**Application of Convolutional Neural Networks Towards the Classification of Human Emotions from Images:**

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**Purpose:**

The purpose of this project is to utilize a Convolutional Neural Network (CNN) to classify the emotions shown on a human face. We have several large datasets that will be used to train the model and then the model will be applied to an image captured once the program detects a face. The program will then display this capture along with the emotion it believes to be most prevalently shown in the image. We implemented this algorithm to work as quickly as possible so that there would only be a short delay between the image capture and the final result displayed.

**Related Work:**

A. Verma et. al used a similar convolutional neural network in order to detect the emotions on the human face. Their network consisted of two convolution layers and two pooling layers and then those layers are flattened and then fully connected in the following layers. They also looked at several different types of architectures in their experiment including rectangular, modified triangular, and Venturi. They showed that the Venturi architecture had the best results with a training accuracy of 98.87%, a training loss of .0224, a testing accuracy of 86.78%, and a testing loss of .9693. For each of the seven emotions they classified their confusion matrix showed that each category had above an 85% correct classification.

M. Zadeh et. al. also used a CNN to classify the emotions on a human face. Their sample set was images of Japanese women showing different emotions. Where their method differed from Verma is that they used two Gabor filters to preprocess the image before feeding it into the CNN. The Gabor filters were used to filter the image by giving a higher response to points where the texture changes and at edges. This allowed for a larger focus to be placed on the more important features of the face that could then be used to determine the emotion. Applied two of these filters lead to a marked improvement in the performance of the CNN across all the epochs that they tested, ending with a thirty-epoch run that gave them an accuracy of 97.16% over the base CNN which had an accuracy of 91.16%. They also showed that this method drastically reduces the time across the higher accuracy ranges.

**Method:**

For our implementation, we used PyTorch and a modified Alexnet architecture for our convolutional neural network. The Alexnet architecture consists of 5 convolutional layers, 3 max pooling layers, and 3 fully connected layers. The only difference between this implementation and Alexnet was that the network was trained on grayscale images. Once they were converted to greyscale, we then resized the data to 224x224 pixels to fit the model, and finally converted that image into a tensor for the CNN to process. The dataset we used was titled “facial emotion recognition” and can be found at the provided link: <https://www.kaggle.com/chiragsoni/ferdata>. Once the data was obtained, we used cross entropy loss to evaluate how far off our prediction was and to change the values. We then trained our data over X epochs and the data trained over X minutes. On the provided test set, our algorithm performed at X percent accuracy.

**EUGENE IMPLEMENTATION**

The facial detection and image capture portion of our program works hand-in-hand with the CNN by capturing an image when a face is detected and passing that image to the CNN to be processed. The same preprocessing done to the test set is done to this image, which includes converting it to greyscale, trimming the image, and then converting it into a tensor. The CNN is then able to take this data and predict the emotion displayed on the face from a live camera feed. This prediction is displayed…

**Experiments and Results:**

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| Number of Epochs: | Test Accuracy: |
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From the data above one can see that we were not able to meet the same accuracy metrics as the relevant work. We saw a drastic decrease in the accuracy when we went from 40 epochs to 100, however this was to be expected since we were overfitting the data. We settled on using X epochs because that gave us the highest accuracy out of the ones that we tested.

What we were able to do was integrate the camera system of our device into the neural network to generate a prediction. This is a new and novel way for the user to interface with the machine and the CNN. While the accuracy of the predictions was lower than we had hoped, we still were able to integrate our facial detection and image capture system into the CNN to generate predictions.

**Bibliography:**

A. Verma, P. Singh and J. S. Rani Alex, "Modified Convolutional Neural Network Architecture Analysis for Facial Emotion Recognition," 2019 International Conference on Systems, Signals and Image Processing (IWSSIP), 2019, pp. 169-173, doi: 10.1109/IWSSIP.2019.8787215.

M. M. Taghi Zadeh, M. Imani and B. Majidi, "Fast Facial emotion recognition Using Convolutional Neural Networks and Gabor Filters," 2019 5th Conference on Knowledge Based Engineering and Innovation (KBEI), 2019, pp. 577-581, doi: 10.1109/KBEI.2019.8734943.