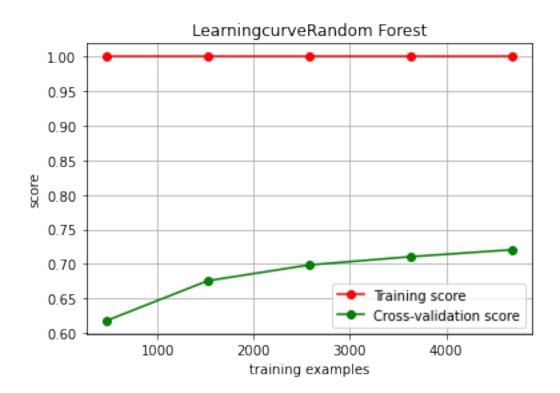
Average runtime per test instance: 0.00046991606451715883

-----









```
-----Task B1: Gird searching on KNN------
KNN Grid search CV on Dataset A:
Training scores:
```

```
0.632 (+/-0.018) for {'n_neighbors': 8}

0.600 (+/-0.008) for {'n_neighbors': 16}

0.574 (+/-0.006) for {'n_neighbors': 32}

0.545 (+/-0.006) for {'n_neighbors': 64}

0.527 (+/-0.011) for {'n_neighbors': 128}
```

## Validation scores:

```
0.484 (+/-0.026) for {'n_neighbors': 8}
0.502 (+/-0.027) for {'n_neighbors': 16}
0.510 (+/-0.026) for {'n_neighbors': 32}
0.509 (+/-0.023) for {'n_neighbors': 64}
0.507 (+/-0.021) for {'n_neighbors': 128}
```

Prediction on a pseudo test set (split from Dataset A):

precision recall f1-score support

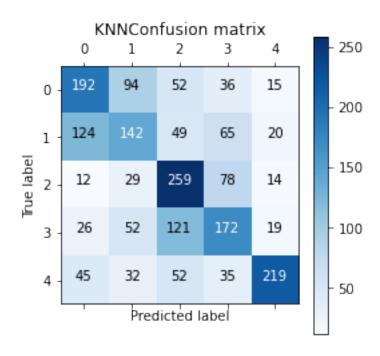
	0	0.48	0.49	0.49	389
	1	0.41	0.35	0.38	400
	2	0.49	0.66	0.56	392
	3	0.45	0.44	0.44	390
	4	0.76	0.57	0.65	383
accurac	у			0.50	1954
macro av	rg	0.52	0.50	0.50	1954
weighted av	rg	0.52	0.50	0.50	1954

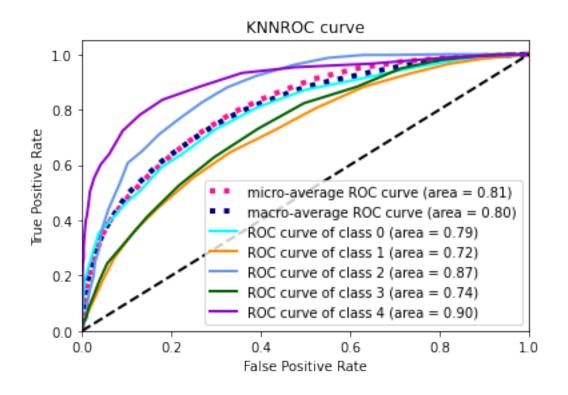
Accuracy: 0.503582395087001

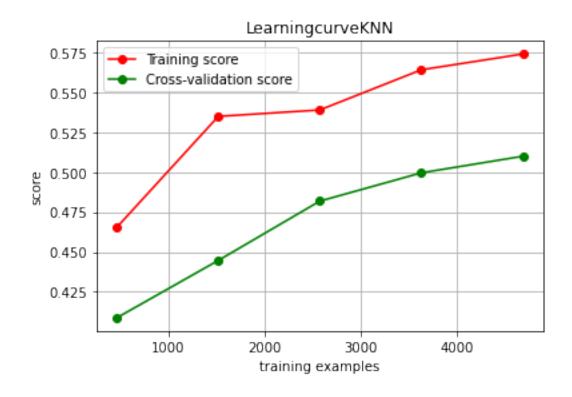
Best parameters found on Dataset A:

{'n\_neighbors': 32}

Average runtime per test instance: 0.003438198700653982





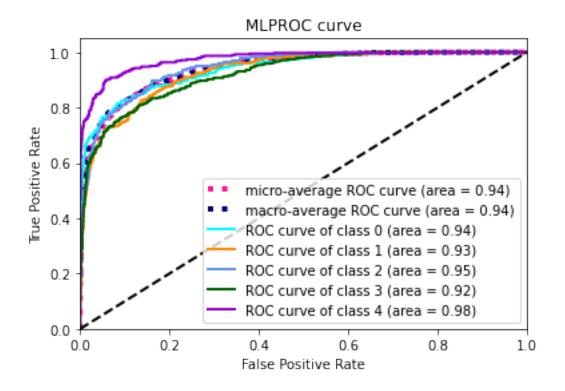


```
-----Task B1: Gird searching on
MT.P-----
Squential with Kfold CV:
Training for fold 1 ...
Adding layer 1:
Adding layer 2:
Adding layer 3:
Model: "sequential_21"
_____
Layer (type)
             Output Shape
______
dense_111 (Dense)
             (None, 2176)
                             298112
_____
dense_112 (Dense) (None, 1088)
                        2368576
dropout_71 (Dropout) (None, 1088)
_____
dense_113 (Dense) (None, 544)
                             592416
dropout_72 (Dropout) (None, 544)
dense_114 (Dense) (None, 272)
                             148240
   -----
dropout_73 (Dropout) (None, 272)
_____
dense_115 (Dense) (None, 5)
                             1365
______
Total params: 3,408,709
Trainable params: 3,408,709
Non-trainable params: 0
-----
Train on 4688 samples, validate on 1173 samples
Epoch 1/50
accuracy: 0.2513 - val_loss: 1.4331 - val_accuracy: 0.4697
Epoch 2/50
accuracy: 0.4985 - val_loss: 1.1148 - val_accuracy: 0.5772
Epoch 3/50
4688/4688 [============= ] - 3s 566us/step - loss: 1.1889 -
accuracy: 0.5105 - val_loss: 1.0000 - val_accuracy: 0.6309
Epoch 4/50
4688/4688 [============= ] - 3s 564us/step - loss: 1.0193 -
```

```
accuracy: 0.5870 - val_loss: 0.9899 - val_accuracy: 0.5678
Epoch 5/50
4688/4688 [============= ] - 3s 557us/step - loss: 0.9510 -
accuracy: 0.6201 - val_loss: 0.9177 - val_accuracy: 0.6377
Epoch 6/50
4688/4688 [=============== ] - 3s 599us/step - loss: 0.9720 -
accuracy: 0.6137 - val_loss: 1.0129 - val_accuracy: 0.5652
Epoch 7/50
4688/4688 [============= ] - 3s 604us/step - loss: 0.8172 -
accuracy: 0.6847 - val_loss: 0.9100 - val_accuracy: 0.6198
Epoch 8/50
4688/4688 [=============== ] - 3s 592us/step - loss: 0.8484 -
accuracy: 0.6685 - val_loss: 0.7664 - val_accuracy: 0.6786
Epoch 9/50
4688/4688 [=============== ] - 3s 615us/step - loss: 0.8175 -
accuracy: 0.6751 - val_loss: 1.0573 - val_accuracy: 0.5797
Epoch 10/50
4688/4688 [============= ] - 3s 628us/step - loss: 0.7459 -
accuracy: 0.7152 - val_loss: 1.0214 - val_accuracy: 0.5729
Epoch 11/50
4688/4688 [============== ] - 3s 605us/step - loss: 0.7657 -
accuracy: 0.7039 - val_loss: 0.7260 - val_accuracy: 0.7212
Epoch 12/50
4688/4688 [============== ] - 3s 605us/step - loss: 0.7245 -
accuracy: 0.7257 - val_loss: 0.6921 - val_accuracy: 0.7357
Epoch 13/50
4688/4688 [============== ] - 3s 620us/step - loss: 0.6477 -
accuracy: 0.7494 - val_loss: 0.7476 - val_accuracy: 0.7059
4688/4688 [============== ] - 3s 598us/step - loss: 0.7013 -
accuracy: 0.7302 - val_loss: 0.7887 - val_accuracy: 0.6905
Epoch 15/50
4688/4688 [============== ] - 3s 612us/step - loss: 0.6537 -
accuracy: 0.7491 - val_loss: 0.6826 - val_accuracy: 0.7340
Epoch 16/50
4688/4688 [=============== ] - 3s 606us/step - loss: 0.6151 -
accuracy: 0.7543 - val loss: 0.8139 - val accuracy: 0.6394
Epoch 17/50
accuracy: 0.7517 - val_loss: 0.6643 - val_accuracy: 0.7485
Epoch 18/50
4688/4688 [============= ] - 3s 596us/step - loss: 0.5593 -
accuracy: 0.7826 - val_loss: 0.6281 - val_accuracy: 0.7647
Epoch 19/50
4688/4688 [============] - 3s 592us/step - loss: 0.6210 -
accuracy: 0.7579 - val_loss: 0.6253 - val_accuracy: 0.7613
Epoch 20/50
4688/4688 [============== ] - 3s 599us/step - loss: 0.5588 -
```

```
accuracy: 0.7818 - val_loss: 0.6548 - val_accuracy: 0.7425
Epoch 21/50
4688/4688 [============= ] - 3s 558us/step - loss: 0.4977 -
accuracy: 0.8010 - val_loss: 1.0502 - val_accuracy: 0.6317
Epoch 22/50
accuracy: 0.7726 - val_loss: 0.8098 - val_accuracy: 0.6760
Epoch 23/50
4688/4688 [============== ] - 3s 562us/step - loss: 0.5076 -
accuracy: 0.8038 - val_loss: 0.6438 - val_accuracy: 0.7374
Epoch 24/50
4688/4688 [============== ] - 3s 579us/step - loss: 0.5060 -
accuracy: 0.8048 - val_loss: 0.6879 - val_accuracy: 0.7460
Epoch 25/50
accuracy: 0.8119 - val_loss: 0.6873 - val_accuracy: 0.7408
Epoch 26/50
4688/4688 [============== ] - 3s 561us/step - loss: 0.5273 -
accuracy: 0.7965 - val_loss: 0.7042 - val_accuracy: 0.7306
Epoch 27/50
4688/4688 [============== ] - 3s 573us/step - loss: 0.4121 -
accuracy: 0.8362 - val_loss: 0.6950 - val_accuracy: 0.7434
Epoch 28/50
4688/4688 [============= ] - 3s 583us/step - loss: 0.4252 -
accuracy: 0.8311 - val_loss: 0.7367 - val_accuracy: 0.7562
Epoch 29/50
4688/4688 [============== ] - 3s 555us/step - loss: 0.5100 -
accuracy: 0.8070 - val_loss: 0.7053 - val_accuracy: 0.7280
accuracy: 0.8242 - val_loss: 0.7412 - val_accuracy: 0.7545
Epoch 31/50
accuracy: 0.8471 - val_loss: 0.6753 - val_accuracy: 0.7553
Epoch 32/50
4688/4688 [============== ] - 3s 570us/step - loss: 0.3937 -
accuracy: 0.8475 - val loss: 0.7678 - val accuracy: 0.6957
Epoch 33/50
accuracy: 0.8319 - val_loss: 0.6486 - val_accuracy: 0.7613
Epoch 34/50
4688/4688 [============= ] - 3s 569us/step - loss: 0.4211 -
accuracy: 0.8364 - val_loss: 0.9098 - val_accuracy: 0.6360
Epoch 35/50
4688/4688 [============= ] - 3s 572us/step - loss: 0.3587 -
accuracy: 0.8637 - val_loss: 0.6792 - val_accuracy: 0.7664
Epoch 36/50
4688/4688 [============== ] - 3s 566us/step - loss: 0.4070 -
```

```
accuracy: 0.8426 - val_loss: 0.6251 - val_accuracy: 0.7698
Epoch 37/50
4688/4688 [============= ] - 3s 571us/step - loss: 0.3393 -
accuracy: 0.8720 - val_loss: 0.7765 - val_accuracy: 0.7357
Epoch 38/50
4688/4688 [=============== ] - 3s 570us/step - loss: 0.3385 -
accuracy: 0.8724 - val_loss: 0.7424 - val_accuracy: 0.7434
Epoch 39/50
4688/4688 [============= ] - 3s 581us/step - loss: 0.3507 -
accuracy: 0.8697 - val_loss: 0.6779 - val_accuracy: 0.7886
Epoch 40/50
4688/4688 [============== ] - 3s 566us/step - loss: 0.3339 -
accuracy: 0.8697 - val_loss: 0.8294 - val_accuracy: 0.7477
Epoch 41/50
accuracy: 0.8793 - val_loss: 0.7567 - val_accuracy: 0.7451
Epoch 42/50
4688/4688 [============= ] - 3s 563us/step - loss: 0.3601 -
accuracy: 0.8635 - val_loss: 0.7133 - val_accuracy: 0.7766
Epoch 43/50
4688/4688 [=============== ] - 3s 567us/step - loss: 0.3821 -
accuracy: 0.8652 - val_loss: 0.7990 - val_accuracy: 0.7400
Epoch 44/50
4688/4688 [============= ] - 3s 571us/step - loss: 0.2712 -
accuracy: 0.8983 - val_loss: 0.8001 - val_accuracy: 0.7741
Epoch 45/50
4688/4688 [============== ] - 3s 569us/step - loss: 0.2338 -
accuracy: 0.9132 - val_loss: 0.8517 - val_accuracy: 0.7400
accuracy: 0.8594 - val_loss: 0.8513 - val_accuracy: 0.7084
Epoch 47/50
4688/4688 [============== ] - 3s 560us/step - loss: 0.2971 -
accuracy: 0.8887 - val_loss: 0.9158 - val_accuracy: 0.7477
Epoch 48/50
accuracy: 0.9168 - val loss: 0.9476 - val accuracy: 0.7519
Epoch 49/50
accuracy: 0.8842 - val_loss: 0.8523 - val_accuracy: 0.7570
Epoch 50/50
4688/4688 [============= ] - 3s 555us/step - loss: 0.2734 -
accuracy: 0.8997 - val_loss: 0.9372 - val_accuracy: 0.7408
1954/1954 [============= ] - 1s 260us/step
```



\_\_\_\_\_\_

Squential with Kfold CV:

Training for fold 2  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

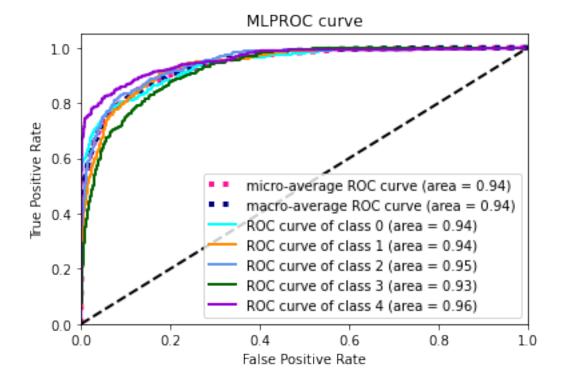
Model: "sequential\_22"

Layer (type)	Output Shape	 Param #
dense_116 (Dense)	(None, 2176)	298112
dense_117 (Dense)	(None, 1088)	2368576
dropout_74 (Dropout)	(None, 1088)	0
dense_118 (Dense)	(None, 544)	592416
dropout_75 (Dropout)	(None, 544)	0
dense_119 (Dense)	(None, 272)	148240
dropout_76 (Dropout)	(None, 272)	0

```
dense_120 (Dense)
                    (None, 5)
                                       1365
______
Total params: 3,408,709
Trainable params: 3,408,709
Non-trainable params: 0
Train on 4689 samples, validate on 1172 samples
Epoch 1/50
4689/4689 [============= ] - 5s 1ms/step - loss: 2.4753 -
accuracy: 0.2696 - val_loss: 2.0849 - val_accuracy: 0.3183
Epoch 2/50
accuracy: 0.3962 - val_loss: 1.2031 - val_accuracy: 0.4608
Epoch 3/50
accuracy: 0.4961 - val_loss: 1.1796 - val_accuracy: 0.5137
Epoch 4/50
4689/4689 [============= ] - 3s 586us/step - loss: 1.1054 -
accuracy: 0.5406 - val_loss: 1.2378 - val_accuracy: 0.5171
Epoch 5/50
accuracy: 0.5835 - val_loss: 1.0320 - val_accuracy: 0.6109
Epoch 6/50
accuracy: 0.6070 - val_loss: 0.9129 - val_accuracy: 0.6621
Epoch 7/50
4689/4689 [============== ] - 3s 596us/step - loss: 0.9064 -
accuracy: 0.6443 - val_loss: 0.9750 - val_accuracy: 0.6271
accuracy: 0.6583 - val_loss: 0.9410 - val_accuracy: 0.6254
4689/4689 [============== ] - 3s 608us/step - loss: 0.8829 -
accuracy: 0.6502 - val_loss: 0.8536 - val_accuracy: 0.6766
Epoch 10/50
4689/4689 [============== ] - 3s 611us/step - loss: 0.8479 -
accuracy: 0.6643 - val loss: 0.7850 - val accuracy: 0.6954
Epoch 11/50
accuracy: 0.7121 - val_loss: 1.3562 - val_accuracy: 0.5819
Epoch 12/50
4689/4689 [============= ] - 3s 586us/step - loss: 0.8446 -
accuracy: 0.6829 - val_loss: 0.9789 - val_accuracy: 0.6032
Epoch 13/50
4689/4689 [============= ] - 3s 580us/step - loss: 0.7213 -
accuracy: 0.7206 - val_loss: 0.8799 - val_accuracy: 0.6792
Epoch 14/50
4689/4689 [============= ] - 3s 590us/step - loss: 0.7315 -
```

```
accuracy: 0.7210 - val_loss: 0.9911 - val_accuracy: 0.6655
Epoch 15/50
4689/4689 [============= ] - 3s 565us/step - loss: 0.7179 -
accuracy: 0.7304 - val_loss: 1.4686 - val_accuracy: 0.5375
Epoch 16/50
accuracy: 0.7129 - val_loss: 0.6841 - val_accuracy: 0.7509
Epoch 17/50
4689/4689 [============== ] - 3s 606us/step - loss: 0.6910 -
accuracy: 0.7400 - val_loss: 0.7560 - val_accuracy: 0.7261
Epoch 18/50
accuracy: 0.7428 - val_loss: 0.9460 - val_accuracy: 0.6817
Epoch 19/50
accuracy: 0.7628 - val_loss: 0.9312 - val_accuracy: 0.6664
Epoch 20/50
4689/4689 [============= ] - 3s 631us/step - loss: 0.7028 -
accuracy: 0.7324 - val_loss: 0.9294 - val_accuracy: 0.6928
Epoch 21/50
4689/4689 [============== ] - 3s 617us/step - loss: 0.6360 -
accuracy: 0.7562 - val_loss: 0.6603 - val_accuracy: 0.7483
Epoch 22/50
4689/4689 [============= ] - 3s 577us/step - loss: 0.5722 -
accuracy: 0.7769 - val_loss: 0.8087 - val_accuracy: 0.6911
Epoch 23/50
4689/4689 [============= ] - 3s 630us/step - loss: 0.5764 -
accuracy: 0.7784 - val_loss: 0.8947 - val_accuracy: 0.6817
accuracy: 0.7722 - val_loss: 0.8724 - val_accuracy: 0.7005
Epoch 25/50
accuracy: 0.7889 - val_loss: 0.8563 - val_accuracy: 0.6604
Epoch 26/50
accuracy: 0.7797 - val_loss: 1.1201 - val_accuracy: 0.6980
Epoch 27/50
accuracy: 0.8017 - val_loss: 0.7057 - val_accuracy: 0.7398
Epoch 28/50
4689/4689 [============== ] - 3s 564us/step - loss: 0.5237 -
accuracy: 0.7923 - val_loss: 0.7805 - val_accuracy: 0.7227
Epoch 29/50
4689/4689 [============ ] - 3s 573us/step - loss: 0.5516 -
accuracy: 0.7897 - val_loss: 1.2664 - val_accuracy: 0.6502
Epoch 30/50
```

```
accuracy: 0.7882 - val_loss: 0.7211 - val_accuracy: 0.7338
Epoch 31/50
4689/4689 [============= ] - 3s 561us/step - loss: 0.4725 -
accuracy: 0.8202 - val_loss: 0.7152 - val_accuracy: 0.7278
Epoch 32/50
accuracy: 0.8174 - val_loss: 1.1517 - val_accuracy: 0.6297
Epoch 33/50
4689/4689 [============= ] - 3s 568us/step - loss: 0.5244 -
accuracy: 0.8091 - val_loss: 0.6905 - val_accuracy: 0.7611
Epoch 34/50
4689/4689 [============== ] - 3s 559us/step - loss: 0.4467 -
accuracy: 0.8258 - val_loss: 1.0056 - val_accuracy: 0.6860
Epoch 35/50
accuracy: 0.8164 - val_loss: 0.8583 - val_accuracy: 0.7295
Epoch 36/50
4689/4689 [============= ] - 3s 569us/step - loss: 0.4136 -
accuracy: 0.8375 - val_loss: 0.9195 - val_accuracy: 0.7150
Epoch 37/50
accuracy: 0.8238 - val_loss: 0.7403 - val_accuracy: 0.7363
Epoch 38/50
4689/4689 [============== ] - 3s 550us/step - loss: 0.3889 -
accuracy: 0.8501 - val_loss: 0.6969 - val_accuracy: 0.7577
Epoch 39/50
4689/4689 [============== ] - 3s 566us/step - loss: 0.4256 -
accuracy: 0.8292 - val_loss: 0.7697 - val_accuracy: 0.7517
accuracy: 0.8524 - val_loss: 1.0102 - val_accuracy: 0.7159
Epoch 41/50
accuracy: 0.8420 - val_loss: 0.7284 - val_accuracy: 0.7662
Epoch 42/50
accuracy: 0.8392 - val loss: 0.7649 - val accuracy: 0.7491
Epoch 43/50
accuracy: 0.8656 - val_loss: 0.9438 - val_accuracy: 0.7449
Epoch 44/50
4689/4689 [============= ] - 3s 576us/step - loss: 0.4331 -
accuracy: 0.8373 - val_loss: 0.8289 - val_accuracy: 0.7338
Epoch 45/50
4689/4689 [============= ] - 3s 585us/step - loss: 0.4307 -
accuracy: 0.8270 - val_loss: 0.8323 - val_accuracy: 0.7398
Epoch 46/50
4689/4689 [============== ] - 3s 572us/step - loss: 0.4017 -
```



```
Squential with Kfold CV:

Training for fold 3 ...

Adding layer 1:

Adding layer 2:

Adding layer 3:

Model: "sequential_23"

Layer (type)

Output Shape

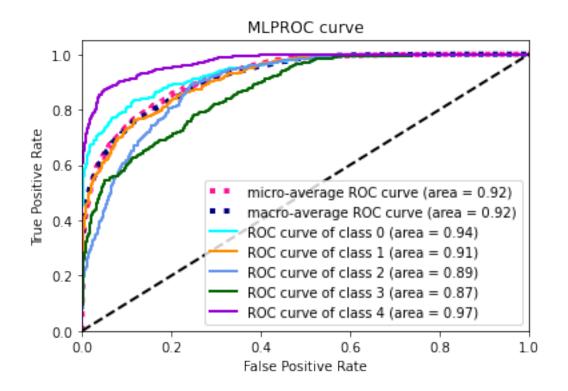
Param #
```

```
______
dense_121 (Dense)
                (None, 2176)
                                 298112
            (None, 1088)
dense_122 (Dense)
                                2368576
       .....
dropout_77 (Dropout) (None, 1088)
-----
dense_123 (Dense) (None, 544)
                           592416
dropout_78 (Dropout) (None, 544)
    _____
dense_124 (Dense) (None, 272)
                                148240
-----
dropout_79 (Dropout) (None, 272)
______
dense_125 (Dense) (None, 5)
                                1365
______
Total params: 3,408,709
Trainable params: 3,408,709
Non-trainable params: 0
Train on 4689 samples, validate on 1172 samples
Epoch 1/50
4689/4689 [============== ] - 5s 1ms/step - loss: 2.6471 -
accuracy: 0.2608 - val_loss: 1.5418 - val_accuracy: 0.2449
Epoch 2/50
4689/4689 [============= ] - 3s 561us/step - loss: 1.3331 -
accuracy: 0.4348 - val_loss: 1.2171 - val_accuracy: 0.5034
accuracy: 0.4873 - val_loss: 1.0170 - val_accuracy: 0.6075
4689/4689 [============== ] - 3s 563us/step - loss: 1.0292 -
accuracy: 0.5747 - val_loss: 1.2940 - val_accuracy: 0.4906
Epoch 5/50
accuracy: 0.5696 - val loss: 0.9984 - val accuracy: 0.5922
Epoch 6/50
accuracy: 0.6355 - val_loss: 1.1330 - val_accuracy: 0.5802
Epoch 7/50
accuracy: 0.6257 - val_loss: 1.3763 - val_accuracy: 0.5794
Epoch 8/50
4689/4689 [=============== ] - 3s 563us/step - loss: 0.8991 -
accuracy: 0.6517 - val_loss: 0.9152 - val_accuracy: 0.6331
Epoch 9/50
4689/4689 [============= ] - 3s 566us/step - loss: 0.8197 -
```

```
accuracy: 0.6818 - val_loss: 1.0712 - val_accuracy: 0.6101
Epoch 10/50
4689/4689 [============= ] - 3s 564us/step - loss: 0.8790 -
accuracy: 0.6729 - val_loss: 0.8213 - val_accuracy: 0.6758
Epoch 11/50
accuracy: 0.6995 - val_loss: 1.0944 - val_accuracy: 0.6254
Epoch 12/50
4689/4689 [============== ] - 3s 567us/step - loss: 0.7915 -
accuracy: 0.6989 - val_loss: 0.8635 - val_accuracy: 0.6869
Epoch 13/50
4689/4689 [=============== ] - 3s 556us/step - loss: 0.7379 -
accuracy: 0.7115 - val_loss: 0.9734 - val_accuracy: 0.5964
Epoch 14/50
accuracy: 0.7388 - val_loss: 1.2883 - val_accuracy: 0.6519
Epoch 15/50
4689/4689 [============= ] - 3s 573us/step - loss: 0.7978 -
accuracy: 0.7080 - val_loss: 0.8306 - val_accuracy: 0.6809
Epoch 16/50
4689/4689 [============== ] - 3s 564us/step - loss: 0.6708 -
accuracy: 0.7364 - val_loss: 1.0543 - val_accuracy: 0.7022
Epoch 17/50
accuracy: 0.7302 - val_loss: 0.9098 - val_accuracy: 0.6536
Epoch 18/50
4689/4689 [============= ] - 3s 561us/step - loss: 0.6452 -
accuracy: 0.7496 - val_loss: 1.0437 - val_accuracy: 0.6459
Epoch 19/50
accuracy: 0.7554 - val_loss: 0.8899 - val_accuracy: 0.7039
Epoch 20/50
accuracy: 0.7396 - val_loss: 0.8768 - val_accuracy: 0.6664
Epoch 21/50
accuracy: 0.7569 - val loss: 0.6599 - val accuracy: 0.7662
Epoch 22/50
accuracy: 0.7910 - val_loss: 1.0192 - val_accuracy: 0.6365
Epoch 23/50
4689/4689 [============= ] - 3s 562us/step - loss: 0.6466 -
accuracy: 0.7462 - val_loss: 0.7320 - val_accuracy: 0.7321
Epoch 24/50
4689/4689 [============ ] - 3s 560us/step - loss: 0.5630 -
accuracy: 0.7746 - val_loss: 0.7698 - val_accuracy: 0.7688
Epoch 25/50
4689/4689 [============== ] - 3s 569us/step - loss: 0.6277 -
```

```
accuracy: 0.7688 - val_loss: 0.8328 - val_accuracy: 0.7304
Epoch 26/50
4689/4689 [============= ] - 3s 560us/step - loss: 0.5471 -
accuracy: 0.7921 - val_loss: 0.9662 - val_accuracy: 0.5896
Epoch 27/50
accuracy: 0.7865 - val_loss: 0.7927 - val_accuracy: 0.7235
Epoch 28/50
4689/4689 [============== ] - 3s 563us/step - loss: 0.5872 -
accuracy: 0.7686 - val_loss: 0.8375 - val_accuracy: 0.7253
Epoch 29/50
accuracy: 0.8036 - val_loss: 0.7648 - val_accuracy: 0.7150
Epoch 30/50
4689/4689 [============== ] - 3s 572us/step - loss: 0.4968 -
accuracy: 0.7965 - val_loss: 0.6435 - val_accuracy: 0.7739
Epoch 31/50
4689/4689 [============= ] - 3s 568us/step - loss: 0.4722 -
accuracy: 0.8089 - val_loss: 0.8470 - val_accuracy: 0.6945
Epoch 32/50
accuracy: 0.7829 - val_loss: 0.9696 - val_accuracy: 0.6834
Epoch 33/50
4689/4689 [============== ] - 3s 576us/step - loss: 0.4842 -
accuracy: 0.8160 - val_loss: 1.2420 - val_accuracy: 0.6971
Epoch 34/50
4689/4689 [============= ] - 3s 564us/step - loss: 0.5060 -
accuracy: 0.8121 - val_loss: 0.7033 - val_accuracy: 0.7415
accuracy: 0.8260 - val_loss: 1.1339 - val_accuracy: 0.6732
Epoch 36/50
accuracy: 0.8010 - val_loss: 0.7691 - val_accuracy: 0.7090
Epoch 37/50
accuracy: 0.8081 - val loss: 0.6896 - val accuracy: 0.7594
Epoch 38/50
accuracy: 0.8264 - val_loss: 0.7794 - val_accuracy: 0.7543
Epoch 39/50
4689/4689 [============= ] - 3s 573us/step - loss: 0.4614 -
accuracy: 0.8202 - val_loss: 1.1558 - val_accuracy: 0.6920
Epoch 40/50
4689/4689 [============= ] - 3s 569us/step - loss: 0.4746 -
accuracy: 0.8224 - val_loss: 1.0884 - val_accuracy: 0.7048
Epoch 41/50
4689/4689 [============== ] - 3s 584us/step - loss: 0.4472 -
```

```
accuracy: 0.8273 - val_loss: 0.7152 - val_accuracy: 0.7372
Epoch 42/50
4689/4689 [============= ] - 3s 573us/step - loss: 0.3950 -
accuracy: 0.8454 - val_loss: 0.9720 - val_accuracy: 0.7065
Epoch 43/50
4689/4689 [============= ] - 3s 579us/step - loss: 0.4640 -
accuracy: 0.8217 - val_loss: 0.9199 - val_accuracy: 0.7210
Epoch 44/50
4689/4689 [============= ] - 3s 575us/step - loss: 0.3967 -
accuracy: 0.8479 - val_loss: 0.6921 - val_accuracy: 0.7662
Epoch 45/50
4689/4689 [============== ] - 3s 570us/step - loss: 0.3698 -
accuracy: 0.8563 - val_loss: 1.0112 - val_accuracy: 0.6800
Epoch 46/50
accuracy: 0.8433 - val_loss: 0.7724 - val_accuracy: 0.7662
Epoch 47/50
4689/4689 [============= ] - 3s 568us/step - loss: 0.3679 -
accuracy: 0.8620 - val_loss: 0.8776 - val_accuracy: 0.7543
Epoch 48/50
4689/4689 [============== ] - 3s 568us/step - loss: 0.4186 -
accuracy: 0.8452 - val_loss: 0.7987 - val_accuracy: 0.7381
Epoch 49/50
4689/4689 [============= ] - 3s 556us/step - loss: 0.3105 -
accuracy: 0.8816 - val_loss: 1.0405 - val_accuracy: 0.6894
Epoch 50/50
accuracy: 0.8403 - val_loss: 0.9245 - val_accuracy: 0.7090
1954/1954 [============ ] - Os 242us/step
```



