

Assignment 1 Basic Linear Algebra in Python

Use Numpy library functions and methods to complete given tasks.

Given: $x \in \mathbb{R}^n$; $A \in \mathbb{R}^{n \times n}$; $B \in \mathbb{R}^{n \times n}$

Read dimension of all variables and their values from keyboard.

Find and Display:

- A^T and B^T
- $A.B$
- Verify $(AB)^T = B^T . A^T$
- $|A|$ and $|B|$
- Check for singularity of A and B
- Find A^{-1} and B^{-1} (if possible) (Read again values of A and B if inverses are not possible and do it till inverses are available)
- Verify $AA^{-1} = I$ and $A^{-1}A = I$
- Trace of A and B, A^{-1} and B^{-1} , A^T and B^T
- Verify trace $(AB) = \text{trace}(BA)$
- Verify $\text{trace}(A^TB) = \text{trace}(AB^T) = \text{trace}(BA^T) = \text{trace}(B^TA)$
- Eigen values and Eigen vectors of A, B, AB, A^T , B^T , A^{-1} , B^{-1}
- Verify trace and determinant property related to Eigen Values for both Matrix A and B
- Check definiteness of Matrices A, B, AB, A^T , B^T , A^{-1} , B^{-1}
- $y = Ax$; $y \in \mathbb{R}^n$
- Retrieve x from y
- Inner product of x and y and verify they are orthogonal or not
- L_2 -Norm of x and y
- Normalize x and y
- Verify CS inequality for x and y
- Verify $x^Ty = y^Tx$
- Verify $y^TAx = x^TA^Ty$

Import Numpy library

In [1]: `import numpy as np`

Take Input and Verify Singularity of Matrix

```
In [2]: n = int(input("Enter dimension (n) for matrix A and B (n x n) and vector x (n x 1): "))

x = list(map(int,input(f"Enter Vector x of size {n}. x 1 in a single line row wise elements: "))
x = np.matrix(x).reshape(n,1)
print("\nVector x:\n", x)

while (True):
    A = list(map(int,input(f"Enter matrix A of size {n}. x {n}. in a single line row wise elements: "))
    A = np.array(A).reshape(n,n)
    print("\nMatrix A:\n", A)
    # |A| and singularity check
    det_A = np.linalg.det(A)
    print(f"\nDeterminant of A: {det_A}")
    if np.isclose(det_A, 0):
        print("\nA is singular, Reading again.")
        continue
    else:
        print("\nA is non-singular")
        break

while (True):
    B = list(map(int,input(f"Enter matrix B of size {n}. x {n}. in a single line row wise elements: "))
    B = np.array(B).reshape(n,n)
    print("\nMatrix B:\n", B)
    # |B| and singularity check
    det_B = np.linalg.det(B)
    print(f"\nDeterminant of B: {det_B}")
    if np.isclose(det_B, 0):
        print("\nB is singular, Reading again.")
        continue
    else:
        print("\nB is non-singular")
        break
```

Vector x:

```
[[1]
 [2]]
```

Matrix A:

```
[[1 2]
 [3 4]]
```

Determinant of A: -2.0000000000000004

A is non-singular

Matrix B:

```
[[5 6]
 [7 8]]
```

Determinant of B: -2.0000000000000005

B is non-singular

Transpose of Matrix

```
In [3]: # Transpose
print("\nTranspose of A:\n", A.T)
print("\nTranspose of B:\n", B.T)
```

Transpose of A:
[[1 3]
[2 4]]

Transpose of B:
[[5 7]
[6 8]]

Matrix Multiplication

```
In [4]: # Matrix multiplication
AB = A @ B
print("\nMatrix AB:\n", AB)
print("\nMatrix AB using numpy:\n", np.dot(A,B))
```

Matrix AB:
[[19 22]
[43 50]]

Matrix AB using numpy:
[[19 22]
[43 50]]

Verify $(AB)^T = B^T \cdot A^T$

```
In [5]: print("\nVerify (AB)^T = B^T A^T:", np.allclose(AB.T, B.T @ A.T))
```

Verify $(AB)^T = B^T A^T$: True

Finding Inverse Of Matrix

```
In [6]: # Inverse
A_inv = np.linalg.inv(A)
print("\nInverse of A:\n", A_inv)
B_inv = np.linalg.inv(B)
print("\nInverse of B:\n", B_inv)
```

Inverse of A:
[[-2. 1.]
[1.5 -0.5]]

Inverse of B:
[[-4. 3.]
[3.5 -2.5]]

Verify $AA^{-1} = I$ and $A^{-1}A = I$

```
In [7]: print("Verify AA^-1 = I:", np.allclose(A @ A_inv, np.eye(n)))
print("Verify A^-1A = I:", np.allclose(A_inv @ A, np.eye(n)))
```

