



Learning Objectives

By the end of this lesson, you will be able to:

- Explain the importance of SciPy
- List the characteristics of SciPy
- Explain sub-packages of SciPy
- Discuss SciPy sub-packages, such as optimization, integration, linear algebra, statistics, weave, and IO



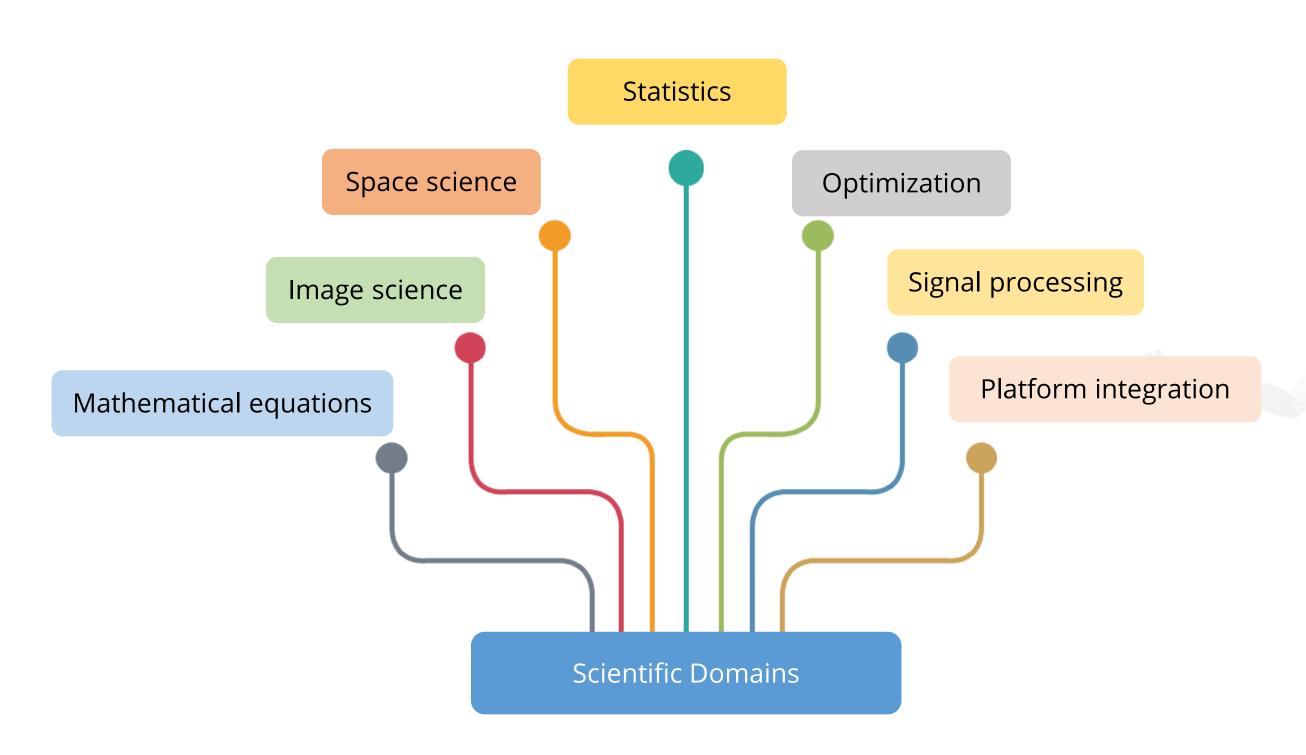


SciPy and Its Characteristics



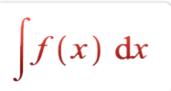
Multiple Scientific Domains

How to handle multiple scientific domains? The solution is SciPy.



SciPy

SciPy has built-in packages that help in handling the scientific domains.



Mathematics integration

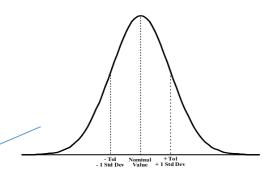
$$5x + 4y = 1$$

$$3x + 2y = 8$$

Linear algebra

Mathematics constants





Statistics (Normal distribution)

