# Assignment 4 YuTing Chiu

# Task1:

#### For Task1 model1:

The dataset is split into training (80%) and validation (20%) subsets by specifying validation split=0.2

The CNN model is created using Sequential, with the following layers:

1. **Input Layer:** Accepts images of shape (100, 100, 3) (3 channels for RGB).

#### 2. Convolutional Layers (Conv2D):

Extract spatial features using 32 and 64 filters of size (3, 3), both activated by ReLU.

#### 3. MaxPooling Layers (MaxPooling2D):

Reduce spatial dimensions using a pool size of (2, 2) after each convolutional layer.

#### 4. Dropout Layers:

Add dropout (20%, 30% rates) after convolution and pooling layers to prevent overfitting.

#### 5. Flatten Layer:

Flatten the 2D feature maps into 1D arrays for fully connected layers.

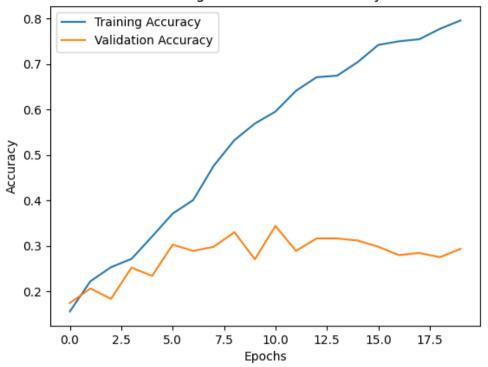
#### 6. Dense Layers:

A fully connected layer with 128 neurons and ReLU activation.

Another dropout (50%) to further reduce overfitting.

The output layer with num\_classes neurons and softmax activation for multi-class classification.

#### Training and Validation Accuracy



```
Epoch 1/20
                        - 103ms/step - accuracy: 0.1558 - loss: 179.7453 - val_accuracy: 0.1743 - val_loss: 1.7760
 28/28 - 3s
Epoch 2/20
28/28 - 1s
Epoch 3/20
28/28 - 1s
                            51ms/step - accuracy: 0.2222 - loss: 1.7919 - val_accuracy: 0.2064 - val_loss: 1.7806
                        - 52ms/step - accuracy: 0.2532 - loss: 1.7596 - val accuracy: 0.1835 - val loss: 1.7754
 Epoch 4/20
4/20
26/28 - 1s
Epoch 5/20
28/28 - 1s
Epoch
                            50ms/step - accuracy: 0.2715 - loss: 1.7025 - val_accuracy: 0.2523 - val_loss: 1.7346
                            51ms/step - accuracy: 0.3207 - loss: 1.6533 - val accuracy: 0.2339 - val loss: 1.7390
 Epoch 6/20
                        - 51ms/step - accuracy: 0.3711 - loss: 1.5478 - val_accuracy: 0.3028 - val_loss: 1.7468
 28/28 - 1s
Epoch 7/20
28/28 - 1s
Epoch 8/20
28/28 - 1s
Epoch
                            50ms/step - accuracy: 0.4009 - loss: 1.4490 - val_accuracy: 0.2890 - val_loss: 1.7803
                        - 49ms/step - accuracy: 0.4765 - loss: 1.3363 - val_accuracy: 0.2982 - val_loss: 1.8350
 Epoch 9/20
20/28 - 1s -
Epoch 10/20
28/28 - 1s
Epoch
                            50ms/step - accuracy: 0.5326 - loss: 1.2099 - val_accuracy: 0.3303 - val_loss: 1.8207
28/28 - 1s -
Epoch 11/20
                            51ms/step - accuracy: 0.5693 - loss: 1.1009 - val_accuracy: 0.2706 - val_loss: 2.0038
 28/28 - 1s -
Epoch 12/20
                            50ms/step - accuracy: 0.5956 - loss: 1.0629 - val_accuracy: 0.3440 - val_loss: 1.9941
 28/28
                            50ms/step - accuracy: 0.6415 - loss: 0.9684 - val_accuracy: 0.2890 - val_loss: 2.3610
             - 1s - 13/20
 Epoch
                            50ms/step - accuracy: 0.6712 - loss: 0.8820 - val_accuracy: 0.3165 - val_loss: 2.4312
 28/28
 Epoch 14/20
Epoch 15/20
28/28 - 1s
                            50ms/step - accuracy: 0.6747 - loss: 0.8492 - val_accuracy: 0.3165 - val_loss: 2.4752
                            50ms/step - accuracy: 0.7045 - loss: 0.8367 - val_accuracy: 0.3119 - val_loss: 2.4860
             16/20
 Epoch
Epoch 17/20
28/28 - 1s
Epoch 17/20
                            50ms/step - accuracy: 0.7423 - loss: 0.7342 - val_accuracy: 0.2982 - val_loss: 2.8075
                            50ms/step - accuracy: 0.7503 - loss: 0.6637 - val_accuracy: 0.2798 - val_loss: 3.0017
 28/28 - 1s
Epoch 18/20
                            51ms/step - accuracy: 0.7549 - loss: 0.6949 - val_accuracy: 0.2844 - val_loss: 2.9286
 28/28
 Epoch 19/20
28/28 - 1s - 50ms/step - accuracy: 0.7961 - loss: 0.6216 - val_accuracy: 0.2936 - val_loss: 3.4041
7/7 - 0s - 10ms/step - accuracy: 0.2936 - loss: 3.4041
Test Accuracy: 0.294
Model saved everges for the control of th
20/28 - 1s -
Epoch 20/20
28/28 - 1s -
 Model saved successfully.
```

