3. Deep learning tools 3.4 SciPy

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Introduction

- Built on top of NumPy.
- Optimized to process multidimensional arrays.
- A tool to work with
 - images and signal processing
 - large data sets.
- This section contains some examples of the capabilities of the library.

Data input and output

- SciPi is used to work with several data types as .mat, .wav and others, with specific commands for each data type.
- In this example we create, save and load a .mat file, that can be also read in MatLab.

```
import scipy.io as sio
import numpy as np

arr = np.array([1,2,3,4])  #create an array

#save the array by the name 'arr_samp'
sio.savemat('sample_data.mat', {'arr_samp': arr})

#load the array into the variable from mat file
sample_arr = sio.loadmat('sample_data.mat')

print(sample_arr['arr_samp'])
```

[[1 2 3 4]]

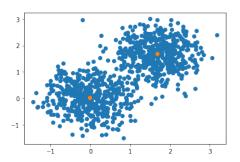
Clustering methods

- Clustering is a procedure that is intended to split a set of data in different points with the purpose of obtaining an interpretation of the data structure.
- A popular rather limited clustering method is K-means, and SciPy has it implemented

Clustering methods

• This is the representation

```
1 out = vq(data, center) #assigns labels to data
2 plt.scatter(whit[:,0], whit[:,1]) #plot the scatterplots
3 plt.scatter(center[:,0], center[:,1]) # plot the centroids
4 plt.show()
```



Constants

- *scipy.constants* gives access to many constants.
- They can be used in mathematical expressions.

```
import scipy.constants
from scipy.constants import find

print(find('light')) # find all constants with keyword 'light'
print(scipy.constants.physical_constants['speed of light in vacuum'])
print(scipy.constants.pi)
```

```
['speed of light in vacuum'] (299792458.0, 'm s^-1', 0.0) 3.141592653589793
```

Linear algebra

• *scipy.linalg* performs linear algebra operations in Python. Faster than BLAS and LAPACK libraries.

```
1 from scipy import linalg
2 import numpy as np
3
4 A = np.array([[1,2],[3,2]]) # create a square matrix
5
6 print(linalg.det(A)) # compute the determinant of a
7 print(linalg.inv(A)) # compute the inverse of the matrix
```

```
-4.0
[[-0.5 0.5]
[ 0.75 -0.25]]
```

Linear algebra

```
val, vect = linalq.eiq(A) # compute the svd of a
2 print('\neigenvalues =')
3 print (val)
4 print('\neigenvectors =')
5 print (vect)
6 b = np.array([2,4]) # 2D vector
7 print(linalq.solve(a,b)) # Solution of equation Ax=b
eigenvalues =
[-1.+0.j 4.+0.j]
eigenvectors =
[[-0.70710678 -0.5547002 ]
  [0.70710678 - 0.8320502911
solution of equation Ax=b:
[1. 0.5]
```

Numeric integrals

- The *integrate* package is used to compute numeric integrals with various methods.
- Example:

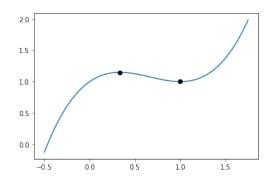
$$\int_{y=0}^{\frac{1}{2}} \int_{x=0}^{1-2y} xy dx dy = \frac{1}{96} \approx 0.0104167$$

```
1 from scipy import integrate
2 def f(x, y):
3    return x*y
4 def bounds_y():
5    return [0, 0.5]
6 def bounds_x(y):
7    return [0, 1-2*y]
8
9 integrate.nquad(f, [bounds_x, bounds_y])
```

(0.0104166666666666666, 4.101620128472366e-16)

Optimization

- Minimizing or maximizing a function with constrained and unconstrained minimization problems.
- The package has the most common optimization approaches such as least squares (least_squares()) and curve fitting techniques (curve_fit()).



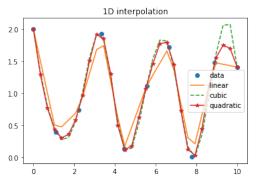
$$f(x) = x^3 - 2x^2 + x$$
$$f'(x) = 3x^2 - 4x + 1$$
$$x_0 = \frac{1}{3}, 1$$

```
1 from scipy import optimize
2 import numpy as np
3
4 def func(x):
5     return x**3 - 2*x**2 + x + 1 # The function
6
7 optimize.minimize_scalar(func) # Find the minimum
```

fun: 1.0
nfev: 12
nit: 8
success: True
x: 1.0

Other stuff

• The package *interpolate* is also useful. It can interpolate using several methods.



- The package *scipy.ndimage* can does image processing operations: display, geometric transformations, filtering, edge detection...
- Library *MISC* is used as a sample to test operations.