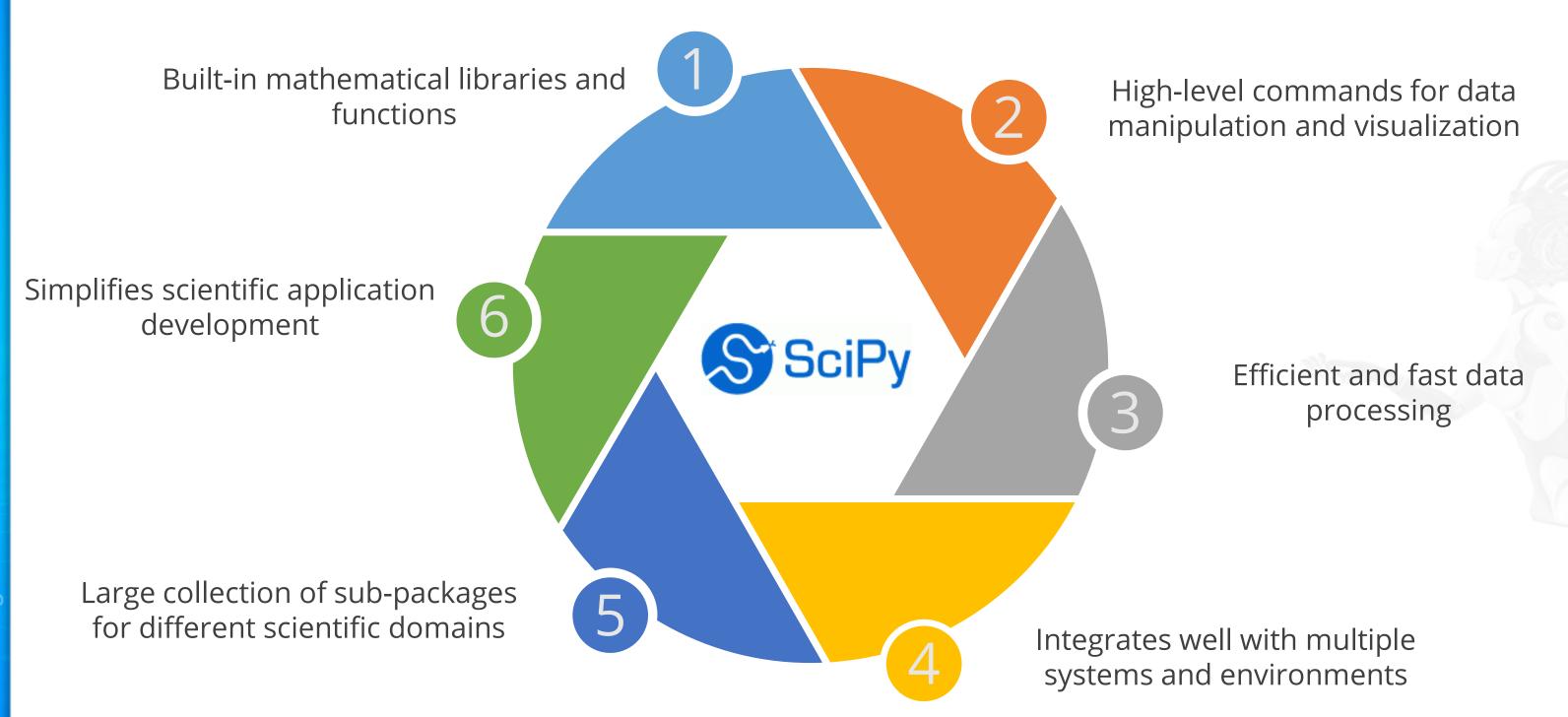


Multidimensional image processing



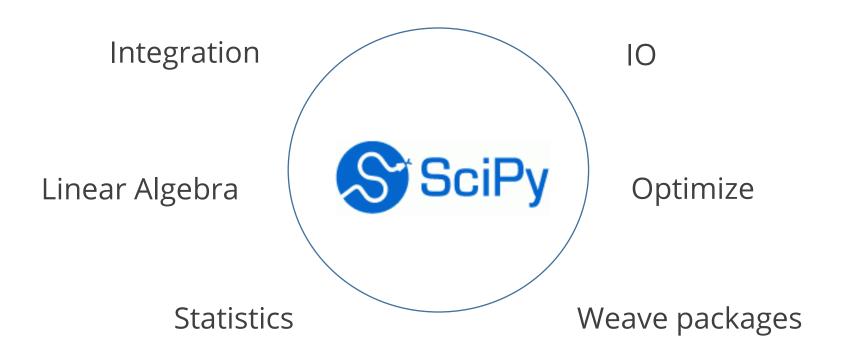
Language integration

SciPy and Its Characteristics



SciPy Packages

Some widely used packages are:





