

Lecture 13

Principal Component Analysis

```
N = 100;
```

```
% Generate data entries for Class 1  
m1 = [3, 3]'; % Mean vector  
cov1 = [2 1; 1 2]; % Covariance matrix  
rng default % For reproducibility  
r1 = mvnrnd(m1,cov1,N);
```

```
data_C1 = zeros(N, 2);  
data_C1 = r1;
```

```
% Generate data entries for Class 2  
m2 = [9, 9]';  
cov2 = [2,1; 1,2];  
rng default % For reproducibility  
r2 = mvnrnd(m2,cov2,N);
```

```
data_C2 = zeros(N, 2);  
data_C2 = r2;
```

```
% Combine data of two classes  
X = vertcat (data_C1, data_C2);
```

```
>> var(X1)
```

```
ans =
```

```
11.7340
```

```
>> X2 = X(:,2);
```

```
>> var(X2)
```

```
ans =
```

```
11.3768
```

```
>> whos X
```

Name	Size	Bytes	Class	Attributes
------	------	-------	-------	------------

X	200x2	3200	double	
---	-------	------	--------	--

```
>> X(1,:)
```

```
ans =
```

```
3.7604 4.4094
```

```
>> X(2,:)
```

```
ans =
```

```
5.5935 3.2091
```

```
>> var(r1)
```

```
ans =
```

```
2.7023 2.3434
```

```
>> X1 = X(:,1);
```

```
>> whos X1
```

Name	Size	Bytes	Class	Attributes
------	------	-------	-------	------------

X1	200x1	1600	double	
----	-------	------	--------	--

```
>> cov(X)
```

```
ans =
```

```
11.7340 10.5415  
10.5415 11.3768
```

Cov(X₁, X₂)

Cov(X₂, X₁)

C - Covariance Matrix

$$C = C^T$$

$$C V_1 = \lambda_1 V_1$$

$(n \times n)$ $(n \times 1)$ λ_1 $(n \times 1)$
 Scalar \leftarrow eigen value

$$C V_2 = \lambda_2 V_2$$

$$V_1^T V_2 = 0 \Rightarrow V_1^T (2V_2) = 0$$

```
>> mu = mean(X);
>> X_mr = X - mu;
>> C = cov(X_mr)
```

C =

```
11.7340 10.5415
10.5415 11.3768
```

```
>> [V,D] = eig(C);
>> V
```

V =

```
0.7011 -0.7131
-0.7131 -0.7011
```

```
>> D
```

D =

```
1.0124 0
0 22.0984
```

```
>> V1 = V(1,:)
```

V1 =

```
0.7011 -0.7131
```

```
>> norm(V1)
```

ans =

```
1.0000
```

```
>> norm(V2)
```

ans =

```
1.0000
```

```
>> V2 = V(2,:)
```

V2 =

```
-0.7131
-0.7011
```

```
>> C*V2
```

ans =

```
-15.7577
-15.4930
```

```
>> D(2,2)*V2
```

ans =

```
-15.7577
-15.4930
```

```
>> V1 = V1'
```

V1 =

```
0.7011
-0.7131
```

```
>> V2
```

V2 =

```
-0.7131
-0.7011
```

```
>> V1'*V2
```

ans =

```
0
```

```
>> C*V1'
```

ans =

```
0.7098
-0.7219
```

```
>> D(1,1)*V1'
```

ans =

```
0.7098
-0.7219
```

SVD

$$X = ASB^T \Rightarrow X^T = (ASB^T)^T = (B^T)^T S^T A^T = BS^T A^T$$

$$X^T X = BS^T A^T \cdot ASB^T = BS^T \underbrace{A^T A}_I \underbrace{S}_\Sigma B^T = B \Sigma B^T \dots (1)$$

where $A^T \cdot A = I$

and $B^T \cdot B = I$

Σ : diagonal matrix

>> S(1:4, 1:2)

ans =

```
138.6237    0
    0 14.1984
    0    0
    0    0
    0    0
    ...
```

$$X = A \cdot S \cdot B^T$$

$N \times m$ $(N \times N)$ $(N \times m)$ $(m \times m)$, $S^T \cdot S = \Sigma$
 200×2 200×2 2×2 2×2
 $B: (m \times m)$

From Eq. (1)

>> B'*B

$$(X^T X) B = B \Sigma \underbrace{B^T \cdot B}_I = B \Sigma$$

$(m \times N)$ $(N \times m)$ $(m \times m)$ I

ans =

```
1.0000  0.0000
0.0000  1.0000
```

B: eigenvector matrix of $X^T X$

>> X'*X*B

>> sigma = S'*S

>> B*sigma

ans =

sigma =

ans =

```
1.0e+04 *
```

```
1.0e+04 *
```

```
1.0e+04 *
```

```
-1.3766 -0.0141
-1.3408  0.0144
```

```
1.9217    0
    0 0.0202
```

```
-1.3766 -0.0141
-1.3408  0.0144
```