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## LIST OF PUBLICATIONS

### **Papers published in International Conference Proceedings**

1. Supervised Learning Approach for Tamil Writer Identity Prediction Using Global and Local Features, *International Conference on IT and Intelligent Systems (ICITIS'2013)*, *International Conference Proceedings of PSRC, Penang (Malaysia)*, August 28-29, 2013
2. Analysis of Tamil Character Writings and Identification of Writer Using Support Vector Machine, *IEEE International Conference on Advanced Communication Control and Computing Technologies (ICACCCT)*, 2014

### **Papers published/accepted in International Journals**

1. Discovering Tamil Writer Identity Using Global and Local Features of Offline Handwritten Text, *International Review on Computers and Software (IRECOS)*, Italy, Vol. 8 No 9, Page 2080 – 2087, 2013
2. Prediction of Writer Using Tamil Handwritten Document Image Based on Pooled Features, *World Academy of Science, Engineering and Technology, International science index*, Vol. 9, No. 6, Page 1481 – 1487, 2015
3. Detection of a Person Using Descriptive Features of Tamil Handwriting and Pattern Learning, *International Journal of Applied Engineering Research*, ISSN 0973-4562, Volume 10, Number 21 (2015) pp 41902-41909, **(Scopus Indexed)**
4. Bayesian Linear Regression Co-efficients for SVM linear kernel to identify writers, *Journal of Advanced Research in Dynamical and Control Systems (JARDCS)*, Vol. 10, 01-Special Issue, 2018, **(Scopus Indexed)**
5. Hybrid Linear Kernel With PCA In SVM Prediction Model of Tamil Writing Pattern, *International Journal of Simulation Systems, Science & Technology*, **(Accepted - Scopus Indexed)**

### **Papers in review - International Journals**

1. Linear Kernel with Weighted Least Square Regression Co-efficient for SVM Based Tamil Writer Identification, *Malaysian Journal of Computer Science (MJCS)*
2. An Efficient Convolutional Neural Network Based Classifier to Predict Tamil Writer, *International Journal of Technology Intelligence and planning*

## Appendix – A

### Sample Tamil Handwritings (Paragraphs)

நீங்கள் எனின் நினைகிறீர்களோ அது நடந்திருந்து  
எனிலே நிலைம். திடீவும்பாலோ குறிச்சால்தான் அதைப்  
நற்றிய உங்கள் ஏண்ணா அல்லது முயிந்து பொருமல்  
அழிந்து வொண்டிட இருக்கிறால், உங்கள் ஏண்ணம்  
நினைவேற்றும்.

நிலைமால்லே நினைப்பு பண்சிரூபிக்கை எமீடு  
ஒஞ்சாக்கிறிரு எணிவதந்து நன் சொந்த அனுமதாவின்றி  
நிலைம் கண்டு கொண்டிவரி பான்ஸி. திலை இந்த அளவுப்பில்  
சுத்திரதை கண்டிஷுநி நண் திடுப்பதற்கிணி கண்டுமிகுநீஷல்கை  
பழப்பிடியான இந்த இருக்கியம் அவருக்கு வந்திய  
வந்து சில அடுதான்றந்து பூன் நமது சமீபதும் கிழ  
பணம் சில செய்தும் எனில் சீதை புற்றிய அர்த்தகைய  
தின்டும் இருக்கு நடவு நீநம் போடுமீ பழக்க சூண்டும்  
ஏற்கென்று பணம் கைக்கு உந்தாயிற்று ஏங்கநாமம்  
பாவிந்து ஒன்று சூண்டும்.  
ஏந்து பொருத்தை அதைய முடிவு இருக்கிறால்  
நமது செயல். நினைவு ஆகிய நினைப்பினால் அமைத்தும்  
அதைப் பற்றியநாக்கு இந்த சூண்டும்.

## Appendix – B

### Sample Datasets

#### Normalized dataset for character level writer identification

Character.csv - Microsoft Excel

Character

Ready

ENG Monday 3:57 PM 03-Sep-18

#### Normalized dataset for word level writer identification

word.csv - Microsoft Excel (Product Activation Failed)

word

Ready

ENG Tuesday 10:44 PM 5/10/2016

## Normalized dataset for paragraph level writer identification

The screenshot shows a Microsoft Excel spreadsheet titled "paragraph.csv". The data consists of 22 rows and 20 columns. The columns are labeled A through T and U. The data appears to be numerical values, likely representing writer identification features. The Excel ribbon is visible at the top, and the status bar at the bottom shows the time as 12:18:05 AM.

## Normalized dataset for Paragraph level writer identification – 422 features

The screenshot shows a Microsoft Excel spreadsheet titled "paragraph.csv". The data consists of 422 rows and 20 columns, starting from row PE126. The columns are labeled PE126 through PG. The data appears to be numerical values, likely representing writer identification features. The Excel ribbon is visible at the top, and the status bar at the bottom shows the identifier as "00420:2".

## Appendix - C

### Sample Code

#### Pre-processing – Character and Word Text Images

```
%reading the image
i=imread('D:\PhD\matlab file\DATASET\K001.jpg');
%converting rgb image into gray scale
c=rgb2gray(i);
%imshow(c);title('GrayScale image');
%Global thresholding the gray level image
level=graythresh(c);
%convert gray scale image to binary
bw=im2bw(c,level);
%thining
%p1=imcomplement(bw);
%new=bwmorph(p1,'thin',1);
%g=imcomplement(new);
%Removal of noise
n=medfilt2(c);
%edge detection using sobel filters
so=edge(c,'sobel');
%dilation of the image
se = strel('line',1,1);
di = imdilate(bw,se);
%calling function "No of black pixel"
[m]=myennoblkSize(so);
%n=sprintf('No of black pixel: %d',m);
%disp(n);
disp(m);
%calling Function "length"
[len]=myenlength(so);
%disp(len)
%calling fuction "edge_x"
[vertical_edges] = edge_x(so);
%disp(so);
%calling function "edge hinge"
[eh] = edgehinge(so, 4);
disp(eh);
%label conn comp
l = bwlabel(so,8);
disp(l);
%number of rows & col
[r c]=size(so);
disp(r); disp(c);
```

```
%skew angle, slope angle etc...
word = preProcessing(i);
disp(word);
subplot(2,3,1);
imshow(i); title('Original image');
subplot(2,3,2);
imshow(bw);title('Binary image');
subplot(2,3,3);
imshow(n);title('Noiseless image');
subplot(2,3,4);
imshow(so);title('Edge Detection');
subplot(2,3,5);
imshow(di);title('Dilation');
```

### **Pre-processing – Paragraph Text Images**

```
clc;
clear all;
close all;
warning off all;
[file,path]=uigetfile('*.jpg','Select an image');
i=imread(strcat(path,file));
figure('name','Original image','numbertitle','off');
imshow(i);
imm=rgb2gray(i);
i1 = rgb2gray(i);
i2 = edge(i1,'canny',0.1);
figure('name','Closed image ','numbertitle','off');
%figure('name','Edge image ','numbertitle','off');
imshow(i2);
% trthh
se = strel('square',2);
i3 = imdilate(i2,se);
figure('name','Filtered image ','numbertitle','off');
%figure('name','Dilate image ','numbertitle','off');
imshow(i3);
i4 = imfill(i3,'holes');
%imshow(i3);
s1 = regionprops(i4, 'BoundingBox');
i5=i4;
figure;imshow(i4);
for i=1:size(i4,1)
    for j=1:size(i4,2)
        if i4(i,j)==1
            for i1=1:5
                i5(i,j+i1)=1;
            end
        end
    end
end
```

```

    end
end
figure;imshow(i4);title('Box Bounding Image');
hold on;
s = regionprops(i5, 'BoundingBox');
for i=1:size(s,1)
    t1=s(i).BoundingBox;
    rectangle('Position',round( s(i).BoundingBox),'LineWidth',2,'Edgecolor','b');
    hold on;
    pause(0.1)
end
subplot(2,3,1);
imshow(i),title('Original image');
subplot(2,3,2);
imshow(i3); title('Closed image');
subplot(2,3,3);
imshow(i2);title('Filtered image');
subplot(2,3,4);
imshow(i4);title('Box Bounding Image');
subplot(2,3,5);
imshow(i5);title('Final normalised Text lines image');

```

### **Feature Extraction – Character and Word Text Images**

```

myFolder = 'D:\PhD\matlab file\DATASET';
jpegFiles = dir('D:\PhD\matlab file\DATASET\*.jpg');
for k = 1:length(jpegFiles)
    baseFileName = jpegFiles(k).name;
    fullFileName = fullfile(myFolder, baseFileName);
    i= imread(fullFileName);
    %pause(0.5);
    %imshow(i); % Display image.
    %drawnow; % Force display to update immediately.
    %converting rgb image into gray scale
    c=rgb2gray(i);
    %imshow(c);title('GrayScale image');
    %Removal of noise
    n=medfilt2(c);
    %Global thresholding the gray level image
    level=graythresh(n);
    %convert gray scale image to binary
    bw=im2bw(n,level);
    %thining
    p1=imcomplement(bw);
    new=bwmorph(p1,'thin',1);
    thi=imcomplement(new);

```

```

%edge detection using sobel filters
so=edge(c,'sobel');
%dilation of the image
se = strel('line',1,1);
di = imdilate(bw,se);
%Length Of an image
[r c]= size(so);
len=0;
l1=c;
l2=0;
for j=1:c
    for k=1:r
        if so(k,j)==1
            if(k<l1)
                l1=k;
            end
        end
        %if b(j,k)==1 && b(j,k+1)==0
        %  l2=k;
        %end
    end
end
for j=c:1
    for k=r:1
        if so(k,j)==1
            if(k>l2)
                l2=k;
            end
        end
        %if b(j,k)==1 && b(j,k+1)==0
        %  l2=k;
        %end
    end
end
len=l1-l2;
%disp(len);
%Height of an image
[r c]= size(n);
hei=0;
l1=c;
l2=0;
for j=1:r
    for k=1:c
        if so(j,k)==1
            if(k<l1)

