```
In [2]: import numpy as np
         import pandas as pd
         import tensorflow as tf
         from sklearn.preprocessing import OneHotEncoder
         from sklearn.metrics import accuracy_score
         import time
         from sklearn.preprocessing import scale
In [4]: | dd_train = pd.read_csv('Training Data.csv')
         dd_test = pd.read_csv('Testing Data.csv')
         dd_real = pd.read_csv('Real Data.csv')
In [51]: in_train = dd_train.values[:,1:129].astype(np.float64)
         in_train = scale(in_train, axis = 1)
         out_train = dd_train.label.astype('category')
         out_train = pd.get_dummies( out_train ).values.astype(np.float64)
         in valid = dd_test.values[:,1:129].astype(np.float64)
         in_valid = scale(in_valid, axis = 1)
         out valid = dd test.label.astype('category')
         out_valid = pd.get_dummies( out_valid ).values.astype(np.float64)
         in real = dd real.values[:,1:129].astype(np.float64)
         in real = scale(in real, axis = 1)
In [52]: | all_labels = np.unique(dd_test.label.astype('category'))
         dd real.label.values
Out[52]: array(['Looter Scooter'], dtype=object)
In [53]: out real = np.reshape((all labels == "Smuggler's Copter") +0,
         [1,264]).astype(np.float64)
         in real.shape
Out[53]: (1, 128)
In [54]: N_train = in_train.shape[0]
         N_feat = in_train.shape[1]
         N cat = out train.shape[1]
         N valid = in valid.shape[0]
         N real = in real.shape[0]
In [55]: N_train
Out[55]: 52800
In [56]: | # hyper-parameters
         EPOCHS = 300
                              # number of training epochs
         N_nodes = 256
                              # nodes in hidden layer
         ALPHA
                              # regularization parameter
                 = 0
         BS
                 = 52800
                                 # batch size
         STD
                 = 0.1
                              # weight initialization standard deviation
```

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In [86]: x train = tf.placeholder( tf.float32, [BS, N feat] )
         y train = tf.placeholder( tf.float32, [BS, N cat] )
         x train f = tf.constant( in_train , tf.float32, [N_train,N_feat])
         y_train_f = tf.constant( out_train , tf.float32, [N_train,N_cat])
         x_valid = tf.constant( in_valid , tf.float32, [N_valid , N_feat] )
         y_valid = tf.constant( out_valid, tf.float32, [N_valid, N_cat])
         x_real = tf.constant( in_real , tf.float32, [N_real , N_feat] )
         y_real = tf.constant( out_real, tf.float32, [N_real, N_cat])
In [71]: w1 = tf.Variable( tf.truncated_normal( [N_feat,N_nodes], stddev = STD,
         seed=0))
         b1 = tf.Variable( tf.truncated_normal( [1,N_nodes], stddev = STD, seed=0))
         y1_train = tf.nn.relu( tf.matmul(x_train,w1) + b1 )
         y1_train_f = tf.nn.relu( tf.matmul(x_train_f,w1) + b1 )
         y1_valid = tf.nn.relu( tf.matmul( x_valid, w1) + b1 )
         y1_real = tf.nn.relu( tf.matmul( x_real, w1) + b1 )
         w2 = tf.Variable( tf.truncated_normal( [N_nodes,N_cat], stddev = STD, seed=0))
         b2 = tf.Variable( tf.truncated_normal( [1,N_cat], stddev = STD, seed=0))
         logits_train = tf.matmul(y1_train,w2) + b2
         logits train f = tf.matmul(y1 train f,w2) + b2
         logits_valid = tf.matmul(y1_valid,w2) + b2
         logits_real = tf.matmul(y1_real,w2) + b2
In [72]: CE = tf.reduce_mean( tf.nn.softmax_cross_entropy_with_logits(
                 logits train, y train) )
         L2 = ALPHA*(tf.nn.12 loss(w1) + tf.nn.12 loss(b1) + 
                        tf.nn.12_loss(w2) + tf.nn.12_loss(b2))
         CE_train_f = tf.reduce_mean( tf.nn.softmax_cross_entropy_with_logits(
                 logits_train_f, y_train_f) )
         CE_valid = tf.reduce_mean( tf.nn.softmax_cross_entropy_with_logits(
                 logits_valid, y_valid) )
         y_train_p = tf.nn.softmax( logits_train )
         y_train_fp = tf.nn.softmax( logits_train_f )
         y_valid_p = tf.nn.softmax( logits_valid )
         y_real_p = tf.nn.softmax( logits_real)
         optimizer = tf.train.AdamOptimizer().minimize(CE+L2)
         init = tf.initialize_all_variables()
In [73]: sess = tf.Session()
         sess.run(init)
```