#### For Task1 model2:

A simple CNN is built with the following layers:

1. **Input Layer:** Accepts images of size (224, 224, 3) (RGB format).

#### 2. Convolutional Layers (Conv2D):

Two convolutional layers with 32 and 64 filters respectively, each using a kernel size of (3, 3) and ReLU activation.

# 3. MaxPooling Layers (MaxPooling2D):

Pooling layers with size (2, 2) to downsample spatial dimensions.

# 4. Dropout Layers:

Dropout (20% and 30%) to reduce overfitting by randomly deactivating neurons during training.

# 5. Flatten Layer:

Converts the 2D feature maps into 1D vectors for the Dense layers.

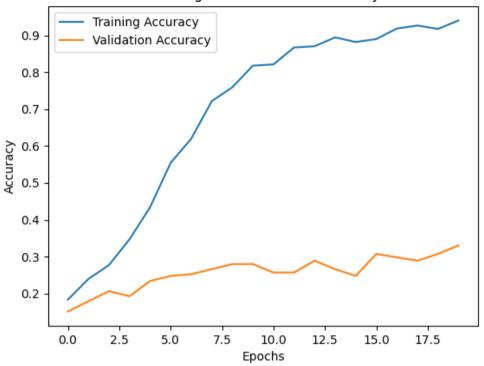
#### 6. Dense Layers:

A fully connected layer with 128 neurons and ReLU activation.

Another dropout (50%) to further reduce overfitting.

The output layer has neurons equal to num\_classes with softmax activation for multi-class classification.

