

# Emotion Detection by Applying CNN

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## Abstract

As the human-robot interaction is catching the eye day by day with the increase in need for automation in every field, personal robots are increasing in every area which may be the coping needs of elderly people, treating autistic patients, or child therapy, even in the area of babysitting the child.

In this project, we are going to build a model which can predict human emotion from an image in real-time on a module based on web design where we can explore user's emotions through surveys.

**Keywords:** Artificial Intelligence, Machine Learning, Deep Learning, Convolutional Neural Network, JavaScript, FER 2013, Facial Expressions, Python, Kaggle, HTML, Web Design

## Introduction

### Emotion Detection using CNN (Machine Learning)

This project can

1. Extract and leverage the facial expressions
2. Gather the information of the facial language traits (i.e., facial posture, facial gestures/movements etc.)

The detection and observation of emotions are important in various situations i.e., to enable naturalistic and personalized human-computer interaction.

Emotion detection often requires modeling of various data inputs such as facial expressions, e.g., anger, sadness, happiness, etc.

From data collected by this technology, provides a unique opportunity for some unimaginable ways in recognising human activities through human physiological sensing.

This could potentially provide better management of chronic diseases such as asthma, diabetes and cardiovascular diseases

Emotion recognition can also be useful in variety of areas such as image processing, cyber security, robotics, psychological studies, and virtual reality applications.

## Objectives

- Project objectives are what you plan to achieve by the end of your project.
- This might include deliverables and assets, or more intangible objectives like increasing productivity or motivation.
- By the end of this project we are going to visualize the emotions of the user through a website module
- For which we are using software like Kaggle, Keras, etc. programming languages like python,

JavaScript, etc. markup languages like HTML and also datasets (FER 2013).

- By using our project we can process images and videos in real-time for monitoring video feeds or automating videos analytics which saves money and makes a better life for necessities.

### Scope of the Project

- The project we are doing is a web design model where user can know their state of mind in a survey mode through a web module.
- In which they are going to select seven random emotions in the website where they act like emotion triggers capturing the facial expressions of the user.
- With a survey module we will show the user their emotional results.
- Whether the user is a sociopath or psychopath by explaining the percentage through an emotion detector.

### Literature Survey

Title	Authors	Year of Publication
An efficient deep learning technique for facial emotion recognition	Asad Khattak, Muhammad Zubair Asghar, Mushtaq Ali, Ulfat Batool	2022
Facial Feature-Based Human Emotion Detection Using Machine Learning	Mritunjay Rai, Agha Asim Husain, Rohit Sharma, Tanmoy Maity, R.K. Yadav	2022
Automated emotion recognition: Current trends and future perspectives	M. Maithri, U. Raghavendra, Anjan Gudigar, Jyothi Samanth, Prabal Datta Barua, Murugappan Murugappan, Yashas Chakole, U. Rajendra Acharya	2022

### Tools Used

- Artificial Intelligence
- Machine Learning
- Deep Learning
- Convolutional neural networks
- JavaScript
- FER 2013
- Facial expressions
- Python
- Kaggle
- HTML
- Web design

### Problem Statement

#### Problem

The main purpose of this project is to know the emotion of the user where it develops the human-computer interaction.

#### Problem Statement

- As robots are helping human beings in most of the cases, robots need to understand human emotion in order to treat human in a more customized manner.
- Predicting human emotion has been a difficult problem for robots which hadn't been solved over a decade's time.

- We are solving that problem by using convolutional neural networking's in which we can detect human emotion for greater use.

### **Objectives**

- For which we are using software like Kaggle, Keras, etc. programming languages like python, JavaScript, etc. markup languages like HTML and also datasets (FER 2013).
- By using our project we can process images and videos in real-time for monitoring video feeds or automating videos analytics which saves money and makes a better life for necessities.

### **Methodology**

This project helps to entertain the user in expressive way where they can find their own emotion.

### **Proposed Methodology**

As per the main methodology used in this project, we need 2 modules:

1. Emotion detector, which works on the back-end.
2. Web design module, which uses emotion detector.

For that, we need some libraries, softwares, languages, etc.

