HO CHI MINH CITY UNIVERSITY OF TECHNOLOGY AND EDUCATION FACULTY FOR HIGH QUALITY TRAINING

MAJOR: ARTIFICIAL INTELLIGENCE ANN TRAINING HOMEWORK



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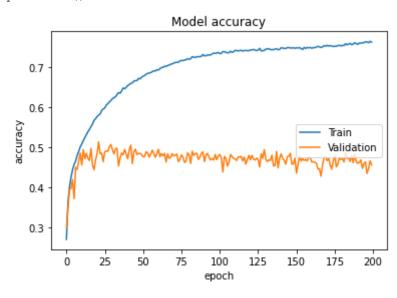
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EXERCISE 1: CIFAR10

```
from keras.datasets import cifar10
                                                                   In [ ]:
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
                                                                   In []:
from tensorflow.keras.utils import to_categorical
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout
from tensorflow.keras.optimizers import RMSprop
from keras.callbacks import EarlyStopping
                                                                   In []:
(x train, y train), (x test, y test) = cifar10.load data()
Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
In []:
for i in range(9):
 plt.subplot(330+i+1)
 plt.imshow(x train[i], cmap = plt.get cmap('gray'))
plt.show()
20
             20
0
              0
20
             20
0
              0
20
             20
                                                                   In []:
x train = x train.reshape(x train.shape[0],-1)
x test = x test.reshape(x test.shape[0],-1)
                                                                   In []:
x train = x train.astype('float32')
x_test = x_test.astype('float32')
                                                                   In []:
x train /= 255
```

```
x test /= 255
                                                      In []:
x train.shape
                                                     Out[]:
(50000, 3072)
                                                      In []:
y train = to categorical(y train, 10)
y test = to categorical(y test,10)
                                                      In []:
model = Sequential()
model.add(Dense(512,activation= 'relu',input shape=(3072,)))
model.add(Dense(512,activation='relu'))
model.add(Dense(10,activation='softmax'))
model.summary()
Model: "sequential"
Layer (type)
                    Output Shape
                                       Param #
______
dense (Dense)
                     (None, 512)
                                        1573376
dense 1 (Dense)
                     (None, 512)
                                        262656
dense_2 (Dense)
                     (None, 10)
                                        5130
______
Total params: 1,841,162
Trainable params: 1,841,162
Non-trainable params: 0
                                                      In []:
model.compile(loss = 'categorical crossentropy',
          optimizer = 'RMSprop',
          metrics = ['accuracy'])
                                                      In []:
history = model.fit(x train, y train, batch size = 128, epochs = 200, verbose
=1, validation_data = (x_test, y_test))
Epoch 1/200
uracy: 0.2567 - val loss: 1.8665 - val accuracy: 0.3352
Epoch 2/200
uracy: 0.3607 - val loss: 1.7801 - val accuracy: 0.3515
Epoch 197/200
uracy: 0.7553 - val_loss: 4.4574 - val_accuracy: 0.4503
Epoch 198/200
uracy: 0.7547 - val loss: 4.6529 - val accuracy: 0.4403
Epoch 199/200
```

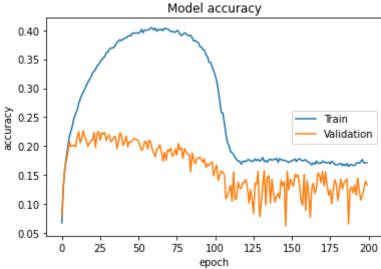
```
391/391 [============== ] - 16s 41ms/step - loss: 0.7736 - acc
uracy: 0.7513 - val loss: 4.3999 - val accuracy: 0.4443
Epoch 200/200
uracy: 0.7518 - val loss: 4.2500 - val accuracy: 0.4677
                                                        In []:
Score = model.evaluate(x test, y test, verbose = 1)
acy: 0.4563
                                                        In []:
plt.plot(history.history['accuracy'])
plt.plot(history.history['val accuracy'])
plt.title('Model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['Train','Validation'], loc ='center right')
plt.show()
```



