# 1 Python Machine Learning Stack (Anaconda)

#### Task 1

In your terminal, run

```
In [2]:
        !conda info
```

```
active environment : dl
```

active env location : C:\Users\kdmen\miniconda3\envs\dl

shell level : 2

user config file : C:\Users\kdmen\.condarc

populated config files : C:\Users\kdmen\.condarc

conda version : 23.3.1

conda-build version : not installed python version: 3.9.16.final.0

virtual packages : \_\_archspec=1=x86\_64

win=0=0

base environment : C:\Users\kdmen\miniconda3 (writable) conda av data dir : C:\Users\kdmen\miniconda3\etc\conda

conda av metadata url : None

channel URLs : https://conda.anaconda.org/conda-forge/win-64

https://conda.anaconda.org/conda-forge/noarch https://repo.anaconda.com/pkgs/main/win-64 https://repo.anaconda.com/pkgs/main/noarch https://repo.anaconda.com/pkgs/r/win-64 https://repo.anaconda.com/pkgs/r/noarch https://repo.anaconda.com/pkgs/msys2/win-64

https://repo.anaconda.com/pkgs/msys2/noarch

package cache : C:\Users\kdmen\miniconda3\pkgs

C:\Users\kdmen\.conda\pkgs

C:\Users\kdmen\AppData\Local\conda\conda\pkgs

envs directories : C:\Users\kdmen\miniconda3\envs

C:\Users\kdmen\.conda\envs

C:\Users\kdmen\AppData\Local\conda\conda\envs

platform : win-64

user-agent : conda/23.3.1 requests/2.29.0 CPython/3.9.16 Windows/10 Win

dows/10.0.22621

administrator : False netrc file : None offline mode : False

# 3 Transition from MATLAB to Python

#### Task 2

Run all of Python commands in the table "Linear Algebra Equivalents" in Numpy for MATLAB Users.

```
In [3]: a = np.array([[1., 2., 3.], [4., 5., 6.]])
         b = np.array([[7., 8., 9.], [10., 11., 12.]])
         c = np.array([[2., 2., 2.], [4., 4., 4.]])
         d = np.array([[1., 1., 1.], [3., 3., 3.]])
         n = 1
         v = np.array([1, 1, 10])
         x = np.array([[5, 5, 5], [11, 12, 13]])
 In [4]: np.ndim(a)
 Out[4]: 2
 In [5]: np.size(a)
 Out[5]: 6
 In [6]: np.shape(a)
Out[6]: (2, 3)
 In [7]: a.shape[n-1]
Out[7]: 2
 In [8]: np.array([[1., 2., 3.], [4., 5., 6.]])
 Out[8]: array([[1., 2., 3.],
                [4., 5., 6.]]
 In [9]: np.block([[a, b], [c, d]])
Out[9]: array([[ 1., 2., 3., 7., 8., 9.],
                [ 4., 5., 6., 10., 11., 12.],
                [ 2., 2., 2., 1., 1., 1.],
                [4., 4., 4., 3., 3., 3.]
In [10]: a[-1]
Out[10]: array([4., 5., 6.])
In [11]: # a[1, 4]
         a[1, 2]
```

```
Out[11]: 6.0
In [12]: # a[1] or a[1, :]
         a[1, :]
Out[12]: array([4., 5., 6.])
In [13]: \# a[0:5] or a[:5] or a[0:5, :]
         a[0:5, :]
Out[13]: array([[1., 2., 3.],
                [4., 5., 6.]])
In [14]: a[-5:]
Out[14]: array([[1., 2., 3.],
                [4., 5., 6.]])
In [15]: a[0:3, 4:9]
Out[15]: array([], shape=(2, 0), dtype=float64)
In [16]: # a[np.ix_([1, 3, 4], [0, 2])]
         a[np.ix_([0, 1], [0, 1, 2])]
Out[16]: array([[1., 2., 3.],
                [4., 5., 6.]]
In [17]: a[2:21:2,:]
Out[17]: array([], shape=(0, 3), dtype=float64)
In [18]: a[::2, :]
Out[18]: array([[1., 2., 3.]])
In [19]: a[::-1,:]
Out[19]: array([[4., 5., 6.],
                [1., 2., 3.]]
In [20]: a[np.r_[:len(a),0]]
Out[20]: array([[1., 2., 3.],
                [4., 5., 6.],
                [1., 2., 3.]])
In [21]: a.T
Out[21]: array([[1., 4.],
                [2., 5.],
                [3., 6.]])
In [22]: a.conj().T
```

```
Out[22]: array([[1., 4.],
                [2., 5.],
                [3., 6.]])
In [23]: # a @ b
         a @ b.T
Out[23]: array([[ 50., 68.],
                [122., 167.]])
In [24]: a * b
Out[24]: array([[ 7., 16., 27.],
                [40., 55., 72.]])
In [25]: a/b
Out[25]: array([[0.14285714, 0.25 , 0.33333333],
                [0.4
                          , 0.45454545, 0.5
                                                   ]])
In [26]: a**3
Out[26]: array([[ 1., 8., 27.],
                [ 64., 125., 216.]])