#### Training and Validation Accuracy



```
26/28 - 10s
Epoch 2/20
28/28 - 0s
Epoc'
Epoch 1/20
              - 375ms/step - accuracy: 0.1833 - loss: 467.2643 - val_accuracy: 0.1514 - val_loss: 1.7957
28/28 - 9s
Epoch 3/20
               321ms/step - accuracy: 0.2394 - loss: 1.7817 - val accuracy: 0.1789 - val loss: 1.7931
28/28 - 9s
Epoch 4/20
             - 314ms/step - accuracy: 0.2772 - loss: 1.7135 - val_accuracy: 0.2064 - val_loss: 1.7952
28/28 - 9s
Epoch 5/20
28/28 - 0
Epoch
             - 313ms/step - accuracy: 0.3471 - loss: 1.5875 - val_accuracy: 0.1927 - val_loss: 1.8313
             - 311ms/step - accuracy: 0.4341 - loss: 1.4085 - val_accuracy: 0.2339 - val_loss: 1.8488
Epoch 6/20
28/28 - 9s
Epoch 7/20
28/28 - 9s
Epoc<sup>1</sup>
             - 317ms/step - accuracy: 0.5544 - loss: 1.2337 - val accuracy: 0.2477 - val loss: 1.9582
             - 315ms/step - accuracy: 0.6197 - loss: 1.0640 - val accuracy: 0.2523 - val loss: 2.1455
Epoch 8/20
               308ms/step - accuracy: 0.7216 - loss: 0.7745 - val_accuracy: 0.2661 - val_loss: 2.3379
28/28
       - 98
Epoch 9/20
20/28 - 9s -
Epoch 10/20
28/28 - 9°
Epoch
               314ms/step - accuracy: 0.7595 - loss: 0.6876 - val_accuracy: 0.2798 - val_loss: 2.8887
               314ms/step - accuracy: 0.8179 - loss: 0.6166 - val accuracy: 0.2798 - val loss: 2.7779
28/28 - 9s
Epoch 11/20
                339ms/step - accuracy: 0.8213 - loss: 0.5115 - val_accuracy: 0.2569 - val_loss: 3.3389
28/28
Epoch 12/20
28/28
               325ms/step - accuracy: 0.8671 - loss: 0.4002 - val_accuracy: 0.2569 - val_loss: 3.4533
       - 9s - 13/20
Epoch
28/28
               315ms/step - accuracy: 0.8706 - loss: 0.4281 - val_accuracy: 0.2890 - val_loss: 3.9898
28/28 - 9s
Epoch 14/20
28/28
               321ms/step - accuracy: 0.8946 - loss: 0.3645 - val accuracy: 0.2661 - val loss: 4.0241
28/28 - 9s -
Epoch 15/20
28/28
               315ms/step - accuracy: 0.8820 - loss: 0.3838 - val_accuracy: 0.2477 - val_loss: 3.0567
Epoch 16/20
               313ms/step - accuracy: 0.8900 - loss: 0.3393 - val_accuracy: 0.3073 - val_loss: 4.5842
28/28 - 9s
Epoch 17/20
20/28 - 9s -
Epoch 18/20
28/28 - 9s
Epoc'
               319ms/step - accuracy: 0.9187 - loss: 0.2915 - val_accuracy: 0.2982 - val_loss: 4.8731
         9s - 315ms/step - accuracy: 0.9267 - loss: 0.3638 - val_accuracy: 0.2890 - val_loss: 4.3165
Epoch 19/20
28/28 - 9s -
Epoch 20/20
             - 313ms/step - accuracy: 0.9175 - loss: 0.2862 - val_accuracy: 0.3073 - val_loss: 5.6307
                387ms/step - accuracy: 0.9404 - loss: 0.3087 - val accuracy: 0.3303 - val loss: 4.8635
28/28 - 11s - 387ms/step - accuracy: 0.9404 - loss: 0.7/7 - 0s - 62ms/step - accuracy: 0.3303 - loss: 4.8635 Test Accuracy: 0.330 Model saved successfully.
         11s
```

# Comparison Between Two Models

Input Shape	(100, 100, 3)	(224, 224, 3)
Convolutional	Conv2D with 32 filters	Conv2D with 32 filters and
Layers	and kernel size (3, 3)	kernel size (3, 3)
	Conv2D with 64 filters	Conv2D with 64 filters and
	and kernel size (3, 3)	kernel size (3, 3)
Pooling Layers	Two MaxPooling2D layers	Two MaxPooling2D layers
	with (2, 2) pooling.	with (2, 2) pooling.
Dropout Layers	Dropout layers after	Dropout layers after
	convolution and dense	convolution and dense layers
	layers with rates: 0.2, 0.3,	with rates: 0.2, 0.3, 0.5.
	0.5.	
Dense Layers	One fully connected layer	One fully connected layer with
	with 128 neurons (ReLU)	128 neurons (ReLU) and a
	and a softmax output	softmax output layer.
	layer.	
Output Layer	Softmax activation for	Softmax activation for multi-
	multi-class classification.	class classification.

