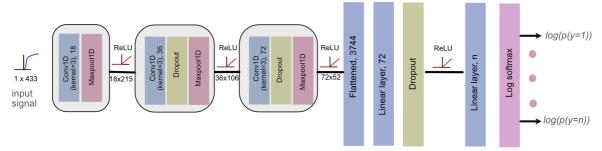
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HSI Analysis ML

Deep learning and machine learning methods for hyper-spectral imaging data



n: number of classes, p: probability

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Requirements

• Miniconda

· Linux/Unix only

• pyTorch >1.5

• CUDA > 10.0

• scikit-learn

matplotlib

• datatable

Setup

- Install miniconda
- Run bash initial setup.sh

Data

- Both training and test data are provided in the data folder
- Training data and test data for noExclusion and balanced datasets

Results

- Computed results are also included in metrics_outputs folder as .json_file produced from the provided codes
- noExclusion dataset (Table 2, Supplementary Tables 2-3) and balanced dataset (Supplementary Tables 4-6)

Training and testing classical machine learning approaches

• To reproduce Supplementary Tables 2 and Table 3, please run: bash Supplementary_Table2_Table3.sh

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• To reproduce Supplementary Tables 5 and Table 6, please run: bash

```
Supplementary_Table5_Table6.sh
```

RESULTS SAVED IN: results Classical ML folder

Includes:

- KNN
- SVM (with rbf kernel set)

Training a deep learning model

• For 3-way and 2-way 1D CNN classsification

```
bash script train.sh
```

Testing a deep learning model

Note: please train the model before running this, seeds are allocated for reproducibility

• To reproduce table 2 in original manuscript, please run: bash Table2.sh

RESULTS SAVED IN: results_noExlusion_test_data* folder

• To reproduce table 4 in supplementary manuscript, please run: bash Supplementary Table4.sh

RESULTS SAVED IN: results balanced test data* folder

Note: Use of pytorch with CUDA is recommended to obtain 1D-CNN results reported in the paper.

If you find any inconsistencies in results or code then please report to Sharib Ali