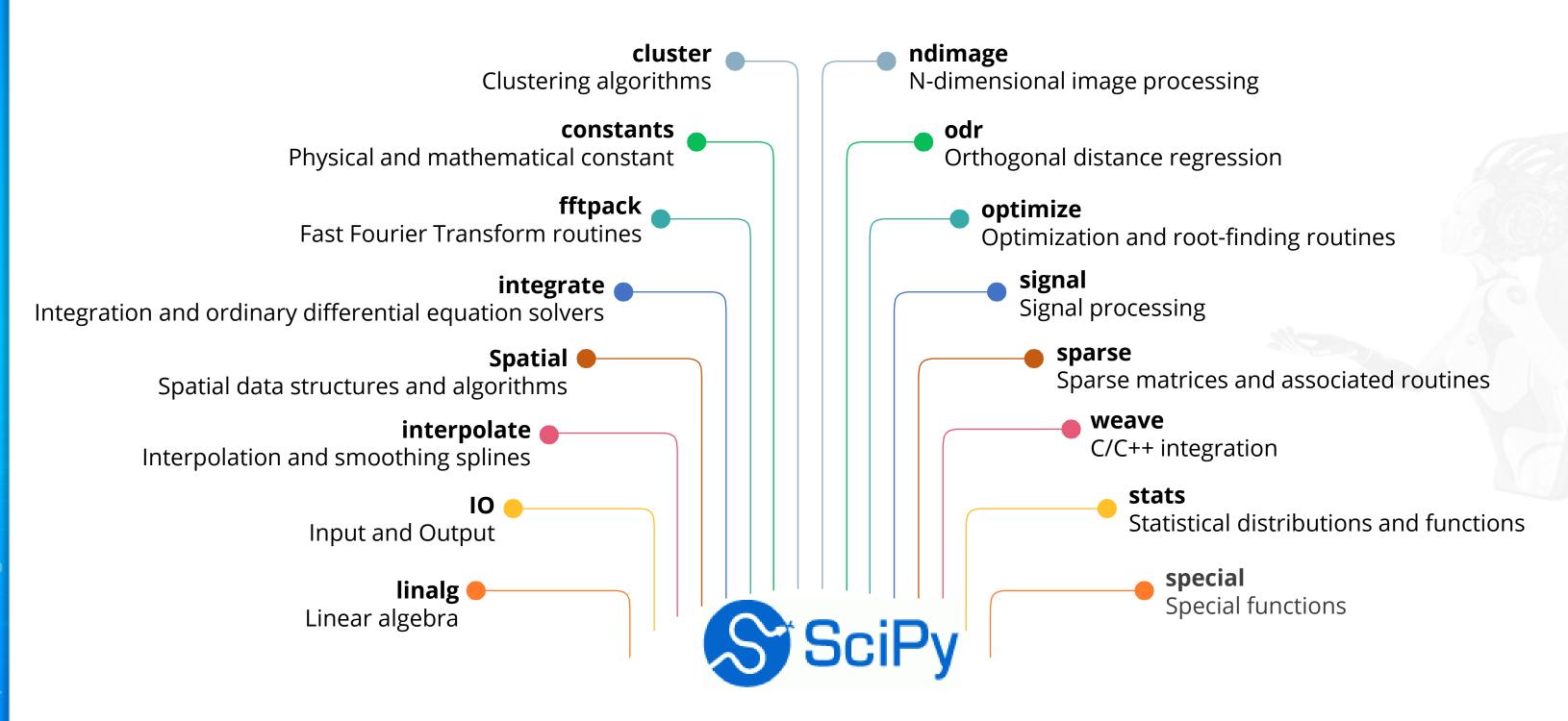


Introduction of SciPy Sub-Package



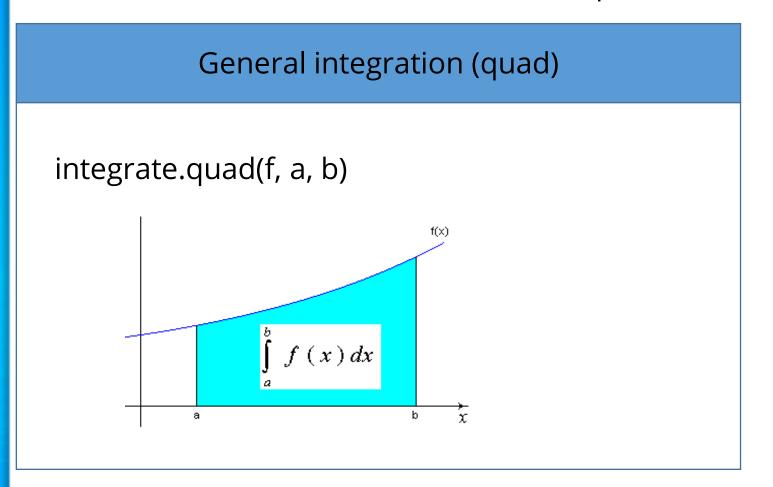
SciPy Sub-Package

SciPy has multiple sub-packages which handle different scientific domains.



SciPy Sub-Package: Integration

SciPy provides integration techniques that solve mathematical sequences and series, or perform function approximation.



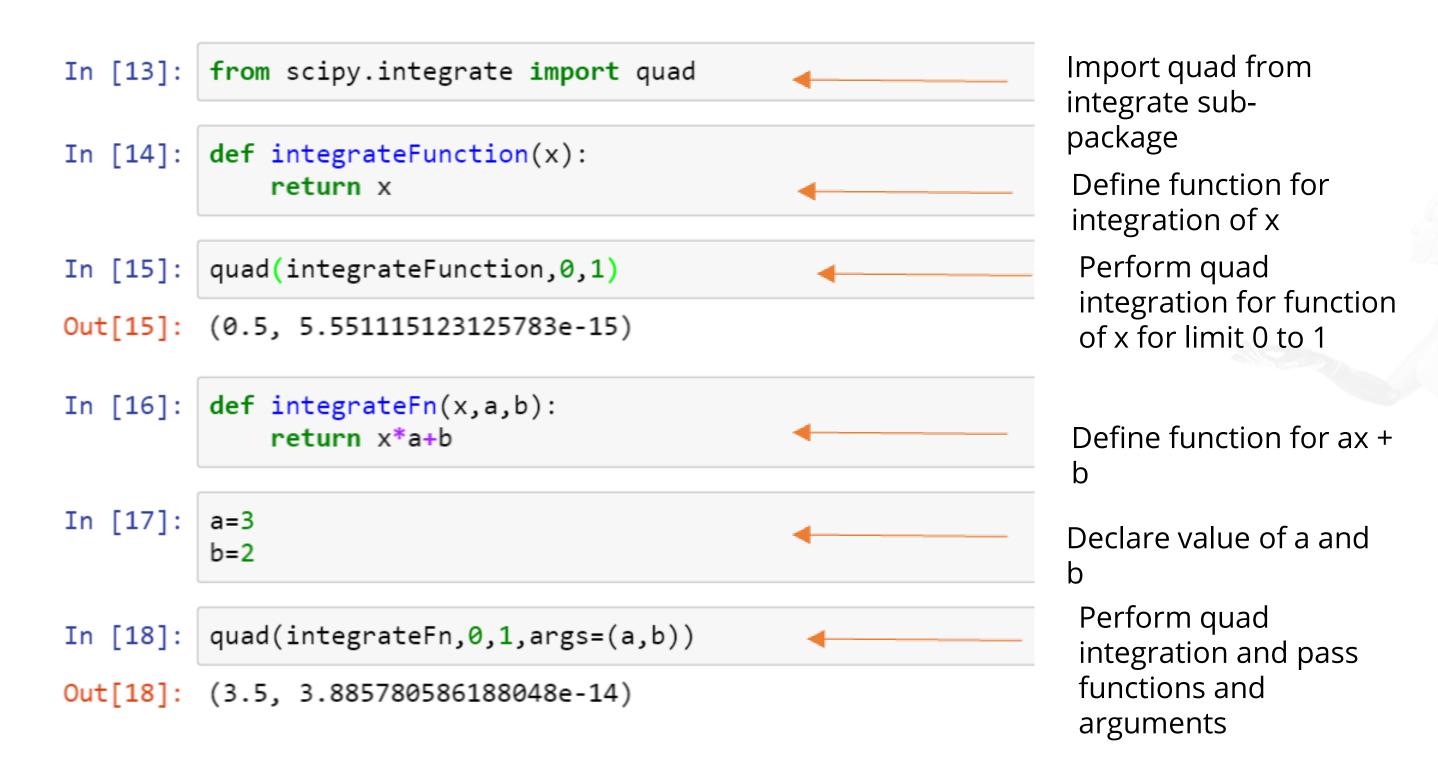
General multiple integration (dblquad, tplquad, nquad)

