



Handwritten Signature Verifier using Image Processing

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Problem definition

- ☐ As said earlier, Signature Recognition is the procedure of determining to whom a particular signature belongs to.
- ☐ System would take as input signature images and tell us
 - To whom the signature belongs to (Author Identification)
 - If the signature is forged or genuine (Signature Verification)





Literature Review

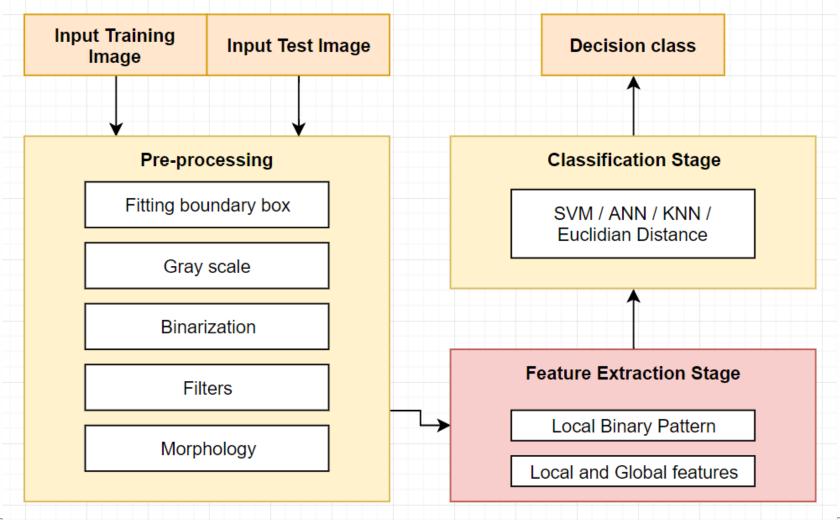
- ☐ An excel sheet of all the Literature Review has been prepared and is been attached with this slide.
- ☐ The literature review contains total of 40 research papers based on the topic Signature Recognition
- ☐ Most of the papers make use of 3 stages:
 - Preprocessing stage
 - Feature extraction stage
 - Classification stage







Proposed Work



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SVKM'S NMIMS Dominal to be UNIVERSITY

Proposed Work

☐ Local Binary Pattern

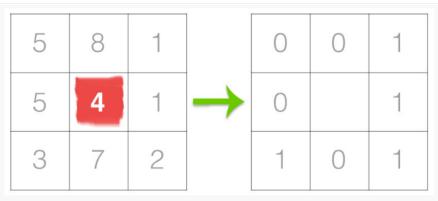


Figure 1: The first step in constructing a LBP is to take the 8 pixel neighborhood surrounding a center pixel and threshold it to construct a set of 8 binary digits. ____

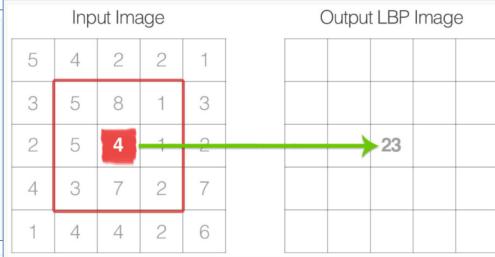


Figure 3: The calculated LBP value is then stored in an output array with the same width and height as the original image.

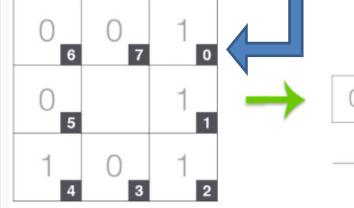




Figure 2: Taking the 8-bit binary neighborhood of the center pixel and converting it into a decimal representation. (Thanks to Bikramjot of Hanzra Tech for the inspiration on this visualization!)





Implementation tool & setup

☐ Python using PyCharm

- Python is a popular programming language used in web & software development, mathematics, system scripting
- PyCharm is a python editor and compiler allows intelligent code completion, on-the-fly error checking and quick-fixes, easy project navigation, and much more.







