

## 1. Gaussian Elimination

$$X+2y-z=3, 2x+y-2z=3, -3x+y+z=-6$$

```
C = [1 2 -1; 2 1 -2; -3 1 1]
```

```
b= [3 3 -6]'
```

```
A = [C b];
```

```
n= size(A,1);
```

```
x = zeros(n,1); %variable matrix [x1 x2
```

```
... xn] column
```

```
for i=1:n-1
```

```
for j=i+1:n
```

```
m = A(j,i)/A(i,i)
```

```
A(j,:) = A(j,:) - m*A(i,:)
```

```
end
```

```
end
```

```
x(n) = A(n,n+1)/A(n,n)
```

```
for i=n-1:-1:1
```

```
summ = 0
```

```
for j=i+1:n
```

```
summ = summ + A(i,j)*x(j,:)
```

```
x(i,:) = (A(i,n+1) - summ)/A(i,i)
```

```
end
```

```
end
```

Output:

x = 3, y = 1, z = 2

## 2. GAUSS JORDAN METHOD

```
A=[1,1,1;4,3,-1;3,5,3];
```

```
n=length(A(1,:));
```

```
Aug =[A,eye(n,n)]
```

```
for j=1:n-1
```

```
for i=j+1:n
```

```
Aug(i,j:2*n)=Aug(i,j:2*n)-Aug(i,j)/Aug(j,j)*Aug(j,j:2*n)
```

```
end
```

```
end
```

```
for j=n:-1:2
```

```
Aug(1:j-1,:)=Aug(1:j-1,:)-Aug(1:j-1,j)/Aug(j,j)*Aug(j,:)
```

```
end
```

```
for j=1:n
```

```
Aug(j,:)=Aug(j,:)/Aug(j,j)
```

```
end
```

```
B=Aug(:,n+1:2*n)
```

### **3.LU DECOMPOSITION**

```
%LU Decomposition
Ab = [1 1 1;1 2 2;1 2 3];
%% Forward Elimination
n= length(A);
L = eye(n);
% With A(1,1) as pivot Element
for i =2:3
    alpha = Ab(i,1)/Ab(1,1);
    L(i,1) = alpha;
    Ab(i,:) = Ab(i,:) - alpha*Ab(1,:);
end
% With A(2,2) as pivot Element
i=3;
alpha = Ab(i,2)/Ab(2,2);
L(i,2) = alpha;
Ab(i,:) = Ab(i,:) - alpha*Ab(2,:);
U = Ab(1:n,1:n)
```

### **4.FOUR FUNDAMENTAL SUBSPACES**

```
Clc;
clear all;
close all;
% Bases of four fundamental vector spaces of matrix A.
A=[1,2,3;2,-1,1];
% Row Reduced Echelon Form
[R, pivot] = rref(A)
% Rank
rank = length(pivot)
% basis of the column space of A
columnsp = A(:,pivot)
% basis of the nullspace of A
nullsp = null(A,'r')
% basis of the row space of A
rowsp = R(1:rank,:)'
% basis of the left nullspace of A
leftnullsp = null(A','r')
```

### **5.Gram Schmidt orthogonalization**

```
A=[0,1,1;1,1,0;1,-1,2;1,0,-1]
Q=zeros(4,3)
R=zeros(3)
for j=1:3
    v=A(:, j);
    for i=1:j-1
        R(i,j)=Q(:,i)'\*A(:,j)
        v=v-R(i,j)*Q(:,i)
    end
    R(j,j)=norm(v)
    Q(:,j)=v/R(j,j) end
```

## 6.Projection by least squares

a)  $A=[1,0;0,1;1,1]$

$b=[1;3;4]$

$x = \text{lsqr}(A,b)$

b) Find the point on the plane  $x+y-z=0$  that is closest to  $(2,1,0)$

`syms c`

$P=[2,1,0]+c*[1,1,-1]$

$s=1*(c+2)+1*(c+1)-1*(-c)==0$

$s1=\text{solve}(s,c)$

$p=[2,1,0]+s1*[1,1,-1]$

c) Find the point on the plane  $3x+4y+z=1$  that is closest to  $(1,0,1)$

`syms c`

$P=[1,0,1]+c*[3,4,1]$

$s=3*(1+3*c)+4*(4*c)+(1+c)==1$

$s1=\text{solve}(s,c)$

$p=[1,0,1]+s1*[3,4,1]$

d) Let onto and find P, the matrix that will project any matrix onto the vector v. Use the result to find projv u. Code:

$u=[1;7]$

$v=[-4;2]$

$P=(v*\text{transpose}(v))/(v*\text{transpose}(v))$

$P*u$

e) Projecting a lot of vectors on a single vector:

Code:

