物聯網實務第十二周作業

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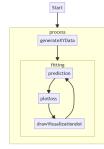
December 5, 2022

Exercise 11-1 Visualize TF gradient descent.

Full detail in https://github.com/belongtothenight/IoTP_Code/blob/main/src/w12/ex11_1.md

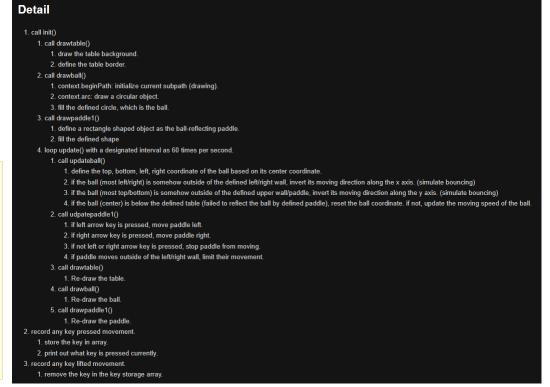
Detail

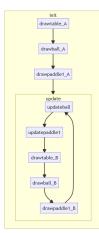
- 1. process()
 - 1. generate initial linear regression coefficient.
 - 2. call generateXYData() to retreive training data sets.
 - 3. feed training datasets to fitting()
- 2. generateXYData()
 - 1. generate x data in the range between 0 and 1.
 - 2. generate y data based on corresponding x value and a random component.
 - 3. return two datasets: x, y.
- 3. fitting()
 - 1. set the method to construct neural network as sequential instead of functional. src
 - 2. add a dense layer
 - 1. tf.layers.dense.units: neuron count
 - 2. tf.layers.dense.inputShape: input node count
 - 3. set training optimizer as tf.SGDOptimizer which utilizes stochastic gradient descent(extreme large dataset).
 - 4. configure model settings for training.
 - 5. add bias node for input layer.
 - 6. convert datasets to tensor sets.
 - 7. start training.
 - 1. tf.keras.model.fit.batchSize: the number of samples to work through before updating the internal model parameters.
 - 2. tf.keras.model.fit.epochs: the number times that the learning algorithm will work through the entire training dataset.
 - 3. tf.keras.model.fit.callbacks: access point of the model during training. In this case, it does the following stuff.
 - tf.keras.callbacks.Callback.onEpochEnd: called at the end of Epoch, and perform the following stuff.
 - 1. print out epoch count and log.
 - 2. call Prediction().
 - 1. prepare x data (input) for prediction (add bias node).
 - 2. turn prepared x data into tensor.
 - 3. tf.keras.model.predict: use input tensor to predict with current iteration of NN.
 - 4. tf.Tensor.dataSync: retrieve data from tensor.
 - 5. tf.keras.model.dispose: delete model.
 - 6. tf. Tensor.dispose: delete predict result from memory.
 - 3. call plotData2()
 - 1. restructure input array.
 - 2. create vegaEmbed data structure containing restructured array as data and plot it on webpage.
 - 4. call plotloss().
 - 1. restructure input array.
 - 2. create vegaEmbed data structure containing restructured array as data and plot it on webpage.
 - 5. extract NN node current weight and log.
 - 6. append newly retrieve data.
 - 7. call drawVisualizationdot()
 - 1. create vis.Graph3d data structure.
 - 2. draw graph.
 - 2. jump back to 1, and loop until the training is finished.



Exercise 11-2 Build the fundation of the pong game.

 $Full \ detail \ in \ https://github.com/belongtothenight/IoTP_Code/blob/main/src/w12/ex11_2.md$





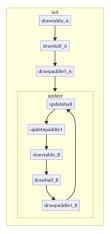
Exercise 11-3 Gather player's control data for later use.

 $Full\ detail\ in\ https://github.com/belongtothenight/IoTP_Code/blob/main/src/w12/ex11_3.md$



Exercise 11-4 Gather more player's control data for later use.

