In [1]:

```
import os
import copy
import time
import shelve
import random
import numpy as np
import pandas as pd
import tensorflow as tf
from pandas import DataFrame
import matplotlib.pyplot as plt
```

In [2]:

```
def load data(name):
    if name == 'mnist':
        (X_train, y_train), (X_test, y_test) = tf.keras.datasets.mnist.load_data()
    elif name == 'fashion mnist':
        (X_train, y_train), (X_test, y_test) = tf.keras.datasets.fashion_mnist.loa
d data()
    else:
        print('Only mnist or fashion mnist.')
        return False
    imageSize = X train.shape[1]*X train.shape[2]
    numClasses = np.max(y train)+1
    X train = np.reshape(X train.astype(float)/255.0, (-1, 784))
    X \text{ test} = \text{np.reshape}(X \text{ test.astype}(float)/255.0, (-1, 784))
    y train = tf.keras.utils.to categorical(y train, num classes=numClasses)
    y test = tf.keras.utils.to categorical(y test, num classes=numClasses)
    X val
            = X train[-10000:]
            = y train[-10000:]
    y val
    X train = X train[:-10000]
    y_{train} = y_{train}[:-10000]
    print('Data Split: ')
    print(f'X train: {X train.shape}, y train: {y train.shape}')
    print(f'X_test : {X_test.shape }, y_test : {y_test.shape }')
    print(f'X_val : {X_val.shape }, y_val : {y_val.shape }')
    data = \{\}
    data['X_train'] = X_train
    data['y_train'] = y_train
    data['X_val'] = X_val
    data['y_val'] = y_val
    data['X_test'] = X_test
    data['y_test'] = y_test
    data['imageSize'] = imageSize
    return data
```

In [3]:

```
class myOptimizer():
            def __init__(self, var):
                        self.m = [np.zeros(np.shape(i)) for i in var]
                        self.v = [np.zeros(np.shape(i)) for i in var]
                        self.u = [np.zeros(np.shape(i)) for i in var]
                                                               = 5e-4
                        self.alpha
                                                               = [0.9, 0.999, 0.999987]
                        self.beta
                        self.epsilon = [1e-8, 1e-6]
            def updateWeights(self, t, grads, var):
                        for i in range(np.shape(var)[0]):
                                    self.m[i] = (self.beta[0]*self.m[i]) + ((1 - self.beta[0])*grads[i])
                                    self.v[i] = (self.beta[1]*self.v[i]) + ((1 - self.beta[1])*(grads[i]*g
rads[i]))
                                    self.u[i] = (self.beta[2]*self.u[i]) + ((1 - self.beta[2])*(grads[i]*grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[i])*(grads[
rads[i]*grads[i]))
                                    mHat = self.m[i]/(1-self.beta[0]**t)
                                    vHat = self.v[i]/(1-self.beta[1]**t)
                                    uHat = self.u[i]/(1-self.beta[2]**t)
                                    # delta = self.alpha*mHat/((tf.sign(vHat)*tf.sqrt(tf.abs(vHat)))+(tf.s
ign(uHat)*tf.math.pow(tf.abs(uHat), 1/3)*self.epsilon[0])+self.epsilon[0])
                                    delta = self.alpha * mHat / (tf.sqrt(vHat)+(tf.sign(uHat)*tf.math.pow(
tf.abs(uHat), 1/3)*self.epsilon[0] + self.epsilon[1]))
                                    var[i].assign(var[i] - delta)
```

