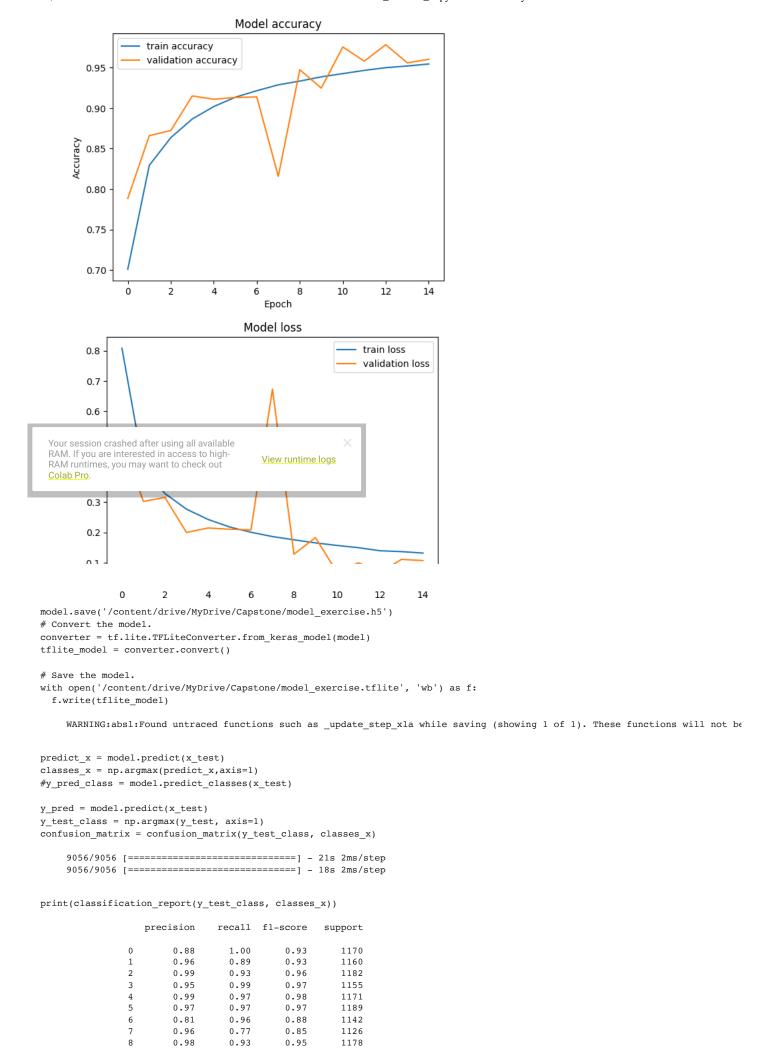
```
from google.colab import drive
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from keras.utils import np_utils
from keras.wrappers.scikit_learn import KerasClassifier
from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
from sklearn.preprocessing import MinMaxScaler
import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.layers import Flatten
from tensorflow.keras.layers import Dense, Dropout
from tensorflow.keras import regularizers
from sklearn.metrics import classification_report, confusion_matrix
from sklearn import metrics
from sklearn.preprocessing import StandardScaler
drive.mount("/content/drive")
path = "/content/drive/MyDrive/Capstone/exercise_datasetV2.csv"
df = pd.read_csv(path)
print(df.head())
banyak_kategori = len(df.index)
    Mounted at /content/drive
      Activity, Exercise or Sport (1 hour) Intensity Description
               Cycling, mountain bike, bmx
                                                               NaN
       Cycling, <10 mph, leisure bicycling
                                                               NaN
                                                               NaN
 Your session crashed after using all available
                                                               NaN
 RAM. If you are interested in access to high-
                                                               NaN
                                        View runtime logs
 RAM runtimes, you may want to check out
 Colab Pro.