Verify $AA^{-1} = I$: True
Verify $A^{-1}A = I$: True

Finding Trace of Matrices

```
In [8]: # Trace
print(f"\nTrace of A: {np.trace(A)}")
print(f"Trace of B: {np.trace(B)}")
print(f"Trace of A^-1: {np.trace(A_inv)}")
print(f"Trace of B^-1: {np.trace(B_inv)}")
print(f"Trace of A^T: {np.trace(A.T)}")
print(f"Trace of B^T: {np.trace(B.T)}")
```

```
Trace of A: 5
Trace of B: 13
Trace of A^-1: -2.4999999999999996
Trace of B^-1: -6.4999999999999982
Trace of A^T: 5
Trace of B^T: 13
```

Trace Verification

```
In [9]: # Trace verification
print(f"\nVerify trace(AB) = trace(BA): {np.trace(A @ B) == np.trace(B @ A)}")
print("\nVerify trace(A^T.B) = trace(A.B^T) = trace(B.A^T) = trace(B^T.A):",
      np.trace(A.T @ B) == np.trace(A @ B.T) == np.trace(B @ A.T) == np.trace(B.T @ A))
```

```
Verify trace(AB) = trace(BA): True
```

```
Verify trace(A^T.B) = trace(A.B^T) = trace(B.A^T) = trace(B^T.A): True
```

Finding Eigenvalues and Eigenvectors

```
In [10]: np.linalg.eig(A)
```

```
Out[10]: EigResult(eigenvalues=array([-0.37228132,  5.37228132]), eigenvectors=array([[ -0.824564
 84, -0.41597356],
 [ 0.56576746, -0.90937671]]))
```

```
In [11]: # Eigenvalues and Eigenvectors
eigvals_A, eigvecs_A = np.linalg.eig(A)
eigvals_B, eigvecs_B = np.linalg.eig(B)
print("\nEigenvalues of A:\n", eigvals_A)
print("Eigenvectors of A:\n", eigvecs_A)
print("\nEigenvalues of B:\n", eigvals_B)
print("Eigenvectors of B:\n", eigvecs_B)
eigvals_AT, eigvecs_AT = np.linalg.eig(A.T)
eigvals_BT, eigvecs_BT = np.linalg.eig(B.T)
print("\nEigenvalues of AT:\n", eigvals_AT)
print("Eigenvectors of AT:\n", eigvecs_AT)
print("\nEigenvalues of BT:\n", eigvals_BT)
print("Eigenvectors of BT:\n", eigvecs_BT)
eigvals_AB, eigvecs_AB = np.linalg.eig(AB)
print("\nEigenvalues of AB:\n", eigvals_AB)
print("Eigenvectors of AB:\n", eigvecs_AB)
eigvals_A_1, eigvecs_A_1 = np.linalg.eig(A_inv)
eigvals_B_1, eigvecs_B_1 = np.linalg.eig(B_inv)
print("\nEigenvalues of A^-1:\n", eigvals_A_1)
print("Eigenvectors of A^-1:\n", eigvecs_A_1)
print("\nEigenvalues of B^-1:\n", eigvals_B_1)
print("Eigenvectors of B^-1:\n", eigvecs_B_1)
```

Eigenvalues of A:

```
[-0.37228132  5.37228132]
```

Eigenvectors of A:

```
[[ -0.82456484 -0.41597356]
 [  0.56576746 -0.90937671]]
```

Eigenvalues of B:

```
[-0.15206735 13.15206735]
```

Eigenvectors of B:

```
[[ -0.75868086 -0.59276441]
 [  0.65146248 -0.80537591]]
```

Eigenvalues of AT:

```
[-0.37228132  5.37228132]
```

Eigenvectors of AT:

```
[[ -0.90937671 -0.56576746]
 [  0.41597356 -0.82456484]]
```

Eigenvalues of BT:

```
[-0.15206735 13.15206735]
```

Eigenvectors of BT:

```
[[ -0.80537591 -0.65146248]
 [  0.59276441 -0.75868086]]
```

Eigenvalues of AB:

```
[5.80198014e-02 6.89419802e+01]
```

Eigenvectors of AB:

```
[[ -0.75781077 -0.40313049]
 [  0.65247439 -0.91514251]]
```

Eigenvalues of A⁻¹:

```
[-2.68614066  0.18614066]
```

Eigenvectors of A⁻¹:

```
[[ -0.82456484 -0.41597356]
 [  0.56576746 -0.90937671]]
```

Eigenvalues of B⁻¹:

```
[-6.57603367  0.07603367]
```

Eigenvectors of B⁻¹:

```
[[ -0.75868086 -0.59276441]
 [  0.65146248 -0.80537591]]
```