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Squential with Kfold CV:

Training for fold  $4 \dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

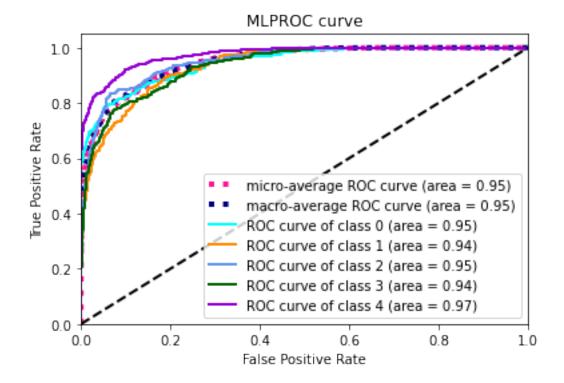
Model: "sequential\_24"

Layer (type)	Output Shape	Param #
dense_126 (Dense)	(None, 2176)	298112
dense_127 (Dense)	(None, 1088)	2368576
dropout_80 (Dropout)	(None, 1088)	0
dense_128 (Dense)	(None, 544)	592416
dropout_81 (Dropout)	(None, 544)	0
dense_129 (Dense)	(None, 272)	148240
dropout_82 (Dropout)	(None, 272)	0

```
(None, 5)
dense_130 (Dense)
                                     1365
______
Total params: 3,408,709
Trainable params: 3,408,709
Non-trainable params: 0
Train on 4689 samples, validate on 1172 samples
Epoch 1/50
4689/4689 [============== ] - 5s 1ms/step - loss: 2.4417 -
accuracy: 0.2583 - val_loss: 2.0449 - val_accuracy: 0.2807
Epoch 2/50
4689/4689 [=============== ] - 3s 570us/step - loss: 1.3734 -
accuracy: 0.4525 - val_loss: 1.8142 - val_accuracy: 0.3131
Epoch 3/50
accuracy: 0.5033 - val_loss: 1.6039 - val_accuracy: 0.3968
Epoch 4/50
4689/4689 [============== ] - 3s 564us/step - loss: 1.1459 -
accuracy: 0.5620 - val_loss: 0.9268 - val_accuracy: 0.6485
Epoch 5/50
accuracy: 0.5658 - val_loss: 1.1693 - val_accuracy: 0.5350
Epoch 6/50
accuracy: 0.6187 - val_loss: 1.0050 - val_accuracy: 0.5896
Epoch 7/50
4689/4689 [============= ] - 3s 561us/step - loss: 0.9202 -
accuracy: 0.6392 - val_loss: 1.0311 - val_accuracy: 0.5853
accuracy: 0.6618 - val_loss: 1.0110 - val_accuracy: 0.5418
accuracy: 0.6822 - val_loss: 0.9298 - val_accuracy: 0.6348
Epoch 10/50
4689/4689 [=============== ] - 3s 556us/step - loss: 0.7965 -
accuracy: 0.6848 - val_loss: 0.9277 - val_accuracy: 0.5734
Epoch 11/50
accuracy: 0.6854 - val_loss: 0.8498 - val_accuracy: 0.6681
Epoch 12/50
4689/4689 [============= ] - 3s 562us/step - loss: 0.7529 -
accuracy: 0.7119 - val_loss: 0.8588 - val_accuracy: 0.6340
Epoch 13/50
4689/4689 [============ ] - 3s 579us/step - loss: 0.7203 -
accuracy: 0.7147 - val_loss: 1.2631 - val_accuracy: 0.5981
Epoch 14/50
```

```
accuracy: 0.7168 - val_loss: 0.8866 - val_accuracy: 0.6800
Epoch 15/50
4689/4689 [============= ] - 3s 566us/step - loss: 0.7179 -
accuracy: 0.7191 - val_loss: 0.7543 - val_accuracy: 0.6997
Epoch 16/50
accuracy: 0.7569 - val_loss: 1.2529 - val_accuracy: 0.6135
Epoch 17/50
4689/4689 [============== ] - 3s 565us/step - loss: 0.7594 -
accuracy: 0.7112 - val_loss: 0.8348 - val_accuracy: 0.6399
Epoch 18/50
4689/4689 [============== ] - 3s 576us/step - loss: 0.6081 -
accuracy: 0.7607 - val_loss: 0.7937 - val_accuracy: 0.6980
Epoch 19/50
4689/4689 [============== ] - 3s 570us/step - loss: 0.6338 -
accuracy: 0.7496 - val_loss: 0.7909 - val_accuracy: 0.6920
Epoch 20/50
4689/4689 [============== ] - 3s 555us/step - loss: 0.6453 -
accuracy: 0.7415 - val_loss: 0.7640 - val_accuracy: 0.7099
Epoch 21/50
accuracy: 0.7688 - val_loss: 0.8225 - val_accuracy: 0.7031
Epoch 22/50
accuracy: 0.7588 - val_loss: 1.0537 - val_accuracy: 0.6561
Epoch 23/50
4689/4689 [============== ] - 3s 576us/step - loss: 0.6073 -
accuracy: 0.7656 - val_loss: 0.7441 - val_accuracy: 0.7398
accuracy: 0.7835 - val_loss: 0.7261 - val_accuracy: 0.7090
Epoch 25/50
accuracy: 0.8017 - val_loss: 0.8745 - val_accuracy: 0.6638
Epoch 26/50
accuracy: 0.7788 - val loss: 0.7999 - val accuracy: 0.7099
Epoch 27/50
accuracy: 0.7829 - val_loss: 1.1382 - val_accuracy: 0.6886
Epoch 28/50
4689/4689 [============= ] - 3s 589us/step - loss: 0.5747 -
accuracy: 0.7855 - val_loss: 0.6948 - val_accuracy: 0.7167
Epoch 29/50
4689/4689 [============ ] - 3s 627us/step - loss: 0.4657 -
accuracy: 0.8128 - val_loss: 0.7436 - val_accuracy: 0.7261
Epoch 30/50
```

```
accuracy: 0.7878 - val_loss: 0.7911 - val_accuracy: 0.7312
Epoch 31/50
4689/4689 [============= ] - 3s 570us/step - loss: 0.4699 -
accuracy: 0.8149 - val_loss: 0.7297 - val_accuracy: 0.7457
Epoch 32/50
4689/4689 [=============== ] - 3s 618us/step - loss: 0.4876 -
accuracy: 0.8070 - val_loss: 0.8710 - val_accuracy: 0.7176
Epoch 33/50
4689/4689 [============= ] - 3s 580us/step - loss: 0.4800 -
accuracy: 0.8100 - val_loss: 0.9452 - val_accuracy: 0.6177
Epoch 34/50
accuracy: 0.7912 - val_loss: 0.7583 - val_accuracy: 0.7201
Epoch 35/50
accuracy: 0.8232 - val_loss: 1.0094 - val_accuracy: 0.7022
Epoch 36/50
4689/4689 [============== ] - 3s 565us/step - loss: 0.4532 -
accuracy: 0.8277 - val_loss: 0.7671 - val_accuracy: 0.7304
Epoch 37/50
accuracy: 0.7997 - val_loss: 0.8137 - val_accuracy: 0.7287
Epoch 38/50
4689/4689 [============= ] - 3s 562us/step - loss: 0.4083 -
accuracy: 0.8379 - val_loss: 0.7561 - val_accuracy: 0.7440
Epoch 39/50
4689/4689 [============= ] - 3s 564us/step - loss: 0.4704 -
accuracy: 0.8123 - val_loss: 1.1343 - val_accuracy: 0.6681
4689/4689 [============== ] - 3s 559us/step - loss: 0.4652 -
accuracy: 0.8181 - val_loss: 0.7783 - val_accuracy: 0.7500
Epoch 41/50
accuracy: 0.8386 - val_loss: 0.9110 - val_accuracy: 0.7133
Epoch 42/50
4689/4689 [============== ] - 3s 579us/step - loss: 0.4062 -
accuracy: 0.8439 - val_loss: 1.2937 - val_accuracy: 0.6254
Epoch 43/50
accuracy: 0.8362 - val_loss: 0.9278 - val_accuracy: 0.7244
Epoch 44/50
4689/4689 [============= ] - 3s 570us/step - loss: 0.3570 -
accuracy: 0.8571 - val_loss: 0.8520 - val_accuracy: 0.7338
Epoch 45/50
4689/4689 [============= ] - 3s 574us/step - loss: 0.4336 -
accuracy: 0.8394 - val_loss: 0.8567 - val_accuracy: 0.7500
Epoch 46/50
4689/4689 [============== ] - 3s 571us/step - loss: 0.3908 -
```



```
Squential with Kfold CV:

Training for fold 5 ...

Adding layer 1:

Adding layer 2:

Adding layer 3:

Model: "sequential_25"

Layer (type)

Output Shape

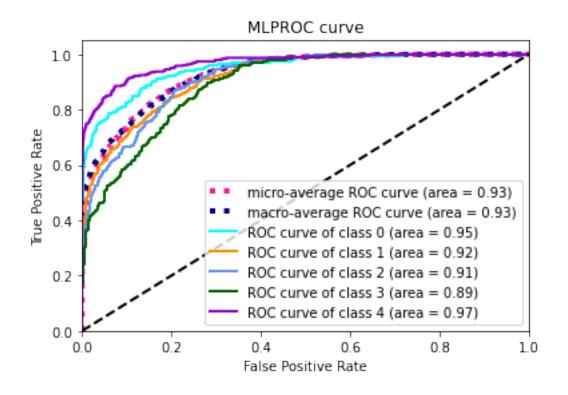
Param #
```

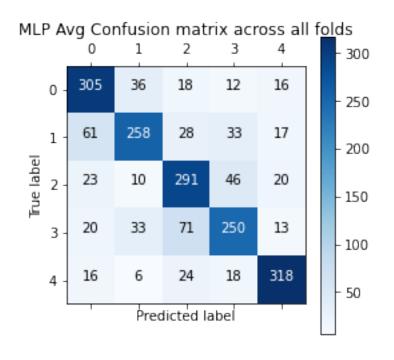
```
______
dense_131 (Dense)
                (None, 2176)
                                298112
            (None, 1088)
dense_132 (Dense)
                                2368576
______
dropout_83 (Dropout) (None, 1088)
_____
dense_133 (Dense) (None, 544)
                          592416
dropout_84 (Dropout) (None, 544)
    -----
dense_134 (Dense) (None, 272)
                                148240
-----
dropout_85 (Dropout) (None, 272)
______
dense_135 (Dense) (None, 5)
                                1365
______
Total params: 3,408,709
Trainable params: 3,408,709
Non-trainable params: 0
Train on 4689 samples, validate on 1172 samples
Epoch 1/50
accuracy: 0.2322 - val_loss: 1.5982 - val_accuracy: 0.3251
Epoch 2/50
4689/4689 [=============== ] - 3s 588us/step - loss: 1.3677 -
accuracy: 0.4316 - val_loss: 1.6841 - val_accuracy: 0.3703
accuracy: 0.5272 - val_loss: 3.2602 - val_accuracy: 0.3840
4689/4689 [============= ] - 3s 580us/step - loss: 1.3174 -
accuracy: 0.5404 - val_loss: 1.4051 - val_accuracy: 0.4514
Epoch 5/50
4689/4689 [============== ] - 3s 571us/step - loss: 1.0409 -
accuracy: 0.5907 - val loss: 1.2064 - val accuracy: 0.5776
Epoch 6/50
accuracy: 0.6270 - val_loss: 1.0321 - val_accuracy: 0.5896
Epoch 7/50
accuracy: 0.6470 - val_loss: 0.9957 - val_accuracy: 0.6237
Epoch 8/50
4689/4689 [============= ] - 3s 562us/step - loss: 0.8832 -
accuracy: 0.6517 - val_loss: 0.9062 - val_accuracy: 0.6340
Epoch 9/50
4689/4689 [============= ] - 3s 558us/step - loss: 0.7903 -
```

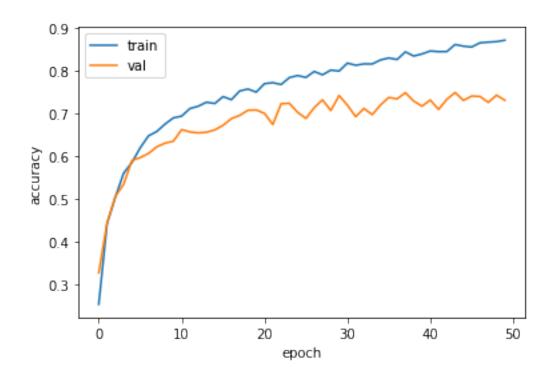
```
accuracy: 0.6897 - val_loss: 0.8547 - val_accuracy: 0.6553
Epoch 10/50
4689/4689 [============= ] - 3s 568us/step - loss: 0.7392 -
accuracy: 0.7140 - val_loss: 0.8292 - val_accuracy: 0.6613
Epoch 11/50
accuracy: 0.6709 - val_loss: 0.7966 - val_accuracy: 0.7176
Epoch 12/50
4689/4689 [============== ] - 3s 567us/step - loss: 0.6729 -
accuracy: 0.7413 - val_loss: 1.1107 - val_accuracy: 0.6271
Epoch 13/50
accuracy: 0.6914 - val_loss: 0.8463 - val_accuracy: 0.6962
Epoch 14/50
accuracy: 0.7279 - val_loss: 1.3926 - val_accuracy: 0.5956
Epoch 15/50
4689/4689 [============= ] - 3s 584us/step - loss: 0.7877 -
accuracy: 0.7125 - val_loss: 0.9539 - val_accuracy: 0.6587
Epoch 16/50
4689/4689 [=============== ] - 3s 566us/step - loss: 0.6800 -
accuracy: 0.7394 - val_loss: 0.9164 - val_accuracy: 0.6587
Epoch 17/50
4689/4689 [============= ] - 3s 572us/step - loss: 0.6782 -
accuracy: 0.7306 - val_loss: 0.8129 - val_accuracy: 0.6741
Epoch 18/50
4689/4689 [============== ] - 3s 591us/step - loss: 0.6890 -
accuracy: 0.7300 - val_loss: 0.8027 - val_accuracy: 0.6903
4689/4689 [============== ] - 3s 604us/step - loss: 0.6005 -
accuracy: 0.7628 - val_loss: 0.7694 - val_accuracy: 0.7167
Epoch 20/50
4689/4689 [=============== ] - 3s 641us/step - loss: 0.6170 -
accuracy: 0.7573 - val_loss: 0.6772 - val_accuracy: 0.7321
Epoch 21/50
accuracy: 0.7686 - val loss: 0.9144 - val accuracy: 0.6544
Epoch 22/50
accuracy: 0.7637 - val_loss: 0.8901 - val_accuracy: 0.7142
Epoch 23/50
4689/4689 [============= ] - 3s 599us/step - loss: 0.6653 -
accuracy: 0.7473 - val_loss: 0.7106 - val_accuracy: 0.7253
Epoch 24/50
4689/4689 [============ ] - 3s 591us/step - loss: 0.5481 -
accuracy: 0.7872 - val_loss: 0.9068 - val_accuracy: 0.6980
Epoch 25/50
4689/4689 [============== ] - 3s 601us/step - loss: 0.5681 -
```

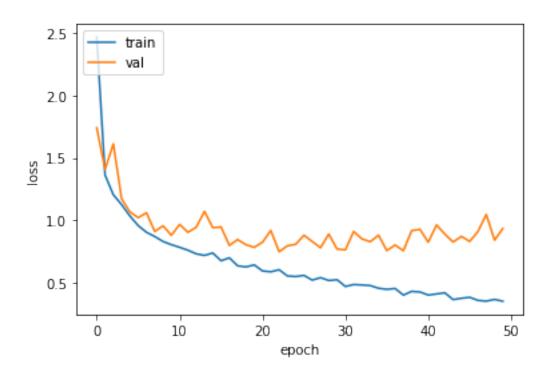
```
accuracy: 0.7748 - val_loss: 0.7723 - val_accuracy: 0.7201
Epoch 26/50
4689/4689 [============= ] - 3s 589us/step - loss: 0.5832 -
accuracy: 0.7771 - val_loss: 0.7964 - val_accuracy: 0.7167
Epoch 27/50
4689/4689 [=============== ] - 3s 590us/step - loss: 0.5319 -
accuracy: 0.7869 - val_loss: 0.8057 - val_accuracy: 0.6758
Epoch 28/50
4689/4689 [============= ] - 3s 590us/step - loss: 0.5728 -
accuracy: 0.7776 - val_loss: 0.8360 - val_accuracy: 0.7415
Epoch 29/50
4689/4689 [============== ] - 3s 600us/step - loss: 0.5257 -
accuracy: 0.7968 - val_loss: 0.9540 - val_accuracy: 0.7176
Epoch 30/50
4689/4689 [============= ] - 3s 592us/step - loss: 0.5255 -
accuracy: 0.8021 - val_loss: 0.9349 - val_accuracy: 0.7193
Epoch 31/50
4689/4689 [============= ] - 3s 567us/step - loss: 0.5248 -
accuracy: 0.8006 - val_loss: 0.8338 - val_accuracy: 0.6766
Epoch 32/50
accuracy: 0.8121 - val_loss: 0.7777 - val_accuracy: 0.7389
Epoch 33/50
4689/4689 [============== ] - 3s 574us/step - loss: 0.4814 -
accuracy: 0.8155 - val_loss: 0.7150 - val_accuracy: 0.7244
Epoch 34/50
4689/4689 [============== ] - 3s 570us/step - loss: 0.4571 -
accuracy: 0.8160 - val_loss: 0.7497 - val_accuracy: 0.7056
accuracy: 0.7995 - val_loss: 0.7096 - val_accuracy: 0.7312
Epoch 36/50
4689/4689 [============== ] - 3s 570us/step - loss: 0.4197 -
accuracy: 0.8433 - val_loss: 0.6907 - val_accuracy: 0.7662
Epoch 37/50
accuracy: 0.8302 - val loss: 0.9756 - val accuracy: 0.7133
Epoch 38/50
accuracy: 0.8364 - val_loss: 0.7898 - val_accuracy: 0.7457
Epoch 39/50
4689/4689 [============= ] - 3s 566us/step - loss: 0.4244 -
accuracy: 0.8430 - val_loss: 0.8364 - val_accuracy: 0.7474
Epoch 40/50
4689/4689 [============ ] - 3s 570us/step - loss: 0.4458 -
accuracy: 0.8360 - val_loss: 0.9183 - val_accuracy: 0.6698
Epoch 41/50
```

```
accuracy: 0.8467 - val_loss: 0.9958 - val_accuracy: 0.6980
Epoch 42/50
4689/4689 [============= ] - 3s 564us/step - loss: 0.4455 -
accuracy: 0.8339 - val_loss: 1.0579 - val_accuracy: 0.6928
Epoch 43/50
accuracy: 0.8379 - val_loss: 0.8423 - val_accuracy: 0.7381
Epoch 44/50
4689/4689 [============= ] - 3s 573us/step - loss: 0.3437 -
accuracy: 0.8686 - val_loss: 0.9433 - val_accuracy: 0.7398
Epoch 45/50
accuracy: 0.8537 - val_loss: 0.7895 - val_accuracy: 0.7457
Epoch 46/50
accuracy: 0.8763 - val_loss: 0.8810 - val_accuracy: 0.7253
Epoch 47/50
4689/4689 [============= ] - 3s 564us/step - loss: 0.3691 -
accuracy: 0.8567 - val_loss: 0.9169 - val_accuracy: 0.7270
Epoch 48/50
accuracy: 0.8631 - val_loss: 0.8885 - val_accuracy: 0.7637
Epoch 49/50
4689/4689 [============= ] - 3s 567us/step - loss: 0.3221 -
accuracy: 0.8731 - val_loss: 0.7822 - val_accuracy: 0.7543
Epoch 50/50
accuracy: 0.8795 - val_loss: 0.8623 - val_accuracy: 0.7073
```









Average scores for pesudo test set across all folds: > Accuracy: 0.7290685892105102 (+- 0.03517560582029268)

> Loss: 0.9378791475491244

> Avg runtime per test instance: 0.0009290013611011642

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## 

## 

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Squential with Kfold CV:

Training for fold 1 ...

