

assignment-7

October 31, 2024

```
[1]: import numpy as np
```

```
# Given matrix A  
A = np.array([  
    [4, -2, 2, 1],  
    [1, 1, 0, 1],  
    [-2, 1, 3, -1],  
    [1, 3, -1, 2]  
])
```

```
[3]: A_inv = np.linalg.inv(A)  
print("Inverse of A:\n", A_inv)
```

Inverse of A:

```
[[ 5.00000000e+00 -3.00000000e+01  1.00000000e+00  1.30000000e+01]  
 [ 4.00000000e+00 -2.50000000e+01  1.00000000e+00  1.10000000e+01]  
 [-1.00000000e+00  7.00000000e+00  5.82867088e-16 -3.00000000e+00]  
 [-9.00000000e+00  5.60000000e+01 -2.00000000e+00 -2.40000000e+01]]
```

```
[5]: eigenvalues, eigenvectors = np.linalg.eig(A)  
print("\nEigenvalues:\n", eigenvalues)  
print("\nEigenvectors:\n", eigenvectors)
```

Eigenvalues:

```
[-0.02231724+0.j          3.6104532 +1.72034086j   3.6104532 -1.72034086j  
 2.80141085+0.j          ]
```

Eigenvectors:

```
[[ -0.43676159+0.j          0.57598271+0.j          0.57598271-0.j  
   0.26748859+0.j          ]  
 [ -0.36900111+0.j          0.00962337-0.2105225j   0.00962337+0.2105225j  
  -0.34893941+0.j          ]  
 [  0.10239683+0.j          -0.00821729+0.55142262j  -0.00821729-0.55142262j  
  -0.06120796+0.j          ]  
 [  0.81399778+0.j          -0.18869091-0.53300366j  -0.18869091+0.53300366j  
  -0.89607183+0.j          ]]
```

```
[7]: P = eigenvectors
D = np.diag(eigenvalues)
P_inv = np.linalg.inv(P)

print("\nMatrix P (Eigenvectors):\n", P)
print("\nDiagonal Matrix D (Eigenvalues):\n", D)
print("\nInverse of Matrix P:\n", P_inv)
```

Matrix P (Eigenvectors):

```
[[ -0.43676159+0.j          0.57598271+0.j          0.57598271-0.j
    0.26748859+0.j          ]
 [ -0.36900111+0.j          0.00962337-0.2105225j    0.00962337+0.2105225j
   -0.34893941+0.j          ]
 [ 0.10239683+0.j          -0.00821729+0.55142262j   -0.00821729-0.55142262j
   -0.06120796+0.j          ]
 [ 0.81399778+0.j          -0.18869091-0.53300366j   -0.18869091+0.53300366j
   -0.89607183+0.j          ]]
```

Diagonal Matrix D (Eigenvalues):

```
[[ -0.02231724+0.j          0.          +0.j          0.          +0.j
    0.          +0.j          ]
 [ 0.          +0.j          3.6104532 +1.72034086j    0.          +0.j
    0.          +0.j          ]
 [ 0.          +0.j          0.          +0.j          3.6104532 -1.72034086j
    0.          +0.j          ]
 [ 0.          +0.j          0.          +0.j          0.          +0.j
    2.80141085+0.j          ]]
```

Inverse of Matrix P:

```
[[ 0.24293955+0.00000000e+00j -1.52243125+6.81957034e-17j
    0.05807612+1.36391407e-16j  0.66140356-6.81957034e-17j]
 [ 1.00207027+1.76334426e-02j -0.26464982-6.27009101e-02j
    0.2698955 -8.46128808e-01j  0.38375201+8.74766701e-02j]
 [ 1.00207027-1.76334426e-02j -0.26464982+6.27009101e-02j
    0.2698955 +8.46128808e-01j  0.38375201-8.74766701e-02j]
 [-0.18035771-0.00000000e+00j -1.34612117-0.00000000e+00j
   -1.06750294+2.75134487e-17j -0.57270992-7.47247360e-17j]]
```

0.0.1 Explanation of Diagonalization Importance

In linear algebra, diagonalizing a matrix simplifies many computations, especially when raising a matrix to a power. A diagonalized matrix represents the system in a simpler form, where each eigenvalue indicates an invariant direction of transformation. This is valuable for understanding the matrix's stability and the long-term behavior of systems.

Diagonalization also reveals insights into the system's properties, such as:

- **Stability:** Eigenvalues can indicate stability in dynamical systems.

- **Matrix powers:** Calculating powers of a diagonal matrix is straightforward, which is useful in various applications.

Thus, diagonalization is a powerful tool in linear algebra and applicable across fields like physics, engineering, and machine learning.

```
[14]: U, sigma, Vt = np.linalg.svd(A)
      Sigma = np.zeros((A.shape[0], A.shape[1]))
      np.fill_diagonal(Sigma, sigma)

      A_reconstructed = U @ Sigma @ Vt
      print("\nReconstructed Matrix A (using SVD):\n", A_reconstructed)
```

Reconstructed Matrix A (using SVD):

```
[[ 4.00000000e+00 -2.00000000e+00  2.00000000e+00  1.00000000e+00]
 [ 1.00000000e+00  1.00000000e+00 -7.77156117e-16  1.00000000e+00]
 [-2.00000000e+00  1.00000000e+00  3.00000000e+00 -1.00000000e+00]
 [ 1.00000000e+00  3.00000000e+00 -1.00000000e+00  2.00000000e+00]]
```

```
[ ]:
```