Using models

Puteaux, Fall/Winter 2020-2021

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##
  Deep Learning in Python
##
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§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)
   4003/4003 [============== ] - 6s 1ms/step - loss: 0.6618 -
   accuracy: 0.6065
[1]: <tensorflow.python.keras.callbacks.History at 0x7f8a92b67e10>
[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([0.4321385 , 0.3558106 , 0.3869378 , ..., 0.41570023, 0.4009153 ,
          0.51102567], dtype=float32)
[3]: my_model.summary()
   Model: "sequential"
   Layer (type)
                 Output Shape
   ______
   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

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:

```
0.6233
 \begin{bmatrix} 0.18206367 & 0.4998302 & 0.67459655 & 0.30947936 & 0.13543367 & 0.1102502 \end{bmatrix} 
0.02050491 0.26318935 0.1280276 0.80353606 0.16114196 0.3158905
0.13176644 0.37345943 0.11816965 0.04890123 0.19254042 0.62895584
0.03364213 0.44898477 0.91410786 0.16338897 0.02260221 0.2199869
0.38381287 0.43739226 0.09307115 0.19783701 0.2853902 0.8136704
0.24513188 0.12899524 0.77318317 0.63659924 0.24298626 0.42589793
0.72395575 \ 0.7790671 \ 0.36263737 \ 0.00200928 \ 0.39457786 \ 0.7270656
0.25519454 0.29238302 0.9481265 0.2391511 0.34779522 0.09307115
0.07399313 0.30995914 0.26053128 0.45542064 0.33843926 0.09703109
0.19939627  0.8168582  0.15182117  0.41155747  0.1612887  0.8324817
0.12674804 0.903114 0.6633173 0.09605091 0.26419607 0.21463478
0.15783155 0.3257305 0.2849766 0.71013194 0.39884084 0.6279598
0.08442548]
```