Implementation tool & setup

Python libraries

- Using number of libraries, which are easy to install & import..
 - OpenCV : Computer vision and machine learning software library.
 - NumPy : Scientific computing & array-processing
 - Imutils : Functions to make basic image processing functions easier
 - Math: Provides access to the mathematical functions
 - MatplotLib : Python 2D plotting library
 - Pymysql : A simple database interface for Python
 - OS : allows easy file handling
 - Scipy: rovides many user-friendly and efficient numerical routines





Implementation tool & setup

☐ SQL using MySQL

- SQL is a standard language for storing, manipulating and retrieving data in databases.
- MySQL is an open source relational database management system, very easy to establish, use and manage



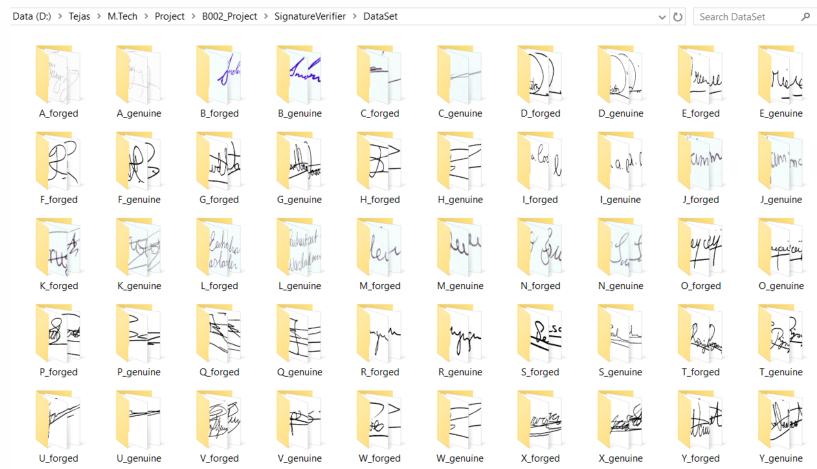






Dataset

25 Authors, 50 Classes, **826 Images** 637 Training images (77.12%) 189 Testing images (22.88%)





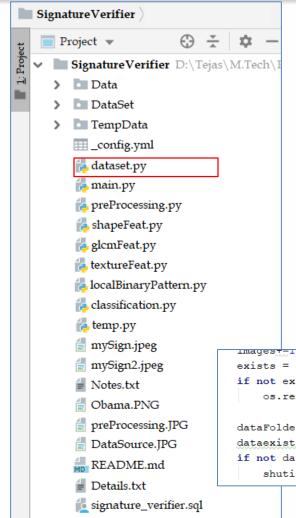


Implementation Work done





Implementation



lbp.JPG

datasetCount.txt

□ Dataset.py

- Our dataset is read, renamed, copied and organized in the correct naming convention to a different folder, from where out system will use
- xyz.png A_genuine_7.png
- Also gives an analysis of the count of dataset

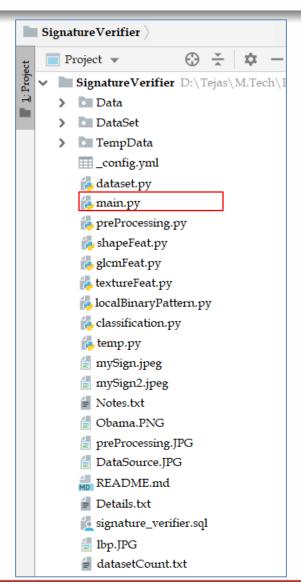
```
mages = 1
exists = os.path.isfile("DataSet/"+folder+"/"+folder+"_"+str(j)+".png")
if not exists:
    os.rename("DataSet/" + folder + "/" + file, "DataSet/" + folder + "/" + folder + "_" + str(j) + ".png")

dataFolder = training_folder if (j < (4*total/5)) else testing_folder
    dataexists = os.path.isfile(dataFolder + "/" + folder + "_" + str(j) + ".png")
if not dataexists:
    shutil.copy("DataSet/" + folder + "/" + folder + "_" + str(j) + ".png", dataFolder)</pre>
```