                                   0.823236
                        60
    2
                        60
                                   3.294974
                                   1.234853
    3
                        60
                                   1.647825
    4
                        60
list berat = []
for i in range(len(df.index)):
 list_berat.append(1)
df['berat'] = list_berat
dict_df = {'Activity, Exercise or Sport (1 hour)' : [], 'Duration (minutes)': [], 'Calories per kg': [], 'berat' : []}
df_new = df
for index, row in df.iterrows():
 print(index)
 menit = row['Duration (minutes)']
 activity = row['Activity, Exercise or Sport (1 hour)']
 calories = row['Calories per kg']
 for i in range(1,menit):
   for j in range(2,101):
     new_calories = calories*1.0/60*i*j
     list activity = dict df.get('Activity, Exercise or Sport (1 hour)')
     list_duration = dict_df.get('Duration (minutes)')
      list_calories = dict_df.get('Calories per kg')
     list berat = dict_df.get('berat')
     list_activity.append(activity)
      list_duration.append(i)
     list_calories.append(new_calories)
      list_berat.append(j)
      #new_row = pd.DataFrame({'Activity, Exercise or Sport (1 hour)' : [activity], 'Duration (minutes)': [i], 'Calories per k
df_curr = pd.DataFrame(dict_df)
df_new = pd.concat([df_curr, df_new.loc[:]]).reset_index(drop=True)
#df2 = pd.concat([new_row,df.loc[:]]).reset_index(drop=True)
print(df new.head())
print(df_new.tail())
```

```
activity durasi calories berat Intensity Description
        0 Cycling, mountain bike, bmx
                                                                                0.058358
                                                                          1
        1 Cycling, mountain bike, bmx
                                                                           1 0.087536
                                                                                                          3
                                                                                                                                              NaN
        2 Cycling, mountain bike, bmx
                                                                          1 0.116715
                                                                                                          4
                                                                                                                                              NaN
                                                                          1 0.145894
        3 Cycling, mountain bike, bmx
                                                                                                          5
                                                                                                                                              NaN
        4 Cycling, mountain bike, bmx
                                                                         1 0.175073
                                                                                                          6
                                                                                                                                              NaN
              durasi calories berat
                                                            1
         Λ
                       1 0.058358
                                                     2
          1
                             0.087536
                                                     3
                       1
          2
                       1
                             0.116715
                                                     4
          3
                       1 0.145894
                                                     5
def get base model():
   model = tf.keras.Sequential([
       normalizer,
       tf.keras.layers.Dense(10, activation='relu'),
       tf.keras.layers.Dense(10, activation='relu'),
       tf.keras.layers.Dense(banyak_kategori, activation = 'softmax')
   model.compile(optimizer=tf.keras.optimizers.Adam(learning rate=2e-3),
                            loss='categorical_crossentropy',
                            metrics=['accuracy'])
   return model
        '\ndef get base model():\n model = tf.keras.Sequential([\n
                                                                                                                         normalizer,\n
                                                                                                                                                        tf.keras.layers.Dense(10, activation='rel
                                                                                                        (banyak_kategori, activation = 'softmax')\n ])\n\n model.compile(c
  Your session crashed after using all available
                                                                                                        gorical_crossentropy',\n
                                                                                                                                                                                metrics=['accuracy'])\n ret
  RAM. If you are interested in access to high-
                                                                       View runtime logs
  RAM runtimes, you may want to check out
  Colab Pro.
encoder = LabelEncoder()
encoder.fit(y)
encoded Y = encoder.transform(y)
# convert integers to dummy variables (i.e. one hot encoded)
dummy_y = np_utils.to_categorical(encoded_Y)
        '\ny = df_new['activity']\nencoder = LabelEncoder()\nencoder.fit(y)\nencoded_Y = encoder.transform(y)\n# convert integers
        hot encoded)\ndummy_y = np_utils.to_categorical(encoded_Y)\n'
#est = KerasClassifier(build fn= get base model, epochs=200, batch size=5, verbose=0)
#kfold = KFold(n_splits=5, shuffle=True)
x = df_new[numeric_feature_names]
results = cross_val_score(est, x, dummy_y, cv=kfold)
print("Baseline: %.2f%% (%.2f%%)" % (results.mean()*100, results.std()*100))
         '\nx = df_new[numeric_feature_names]\n\nresults = cross_val_score(est, x, dummy_y, cv=kfold)\nprint("Baseline: %.2f%% (%.
