

CNN 2023

December 17

Facial Expression Recognition (FER) is a computer vision task aimed at identifying and categorizing emotional expressions depicted on a human face. The goal is to automate the process of determining emotions in real-time, by analyzing the various features of a face such as eyebrows, eyes, mouth, and other features, and mapping them to a set of emotions such as anger, fear, surprise, sadness and happiness.

Face
emotional

Used Function & Library:

- ✓ `cv2`
- ✓ `DeepFace`
- ✓ `matplotlib.pyplot` as `plt`
- ✓ `cv2.imread`
- ✓ `plt.imshow`
- ✓ `cv2.cvtColor`
- ✓ `DeepFace.analyze`
- ✓ `type()`
- ✓ `cv2.CascadeClassifier`
- ✓ `cv2.data.harcascades`
- ✓ `cv2.COLOR_BGR2RGB`
- ✓ `cv2.COLOR_BGR2GRAY`
- ✓ `FaceCasCade.detectMultiScale`

Section 1:

```
In [1]: import cv2

In [2]: from deepface import DeepFace
WARNING:tensorflow:From C:\Users\pc\anaconda3\Lib\site-packages\keras\src\losses.py:2976: The name tf.losses.sparse_softmax_cross_entropy is deprecated. Please use tf.compat.v1.losses.sparse_softmax_cross_entropy instead.

In [3]: img = cv2.imread('cry.jpg')

In [4]: import matplotlib.pyplot as plt

In [5]: plt.imshow(img)

Out[5]: <matplotlib.image.AxesImage at 0x1ff4f332bd0>
```

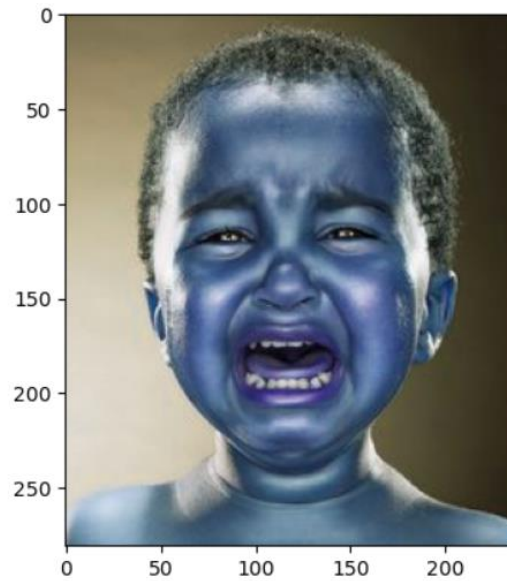
✓ Import cv2 → which is a popular and powerful tool for **computer vision, Image processing , face detection , object recognition, resize , preprocess the images.**

✓ from deepface import DeepFace → we use It **in Face recognition, Facial attribute analysis tool , pre-trained models . it used to access the emotion detection model.**

✓ `img = cv2.imread('cry.jpg')` → **read the image from the path.**

Import matplotlib.pyplot as plt → creating various types of **plots and Graphs.**

✓ `plt.imshow(img)` → **display img as graph.**



✓ The axis refer to pixels in image , but the image with not correct colors!!

```
In [6]: plt.imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
```

```
Out[6]: <matplotlib.image.AxesImage at 0x1ff4f407810>
```



✓ `cv2.cvtColor(img, cv2.COLOR_BGR2RGB)` → convert the image color to correct.

Section2:

```
In [7]: predictions = DeepFace.analyze(img)
Action: race: 100%|██████████| 4/4 [00:23<00:00, 5.87s/it]
```

```
In [8]: predictions
Out[8]: [{'emotion': {'angry': 2.0432965829968452,
  'disgust': 0.0208952886168845,
  'fear': 5.043782666325569,
  'happy': 0.1652681501582265,
  'sad': 92.71451234817505,
  'surprise': 3.620834831963293e-05,
  'neutral': 0.012206201063236222},
  'dominant_emotion': 'sad',
  'region': {'x': 41, 'y': 58, 'w': 154, 'h': 154},
  'age': 27,
  'gender': {'Woman': 0.1291487831622362, 'Man': 99.87084865570068},
  'dominant_gender': 'Man',
  'race': {'asian': 0.7353193099242038,
  'indian': 3.3545876058690314,
  'black': 93.12836486866264,
  'white': 0.1470654708446843,
  'middle eastern': 0.1014742478675692,
  'latino hispanic': 2.533191220950557},
  'dominant_race': 'black'}]
```

Predictions = DeepFace.analyze(img) → display all features of pre-trained model and save it predictions var.

```
In [9]: type(predictions)
```

```
Out[9]: list
```

```
In [10]: predictions[0]['dominant_emotion']
```

```
Out[10]: 'sad'
```