mlp optimizer

In [4]:

```
class MLP(object):
    def init (self, name, size input, size hidden, size output, learning rate=
0.01, weight coeff=1,\
                  Reg=None, RegC=0, training=None, validation=None, accuracy=0, de
vice=None):
        self.name
                           = name
        self.size input
                           = size input
        self.size hidden
                           = size hidden
        self.size output
                           = size output
        self.Req
                           = Req
        self.ReqC
                           = ReqC
        self.training
                           = training
        self.validation
                          = validation
        self.accuracy
                           = accuracy
        self.device
                           = device
        self.learning rate = learning rate
        self.weight coeff = weight coeff
        self.W1 = self.initWeights(self.size input, self.size hidden[0], self.weig
ht coeff)
        self.b1 = self.initWeights(1, self.size hidden[0], self.weight coeff)
        self.W2 = self.initWeights(self.size hidden[0], self.size hidden[1], self.
weight coeff)
        self.b2 = self.initWeights(1, self.size hidden[1], self.weight coeff)
        self.W3 = self.initWeights(self.size_hidden[1], self.size_hidden[2], self.
weight coeff)
        self.b3 = self.initWeights(1, self.size hidden[2], self.weight coeff)
        self.W4 = self.initWeights(self.size hidden[2], self.size output, self.wei
ght_coeff)
        self.b4 = self.initWeights(1, self.size output, self.weight coeff)
        self.variables = [self.W1, self.b1, self.W2, self.b2, self.W3, self.b3, se
lf.W4, self.b41
        self.optimizer = myOptimizer(self.variables)
    def initWeights(self, rows, columns, multFactor=1):
        return tf.Variable(multFactor*tf.random.normal([rows, columns]))
    def forward(self, X):
        if self.device is not None:
            with tf.device('qpu:0' if self.device=='qpu' else 'cpu'):
                self.y = self.compute output(X)
        else:
            self.y = self.compute output(X)
        return self.y
```

```
def getRegLoss(self, X_train):
        if self.Req=='L2':
            return (self.RegC/X_train.shape[0])*(tf.reduce_sum(tf.math.square(self))
.variables[0])) +
                                                  tf.reduce sum(tf.math.square(self
.variables[2])) +
                                                  tf.reduce sum(tf.math.square(self
.variables[4])) +
                                                  tf.reduce sum(tf.math.square(self
.variables[6])))
        elif self.Reg=='L1':
            return (self.RegC/X train.shape[0])*tf.abs(tf.reduce sum(self.variable
s[0]) +
                                                        tf.reduce sum(self.variable
s[2]) +
                                                        tf.reduce sum(self.variable
s[4]) +
                                                        tf.reduce sum(self.variable
s[6]))
        elif self.Reg=='L1+L2':
            L2 = (self.RegC/X train.shape[0])*(tf.reduce sum(tf.math.square(self.
variables[0])) +
                                                 tf.reduce sum(tf.math.square(self.
variables[2])) +
                                                 tf.reduce sum(tf.math.square(self.
variables[4])) +
                                                 tf.reduce_sum(tf.math.square(self.
variables[6])))
            L1 = (self.RegC/X train.shape[0])*tf.abs(tf.reduce sum(self.variables[
0]) +
                                                      tf.reduce_sum(self.variables[
21) +
                                                      tf.reduce sum(self.variables[
4]) +
                                                      tf.reduce sum(self.variables[
6]))
            return L1+L2
        else:
            return 0
    def loss(self, y pred, y true):
        y_true_tf = tf.cast(tf.reshape(y_true, (-1, self.size_output)), dtype=tf.f
loat32)
        y_pred_tf = tf.cast(y_pred, dtype=tf.float32)
        loss = tf.keras.losses.CategoricalCrossentropy()(y true tf, y pred tf)
        return loss
    def backward(self, X train, y train, t):
```

```
with tf.GradientTape() as tape:
        predicted = self.forward(X train)
        current loss = self.loss(predicted, y train)
        current_loss += self.getRegLoss(X_train)
   grads = tape.gradient(current_loss, self.variables)
    self.optimizer.updateWeights(t, grads, self.variables)
def compute output(self, X):
   X tf = tf.cast(X, dtype=tf.float32)
   w1Hat = tf.matmul(X tf, self.variables[0]) + self.variables[1]
    h1Hat = tf.nn.relu(w1Hat)
   w2Hat = tf.matmul(h1Hat, self.variables[2]) + self.variables[3]
   h2Hat = tf.nn.relu(w2Hat)
   w3Hat = tf.matmul(h2Hat, self.variables[4]) + self.variables[5]
   h3Hat = tf.nn.relu(w3Hat)
   w4Hat = tf.matmul(h3Hat, self.variables[6]) + self.variables[7]
    output = tf.nn.softmax(w4Hat)
    return output
def getAccuracy(self, predictions, outputs):
    preds = np.argmax(predictions, axis=1)
   y_true = np.argmax(outputs, axis=1)
    return (preds==y_true).mean()
```

