neural_network

May 4, 2023

[ ]:
```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
import tensorflow as tf
from keras import Sequential, layers, optimizers, models
from random import randint

#* for small network, using GPU will not provide a lot of speed improvement ->
 ↪the following line limits the machine to use only CPU instead
tf.config.set_visible_devices([], 'GPU')
```

1 Data Processing

data from UCI Machine Learning Repositories: https://archive.ics.uci.edu/ml/datasets/SkillCraft1+Master+Table

source: Thompson JJ, Blair MR, Chen L, Henrey AJ (2013) Video Game Telemetry as a Critical Tool in the Study of Complex Skill Learning. PLoS ONE 8(9): e75129.

[ ]:
```
data = pd.read_csv('../data/SkillCraft1_Dataset.csv', na_values='?')
data = data.drop(['GameID'], axis=1)

filtered_data = data[data[data.columns].notnull().all(1)] # filter out any row
 ↪that contains missing value
filtered_data
```

[ ]:

	LeagueIndex	Age	HoursPerWeek	TotalHours	APM	SelectByHotkeys	\
0	5	27.0	10.0	3000.0	143.7180	0.003515	
1	5	23.0	10.0	5000.0	129.2322	0.003304	
2	4	30.0	10.0	200.0	69.9612	0.001101	
3	3	19.0	20.0	400.0	107.6016	0.001034	
4	3	32.0	10.0	500.0	122.8908	0.001136	
…	…	…	…	…	…	…	
3335	4	20.0	8.0	400.0	158.1390	0.013829	
3336	5	16.0	56.0	1500.0	186.1320	0.006951	
3337	4	21.0	8.0	100.0	121.6992	0.002956	
3338	3	20.0	28.0	400.0	134.2848	0.005424	
3339	4	22.0	6.0	400.0	88.8246	0.000844	

	AssignToHotkeys	UniqueHotkeys	MinimapAttacks	MinimapRightClicks	\
0	0.000220	7	0.000110	0.000392	
1	0.000259	4	0.000294	0.000432	
2	0.000336	4	0.000294	0.000461	
3	0.000213	1	0.000053	0.000543	
4	0.000327	2	0.000000	0.001329	
…	…	…	…	…	
3335	0.000504	7	0.000217	0.000313	
3336	0.000360	6	0.000083	0.000166	
3337	0.000241	8	0.000055	0.000208	
3338	0.000182	5	0.000000	0.000480	
3339	0.000108	2	0.000000	0.000341	

	NumberOfPACs	GapBetweenPACs	ActionLatency	ActionsInPAC	\
0	0.004849	32.6677	40.8673	4.7508	
1	0.004307	32.9194	42.3454	4.8434	
2	0.002926	44.6475	75.3548	4.0430	
3	0.003783	29.2203	53.7352	4.9155	
4	0.002368	22.6885	62.0813	9.3740	
…	…	…	…	…	
3335	0.003583	36.3990	66.2718	4.5097	
3336	0.005414	22.8615	34.7417	4.9309	
3337	0.003690	35.5833	57.9585	5.4154	
3338	0.003205	18.2927	62.4615	6.0202	
3339	0.003099	45.1512	63.4435	5.1913	

	TotalMapExplored	WorkersMade	UniqueUnitsMade	ComplexUnitsMade	\
0	28	0.001397	6	0.0	
1	22	0.001193	5	0.0	
2	22	0.000745	6	0.0	
3	19	0.000426	7	0.0	
4	15	0.001174	4	0.0	
…	…	…	…	…	
3335	30	0.001035	7	0.0	
3336	38	0.001343	7	0.0	
3337	23	0.002014	7	0.0	
3338	18	0.000934	5	0.0	
3339	20	0.000476	8	0.0	

	ComplexAbilitiesUsed
0	0.000000
1	0.000208
2	0.000189
3	0.000384
4	0.000019
…	…
3335	0.000287

```
3336              0.000388
3337              0.000000
3338              0.000000
3339              0.000054

[3338 rows x 19 columns]
```

2 Model Construction

[ ]:
```
predict = 'LeagueIndex'
rank: dict = {1: 'Bronze', 2: 'Silver', 3: 'Gold', 4: 'Platinum', 5: 'Diamond',␣
 ↪6: 'Master', 7: 'GrandMaster', 8: 'Professional'}

x = np.array(filtered_data.drop([predict], axis=1))
y = np.array(filtered_data[predict])

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.1)
```