First, in the emotion detection, we use a dataset called FER (Facial Expression Recognition) to analyze the facial expressions (i.e., emotion-detection.py) for classifying the emotion with CNN using Keras on the Kaggle. After completion of the code, it will give us an h5 file to download. Later we use that file on the main python code as a dataset directory on PyCharm. We also use HAAR cascade frontal-face for the Object Detection Algorithm in the main code (main.py). After giving the directories and attaching the dataset to the (main.py) code it will display the result by showing the emotions by using a webcam

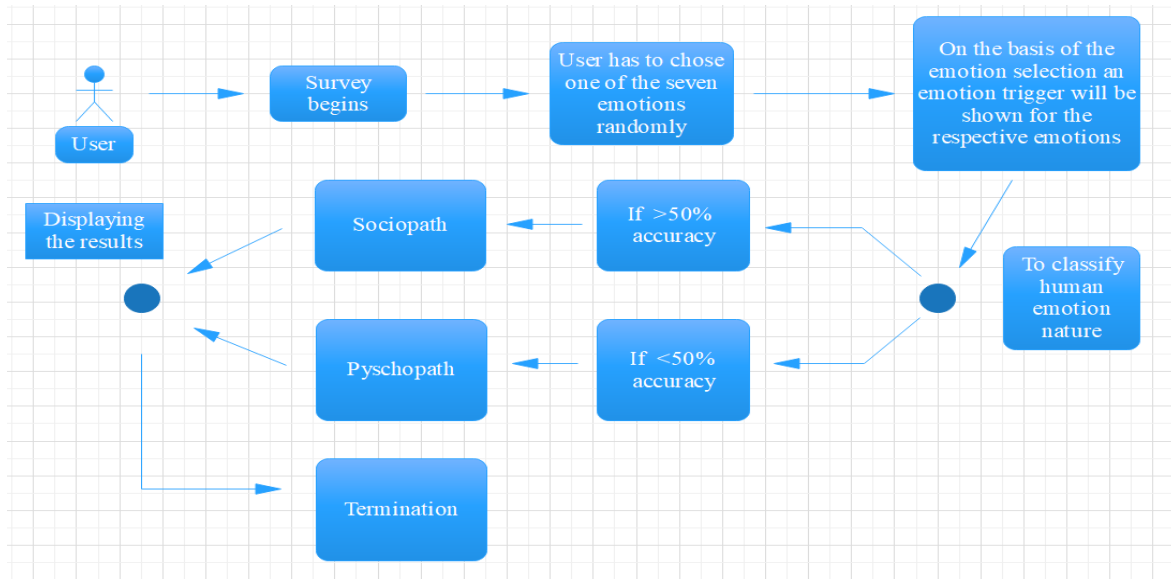
From there we build a web application by using the templates folder containing HTML files and web design, for implementing our second module. For that we need a library called Flask which helps you to provide tools and technologies that allow you to build a web application in python. By using flask, we write a code with i.e., app.py which also contains files to build a web app in python which uses emotion detection files from the module to run in the web application. By this, our second module will be completed and that is how we implement our project.

### **Architecture**

The main objective of this system architecture is to detect the emotion so the approach of that process goes like this: AI - ML - Deep Learning - CNN - Data Sets - Kaggle - Python - Web Design - Facial Recognition.

As can be seen, there are 9 stages to approach this project.