```

```

l1=k;
end
end
%if b(j,k)==1 && b(j,k+1)==0
% l2=k;
%end
end
end
for j=r:1
for k=c:1
if so(j,k)==1
if(k>l2)
l2=k;
end
end
%if b(j,k)==1 && b(j,k+1)==0
% l2=k;
%end
end
hei=l1-l2;
%disp(hei);
%area
a=len*hei;
%disp(a);
%number of black pixel
[m]=myennoblkSize(so);
%str=sprintf("\nNumber Of black Pixel: %d",m);
%disp(m);
%entropy of gray values
ent= entropy(so);
%str=sprintf("\tEntropy of gray value: %f ",ent);
%disp(ent);
%moment invaraint
%[M]=feature_vec(so);
%mome=sprintf("\n\tMoment invariant: %f",M);
%disp(mome);
[M1 M2 M3 M4 M5 M6 M7]=feature_vec(so);
%disp(M1);disp(M2); disp(M3);disp(M4);disp(M5);disp(M6);disp(M7);
%loops
[loops, pixels] = getLoops(so);
%disp(loops);
%junction
[junctions, pixels] = getJunctions(thi);
%disp(junctions);
%aspect ratio

```

```

ap=c/r;
%disp(ap);
lhbemv=sprintf('%d %d %d %d %d %8.4f %8.4f %8.4f %8.4f %8.4f %8.4f
%8.4f',len,hei,a,loops,junctions,ap,ent,M1,M2,M3,M4,M5,M6,M7);
%disp(lhbemv);
%slant image,slope etc..
%[word] = preProcessing(i);
%disp(word);
%edge hinge
eh = edgehinge(so, 4);
%disp(eh);
%to write into notepad
%sm=0;cl=4;
%fileid=fopen('Z:\mphil\matlab file\Datasets\trail.dat','a');
%fprintf(fileid,'%d %d:%d %d:%d %d:%d %d:%d %d:%d.%0.4f %d:%0.4f %d:%0.4f
%d:%0.4f %d:%0.4f %d:%0.4f %d:%0.4f %d:%0.4f\r\n',cl,sm+1,len, sm+2,hei,
sm+3,a, sm+4,loops, sm+5, junctions, sm+6, ap, sm+7,ent, sm+8,M1, sm+9,M2, sm+10,M3,
sm+11,M4, sm+12,M5, sm+13,M6, sm+14,M7);
fclose(fileid);
axa=[num2str(len) '' num2str(hei) '' num2str(a) '' num2str(m) '' num2str(loops) ''
num2str(junctions) '' num2str(ap) '' num2str(word.lowerBaseline) ''
num2str(word.upperBaseline) '' num2str(word.ascenderBaseline) ''
num2str(word.descenderBaseline) '' num2str(word.angleSlope) '' num2str(word. angleSlant) ''
num2str(ent) '' num2str(M1) '' num2str(M2) '' num2str(M3) '' num2str(M4) '' num2str(M5) ''
num2str(M6) '' num2str(M7)];
ReportGeneratorAdd(axa,'D:\PhD\matlab file\DATASET\bn.csv');
%axa=[horzcat(eh)];
%ReportGeneratorAdd(axa,'Z:\mphil\matlab file\Datasets\hori.arff');
end

```

## Feature Extraction – Paragraph Text Images

```

clc;
clear all;
close all;
warning off all;
[file,path]=uigetfile('10 writer 15 each\*.jpg','Select an image');
i=imread(strcat(path,file));
figure('name','Original image','numbertitle','off')
%imshow(i);
imm=rgb2gray(i);
i1 = rgb2gray(i);
imor=imresize(i1,[256,256]);
i2 = edge(i1,'canny',0.1);
figure('name','Edge image image','numbertitle','off')
imshow(i2)

```

```

% trthh
se = strel('square',2);
i3 = imdilate(i2,se);
figure('name','Dilate image','numbertitle','off')
imshow(i3)
i4 = imfill(i3,'holes');
imshow(i3);
s1 = regionprops(i4, 'BoundingBox');
i5=i4;
figure;imshow(i4);
for i=1:size(i4,1)
    for j=1:size(i4,2)
        if i4(i,j)==1
            for i1=1:30
                i5(i,j+i1)=1;
            end
        end
    end
end
figure;imshow(i5);
figure;imshow(i4);
hold on;
s = regionprops(i5, 'BoundingBox');
t2=1;
for i=1:size(s,1)
    t1=s(i).BoundingBox;
    rectangle('Position',round( s(i).BoundingBox),'LineWidth',2,'Edgecolor','b');
    hold on;
    pause(0.1)
    y(i)=round(t1(2));
end
y1=sort(y);
for i=1:size(y1,2)
    y2=find(y==y1(i));
    for j=1:size(y2,2)
        y3(i+(j-1),:)=round( s(y2(j)).BoundingBox);
        i=i+(j-1);
    end
end
tmp=1;
for i=1:size(s,1)
    t1=y3(i,:);
    if t1(4)>30
        tm2(tmp,:)=t1;
        tmp=tmp+1;
    end

```

```

end
figure;imshow(i4);
hold on;
s = regionprops(i5, 'BoundingBox');
for i=1:size(tm2,1)
%   t1=s(i).BoundingBox;
    rectangle('Position',tm2(i,:),'LineWidth',2,'Edgecolor','b');
    hold on;
    pause(0.1)
%   y(i)=round(t1(2));
end
k=imor;
k1=double(k);
img=imHistogram(k1);
GLCM2 = graycomatrix(k1,'Offset',[2 0;0 2]);
out = GLCM_Features1(GLCM2,0);
f1=(out.contr);
f2=(out.corrn);
f3=(out.corrp);
f4=(out.cprom);
f5=(out.cshad);
f6=(out.dissi);
f7=(out.energ);
f8=(out.entro);
f9=(out.homom);
f10=(out.homop);
f11=(out.maxpr);
f12=(out.sosvh);
f13=(out.savgh);
f14=(out.svarh);
f15=(out.senth);
f16=(out.dvarh);
f17=(out.denth);
f18=(out.inf1h);
f19=(out.inf2h);
f20=(out.homom);
f21=(out.indnc);
f22=(out.idmnc);
f23=(out.autoc);
X=[f1,f2,f3,f4,f5,f6,f7,f9,f10,f11,f12,f13,f14,f15,f16,f17,f18,f19,f20,f21,f22,f23];
img_out_disp=mean(gabor(k));
C = mean(ContCode(k,2));% directional features
d=GGD16(k);
pfilt = '9-7';
dfilt = 'pkva';
nlevs = [0, 0, 4, 4, 5];

```

```

kr=imresize(i1,[512,512]);
y = pdfbdec(double(kr), pfilt, dfilt, nlevs);
c1=GGD16(y{1});
test=[X,img_out_disp,C];
on=1;
if on==1
for it=1:150
    img =imread(strcat('10 writer 15 each\',int2str(it),'.jpg'));
    i=imresize(img,[256 256]);
    imm=rgb2gray(i);
    i1 = rgb2gray(i);
    imor=i1;
    i2 = edge(i1,'canny',0.1);
    se = strel('square',2);
    i3 = imdilate(i2,se);
    i4 = imfill(i3,'holes');
    s1 = regionprops(i4, 'BoundingBox');
    i5=i4;
    % figure;imshow(i4);
    %figure;imshow(i4);
    for i=1:size(i4,1)
        for j=1:size(i4,2)
            if i4(i,j)==1
                for i1=1:30
                    i5(i,j+i1)=1;
                end
            end
        end
    end
    %figure;imshow(i5);
    %figure;imshow(i4);
    k=imor;
    k1=double(k);
    img=imHistogram(k1);
    GLCM2 = graycomatrix(k1,'Offset',[2 0;0 2]);
    out = GLCM_Features1(GLCM2,0);
    f1=(out.contr);
    f2=(out.corrn);
    f3=(out.corrp);
    f4=(out.cprom);
    f5=(out.cshad);
    f6=(out.dissi);
    f7=(out.energ);
    f8=(out.entro);
    f9=(out.homom);
    f10=(out.homop);

```

```

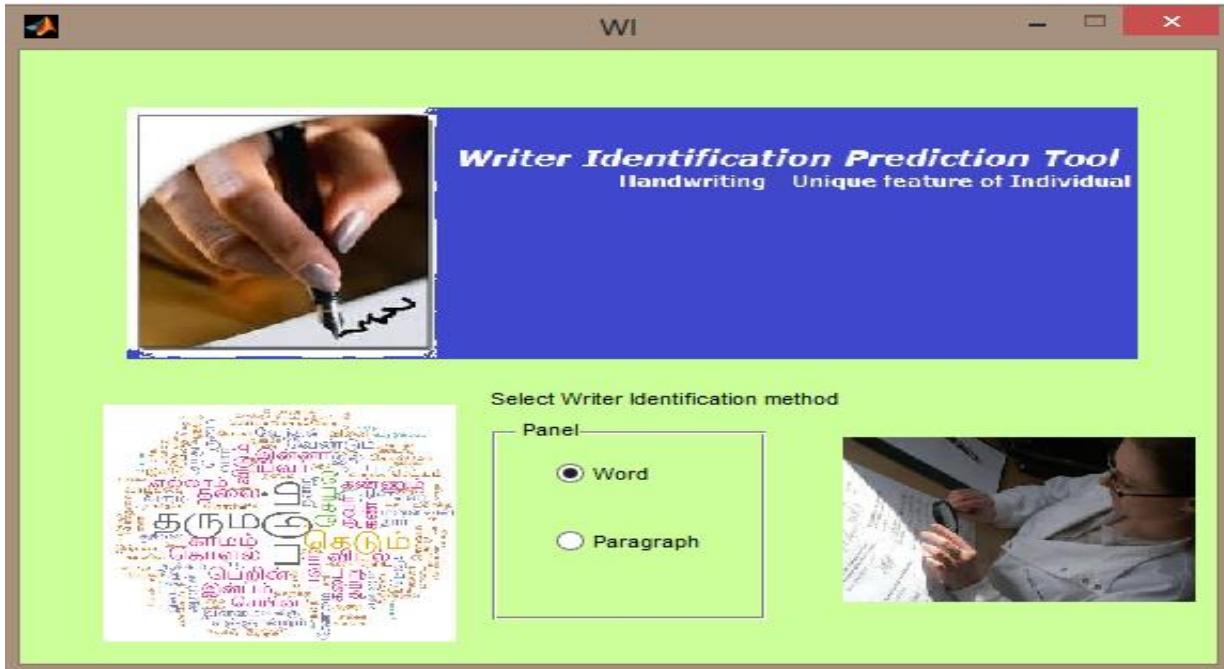
f11=(out.maxpr);
f12=(out.sosvh);
f13=(out.savgh);
f14=(out.svarh);
f15=(out.senth);
f16=(out.dvarh);
f17=(out.denth);
f18=(out.inf1h);
f19=(out.inf2h);
f20=(out.homom);
f21=(out.indnc);
f22=(out.idmnc);
f23=(out.autoc);
X=[f1,f2,f3,f4,f5,f6,f7,f9,f10,f11,f12,f13,f14,f15,f16,f17,f18,f19,f20,f21,f22,f23];
img_out_disp=mean(gabor(k));
% directional features
C = mean(ContCode(k,2));
%GGD FEATURE
d=GGD16(k);
pfilt = '9-7';
dfilt = 'pkva';
nlevs = [0, 0, 4, 4, 5];
kr=imresize(i1,[512,512]);
y = pdfbdec(double(kr), pfilt, dfilt, nlevs);
%CONTOURLET GGD FEATURE
%c1=GGD16(y{1});
train(it,:)=[X,img_out_disp,C];
it
clear y
end
%save trs1.xls train;
end
%load trs;
for i=1:size(train,1)
    for j=1:size(train,2)
        if isnan(train(i,j))==1
            train(i,j)=1;
        end
    end
end
[eivec, prodata, eival] = princomp(train);
[fo, fdx] =sort(eival,'descend');
sc1=train(:, fdx(1:222));
sc2=test(:, fdx(1:222));
train1=xlswrite('D:\PhD\matlab file\Datasets\paragraph\dataset\datasetII.xlsx',train);
sc11=xlswrite('D:\PhD\matlab file\Datasets\paragraph\dataset\datasetIII.xlsx',sc1);

