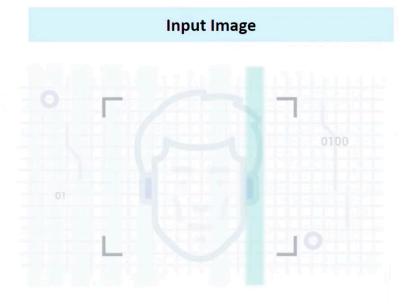


Emotion Recognition using Deep Learning

Web Application Development

By: Eli Daniels

Our objective is - to predict emotion for face images



Emotion Detection Model



Output



The Datasets:

FER2013 (Facial Expression Recognition 2013):

- FER2013 is one of the most widely used datasets for emotion recognition.
 It contains over 35,000 grayscale images of faces categorized into seven
 different emotions: anger, disgust, fear, happiness, sadness, surprise, and
 neutral.
- This dataset is a good starting point for research and experimentation due to its size and availability.

CK+ (Cohn-Kanade):

 CK+ is a dataset specifically designed for facial expression analysis. It includes over 500 image sequences of subjects displaying various facial expressions. It's often used for more detailed studies.

The Process:

Model Selection

Suitable deep learning architecture CNN

Data Preparation

Dataset: FER2013 (Facial Expression Recognition) and CK+ (Cohn-Kanade):

Preprocess the data, including resizing, normalizing pixel values

Model Training

Validation and Tuning:

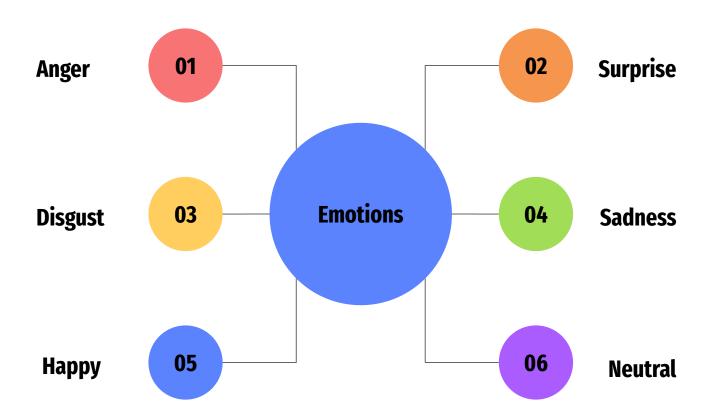
✓ Fine-tune hyperparameters, model architecture, and apply techniques like early stopping or learning rate scheduling.

- ✓ Initialize the chosen model.
- ✓ Compile the model with appropriate loss and optimization functions.
- Train the model using the training dataset for a specified number of epochs.

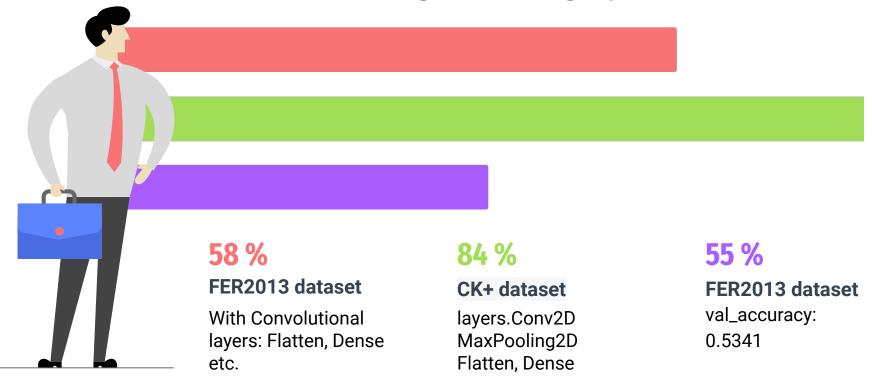




Emotional Breakdown



Emotional Intelligence Infographics



Algorithm: CK+ dataset, CNN, optimizer='adam', activation='softmax'

Quick Dive Into Metrics

Loss (Training vs.

• Training Loss: 0.5304

Validation):
 Validation Loss: 0.5220
 Accuracy (Training vs.
 Training Accuracy: 0.8418 (84.18%)

Validation):
Validation Accuracy: 0.8311 (83.11%)
The model's performance appears to be quite good.
It has learned from the training data and is generalized.

It has learned from the training data and is generalizing well to the validation data.
 The small difference between training and validation metrics indicates that the model is not suffering from significant overfitting or underfitting.

Demonstration





Emotion Scores:

angry: 0.005386296194046736

disgust: 0.013322117738425732

fear: 0.11439087241888046

happy: 0.258876234292984

sad: 0.010991943068802357

surprise: 0.15653066337108612

neutral: 0.35665422677993774

dominant emotion: It Depends

Real World Application!

1. Virtual Meetings:

- a. **Emotion Analysis in Virtual Meetings**: During virtual meetings, the program can help analyze participants' emotions in real-time. This information can be valuable for gauging engagement and adjusting the meeting's content accordingly.
- b. **Face Detection for Participant Highlighting**: You can use face detection to highlight the active speaker's face or participants' faces, improving the visual experience for attendees.

2. **Education**:

- a. Student Engagement: In online education, gauge students' engagement and understanding by analyzing their facial expressions.
- b. **Attention Tracking**: Monitor student attentiveness during virtual classes or lectures.

3. **Healthcare**:

- a. **Remote Patient Monitoring**: In telehealth, analyze patients' emotions and vital signs for remote diagnosis and treatment
- 4. Market Research:
 - a. **Consumer Sentiment Analysis**: Analyze facial expressions in focus groups or product testing to gauge consumer sentiment toward products or advertisements.
- 5. **Gaming**:
- **Game Interaction**: Create games that adapt based on the player's emotions, making the gaming experience more immersive

Pie Chart