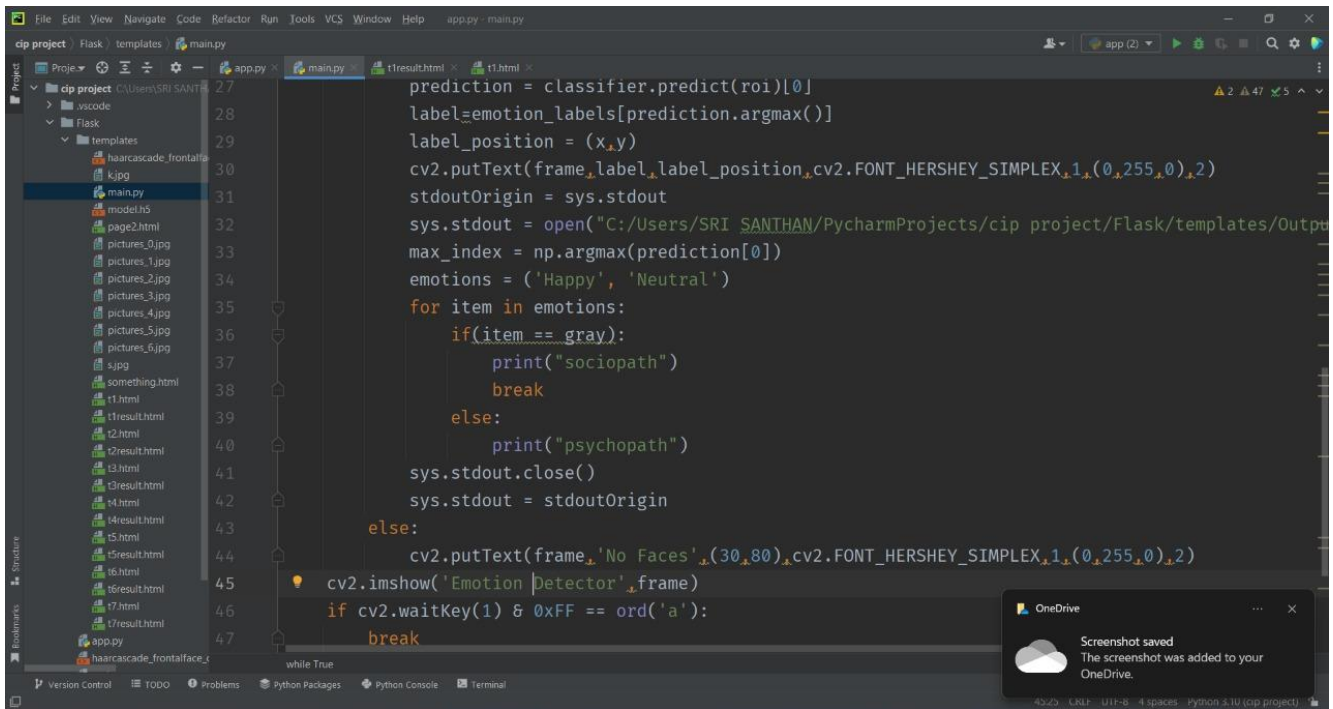
### UML Diagram



### Result Sample Coding

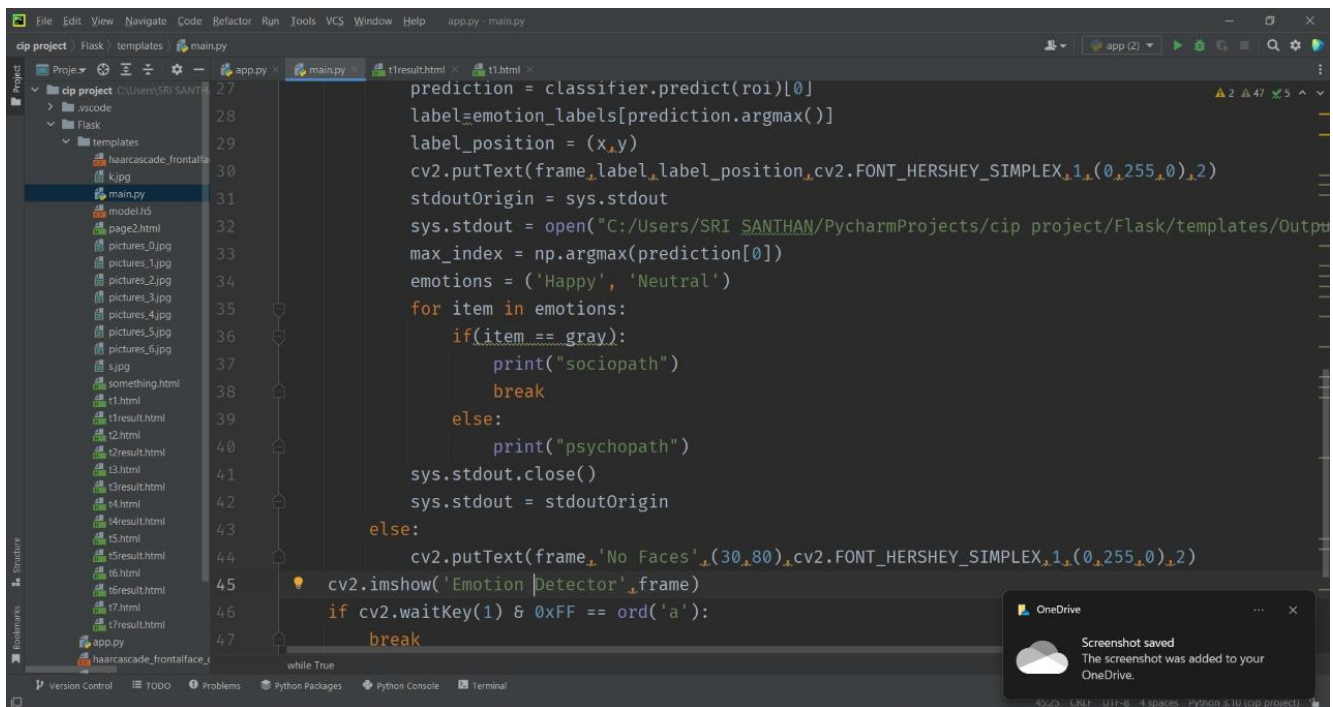
```
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cip project Flask templates main.py
Project
  cip project CAUsers\SRI SANTHAN
  |
  |__ vscod
  |__ Flask
  |__ templates
  |   |__ haarcascade_frontalface
  |   |__ k.jpg
  |   |__ main.py
  |   |__ model.h5
  |   |__ page2.html
  |   |__ pictures_0.jpg
  |   |__ pictures_1.jpg
  |   |__ pictures_2.jpg
  |   |__ pictures_3.jpg
  |   |__ pictures_4.jpg
  |   |__ pictures_5.jpg
  |   |__ pictures_6.jpg
  |   |__ s.jpg
  |   |__ something.html
  |   |__ t1.html
  |   |__ t1result.html
  |   |__ t2.html
  |   |__ t2result.html
  |   |__ t3.html
  |   |__ t3result.html
  |   |__ t4.html
  |   |__ t4result.html
  |   |__ t5.html
  |   |__ t5result.html
  |   |__ t6.html
  |   |__ t6result.html
  |   |__ t7.html
  |   |__ t7result.html
  |   |__ app.py
  |   |__ haarcascade_frontalface_x
  |
  |__ Structure
  |
  |__ Bookmarks
  |
  |__ Version Control
  |
  |__ TODO
  |
  |__ Problems
  |
  |__ Python Packages
  |
  |__ Python Console
  |
  |__ Terminal
  |
  |__ Event Log
  |
  |__ 32:25 CRIF UTF-8 4 spaces Python 3.10 (cip project)

1 from keras.models import load_model
2 from time import sleep
3 from keras.preprocessing.image import img_to_array
4 from keras.preprocessing import image
5 import cv2
6 import numpy as np
7 import sys
8 face_classifier = cv2.CascadeClassifier(r'C:\Users\SRI_SANTHAN\PycharmProjects\cip project\Flask\haa
9 classifier =load_model(r'C:\Users\SRI_SANTHAN\PycharmProjects\cip project\Flask\model.h5')
10 emotion_labels = ['Angry','Disgust','Fear','Happy','Neutral', 'Sad', 'Surprise']
11 cap = cv2.VideoCapture(0)
12 while True:
13     check_frame = cap.read()
14     labels = []
15     gray = cv2.cvtColor(frame,cv2.COLOR_BGR2GRAY)
