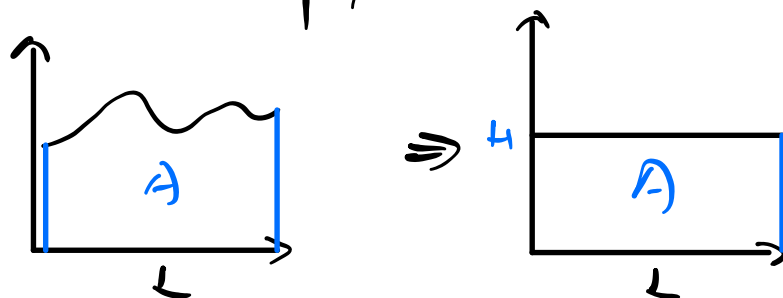


1. Monte Carlo .

随机抽样 / 统计模拟



$$H = \text{Average} [\text{Sum}(\text{曲线的所有值})]$$

2. Newton - Cotes .

`scipy.integrate.newton_cotes(rn, equal=0)`

[\[source\]](#)

Return weights and error coefficient for Newton-Cotes integration.

Suppose we have (N+1) samples of f at the positions x_0, x_1, \dots, x_N . Then an N-point Newton-Cotes formula for the integral between x_0 and x_N is:

$$\int_{x_0}^{x_N} f(x) dx = \Delta x \sum_{i=0}^N a_i f(x_i) + B_N(\Delta x)^{N+2} f^{N+1}(\xi)$$

where $\xi \in [x_0, x_N]$ and $\Delta x = \frac{x_N - x_0}{N}$ is the average samples spacing.

If the samples are equally-spaced and N is even, then the error term is $B_N(\Delta x)^{N+3} f^{N+2}(\xi)$.

Parameters: **rn** : *int*

The integer order for equally-spaced data or the relative positions of the samples with the first sample at 0 and the last at N, where N+1 is the length of *rn*. N is the order of the Newton-Cotes integration.

equal : *int, optional*

Set to 1 to enforce equally spaced data.

Returns: **an** : *ndarray*

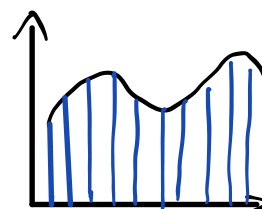
1-D array of weights to apply to the function at the provided sample positions.



B : *float*

Error coefficient.

返回值注意一下.



Spacing : N

$$\int f(x) dx =$$

$$\Delta x \sum w_i \cdot f(x_i) + \text{Error}$$

每部分的权重

3. Simpson / Gauss (Fixed)

`scipy.integrate.simpson(y, x=None, dx=1.0, axis=-1, even='avg')`

[source]

Integrate $y(x)$ using samples along the given axis and the composite Simpson's rule. If x is None, spacing of dx is assumed.

If there are an even number of samples, N , then there are an odd number of intervals ($N-1$), but Simpson's rule requires an even number of intervals. The parameter 'even' controls how this is handled.

$$\int_{x_1}^{x_3} f(x) dx = h \left[\frac{1}{3} f_1 + \frac{4}{3} f_2 + \frac{1}{3} f_3 \right] + O(h^5 f^{(4)})$$

3 points - 2 intervals
in the integration

even : str {'avg', 'first', 'last'}, optional

'avg' : Average two results: 1) use the first $N-2$ intervals with

a trapezoidal rule on the last interval and 2) use the last $N-2$ intervals with a trapezoidal rule on the first interval.

'first' : Use Simpson's rule for the first $N-2$ intervals with

a trapezoidal rule on the last interval.

'last' : Use Simpson's rule for the last $N-2$ intervals with a

trapezoidal rule on the first interval.

实现: $x = \text{np.arange}(0, 10)$
 $y = \text{np.power}(x, 3)$
 $\text{integrate.simpson}(y, x)$
 $\rightarrow 1642.5$

`scipy.integrate.fixed_quad(func, a, b, args=(), n=5)`

[source]

Compute a definite integral using fixed-order Gaussian quadrature.

Integrate $func$ from a to b using Gaussian quadrature of order n .

Parameters: **func** : callable

A Python function or method to integrate (must accept vector inputs). If integrating a vector-valued function, the returned array must have shape $(..., \text{len}(x))$.

a : float

Lower limit of integration.

b : float

Upper limit of integration.

args : tuple, optional

Extra arguments to pass to function, if any.

n : int, optional

Order of quadrature integration. Default is 5.