In [5]:

```
def trainModel(model, data, NUM EPOCHS=10, batchSize=50, seedVal=1234):
    X_train = data['X_train']
    y_train = data['y_train']
            = data['X val']
    X val
            = data['y_val']
    y_val
    training = np.zeros(shape=(NUM EPOCHS, 3))
    validation = np.zeros(shape=(NUM EPOCHS, 3))
    train_ds = tf.data.Dataset.from_tensor_slices((X_train, y_train)).batch(batchS
ize)
    val ds
             = tf.data.Dataset.from_tensor_slices((X_val, y_val)).batch(batchSize)
    print(f'\n\n******************** Training model: {model.name} with seed: {seedVal}
*************\n')
    time start = time.time()
    for epoch in range(NUM EPOCHS):
        train loss = tf.zeros([1, 1], dtype=tf.float32)
                 = tf.zeros([1, 1], dtype=tf.float32)
        val loss
        train ds = tf.data.Dataset.from tensor slices((X train, y train)).shuffle(
25, seed = epoch*(seedVal)).batch(batchSize)
        val ds
               = tf.data.Dataset.from_tensor_slices((X_val, y_val)).shuffle(25,
seed = epoch*(seedVal)).batch(batchSize)
        t = 0
        for inputs, outputs in train ds:
            t+=1
            train pred = model.forward(inputs)
            train loss = train loss + model.loss(train pred, outputs)
            model.backward(inputs, outputs, t)
            train acc = model.getAccuracy(train pred, outputs)
        for inputs, outputs in val ds:
            val pred = model.forward(inputs)
            val_loss = val_loss + model.loss(val_pred, outputs)
            val acc = model.getAccuracy(val pred, outputs)
        # train_loss = np.array(train loss)
        # val loss = np.array(val loss)
        training[epoch] = [epoch+1, train acc, np.sum(train loss)/X train.shape[0
]]
        validation[epoch] = [epoch+1, val acc, np.sum(train loss)/X train.shape[0
]]
        print('# Epoch:={}/{} - train loss:={:.4f} - val loss:={:.4f}, train acc:
={:.2f} - val acc:={:.2f}'\
              .format(epoch+1, NUM_EPOCHS, np.sum(train_loss)/X_train.shape[0], np
.sum(val loss)/X val.shape[0], train acc, val acc))
    time taken = time.time()-time start
```

```
print(f'\nTotal time taken (in seconds): {time taken: .2f}')
   print(f'\nFinished training model: {model.name}\n')
   model.training = training
   model.validation = validation
def testModel(model, data):
   X test = data['X test']
   y_test = data['y_test']
   preds = model.forward(X test)
   pred = np.argmax(preds, axis=1)
   y true= np.argmax(y_test, axis=1)
   model.accuracy = (pred==y true).mean()*100
   print(f'************* Testing ****************)
   print(f'{model.name} model accuracy = {model.accuracy:.2f}%')
   def plotAccuracyAndLoss(model):
   training = model.training
   validation = model.validation
   fig, (ax1, ax2) = plt.subplots(1, 2)
   training[:, -1] = training[:, -1]/np.linalg.norm(training[:, -1])
   ax1.plot(training[:,0], training[:,1], 'g')
   ax1.plot(training[:,0], training[:,2], 'b')
   ax1.set title('Training')
   ax1.legend(["Accuracy", "Loss"])
   validation[:, -1] = validation[:, -1]/np.linalg.norm(validation[:, -1])
   ax2.plot(validation[:,0], validation[:,1], 'g')
   ax2.plot(validation[:,0], validation[:,2], 'b')
   ax2.set title('Validation')
   ax2.legend(["Accuracy", "Loss"])
   plt.show()
```

In [6]:

```
def main():
    for j in range(2):
        j=1
        if j==0:
            data = load data('mnist')
            size_hidden = [512, 256, 128]
            learning rate = 5e-4
            weight coeff = 0.1
        if j==1:
            print('fashion mnist is running')
            data = load data('fashion mnist')
            size hidden = [1024, 512, 256]
            learning rate = 5e-4
            weight coeff = 0.1
        imageSize = data['imageSize']
        size input = imageSize
        size output = 10
        allModels = {}
        # allModels['mlp on gpu default'] = {}
        allModels['mlp_on_gpu_RegL1'] = {}
        allModels['mlp on gpu RegL2']
                                      = {}
        for model name in allModels:
            model = allModels[model name]
            cnt = -1
            numEpochs = 10
            batchSize = 50
            numTrials = 10
            seeds = random.sample(range(1000, 9999), numTrials)
                           = np.zeros(shape=(numEpochs, 1))
            accuracy = np.zeros(shape=(numTrials, 1))
            for i in seeds:
                cnt += 1
                np.random.seed(i)
                tf.random.set_seed(i)
                print(f'Count: {cnt}, j=: {j}')
                if model_name == 'mlp_on_gpu_default':
                    model['name'] = MLP('mlp_on_gpu_default', size_input, size_hid
den, size output, learning rate, weight coeff,\
                                              device='qpu')
```

```
elif model_name == 'mlp_on_gpu_RegL1':
                    model['name'] = MLP('mlp_on_gpu_RegL1', size_input, size_hidde
n, size output, learning rate, weight coeff,\
                                             'L1', 1e-4, device='gpu')
                elif model_name == 'mlp_on_gpu_RegL2':
                    model['name'] = MLP('mlp_on_gpu_RegL2', size_input, size_hidde
n, size output, learning rate, weight coeff,\
                                             'L2', 1e-4, device='gpu')
                else:
                    pass
                trainModel(model['name'], data, numEpochs, batchSize, i)
                testModel(model['name'], data)
                accuracy[cnt] = model['name'].accuracy
                plotAccuracyAndLoss(model['name'])
                allModels[model name][i] = model['name']
                allModels[model name]['Accuracy'] = [np.mean(accuracy), np.var(acc
uracy)]
        if j==0:
            mnist = allModels
        elif j==1:
            fashion mnist = allModels
        else:
            pass
    return mnist, fashion_mnist
```

```
In [ ]:
```

```
if __name__ == "__main__":
    mnist = main()
```

```
fashion_mnist is running
```

Data Split:

X_train: (50000, 784), y_train: (50000, 10) X_test: (10000, 784), y_test: (10000, 10) X_val: (10000, 784), y_val: (10000, 10) Count: 0, j=: 1

2022-03-05 13:16:45.647164: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.651793: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.652128: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.654032: I tensorflow/core/platform/cpu_feature_guard.cc:151] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 FMA

To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.