        results.std()*100))\n
https://machinelearningmastery.com/multi-class-classification-tutorial-keras-deep-learning-library/
\verb|https://www.tensorflow.org/tutorials/load_data/pandas_dataframe| \\
https://regenerativetoday.com/a-step-by-step-tutorial-to-develop-a-multi-output-model-in-tensorflow/
        \nhttps://machinelearningmastery.com/multi-class-classification-tutorial-keras-deep-learning-library/\nhttps://www.tensc
        a/pandas\_dataframe \verb|\nhttps://regenerativetoday.com/a-step-by-step-tutorial-to-develop-a-multi-output-model-in-tensorflow/\end{a} is a simple of the context of the cont
jumlah_class = len(df_new['activity'].value_counts())
print(jumlah class)
```

248

```
df_new['activity'] = df_new['activity'].astype('category')
df_new['activity_category'] = df_new['activity'].cat.codes.astype('category')
print(df_new.head())
                          activity durasi calories berat Intensity Description
    0 Cycling, mountain bike, bmx
                                         1 0.058358
                                                           2
                                                                               NaN
                                         1 0.087536
    1
      Cycling, mountain bike, bmx
                                                           3
                                                                               NaN
    2 Cycling, mountain bike, bmx
                                         1 0.116715
                                                           4
                                                                               NaN
       Cycling, mountain bike, bmx
                                         1 0.145894
                                                           5
                                                                               NaN
    4 Cycling, mountain bike, bmx
                                         1 0.175073
                                                           6
                                                                               NaN
      activity_category
    0
                     61
    1
    2
                     61
    3
                     61
    4
                     61
df new 2 = df new.drop(columns = ['activity', 'Intensity Description'])
sc = StandardScaler()
x = pd.DataFrame(sc.fit_transform(df_new_2))
df new 2['durasi'] = MinMaxScaler().fit transform(np.array(df new 2['durasi']).reshape(-1,1))
df_new_2['calories'] = MinMaxScaler().fit_transform(np.array(df_new_2['calories']).reshape(-1,1))
df_new_2['berat'] = MinMaxScaler().fit_transform(np.array(df_new_2['berat']).reshape(-1,1))
y = tf.keras.utils.to_categorical(df_new["activity_category"].values, num_classes=jumlah_class)
x_train, x_test, y_train, y_test = train_test_split(x.values, y, test_size=0.2)
 Your session crashed after using all available
 RAM. If you are interested in access to high-
                                       View runtime logs
 RAM runtimes, you may want to check out
 Colab Pro.
    [[-1.35066498e+00 -4.83891605e-01 1.22482836e+00 -7.61270455e-01]
     [ 6.45524518e-01 -3.50932742e-01 2.99439752e-04 -8.45080046e-01]
     [-5.90129526e-02 -4.60763979e-01 -1.74633263e-01 1.40381066e+00]
     [-4.69993144e-01 2.35554652e-01 7.35016792e-01 -1.38984239e+00]
     [-1.64422226e+00 -9.10832216e-01 -1.01431024e+00 -4.81905150e-01]
     [-1.76435864e-01 4.43372659e-02 -6.96736414e-02 -1.13841362e+00]]
    [[0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
[0. 0. 0. ... 0. 0. 0.]]
    [[-1.70293372 -0.89776039 1.67965339 0.16063505]
     [-0.46999314 -0.44537205 0.59507063 -1.264128
     [-0.76355042 -0.46448915 0.80498987 -0.71936566]
     [-0.58741606 0.61334078 1.15485528 1.11047709]
     [ 1.17392762 -0.02819828 -0.06967364 1.32000106]]
    [[0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]
     [0. 0. 0. ... 0. 0. 0.]]
from keras.engine import sequential
def get model():
   model = tf.keras.Sequential([
       Dense(50, activation='relu'),
       Dense(50, activation='relu'),
       Dense(60, activation='relu'),
       Dense(70, activation='relu'),
       Dense(80, activation='relu'),
       Dense(90, activation='relu'),
       Dense(100, activation='relu'),
        Dense(banyak_kategori, activation='softmax')
    1)
   model.compile(optimizer='adam',
                  loss='categorical crossentropy',
```

```
metrics=['accuracy'])
  return model
#x train=np.asarray(x train).astype(np.int)
#y_train=np.asarray(y_train).astype(np.int)
my callbacks = [
  tf.keras.callbacks.EarlyStopping(patience=2),
  tf.keras.callbacks.ModelCheckpoint(filepath='model.{epoch:02d}-{val loss:.2f}.h5'),
  tf.keras.callbacks.TensorBoard(log_dir='./logs'),
model = get_model()
model_fit = model.fit(x_train,
            y train,
            epochs = 15,
            validation_data = (x_test, y_test))
  Epoch 1/15
  Epoch 2/15
  Epoch 3/15
  Epoch 4/15
  Epoch 5/15
  ims/step - loss: 0.2181 - accuracy: 0.9134 - val_loss: 0.2106 - val_&
 Your session crashed after using all available
RAM. If you are interested in access to high-
                      View runtime logs
RAM runtimes, you may want to check out
                                 ms/step - loss: 0.2005 - accuracy: 0.9215 - val loss: 0.2084 - val &
Colab Pro.