- integrate.dblquad()
- integrate.tplquad()
- integrate.nquad()

The limits of all inner integrals need to be defined as functions.

SciPy Sub-Package: Integration

This example shows how to perform quad integration.



SciPy Sub-Package: Integration

This example shows you how to perform multiple integration.

```
In [20]: import scipy.integrate as integrate

Import integrate package sub-package

In [21]: def f(x, y):
    return x + y
    integrate.dblquad(f, 0, 1,lambda x: 0, lambda x: 2)

Out[21]: (3.0, 3.3306690738754696e-14)

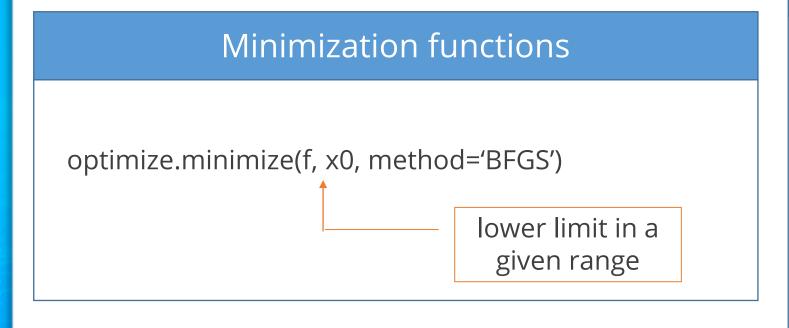
Perform multiple integration using the lambda built-in function
```

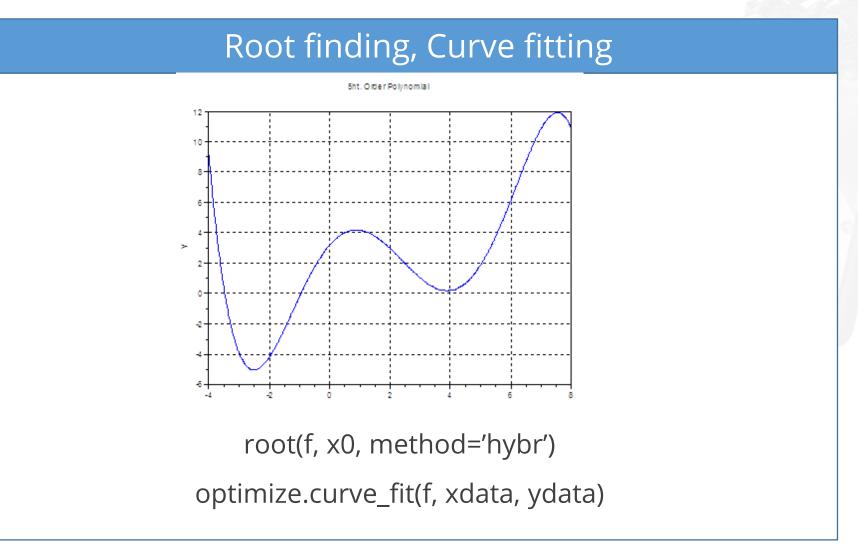




Optimization is a process to improve performance of a system mathematically by fine-tuning the process parameters.

