

# Structure and Interpretation of Classical Mechanics with Python and Sagemath

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May 23, 2025

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## PRELIMINARIES

---

### 0.1 README

This is a translation to Python and Sagemath of (most of) the Scheme code of the book ‘Structure and interpretation of classical mechanics’ by Sussman and Wisdom. When referring to *the book*, I mean their book. I expect the reader to read the related parts of the book, and use the Python code to understand the Scheme code of the book (and vice versa). I therefore don’t explain much of the logic of the code in this document. I’ll try to stick to the naming of functions and variables as used in the book. I also try to keep the functional programming approach of the book; consequently, I don’t strive to the most pythonic code possible. To keep the code clean, I never protect functions against stupid input; realize that this is research project, the aim is not to produce a fool-proof software product.

- The file `sicm_sagemath.pdf` shows all code samples together with the output when running the code.
- The directory `org` contains the org files.
- The directory `sage` contains all sage files obtained from tangling the org files.

In the pdf file I tend to place explanations, comments, and observations about the code and the results *above* the code blocks.

I wrote this document in Emacs and Org mode. When developing, I first made a sage file with all code for a specific section of the book. Once all worked, I copied the code to an Org file and make code blocks. Then I tangled, for instance, generally useful code of `section1.4.org` to `utils1.4.sage` and to `section1.4.sage` for code specific for Section 1.4 of the book. This way I can load the utils files at later stages.

I found it convenient to test things in a `tests.sage` file. Then, I could edit within emacs and see the consequences directly in the sage session by opening a sage session on the command prompt and attaching the session to the file like so:

```
sage: attach("tests.sage")
```

Finally, here are some resources that were helpful to me:

- An online version of the book: <https://tgvaughan.github.io/sicm/>
- An org file of the book with Scheme: <https://github.com/mentat-collective/sicm-book/blob/main/org/chapter001.org>

- A port to Clojure: <https://github.com/sicmutils/sicmutils>
- The Sagemath reference guide: <https://doc.sagemath.org/html/en/reference/>
- Handy tuples: <https://github.com/jtauber/functional-differential-geometry>
- ChatGPT proved to be a great help in the process of becoming familiar with Scheme and Sagemath.
- Some solutions to problems: <https://github.com/hnarayanan/sicm>

In the next sections we provide Python and Sagemath code for background functions that are used, but not defined, in the book.

## 0.2 `todo` OUTPUT TO L<sup>A</sup>T<sub>E</sub>X

We need some tricks to adapt the L<sup>A</sup>T<sub>E</sub>X output of Sagemath to our liking.

We use `re` to modify LaTeX strings. I discovered the two latex options from this site: [Sage, LaTeX and Friends](#).

---

```

1 import re
2
3 latex.matrix_delimiters(left='[', right=']')
4 latex.matrix_column_alignment("c")

```

---

Note in passing that the title of the code block shows the file to which the code is tangled, and if a code block is not tangled, the title says "don't tangle".

To keep the formulas short in L<sup>A</sup>T<sub>E</sub>X, I remove all strings like  $(t)$ , and replace  $\partial x / \partial t$  by  $\dot{x}$ . This is the job of the regular expressions below.

---

```

5 def simplify_latex(s):
6     s = re.sub(r"\\frac{\\partial}{\\partial t}", r"\\dot ", s)
7     s = re.sub(r"\\left(t\\right)", r"", s)
8     s = re.sub(
9         r"\\frac{\\partial^{2}}{\\partial t^{2}}",
10        r"\\ddot ",
11        s,
12    )
13    return s

```

---

The function `show_expression` prints expressions to L<sup>A</sup>T<sub>E</sub>X. There is a caveat, though. When `show_expression` would return a string, org mode (or perhaps Python) adds many escape symbols for the `\` character, which turns out to ruin the L<sup>A</sup>T<sub>E</sub>X output in an org file. For this reason, I just call `print` ; for my purposes (writing these files in emacs and org mode) it works the way I want.

---

```

14 def show_expression(s, simplify=True):
15     s = latex(s)
16     if simplify:
17         s = simplify_latex(s)
18     res = r"\begin{dmath*}"
19     res += "\n" + s + "\n"
20     res += r"\end{dmath*}"
21     print(res)

```

---

### 0.2.1 Printing with org mode

There is a subtlety with respect to printing in org mode and in tangled files. When working in sage files, and running them from the prompt, I call `show(expr)` to have some expression printed to the screen. So, when running Sage from the prompt, I do *not* want to see L<sup>A</sup>T<sub>E</sub>X output. However, when executing a code block in org mode, I *do* want to get L<sup>A</sup>T<sub>E</sub>X output. For this, I could use the book's `show_expression` in the code blocks in the org file. So far so good, but now comes the subtlety. When I *tangle* the code from the org file to a sage file, I don't want to see `show_expression`, but just `show`. Thus, I should use `show` throughout, but in the org mode file, `show` should call `show_expression`. To achieve this, I include the following `show` function in org mode, but I don't tangle it to the related sage files.

---

```

22 def show(s, simplify=True):
23     return show_expression(s, simplify)

```

---

## 0.3 THE TUPLE CLASS

The book uses up tuples quite a bit. This code is a copy of `tuples.py` from <https://github.com/jtauber/functional-differential-geometry>. See `tuples.rst` in that repo for further explanations.

---

```

24 """
25 This is a copy of tuples.py from
26 https://github.com/jtauber/functional-differential-geometry.
27 """
28
29 from sage.structure.element import Matrix, Vector
30
31 class Tuple:
32     def __init__(self, *components):
33         self._components = components
34
35     def __getitem__(self, index):

```

---

```

36         return self._components[index]
37
38     def __len__(self):
39         return len(self._components)
40
41     def __eq__(self, other):
42         if (
43             isinstance(other, self.__class__)
44             and self._components == other._components
45         ):
46             return True
47         else:
48             return False
49
50     def __ne__(self, other):
51         return not (self.__eq__(other))
52
53     def __add__(self, other):
54         if isinstance(self, Tuple):
55             if not isinstance(other, self.__class__) or len(self) != len(
56                 other
57             ):
58                 raise TypeError("can't add incompatible Tuples")
59             else:
60                 return self.__class__(
61                     *(
62                         s + o
63                         for (s, o) in zip(self._components, other._components)
64                     )
65                 )
66         else:
67             return self + other
68
69     def __iadd__(self, other):
70         return self + other
71
72     def __neg__(self):
73         return self.__class__(*(-s for s in self._components))
74
75     def __sub__(self, other):
76         return self + (-other)
77
78     def __isub__(self, other):
79         return self - other
80
81     def __call__(self, **kwargs):
82         return self.__class__(
83             *(
84                 (c(**kwargs) if isinstance(c, Expr) else c)
85                 for c in self._components
86             )
87         )

```

```

88
89 def subs(self, args):
90     # substitute variables with args
91     return self.__class__(*(c.subs(args) for c in self._components))
92
93 def list(self):
94     "convert tuple and its components to one list."
95     result = []
96     for comp in self._components:
97         if isinstance(comp, (Tuple, Matrix, Vector)):
98             result.extend(comp.list())
99         else:
100             result.append(comp)
101     return result
102
103 def derivative(self, var):
104     "Compute the derivative of all components and put the result in a tuple."
105     return self.__class__(
106         *[derivative(comp, var) for comp in self._components]
107     )

```

---

We have up tuples and down tuples. They differ in the way they are printed.

---

```

108 class UpTuple(Tuple):
109     def __repr__(self):
110         return "up({})".format(", ".join(str(c) for c in self._components))
111
112     def _latex_(self):
113         "Print up tuples vertically."
114         res = r"\begin{array}{c}"
115         for comp in self._components:
116             res += r"\begin{array}{c}"
117             res += latex(comp)
118             res += r"\end{array}"
119             res += r" \\"
120         res += r"\end{array}"
121         return res
122
123 class DownTuple(Tuple):
124     def __repr__(self):
125         return "down({})".format(", ".join(str(c) for c in self._components))
126
127     def _latex_(self):
128         "Print down tuples horizontally."
129         res = r"\begin{array}{c}"
130         for comp in self._components:
131             res += r"\begin{array}{c}"
132             res += latex(comp)
133             res += r"\end{array}"
134             res += r" & "
135         res += r"\end{array}"

```

```

136         return res
137
138     up = UpTuple
139     down = DownTuple
140
141     up._dual = down
142     down._dual = up

```

---

Here is some functionality to unpack tuples. I don't use it for the moment, but it is provided by the `tuples.py` package that I donwloaded from the said github repo.

```

----- ../sage/tuples.sage -----
143 def ref(tup, *indices):
144     if indices:
145         return ref(tup[indices[0]], *indices[1:])
146     else:
147         return tup
148
149
150 def component(*indices):
151     def _(tup):
152         return ref(tup, *indices)
153
154     return _

```

---

## 0.4 FUNCTIONAL PROGRAMMING WITH PYTHON FUNCTIONS

In this section we set up some generic functionality to support the summation, product, and composition of functions:

$$\begin{aligned}
 (f + g)(x) &= f(x) + g(x), \\
 (fg)(x) &= f(x)g(x), \\
 (f \circ g)(x) &= f(g(x)).
 \end{aligned}$$

This is easy to code with recursion.

### 0.4.1 Standard imports

```

----- ../sage/functions.sage -----
155 load("tuples.sage")

```

---

We need to load `functions.sage` to run the examples in the test file.

```

----- ../sage/functions_tests.sage -----
156 load("functions.sage")

```

---

We load `show_expression` to control the  $\text{\LaTeX}$  output in this org file.



---

don't tangle

---

```
157 load("show_expression.sage")
```

---

### 0.4.2 The Function class

The Function class provides the functionality we need for functional programming.

---

```
158 class Function:
159     def __init__(self, func):
160         self._func = func
161
162     def __call__(self, *args):
163         return self._func(*args)
164
165     def __add__(self, other):
166         return Function(lambda *args: self(*args) + other(*args))
167
168     def __neg__(self):
169         return Function(lambda *args: -self(*args))
170
171     def __sub__(self, other):
172         return self + (-other)
173
174     def __mul__(self, other):
175         if isinstance(other, Function):
176             return Function(lambda *args: self(*args) * other(*args))
177         return Function(lambda *args: other * self(*args))
178
179     def __rmul__(self, other):
180         return self * other
181
182     def __pow__(self, exponent):
183         if exponent == 0:
184             return Function(lambda x: 1)
185         else:
186             return self * (self ** (exponent - 1))
```

---

The next function decorates a function `f` that returns another function `inner_f`, so that `inner_f` becomes a `Function`.

---

```
187 def Func(f):
188     def wrapper(*args, **kwargs):
189         return Function(f(*args, **kwargs))
190
191     return wrapper
```

---

Below I include an example to see how to use, and understand, this decorator. Composition is just a recursive call of functions.

---

```

192 @Func
193 def compose(*funcs):
194     if len(funcs) == 1:
195         return lambda x: funcs[0](x)
196     return lambda x: funcs[0](compose(*funcs[1:])(x))

```

---

### 0.4.3 Some standard functions

To use python functions as Functions, use lambda like this.

---

```

197 def f(x):
198     return 5 * x
199
200
201 F = Function(lambda x: f(x))

```

---

The identity is just interesting. Perhaps we'll use it later.

---

```

202 identity = Function(lambda x: x)

```

---

To be able to code things like  $(\sin + \cos)(x)$  we need to postpone the application of  $\sin$  and  $\cos$  to their arguments. Therefore we override their definitions.

---

```

203 sin = Function(lambda x: sage.functions.trig.sin(x))
204 cos = Function(lambda x: sage.functions.trig.cos(x))

```

---

We will use quadratic functions often.

---

```

205 from functools import singledispatch
206
207
208 @singledispatch
209 def _square(x):
210     raise TypeError(f"Unsupported type: {type(x)}")
211
212
213 @_square.register(int)
214 @_square.register(float)
215 @_square.register(Expression)
216 @_square.register(Integer)
217 def _(x):
218     return x ^ 2
219
220

```

---

```

221 @_square.register(Vector)
222 @_square.register(list)
223 @_square.register(tuple)
224 def _(x):
225     v = vector(x)
226     return v.dot_product(v)
227
228
229 @_square.register(Matrix)
230 def _(x):
231     if x.ncols() == 1:
232         return (x.T * x)[0, 0]
233     elif x.nrows() == 1:
234         return (x * x.T)[0, 0]
235     else:
236         raise TypeError(
237             f"Matrix must be a row or column vector, got shape {x.nrows()}x{x.ncols()}"
238         )
239
240
241 square = Function(lambda x: _square(x))

```

---

To use Sagemath functions we make an abbreviation.

```

..... ../sage/functions.sage .....
242 function = sage.symbolic.function_factory.function

```

---

Now we can make symbolic functions like so.

```

..... ../sage/functions_tests.sage .....
243 V = Function(lambda x: function("V")(x))

```

---

#### 0.4.4 Examples

```

..... ../sage/functions_tests.sage .....
244 x, y = var("x y", domain = RR)
245
246 show((square)(x + y).expand())

```

---

$$x^2 + 2xy + y^2$$

```

..... ../sage/functions_tests.sage .....
247 show((square + square)(x + y))

```

---

$$2(x + y)^2$$

```

..... ../sage/functions_tests.sage .....
248 show((square * square)(x))

```

---

$$x^4$$

---

```

249  show((sin + cos)(x))

```

---

$$\cos(x) + \sin(x)$$

---

```

250  show((square + V)(x))

```

---

$$x^2 + V(x)$$

---

```

251  hh = compose(square, sin)
252  show((hh + hh)(x))

```

---

$$2 \sin(x)^2$$

We know that  $2 \sin x \cos x = \sin(2x)$ .

---

```

253  show((2 * (sin * cos)(x) - sin(2 * x)).simplify_full())

```

---

$$0$$

Next, we test differentiation and integration.

---

```

254  show(diff(-compose(square, cos)(x), x))
255  show(integrate((2 * sin * cos)(x), x))

```

---

$$2 \cos(x) \sin(x)$$

$$-\cos(x)^2$$

Arithmetic with symbolic functions works too.

