

# **Facial Expression Recognition Using Convolutional Neural Networks**

**Presented By Joey Navarro, Data Scientist**

# Agenda

- **Introduction**
- **Problem Statement**
- **Required Toolkit**
- **Dataset Used**
- **Neural Network Framework Used**
- **Data Preprocessing and Model Preparation**
- **Model Training Loop**
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# Introduction.

- Individuals on the autism spectrum often struggle understand others emotions through their facial expressions and other features.
- What if there was a way to detect emotion based off of facial expressions to help individuals on the autism spectrum identify how others feel?



# Problem Statement

Utilize pre-classified images to train a model for the purpose of facial expression recognition using a deep convolutional neural network.

## Why use a deep convolutional neural network for this task?

Deep learning excels in recognizing objects in images as it's implemented using 3 or more layers of artificial neural networks where each layer is responsible for extracting one or more feature of the image.

# What I Need - Toolkit

- Pre-Classified Data (CK+ Data Set)
- Neural Network Framework (Pytorch)
- OpenCV (Computer Vision Library)
- Matplotlib library
- Numpy library
- Model Architecture (VGG19 & ResNet18)
- NVIDIA CUDA Capable Graphics Card with at least 6GB Memory
- Windows or Linux OS Computer



# Pre-Classified Data - Cohn-Kanade+ (CK+) Data Set

- Compiled by Jeffrey Cohn, and Takeo Kanade at Carnegie Mellon University Robotics Institute and The University of Pittsburgh Department of Psychology Respectively for Disney Research.
- The CK+ dataset is an extension of the CK dataset.
- The original dataset contains 327 labeled facial videos.
- The dataset used here is composed of the last three frames from each video in the CK+ dataset, therefore it contains a total of 981 facial expressions.
- Not original dataset used in project modeling.



# Neural Network Framework - Pytorch

1. PyTorch is Pythonic
2. Easy to learn
3. Higher developer productivity
4. Easy debugging
5. Data Parallelism



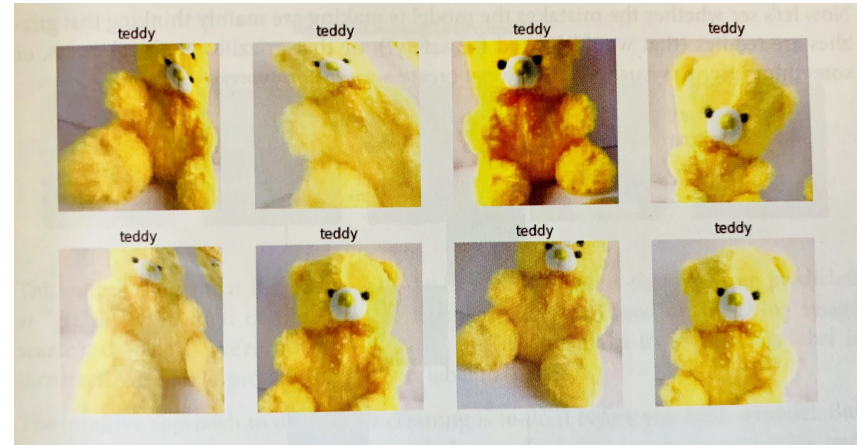
## NVIDIA CUDA Capable Graphics Card with at least 6GB Memory