```

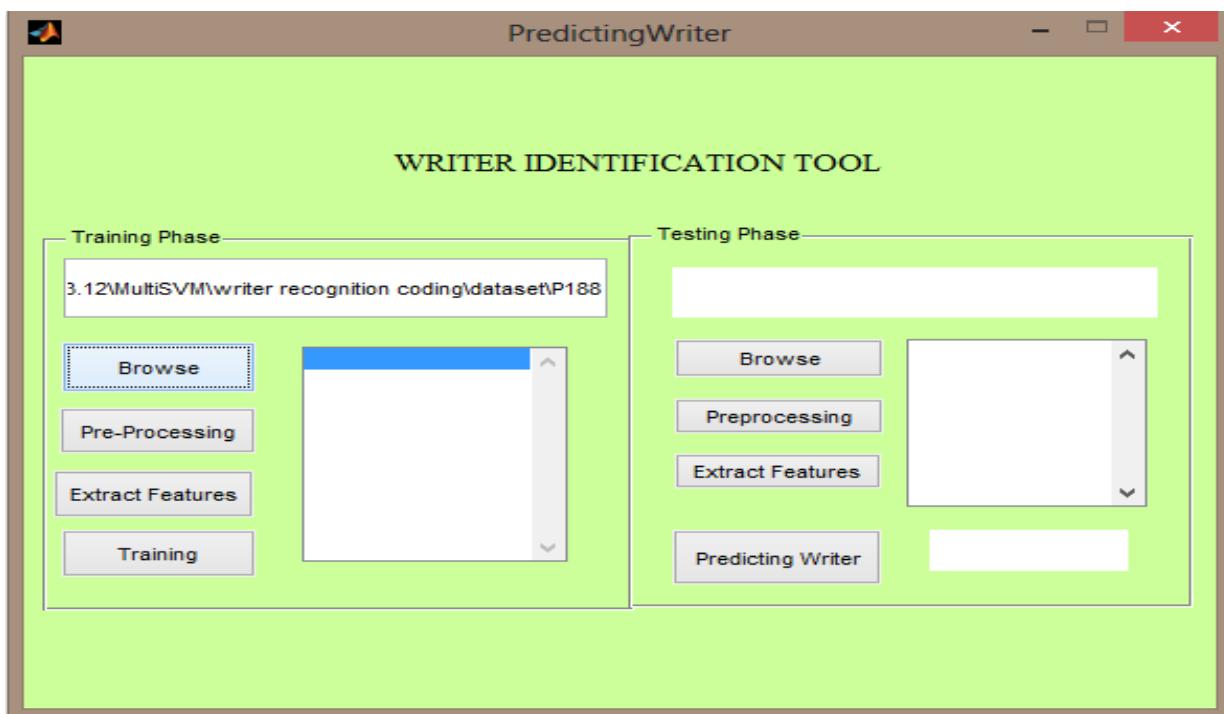
## Appendix - D

### Sample Screen Shots - Writer Identification Tool

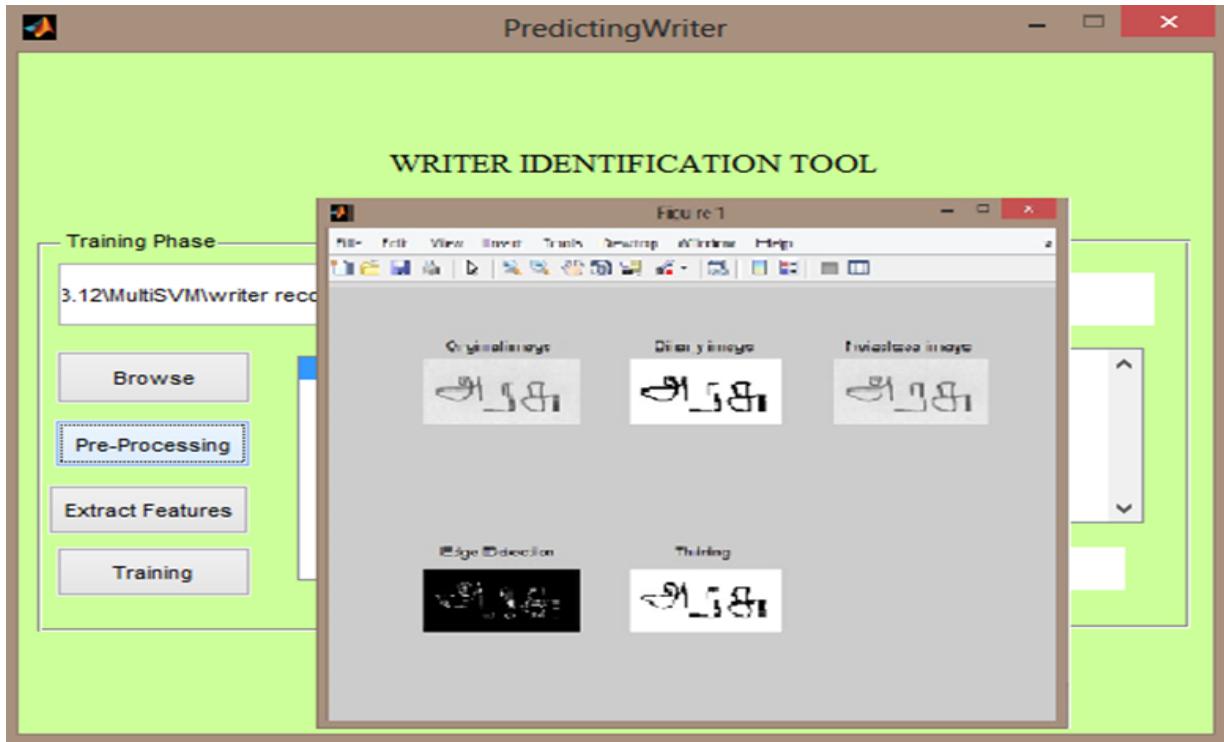
Writer identification tool design – word text image