2022-03-05 13:16:45.655331: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.655851: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.656156: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.982149: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.982461: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:936] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero

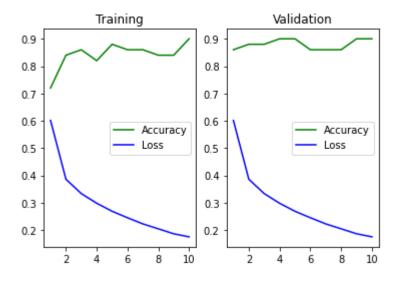
2022-03-05 13:16:45.982718: I tensorflow/stream_executor/cuda/cuda_gpu _executor.cc:936] successful NUMA node read from SysFS had negative va lue (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-03-05 13:16:45.982970: I tensorflow/core/common_runtime/gpu/gpu_d evice.cc:1525] Created device /job:localhost/replica:0/task:0/device:G PU:0 with 7157 MB memory: -> device: 0, name: NVIDIA GeForce GTX 107 0, pci bus id: 0000:01:00.0, compute capability: 6.1

```
********* Training model: mlp_on_gpu_RegL1 with seed: 7879 ****
********
```

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0105 - val loss:=0.0096, train acc:=0.7
2 - val acc:=0.86
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0085, train acc:=0.8
4 - val acc:=0.88
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0081, train acc:=0.8
6 - val acc:=0.88
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0072, train acc:=0.8
2 - val acc:=0.90
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0079, train acc:=0.8
8 - val acc:=0.90
# Epoch:=6/10 - train loss:=0.0043 - val loss:=0.0078, train acc:=0.8
6 - val acc:=0.86
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0085, train acc:=0.8
6 - val acc:=0.86
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0079, train acc:=0.8
4 - val acc:=0.86
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0080, train acc:=0.8
4 - val acc:=0.90
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0077, train acc:=0.
90 - val acc:=0.90
Total time taken (in seconds): 223.45
Finished training model: mlp on gpu RegL1
*********** Testing **********
mlp on gpu RegL1 model accuracy = 87.37%
************
```



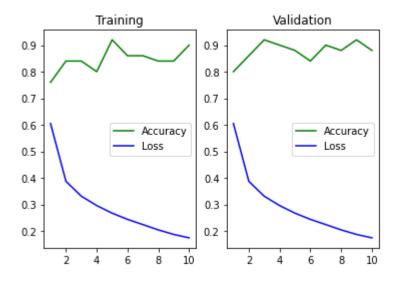
Count: 1, j=: 1

******** Training model: mlp_on_gpu_RegL1 with seed: 6511 ****

/home/skanda/Softwares/miniconda3/lib/python3.9/site-packages/numpy/co re/fromnumeric.py:1970: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tup les-or ndarrays with different lengths or shapes) is deprecated. If yo u meant to do this, you must specify 'dtype=object' when creating the ndarray.

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0106 - val loss:=0.0094, train acc:=0.7
6 - val acc:=0.80
# Epoch:=2/10 - train loss:=0.0068 - val loss:=0.0079, train acc:=0.8
4 - val acc:=0.86
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0075, train acc:=0.8
4 - val acc:=0.92
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0073, train acc:=0.8
0 - val acc:=0.90
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0069, train acc:=0.9
2 - val acc:=0.88
# Epoch:=6/10 - train loss:=0.0043 - val loss:=0.0077, train acc:=0.8
6 - val acc:=0.84
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0084, train acc:=0.8
6 - val acc:=0.90
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0081, train acc:=0.8
4 - val acc:=0.88
# Epoch:=9/10 - train loss:=0.0033 - val loss:=0.0089, train acc:=0.8
4 - val acc:=0.92
# Epoch:=10/10 - train loss:=0.0031 - val loss:=0.0087, train acc:=0.
90 - val acc:=0.88
Total time taken (in seconds): 215.51
Finished training model: mlp on gpu RegL1
********** Testing **********
mlp on gpu RegL1 model accuracy = 87.45%
```



Count: 2, j=: 1

******* Training model: mlp_on_gpu_RegL1 with seed: 9175 ****

/home/skanda/Softwares/miniconda3/lib/python3.9/site-packages/numpy/core/fromnumeric.py:1970: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray.

