

numpy-3

November 16, 2025

[2]: # Inverse

```
[ ]: # Q. Make a matrix, A = 3x3
# Make a Identity matrix, I = 3x3

# Check for

# A * I = A
# A * invA = I
```

[7]: import numpy as np

```
A = np.array([1, 2, 3, 4, 5, 3, 1, 7, 9]).reshape(3, 3)
I = np.eye(3)

print(A)
print(I)
```

```
[[1 2 3]
 [4 5 3]
 [1 7 9]]
 [[1. 0. 0.]
 [0. 1. 0.]
 [0. 0. 1.]]
```

[8]: np.matmul(A, I)

```
[8]: array([[1., 2., 3.],
 [4., 5., 3.],
 [1., 7., 9.]])
```

[]: # Determinant of matrix

```
np.linalg.det(A)
```

[]: np.float64(27.0)

```
[10]: invA = np.linalg.inv(A)
invA
```

```
[10]: array([[ 0.88888889,  0.11111111, -0.33333333],
   [-1.22222222,  0.22222222,  0.33333333],
   [ 0.85185185, -0.18518519, -0.11111111]])
```

```
[11]: np.matmul(A, invA)
```

```
[11]: array([[ 1.00000000e+00,  2.77555756e-17,  4.16333634e-17],
   [-1.11022302e-16,  1.00000000e+00,  4.16333634e-17],
   [-3.33066907e-16, -2.77555756e-17,  1.00000000e+00]])
```

```
[ ]: # Transpose of Matrix
A.T
```

```
[ ]: array([[1, 4, 1],
   [2, 5, 7],
   [3, 3, 9]])
```

```
[13]: A.sum()
```

```
[13]: np.int64(35)
```

```
[14]: A.mean()
```

```
[14]: np.float64(3.888888888888889)
```

```
[15]: A.std()
```

```
[15]: np.float64(2.5579698740491863)
```

```
[16]: A.min()
```

```
[16]: np.int64(1)
```

```
[17]: A.max()
```

```
[17]: np.int64(9)
```

```
[18]: # Row sum
```

```
A.sum(axis=0)
```

```
[18]: array([ 6, 14, 15])
```

```
[19]: # Col sum
```

```
A.sum(axis=1)
```

```
[19]: array([ 6, 12, 17])

[ ]: np.linalg.eig(A)
      # A.v = lemma.v

[ ]: EigResult(eigenvalues=array([13.09569004+0.j           , 0.95215498+1.07477808j,
      0.95215498-1.07477808j]), eigenvectors=array([[[-0.28447549+0.j           ,
      0.2940061 -0.27326013j,
      0.2940061 +0.27326013j],
     [-0.45354298+0.j           , -0.71082121+0.j           ,
     -0.71082121-0.j           ],
     [-0.8446138 +0.j           , 0.5670899 +0.10968848j,
     0.5670899 -0.10968848j]]))

[ ]: arr = np.array([1, 2, 3, 4])
      # Cumulative prod
      np.cumprod(arr)

[ ]: array([ 1,  2,  6, 24])

[ ]: # Cumulative sum
      np.cumsum(arr)

[ ]: array([ 1,  3,  6, 10])

[23]: np.nan

[23]: nan

[24]: np.inf

[24]: inf

[25]: arr = np.array([1, 2, 3, np.nan, 4, 5, np.nan])
      arr

[25]: array([ 1.,  2.,  3., nan,  4.,  5., nan])

[27]: # Handling missing data
      mask = np.isnan(arr)
      arr[mask] = np.nanmean(arr)

[28]: arr
```

```
[28]: array([1., 2., 3., 3., 4., 5., 3.])
```

```
[29]: # Removing missing data
```

```
arr = np.array([1, 2, 3, np.nan, 4, 5, np.nan])  
  
mask = ~np.isnan(arr)  
  
arr[mask]
```

```
[29]: array([1., 2., 3., 4., 5.])
```

```
[30]: # 3 1 + 5 2 - 9 3 = 6  
# -3 1 + 7 3 = -2  
# - 2 + 4 3 = 8
```

```
[32]: A = np.array([[3, 5, -9], [-3, 0, 7], [0, -1, 4]])  
B = np.array([[6], [-2], [8]])
```

```
[33]: invA = np.linalg.inv(A)
```

```
[34]: np.matmul(invA, B)
```

```
[34]: array([[6.37037037],  
           [1.77777778],  
           [2.44444444]])
```

```
[35]: np.linalg.solve(A, B)
```

```
[35]: array([[6.37037037],  
           [1.77777778],  
           [2.44444444]])
```

1 Statistics and Probability

1.0.1 Types of stats

1. Descriptive stats
 - summarize data (mean, median, mode)
2. Inferential stats
 - Make predictions or inference (hypothesis testing, confidence interval)
3. Parametric stats
 - Assumes data follows a distribution like normal (t-test, ANOVA)
4. Non-parametric stats
 - No assumption of data distribution (median test, Wilcoxon test)

```
[38]: # Measure of central tendency
```

```
# mean -> if there is outlier we dont use mean,  
# median,  
# mode  
  
data = np.array([1, 2, 1, 4, 5, 6, 5, 5, 9])  
  
mean = np.mean(data)  
median = np.median(data)  
mode = np.bincount(data).argmax()
```

```
[39]: print(mean)  
print(median)  
print(mode)
```

```
4.222222222222222  
5.0  
5
```

```
[ ]:
```