

## Weekly Progress Report - 6

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<b>Project Title</b>	Combating Digital Misinformation: Deepfake Detection Using Deep Neural Networks
<b>Name of the Supervisor (Mentor) at PDEU</b>	Dr. Samir Patel
<b>Week Number</b>	Week 6

### **Progress made in Week:**

#### 1. Dataset Organization and Splitting

- Created a structured dataset of 400 videos (200 real and 200 fake), each containing 200 frames.
- Implemented a video-wise train/validation/test split to prevent data leakage:
  - 70% train, 15% validation, 15% test.
- Final frame count:
  - Real frames → 50 per video sampled evenly.
  - Fake frames → 50 per video sampled evenly.

#### 2. Frame Sampling & Redundancy Reduction

- Developed two approaches to reduce redundant frames:
  1. Step Sampling: Selects evenly spaced frames from each video.
  2. Laplacian Variance (Sharpness-based): Filters out blurred or low-quality frames.
- Final pipeline uses step sampling for uniform coverage.

#### 3. Data Augmentation

- Applied Albumentations library for advanced transformations:
  - Horizontal flip, random brightness/contrast.
  - Gaussian noise for robustness to artifacts.
  - JPEG compression to simulate deepfake-specific distortions.
  - Coarse dropout to mimic occlusions.
- Normalized frames using ImageNet statistics to match XceptionNet pretraining.

#### 4. XceptionNet Integration

- Loaded pretrained XceptionNet using timm.
- Replaced the final classifier layer with a binary classification head for deepfake detection.

#### 5. Training Pipeline Setup

- Implemented custom PyTorch training loop:
  - Optimizer → Adam (LR = 1e-4, weight decay = 1e-5)
  - Loss function → CrossEntropyLoss
  - Batch size → 32
- Integrated DataLoaders for efficient data handling with augmentation applied only on the training set.

#### 6. Validation and Visualization

- Created scripts to visualize:
  - Sampled frames.
  - Mini-batch outputs to verify augmentations.
- Implemented functions to track accuracy and loss per epoch.

#### Future Steps

- Begin full training with XceptionNet and evaluate on validation and test sets.
- Implement Optuna-based hyperparameter tuning for optimizing learning rate, batch size, and augmentation strength.
- Start integrating attention modules into XceptionNet.

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