

# EEG Classification Project Report

## Project Overview

This project focuses on classifying EEG signals using two distinct approaches:

- **Logistic Regression** with log-variance features
- **EEGNet**, a compact deep learning architecture tailored for EEG data

The goal is to evaluate and compare their performance across multiple cross-validation folds.

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## Dataset & Preprocessing

- **Input shape:** EEG epochs of shape (N, C, T)
    - N: number of samples
    - C: number of channels
    - T: number of time points
  - **Preprocessing:**
    - Band-pass filtering
    - Epoch extraction
    - Feature extraction (log-variance for Logistic Regression)
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## Models

### 1. Logistic Regression

- Features: Log-variance across channels
- Optimizer: Momentum-based gradient descent
- Regularization: L2 penalty
- Hyperparameters:
  - Epochs: 1000
  - Learning rate: 0.01
  - Batch size: 8

- L2: 1e-5

## 2. EEGNet

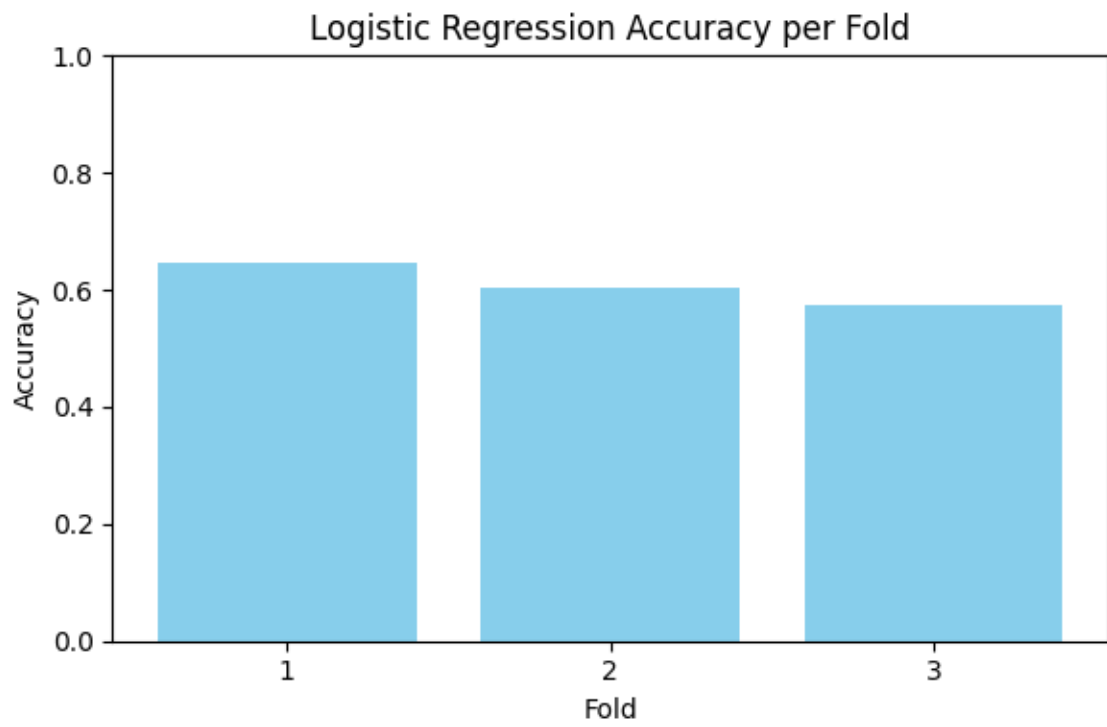
- Architecture: Depthwise and separable convolutions
  - Input: Raw EEG epochs
  - Optimizer: Adam
  - Loss: CrossEntropy with label smoothing
  - Hyperparameters:
    - Epochs: 100
    - Learning rate: 5e-4
    - Batch size: 32
    - Dropout: 0.0
    - Early stopping: Patience = 20
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## Results