In [27]: (a > 0.5)
Out[27]: array([[ True, True, True],
                [ True, True, True]])
In [28]: np.nonzero(a > 0.5)
Out[28]: (array([0, 0, 0, 1, 1, 1], dtype=int64),
          array([0, 1, 2, 0, 1, 2], dtype=int64))
In [29]: a[:,np.nonzero(v > 0.5)[0]]
Out[29]: array([[1., 2., 3.],
                [4., 5., 6.]]
In [30]: a[:, v.T > 0.5]
Out[30]: array([[1., 2., 3.],
                [4., 5., 6.]]
In [31]: a[a < 0.5] = 0
In [32]: a * (a > 0.5)
Out[32]: array([[1., 2., 3.],
                [4., 5., 6.]]
In [33]: a[:] = 3
In [34]: y = x \cdot copy()
```

```
In [35]: y = x[1, :].copy()
In [36]: y = x.flatten()
In [37]: # np.arange(1., 11.) or np.r_[1.:11.] or np.r_[1:10:10j]
         np.arange(1., 11.)
Out[37]: array([ 1., 2., 3., 4., 5., 6., 7., 8., 9., 10.])
In [38]: # np.arange(10.) or np.r_[:10.] or np.r [:9:10j]
         np.arange(10.)
Out[38]: array([0., 1., 2., 3., 4., 5., 6., 7., 8., 9.])
In [39]: np.arange(1.,11.)[:, np.newaxis]
Out[39]: array([[ 1.],
                [ 2.],
                [ 3.],
                [ 4.],
                [5.],
                [ 6.],
                [ 7.],
                [ 8.],
                [ 9.],
                [10.]])
In [40]: np.zeros((3, 4))
Out[40]: array([[0., 0., 0., 0.],
                [0., 0., 0., 0.]
                [0., 0., 0., 0.]])
In [41]: np.zeros((3, 4, 5))
Out[41]: array([[[0., 0., 0., 0., 0.],
                 [0., 0., 0., 0., 0.]
                 [0., 0., 0., 0., 0.]
                 [0., 0., 0., 0., 0.]],
                 [[0., 0., 0., 0., 0.],
                 [0., 0., 0., 0., 0.],
                 [0., 0., 0., 0., 0.]
                 [0., 0., 0., 0., 0.]
                 [[0., 0., 0., 0., 0.],
                 [0., 0., 0., 0., 0.]
                 [0., 0., 0., 0., 0.]
                 [0., 0., 0., 0., 0.]]
In [42]: np.ones((3, 4))
```

```
Out[42]: array([[1., 1., 1., 1.],
                [1., 1., 1., 1.],
                [1., 1., 1., 1.]])
In [43]: np.eye(3)
Out[43]: array([[1., 0., 0.],
                [0., 1., 0.],
                 [0., 0., 1.]])
In [44]: np.diag(a)
Out[44]: array([3., 3.])
In [45]: np.diag(v, 0)
Out[45]: array([[ 1, 0, 0],
                [ 0, 1, 0],
                 [ 0, 0, 10]])
In [46]: from numpy.random import default rng
         rng = default rng(42)
         rng.random((3, 4))
Out[46]: array([[0.77395605, 0.43887844, 0.85859792, 0.69736803],
                 [0.09417735, 0.97562235, 0.7611397, 0.78606431],
                 [0.12811363, 0.45038594, 0.37079802, 0.92676499]])
In [47]: np.linspace(1,3,4)
Out[47]: array([1.
                          , 1.66666667, 2.33333333, 3.
                                                               ])
In [48]: # np.mgrid[0:9.,0:6.] or np.meshgrid(r_[0:9.],r_[0:6.])
         np.mgrid[0:9.,0:6.]
Out[48]: array([[[0., 0., 0., 0., 0., 0.],
                  [1., 1., 1., 1., 1., 1.],
                  [2., 2., 2., 2., 2., 2.],
                  [3., 3., 3., 3., 3., 3.]
                  [4., 4., 4., 4., 4., 4.]
                  [5., 5., 5., 5., 5., 5.]
                  [6., 6., 6., 6., 6., 6.]
                  [7., 7., 7., 7., 7., 7.]
                 [8., 8., 8., 8., 8., 8.]],
                 [[0., 1., 2., 3., 4., 5.],
                 [0., 1., 2., 3., 4., 5.],
                 [0., 1., 2., 3., 4., 5.],
                  [0., 1., 2., 3., 4., 5.],
                  [0., 1., 2., 3., 4., 5.],
                  [0., 1., 2., 3., 4., 5.],
                  [0., 1., 2., 3., 4., 5.],
                  [0., 1., 2., 3., 4., 5.],
                  [0., 1., 2., 3., 4., 5.]])
```

```
In [49]: # ogrid[0:9.,0:6.] or np.ix_(np.r_[0:9.],np.r_[0:6.]
         np.ogrid[0:9.,0:6.]
Out[49]: [array([[0.],
                  [1.],
                  [2.],
                  [3.],
                  [4.],
                  [5.],
                  [6.],