EVGA GeForce GTX 1660 SC Ultra



# Data Preprocessing and Preparing Model

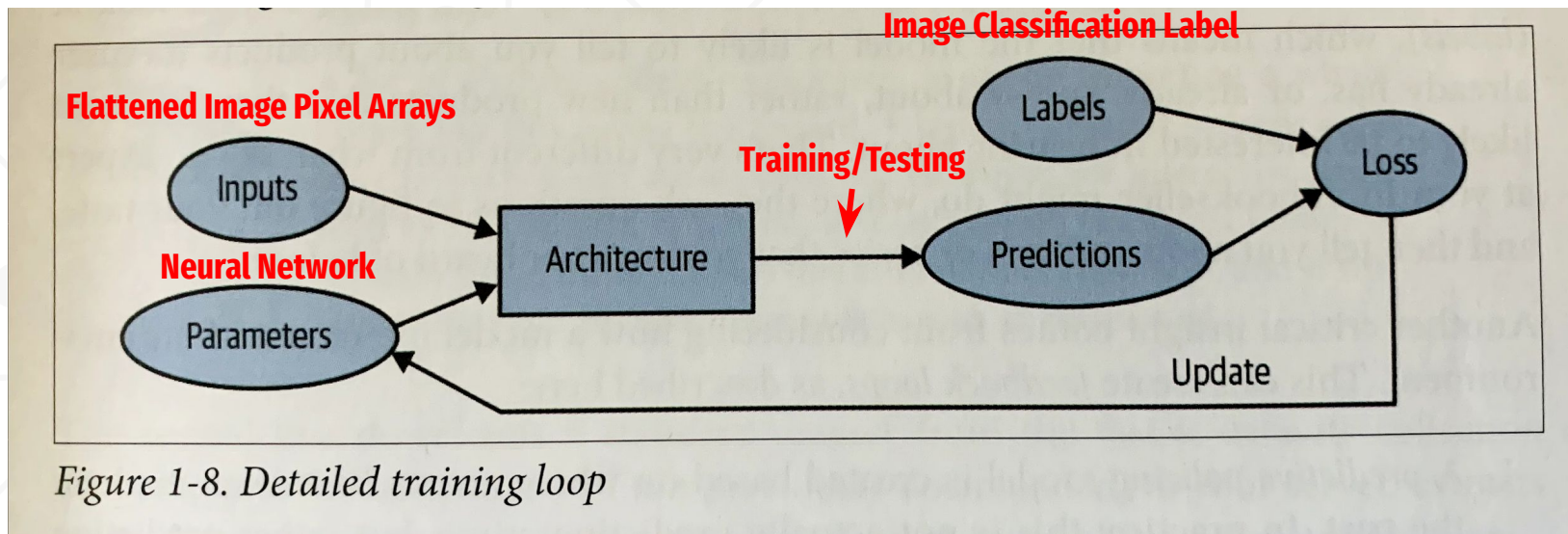
- Read images as pixel arrays for 48 x 48 pixel images for each emotion.
- Flattened the arrays into an integer list.
- Scaled data by a factor of 3 (limited by GPU memory).
- Create training dataset from a random  $\frac{1}{3}$  of all images.
- Setup augmentation by creating random variation transformations.
- Set up 10 Fold K-Fold cross validation.



Example of random image data transformations

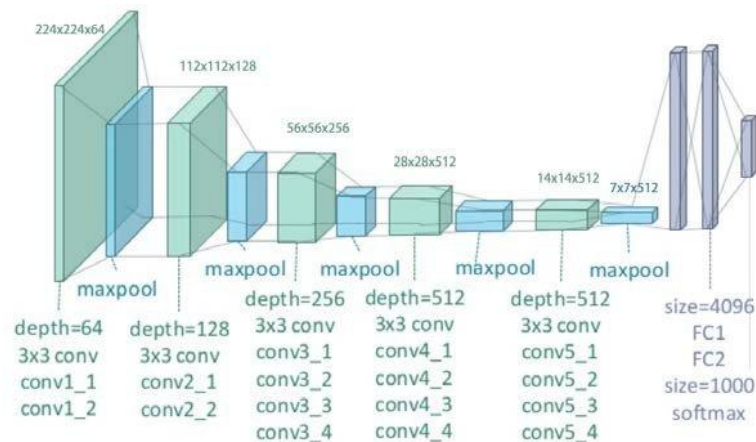
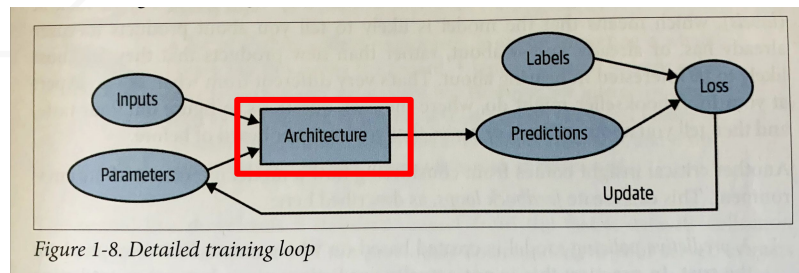