16     faces = face_classifier.detectMultiScale(gray)
17     for (x,y,w,h) in faces:
18         cv2.rectangle(frame,(x,y),(x+w,y+h),(0,255,255),2)
19         roi_gray = gray[y:y+h,x:x+w]
20         roi_gray = cv2.resize(roi_gray,(48,48),interpolation=cv2.INTER_AREA)
21
```



```
File Edit View Navigate Code Refactor Run Tools VCS Window Help app.py - main.py
cip project / Flask / templates / main.py
Project
  cip project C:\Users\SRI SANTHAN
  vscode
  Flask
    templates
      haarcascade_frontalface
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      main.py
      model.h5
      page2.html
      pictures_0.jpg
      pictures_1.jpg
      pictures_2.jpg
      pictures_3.jpg
      pictures_4.jpg
      pictures_5.jpg
      pictures_6.jpg
      s.jpg
      something.html
      t1.html
      t1result.html
      t2.html
      t2result.html
      t3.html
      t3result.html
      t4.html
      t4result.html
      t5.html
      t5result.html
      t6.html
      t6result.html
      t7.html
      t7result.html
      app.py
      haarcascade_frontalface_x
      while True
Python Console
Terminal
prediction = classifier.predict(roi)[0]
label=emotion_labels[prediction.argmax()]
label_position = (x,y)
cv2.putText(frame,label,label_position,cv2.FONT_HERSHEY_SIMPLEX,1,(0,255,0),2)
stdoutOrigin = sys.stdout
sys.stdout = open("C:/Users/SRI SANTHAN/PycharmProjects/cip project/Flask/templates/Output.txt", "a")
max_index = np.argmax(prediction[0])
emotions = ('Happy', 'Neutral')
for item in emotions:
    if(item == gray):
        print("sociopath")
        break
    else:
        print("psychopath")
sys.stdout.close()
sys.stdout = stdoutOrigin
else:
    cv2.putText(frame,'No Faces',(30,80),cv2.FONT_HERSHEY_SIMPLEX,1,(0,255,0),2)
cv2.imshow('Emotion Detector',frame)
if cv2.waitKey(1) & 0xFF == ord('a'):
    break
```

