

## Diff:

### Differences between given skeleton and solution

In order to make the sample solution easier to understand, the differences between it and the given skeleton source code were highlighted with the help of the program `diff`.

### Legend:

- Gray: unchanged text (only excerpts).
- Green: new lines
- Yellow: changed lines
- Red: deleted lines

Note: Files not listed have not been changed.

This document was created with the help of [diff2html](#) erstellt.

```
diff -u ../course03-numerical-computation/exercise/code/01_simulation.py ../course03-numerical-computation/exercise/solution/01_simulation.py
```

```
../course03-numerical-computation/exercise/code/01_simulation.py
```

```
../course03-numerical-computation/exercise/solution/01_simulation.py
```

```
1 import sys
2 import numpy as np
3 from numpy import r_, pi
4 from matplotlib import pyplot as plt # used for the plotting at the end
5
6
7 # Move the following line further down as you are advancing.
8 # Background: actively exiting the program here prevents errors,
9 # due to usage of undefined names like `XXX`.
10
11
12 sys.exit() # Ends the program here. Otherwise: error messages
13
14
15
16
17 # Task 1:
18
19 # import the function solve_ivp from the package scipy.integrate
20 from scipy.XXX import XXX
21
22 # import two functions for calculating the accelerations
23 # (look inside the file `equations_of_motion.py`!)
24 from equations_of_motion import xdd_fnc #, XXX
25
26
27 # Task 2:
28
29 def rhs(XXX, XXX):
30     # This function calculates the time derivative z_dot from the state z
31     # the 1st argument (the time t) is not needed here
32
33     x, phi, xd, phid = z # unpacking (see overview slides in course01)
34     F = 0
35
36     xdd = xdd_fnc(XXX, XXX, ...)
37     phidd = XXX
38
39
40     # Return the derivative of the state vector
41     z_dot = r_[xd, phid, XXX, XXX]
42     return z_dot
43
44
45
46
47
48 # Task 3:
49
50 zz0 = np.array([XXX, pi*0.5, XXX, XXX])
51
52
53 # Task 4:
54
55 # do the numerical integration
56 res = solve_ivp(XXX, (tt[0], tt[-1]), XXX, t_eval=tt, rtol=1e-5)
```

```
1 import sys
2 import numpy as np
3 from numpy import r_, pi
4 from matplotlib import pyplot as plt # used for the plotting at the end
5
6
7 # Task 1:
8
9 # import the function solve_ivp from the package scipy.integrate
10 from scipy.integrate import solve_ivp
11
12 # import two functions for calculating the accelerations
13 # (look inside the file `equations_of_motion.py`!)
14 from equations_of_motion import xdd_fnc, phidd_fnc
15
16
17 # Task 2:
18
19 def rhs(t, z):
20     # This function calculates the time derivative z_dot from the state z
21     # the 1st argument (the time t) is not needed here
22
23     x, phi, xd, phid = z # unpacking (see overview slides in course01)
24     F = 0
25
26     xdd = xdd_fnc(x, phi, xd, phid, F)
27     phidd = phidd_fnc(x, phi, xd, phid, F)
28
29     # Return the derivative of the state vector
30     z_dot = r_[xd, phid, xdd, phidd]
31     return z_dot
32
33
34
35
36
37 # Task 3:
38
39 zz0 = np.array([0, pi*0.5, 0, 0])
40
41
42 # Task 4:
43
44 # do the numerical integration
45 res = solve_ivp(rhs, (tt[0], tt[-1]), zz0, t_eval=tt, rtol=1e-5)
```



```

24 # import two functions for calculating the accelerations
25 # (look inside the file `equations_of_motion.py`)
26 from equations_of_motion import xdd_fnc, XXX
27
28
29 # load pseudo measurement data in binary format
30 zz_res_target = np.load(XXX)
31
32 # -> this is a 2d array with shape = (4, 1001), i.e. 4 rows, 1001 cols.
33 # meaning of rows: x, phi, xd, phid
34 # meaning of cols: time instant
35
36
37 ## alternatively: load data in text format:
38 # zz_res_target = np.loadtxt(XXX)
39
40
41 # Task 2:
42 :
43 :returns: err - non-negative real valued error measure
44 :         (how "wrong the simulation result is")
45 :
46 :
47 m2, l = XXX # unpacking the parameter vector
48
49 # Task 3:
50 :
51 :
52 Righthand side of the equations of motion
53 (Note: this depends on m2 and l from the surrounding namespace).
54 :
55 x, phi, xd, phid = XXX # unpacking
56 F = 0
57
58 # m2 and l come from the namespace one level higher
59 # (you might want to look again into `equations_of_motion.py`)
60 # to check the signature of these functions:
61 xdd = xdd_fnc(XXXXXXXXX, m2, l)
62 phidd = phidd_fnc(XXXXXXXXXX, XXX, XXX)
63
64 # return derivative of the state vector
65 return XXX
66
67 # end of the inner function definition of rhs
68 :
69 :
70 # them here:
71 :
72 # array with evaluation times (should be consistent with the measured data)
73 tt = np.linspace(0, 10, XXX)
74
75 # select a consistent initial state (4 values) for the simulation
76 # from the measurement data (-> choose the first column)
77 zz0 = XXX
78
79 # do the simulation (get result container)
80 sim_res = solve_ivp(XXX, (XXXX, XX), YYY, t_eval=tt, rtol=1e-5)
81
82

