

Q6)

What we did in this question was, that we first found our svd from our training data set.

Then we used those parameters on our test model, in order to get the minimum errors of our test dataset.

Then we used those minimum values from our test model, in order to calculate a value of threshold, below which our model would classify an image as recognised, and above which our model would classify an image as not recognisable.

The metric we used was-
threshold= Mean error of our dataset + k*(max error in our test dataset- mean error of our dataset).

(Here we varied the values of k, such that we get the best values of false positive and false negative).

We then found the errors on our test dataset.

Whenever the minimum error for an image was greater than the threshold, we classified it as unrecognisable. But as its part of our test dataset, we know that the image is recognisable. Thus we called it a false negative.

We then found the errors on our unseen images. Whenever the error for an image turned out lesser than the threshold, we classify them as recognisable. But since this is the unseen dataset, we know that these images are unclassifiable. Hence we term all these instances as false positives.

We varied k manually. As k increased, we saw that the amount of false positives increased and false negatives decreased. And vice-versa. As our k varied, we saw that either one of falsepositive or falsenegative became fairly large. For our value of k as 0.55, we see that we get very good values for both falsepositive and falsenegative.


















Thus for k=0.55, we saw that 25 (k2) out of the 128 (k1) images were classified as unrecognised, hence false negatives. Thus our net rate of false negatives, i.e. falsenegative came out to be 0.1953.

We also see for the above k, that 6 (k4) out of the 32 (k3) unseen images were classified as recognisable, hence false positives. Thus rate of false positives was 0.1875.

(Rate of false positives= falsepositive= number of false positives/ total instances tested).

(Rate of false negatives= falsenegative= number of false negatives/ total instances tested).

Note: whatever I have referred to as unseen images means images of people that aren't in the dataset.

	distance	1x192 double
	falsenegative	0.1953
	falsepositive	0.1875
	final_rr	13x1 double
	i	96
	l	14
	image	112x92 double
	image_files	10x1 struct
	image_size	10304
	is_directory	1x50 logical
	j	8
	k	4
	K	1x13 double
	k1	128
	k2	25
	k3	32
	k4	6

Code-

% INITIALIZATION

```
image_size = 112*92;
train_image_array = zeros(image_size,32,6);
test_image_array = zeros(image_size,32,4);
not_in_gallery_image_array = zeros(image_size,8,4);
K = [1 2 3 5 10 15 20 30 50 75 100 150 170];
final_rr = zeros([length(K),1]);
```

%FILE READING

```
directory_name = dir('ORL/');
is_directory = [directory_name.isdir] & ~strcmp({directory_name.name},'.') &
~strcmp({directory_name.name},'..');
person_directory = directory_name(is_directory);

for i= 1:length(person_directory)
    image_files = dir(['ORL/' person_directory(i).name '/*.pgm']);
    for j=1:length(image_files)
        image = double(imread(['ORL/' person_directory(i).name '/' image_files(j).name]));
        if (i > 32) && (j<=4)
            not_in_gallery_image_array(:,i-32,j) = image(:);
        elseif (i <= 32) && (j>6)
            test_image_array(:,i,j-6) = image(:);
        elseif (i <= 32) && (j<=6)
            train_image_array(:,i,j) = image(:);
        end
    end
end
end
```

```
train_image_array = reshape(train_image_array,image_size,[]);
test_image_array = reshape(test_image_array,image_size,[]);
not_in_gallery_image_array = reshape(not_in_gallery_image_array,image_size,[]);
%imshow(reshape(train_image_array(:,1),112,92),[])
```

%MEAN

```
mean_images = mean(train_image_array(:,:), 'all');
train_image_array= train_image_array-mean_images;
```

%SVD

```
[U,S,V] = svd(train_image_array);
```

%TESTING

```
thres= zeros(32*3);
```

```
U_k = U(:,1:50);
```

```
rr =0;
```

```
projected_train_image_array = transpose(U_k)*train_image_array;
```

```
for j= 1:32
```

```
    for k = 1:4
```

```
        test_image = test_image_array(:,32*(k-1)+j)-mean_images;
```

```
        projected_test_vector = transpose(U_k)*test_image(:);
```

```
        distance = vecnorm(projected_train_image_array-projected_test_vector,2,1);
```

```
        [M,l] = min(distance);
```

```

        if mod(l-1,32)+1 == j
            thres((j-1)*3 + k)= M;
        end
    end
end
end

```

```

num=0;
val=0;
max=0;
for i=1:length(thres)
    if thres(i)>0
        num=num+1;
        val=val+thres(i);
        if thres(i)>max
            max= thres(i);
        end
    end
end
end

```

```

meann= val*1.0/num;
threshold= meann + 0.55*(max- meann);

```

```

k1=0;
k2=0;
projected_train_image_array = transpose(U_k)*train_image_array;
for j= 1:32
    for k=1:4
        test_image = test_image_array(:,32*(k-1)+ j)-mean_images;
        projected_test_vector = transpose(U_k)*test_image(:);
        distance = vecnorm(projected_train_image_array-projected_test_vector,2,1);
        [M,l] = min(distance);
        k1= k1+1;
        if M>threshold
            k2= k2+1;
        end
    end
end
end

```

```

falsenegative= k2*1.0/k1;

```

```

k3=0;
k4=0;
projected_train_image_array = transpose(U_k)*train_image_array;

for j= 1:8
    for k = 1:4
        test_image = not_in_gallery_image_array(:,8*(k-1)+j)-mean_images;
        projected_test_vector = transpose(U_k)*test_image(:);
        distance = vecnorm(projected_train_image_array-projected_test_vector,2,1);
        [M,l] = min(distance);
        k3= k3+1;
        if M<threshold
            k4= k4+1;
        end
    end
end
end

falsepositive= k4*1.0/k3;

```