Returns:

val : float

Gaussian quadrature approximation to the integral

none : None

Statically returned value of None

y - 要集成的阵列.

x - 采样 y 的点

dx - 间距

Google:

对于间隔相等的奇数

个样本. 若 $order \leq 3$.

结果是精确的.

若不相等, 则 $order \leq 2$
才有精确结果.

分为 n 段.

每段用 Simpson 推比一个近似值再加在一起.

Simpson = 梯形法外推.
 \uparrow Accuracy.

\Rightarrow 注意返回值. `fixed_quad(1)[0]`.

4. BVP ODES - shooting Method.

两点值

这里和 IVP 一起写比较好.

单点值.

IVP:

看它会不会给 Method 解释.

```
scipy.integrate.solve_ivp(fun, t_span, y0, method='RK45', t_eval=None,
dense_output=False, events=None, vectorized=False, args=None, **options) [source]
```

Solve an initial value problem for a system of ODEs.

This function numerically integrates a system of ordinary differential equations given an initial value:

```
dy / dt = f(t, y)
y(t0) = y0
```

Here t is a 1-D independent variable (time), $y(t)$ is an N-D vector-valued function (state), and an N-D vector-valued function $f(t, y)$ determines the differential equations. The goal is to find $y(t)$ approximately satisfying the differential equations, given an initial value $y(t_0)=y_0$.

fun: right-hand side of the differential equation $f(t, y)$.

t_span:

Interval of integration (t_0, t_f) . The solver starts with $t=t_0$ and integrates until it reaches $t=t_f$.

y0: an array specifying the initial state.

$\rightarrow \frac{dy}{dt}$ B.C $[0, 20]$
 $y(0) \rightarrow y(20)$
 \rightarrow I.C $[4]$ $y(0)=4$ $y(20)=?$

Return:

t - 时间点 y - t 处的解值.

shooting method.

考虑 BVP: $y'' = f(x, y, y')$ $a < x < b$

Given: $y(a) = \alpha$ $y(b) = \beta$ known

Assume: $y'(a) = \underline{t}$ \leftarrow unknown

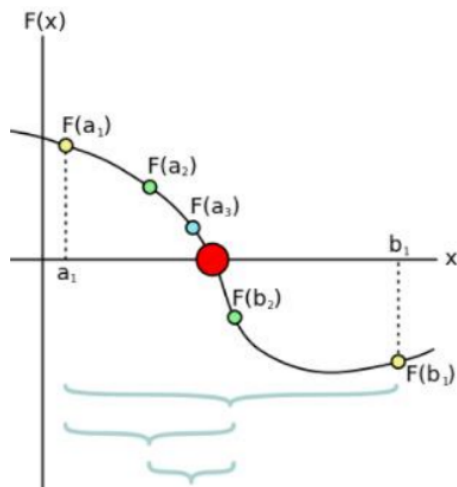
通过选取不同的

t 值, 我们可以得到不同的 $y(b)$ 值.

当 $y(b) \rightarrow \beta$ 时, 此时的 t 即为正确的初值.

5. Bisection / Newton-Raphson .

Bisection . 两点法 .



```
#Bisection
def func(x):
    return x**2-1
print(op.bisect(func, 0, 2)) #0-2区间
```

Raphson

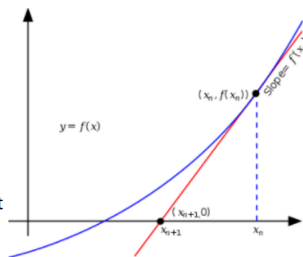
With a single starting point x_0

We know $f(x)$

So, we also know $f'(x_0)$

Find the intersection of the derivative line with the x axis

Use the intersection x coordinate as the starting point
Repeat the algorithm



```
#Newton-Raphson
def f0(x):
    return x**3-1

def f1(x):
    return 3*x**2 #f0的导数
print(op.newton(f0, 1.5, f1)) #估计0点在x=1.5处
```

$$f'(x_n) = \frac{f(x_n) - 0}{x_n - x_{n+1}} \Rightarrow x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} \quad \text{Newton Raphson}$$

Can be generalize to higher order derivatives – Householder's method