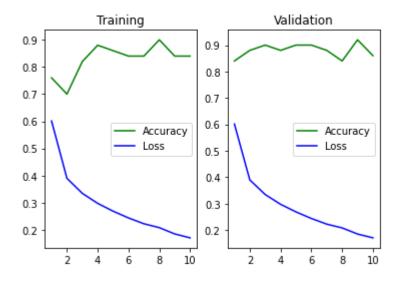
result = asarray(a).shape

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```
# Epoch:=1/10 - train loss:=0.0104 - val loss:=0.0090, train acc:=0.7
6 - val acc:=0.84
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0083, train acc:=0.7
0 - val acc:=0.88
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0075, train acc:=0.8
2 - val acc:=0.90
# Epoch:=4/10 - train loss:=0.0051 - val loss:=0.0079, train acc:=0.8
8 - val acc:=0.88
# Epoch:=5/10 - train loss:=0.0046 - val loss:=0.0080, train acc:=0.8
6 - val acc:=0.90
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0077, train acc:=0.8
4 - val acc:=0.90
# Epoch:=7/10 - train loss:=0.0038 - val loss:=0.0079, train acc:=0.8
4 - val acc:=0.88
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0078, train acc:=0.9
0 - val acc:=0.84
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0079, train acc:=0.8
4 - val acc:=0.92
# Epoch:=10/10 - train loss:=0.0029 - val loss:=0.0077, train acc:=0.
84 - val acc:=0.86
```

Total time taken (in seconds): 223.33

Finished training model: mlp_on_gpu_RegL1



Count: 3, j=: 1

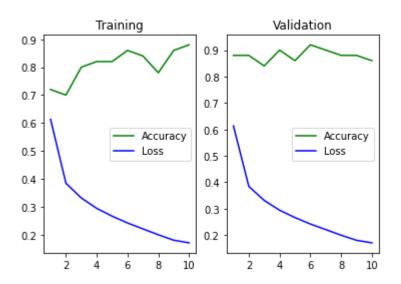
******** Training model: mlp_on_gpu_RegL1 with seed: 8033 ****

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0107 - val loss:=0.0089, train acc:=0.7
2 - val acc:=0.88
              - train loss:=0.0067 - val loss:=0.0084, train acc:=0.7
# Epoch:=2/10
0 - val acc:=0.88
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0081, train acc:=0.8
0 - val acc:=0.84
# Epoch:=4/10 - train loss:=0.0051 - val loss:=0.0073, train acc:=0.8
2 - val acc:=0.90
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0076, train acc:=0.8
2 - val acc:=0.86
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0071, train acc:=0.8
6 - val acc:=0.92
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0073, train acc:=0.8
4 - val acc:=0.90
# Epoch:=8/10 - train loss:=0.0035 - val loss:=0.0080, train acc:=0.7
8 - val acc:=0.88
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0080, train acc:=0.8
6 - val acc:=0.88
               - train loss:=0.0030 - val loss:=0.0075, train acc:=0.
# Epoch:=10/10
88 - val acc:=0.86
```

Total time taken (in seconds): 222.12

Finished training model: mlp on gpu RegL1

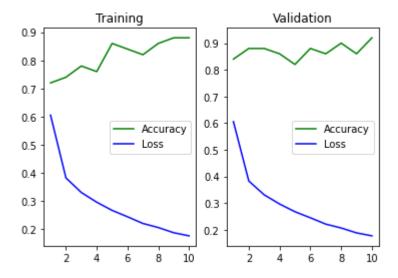


```
Count: 4, j=: 1
```

```
********* Training model: mlp_on_gpu_RegL1 with seed: 8650 ****
********
```

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0106 - val loss:=0.0095, train acc:=0.7
2 - val acc:=0.84
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0084, train acc:=0.7
4 - val acc:=0.88
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0077, train acc:=0.7
8 - val acc:=0.88
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0073, train acc:=0.7
6 - val acc:=0.86
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0074, train acc:=0.8
6 - val acc:=0.82
# Epoch:=6/10 - train loss:=0.0043 - val loss:=0.0073, train acc:=0.8
4 - val acc:=0.88
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0080, train acc:=0.8
2 - val acc:=0.86
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0079, train acc:=0.8
6 - val acc:=0.90
# Epoch:=9/10 - train loss:=0.0033 - val loss:=0.0077, train acc:=0.8
8 - val acc:=0.86
# Epoch:=10/10 - train loss:=0.0031 - val loss:=0.0074, train acc:=0.
88 - val acc:=0.92
Total time taken (in seconds): 216.44
Finished training model: mlp on gpu RegL1
*********** Testing **********
mlp on gpu RegL1 model accuracy = 87.84%
```



Count: 5, j=: 1

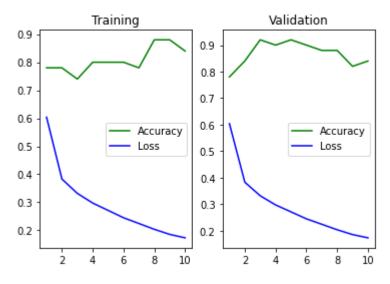
/home/skanda/Softwares/miniconda3/lib/python3.9/site-packages/numpy/core/fromnumeric.py:1970: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray.

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0106 - val loss:=0.0096, train acc:=0.7
8 - val acc:=0.78
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0080, train acc:=0.7
8 - val acc:=0.84
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0079, train acc:=0.7
4 - val acc:=0.92
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0072, train acc:=0.8
0 - val acc:=0.90
# Epoch:=5/10 - train loss:=0.0048 - val loss:=0.0075, train acc:=0.8
0 - val acc:=0.92
# Epoch:=6/10 - train loss:=0.0043 - val loss:=0.0073, train acc:=0.8
0 - val acc:=0.90
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0072, train acc:=0.7
8 - val acc:=0.88
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0075, train acc:=0.8
8 - val acc:=0.88
# Epoch:=9/10 - train loss:=0.0033 - val loss:=0.0079, train acc:=0.8
8 - val acc:=0.82
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0080, train acc:=0.
84 - val acc:=0.84
```