Aspect	Model 1	Model 2
Training Accuracy	79.6%	94.0%
Validation Accuracy	29.3%	33.0%
Test Accuracy	29.4%	33.0%
Test Loss	3.4041	4.8635
Training Speed	Faster (~50ms/step)	Slower (~315ms/step)
Input Size	(100x100)	(224x224)

Model 2 shows slightly better performance in terms of validation and test accuracy (33% vs. 29.3% for Model 1) due to its higher input resolution, but it suffers from significantly higher computational cost and overfitting, similar to Model 1. Both models have limited generalization, as indicated by the low validation and test accuracy. Improving data preprocessing (e.g., augmentation) and using transfer learning could help boost performance while addressing overfitting. For resource

efficiency, Model 1 is preferable unless further optimizations are applied to Model 2.

# Task2:

#### (i)Pre-trained Model and Added Layers

# **Pre-trained Model:**

EfficientNetV2B3 with ImageNet weights.

**Model Size:** ~12 million parameters.

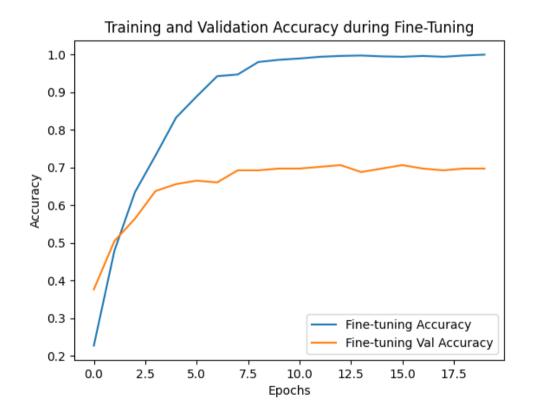
# **Added Layers:**

GlobalAveragePooling2D: To reduce spatial dimensions to a vector.

Dense (128 neurons): Fully connected layer with ReLU activation.

**Dropout** (rate=0.5): Regularization to prevent overfitting.

**Dense (output neurons = number of classes):** Softmax layer for classification.



```
1/20
- 114s - 4s/step - accuracy: 0.2268 - loss: 1.7780 - val_accuracy: 0.3761 - val_loss: 1.6326
28/28
28/28 - 57s - 2s/step - accuracy: 0.4788 - loss: 1.4774 - val_accuracy: 0.5046 - val_loss: 1.4783
Epoch 3/20
.
28/28 - 56s - 2s/step - accuracy: 0.6346 - loss: 1.1939 - val_accuracy: 0.5642 - val_loss: 1.3040
Epoch 4/20
28/28 - 56s - 2s/step - accuracy: 0.7320 - loss: 0.9454 - val_accuracy: 0.6376 - val_loss: 1.1510
Epoch 5/20
              2s/step - accuracy: 0.8328 - loss: 0.6715 - val_accuracy: 0.6560 - val_loss: 1.0422
     - 61s
28/28
Epoch 6/20
              2s/step - accuracy: 0.8889 - loss: 0.4776 - val_accuracy: 0.6651 - val_loss: 0.9760
     - 63s
Epoch 7/20
28/28 - 60s - 2s/step - accuracy: 0.9427 - loss: 0.3100 - val_accuracy: 0.6606 - val_loss: 0.9485
Epoch 8/20
              2s/step - accuracy: 0.9473 - loss: 0.2270 - val_accuracy: 0.6927 - val_loss: 0.9553
28/28 - 64s -
Epoch 9/20
              2s/step - accuracy: 0.9805 - loss: 0.1409 - val_accuracy: 0.6927 - val_loss: 0.9566
28/28
     - 62s
Epoch 10/20
28/28
              2s/step - accuracy: 0.9863 - loss: 0.1072 - val_accuracy: 0.6972 - val_loss: 0.9908
Epoch 11/20
     - 61s - 2s/step - accuracy: 0.9897 - loss: 0.0755 - val_accuracy: 0.6972 - val_loss: 1.0193
28/28
Epoch 12/20
             2s/step - accuracy: 0.9943 - loss: 0.0544 - val_accuracy: 0.7018 - val_loss: 1.0695
28/28 - 60s -
Epoch 13/20
             2s/step - accuracy: 0.9966 - loss: 0.0405 - val_accuracy: 0.7064 - val_loss: 1.0846
28/28
Epoch 14/20
28/28 - 62s - 2s/step - accuracy: 0.9977 - loss: 0.0398 - val_accuracy: 0.6881 - val_loss: 1.1785
Epoch 15/20
28/28 - 60s - 2s/step - accuracy: 0.9954 - loss: 0.0303 - val_accuracy: 0.6972 - val_loss: 1.1621
Epoch 16/20
28/28 - 61s - 2s/step - accuracy: 0.9943 - loss: 0.0327 - val_accuracy: 0.7064 - val_loss: 1.1764
Epoch 17/20
28/28
              2s/step - accuracy: 0.9966 - loss: 0.0300 - val_accuracy: 0.6972 - val_loss: 1.2491
Epoch 18/20
     - 62s - 2s/step - accuracy: 0.9943 - loss: 0.0262 - val_accuracy: 0.6927 - val_loss: 1.2777
28/28
Epoch 19/20
28/28 - 62s - 2s/step - accuracy: 0.9977 - loss: 0.0208 - val_accuracy: 0.6972 - val_loss: 1.3347
Epoch 20/20
28/28 - 61s - 2s/step - accuracy: 1.0000 - loss: 0.0164 - val_accuracy: 0.6972 - val_loss: 1.3286
7/7 - 3s - 365ms/step - accuracy: 0.6972 - loss: 1.3286
Test Accuracy after fine-tuning: 0.697
Model saved successfully
```