EXERCISE 2: CIFAR100

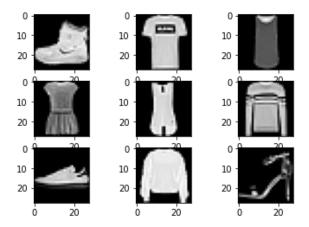
```
from keras.datasets import cifar100
                                                                    In []:
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
                                                                    In []:
from tensorflow.keras.utils import to_categorical
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout
from tensorflow.keras.optimizers import RMSprop
from keras.callbacks import EarlyStopping
                                                                    In []:
(x_train, y_train), (x_test, y_test) = cifar100.load_data()
Downloading data from https://www.cs.toronto.edu/~kriz/cifar-100-python.tar.g
In []:
for i in range(9):
 plt.subplot(330+i+1)
 plt.imshow(x_train[i], cmap = plt.get_cmap('gray'))
plt.show()
20
             20
                           20
0
              0
                            0
             20
                           20
0
              0
                            0
20
             20
                           20
               0
                             0
                                 20
      20
                    20
x train = x train.reshape(x train.shape[0],-1)
x_{test} = x_{test.reshape}(x_{test.shape}[0],-1)
                                                                    In []:
x train = x train.astype('float32')
x test = x test.astype('float32')
                                                                    In []:
```

```
x train /= 255
x test /= 255
                                                      In []:
x train.shape
                                                     Out[]:
(50000, 3072)
                                                      In []:
y train = to categorical(y train, 100)
y_test = to_categorical(y_test,100)
                                                      In []:
model = Sequential()
model.add(Dense(512,activation= 'relu',input_shape=(3072,)))
model.add(Dense(512,activation='relu'))
model.add(Dense(100, activation='softmax'))
model.summary()
Model: "sequential"
Layer (type)
                    Output Shape
______
                     (None, 512)
                                        1573376
dense (Dense)
dense 1 (Dense)
                     (None, 512)
                                        262656
dense 2 (Dense)
                     (None, 100)
                                        51300
______
Total params: 1,887,332
Trainable params: 1,887,332
Non-trainable params: 0
                                                      In [ ]:
model.compile(loss = 'categorical crossentropy',
          optimizer = 'RMSprop',
          metrics = ['accuracy'])
                                                      In []:
history = model.fit(x_train, y_train, batch_size = 128, epochs = 200, verbose
=1, validation data = (x test, y test))
Epoch 1/200
acy: 0.0676 - val loss: 4.0825 - val accuracy: 0.0799
Epoch 2/200
acy: 0.1300 - val loss: 3.7979 - val accuracy: 0.1307
Epoch 198/200
acy: 0.1711 - val_loss: 4.2361 - val_accuracy: 0.1252
Epoch 199/200
acy: 0.1708 - val loss: 4.1416 - val accuracy: 0.1402
```



EXERCISE 3: FASHION_MNIST

```
from keras.datasets import fashion mnist
                                                      In [ ]:
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LinearRegression
                                                      In []:
from tensorflow.keras.utils import to categorical
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout
from tensorflow.keras.optimizers import RMSprop
from keras.callbacks import EarlyStopping
                                                      In []:
(x train, y train), (x test, y test) = fashion mnist.load data()
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-data
sets/train-labels-idx1-ubyte.gz
40960/29515 [====================] - Os Ous/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-data
sets/train-images-idx3-ubyte.gz
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-data
sets/t10k-labels-idx1-ubyte.gz
======] - Os Ous/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-data
sets/t10k-images-idx3-ubyte.gz
In []:
for i in range(9):
 plt.subplot(330+i+1)
 plt.imshow(x train[i], cmap = plt.get cmap('gray'))
plt.show()
```



x train.shape

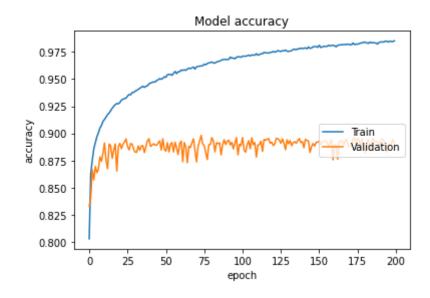
```
Out[]:
(60000, 28, 28)
                                                                             In []:
x_train = x_train.reshape(x_train.shape[0],-1)
x_test = x_test.reshape(x_test.shape[0],-1)
                                                                             In []:
x_train = x_train.astype('float32')
x_test = x_test.astype('float32')
                                                                             In []:
x train /= 255
x test /= 255
                                                                             In []:
y_train = to_categorical(y_train,10)
y_test = to_categorical(y_test,10)
                                                                             In []:
model = Sequential()
model.add(Dense(512,activation= 'relu',input_shape=(784,)))
model.add(Dense(512,activation='relu'))
model.add(Dense(10,activation='softmax'))
model.summary()
Model: "sequential"
```

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 512)	401920
dense_1 (Dense)	(None, 512)	262656
dense_2 (Dense)	(None, 10)	5130