Total time taken (in seconds): 218.29

Finished training model: mlp_on_gpu_RegL1



Count: 6, j=: 1

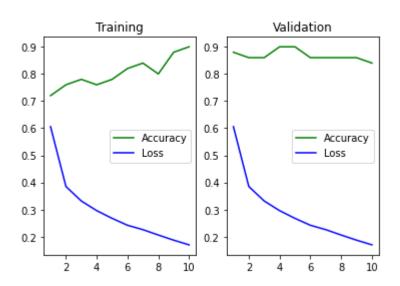
********** Training model: mlp_on_gpu_RegL1 with seed: 6365 ****

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0106 - val loss:=0.0094, train acc:=0.7
2 - val acc:=0.88
# Epoch:=2/10 - train loss:=0.0068 - val loss:=0.0090, train acc:=0.7
6 - val acc:=0.86
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0079, train acc:=0.7
8 - val acc:=0.86
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0074, train acc:=0.7
6 - val acc:=0.90
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0071, train acc:=0.7
8 - val acc:=0.90
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0079, train acc:=0.8
2 - val acc:=0.86
# Epoch:=7/10 - train loss:=0.0040 - val loss:=0.0083, train acc:=0.8
4 - val acc:=0.86
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0090, train acc:=0.8
0 - val acc:=0.86
# Epoch:=9/10 - train loss:=0.0033 - val loss:=0.0093, train acc:=0.8
8 - val acc:=0.86
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0104, train acc:=0.
90 - val acc:=0.84
```

Total time taken (in seconds): 218.36

Finished training model: mlp on gpu RegL1

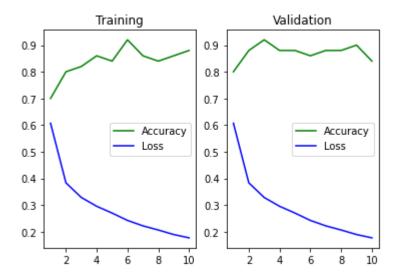


```
Count: 7, j=: 1
```

```
******** Training model: mlp_on_gpu_RegL1 with seed: 4266 ****
********
```

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0107 - val loss:=0.0099, train acc:=0.7
0 - val acc:=0.80
# Epoch:=2/10 - train loss:=0.0068 - val loss:=0.0083, train acc:=0.8
0 - val acc:=0.88
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0079, train acc:=0.8
2 - val acc:=0.92
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0078, train acc:=0.8
6 - val acc:=0.88
# Epoch:=5/10 - train loss:=0.0048 - val loss:=0.0072, train acc:=0.8
4 - val acc:=0.88
# Epoch:=6/10 - train loss:=0.0043 - val loss:=0.0072, train acc:=0.9
2 - val acc:=0.86
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0076, train acc:=0.8
6 - val acc:=0.88
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0076, train acc:=0.8
4 - val acc:=0.88
# Epoch:=9/10 - train loss:=0.0033 - val loss:=0.0079, train acc:=0.8
6 - val acc:=0.90
# Epoch:=10/10 - train loss:=0.0031 - val loss:=0.0080, train acc:=0.
88 - val acc:=0.84
Total time taken (in seconds): 214.17
Finished training model: mlp_on_gpu_RegL1
*********** Testing **********
mlp on gpu RegL1 model accuracy = 87.72%
***********
```



Count: 8, j=: 1

******** Training model: mlp_on_gpu_RegL1 with seed: 1980 ****

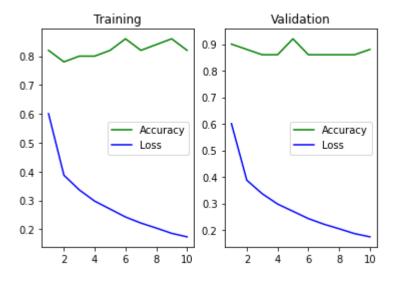
/home/skanda/Softwares/miniconda3/lib/python3.9/site-packages/numpy/co re/fromnumeric.py:1970: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tup les-or ndarrays with different lengths or shapes) is deprecated. If yo u meant to do this, you must specify 'dtype=object' when creating the ndarray.

