

C1W3_Assignment

May 30, 2021

1 Week 3 Assignment: Implement a Quadratic Layer

In this week's programming exercise, you will build a custom quadratic layer which computes $y = ax^2 + bx + c$. Similar to the ungraded lab, this layer will be plugged into a model that will be trained on the MNIST dataset. Let's get started!

1.0.1 Imports

```
[1]: import tensorflow as tf
from tensorflow.keras.layers import Layer

import utils
```

1.0.2 Define the quadratic layer (TODO)

Implement a simple quadratic layer. It has 3 state variables: a , b and c . The computation returned is $ax^2 + bx + c$. Make sure it can also accept an activation function.

`__init__`

- call `super(my_fun, self)` to access the base class of `my_fun`, and call the `__init__()` function to initialize that base class. In this case, `my_fun` is `SimpleQuadratic` and its base class is `Layer`.
- `self.units`: set this using one of the function parameters.
- `self.activation`: The function parameter `activation` will be passed in as a string. To get the tensorflow object associated with the string, please use `tf.keras.activations.get()`

build The following are suggested steps for writing your code. If you prefer to use fewer lines to implement it, feel free to do so. Either way, you'll want to set `self.a`, `self.b` and `self.c`.

- `a_init`: set this to tensorflow's `random_normal_initializer()`
- `a_init_val`: Use the `random_normal_initializer()` that you just created and invoke it, setting the `shape` and `dtype`.

- The `shape` of `a` should have its row dimension equal to the last dimension of `input_shape`, and its column dimension equal to the number of units in the layer.
- This is because you'll be matrix multiplying $x^2 * a$, so the dimensions should be compatible.
- set the dtype to 'float32'
- `self.a`: create a tensor using `tf.Variable`, setting the `initial_value` and set trainable to `True`.
- `b_init`, `b_init_val`, and `self.b`: these will be set in the same way that you implemented `a_init`, `a_init_val` and `self.a`
- `c_init`: set this to `tf.zeros_initializer`.
- `c_init_val`: Set this by calling the `tf.zeros_initializer` that you just instantiated, and set the `shape` and `dtype`
 - `shape`: This will be a vector equal to the number of units. This expects a tuple, and remember that a tuple `(9,)` includes a comma.
 - `dtype`: set to 'float32'.
- `self.c`: create a tensor using `tf.Variable`, and set the parameters `initial_value` and `trainable`.

call The following section performs the multiplication $x^2a + xb + c$. The steps are broken down for clarity, but you can also perform this calculation in fewer lines if you prefer.

- `x_squared`: use `tf.math.square()`
- `x_squared_times_a`: use `tf.matmul()`.
- If you see an error saying `InvalidArgumentError: Matrix size-incompatible`, please check the order of the matrix multiplication to make sure that the matrix dimensions line up.
- `x_times_b`: use `tf.matmul()`.
- `x2a_plus_xb_plus_c`: add the three terms together.
- `activated_x2a_plus_xb_plus_c`: apply the class's `activation` to the sum of the three terms.

```
[9]: ## Please uncomment all lines in this cell and replace those marked with `#`  

    ↪YOUR CODE HERE`.  

## You can select all lines in this code cell with Ctrl+A (Windows/Linux) or  

    ↪Cmd+A (Mac), then press Ctrl+/ (Windows/Linux) or Cmd+/ (Mac) to uncomment.  

  

class SimpleQuadratic(Layer):  

    def __init__(self, units=32, activation=None):  

#         '''Initializes the class and sets up the internal variables'''  

        super(SimpleQuadratic, self).__init__()  

        self.units = units  

        self.activation = tf.keras.activations.get(activation)  

  

    def build(self, input_shape):  

#         '''Create the state of the layer (weights)'''
```

```

#         # a and b should be initialized with random normal, c (or the bias)
#         ↪with zeros.
#         # remember to set these as trainable.
a_init = tf.random_normal_initializer()
b_init = tf.random_normal_initializer()
c_init = tf.zeros_initializer()

a_init_val = a_init(shape = (input_shape[-1], self.units),
#         ↪dtype='float32')
b_init_val = b_init(shape = (input_shape[-1], self.units),
#         ↪dtype='float32')
c_init_val = c_init(shape = (self.units,), dtype='float32')

self.a = tf.Variable(a_init_val, trainable =True)
self.b = tf.Variable(b_init_val, trainable =True)
self.c = tf.Variable(c_init_val, trainable =True)

def call(self, inputs):
#         '''Defines the computation from inputs to outputs'''
    return self.activation( tf.matmul( tf.square(inputs), self.a) + tf.
#         ↪matmul(inputs, self.b) + self.c)

```

Test your implementation

```
[10]: utils.test_simple_quadratic(SimpleQuadratic)
```

All public tests passed

Train your model with the SimpleQuadratic layer that you just implemented.

```

[11]: # THIS CODE SHOULD RUN WITHOUT MODIFICATION
# AND SHOULD RETURN TRAINING/TESTING ACCURACY at 97%+

mnist = tf.keras.datasets.mnist

(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_train, x_test = x_train / 255.0, x_test / 255.0

model = tf.keras.models.Sequential([
    tf.keras.layers.Flatten(input_shape=(28, 28)),
    SimpleQuadratic(128, activation='relu'),
    tf.keras.layers.Dropout(0.2),
    tf.keras.layers.Dense(10, activation='softmax')
])

model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])

```

```
model.fit(x_train, y_train, epochs=5)
model.evaluate(x_test, y_test)
```

Train on 60000 samples

Epoch 1/5

60000/60000 [=====] - 13s 209us/sample - loss: 0.2728 -
accuracy: 0.9189

Epoch 2/5

60000/60000 [=====] - 12s 203us/sample - loss: 0.1364 -
accuracy: 0.9596

Epoch 3/5

60000/60000 [=====] - 12s 203us/sample - loss: 0.1006 -
accuracy: 0.9688

Epoch 4/5

60000/60000 [=====] - 12s 202us/sample - loss: 0.0828 -
accuracy: 0.9738

Epoch 5/5

60000/60000 [=====] - 12s 203us/sample - loss: 0.0718 -
accuracy: 0.9767

10000/10000 [=====] - 1s 68us/sample - loss: 0.0699 -
accuracy: 0.9774

[11]: [0.06991157725492958, 0.9774]

[]: