

Soft Computing Digital Assignment (ITA6004)

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Q4. Use the following

datasets: https://www.kaggle.com/datasets/muratkokludataset/rice-image-dataset/code

Train the model using AlexNet, ResNet and VGGNet.

Code (All together):

```
# code cell 1
import os
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense,
Dropout
from tensorflow.keras.applications import ResNet50, VGG16
# code cell 2
data_directory = r'archive\train'
train datagen = ImageDataGenerator(rescale=1./255, validation split=0.3)
train_generator = train_datagen.flow_from_directory(
    data_directory,
    target size=(224, 224),
    batch size=32,
    class mode='categorical',
    subset='training'
validation generator = train datagen.flow from directory(
    data directory,
    target_size=(224, 224),
    batch_size=32,
    class mode='categorical',
    subset='validation'
# code cell 3
alexnet_model = Sequential()
alexnet_model.add(Conv2D(96, kernel_size=(11, 11), strides=(4, 4),
activation='relu', input_shape=(224, 224, 3)))
alexnet_model.add(MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))
alexnet_model.add(Conv2D(256, kernel_size=(5, 5), activation='relu'))
alexnet_model.add(MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))
```

```
alexnet_model.add(Conv2D(384, kernel_size=(3, 3), activation='relu'))
alexnet model.add(Conv2D(384, kernel size=(3, 3), activation='relu'))
alexnet model.add(Conv2D(256, kernel size=(3, 3), activation='relu'))
alexnet_model.add(MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))
alexnet model.add(Flatten())
alexnet model.add(Dense(4096, activation='relu'))
alexnet_model.add(Dropout(0.5))
alexnet_model.add(Dense(4096, activation='relu'))
alexnet model.add(Dropout(0.5))
alexnet_model.add(Dense(5, activation='softmax'))
alexnet_model.compile(optimizer='adam', loss='categorical_crossentropy',
metrics=['accuracy'])
alexnet history = alexnet model.fit(train generator,
validation data=validation generator, epochs=3)
# code cell 4
# Save the trained model
alexnet model.save('alexnet model.h5')
# code cell 5
saved_model = tf.keras.models.load_model('alexnet_model.h5')
# Test the model on new data
test_directory = r'archive\test'
test_datagen = ImageDataGenerator(rescale=1./255)
test_generator = test_datagen.flow_from_directory(
    test_directory,
    target_size=(224, 224),
    batch_size=32,
    class_mode='categorical',
    shuffle=False
# Evaluate the saved model
scores = saved_model.evaluate(test_generator)
print("Test Loss:", scores[0])
print("Test Accuracy:", scores[1])
# code cell 6
# Make predictions using the saved model
predictions = saved_model.predict(test_generator)
predictions
```

```
# code cell 7
resnet model = Sequential()
resnet model.add(ResNet50(include top=False, pooling='avg',
weights='imagenet'))
resnet model.add(Dense(5, activation='softmax'))
resnet_model.layers[0].trainable = False
resnet_model.compile(optimizer='adam', loss='categorical_crossentropy',
metrics=['accuracy'])
resnet history = resnet model.fit(train generator,
validation_data=validation_generator, epochs=3)
# code cell 8
resnet model.save('resnet model.h5')
# code cell 9
# Load the saved model
saved model = tf.keras.models.load model('resnet model.h5')
# Test the model on new data
test_directory = r'archive\test'
test_datagen = ImageDataGenerator(rescale=1./255)
test_generator = test_datagen.flow_from_directory(
    test_directory,
   target_size=(224, 224),
    batch_size=32,
    class_mode='categorical',
    shuffle=False
# Evaluate the saved model
scores = saved_model.evaluate(test_generator)
print("Test Loss:", scores[0])
print("Test Accuracy:", scores[1])
# code cell 10
vgg_model = Sequential()
vgg_model.add(VGG16(include_top=False, pooling='avg', weights='imagenet'))
vgg_model.add(Dense(5, activation='softmax'))
vgg_model.layers[0].trainable = False
vgg_model.compile(optimizer='adam', loss='categorical_crossentropy',
metrics=['accuracy'])
```

```
vgg_history = vgg_model.fit(train_generator,
validation_data=validation_generator, epochs=3)
# code cell 11
vgg_model.save('vgg_model.h5')
# code cell 12
# Load the saved model
saved_model = tf.keras.models.load_model('vgg_model.h5')
# Test the model on new data
test_directory = r'archive\test'
test_datagen = ImageDataGenerator(rescale=1./255)
test_generator = test_datagen.flow_from_directory(
   test_directory,
    target_size=(224, 224),
   batch_size=32,
    class_mode='categorical',
    shuffle=False
# Evaluate the saved model
scores = saved_model.evaluate(test_generator)
print("Test Loss:", scores[0])
print("Test Accuracy:", scores[1])
```