[ ]:
```
model = Sequential()
model.add(layers.Dense(16, input_shape=(18,), activation='sigmoid'))
model.add(layers.Dense(8, activation='sigmoid'))
# model.add(layers.Dense(8, activation='relu'))
model.add(layers.Dense(8, activation='softmax')) #? Dense(8) works
model.summary()

#! using optimizer = 'adam' does not work for M1 architecture
model.compile(optimizer=optimizers.Adam(),␣
 ↪loss='sparse_categorical_crossentropy', metrics=['accuracy'])

model.fit(x_train, y_train, epochs=10)

test_loss, test_acc = model.evaluate(x_test, y_test)
print(f'Test accuracy: {test_acc}')

model.save('../model/8LeagueSkills_NeuralNetwork.h5')
```

```
Model: "sequential"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 dense (Dense)               (None, 16)                304

 dense_1 (Dense)             (None, 8)                 136

 dense_2 (Dense)             (None, 8)                 72

=================================================================
```

```
Total params: 512
Trainable params: 512
Non-trainable params: 0
_________________________________________________________________
Epoch 1/10
2023-05-04 01:10:20.963508: W
tensorflow/tsl/platform/profile_utils/cpu_utils.cc:128] Failed to get CPU
frequency: 0 Hz
94/94 [==============================] - 0s 619us/step - loss: 2.0410 -
accuracy: 0.1771
Epoch 2/10
94/94 [==============================] - 0s 509us/step - loss: 1.8722 -
accuracy: 0.1984
Epoch 3/10
94/94 [==============================] - 0s 503us/step - loss: 1.7758 -
accuracy: 0.2423
Epoch 4/10
94/94 [==============================] - 0s 503us/step - loss: 1.7183 -
accuracy: 0.2603
Epoch 5/10
94/94 [==============================] - 0s 783us/step - loss: 1.6763 -
accuracy: 0.3093
Epoch 6/10
94/94 [==============================] - 0s 494us/step - loss: 1.6412 -
accuracy: 0.3259
Epoch 7/10
94/94 [==============================] - 0s 481us/step - loss: 1.6041 -
accuracy: 0.3322
Epoch 8/10
94/94 [==============================] - 0s 484us/step - loss: 1.5644 -
accuracy: 0.3435
Epoch 9/10
94/94 [==============================] - 0s 489us/step - loss: 1.5230 -
accuracy: 0.3425
Epoch 10/10
94/94 [==============================] - 0s 490us/step - loss: 1.4944 -
accuracy: 0.3472
11/11 [==============================] - 0s 677us/step - loss: 1.5438 -
accuracy: 0.3024
Test accuracy: 0.30239519476890564
```

3 Prediction

[ ]:
```
model: Sequential = models.load_model('../model/8LeagueSkills_NeuralNetwork.h5')
predictions = model.predict(x_train)

for _ in range(10):
    i = randint(0, len(predictions))
    print(f'Prediction: {rank[np.argmax(predictions[i])]}, Actual:␣
 ↪{rank[y_train[i]]}')
```

```
94/94 [==============================] - 0s 403us/step
Prediction: Diamond, Actual: Diamond
Prediction: Diamond, Actual: Platinum
Prediction: Diamond, Actual: Master
Prediction: Diamond, Actual: Diamond
Prediction: Diamond, Actual: Diamond
Prediction: Silver, Actual: Silver
Prediction: Gold, Actual: Gold
Prediction: Diamond, Actual: Diamond
Prediction: Diamond, Actual: Platinum
Prediction: Diamond, Actual: Diamond
```