---

```

256  U = Function(lambda x: function("U")(x))
257  V = Function(lambda x: function("V")(x))

```

---



---

```

258  show((U + V)(x))
259  show((V + V)(x))
260  show((V(U(x))))
261  show((compose(V, U)(x)))

```

---

$$U(x) + V(x)$$

$$2V(x)$$

$$V(U(x))$$

$$V(U(x))$$

---

```

262 def f(x):
263     def g(y):
264         return x * y ^ 2
265
266     return g

```

---



---

```

267 show(f(3)(5))

```

---

75

However, we cannot apply algebraic operations on `f`. For instance, this does not work; it gives `TypeError: unsupported operand type(s) for +: 'function' and 'function'`.

---

```

268 show((f(3) + f(2))(4))

```

---

By decoration with `@Func` we get what we need.

---

```

269 @Func
270 def f(x):
271     def g(y):
272         return x * y ^ 2
273
274     return g

```

---



---

```

275 show((f(3) + f(2))(4))

```

---

80

Indeed:  $(3 + 2) * 4^2 = 80$ .

Decorating with `@Func` is the same as this.

---

```

276 def f(x):
277     def g(y):
278         return x * y ^ 2
279
280     return Function(lambda y: g(y))

```

---



---

```

281 show((f(3) + f(2))(4))

```

---

80

## 0.5 DIFFERENTIATION

0.5.1 *Standard imports*

```

_____/sage/differentiation.sage_____
282 load(
283     "functions.sage",
284     "tuples.sage",
285 )
_____

_____/sage/differentiation_tests.sage_____
286 load("differentiation.sage")
287
288 var("t", domain="real")
_____

_____ don't tangle _____
289 load("show_expression.sage")
_____

```

0.5.2 *Examples with matrices, functions and tuples*

```

_____/sage/differentiation_tests.sage_____
290 _ = var("a b c x y", domain=RR)
291 M = matrix([[a, b], [b, c]])
292 b = vector([a, b])
293 v = vector([x, y])
294 F = 1 / 2 * v * M * v + b * v + c
_____

_____/sage/differentiation_tests.sage_____
295 show(F)
_____


$$\frac{1}{2}(ax + by)x + ax + \frac{1}{2}(bx + cy)y + by + c$$


_____/sage/differentiation_tests.sage_____
296 show(F.expand())
_____


$$\frac{1}{2}ax^2 + bxy + \frac{1}{2}cy^2 + ax + by + c$$


_____/sage/differentiation_tests.sage_____
297 show(diff(F, x))
_____

```

$$ax + by + a$$

Repeated differentiation works nicely.

---

```

298 ..sage/differentiation_tests.sage
show(diff(F, [x, y]))

```

---

$$b$$

This is the Jacobian.

---

```

299 ..sage/differentiation_tests.sage
show(jacobian(F, [x, y]))

```

---

$$\begin{bmatrix} ax + by + a & bx + cy + b \end{bmatrix}$$

---

```

300 ..sage/differentiation_tests.sage
show(jacobian(F, v.list())) # convert the column matrix to a list

```

---

$$\begin{bmatrix} ax + by + a & bx + cy + b \end{bmatrix}$$

This expression gives an error.

---

```

301 ..sage/differentiation_tests.sage
diff(F, v) # v is not a list, but a vector

```

---

To differentiate a Python function we need to provide the arguments to the function.

---

```

302 ..sage/differentiation_tests.sage
def F(v):
303     return 1 / 2 * v * M * v + b * v + c

```

---



---

```

304 ..sage/differentiation_tests.sage
show(diff(F(v), x)) # add the arguments to F
305 show(jacobian(F(v), v.list()))

```

---

$$ax + by + a$$

$$\begin{bmatrix} ax + by + a & bx + cy + b \end{bmatrix}$$

The next two examples do not work.

---

```

306 ..sage/differentiation_tests.sage
jacobian(F, v) # F has no arguments
307 jacobian(F(v), v) # v is not a list

```

---

The Tuple class supports differentiation.

---

```

308 ..sage/differentiation_tests.sage
T = up(t, t ^ 2, t ^ 3, sin(3 * t))
309 show(diff(T, t))

```

---

$$\begin{matrix} 1 \\ 2t \\ 3t^2 \\ 3\cos(3t) \end{matrix}$$

### 0.5.3 Differentiation with respect to time

The function  $D$  takes a function (of time) as argument, and returns the derivative with respect to time:

$$D(f(\cdot)): t \rightarrow f'(t).$$

---

```

310 @Func
311 def D(f):
312     return lambda t: diff(f(t), t)
313     #return derivative(expr, t)

```

---

Here is an example.

---

```

314 q = Function(lambda t: function("q")(t))
315
316 show(D(q)(t))

```

---

$$\dot{q}$$

### 0.5.4 Differentiation with respect to function arguments

The Euler-Lagrange equations depend on the partial derivative of a Lagrangian  $L$  with respect to  $q$  and  $v$ , and a total derivative with respect to time. Now  $q$  and  $v$  will often be functions of time, so we need to find a way to differentiate with respect to *functions*, like  $q(\cdot)$ , rather than just symbols, like  $x$ . To implement this in Sagemath turned out to be far from easy, at least for me.

First, observe that the Jacobian in Sagemath takes as arguments a function and the variables with respect to which to take the derivatives. So, I tried this first:

---

```

317 q = Function(lambda t: function("q")(t))

```

---

But the next code gives errors saying that the argument  $q$  should be a symbolic function, which it is not.

---

```

318 F = 5 * q + 3 * t
319
320 show(diff(F, r)) # does not work
321 show(jacobian(F, [q, t])) # does not work

```

---

To get around this problem, I use the following strategy to differentiate a function  $F$  with respect to functions.



1. Make a list of dummy symbols, one for *each argument* of  $F$  that is *not a symbol*. To understand this in detail, observe that arguments like  $t$  or  $x$  are symbols, but such symbols need not be protection. In other words: we don't have to replace a symbol by another symbol, because Sagemath can already differentiate wrt symbols; it's the other 'things' are the things that have to be replaced by a variable. Thus, arguments like  $q(t)$  that are *not* symbols have to be protected by replacing them with dummy symbols.
2. Replace in  $F$  the arguments by their dummy variables. We use the Sagemath `subs` functionality of Sagemath to substitute the dummy variables for the functions. Now there is one further problem: `subs` does not work on lists or tuples. However, `subs` *does work* on vectors and matrices. Therefore, we cast all relevant lists to vectors, which suffices for our goal.
3. Take the Jacobian of  $F$  with respect to the dummy symbols. We achieve this by substituting the dummy symbols in the vector of arguments and the vector of variables.
4. Invert: Replace in the final result the dummy symbols by their related arguments.

We use `id(v)` to create a unique variable name for each dummy variable and store the mapping from the functions to the dummy variables in a dictionary `subs`. (As these are internal names, the actual variable names are irrelevant; as long as they are unique, it's OK.)

We know from the above that `jacobian` expects a *list* with the variables with respect to which to differentiate. Therefore, we turn the vector with substituted variables to a list.

---

```

322 def Jacobian(F):
323     def f(args, vrs):
324         if isinstance(args, (list, tuple)):
325             args = vector(args)
326         if isinstance(vrs, (list, tuple)):
327             vrs = vector(vrs)
328         subs = {
329             v: var(f"v{id(v)}", domain=RR)
330             for v in args.list()
331             if not v.is_symbol()
332         }
333         result = jacobian(F(args.subs(subs)), vrs.subs(subs).list())
334         inverse_subs = {v: k for k, v in subs.items()}
335         return result.subs(inverse_subs)
336
337     return f

```

---

Here are some examples to see how to use this Jacobian. Note that Jacobian expects the arguments and variables to be *lists*, or list like. As a result, in the function  $F$  we have to unpack the list.

---

```

338 v = var("v", domain=RR)
339
340
341 def F(v):
342     r, t = v.list()
343     return 5 * r ^ 3 + 3 * t ^ 2 * r
344
345
346 show(Jacobian(F)([v, t], [t]))
347 show(Jacobian(F)([v, t], [v, t]))

```

---

$$\begin{bmatrix} 6tv \end{bmatrix}$$

$$\begin{bmatrix} 3t^2 + 15v^2 & 6tv \end{bmatrix}$$

This works. Now we try the same with a function like argument. Recall,  $v$  must be a list for partial on which gradient depends.

---

```

348 q = Function(lambda t: function("q")(t))
349 v = [q(t), t]
350 show(Jacobian(F)(v, v))

```

---

$$\begin{bmatrix} 3t^2 + 15q^2 & 6tq \end{bmatrix}$$

### 0.5.5 Gradient and Hessian

Next we build the gradient. We can use Sagemath's jacobian, but as is clear from above, we need to indicate explicitly the variable names with respect to which to differentiate. Moreover, we like to be able to take the gradient with respect to literal functions. Thus, we use the Jacobian defined above.

One idea for the gradient is like this. However, this does not allow to use gradient as a function in functional composition.

---

```

351 def gradient(F, v):
352     return Jacobian(F)(v, v).T

```

---

We therefore favor the next implementation. BTW, note that the gradient is a vector in a tangent space, hence it is column vector. For that reason we transpose the Jacobian.

---

```

353 def gradient(F):
354     return lambda v: Jacobian(F)(v, v).T

```

---

---

```

355  ..../sage/differentiation_tests.sage
show(gradient(F)(v))

```

---

$$\begin{bmatrix} 3t^2 + 15q^2 \\ 6tq \end{bmatrix}$$

When differentiating a symbolic function, wrap such a function in a Function.

---

```

356  ..../sage/differentiation_tests.sage
U = Function(lambda x: function("U")(square(x)))
357  show(gradient(U)(v))

```

---

$$\begin{bmatrix} 2qD_0(U)(t^2 + q^2) \\ 2tD_0(U)(t^2 + q^2) \end{bmatrix}$$

The Hessian can now be defined as the composition of the gradient with itself.

---

```

358  ..../sage/differentiation.sage
def Hessian(F):
359      return lambda v: compose(gradient, gradient)(F)(v)

```

---

```

360  ..../sage/differentiation_tests.sage
show(Hessian(F)(v))

```

---

$$\begin{bmatrix} 30q & 6t \\ 6t & 6q \end{bmatrix}$$

### 0.5.6 Differentiation with respect to slots

To follow the notation of the book, we need to define a python function that computes partial derivatives with respect to the slot of a function; for example, in  $\partial_1 L$  the 1 indicates that the partial derivatives are supposed to be taken wrt the coordinate variables. The Jacobian function built above allows us a very simple solution. Note that we return a Function so that we can use this operator in functional composition if we like.

---

```

361  ..../sage/differentiation.sage
@Func
362  def partial(f, slot):
363      def wrapper(local):
364          if slot == 0:
365              selection = [time(local)]
366          elif slot == 1:
367              selection = coordinate(local)
368          elif slot == 2:
369              selection = velocity(local)
370          return Jacobian(f)(local, selection)
371
372  return wrapper

```

---

The main text contains many examples.

## CHAPTER 1

---

### 1.4 COMPUTING ACTIONS

#### 1.4.1 *Standard setup*

I create an Org file for each separate section of the book; for this section it's `section1.4.org`. Code that is useful for later sections is tangled to `utils1.4.sage` and otherwise to `section1.4.sage`. This allows me to run the sage scripts on the prompt. Note that the titles of the code blocks correspond to the file to which the code is written when tangled.

---

```

373 import numpy as np
374
375 load("functions.sage", "differentiation.sage", "tuples.sage")

```

---

BTW, don't do `from sage.all import *` because that will lead to name space conflicts, for instance with the Gamma function which we define below.

---

```

376 load("utils1.4.sage")
377
378 t = var("t", domain="real")

```

---

The next module is used for nice printing in org mode files; it should only be loaded in org mode files.

---

```

379 load("show_expression.sage")

```

---

#### 1.4.2 *The Lagrangian for a free particle.*

The function `L_free_particle` takes `mass` as an argument and returns the (curried) function `Lagrangian` that takes a local tuple as an argument.

---

```

380 def L_free_particle(mass):
381     def Lagrangian(local):
382         v = velocity(local)
383         return 1 / 2 * mass * square(v)
384
385     return Lagrangian

```

---

For the next step, we need a *literal functions* and *coordinate paths*.

### 1.4.3 Literal functions

A `literal_function` maps the time  $t$  to a coordinate or velocity component of the path, for instance,  $t \rightarrow x(t)$ . Since we need to perform arithmetic with literal functions, see below for some examples, we encapsulate it in a Function.

---

```

386 @Func
387 def literal_function(name):
388     return lambda t: function(name)(t)

```

---

It's a function.

---

```

389 x = literal_function("x")
390 print(x)

```

---

<\_\_main\_\_.Function object at 0x71122066e470>

Here are some operations on  $x$ .

---

```

391 show(x(t))
392 show((x+x)(t))
393 show(square(x)(t))

```

---

Note that, to keep the notation brief, the  $t$  is suppressed in the  $\text{\LaTeX}$  output.

### 1.4.4 Paths

We will represent coordinate path functions  $q$  and velocity path functions  $v$  as functions that map time to vectors. Thus, `column_path` returns a function of time, not yet a path. We also need to perform arithmetic on paths, like  $3q$ , therefore we encapsulate the path in a Function.

---

```

394 @Func
395 def column_path(lst):
396     #return lambda t: vector([l(t) for l in lst])
397     return lambda t: column_matrix([l(t) for l in lst])

```

---

---

```

398 q = column_path(
399     [
400         literal_function("x"),
401         literal_function("y"),
402     ]
403 )

```

---

Here is an example to see how to use q.

---

```

404 show(q(t))

```

---

$$\begin{bmatrix} x \\ y \end{bmatrix}$$

---

```

405 show((q + q)(t))

```

---

$$\begin{bmatrix} 2x \\ 2y \end{bmatrix}$$

---

```

406 show((2 * q)(t))

```

---

$$\begin{bmatrix} 2x \\ 2y \end{bmatrix}$$

---

```

407 show((q * q)(t))

```

---

### 1.4.5 Gamma function

The Gamma function lifts a coordinate path to a function that maps time to a local tuple of the form  $(t, q(t), v(t), \dots)$ . That is,

$$\begin{aligned}\Gamma[q](\cdot) &= (\cdot, q(\cdot), v(\cdot), \dots), \\ \Gamma[q](t) &= (t, q(t), v(t), \dots).\end{aligned}$$

To follow the conventions of the book, we use an up tuple for Gamma. However, I don't build the coordinate path nor the velocity as up tuples because I find Sagemath vectors more convenient.

$\Gamma$  just receives  $q$  as an argument. Then it computes the velocity  $v = Dq$ , from which the acceleration follows recursively as  $a = Dv, \dots$ . Recall that  $D$  computes the derivative (wrt time) of a function that depends on time.