SciPy provides several optimization algorithms, such as bfgs, Nelder-Mead simplex, Newton Conjugate Gradient, COBYLA, or SLSQP.





```
Import numpy and
In [32]:
         import numpy as np
                                                                                     optimize from SciPy
         from scipy import optimize
                                                                                        Define function for
In [33]:
         def f(x):
             return x^{**2} + 5*np.sin(x)
                                                                                        X^2 + 5 \sin x
In [34]:
         minimaValue = optimize.minimize(f,x0=2,method='bfgs',options={'disp':True})
         Optimization terminated successfully.
                  Current function value: -3.246394
                                                                                         Perform optimize
                  Iterations: 4
                                                                                         minimize function
                  Function evaluations: 24
                                                                                         using bfgs method
                  Gradient evaluations: 8
                                                                                         and options
         minimaValueWithoutOpt = optimize.minimize(f,x0=2,method='bfgs')
In [35]:
In [36]:
         minimaValueWithoutOpt
Out[36]:
               fun: -3.2463942726915382
                                                                                    Perform optimize minimize
          hess_inv: array([[ 0.15430551]])
                                                                                    function using bfgs method and
               jac: array([ -8.94069672e-08])
                                                                                    without options
           message: 'Optimization terminated successfully.'
              nfev: 24
               nit: 4
              njev: 8
            status: 0
           success: True
                 x: array([-1.11051051])
```

```
In [118]: import numpy as np
           from scipy.optimize import root
                                                                     Define function for
           def rootfunc(x):
                                                                     X + 3.5 \cos x
               return x + 3.5 * np.cos(x)
                                                                  Pass x value in argument for
In [119]: rootValue = root(rootfunc, 0.3)
                                                                  root
In [120]:
           rootValue
Out[120]:
              fjac: array([[-1.]])
                                                                     Function value and array
                fun: array([ 2.22044605e-16])
            message: 'The solution converged.'
                                                                     values
               nfev: 14
                qtf: array([ -8.32889313e-13])
                  r: array([-4.28198145])
             status: 1
            success: True
                  x: array([-1.21597614])
```





SciPy provides rapid linear algebra capabilities and contains advanced algebraic functions.

Inverse of matrixFinding DeterminantSolve LinearSingle ValuesystemsDecomposition (SVD)

This function is used to compute the inverse of the given matrix. Let's look at the inverse matrix operation.

```
In [65]: import numpy as np
         from scipy import linalg
                                                               Import linalg and
         matrix = np.array([[10,6],[2,7]])
                                                                Define a numpy
         matrix
                                                               matrix or array
Out[65]: array([[10, 6],
                [ 2, 7]])
In [66]: type(matrix)
                                                                   View the type
Out[66]: numpy.ndarray
                                                             Use inv function to
In [67]: linalg.inv(matrix)
                                                             inverse the matrix
Out[67]: array([[ 0.12068966, -0.10344828],
                [-0.03448276, 0.17241379]])
```

SciPy provides rapid linear algebra capabilities and contains advanced algebraic functions.

Inverse of matrix

Finding Determinant

Solve Linear

Systems

Single Value

Decomposition (SVD)

With this function you can compute the value of the determinant for the given matrix.

```
In [68]:
          import numpy as np
                                                                Import linalg and
          from scipy import linalg
                                                                Define an numpy matrix or
          matrix = np.array([[4,9],[3,5]])
                                                                array
          matrix
Out[68]: array([[4, 9],
                 [3, 5]])
                                                            Use det function to find the
          linalg.det(matrix)
In [69]:
                                                            determinant value of the
Out[69]: -7.000000000000001
                                                            matrix
```

SciPy provides rapid linear algebra capabilities and contains advanced algebraic functions.

Inverse of matrix

Finding Determinant

Solve Linear systems

Single Value Decomposition (SVD)

Linear equations

$$2x + 3y + z = 21$$

 $-x + 5y + 4z = 9$
 $3x + 2y + 9z = 6$

```
In [88]:
         import numpy as np
                                             Import linalg
         from scipy import linalg ←
In [89]: numArray = np.array([[2,3,1],[-1,5,4],[3,2,9]])
In [90]: numArrValue = np.array([21,9,6])
In [91]: linalg.solve(numArray,numArrValue)
                                                Use solve
Out[91]: array([ 4.95, 4.35, -1.95])
                                               method
```

SciPy provides rapid linear algebra capabilities and contains advanced algebraic functions.

Inverse of matrix		Finding Determinant	Solve Linear systems		Single Value Decomposition (SVD)
In [103]:	_	numpy as np ipy import linalg	· · · · · · · · · · · · · · · · · · ·	4	Import linalg
In [104]:	numSvdArr = np.array([[3,5,1],[9,5,7]])				Define matrix
In [105]: Out[105]:		•		•	Find shape of ndarray which is 2X3 matrix
In [106]:	linalg.s	svd(numSvdArr)		4	Use svd function
Out[106]:	<pre>(array([[-0.37879831, -0.92547925],</pre>			, ←	Jnitary matrix) Sigma or square root of eigenvalues VH is values collected into unitary matrix

Calculate Eigenvalues and Eigenvectors



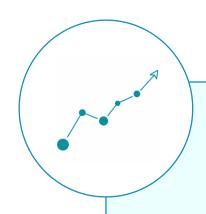
Problem Statement: Demonstrate how to calculate eigenvalues and eigenvectors

Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login**.





SciPy provides a very rich set of statistical functions which are:

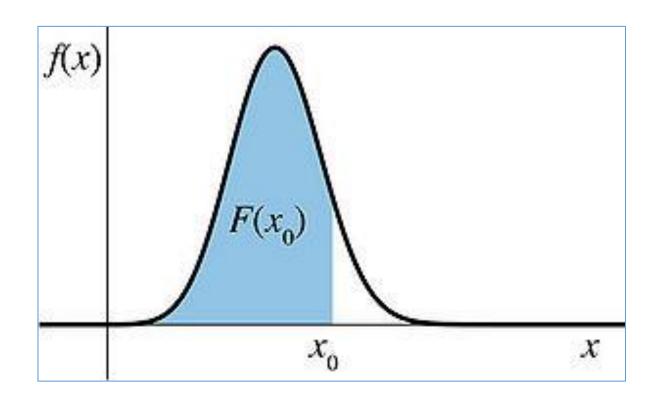


- This package contains distributions for which random variables are generated.
- These packages enable the addition of new routines and distributions. It also offers convenience methods such as pdf(), cdf()
- Following are the statistical functions for a set of data:
 - linear regression: linregress()
 - describing data: describe(), normaltest()

CDF or Cumulative Distribution Function provides the cumulative probability associated with a function.