# Model Training Loop



# Model Architecture - VGG19

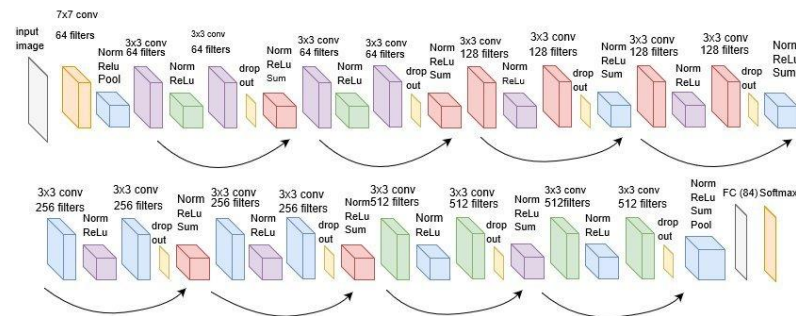
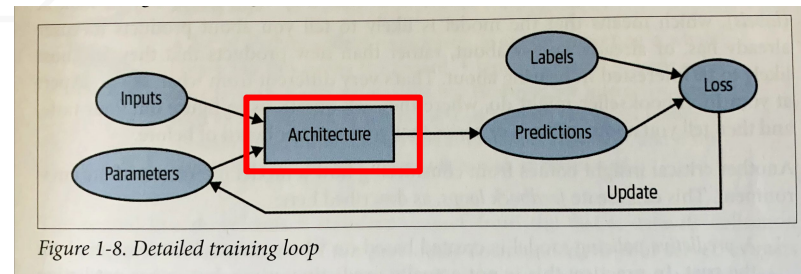
- This architecture's name comes from its founding group at Oxford, namely Visual Geometry Group (VGG).
- The 19 after the name stands for the number of layers in the network.
- This network can be used for transfer learning, therefore it may also be used for facial recognition tasks.
- It is built by stacking convolutions together but the model's depth is limited because of an issue called diminishing gradient. This issue makes deep convolutional networks difficult to train.



VGG19 Architecture

# Model Architecture - ResNet18

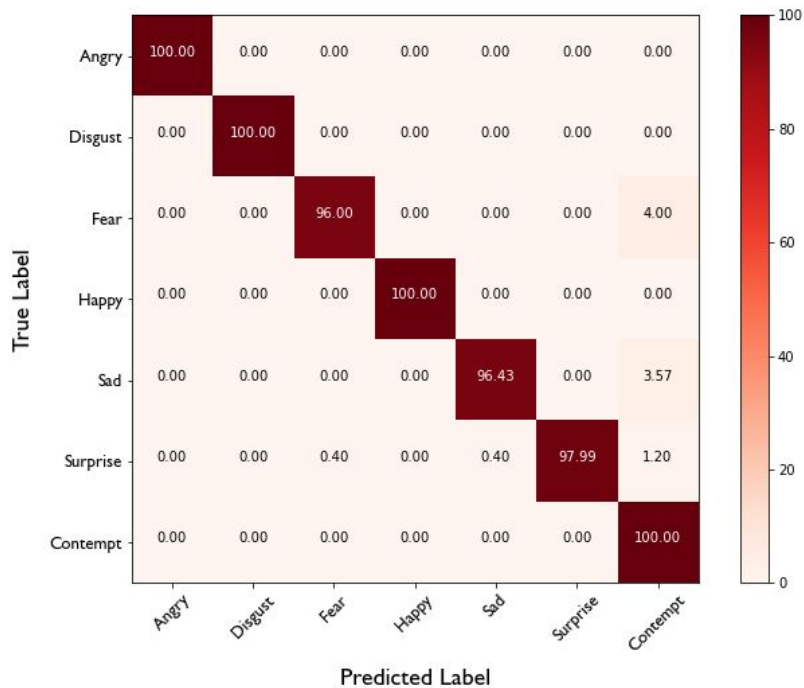
- This architecture's name comes from the functions used by the network, namely residual learning functions or Residual learning Network (ResNet).
- The 18 after the name stands for the number of layers in the network. Resnet model was proposed to solve the issue of diminishing gradient.
- The idea is to skip the connection and pass the residual to the next layer so that the model can continue to train.
- With Resnet models, CNN models can go deeper and deeper.