Jupyter Notebook screenshots (code and outputs)

```
In [2]: import os
        import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import tensorflow as tf
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout
       from tensorflow.keras.applications import ResNet50, VGG16
In [10]: data_directory = r'archive\train'
          train_datagen = ImageDataGenerator(rescale=1./255, validation_split=0.3)
          train_generator = train_datagen.flow_from_directory(
              data_directory,
              target_size=(224, 224),
              batch_size=32,
              class_mode='categorical',
              subset='training'
          validation_generator = train_datagen.flow_from_directory(
              data_directory,
              target_size=(224, 224),
              batch_size=32,
              class_mode='categorical',
              subset='validation'
          )
```

Found 52493 images belonging to 5 classes. Found 22497 images belonging to 5 classes.

```
In [11]: alexnet_model = Sequential()
             alexnet_model.add(Conv2D(96, kernel_size=(11, 11), strides=(4, 4), activation='relu', input_shape=(224, 224, 3)))
            alexnet_model.add(MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))
alexnet_model.add(Conv2D(256, kernel_size=(5, 5), activation='relu'))
            alexnet_model.add(MaxPooling2D(pool_size=(3, 3), activation='relu'))
alexnet_model.add(Conv2D(384, kernel_size=(3, 3), activation='relu'))
alexnet_model.add(Conv2D(384, kernel_size=(3, 3), activation='relu'))
alexnet_model.add(Conv2D(256, kernel_size=(3, 3), activation='relu'))
alexnet_model.add(MaxPooling2D(pool_size=(3, 3), activation='relu'))
alexnet_model.add(MaxPooling2D(pool_size=(3, 3), strides=(2, 2)))
            alexnet_model.add(Flatten())
alexnet_model.add(Dense(4096,
                                                     activation='relu'))
            alexnet_model.add(Dropout(0.5))
alexnet_model.add(Dropout(0.5))
alexnet_model.add(Dropout(0.5))
alexnet_model.add(Dropout(0.5))
alexnet_model.add(Dropout(0.5))
alexnet_model.add(Dropout(0.5))
            alexnet model.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy'])
            a lexnet\_history = alexnet\_model.fit(train\_generator, validation\_data=validation\_generator, epochs=3)
             Epoch 1/3
             1641/1641 [=
                                   cy: 0.9279
Epoch 2/3
             1641/1641 |
                                            ==========] - 1279s 779ms/step - loss: 0.1185 - accuracy: 0.9628 - val_loss: 0.1359 - val_accura
             cv: 0.9586
                                 1641/1641 [
            cy: 0.9560
```

```
In [12]: # Save the trained model
        alexnet_model.save('alexnet_model.h5')
In [3]: # Load the saved model
        saved_model = tf.keras.models.load_model('alexnet_model.h5')
        # Test the model on new data
        test_directory = r'archive\test'
        test_datagen = ImageDataGenerator(rescale=1./255)
        test_generator = test_datagen.flow_from_directory(
            test directory,
            target_size=(224, 224),
            batch_size=32,
            class_mode='categorical',
            shuffle=False
        # Evaluate the saved model
        scores = saved_model.evaluate(test_generator)
        print("Test Loss:", scores[0])
        print("Test Accuracy:", scores[1])
        Found 50 images belonging to 5 classes.
        2/2 [================== ] - 0s 63ms/step - loss: 0.0904 - accuracy: 0.9600
