### Project 1.

# Is that you? Metric Learning Approaches for Face Identification

#### Team Ikshana

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#### Introduction

Paper: Is that you? Metric Learning Approaches for Face Identification

By Matthieu Guillaumin, Jakob Verbeek and Cordelia Schmid

The paper, talks about the problem of **face recognition and identification**. That is to determine if two face images depict the same person or not.

**Challenges observed:** This is difficult due to variations in scale, pose, lighting, background, expression, hairstyle, and glasses

**Approach:** Proposed two methods for learning robust distance metrics (Mahalanobis distance) that outperform the various state of art techniques in face Identification which are 1) LDML 2) MkNN

$$d_{\mathbf{M}}(\mathbf{x}_i, \mathbf{x}_j) = (\mathbf{x}_i - \mathbf{x}_j)^{\top} \mathbf{M} (\mathbf{x}_i - \mathbf{x}_j)$$

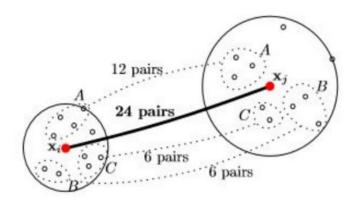
#### Logistic discriminant approach (LDML):

LDML is a logistic discriminant approach which learns the metric from a set of labelled image pairs. Its objective is to make distance between images in positive pairs to be smaller than the distances corresponding to negative pairs.

$$p_n = p(y_i = y_j | \mathbf{x}_i, \mathbf{x}_j; \mathbf{M}, b) = \sigma(b - d_{\mathbf{M}}(\mathbf{x}_i, \mathbf{x}_j)),$$
  
$$\sigma(z) = (1 + \exp(-z))^{-1}$$

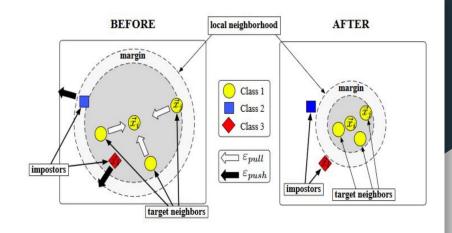
#### Nearest neighbour approach (MkNN):

Computes the probability for two images to belong to the same class. The score of our Marginalized kNN (MkNN) binary classifier for a pair of images  $(x_i, x_j)$  is based on how many positive neighbour pairs we can form from neighbours of  $x_i$  and  $x_j$ .



#### State of Art Model:

Large Margin Nearest Neighbour Metrics (LMNN)



The intuition is that for each data point we need a metric which makes the k nearest neighbours of its own class – target neighbours – closer than points from other classes. The objective has two term, one minimises the distances between target neighbours, while the other is a hinge-loss that encourages target neighbours to be at least one distance unit closer than points from other classes.

## Objectives:

- 1. To implement different methods of finding distance metrics like LMNN, LDML, MKNN based Face Classification on unrestricted data of LFW Dataset.
- We experiment different combinations of these methods for better estimation of results.
- 3. We also experiment the accuracies of Face Identification with different distance metrics for MKNN.

#### LFW Dataset

#### **Labeled Faces in the Wild Dataset**

- 13233 images
- 5749 people
- 1680 people with two or more images





Elisabeth Schumacher, 1 Elisabeth Schumacher, 2





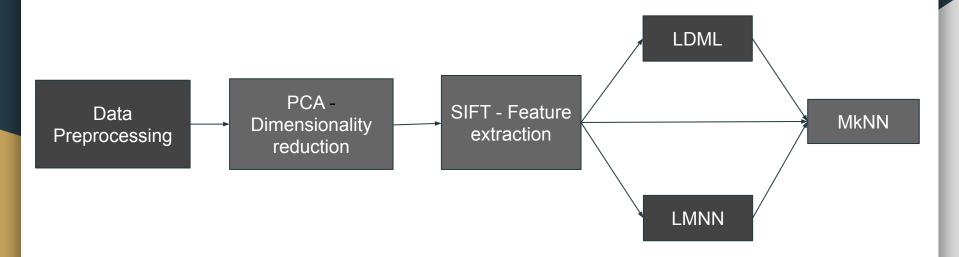
Debra Messing, 1

Debra Messing, 2

The paper runs the experiment in two different settings

- Restricted fixed set of positive and negative image pairs given
- Unrestricted faces labeled by their identity

# Pipeline



## Project Deliverables - Timeline

Mid Evaluation	Feature Extraction from Images using SIFT
	LMNN Implementation
End Evaluation	MkNN Implementation
	Run different experiments to compare between L2, LMNN and different k values
	LDML (if time permits)