

Question 1 Part A) Implementing PCA

```
load 0_1_2.mat;
```

Centrolize X Below as Xt

```
CMean = mean(X,2);  
ss = ones(1,300);  
Xt = (X - gmultiply(ss, CMean));  
Ccov = (Xt)*(Xt');
```

Singular Decomposition - The first 2 vectors are the largest eigenvectors

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

Project 2 vectors

```
Ypca = U(:,1:2)' * -Xt;
```

Reshape and plot images

```
images = reshape(X,8,8,300);  
plotimages(images,Ypca,0.08,0.25);
```

Q1 Part B Implementing FCA

Define X1, X2 and X3

```
X1 = X(:,1:100);  
X2 = X(:,101:200);  
X3 = X(:,201:300);  
Sw = cov(X1')+cov(X2')+ cov(X3');  
m1 = mean(X1,2);  
m2 = mean(X2,2);  
m3 = mean(X3,2);  
Sb = (m1 - m2)*(m1 - m2)' + (m1 - m3)*(m1 - m3)' + (m2 - m3)*(m2 - m3)';
```

Find Largest Eigenvalue

```
Q = inv(Sw)*Sb;  
[V,D] = eig(Q);  
eigval=diag(D);  
[sort_eigval,sort_eigval_index]=sort(eigval,'descend');  
W=V(:,sort_eigval_index(1:2));
```

Then find y and add 2D presentation

```
y = W'*X;  
images = reshape(X,8,8,300);  
Yfca = [y; randn(1,300)];  
plotimages(images,Yfca,0.005,0.2);
```

Q1 Part c

Elements

```
C1 = Ypca(:,1:100);  
C2 = Ypca(:,101:200);  
C3 = Ypca(:,201:300);  
C1 = C1';  
C2 = C2';  
C3 = C3';  
C = Ypca';  
n1 = 100;  
n2 = 100;  
n3 = 100;  
N = 300;
```

```

mu1=mean(C1);
mu2=mean(C2);
mu3=mean(C3);
mu = mean(C);
D1=C1-repmat(mu1,n1,1);
D2=C2-repmat(mu2,n2,1);
D3=C3-repmat(mu3,n3,1);
D = C - repmat(mu,N,1);
Cov1=D1'*D1;
Cov2=D2'*D2;
Cov3=D3'*D3;
Cov = D'*D;
Cov1 = 1/n1 .* Cov1;
Cov2 = 1/n2 .* Cov2;
Cov3 = 1/n3 .* Cov3;
Cov = 1/N * Cov;
invCov1=inv(Cov1);
invCov2=inv(Cov2);
invCov3=inv(Cov3);
invCov = inv(Cov);
P1=n1/N;
P2=n2/N;
P3=n3/N;

```

LDA

We want to find A, B, where $ATx + B = 0$, $x = [x,y]$

Assume $Cov1 = Cov2 = Cov3 = Cov$

LDAW₀ represents the bias term

LDAW_i is the coefficients of the linear term

```

LDAW10=log(P1)-0.5*mu1*((invCov)*mu1');
LDAW1=(invCov)*mu1';

LDAW20=log(P2)-0.5*mu2*((invCov)*mu2');
LDAW2=(invCov2)*mu2';

LDAW30=log(P3)-0.5*mu3*((invCov)*mu3');
LDAW3=(invCov)*mu3';

```

Below is the functions

```

SW12 = LDAW1 - LDAW2;
SW13 = LDAW1 - LDAW3;
SW23 = LDAW2 - LDAW3;
A = [SW12';SW13';SW23']

```

```

A = 3×2
    4.7368    1.2140
    0.8988    0.7053
   -3.8380   -0.5087

```

```

B = [LDAW10 - LDAW20;LDAW10 - LDAW30;LDAW20 - LDAW30]

```

```

B = 3×1
   -0.0458
   -0.5915
   -0.5456

```

```

LDAS12 = @(x,y) SW12(1)* x + SW12(2)* y + B(1)==0;
LDAS13 = @(x,y) SW13(1)* x + SW13(2)* y + B(2)== 0;
LDAS23 = @(x,y) SSW23(1)* x + SW23(2)* y + B(3)== 0;

LDA = @(x,y) A * [x;y] + B

```

LDA = function_handle with value:

$$@(\mathbf{x}, \mathbf{y}) \mathbf{A} * [\mathbf{x}; \mathbf{y}] + \mathbf{B}$$

QDA

We want to find D, F, C, where $\mathbf{F} \mathbf{T} \mathbf{x} + \mathbf{C}$, $\mathbf{x} = [\mathbf{x}; \mathbf{y}; \mathbf{x}^2; \mathbf{x} * \mathbf{y}; \mathbf{y}^2]$

