# Lab 3

#### Gil Cohen

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### 1 Introduction

This data set consists of 50 samples from each of three species of Iris (Iris setosa, Iris virginica and Iris versicolor). Four features were measured from each sample: the length and the width of the sepals and petals, in centimeters.

# 2 Data Visualization and Preparation

```
!pip install git+https://github.com/williamedwardhahn/mpcr
from mpcr import *
drive.mount('/content/drive')
#reading the Iris.cvs dataset into a dataframe called dataset
dataset = pd.read_csv('/content/drive/My Drive/Iris.csv')
''' Plotting pairwise relationships between features as scatter plots
   and the marginal distributions of each feature as univariate plots
   along the diagonal (Figure 1),,,
g = sns.pairplot(dataset)
#Plotting 5 by 5 subplot grid for pairwise relationships (Figure 2)
g = sns. PairGrid (dataset, hue="Species")
#Setting the diagonal plots to be histograms
g = g.map_diag(plt.hist)
#Setting the non-diagonal plots to be scatterplots
g = g.map_offdiag(plt.scatter)
#adding a legend
g = g.add_legend()
```

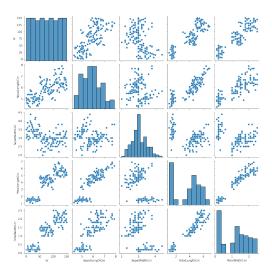


Figure 1: Pairwise relationships between features

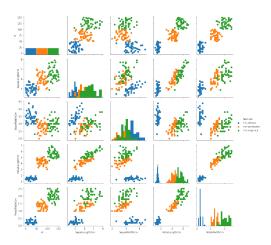


Figure 2: Pairwise relationships with legend

```
#converting input_data into an array input_data = np.array(input_data)
```

#converting input\_data into an array
input\_data = np.array(input\_data)

<sup>&#</sup>x27;'' creating a new dataframe from dataset called target\_data that contains only the 'Species' column''' target\_data = dataset [['Species']]

''' creating a new dataframe called input\_data from the dataset

dataframe by dropping the "Id" and 'Species' columns'' input\_data = dataset.drop(['Id', 'Species'], axis=1)

```
''', displaying the first 10 instances in the input_data array as
a colormap image','
plt.imshow(input_data[0:10,:])
''', Converting the categorical data 'Species' into a dataframe
of indicator variables ',','
target_data = pd.get_dummies(target_data.Species)
''', Changing target_data to be an array of the indices of the
columns of target_data at which each instance has a value of 1'''
_{-}, target_data = np.where(target_data==1)
#Random permutation of indices 0, 1, ..., 149 as an array
r = np.random.permutation(input_data.shape[0])
#Cutting the length of r by 20 percent and assigning the value to cut
cut = int(0.8*len(r))
input_data.shape
-> (150, 4)
#Setting X to be the input_data for a random selection of 120 instances
X = input_data[r[:cut],:]
#Setting X_test to be the input_data for the remaining 30 instances
X_{test} = input_{data}[r[cut:],:]
#Setting Y to be the target_data for the same instances used in X
Y = target_data[r[:cut]]
#Setting Y_test to be the target_data for the remaining 30 instances
Y_test = target_data[r[cut:]]
    Defining Functions
3
''', Defining a function called softmax that takes a tensor as input
and returns one containing column values that sum to 1 for each row','
def softmax(x):
    s1 = torch.exp(x - torch.max(x,1)[0][:,None])
    s = s1 / s1.sum(1)[:,None]
    return s
#Defining a cross entropy loss function
def cross_entropy(outputs, labels):
```

labels.long()])/outputs.size()[0]

