

# 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

§1.3.4 Using models

## 1. How to use models?

- Save.
- Reload.
- Make predictions.

## 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)

```

```

1/4003 [...] - ETA: 41:44 - loss: 1.8021 -
accuracy: 0.3438WARNING:tensorflow:Callbacks method `on_train_batch_end` is slow
compared to the batch time (batch time: 0.0015s vs `on_train_batch_end` time:
0.0078s). Check your callbacks.
4003/4003 [=====] - 7s 2ms/step - loss: 0.6672 -
accuracy: 0.6066

```

[1]: <tensorflow.python.keras.callbacks.History at 0x7fa22b48cc90>

```

[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([0.4429109 , 0.24186552, 0.33746815, ..., 0.41122115, 0.392472 ,  
0.44569722], 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

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### 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 [=====] - 1s 19ms/step - loss: 3.1156 - accuracy:  
0.6294

```
[4.5956634e-02 2.5045204e-01 5.1434052e-01 5.0341493e-01 5.2601665e-02  
3.2252628e-02 2.0135791e-04 2.3232634e-01 2.2102144e-02 1.7548123e-01  
7.0456423e-02 1.3825163e-01 2.5755202e-02 2.1712604e-01 3.7076745e-02  
1.4771691e-03 1.2709653e-01 2.1479535e-01 2.9029315e-03 1.5513359e-01  
1.0916419e-01 6.9038771e-02 2.5168591e-04 1.5069677e-01 3.2014742e-01  
4.1736446e-02 1.8013060e-01 7.2635961e-01 4.9539909e-02 1.6717485e-01  
3.5763359e-01 4.2036524e-01 3.4595866e-02 1.0609812e-01 2.1416102e-01  
1.7280167e-01 1.5531802e-01 4.4770289e-02 1.6887575e-01 2.8204811e-01  
1.5904248e-01 3.2595447e-01 3.0528268e-01 1.2362610e-02 2.5197452e-01  
4.8394287e-03 2.4139930e-02 1.5481015e-02 2.2863553e-01 2.0442224e-01  
2.2514185e-01 5.5243977e-06 3.4189633e-01 2.9824379e-01 8.2491070e-02  
2.9127419e-01 1.4131288e-01 3.1110527e-02 3.7589970e-01 3.4595866e-02  
1.3674601e-02 1.7604549e-01 5.2362669e-02 2.9507169e-02 1.7952479e-01  
1.5854310e-02 1.2689959e-01 1.9330798e-01 5.6352358e-02 2.6672539e-01  
7.0547290e-02 1.5474878e-01 8.2864873e-03 2.8222287e-03 2.7167645e-01  
2.8969306e-01 2.0678326e-01 1.7962284e-01 4.3490294e-02 1.7901182e-01  
2.9755443e-01 2.3739902e-02 2.0710088e-01 7.2920546e-02 7.1207143e-02  
3.0904174e-01 1.2200248e-01 2.9367533e-01 3.0490071e-01 2.8705773e-01  
3.2373380e-02]
```