Total params: 669,706 Trainable params: 669,706 Non-trainable params: 0

In []:

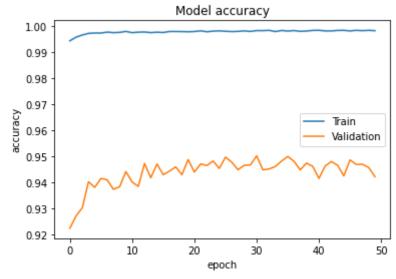
```
model.compile(loss = 'categorical crossentropy',
         optimizer = 'RMSprop',
         metrics = ['accuracy'])
                                               In []:
history = model.fit(x train, y train, batch size = 128, epochs = 200, verbose
=1, validation_data = (x_test, y_test))
Epoch 1/200
acy: 0.8027 - val loss: 0.4626 - val accuracy: 0.8330
Epoch 2/200
acy: 0.8627 - val loss: 0.4402 - val accuracy: 0.8399
Epoch 197/200
acy: 0.9849 - val loss: 2.9288 - val accuracy: 0.8894
Epoch 198/200
acy: 0.9847 - val loss: 2.8384 - val accuracy: 0.8910
Epoch 199/200
acy: 0.9845 - val loss: 3.1128 - val accuracy: 0.8943
Epoch 200/200
acy: 0.9853 - val loss: 3.0674 - val accuracy: 0.8903
                                               In []:
Score = model.evaluate(x test, y test, verbose = 1)
acy: 0.8903
                                               In []:
plt.plot(history.history['accuracy'])
plt.plot(history.history['val accuracy'])
plt.title('Model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['Train','Validation'], loc ='center right')
plt.show()
```



EXERCISE 4: 2ND DEGREES ROBOTIC ARM

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear model import LinearRegression
                                                                            In [31]:
from tensorflow.keras.utils import to categorical
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout
from tensorflow.keras.optimizers import RMSprop
from keras.callbacks import EarlyStopping
from sklearn.model selection import train test split
from tensorflow import keras
                                                                            In [32]:
11 = 10
12 = 40
px = []
py= []
tt1 =[]
tt2 =[]
                                                                            In [33]:
for i in np.arange(0, 360, 1):
    for j in np.arange(0, 360, 1):
      x = 11*np.cos(np.radians(i)) + 12*np.cos(np.radians(i+j))
      y = 11*np.sin(np.radians(i)) + 12*np.sin(np.radians(i+j))
      px.append(x)
      py.append(y)
      tt1.append(i)
      tt2.append(j)
                                                                            In [34]:
Px = np.array([px])
Py = np.array([py])
Tt1= np.array([tt1])
Tt2 = np.array([tt2])
                                                                            In [35]:
Tt1 = Tt1.astype('float32')
Tt2 = Tt2.astype('float32')
                                                                            In [36]:
Tt1 /= 360
Tt2 /= 360
                                                                            In [37]:
result = np.concatenate((Tt1.T, Tt2.T),axis =1)
train = np.concatenate((Px.T, Py.T),axis =1)
result.shape
                                                                           Out[37]:
(129600, 2)
                                                                                12
```