W_{i0} represents the bias term

W_i is the coefficients of the linear term

WW_i is the quadratic term

```
W10=log(P1)-0.5*mu1*((invCov1)*mu1');
W1=(invCov1)*mu1';
WW1=-0.5 * invCov1;

W20=log(P2)-0.5*mu2*((invCov2)*mu2');
W2=(invCov2)*mu2';
WW2=-0.5 * invCov2;

W30=log(P3)-0.5*mu3*((invCov3)*mu3');
W3=(invCov3)*mu3';
WW3=-0.5 * invCov3;

SWW12 = WW1 - WW2;
SWW13 = WW1 - WW3;
SWW23 = WW2 - WW3;
QSW12 = W1 - W2;
QSW13 = W1 - W3;
QSW23 = W2 - W3;
C = [W10 - W20; W10 - W30; W20 - W30]
```

```
C = 3×1
    -0.0319
    -2.4484
    -2.4165
```

```
F = [QSW12(1), QSW12(2), SWW12(1,1), SWW12(1,2)+SWW12(2,1), SWW12(2,2);
      QSW13(1), QSW13(2), SWW13(1,1), SWW13(1,2)+SWW13(2,1), SWW13(2,2);
      QSW23', SWW23(1,1), SWW23(1,2)+SWW23(2,1), SWW23(2,2)]
```

```
F = 3×5
    7.5636    1.6946    0.0584    0.4560   -0.6589
    3.8912    1.6291    1.5717    0.4821    0.4353
   -3.6724   -0.0654    1.5133    0.0261    1.0942
```

```
D = [SWW12; SWW13; SWW23]
```

```
D = 6×2
    0.0584    0.2280
    0.2280   -0.6589
    1.5717    0.2411
    0.2411    0.4353
    1.5133    0.0131
    0.0131    1.0942
```

```
QDA = @(x,y) F * [x;y;x^2;x*y;y^2] + C
```

```
QDA = function_handle with value:
    @(x,y)F*[x;y;x^2;x*y;y^2]+C
```

Question 1 Part D

```
LDAS12 = @(x,y) SW12(1)* x + SW12(2)* y + B(1);
LDAS13 = @(x,y) SW13(1)* x + SW13(2)* y + B(2);
LDAS23 = @(x,y) SW23(1)* x + SW23(2)* y + B(3);
QDAS12 = @(x,y) QSW12(1) * x + QSW12(2) * y + SWW12(1,1) * x^2 + (SWW12(1,2)+SWW12(2,1))*x*y + SWW
QDAS13 = @(x,y) QSW13(1) * x + QSW13(2) * y + SWW13(1,1) * x^2 + (SWW13(1,2)+SWW13(2,1))*x*y + SWW
QDAS23 = @(x,y) QSW23(1) * x + QSW23(2) * y + SWW23(1,1) * x^2 + (SWW23(1,2)+SWW23(2,1))*x*y + SWW
```

```

plotimages(images,Ypca,0.03,0.25);
hold on
fimplicit(LDAS12,[-2.5 2.5 -2.5 2]);
hold on
fimplicit(LDAS13,[-2.5 2.5 -2.5 2]);
hold on
fimplicit(LDAS23,[-2.5 2.5 -2.5 2]);
hold on
fimplicit(QDAS12,[-2.5 2.5 -2.5 2]);

```

Warning: Function behaves unexpectedly on array inputs. To improve performance, properly vectorize your function to return an output with the

```

hold on
fimplicit(QDAS13,[-2.5 2.5 -2.5 2]);

```

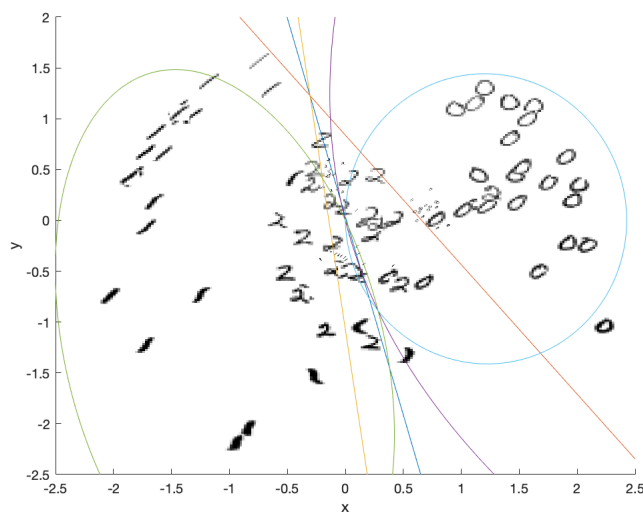
Warning: Function behaves unexpectedly on array inputs. To improve performance, properly vectorize your function to return an output with the

```

hold on
fimplicit(QDAS23,[-2.5 2.5 -2.5 2]);

```

Warning: Function behaves unexpectedly on array inputs. To improve performance, properly vectorize your function to return an output with the



Question 1 Part e

apply LDA to Ypca

```

Class12 = bsxfun(LDAS12,Ypca(1,:),Ypca(2,:));
Class13 = bsxfun(LDAS13,Ypca(1,:),Ypca(2,:));
Class23 = bsxfun(LDAS23,Ypca(1,:),Ypca(2,:));
x1 = Class12(1,1:100);
countC1 = length(nonzeros(x1(x1>0)));
x2 = Class23(1,101:200);
countC2 = length(nonzeros(x2(x2>0)));
x3 = -Class13(1,201:300);
countC3 = length(nonzeros(x3(x3>0)));
Correct = countC1 +countC2+countC3;
Error_rate = 1- Correct / 300

```

Error_rate = 0.0600

The Error Rate is 6% using LDA