Implementation



☐ Main.py

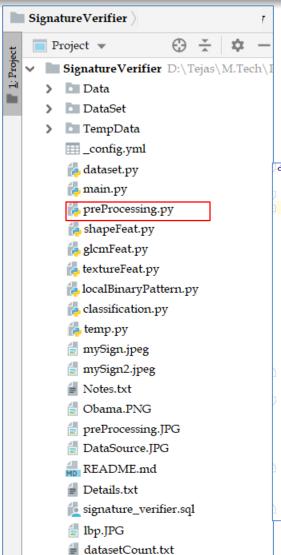
 This is the main python file where the system working starts, calls the other functions, gives the appropriate results and ends.

```
for filename in os.listdir(testing folder):
   img = cv.imread(os.path.join(testing folder, filename), 0)
   if img is not None:
       isLBP = False
       f = open("Data/" + datafile, "a")
       f.write("\n")
       print("Image file : ", filename)
       f.write("\nImage file : "+ str(filename) +"\n")
       f.close()
       orgImg = img
       # cv.imshow(filename, img)
       proImg = pr.preprocess(orgImg, datafile)
       # cv.imshow(filename, myImg)
       sf.shapeFeat(proImg, testingFeatures, isLBP, datafile, filename)
       gf.glcm(proImg, testingFeatures, datafile)
       tf.textFeat(proImg, testingFeatures, datafile)
       isLBP = True
       lbpImg = lbp.lbp(orgImg, datafile)
       sf.shapeFeat(lbpImg, testingFeatures, isLBP, datafile, filename)
       qf.qlcm(lbpImq, testingFeatures, datafile)
       tf.textFeat(lbpImg, testingFeatures, datafile)
       cl.actualclass(filename, testingClasses, datafile)
       cl.knn(trainingFeatures, testingFeatures, trainingClasses, decisionClasses, datafile)
```





Implementation



Preprocessing.py

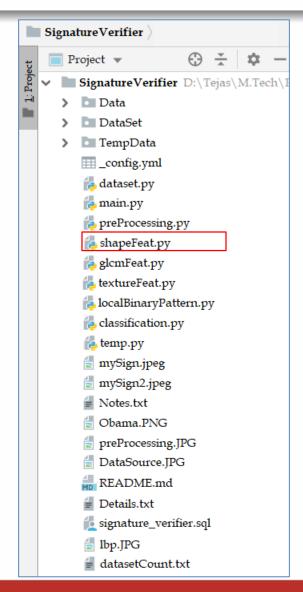
- This function takes the image in the raw format and converts into a pre-processed format.
- Resize, RGB to Grey, Otsu thresholding, boundary Box cropping

```
def preprocess(img, datafile):
        print()
        if (len (img.shape) >2):
            img = cv.cvtColor(img, cv.COLOR RGB2GRAY)
        img = imutils.resize(img, 720)
        # ~~ Otsu's thresholding after Gaussian filtering ~~
        blur = cv.GaussianBlur(imq, (3,3),0)
        ret, img = cv.threshold(blur, 0, 255, cv.THRESH OTSU)
        img = boundaryBox(img, datafile)
    except Exception as error:
        print("An exception was thrown in " + inspect.stack()[0][3])
        f = open("Data/"+datafile, "a")
        print("Error: "+ str(error))
        f.write("\nError: "+ str(error))
        f.close()
    finally:
        return img
```





Implementation



☐ shapeFeat.py

- This function extracts shape based features of the signature from the images.
- These includes:
 - Aspect Ratio
 - Center of Gravity
 - Normalized Area
 - Baseline Shift
 - Eccentricity
 - HuMoments
 - No of Corners