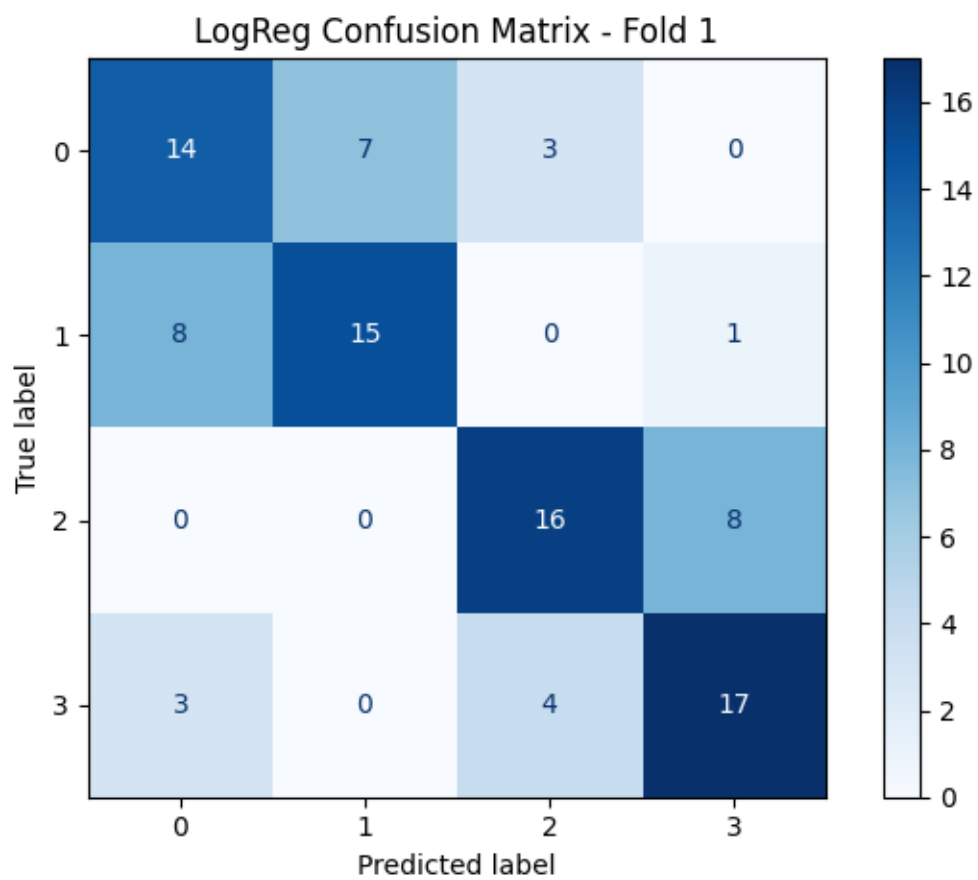
### Logistic Regression

- **Mean Accuracy:**  $0.6076 \pm 0.0299$