When  $n = 3$ , it returns a function of time that produces the first three elements of the local tuple  $(t, q(t), v(t))$ . This is the default. Once all derivatives are computed, we convert the result to a function that maps time to an up tuple.

---

```

408 def Gamma(q, n=3):
409     # if isinstance(q, np.ndarray):
410     #     q = vector(q.tolist()) # todo, is this still needed?
411
412     if n < 2:
413         raise ValueError("n must be > 1")
414     Dq = [q]
415     for k in range(2, n):
416         Dq.append(D(Dq[-1]))
417     return lambda t: up(t, *[v(t) for v in Dq])

```

---

When applying Gamma to a path, we get this.

---

```

418 local = Gamma(q)(t)
419 show(local)

```

---

$$\begin{matrix} t \\ \left[ \begin{matrix} x \\ y \\ \dot{x} \\ \dot{y} \end{matrix} \right] \end{matrix}$$

We can include the acceleration too.

---

```

420 show(Gamma(q, 4)(t))

```

---

$$\begin{matrix} t \\ \left[ \begin{matrix} x \\ y \\ \dot{x} \\ \dot{y} \\ \ddot{x} \\ \ddot{y} \end{matrix} \right] \end{matrix}$$

todo: revise the definitions of time, coordiante, velocity, below.

Finally, here are some projections operators from the local tuple to supspaces.

---

```

421 time = Function(lambda local: local[0])
422 coordinate = Function(lambda local: local[1])
423 velocity = Function(lambda local: local[2])

```

---



---

```

424 show(compose(velocity, Gamma(q))(t))

```

---

$$\left[ \begin{matrix} \dot{x} \\ \dot{y} \end{matrix} \right]$$

## 1.4.6 Continuation with the free particle.

Now we know how to build literal functions and  $\Gamma$ , we can continue with the Lagrangian of the free particle.

---

```

425 q = column_path(
426     [
427         literal_function("x"),
428         literal_function("y"),
429         literal_function("z"),
430     ]
431 )

```

---



---

```

432 show(q(t))

```

---

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

---

```

433 show(D(q)(t))

```

---

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{z} \end{bmatrix}$$

---

```

434 show(Gamma(q)(t))

```

---

$$\begin{bmatrix} t \\ x \\ y \\ z \\ \dot{x} \\ \dot{y} \\ \dot{z} \end{bmatrix}$$

The Lagrangian of a free particle with mass  $m$  applied to the path  $\Gamma$  gives this. Our first implementation is like this:  $L(\Gamma[q](t))$ , that is,  $\Gamma[q](t)$  makes a local tuple, and this is given as argument to  $L$ .

---

```

435 load("functions.sage")
436 m = var('m', domain='positive')
437 show(L_free_particle(m)(Gamma(q)(t)))

```

---



$$\frac{1}{2}(\dot{x}^2 + \dot{y}^2 + \dot{z}^2)m$$

Here is the implementation of the book:  $(L \circ \Gamma[q])(t)$ , that is,  $L \circ \Gamma[q]$  is a function that depends on  $t$ . Note how the brackets are placed after `Gamma(q)`.

---

```
438  ..../sage/section1.4.sage
show(compose(L_free_particle(m), Gamma(q))(t))
```

---

$$\frac{1}{2}(\dot{x}^2 + \dot{y}^2 + \dot{z}^2)m$$

We now compute the integral of Lagrangian  $L$  along the path  $q$ , but for this we need a function to carry out 1D integration (along time in our case). Of course, Sagemath already supports a definite integral in a library.

---

```
439  ..../sage/utills1.4.sage
from sage.symbolic.integration.integral import definite_integral
```

---

I don't like to read  $dt$  at the end of the integral because  $dt$  reads like the product of the variables  $d$  and  $t$ . Instead, I prefer to read  $dt$ ; for this reason I overwrite the L<sup>A</sup>T<sub>E</sub>X formatting of `definite_integral`.

---

```
440  ..../sage/utills1.4.sage
def integral_latex_format(*args):
441      expr, var, a, b = args
442      return (
443          fr"\int_{{{a}}}}^{{{b}}}} "
444          + latex(expr)
445          + r"\, \text{rm}{d}\,",
446          + latex(var)
447      )
448
449
450  definite_integral._print_latex_ = integral_latex_format
```

---

Here is the action along a generic path  $q$ .

---

```
451  ..../sage/section1.4.sage
T = var("T", domain="positive")
452
453  def Lagrangian_action(L, q, t1, t2):
454      return definite_integral(compose(L, Gamma(q))(t), t, t1, t2)
455
456  show(Lagrangian_action(L_free_particle(m), q, 0, T))
```

---

$$\frac{1}{2}m\left(\int_0^T \dot{x}^2 dt + \int_0^T \dot{y}^2 dt + \int_0^T \dot{z}^2 dt\right)$$

To get a numerical answer, we take the test path of the book. Below we'll do some arithmetic with `test_path`; therefore we encapsulate it in a Function.

---

```

457 test_path = Function(lambda t: vector([4 * t + 7, 3 * t + 5, 2 * t + 1]))
458 show(Lagrangian_action(L_free_particle(mass=3), test_path, 0, 10))

```

---

435

Let's try a harder path. We don't need this later, so the encapsulation in Function is not necessary.

---

```

459 hard_path = lambda t: vector([4 * t + 7, 3 * t + 5, 2 * exp(-t) + 1])
460
461 result = Lagrangian_action(L_free_particle(mass=3), hard_path, 0, 10)
462 show(result)
463 show(float(result))

```

---

$$3(125e^{20} - 1)e^{(-20)} + 3$$

377.9999999938165

The value of the integral is different from 435 because the end points of this harder path are not the same as the end points of the test path.

#### 1.4.7 Path of minimum action

First some experiments to see whether my code works as intended.

---

```

464 @Func
465 def make_eta(nu, t1, t2):
466     return lambda t: (t - t1) * (t - t2) * nu(t)
467
468
469 nu = Function(lambda t: vector([sin(t), cos(t), t ^ 2]))
470
471 show((1 / 3 * make_eta(nu, 3, 4) + test_path)(t))

```

---

$$\left( \frac{1}{3}(t-3)(t-4)\sin+4t+7, \frac{1}{3}(t-3)(t-4)\cos+3t+5, \frac{1}{3}(t-3)(t-4)t^2+2t+1 \right)$$

In the next code, I add the `n()` to force the result to a floating point number. (Without this, the result is a long expression with lots of cosines and sines.)

---

```

472 def varied_free_particle_action(mass, q, nu, t1, t2):
473     eta = make_eta(nu, t1, t2)
474

```

---

```

475     def f(eps):
476         return Lagrangian_action(L_free_particle(mass), q + eps * eta, t1, t2).n()
477
478     return f
479
480 show(varied_free_particle_action(3.0, test_path, nu, 0.0, 10.0)(0.001))

```

---

436.291214285714

By comparing our result with that of the book, we see we are still on track.  
Now use Sagemath's `find_local_minimum` to minimize over  $\epsilon$ .

```

_____/sage/section1.4.sage_____
481 res = find_local_minimum(
482     varied_free_particle_action(3.0, test_path, nu, 0.0, 10.0), -2.0, 1.0
483 )
484 show(res)

```

---

(435.0000000000000,0.0)

We see that the optimal value for  $\epsilon$  is 0, and we retrieve our earlier value of the Lagrangian action.

#### 1.4.8 Finding minimal trajectories

The `make_path` function uses a Lagrangian polynomial to interpolate a given set of data.

```

_____/sage/utlils1.4.sage_____
485 def Lagrangian_polynomial(ts, qs):
486     return RR['x'].lagrange_polynomial(list(zip(ts, qs)))

```

---

While a Lagrangian polynomial gives an excellent fit on the fitted points, its behavior in between these points can be quite wild. Let us test the quality of the fit before using this interpolation method. From the book we know we need to fit  $\cos(t)$  on  $t \in [0, \pi/2]$ , so let us try this first before trying to find the optimal path for the harmonic Lagrangian. Since  $\cos^2 x + \sin^2 x = 1$ , we can use this relation to check the quality of derivative of the fitted polynomial at the same time. The result is better than I expected.

```

_____/sage/section1.4.sage_____
487 ts = np.linspace(0, pi / 2, 5)
488 qs = [cos(t).n() for t in ts]
489 lp = Lagrangian_polynomial(ts, qs)
490 ts = np.linspace(0, pi / 2, 20)
491 Cos = [lp(x=t).n() for t in ts]
492 Sin = [lp.derivative(x)(x=t).n() for t in ts]
493 Zero = [abs(Cos[i] ^ 2 + Sin[i] ^ 2 - 1) for i in range(len(ts))]
494 show(max(Zero))

```

---

In the function `make_path` we use numpy's `linspace` instead of the linear interpolants of the book. Note that the coordinate paths above are column-vector functions, so `make_path` should return the same type.

---

```

495 def make_path(t0, q0, t1, q1, qs):
496     ts = np.linspace(t0, t1, len(qs) + 2)
497     qs = np.r_[q0, qs, q1]
498     return lambda t: vector([Lagrangian_polynomial(ts, qs)(t)])

```

---

Here is the harmonic Lagrangian.

---

```

499 def L_harmonic(m, k):
500     def Lagrangian(local):
501         q = coordinate(local)
502         v = velocity(local)
503         return (1 / 2) * m * square(v) - (1 / 2) * k * square(q)
504
505     return Lagrangian

```

---



---

```

506 def parametric_path_action(Lagrangian, t0, q0, t1, q1):
507     def f(qs):
508         path = make_path(t0, q0, t1, q1, qs=qs)
509         return Lagrangian_action(Lagrangian, path, t0, t1)
510
511     return f

```

---

Let's try this on the path  $\cos(t)$ . The intermediate values `qs` will be optimized below, whereas `q0` and `q1` remain fixed. Thus, we strip the first and last element of `linspace` to make `qs`. The result tells us what we can expect for the minimal value for the integral over the Lagrangian along the optimal path.

---

```

512 t0, t1 = 0, pi / 2
513 q0, q1 = cos(t0), cos(t1)
514 T = np.linspace(0, pi / 2, 5)
515 initial_qs = [cos(t).n() for t in T][1:-1]
516 parametric_path_action(L_harmonic(m=1, k=1), t0, q0, t1, q1)(initial_qs)

```

---

What is the quality of the path obtained by the Lagrangian interpolation? (Recall that a path is a vector; to extract the value of the element that corresponds to the path, we need to write `best_path(t=t)[0]`.)

---

```

517 def find_path(Lagrangian, t0, q0, t1, q1, n):
518     ts = np.linspace(t0, t1, n)
519     initial_qs = np.linspace(q0, q1, n)[1:-1]

```

---

---

```

520     minimizing_qs = minimize(
521         parametric_path_action(Lagrangian, t0, q0, t1, q1),
522         initial_qs,
523     )
524     return make_path(t0, q0, t1, q1, minimizing_qs)
525
526 best_path = find_path(L_harmonic(m=1, k=1), t0=0, q0=1, t1=pi / 2, q1=0, n=5)
527 result = [
528     abs(best_path(t)[0].n() - cos(t).n()) for t in np.linspace(0, pi / 2, 10)
529 ]
530 show(max(result))

```

---

0.000172462354236957

Great. All works!

Finally, here is a plot of the Lagrangian as a function of  $q(t)$ .

---

```

531 T = np.linspace(0, pi / 2, 20)
532 q = lambda t: vector([cos(t)])
533 lvalues = [L_harmonic(m=1, k=1)(Gamma(q)(t))(t=ti).n() for ti in T]
534 points = list(zip(ts, lvalues))
535 plot = list_plot(points, color="black", size=30)
536 plot.axes_labels(["$t$", "$L$"])
537 plot.save("../figures/Lagrangian.png", figsize=(4, 2))

```

---

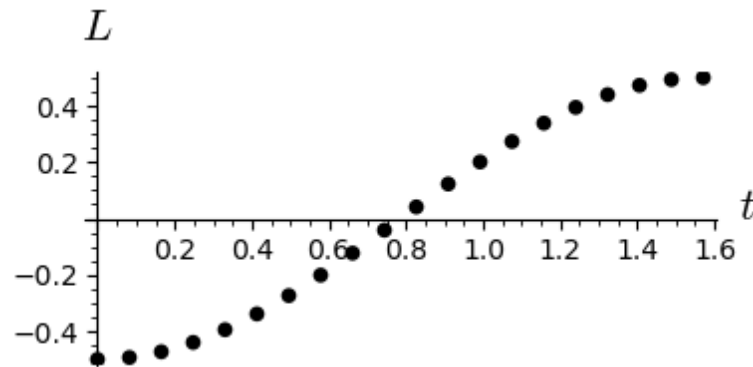


Figure 1.1: The harmonic Lagrangian as a function of the optimal path  $q(t) = \cos t$ ,  $t \in [0, \pi/2]$ .

## 1.5 THE EULER-LAGRANGE EQUATIONS

### 1.5.1 Standard imports

---

```

538 load("utils1.4.sage")

```

---

---

```

539 load("utils1.5.sage")
540
541 t = var("t", domain="real")

```

---



---

```

542 load("show_expression.sage")

```

---

### 1.5.2 Derivation of the Lagrange equations

#### Harmonic oscillator

Here is a test on the harmonic oscillator.

---

```

543 load("utils1.4.sage")
544 k, m = var('k m', domain="positive")
545 q = column_path([literal_function("x")])

```

---



---

```

546 L = L_harmonic(m, k)
547 show(L(Gamma(q)(t)))

```

---

$$-\frac{1}{2}kx^2 + \frac{1}{2}m\dot{x}^2$$

We can apply  $\partial_1 L$  and  $\partial_2 L$  to a configuration path  $q$  that we lift to a local tuple by means of  $\Gamma$ . Realize therefore that  $\text{partial}(L\_harmonic(m, k), 1)$  maps a local tuple to a real number, and  $\text{Gamma}(q)$  maps a time  $t$  to a local tuple. The next code implements  $\partial_1 L(\Gamma(q)(t))$  and  $\partial_2 L(\Gamma(q)(t))$ . (Check how the brackets are organized.)