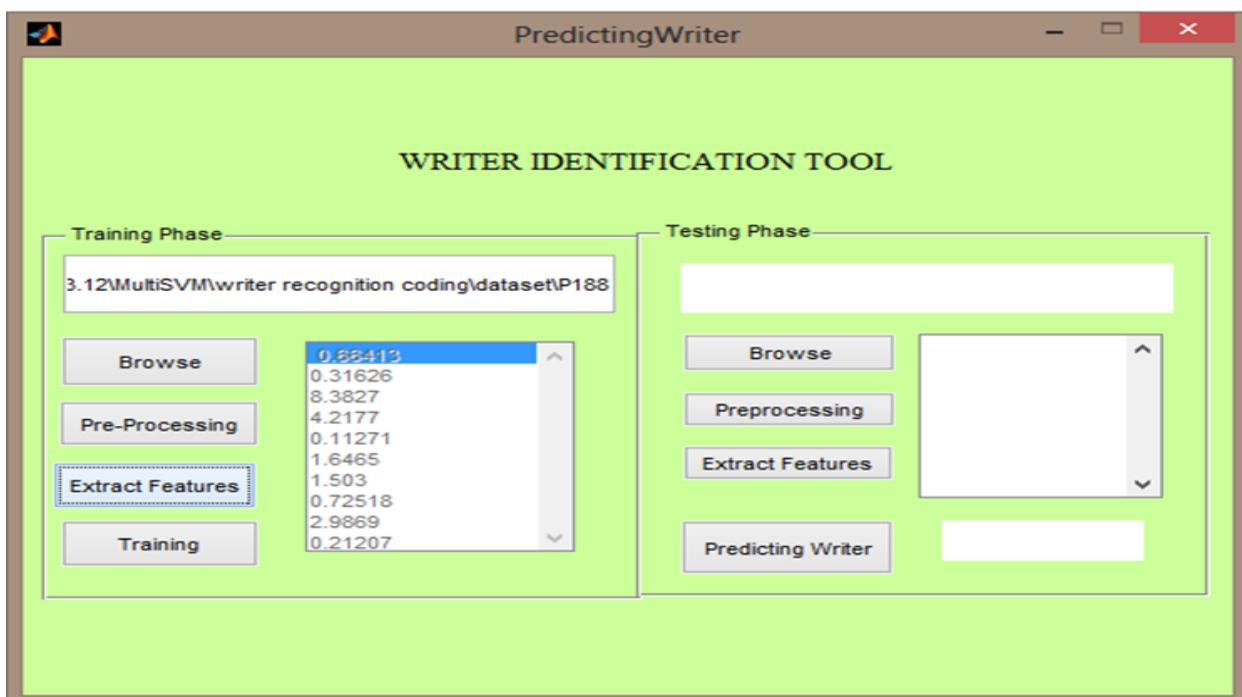
Loading dataset – Training Phase



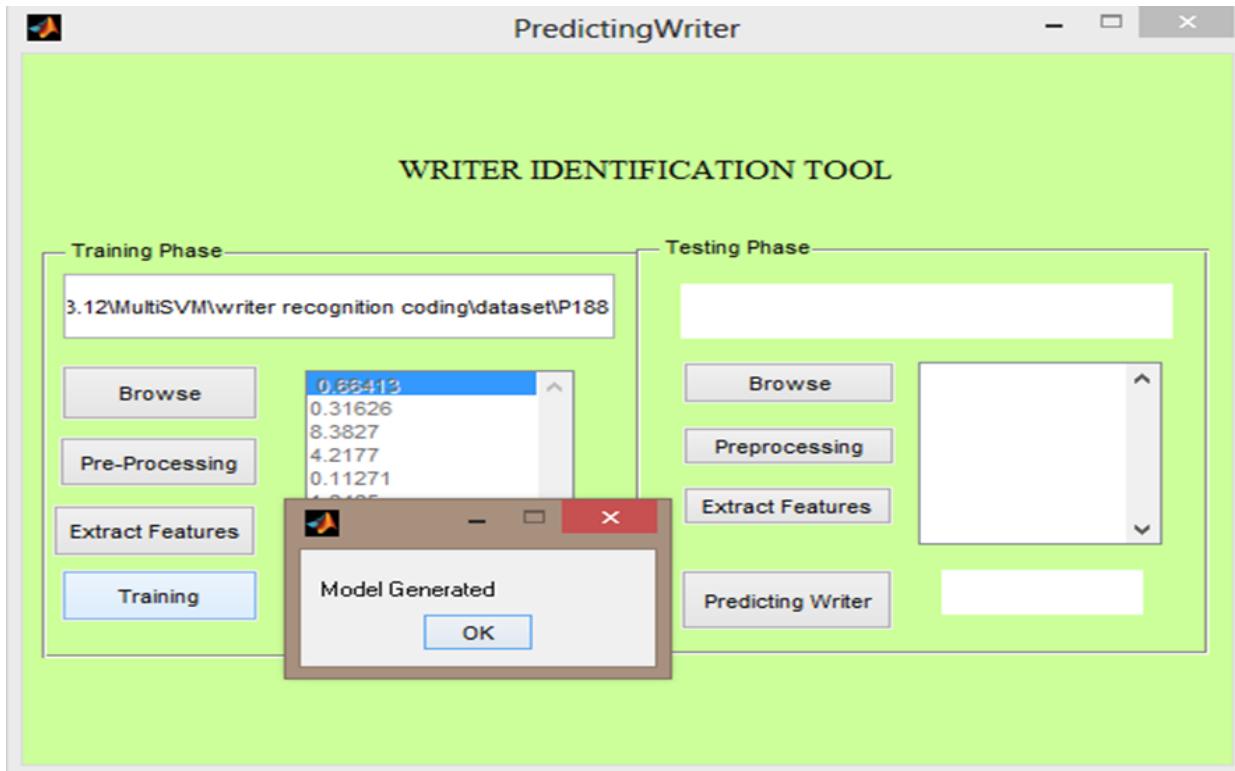
## Preprocessing - Training Phase



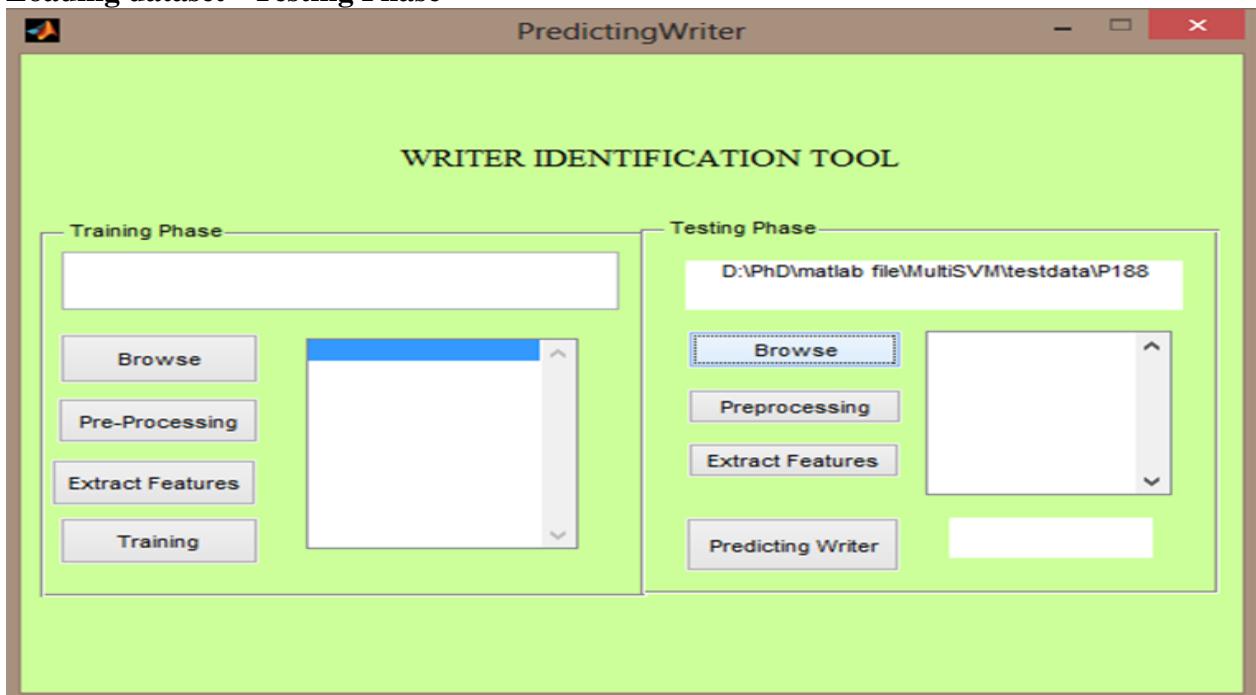
## Feature extraction - Training Phase



## Model generated - Training Phase



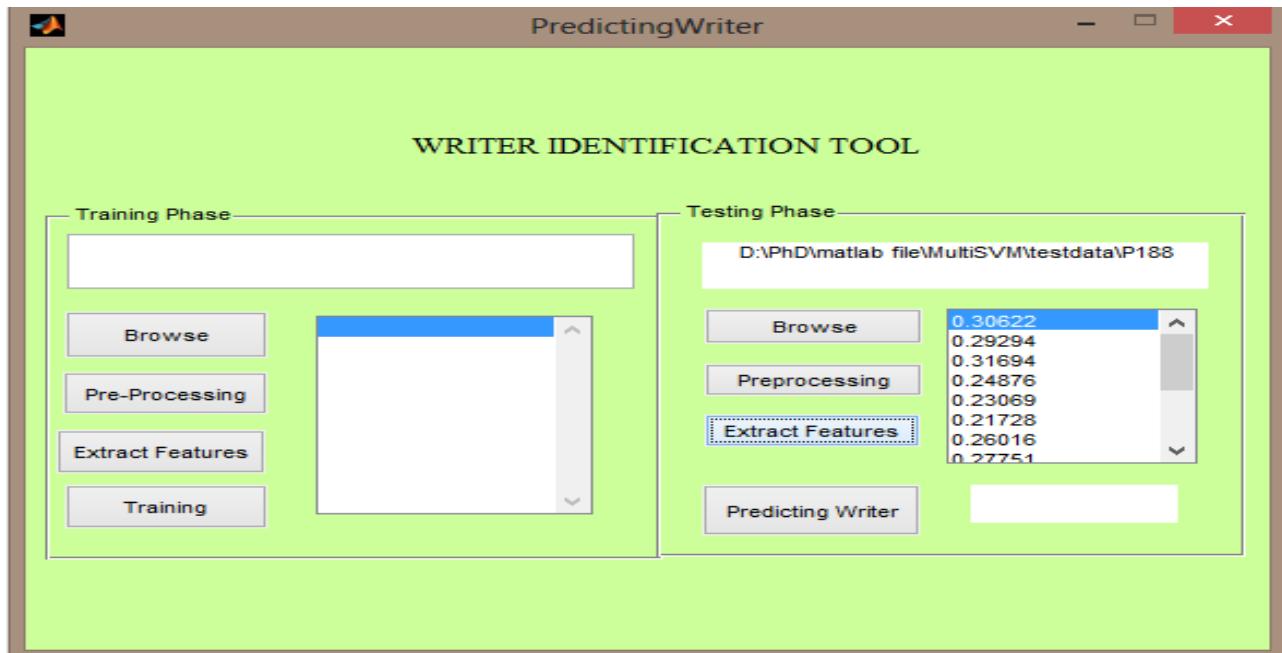
