grade 100%

Custom Layers

LATEST SUBMISSION GRADE

100%		
1.	Lambda layer allows to execute an arbitrary function only within a Sequential API model. True False	1/1 point
	✓ Correct Correct!	
2.	Which one of the following is the correct syntax for mapping an increment of 2 to the value of "x" using a Lambda layer? (tf = Tensorflow) tf.keras.layers.Lambda(lambda x: tf.math.add(x, 2.0)) tf.keras.layers(lambda x: tf.math.add(x, 2.0)) tf.keras.layers(lambda x: tf.math.add(x, 2.0)) tf.keras.layers(lambda x: tf.math.add(x, 2.0))	1/1 point
	✓ Correct Correct!	
3.	One drawback of Lambda layers is that you cannot call a custom built function from within them. True False	1/1 point
	✓ Correct Correct!	
4.	A Layer is defined by having "States" and "Computation". Consider the following code and check all that are true: class SimpleDense(Layer): definit(self, units=32): super(SimpleDense, self)init() self.units = units def build(self, input_shape): w_init = tf.random_normal_initializer() self.w = tf.Variable(name="kernel",	1/1 point
	<pre>dtype='float32'), trainable=True) b_init = tf.zeros_initializer() self.b = tf.Variable(name="bias",</pre>	
	<pre>def call(self, inputs): return tf.matmul(inputs, self.w) + self.b</pre>	
	def call(self, inputs): performs the computation and is called when the Class is instantiated. In definit(self, units=32): you use the <i>super</i> keyword to initialize all of the custom layer attributes	
	You use def build(self, input_shape): to create the state of the layers and specify local input states.	
	✓ Correct Correct!	
	After training, this class will return a w*X + b computation, where X is the input, w is the weight/kernel tensor with trained values, and b is the bias tensor with trained values.	

```
class SimpleDense(Layer):
   def __init__(self, units=32):
    super(SimpleDense, self).__init__()
      self.units = units
   def build(self, input_shape):
      w_init = tf.random_normal_initializer()
self.w = tf.Variable(name="kernel",
                           b_init = tf.zeros_initializer()
      self.b = tf.Variable(name="bias",
                           initial_value=b_init(shape=(self.units,), dtype='float32'),
                                                                                  trainable=False)
   def call(self, inputs):
    return tf.matmul(inputs, self.w) + self.b
What are the function modifications that are needed for passing an activation function to this custom layer
implementation?
def build(self, units=32, activation=None):
     self.activation = activation
   def call(self, inputs):
     return self.activation(tf.matmul(inputs, self.w) + self.b)
def __init__(self, units=32, activation=None):
     self.activation = tf.keras.activations.get(activation)
   def call(self, inputs):
     return self.activation(tf.matmul(inputs, self.w) + self.b)
def __init__(self, units=32):
     self.activation = tf.keras.activations.get(activation)
    def call(self, inputs):
      return self.activation(tf.matmul(inputs, self.w) + self.b)
def build(self, input_shape):
     self.activation = tf.keras.activations.get(activation)
   def call(self, inputs):
    return self.activation(tf.matmul(inputs, self.w) + self.b)
   ✓ Correct
```

Correct!