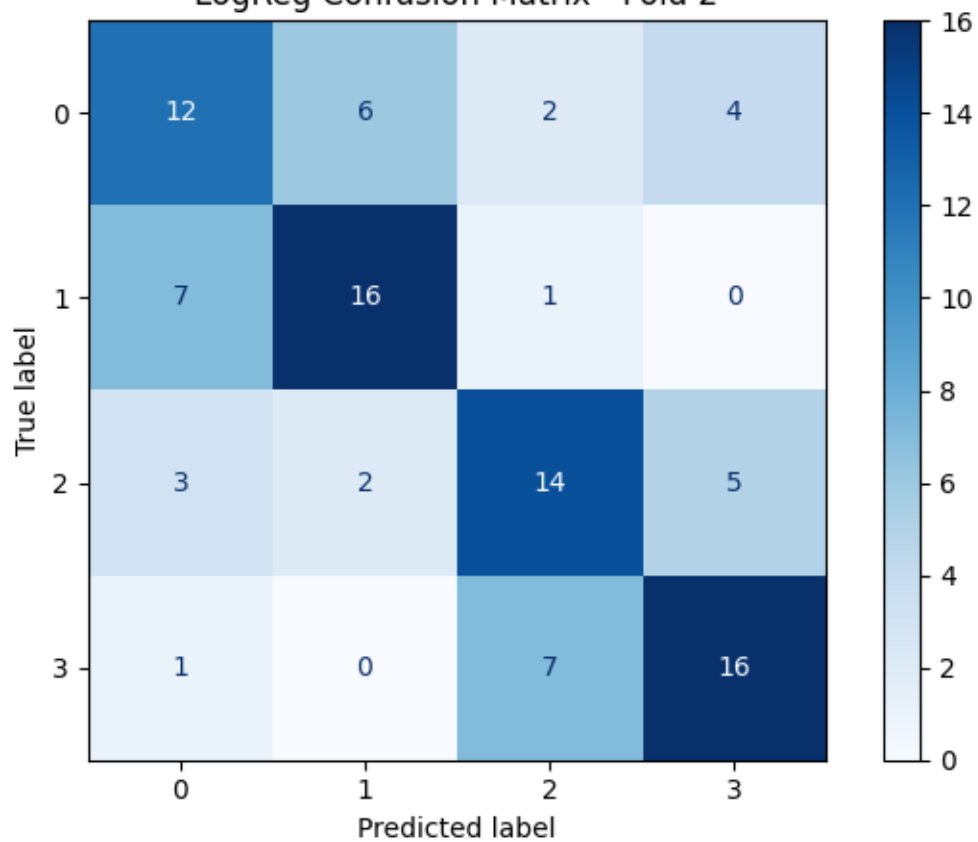
- **Accuracy per Fold:**

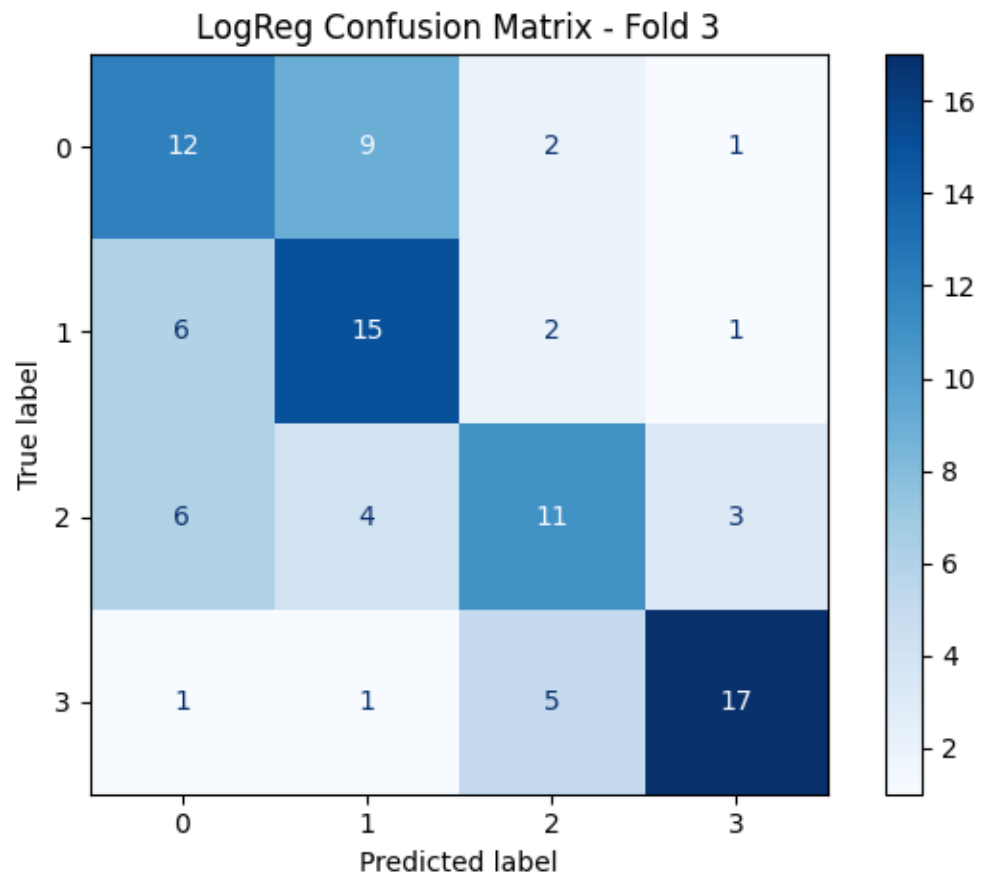


- **Confusion Matrices:**



LogReg Confusion Matrix - Fold 2



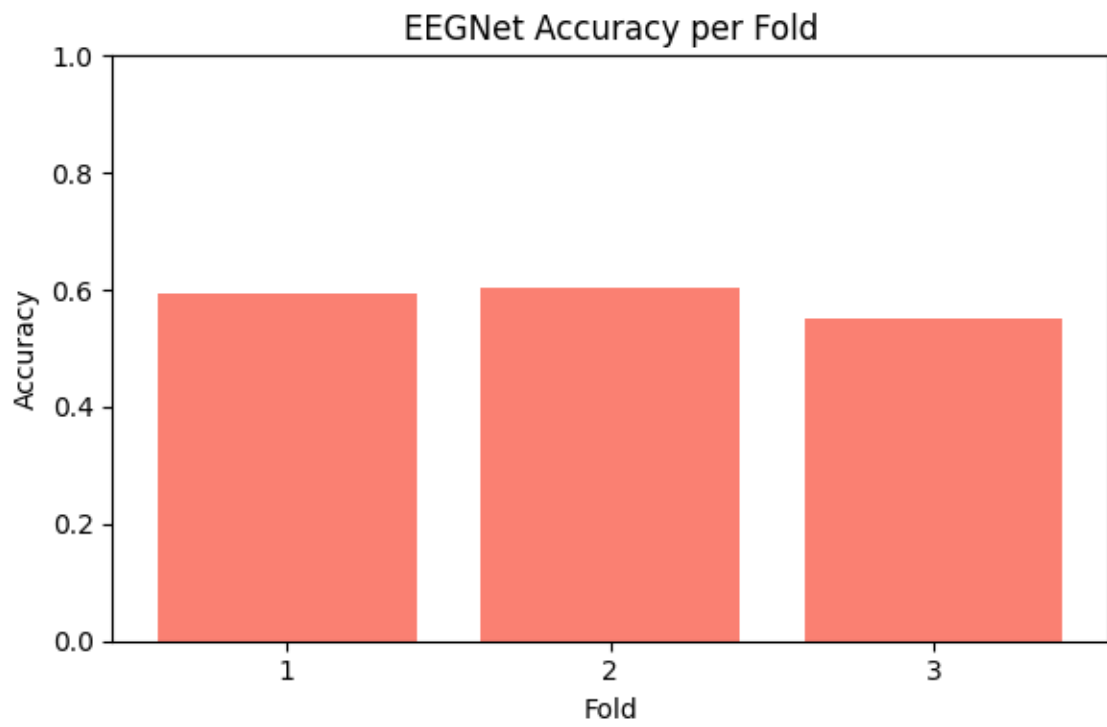