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0104 - val loss:=0.0089, train acc:=0.8
2 - val acc:=0.90
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0086, train acc:=0.7
8 - val acc:=0.88
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0078, train acc:=0.8
0 - val acc:=0.86
# Epoch:=4/10 - train loss:=0.0051 - val loss:=0.0077, train acc:=0.8
0 - val acc:=0.86
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0076, train acc:=0.8
2 - val acc:=0.92
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0072, train acc:=0.8
6 - val acc:=0.86
# Epoch:=7/10 - train loss:=0.0038 - val loss:=0.0071, train acc:=0.8
2 - val acc:=0.86
# Epoch:=8/10 - train loss:=0.0035 - val loss:=0.0075, train acc:=0.8
4 - val acc:=0.86
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0080, train acc:=0.8
6 - val acc:=0.86
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0082, train acc:=0.
82 - val acc:=0.88
```

Total time taken (in seconds): 220.86

Finished training model: mlp_on_gpu_RegL1



Count: 9, j=: 1

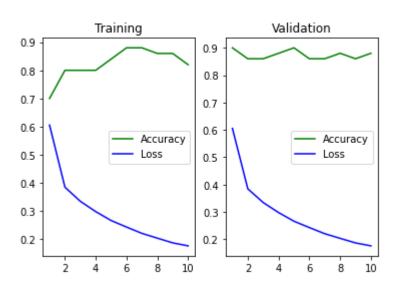
*********** Training model: mlp_on_gpu_RegL1 with seed: 6245 ****

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0104 - val loss:=0.0098, train acc:=0.7
0 - val acc:=0.90
              - train loss:=0.0066 - val loss:=0.0082, train acc:=0.8
# Epoch:=2/10
0 - val acc:=0.86
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0073, train acc:=0.8
0 - val acc:=0.86
# Epoch:=4/10 - train loss:=0.0051 - val loss:=0.0074, train acc:=0.8
0 - val acc:=0.88
# Epoch:=5/10 - train loss:=0.0046 - val loss:=0.0069, train acc:=0.8
4 - val acc:=0.90
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0074, train acc:=0.8
8 - val acc:=0.86
# Epoch:=7/10 - train loss:=0.0038 - val loss:=0.0079, train acc:=0.8
8 - val acc:=0.86
# Epoch:=8/10 - train loss:=0.0035 - val loss:=0.0083, train acc:=0.8
6 - val acc:=0.88
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0081, train acc:=0.8
6 - val acc:=0.86
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0077, train acc:=0.
82 - val acc:=0.88
```

Total time taken (in seconds): 215.41

Finished training model: mlp_on_gpu_RegL1



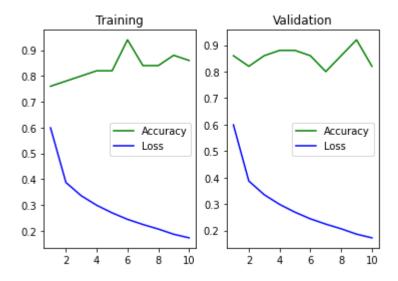
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```
Count: 0, j=: 1
```

```
******** Training model: mlp_on_gpu_RegL2 with seed: 9233 ****
********
```

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0104 - val loss:=0.0089, train acc:=0.7
6 - val acc:=0.86
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0084, train acc:=0.7
8 - val acc:=0.82
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0078, train acc:=0.8
0 - val acc:=0.86
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0075, train acc:=0.8
2 - val acc:=0.88
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0076, train acc:=0.8
2 - val acc:=0.88
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0080, train acc:=0.9
4 - val acc:=0.86
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0080, train acc:=0.8
4 - val acc:=0.80
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0079, train acc:=0.8
4 - val acc:=0.86
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0073, train acc:=0.8
8 - val acc:=0.92
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0077, train acc:=0.
86 - val acc:=0.82
Total time taken (in seconds): 217.69
Finished training model: mlp on gpu RegL2
*********** Testing **********
mlp on qpu RegL2 model accuracy = 87.87%
************
```



Count: 1, j=: 1

******** Training model: mlp_on_gpu_RegL2 with seed: 5617 ****

/home/skanda/Softwares/miniconda3/lib/python3.9/site-packages/numpy/core/fromnumeric.py:1970: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray.

result = asarray(a).shape

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```
# Epoch:=1/10 - train loss:=0.0109 - val loss:=0.0100, train acc:=0.7
0 - val acc:=0.92
# Epoch:=2/10 - train loss:=0.0068 - val loss:=0.0084, train acc:=0.6
8 - val acc:=0.86
# Epoch:=3/10 - train loss:=0.0059 - val loss:=0.0083, train acc:=0.7
2 - val acc:=0.84
# Epoch:=4/10 - train loss:=0.0053 - val loss:=0.0086, train acc:=0.7
6 - val acc:=0.80
# Epoch:=5/10 - train loss:=0.0047 - val loss:=0.0080, train acc:=0.8
6 - val acc:=0.90
# Epoch:=6/10 - train loss:=0.0043 - val loss:=0.0076, train acc:=0.8
6 - val acc:=0.88
# Epoch:=7/10 - train loss:=0.0039 - val loss:=0.0076, train acc:=0.8
6 - val acc:=0.90
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0080, train acc:=0.8
4 - val acc:=0.88
# Epoch:=9/10 - train loss:=0.0033 - val loss:=0.0084, train acc:=0.8
6 - val acc:=0.88
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0089, train acc:=0.
86 - val acc:=0.86
```

Total time taken (in seconds): 220.92

Finished training model: mlp_on_gpu_RegL2



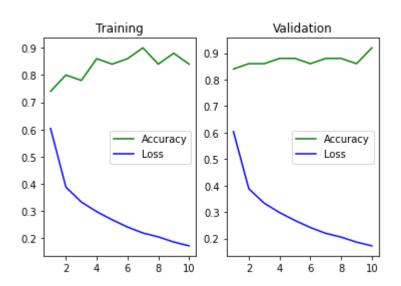
Count: 2, j=: 1

result = asarray(a).shape

