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## CS771A : Assignment 2

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E-bot

**Submitted to:**  
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# 1 Task- 1

## 2 1.1 Identifying Foreground and Background pixels

3 In the given dataset of the images we have to differentiate between  
4 the background and the character inside the images. As mentioned  
5 in the problem statement background pixels has a light shade  
6 whereas the obfuscating lines has a darker shade and the borders  
7 of the characters are even darker.

8 To differentiate the character and the background of the image on  
9 the basis of the brightness, we used HSV color format. To know the  
10 average values of the background pixels, we took last column of the  
11 image as it mostly contains background pixels. Now we have the  
12 avg value of brightness( $v$ ) of the background pixels. We used it to  
13 create a mask on character outlines. We selected the pixels which  
14 have brightness value less than  $0.9*v$ . We are taking 0.9 times  $v$  to  
15 make sure that all background segments will be excluded.

16 We are considering the fillings of character outlines also as  
17 background. Now, we applied the mask on the image, which gives  
18 us only character outlines and obfuscating lines.

## 19 1.2 Dealing with obfuscating lines

20 Removing the obfuscating lines becomes easy using a image  
21 processing technique called **Erosion**. We simply converted the  
22 masked image into gray scale. Then using a small threshold value  
23 converted it to binary images. Then applied the erosion technique  
24 using a kernel size of (3X3). Larger kernel size might erase  
25 character outline itself.

## 26 1.3 Segmenting image into pieces

27 As mentioned in the problem statement we are only interested in  
28 the last character of the image. First we took the last section  
29 (350X500) of the whole image. Than we trimmed the image into  
30 pieces by looking for vertical columns of pixels that contain very  
31 few non-background pixels.

## 32 **1.4 Learning ML model**

### 33 **1.4.1 Model used**

34 The model used is a logistic regression model, a linear classification  
35 algorithm. Here we have two classes 'EVEN' and 'ODD'. We tried a  
36 bunch of models like KNN, SVM, Neural Networks etc. all models  
37 performed perfectly while the prediction (accuracy/ parity match  
38 score = 1.0). We choose logistic regression because it is simple,  
39 fast and taking very less space to store the trained model.

### 40 **1.4.2 Pre-processing of the data**

41 We have data as images of size 500X100 pixels. We already  
42 trimmed the images such that it only contains last digit (minimum  
43 blank space) of the hexadecimal number. The final size of the  
44 segmented image comes out 80X80, i.e. we have 6400 features for  
45 a ML model.

46 Now, the dimensions of our dataset is 3 (2000X80X80, here we  
47 are using binary image which has only one channel). But most of  
48 ML models work on datasets with less than or equal to 2  
49 dimensions. So we flatten each image into linear array. Now,  
50 our dataset is size of 2000X6400, 2 dimensions.

### 51 **1.4.3 Training algorithm and hyper-parameters**

52 We used default training algorithm and hyper-parameters from the  
53 sk-learn library for the logistic regression model.

### 54 **1.4.4 validation procedure**

55 We performed a train-test split of the data into training and testing  
56 sets, 30% of the data was used for testing, while the remaining 70%  
57 was used for training. We also shuffled the data before splitting as  
58 it is a common practice to ensure randomness in the data split.