---

```

548 show(partial(L, 1)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} -kx \end{bmatrix}$$

---

```

549 show(partial(L, 2)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} m\dot{x} \end{bmatrix}$$

Here are the same results, but now with functional composition.

$$(\partial_1 L \circ \Gamma(q))(t), \quad (\partial_2 L \circ \Gamma(q))(t).$$

---

```

550 show(compose(partial(L, 1), Gamma(q))(t))
551 show(compose(partial(L, 2), Gamma(q))(t))

```

---

$$\begin{bmatrix} -kx \end{bmatrix}$$

$$\begin{bmatrix} m\dot{x} \end{bmatrix}$$

These results are functions of  $t$ , so we can take the derivative with respect to  $t$ , which forms the last step to check before building the Euler-Lagrange equations. To understand this, note the following function mappings, where we write  $t$  for time,  $l$  for a local tuple,  $v$  a velocity-like vector, and  $a$  an acceleration-like vector:

$$\begin{aligned} \Gamma[q] &: t \rightarrow l, \\ \partial_2 L &: l \rightarrow v \\ \partial_2 L \circ \Gamma[q] &: t \rightarrow v \\ D(v) &: t \rightarrow a \\ D(\partial_2 L \circ \Gamma[q]) &: t \rightarrow a. \end{aligned}$$

In more classical notation, we compute this:

$$\frac{d}{dt} \left( \frac{\partial}{\partial \dot{q}} L(\Gamma(q)) \right) (t)$$

---

```

552 show(D(compose(partial(L, 2), Gamma(q)))(t))

```

---

$$\begin{bmatrix} m\ddot{x} \end{bmatrix}$$

There we are! We can now try the other examples of the book.

### *Orbital motion*

---

```

553 q = column_path([literal_function("xi"), literal_function("eta")])

```

---



---

```

554 var("mu", domain="positive")
555
556 def L_orbital(m, mu):
557     def Lagrangian(local):
558         q = coordinate(local)
559         v = velocity(local)
560         return (1 / 2) * m * square(v) + mu / sqrt(square(q))
561
562     return Lagrangian

```

---

---

```

563 L = L_orbital(m, mu)
564 show(L(Gamma(q)(t)))

```

---

$$\frac{1}{2} (\dot{\eta}^2 + \dot{\zeta}^2) m + \frac{\mu}{\sqrt{\eta^2 + \zeta^2}}$$

---

```

565 show(partial(L, 1)(Gamma(q)(t)))

```

---

$$\left[ -\frac{\mu \dot{\zeta}}{(\eta^2 + \zeta^2)^{\frac{3}{2}}}, -\frac{\mu \dot{\eta}}{(\eta^2 + \zeta^2)^{\frac{3}{2}}} \right]$$

---

```

566 show(partial(L, 2)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} m \ddot{\zeta} & m \ddot{\eta} \end{bmatrix}$$

*An ideal planar pendulum, Exercise 1.9.a of the book*

We need a new path in terms of  $\theta$  and  $\dot{\theta}$ .

---

```

567 q = column_path([literal_function("theta")])

```

---

Here is the Lagrangian. Recall that the coordinates of the space form a vector. Here, theta is the only element of the vector, which we can extract by considering element 0. For thetadot we don't have to do this since we consider  $\dot{\theta}^2$ , and the square function accepts vectors as input and returns a real. However, for reasons of consistency, we choose to do this nonetheless.

---

```

568 var("m g l", domain="positive")
569
570
571 def L_planar_pendulum(m, g, l):
572     def Lagrangian(local):
573         theta = coordinate(local).list()[0]
574         thetadot = velocity(local).list()[0]
575         T = (1 / 2) * m * l ^ 2 * square(thetadot)
576         V = m * g * l * (1 - cos(theta))
577         return T - V
578
579     return Lagrangian

```

---



---

```

580 L = L_planar_pendulum(m, g, l)
581 show(L(Gamma(q)(t)))

```

---



---

```

582 ..sage/section1.5.sage
show(partial(L, 1)(Gamma(q)(t)))

```

---



---

```

583 ..sage/section1.5.sage
show(partial(L, 2)(Gamma(q)(t)))

```

---

*Henon Heiles potential, Exercise 1.9.b of the book*

As the potential depends on the  $x$  and  $y$  coordinate separately, we need to unpack the coordinate vector.

---

```

584 def L_Henon_Heiles(m):
585     def Lagrangian(local):
586         x, y = coordinate(local).list()
587         v = velocity(local)
588         T = (1 / 2) * square(v)
589         V = 1 / 2 * (square(x) + square(y)) + square(x) * y - y**3 / 3
590         return T - V
591
592     return Lagrangian

```

---



---

```

593 ..sage/section1.5.sage
L = L_Henon_Heiles(m)
594 q = column_path([literal_function("x"), literal_function("y")])
595 show(L(Gamma(q)(t)))

```

---

$$-x^2y + \frac{1}{3}y^3 - \frac{1}{2}x^2 - \frac{1}{2}y^2 + \frac{1}{2}\dot{x}^2 + \frac{1}{2}\dot{y}^2$$

---

```

596 ..sage/section1.5.sage
show(partial(L, 1)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} -2xy - x & -x^2 + y^2 - y \end{bmatrix}$$

---

```

597 ..sage/section1.5.sage
show(partial(L, 2)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} \dot{x} & \dot{y} \end{bmatrix}$$

*Motion on the 2d sphere, Exercise 1.9.c of the book*

---

```

598 var('R', domain="positive")
599
600
601 def L_sphere(m, R):
602     def Lagrangian(local):
603         theta, phi = coordinate(local).list()
604         alpha, beta = velocity(local).list()
605         L = m * R * (square(alpha) + square(beta * sin(theta))) / 2
606         return L
607
608     return Lagrangian

```

---

```

609 q = column_path([literal_function("phi"), literal_function("theta")])
610 L = L_sphere(m, R)
611
612 show(L(Gamma(q)(t)))

```

---

$$\frac{1}{2} (\sin(\phi)^2 \dot{\theta}^2 + \dot{\phi}^2) Rm$$

---

```

613 show(partial(L, 1)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} Rm \cos(\phi) \sin(\phi) \dot{\theta}^2 & 0 \end{bmatrix}$$

---

```

614 show(partial(L, 2)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} Rm \dot{\phi} & Rm \sin(\phi)^2 \dot{\theta} \end{bmatrix}$$

### *Higher order Lagrangians*

I recently read the books of Larry Susskind on the theoretical minimum for physics. He claims that Lagrangians up to first order derivatives suffice to understand nature, so I skip this part.

### 1.5.3 *Computing Lagrange's equation*

The Euler-Lagrange equations are simple to implement now that we have a good function for computing partial derivatives.

### *The Euler Lagrange Equations*

We work in steps to see how all components tie together.

---

```

615 q = column_path(
616     [
617         literal_function("x"),
618         literal_function("y"),
619     ]
620 )
621
622 L = L_free_particle(m)
623 show(compose(partial(L, 1), Gamma(q))(t))
624 show(compose(partial(L, 2), Gamma(q))(t))
625 show(D(compose(partial(L, 2), Gamma(q)))(t))
626 show(
627     (D(compose(partial(L, 2), Gamma(q))) - compose(partial(L, 1), Gamma(q)))(t)
628 )

```

---

$$\begin{bmatrix} 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} m\dot{x} & m\dot{y} \end{bmatrix}$$

$$\begin{bmatrix} m\ddot{x} & m\ddot{y} \end{bmatrix}$$

$$\begin{bmatrix} m\ddot{x} & m\ddot{y} \end{bmatrix}$$

The last step forms the Euler-Lagrange equation, which we can now implement as a function.

---

```

629 def Lagrange_equations(L):
630     def f(q):
631         return D(compose(partial(L, 2), Gamma(q))) - compose(
632             partial(L, 1), Gamma(q)
633         )
634
635     return f

```

---

### *The free particle*

We compute the Lagrange equation for a path linear in  $t$  for the Lagrangian of a free particle..

---

```

636 var("a b c a0 b0 c0", domain="real")
637 test_path = lambda t: column_matrix([a * t + a0, b * t + b0, c * t + c0])

```

---

Note that if we do not provide the argument `t` to `l_eq` we receive a function instead of vector.

---

```

638 l_eq = Lagrange_equations(L_free_particle(m))(test_path)
639 show(l_eq(t))

```

---

$$\begin{bmatrix} 0 & 0 & 0 \end{bmatrix}$$

This is correct since a free particle is not moving in a potential field, hence only depends on the velocity but not the coordinates of the path. But since the velocity is linear in  $t$ , all components along the test path become zero.

Here are the EL equations for a generic 1D path.

---

```

640 q = column_path([literal_function("x")])
641 l_eq = Lagrange_equations(L_free_particle(m))(q)
642 show(l_eq(t))

```

---

$$\begin{bmatrix} m\ddot{x} \end{bmatrix}$$

Equating this to (0) shows that the solution of these differential equations is linear in  $t$ .

### *The harmonic oscillator*

---

```

643 var("A phi omega", domain="real")
644 assume(A > 0)
645
646 proposed_path = lambda t: vector([A * cos(omega * t + phi)])

```

---

`Lagrange_equations` returns a matrix whose elements correspond to the components of the configuration path  $q$ .

---

```

647 l_eq = Lagrange_equations(L_harmonic(m, k))(proposed_path)(t)
648 show(l_eq)

```

---

$$\begin{bmatrix} -Am\omega^2 \cos(\omega t + \phi) + Ak \cos(\omega t + \phi) \end{bmatrix}$$

To obtain the contents of this  $1 \times 1$  matrix, we take the element `[0][0]`.

---

```

649 show(l_eq[0][0])

```

---

$$-Am\omega^2 \cos(\omega t + \phi) + Ak \cos(\omega t + \phi)$$

Let's factor out the cosine.

---

```

650 show(l_eq[0, 0].factor())

```

---

$$-(m\omega^2 - k)A \cos(\omega t + \phi)$$

*Kepler's third law*

Recall that to unpack the coordinates, we have to convert the vector to a Python list.

---

```

651 var("G m m1 m2", domain="positive")
652
653
654 def L_central_polar(m, V):
655     def Lagrangian(local):
656         r, phi = coordinate(local).list()
657         rdot, phidot = velocity(local).list()
658         T = 1 / 2 * m * (square(rdot) + square(r * phidot))
659         return T - V(r)
660
661     return Lagrangian
662
663
664 def gravitational_energy(G, m1, m2):
665     def f(r):
666         return -G * m1 * m2 / r
667
668     return f

```

---

```

669 q = column_path([literal_function("r"), literal_function("phi")])
670 V = gravitational_energy(G, m1, m2)
671 L = L_central_polar(m, V)
672 show(L(Gamma(q)(t)))

```

---

$$\frac{1}{2} (r^2 \dot{\phi}^2 + \dot{r}^2) m + \frac{G m_1 m_2}{r}$$

---

```

673 l_eq = Lagrange_equations(L)(q)(t)

```

---

```

674 show(l_eq[0, 1] == 0)

```

---

$$mr^2\ddot{\phi} + 2mr\dot{\phi}\dot{r} = 0$$

In this equation, let's divide by  $mr$  to get  $r\ddot{\phi} + 2\dot{\phi}\dot{r} = 0$ , which is equal to  $\partial_t(\dot{\phi}r^2) = 0$ . This implies that  $\dot{\phi}r^2 = C$ , i.e., a constant. If  $r \neq 0$  and constant, which we should assume according to the book, then we see that  $\dot{\phi}$  is constant, so the two bodies rotate with constant angular speed around each other.

What can we say about the other equation?

---

```

675 show(l_eq[0, 0] == 0)

```

---

$$-mr\dot{\phi}^2 + m\ddot{r} + \frac{Gm_1m_2}{r^2} = 0$$

As  $r$  is constant according to the book,  $\ddot{r} = 0$ . By dividing by  $m := m_1m_2/(m_1 + m_2)$ , this equation reduces to  $r^3\dot{\phi}^2 = G(m_1 + m_2)$ , which is the form we were to find according to the exercise.

## 1.6 HOW TO FIND LAGRANGIANS

### 1.6.1 *Standard imports*

```

676  _____ ../sage/utis1.6.sage _____
      load("utis1.5.sage")
      _____

677  _____ ../sage/section1.6.sage _____
      load("utis1.6.sage")
      _____

678  _____ don't tangle _____
      load("show_expression.sage")
      _____

```

### 1.6.2 *Constant acceleration*

We start with a point in a uniform gravitational field.

```

679  _____ ../sage/utis1.6.sage _____
      var("t", domain="real")
680  var("g m", domain="positive")
681
682
683  def L_uniform_acceleration(m, g):
684      def Lagrangian(local):
685          x, y = coordinate(local).list()
686          v = velocity(local)
687          T = 1 / 2 * m * square(v)
688          V = m * g * y
689          return T - V
690
691      return Lagrangian
      _____

692  _____ ../sage/section1.6.sage _____
      q = column_path([literal_function("x"), literal_function("y")])
693  l_eq = Lagrange_equations(L_uniform_acceleration(m, g))(q)
694  show(l_eq(t))
      _____

```

$$\begin{bmatrix} m\ddot{x} & gm + m\ddot{y} \end{bmatrix}$$

### 1.6.3 Central force field

---

```

695 def L_central_rectangular(m, U):
696     def Lagrangian(local):
697         q = coordinate(local)
698         v = velocity(local)
699         T = 1 / 2 * m * square(v)
700         return T - U(sqrt(square(q)))
701
702     return Lagrangian

```

---

Let us first try this on a concrete potential function.

---

```

703 def U(r):
704     return 1 / r

```

---



---

```

705 show(Lagrange_equations(L_central_rectangular(m, U))(q)(t))

```

---

$$\left[ m\ddot{x} - \frac{x}{(x^2+y^2)^{\frac{3}{2}}} \quad m\ddot{y} - \frac{y}{(x^2+y^2)^{\frac{3}{2}}} \right]$$

Now we try it on a general central potential.

---

```

706 U = Function(lambda x: function("U")(x))
707 show(Lagrange_equations(L_central_rectangular(m, U))(q)(t))

```

---

$$\left[ m\ddot{x} + \frac{x D_0(U)(\sqrt{x^2+y^2})}{\sqrt{x^2+y^2}} \quad m\ddot{y} + \frac{y D_0(U)(\sqrt{x^2+y^2})}{\sqrt{x^2+y^2}} \right]$$

### 1.6.4 Coordinate transformations

To get things straight: the function  $F$  is the transformation of the coordinates  $x'$  to  $x$ , i.e.,  $x = F(t, x')$ . The function  $C$  lifts the transformation  $F$  to the phase space, so it transforms  $\Gamma(q')$  to  $\Gamma(q)$ .

The result of  $\partial_1 F v$  is a vector, because  $v$  is a vector. We have to cast  $\partial_0 F$  into a vector to enable the summation of these two terms.

---