			One standard
Age Range	Frequency	Cumulative Frequency	deviation
0-10	19	19	
10-20	55	74	←68% of data
21-30	23	97	Total number of persons within 95% of data
31-40	36	133	this age 99.7% of data
41-50	10	143	
51-60	17	160	-3 -2 -1 0 1 2 3
			F(x) = P(X≤x) negative infinity

Probability Density Function, or **PDF**, of a continuous random variable is the derivative of its Cumulative Distribution Function, or CDF.



$$f(x) = rac{dF(x)}{dx}$$
 Derivative of CDF

Shown here are functions used to perform Normal Distribution:

```
Import norm for normal
          from scipy.stats import norm
In [108]:
                                                                  distribution
                                                                      rvs for Random variables
          norm.rvs(loc=0,scale=1,size=10)
In [110]:
Out[110]: array([-0.16337774, 0.39039561, 0.85642826, 0.30134358, -1.86009474,
                 -0.29621603, 0.03863757, 0.23727056, -1.42395316, -0.5730162 ])
                                                     cdf for Cumulative Distribution Function
          norm.cdf(5,loc=1,scale=2)
In [112]:
Out[112]: 0.97724986805182079
                                                         pdf for Probability Density
          norm.pdf(9,loc=0,scale=1)
In [113]:
                                                         Function for random
Out[113]: 1.0279773571668917e-18
                                                         distribution
```



loc and **scale** are used to adjust the location and scale of the data distribution.



SciPy Sub-Package: Weave and IO



SciPy Sub-Package: Weave

The weave package provides ways to modify and extend any supported extension libraries.



Features of Weave Package:

- Includes C/C++ code within Python code
- Speed ups of 1.5x to 30x compared to algorithms written in pure Python

Two main functions of weave::

- inline() compiles and executes C/C++ code on the fly
- blitz() compiles NumPy Python expressions for fast execution

SciPy Sub-Package: IO

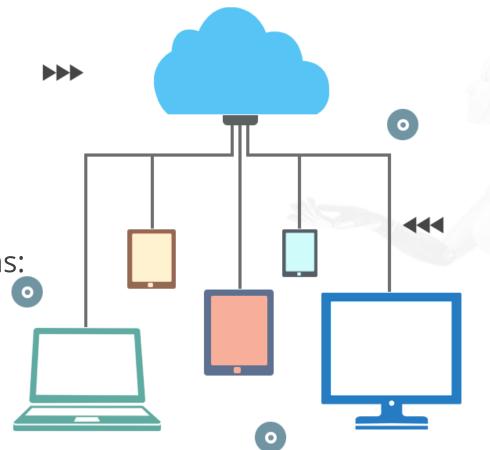
The IO package provides a set of functions to deal with several kinds of file formats.

It offers a set of functions to deal with file formats that include:

- MatLab file
- IDL files
- Matrix market files
- Wav sound files
- Arff files
- Netcdf files

Package provides additional files and its corresponding methods such as:

- Numpy.loadtxt()/Numpy.savetxt()
- Numpy.genfromtxt()/Numpy.recfromcsv()
- Numpy.save()/Numpy.load()



Using SciPy to Solve a Linear Algebra Problem



Problem Statement:

There is a test with 30 questions worth 150 marks. The test has two types of questions:

- 1. True or false carries 4 marks each
- 2. Multiple choice carries 9 marks each

Find the number of true or false and multiple-choice questions.

Common instructions:

- •If you are new to Python, download the "Anaconda Installation Instructions" document from the "Resources" tab to view the steps for installing Anaconda and the Jupyter notebook.
- •Download the "Assignment 01" notebook and upload it on the Jupyter notebook to access it.
- •Follow the cues provided to complete the assignment.

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Using SciPy to Declare Random Values



Problem Statement:

Use SciPy to declare 20 random values for random values and perform the following:

- 1. CDF Cumulative Distribution Function for 10 random variables.
- 2. PDF Probability Density Function for 14 random variables.

Common instructions:

- •If you are new to Python, download the "Anaconda Installation Instructions" document from the "Resources" tab to view the steps for installing Anaconda and the Jupyter notebook.
- •Download the "Assignment 02" notebook and upload it on the Jupyter notebook to access it.
- •Follow the cues provided to complete the assignment.

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Key Takeaways

You are now able to:

- Explain the importance of SciPy
- List the characteristics of SciPy
- Explain sub-packages of SciPy
- Discuss SciPy sub-packages, such as optimization, integration, linear algebra, statistics, weave, and IO



DATA AND ARTIFICIAL INTELLIGENCE



Knowledge Check

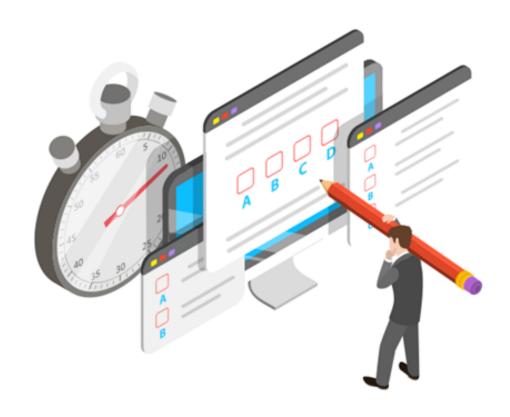


Knowledge Check

4

What are the specification limits provided for curve fitting function (optimize.curve.fit), during the optimization process?