                  [7.],
                  [8.]]),
           array([[0., 1., 2., 3., 4., 5.]])]
In [50]: | np.meshgrid([1,2,4],[2,4,5])
Out[50]: [array([[1, 2, 4],
                  [1, 2, 4],
                  [1, 2, 4]]),
           array([[2, 2, 2],
                 [4, 4, 4],
                 [5, 5, 5]])]
In [51]: np.ix_([1,2,4],[2,4,5])
Out[51]: (array([[1],
                  [2],
                  [4]]),
           array([[2, 4, 5]]))
In [52]: m = 3
         np.tile(a, (m, n))
Out[52]: array([[3., 3., 3.],
                 [3., 3., 3.],
                 [3., 3., 3.],
                 [3., 3., 3.],
                [3., 3., 3.],
                 [3., 3., 3.]])
In [53]: # np.concatenate((a,b),1) or np.hstack((a,b)) or np.column_stack((a,b)) or np.c_[a,
         np.concatenate((a,b),1)
Out[53]: array([[ 3., 3., 3., 7., 8., 9.],
                 [ 3., 3., 3., 10., 11., 12.]])
In [54]: # np.concatenate((a,b)) or np.vstack((a,b)) or np.r_[a,b]
         np.concatenate((a,b))
Out[54]: array([[ 3., 3., 3.],
                 [3., 3., 3.],
                 [7., 8., 9.],
                 [10., 11., 12.]])
In [55]: a.max()
```

```
Out[55]: 3.0
In [56]: a.max(0)
Out[56]: array([3., 3., 3.])
In [57]: a.max(1)
Out[57]: array([3., 3.])
In [58]: np.maximum(a, b)
Out[58]: array([[ 7., 8., 9.],
                [10., 11., 12.]])
In [59]: # np.sqrt(v @ v) or np.linalg.norm(v)
         np.linalg.norm(v)
Out[59]: 10.099504938362077
In [60]: np.logical_and(a,b)
Out[60]: array([[ True, True, True],
                [ True, True, True]])
In [61]: np.logical_or(a,b)
Out[61]: array([[ True, True, True],
                [ True, True, True]])
In [62]: aa = np.array([5, 3, 7], dtype=np.uint8)
         bb = np.array([3, 6, 7], dtype=np.uint8)
         #a & b
         np.bitwise_and(aa, bb)
Out[62]: array([1, 2, 7], dtype=uint8)
In [63]: #a / b
         np.bitwise_or(aa, bb)
Out[63]: array([7, 7, 7], dtype=uint8)
In [64]: sq_mtrx = np.array([[1,2,3], [4,5,6], [7,8,9]])
         np.linalg.inv(sq_mtrx)
Out[64]: array([[-4.50359963e+15, 9.00719925e+15, -4.50359963e+15],
                [ 9.00719925e+15, -1.80143985e+16, 9.00719925e+15],
                [-4.50359963e+15, 9.00719925e+15, -4.50359963e+15]])
In [65]: np.linalg.pinv(sq_mtrx)
```

```
Out[65]: array([[-6.38888889e-01, -1.66666667e-01, 3.05555556e-01],
                [-5.5555556e-02, 3.46944695e-17, 5.5555556e-02],
                [ 5.27777778e-01, 1.66666667e-01, -1.94444444e-01]])
In [66]: np.linalg.matrix_rank(sq_mtrx)
Out[66]: 2
In [67]: # linalg.solve(a, b) if a is square; linalg.lstsq(a, b) otherwise
         np.linalg.solve(sq_mtrx, sq_mtrx)
Out[67]: array([[1., 0.5, 0.],
                [0., 0., 0.],
                [0., 0.5, 1.]])
In [68]: # MATLAB: b/a --> NUMPY: Solve a.T x.T = b.T instead
In [69]: U, S, Vh = np.linalg.svd(a); V = Vh.T
In [70]: A = np.array([[4, -1, 2],
                      [-1, 5, 3],
                      [2, 3, 6]])
         # Check if the matrix is positive definite
         is_positive_definite = np.all(np.linalg.eigvals(A) > 0)
         np.linalg.cholesky(A)
Out[70]: array([[ 2.
                           , 0. , 0.
                                                     ],
                         , 2.17944947, 0.
                [-0.5
                [ 1.
                          , 1.60591014, 1.55597321]])
In [71]: D,V = np.linalg.eig(sq_mtrx)
In [72]: D,V = np.linalg.eig((sq_mtrx, sq_mtrx))
In [73]: # Compute all eigenvalues and eigenvectors
         D, V = np.linalg.eigh(A)
         # Get the k (three) largest eigenvalues and their corresponding eigenvectors
         k = 3
         top_eigenvalues = D[-k:]
         top_eigenvectors = V[:, -k:]
In [74]: Q,R = np.linalg.qr(A)
In [75]: # P,L,U = linalg.lu(a) where a == P@L@U
         import scipy
         P,L,U = scipy.linalg.lu(A)
In [76]: np.fft.fft(a)
```