```

708 def F_to_C(F):
709     def f(local):
710         return up(
711             time(local),
712             F(local),
713             partial(F, 0)(local) + partial(F, 1)(local) * velocity(local),
714         )
715
716     return f

```

---

## 1.6.5 polar coordinates

---

```

717 def p_to_r(local):
718     r, phi = coordinate(local).list()
719     return column_matrix([r * cos(phi), r * sin(phi)])

```

---

We apply `F_to_C` and `p_to_r` to several examples, to test and to understand how they collaborate. We need to make the appropriate variables for the space in terms of  $r$  and  $\phi$ .

---

```

720 r = literal_function("r")
721 phi = literal_function("phi")
722 q = column_path([r, phi])
723 show(p_to_r(Gamma(q)(t)))

```

---

$$\begin{bmatrix} \cos(\phi) r \\ r \sin(\phi) \end{bmatrix}$$

This is the derivative wrt  $t$ . As the transformation `p_to_r` does not depend explicitly on  $t$ , the result should be a column matrix of zeros.

---

```

724 show((partial(p_to_r, 0)(Gamma(q)(t))))

```

---

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Next is the derivative wrt  $r$  and  $\phi$ .

---

```

725 show((partial(p_to_r, 1)(Gamma(q)(t))))

```

---

$$\begin{bmatrix} \cos(\phi) & -r \sin(\phi) \\ \sin(\phi) & \cos(\phi) r \end{bmatrix}$$

---

```

726 show(F_to_C(p_to_r)(Gamma(q)(t)))

```

---

$$\begin{bmatrix} t \\ \cos(\phi) r \\ r \sin(\phi) \\ -r \sin(\phi) \dot{\phi} + \cos(\phi) \dot{r} \\ \cos(\phi) r \dot{\phi} + \sin(\phi) \dot{r} \end{bmatrix}$$

We can see what happens for the Lagrangian for the central force in polar coordinates.



---

```

727 def L_central_polar(m, U):
728     def Lagrangian(local):
729         return compose(L_central_rectangular(m, U), F_to_C(p_to_r))(local)
730
731     return Lagrangian

```

---



---

```

732 # show(L_central_polar(m, U)(Gamma(q)(t)))
733 show(L_central_polar(m, U)(Gamma(q)(t)).simplify_full())

```

---

$$\frac{1}{2}mr^2\dot{\phi}^2 + \frac{1}{2}m\dot{r}^2 - U(\sqrt{r^2})$$

---

```

734 expr = Lagrange_equations(L_central_polar(m, U))(q)(t)
735 show(expr.simplify_full().expand())

```

---

$$\left[ -m r \dot{\phi}^2 + m \ddot{r} + \frac{r D_0(U)(\sqrt{r^2})}{\sqrt{r^2}} \quad m r^2 \ddot{\phi} + 2 m r \dot{\phi} \dot{r} \right]$$

### 1.6.6 Coriolis and centrifugal forces

---

```

736 def L_free_rectangular(m):
737     def Lagrangian(local):
738         v = velocity(local)
739         return 1 / 2 * m * square(v)
740
741     return Lagrangian
742
743
744 def L_free_polar(m):
745     def Lagrangian(local):
746         return L_free_rectangular(m)(F_to_C(p_to_r)(local))
747
748     return Lagrangian
749
750
751 def F(Omega):
752     def f(local):
753         t = time(local)
754         r, theta = coordinate(local).list()
755         return vector([r, theta + Omega * t])
756
757     return f
758
759
760 def L_rotating_polar(m, Omega):

```

```

761     def Lagrangian(local):
762         return L_free_polar(m)(F_to_C(F(0mega))(local))
763
764     return Lagrangian
765
766
767
768     def r_to_p(local):
769         x, y = coordinate(local).list()
770         return column_matrix([sqrt(x * x + y * y), atan(y / x)])
771
772
773     def L_rotating_rectangular(m, 0mega):
774         def Lagrangian(local):
775             return L_rotating_polar(m, 0mega)(F_to_C(r_to_p)(local))
776
777         return Lagrangian

```

---

```

778     _ = var("0mega", domain="positive")
779     q_xy = column_path([literal_function("x"), literal_function("y")])
780     expr = L_rotating_rectangular(m, 0mega)(Gamma(q_xy)(t)).simplify_full()

```

---

```

781     show(expr)

```

---

$$\frac{1}{2}\Omega^2 mx^2 + \frac{1}{2}\Omega^2 my^2 - \Omega my\dot{x} + \Omega mx\dot{y} + \frac{1}{2}m\dot{x}^2 + \frac{1}{2}m\dot{y}^2$$

The simplification of the Lagrange equations takes some time.

```

782     expr = Lagrange_equations(L_rotating_rectangular(m, 0mega))(q)(t)
783     show(expr.simplify_full())

```

---

I edited the result a bit by hand.

$$-m\Omega^2 x - 2m\Omega\dot{y} + m\ddot{x}, -m\Omega^2 y + 2m\Omega\dot{x} + m\ddot{y}.$$

### 1.6.7 Constraints, a driven pendulum

Rather than implementation the formulas of the book at this place, we follow the idea they explain a bit later in the book: formulate a Lagrangian in practical coordinates, then formulate the problem in practical coordinates *for that problem*, and then use a coordinate transformation from the problem's coordinates to the Lagrangian coordinates.

For the driven pendulum, the Lagrangian is easiest to express in terms of  $x$  and  $y$  coordinates, while the pendulum needs an angle  $\theta$ . So, we need a transformation from

$\theta$  to  $x$  and  $y$ . Note that the function `coordinate` returns a  $(1 \times 1)$  column matrix which just contains  $\theta$ . So, we have to pick element  $(0,0)$ . Another point is that here `ys` needs to be evaluated at `t`; in the other functions `ys` is just passed on as a function.

---

```

784 def dp_coordinates(l, ys):
785     "From theta to x, y coordinates."
786     def f(local):
787         t = time(local)
788         theta = coordinate(local)[0, 0]
789         return column_matrix([l * sin(theta), ys(t) - l * cos(theta)])
790
791     return f

```

---



---

```

792 def L_pend(m, l, g, ys):
793     def Lagrangian(local):
794         return L_uniform_acceleration(m, g)(
795             F_to_C(dp_coordinates(l, ys))(local)
796         )
797
798     return Lagrangian

```

---



---

```

799 _ = var("l", domain="positive")
800
801 theta = column_path([literal_function("theta")])
802 ys = literal_function("y")
803
804 expr = L_pend(m, l, g, ys)(Gamma(theta)(t)).simplify_full()
805 show(expr)

```

---

$$\frac{1}{2} l^2 m \dot{\theta}^2 + l m \sin(\theta) \dot{\theta} \dot{y} + g l m \cos(\theta) - g m y + \frac{1}{2} m \dot{y}^2$$

## 1.7 EVOLUTION OF DYNAMICAL STATE

### 1.7.1 Standard imports

---

```

806 load("utils1.6.sage")

```

---

```

807 load("utils1.7.sage")
808
809 var("t", domain=RR)

```

---

```

810 load("show_expression.sage")

```

---

don't tangle

### 1.7.2 Acceleration and state derivative

We build the functions `Lagrangian_to_acceleration` and `Lagrangian_to_state_derivative` in steps.

---

```

811 q = column_path([literal_function("x"), literal_function("y")])
812 local = Gamma(q)(t)
813 m, k = var("m k", domain="positive")
814 L = L_harmonic(m, k)
815 show(L(local))

```

---

$$-\frac{1}{2}(x^2 + y^2)k + \frac{1}{2}(\dot{x}^2 + \dot{y}^2)m$$

---

```

816 F = compose(transpose, partial(L, 1))
817 show(F(local))
818 P = partial(L, 2)
819 show((F - partial(P, 0))(local))

```

---

$$\begin{pmatrix} -kx \\ -ky \end{pmatrix}$$

$$\begin{pmatrix} -kx \\ -ky \end{pmatrix}$$

---

```

820 show((partial(P, 1) * velocity)(local))

```

---

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Convert to vector.

---

```

821 show((F - partial(P, 0) - partial(P, 1) * velocity)(local))

```

---

$$\begin{pmatrix} -kx \\ -ky \end{pmatrix}$$

---

```

822 def Lagrangian_to_acceleration(L):
823     def f(local):
824         P = partial(L, 2)
825         F = compose(transpose, partial(L, 1))
826         M = (F - partial(P, 0)) - partial(P, 1) * velocity
827         return partial(P, 2)(local).solve_right(M(local))
828
829     return f

```

---

We apply this to the harmonic oscillator.

---

```

830  ..../sage/section1.7.sage
show(Lagrangian_to_acceleration(L)(local))

```

---

$$\begin{pmatrix} -\frac{kx}{m} \\ -\frac{ky}{m} \end{pmatrix}$$

### 1.7.3 *Intermezzo, numerically integrating ODEs with Sagemath*

At a later stage, we want to numerically integrate the system of ODEs that result from the Lagrangian. This works a bit different from what I expected; here are two examples to see the problem.

Consider the system of DEs for the circle:  $\dot{x} = y$ ,  $\dot{y} = -x$ . This code implements the rhs:

---

```

831  def de_rhs(x, y):
832      return [y, -x]
833
834
835  sol = desolve_odeint(de_rhs(x, y), [1, 0], srange(0, 100, 0.05), [x, y])
836  pp = list(zip(sol[:, 0], sol[:, 1]))
837  p = points(pp, color='blue', size=3)
838  p.save('circle.png')

```

---

However, if I replace the RHS of the DE by by constants,, I get an error that the integration variables are unknown.

---

```

839  def de_rhs(x, y):
840      return [1, -1]

```

---

The solution is to replace the numbers by expressions.

---

```

841  def convert_to_expr(n):
842      return SR(n)

```

---

And then define the function of differentials like this.

---

```

843  def de_rhs(x, y):
844      return [convert_to_expr(1), convert_to_expr(-1)]

```

---

Now things work as they should.

## 1.7.4 Continuing with the oscillator

The next function computes the state derivative of the Lagrangian. For the purpose of numerical integration, we cast the result of the derivative of  $dt/dt = 1$  to an expression, more specifically, by the above intermezzo we should set the derivative of  $t$  to `convert_to_expr(1)`.

---

```

845 def Lagrangian_to_state_derivative(L):
846     acceleration = Lagrangian_to_acceleration(L)
847     return lambda state: up(
848         convert_to_expr(1), velocity(state), acceleration(state)
849     )

```

---



---

```

850 show(Lagrangian_to_state_derivative(L)(local))

```

---

$$\begin{pmatrix} 1 \\ \dot{x} \\ \dot{y} \\ -\frac{kx}{m} \\ -\frac{ky}{m} \end{pmatrix}$$

---

```

851 def harmonic_state_derivative(m, k):
852     return Lagrangian_to_state_derivative(L_harmonic(m, k))

```

---



---

```

853 show(harmonic_state_derivative(m, k)(local))

```

---

$$\begin{pmatrix} 1 \\ \dot{x} \\ \dot{y} \\ -\frac{kx}{m} \\ -\frac{ky}{m} \end{pmatrix}$$

---

```

854 def qv_to_state_path(q, v):
855     return lambda t: up(t, q(t), v(t))

```

---



---

```

856 def Lagrange_equations_first_order(L):
857     def f(q, v):
858         state_path = qv_to_state_path(q, v)
859         res = D(state_path)
860         res -= compose(Lagrangian_to_state_derivative(L), state_path)
861         return res
862
863     return f

```

---

---

```

864 res = Lagrange_equations_first_order(L_harmonic(m, k))(
865     column_path([literal_function("x"), literal_function("y")]),
866     column_path([literal_function("v_x"), literal_function("v_y")]),
867 )
868 show(res(t))

```

---

$$\begin{pmatrix} 0 \\ -v_x + \dot{x} \\ -v_y + \dot{y} \\ \frac{kx}{m} + \dot{v}_x \\ \frac{ky}{m} + \dot{v}_y \end{pmatrix}$$

### 1.7.5 Numerical integration

For the numerical integrator we have to specify the variables that appear in the differential equations. For this purpose we use dummy vectors.

---

```

869 def make_dummy_vector(name, dim):
870     return column_matrix([var(f"{name}{i}", domain=RR) for i in range(dim)])

```

---

The state\_advancer needs an evolve function. We use the initial conditions ics to figure out the dimension of the coordinate space. Once we have the dimension, we construct a dummy up tuple with coordinate and velocity variables. The ode solver need plain lists; since space is an up tuple, the list method of Tuple can provide for this.

---

```

871 def evolve(state_derivative, ics, times):
872     dim = coordinate(ics).nrows()
873     coordinates = make_dummy_vector("q", dim)
874     velocities = make_dummy_vector("v", dim)
875     space = up(t, coordinates, velocities)
876     soln = desolve_odeint(
877         des=state_derivative(space).list(),
878         ics=ics.list(),
879         times=times,
880         dvars=space.list(),
881         atol=1e-13,
882     )
883     return soln

```

---

The state advancer integrates the orbit for a time T and starting at the initial conditions.

---

```

884 def state_advancer(state_derivative, ics, T):
885     init_time = time(ics)
886     times = [init_time, init_time + T]
887     soln = evolve(state_derivative, ics, times)
888     return soln[-1]

```

---

As a test, let's apply it to the one D harmonic oscillator.

---

```

889 state_advancer(
890     harmonic_state_derivative(m=2, k=1),
891     ics=up(0, column_matrix([1, 2]), column_matrix([3, 4])),
892     T=10,
893 )

```

---

array([10. , 3.71279102, 5.42061989, 1.61480284, 1.8189101 ])

These are (nearly) the same results as in the book.

### 1.7.6 The driven pendulum

Here is the driver for the pendulum.

---

```

894 def periodic_drive(amplitude, frequency, phase):
895     def f(t):
896         return amplitude * cos(frequency * t + phase)
897
898     return f

```

---

With this we make the Lagrangian.

---

```

899 _ = var("m l g A omega")
900
901
902 def L_periodically_driven_pendulum(m, l, g, A, omega):
903     ys = periodic_drive(A, omega, 0)
904
905     def Lagrangian(local):
906         return L_pend(m, l, g, ys)(local)
907
908     return Lagrangian

```

---



---