```
In [38]:
x train = result[0:90720,:]
y train = train[0:90720,:]
x \text{ test} = \text{result}[90720:129600,:]
y test = train[90720:129600,:]
                                                               In [39]:
model=Sequential()
model.add(Dense(512, activation='relu', input_shape=(2,)))
model.add(Dense(512, activation='relu'))
model.add(Dense(2, activation='linear'))
model.summary()
Model: "sequential 1"
Layer (type)
                       Output Shape
                                              Param #
_____
dense 3 (Dense)
                         (None, 512)
                                               1536
dense 4 (Dense)
                        (None, 512)
                                               262656
dense 5 (Dense)
                         (None, 2)
                                               1026
______
Total params: 265,218
Trainable params: 265,218
Non-trainable params: 0
                                                               In [40]:
model.compile(loss = ['mae'],
            optimizer = 'Adam',
            metrics = ['accuracy'])
                                                               In [42]:
history = model.fit(x train, y train, batch size=128, epochs=50, verbose=1,
validation_data=(x_test,y_test))
Epoch 1/50
racy: 0.9942 - val loss: 11.2612 - val accuracy: 0.9222
Epoch 2/50
709/709 [============= ] - 7s 10ms/step - loss: 0.3756 - accu
racy: 0.9956 - val loss: 10.6547 - val_accuracy: 0.9270
Epoch 48/50
709/709 [=============== ] - 7s 10ms/step - loss: 0.2094 - accu
racy: 0.9981 - val loss: 6.1848 - val accuracy: 0.9468
Epoch 49/50
709/709 [=========== ] - 7s 10ms/step - loss: 0.1991 - accu
racy: 0.9982 - val loss: 6.1670 - val accuracy: 0.9456
Epoch 50/50
709/709 [============ ] - 8s 11ms/step - loss: 0.2009 - accu
racy: 0.9981 - val_loss: 6.4274 - val_accuracy: 0.9421
                                                               In [43]:
Score = model.evaluate(x test, y test, verbose = 1)
```



EXERCISE 5: 3RD DEGREES ROBOTIC ARM

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear model import LinearRegression
                                                                             In [7]:
from tensorflow.keras.utils import to categorical
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout
from tensorflow.keras.optimizers import RMSprop
from keras.callbacks import EarlyStopping
from sklearn.model selection import train test split
from tensorflow import keras
                                                                             In [8]:
11 = 10
12 = 40
L3 = 20
px = []
py= []
theta= []
tt1 =[]
tt2 =[]
tt3 =[]
                                                                             In [9]:
for i in np.arange(0, 360, 5):
    for j in np.arange(0, 360, 5):
      for k in np.arange(0, 360, 5):
              x = 11*np.cos(np.radians(i)) + 12*np.cos(np.radians(i+j)) +
L3*np.cos(np.radians(i+j+k))
              y = 11*np.sin(np.radians(i)) + 12*np.sin(np.radians(i+j)) +
L3*np.sin(np.radians(i+j+k))
              tt_sum = (i+j+k) %360
              px.append(x)
              py.append(y)
              theta.append(tt sum)
              tt1.append(i)
              tt2.append(j)
              tt3.append(k)
                                                                            In [10]:
Px = np.array([px])
Py = np.array([py])
Tt sum = np.array([theta])
Tt1= np.array([tt1])
Tt2 = np.array([tt2])
Tt3 = np.array([tt3])
                                                                            In [11]:
Px.shape
                                                                           Out[11]:
                                                                                15
```

```
(1, 373248)
                                                                    In [12]:
Tt1 = Tt1.astype('float32')
Tt2 = Tt2.astype('float32')
Tt3 = Tt3.astype('float32')
                                                                    In [13]:
Tt1 /= 360
Tt2 /= 360
Tt3 /= 360
                                                                    In [14]:
result = np.concatenate((Tt1.T, Tt2.T, Tt3.T),axis =1)
train = np.concatenate((Px.T, Py.T, Tt sum.T), axis =1)
result.shape
                                                                   Out[14]:
(373248, 3)
                                                                    In [15]:
x train = result[0:261273,:]
y train = train[0:261273,:]
x \text{ test} = \text{result}[261273:373248,:]
y test = train[261273:373248,:]
                                                                    In [16]:
model=Sequential()
model.add(Dense(512, activation='relu', input_shape=(3,)))
model.add(Dense(512, activation='relu'))
model.add(Dense(3, activation='linear'))
model.summary()
Model: "sequential"
Layer (type)
                        Output Shape
                                                  Param #
_____
                           (None, 512)
 dense (Dense)
                                                    2048
                          (None, 512)
dense 1 (Dense)
                                                    262656
dense 2 (Dense)
                           (None, 3)
                                                    1539
______
Total params: 266,243
Trainable params: 266,243
Non-trainable params: 0
                                                                    In [17]:
model.compile(loss = ['mae'],
             optimizer = 'Adam',
             metrics = ['accuracy'])
                                                                    In [19]:
history = model.fit(x train, y train, batch size=128, epochs=30, verbose=1,
validation_data=(x_test,y_test))
Epoch 1/30
```

```
curacy: 0.9369 - val loss: 24.1546 - val accuracy: 0.9246
Epoch 2/30
curacy: 0.9372 - val loss: 22.8380 - val_accuracy: 0.9253
Epoch 29/30
curacy: 0.9791 - val loss: 7.8676 - val accuracy: 0.9638
Epoch 30/30
curacy: 0.9796 - val loss: 7.6863 - val accuracy: 0.9698
                                                 In [20]:
Score = model.evaluate(x test, y test, verbose = 1)
3500/3500 [============== ] - 11s 3ms/step - loss: 7.6863 - ac
curacy: 0.9698
                                                 In [21]:
plt.plot(history.history['accuracy'])
plt.plot(history.history['val accuracy'])
plt.title('Model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['Train','Validation'], loc ='center right')
plt.show()
```