$u=8*\text{rand}(2,100)-4;$

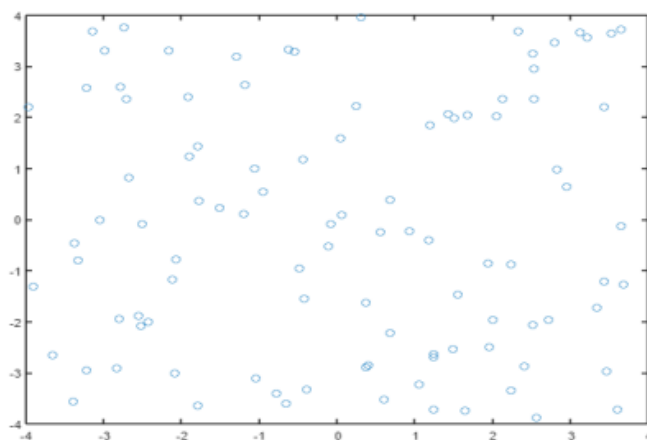
$x=u(1,:)$

$y=u(2,:)$

`plot(x,y,'o')`

In the below figure I have generated a 100 random vectors.

In this figure each circle represents the tip of a vector whose tail begins at the origin.

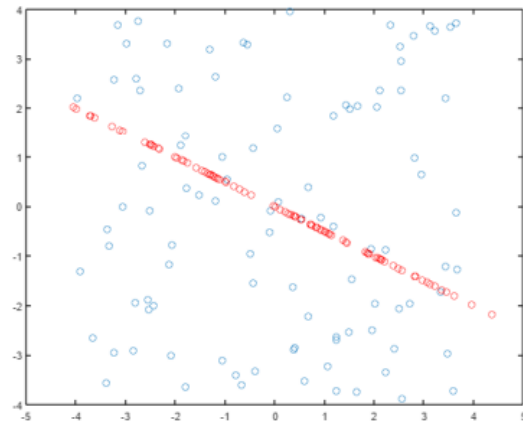


Next , I will take the projection matrix P to project each of the 100 2 by 1 vectors in matrix U onto the vector v, Projection matrices and least squares: Code:

$P=[0.8,-0.4;-0.4,0.2]$

$Pu=P*u;$

```
x=Pu(1,:)
y=Pu(2,:)
hold on
plot(x,y,'o')
```



Here each vector in the matrix u is projected onto a line in the direction of the vector  $v = [-1; 2]$

## **7.QR FACTORIZATION**

a) Find QR factorization of the matrix

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

```
A=[1,1,0;1,0,1;0,1,1]
```

```
[Q,R]=qr(A)
```

b)QR Factorization of Pascal Matrix

```
A = sym(pascal(3))
```

```
[Q,R] = qr(A)
```

```
isAlways(A == Q*R)
```

c) QR Decomposition to Solve Matrix Equation of the form  $Ax=b$

```
A = sym(invhilb(5))
```

```
b = sym([1:5])
```

```
[C,R] = qr(A,b);
```

```
X = R\C
```

```
isAlways(A*X == b)
```

## **8.Eigenvalues and Eigenvectors**

a).computing detmrinant and trace

```
A=[1,1,3;1,5,1;3,1,1]
```

```
e=eig(A)
```

```

det(A)
prod(eig(A))
det(A)=prod(eig(A))
sum(eig(A))
trace(A)
[V,D]=eig(A)
A*V-V*D

```

#### b).Computing Eigenvalues and Eigenvectors

```

A=[2,2,1;1,3,1;1,2,2]
e=eig(A)
[V,D]=eig(A)

```

$A*V(:,1)$       % This command will gives you First eigenvectore

$D(1,1)*V(:,1)$     % First eigenvalue

If you replace 2 or 3 in place of 1, that will create second or third eigenvalue and eigenvectors respectively.

#### c)For similar Matrices

```

A=[2,2,1;1,3,1;1,2,2]
B=[1,-1,1;1,0,0;-1,1,-1]
e=eig(A,B)
[V,D]=eig(A,B)
[V,D,W]=eig(A,B)

```

$e = \text{eig}(A,B)$  returns a column vector containing the generalized eigenvalues of square matrices A and B.

$[V,D] = \text{eig}(A,B)$  returns diagonal matrix D of generalized eigenvalues and full matrix V whose columns are the corresponding right eigenvectors,

so that  $A*V = B*V*D$ .

$[V,D,W] = \text{eig}(A,B)$  also returns full matrix W whose columns are the

corresponding left eigenvectors, so that  $W'*A = D*W'*B$ .

The generalized eigenvalue problem is to determine the solution to the equation  $Av = \lambda Bv$ , where  $A$  and  $B$  are  $n$ -by- $n$  matrices,  $v$  is a column vector of length  $n$ , and  $\lambda$  is a scalar.

The values of  $\lambda$  that satisfy the equation are the generalized eigenvalues.

The corresponding values of  $v$  are the generalized right eigenvectors.

The left eigenvectors,  $w$ , satisfy the equation  $w'A = \lambda w'B$ .