- a. Upper limit value
- b. Lower limit value
- C. Upper and lower limit values
- d. Only the optimization method



1

What are the specification limits provided for curve fitting function (optimize.curve.fit), during the optimization process?

- a. Upper limit value
- b. Lower limit value
- C. Upper and lower limit values
- d. Only the optimization method



The correct answer is **c**

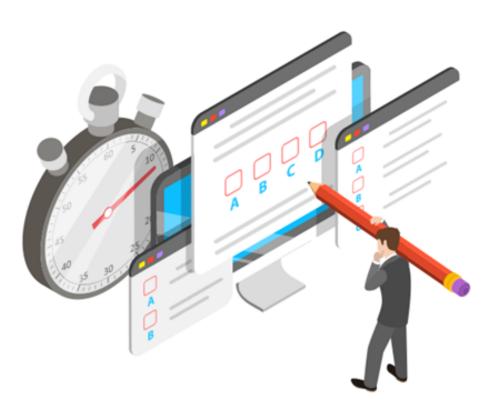
Both the upper and lower limit values should be specified for optimize.curve.fit function.



Which of the following function is used for inversing the matrix?

2

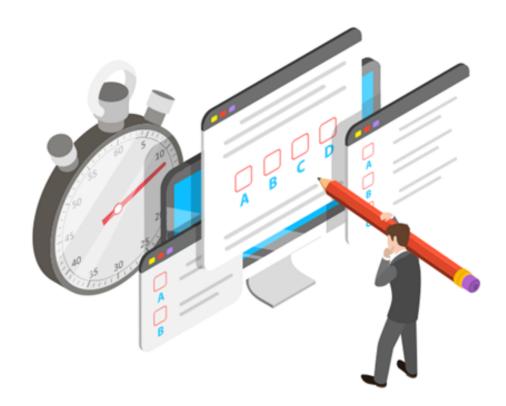
- a. SciPy.special
- b. SciPy.linalg
- ^{C.} SciPy.signal
- d. SciPy.stats



Which of the following function is used for inversing the matrix?

2

- a. SciPy.special
- b. SciPy.linalg
- ^{C.} SciPy.signal
- d. SciPy.stats



The correct answer is **b**

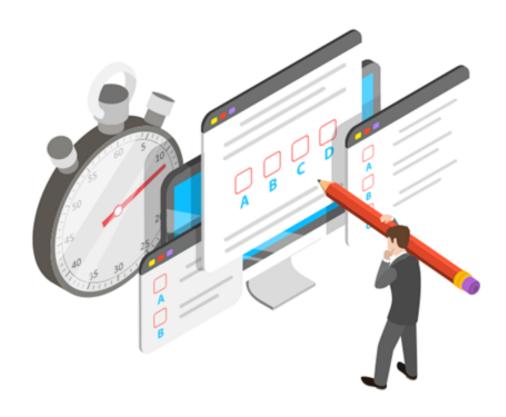
SciPy.linalg is used to inverse the matrix.



Which of the following is performed using SciPy?

3

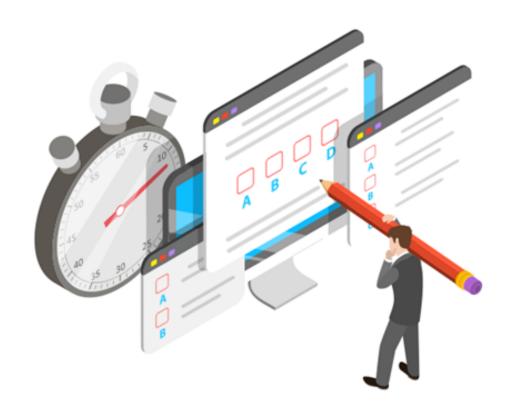
- a. Website
- b. Plot data
- C. Scientific calculations
- d. System administration



Which of the following is performed using SciPy?

3

- a. Website
- b. Plot data
- C. Scientific calculations
- d. System administration



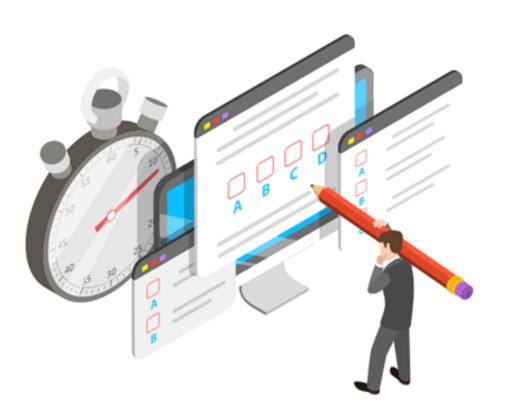
The correct answer is **c**

SciPy has been specially made to perform scientific calculations. Generally, Python is the programming language that has libraries to perform all listed activities.

4

Which of the following functions is used to calculate minima?

