# **PROJECT 1**

```
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```

## Annex C

#### **Prediction Function for CNN**

### Importing useful Libraries

```
import cv2 as cv
import matplotlib.pyplot as plt
import numpy as np
```

### Loading the trained parameters and reshaping them as per model.

In this model, there are 08 parameters:

- 1) f1- filters of first layer (10 filters)
- 2) f2- filters of 2nd layer (10 filters)
- 3) w3- weight matrix in fc layer
- 4) w4- weight matrix in fc layer
- 5) b1- bias terms for first convolution
- 6) b2- bias terms for first convolution
- 7) b3- bias for first hidden layer
- 8) b4- bias for 2nd hidden layer

```
In [24]:
    f1 = np.loadtxt("filter1.txt").reshape(10, 1,5,5)
    f2 = np.loadtxt("filter2.txt").reshape(10, 10,5,5)
    w3 = np.loadtxt("weight3.txt").reshape(96,4410)
    w4 = np.loadtxt("weight4.txt").reshape(8,96)
    b1 = np.loadtxt("bias1.txt").reshape(10, 1)
    b2 = np.loadtxt("bias2.txt").reshape(10, 1)
    b3 = np.loadtxt("bias3.txt").reshape(96,1)
    b4 = np.loadtxt("bias4.txt").reshape(8,1)
```

# Checking shapes of parameters

```
In [25]:
    print("f1=", f1.shape)
    print("f2=", f2.shape)
    print("w3=", w3.shape)
    print("w4=", w4.shape)
    print("b1=", b1.shape)
    print("b2=", b2.shape)
    print("b3=", b3.shape)
    print("b4=", b4.shape)

f1= (10, 1, 5, 5)
    f2= (10, 10, 5, 5)
    w3= (96, 4410)
    w4= (8, 96)
    b1= (10, 1)
```

```
b2= (10, 1)
b3= (96, 1)
b4= (8, 1)
```

#### Creating a list of paramters

```
In [26]: parameters= [f1, f2, w3, w4, b1, b2, b3, b4]
```

#### **Prediction Function**

```
In [27]:
          def Prediction_Function(image, label, parameters, conv_stride, pool_filter, pool_stride):
              first convolution = scratch convolution(image, f1, b1, conv stride)
              first convolution[first convolution<=0] = 0
              pooled1 = scratch_maxpool(first_convolution, pool_filter, pool_stride)
              second_convolution = scratch_convolution(first_convolution, f2, b2, conv_stride)
              second_convolution[second_convolution<=0] = 0</pre>
              pooled2 = scratch_maxpool(second_convolution, pool_filter, pool_stride)
              (nf2, dim2, _) = pooled2.shape
              fully_connected = pooled2.reshape((nf2 * dim2 * dim2, 1))
              z = w3.dot(fully\_connected) + b3
              z[z \le 0] = 0
              output = w4.dot(z) + b4
              probs = scratch_softmax(output)
              loss = scratch_CrossEntropyLoss(probs, label)
              return probs, np.argmax(probs), loss
```

## Defining functions needed for CNN prediction function.

```
In [28]:
          def scratch_convolution(image, filter, bias, stride=1):
              (no_of_filters, number_of_channels_f, f, _) = filter.shape
              number_of_channels, image_dim, _ = image.shape
              out_dim = int((image_dim - f)/stride)+1
              assert number of channels == number of channels f
              convolved = np.zeros((no_of_filters,out_dim,out_dim))
              #moving window over image
              for curr_filter in range(no_of_filters):
                   curr_y = out_y = 0
                   while curr_y + f <= image_dim:</pre>
                       curr_x = out_x = 0
                       while curr_x + f <= image_dim:</pre>
                           convolved[curr_filter, out_y, out_x] = np.sum(filter[curr_filter] * image
                           curr_x += stride
                           out_x += 1
                       curr_y += stride
                       out y += 1
               return convolved
```

```
In [29]:
          def scratch_maxpool(image, filter=2, stride=2):
               number_of_channels, h_prev, w_prev = image.shape
              height = int((h_prev - filter)/stride)+1
              width = int((w_prev - filter)/stride)+1
              downsampled = np.zeros((number_of_channels, height, width))
               #moving window over image
               for i in range(number_of_channels):
                   curr_y = out_y = 0
                   while curr_y + filter <= h_prev:</pre>
                       curr_x = out_x = 0
                       while curr_x + filter <= w_prev:</pre>
                           downsampled[i, out_y, out_x] = np.max(image[i, curr_y:curr_y+filter, curr]
                           curr x += stride
                           out_x += 1
                       curr_y += stride
```

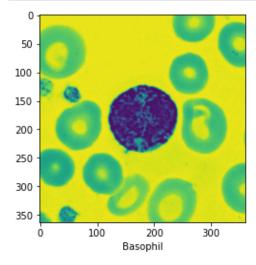
```
out_y += 1
return downsampled

In [30]:

def scratch_softmax(scores):
    out = np.exp(scores)
    return out/np.sum(out)
def scratch_CrossEntropyLoss(probs, label):
    return -np.sum(label * np.log(probs))
```

# Loading a test image from given dataset

```
image= cv.imread("BA_47.jpg",0)
plt.imshow(image)
plt.xlabel("Basophil")
image= cv.resize(image,(50,50),interpolation = cv.INTER_AREA)
image=np.reshape(image, (1,50,50))
```



#### Finding loss and probability using trained parameters

```
In [53]:
          Probabilities, Label, Loss = Prediction Function(image, label=6, parameters = parameters, co
In [54]:
          print("Probabilities:",Probabilities)
          print("Predicted Label:",Label)
          print("Loss:",Loss)
          Probabilities: [[9.61201676e-07]
           [8.20710788e-07]
           [8.76255410e-08]
           [9.15658616e-07]
           [1.20106748e-07]
           [5.91051855e-07]
           [9.36971976e-01]
           [6.30245274e-02]]
          Predicted Label: 6
          Loss: 546.7654344830379
 In [ ]:
 In [ ]:
```