```
In [74]: t0 = time.time()
         np.random.seed(0)
         print( 'Epochs = %3d, Training Set Size = %4d, Nodes = %5d, Alpha = %3.4f, Batch
          Size = %4d, STD = %5.3f' %
               (EPOCHS, N_train, N_nodes, ALPHA, BS, STD))
         print()
         print('%15s%24s%24s' % (' ','cross-entropy','error-rate'))
         print('%15s%12s%12s%12s%12s%12s' %
               ('epoch', 'training', 'validation', 'training', 'validation', 'L2', 'time (mi
         n)'))
         for i in range(EPOCHS+1): # For every Epoch
             ran = np.random.permutation(N_train) # Order the data
             reran = np.reshape( ran, [ int(N_train/BS) ,BS] ) #Reshape ordering as a m
         atrix
             for j in range( int(N_train/BS) ): # For every batch
                 mini = reran[j,:].astype(int) # Find the batch indices
                 x_batch = in_train[mini,:] # Call the batch features
                 y_batch = out_train[mini] # Call the batch labels
                 # Do a step with a batch
                 sess.run([optimizer],feed_dict = {x_train:x_batch, y_train:y_batch})
             if (i % int(EPOCHS/10)) == 0:
                 ( ce_train, ce_valid, out_train_pf, out_valid_p, 12 ) = \
                 sess.run( [CE_train_f, CE_valid, y_train_fp, y_valid_p,L2])
                 err_train = 1-accuracy_score( out_train.argmax(axis=1),
                                            out_train_pf.argmax(axis=1))
                 err_valid = 1-accuracy_score( out_valid.argmax(axis=1),
                                             out_valid_p.argmax(axis=1))
                 t = (time.time() - t0)/60
                 print('%7d %7d %12.5f %12.5f %12.6f %12.3f %12.1f' %
                       (EPOCHS,i,ce_train,ce_valid,err_train,err_valid,12,t))
```

Epochs =300, Training Set Size =52800, Nodes = 256, Alpha = 0.0000, Batch S ize = 52800, STD = 0.100

			cross-entropy		error-rate		
		epoch	training	validation	training	validation	L2
time (min)							
	300	0	5.85669	5.85707	0.99744	13 0.998030	0.
000		0.0					
	300	30	1.45093	1.45758	0.08126	59 0.084508	0.
000		0.2					
	300	60	0.09357	0.09711	0.00329	95 0.003750	0.
000		0.3					
	300	90	0.03798	0.04135	0.00132	26 0.001932	0.
000		0.5					
	300	120	0.02485	0.02827	0.00079	95 0.001515	0.
000		0.7					
	300	150	0.01810	0.02155	0.00043	36 0.001174	0.
000		0.8					
	300	180	0.01389	0.01733	0.00022	27 0.000985	0.
000		1.0					
	300	210	0.01104	0.01445	0.00013	33 0.000871	0.
000		1.1					
	300	240	0.00901	0.01238	0.00011	L4 0.000795	0.
000		1.3					
	300	270	0.00751	0.01083	0.00005	0.000720	0.
000		1.4					
	300	300	0.00636	0.00963	0.00003	38 0.000682	0.
000		1.6					

```
In [75]: (out_real_p) = sess.run( [y_real_p])
```

In [89]: all\_labels[out\_real\_p[0].argmax()]

Out[89]: 'Demolish'

In [92]: out\_real\_p = out\_real\_p[0]
 out\_real\_p.argmax()

Out[92]: 60

In [93]: all\_labels[60]

Out[93]: 'Demolish'

In [94]: out\_real\_p[60]

Out[94]: 0.34933463

In [ ]: