Lab 3

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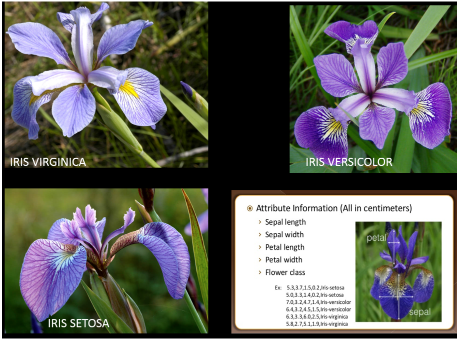
This data set consists of 50 samples from each of three species of Iris (Iris setosa, Iris virginica and Iris versicolor). Four features were measured from each sample: the length and the width of the sepals and petals, in centimeters.



Data Visualization and Preparation

```
#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'''
g = sns.pairplot(dataset)
```

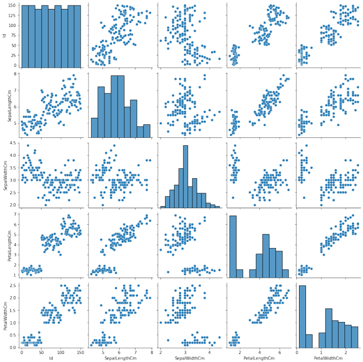


Figure: Pairwise relationships between features

Data Visualization and Preparation

```
#Plotting 5 by 5 subplot grid for pairwise relationships
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()
```

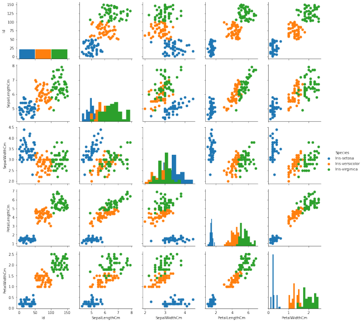


Figure: Pairwise relationships with legend

Data Visualization and Preparation

```
'''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)

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

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input_data = np.array(input_data)

'''displaying the first 10 instances in the input_data array as
a colormap image'''
plt.imshow(input_data[0:10,:])
```

Data Visualization and Preparation

```
'''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

```
'''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):
    return -torch.sum(softmax(outputs).log()[range(outputs.size()[0]),
    labels.long()])/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 (torch.sum(torch.max(out,1)[1] == y).item())/y.shape[0]
```

Defining Functions

```
#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):
    b = c.b  #size of batch
    if mode == "train":
        r = np.random.randint(X.shape[0]-b)
        x = X[r:r+b,:] #getting batch from training input_data
        y = 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].data = w[j].data - c.h*w[j].grad.data
            w[j].grad.data.zero_()
```

Defining Functions

```
'''Defining a function that makes plots for the accuracy of a
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)

#Function that returns the input value if its greater than 0, and 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
```

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()
```

Training and Testing

```
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
−> 0.5666666666666667

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
−> 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()
```

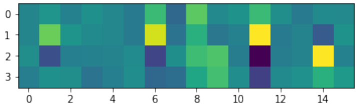

Figure: w[0]

Training and Testing

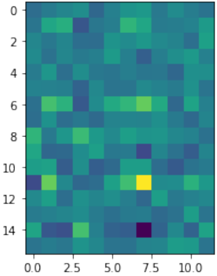

Figure: w[1]

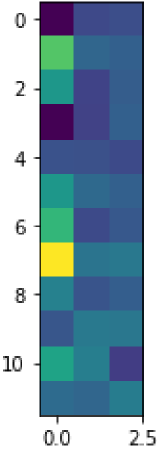

Figure: w[2]