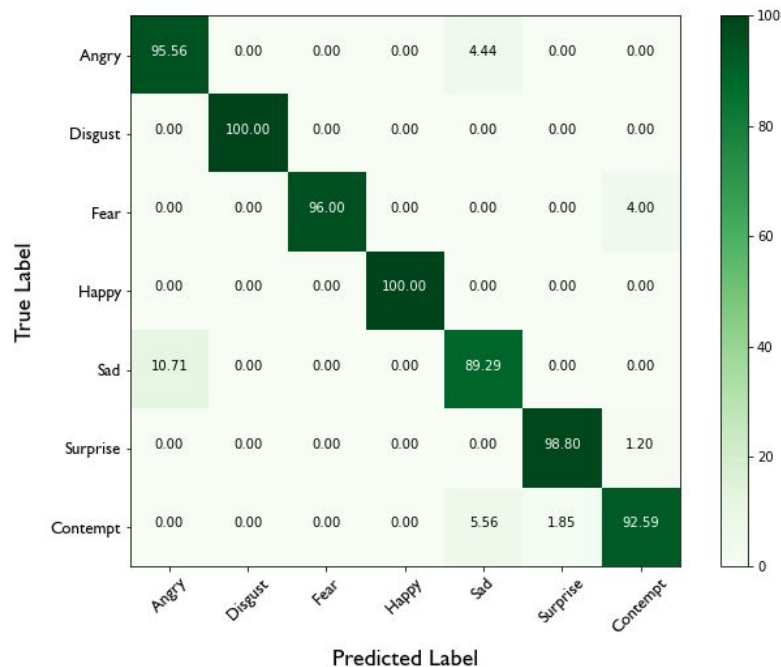
ResNet18 Architecture

# Running the Models - Confusion Matrices

VGG19 Convolutional Neural Network Model  
Normalized Confusion Matrix (Model Accuracy: 99.185%)

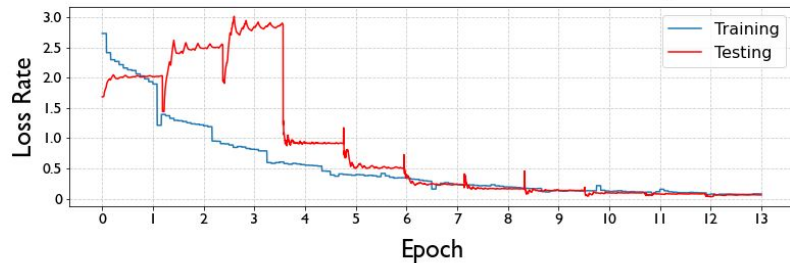


ResNet18 Convolutional Neural Network Model  
Normalized Confusion Matrix (Model Accuracy: 97.452%)

