Creating a keras model

Puteaux, Fall/Winter 2020-2021

- §1 Introduction to Deep Learning in Python
- §1.3 Building deep learning models with keras

1 Creating a keras model

- 1.1 What are the model-building steps?
 - The Keras workflow has four steps:
 - specify Architecture
 - compile
 - fit
 - predict

1.2 Code of model specification:

[5]:

count

wage_per_hour

534.000000

1.3 Practice question for understanding the data:

- It will be started soon to building models in Keras to predict wages based on various professional and demographic factors by the next steps. Before starting building a model, it's good to understand the data by performing some exploratory analysis.
- It is recommended to use the .head() and .describe() methods in the IPython Shell to quickly overview the DataFrame.
- The target variable which will be predicting is wage_per_hour. Some of the predictor variables are binary indicators, where a value of 1 represents True, and 0 represents False.
- Of the nine predictor variables in the DataFrame, how many are binary indicators? The min and max values, as shown by .describe() will be informative here. How many binary indicator predictors are there?

```
\square 0.
          \square 5.
          \boxtimes 6.
     ► Package pre-loading:
[2]: import pandas as pd
     ▶ Data pre-loading:
[3]: df = pd.read_csv('ref2. Hourly wages.csv')
     ▶ Question-solving method:
[4]: df.head()
[4]:
         wage_per_hour
                           union
                                   education_yrs
                                                     experience_yrs
                                                                        age
                                                                              female
                                                                                       marr
                    5.10
                                                                                    1
                               0
                                                 8
                                                                   21
                                                                         35
                                                                                           1
                    4.95
                               0
                                                 9
                                                                         57
                                                                                    1
     1
                                                                   42
                                                                                           1
     2
                    6.67
                                                12
                                                                         19
                                                                                    0
                                                                                           0
                               0
                                                                    1
     3
                    4.00
                                                                                    0
                                                                                           0
                               0
                                                12
                                                                    4
                                                                         22
     4
                    7.50
                               0
                                                12
                                                                   17
                                                                         35
                                                                                    0
                                                                                           1
         south
                 manufacturing
                                   construction
     0
              0
     1
              0
                               1
                                                0
     2
              0
                                                0
                               1
     3
                                                0
              0
                               0
     4
              0
                               0
                                                0
     df.describe()
[5]:
```

education_yrs

534.000000

union

534.000000

experience_yrs

534.000000

age

534.000000

```
9.024064
                         0.179775
                                        13.018727
                                                         17.822097
                                                                     36.833333
mean
std
            5.139097
                         0.384360
                                                         12.379710
                                                                     11.726573
                                         2.615373
min
            1.000000
                         0.000000
                                         2.000000
                                                         0.000000
                                                                     18.000000
25%
            5.250000
                         0.000000
                                        12.000000
                                                         8.000000
                                                                     28.000000
50%
                                                         15.000000
                                                                     35.000000
            7.780000
                         0.000000
                                        12.000000
75%
           11.250000
                         0.000000
                                        15.000000
                                                         26.000000
                                                                     44.000000
           44.500000
                         1.000000
                                        18.000000
                                                        55.000000
                                                                     64.000000
max
           female
                                      south
                                             manufacturing
                                                             construction
                          marr
       534.000000 534.000000
                                534.000000
                                                534.000000
                                                               534.000000
count
         0.458801
                      0.655431
                                  0.292135
                                                  0.185393
mean
                                                                 0.044944
std
         0.498767
                      0.475673
                                  0.455170
                                                  0.388981
                                                                 0.207375
min
         0.000000
                      0.000000
                                  0.000000
                                                  0.000000
                                                                 0.000000
25%
         0.000000
                      0.000000
                                  0.000000
                                                  0.000000
                                                                 0.000000
50%
         0.000000
                      1.000000
                                  0.000000
                                                  0.000000
                                                                 0.000000
75%
                      1.000000
         1.000000
                                   1.000000
                                                  0.000000
                                                                 0.000000
max
         1.000000
                      1.000000
                                   1.000000
                                                  1.000000
                                                                 1.000000
```

```
[6]: cols = df.columns
    count = 0
    for i in range(len(cols)):
        if ((df.iloc[:, i].unique()[0] in [0, 1])
             and (df.iloc[:, i].unique()[1] in [0, 1])):
             count += 1
        else:
            pass
    print('There are {} binary indicator predictors here.'.format(count))
```

There are 6 binary indicator predictors here.

1.4 Practice exercises for creating a Keras model:

▶ Package pre-loading:

```
[7]: import pandas as pd
```

▶ Data pre-loading:

```
[8]: df = pd.read_csv('ref2. Hourly wages.csv')

target = df.iloc[:, 0].to_numpy()
predictors = df.iloc[:, 1:].to_numpy()
```

▶ Model specifying practice:

```
[9]: # Import necessary modules
import keras
from keras.layers import Dense
```

```
from keras.models import Sequential

# Save the number of columns in predictors: n_cols
n_cols = predictors.shape[1]

# Set up the model: model
model = Sequential()

# Add the first layer
model.add(Dense(50, activation='relu', input_shape=(n_cols, )))

# Add the second layer
model.add(Dense(32, activation='relu'))

# Add the output layer
model.add(Dense(1))
```