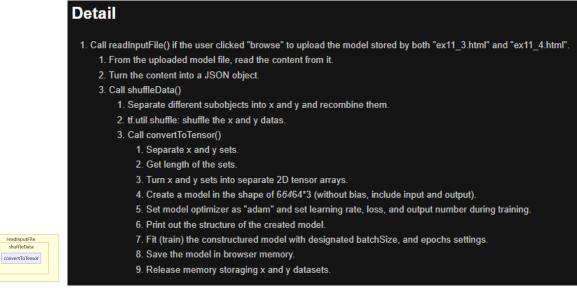
 $Full \ detail \ in \ https://github.com/belongtothenight/IoTP_Code/blob/main/src/w12/ex11_4.md$



Gather more data then exercise 11-3.

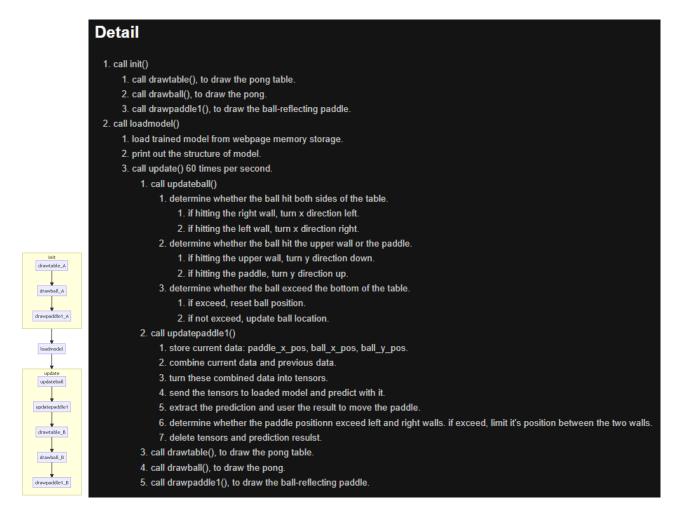
Exercise 11-5 Use gathered data to train NN model.

Full detail in https://github.com/belongtothenight/IoTP_Code/blob/main/src/w12/ex11_5.md



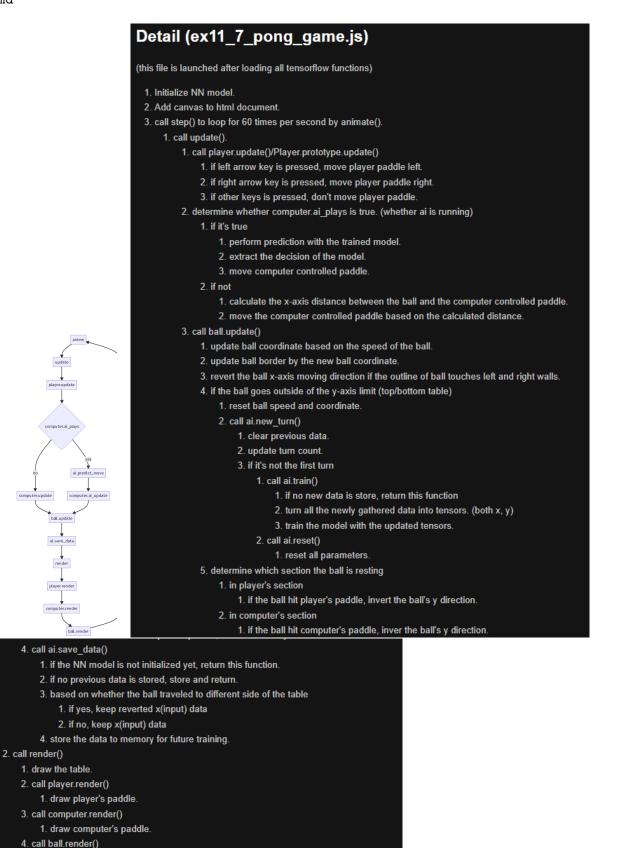
Exercise 11-6 Use trained model to interact with player.

Full detail in https://github.com/belongtothenight/IoTP_Code/blob/main/src/w12/ex11_6.md



Exercise 11-7 Gather player data while playing and use them to train NN realtime.

 $Full \ detail \ in \ https://github.com/belongtothenight/IoTP_Code/blob/main/src/w12/ex11_7.md$



1. draw the ball.