Adding layer 1: Adding layer 2: Adding layer 3: Adding layer 4:

Model: "sequential\_26"

Layer (type)	Output	Shape	Param #
dense_136 (Dense)	(None,	4352)	596224
dense_137 (Dense)	(None,	2176)	9472128
dropout_86 (Dropout)	(None,	2176)	0
dense_138 (Dense)	(None,	1088)	2368576
dropout_87 (Dropout)	(None,	1088)	0
dense_139 (Dense)	(None,	544)	592416
dropout_88 (Dropout)	(None,	544)	0
dense_140 (Dense)	(None,	272)	148240
dropout_89 (Dropout)	(None,	272)	0
dense_141 (Dense)	(None,	5)	1365

Total params: 13,178,949 Trainable params: 13,178,949 Non-trainable params: 0

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Train on 4688 samples, validate on 1173 samples

Epoch 1/50

accuracy: 0.2080 - val\_loss: 1.6381 - val\_accuracy: 0.1927

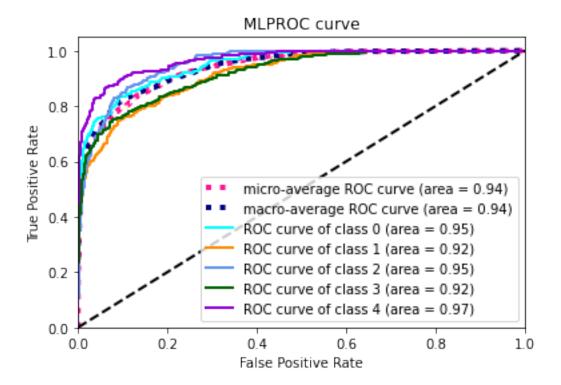
Epoch 2/50

accuracy: 0.2261 - val\_loss: 1.5656 - val\_accuracy: 0.2361

```
Epoch 3/50
accuracy: 0.3140 - val_loss: 1.2868 - val_accuracy: 0.4808
4688/4688 [============== ] - 8s 2ms/step - loss: 1.3801 -
accuracy: 0.4230 - val_loss: 1.2946 - val_accuracy: 0.4425
4688/4688 [============= ] - 8s 2ms/step - loss: 1.2720 -
accuracy: 0.4723 - val_loss: 1.1768 - val_accuracy: 0.5115
Epoch 6/50
4688/4688 [============== ] - 8s 2ms/step - loss: 1.1505 -
accuracy: 0.5109 - val_loss: 1.0828 - val_accuracy: 0.5507
Epoch 7/50
accuracy: 0.5482 - val_loss: 1.0192 - val_accuracy: 0.6206
Epoch 8/50
4688/4688 [============= ] - 8s 2ms/step - loss: 1.0036 -
accuracy: 0.5872 - val_loss: 0.9291 - val_accuracy: 0.6300
Epoch 9/50
accuracy: 0.6094 - val_loss: 0.8415 - val_accuracy: 0.6684
Epoch 10/50
accuracy: 0.6067 - val_loss: 0.8176 - val_accuracy: 0.6931
Epoch 11/50
4688/4688 [============== ] - 8s 2ms/step - loss: 0.9224 -
accuracy: 0.6416 - val_loss: 0.7930 - val_accuracy: 0.7059
Epoch 12/50
accuracy: 0.6734 - val_loss: 0.9875 - val_accuracy: 0.6309
Epoch 13/50
accuracy: 0.6811 - val_loss: 0.8496 - val_accuracy: 0.6777
Epoch 14/50
accuracy: 0.7003 - val_loss: 0.8860 - val_accuracy: 0.6462
Epoch 15/50
accuracy: 0.6736 - val_loss: 1.0178 - val_accuracy: 0.6198
Epoch 16/50
4688/4688 [============== ] - 8s 2ms/step - loss: 0.7384 -
accuracy: 0.7216 - val_loss: 0.9540 - val_accuracy: 0.6684
Epoch 17/50
accuracy: 0.6962 - val_loss: 0.7126 - val_accuracy: 0.7400
Epoch 18/50
4688/4688 [============== ] - 8s 2ms/step - loss: 0.7250 -
accuracy: 0.7229 - val_loss: 0.7854 - val_accuracy: 0.6965
```

```
Epoch 19/50
accuracy: 0.7186 - val_loss: 0.6936 - val_accuracy: 0.7425
Epoch 20/50
accuracy: 0.7651 - val_loss: 0.8724 - val_accuracy: 0.6684
4688/4688 [============ ] - 8s 2ms/step - loss: 0.7168 -
accuracy: 0.7186 - val_loss: 0.8075 - val_accuracy: 0.7093
Epoch 22/50
accuracy: 0.7351 - val_loss: 0.7483 - val_accuracy: 0.7400
Epoch 23/50
accuracy: 0.7541 - val_loss: 0.8198 - val_accuracy: 0.7263
Epoch 24/50
4688/4688 [============= ] - 8s 2ms/step - loss: 0.7021 -
accuracy: 0.7363 - val_loss: 0.6572 - val_accuracy: 0.7613
Epoch 25/50
accuracy: 0.7692 - val_loss: 0.7760 - val_accuracy: 0.7357
Epoch 26/50
accuracy: 0.7722 - val_loss: 0.7308 - val_accuracy: 0.7417
Epoch 27/50
accuracy: 0.7385 - val_loss: 0.6560 - val_accuracy: 0.7562
Epoch 28/50
accuracy: 0.7867 - val_loss: 0.8840 - val_accuracy: 0.6641
Epoch 29/50
accuracy: 0.7632 - val_loss: 0.7125 - val_accuracy: 0.7519
Epoch 30/50
accuracy: 0.7818 - val_loss: 0.7746 - val_accuracy: 0.7059
Epoch 31/50
4688/4688 [============= ] - 8s 2ms/step - loss: 0.5694 -
accuracy: 0.7873 - val_loss: 0.6795 - val_accuracy: 0.7681
Epoch 32/50
accuracy: 0.7929 - val_loss: 0.6806 - val_accuracy: 0.7579
accuracy: 0.7952 - val_loss: 0.8198 - val_accuracy: 0.7050
Epoch 34/50
4688/4688 [=============== ] - 8s 2ms/step - loss: 0.5905 -
accuracy: 0.7750 - val_loss: 0.7037 - val_accuracy: 0.7511
```

```
Epoch 35/50
accuracy: 0.7863 - val_loss: 0.8252 - val_accuracy: 0.7366
Epoch 36/50
4688/4688 [============== ] - 8s 2ms/step - loss: 0.5030 -
accuracy: 0.8108 - val_loss: 0.7484 - val_accuracy: 0.7442
Epoch 37/50
4688/4688 [============ ] - 8s 2ms/step - loss: 0.4883 -
accuracy: 0.8095 - val_loss: 0.9270 - val_accuracy: 0.7008
Epoch 38/50
accuracy: 0.8031 - val_loss: 0.7708 - val_accuracy: 0.7280
Epoch 39/50
accuracy: 0.8110 - val_loss: 0.7021 - val_accuracy: 0.7673
Epoch 40/50
4688/4688 [============= ] - 8s 2ms/step - loss: 0.4682 -
accuracy: 0.8234 - val_loss: 0.8232 - val_accuracy: 0.7323
Epoch 41/50
accuracy: 0.8035 - val_loss: 0.7145 - val_accuracy: 0.7621
Epoch 42/50
accuracy: 0.8370 - val_loss: 0.7317 - val_accuracy: 0.7664
Epoch 43/50
4688/4688 [============== ] - 8s 2ms/step - loss: 0.4860 -
accuracy: 0.8187 - val_loss: 0.9048 - val_accuracy: 0.7059
Epoch 44/50
accuracy: 0.8232 - val_loss: 0.7772 - val_accuracy: 0.7656
Epoch 45/50
accuracy: 0.8091 - val_loss: 0.7145 - val_accuracy: 0.7818
Epoch 46/50
4688/4688 [============= ] - 8s 2ms/step - loss: 0.3957 -
accuracy: 0.8560 - val_loss: 0.8852 - val_accuracy: 0.7204
Epoch 47/50
accuracy: 0.8419 - val_loss: 0.7729 - val_accuracy: 0.7562
Epoch 48/50
4688/4688 [============== ] - 8s 2ms/step - loss: 0.4341 -
accuracy: 0.8381 - val_loss: 0.9681 - val_accuracy: 0.7008
Epoch 49/50
4688/4688 [=============== ] - 8s 2ms/step - loss: 0.4138 -
accuracy: 0.8362 - val_loss: 0.8110 - val_accuracy: 0.7587
Epoch 50/50
4688/4688 [============== ] - 8s 2ms/step - loss: 0.4519 -
accuracy: 0.8232 - val_loss: 0.7668 - val_accuracy: 0.7468
```



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Squential with Kfold CV:

Training for fold 2 ...

Adding layer 1:

Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_27"

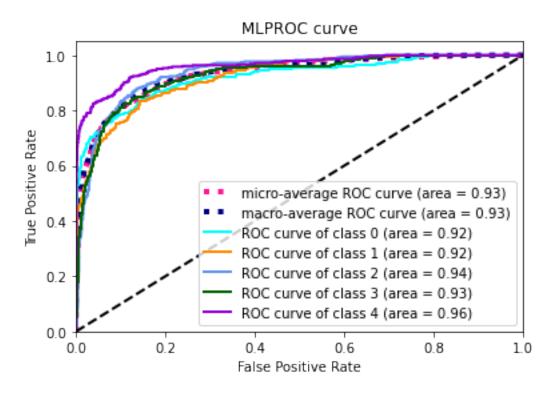
Layer (type)	Output Shape	Param #
dense_142 (Dense)	(None, 4352)	596224
dense_143 (Dense)	(None, 2176)	9472128
dropout_90 (Dropout)	(None, 2176)	0
dense_144 (Dense)	(None, 1088)	2368576
dropout_91 (Dropout)	(None, 1088)	0
dense_145 (Dense)	(None, 544)	592416

```
dropout_92 (Dropout) (None, 544)
             (None, 272)
dense_146 (Dense)
                                 148240
dropout_93 (Dropout) (None, 272)
_____
dense_147 (Dense) (None, 5) 1365
______
Total params: 13,178,949
Trainable params: 13,178,949
Non-trainable params: 0
Train on 4689 samples, validate on 1172 samples
Epoch 1/50
4689/4689 [============= ] - 11s 2ms/step - loss: 4.0237 -
accuracy: 0.2103 - val_loss: 2.0487 - val_accuracy: 0.2005
accuracy: 0.2190 - val_loss: 2.0395 - val_accuracy: 0.2014
accuracy: 0.2563 - val_loss: 2.3217 - val_accuracy: 0.2005
Epoch 4/50
4689/4689 [============== ] - 8s 2ms/step - loss: 1.6178 -
accuracy: 0.3137 - val_loss: 2.3415 - val_accuracy: 0.3379
Epoch 5/50
accuracy: 0.4182 - val_loss: 1.4950 - val_accuracy: 0.4206
Epoch 6/50
4689/4689 [============= ] - 8s 2ms/step - loss: 1.1945 -
accuracy: 0.4907 - val_loss: 1.3545 - val_accuracy: 0.4778
Epoch 7/50
accuracy: 0.5285 - val loss: 2.1440 - val accuracy: 0.3814
Epoch 8/50
4689/4689 [============== ] - 8s 2ms/step - loss: 1.2197 -
accuracy: 0.5451 - val_loss: 1.3135 - val_accuracy: 0.5034
Epoch 9/50
accuracy: 0.5841 - val_loss: 0.9585 - val_accuracy: 0.6314
Epoch 10/50
accuracy: 0.5865 - val_loss: 1.2158 - val_accuracy: 0.5811
Epoch 11/50
accuracy: 0.5984 - val_loss: 1.4115 - val_accuracy: 0.6169
Epoch 12/50
```

```
accuracy: 0.6298 - val_loss: 1.7074 - val_accuracy: 0.5265
Epoch 13/50
accuracy: 0.6351 - val_loss: 1.1789 - val_accuracy: 0.5111
Epoch 14/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.9125 -
accuracy: 0.6485 - val_loss: 0.9778 - val_accuracy: 0.6706
Epoch 15/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8228 -
accuracy: 0.7016 - val_loss: 1.1151 - val_accuracy: 0.5939
Epoch 16/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8526 -
accuracy: 0.6588 - val_loss: 1.1300 - val_accuracy: 0.5717
Epoch 17/50
accuracy: 0.6940 - val_loss: 0.9140 - val_accuracy: 0.6340
Epoch 18/50
accuracy: 0.6989 - val_loss: 0.8501 - val_accuracy: 0.6749
Epoch 19/50
accuracy: 0.6844 - val_loss: 1.0098 - val_accuracy: 0.6195
Epoch 20/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.7334 -
accuracy: 0.7187 - val_loss: 1.0968 - val_accuracy: 0.6578
Epoch 21/50
accuracy: 0.7016 - val_loss: 0.9364 - val_accuracy: 0.6664
Epoch 22/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.7162 -
accuracy: 0.7270 - val_loss: 0.9176 - val_accuracy: 0.6493
Epoch 23/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8106 -
accuracy: 0.6933 - val loss: 0.9856 - val accuracy: 0.6510
Epoch 24/50
accuracy: 0.7095 - val_loss: 1.1486 - val_accuracy: 0.6553
Epoch 25/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.7592 -
accuracy: 0.7189 - val_loss: 1.0055 - val_accuracy: 0.6101
Epoch 26/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.7202 -
accuracy: 0.7281 - val_loss: 0.8178 - val_accuracy: 0.6997
Epoch 27/50
accuracy: 0.7573 - val_loss: 0.7484 - val_accuracy: 0.7235
Epoch 28/50
```

```
accuracy: 0.7537 - val_loss: 0.8239 - val_accuracy: 0.7082
Epoch 29/50
accuracy: 0.7334 - val loss: 1.0383 - val accuracy: 0.6254
Epoch 30/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.6406 -
accuracy: 0.7579 - val_loss: 1.1286 - val_accuracy: 0.6203
Epoch 31/50
accuracy: 0.7298 - val_loss: 1.1251 - val_accuracy: 0.6254
Epoch 32/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6666 -
accuracy: 0.7633 - val_loss: 0.8580 - val_accuracy: 0.6894
Epoch 33/50
accuracy: 0.7646 - val_loss: 0.6853 - val_accuracy: 0.7346
Epoch 34/50
accuracy: 0.7899 - val_loss: 1.0711 - val_accuracy: 0.6101
Epoch 35/50
accuracy: 0.7498 - val_loss: 0.8397 - val_accuracy: 0.6962
Epoch 36/50
accuracy: 0.7695 - val_loss: 0.7886 - val_accuracy: 0.7125
Epoch 37/50
accuracy: 0.7759 - val_loss: 0.7944 - val_accuracy: 0.7022
Epoch 38/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.5855 -
accuracy: 0.7801 - val_loss: 0.9283 - val_accuracy: 0.7065
Epoch 39/50
accuracy: 0.7778 - val loss: 0.9900 - val accuracy: 0.7116
Epoch 40/50
accuracy: 0.7863 - val_loss: 1.0926 - val_accuracy: 0.6689
Epoch 41/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.6561 -
accuracy: 0.7639 - val_loss: 0.7815 - val_accuracy: 0.7201
Epoch 42/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5367 -
accuracy: 0.8019 - val_loss: 1.3795 - val_accuracy: 0.6237
Epoch 43/50
accuracy: 0.7626 - val_loss: 0.8611 - val_accuracy: 0.7108
Epoch 44/50
```

```
accuracy: 0.7831 - val_loss: 0.8333 - val_accuracy: 0.7048
Epoch 45/50
accuracy: 0.7720 - val_loss: 0.8343 - val_accuracy: 0.7090
Epoch 46/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5336 -
accuracy: 0.7919 - val_loss: 0.7028 - val_accuracy: 0.7628
Epoch 47/50
4689/4689 [=======
               ========= ] - 8s 2ms/step - loss: 0.5552 -
accuracy: 0.7893 - val_loss: 0.7527 - val_accuracy: 0.7466
Epoch 48/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5359 -
accuracy: 0.7961 - val_loss: 0.9366 - val_accuracy: 0.7039
Epoch 49/50
accuracy: 0.8049 - val_loss: 0.9575 - val_accuracy: 0.7099
Epoch 50/50
accuracy: 0.8025 - val_loss: 1.0556 - val_accuracy: 0.7218
```



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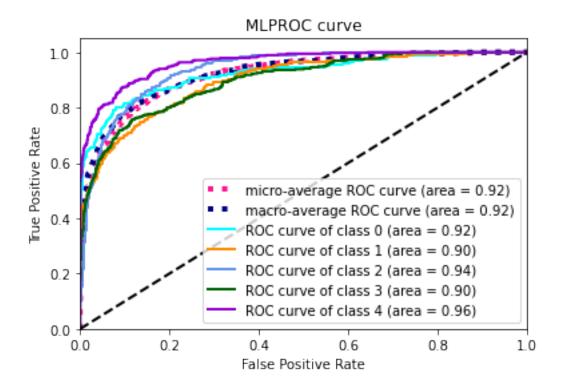
Squential with Kfold CV:

```
Training for fold 3 ...
Adding layer 1:
Adding layer 2:
Adding layer 3:
Adding layer 4:
Model: "sequential_28"
_____
Layer (type) Output Shape Param #
______
                 (None, 4352)
dense_148 (Dense)
                                  596224
dense_149 (Dense) (None, 2176)
                                 9472128
dropout_94 (Dropout) (None, 2176)
_____
dense_150 (Dense) (None, 1088)
                                  2368576
  -----
dropout_95 (Dropout) (None, 1088)
_____
dense_151 (Dense) (None, 544)
                                 592416
_____
dropout_96 (Dropout) (None, 544)
_____
dense_152 (Dense)
                 (None, 272)
                                  148240
_____
dropout_97 (Dropout) (None, 272)
dense_153 (Dense) (None, 5) 1365
______
Total params: 13,178,949
Trainable params: 13,178,949
Non-trainable params: 0
Train on 4689 samples, validate on 1172 samples
Epoch 1/50
4689/4689 [============== ] - 11s 2ms/step - loss: 3.8730 -
accuracy: 0.2054 - val loss: 2.3876 - val accuracy: 0.1971
Epoch 2/50
4689/4689 [============= ] - 8s 2ms/step - loss: 1.7506 -
accuracy: 0.2378 - val_loss: 2.6762 - val_accuracy: 0.2184
Epoch 3/50
4689/4689 [============= ] - 8s 2ms/step - loss: 1.6635 -
accuracy: 0.3210 - val_loss: 2.7767 - val_accuracy: 0.1988
Epoch 4/50
4689/4689 [=========== ] - 8s 2ms/step - loss: 1.6333 -
accuracy: 0.3312 - val_loss: 1.6755 - val_accuracy: 0.2210
Epoch 5/50
4689/4689 [============= ] - 8s 2ms/step - loss: 1.2763 -
```

```
accuracy: 0.4557 - val_loss: 1.3432 - val_accuracy: 0.3515
Epoch 6/50
4689/4689 [============== ] - 8s 2ms/step - loss: 1.2552 -
accuracy: 0.4715 - val_loss: 1.5296 - val_accuracy: 0.5324
Epoch 7/50
accuracy: 0.5146 - val_loss: 1.5151 - val_accuracy: 0.4343
Epoch 8/50
4689/4689 [============== ] - 8s 2ms/step - loss: 1.1778 -
accuracy: 0.5187 - val_loss: 1.2024 - val_accuracy: 0.5128
Epoch 9/50
accuracy: 0.5545 - val_loss: 1.2547 - val_accuracy: 0.4386
Epoch 10/50
accuracy: 0.5528 - val_loss: 1.5536 - val_accuracy: 0.4863
Epoch 11/50
4689/4689 [============== ] - 8s 2ms/step - loss: 1.0763 -
accuracy: 0.5760 - val_loss: 1.0940 - val_accuracy: 0.5469
Epoch 12/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.9794 -
accuracy: 0.6102 - val_loss: 1.0362 - val_accuracy: 0.5333
Epoch 13/50
accuracy: 0.6082 - val_loss: 1.0005 - val_accuracy: 0.6425
Epoch 14/50
accuracy: 0.6238 - val_loss: 0.9790 - val_accuracy: 0.6015
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8453 -
accuracy: 0.6773 - val_loss: 1.1148 - val_accuracy: 0.5563
Epoch 16/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8860 -
accuracy: 0.6520 - val_loss: 1.1445 - val_accuracy: 0.6246
Epoch 17/50
accuracy: 0.6665 - val loss: 0.8444 - val accuracy: 0.6664
Epoch 18/50
accuracy: 0.6801 - val_loss: 1.1690 - val_accuracy: 0.5145
Epoch 19/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8256 -
accuracy: 0.6831 - val_loss: 0.8050 - val_accuracy: 0.6988
Epoch 20/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.7852 -
accuracy: 0.6957 - val_loss: 0.9902 - val_accuracy: 0.6109
Epoch 21/50
4689/4689 [============== ] - 9s 2ms/step - loss: 0.7514 -
```

```
accuracy: 0.7123 - val_loss: 0.8125 - val_accuracy: 0.7014
Epoch 22/50
accuracy: 0.7127 - val_loss: 1.3133 - val_accuracy: 0.6152
Epoch 23/50
accuracy: 0.6884 - val_loss: 1.0505 - val_accuracy: 0.6485
Epoch 24/50
4689/4689 [============= ] - 9s 2ms/step - loss: 0.7433 -
accuracy: 0.7208 - val_loss: 1.0683 - val_accuracy: 0.6340
Epoch 25/50
accuracy: 0.7245 - val_loss: 1.2188 - val_accuracy: 0.5759
Epoch 26/50
accuracy: 0.7287 - val_loss: 0.9219 - val_accuracy: 0.6706
Epoch 27/50
4689/4689 [============== ] - 9s 2ms/step - loss: 0.6756 -
accuracy: 0.7430 - val_loss: 0.8099 - val_accuracy: 0.6903
Epoch 28/50
4689/4689 [=============== ] - 8s 2ms/step - loss: 0.6455 -
accuracy: 0.7496 - val_loss: 0.8854 - val_accuracy: 0.7176
Epoch 29/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.7017 -
accuracy: 0.7317 - val_loss: 0.8890 - val_accuracy: 0.6613
Epoch 30/50
accuracy: 0.7351 - val_loss: 0.6914 - val_accuracy: 0.7440
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6061 -
accuracy: 0.7731 - val_loss: 0.7483 - val_accuracy: 0.7218
Epoch 32/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6475 -
accuracy: 0.7575 - val_loss: 0.8727 - val_accuracy: 0.6928
Epoch 33/50
accuracy: 0.7592 - val loss: 1.3000 - val accuracy: 0.5768
Epoch 34/50
accuracy: 0.7334 - val_loss: 1.0099 - val_accuracy: 0.6519
Epoch 35/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6558 -
accuracy: 0.7520 - val_loss: 1.2856 - val_accuracy: 0.6032
Epoch 36/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.6722 -
accuracy: 0.7462 - val_loss: 0.6993 - val_accuracy: 0.7389
Epoch 37/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5919 -
```

```
accuracy: 0.7746 - val_loss: 0.7120 - val_accuracy: 0.7398
Epoch 38/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6239 -
accuracy: 0.7607 - val_loss: 1.1380 - val_accuracy: 0.6246
Epoch 39/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.7002 -
accuracy: 0.7364 - val_loss: 0.8770 - val_accuracy: 0.7005
Epoch 40/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6310 -
accuracy: 0.7592 - val_loss: 0.8037 - val_accuracy: 0.7167
Epoch 41/50
accuracy: 0.7754 - val_loss: 1.0212 - val_accuracy: 0.6758
Epoch 42/50
accuracy: 0.7550 - val_loss: 0.6764 - val_accuracy: 0.7406
Epoch 43/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.5984 -
accuracy: 0.7735 - val_loss: 0.7936 - val_accuracy: 0.6988
Epoch 44/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6386 -
accuracy: 0.7603 - val_loss: 0.8009 - val_accuracy: 0.6920
Epoch 45/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5661 -
accuracy: 0.7835 - val_loss: 0.8133 - val_accuracy: 0.7082
Epoch 46/50
accuracy: 0.7639 - val_loss: 0.9571 - val_accuracy: 0.6664
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5703 -
accuracy: 0.7857 - val_loss: 0.9087 - val_accuracy: 0.7125
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5491 -
accuracy: 0.7897 - val_loss: 0.6590 - val_accuracy: 0.7696
Epoch 49/50
accuracy: 0.8085 - val loss: 0.9966 - val accuracy: 0.7022
Epoch 50/50
accuracy: 0.7904 - val_loss: 1.0950 - val_accuracy: 0.7099
1954/1954 [========== ] - 1s 701us/step
```



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Squential with Kfold CV:

Training for fold 4  $\dots$ 

Adding layer 1: Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_29"