Model	Test Accuracy
Fine-Tuned Model (Task 2)	69.7%
Better Model (Task 1)	33.0%

The fine-tuned model using EfficientNetV2B3 significantly outperformed the Task 1 models, achieving a much higher test accuracy (69.7% vs. 33.0%). This improvement highlights the effectiveness of transfer learning, leveraging pre-trained features for better generalization. The added layers complemented the pre-trained model, adapting it to the specific dataset while preventing overfitting. This model is recommended for use, given sufficient computational resources.

# Task3:



Pic 1: Sad



Pic 2 happy



Pic 3 happy



Pic 4 happy



Pic 5 sad



Pic 6 happy



Pic 7 happy



Pic 8 happy



Pic 9 sad



#### Pic 10 happy

```
import image_test_task2_model
Model loaded successfully.
Class names: ['anger', 'happy', 'disgust', 'sad', 'pain', 'fear']
File: ./Facial Pics\l.jpg, Predicted Class: sad
File: ./Facial Pics\l.jpg, Predicted Class: pain
File: ./Facial Pics\l.jpg, Predicted Class: disgust
```

#### 1. Task 1 Model Errors

**Simpler Architecture:** Task 1 model is too simple to capture the subtle features of emotions, leading to frequent misclassifications like "Pain" or "Disgust."

**Poor Generalization:** It struggles to generalize well to unseen data, likely overfitting to the training set.

**Low Input Resolution:** Even with 224x224, the model lacks the capacity to extract detailed features needed for emotion recognition.

#### 2. Task 2 Model Errors

Class Bias: The Task 2 model predicts "Sad" for most images, possibly due to class imbalance in the training dataset.

**Pretrained Features Limitation:** Although it uses EfficientNetV2B3, which is pretrained on ImageNet, those features are better suited for object detection rather than subtle facial expressions.

#### Conclusion:

The original dataset may be too small to effectively train the models, resulting in both Task 1 and Task 2 models having low accuracy and struggling to distinguish images in the new dataset. The limited data likely caused insufficient learning and poor generalization, leading to biases and errors in predictions. Expanding the dataset with more diverse and balanced samples or using specialized emotion recognition datasets could significantly improve the models' performance.