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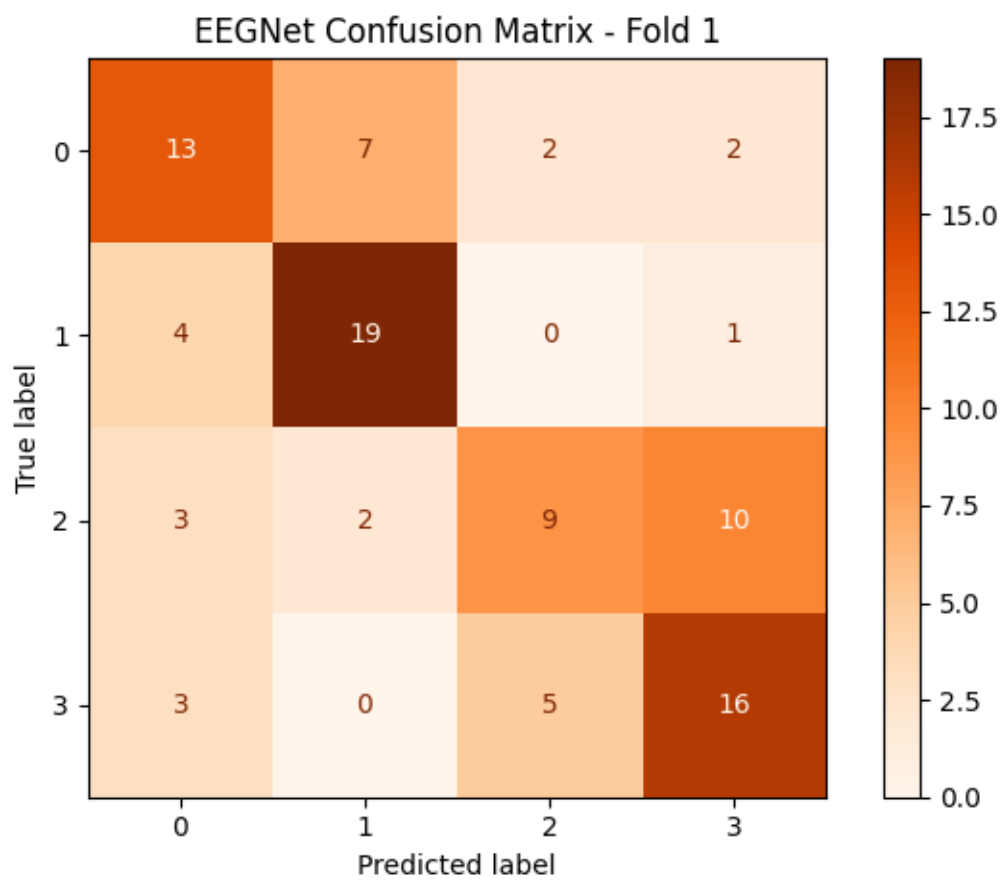
#### EEGNet