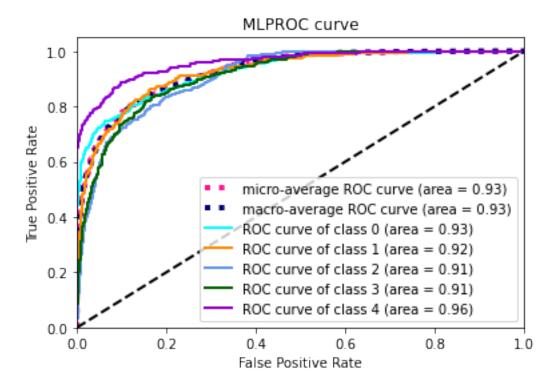
Layer (type)	Output Shape	Param #
dense_154 (Dense)	(None, 4352)	596224
dense_155 (Dense)	(None, 2176)	9472128
dropout_98 (Dropout)	(None, 2176)	0
dense_156 (Dense)	(None, 1088)	2368576
dropout_99 (Dropout)	(None, 1088)	0
dense_157 (Dense)	(None, 544)	592416
dropout_100 (Dropout)	(None, 544)	0

```
(None, 272)
dense_158 (Dense)
                                 148240
dropout_101 (Dropout) (None, 272)
                                 Ο
______
dense_159 (Dense) (None, 5)
______
Total params: 13,178,949
Trainable params: 13,178,949
Non-trainable params: 0
Train on 4689 samples, validate on 1172 samples
Epoch 1/50
4689/4689 [============ ] - 12s 3ms/step - loss: 4.3956 -
accuracy: 0.2024 - val_loss: 2.0463 - val_accuracy: 0.2048
Epoch 2/50
accuracy: 0.2273 - val_loss: 2.0997 - val_accuracy: 0.1724
Epoch 3/50
accuracy: 0.2917 - val_loss: 1.8044 - val_accuracy: 0.3379
Epoch 4/50
accuracy: 0.3764 - val_loss: 1.1498 - val_accuracy: 0.5614
Epoch 5/50
accuracy: 0.4519 - val_loss: 1.3095 - val_accuracy: 0.4130
Epoch 6/50
accuracy: 0.4803 - val_loss: 1.3498 - val_accuracy: 0.4727
Epoch 7/50
accuracy: 0.5231 - val_loss: 1.1543 - val_accuracy: 0.5427
Epoch 8/50
4689/4689 [============= ] - 8s 2ms/step - loss: 1.0974 -
accuracy: 0.5408 - val_loss: 1.0557 - val_accuracy: 0.5657
Epoch 9/50
accuracy: 0.5654 - val_loss: 1.4051 - val_accuracy: 0.4676
Epoch 10/50
4689/4689 [============== ] - 8s 2ms/step - loss: 1.1015 -
accuracy: 0.5666 - val_loss: 1.2692 - val_accuracy: 0.5452
4689/4689 [============ ] - 8s 2ms/step - loss: 1.0111 -
accuracy: 0.6006 - val_loss: 1.2174 - val_accuracy: 0.4582
Epoch 12/50
accuracy: 0.6206 - val_loss: 0.9386 - val_accuracy: 0.6288
```

```
Epoch 13/50
accuracy: 0.6488 - val_loss: 1.1407 - val_accuracy: 0.5717
Epoch 14/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8597 -
accuracy: 0.6799 - val_loss: 1.4098 - val_accuracy: 0.5469
Epoch 15/50
4689/4689 [=========== ] - 8s 2ms/step - loss: 0.9962 -
accuracy: 0.6221 - val_loss: 1.1353 - val_accuracy: 0.5776
Epoch 16/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.8611 -
accuracy: 0.6652 - val_loss: 0.7631 - val_accuracy: 0.7056
Epoch 17/50
accuracy: 0.6761 - val_loss: 1.7338 - val_accuracy: 0.5742
Epoch 18/50
accuracy: 0.6897 - val_loss: 1.1455 - val_accuracy: 0.6280
Epoch 19/50
accuracy: 0.6946 - val_loss: 0.6716 - val_accuracy: 0.7509
Epoch 20/50
accuracy: 0.6882 - val_loss: 0.8644 - val_accuracy: 0.6416
Epoch 21/50
accuracy: 0.6965 - val_loss: 0.7173 - val_accuracy: 0.7543
Epoch 22/50
accuracy: 0.7185 - val_loss: 0.8348 - val_accuracy: 0.7056
Epoch 23/50
accuracy: 0.7112 - val_loss: 0.7204 - val_accuracy: 0.7295
Epoch 24/50
accuracy: 0.6989 - val_loss: 0.9831 - val_accuracy: 0.6741
Epoch 25/50
accuracy: 0.7025 - val_loss: 0.8246 - val_accuracy: 0.7039
Epoch 26/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6846 -
accuracy: 0.7362 - val_loss: 0.9533 - val_accuracy: 0.6706
Epoch 27/50
accuracy: 0.7164 - val_loss: 1.5720 - val_accuracy: 0.5794
Epoch 28/50
accuracy: 0.7038 - val_loss: 0.8346 - val_accuracy: 0.7449
```

```
Epoch 29/50
accuracy: 0.7586 - val_loss: 0.7435 - val_accuracy: 0.7099
Epoch 30/50
accuracy: 0.7315 - val_loss: 0.6653 - val_accuracy: 0.7534
Epoch 31/50
4689/4689 [=========== ] - 8s 2ms/step - loss: 0.6983 -
accuracy: 0.7343 - val_loss: 0.7678 - val_accuracy: 0.7167
Epoch 32/50
accuracy: 0.7466 - val_loss: 0.9087 - val_accuracy: 0.6510
Epoch 33/50
accuracy: 0.7481 - val_loss: 0.6884 - val_accuracy: 0.7474
Epoch 34/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6371 -
accuracy: 0.7524 - val_loss: 0.7420 - val_accuracy: 0.7713
Epoch 35/50
accuracy: 0.7407 - val_loss: 0.6400 - val_accuracy: 0.7577
Epoch 36/50
accuracy: 0.7754 - val_loss: 1.1193 - val_accuracy: 0.6826
Epoch 37/50
accuracy: 0.7511 - val_loss: 0.8044 - val_accuracy: 0.7031
Epoch 38/50
accuracy: 0.7639 - val_loss: 0.8826 - val_accuracy: 0.6962
Epoch 39/50
accuracy: 0.7635 - val_loss: 0.9175 - val_accuracy: 0.6544
Epoch 40/50
accuracy: 0.7633 - val_loss: 0.7371 - val_accuracy: 0.7440
Epoch 41/50
accuracy: 0.7784 - val_loss: 0.6097 - val_accuracy: 0.7688
Epoch 42/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5721 -
accuracy: 0.7844 - val_loss: 0.8690 - val_accuracy: 0.6971
accuracy: 0.7724 - val_loss: 0.8957 - val_accuracy: 0.7278
Epoch 44/50
accuracy: 0.7852 - val_loss: 0.6552 - val_accuracy: 0.7841
```

```
Epoch 45/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5519 -
accuracy: 0.7855 - val_loss: 0.7807 - val_accuracy: 0.7073
Epoch 46/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.5342 -
accuracy: 0.7861 - val_loss: 0.6627 - val_accuracy: 0.7756
Epoch 47/50
accuracy: 0.7724 - val_loss: 0.7850 - val_accuracy: 0.7491
Epoch 48/50
4689/4689 [============== ] - 8s 2ms/step - loss: 0.6114 -
accuracy: 0.7754 - val_loss: 0.8071 - val_accuracy: 0.7594
Epoch 49/50
accuracy: 0.8136 - val_loss: 0.6622 - val_accuracy: 0.7722
Epoch 50/50
4689/4689 [============= ] - 8s 2ms/step - loss: 0.5701 -
accuracy: 0.7884 - val_loss: 0.9241 - val_accuracy: 0.7406
```



Squential with Kfold CV: Training for fold 5 ... Adding layer 1: Adding layer 2: Adding layer 3: Adding layer 4:

Epoch 6/50

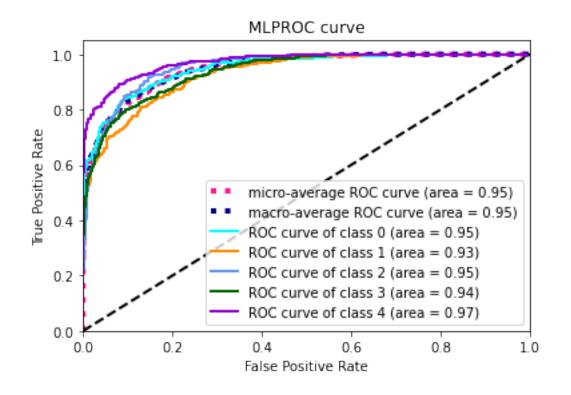
Model: "sequential\_30"

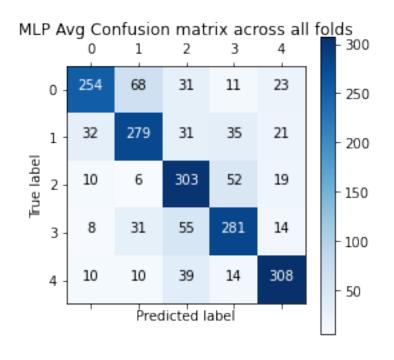
Layer (type)				
dense_160 (Dense)				
dense_161 (Dense)				
dropout_102 (Dropout)	(None,		0	
dense_162 (Dense)	(None,	1088)	2368576	
dropout_103 (Dropout)	(None,	1088)	0	
dense_163 (Dense)	(None,	544)	592416	
dropout_104 (Dropout)	(None,		0	
dense_164 (Dense)	(None,		148240	
dropout_105 (Dropout)	(None,	272)	0	
dense_165 (Dense)	(None,	5)	1365	
Total params: 13,178,949 Trainable params: 13,178,949 Non-trainable params: 0				
Train on 4689 samples, validate on 1172 samples Epoch 1/50				
4689/4689 [====================================			_	4.0180 -
4689/4689 [====================================				1.7492 -
4689/4689 [====================================			-	1.7192 -
4689/4689 [====================================			-	1.6015 -
4689/4689 [====================================			-	1.4457 -

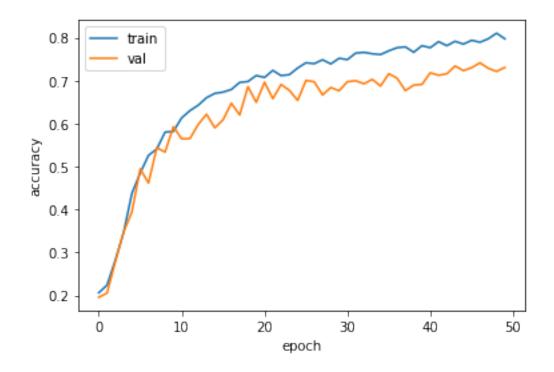
```
4689/4689 [=============== ] - 11s 2ms/step - loss: 1.3009 -
accuracy: 0.4722 - val_loss: 1.6291 - val_accuracy: 0.4403
Epoch 7/50
accuracy: 0.5165 - val_loss: 1.6820 - val_accuracy: 0.3302
Epoch 8/50
4689/4689 [============== ] - 11s 2ms/step - loss: 1.1925 -
accuracy: 0.5135 - val_loss: 1.3593 - val_accuracy: 0.5085
Epoch 9/50
accuracy: 0.5882 - val_loss: 1.4819 - val_accuracy: 0.4633
Epoch 10/50
4689/4689 [============== ] - 11s 2ms/step - loss: 1.0054 -
accuracy: 0.5967 - val_loss: 0.8600 - val_accuracy: 0.6587
Epoch 11/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.9332 -
accuracy: 0.6505 - val_loss: 1.3187 - val_accuracy: 0.4957
Epoch 12/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.9841 -
accuracy: 0.6174 - val_loss: 1.3824 - val_accuracy: 0.5068
Epoch 13/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.9523 -
accuracy: 0.6415 - val_loss: 1.2479 - val_accuracy: 0.5879
Epoch 14/50
accuracy: 0.6494 - val_loss: 1.0385 - val_accuracy: 0.6442
Epoch 15/50
accuracy: 0.6771 - val_loss: 1.0959 - val_accuracy: 0.6041
Epoch 16/50
accuracy: 0.6694 - val_loss: 1.2717 - val_accuracy: 0.4753
Epoch 17/50
accuracy: 0.6654 - val_loss: 1.0818 - val_accuracy: 0.6237
Epoch 18/50
accuracy: 0.6880 - val_loss: 1.2795 - val_accuracy: 0.5870
Epoch 19/50
accuracy: 0.7095 - val_loss: 1.1934 - val_accuracy: 0.6186
Epoch 20/50
accuracy: 0.6925 - val_loss: 0.8841 - val_accuracy: 0.6698
Epoch 21/50
4689/4689 [=============== ] - 11s 2ms/step - loss: 0.7732 -
accuracy: 0.7083 - val_loss: 0.9254 - val_accuracy: 0.6510
Epoch 22/50
```

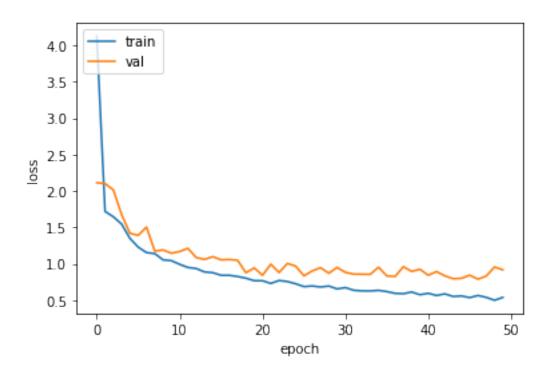
```
4689/4689 [=============== ] - 11s 2ms/step - loss: 0.7234 -
accuracy: 0.7266 - val_loss: 1.1501 - val_accuracy: 0.5802
Epoch 23/50
accuracy: 0.7121 - val_loss: 0.8112 - val_accuracy: 0.7005
Epoch 24/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.7706 -
accuracy: 0.7044 - val_loss: 1.1555 - val_accuracy: 0.6630
Epoch 25/50
accuracy: 0.7321 - val_loss: 1.0051 - val_accuracy: 0.6442
Epoch 26/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.6922 -
accuracy: 0.7432 - val_loss: 0.7365 - val_accuracy: 0.7210
Epoch 27/50
accuracy: 0.7428 - val_loss: 0.7138 - val_accuracy: 0.7381
Epoch 28/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.6794 -
accuracy: 0.7496 - val_loss: 1.2956 - val_accuracy: 0.4991
Epoch 29/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.7737 -
accuracy: 0.7078 - val_loss: 0.9559 - val_accuracy: 0.6724
Epoch 30/50
accuracy: 0.7541 - val_loss: 1.4787 - val_accuracy: 0.5589
Epoch 31/50
accuracy: 0.7206 - val_loss: 1.0782 - val_accuracy: 0.6561
Epoch 32/50
accuracy: 0.7605 - val_loss: 0.9620 - val_accuracy: 0.7082
Epoch 33/50
accuracy: 0.7624 - val_loss: 0.7823 - val_accuracy: 0.6997
Epoch 34/50
accuracy: 0.7624 - val_loss: 0.7374 - val_accuracy: 0.7312
Epoch 35/50
accuracy: 0.7754 - val_loss: 1.1648 - val_accuracy: 0.6442
Epoch 36/50
accuracy: 0.7456 - val_loss: 0.7984 - val_accuracy: 0.7031
Epoch 37/50
4689/4689 [=============== ] - 11s 2ms/step - loss: 0.5900 -
accuracy: 0.7731 - val_loss: 0.8834 - val_accuracy: 0.6826
Epoch 38/50
```

```
4689/4689 [=============== ] - 11s 2ms/step - loss: 0.5781 -
accuracy: 0.7852 - val_loss: 1.0709 - val_accuracy: 0.6288
Epoch 39/50
4689/4689 [============ ] - 11s 2ms/step - loss: 0.6650 -
accuracy: 0.7419 - val_loss: 0.9787 - val_accuracy: 0.6143
Epoch 40/50
accuracy: 0.7752 - val_loss: 1.1569 - val_accuracy: 0.5947
Epoch 41/50
accuracy: 0.7622 - val_loss: 1.0761 - val_accuracy: 0.6647
Epoch 42/50
accuracy: 0.7761 - val_loss: 0.7994 - val_accuracy: 0.7346
Epoch 43/50
4689/4689 [============= ] - 11s 2ms/step - loss: 0.5815 -
accuracy: 0.7799 - val_loss: 0.7064 - val_accuracy: 0.7372
Epoch 44/50
accuracy: 0.8066 - val_loss: 0.8968 - val_accuracy: 0.7244
Epoch 45/50
4689/4689 [============== ] - 11s 2ms/step - loss: 0.5742 -
accuracy: 0.7752 - val_loss: 0.8467 - val_accuracy: 0.7108
Epoch 46/50
accuracy: 0.7720 - val_loss: 0.9985 - val_accuracy: 0.7261
Epoch 47/50
accuracy: 0.7579 - val_loss: 0.7133 - val_accuracy: 0.7440
Epoch 48/50
accuracy: 0.7882 - val_loss: 0.7787 - val_accuracy: 0.7099
Epoch 49/50
accuracy: 0.7893 - val loss: 1.3433 - val accuracy: 0.6647
Epoch 50/50
accuracy: 0.7827 - val_loss: 0.7327 - val_accuracy: 0.7346
```





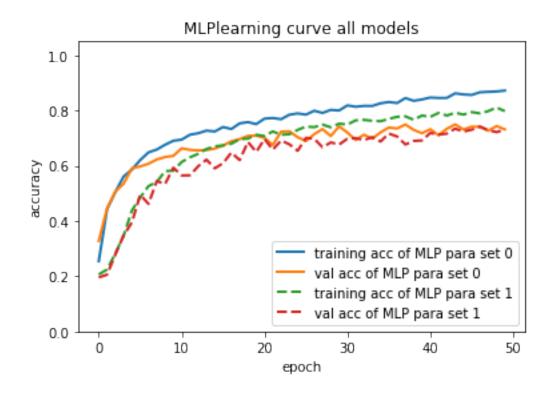


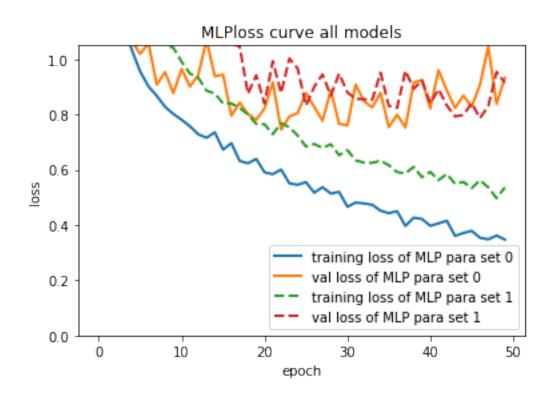


Average scores for pesudo test set across all folds: > Accuracy: 0.7301944732666016 (+- 0.01786776206097665)

> Loss: 0.9133736537081193

## 





```
Opt model parameter found on the pseudo test set:
     {'num_hidden_layer': 4, 'hidden_layer_activation': ['relu', 'tanh', 'relu',
     'tanh'], 'dropout': [0.5, 0.25, 0.125, 0.0625], 'last_activation': 'softmax'}
     Best average pseudo test set accuracy score with the opt model:
     0.7301944732666016
     Average runtime per test instance: 0.0022417235398853813
 []:
[14]: from B2 import B2_functions_all as B2_functions
[15]: TaskB2_opt_models_dict, TaskB2_res_dict = B2_functions.get_B2_results()
                   -----Reading Task B2 dataset-----
     -----Task B2: Gird searching on
     SVM-----
     SVM Grid search CV on Dataset A:
     Training scores:
     0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
     0.846 (+/-0.005) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
     0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'sigmoid'}
     0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
     0.846 (+/-0.005) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
     0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'sigmoid'}
     0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
     0.846 (+/-0.005) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
     0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'sigmoid'}
     0.472 (+/-0.120) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
     0.846 (+/-0.005) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
     0.367 (+/-0.001) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'sigmoid'}
     0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
     0.846 (+/-0.005) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
     0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'sigmoid'}
     0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
     0.846 (+/-0.005) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
     0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'sigmoid'}
```

```
0.368 (+/-0.002) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
0.846 (+/-0.005) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
0.368 (+/-0.002) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.848 (+/-0.005) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
0.846 (+/-0.005) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}
0.845 (+/-0.005) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.207 (+/-0.000) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.844 (+/-0.005) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
0.207 (+/-0.000) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.368 (+/-0.002) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
0.844 (+/-0.005) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
0.368 (+/-0.002) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.848 (+/-0.005) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
0.844 (+/-0.005) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
0.846 (+/-0.005) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.848 (+/-0.005) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
0.844 (+/-0.005) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}
0.845 (+/-0.005) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.368 (+/-0.002) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
0.368 (+/-0.002) for {'C': 1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.848 (+/-0.005) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
0.846 (+/-0.005) for {'C': 1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.847 (+/-0.004) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
0.846 (+/-0.005) for {'C': 1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.848 (+/-0.005) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.843 (+/-0.006) for {'C': 1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.847 (+/-0.005) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.846 (+/-0.005) for {'C': 10, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.846 (+/-0.005) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 10, 'gamma': 0.001, 'kernel': 'linear'}
0.846 (+/-0.005) for {'C': 10, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.847 (+/-0.006) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 10, 'gamma': 0.01, 'kernel': 'linear'}
0.844 (+/-0.005) for {'C': 10, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.849 (+/-0.003) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.845 (+/-0.005) for {'C': 10, 'gamma': 0.1, 'kernel': 'linear'}
0.836 (+/-0.005) for {'C': 10, 'gamma': 0.1, 'kernel': 'sigmoid'}
Validation scores:
0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'rbf'}
0.846 (+/-0.019) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'linear'}
0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.0001, 'kernel': 'sigmoid'}
```

```
0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'rbf'}
0.846 (+/-0.019) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'linear'}
0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'rbf'}
0.846 (+/-0.019) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'linear'}
0.207 (+/-0.000) for {'C': 0.001, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.469 (+/-0.112) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'rbf'}
0.846 (+/-0.019) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'linear'}
0.367 (+/-0.006) for {'C': 0.001, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'rbf'}
0.846 (+/-0.021) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'linear'}
0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'rbf'}
0.846 (+/-0.021) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'linear'}
0.207 (+/-0.000) for {'C': 0.01, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.368 (+/-0.007) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'rbf'}
0.846 (+/-0.021) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'linear'}
0.368 (+/-0.006) for {'C': 0.01, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.848 (+/-0.019) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'rbf'}
0.846 (+/-0.021) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'linear'}
0.845 (+/-0.021) for {'C': 0.01, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.207 (+/-0.000) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.845 (+/-0.020) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'linear'}
0.207 (+/-0.000) for {'C': 0.1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.368 (+/-0.008) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'rbf'}
0.845 (+/-0.020) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'linear'}
0.368 (+/-0.006) for {'C': 0.1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.848 (+/-0.018) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'rbf'}
0.845 (+/-0.020) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'linear'}
0.846 (+/-0.019) for {'C': 0.1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.848 (+/-0.019) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}
0.845 (+/-0.020) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'linear'}
0.845 (+/-0.019) for {'C': 0.1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.368 (+/-0.008) for {'C': 1, 'gamma': 0.0001, 'kernel': 'rbf'}
0.845 (+/-0.022) for {'C': 1, 'gamma': 0.0001, 'kernel': 'linear'}
0.368 (+/-0.006) for {'C': 1, 'gamma': 0.0001, 'kernel': 'sigmoid'}
0.848 (+/-0.019) for {'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
0.845 (+/-0.022) for {'C': 1, 'gamma': 0.001, 'kernel': 'linear'}
0.846 (+/-0.019) for {'C': 1, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.847 (+/-0.019) for {'C': 1, 'gamma': 0.01, 'kernel': 'rbf'}
0.845 (+/-0.022) for {'C': 1, 'gamma': 0.01, 'kernel': 'linear'}
0.846 (+/-0.021) for {'C': 1, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.848 (+/-0.020) for {'C': 1, 'gamma': 0.1, 'kernel': 'rbf'}
0.845 (+/-0.022) for {'C': 1, 'gamma': 0.1, 'kernel': 'linear'}
0.845 (+/-0.019) for {'C': 1, 'gamma': 0.1, 'kernel': 'sigmoid'}
0.847 (+/-0.019) for {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
0.844 (+/-0.021) for {'C': 10, 'gamma': 0.0001, 'kernel': 'linear'}
0.846 (+/-0.019) for {'C': 10, 'gamma': 0.0001, 'kernel': 'sigmoid'}
```

```
0.845 (+/-0.021) for {'C': 10, 'gamma': 0.001, 'kernel': 'rbf'}
0.844 (+/-0.021) for {'C': 10, 'gamma': 0.001, 'kernel': 'linear'}
0.846 (+/-0.021) for {'C': 10, 'gamma': 0.001, 'kernel': 'sigmoid'}
0.845 (+/-0.018) for {'C': 10, 'gamma': 0.01, 'kernel': 'rbf'}
0.844 (+/-0.021) for {'C': 10, 'gamma': 0.01, 'kernel': 'linear'}
0.844 (+/-0.020) for {'C': 10, 'gamma': 0.01, 'kernel': 'sigmoid'}
0.847 (+/-0.019) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.844 (+/-0.021) for {'C': 10, 'gamma': 0.1, 'kernel': 'rbf'}
0.847 (+/-0.021) for {'C': 10, 'gamma': 0.1, 'kernel': 'linear'}
0.837 (+/-0.022) for {'C': 10, 'gamma': 0.1, 'kernel': 'sigmoid'}
```