```

```

10 # import two functions for calculating the accelerations
11 # (look inside the file `equations_of_motion.py`)
12 from equations_of_motion import xdd_fnc, phidd_fnc
13
14 # load pseudo measurement data in binary format
15 zz_res_target = np.load('measurement-data.npy')
16
17 # -> this is a 2d array with shape = (4, 1001), i.e. 4 rows, 1001 cols.
18 # meaning of rows: x, phi, xd, phid
19 # meaning of cols: time instant
20
21 ## alternatively: load data in text format:
22 # zz_res_target = np.loadtxt('measurement-data.txt')
23
24
25 # Task 2:
26 :
27 :returns: err - non-negative real valued error measure
28 :         (how "wrong the simulation result is")
29 :
30 :
31 m2, l = p # unpacking the parameter vector
32
33 # Task 3:
34 :
35 :
36 Righthand side of the equations of motion
37 (Note: this depends on m2 and l from the surrounding namespace).
38 :
39 x, phi, xd, phid = z # unpacking
40 F = 0
41
42 # m2 and l come from the namespace one level higher
43 # (you might want to look again into `equations_of_motion.py`)
44 # to check the signature of these functions:
45 xdd = xdd_fnc(x, phi, xd, phid, F, m2, l)
46 phidd = phidd_fnc(x, phi, xd, phid, F, m2, l)
47
48 # return derivative of the state vector
49 return np.array([xd, phid, xdd, phidd])
50
51 # end of the inner function definition of rhs
52 :
53 :
54 # them here:
55 :
56 # array with evaluation times (should be consistent with the measured data)
57 tt = np.linspace(0, 10, 1001)
58
59 # select a consistent initial state (4 values) for the simulation
60 # from the measurement data (-> choose the first column)
61 zz0 = zz_res_target[:, 0]
62
63 # do the simulation (get result container)
64 sim_res = solve_ivp(rhs, (tt[0], tt[-1]), zz0, t_eval=tt, rtol=1e-5)
65
66

```

```

102 # select the state vector (which we call "z" but scipy calls "y")
103 zz_res = XXX.y
104
105 # Task 5:
106
107 # Calculate the difference of the x-positions (first line in each case)
108 # then square (...**2),
109 # then add up (applying np.sum)
110 err = np.sum( (XXX[YYY, ZZZ] - XXX[YYY, ZZZ])**XXX )
111
112 # Status message and output (to assess progress of optimization)
113 print("simulation ready. p =", p, " equation error:", err)
:
119
120 # Task 6:
121
122 p0 = np.array([.5, .7]) # Startschätzung für m2 und l
123
124 # import the function minimize from the module scipy.optimize
125 from XXX.XXX import XXX
126
127 # do the optimization (call the algorithm, which internally repeatedly calls min_target)
128 min_res = minimize(XXX, XXX, method="Nelder-Mead")
129
130 print("\n", "minimization result (data structure):", min_res, "\n")
131 print("estimated parameters (m2, l):", min_res.x, "\n")

85 # select the state vector (which we call "z" but scipy calls "y")
86 zz_res = sim_res.y
87
88 # Task 5:
89
90 # Calculate the difference of the x-positions (first line in each case)
91 # then square (...**2),
92 # then add up (applying np.sum)
93 err = np.sum( (zz_res[0, :] - zz_res_target[0, :])**2 )
94
95 # Status message and output (to assess progress of optimization)
96 print("simulation ready. p =", p, " equation error:", err)
:
102
103 # Task 6:
104
105 p0 = np.array([.5, .7]) # initial guess for m2 and l
106
107 # import the function minimize from the module scipy.optimize
108 from scipy.optimize import minimize
109
110 # do the optimization (call the algorithm, which internally repeatedly calls min_target)
111 min_res = minimize(min_target, p0, method="Nelder-Mead")
112
113
114 print("\n", "minimization result (data structure):", min_res, "\n")
115 print("estimated parameters (m2, l):", min_res.x, "\n")

```