

Exploring Facial Recognition Techniques: LBP versus Deep Learning on LFW Dataset

M.Sc. of Cybersecurity

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Objective

1 Introduction

- **Local Binary Pattern**

- captures pixel-wise patterns and transitions within small regions
- computationally efficient for facial recognition

- **Deep learning**

- Deep neural networks for the extraction of facial features
- Raise the level of accuracy in face recognition

- **Main aim:**

- Highlights differences between Face detection and Recognition with LBP and Deep Learning
- Use an unconstrained dataset like **LFW**.



Identification Open Set

1 Introduction

- **Tasks**

- Determine whether a probe's biometric signature matches someone's in the gallery

- **Threshold settings**

- customized to achieve ideal rates according to application purpose





Dataset

1 Introduction

- **Labeled Faces in the Wild**

- Images captured under uncontrolled conditions
- Variations in terms of PIE (Pose, Illumination and Expression)
- Partial Occlusion





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2 Project Implementation

► Introduction

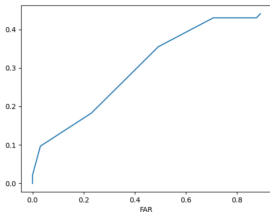
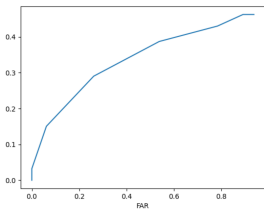
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Local Binary Pattern

2 Project Implementation



- **Face Detection** carried out with the *face_detect()* function in which we could use either **Haar Classifier** or **LBP Classifier**
- **Haar Cascade Classifier**
 - Effective and highly accurate for detection in various scenarios but require significant memory resources and may struggle with image variations.
- **LBP Classifier**
 - Efficient and faster but sensitive to noise in the data, leading to false positives or decreased accuracy



Local Binary Pattern

2 Project Implementation

- **Create Face recognizer**
- **Face Recognition** The face Recognition happens inside the predict function
- **Label** is the label predicted that is inside the gallery while **the confidence** is a measure of distance.

```
#create the LBP face recognizer  
face_recognizer = cv2.face.LBPHFaceRecognizer_create()
```

```
label, confidence = face_recognizer.predict(face)
```




Local Binary Pattern

2 Project Implementation

- Rates

- Confidence is the value to compare to the threshold in order to increment **GA**, **FR**, **FA**, **GR**.

```
def calculate_metrics(t, test_img, subfolder, predicted_img, subjects, DI, FR, FA, GR):  
    (label_pred, predicted, confidence) = predict(test_img)  
    if predicted is not None:  
        if confidence <= t:  
            if label_pred == subfolder:  
                predicted_img.append((label_pred, predicted))  
                DI += 1  
            else:  
                if subfolder not in subjects:  
                    FA += 1  
        else:  
            if label_pred != subfolder:  
                GR += 1  
    else:  
        GR += 1  
  
    return (DI, FR, FA, GR)
```



Deep Learning

2 Project Implementation

- **Deep Face**

- **Find** function
- VGG model as recognition model
- OpenCV for face detection
- Thresholds varies from 0.01 to 0.99

For deep learning we used the same logic to calculate rates.

- **Cosine Similarity**

- used to measure the or similarity or distance between two face embedding, treated as vectors, calculating the cosine of the angle between them.



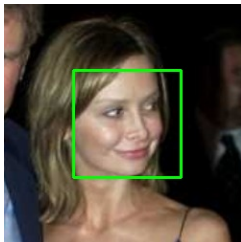
Deep Learning

2 Project Implementation

- **Deep Face**

- Organization of probes and gallery is the same used for LBP
- Only variation in reprocessing: *find* function, which search for the identity in the gallery and return list of pandas data frame
- Calculate the number of GA, FR, FA and GR

- Due to the different nature of DeepFace, we're now able to correctly detect the face of Calista





Performance Evaluations

2 Project Implementation

- **Assessment**

- Calculate the following rates:

- Detection and Identification Rate $DIR(t,1)$: rank k is 1 for both LBP and DF
- False Acceptance Rate (FAR)
- False Rejection Rate (FRR)
- Equal Error Rate (ERR)

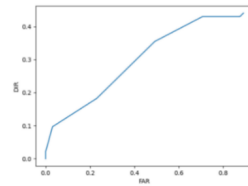
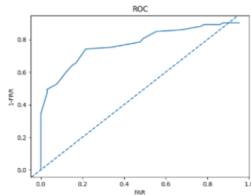
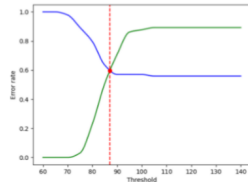
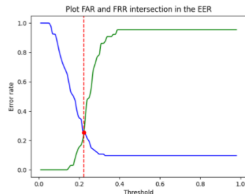




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Considerations

3 Conclusion

- **Deep Face vs LBP**

- LBP ability to address complex variations may be limited, especially with diverse lighting and angles
- Deep Learning excelled in flexibility even with complex patterns
- LBP approach is faster computational cheaper

- **Future Works**

- Multi biometric system by combining scores (normalization of scores since we have different ranges)
- Live Detection



Exploring Facial Recognition Techniques: *Thank you for listening!*

Any questions?