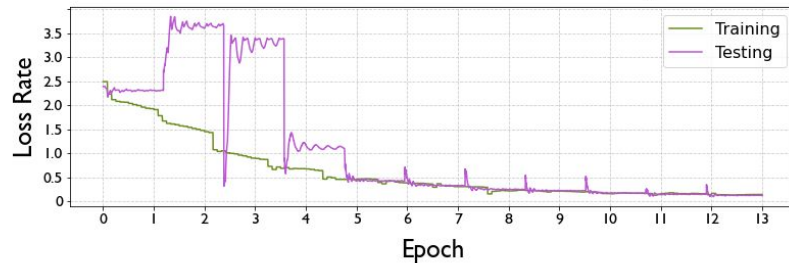


# Running the Models - Loss

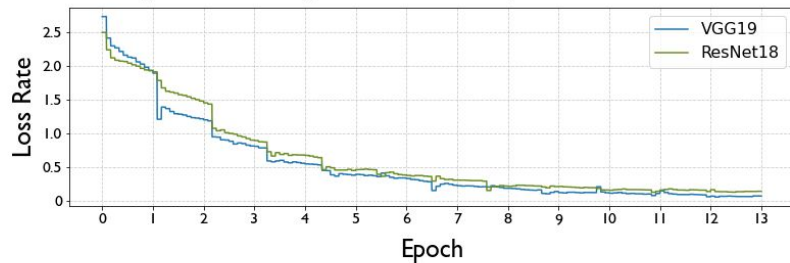
VGG19 Model Loss On CK+ Data For 12 Epochs



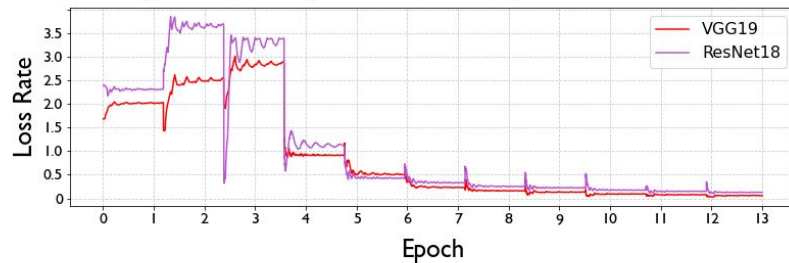
ResNet18 Model Loss On CK+ for 12 Epochs



Model Loss on CK+ Training Data For 12 Epochs  
[Lowest Loss] VGG19: 0.063 | ResNet18: 0.133



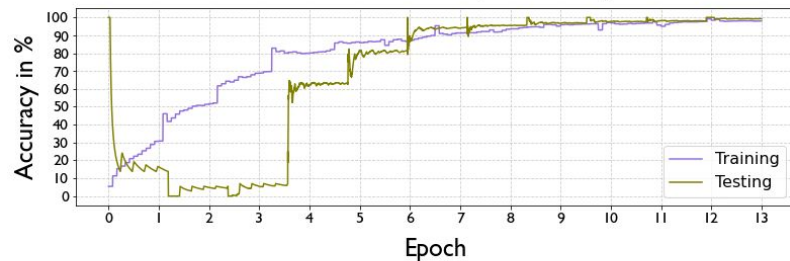
Model Loss on CK+ Testing Data For 12 Epochs  
[Lowest Loss] VGG19: 0.027 | ResNet18: 0.100



