```
import numpy as np
list=[1,2,3,4]
np.array(list)
list2=[[5,6,7],[6,9,2],[4,1,0]]
np.array(list2)
np.arange(0,10)
np.arange(0,11,2)
np.zeros(3)
np.zeros((3,3))
np.ones(3)
np.ones((3,3))
np.eye(3)
np.linspace(0,3,6)
np.linspace(0,10,100)
np.random.rand(3)
np.random.rand(3,4)
np.random.randn(3)
np.random.randint(4,8,2)
l=np.arange(0,25)
1.reshape(5,5)
1.max()
1.min()
1.argmax()
1.argmin()
1.shape
1.reshape(25,1)
1.dtype
k=[[5,10,15],[20,25,30],[35,40,45]]
arr1=np.array(k)
arr1[1]
arr1[1:3]
arr1[1:2]=40
arr1[:]=40
arr1[1:3,0:2]
kk=arr1>20
arr1[kk]
     array([40, 40, 40, 40, 40, 40, 40, 40, 40])
import numpy as np
arr=np.arange(0,10)
arr+arr
```

```
array([ 0, 2, 4, 6, 8, 10, 12, 14, 16, 18])
import numpy as np
np.zeros(10)
    array([0., 0., 0., 0., 0., 0., 0., 0., 0.])
v=np.arange(0,9)
v.reshape(3,3)
     array([[0, 1, 2],
           [3, 4, 5],
            [6, 7, 8]])
np.random.randn(1,50,25)
    array([[[ 0.79402049, 0.2404533 , -0.32466454, ..., 1.19554394,
              -1.70015004, 0.10870978],
             [0.76304578, -0.16212469, 0.7451845, ..., -1.11951122,
              -0.26102813, -0.21624645],
             [-0.78370119, 0.17046067, -0.10002257, ..., -0.34728315,
              0.34957869, -0.63081159],
             [0.30988196, -0.56864689, -0.3923799, ..., -0.63787804,
              1.9911658 , -0.87374272],
             [1.7547398, 0.72972474, -0.71352036, ..., 1.16784415,
             -1.13838518, 0.8487579 ],
             [1.22755096, -0.88741545, 0.17338364, ..., 0.21634336,
              0.22685185, -2.03982837]])
ar=np.arange(1,101)
d=ar.reshape(10,10)
d/100
    array([[0.01, 0.02, 0.03, 0.04, 0.05, 0.06, 0.07, 0.08, 0.09, 0.1],
            [0.11, 0.12, 0.13, 0.14, 0.15, 0.16, 0.17, 0.18, 0.19, 0.2],
            [0.21, 0.22, 0.23, 0.24, 0.25, 0.26, 0.27, 0.28, 0.29, 0.3],
            [0.31, 0.32, 0.33, 0.34, 0.35, 0.36, 0.37, 0.38, 0.39, 0.4],
            [0.41, 0.42, 0.43, 0.44, 0.45, 0.46, 0.47, 0.48, 0.49, 0.5],
            [0.51, 0.52, 0.53, 0.54, 0.55, 0.56, 0.57, 0.58, 0.59, 0.6],
            [0.61, 0.62, 0.63, 0.64, 0.65, 0.66, 0.67, 0.68, 0.69, 0.7],
            [0.71, 0.72, 0.73, 0.74, 0.75, 0.76, 0.77, 0.78, 0.79, 0.8],
            [0.81, 0.82, 0.83, 0.84, 0.85, 0.86, 0.87, 0.88, 0.89, 0.9],
            [0.91, 0.92, 0.93, 0.94, 0.95, 0.96, 0.97, 0.98, 0.99, 1. ]])
```

SVD

import numpy as np

```
from numpy.linalg import svd
# this matrix has rank=2, since col3 = col1+co2,
## but col1 and col2 are independent from each other
A = np.array([[1,2,3], [4,5,6], [5,7,9]])
U, S, VT = svd(A)
print("Left Singular Vectors:")
print(U)
print("Singular Values:")
print(np.diag(S))
print("Right Singular Vectors:")
print(VT)
    Left Singular Vectors:
     [[-0.2354116  0.78182354  -0.57735027]
     [-0.55937325 -0.5947842 -0.57735027]
      [-0.79478485 0.18703934 0.57735027]]
    Singular Values:
     [[1.56633231e+01 0.00000000e+00 0.00000000e+00]
     [0.00000000e+00 8.12593979e-01 0.00000000e+00]
      [0.00000000e+00 0.00000000e+00 1.13716384e-15]]
     Right Singular Vectors:
     [[-0.41158755 -0.56381288 -0.71603821]
      [-0.8148184 -0.12429146 0.56623547]
      # Return the original matrix A
# @ is used for matrix multiplication in Py3, use np.matmul with Py2
print(U @ np.diag(S) @ VT)
     [[1. 2. 3.]
     [4. 5. 6.]
      [5. 7. 9.]]
TRUNCATED SVD
import numpy as np
from sklearn.decomposition import TruncatedSVD
A = np.array([[1,2,3], [4,5,6], [5,7,9]])
print("Original Matrix:")
Α
    Original Matrix:
     array([[1, 2, 3],
```

```
[4, 5, 6],
svd = TruncatedSVD(n_components = 2) # reduce to 2 features
A_transf = svd.fit_transform(A)

print("Singular values:")
print(svd.singular_values_)
print()

print("Transformed Matrix after reducing to 2 features:")
print(A_transf)

Singular values:
  [15.66332312 0.81259398]

Transformed Matrix after reducing to 2 features:
  [[ 3.68732795 0.6353051 ]
  [ 8.76164389 -0.48331806]
  [12.44897184 0.15198704]]
```

RESULT: Program executed successfully and output is obtained

×