- **Mean Accuracy:**  $0.5833 \pm 0.0225$

- **Accuracy per Fold:**

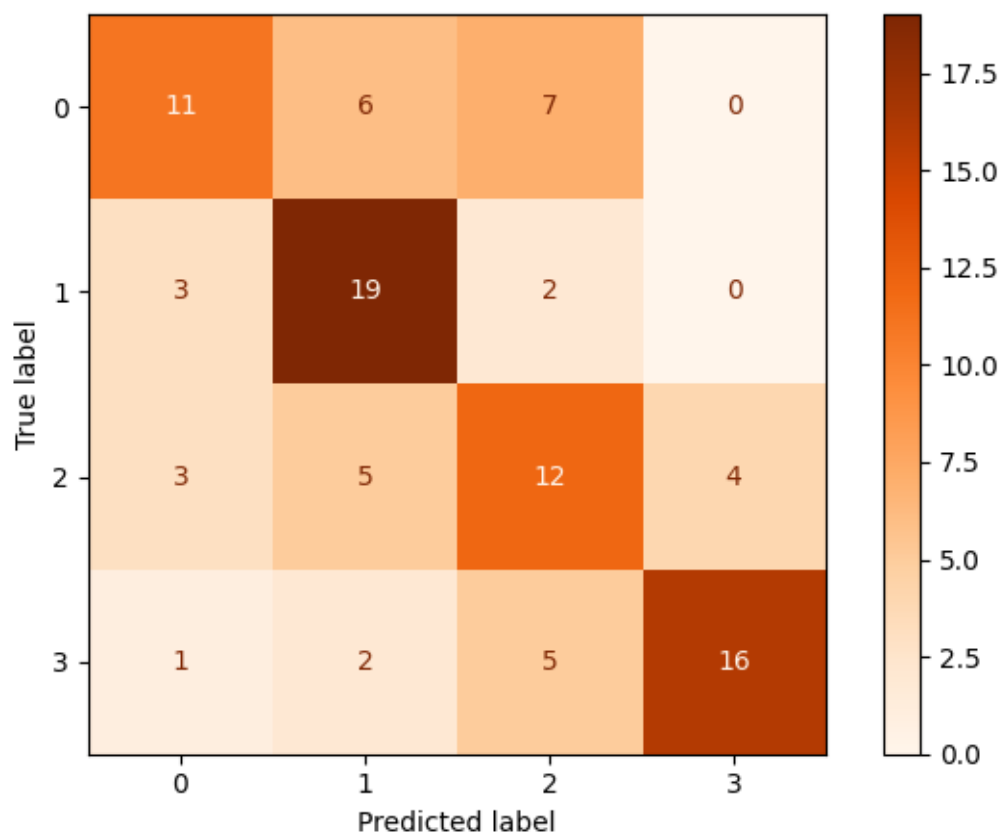


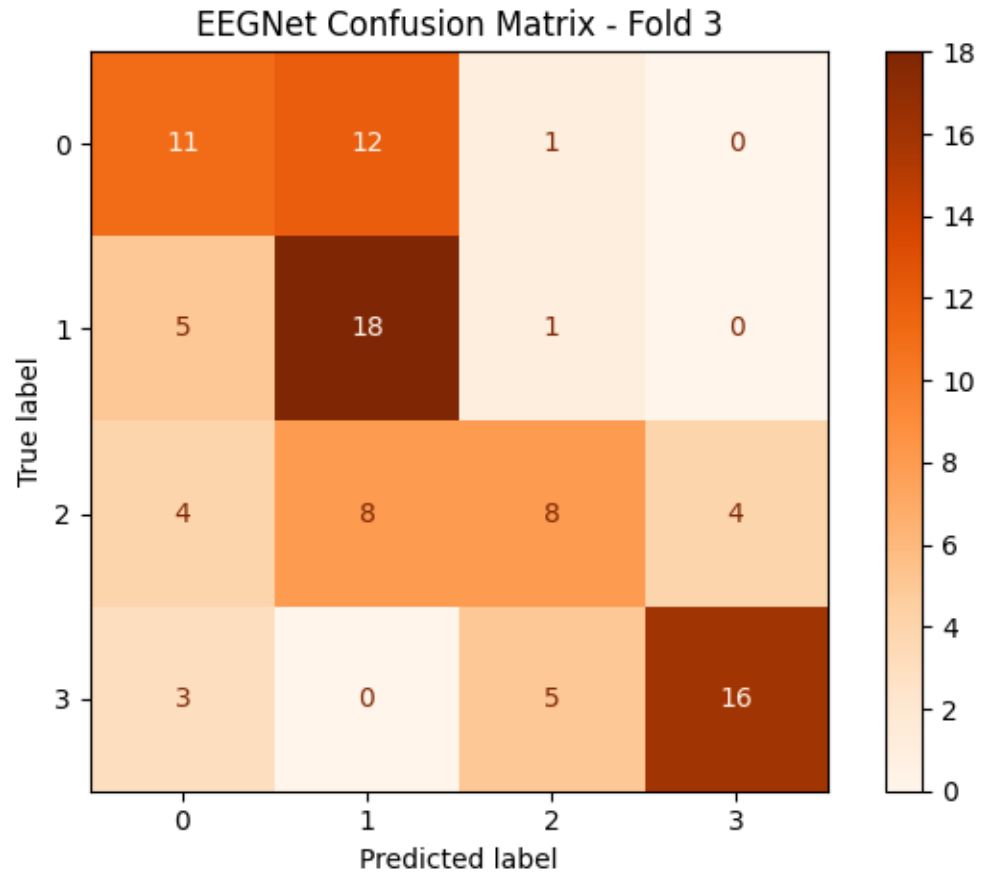
- **Confusion Matrices:**





EEGNet Confusion Matrix - Fold 2





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### Comparison & Insights

Model	Mean Accuracy	Training Time
Logistic Regression	0.6076 ± 0.0299	Fast
EEGNet	0.5833 ± 0.0225	Longer

- Logistic Regression is lightweight and interpretable.
  - EEGNet captures temporal and spatial patterns more effectively.
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