Your grade: 100%

Your latest: 100% • Your highest: 100% • To pass you need at least 80%. We keep your highest score.

Next item →

1.	Lambda layer allows to execute an arbitrary function only within a Sequential API model.	1/1 point
	True False	
	⊘ 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)	1/1 point
	O tf.keras.layers(lambda x: tf.math.add(x, 2.0))	
	O tf.keras.layers.Lambda(x: tf.math.add(x, 2.0))	
	(tf.keras.Lambda(x: tf.math.add(x, 2.0))	
	(ambda x: tf.math.add(x, 2.0))	
	⊘ correct Correct!	
3.	One drawback of Lambda layers is that you cannot call a custom built function from within them.	1/1 point
	O True	
	False	
	⊘ correct Correct!	
4.	A <i>Layer</i> is defined by having "States" and "Computation". Consider the following code and check all that are true:	1/1 point
	class SimpleDense(Layer):	
	<pre>definit(self, units=32): super(SimpleDense, self)init() self.units = units</pre>	
	<pre>def build(self, input_shape): w_init = tf.random_normal_initializer() self.w = tf.Variable(name="kernel",</pre>	
	initial_value=w_init(shape=(input_shape[-1], self.units), dtype='float32'), trainable=True)	
	<pre>b_init = tf.zeros_initializer()</pre>	
	<pre>self.b = tf.Variable(name="bias",</pre>	
	<pre>def call(self, inputs): return tf.matmul(inputs, self.w) + self.b</pre>	
	✓ You use def build(self, input_shape): to create the state of the layers and specify local input states.	
	○ Correct	
	Correct!	
	def call(self, inputs): performs the computation and is called when the Class is instantiated.	
	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.	
	☐ In definit(self, units=32): you use the <i>super</i> keyword to initialize all of the custom layer attributes	

```
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)
O 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)
O def build(self, input_shape):
     {\tt self.activation = tf.keras.activations.get(activation)}
    def call(self, inputs):
     return self.activation(tf.matmul(inputs, self.w) + self.b)
  ⊘ Correct
     Correct!
```