3.1 4 Leagues Prediction

[ ]:
```
fourLeague_data = filtered_data.copy()

rank: dict = {1: 'Bronze-Silver', 2: 'Gold-Platinum', 3: 'Diamond-Master', 4:␣
 ↪'GrandMaster-Professional'}
fourLeague_data.loc[fourLeague_data['LeagueIndex'] == 2, 'LeagueIndex'] = 1
fourLeague_data.loc[fourLeague_data['LeagueIndex'] == 3, 'LeagueIndex'] = 2
fourLeague_data.loc[fourLeague_data['LeagueIndex'] == 4, 'LeagueIndex'] = 2
fourLeague_data.loc[fourLeague_data['LeagueIndex'] == 5, 'LeagueIndex'] = 3
fourLeague_data.loc[fourLeague_data['LeagueIndex'] == 6, 'LeagueIndex'] = 4
fourLeague_data.loc[fourLeague_data['LeagueIndex'] == 7, 'LeagueIndex'] = 4
fourLeague_data.loc[fourLeague_data['LeagueIndex'] == 8, 'LeagueIndex'] = 4

predict = 'LeagueIndex'

x = np.array(fourLeague_data.drop([predict], axis=1))
y = np.array(fourLeague_data[predict])

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.1)

model = Sequential()
model.add(layers.Dense(16, input_shape=(18,), activation='sigmoid'))
model.add(layers.Dense(8, activation='sigmoid'))
# model.add(layers.Dense(8, activation='relu'))
model.add(layers.Dense(5, activation='softmax')) #? Dense(4) doesn't work
```

```
model.summary()

#! using optimizer = 'adam' does not work for M1 architecture
model.compile(optimizer=optimizers.Adam(),␣
 ↪loss='sparse_categorical_crossentropy', metrics=['accuracy'])

model.fit(x_train, y_train, epochs=10)

test_loss, test_acc = model.evaluate(x_test, y_test)
print(f'Test accuracy: {test_acc}')

model.save('../model/4LeagueSkills_NeuralNetwork.h5')
```

```
Model: "sequential_1"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 dense_3 (Dense)             (None, 16)                304

 dense_4 (Dense)             (None, 8)                 136

 dense_5 (Dense)             (None, 5)                 45

=================================================================
Total params: 485
Trainable params: 485
Non-trainable params: 0
_________________________________________________________________
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 dense_3 (Dense)             (None, 16)                304

 dense_4 (Dense)             (None, 8)                 136

 dense_5 (Dense)             (None, 5)                 45

=================================================================
Total params: 485
Trainable params: 485
Non-trainable params: 0
_________________________________________________________________
Epoch 1/10
94/94 [==============================] - 0s 608us/step - loss: 1.5841 -
accuracy: 0.2400
Epoch 2/10
94/94 [==============================] - 0s 496us/step - loss: 1.4240 -
accuracy: 0.4078
```

```
Epoch 3/10
94/94 [==============================] - 0s 497us/step - loss: 1.3608 -
accuracy: 0.4078
Epoch 4/10
94/94 [==============================] - 0s 502us/step - loss: 1.3274 -
accuracy: 0.4078
Epoch 5/10
94/94 [==============================] - 0s 497us/step - loss: 1.2878 -
accuracy: 0.4078
Epoch 6/10
94/94 [==============================] - 0s 576us/step - loss: 1.2517 -
accuracy: 0.4075
Epoch 7/10
94/94 [==============================] - 0s 1ms/step - loss: 1.2158 - accuracy:
0.4081
Epoch 8/10
94/94 [==============================] - 0s 573us/step - loss: 1.1729 -
accuracy: 0.4095
Epoch 9/10
94/94 [==============================] - 0s 511us/step - loss: 1.1373 -
accuracy: 0.4291
Epoch 10/10
94/94 [==============================] - 0s 506us/step - loss: 1.1082 -
accuracy: 0.4670
11/11 [==============================] - 0s 665us/step - loss: 1.0933 -
accuracy: 0.4731
Test accuracy: 0.473053902387619
```

[ ]:
```
predictions = model.predict(x_train)

for _ in range(10):
    i = randint(0, len(predictions))
    print(f'Prediction: {rank[np.argmax(predictions[i])]}, Actual:␣
  ↪{rank[y_train[i]]}')
```

```
94/94 [==============================] - 0s 434us/step
Prediction: Gold-Platinum, Actual: GrandMaster-Professional
Prediction: Gold-Platinum, Actual: Bronze-Silver
Prediction: Gold-Platinum, Actual: Bronze-Silver
Prediction: Gold-Platinum, Actual: Bronze-Silver
Prediction: Gold-Platinum, Actual: Diamond-Master
Prediction: Gold-Platinum, Actual: Gold-Platinum
Prediction: Diamond-Master, Actual: GrandMaster-Professional
Prediction: Gold-Platinum, Actual: GrandMaster-Professional
Prediction: Gold-Platinum, Actual: GrandMaster-Professional
Prediction: Diamond-Master, Actual: Gold-Platinum
```

