

Deepfake video detection using CNN

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Introduction

- Image and video forgery are posing a threat to society in today's world. People can artificially create any audio or video clip.
- In a deepfake video, a person's face, emotion, or speech is replaced by someone else's face, different emotion, or speech, using deep learning technology.
- Deepfake videos are often visually indistinguishable from real ones.
- They can have a heavy social, political, and emotional impact on individuals, as well as on society.

Current state of art

- Deepfake technology is rapidly advancing and becoming more difficult to detect.
- Current methods for detecting Deepfakes involve analyzing artifacts and inconsistencies in the video.
- However, Deepfake creators are becoming more skilled at avoiding these artifacts and making their videos appear more realistic.
- Advanced machine learning techniques, such as GANs, are being used to create more convincing Deepfakes.
- To stay up with the developing technologies, there is a need for more advanced and precise detecting techniques.

Motivation

- The increasing number of deepfake videos is a growing concern
- Advancements in technology have made it easier to create deepfake videos
- Deepfake videos have harmful impacts on people
- The project aims to detect deepfake videos to reduce their impact
- Developing a deepfake video detector is a proposed solution
- The detector will be able to identify and reduce the spread of deepfake videos

Objectives

- To study the existing models for detecting deepfake videos.
- To develop an optimal solution for deepfake detection with improved accuracy.
- To compare the impact of different feature extractors on the model and choose the best among them for implementation
- To develop a model with low computational complexity and memory requirements so it could be used in edge devices.
- To prevent the propagation of false information through deepfake media

Literature Review

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Proposed Methodology

The proposed solution for deepfake detection consists of five steps:

- Frame Extraction Facial Landmarks Detection
- Temporal Facial Feature Analysis
- Data preprocessing
- Data split: Training, validation and Testing
- Customized Convolutional Neural Network (CNN)

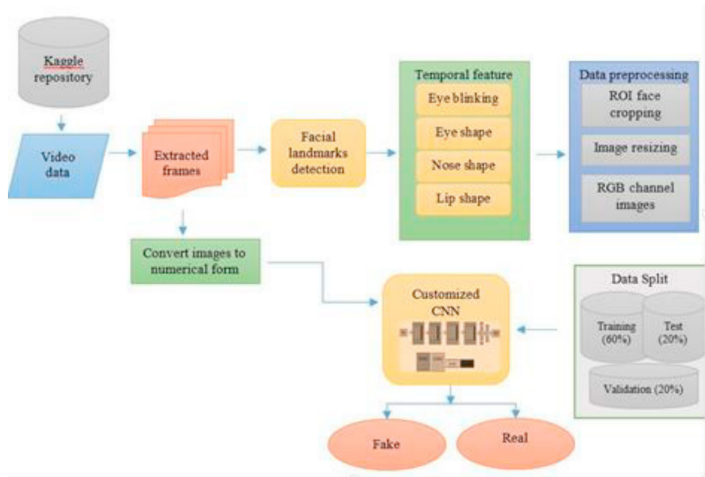


Fig.1. Block diagram of proposed model

Algorithm:

Input: Deepfake video dataset

Output: Good classification results

Strategy:

- Step 1. Input video dataset from Kaggle
- Step 2. Frames extraction from videos
- Step 3. Detection of the all-face features using Facial landmark predictor model
 - a. Eyes blink detection
 - b. Eyes shape detection
 - c. Lips shape detection
 - d. Nose shape detection
- Step 4. Perform data preprocessing on frames
 - a. Crop the face region of interest
 - b. Image resize into 224 x 224
 - c. Ensure all images in the RGB channel
- Step 5. Data splitting into three parts
 - a. Training set (60%)
 - b. Validation set (20%)
 - c. Testing test (20%)
- Step 6. Hyper Parameters setting
- Step 7. Apply Customized CNN model by adding some layer for training
 - a. conv2d layer
 - b. Batch normalization
 - c. Max pooling layer
 - d. Drop out layer
 - e. Flatten
 - f. Dense layer
- Step 8. Perform testing on a test set
- Step 9. Calculate performance metrics
- Step 10. Classification results whether it is fake or real

Implementation Status and Plan

- Proposed method has been implemented and tested on a dataset of both real and deepfake videos.
- Increase the accuracy of the system.
- Future plans include further refinement of the model by incorporating more advanced techniques and increasing the size of the dataset for training.
- Evaluate the effectiveness of the model on a larger and more diverse dataset in the future.

Thank you!