## Deepfake Detection Using Deep learning

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#### Introduction I

- Deep fakes are created by combing and superimposing existing images and videos onto source images or videos using a deep learning technique known as generative adversarial network.
- Deepfake systems typically need various photos and video data to develop models to create realistic videos and images.
- Higher deep networks and freely accessible data have made fake videos and images almost equal to humans and even advanced computer algorithms.
- However, when the concept of neural networks grew big, and understanding improved its computational abilities, people started using this technology in their daily lives.
- Researchers chose to find methods of deep face detection to help people from this danger.
- Video manipulation and Digital image technologies are developing quickly for many decades.

## Motivation I

- As deepfakes are fake videos generated using some digital software like faceswap, faceit, deeplab etc. in which computer creates an Al video which lead to new video that shows the actions that wasnât happened. but the results looks like acceptable. they are so difficult to identify that they are false.
- ② Developing these techniques made a great impact on social and politics. The real danger is because of false information it creates mistrust or lack of interest in people about what they hear in online.

## Motivation II

These problems may leads to worst conditions like

- Fake News
- Malicious hoaxes
- Financial fraud
- Politician videos.
- facial reanimation

are some of them this is why deepfake detection is needed.

# Literature Survey

Author	Title	Year	Publish	Work
Karnouskos,	Digital	2020	IEEE	considering the how
Stamatis	Media:			deepfakes effects the
	The Era of			media
	Deepfakes			
Nguyen,	Deep Learn-	2019	IEEE	the objective is survey of
Thanh and	ing for			how to detect and create
Nguyen,	Deepfakes			a deep fake
Cuong M.	Creation			
and Nguyen	and De-			
	tection: A			
	Survey			
Meng, Cun-	Video En-	2010	IEEE	the objective here is to
yan and	cryption			generate a dataset from
Zhang,	Based on			input video using opencv
Xinghui	OpenCV			

## Literature Survey

	Author	Title	Year	Publish	Work
ĺ	Morozov, Artem Deepfake		2020	IEEE	the objective here is to
	A., Maksutov,	Detection			create fake video using
	and	Methods us-			gan and model experi-
		ing Machine			ments
		Learning			
	Piotr Kawa	Deepfake	2020	IEEE	deep fake detection
	and P. Syga	Detection			experimentst using
		with Low-			mesonet-4 and mesonet-
		Resources			inception -4
	Pokroy,	DeepFake	2021	IEEE	the study conducted is
	Artem A.	Detection:			to compare the accura-
	and Egorov,	Comparing			cies of various deep net-
	Alexey D	Pretrained			work models
		Models			
	Mahamkali,	OpenCV for	2015	IEEE	feature detection&object
	Naveenku-	CV Applica-			tracking using opency/19

## **OBJECTIVES**

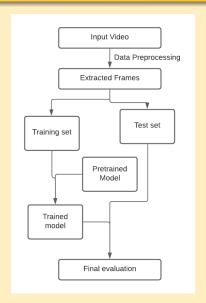
- In this system,a pretrained model is used on training set and model with best accuracy is used to determine wheather the video is fradulent or nonfraudulent
- 2 To train the model, dataset must be retrieved and data must be preprocessed in following steps:
  - extract frames from video
  - filtering the extracted frames
  - splitting of data into train and test

## System Architecture

The final purpose of this work is to recognize whether a video obtained real or knowing whether it was created using some other deepfake methods. As the input to the model or the system is a video

But, deep learning models do take pictures as input, so there needs to be a change in the input to model from the system. This can be achieved through pre-processing.

## System Architecture Cont..



## System Architecture Cont..

#### Data Preprocessing:

module need to check how other factors impact on the model training. Most of the video frames not only contains faces but also person's body parts and background area of the image. Thus unrelated data can highly impact our model training. So we need to focus on the face area and our preprocessing module need to capture the face as input.preprocessing consists of three steps:

- collect frames from the video
- detecting faces from the individual images.
- resizing and saving locally.

## System Architecture Cont..

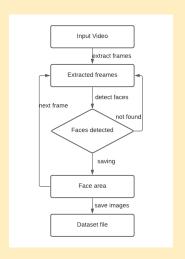


Figure: preprocessing

Importing the dataset and required libraries, the data is fetched using a opency, the data is analyzed and preprocess in steps:

1) Extraction of frames: Using opency tool(imread, imwrite) we can extract the frames from the input video with certain frame rate depending how no of images we want to store.



**2)size reduction:** the dimensions of each of the image that we have extracted from video are much large, so for better output and to take less time we have scale down each image.



Figure: size reduction

**3) face identification:** Now that we have resized our image, Apply the CascadeClassifier function, an OpenCV function, to an XML file location for frontal face detection, i.e.,

haarcascade\_frontalface\_default.XML, in our project. I have used path in the machine that I have downloaded.

Generally the images are of RGB channel that we see. But OpenCv reads the image in BGR format, so we have to convert into gray channel because it is simple to process also takes less time for computing as it holds just one channel of white-black.



With the help of function detectMultiScale,we get the coordinates of face, as x-coordinate, y-coordinate, width and height. With the help of x and y coordinated a square is drawn.

**4)modifying the data:** for an image if we found a face then we will replace it with updated image else remove the image from dataset which is initially created.

```
1 if faces is ():
2    print("No faces found")
3 for (x,y,w,h) in faces:
4    cv2.rectangle(resized, (x,y), (x+w,y+h), (127,0,255), 2)
5    image2=resized[y:y+h,x:x+w]
6    cv2_imshow(image2)
7    cv2.imwrite("/content/drive/MyDrive/IIITM/BTP/data2/frame6.jpg",image2)
8    cv2.waitKey(0)
9    break
10
11 cv2.destroyAllWindows()
```

## Tasks to be completede

- To study the deep learning model used to detect the deepfake video
- 2 To implement the code and visualize the output based on input video

#### References I

- (1),(2),(3),(4),(5),(6),(7),(8)
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