Prediction on a pseudo test set (split from Dataset A):

precision recall f1-score support

	precipion	rccarr	II BCOIC	Buppor
0	0.91	0.82	0.86	420
1	1.00	0.80	0.89	411
2	0.99	0.77	0.87	402
3	0.96	0.81	0.87	401
4	0.56	0.95	0.70	389
accuracy			0.83	2023
macro avg	0.88	0.83	0.84	2023
weighted avg	0.89	0.83	0.84	2023

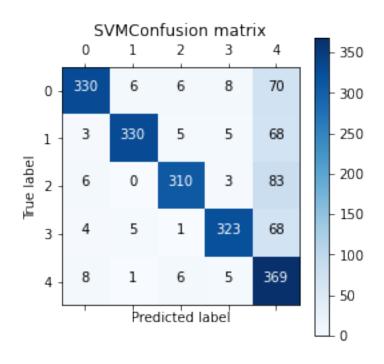
Accuracy: 0.8284725654967869

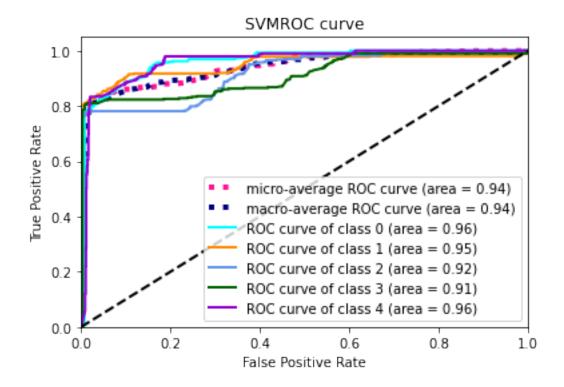
Best parameters found on Dataset A:

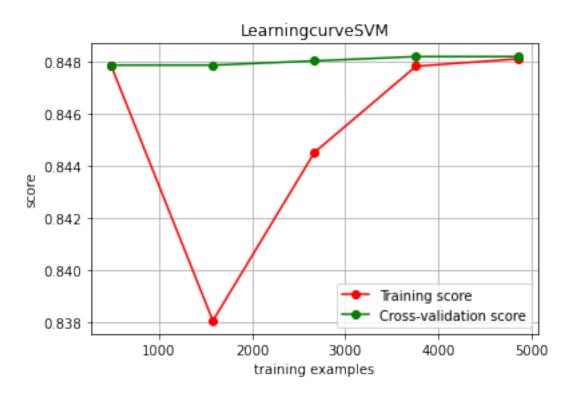
{'C': 0.1, 'gamma': 0.1, 'kernel': 'rbf'}

Average runtime per test instance: 0.00011334758669918737

-----







```
-----Task B2: Gird searching on Rando Forest-----
```

RF Grid search CV on Dataset A:

## Training scores:

```
0.935 (+/-0.003) for {'max_depth': 64, 'n_estimators': 64}
0.935 (+/-0.003) for {'max_depth': 64, 'n_estimators': 128}
0.935 (+/-0.003) for {'max_depth': 64, 'n_estimators': 256}
0.935 (+/-0.003) for {'max_depth': 64, 'n_estimators': 512}
0.935 (+/-0.003) for {'max_depth': 64, 'n_estimators': 1024}
0.935 (+/-0.003) for {'max_depth': 128, 'n_estimators': 64}
0.935 (+/-0.003) for {'max_depth': 128, 'n_estimators': 128}
0.935 (+/-0.003) for {'max_depth': 128, 'n_estimators': 256}
0.935 (+/-0.003) for {'max_depth': 128, 'n_estimators': 512}
0.935 (+/-0.003) for {'max_depth': 128, 'n_estimators': 512}
0.935 (+/-0.003) for {'max_depth': 256, 'n_estimators': 64}
0.935 (+/-0.003) for {'max_depth': 256, 'n_estimators': 128}
0.935 (+/-0.003) for {'max_depth': 256, 'n_estimators': 256}
0.935 (+/-0.003) for {'max_depth': 256, 'n_estimators': 512}
0.935 (+/-0.003) for {'max_depth': 256, 'n_estimators': 512}
```

```
0.935 (+/-0.003) for {'max_depth': 256, 'n_estimators': 1024}
0.935 (+/-0.003) for {'max_depth': 512, 'n_estimators': 64}
0.935 (+/-0.003) for {'max_depth': 512, 'n_estimators': 128}
0.935 (+/-0.003) for {'max_depth': 512, 'n_estimators': 256}
0.935 \ (+/-0.003)  for {'max depth': 512, 'n estimators': 512}
0.935 (+/-0.003) for {'max_depth': 512, 'n_estimators': 1024}
0.935 \ (+/-0.003)  for {'max depth': 1024, 'n estimators': 64}
0.935 (+/-0.003) for {'max_depth': 1024, 'n_estimators': 128}
0.935 (+/-0.003) for {'max depth': 1024, 'n estimators': 256}
0.935 \ (+/-0.003) \ for \ {\rm 'max\_depth': 1024, 'n\_estimators': 512}
0.935 (+/-0.003) for {'max_depth': 1024, 'n_estimators': 1024}
Validation scores:
0.867 (+/-0.016) for {'max_depth': 64, 'n_estimators': 64}
0.867 (+/-0.015) for {'max_depth': 64, 'n_estimators': 128}
0.866 (+/-0.019) for {'max_depth': 64, 'n_estimators': 256}
0.866 (+/-0.017) for {'max_depth': 64, 'n_estimators': 512}
0.866 (+/-0.017) for {'max_depth': 64, 'n_estimators': 1024}
0.868 (+/-0.016) for {'max depth': 128, 'n estimators': 64}
0.867 (+/-0.017) for {'max_depth': 128, 'n_estimators': 128}
0.866 (+/-0.018) for {'max_depth': 128, 'n_estimators': 256}
0.866 (+/-0.016) for {'max_depth': 128, 'n_estimators': 512}
0.868 (+/-0.016) for {'max_depth': 128, 'n_estimators': 1024}
0.865 (+/-0.018) for {'max_depth': 256, 'n_estimators': 64}
0.866 (+/-0.020) for {'max_depth': 256, 'n_estimators': 128}
0.865 (+/-0.020) for {'max_depth': 256, 'n_estimators': 256}
0.867 (+/-0.021) for {'max_depth': 256, 'n_estimators': 512}
0.866 (+/-0.019) for {'max_depth': 256, 'n_estimators': 1024}
0.866 (+/-0.016) for {'max_depth': 512, 'n_estimators': 64}
0.866 (+/-0.018) for {'max_depth': 512, 'n_estimators': 128}
0.866 (+/-0.019) for {'max_depth': 512, 'n_estimators': 256}
0.866 \ (+/-0.015) \ for \ {\rm 'max\_depth': 512, 'n\_estimators': 512}
0.866 \ (+/-0.018)  for {\max_{depth'}: 512, 'n_{estimators'}: 1024}
0.865 \ (+/-0.016)  for {'max depth': 1024, 'n estimators': 64}
0.867 (+/-0.017) for {'max_depth': 1024, 'n_estimators': 128}
0.866 (+/-0.016) for {'max depth': 1024, 'n estimators': 256}
0.867 (+/-0.020) for {'max_depth': 1024, 'n_estimators': 512}
0.867 (+/-0.018) for {'max_depth': 1024, 'n_estimators': 1024}
Prediction on a pseudo test set (split from Dataset A):
              precision
                           recall f1-score
                                               support
           0
                   0.89
                             0.87
                                        0.88
                                                   420
           1
                   0.93
                             0.85
                                        0.89
                                                   411
```

2

3

0.89

0.87

0.74

0.82

0.85

0.92

195

0.86

0.86

0.82

402

401

389

accuracy			0.86	2023
macro avg	0.87	0.86	0.86	2023
weighted avg	0.87	0.86	0.86	2023

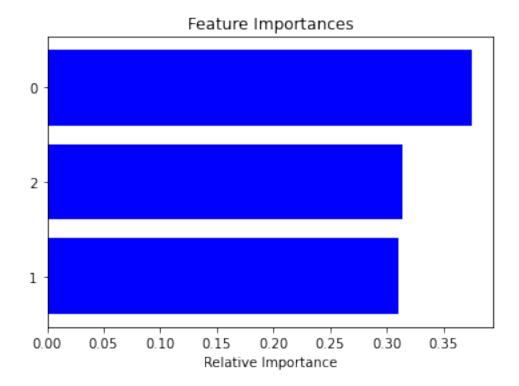
Accuracy: 0.8601087493821058

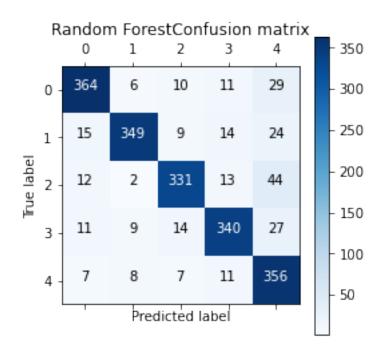
Best parameters found on Dataset A:

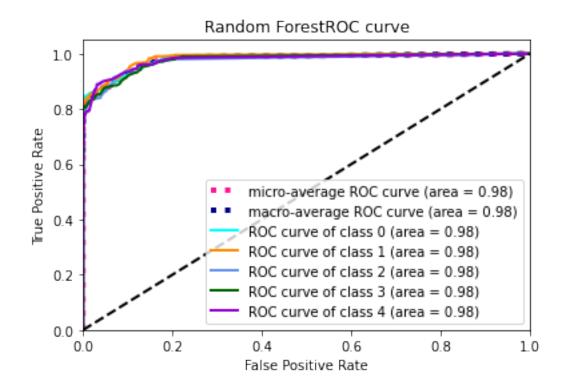
{'max\_depth': 128, 'n\_estimators': 64}

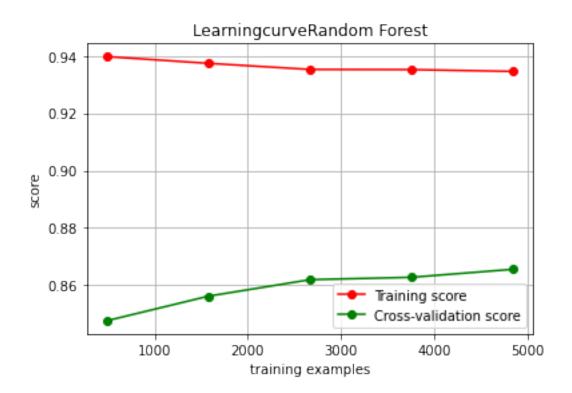
Average runtime per test instance: 1.924119252588156e-05

-----









```
------Task B2: Gird searching on KNN------
KNN Grid search CV on Dataset A:

Training scores:

0.879 (+/-0.004) for {'n_neighbors': 8}
0.864 (+/-0.004) for {'n_neighbors': 16}
```

## Validation scores:

```
0.853 (+/-0.030) for {'n_neighbors': 8}
0.847 (+/-0.022) for {'n_neighbors': 16}
0.843 (+/-0.023) for {'n_neighbors': 32}
0.842 (+/-0.019) for {'n_neighbors': 64}
0.840 (+/-0.019) for {'n_neighbors': 128}
```

0.860 (+/-0.005) for {'n\_neighbors': 32} 0.852 (+/-0.005) for {'n\_neighbors': 64} 0.847 (+/-0.005) for {'n\_neighbors': 128}

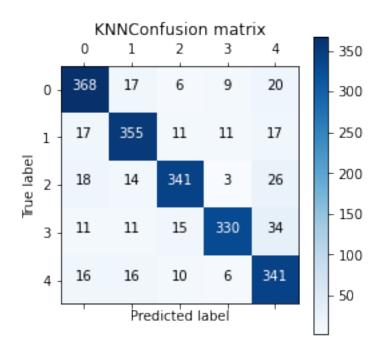
0 0.86 0.88 0.87 420 1 0.86 0.86 0.86 411 2 0.89 0.85 0.87 402 3 0.92 0.82 0.87 401 4 0.78 0.88 0.82 389  accuracy 0.86 0.86 0.86 2023 weighted avg 0.86 0.86 0.86 0.86 2023						
2 0.89 0.85 0.87 402 3 0.92 0.82 0.87 401 4 0.78 0.88 0.82 389  accuracy 0.86 2023 macro avg 0.86 0.86 0.86 2023		0	0.86	0.88	0.87	420
3 0.92 0.82 0.87 401 4 0.78 0.88 0.82 389 accuracy 0.86 2023 macro avg 0.86 0.86 0.86 2023		1	0.86	0.86	0.86	411
4 0.78 0.88 0.82 389  accuracy 0.86 2023 macro avg 0.86 0.86 0.86 2023		2	0.89	0.85	0.87	402
accuracy 0.86 2023 macro avg 0.86 0.86 0.86 2023		3	0.92	0.82	0.87	401
macro avg 0.86 0.86 0.86 2023		4	0.78	0.88	0.82	389
macro avg 0.86 0.86 0.86 2023						
	accura	су			0.86	2023
weighted avg 0.86 0.86 0.86 2023	macro a	vg	0.86	0.86	0.86	2023
	weighted av	vg	0.86	0.86	0.86	2023

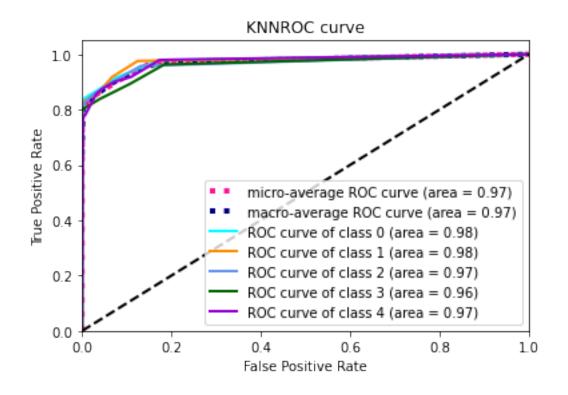
Accuracy: 0.8576371725160653

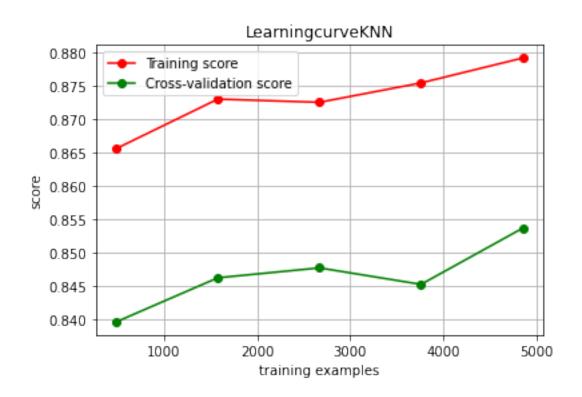
Best parameters found on Dataset A:

{'n\_neighbors': 8}

Average runtime per test instance: 6.944144414429375e-05





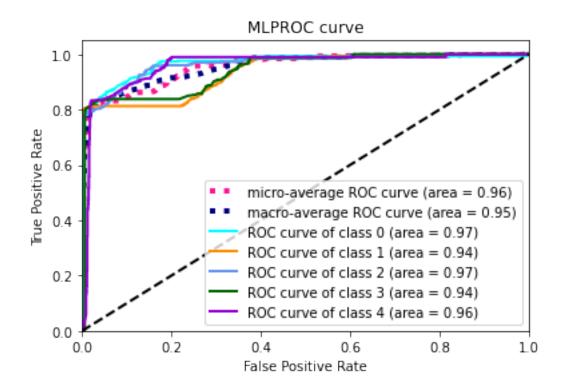


```
-----Task B2: Gird searching on
MT.P-----
Squential with Kfold CV:
Training for fold 1 ...
Adding layer 1:
Adding layer 2:
Adding layer 3:
Model: "sequential_31"
-----
Layer (type)
             Output Shape
______
dense_166 (Dense)
             (None, 48)
                          192
_____
dense_167 (Dense) (None, 24)
                          1176
_____
dropout_106 (Dropout) (None, 24)
_____
dense_168 (Dense) (None, 12)
                          300
dropout_107 (Dropout) (None, 12)
dense_169 (Dense)
             (None, 6)
                          78
  _____
dropout_108 (Dropout) (None, 6)
_____
dense_170 (Dense) (None, 5)
______
Total params: 1,781
Trainable params: 1,781
Non-trainable params: 0
-----
Train on 4853 samples, validate on 1214 samples
Epoch 1/50
accuracy: 0.2475 - val_loss: 1.4781 - val_accuracy: 0.3616
Epoch 2/50
4853/4853 [============== ] - Os 41us/step - loss: 1.4961 -
accuracy: 0.3433 - val_loss: 1.4022 - val_accuracy: 0.5214
Epoch 3/50
accuracy: 0.3991 - val_loss: 1.3422 - val_accuracy: 0.6853
Epoch 4/50
```

```
accuracy: 0.4731 - val_loss: 1.2817 - val_accuracy: 0.6862
Epoch 5/50
accuracy: 0.4859 - val_loss: 1.2291 - val_accuracy: 0.6853
Epoch 6/50
accuracy: 0.5170 - val_loss: 1.1768 - val_accuracy: 0.8418
Epoch 7/50
accuracy: 0.5314 - val_loss: 1.1290 - val_accuracy: 0.8443
Epoch 8/50
accuracy: 0.5308 - val_loss: 1.0957 - val_accuracy: 0.8443
Epoch 9/50
accuracy: 0.5677 - val_loss: 1.0565 - val_accuracy: 0.8451
Epoch 10/50
accuracy: 0.5683 - val_loss: 1.0260 - val_accuracy: 0.8476
Epoch 11/50
accuracy: 0.5943 - val_loss: 0.9905 - val_accuracy: 0.8451
Epoch 12/50
accuracy: 0.6038 - val_loss: 0.9599 - val_accuracy: 0.8443
Epoch 13/50
accuracy: 0.6153 - val_loss: 0.9324 - val_accuracy: 0.8443
accuracy: 0.6198 - val_loss: 0.9030 - val_accuracy: 0.8468
Epoch 15/50
accuracy: 0.6192 - val_loss: 0.8766 - val_accuracy: 0.8435
Epoch 16/50
4853/4853 [=============== ] - Os 37us/step - loss: 1.0934 -
accuracy: 0.6322 - val loss: 0.8578 - val accuracy: 0.8427
Epoch 17/50
accuracy: 0.6513 - val_loss: 0.8330 - val_accuracy: 0.8443
Epoch 18/50
accuracy: 0.6833 - val_loss: 0.8071 - val_accuracy: 0.8451
Epoch 19/50
accuracy: 0.6765 - val_loss: 0.7907 - val_accuracy: 0.8451
Epoch 20/50
```

```
accuracy: 0.6893 - val_loss: 0.7738 - val_accuracy: 0.8443
Epoch 21/50
accuracy: 0.6973 - val_loss: 0.7557 - val_accuracy: 0.8451
Epoch 22/50
accuracy: 0.7020 - val_loss: 0.7412 - val_accuracy: 0.8443
Epoch 23/50
accuracy: 0.7142 - val_loss: 0.7253 - val_accuracy: 0.8443
Epoch 24/50
accuracy: 0.7183 - val_loss: 0.7081 - val_accuracy: 0.8435
Epoch 25/50
accuracy: 0.7216 - val_loss: 0.6950 - val_accuracy: 0.8443
Epoch 26/50
accuracy: 0.7140 - val_loss: 0.6842 - val_accuracy: 0.8427
Epoch 27/50
accuracy: 0.7239 - val_loss: 0.6750 - val_accuracy: 0.8435
Epoch 28/50
accuracy: 0.7264 - val_loss: 0.6638 - val_accuracy: 0.8418
Epoch 29/50
accuracy: 0.7268 - val_loss: 0.6534 - val_accuracy: 0.8418
4853/4853 [============== ] - Os 39us/step - loss: 0.8934 -
accuracy: 0.7309 - val_loss: 0.6433 - val_accuracy: 0.8418
Epoch 31/50
accuracy: 0.7255 - val_loss: 0.6353 - val_accuracy: 0.8402
Epoch 32/50
4853/4853 [=============== ] - Os 41us/step - loss: 0.8691 -
accuracy: 0.7313 - val_loss: 0.6307 - val_accuracy: 0.8410
Epoch 33/50
accuracy: 0.7344 - val_loss: 0.6217 - val_accuracy: 0.8410
Epoch 34/50
accuracy: 0.7224 - val_loss: 0.6149 - val_accuracy: 0.8418
Epoch 35/50
accuracy: 0.7311 - val_loss: 0.6087 - val_accuracy: 0.8418
Epoch 36/50
```

```
accuracy: 0.7317 - val_loss: 0.6023 - val_accuracy: 0.8418
Epoch 37/50
accuracy: 0.7377 - val_loss: 0.5953 - val_accuracy: 0.8418
Epoch 38/50
accuracy: 0.7402 - val_loss: 0.5905 - val_accuracy: 0.8418
Epoch 39/50
accuracy: 0.7412 - val_loss: 0.5830 - val_accuracy: 0.8418
Epoch 40/50
accuracy: 0.7418 - val_loss: 0.5810 - val_accuracy: 0.8418
Epoch 41/50
4853/4853 [============== ] - Os 22us/step - loss: 0.7986 -
accuracy: 0.7457 - val_loss: 0.5788 - val_accuracy: 0.8410
Epoch 42/50
accuracy: 0.7342 - val_loss: 0.5736 - val_accuracy: 0.8418
Epoch 43/50
accuracy: 0.7367 - val_loss: 0.5694 - val_accuracy: 0.8410
Epoch 44/50
accuracy: 0.7465 - val_loss: 0.5654 - val_accuracy: 0.8427
Epoch 45/50
accuracy: 0.7439 - val_loss: 0.5608 - val_accuracy: 0.8418
accuracy: 0.7525 - val_loss: 0.5569 - val_accuracy: 0.8427
Epoch 47/50
accuracy: 0.7501 - val_loss: 0.5532 - val_accuracy: 0.8418
Epoch 48/50
4853/4853 [=============== ] - Os 30us/step - loss: 0.7848 -
accuracy: 0.7449 - val loss: 0.5511 - val accuracy: 0.8427
Epoch 49/50
accuracy: 0.7459 - val_loss: 0.5490 - val_accuracy: 0.8427
Epoch 50/50
accuracy: 0.7443 - val_loss: 0.5447 - val_accuracy: 0.8410
2023/2023 [=========== ] - 0s 84us/step
```



Squential with Kfold CV:

Training for fold  $2 \dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

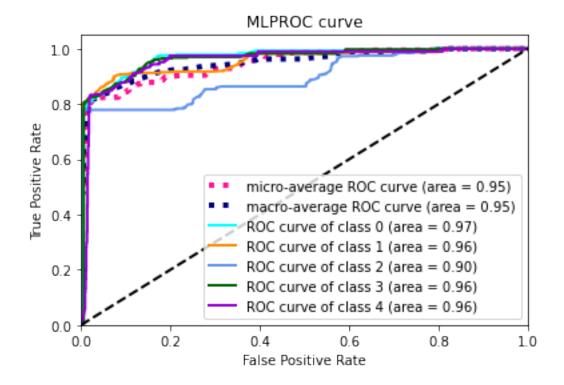
Model: "sequential\_32"