3.2 3 Leagues Prediction

[ ]:
```
threeLeague_data = filtered_data.copy()

rank: dict = {1: 'Bronze-Silver-Gold', 2: 'Platinum-Diamond-Master', 3:␣
 ↪'GrandMaster-Professional'}
threeLeague_data.loc[threeLeague_data['LeagueIndex'] == 2, 'LeagueIndex'] = 1
threeLeague_data.loc[threeLeague_data['LeagueIndex'] == 3, 'LeagueIndex'] = 1
threeLeague_data.loc[threeLeague_data['LeagueIndex'] == 4, 'LeagueIndex'] = 2
threeLeague_data.loc[threeLeague_data['LeagueIndex'] == 5, 'LeagueIndex'] = 2
threeLeague_data.loc[threeLeague_data['LeagueIndex'] == 6, 'LeagueIndex'] = 2
threeLeague_data.loc[threeLeague_data['LeagueIndex'] == 7, 'LeagueIndex'] = 3
threeLeague_data.loc[threeLeague_data['LeagueIndex'] == 8, 'LeagueIndex'] = 3

predict = 'LeagueIndex'

x = np.array(threeLeague_data.drop([predict], axis=1))
y = np.array(threeLeague_data[predict])

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.1)

model = Sequential()
model.add(layers.Dense(16, input_shape=(18,), activation='sigmoid'))
model.add(layers.Dense(8, activation='sigmoid'))
# model.add(layers.Dense(8, activation='relu'))
model.add(layers.Dense(4, activation='softmax')) #? Dense(3) doesn't work
model.summary()

#! using optimizer = 'adam' does not work for M1 architecture
model.compile(optimizer=optimizers.Adam(),␣
 ↪loss='sparse_categorical_crossentropy', metrics=['accuracy'])

model.fit(x_train, y_train, epochs=10)

test_loss, test_acc = model.evaluate(x_test, y_test)
print(f'Test accuracy: {test_acc}')

model.save('../model/3LeagueSkills_NeuralNetwork.h5')
```

```
Model: "sequential_2"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 dense_6 (Dense)             (None, 16)                304

 dense_7 (Dense)             (None, 8)                 136

 dense_8 (Dense)             (None, 4)                 36
```

```
=================================================================
Total params: 476
Trainable params: 476
Non-trainable params: 0
_________________________________________________________________
_________________________________________________________________
 Layer (type)                 Output Shape              Param #
=================================================================
 dense_6 (Dense)              (None, 16)                304

 dense_7 (Dense)              (None, 8)                 136

 dense_8 (Dense)              (None, 4)                 36

=================================================================
Total params: 476
Trainable params: 476
Non-trainable params: 0
_________________________________________________________________
Epoch 1/10
94/94 [==============================] - 0s 702us/step - loss: 0.8374 -
accuracy: 0.6654
Epoch 2/10
94/94 [==============================] - 0s 602us/step - loss: 0.7036 -
accuracy: 0.6654
Epoch 3/10
94/94 [==============================] - 0s 983us/step - loss: 0.6362 -
accuracy: 0.7004
Epoch 4/10
94/94 [==============================] - 0s 613us/step - loss: 0.5888 -
accuracy: 0.7490
Epoch 5/10
94/94 [==============================] - 0s 622us/step - loss: 0.5570 -
accuracy: 0.7600
Epoch 6/10
94/94 [==============================] - 0s 589us/step - loss: 0.5343 -
accuracy: 0.7790
Epoch 7/10
94/94 [==============================] - 0s 513us/step - loss: 0.5162 -
accuracy: 0.7886
Epoch 8/10
94/94 [==============================] - 0s 508us/step - loss: 0.5030 -
accuracy: 0.7923
Epoch 9/10
94/94 [==============================] - 0s 521us/step - loss: 0.4982 -
accuracy: 0.7889
Epoch 10/10
```

```
94/94 [==============================] - 0s 582us/step - loss: 0.4933 -
accuracy: 0.7893
11/11 [==============================] - 0s 730us/step - loss: 0.4267 -
accuracy: 0.7904
Test accuracy: 0.7904191613197327
```

[ ]:
```
predictions = model.predict(x_train)

for _ in range(10):
    i = randint(0, len(predictions))
    print(f'Prediction: {rank[np.argmax(predictions[i])]}, Actual:␣
 ↪{rank[y_train[i]]}')
```

```
94/94 [==============================] - 0s 439us/step
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Bronze-Silver-Gold, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
Prediction: Platinum-Diamond-Master, Actual: Platinum-Diamond-Master
```

