

Using models

Puteaux, Fall/Winter 2020-2021

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#####  
##                               ##  
## Deep Learning in Python    ##  
##                               ##  
#####
```

\$1 Introduction to Deep Learning in Python

\$1.3 Building deep learning models with keras

1 Using models

1.1 How to use models?

- Save.
- Reload.
- Make predictions.

1.2 Code of saving, reloading, and using the model reloaded:

```
[1]: import pandas as pd  
from keras.layers import Dense  
from keras.models import Sequential  
from keras.utils.np_utils import to_categorical  
  
data = pd.read_csv('ref5. Basketball shot log.csv')  
  
def data_preparation(df):  
    df = df.reindex(columns=[  
        'SHOT_CLOCK', 'DRIBBLES', 'TOUCH_TIME', 'SHOT_DIST', 'CLOSE_DEF_DIST',  
        'SHOT_RESULT'  
    ])  
    df['SHOT_CLOCK'] = df['SHOT_CLOCK'].fillna(0)  
    df['SHOT_RESULT'].replace('missed', 0, inplace=True)  
    df['SHOT_RESULT'].replace('made', 1, inplace=True)  
    df.columns = df.columns.str.lower()
```

```
return df

df = data_preparation(data)
predictors = df.drop(['shot_result'], axis=1).to_numpy()
n_cols = predictors.shape[1]
target = to_categorical(df.shot_result)

model = Sequential()
model.add(Dense(100, activation='relu', input_shape=(n_cols, )))
model.add(Dense(100, activation='relu'))
model.add(Dense(100, activation='relu'))
model.add(Dense(2, activation='softmax'))
model.compile(optimizer='adam',
              loss='categorical_crossentropy',
              metrics=['accuracy'])

model.fit(predictors, target)
```

```
4003/4003 [=====] - 4s 988us/step - loss: 0.6641 -
accuracy: 0.6055
```

```
[1]: <tensorflow.python.keras.callbacks.History at 0x7f9086403310>
```

```
[2]: from keras.models import load_model

model.save('ref7. Model file.h5')
my_model = load_model('ref7. Model file.h5')

predictions = my_model.predict(predictors)
probability_true = predictions[:, 1]
probability_true
```

```
[2]: array([0.43192425, 0.3401706 , 0.35761935, ..., 0.40975374, 0.40411907,
          0.47777078], dtype=float32)
```

```
[3]: my_model.summary()
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 100)	600
dense_1 (Dense)	(None, 100)	10100
dense_2 (Dense)	(None, 100)	10100

```
dense_3 (Dense)                (None, 2)                202
=====
Total params: 21,002
Trainable params: 21,002
Non-trainable params: 0
-----
```

1.3 Practice exercises for using models:

► Package pre-loading:

```
[4]: import pandas as pd
      from keras.layers import Dense
      from keras.models import Sequential
      from keras.utils import to_categorical
```

► Data pre-loading:

```
[5]: df = pd.read_csv('ref6. Titanic.csv')

df.replace(False, 0, inplace=True)
df.replace(True, 1, inplace=True)

predictors = df.drop(['survived'], axis=1).to_numpy()
n_cols = predictors.shape[1]
target = to_categorical(df.survived)

pred_data = pd.read_csv('ref8. Titanic predictors data.csv')
pred_data.replace(False, 0, inplace=True)
pred_data.replace(True, 1, inplace=True)
```

► Making predictions practice:

```
[6]: # Specify, compile, and fit the model
model = Sequential()
model.add(Dense(32, activation='relu', input_shape=(n_cols, )))
model.add(Dense(2, activation='softmax'))
model.compile(optimizer='sgd',
              loss='categorical_crossentropy',
              metrics=['accuracy'])
model.fit(predictors, target)

# Calculate predictions: predictions
predictions = model.predict(pred_data)

# Calculate predicted probability of survival: predicted_prob_true
predicted_prob_true = predictions[:, 1]
```

```
# print predicted_prob_true  
print(predicted_prob_true)
```

28/28 [=====] - 0s 8ms/step - loss: 4.6189 - accuracy:
0.5947

```
[0.33328182 0.46244395 0.98904455 0.45726845 0.23479225 0.23322007  
0.19821607 0.3631005 0.2882916 0.43549198 0.2729657 0.42995116  
0.2880962 0.7672388 0.23744662 0.21667662 0.32192716 0.4371153  
0.17939065 0.6097299 0.35813093 0.2831361 0.20222539 0.385043  
0.8812879 0.23198238 0.3695973 0.8302557 0.23997924 0.32296962  
0.4527913 0.6848475 0.21921225 0.29762074 0.34601828 0.37780848  
0.32957593 0.20942448 0.32904848 0.45084354 0.32383007 0.4500415  
0.42743757 0.2130326 0.36068285 0.19255604 0.80995643 0.22487696  
0.42362073 0.7379827 0.8083559 0.10151579 0.45197144 0.49407145  
0.47167435 0.39426166 0.922086 0.43898413 0.49883062 0.21921225  
0.21931268 0.44096443 0.46569982 0.80711687 0.43567106 0.35603607  
0.4742675 0.388488 0.2390165 0.4645945 0.27332 0.4191591  
0.2877974 0.18710254 0.45809928 0.41166952 0.35927066 0.33088225  
0.20766094 0.3508171 0.44333902 0.22947581 0.3878728 0.34814742  
0.25498232 0.4627092 0.40392458 0.45070586 0.47427937 0.4299565  
0.25209993]
```