```

909 q = column_path([literal_function("theta")])
910 show(
911     L_periodically_driven_pendulum(m, l, g, A, omega)(
912         Gamma(q)(t)
913     ).simplify_full()
914 )

```

---



$$\frac{1}{2} A^2 m \omega^2 \sin(\omega t)^2 - A l m \omega \sin(\omega t) \sin(\theta) \dot{\theta} + \frac{1}{2} l^2 m \dot{\theta}^2 - A g m \cos(\omega t) + g l m \cos(\theta)$$

---

```

915 expr = Lagrange_equations(L_periodically_driven_pendulum(m, l, g, A, omega))(
916     q
917 )(t).simplify_full()
918 show(expr)

```

---

$$\left( l^2 m \ddot{\theta} - (A l m \omega^2 \cos(\omega t) - g l m) \sin(\theta) \right)$$

---

```

919 show(
920     Lagrangian_to_acceleration(
921         L_periodically_driven_pendulum(m, l, g, A, omega)
922     )(Gamma(q)(t)).simplify_full()
923 )

```

---

$$\left( \frac{(A \omega^2 \cos(\omega t) - g) \sin(\theta)}{l} \right)$$

---

```

924 def pend_state_derivative(m, l, g, A, omega):
925     return Lagrangian_to_state_derivative(
926         L_periodically_driven_pendulum(m, l, g, A, omega)
927     )

```

---



---

```

928 expr = pend_state_derivative(m, l, g, A, omega)(Gamma(q)(t))
929 show(time(expr))
930 show(coordinate(expr).simplify_full())
931 show(velocity(expr).simplify_full())

```

---

$$1$$

$$\left( \dot{\theta} \right)$$

$$\left( \frac{(A \omega^2 \cos(\omega t) - g) \sin(\theta)}{l} \right)$$

---

```

932 def principal_value(cut_point):
933     def f(x):
934         return (x + cut_point) % (2 * np.pi) - cut_point
935
936     return f

```

---

---

```

937 def plot_driven_pendulum(A, T, step_size=0.01):
938     times = srange(0, T, step_size, include_endpoint=True)
939     soln = evolve(
940         pend_state_derivative(m=1, l=1, g=9.8, A=A, omega=2 * sqrt(9.8)),
941         ics=up(0, column_matrix([1]), column_matrix([0])),
942         times=times,
943     )
944     thetas = soln[:, 1]
945     pp = list(zip(times, thetas))
946     p = points(pp, color='blue', size=3)
947     p.save(f'../figures/driven_pendulum_{A:.2f}.png')
948
949     thetas = principal_value(np.pi)(thetas)
950     pp = list(zip(times, thetas))
951     p = points(pp, color='blue', size=3)
952     p.save(f'../figures/driven_pendulum_{A:.2f}_principal_value.png')
953
954     thetadots = soln[:, 2]
955     pp = list(zip(thetas, thetadots))
956     p = points(pp, color='blue', size=3)
957     p.save(f'../figures/driven_pendulum_{A:.2f}_trajectory.png')
958

```

---

So now we make the plot.

---

```

959 plot_driven_pendulum(A=0.1, T=100, step_size=0.005)

```

---

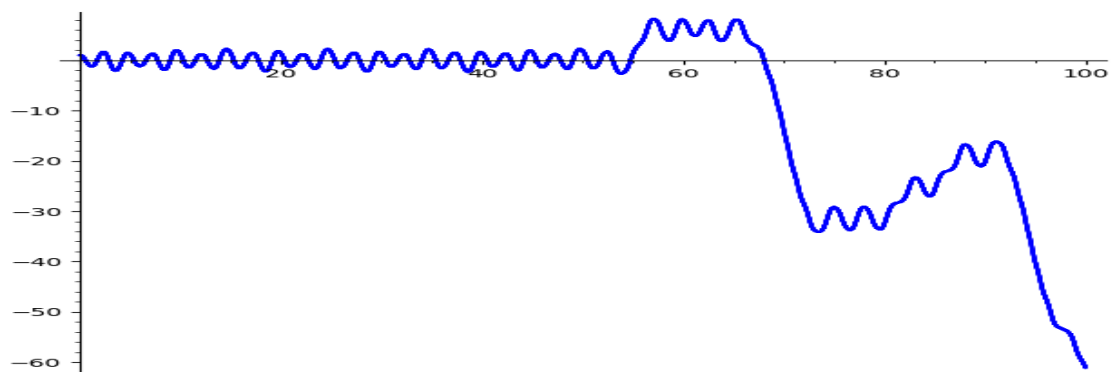
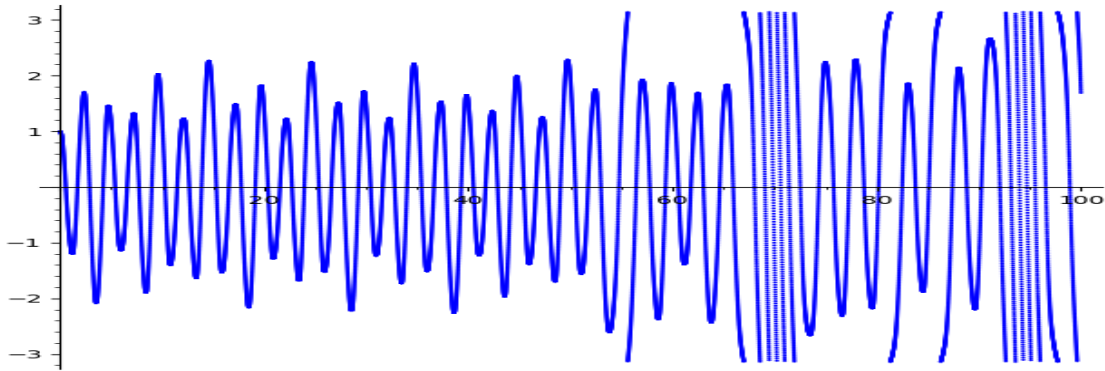
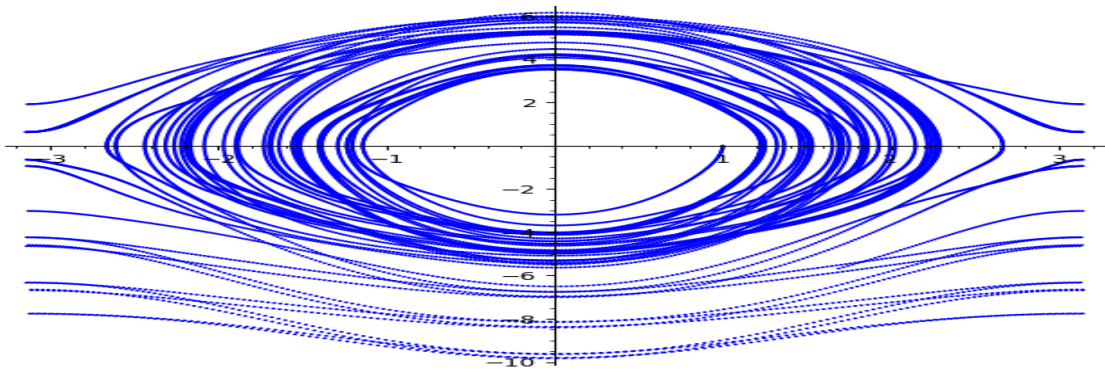


Figure 1.2: The angle of the vertically driven pendulum as a function of time. Obviously, around  $t = 80$ , the pendulum makes a few revolutions, and then starts to wobble again.

Figure 1.3: The angle on  $(-\pi, \pi]$ .Figure 1.4: The trajectory of  $\theta$  and  $\dot{\theta}$ .

## 1.8 CONSERVED QUANTITIES

### 1.8.1 *Standard imports*

```

_____ ../sage/utils1.8.sage _____
960 load("utils1.6.sage")
_____

_____ ../sage/section1.8.sage _____
961 load("utils1.8.sage")
962
963 var("t", domain=RR)
_____

_____ don't tangle _____
964 load("show_expression.sage")
_____

```

### 1.8.2 *Energy Conservation*

From the Lagrangian we can construct the energy function. Note that we should cast  $P = \partial_2 L$  to a vector so that  $P * v$  becomes a number instead of a  $1 \times 1$  matrix. As we use the Lagrangian in functional arithmetic, we convert  $L$  into a Function.

---

```

965 def Lagrangian_to_energy(L):
966     P = partial(L, 2)
967     LL = Function(lambda local: L(local))
968     return lambda local: (P * velocity - LL)(local)

```

---

### 1.8.3 Central Forces in Three Dimensions

Instead of building the kinetic energy in spherical coordinates, as in Section 1.8.3 of the book, I am going to use the ideas that have been expounded book in earlier sections: define the Lagrangian in convenient coordinates, and then use a coordinate transform to obtain it in coordinates that show the symmetries of the system.

---

```

969 q = column_path(
970     [
971         literal_function("r"),
972         literal_function("theta"),
973         literal_function("phi"),
974     ]
975 )

```

---

Next the transformation from spherical to 3D rectangular coordinates.

---

```

976 def s_to_r(spherical_state):
977     r, theta, phi = coordinate(spherical_state).list()
978     return vector(
979         [r * sin(theta) * cos(phi), r * sin(theta) * sin(phi), r * cos(theta)]
980     )

```

---

For example, here is are the velocities expressed in spherical coordinates.

---

```

981 show(velocity(F_to_C(s_to_r)(Gamma(q)(t))).simplify_full())

```

---

$$\begin{bmatrix} \cos(\phi)\cos(\theta)r\dot{\theta} - (r\sin(\phi)\dot{\phi} - \cos(\phi)\dot{r})\sin(\theta) \\ \cos(\theta)r\sin(\phi)\dot{\theta} + (\cos(\phi)r\dot{\phi} + \sin(\phi)\dot{r})\sin(\theta) \\ -r\sin(\theta)\dot{\theta} + \cos(\theta)\dot{r} \end{bmatrix}$$

Now we are ready to check the code examples of the book.

---

```

982 V = Function(lambda r: function("V")(r))
983
984 def L_3D_central(m, V):
985     def Lagrangian(local):
986         return L_central_rectangular(m, V)(F_to_C(s_to_r)(local))
987
988     return Lagrangian

```

---

---

```

989 ..sage/section1.8.sage
show(partial(L_3D_central(m, V), 1)(Gamma(q)(t)).simplify_full())

```

---

$$\left[ -\frac{rD_0(V)(\sqrt{r^2}) - (mr\sin(\theta)^2\dot{\phi}^2 + mr\dot{\theta}^2)\sqrt{r^2}}{\sqrt{r^2}} \quad m\cos(\theta)r^2\sin(\theta)\dot{\phi}^2 \quad 0 \right]$$

---

```

990 ..sage/section1.8.sage
show(partial(L_3D_central(m, V), 2)(Gamma(q)(t)).simplify_full())

```

---

$$\left[ m\dot{r} \quad mr^2\dot{\theta} \quad mr^2\sin(\theta)^2\dot{\phi} \right]$$

---

```

991 ..sage/section1.8.sage
def ang_mom_z(m):
992     def f(rectangular_state):
993         xyx = vector(coordinate(rectangular_state))
994         v = vector(velocity(rectangular_state))
995         return xyx.cross_product(m * v)[2]
996
997     return f
998
999
1000 show(compose(ang_mom_z(m), F_to_C(s_to_r))(Gamma(q)(t)).simplify_full())

```

---

$$mr^2\sin(\theta)^2\dot{\phi}$$

This is the check that  $E = T + V$ .

---

```

1001 ..sage/section1.8.sage
show(Lagrangian_to_energy(L_3D_central(m, V))(Gamma(q)(t)).simplify_full())

```

---

$$\left[ \frac{1}{2}mr^2\sin(\theta)^2\dot{\phi}^2 + \frac{1}{2}mr^2\dot{\theta}^2 + \frac{1}{2}m\dot{r}^2 + V(\sqrt{r^2}) \right]$$

#### 1.8.4 The Restricted Three-Body Problem

I decompose the potential energy function into smaller functions; I find the implementation in the book somewhat heavy.

---

```

1002 ..sage/section1.8.sage
var("G M0 M1 a", domain="positive")
1003
1004
1005 def distance(x, y):
1006     return sqrt(square(x - y))
1007
1008
1009 def angular_freq(M0, M1, a):

```

```

1010     return sqrt(G * (M0 + M1) / a ^ 3)
1011
1012
1013 def V(a, M0, M1, m):
1014     Omega = angular_freq(M0, M1, a)
1015     a0, a1 = M1 / (M0 + M1) * a, M0 / (M0 + M1) * a
1016
1017     def f(t, origin):
1018         pos0 = -a0 * column_matrix([cos(Omega * t), sin(Omega * t)])
1019         pos1 = a1 * column_matrix([cos(Omega * t), sin(Omega * t)])
1020         r0 = distance(origin, pos0)
1021         r1 = distance(origin, pos1)
1022         return -G * m * (M0 / r0 + M1 / r1)
1023
1024     return f
1025
1026 def L0(m, V):
1027     def f(local):
1028         t, q, v = time(local), coordinate(local), velocity(local)
1029         return 1 / 2 * m * square(v) - V(t, q)
1030
1031     return f

```

For the computer it's easy to compute the energy, but the formula is pretty long.

```

----- ../sage/section1.8.sage -----
1032 q = column_path([literal_function("x"), literal_function("y")])
1033 expr = (sqrt(G*M0 + G*M1)*t) / a^(3/2)
1034 A = var('A')
1035
1036 show(
1037     Lagrangian_to_energy(L0(m, V(a, M0, M1, m)))(Gamma(q)(t))
1038     .simplify_full()
1039     .expand()
1040     .subs({expr: A})
1041 )

```

$$\left[ -\frac{\sqrt{M_0^2 + 2M_0M_1 + M_1^2}GM_0m}{\sqrt{2M_0M_1a\cos(A)x + 2M_1^2a\cos(A)x + 2M_0M_1a\sin(A)y + 2M_1^2a\sin(A)y + M_1^2a^2 + M_0^2x^2 + 2M_0M_1x^2 + M_1^2x^2 + M_0^2y^2 + 2M_0M_1y^2 + M_1^2y^2}}$$

I skip the rest of the code of this part as it is just copy work from the mathematical formulas.

### 1.8.5 Noether's theorem

We need to rotate around a given axis in 3D space. ChatGPT gave me the code right away.