3.3 2 Leagues Prediction

[ ]:
```
twoLeague_data = filtered_data.copy()

rank: dict = {1: 'Bronze-Silver-Gold-Platinum', 2:␣
 ↪'Diamond-Master-GrandMaster-Professional'}
twoLeague_data.loc[twoLeague_data['LeagueIndex'] == 2, 'LeagueIndex'] = 1
twoLeague_data.loc[twoLeague_data['LeagueIndex'] == 3, 'LeagueIndex'] = 1
twoLeague_data.loc[twoLeague_data['LeagueIndex'] == 4, 'LeagueIndex'] = 1
twoLeague_data.loc[twoLeague_data['LeagueIndex'] == 5, 'LeagueIndex'] = 2
twoLeague_data.loc[twoLeague_data['LeagueIndex'] == 6, 'LeagueIndex'] = 2
twoLeague_data.loc[twoLeague_data['LeagueIndex'] == 7, 'LeagueIndex'] = 2
twoLeague_data.loc[twoLeague_data['LeagueIndex'] == 8, 'LeagueIndex'] = 2

predict = 'LeagueIndex'

x = np.array(twoLeague_data.drop([predict], axis=1))
y = np.array(twoLeague_data[predict])

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.1)

model = Sequential()
model.add(layers.Dense(16, input_shape=(18,), activation='sigmoid'))
```

```
model.add(layers.Dense(8, activation='sigmoid'))
# model.add(layers.Dense(8, activation='relu'))
model.add(layers.Dense(3, activation='softmax')) #? Dense(2) doesn't work
model.summary()

#! using optimizer = 'adam' does not work for M1 architecture
model.compile(optimizer=optimizers.Adam(),␣
 ↪loss='sparse_categorical_crossentropy', metrics=['accuracy'])

model.fit(x_train, y_train, epochs=10)

test_loss, test_acc = model.evaluate(x_test, y_test)
print(f'Test accuracy: {test_acc}')

model.save('../model/2LeagueSkills_NeuralNetwork.h5')
```

```
Model: "sequential_3"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 dense_9 (Dense)             (None, 16)                304

 dense_10 (Dense)            (None, 8)                 136

 dense_11 (Dense)            (None, 3)                 27

=================================================================
Total params: 467
Trainable params: 467
Non-trainable params: 0
_________________________________________________________________
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 dense_9 (Dense)             (None, 16)                304

 dense_10 (Dense)            (None, 8)                 136

 dense_11 (Dense)            (None, 3)                 27

=================================================================
Total params: 467
Trainable params: 467
Non-trainable params: 0
_________________________________________________________________
Epoch 1/10
94/94 [==============================] - 0s 804us/step - loss: 0.7693 -
accuracy: 0.5320
```

```
Epoch 2/10
94/94 [==============================] - 0s 506us/step - loss: 0.6785 -
accuracy: 0.6109
Epoch 3/10
94/94 [==============================] - 0s 514us/step - loss: 0.6098 -
accuracy: 0.7617
Epoch 4/10
94/94 [==============================] - 0s 506us/step - loss: 0.5581 -
accuracy: 0.7733
Epoch 5/10
94/94 [==============================] - 0s 507us/step - loss: 0.5237 -
accuracy: 0.7776
Epoch 6/10
94/94 [==============================] - 0s 497us/step - loss: 0.5007 -
accuracy: 0.7766
Epoch 7/10
94/94 [==============================] - 0s 501us/step - loss: 0.4841 -
accuracy: 0.7853
Epoch 8/10
94/94 [==============================] - 0s 497us/step - loss: 0.4716 -
accuracy: 0.7919
Epoch 9/10
94/94 [==============================] - 0s 497us/step - loss: 0.4677 -
accuracy: 0.7933
Epoch 10/10
94/94 [==============================] - 0s 500us/step - loss: 0.4647 -
accuracy: 0.7909
11/11 [==============================] - 0s 687us/step - loss: 0.5031 -
accuracy: 0.7695
Test accuracy: 0.7694610953330994
```

[ ]:
```
predictions = model.predict(x_train)

for _ in range(10):
    i = randint(0, len(predictions))
    print(f'Prediction: {rank[np.argmax(predictions[i])]}, Actual:␣
 ↪{rank[y_train[i]]}')
```

```
94/94 [==============================] - 0s 400us/step
Prediction: Diamond-Master-GrandMaster-Professional, Actual: Diamond-Master-
GrandMaster-Professional
Prediction: Bronze-Silver-Gold-Platinum, Actual: Bronze-Silver-Gold-Platinum
Prediction: Diamond-Master-GrandMaster-Professional, Actual: Diamond-Master-
GrandMaster-Professional
Prediction: Bronze-Silver-Gold-Platinum, Actual: Bronze-Silver-Gold-Platinum
Prediction: Bronze-Silver-Gold-Platinum, Actual: Bronze-Silver-Gold-Platinum
Prediction: Bronze-Silver-Gold-Platinum, Actual: Bronze-Silver-Gold-Platinum
Prediction: Bronze-Silver-Gold-Platinum, Actual: Bronze-Silver-Gold-Platinum
```

```
Prediction: Diamond-Master-GrandMaster-Professional, Actual: Diamond-Master-
GrandMaster-Professional
Prediction: Diamond-Master-GrandMaster-Professional, Actual: Diamond-Master-
GrandMaster-Professional
Prediction: Diamond-Master-GrandMaster-Professional, Actual: Diamond-Master-
GrandMaster-Professional
```