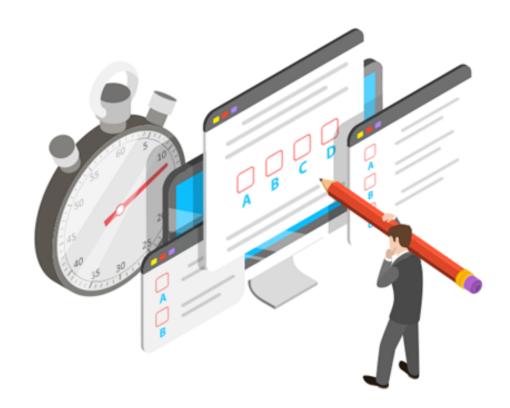
- a. optimize.minimize()
- b. integrate.quad()
- c. stats.linregress()
- d. linalg.solve()



4

Which of the following functions is used to calculate minima?

- a. optimize.minimize()
- b. integrate.quad()
- c. stats.linregress()
- d. linalg.solve()



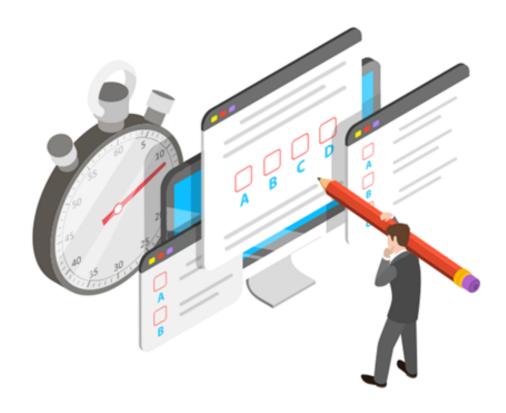
The correct answer is a

The function optimize.minimize() is used to calculate minima. integrate.quad () is used for integral calculation, stats.linregress() is used for linear regression, and linalg.solve() is used to solve a linear system.

5

Which of the following syntaxes is used to generate 100 random variables from a t-distribution with df = 10?

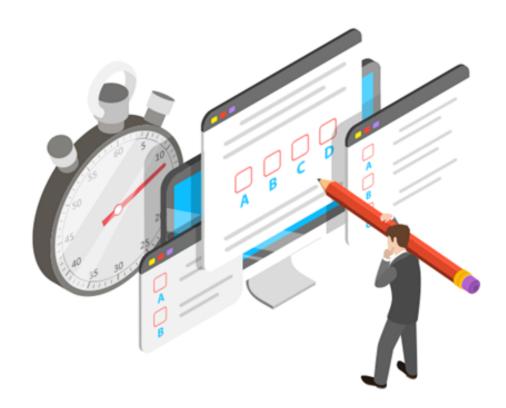
- a. stats.t.pmf(df=10, size=100)
- b. stats.t.pdf(df=10, size=100)
- c. stats.t.rvs(df=10, size=100)
- d. stats.t.rand(df=10, size=100)



5

Which of the following syntaxes is used to generate 100 random variables from a t-distribution with df = 10?

- a. stats.t.pmf(df=10, size=100)
- b. stats.t.pdf(df=10, size=100)
- c. stats.t.rvs(df=10, size=100)
- d. stats.t.rand(df=10, size=100)



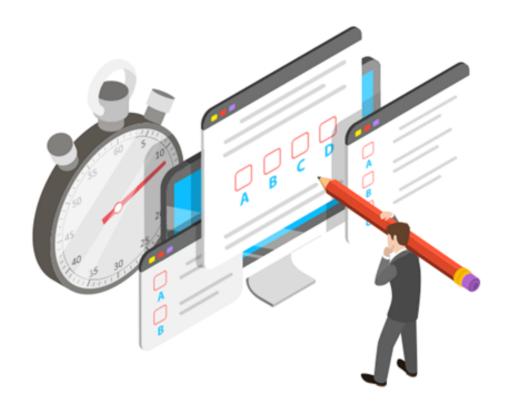
The correct answer is **c**

The stats.t.rvs() function is used to generate random variables. stats.t.pmf() function is used to generate the probability of mass function, and stats.t.pdf() is used to generate probability density function. Note that stats.t.rand () does not exist.

Which of the following functions is used to run C or C++ codes in SciPy?

6

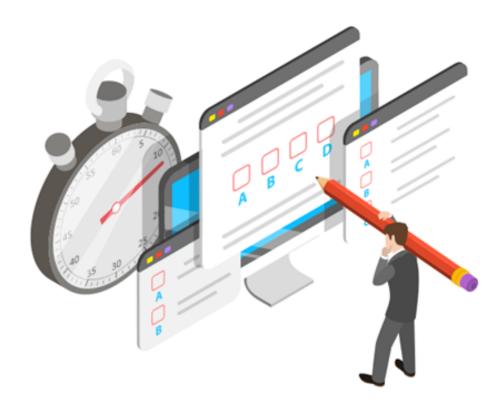
- a. io.loadmat()
- b. weave.inline()
- c. weave.blitz()
- d. io.whosmat()



Which of the following functions is used to run C or C++ codes in SciPy?

6

- io.loadmat() a.
- weave.inline()
- weave.blitz()
- io.whosmat()



The correct answer is **b**

inline() function accepts C codes as string and compiles them for later use. loadmat() loads variables from .mat file. whosmat() checks the variables inside a .mat file.blitz(), and then compiles NumPy expressions for faster running, but it can't accept C codes. simplilearn



Thank You