```

----- ../sage/utils1.8.sage -----
1042 def rotation_matrix(axis, theta):

```

```

1043 """
1044 Return the 3x3 rotation matrix for a rotation of angle theta (in radians)
1045 about the given axis. The axis is specified as an iterable of 3 numbers.
1046 """
1047 # Convert the axis to a normalized vector
1048 axis = vector(axis).normalized()
1049 x, y, z = axis
1050 c = cos(theta)
1051 s = sin(theta)
1052 t = 1 - c # common factor
1053
1054 # Construct the rotation matrix using Rodrigues' formula
1055 R = matrix(
1056     [
1057         [c + x**2 * t, x * y * t - z * s, x * z * t + y * s],
1058         [y * x * t + z * s, c + y**2 * t, y * z * t - x * s],
1059         [z * x * t - y * s, z * y * t + x * s, c + z**2 * t],
1060     ]
1061 )
1062 return R

```

---

```

..... ../sage/section1.8.sage .....
1063 def F_tilde(angle_x, angle_y, angle_z):
1064     def f(local):
1065         return (
1066             rotation_matrix([1, 0, 0], angle_x)
1067             * rotation_matrix([0, 1, 0], angle_y)
1068             * rotation_matrix([0, 0, 1], angle_z)
1069             * coordinate(local)
1070         )
1071
1072     return f

```

---

```

..... ../sage/section1.8.sage .....
1073 q = column_path(
1074     [literal_function("x"), literal_function("y"), literal_function("z")]
1075 )

```

---

Let's see what we get when we exercise a rotation of  $s$  radians round the  $x$  axis.

```

..... ../sage/section1.8.sage .....
1076 def Rx(s):
1077     return lambda local: F_tilde(s, 0, 0)(local)
1078
1079
1080 s, u, v = var("s u v")
1081 latex.matrix_delimiters(left='[', right=']')
1082 latex.matrix_column_alignment("c")
1083 show(Rx(s)(Gamma(q)(t)))
1084 show(diff(Rx(s)(Gamma(q)(t)), s)(s=0))

```

---

$$\begin{bmatrix} x \\ \cos(s)y - \sin(s)z \\ \sin(s)y + \cos(s)z \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ -z \\ y \end{bmatrix}$$

And now we check the result of the book. The computation of  $D \text{ F\_tilde}$  is somewhat complicated. Observe that  $\text{F\_tilde}$  is a function of the rotation angles, and returns a function that takes `local` as argument. Now we want to differentiate  $\text{F\_tilde}$  with respect to the angles, so these are the variables we need to provide to the Jacobian. For this reason, we bind the result of  $\text{F\_tilde}$  to `local`, and use a lambda function to provide the angles as the variables. This gives us  $\text{Ftilde}$  (note that I drop the underscore in this name). There is one further point:  $\text{F\_tilde}$  expects three angles, while the Jacobian provides the list  $[s, u, v]$  as the argument to  $\text{Ftilde}$ . Therefore we unpack the argument  $x$  of the lambda function to convert the list  $[s, u, v]$  into three separate arguments. The last step is to fill in  $s = u = v = 0$ .

Note that we differentiate wrt  $s, u, v$  and not wrt  $t$ . In itself, using  $t$  would not be a problem, but since we pass  $\text{Gamma}(q)(t)$  to  $\text{F\_tilde}$ , the function depends also on  $t$  via the path  $t \rightarrow \Gamma(q, t)$  which we should avoid.

As for the result, I don't see why my result differs by a minus sign from the result in the book.

---

```

1085 U = Function(lambda r: function("U")(r))
1086
1087
1088 def the_Noether_integral(local):
1089     L = L_central_rectangular(m, U)
1090     Ftilde = lambda x: F_tilde(*x)(local)
1091     DF0 = Jacobian(Ftilde)([s,u, v], [s, u, v])(s=0, u=0, v=0)
1092     return partial(L, 2)(local) * DF0

```

---

```

1093 show(the_Noether_integral(Gamma(q)(t)).simplify_full())

```

---

$$\begin{bmatrix} -mz\dot{y} + my\dot{z} & mz\dot{x} - mx\dot{z} & -my\dot{x} + mx\dot{y} \end{bmatrix}$$



## CONTENTS

---

### 1.9 ABSTRACTION OF PATH FUNCTIONS

I found this section difficult to understand, so I work in small steps to the final result, and include checks to see what goes on.

#### 1.9.1 *Standard imports*

```
_____/sage/utills1.9.sage_____  
1094 load("utills1.6.sage")  
_____  
_____/sage/section1.9.sage_____  
1095 load("utills1.9.sage")  
1096  
1097 var("t", domain=RR)  
_____  
_____/sage/section1.9.sage_____  
1098 load("show_expression.sage")  
_____
```

#### 1.9.2 *Understanding $F_{to_C}$*

The Scheme code starts with defining `Gamma_bar` in terms of `f_bar` and `osculating_path`. We build `f_bar` first and apply it to the example in which polar coordinates are converted to rectilinear coordinates.

Next, let's spell out the arguments of all functions to see how everything works together. A literal function maps time  $t$  to some part of the space, often to a coordinate,  $x$  say.

```
_____/sage/section1.9.sage_____  
1099 r, theta = literal_function("r"), literal_function("theta")  
1100 show(r)  
_____
```

```
<__main__.Function object at 0x752ed4eb27a0>
```

So, `r` is a Function. We can evaluate `r` at  $t$ . I pass `simplify=False` to show to *not* suppress the dependence on  $t$ .

```
_____/sage/section1.9.sage_____  
1101 show((r(t), theta(t)), simplify=False)  
_____
```

$$(r(t), \theta(t))$$

A `column_path` takes literal functions as arguments and returns a coordinate path. Hence, it is a function of  $t$  and returns  $q(t)$ . (I use the notation of the code examples of the book such as `q_prime` so that I can copy the examples into the functions I build later.)

---

```

1102 q_prime = column_path([r, theta])
1103 show(q_prime(t), simplify=False)

```

---

$$\begin{bmatrix} r(t) \\ \theta(t) \end{bmatrix}$$

The function  $\Gamma$  takes a coordinate path  $q$  (which is a function of time) as input, and returns a function of  $t$  that maps to a local up tuple  $l$ :

$$\Gamma[q] : t \rightarrow l = (t, q(t), v(t), \dots).$$

---

```

1104 show(Gamma(q_prime))

```

---

<function Gamma.<locals>.<lambda> at 0x752ed4ba0cc0>

Indeed, `Gamma` is a function, and has to be applied to some argument to result into a value. In fact, when  $\Gamma(q)$  is applied to  $t$ , we get the local up tuple  $l$ . Observe, that a local tuple is *not* a functions of time, by that I mean, a local is not a Python function of time, and therefore does not take any further arguments.

---

```

1105 show(Gamma(q_prime)(t), simplify=False)

```

---

$$\begin{matrix} t \\ \begin{bmatrix} r(t) \\ \theta(t) \end{bmatrix} \\ \begin{bmatrix} \frac{\partial}{\partial t} r(t) \\ \frac{\partial}{\partial t} \theta(t) \end{bmatrix} \end{matrix}$$

The coordinate transformation  $F$  in the example that transforms polar coordinates to rectilinear coordinates is `p_to_r`. This transform  $F$  maps a local tuple  $l$  to coordinates  $q(t)$ . Therefore, we can apply  $F$  to  $\Gamma[q](t)$ , and use composition like this:

$$F(\Gamma[q](t)) = (F \circ \Gamma[q])(t).$$

Observe that  $F \circ \Gamma[q]$  is a function of  $t$ .

---

```

1106 F = p_to_r
1107 show(compose(F, Gamma(q_prime))(t), simplify=False)

```

---

$$\begin{bmatrix} \cos(\theta(t))r(t) \\ r(t)\sin(\theta(t)) \end{bmatrix}$$

Since  $F \circ \Gamma[q]$  is a function of  $t$  to a coordinate path  $q(t)$ , this function has the same ‘protocol’ as a coordinate path function. We can therefore apply  $\Gamma$  to the composite function  $F \circ \Gamma[q]$  to obtain a function that maps  $t$  to a local tuple in the transformed space.

$$Q: t \rightarrow \Gamma[F \circ \Gamma[q]](t).$$

---

```

1108 Q = lambda t: compose(p_to_r, Gamma(q_prime))(t)
1109 show(Gamma(Q)(t), simplify=False)

```

---

$$\begin{bmatrix} \cos(\theta(t))r(t) \\ r(t)\sin(\theta(t)) \\ -r(t)\sin(\theta(t))\frac{\partial}{\partial t}\theta(t) + \cos(\theta(t))\frac{\partial}{\partial t}r(t) \\ \cos(\theta(t))r(t)\frac{\partial}{\partial t}\theta(t) + \sin(\theta(t))\frac{\partial}{\partial t}r(t) \end{bmatrix}$$

Now that we have analyzed all steps, we can make `f_bar`.

---

```

1110 def f_bar(q_prime):
1111     q = lambda t: compose(F, Gamma(q_prime))(t)
1112     return lambda t: Gamma(q)(t)

```

---

Here is the check. I suppress the dependence on  $t$  again to keep the result easier to read.

---

```

1113 show(f_bar(q_prime)(t))

```

---

$$\begin{bmatrix} \cos(\theta)r \\ r\sin(\theta) \\ -r\sin(\theta)\dot{\theta} + \cos(\theta)\dot{r} \\ \cos(\theta)r\dot{\theta} + \sin(\theta)\dot{r} \end{bmatrix}$$

The second function to build is `osculating_path`. This is the Taylor series of the book in which a local tuple is mapped to coordinate space:

$$O(t, q, v, a, \dots)(\cdot) = q + v(\cdot - t) + a/2(\cdot - t)^2 + \dots$$

I write  $\cdot$  instead of  $t'$  to make explicit that  $O(l)$  is still a function, of  $t'$  in this case.

Clearly, the RHS is a sum of vectors all of which have the same dimension as the space of coordinates.

Rather than computing  $dt^n$  as  $(t - t')^n$ , and  $n!$  for each  $n$ , I compute these values recursively. The implementation assumes that the local tuple  $\Gamma[q](t)$  contains at least the elements  $t$  and  $q$ , that is  $\Gamma[q](t) = (t, q, \dots)$ . This local tuple has length 2; the local tuple  $l = (t, q, v)$  has length 3.

---

```

1114 def osculating_path(local):
1115     t = time(local)
1116     q = coordinate(local)
1117
1118     def wrapper(t_prime):
1119         res = q
1120         dt = 1
1121         factorial = 1
1122         for k in range(2, len(local)):
1123             factorial *= k
1124             dt *= t_prime - t
1125             res += local[k] * dt / factorial
1126         return res
1127
1128     return wrapper

```

---

Here is an example.

---

```

1129 t_prime = var("tt", domain="positive", latex_name="t'")
1130 q = column_path([literal_function("r"), literal_function("theta")])
1131 local = Gamma(q)(t)
1132 show(osculating_path(local)(t_prime))

```

---

$$\begin{bmatrix} -\frac{1}{2}(t - t')\dot{r} + r \\ -\frac{1}{2}(t - t')\dot{\theta} + \theta \end{bmatrix}$$

With the above pieces we can finally build `Gamma_bar`.

---

```

1133 def Gamma_bar(f_bar):
1134     def wrapped(local):
1135         t = time(local)
1136         q_prime = osculating_path(local)
1137         return f_bar(q_prime)(t)
1138
1139     return wrapped

```

---

```

1140 show(Gamma_bar(f_bar)(local))

```

---

$$\begin{matrix} t \\ \begin{bmatrix} \cos(\theta)r \\ r\sin(\theta) \end{bmatrix} \\ \begin{bmatrix} -r\sin(\theta)\dot{\theta} + \cos(\theta)\dot{r} \\ \cos(\theta)r\dot{\theta} + \sin(\theta)\dot{r} \end{bmatrix} \end{matrix}$$

We can use `Gamma_bar` in to produce the transformation for polar to rectilinear coordinates.

---

```

1141 def F_to_C(F):
1142     def C(local):
1143         n = len(local)
1144
1145         def f_bar(q_prime):
1146             q = lambda t: compose(F, Gamma(q_prime))(t)
1147             return lambda t: Gamma(q, n)(t)
1148
1149         return Gamma_bar(f_bar)(local)
1150
1151     return C

```

---

```

1152 show(F_to_C(p_to_r)(local))

```

---

$$\begin{matrix} t \\ \begin{bmatrix} \cos(\theta)r \\ r\sin(\theta) \end{bmatrix} \\ \begin{bmatrix} -r\sin(\theta)\dot{\theta} + \cos(\theta)\dot{r} \\ \cos(\theta)r\dot{\theta} + \sin(\theta)\dot{r} \end{bmatrix} \end{matrix}$$

Here is the total time derivative.

---

```

1153 @Func
1154 def Dt(F):
1155     def DtF(local):
1156         n = len(local)
1157
1158         def DF_on_path(q):
1159             return D(lambda t: F(Gamma(q, n - 1)(t)))
1160
1161         return Gamma_bar(DF_on_path)(local)
1162
1163     return lambda state: DtF(local)

```

---

## 1.9.3 Lagrange equations at a moment

---

```

1164 def Euler_Lagrange_operator(L):
1165     return lambda local: (Dt(partial(L, 2)) - partial(L, 1))(local)

```

---

To apply this operator to a local tuple, we need to include the acceleration.

---

```

1166 q = column_path([literal_function("x")])
1167 local = Gamma(q, 4)(t)
1168 show(local)

```

---

$$\begin{matrix} t \\ \left[ \begin{matrix} x \\ \dot{x} \\ \ddot{x} \end{matrix} \right] \end{matrix}$$

---

```

1169 m, k = var("m k", domain="positive")
1170 L = L_harmonic(m, k)
1171 show(Euler_Lagrange_operator(L)(local))

```

---

$$\left[ kx + m\ddot{x} \right]$$

## CHAPTER 2

---

I skipped this one.

## CHAPTER 3

---

### 3.1 HAMILTON'S EQUATIONS

#### 3.1.1 *Standard imports*

---

```

1172 load("../sage/utils3.1.sage")
1173 load("../sage/section3.1.sage")
1174
1175 t = var("t", domain="real")
1176
1176 load("../sage/don't tangle")
1176 load("show_expression.sage")

```

---

#### 3.1.2 *Computing Hamilton's equations*

The code in Section 3.1 of the book starts with the following function.

---

```

1177 def Hamilton_equations(Hamiltonian):
1178     def f(q, p):
1179         state_path = qp_to_H_state_path(q, p)
1180         return D(state_path) - compose(
1181             Hamiltonian_to_state_derivative(Hamiltonian), state_path
1182         )
1183
1184     return f

```

---

This needs the next function.

---