## Loading dataset - Testing Phase



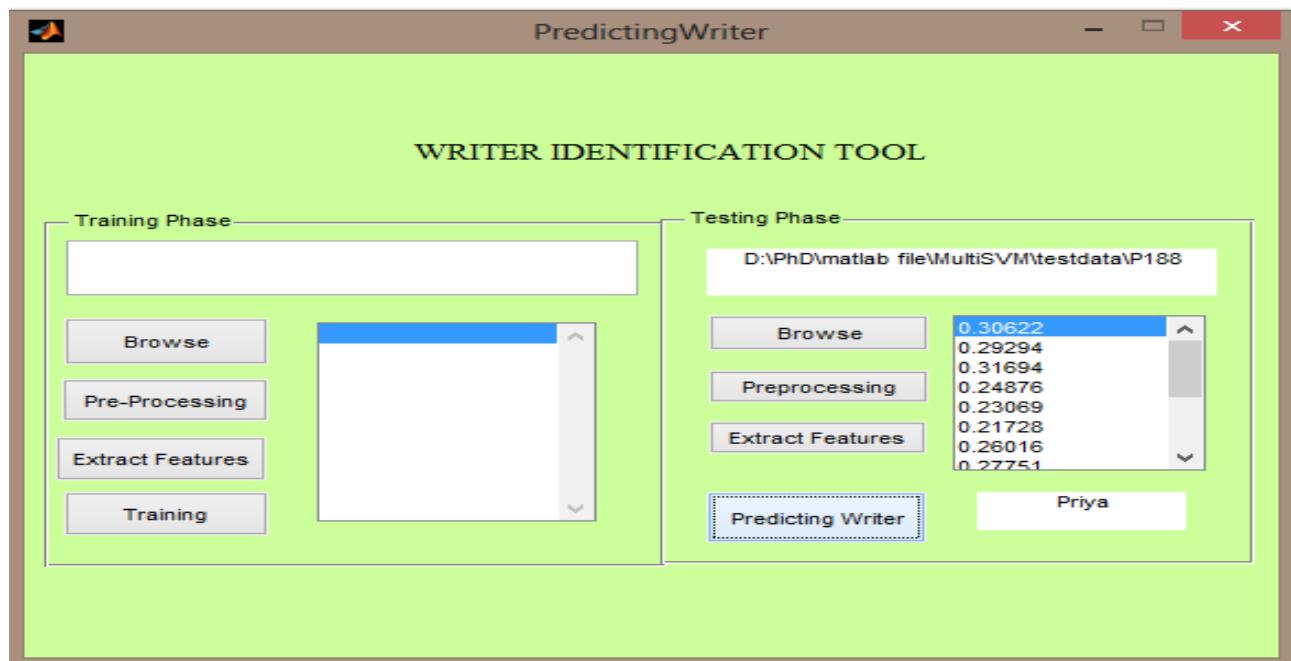
## Preprocessing – Testing Phase



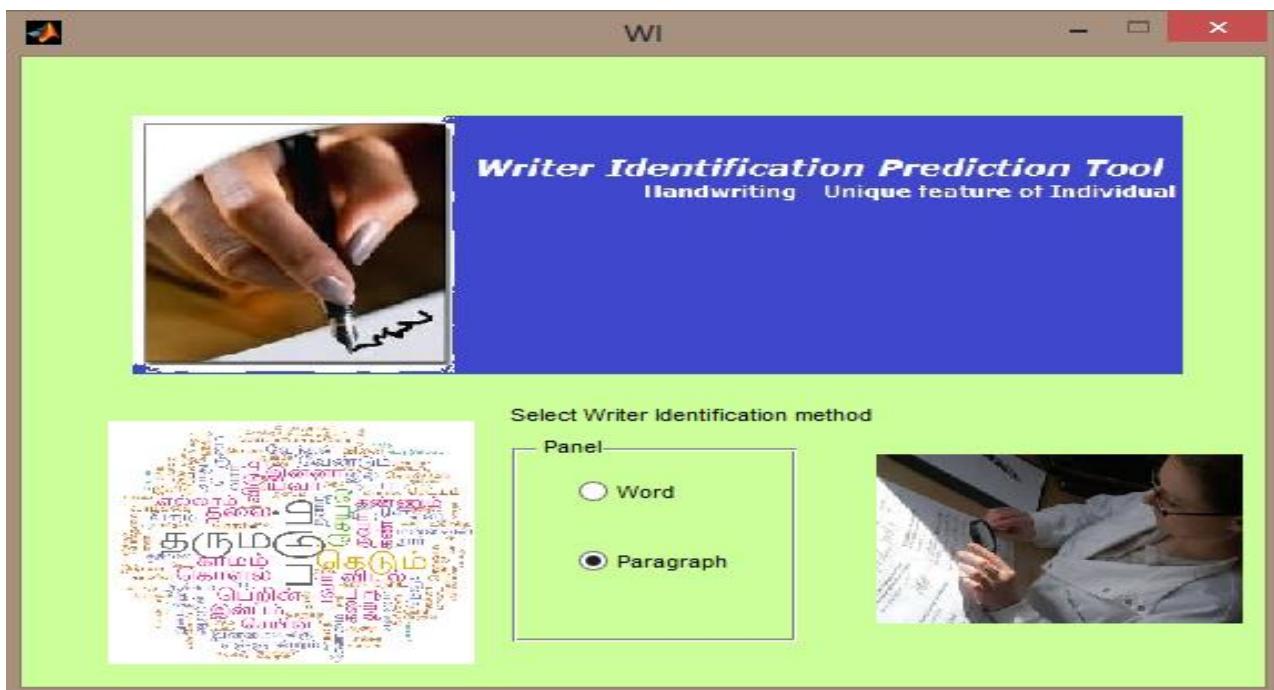
## Feature Extraction - Testing Phase



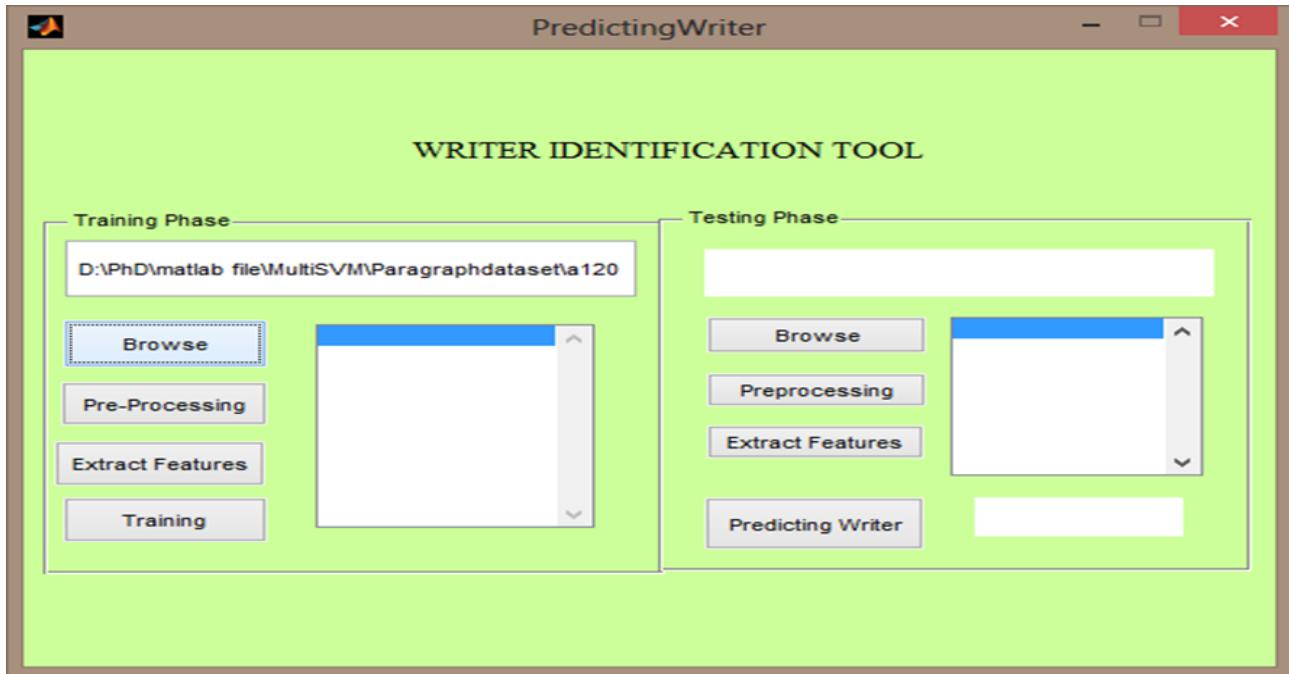
## Predicting Writer using word text image



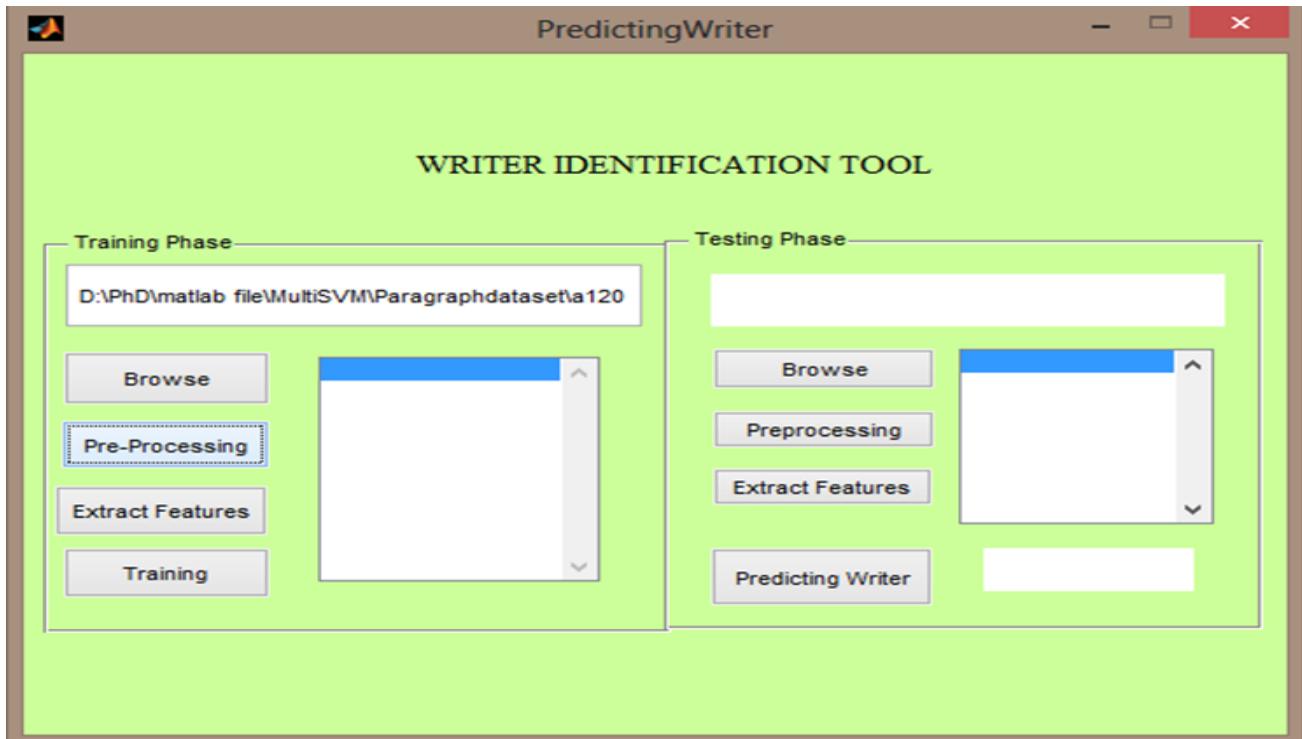
## Writer identification tool - paragraph text image