Layer (type)	Output Shape	Param #
dense_171 (Dense)	(None, 48)	192
dense_172 (Dense)	(None, 24)	1176
dropout_109 (Dropout)	(None, 24)	0
dense_173 (Dense)	(None, 12)	300
dropout_110 (Dropout)	(None, 12)	0
dense_174 (Dense)	(None, 6)	78
dropout_111 (Dropout)	(None, 6)	0

```
dense_175 (Dense)
               (None, 5)
                              35
______
Total params: 1,781
Trainable params: 1,781
Non-trainable params: 0
Train on 4853 samples, validate on 1214 samples
Epoch 1/50
accuracy: 0.2155 - val_loss: 1.5017 - val_accuracy: 0.3649
Epoch 2/50
accuracy: 0.3128 - val_loss: 1.4313 - val_accuracy: 0.3682
Epoch 3/50
4853/4853 [=============== ] - Os 40us/step - loss: 1.4580 -
accuracy: 0.3538 - val_loss: 1.3830 - val_accuracy: 0.5445
Epoch 4/50
accuracy: 0.3995 - val_loss: 1.3190 - val_accuracy: 0.5445
Epoch 5/50
accuracy: 0.4379 - val_loss: 1.2627 - val_accuracy: 0.5437
Epoch 6/50
accuracy: 0.4719 - val_loss: 1.2062 - val_accuracy: 0.6269
Epoch 7/50
accuracy: 0.5114 - val_loss: 1.1497 - val_accuracy: 0.7076
accuracy: 0.5370 - val_loss: 1.0933 - val_accuracy: 0.7084
Epoch 9/50
accuracy: 0.5549 - val_loss: 1.0539 - val_accuracy: 0.7076
Epoch 10/50
4853/4853 [=============== ] - Os 33us/step - loss: 1.1907 -
accuracy: 0.5735 - val loss: 1.0189 - val accuracy: 0.7084
Epoch 11/50
accuracy: 0.5862 - val_loss: 0.9771 - val_accuracy: 0.7076
Epoch 12/50
accuracy: 0.5899 - val_loss: 0.9480 - val_accuracy: 0.7076
Epoch 13/50
accuracy: 0.5920 - val_loss: 0.9199 - val_accuracy: 0.7076
Epoch 14/50
```

```
accuracy: 0.5902 - val_loss: 0.8922 - val_accuracy: 0.7076
Epoch 15/50
accuracy: 0.6073 - val_loss: 0.8582 - val_accuracy: 0.7076
Epoch 16/50
accuracy: 0.6042 - val_loss: 0.8361 - val_accuracy: 0.7084
Epoch 17/50
accuracy: 0.6052 - val_loss: 0.8111 - val_accuracy: 0.7076
Epoch 18/50
accuracy: 0.6155 - val_loss: 0.7938 - val_accuracy: 0.7076
Epoch 19/50
4853/4853 [=============== ] - Os 35us/step - loss: 1.0039 -
accuracy: 0.6211 - val_loss: 0.7817 - val_accuracy: 0.7084
Epoch 20/50
accuracy: 0.6227 - val_loss: 0.7638 - val_accuracy: 0.7084
Epoch 21/50
accuracy: 0.6295 - val_loss: 0.7502 - val_accuracy: 0.7076
Epoch 22/50
4853/4853 [=============== ] - Os 50us/step - loss: 0.9605 -
accuracy: 0.6279 - val_loss: 0.7345 - val_accuracy: 0.7084
Epoch 23/50
accuracy: 0.6371 - val_loss: 0.7220 - val_accuracy: 0.7084
4853/4853 [=============== ] - Os 26us/step - loss: 0.9403 -
accuracy: 0.6400 - val_loss: 0.7211 - val_accuracy: 0.7084
Epoch 25/50
accuracy: 0.6474 - val_loss: 0.7079 - val_accuracy: 0.7084
Epoch 26/50
4853/4853 [=============== ] - Os 34us/step - loss: 0.9362 -
accuracy: 0.6357 - val loss: 0.6999 - val accuracy: 0.7084
Epoch 27/50
accuracy: 0.6410 - val_loss: 0.6828 - val_accuracy: 0.7092
Epoch 28/50
accuracy: 0.6610 - val_loss: 0.6733 - val_accuracy: 0.7100
Epoch 29/50
accuracy: 0.6520 - val_loss: 0.6654 - val_accuracy: 0.7084
Epoch 30/50
```

```
accuracy: 0.6703 - val_loss: 0.6579 - val_accuracy: 0.7109
Epoch 31/50
accuracy: 0.6606 - val_loss: 0.6528 - val_accuracy: 0.7100
Epoch 32/50
accuracy: 0.6654 - val_loss: 0.6461 - val_accuracy: 0.7109
Epoch 33/50
accuracy: 0.6722 - val_loss: 0.6351 - val_accuracy: 0.8674
Epoch 34/50
accuracy: 0.6823 - val_loss: 0.6294 - val_accuracy: 0.7191
Epoch 35/50
accuracy: 0.6872 - val_loss: 0.6271 - val_accuracy: 0.8682
Epoch 36/50
accuracy: 0.6938 - val_loss: 0.6197 - val_accuracy: 0.8682
Epoch 37/50
accuracy: 0.6948 - val_loss: 0.6157 - val_accuracy: 0.8674
Epoch 38/50
4853/4853 [=============== ] - Os 33us/step - loss: 0.8324 -
accuracy: 0.7107 - val_loss: 0.6118 - val_accuracy: 0.8674
Epoch 39/50
accuracy: 0.7039 - val_loss: 0.5988 - val_accuracy: 0.8674
4853/4853 [=============== ] - Os 53us/step - loss: 0.8422 -
accuracy: 0.7189 - val_loss: 0.5915 - val_accuracy: 0.8674
Epoch 41/50
accuracy: 0.7226 - val_loss: 0.5833 - val_accuracy: 0.8674
Epoch 42/50
4853/4853 [=============== ] - Os 27us/step - loss: 0.8314 -
accuracy: 0.7268 - val loss: 0.5802 - val accuracy: 0.8674
Epoch 43/50
accuracy: 0.7253 - val_loss: 0.5850 - val_accuracy: 0.8600
Epoch 44/50
accuracy: 0.7373 - val_loss: 0.5689 - val_accuracy: 0.8674
Epoch 45/50
accuracy: 0.7350 - val_loss: 0.5743 - val_accuracy: 0.8649
Epoch 46/50
```



```
Squential with Kfold CV:

Training for fold 3 ...

Adding layer 1:

Adding layer 2:

Adding layer 3:

Model: "sequential_33"

Layer (type)

Output Shape

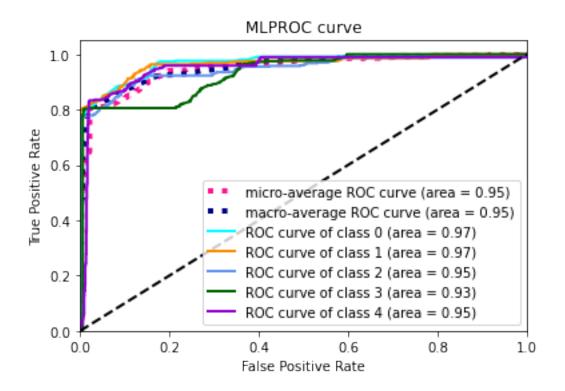
Param #
```

```
______
dense_176 (Dense)
             (None, 48)
                          192
dense_177 (Dense)
          (None, 24)
                          1176
      -----
dropout_112 (Dropout) (None, 24)
_____
dense_178 (Dense) (None, 12)
                          300
______
dropout_113 (Dropout) (None, 12)
   ._____
dense_179 (Dense) (None, 6)
                          78
dropout_114 (Dropout) (None, 6)
_____
dense_180 (Dense) (None, 5)
                          35
______
Total params: 1,781
Trainable params: 1,781
Non-trainable params: 0
Train on 4854 samples, validate on 1213 samples
Epoch 1/50
accuracy: 0.1448 - val_loss: 1.5211 - val_accuracy: 0.1022
Epoch 2/50
accuracy: 0.2340 - val_loss: 1.4780 - val_accuracy: 0.5243
accuracy: 0.3068 - val_loss: 1.4367 - val_accuracy: 0.6727
accuracy: 0.3754 - val_loss: 1.3880 - val_accuracy: 0.6719
Epoch 5/50
4854/4854 [=============== ] - Os 38us/step - loss: 1.4240 -
accuracy: 0.4178 - val loss: 1.3549 - val accuracy: 0.6777
Epoch 6/50
accuracy: 0.4596 - val_loss: 1.3158 - val_accuracy: 0.6752
Epoch 7/50
accuracy: 0.4813 - val_loss: 1.2781 - val_accuracy: 0.7964
Epoch 8/50
accuracy: 0.5000 - val_loss: 1.2359 - val_accuracy: 0.6834
Epoch 9/50
```

```
accuracy: 0.5072 - val_loss: 1.1930 - val_accuracy: 0.6810
Epoch 10/50
accuracy: 0.5200 - val_loss: 1.1550 - val_accuracy: 0.6810
Epoch 11/50
accuracy: 0.5383 - val_loss: 1.1292 - val_accuracy: 0.6810
Epoch 12/50
accuracy: 0.5400 - val_loss: 1.0978 - val_accuracy: 0.6785
Epoch 13/50
accuracy: 0.5455 - val_loss: 1.0690 - val_accuracy: 0.6826
Epoch 14/50
accuracy: 0.5612 - val_loss: 1.0328 - val_accuracy: 0.6785
Epoch 15/50
accuracy: 0.5682 - val_loss: 1.0001 - val_accuracy: 0.6785
Epoch 16/50
accuracy: 0.5688 - val_loss: 0.9693 - val_accuracy: 0.6785
Epoch 17/50
accuracy: 0.5820 - val_loss: 0.9463 - val_accuracy: 0.6801
Epoch 18/50
accuracy: 0.5906 - val_loss: 0.9202 - val_accuracy: 0.6785
Epoch 19/50
accuracy: 0.5960 - val_loss: 0.8962 - val_accuracy: 0.6801
Epoch 20/50
accuracy: 0.5989 - val_loss: 0.8771 - val_accuracy: 0.6785
Epoch 21/50
accuracy: 0.6063 - val loss: 0.8629 - val accuracy: 0.6818
Epoch 22/50
accuracy: 0.6055 - val_loss: 0.8459 - val_accuracy: 0.6801
Epoch 23/50
accuracy: 0.6224 - val_loss: 0.8301 - val_accuracy: 0.7519
Epoch 24/50
accuracy: 0.6339 - val_loss: 0.8118 - val_accuracy: 0.6892
Epoch 25/50
```

```
accuracy: 0.6345 - val_loss: 0.8010 - val_accuracy: 0.6826
Epoch 26/50
accuracy: 0.6405 - val_loss: 0.7891 - val_accuracy: 0.6834
Epoch 27/50
accuracy: 0.6382 - val_loss: 0.7796 - val_accuracy: 0.6818
Epoch 28/50
accuracy: 0.6397 - val_loss: 0.7694 - val_accuracy: 0.6966
Epoch 29/50
accuracy: 0.6452 - val_loss: 0.7611 - val_accuracy: 0.7007
Epoch 30/50
accuracy: 0.6590 - val_loss: 0.7568 - val_accuracy: 0.8285
Epoch 31/50
accuracy: 0.6562 - val_loss: 0.7460 - val_accuracy: 0.6999
Epoch 32/50
accuracy: 0.6539 - val_loss: 0.7372 - val_accuracy: 0.6826
Epoch 33/50
accuracy: 0.6568 - val_loss: 0.7281 - val_accuracy: 0.7560
Epoch 34/50
accuracy: 0.6607 - val_loss: 0.7175 - val_accuracy: 0.7049
accuracy: 0.6733 - val_loss: 0.7079 - val_accuracy: 0.6958
Epoch 36/50
accuracy: 0.6677 - val_loss: 0.7068 - val_accuracy: 0.8508
Epoch 37/50
accuracy: 0.7015 - val loss: 0.6951 - val accuracy: 0.7815
Epoch 38/50
accuracy: 0.6854 - val_loss: 0.6876 - val_accuracy: 0.8203
Epoch 39/50
accuracy: 0.6902 - val_loss: 0.6813 - val_accuracy: 0.7865
Epoch 40/50
accuracy: 0.6877 - val_loss: 0.6753 - val_accuracy: 0.7741
Epoch 41/50
```

```
accuracy: 0.6893 - val_loss: 0.6746 - val_accuracy: 0.8277
Epoch 42/50
accuracy: 0.6963 - val_loss: 0.6653 - val_accuracy: 0.8376
Epoch 43/50
accuracy: 0.6943 - val_loss: 0.6584 - val_accuracy: 0.7675
Epoch 44/50
accuracy: 0.6924 - val_loss: 0.6542 - val_accuracy: 0.8392
Epoch 45/50
accuracy: 0.6930 - val_loss: 0.6589 - val_accuracy: 0.8508
Epoch 46/50
accuracy: 0.7011 - val_loss: 0.6581 - val_accuracy: 0.8417
Epoch 47/50
accuracy: 0.7134 - val_loss: 0.6431 - val_accuracy: 0.8442
Epoch 48/50
accuracy: 0.7023 - val_loss: 0.6418 - val_accuracy: 0.8467
Epoch 49/50
accuracy: 0.7083 - val_loss: 0.6327 - val_accuracy: 0.8508
Epoch 50/50
accuracy: 0.7260 - val_loss: 0.6379 - val_accuracy: 0.8483
2023/2023 [============ ] - 0s 88us/step
```



Squential with Kfold  ${\tt CV}$ :

Training for fold 4  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

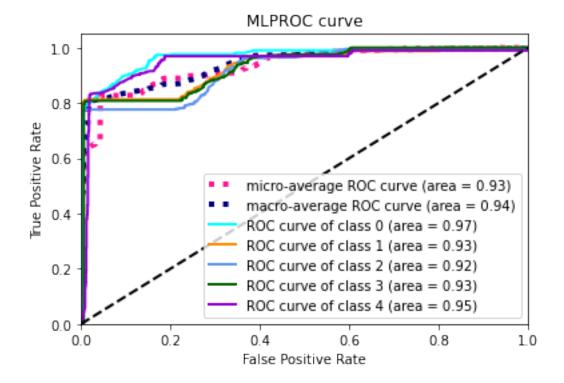
Model: "sequential\_34"

Layer (type)	Output Shape	Param #
dense_181 (Dense)	(None, 48)	192
dense_182 (Dense)	(None, 24)	1176
dropout_115 (Dropout)	(None, 24)	0
dense_183 (Dense)	(None, 12)	300
dropout_116 (Dropout)	(None, 12)	0
dense_184 (Dense)	(None, 6)	78
dropout_117 (Dropout)	(None, 6)	0
dropout_116 (Dropout) dense_184 (Dense)	(None, 12)	0 78

```
dense_185 (Dense)
              (None, 5)
                           35
______
Total params: 1,781
Trainable params: 1,781
Non-trainable params: 0
Train on 4854 samples, validate on 1213 samples
Epoch 1/50
accuracy: 0.2188 - val_loss: 1.5573 - val_accuracy: 0.3570
Epoch 2/50
accuracy: 0.2936 - val_loss: 1.5127 - val_accuracy: 0.5161
Epoch 3/50
accuracy: 0.3603 - val_loss: 1.4570 - val_accuracy: 0.6702
Epoch 4/50
accuracy: 0.4063 - val_loss: 1.4097 - val_accuracy: 0.5556
Epoch 5/50
accuracy: 0.4236 - val_loss: 1.3669 - val_accuracy: 0.5194
Epoch 6/50
accuracy: 0.4501 - val_loss: 1.3231 - val_accuracy: 0.5169
Epoch 7/50
accuracy: 0.4771 - val_loss: 1.2737 - val_accuracy: 0.6801
accuracy: 0.4893 - val_loss: 1.2286 - val_accuracy: 0.6892
accuracy: 0.5068 - val_loss: 1.1843 - val_accuracy: 0.6801
Epoch 10/50
accuracy: 0.5126 - val loss: 1.1416 - val accuracy: 0.6801
Epoch 11/50
accuracy: 0.5157 - val_loss: 1.1082 - val_accuracy: 0.6810
Epoch 12/50
accuracy: 0.5299 - val_loss: 1.0706 - val_accuracy: 0.6801
Epoch 13/50
accuracy: 0.5352 - val_loss: 1.0415 - val_accuracy: 0.6810
Epoch 14/50
```

```
accuracy: 0.5525 - val_loss: 1.0190 - val_accuracy: 0.6810
Epoch 15/50
accuracy: 0.5606 - val_loss: 0.9975 - val_accuracy: 0.6801
Epoch 16/50
accuracy: 0.5816 - val_loss: 0.9792 - val_accuracy: 0.6793
Epoch 17/50
accuracy: 0.5859 - val_loss: 0.9635 - val_accuracy: 0.6801
Epoch 18/50
accuracy: 0.5921 - val_loss: 0.9474 - val_accuracy: 0.6785
Epoch 19/50
accuracy: 0.6131 - val_loss: 0.9339 - val_accuracy: 0.6793
Epoch 20/50
accuracy: 0.6108 - val_loss: 0.9219 - val_accuracy: 0.6793
Epoch 21/50
accuracy: 0.6205 - val_loss: 0.9125 - val_accuracy: 0.6801
Epoch 22/50
accuracy: 0.6286 - val_loss: 0.9026 - val_accuracy: 0.6793
Epoch 23/50
accuracy: 0.6286 - val_loss: 0.8919 - val_accuracy: 0.6801
accuracy: 0.6298 - val_loss: 0.8812 - val_accuracy: 0.6801
Epoch 25/50
accuracy: 0.6409 - val_loss: 0.8747 - val_accuracy: 0.6801
Epoch 26/50
accuracy: 0.6457 - val_loss: 0.8658 - val_accuracy: 0.6801
Epoch 27/50
accuracy: 0.6593 - val_loss: 0.8587 - val_accuracy: 0.6801
Epoch 28/50
accuracy: 0.6454 - val_loss: 0.8487 - val_accuracy: 0.6801
Epoch 29/50
accuracy: 0.6457 - val_loss: 0.8385 - val_accuracy: 0.6810
Epoch 30/50
```

```
accuracy: 0.6508 - val_loss: 0.8286 - val_accuracy: 0.6818
Epoch 31/50
accuracy: 0.6603 - val_loss: 0.8218 - val_accuracy: 0.6810
Epoch 32/50
accuracy: 0.6712 - val_loss: 0.8162 - val_accuracy: 0.7370
Epoch 33/50
accuracy: 0.6912 - val_loss: 0.8098 - val_accuracy: 0.7378
Epoch 34/50
accuracy: 0.6852 - val_loss: 0.8027 - val_accuracy: 0.7683
Epoch 35/50
accuracy: 0.6976 - val_loss: 0.7948 - val_accuracy: 0.6908
Epoch 36/50
accuracy: 0.6840 - val_loss: 0.7873 - val_accuracy: 0.7980
Epoch 37/50
accuracy: 0.6976 - val_loss: 0.7827 - val_accuracy: 0.8335
Epoch 38/50
accuracy: 0.7048 - val_loss: 0.7767 - val_accuracy: 0.8326
Epoch 39/50
accuracy: 0.7105 - val_loss: 0.7719 - val_accuracy: 0.8318
accuracy: 0.7279 - val_loss: 0.7644 - val_accuracy: 0.8318
Epoch 41/50
accuracy: 0.7367 - val_loss: 0.7600 - val_accuracy: 0.8318
Epoch 42/50
accuracy: 0.7349 - val loss: 0.7561 - val accuracy: 0.8335
Epoch 43/50
accuracy: 0.7501 - val_loss: 0.7539 - val_accuracy: 0.8335
Epoch 44/50
accuracy: 0.7528 - val_loss: 0.7475 - val_accuracy: 0.8335
Epoch 45/50
accuracy: 0.7678 - val_loss: 0.7414 - val_accuracy: 0.8335
Epoch 46/50
```



```
Squential with Kfold CV:

Training for fold 5 ...

Adding layer 1:

Adding layer 2:

Adding layer 3:

Model: "sequential_35"

Layer (type)

Output Shape

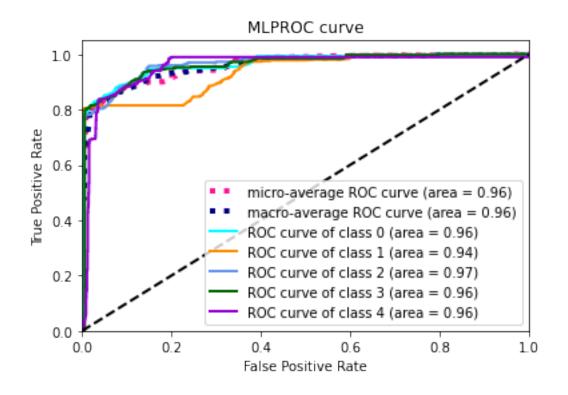
Param #
```

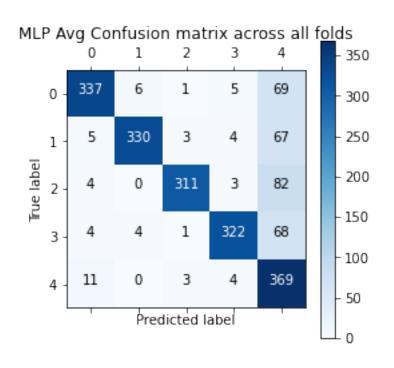
```
______
dense_186 (Dense)
            (None, 48)
                        192
dense_187 (Dense)
         (None, 24)
                        1176
     _____
dropout_118 (Dropout) (None, 24)
_____
dense_188 (Dense) (None, 12)
                        300
______
dropout_119 (Dropout) (None, 12)
-----
dense_189 (Dense) (None, 6)
                        78
_____
dropout_120 (Dropout) (None, 6)
_____
dense_190 (Dense) (None, 5)
                        35
______
Total params: 1,781
Trainable params: 1,781
Non-trainable params: 0
Train on 4854 samples, validate on 1213 samples
Epoch 1/50
accuracy: 0.2382 - val_loss: 1.4726 - val_accuracy: 0.1797
Epoch 2/50
accuracy: 0.3164 - val_loss: 1.4051 - val_accuracy: 0.3528
accuracy: 0.3793 - val_loss: 1.3582 - val_accuracy: 0.5268
Epoch 4/50
accuracy: 0.4157 - val_loss: 1.3091 - val_accuracy: 0.5243
Epoch 5/50
accuracy: 0.4330 - val loss: 1.2713 - val accuracy: 0.5251
Epoch 6/50
accuracy: 0.4557 - val_loss: 1.2326 - val_accuracy: 0.5251
Epoch 7/50
accuracy: 0.4639 - val_loss: 1.1982 - val_accuracy: 0.5251
Epoch 8/50
accuracy: 0.4813 - val_loss: 1.1669 - val_accuracy: 0.5251
Epoch 9/50
```

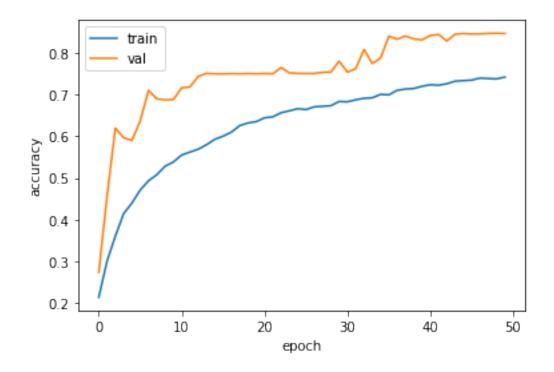
```
accuracy: 0.5037 - val_loss: 1.1302 - val_accuracy: 0.5251
Epoch 10/50
accuracy: 0.5169 - val_loss: 1.1065 - val_accuracy: 0.5251
Epoch 11/50
accuracy: 0.5408 - val_loss: 1.0766 - val_accuracy: 0.6686
Epoch 12/50
accuracy: 0.5476 - val_loss: 1.0538 - val_accuracy: 0.6843
Epoch 13/50
accuracy: 0.5567 - val_loss: 1.0381 - val_accuracy: 0.8046
Epoch 14/50
accuracy: 0.5754 - val_loss: 1.0089 - val_accuracy: 0.8425
Epoch 15/50
accuracy: 0.6086 - val_loss: 0.9800 - val_accuracy: 0.8425
Epoch 16/50
accuracy: 0.6154 - val_loss: 0.9649 - val_accuracy: 0.8425
Epoch 17/50
accuracy: 0.6257 - val_loss: 0.9390 - val_accuracy: 0.8417
Epoch 18/50
accuracy: 0.6471 - val_loss: 0.9146 - val_accuracy: 0.8417
Epoch 19/50
accuracy: 0.6539 - val_loss: 0.8998 - val_accuracy: 0.8417
Epoch 20/50
accuracy: 0.6549 - val_loss: 0.8748 - val_accuracy: 0.8409
Epoch 21/50
accuracy: 0.6685 - val loss: 0.8523 - val accuracy: 0.8409
Epoch 22/50
accuracy: 0.6704 - val_loss: 0.8316 - val_accuracy: 0.8409
Epoch 23/50
accuracy: 0.6817 - val_loss: 0.8093 - val_accuracy: 0.8417
Epoch 24/50
accuracy: 0.6850 - val_loss: 0.7865 - val_accuracy: 0.8417
Epoch 25/50
```

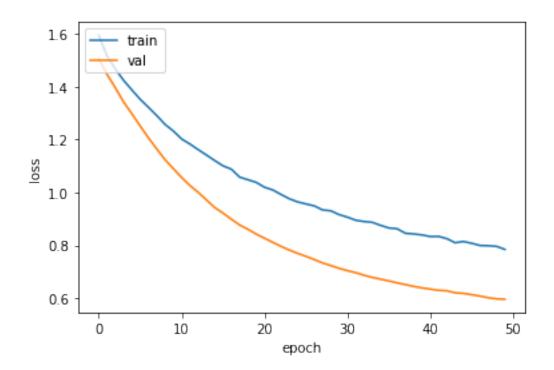
```
accuracy: 0.6879 - val_loss: 0.7684 - val_accuracy: 0.8409
Epoch 26/50
accuracy: 0.6860 - val_loss: 0.7523 - val_accuracy: 0.8409
Epoch 27/50
accuracy: 0.6922 - val_loss: 0.7345 - val_accuracy: 0.8409
Epoch 28/50
accuracy: 0.6885 - val_loss: 0.7113 - val_accuracy: 0.8409
Epoch 29/50
accuracy: 0.6984 - val_loss: 0.6980 - val_accuracy: 0.8409
Epoch 30/50
accuracy: 0.7079 - val_loss: 0.6781 - val_accuracy: 0.8409
Epoch 31/50
accuracy: 0.7126 - val_loss: 0.6643 - val_accuracy: 0.8409
Epoch 32/50
accuracy: 0.7178 - val_loss: 0.6521 - val_accuracy: 0.8417
Epoch 33/50
accuracy: 0.7025 - val_loss: 0.6390 - val_accuracy: 0.8417
Epoch 34/50
accuracy: 0.7132 - val_loss: 0.6283 - val_accuracy: 0.8417
accuracy: 0.7159 - val_loss: 0.6208 - val_accuracy: 0.8434
Epoch 36/50
4854/4854 [=============== ] - Os 31us/step - loss: 0.8320 -
accuracy: 0.7217 - val_loss: 0.6101 - val_accuracy: 0.8442
Epoch 37/50
accuracy: 0.7215 - val_loss: 0.6007 - val_accuracy: 0.8442
Epoch 38/50
accuracy: 0.7274 - val_loss: 0.5902 - val_accuracy: 0.8442
Epoch 39/50
accuracy: 0.7295 - val_loss: 0.5873 - val_accuracy: 0.8442
Epoch 40/50
accuracy: 0.7252 - val_loss: 0.5811 - val_accuracy: 0.8450
Epoch 41/50
```

```
accuracy: 0.7262 - val_loss: 0.5754 - val_accuracy: 0.8450
Epoch 42/50
accuracy: 0.7225 - val_loss: 0.5729 - val_accuracy: 0.8442
Epoch 43/50
accuracy: 0.7260 - val_loss: 0.5729 - val_accuracy: 0.8450
Epoch 44/50
accuracy: 0.7344 - val_loss: 0.5654 - val_accuracy: 0.8450
Epoch 45/50
accuracy: 0.7301 - val_loss: 0.5553 - val_accuracy: 0.8450
Epoch 46/50
accuracy: 0.7359 - val_loss: 0.5508 - val_accuracy: 0.8450
Epoch 47/50
accuracy: 0.7353 - val_loss: 0.5463 - val_accuracy: 0.8450
Epoch 48/50
accuracy: 0.7373 - val_loss: 0.5383 - val_accuracy: 0.8450
Epoch 49/50
accuracy: 0.7349 - val_loss: 0.5365 - val_accuracy: 0.8450
Epoch 50/50
accuracy: 0.7384 - val_loss: 0.5347 - val_accuracy: 0.8450
2023/2023 [============ ] - 0s 88us/step
```









Average scores for pesudo test set across all folds: > Accuracy: 0.8255066752433777 (+- 0.001210805477908015)

> Loss: 0.63940720067894

## 

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Squential with Kfold CV:

Training for fold 1 ...