Image file : Y_genuine_12.png

~~~ Shape features ~~~

Aspect Ratio : 1.8882175226586102

X\_COG: 314.35614528971786 Y COG: 160.09018323098957

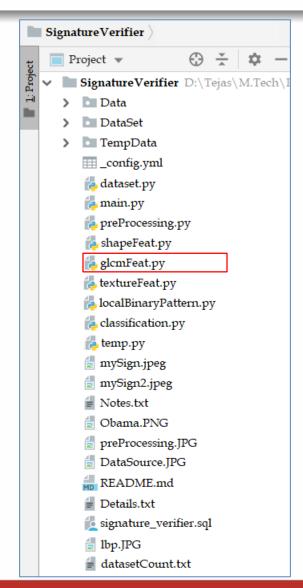
Normalized area: 0.8808652567975831 Baseline shift: 5.49301214401541 Eccentricity: 0.844747849543066 HuMoments: 13.06358045823279

Corners: 317





# **Implementation**



## □ glcmFeat.py

- This function extracts glcm based features of the signature from the images.
- Gray Level Coherence Matrix
- These includes:
  - Contrast
  - Dissimilarity
  - Homogeneity
  - Energy
  - Correlation
  - ASM

Image file : Y\_genuine\_12.png

~~~ GLCM features ~~~

Contrast: 1619.1396264234256

Dissimilarity: 6.3495671624448065

Homogeneity: 0.9751001195456674

Energy: 0.8751210985740953

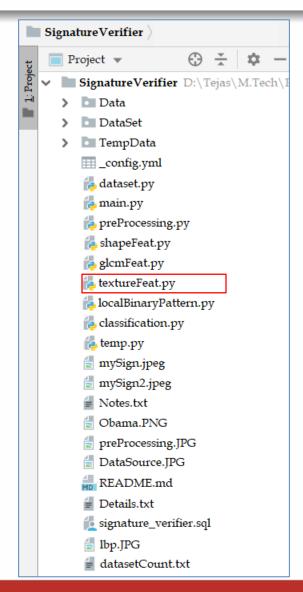
Correlation: 0.8813611181808606

ASM: 0.7658369371695313





Implementation



□ textureFeat.py

- This function extracts texture based features of the signature from the images.
- These includes:
 - Mean
 - Variance
 - Skewness
 - Kurtosis
 - Energy
 - Haralick

```
Image file: Y_genuine_12.png

--- Texture features ---
Mean: 224.62064048338368

Variance: 132.49999999976717

Skewness: -2.67989282143756

Kurtosis: 6.181647412206975

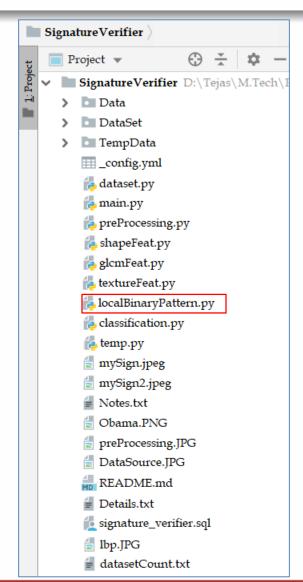
Entropy: 5.672146449324859

Haralick: 2663.808568037615
```





Implementation



- □ localBinaryPattern.py
 - This python file contains function that converts the image to an LBP image.
 - Local Binary Pattern