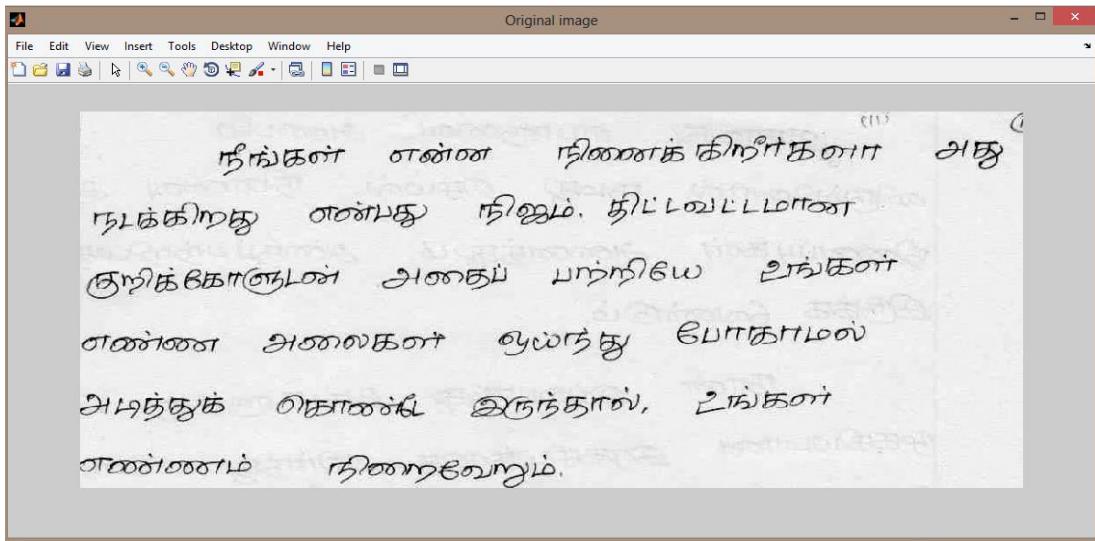
## Loading dataset - Training Phase



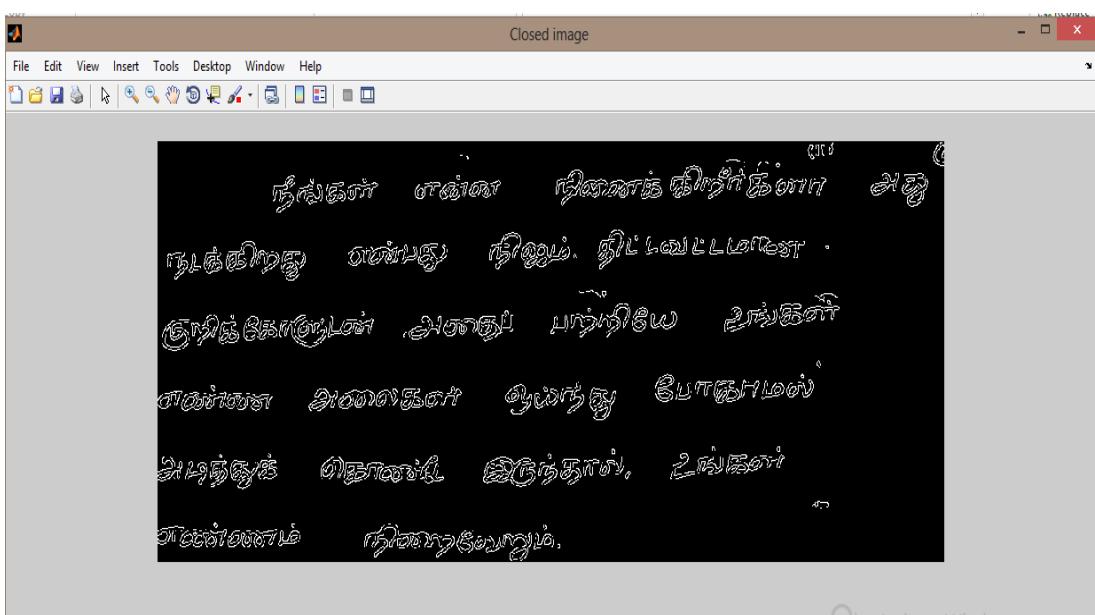
## Preprocessing – Training Phase



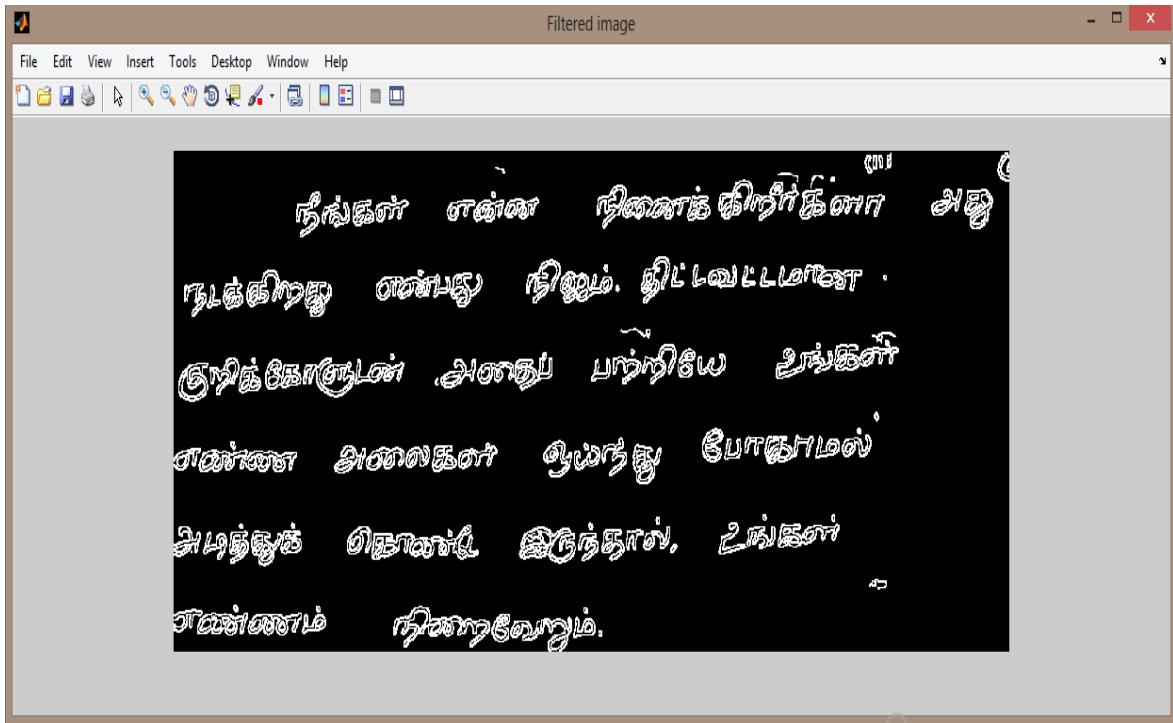
## Original image



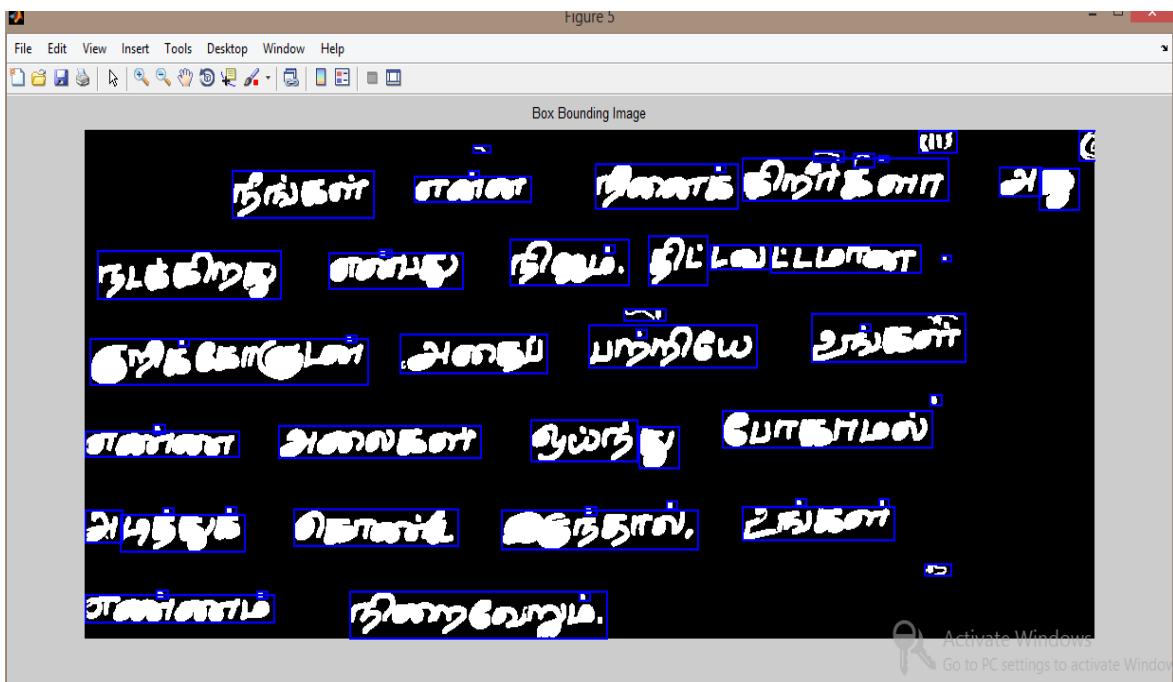
## Dilation



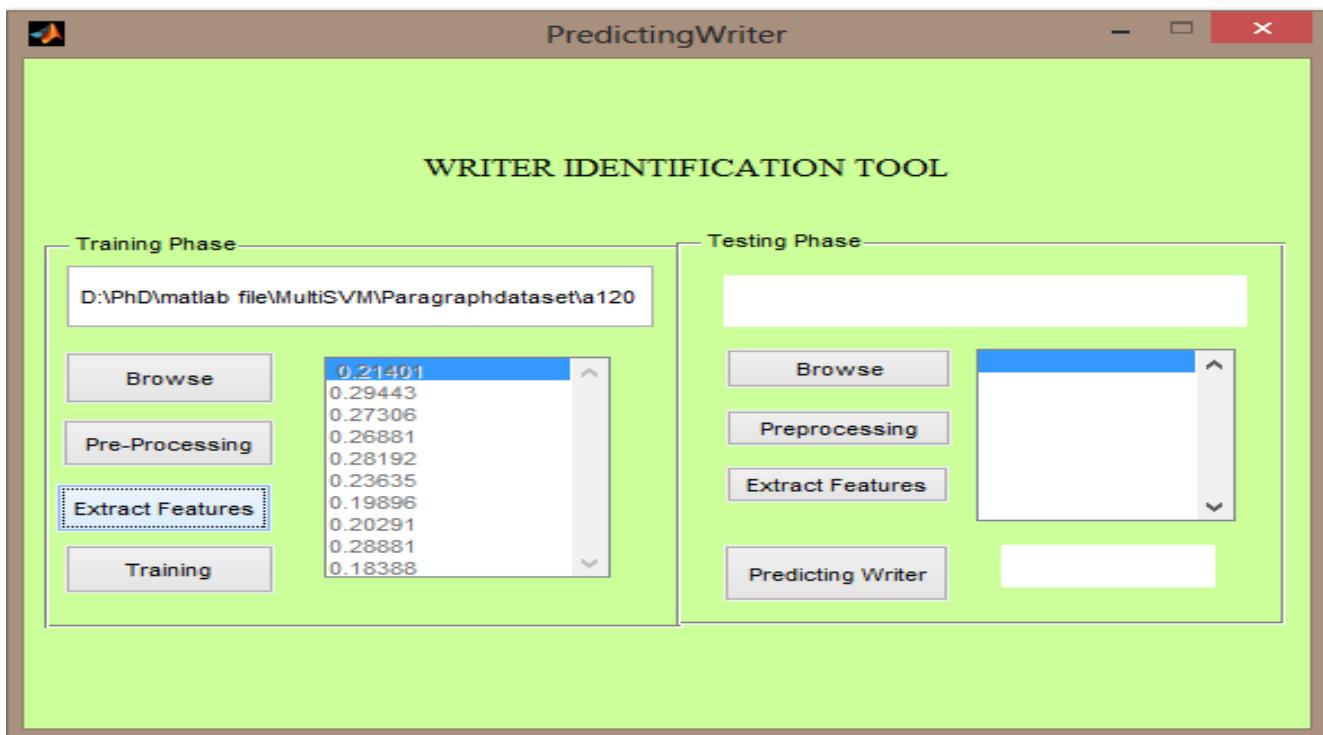