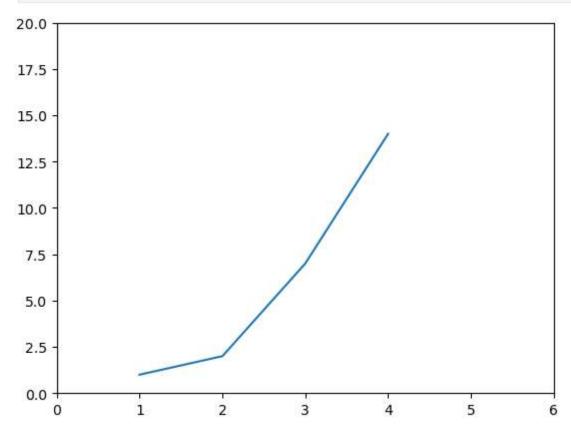
```
Out[76]: array([[9.+0.j, 0.+0.j, 0.+0.j],
                [9.+0.j, 0.+0.j, 0.+0.j]
In [77]: np.fft.ifft(a)
Out[77]: array([[3.+0.j, 0.+0.j, 0.+0.j],
                [3.+0.j, 0.+0.j, 0.+0.j]]
In [78]: # np.sort(a) or a.sort(axis=0)
         np.sort(a)
Out[78]: array([[3., 3., 3.],
                [3., 3., 3.]])
In [79]: # np.sort(a, axis=1) or a.sort(axis=1)
         np.sort(a, axis=1)
Out[79]: array([[3., 3., 3.],
                [3., 3., 3.]])
In [80]: I = np.argsort(a[:, 0]); b = a[I,:]
In [81]: Z = np.array([[1, 2], [3, 4], [5, 6]])
         y = np.array([7, 8, 9])
         np.linalg.lstsq(Z, y, rcond=None)[0]
Out[81]: array([-6., 6.5])
In [82]: from scipy.signal import resample
         x = np.array([1, 2, 3, 4, 5])
         resample(x, int(np.ceil(len(x) / 2)))
Out[82]: array([2.
                          , 2.30801829, 4.69198171])
In [83]: np.unique(a)
Out[83]: array([3.])
In [84]: a.squeeze()
Out[84]: array([[3., 3., 3.],
                [3., 3., 3.]])
```