                                 ms/step - loss: 0.1864 - accuracy: 0.9287 - val loss: 0.6731 - val &
  Epoch 9/15
  Epoch 10/15
  Epoch 11/15
  Epoch 12/15
  36221/36221 [=
          Epoch 13/15
  Epoch 14/15
  36221/36221 [=
           ================================ ] - 182s 5ms/step - loss: 0.1367 - accuracy: 0.9520 - val_loss: 0.1110 - val_&
  Epoch 15/15
  def plot_accuracy(history):
  plt.plot(history.history['accuracy'],label='train accuracy')
  plt.plot(history.history['val_accuracy'],label='validation accuracy')
  plt.title('Model accuracy')
  plt.ylabel('Accuracy')
  plt.xlabel('Epoch')
  plt.legend(loc='best')
  plt.savefig('Accuracy_v1_model_inceptionv3')
  plt.show()
def plot_loss(history):
  plt.plot(history.history['loss'],label="train loss")
  plt.plot(history.history['val_loss'],label="validation loss")
  plt.title('Model loss')
  plt.ylabel('Loss')
  plt.xlabel('Epoch')
  plt.legend(loc='best')
  plt.savefig('Loss_v1_model_inceptionv3')
  plt.show()
plot accuracy(model fit)
plot_loss(model_fit)
Гэ
```



```
q
          0.94
                     0.99
                                 0.97
                                            1204
10
          1.00
                     0.98
                                 0.99
                                            1122
11
          0.89
                     1.00
                                 0.94
                                            1167
12
          0.98
                     0.89
                                 0.93
                                            1203
13
          0.97
                     0.96
                                 0.96
                                            1181
          0.92
                     1.00
                                 0.96
                                            1182
14
15
          0.90
                     0.92
                                 0.91
                                            1143
16
          0.96
                     0.84
                                 0.90
                                            1177
17
          0.87
                     0.94
                                 0.90
                                            1169
18
          1.00
                     0.73
                                 0.84
                                            1192
19
          0.84
                     0.97
                                 0.90
                                            1155
20
          0.99
                     1.00
                                 0.99
                                            1147
21
          0.96
                     0.99
                                 0.98
                                            1200
22
          0.99
                     1.00
                                 1.00
                                            1129
          1.00
                     0.96
                                 0.98
                                            1177
23
24
          0.98
                     1.00
                                 0.99
                                            1183
25
          0.97
                     1.00
                                 0.99
                                            1166
26
          1.00
                     0.97
                                 0.99
                                            1091
                                 0.99
27
          0.98
                     0.99
                                            1207
28
          0.99
                     1.00
                                 0.99
                                            1156
29
          1.00
                     0.94
                                 0.97
                                            1187
30
          0.95
                     0.96
                                 0.96
                                            1152
31
          0.98
                     0.84
                                 0.90
                                            1109
          0.85
                     1.00
                                 0.92
32
                                            1180
33
          1.00
                     0.93
                                 0.97
                                            1161
34
          0.94
                     0.98
                                 0.96
                                            1120
35
          0.98
                     1.00
                                 0.99
                                            1159
                     1.00
                                 1.00
36
          1.00
                                            1151
37
          1.00
                     0.97
                                 0.98
                                            1215
38
          0.96
                                 0.98
                     1.00
                                            1144
39
          0.97
                     0.99
                                 0.98
                                            1119
40
          1.00
                     0.96
                                 0.98
                                            1135
41
          0.83
                     1.00
                                 0.90
                                            1165
                                             1138
42
          0.89
                     0.93
                                 0.91
          0.95
                     0.99
                                 0.97
                                             1150
43
          0.87
                                             1198
```

Your session crashed after using all available RAM. If you are interested in access to high-RAM runtimes, you may want to check out Colab Pro.