## Edge detection



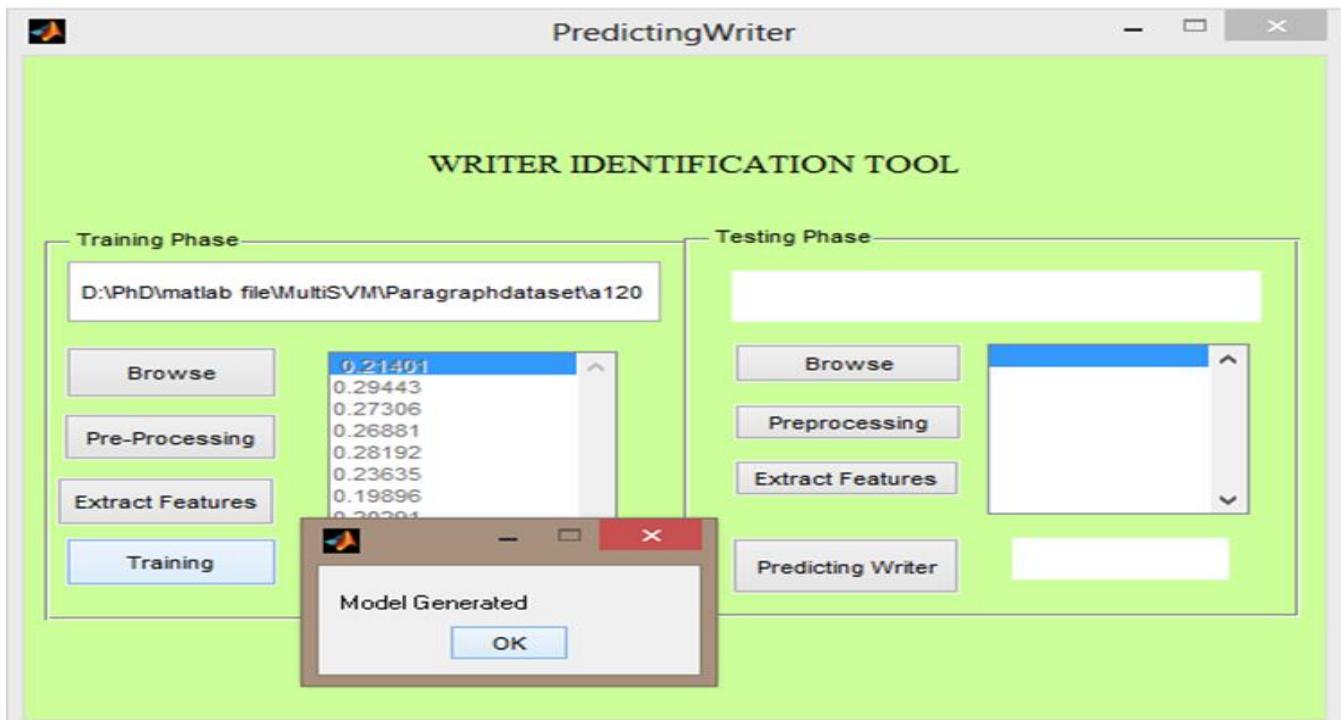
## Box bounding



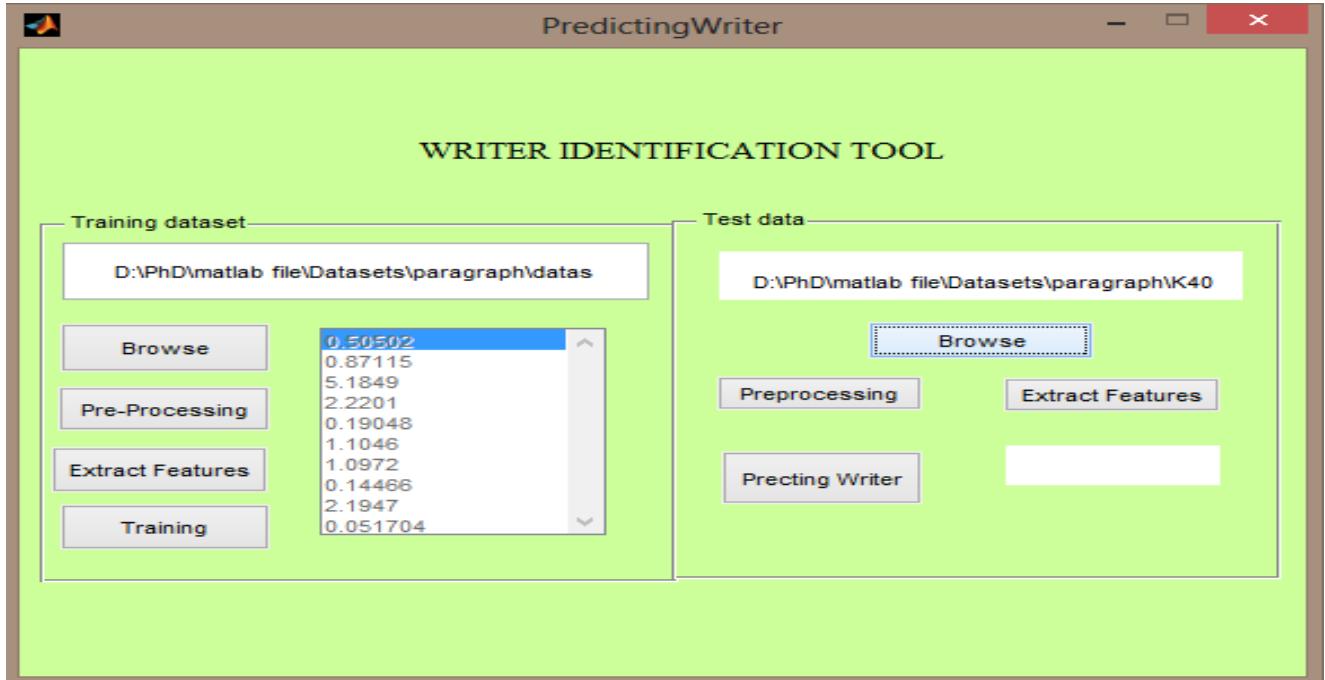
## Feature extraction - Training Phase



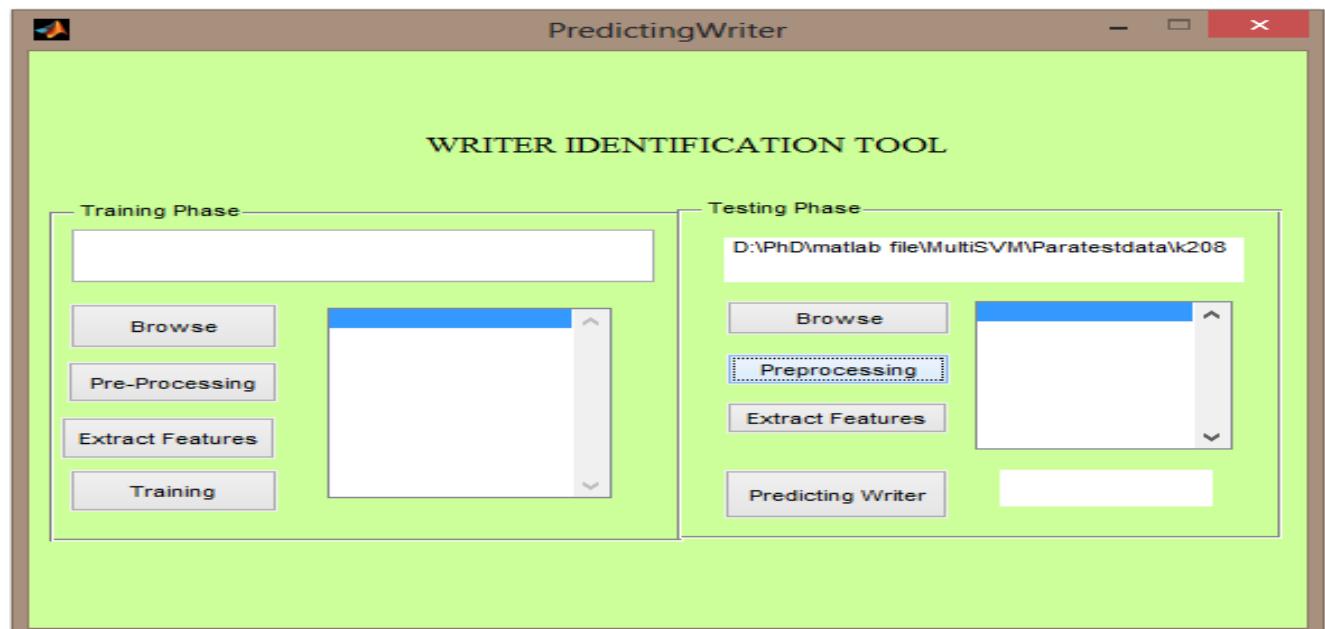
## Model generated - Training Phase



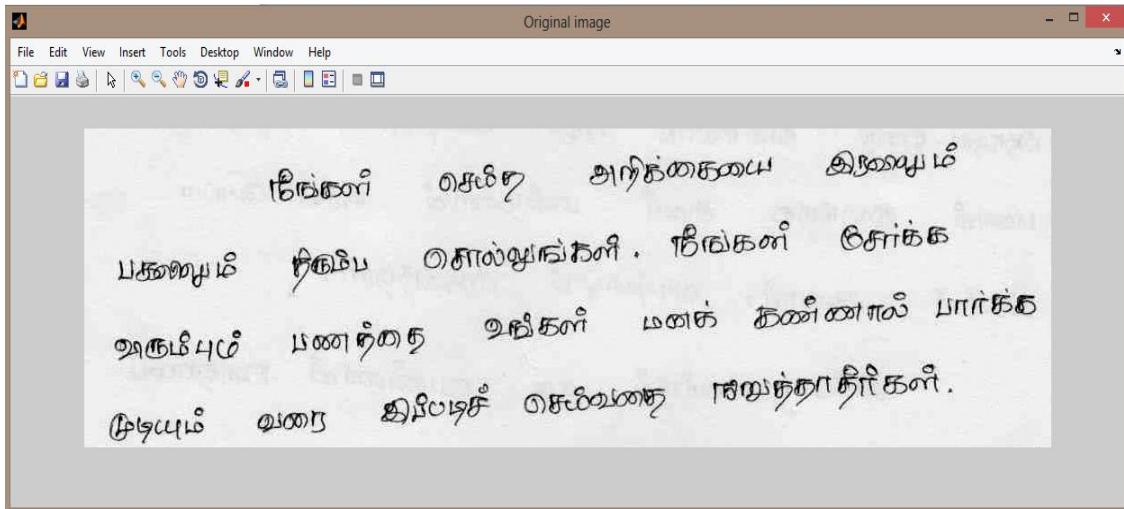