# 4 Plotting (MatPlotLib / PyPlot)

### Task 3

Run the following script in IPython and paste the figure created by the script into your report.

```
In [85]: import matplotlib.pyplot as plt
plt.plot([1,2,3,4], [1,2,7,14])
plt.axis([0, 6, 0, 20])
plt.show()
```



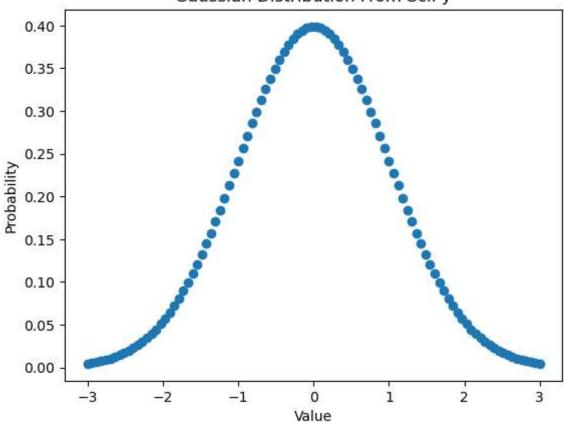
## Task 4

Use Matplotlib to create a figure of your choice in IPython. Paste your code and figure into your report.

```
In [86]: import scipy.stats as stats

mean = 0
    variance = 1
    std = np.sqrt(variance)
    x = np.linspace(mean-3*std, mean+3*std, 100)
    plt.scatter(x, stats.norm.pdf(x, mean, std))
    plt.title("Gaussian Distribution From SciPy")
    plt.xlabel("Value")
    plt.ylabel("Probability")
    plt.show()
```

#### Gaussian Distribution From SciPy



# 5 Version Control System (BitBucket/GitHub)

## Task 5:

- Paste your VCS account into your report.
- Platform: Github
- Username: kdmalc
- Link: https://github.com/kdmalc

# 6 Integrated Development Environment (PyCharm)

## Task 6:

Start a new project in Pycharm. Commit and push your project to Bitbucket/GitHub as a public project. Paste the link of your project in your report.

• https://github.com/kdmalc/deeplearning576