assignment5_3

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[1]: from keras.datasets import boston_housing

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(train_data, train_targets), (test_data, test_targets) = boston_housing.
      →load_data()
[2]: train_data.shape
[2]: (404, 13)
[3]: test_data.shape
[3]: (102, 13)
[4]: mean = train_data.mean(axis=0)
     train_data -= mean
     std = train_data.std(axis=0)
     train_data /= std
     test_data -= mean
     test_data /= std
[5]: def build model():
         model = models.Sequential()
         model.add(layers.Dense(64, activation='relu',
                               input_shape=(train_data.shape[1],)))
         model.add(layers.Dense(64, activation='relu'))
         model.add(layers.Dense(1))
         model.compile(optimizer='rmsprop', loss='mse', metrics=['mae'])
         return model
[8]: import numpy as np
     from keras import models
     from keras import layers
     k = 4
     num_val_samples = len(train_data) // k
     num_epochs = 100
     all_scores = []
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for i in range(k):
          print('processing fold #', i)
          val_data = train_data[i * num_val_samples: (i + 1) * num_val_samples]
          val_targets = train_targets[i * num_val_samples: (i + 1) * num_val_samples]
          partial_train_data = np.concatenate(
              [train_data[:i * num_val_samples],
              train_data[(i + 1) * num_val_samples:]],
          partial_train_targets = np.concatenate(
              [train_targets[:i * num_val_samples],
              train_targets[(i + 1) * num_val_samples:]],
              axis=0)
          model = build_model()
          model.fit(partial_train_data, partial_train_targets,
                  epochs=num_epochs, batch_size=1, verbose=0)
          val_mse, val_mae = model.evaluate(val_data, val_targets, verbose=0)
          all_scores.append(val_mse)
     processing fold # 0
     processing fold # 1
     processing fold # 2
     processing fold # 3
 [9]: all_scores
 [9]: [9.317792892456055, 10.8341064453125, 19.7426815032959, 10.211949348449707]
[10]: np.mean(all_scores)
[10]: 12.52663254737854
[13]: print(history.history.keys())
     dict_keys(['loss', 'mae', 'val_loss', 'val_mae'])
[15]: num_epochs = 500
      all_mae_histories = []
      for i in range(k):
          print('processing fold #', i)
          val_data = train_data[i * num_val_samples: (i + 1) * num_val_samples]
          val_targets = train_targets[i * num_val_samples: (i + 1) * num_val_samples]
          partial_train_data = np.concatenate(
              [train_data[:i * num_val_samples],
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train_data[(i + 1) * num_val_samples:]],
              axis=0)
          partial_train_targets = np.concatenate(
              [train_targets[:i * num_val_samples],
              train_targets[(i + 1) * num_val_samples:]],
              axis=0)
          model = build_model()
          model.fit(partial_train_data, partial_train_targets,
                  epochs=num_epochs, batch_size=1, verbose=0)
          history = model.fit(partial_train_data, partial_train_targets,
                             validation_data=(val_data, val_targets),
                             epochs=num_epochs, batch_size=1, verbose=0)
          mae_history = history.history['val_mae']
          all_mae_histories.append(mae_history)
     processing fold # 0
     processing fold # 1
     processing fold # 2
     processing fold # 3
[16]: average_mae_history = [
          np.mean([x[i] for x in all_mae_histories]) for i in range(num_epochs)]
[17]: import matplotlib.pyplot as plt
      plt.plot(range(1, len(average_mae_history) + 1), average_mae_history)
      plt.xlabel('Epochs')
      plt.ylabel('Validation MAE')
      plt.show()
```