## Loading dataset - Testing Phase



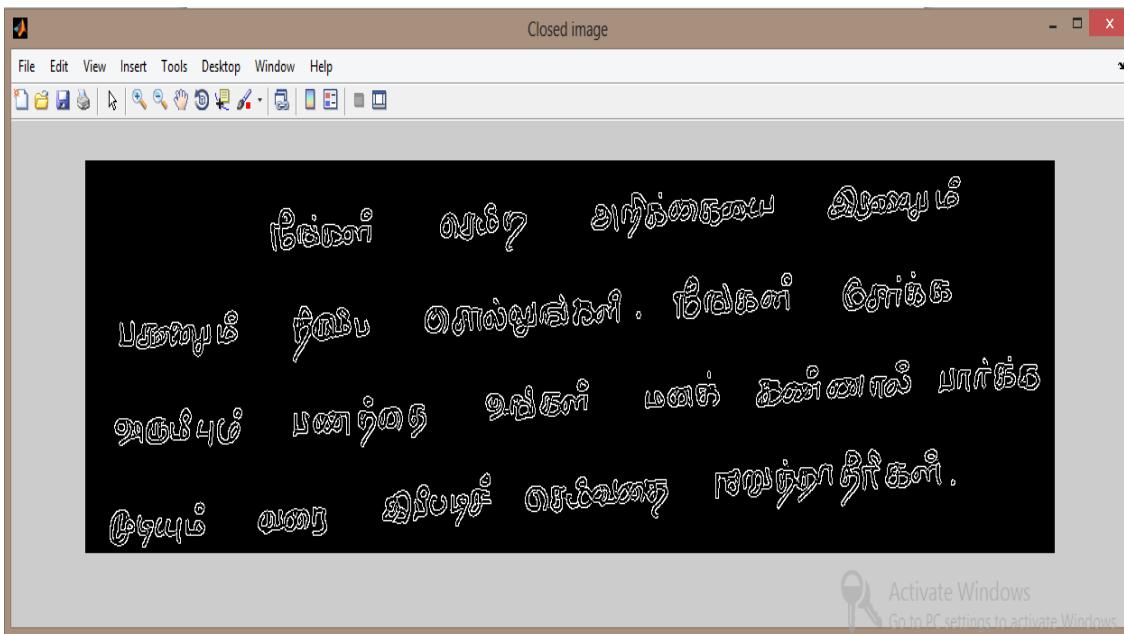
## Preprocessing - Testing Phase



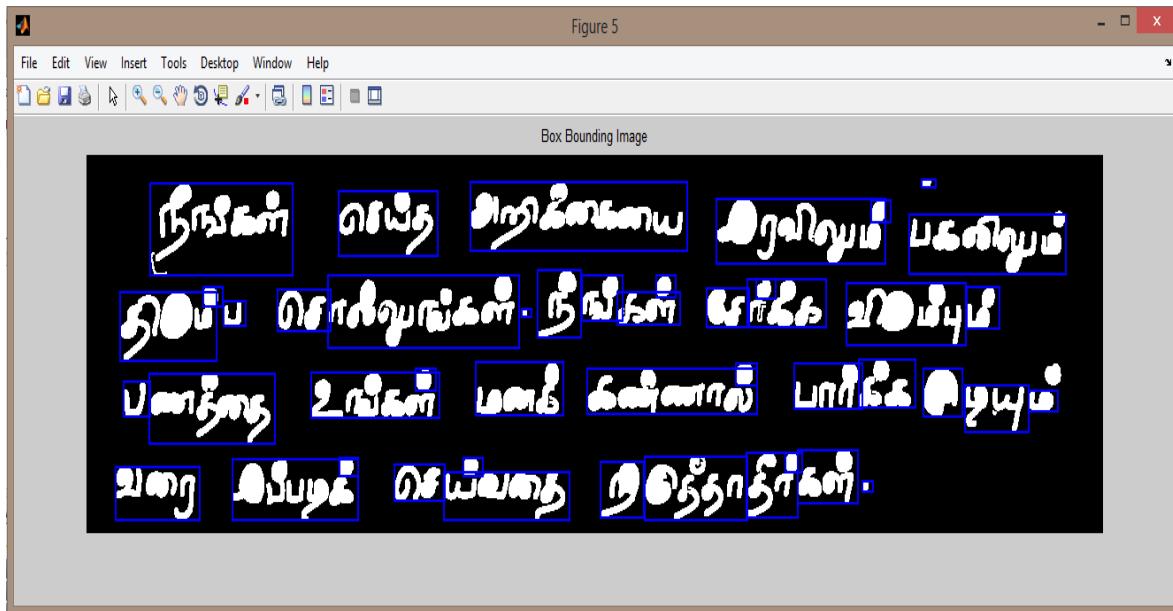
## Original Image



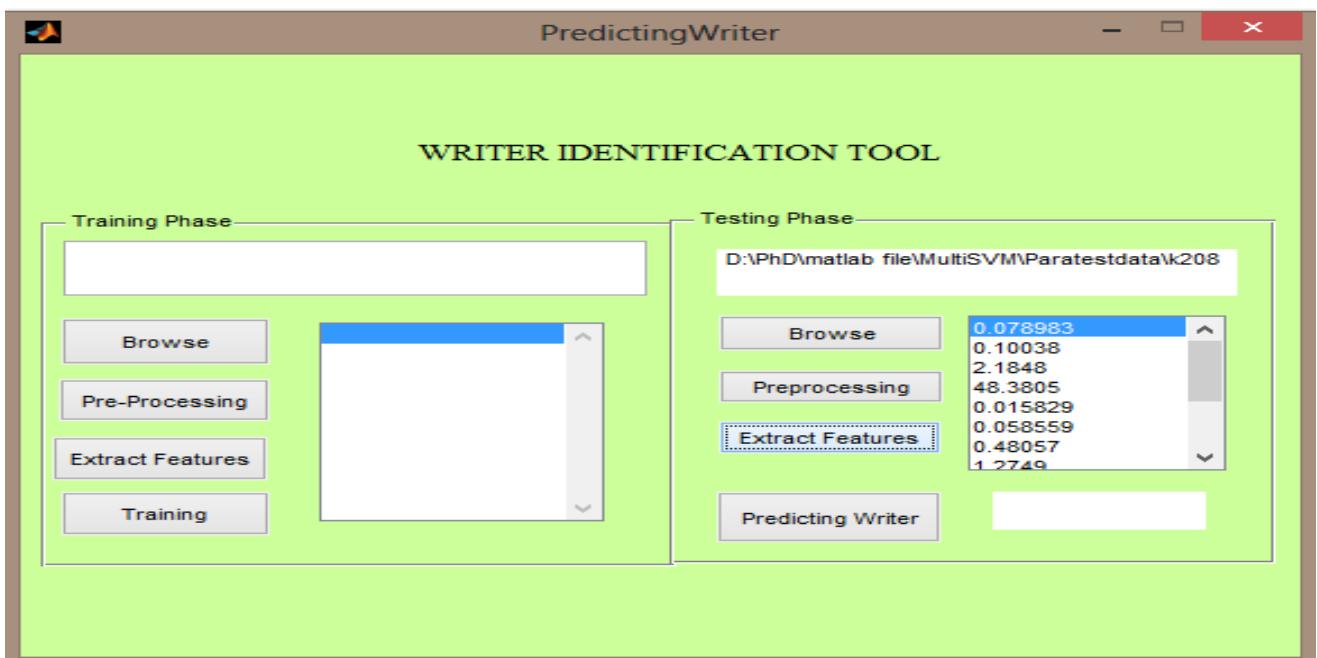
## Dilation



## Boxbounding



## Feature Extraction – Testing Phase



## Predicting Writer using paragraph text image