return -torch.sum(softmax(outputs).log()[range(outputs.size()[0]),

```
def randn_trunc(s): #Truncated Normal Random Numbers
    mu = 0
    sigma = 0.1
    R = stats.truncnorm((-2*sigma - mu) / sigma, (2*sigma - mu) / sigma,
    loc=mu, scale=sigma)
    return R.rvs(s)
def acc(out,y):
    with torch.no_grad():
         return (\text{torch.sum}(\text{torch.max}(\text{out}, 1)[1] = \text{y}).\text{item}())/\text{y.shape}[0]
#Defining functions that each construct a tensor from "data"
def GPU(data):
    return torch.tensor(data, requires_grad=True, dtype=torch.float,
    device=torch.device('cuda'))
def GPU_data(data):
    return torch.tensor(data, requires_grad=False, dtype=torch.float,
    device=torch.device('cuda'))
'''defining a function that returns a batch of size b for the
input_data and target_data ','
def get_batch (mode):
               #size of batch
    b = c.b
    if mode == "train":
         r = np.random.randint(X.shape[0]-b)
         x = X[r:r+b,:] #getting batch from training input_data
        v = Y[r:r+b]
    elif mode == "test":
         r = np.random.randint(X_test.shape[0] - b)
        x = X_{test}[r:r+b,:] #getting batch from test input_data
         y = Y_test[r:r+b]
    return x,y
#defining a function that performs a gradient step on w
def gradient_step(w):
    for j in range (len (w)):
             w[j] \cdot data = w[j] \cdot data - c \cdot h*w[j] \cdot grad \cdot data
             w[j].grad.data.zero_()
```

```
model over its training data and a batch of its test data','
def make_plots():
    acc_train = acc(model(x, w), y)
    xt, yt = get_batch('test')
    acc_test = acc(model(xt, w), yt)
    wb.log({"acc_train": acc_train, "acc_test": acc_test})
''' calling GPU_data function on the input data and target data
for the training and testing instances to form tensors','
X = GPU_{-}data(X)
Y = GPU_data(Y)
X_test = GPU_data(X_test)
Y_test = GPU_data(Y_test)
''', Defining a function that returns the input value if its greater
than 0, and returns 0 otherwise ''
def relu(x):
    return x * (x > 0)
def model(x,w):
    for j in range (len (w)):
        ''', re-assign x to be the matrix product x * w[j] if
        its positive and 0 otherwise, ','
        x = relu(matmul(x, w[j]))
    return x
```

## 4 Training and Testing

```
wb.init(project="Iris");
c = wb.config

c.h = 0.05  #increment value for gradient step
c.b = 20  #size of batch
c.layers = 3  #number of layers for w
c.epochs = 2500  #number of epochs for training

c.f_n = [4,16,16,3]  #initial values for randn_trunc

#initial weights
w = [GPU(randn_trunc((c.f_n[i],c.f_n[i+1]))) for i in range(c.layers)]
```

```
for i in range (c.epochs):
    x, y = get_batch('train')
                               #getting a batch of training instances
    loss = cross\_entropy(softmax(model(x,w)),y)
    loss.backward()
    gradient_step(w)
                       #performing a weight update
    if (i+1) \% 1 = 0:
        make_plots()
acc(model(X,w),Y) #accuracy of model over the training data
-> 0.675
acc(model(X_test,w), Y_test) #accuracy of model over the test data
X[0] #first instance in training data
-> tensor([5.6000, 3.0000, 4.1000, 1.3000], device='cuda:0')
model(X[0],w) #output of the model for the first instance
\rightarrow tensor([-0., -0., -0.], device='cuda:0', grad_fn=<MulBackward0>)
#returns index of maximum value of model output for the first instance
torch.argmax(model(X[0],w))
#Displaying the three tensors in w as color images
for i in range (len (w)):
    plt.imshow(w[i].cpu().detach().numpy())
    plt.show()
```

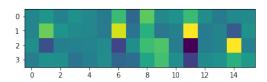


Figure 3: w[0]

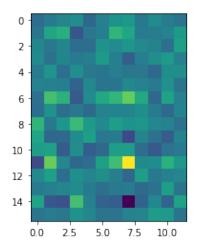


Figure 4: w[1]

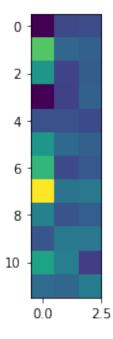


Figure 5: w[2]