```

1185 def qp_to_H_state_path(q, p):
1186     def f(t):
1187         return up(t, q(t), p(t))
1188
1189     return f

```

---

Here  $p$  is a function that maps  $t$  to a momentum vector. Such vectors are represented as lying vectors (or down tuples in the book). To implement this, `row_path` takes a list



and returns a function that maps time to the transpose of a column path. In passing we define `row_matrix` as the transpose of `column_matrix`, which is the function provided by Sagemath, and a transpose function.

---

```

1190 def transpose(M):
1191     return M.T
1192
1193
1194 def row_path(lst):
1195     return lambda t: transpose(column_path(lst)(t))
1196
1197
1198 def row_matrix(lst):
1199     return transpose(column_matrix(lst))

```

---

Let's try what we built.

---

```

1200 q = column_path([literal_function("q_x"), literal_function("q_y")])
1201 p = row_path([literal_function("p_x"), literal_function("p_y")])

```

---



---

```

1202 show(p(t))

```

---

$$\begin{bmatrix} p_x & p_y \end{bmatrix}$$

---

```

1203 H_state = qp_to_H_state_path(q, p)(t)
1204 show(H_state)

```

---

$$\begin{bmatrix} t \\ q_x \\ q_y \\ p_x & p_y \end{bmatrix}$$

The next function on which Hamiltonian equations depends is `Hamiltonian_to_state_derivative`. The book prints the system of differential equations as a column vector. Therefore we transpose  $\partial_2 H$ .

---

```

1205 def Hamiltonian_to_state_derivative(Hamiltonian):
1206     def f(H_state):
1207         return up(
1208             SR(1),
1209             partial(Hamiltonian, 2)(H_state).T,
1210             -partial(Hamiltonian, 1)(H_state),
1211         )
1212
1213     return f

```

---

Here is an example with `H_rectangular`. For some reason, the book takes just the first and second component of `q`, i.e. `(req q 0)` and `(ref q 1)`, to pass to the potential, but the general formula works just as well.

---

```

1214 var("m")
1215
1216 def H_rectangular(m, V):
1217     def f(state):
1218         q, p = coordinate(state), momentum(state)
1219         return square(p) / 2 / m + V(q)
1220
1221     return f

```

---

For this to work, we need a momentum projection operator. It's the same as the velocity projection.

---

```

1222 momentum = Function(lambda H_state: H_state[2])

```

---

Recall, to use symbolic functions in differentiation, the symbolic function requires an unpacked list of arguments.

---

```

1223 V = Function(lambda x: function("V")(*x.list()))

```

---

This is the Hamiltonian.

---

```

1224 H = H_rectangular
1225 show(H(m, V)(H_state))

```

---

$$\frac{p_x^2 + p_y^2}{2m} + V(q_x, q_y)$$

Partial derivatives work.

---

```

1226 show(partial(H(m, V), 1)(H_state))

```

---

$$\begin{bmatrix} D_0(V)(q_x, q_y) & D_1(V)(q_x, q_y) \end{bmatrix}$$

---

```

1227 show(Hamiltonian_to_state_derivative(H(m, V))(H_state))

```

---

$$\begin{bmatrix} 1 \\ \frac{p_x}{m} \\ \frac{p_y}{m} \end{bmatrix} \begin{bmatrix} -D_0(V)(q_x, q_y) & -D_1(V)(q_x, q_y) \end{bmatrix}$$

---

```

1228  ..../sage/section3.1.sage
show(Hamilton_equations(H(m, V))(q, p)(t))

```

---

$$\begin{bmatrix} 0 \\ -\frac{p_x}{m} + \dot{q}_x \\ -\frac{p_y}{m} + \dot{q}_y \end{bmatrix}$$

$$\begin{bmatrix} D_0(V)(q_x, q_y) + \dot{p}_x & D_1(V)(q_x, q_y) + \dot{p}_y \end{bmatrix}$$

### 3.1.3 The Legendre Transformation

To understand the code of the book, observe the following.

$$\begin{aligned} F(v) &= 1/2v^T Mv + b^t v + c, \\ \partial_v F(v) &= Mv + b, \\ \partial_v F(0) &= b, \\ \partial_v^2 F(v) &= M. \end{aligned}$$

Clearly,  $\partial_v F$  is the gradient, and  $\partial_v^2 F$  is the Hessian. Observe that under the operation of the gradient, the vector  $b$  changes shape: from  $b^t$  to  $b$ .

In the code, the argument  $w$  corresponds to a moment, hence is a lying vector. We need some dummy symbols with respect to which to differentiate, and then we set the dummy variables to 0 in the gradient and the Hessian. For this second step, Sagemath uses substitution with a dictionary when multiple arguments are involved, which is the case here because  $w$  is a vector. So, by making a zeros dictionary that maps symbols to 0, we can use the keys of zeros as the dummy symbols, and then use zeros itself in the substitution. Then, to solve for  $v$  such that  $Mv = w^t - b$ , the lying vector  $w$  has to be transposed.

---

```

1229  ..../sage/utils3.1.sage
def Legendre_transform(F):
1230      def G(w):
1231          zeros = {var(f"v_{i}") : 0 for i in range(w.ncols())}
1232          b = gradient(F)(list(zeros.keys())).subs(zeros)
1233          M = Hessian(F)(list(zeros.keys())).subs(zeros)
1234          v = M.solve_right(w.T - b)
1235          return w * v - F(v)
1236
1237      return G

```

---

Now we are equipped to convert a Lagrangian into a Hamiltonian.

---

```

1238  ..../sage/utils3.1.sage
def Lagrangian_to_Hamiltonian(Lagrangian):
1239     def f(H_state):
1240         t = time(H_state)

```

---

```

1241     q = coordinate(H_state)
1242     p = momentum(H_state)
1243
1244     def L(qdot):
1245         return Lagrangian(up(t, q, qdot))
1246
1247     return Legendre_transform(L)(p)
1248
1249     return f

```

---

```

1250 res = Lagrangian_to_Hamiltonian(L_central_rectangular(m, V))(H_state)
1251 show(res)

```

---

$$\left[ -\frac{1}{2}m\left(\frac{p_x^2}{m^2} + \frac{p_y^2}{m^2}\right) + \frac{p_x^2}{m} + \frac{p_y^2}{m} + V\left(\sqrt{q_x^2 + q_y^2}\right) \right]$$

```

1252 show(res.simplify_full())

```

---

$$\left[ \frac{2mV(\sqrt{q_x^2 + q_y^2}) + p_x^2 + p_y^2}{2m} \right]$$

```

1253 var("m g l")
1254 q = column_path([literal_function("theta")])
1255 p = row_path([literal_function("p")])

```

---

Here is exercise 3.1.

```

1256 # space = make_named_space(["\\theta"])
1257 H_state = qp_to_H_state_path(q, p)(t)
1258 show(Lagrangian_to_Hamiltonian(L_planar_pendulum(m, g, l))(H_state))

```

---

$$\left[ -glm(\cos(\theta) - 1) + \frac{p^2}{2l^2m} \right]$$

```

1259 q = column_path([literal_function("q_x"), literal_function("q_y")])
1260 p = row_path([literal_function("p_x"), literal_function("p_y")])
1261 H_state = qp_to_H_state_path(q, p)(t)
1262 show(Lagrangian_to_Hamiltonian(L_Henon_Heiles(m))(H_state))

```

---

$$\left[ q_x^2 q_y - \frac{1}{3} q_y^3 + \frac{1}{2} p_x^2 + \frac{1}{2} p_y^2 + \frac{1}{2} q_x^2 + \frac{1}{2} q_y^2 \right]$$

---

```

1263 def L_sphere(m, R):
1264     def Lagrangian(local):
1265         theta, phi = coordinate(local).list()
1266         thetadot, phidot = velocity(local).list()
1267         return 1 / 2 * m * R ^ 2 * (
1268             square(thetadot) + square(phidot * sin(theta))
1269         )
1270
1271     return Lagrangian
1272
1273
1274 var("R", domain="positive")

```

---

```

1275 q = column_path([literal_function("theta"), literal_function("phi")])
1276 p = row_path([literal_function("p_x"), literal_function("p_y")])
1277 H_state = qp_to_H_state_path(q, p)(t)
1278 show(Lagrangian_to_Hamiltonian(L_sphere(m, R))(H_state).simplify_full())

```

---

$$\left[ \frac{p_x^2 \sin(\theta)^2 + p_y^2}{2 R^2 m \sin(\theta)^2} \right]$$

## 3.2 POISSON BRACKETS

### 3.2.1 The standard imports.

### 3.2.2 Standard imports

---

```

1279 load("utils3.1.sage")

```

---

```

1280 load("utils3.2.sage")
1281
1282 t = var("t", domain="real")

```

---

```

1283 load("show_expression.sage")

```

---

### 3.2.3 The Poisson Bracket

This is the Poisson bracket.

---

```

1284 @Func

```

---

```

1285 def Poisson_bracket(F, G):
1286     def f(state):
1287         left = (partial(F, 1) * compose(transpose, partial(G, 2)))(state)
1288         right = (partial(F, 2) * compose(transpose, partial(G, 1)))(state)
1289         return (left - right).simplify_full()
1290
1291     return f

```

---

We can make general state functions like so.

```

1292 @Func
1293 def state_function(name):
1294     return lambda H_state: function(name)(
1295         time(H_state), *coordinate(H_state).list(), *momentum(H_state).list()
1296     )

```

---

The first test is to see whether  $\{Q, H\} = \partial_2 H$  and  $\{P, H\} = -\partial_1 H$ , where  $Q$  and  $P$  are the coordinate and momentum selectors, and  $H$  is a general state function.

```

1297 q = column_matrix([var("q_x"), var("q_y")])
1298 p = row_matrix([var("p_x"), var("p_y")])
1299 sigma = up(t, q, p)
1300 H = state_function("H")
1301
1302 show(Poisson_bracket(coordinate, H)(sigma))
1303 show(Poisson_bracket(momentum, H)(sigma))

```

---

$$\begin{bmatrix} \frac{\partial}{\partial p_x} H(t, q_x, q_y, p_x, p_y) \\ \frac{\partial}{\partial p_y} H(t, q_x, q_y, p_x, p_y) \end{bmatrix}$$

$$\begin{bmatrix} -\frac{\partial}{\partial q_x} H(t, q_x, q_y, p_x, p_y) \\ -\frac{\partial}{\partial q_y} H(t, q_x, q_y, p_x, p_y) \end{bmatrix}$$

All is correct. Note that both results are standing vectors.

### 3.2.4 Properties of the Poisson bracket

We know that  $\{H, H\} = 0$  for any function. Let's test this for our implementation.

```

1304 show(Poisson_bracket(H, H)(sigma))

```

---

$$\begin{bmatrix} 0 \end{bmatrix}$$

The property  $\{F, F\} = 0$  is actually implied when we can show that the Poisson bracket is anti-symmetric.

---

```

1305 F = state_function("F")
1306 G = state_function("G")
1307
1308 show((Poisson_bracket(F, G) + Poisson_bracket(G, F))(sigma))

```

---

$$\begin{bmatrix} 0 \end{bmatrix}$$

How about  $\{F, G + H\} = \{F, G\} + \{F, H\}$ ?

---

```

1309 show(
1310     (
1311         Poisson_bracket(F, G + H)
1312         - Poisson_bracket(F, G)
1313         - Poisson_bracket(F, H)
1314     )(sigma)
1315 )

```

---

$$\begin{bmatrix} 0 \end{bmatrix}$$

To check the rule  $\{F, cG\} = c\{F, G\}$  we need a constant function. By making the next function independent of any argument, it becomes constant.

---

```

1316 constant = Function(Lambda H_state: function("c"))

```

---

Is it indeed constant?

---

```

1317 show(Jacobian(constant)(sigma, sigma))

```

---

$$\begin{bmatrix} 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

So, next we can check  $\{F, cG\} = c\{F, G\}$ .

---

```

1318 show(
1319     (Poisson_bracket(F, constant * G) - constant * Poisson_bracket(F, G))(
1320         sigma
1321     ).simplify_full()
1322 )

```

---

$$\begin{bmatrix} 0 \end{bmatrix}$$

Finally, here is the check on Jacobi's identity.

---

```

1323 jacobi = (
1324     Poisson_bracket(F, Poisson_bracket(G, H))
1325     + Poisson_bracket(G, Poisson_bracket(H, F))
1326     + Poisson_bracket(H, Poisson_bracket(F, G))
1327 )
1328
1329 show(jacobi(sigma).simplify_full())

```

---

$$\begin{bmatrix} 0 \end{bmatrix}$$

### 3.2.5 Poisson bracket of a conserved quantity

To check that the Poisson bracket of a conserved quantity is conserved we need a function that does not depend on time.

---

```

1330 def f(H_state):
1331     return function("f")(
1332         *coordinate(H_state).list(), *momentum(H_state).list()
1333     )

```

---

Clearly, the derivative with respect to time of this function is zero, so it does what we need.

---

```

1334 show(diff(f(sigma), time(sigma)))

```

---

$$0$$

Now consider  $\{F, H\}$  where  $H$  is the rectangular Hamiltonian.

---

```

1335 V = Function(lambda q: function("V")(*q.list()))
1336
1337 var(m, domain="positive")
1338
1339 H = H_rectangular(m, V)

```

---

I compute the Poisson bracket of  $F$  and  $H$  for one dimension so that the result remains small.



---

```

1340 q = column_matrix([var("q")])
1341 p = row_matrix([var("p")])
1342 sigma = up(t, q, p)
1343
1344 show(Poisson_bracket(f, H)(sigma).expand())

```

---

$$\left[ -\frac{\partial}{\partial q} V(q) \frac{\partial}{\partial p} f(q, p) + \frac{p \frac{\partial}{\partial q} f(q, p)}{m} \right]$$

To complete the check, note that, by Hamilton's equation,  $\dot{q} = \partial H / \partial p$ ,  $\dot{p} = -\partial H / \partial q = -\partial V / \partial q$ . If we replace that in the above equation we obtain

$$\dot{p} \frac{\partial f}{\partial p} + \dot{q} \frac{\partial f}{\partial q} = \frac{df}{dt}.$$

Since  $f$  is conserved, the total time derivative of  $F$  is zero, hence  $f$  and  $H$  commute.

## CHAPTER 4

---

I skipped this one.