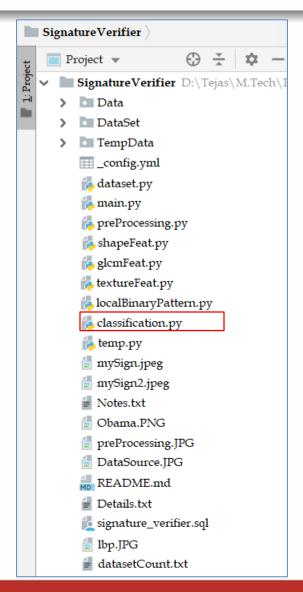
Campus

```
def lbp(img):
   if (len (img.shape) >2):
       img = cv.cvtColor(img, cv.COLOR RGB2GRAY)
   binary = img.copy()
   lbpImg = img.copy()
   height = img.shape[0]
   width = img.shape[1]
   while (y<height-1):
       x=1
       while (x < width-1):
           binary[y-1][x-1] = 1 if(img[y-1][x-1] < img[y][x]) else 0
           binary[y-1][x] = 1 if(img[y-1][x] < img[y][x]) else 0
           binary[y-1][x+1] = 1 if (img[y-1][x+1] < img[y][x]) else 0
           binary[y][x-1] = 1 if(img[y][x-1] < img[y][x]) else 0
           binary[y][x+1] = 1 if(img[y][x+1] < img[y][x]) else 0
           binary[y+1][x-1] = 1 if (img[y+1][x-1] < img[y][x]) else 0
           binary[y+1][x] = 1 if(img[y+1][x] < img[y][x]) else 0
           binary[y+1][x+1] = 1 if (img[y+1][x+1] < img[y][x]) else 0
           lbpImg[y][x] = (binary[y][x-1]*128) + (binary[y-1]
           v+=1
   return lbpImg
   Computer Engineering Dept. MPSTME, Mumbai
```





Implementation



- ☐ classification.py
 - This python file contains KNN classification function which classifies the given test data to a particular class.
 - There are two functions one that gets the actual class of the function by merely the name of the file and other is the KNN classifier.

```
Image file : A_forged_21.png

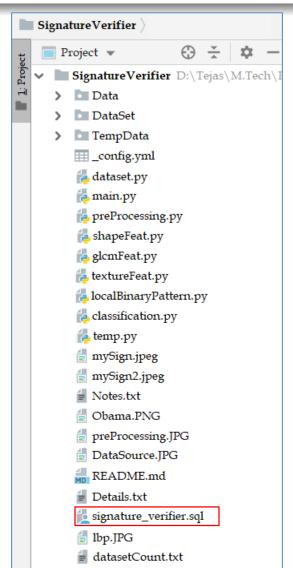
Actual Class: A_forged

Decision: A_forged
```





Implementation



□ DATABASE

- This has SQL queries for creating database, creating tables, inserting entries into the tables.
- Main.py also inserts and reads each feature vector into the database

| myadmin/db_s | structure.pl | np?server=1&db | =signature | _verifier | | | | | | | | |
|---|--------------|------------------|---------------|--------------|-------------------|----------------|------|-----------|------------|-----------|-----------|---------|
| Server: 12 | 27.0.0.1 » | 🗑 Database: sign | ature_verific | er | | | | | | | | |
| Structure 📗 SQL | | _ | Que | ry 🖶 I | Export | 🛺 Import 🥜 | | perations | Privileges | | Routines | |
| Filters Containing the word: | | | | | | | | | | | | |
| Table 🔺 | A | ction | | | | | | Rows 🕢 | Туре | Collation | | Size |
| testing_cla | asses 📹 | Browse J | Structure | 🕦 Search | ⊪ i Insert | 🚍 Empty | Drop | 189 | InnoDB | latin1_sw | edish_ci | 16 KiB |
| testing_fe | atures 🧌 | Browse J | Structure | Search | ⊪ i Insert | 🖷 Empty | Drop | 189 | InnoDB | latin1_sw | edish_ci | 64 KiB |
| training_c | lasses 🤺 | Browse J | Structure | 🕦 Search | ⊪ i Insert | Empty | Drop | 637 | InnoDB | latin1_sw | edish_ci | 64 KiB |
| training_fe | eatures 🧌 | Browse J | Structure | Search | ≟ i Insert | ⊞ Empty | Drop | 637 | InnoDB | latin1_sw | edish_ci | 144 KiB |
| 4 tables | Sı | ım | | | | | | 1,652 | InnoDB | latin1_sv | /edish_ci | 288 KiB |
| ↑ □ Che | eck all | With selected | : | | • | | | | | | | |
| Print 📠 Data | | , | | | | | | | | | | |
| Name: | | | Numl | per of colur | mns: 4 | | | | | | | |
| Computer Engineering Dept. MPSTME, Mumbai | | | | | | | | | | 20 | | |





Gantt Chart







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Thank You ©

Any Questions?