OneDrive  
Screenshot saved  
The screenshot was added to your OneDrive.



```
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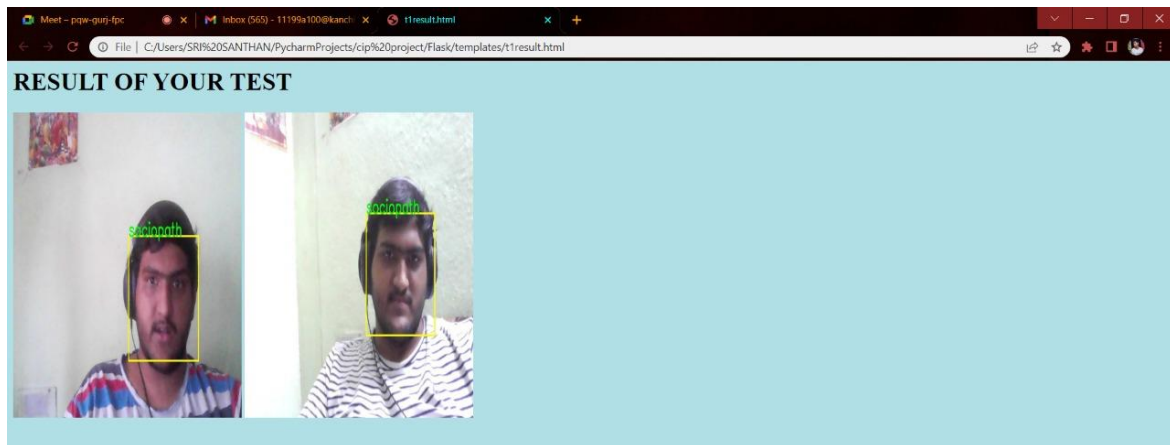
```
app.py x t1result.html x t1.html x
22
23 @app.route('/video1',methods=['get','post'])
24 def video1():...
72
73 @app.route('/video2',methods=['get','post'])
74 def video2():...
123
124 @app.route('/video3',methods=['get','post'])
125 def video3():...
174
175 @app.route('/video4',methods=['get','post'])
176 def video4():...
225
226 @app.route('/video5',methods=['get','post'])
227 def video5():...
275
276 @app.route('/video6',methods=['get','post'])
277 def video6():...
326
327 @app.route('/video7',methods=['get','post'])
328 def video7():...
```

```
app.py x t1result.html x t1.html x
384 @app.route('/third',methods=['get','post'])
385 def third():
386     return render_template("t2.html")
387
388
389 @app.route('/fourth',methods=['get','post'])
390 def fourth():
391     return render_template("t2result.html")
392
393 @app.route('/fifth',methods=['get','post'])
394 def fifth():
395     return render_template("t3.html")
396
397 @app.route('/sixth',methods=['get','post'])
398 def sixth():
399     return render_template("t3result.html")
400
401 @app.route('/seventh',methods=['get','post'])
402 def seventh():
403     return render_template("t4.html")
404
```

```
app.py x t1result.html x t1.html x
405 @app.route('/eighth',methods=['get','post'])
406 def eight():
407
408     return render_template("t4result.html")
409
410 @app.route('/nine',methods=['get','post'])
411 def nine():
412     return render_template("t5.html")
413
414 @app.route('/tenth',methods=['get','post'])
415 def tenth():
416     return render_template("t5result.html")
417
418 @app.route('/eleven',methods=['get','post'])
419 def eleven():
420     return render_template("t6.html")
421
422 @app.route('/tweleve',methods=['get','post'])
423 def tweleve():
424     return render_template("t6result.html")
425
```

```
app.py x t1result.html x t1.html x
420     return render_template("t6.html")
421
422 @app.route('/tweleve',methods=['get','post'])
423 def tweleve():
424     return render_template("t6result.html")
425
426 @app.route('/thirteen',methods=['get','post'])
427 def thirteen():
428     return render_template("t7.html")
429
430 @app.route('/fourteen',methods=['get','post'])
431 def fourteen():
432     return render_template("t7result.html")
433
434 if __name__ == "__main__":
435     app.run(debug=True)
```

## Sample Output



This is how it is going to show the result.

## Conclusion

- Emotion detection is a great technology, which develops human-computer interaction, in many ways.
- Emotion recognition can also be useful in a variety of areas such as image processing, cyber security, robotics, psychological studies, and virtual reality applications.
- From data collected by this technology, provides a unique opportunity for some unimaginable ways in recognising human activities through human physiological sensing.

Hence, this is our project on emotion detection.

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### Emotion Detection

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