Verify trace and determinant properties of Eigen values for A and B

```
In [12]: print("\nVerify det(A) = Multiplication of eigvals_A:", np.isclose(det_A, np.prod(eigva
print("\nVerify det(B) = Multiplication of eigvals_B:", np.isclose(det_B, np.prod(eigva
print("\nVerify trace(A) = Summation of eigvals_A:", np.isclose(np.trace(A), np.sum(eig
print("\nVerify trace(B) = Summation of eigvals_B:", np.isclose(np.trace(B), np.sum(eig
```

Verify det(A) = Multiplication of eigvals_A: True

Verify det(B) = Multiplication of eigvals_B: True

Verify trace(A) = Summation of eigvals_A: True

Verify trace(B) = Summation of eigvals_B: True

Definiteness Test for A and B (In general not symmetric)

```
In [13]: Asym = A+A.T
Bs = B+B.T
eigvals_Asym, eigvecs_Asym = np.linalg.eig(Asym)
```

```

eigvals_Bsym, eigvecs_Bsym = np.linalg.eig(Bsym)
if(np.all(eigvals_Asym > np.finfo(np.float32).eps)):
    print("\nA is Positive definite")
elif(np.all(eigvals_Asym >= 0)):
    print("\nA is Positive semi-definite")
elif(np.all(eigvals_Asym < 0)):
    print("\nA is Negative semi-definite")
elif(np.all(eigvals_Asym <= 0)):
    print("\nA is Negative semi-definite")
else:
    print("\nA is indefinite")

if(np.all(eigvals_Bsym > 0)):
    print("\nB is Positive definite")
elif(np.all(eigvals_Bsym >= 0)):
    print("\nB is Positive semi-definite")
elif(np.all(eigvals_Bsym < 0)):
    print("\nB is Negative semi-definite")
elif(np.all(eigvals_Bsym <= 0)):
    print("\nB is Negative semi-definite")
else:
    print("\nB is indefinite")

```

A is indefinite

B is indefinite

Vector Operations

In [14]: A

Out[14]: array([[1, 2],
[3, 4]])

In [15]: x

Out[15]: matrix([[1],
[2]])

$$y = Ax; y \in \mathbb{R}^n$$

```

In [16]: # Vector operations
y = A @ x
print("\ny = Ax:\n", y)

```

```

y = Ax:
[[ 5]
 [11]]

```

Retrieved x from y

In [17]: np.linalg.solve(A,y)

Out[17]: matrix([[1.],
[2.]])

```

In [18]: x_retrieved = A_inv @ y
print("Retrieved x from y:\n", x_retrieved)

```

```

Retrieved x from y:
[[1.]
 [2.]]

```

Inner product, norm, and orthogonality

```
In [19]: inner_product = x.T @ y
print(f"\nInner product of x and y: {inner_product}")
print("x and y are orthogonal:" if np.isclose(inner_product, 0) else "x and y are not o
print(f"\nNorm of x: {np.linalg.norm(x)}")
print(f"Norm of y: {np.linalg.norm(y)}") #L2-norm by default
```

Inner product of x and y: [[27]]

x and y are not orthogonal.

Norm of x: 2.23606797749979

Norm of y: 12.083045973594572

L₁-Norm of x and y

```
In [20]: print(f"\nNorm of x: {np.linalg.norm(x,ord=1)}")
print(f"Norm of y: {np.linalg.norm(y,ord=1)}")
```

Norm of x: 3.0

Norm of y: 16.0

Normalizing Vector x & y

```
In [21]: # Normalize
x_normalized = x / np.linalg.norm(x)
y_normalized = y / np.linalg.norm(y)
print("\nNormalized x:\n", x_normalized)
print("Normalized y:\n", y_normalized)
```

Normalized x:

```
[[0.4472136 ]
 [0.89442719]]
```

Normalized y:

```
[[0.41380294]
 [0.91036648]]
```

Verifying CS(Cauchy-Schwarz\Cauchy-Bunyakovsky-Schwarz inequality) inequality i.e. $|\langle a, b \rangle| \leq \|a\| \cdot \|b\|$

```
In [22]: print("\nVerify CS inequality for x and y:",
inner_product <= np.linalg.norm(x) * np.linalg.norm(y))
if abs(inner_product) <= np.linalg.norm(x) * np.linalg.norm(y) :
    print("CS inequality ( $|\langle a, b \rangle| \leq \|a\| \cdot \|b\|$ ) verified")
else :
    print("CS inequality not verified")
```

Verify CS inequality for x and y: [[True]]

CS inequality ($|\langle a, b \rangle| \leq \|a\| \cdot \|b\|$) verified

Additional verifications

```
In [23]: print("\nVerify  $x^T y = y^T x$ :", np.isclose(inner_product, y.T @ x))
print("\nVerify  $y^T A x = x^T A^T y$ :",
np.isclose((y.T @ A) @ x, (x.T @ A.T) @ y))
```

Verify $x^T y = y^T x$: [[True]]

Verify $y^T A x = x^T A^T y$: [[True]]