# Extract the metrics

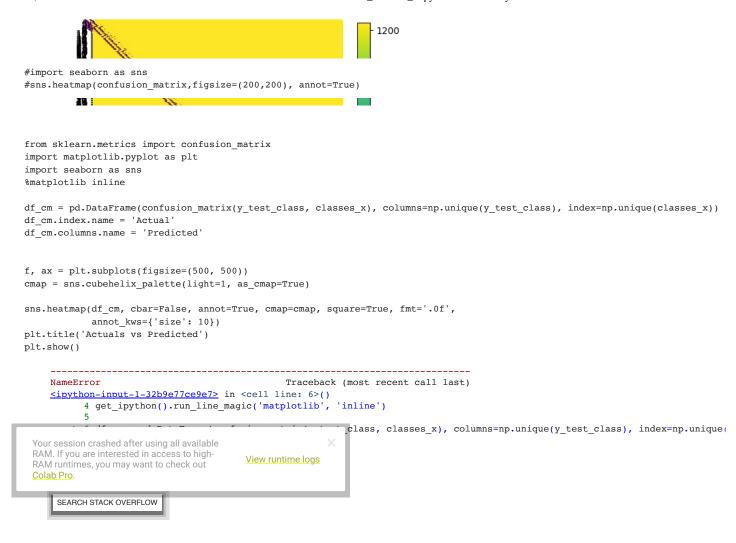
View runtime logs

51	0.99	0.98	0.98	1171
52	0.96	0.98	0.97	1205
53	0.67	0.91	0.77	1167
54	1.00	0.94	0.97	1166
EE	1 00	0 60	0 76	1011

report = classification\_report(y\_test\_class, classes\_x, output\_dict=True, zero\_division=0)

```
precision = report['macro avg']['precision']
recall = report['macro avg']['recall']
f1_score = report['macro avg']['f1-score']
support = report['macro avg']['support']
accuracy = report['accuracy']
print("accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("F1-score:", f1_score)
print("support" , support)
    accuracy: 0.9603090791126572
    Precision: 0.9660990410937471
    Recall: 0.9608365246358126
    F1-score: 0.9592716520206644
    support 289764
def plot_confusion_matrix(matrix, labels, title='Confusion matrix'):
   fig, ax = plt.subplots()
   ax.set_xticks([x for x in range(len(labels))])
   ax.set_yticks([y for y in range(len(labels))])
   # Place labels on minor ticks
   ax.set_xticks([x + 0.5 for x in range(len(labels))], minor=True)
   ax.set_xticklabels(labels, rotation='90', fontsize=10, minor=True)
   ax.set_yticks([y + 0.5 for y in range(len(labels))], minor=True)
   ax.set_yticklabels(labels[::-1], fontsize=10, minor=True)
   # Hide major tick labels
   ax.tick_params(which='major', labelbottom='off', labelleft='off')
   # Finally, hide minor tick marks
   ax.tick_params(which='minor', width=0)
   # Plot heat map
   proportions = [1. * row / sum(row) for row in matrix]
```

```
ax.pcolor(np.array(proportions[::-1]), cmap=plt.cm.Blues)
    # Plot counts as text
    for row in range(len(matrix)):
        for col in range(len(matrix[row])):
            confusion = matrix[::-1][row][col]
            if confusion != 0:
                ax.text(col + 0.5, row + 0.5, confusion, fontsize=9,
                    horizontalalignment='center',
                    verticalalignment='center')
    # Add finishing touches
    ax.grid(True, linestyle=':')
    ax.set_title(title)
    fig.tight_layout()
    plt.show()
#print(type(confusion_matrix(y_test_class, classes_x)))
#print(y_test_class)
#print(y_test)
#print(len(y_test_class))
dict_activity = dict(enumerate(df_new['activity'].cat.categories))
df_new['activity_code'] = df_new['activity'].cat.codes
print(df_new['activity_code'])
print(dict_activity)
df_new['activity_reversed'] = df_new['activity_code'].map(dict_activity)
df y test class = pd.DataFrame(y test class, columns = ['activity class'])
df_y_test_class['activity_class_reversed'] = df_y_test_class['activity_class'].map(dict_activity)
print(df_y_test_class)
              61
 Your session crashed after using all available
 RAM. If you are interested in access to high-
                                         View runtime logs
 RAM runtimes, you may want to check out
 Colab Pro.
     1448811
                13
     1448812
                 40
     1448813
                207
     1448814
     1448815
                208
     Name: activity_code, Length: 1448816, dtype: int16
     {0: 'Aerobics, general', 1: 'Aerobics, high impact', 2: 'Aerobics, low impact', 3: 'Aerobics, step aerobics', 4: 'Archery
             activity_class
                                               activity class reversed
     0
                        135
                                                   Riding, snow blower
     1
                         33
                                         Carrying infant, level ground
     2
                        112
                                  Martial arts, judo, karate, jujitsu
     3
                        123
                                                    Playing paddleball
     4
                         54
                                                                Curling
     289759
                        199
                                        Swimming laps, freestyle, slow
     289760
                         71
                                             Frisbee, ultimate frisbee
     289761
                         72
                                                    Gardening, general
     289762
                             Swimming, treading water, fast, vigorous
                        203
     289763
                        218
                                      Walk / run, playing with animals
     [289764 rows x 2 columns]
#cm_display = metrics.ConfusionMatrixDisplay(confusion_matrix = confusion_matrix)
#cm_display.plot()
#plt.show()
```



6s completed at 10:32 PM

X