```
- train loss:=0.0105 - val loss:=0.0093, train acc:=0.7
# Epoch:=1/10
4 - val acc:=0.84
              - train loss:=0.0067 - val loss:=0.0086, train acc:=0.8
# Epoch:=2/10
0 - val acc:=0.86
              - train loss:=0.0058 - val loss:=0.0074, train acc:=0.7
# Epoch:=3/10
8 - val acc:=0.86
# Epoch:=4/10 - train loss:=0.0052 - val loss:=0.0075, train acc:=0.8
6 - val acc:=0.88
# Epoch:=5/10 - train loss:=0.0046 - val loss:=0.0074, train acc:=0.8
4 - val acc:=0.88
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0074, train acc:=0.8
6 - val acc:=0.86
# Epoch:=7/10 - train loss:=0.0038 - val loss:=0.0080, train acc:=0.9
0 - val acc:=0.88
# Epoch:=8/10 - train loss:=0.0036 - val loss:=0.0077, train acc:=0.8
4 - val acc:=0.88
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0095, train acc:=0.8
8 - val acc:=0.86
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0078, train acc:=0.
84 - val acc:=0.92
```

Total time taken (in seconds): 218.62

Finished training model: mlp on gpu RegL2



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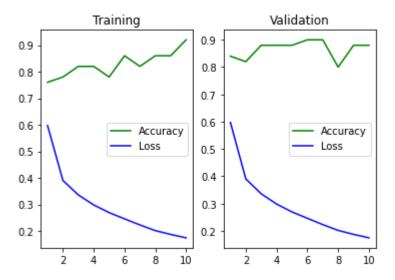
```
Count: 3, j=: 1
```

```
******** Training model: mlp_on_gpu_RegL2 with seed: 8668 ****
********
```

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0102 - val loss:=0.0092, train acc:=0.7
6 - val acc:=0.84
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0092, train acc:=0.7
8 - val acc:=0.82
# Epoch:=3/10 - train loss:=0.0057 - val loss:=0.0082, train acc:=0.8
2 - val acc:=0.88
# Epoch:=4/10 - train loss:=0.0051 - val loss:=0.0082, train acc:=0.8
2 - val acc:=0.88
# Epoch:=5/10 - train loss:=0.0046 - val loss:=0.0077, train acc:=0.7
8 - val acc:=0.88
# Epoch:=6/10 - train loss:=0.0042 - val loss:=0.0075, train acc:=0.8
6 - val acc:=0.90
# Epoch:=7/10 - train loss:=0.0038 - val loss:=0.0082, train acc:=0.8
2 - val acc:=0.90
# Epoch:=8/10 - train loss:=0.0034 - val loss:=0.0080, train acc:=0.8
6 - val acc:=0.80
# Epoch:=9/10 - train loss:=0.0032 - val loss:=0.0083, train acc:=0.8
6 - val acc:=0.88
# Epoch:=10/10 - train loss:=0.0030 - val loss:=0.0079, train acc:=0.
92 - val acc:=0.88
Total time taken (in seconds): 231.25
Finished training model: mlp on gpu RegL2
*********** Testing **********
mlp on qpu RegL2 model accuracy = 87.37%
**************
```

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Count: 4, j=: 1

********** Training model: mlp_on_gpu_RegL2 with seed: 3349 ****

/home/skanda/Softwares/miniconda3/lib/python3.9/site-packages/numpy/co re/fromnumeric.py:1970: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tup les-or ndarrays with different lengths or shapes) is deprecated. If yo u meant to do this, you must specify 'dtype=object' when creating the ndarray.

result = asarray(a).shape

```
# Epoch:=1/10 - train loss:=0.0104 - val loss:=0.0098, train acc:=0.7
4 - val acc:=0.90
# Epoch:=2/10 - train loss:=0.0067 - val loss:=0.0084, train acc:=0.8
2 - val acc:=0.90
# Epoch:=3/10 - train loss:=0.0058 - val loss:=0.0078, train acc:=0.8
2 - val acc:=0.88
# Epoch:=4/10 - train loss:=0.0051 - val loss:=0.0076, train acc:=0.8
4 - val acc:=0.94
```

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