        Test Loss: 0.09042396396398544
        Test Accuracy: 0.9599999785423279
In [16]: # Make predictions using the saved model
         predictions = saved_model.predict(test_generator)
         predictions
         2/2 [======] - 0s 74ms/step
Out[16]: array([[9.91539896e-01, 2.13779754e-06, 3.16425030e-05, 2.86038521e-05,
                  8.39765277e-03],
                 [9.68086481e-01, 4.26077080e-04, 1.10732147e-03, 3.04619316e-03,
                  2.73338556e-02],
                [1.84629500e-01, 2.91937922e-05, 1.07955157e-05, 1.06019324e-04,
                  8.15224528e-01],
                 [9.92770493e-01, 1.38189068e-06, 8.48506606e-05, 1.14904969e-05,
                  7.13177538e-03],
                 [9.71647382e-01, 6.52114477e-06, 1.68600556e-04, 3.72680952e-05,
                  2.81401202e-02],
                 [9.77940023e-01, 7.85245429e-05, 9.74769413e-04, 4.64776211e-04,
                  2.05419753e-02],
                 [5.95463812e-01, 1.98034148e-04, 7.52603577e-04, 6.65900472e-04,
                  4.02919620e-01],
                 [2.14261264e-01, 6.28716271e-06, 2.46522495e-06, 3.50250666e-05,
                  7.85695016e-01],
                 [8.83774757e-01, 4.22853074e-04, 3.00153741e-03, 1.62796467e-03,
                  1.11172900e-01],
                 [9.61094558e-01, 1.05437357e-04, 2.43932661e-03, 5.15449152e-04,
                  3.58452573e-02],
                 [6.51392135e-12, 9.99985337e-01, 1.66503859e-11, 1.46579350e-05,
                  4.59637381e-14],
                 [4.97236306e-06, 9.91396129e-01, 8.69758060e-06, 8.58986285e-03,
                  2.99155744e-07],
                 [3.94106365e-07, 9.96886313e-01, 8.35139929e-07, 3.11251241e-03,
                  1.30491982e-08],
```

```
[5.53318591e-09, 9.99710739e-01, 1.23903039e-08, 2.89284420e-04,
  1.51993293e-10],
  [1.00528279e-08, 9.99581516e-01, 2.04353778e-08, 4.18540876e-04,
   2.47467880e-10],
  [5.00998283e-07, 9.96948302e-01, 9.46903754e-07, 3.05032823e-03,
   2.01894803e-08],
  [1.26276814e-06, 9.95135248e-01, 2.37248219e-06, 4.86111687e-03,
   5.60989157e-08],
  [2.49305185e-06, 9.94105935e-01, 4.22009089e-06, 5.88715030e-03,
   1.35478899e-07],
  [6.53380861e-09, 9.99644756e-01, 1.38838523e-08, 3.55280266e-04,
  1.47064680e-10],
  [1.04061464e-05, 9.87655222e-01, 1.77371221e-05, 1.23159969e-02,
  6.62076218e-07],
  [1.90635352e-09, 1.36068336e-15, 9.99999881e-01, 9.75753238e-08,
   1.57420957e-17],
  [4.53319984e-24, 0.00000000e+00, 1.00000000e+00, 6.28217301e-21,
  0.00000000e+00],
  [2.36584955e-23, 0.00000000e+00, 1.00000000e+00, 1.37476640e-21,
  0.00000000e+00],
  [2.49291220e-16, 4.82958904e-28, 1.00000000e+00, 5.33366415e-14,
  1.32019394e-30],
  [1.54917579e-04, 7.38811623e-10, 9.99816835e-01, 2.82869423e-05,
  4.34786829e-09],
  [4.01806233e-09, 8.08653835e-18, 1.00000000e+00, 1.20017307e-09,
   5.63361349e-17],
  [9.82036067e-12, 8.50069948e-22, 1.00000000e+00, 4.79881516e-11,
  2.26117348e-22],
  [3.96369160e-08, 1.77635633e-14, 9.99999762e-01, 2.39802006e-07,
   2.26502243e-15],
  [2.28995667e-08, 9.12323412e-20, 1.00000000e+00, 4.81678863e-11,
   6.07954986e-17],
  [6.95464042e-19, 1.61641111e-32, 1.00000000e+00, 1.80785246e-16,
In [18]: resnet_model = Sequential()
     resnet_model.add(ResNet50(include_top=False, pooling='avg', weights='imagenet'))
     resnet_model.add(Dense(5, activation='softmax'))
     resnet model.layers[0].trainable = False
     resnet model.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy'])