- ✓ The type of features that model display is in **list of dictionary** format.
- ✓ List that displayed has **two index**, so we access the target emotion from **dominant_emotion**.

```

In [11]: faceCascade = cv2.CascadeClassifier(cv2.data.harcascades + 'haarcascade_frontalface_default.xml')

In [12]: gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
        faces = faceCascade.detectMultiScale(gray, scaleFactor=1.1, minNeighbors=5, minSize=(30, 30))

        for (x, y, w, h) in faces:
            cv2.rectangle(img, (x, y), (x+w, y+h), (0, 255, 0), 2)

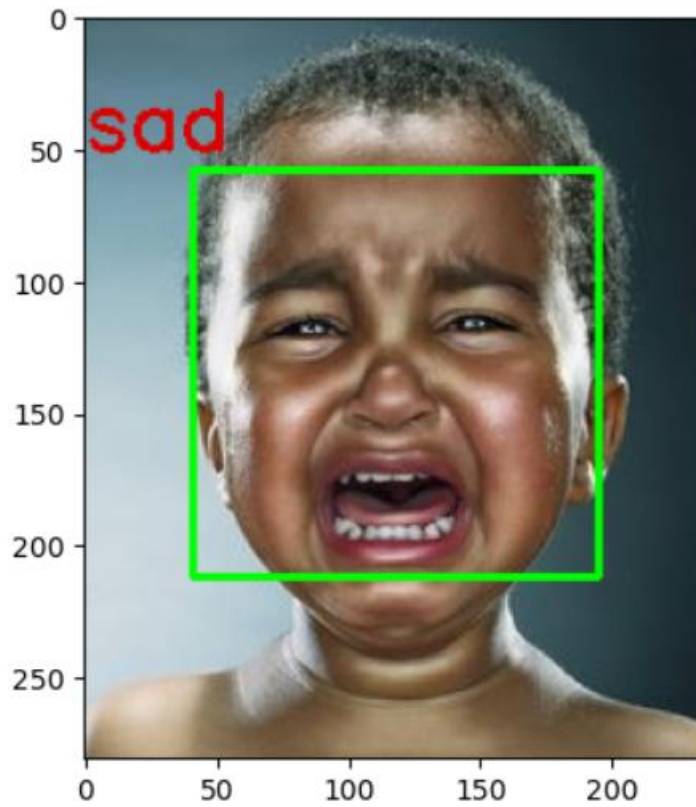
In [13]: font = cv2.FONT_HERSHEY_SIMPLEX
        cv2.putText(img,
                    predictions[0]['dominant_emotion'],
                    (0, 50),
                    font, 1,
                    (0,0,225),
                    2,
                    cv2.LINE_4);

```

- ✓ We use **cv2.CascadeClassifier** as a model which has multiple files of detections so we **'haarcascade_frontal_default.xml'** to detect face emotion and we get the path of this file using **cv2.data.harcascades**.
- ✓ We convert the color of image to **gray** to **reduce the complexity** of image.
- ✓ Next the faces store **four** value as scale of image **X Y W H** as **X** and **Y** refers to **the top left corner** of the image, **W** refers to **width of image** and **H** refers to **height of image**.
- ✓ Next we use function to draw rectangle in the detected face and display his emotion in the top left corner of the image.

```
In [14]: plt.imshow(cv2.cvtColor(img, cv2.COLOR_BGR2RGB))
```

```
Out[14]: <matplotlib.image.AxesImage at 0x1ffbd5427d0>
```



The last Section (WebCam):

```
In [16]: import cv2
from deepface import DeepFace

faceCascade = cv2.CascadeClassifier(cv2.data.haarcascades + 'haarcascade_frontalface_default.xml')

cap = cv2.VideoCapture(1)
if not cap.isOpened():
    cap = cv2.VideoCapture(0)
if not cap.isOpened():
    raise IOError("Cannot open webcam")

while True:
    ret, frame = cap.read()
    result = DeepFace.analyze(frame, actions=['emotion'])

    gray = cv2.cvtColor(frame, cv2.COLOR_BGR2GRAY)
    faces = faceCascade.detectMultiScale(gray, scaleFactor=1.1, minNeighbors=4)

    for (x, y, w, h) in faces:
        cv2.rectangle(frame, (x, y), (x+w, y+h), (0, 255, 0), 2)

    font = cv2.FONT_HERSHEY_SIMPLEX
```

```
font = cv2.FONT_HERSHEY_SIMPLEX

cv2.putText(frame,
            result[0]['dominant_emotion'],
            (50, 50),
            font, 3,
            (0, 0, 255), # Fixed the color (0,0,255) corresponds to red
            2,
            cv2.LINE_4)

cv2.imshow('Demo video', frame)

if cv2.waitKey(2) & 0xFF == ord('q'):
    break

cap.release()
cv2.destroyAllWindows()
```

```
cap.release()
cv2.destroyAllWindows()

Action: emotion: 100% | 1/1 [00:00<00:00, 6.76it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 11.54it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 16.72it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 18.29it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 19.13it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 15.52it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 17.17it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 17.96it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 18.24it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 18.47it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 18.12it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 18.33it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 17.77it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 17.97it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 17.82it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 17.59it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 17.54it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 18.31it/s]
Action: emotion: 100% | 1/1 [00:00<00:00, 18.31it/s]
```