

C1W2_Assignment

December 13, 2020

1 W2 Assignment: Creating a Custom Loss Function

This short exercise will require you to write a simple linear regression neural network that is trained on two arrays: xs (inputs) and ys (labels), where the relationship between each corresponding element is $y = 2x - 1$.

$xs = [-1.0, 0.0, 1.0, 2.0, 3.0, 4.0]$

$ys = [-3.0, -1.0, 1.0, 3.0, 5.0, 7.0]$

You will need to implement a custom loss function that returns the root mean square error (RMSE) of $y_{true} - y_{pred}$. Let's begin!

```
[1]: import tensorflow as tf
import numpy as np
from tensorflow import keras
from tensorflow.keras import backend as K

import utils
```

```
[2]: # inputs
xs = np.array([-1.0, 0.0, 1.0, 2.0, 3.0, 4.0], dtype=float)

# labels. relationship with the inputs above is y=2x-1.
ys = np.array([-3.0, -1.0, 1.0, 3.0, 5.0, 7.0], dtype=float)
```

1.0.1 Define the custom loss function (TODO)

Define the custom loss function below called `my_rmse()` that returns the RMSE between the target (`y_true`) and prediction (`y_pred`).

You will return $\sqrt{\text{error}}$, where $\text{error} = \text{mean}((y_{true} - y_{pred})^2)$ - error: the difference between the true label and predicted label. - `sqr_error`: the square of the error. - `mean_sqr_error`: the mean of the square of the error - `sqrt_mean_sqr_error`: the square root of the mean of the square of the error (the root mean squared error). - Please use `K.mean`, `K.square`, and `K.sqrt` - The steps are broken down into separate lines of code for clarity. Feel free to combine them, and just remember to return the root mean squared error.

```
[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.  

  

def my_rmse(y_true, y_pred):  

    error = y_true-y_pred # YOUR CODE HERE  

    sqr_error = K.square(error)# YOUR CODE HERE  

  

    mean_sqr_error =K.mean(sqr_error) # YOUR CODE HERE  

  

    sqrt_mean_sqr_error =K.sqrt(mean_sqr_error) # YOUR CODE HERE  

    return sqrt_mean_sqr_error
```

```
[10]: utils.test_my_rmse(my_rmse)
```

All public tests passed

1.0.2 Define a model using the custom loss function (TODO)

Similar to the ungraded labs, you will define a simple model and pass the function you just coded as the loss. - When compiling the model, you'll choose the `sgd` optimizer and set the `loss` parameter to the custom loss function that you just defined. - For grading purposes, please leave the other parameter values as is.

```
[11]: # 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.  

  

# define the model architecture  

model = tf.keras.Sequential([keras.layers.Dense(units=1, input_shape=[1])])  

  

# use the function you just coded as the loss  

model.compile(optimizer='sgd', loss=my_rmse)  

  

# train the model  

model.fit(xs, ys, epochs=500,verbose=0)  

  

# test with a sample input  

print(model.predict([10.0]))
```

```
[[19.093954]]
```

```
[12]: utils.test_model_loss(model.loss)
```

```
All public tests passed
```

```
[ ]:
```