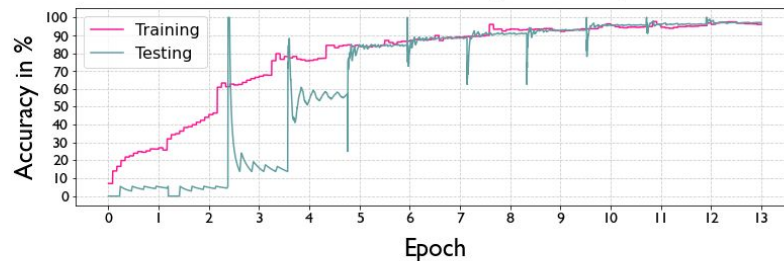


# Running the Models - Accuracy

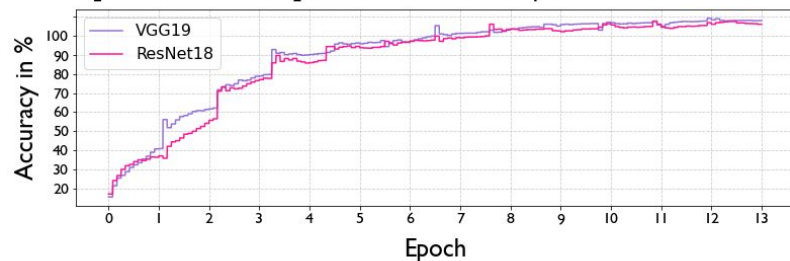
VGG19 Model Accuracy For 12 Epochs



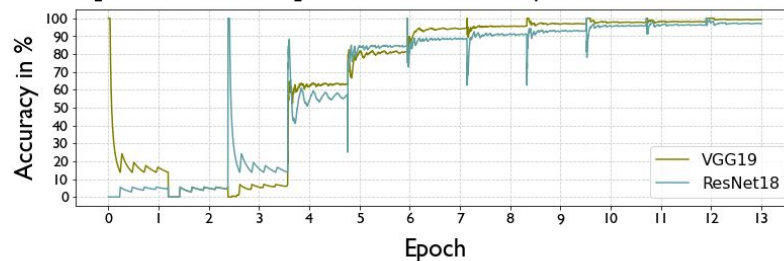
ResNet18 Model Accuracy For 12 Epochs



Model Accuracy on Training Data For 12 Epochs  
[Best Accuracies] VGG19: 99.219% | ResNet18: 97.656%



Model Accuracy on Test Data For 12 Epochs  
[Best Accuracies] VGG19: 99.185% | ResNet18: 97.452%

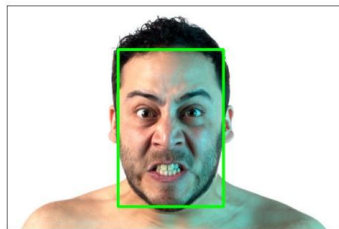


# Prediction - Example

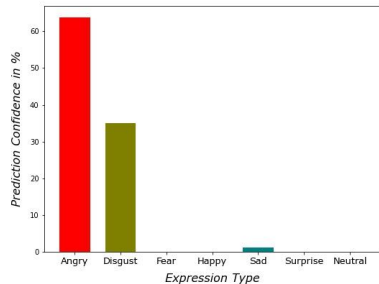
Original Image



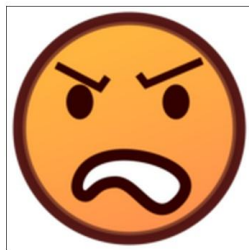
Image Face Detection



Expression Prediction Scores



Emoji Expression



**Angry Face is Being Expressed**

Original Image

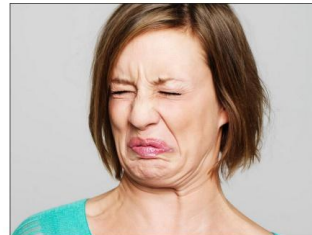
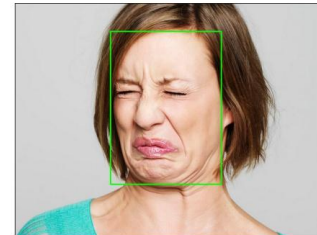
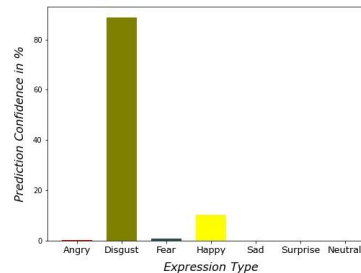