Adding layer 1:

Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_36"

Layer (type)	Output	Shape	Param #
dense_191 (Dense)	(None,	96)	384
dense_192 (Dense)	(None,	48)	4656
dropout_121 (Dropout)	(None,	48)	0
dense_193 (Dense)	(None,	24)	1176
dropout_122 (Dropout)	(None,	24)	0
dense_194 (Dense)	(None,	12)	300
dropout_123 (Dropout)	(None,	12)	0
dense_195 (Dense)	(None,	6)	78
dropout_124 (Dropout)	(None,	6)	0
dense_196 (Dense)	(None,	5)	35
Tatal manager 6 600			

Total params: 6,629 Trainable params: 6,629 Non-trainable params: 0

-----

Train on 4853 samples, validate on 1214 samples

Epoch 1/50

4853/4853 [============= ] - 4s 924us/step - loss: 1.5452 -

accuracy: 0.3179 - val\_loss: 1.3924 - val\_accuracy: 0.7636

Epoch 2/50

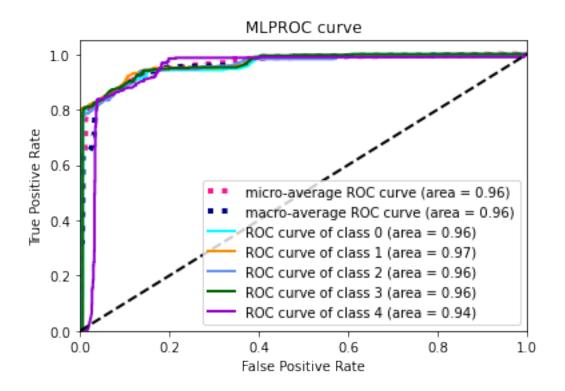
accuracy: 0.5263 - val\_loss: 1.3136 - val\_accuracy: 0.8460

Epoch 3/50

```
accuracy: 0.6351 - val_loss: 1.2352 - val_accuracy: 0.8460
Epoch 4/50
4853/4853 [=============== ] - Os 43us/step - loss: 1.3049 -
accuracy: 0.6693 - val_loss: 1.1787 - val_accuracy: 0.8460
Epoch 5/50
accuracy: 0.7057 - val_loss: 1.1164 - val_accuracy: 0.8460
Epoch 6/50
accuracy: 0.7373 - val_loss: 1.0607 - val_accuracy: 0.8468
Epoch 7/50
accuracy: 0.7667 - val_loss: 1.0039 - val_accuracy: 0.8468
accuracy: 0.7657 - val_loss: 0.9528 - val_accuracy: 0.8468
Epoch 9/50
accuracy: 0.7744 - val_loss: 0.8992 - val_accuracy: 0.8468
Epoch 10/50
4853/4853 [=============== ] - Os 42us/step - loss: 1.0160 -
accuracy: 0.7725 - val_loss: 0.8519 - val_accuracy: 0.8468
Epoch 11/50
accuracy: 0.7799 - val_loss: 0.8168 - val_accuracy: 0.8468
Epoch 12/50
accuracy: 0.7888 - val_loss: 0.7836 - val_accuracy: 0.8468
Epoch 13/50
accuracy: 0.7882 - val_loss: 0.7561 - val_accuracy: 0.8468
Epoch 14/50
accuracy: 0.7830 - val_loss: 0.7310 - val_accuracy: 0.8476
Epoch 15/50
accuracy: 0.7915 - val_loss: 0.7105 - val_accuracy: 0.8476
Epoch 16/50
accuracy: 0.7894 - val_loss: 0.6931 - val_accuracy: 0.8476
Epoch 17/50
4853/4853 [=============== ] - Os 49us/step - loss: 0.8161 -
accuracy: 0.7968 - val_loss: 0.6773 - val_accuracy: 0.8484
Epoch 18/50
accuracy: 0.8018 - val_loss: 0.6630 - val_accuracy: 0.8484
Epoch 19/50
```

```
accuracy: 0.8009 - val_loss: 0.6497 - val_accuracy: 0.8493
Epoch 20/50
4853/4853 [============ ] - Os 49us/step - loss: 0.7674 -
accuracy: 0.8028 - val_loss: 0.6374 - val_accuracy: 0.8493
Epoch 21/50
4853/4853 [=============== ] - Os 43us/step - loss: 0.7547 -
accuracy: 0.8061 - val_loss: 0.6246 - val_accuracy: 0.8493
Epoch 22/50
accuracy: 0.8053 - val_loss: 0.6143 - val_accuracy: 0.8493
Epoch 23/50
accuracy: 0.8040 - val_loss: 0.6067 - val_accuracy: 0.8493
Epoch 24/50
accuracy: 0.8096 - val_loss: 0.5992 - val_accuracy: 0.8493
Epoch 25/50
4853/4853 [=============== ] - Os 47us/step - loss: 0.7051 -
accuracy: 0.8036 - val_loss: 0.5908 - val_accuracy: 0.8493
Epoch 26/50
4853/4853 [=============== ] - Os 50us/step - loss: 0.6943 -
accuracy: 0.8160 - val_loss: 0.5823 - val_accuracy: 0.8493
Epoch 27/50
accuracy: 0.8121 - val_loss: 0.5791 - val_accuracy: 0.8493
Epoch 28/50
4853/4853 [============== ] - Os 40us/step - loss: 0.6854 -
accuracy: 0.8113 - val_loss: 0.5716 - val_accuracy: 0.8493
Epoch 29/50
accuracy: 0.8123 - val_loss: 0.5630 - val_accuracy: 0.8493
Epoch 30/50
accuracy: 0.8191 - val loss: 0.5618 - val accuracy: 0.8493
Epoch 31/50
accuracy: 0.8160 - val_loss: 0.5584 - val_accuracy: 0.8493
Epoch 32/50
4853/4853 [=============== ] - Os 45us/step - loss: 0.6460 -
accuracy: 0.8236 - val_loss: 0.5558 - val_accuracy: 0.8493
Epoch 33/50
accuracy: 0.8178 - val_loss: 0.5430 - val_accuracy: 0.8501
Epoch 34/50
accuracy: 0.8220 - val_loss: 0.5451 - val_accuracy: 0.8493
Epoch 35/50
```

```
accuracy: 0.8218 - val_loss: 0.5327 - val_accuracy: 0.8509
Epoch 36/50
4853/4853 [============ ] - Os 36us/step - loss: 0.6339 -
accuracy: 0.8191 - val_loss: 0.5348 - val_accuracy: 0.8501
Epoch 37/50
accuracy: 0.8246 - val_loss: 0.5326 - val_accuracy: 0.8501
Epoch 38/50
accuracy: 0.8216 - val_loss: 0.5284 - val_accuracy: 0.8501
Epoch 39/50
accuracy: 0.8294 - val_loss: 0.5168 - val_accuracy: 0.8509
Epoch 40/50
accuracy: 0.8234 - val_loss: 0.5127 - val_accuracy: 0.8501
Epoch 41/50
accuracy: 0.8281 - val_loss: 0.5100 - val_accuracy: 0.8509
Epoch 42/50
accuracy: 0.8218 - val_loss: 0.5061 - val_accuracy: 0.8509
Epoch 43/50
accuracy: 0.8298 - val_loss: 0.5141 - val_accuracy: 0.8484
Epoch 44/50
accuracy: 0.8257 - val_loss: 0.4999 - val_accuracy: 0.8509
Epoch 45/50
accuracy: 0.8257 - val_loss: 0.4983 - val_accuracy: 0.8517
Epoch 46/50
accuracy: 0.8349 - val_loss: 0.4956 - val_accuracy: 0.8509
Epoch 47/50
accuracy: 0.8271 - val_loss: 0.5143 - val_accuracy: 0.8402
Epoch 48/50
accuracy: 0.8261 - val_loss: 0.4934 - val_accuracy: 0.8501
Epoch 49/50
accuracy: 0.8302 - val_loss: 0.4899 - val_accuracy: 0.8509
Epoch 50/50
accuracy: 0.8296 - val_loss: 0.4874 - val_accuracy: 0.8493
2023/2023 [=========== ] - 0s 113us/step
```



Squential with Kfold CV:

Training for fold 2  $\dots$ 

Adding layer 1: Adding layer 2: Adding layer 3:

Adding layer 4:

Model: "sequential\_37"

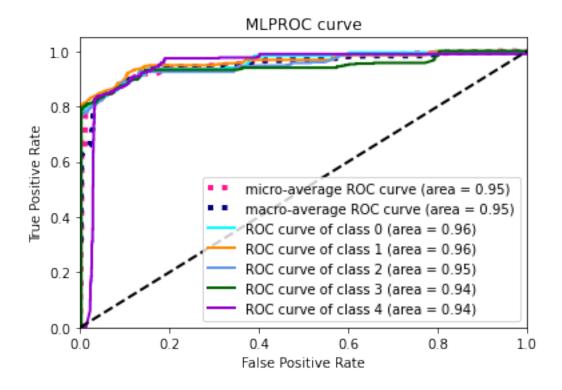
Layer (type)	Output Shape	Param #
dense_197 (Dense)	(None, 96)	384
dense_198 (Dense)	(None, 48)	4656
dropout_125 (Dropout)	(None, 48)	0
dense_199 (Dense)	(None, 24)	1176
dropout_126 (Dropout)	(None, 24)	0
dense_200 (Dense)	(None, 12)	300
dropout_127 (Dropout)	(None, 12)	0

```
dense_201 (Dense)
           (None, 6)
                            78
dropout_128 (Dropout) (None, 6)
                             Ω
_____
dense_202 (Dense) (None, 5)
______
Total params: 6,629
Trainable params: 6,629
Non-trainable params: 0
Train on 4853 samples, validate on 1214 samples
Epoch 1/50
accuracy: 0.4179 - val_loss: 1.2784 - val_accuracy: 0.6993
Epoch 2/50
accuracy: 0.5671 - val_loss: 1.1660 - val_accuracy: 0.8221
Epoch 3/50
accuracy: 0.5996 - val_loss: 1.0755 - val_accuracy: 0.6985
Epoch 4/50
accuracy: 0.6336 - val_loss: 1.0061 - val_accuracy: 0.7010
Epoch 5/50
accuracy: 0.6571 - val_loss: 0.9470 - val_accuracy: 0.7010
Epoch 6/50
accuracy: 0.6858 - val_loss: 0.9099 - val_accuracy: 0.7018
Epoch 7/50
accuracy: 0.6985 - val_loss: 0.8744 - val_accuracy: 0.8526
Epoch 8/50
accuracy: 0.7148 - val_loss: 0.8453 - val_accuracy: 0.8517
Epoch 9/50
accuracy: 0.7416 - val_loss: 0.8211 - val_accuracy: 0.8534
Epoch 10/50
accuracy: 0.7538 - val_loss: 0.7930 - val_accuracy: 0.8550
4853/4853 [============= ] - Os 68us/step - loss: 0.9233 -
accuracy: 0.7717 - val_loss: 0.7702 - val_accuracy: 0.8558
Epoch 12/50
accuracy: 0.7830 - val_loss: 0.7513 - val_accuracy: 0.8542
```

```
Epoch 13/50
accuracy: 0.7902 - val_loss: 0.7355 - val_accuracy: 0.8526
Epoch 14/50
accuracy: 0.8022 - val_loss: 0.7254 - val_accuracy: 0.8542
Epoch 15/50
accuracy: 0.8049 - val_loss: 0.7147 - val_accuracy: 0.8542
Epoch 16/50
4853/4853 [=============== ] - 0s 71us/step - loss: 0.8309 -
accuracy: 0.8020 - val_loss: 0.7029 - val_accuracy: 0.8575
Epoch 17/50
accuracy: 0.8088 - val_loss: 0.6916 - val_accuracy: 0.8575
Epoch 18/50
accuracy: 0.8139 - val_loss: 0.6766 - val_accuracy: 0.8575
Epoch 19/50
accuracy: 0.8166 - val_loss: 0.6673 - val_accuracy: 0.8575
Epoch 20/50
accuracy: 0.8195 - val_loss: 0.6610 - val_accuracy: 0.8575
Epoch 21/50
accuracy: 0.8236 - val_loss: 0.6491 - val_accuracy: 0.8575
Epoch 22/50
accuracy: 0.8255 - val_loss: 0.6428 - val_accuracy: 0.8575
Epoch 23/50
4853/4853 [============== ] - Os 54us/step - loss: 0.7507 -
accuracy: 0.8211 - val_loss: 0.6303 - val_accuracy: 0.8583
Epoch 24/50
accuracy: 0.8242 - val_loss: 0.6304 - val_accuracy: 0.8583
Epoch 25/50
accuracy: 0.8312 - val_loss: 0.6168 - val_accuracy: 0.8583
Epoch 26/50
accuracy: 0.8251 - val_loss: 0.6065 - val_accuracy: 0.8583
Epoch 27/50
accuracy: 0.8300 - val_loss: 0.5983 - val_accuracy: 0.8575
Epoch 28/50
accuracy: 0.8310 - val_loss: 0.5916 - val_accuracy: 0.8583
```

```
Epoch 29/50
accuracy: 0.8317 - val_loss: 0.5972 - val_accuracy: 0.8583
Epoch 30/50
accuracy: 0.8308 - val_loss: 0.5817 - val_accuracy: 0.8583
Epoch 31/50
accuracy: 0.8345 - val_loss: 0.5775 - val_accuracy: 0.8583
Epoch 32/50
accuracy: 0.8372 - val_loss: 0.5716 - val_accuracy: 0.8583
Epoch 33/50
accuracy: 0.8358 - val_loss: 0.5633 - val_accuracy: 0.8583
Epoch 34/50
accuracy: 0.8349 - val_loss: 0.5648 - val_accuracy: 0.8583
Epoch 35/50
accuracy: 0.8349 - val_loss: 0.5523 - val_accuracy: 0.8583
Epoch 36/50
4853/4853 [============== ] - Os 43us/step - loss: 0.6384 -
accuracy: 0.8393 - val_loss: 0.5479 - val_accuracy: 0.8583
Epoch 37/50
4853/4853 [============== ] - Os 49us/step - loss: 0.6344 -
accuracy: 0.8339 - val_loss: 0.5505 - val_accuracy: 0.8583
Epoch 38/50
accuracy: 0.8331 - val_loss: 0.5402 - val_accuracy: 0.8583
Epoch 39/50
accuracy: 0.8345 - val_loss: 0.5339 - val_accuracy: 0.8583
Epoch 40/50
4853/4853 [=============== ] - Os 65us/step - loss: 0.6194 -
accuracy: 0.8356 - val_loss: 0.5421 - val_accuracy: 0.8583
Epoch 41/50
accuracy: 0.8393 - val_loss: 0.5381 - val_accuracy: 0.8583
Epoch 42/50
accuracy: 0.8360 - val_loss: 0.5231 - val_accuracy: 0.8583
Epoch 43/50
accuracy: 0.8339 - val_loss: 0.5250 - val_accuracy: 0.8583
Epoch 44/50
accuracy: 0.8337 - val_loss: 0.5183 - val_accuracy: 0.8583
```

```
Epoch 45/50
accuracy: 0.8382 - val_loss: 0.5204 - val_accuracy: 0.8583
Epoch 46/50
accuracy: 0.8399 - val_loss: 0.5216 - val_accuracy: 0.8583
Epoch 47/50
accuracy: 0.8349 - val_loss: 0.5094 - val_accuracy: 0.8575
Epoch 48/50
accuracy: 0.8407 - val_loss: 0.5064 - val_accuracy: 0.8583
Epoch 49/50
accuracy: 0.8370 - val_loss: 0.5074 - val_accuracy: 0.8583
Epoch 50/50
accuracy: 0.8391 - val_loss: 0.4953 - val_accuracy: 0.8575
2023/2023 [============ ] - 0s 103us/step
```



Squential with Kfold CV: Training for fold 3 ... Adding layer 1: Adding layer 2: Adding layer 3: Adding layer 4:

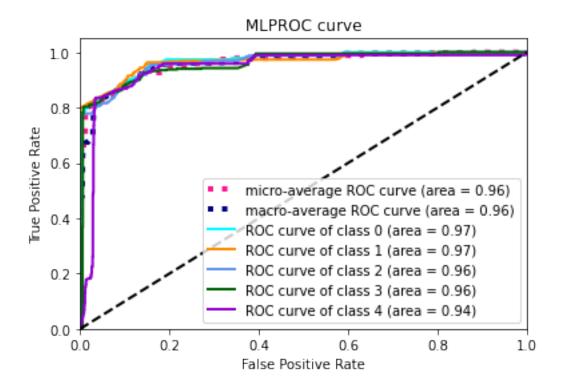
Model: "sequential\_38"

Model: "sequential_38"							
Layer (type)	_	Shape	Param #				
dense_203 (Dense)	(None,		384				
dense_204 (Dense)	(None,	48)	4656				
dropout_129 (Dropout)		48)	0				
dense_205 (Dense)	(None,		1176				
dropout_130 (Dropout)			0				
dense_206 (Dense)	(None,	12)	300				
dropout_131 (Dropout)	(None,		0				
dense_207 (Dense)			78				
dropout_132 (Dropout)			0				
dense_208 (Dense)	(None,	5)	35				
Total params: 6,629 Trainable params: 6,629 Non-trainable params: 0 Train on 4854 samples, valid	Trainable params: 6,629						
Epoch 1/50 4854/4854 [=========	======	=====] - 4s 840us/	step – loss	: 1.5568 -			
accuracy: 0.3105 - val_loss: Epoch 2/50			_				
4854/4854 [====================================			-	1.4183 -			
4854/4854 [====================================			-	1.3340 -			
4854/4854 [====================================			-	1.2613 -			
Epoch 5/50 4854/4854 [====================================			-	1.2006 -			

```
accuracy: 0.6316 - val_loss: 0.9914 - val_accuracy: 0.6843
Epoch 7/50
4854/4854 [============ ] - Os 42us/step - loss: 1.1049 -
accuracy: 0.6364 - val_loss: 0.9477 - val_accuracy: 0.6834
Epoch 8/50
accuracy: 0.6609 - val_loss: 0.9071 - val_accuracy: 0.6834
Epoch 9/50
accuracy: 0.6761 - val_loss: 0.8765 - val_accuracy: 0.6834
Epoch 10/50
accuracy: 0.6774 - val_loss: 0.8508 - val_accuracy: 0.6843
Epoch 11/50
accuracy: 0.6996 - val_loss: 0.8237 - val_accuracy: 0.8434
Epoch 12/50
accuracy: 0.7037 - val_loss: 0.7989 - val_accuracy: 0.8425
Epoch 13/50
accuracy: 0.7159 - val_loss: 0.7787 - val_accuracy: 0.8425
Epoch 14/50
4854/4854 [=============== ] - Os 41us/step - loss: 0.9137 -
accuracy: 0.7340 - val_loss: 0.7585 - val_accuracy: 0.8467
Epoch 15/50
accuracy: 0.7396 - val_loss: 0.7417 - val_accuracy: 0.8475
Epoch 16/50
accuracy: 0.7435 - val_loss: 0.7220 - val_accuracy: 0.8467
Epoch 17/50
accuracy: 0.7678 - val_loss: 0.7012 - val_accuracy: 0.8467
Epoch 18/50
accuracy: 0.7746 - val_loss: 0.6904 - val_accuracy: 0.8475
Epoch 19/50
accuracy: 0.7833 - val_loss: 0.6749 - val_accuracy: 0.8475
Epoch 20/50
accuracy: 0.7930 - val_loss: 0.6615 - val_accuracy: 0.8475
Epoch 21/50
accuracy: 0.7977 - val_loss: 0.6515 - val_accuracy: 0.8483
Epoch 22/50
```

```
accuracy: 0.8022 - val_loss: 0.6415 - val_accuracy: 0.8475
Epoch 23/50
accuracy: 0.8028 - val loss: 0.6342 - val accuracy: 0.8491
Epoch 24/50
accuracy: 0.8024 - val_loss: 0.6229 - val_accuracy: 0.8483
Epoch 25/50
accuracy: 0.8094 - val_loss: 0.6157 - val_accuracy: 0.8491
Epoch 26/50
accuracy: 0.8086 - val_loss: 0.6095 - val_accuracy: 0.8483
Epoch 27/50
accuracy: 0.8156 - val_loss: 0.6019 - val_accuracy: 0.8483
Epoch 28/50
accuracy: 0.8129 - val_loss: 0.5946 - val_accuracy: 0.8491
Epoch 29/50
accuracy: 0.8191 - val_loss: 0.5899 - val_accuracy: 0.8483
Epoch 30/50
accuracy: 0.8189 - val_loss: 0.5878 - val_accuracy: 0.8475
Epoch 31/50
accuracy: 0.8173 - val_loss: 0.5825 - val_accuracy: 0.8475
Epoch 32/50
accuracy: 0.8226 - val_loss: 0.5773 - val_accuracy: 0.8475
Epoch 33/50
accuracy: 0.8193 - val_loss: 0.5724 - val_accuracy: 0.8491
Epoch 34/50
accuracy: 0.8232 - val_loss: 0.5712 - val_accuracy: 0.8475
Epoch 35/50
accuracy: 0.8228 - val_loss: 0.5693 - val_accuracy: 0.8467
Epoch 36/50
accuracy: 0.8257 - val_loss: 0.5644 - val_accuracy: 0.8475
Epoch 37/50
accuracy: 0.8280 - val_loss: 0.5572 - val_accuracy: 0.8475
Epoch 38/50
```

```
0.82 - 0s 41us/step - loss: 0.6515 - accuracy: 0.8300 - val_loss: 0.5553 -
val_accuracy: 0.8483
Epoch 39/50
accuracy: 0.8263 - val_loss: 0.5485 - val_accuracy: 0.8491
Epoch 40/50
accuracy: 0.8274 - val_loss: 0.5440 - val_accuracy: 0.8483
Epoch 41/50
accuracy: 0.8269 - val_loss: 0.5402 - val_accuracy: 0.8491
Epoch 42/50
accuracy: 0.8292 - val_loss: 0.5471 - val_accuracy: 0.8483
Epoch 43/50
accuracy: 0.8300 - val_loss: 0.5437 - val_accuracy: 0.8483
Epoch 44/50
accuracy: 0.8350 - val_loss: 0.5354 - val_accuracy: 0.8483
Epoch 45/50
accuracy: 0.8267 - val_loss: 0.5361 - val_accuracy: 0.8483
Epoch 46/50
accuracy: 0.8282 - val_loss: 0.5271 - val_accuracy: 0.8491
Epoch 47/50
accuracy: 0.8346 - val_loss: 0.5153 - val_accuracy: 0.8491
Epoch 48/50
accuracy: 0.8321 - val_loss: 0.5259 - val_accuracy: 0.8483
Epoch 49/50
accuracy: 0.8315 - val_loss: 0.5144 - val_accuracy: 0.8500
Epoch 50/50
accuracy: 0.8282 - val_loss: 0.5168 - val_accuracy: 0.8500
2023/2023 [============ ] - 0s 104us/step
```



Squential with Kfold CV:

Training for fold  $4 \dots$ 

Adding layer 1:

Adding layer 2:

Adding layer 3:

Adding layer 4:

Model: "sequential\_39"

