

Deepfake Detection System

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Abstract: Deepfake can distort our perception of the truth and we need to develop a strategy to improve their detection. Deep Fakes are increasingly detrimental to privacy, social security, and democracy. We plan to achieve better accuracy in predicting real and fake videos.

Keywords: Deepfake, Deepfake detection System, Face Forensic, AI arms race, Computer Vision

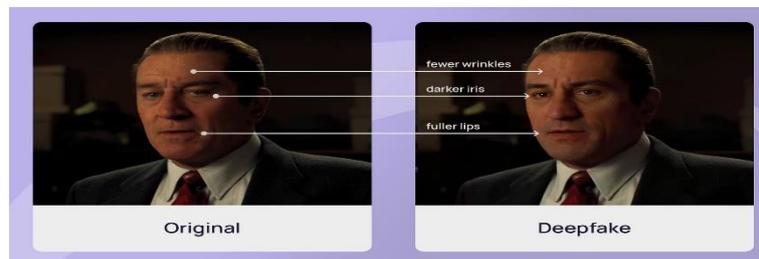
I. INTRODUCTION

Deepfake algorithm is used to examine the facial expression and the body moments, and then it compares the other human facial expression with the same body moments. It also checks the voice that is spreading at the internet with the name of any celebrity is original or not? Because many celebrities become the victim of such cases where wrong video or voice messages spreading in the wrong way. Worries regarding deepfakes have resulted in a growth of countermeasures, prompting researchers to conduct extensive research. And lectures describing strategies to defend against them abound at computer vision and graphics conferences.

II. OBJECTIVE

The objective of this paper is to give the reader a better knowledge of:

- How Deepfake in media is generated and identified.
- The latest developments and breakthroughs in this realm.
- Weaknesses of existing security methods.
- Areas requiring more investigation and consideration.



III. MODEL DESIGN AND SELECTION

Due to the nature of deep neural networks being data-driven, it is necessary to acquire massive deepfake datasets with various different synthesis methods in order to achieve promising results. The following deepfake datasets were used in the final model at DF Detect:

- DeepFake-TIMIT
- FaceForensics++
- Google Deep Fake Detection (DFD)
- Celeb-DF
- Facebook Deepfake Detection Challenge (DFDC)

Prerequisites

- Python 3
- Keras
- TensorFlow
- EfficientNet for TensorFlow Keras

- OpenCV on Wheels
- MTCNN
- Django

Global Deepfake Detection Market

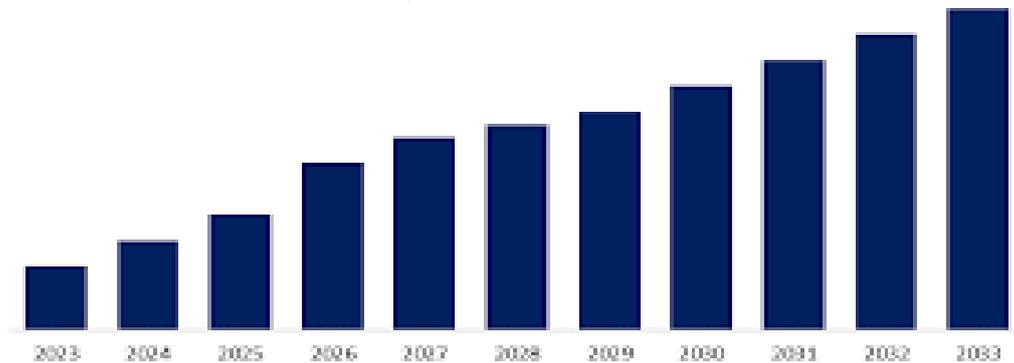


Fig. 1 Global Deepfake Detection Market

Market Growth Projection	The global deepfake detection market is projected to grow significantly, with some reports forecasting a compound annual growth rate (CAGR) of over 37% between 2023 and 2033.
Market Drivers	The market is driven by the increasing need to prevent misuse of AI-generated media, such as videos and images, in various sectors
Fastest-Growing Verticals	Industries such as banking, financial services, and insurance (BFSI) are expected to be the fastest-growing verticals for deepfake detection, due to a rise in synthetic fraud and regulatory pressure.
Largest Market Share	Video deepfake detection currently holds the largest share of the market, as the widespread use of video content increases the potential for manipulation.
Deepfake Detection System Performance	Deepfake Detection System learning models can achieve detection rates of 91% to 98% on benchmark datasets like the FaceForensics+ and Celeb-DFv2, provided the fakes are similar to those in the training data.
Real-time Detection Technology	Technology used in identity verification can detect deepfakes in real-time by analyzing responses to challenges like blinking on command. A forged video or audio would fail such a test.

IV. CONCLUSION

Deepfake Detection System learning models can achieve detection rates of 91% to 98% on benchmark datasets like the FaceForensics++ and Celeb-DFv2, provided the fakes are similar to those in the training data. Technology used in identity verification can detect deepfakes in real-time by analyzing responses to challenges like blinking on command. A forged video or audio would fail such a test. Deepfake Detection Systems represent a critical and ongoing effort in the "AI arms race," where detection technologies must constantly evolve to keep pace with increasingly sophisticated deepfake generation methods. While deep learning-based systems have achieved high accuracy in controlled environments, significant challenges remain concerning real-world effectiveness, ethical implications, and legal frameworks.

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