     resnet_history = resnet_model.fit(train_generator, validation_data=validation_generator, epochs=3)
             0.8757
     Epoch 2/3
     1641/1641 [
                 0.8892
     1641/1641 [==========] - 3004s 2s/step - loss: 0.2066 - accuracy: 0.9475 - val loss: 0.3254 - val accuracy:
```

```
In [19]: resnet_model.save('resnet_model.h5')
 In [4]: # Load the saved model
          saved_model = tf.keras.models.load_model('resnet_model.h5')
          # Test the model on new data
          test_directory = r'archive\test'
          test_datagen = ImageDataGenerator(rescale=1./255)
          test_generator = test_datagen.flow_from_directory(
             test_directory,
             target_size=(224, 224),
             batch_size=32,
             class_mode='categorical',
             shuffle \hbox{\tt =} \textbf{False}
          # Evaluate the saved model
          scores = saved_model.evaluate(test_generator)
          print("Test Loss:", scores[0])
         print("Test Accuracy:", scores[1])
          Found 50 images belonging to 5 classes.
          2/2 [========== ] - 3s 717ms/step - loss: 0.1599 - accuracy: 0.9200
          Test Loss: 0.15987545251846313
          Test Accuracy: 0.9200000166893005
In [22]: vgg model = Sequential()
       vgg_model.add(VGG16(include_top=False, pooling='avg', weights='imagenet'))
      vgg_model.add(Dense(5, activation='softmax'))
      vgg model.layers[0].trainable = False
      vgg_model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
      vgg_history = vgg_model.fit(train_generator, validation_data=validation_generator, epochs=3)
      Epoch 1/3
       1641/1641 [
                   0.8873
      Epoch 2/3
      1641/1641 [=
                   ====================== - 5091s 3s/step - loss: 0.2076 - accuracy: 0.9661 - val loss: 0.2755 - val accuracy:
      Epoch 3/3
      0.9085
```

```
In [23]: vgg_model.save('vgg_model.h5')
In [5]: # Load the saved model
        saved_model = tf.keras.models.load_model('vgg_model.h5')
        # Test the model on new data
        test_directory = r'archive\test'
        test_datagen = ImageDataGenerator(rescale=1./255)
        test_generator = test_datagen.flow_from_directory(
           test directory,
           target_size=(224, 224),
           batch_size=32,
           class_mode='categorical',
           shuffle=False
        # Evaluate the saved model
        scores = saved model.evaluate(test generator)
        print("Test Loss:", scores[0])
        print("Test Accuracy:", scores[1])
        Found 50 images belonging to 5 classes.
        Test Loss: 0.08350679278373718
        Test Accuracy: 0.9800000190734863
```

Conclusion:

This assignment involves the comparative study of AlexNet, ResNet and VGGNet models on Rice Image Dataset. From the above analysis (code) it was observed that:

- AlexNet model: gives a accuracy of 95.99% with a loss of 9%
- ResNet model: gives a accuracy of 92% with a loss of 15%
- VGGNet model: gives a accuracy of 98% with a loss of 8.3%

The number of epochs used is 3. This was done to quickly train the model as the dataset was big. For the first model, each epoch took 20 minutes, for the second it took 40 minutes and for the third it took 60 minutes.

If the number of epochs are increased than the loss can be decreased significantly

At the end, it is observed that the VGGNet model gives comparatively higher accuracy and has less test loss than the other 2 models. So VGGNet deep learning model best fits for this Rice Image Dataset.