

# Linear Algebra for Machine Learning

Reference: Chapter 2 (Linear Algebra) of the [Deep Learning Book](#) by Aaron Courville, Ian Goodfellow, and Yoshua Bengio (Attached)

Find the necessary files here > [CSE 472 Assignment 1 Files](#)

## Task 1: Matrix Transformation

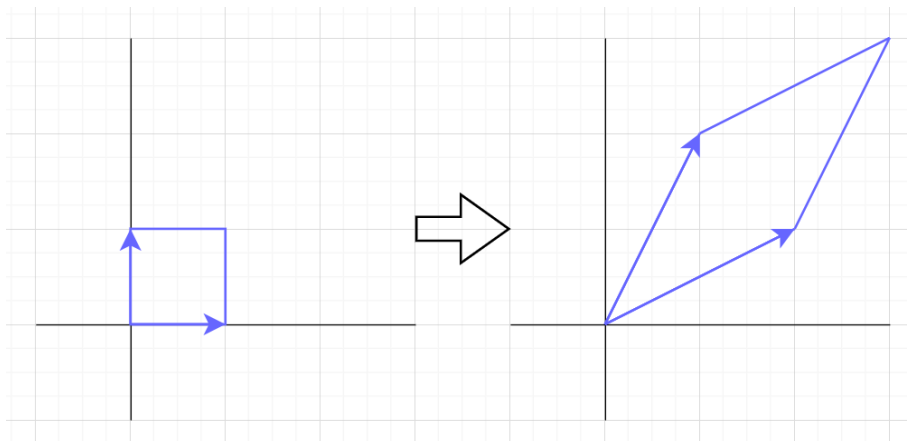
Go through and run the notebook [“matrix-transformations-and-eigen-decomposition”](#) to get an intuition about

- How a [matrix](#) can transform [a vector](#)?
- What do [columns of matrices](#) mean in [terms of transformation](#)?
- What does [eigenvector](#) mean?

(We recommend you also read the whole of Chapter 2 of the Deep Learning Book.)

Then,

- Change the [cell values of matrix  \$M\$](#)  so that it does the [following shear transformation](#)



- Run the whole notebook [again and submit](#)

## Task 2: Eigen Decomposition

### SubTask 2A: Random Matrix (random\_eigen.py)

- Take the dimensions of matrix  $n$  as input.
- Produce a random  $n \times n$  invertible matrix  $A$ . For the purpose of demonstrating, every cell of  $A$  will be an integer.
- Perform Eigen Decomposition using NumPy's library function
- Reconstruct  $A$  from eigenvalue and eigenvectors (Refer to Section 2.7).
- Check if the reconstruction is perfect. (`np.allclose` will come in handy)

## SubTask 2B: Symmetric Matrix (symmetric\_eigen.py)

- Take the dimensions of matrix  $n$  as input.
- Produce a random  $n \times n$  invertible symmetric matrix  $A$ . For the purpose of demonstrating, every cell of  $A$  will be an integer.
- Perform Eigen Decomposition using NumPy's library function
- Reconstruct  $A$  from eigenvalue and eigenvectors (Refer to Section 2.7).
- Check if the reconstruction is perfect. (`np.allclose` will come in handy)
- Please be mindful of applying efficient methods

## Task 3: Singular Value Decomposition

(moore-penrose.py)

- Take the dimensions of matrix  $n, m$  as input.
- Produce a random  $n \times m$  matrix  $A$ . For the purpose of demonstrating, every cell of  $A$  must be an integer.
- Perform Singular Value Decomposition using NumPy's library function
- Calculate the Moore-Penrose Pseudoinverse using NumPy's library function
- Calculate the Moore-Penrose Pseudoinverse *again* using Eq. 2.47
- Check if these two inverses are equal (`np.allclose` will come in handy)

## Submission

```
1705123
|-- matrix-transformations-and-eigen-decomposition.ipynb
|-- random_eigen.py
|-- symmetric_eigen.py
|-- moore-penrose.py
```

Zip the folder and rename it to **[Student\_ID].zip**

**Deadline: 02 December 2022, Friday 11.55 PM**