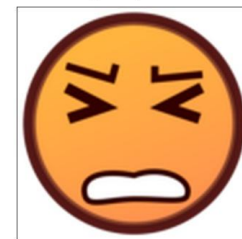
Image Face Detection



Expression Prediction Scores



Emoji Expression



**Disgust Face is Being Expressed**

# Prediction - Example

Original Image



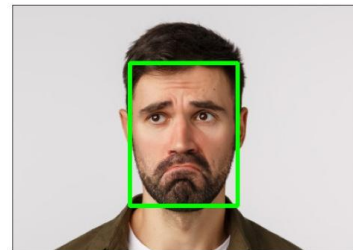
Image Face Detection



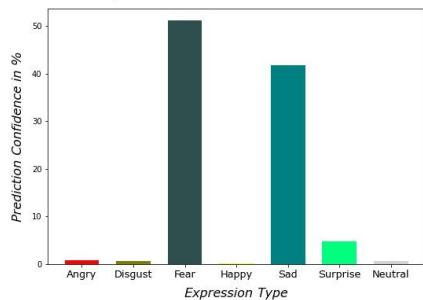
Original Image



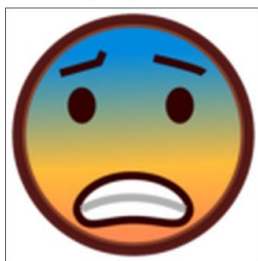
Image Face Detection



Expression Prediction Scores

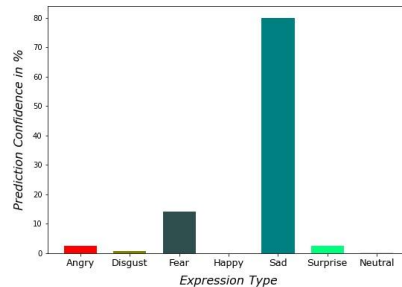


Emoji Expression

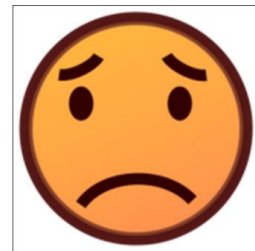


**Fear Face is Being Expressed**

Expression Prediction Scores



Emoji Expression



**Sad Face is Being Expressed**

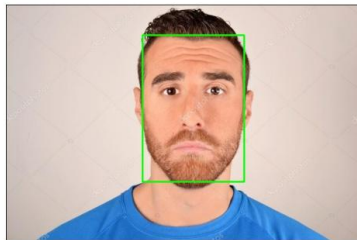


# Prediction - Example

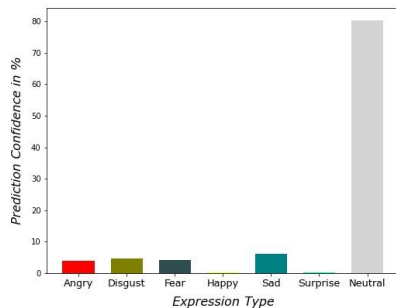
Original Image



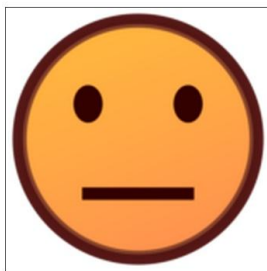
Image Face Detection



Expression Prediction Scores



Emoji Expression

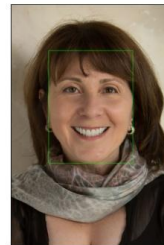


**Neutral Face is Being Expressed**

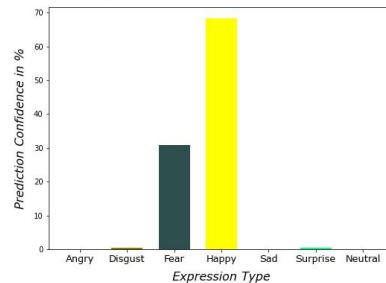
Original Image



Image Face Detection



Expression Prediction Scores



Emoji Expression



**Happy Face is Being Expressed**

# Conclusion

- The CK+ dataset proved to be a very useful dataset to use for creating a model using deep convolutional neural networks.
- Both VGG19 and ResNet18 Model Architecture led to the creation of very accurate models with low loss rates.
- The models were not perfect as both neural networks struggled slightly when distinguishing between some emotions. VGG19 struggled slightly when learning disgust, sadness, and surprise.
- ResNet18 struggled slightly when learning anger, disgust, fear, sadness, surprise, and contempt. Therefore, when deciding which model to use for making expression predictions, VGG19 was the natural choice.
- When predicting expressions the model struggled when the image did not have a plain background and it also struggled to distinguish between subtly different emotions when in practice on real world images.
- Future work would focus on obtaining a larger dataset with more training data that shows a bit more variation in the expressions for each emotion and also creating a webcam app to use in real time for individuals such as my sister-in-law.



**Questions?**