Machine Learning Internship - ImagoAl

Short Report

1. Preprocessing Steps & Rationale

- Missing Values: Filled missing values with the median to ensure robustness.
- **Feature Scaling:** Used **MinMaxScaler** to normalize spectral data between 0 and 1 for stable training.
- Dimensionality Reduction: Applied PCA (n_components=50) to retain 95% variance and reduce noise.

2. Insights from Dimensionality Reduction (PCA)

- PCA reduced **original feature dimensions** from **450** to **50** components.
- Retained 95% of variance, improving computational efficiency while keeping essential information.
- Scatter plot analysis showed clear patterns between PCA components and DON concentration.

3. Model Selection, Training & Evaluation

- Selected Model: Convolutional Neural Network (CNN-Conv1D) for spectral data feature extraction.
- Loss Function: Mean Squared Error (MSE) for stable regression performance.
- Hyperparameters Optimized: Filters (128, 64, 32), Kernel Size (5, 3, 3), Dropout (0.3, 0.2).
- Training Setup: 80% training, 20% testing, batch size = 16, epochs = 80.

4. Key Findings & Suggestions for Improvement

Performance Metrics:

- Mean Absolute Error (MAE): 0.0338
- Root Mean Squared Error (RMSE): 0.0798
- R² Score: 0.6092

Limitations & Improvements

- **PCA Feature Loss:** Some spectral details might be lost; **alternative methods like Autoencoders** could be explored.
- Alternative Models: LSTM or Transformer-based models could capture sequential dependencies better.

- More Data Needed: The dataset size is limited, data augmentation or more samples could improve generalization.
- **Hyperparameter Tuning:** Using **Grid Search or Bayesian Optimization** could fine-tune the CNN architecture.