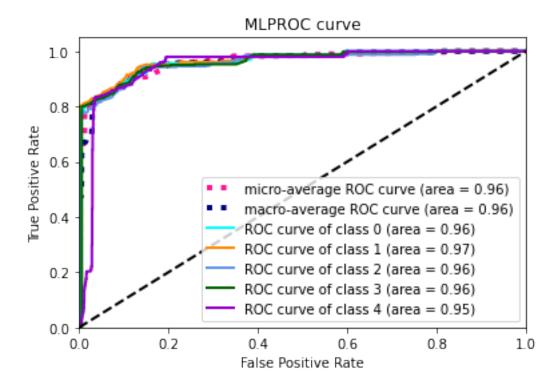
Layer (type)	Output Shape	Param #
dense_209 (Dense)	(None, 96)	384
dense_210 (Dense)	(None, 48)	4656
dropout_133 (Dropout)	(None, 48)	0
dense_211 (Dense)	(None, 24)	1176
dropout_134 (Dropout)	(None, 24)	0
dense_212 (Dense)	(None, 12)	300
dropout_135 (Dropout)	(None, 12)	0

```
dense_213 (Dense)
           (None, 6)
                              78
dropout_136 (Dropout) (None, 6)
                              Ω
_____
dense_214 (Dense) (None, 5)
______
Total params: 6,629
Trainable params: 6,629
Non-trainable params: 0
Train on 4854 samples, validate on 1213 samples
Epoch 1/50
accuracy: 0.3113 - val_loss: 1.3401 - val_accuracy: 0.5251
Epoch 2/50
accuracy: 0.4524 - val_loss: 1.2396 - val_accuracy: 0.7840
Epoch 3/50
accuracy: 0.5610 - val_loss: 1.1484 - val_accuracy: 0.8228
Epoch 4/50
accuracy: 0.6650 - val_loss: 1.0528 - val_accuracy: 0.8244
Epoch 5/50
accuracy: 0.7105 - val_loss: 0.9805 - val_accuracy: 0.8228
Epoch 6/50
4854/4854 [============= ] - Os 43us/step - loss: 1.0597 -
accuracy: 0.7305 - val_loss: 0.9237 - val_accuracy: 0.8252
Epoch 7/50
accuracy: 0.7505 - val_loss: 0.8795 - val_accuracy: 0.8261
Epoch 8/50
accuracy: 0.7575 - val_loss: 0.8409 - val_accuracy: 0.8252
Epoch 9/50
accuracy: 0.7748 - val_loss: 0.8056 - val_accuracy: 0.8252
Epoch 10/50
accuracy: 0.7851 - val_loss: 0.7840 - val_accuracy: 0.8252
4854/4854 [============= ] - Os 29us/step - loss: 0.8681 -
accuracy: 0.7857 - val_loss: 0.7577 - val_accuracy: 0.8252
Epoch 12/50
accuracy: 0.8000 - val_loss: 0.7420 - val_accuracy: 0.8244
```

```
Epoch 13/50
accuracy: 0.8004 - val_loss: 0.7293 - val_accuracy: 0.8244
Epoch 14/50
accuracy: 0.7998 - val_loss: 0.7177 - val_accuracy: 0.8261
Epoch 15/50
accuracy: 0.8115 - val_loss: 0.7067 - val_accuracy: 0.8261
Epoch 16/50
accuracy: 0.8177 - val_loss: 0.6929 - val_accuracy: 0.8252
Epoch 17/50
accuracy: 0.8123 - val_loss: 0.6844 - val_accuracy: 0.8236
Epoch 18/50
accuracy: 0.8179 - val_loss: 0.6742 - val_accuracy: 0.8244
Epoch 19/50
accuracy: 0.8226 - val_loss: 0.6687 - val_accuracy: 0.8236
Epoch 20/50
accuracy: 0.8276 - val_loss: 0.6599 - val_accuracy: 0.8244
Epoch 21/50
accuracy: 0.8267 - val_loss: 0.6520 - val_accuracy: 0.8236
Epoch 22/50
accuracy: 0.8321 - val_loss: 0.6451 - val_accuracy: 0.8252
Epoch 23/50
accuracy: 0.8340 - val_loss: 0.6406 - val_accuracy: 0.8252
Epoch 24/50
accuracy: 0.8288 - val_loss: 0.6335 - val_accuracy: 0.8244
Epoch 25/50
accuracy: 0.8352 - val_loss: 0.6264 - val_accuracy: 0.8244
Epoch 26/50
accuracy: 0.8329 - val_loss: 0.6248 - val_accuracy: 0.8244
Epoch 27/50
accuracy: 0.8360 - val_loss: 0.6183 - val_accuracy: 0.8236
Epoch 28/50
accuracy: 0.8381 - val_loss: 0.6126 - val_accuracy: 0.8244
```

```
Epoch 29/50
accuracy: 0.8385 - val_loss: 0.6014 - val_accuracy: 0.8244
Epoch 30/50
accuracy: 0.8407 - val_loss: 0.6021 - val_accuracy: 0.8236
Epoch 31/50
accuracy: 0.8381 - val_loss: 0.5916 - val_accuracy: 0.8244
Epoch 32/50
accuracy: 0.8407 - val_loss: 0.5846 - val_accuracy: 0.8244
Epoch 33/50
accuracy: 0.8424 - val_loss: 0.5867 - val_accuracy: 0.8261
Epoch 34/50
accuracy: 0.8430 - val_loss: 0.5788 - val_accuracy: 0.8244
Epoch 35/50
accuracy: 0.8463 - val_loss: 0.5702 - val_accuracy: 0.8252
Epoch 36/50
accuracy: 0.8418 - val_loss: 0.5662 - val_accuracy: 0.8261
Epoch 37/50
accuracy: 0.8422 - val_loss: 0.5642 - val_accuracy: 0.8261
Epoch 38/50
accuracy: 0.8424 - val_loss: 0.5661 - val_accuracy: 0.8261
Epoch 39/50
accuracy: 0.8447 - val_loss: 0.5572 - val_accuracy: 0.8269
Epoch 40/50
accuracy: 0.8447 - val_loss: 0.5567 - val_accuracy: 0.8269
Epoch 41/50
accuracy: 0.8461 - val_loss: 0.5497 - val_accuracy: 0.8252
Epoch 42/50
accuracy: 0.8451 - val_loss: 0.5458 - val_accuracy: 0.8269
accuracy: 0.8428 - val_loss: 0.5456 - val_accuracy: 0.8269
Epoch 44/50
accuracy: 0.8449 - val_loss: 0.5433 - val_accuracy: 0.8269
```

```
Epoch 45/50
accuracy: 0.8455 - val_loss: 0.5365 - val_accuracy: 0.8269
Epoch 46/50
accuracy: 0.8478 - val_loss: 0.5384 - val_accuracy: 0.8269
Epoch 47/50
accuracy: 0.8438 - val_loss: 0.5392 - val_accuracy: 0.8261
Epoch 48/50
accuracy: 0.8451 - val_loss: 0.5353 - val_accuracy: 0.8269
Epoch 49/50
accuracy: 0.8473 - val_loss: 0.5276 - val_accuracy: 0.8269
Epoch 50/50
accuracy: 0.8453 - val_loss: 0.5295 - val_accuracy: 0.8269
2023/2023 [============ ] - 0s 199us/step
```



Squential with Kfold CV: Training for fold 5 ... Adding layer 1: Adding layer 2: Adding layer 3: Adding layer 4:

Epoch 6/50

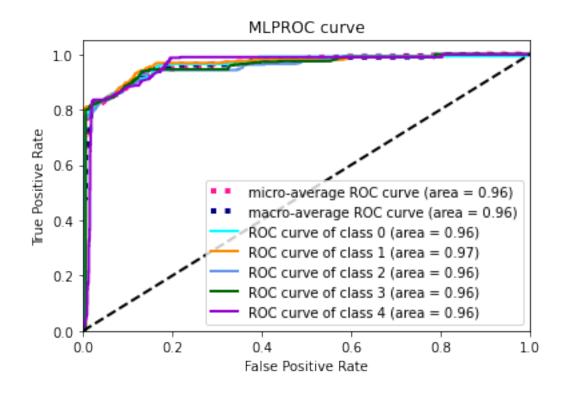
Model: "sequential\_40"

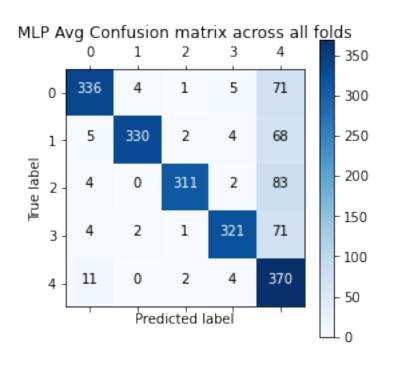
Layer (type)		Shape		
	(None,		384	
dense_216 (Dense)		48)	4656	
dropout_137 (Dropout)	(None,	48)	0	
dense_217 (Dense)	(None,	24)	1176	
dropout_138 (Dropout)	(None,	24)	0	
dense_218 (Dense)	(None,	12)	300	
dropout_139 (Dropout)			0	
dense_219 (Dense)	(None,		78	
<pre>dropout_140 (Dropout)</pre>	(None,		0	
dense_220 (Dense)			35	
Total params: 6,629 Trainable params: 6,629 Non-trainable params: 0				
Train on 4854 samples, valid	ate on	1213 samples		
4854/4854 [====================================			_	709 -
4854/4854 [====================================			_	2820 -
4854/4854 [====================================			_	1968 -
4854/4854 [====================================			_	1223 -
4854/4854 [====================================			_	0645 -

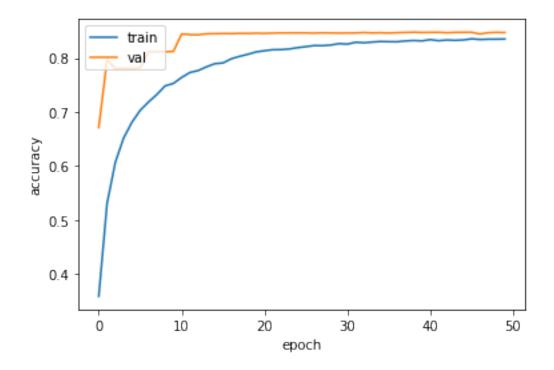
```
accuracy: 0.7316 - val_loss: 0.8502 - val_accuracy: 0.8508
Epoch 7/50
4854/4854 [============= ] - Os 57us/step - loss: 0.9879 -
accuracy: 0.7417 - val_loss: 0.8135 - val_accuracy: 0.8508
Epoch 8/50
accuracy: 0.7633 - val_loss: 0.7859 - val_accuracy: 0.8516
Epoch 9/50
accuracy: 0.7756 - val_loss: 0.7604 - val_accuracy: 0.8500
Epoch 10/50
accuracy: 0.7769 - val_loss: 0.7375 - val_accuracy: 0.8508
Epoch 11/50
accuracy: 0.7839 - val_loss: 0.7169 - val_accuracy: 0.8524
Epoch 12/50
accuracy: 0.7915 - val_loss: 0.7008 - val_accuracy: 0.8500
Epoch 13/50
accuracy: 0.7903 - val_loss: 0.6837 - val_accuracy: 0.8500
Epoch 14/50
accuracy: 0.8004 - val_loss: 0.6689 - val_accuracy: 0.8508
Epoch 15/50
accuracy: 0.8010 - val_loss: 0.6538 - val_accuracy: 0.8508
Epoch 16/50
accuracy: 0.8030 - val_loss: 0.6434 - val_accuracy: 0.8508
Epoch 17/50
accuracy: 0.8084 - val loss: 0.6343 - val accuracy: 0.8508
Epoch 18/50
accuracy: 0.8084 - val_loss: 0.6193 - val_accuracy: 0.8516
Epoch 19/50
accuracy: 0.8121 - val_loss: 0.6129 - val_accuracy: 0.8508
Epoch 20/50
accuracy: 0.8148 - val_loss: 0.6007 - val_accuracy: 0.8524
Epoch 21/50
accuracy: 0.8142 - val_loss: 0.5932 - val_accuracy: 0.8508
Epoch 22/50
```

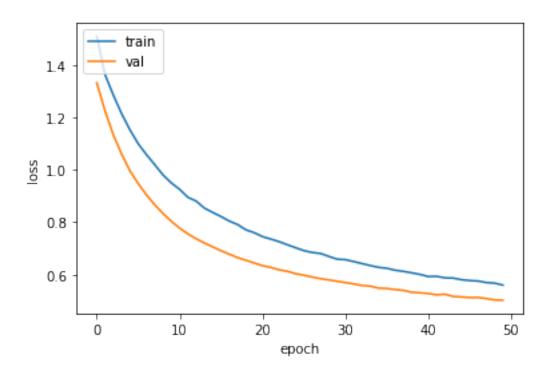
```
accuracy: 0.8127 - val_loss: 0.5927 - val_accuracy: 0.8524
Epoch 23/50
4854/4854 [============ ] - Os 55us/step - loss: 0.6961 -
accuracy: 0.8169 - val_loss: 0.5797 - val_accuracy: 0.8516
Epoch 24/50
accuracy: 0.8199 - val_loss: 0.5756 - val_accuracy: 0.8533
Epoch 25/50
accuracy: 0.8175 - val_loss: 0.5673 - val_accuracy: 0.8533
Epoch 26/50
accuracy: 0.8237 - val_loss: 0.5649 - val_accuracy: 0.8533
Epoch 27/50
accuracy: 0.8228 - val_loss: 0.5555 - val_accuracy: 0.8533
Epoch 28/50
accuracy: 0.8220 - val_loss: 0.5518 - val_accuracy: 0.8533
Epoch 29/50
accuracy: 0.8197 - val_loss: 0.5470 - val_accuracy: 0.8533
Epoch 30/50
accuracy: 0.8245 - val_loss: 0.5391 - val_accuracy: 0.8533
Epoch 31/50
accuracy: 0.8239 - val_loss: 0.5375 - val_accuracy: 0.8533
Epoch 32/50
accuracy: 0.8228 - val_loss: 0.5328 - val_accuracy: 0.8541
Epoch 33/50
accuracy: 0.8272 - val_loss: 0.5262 - val_accuracy: 0.8533
Epoch 34/50
accuracy: 0.8255 - val_loss: 0.5202 - val_accuracy: 0.8541
Epoch 35/50
accuracy: 0.8282 - val_loss: 0.5167 - val_accuracy: 0.8541
Epoch 36/50
accuracy: 0.8263 - val_loss: 0.5219 - val_accuracy: 0.8508
Epoch 37/50
accuracy: 0.8228 - val_loss: 0.5098 - val_accuracy: 0.8541
Epoch 38/50
```

```
accuracy: 0.8321 - val_loss: 0.5094 - val_accuracy: 0.8549
Epoch 39/50
accuracy: 0.8282 - val loss: 0.5062 - val accuracy: 0.8549
Epoch 40/50
accuracy: 0.8286 - val_loss: 0.4975 - val_accuracy: 0.8541
Epoch 41/50
accuracy: 0.8309 - val_loss: 0.5006 - val_accuracy: 0.8549
Epoch 42/50
accuracy: 0.8294 - val_loss: 0.4912 - val_accuracy: 0.8549
Epoch 43/50
4854/4854 [============ ] - Os 51us/step - loss: 0.5622 -
accuracy: 0.8313 - val_loss: 0.4968 - val_accuracy: 0.8533
Epoch 44/50
accuracy: 0.8255 - val_loss: 0.4858 - val_accuracy: 0.8541
Epoch 45/50
accuracy: 0.8321 - val_loss: 0.4836 - val_accuracy: 0.8541
Epoch 46/50
accuracy: 0.8280 - val_loss: 0.4788 - val_accuracy: 0.8541
Epoch 47/50
accuracy: 0.8323 - val_loss: 0.4850 - val_accuracy: 0.8500
Epoch 48/50
accuracy: 0.8315 - val_loss: 0.4821 - val_accuracy: 0.8524
Epoch 49/50
accuracy: 0.8298 - val loss: 0.4772 - val accuracy: 0.8524
Epoch 50/50
accuracy: 0.8348 - val_loss: 0.4814 - val_accuracy: 0.8533
2023/2023 [============ ] - 0s 137us/step
```





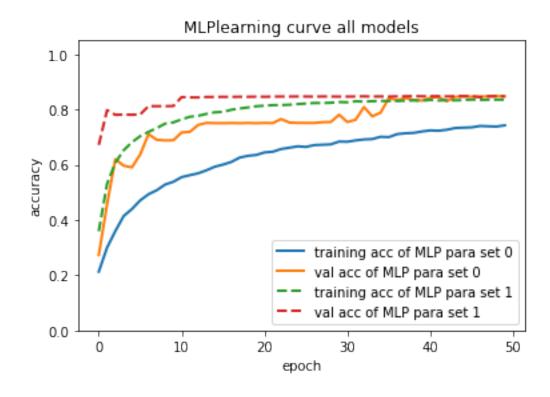


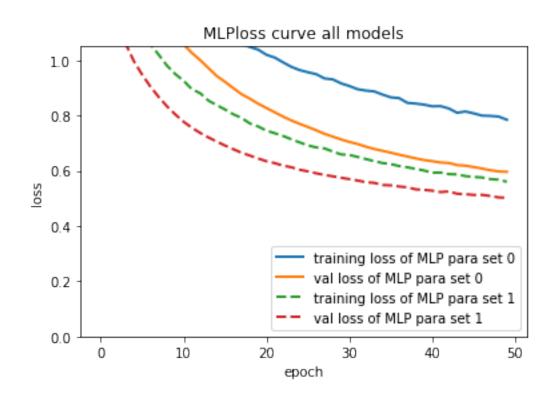


Average scores for pesudo test set across all folds: > Accuracy: 0.8255066752433777 (+- 0.0013981374057281189)

> Loss: 0.5529865491172199

## 





```
Opt model parameter found on the pseudo test set:
    {'num_hidden_layer': 3, 'hidden_layer_activation': ['relu', 'tanh', 'relu'],
    'dropout': [0.5, 0.25, 0.125], 'last_activation': 'softmax'}
    Best average pseudo test set accuracy score with the opt model:
    0.8255066752433777
    Average runtime per test instance: 0.0011530807568658065
[20]: import importlib
[]: #----- read true test set Task
[24]: importlib.reload(A1_functions)
     TaskA1_true_test_res_dict = A1_functions.get_A1_true_test_res('celeba_test',_
     →'gender', TaskA1_opt_models_dict)
    (969,)
    -----Task A1: True test performance with
    LR-----
    Prediction on true test set:
                precision recall f1-score
                                            support
            0.0
                    0.90
                            0.93
                                      0.91
                                               488
            1.0
                    0.92
                             0.90
                                      0.91
                                               481
                                      0.91
                                               969
        accuracy
       macro avg
                    0.91
                             0.91
                                      0.91
                                               969
    weighted avg
                    0.91
                             0.91
                                      0.91
                                               969
    True test set Accuracy: 0.9112487100103199
    Average runtime per test instance: 9.349748200061274e-07
    _____
    -----Task A1: True test performance with
    SVM-----
    Prediction on true test set:
                precision recall f1-score support
```

0.0	0.90	0.93	0.91	488
1.0	0.92	0.90	0.91	481
accuracy			0.91	969
macro avg	0.91	0.91	0.91	969
weighted avg	0.91	0.91	0.91	969

True test set Accuracy: 0.913312693498452

Average runtime per test instance: 0.00021770888683842677

-----Task A1: True test performance with RF-----

Prediction on true test set:

	precision	recall	f1-score	support
0.0	0.84	0.87	0.86	488
1.0	0.87	0.83	0.85	481
accuracy	•		0.85	969
macro avg		0.85	0.85	969
weighted avg	0.85	0.85	0.85	969
weighted ave	0.85	0.85	0.85	S

True test set Accuracy: 0.8534571723426213

Average runtime per test instance: 0.000554773214554762

-----

-----Task A1: True test performance with KNN------

Prediction on true test set:

	precision	recall	f1-score	support
0.0	0.77	0.93	0.85	488
1.0	0.91	0.73	0.81	481
accuracy			0.83	969
macro avg	0.84	0.83	0.83	969
weighted avg	0.84	0.83	0.83	969

True test set Accuracy: 0.8286893704850361

Average runtime per test instance: 0.001915088748046119

	Task A1: True test performance with
	MLP
	969/969 [===================================
	scores for true test set:
	> Accuracy: 0.9205366373062134
	> Loss: 0.37550388945514573
	> Avg runtime per test instance: 0.0005101350561876169
:	<pre>importlib.reload(A2_functions)</pre>
	TaskA2_true_test_res_dict = A2_functions.get_A2_true_test_res('celeba_test',
	<pre>→'smiling', TaskA2_opt_models_dict)</pre>

, ⊔

(969,)

[25]

-----Task A2: True test performance with LR------

Prediction on true test set:

	precision	recall	f1-score	support
0.0	0.89	0.92	0.90	472
1.0	0.92	0.89	0.90	497
accuracy			0.90	969
macro avg	0.90	0.90	0.90	969
weighted avg	0.90	0.90	0.90	969

True test set Accuracy: 0.9029927760577915

Average runtime per test instance: 1.8669970879490778e-06

-----

-----Task A2: True test performance with  ${\tt SVM------}$ 

Prediction on true test set:

	precision	recall	f1-score	support
0.0	0.90	0.92	0.91	472
1.0	0.92	0.90	0.91	497
accuracy			0.91	969
macro avg	0.91	0.91	0.91	969
weighted avg	0.91	0.91	0.91	969

True test set Accuracy: 0.9112487100103199

Average runtime per test instance: 0.0003592611220352913 -----Task A2: True test performance with RF-----Prediction on true test set: precision recall f1-score support 0.0 0.88 0.90 0.89 1.0 0.91 0.88 0.89 472 0.89 497 0.89 969 accuracy 0.89 0.89 0.89 macro avg 969 weighted avg 0.89 0.89 0.89 969 True test set Accuracy: 0.8926728586171311 Average runtime per test instance: 0.00012063168150721689 \_\_\_\_\_ -----Task A2: True test performance with Prediction on true test set: precision recall f1-score support 0.87 0.91 0.89 0.91 0.87 0.89 0.0 472 1.0 0.89 497 accuracy 0.89 969 macro avg 0.89 0.89 0.89 969 0.89 0.89 0.89 969 weighted avg True test set Accuracy: 0.890608875128999 Average runtime per test instance: 0.002542159628695871 -----Task A2: True test performance with MLP-----969/969 [======= ] - 1s 1ms/step scores for true test set: > Accuracy: 0.884416937828064 > Loss: 0.4281818089839475 > Avg runtime per test instance: 0.0012556738897742871

```
[26]: importlib.reload(B1_functions)

TaskB1_true_test_res_dict = B1_functions.

→get_B1_true_test_res('cartoon_set_test', 'face_shape', 
→TaskB1_opt_models_dict)
```

(1939,)

-----Task B1: True test performance with  ${\tt SVM------}$ 

Prediction on true test set:

	precision	recall	f1-score	support
0	0.82	0.77	0.79	362
1	0.72	0.68	0.70	378
2	0.76	0.81	0.78	424
3	0.69	0.72	0.71	394
4	0.82	0.82	0.82	381
accuracy			0.76	1939
macro avg	0.76	0.76	0.76	1939
weighted avg	0.76	0.76	0.76	1939

True test set Accuracy: 0.7601856627127386

Average runtime per test instance: 0.0013159975186643556

-----

-----Task B1: True test performance with  $\ensuremath{\text{RF}}\xspace------$ 

Prediction on true test set:

	precision	recall	f1-score	support
0	0.81	0.70	0.75	362
1	0.65	0.57	0.61	378
2	0.72	0.82	0.76	424
3	0.64	0.68	0.66	394
4	0.80	0.83	0.81	381
accuracy			0.72	1939
macro avg	0.72	0.72	0.72	1939
weighted avg	0.72	0.72	0.72	1939

True test set Accuracy: 0.7209902011346054

		-Task B1:	True test	performance	With
KNN					
Prediction on	true test :	set:			
	precision	recall	f1-score	support	
0	0.52	0.55	0.54	362	
1	0.41	0.35	0.38	378	
2	0.55	0.70	0.61	424	
3	0.46				
4	0.76	0.61	0.68	381	
accuracy			0.53	1939	
macro avg	0.54	0.53			
weighted avg	0.54	0.53	0.53	1939	
_	ne per test	instance	0.0036276	6740595014394	4
MLP		-Task B1:	True test	performance	with
MLP	eetest set	-Task B1: 	True test	performance	with
MLP	ne test set:	-Task B1:  =========== : :	True test	performance	with
MLP		-Task B1:  ======== : : : : : : : :	True test	performance - 2s 884us/st	with
MLP	re test set: .74522948268 .73460119492	-Task B1:  : : 507568 23 nstance: (	True test	performance - 2s 884us/st	with
MLP	re test set: .74522948268 .73460119492	-Task B1:  : : 507568 23 nstance: (	True test	performance - 2s 884us/st	with
MLP	ne test set: .74522948268 .73460119492 per test in .oad(B2_func	-Task B1: : :507568 23 nstance: ( :tions)	True test	performance - 2s 884us/st 20538238793	with
MLP	ne test set: .74522948268 .73460119492 per test in .oad(B2_func	-Task B1: : :507568 23 nstance: ( :tions)	True test	performance - 2s 884us/st 20538238793	with

0	0.93	0.78	0.85	410
1	0.99	0.78	0.87	404
2	0.99	0.81	0.89	420
3	0.97	0.81	0.88	401
4	0.54	0.97	0.70	382
accuracy			0.83	2017
macro avg	0.88	0.83	0.84	2017
weighted avg	0.89	0.83	0.84	2017

True test set Accuracy: 0.8264749628160635

Average runtime per test instance: 6.300988147680749e-05

-----Task B2: True test performance with RF-----

## Prediction on true test set:

	precision	recall	f1-score	support
0 1	0.89	0.83	0.86	410 404
2 3 4	0.81 0.80 0.47	0.88 0.82 0.20	0.84 0.81 0.28	420 401 382
accuracy macro avg weighted avg	0.71 0.71	0.72 0.72	0.72 0.69 0.70	2017 2017 2017

True test set Accuracy: 0.7223599405057015

Average runtime per test instance: 4.75632653829776e-05

-----Task B2: True test performance with KNN-----

## Prediction on true test set:

	precision	recall	f1-score	support
0	0.78	0.89	0.83	410
1	0.79	0.84	0.82	404
2	0.95	0.85	0.90	420
3	0.58	0.82	0.68	401
4	0.49	0.23	0.32	382

			0.73	2017	
macro avg	0.72	0.72	0.71	2017	
weighted avg	0.72	0.73	0.71	2017	
True test set A	ccuracy: 0.	7312840852	751611		
Average runtime	per test i	nstance: 3	.682249066	376958e-05	
	T	ask B2: Tr	ue test pe	rformance with	
MLP					
0017/0017 [		=======	====] - 0	s 132us/step	
2017/2017 [====					
scores for true	test set:				
		8774			
scores for true	23500275611				