

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 OUTPUT TO L^AT_EX

We need some tricks to adapt the L^AT_EX 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^AT_EX, 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^AT_EX. 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^AT_EX 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^AT_EX output. However, when executing a code block in org mode, I *do* want to get L^AT_EX 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 \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")
_____/sage/differentiation_tests.sage_____
289 load("show_expression.sage")
_____don't tangle_____

```

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 show(diff(F, [x, y]))

```

$$b$$

This is the Jacobian.

```

299 show(jacobian(F, [x, y]))

```

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

```

300 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 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 def F(v):
303     return 1 / 2 * v * M * v + b * v + c

```

```

304 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 jacobian(F, v) # F has no arguments
307 jacobian(F(v), v) # v is not a list

```

The Tuple class supports differentiation.

```

308 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 \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.

Γ 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 n < 2:
410         raise ValueError("n must be > 1")
411     Dq = [q]
412     for k in range(2, n):
413         Dq.append(D(Dq[-1]))
414     return lambda t: up(t, *[v(t) for v in Dq])

```

When applying Gamma to a path, we get this.

```

415 local = Gamma(q)(t)
416 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.

```

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

```

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

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

```

418 time = Function(lambda local: local[0])
419 coordinate = Function(lambda local: local[1])
420 velocity = Function(lambda local: local[2])

```

```

421 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 Γ , we can continue with the Lagrangian of the free particle.

```

422 q = column_path(
423     [
424         literal_function("x"),
425         literal_function("y"),
426         literal_function("z"),
427     ]
428 )

```

```

429 show(q(t))

```

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

```

430 show(D(q)(t))

```

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

```

431 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 Γ 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 .

```

432 load("functions.sage")
433 m = var('m', domain='positive')
434 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)`.

```
435  ..../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.

```
436  ..../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^AT_EX formatting of `definite_integral`.

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

Here is the action along a generic path q .

```
448  ..../sage/section1.4.sage
T = var("T", domain="positive")
449
450  def Lagrangian_action(L, q, t1, t2):
451      return definite_integral(compose(L, Gamma(q))(t), t, t1, t2)
452
453  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.

```

454 test_path = Function(lambda t: vector([4 * t + 7, 3 * t + 5, 2 * t + 1]))
455 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.

```

456 hard_path = lambda t: vector([4 * t + 7, 3 * t + 5, 2 * exp(-t) + 1])
457
458 result = Lagrangian_action(L_free_particle(mass=3), hard_path, 0, 10)
459 show(result)
460 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.

```

461 @Func
462 def make_eta(nu, t1, t2):
463     return lambda t: (t - t1) * (t - t2) * nu(t)
464
465
466 nu = Function(lambda t: vector([sin(t), cos(t), t ^ 2]))
467
468 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.)

```

469 def varied_free_particle_action(mass, q, nu, t1, t2):
470     eta = make_eta(nu, t1, t2)
471

```

```

472     def f(eps):
473         return Lagrangian_action(L_free_particle(mass), q + eps * eta, t1, t2).n()
474
475     return f
476
477 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 ϵ .

```

_____/sage/section1.4.sage_____
478 res = find_local_minimum(
479     varied_free_particle_action(3.0, test_path, nu, 0.0, 10.0), -2.0, 1.0
480 )
481 show(res)

```

(435.0000000000000,0.0)

We see that the optimal value for ϵ 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/utls1.4.sage_____
482 def Lagrangian_polynomial(ts, qs):
483     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_____
484 ts = np.linspace(0, pi / 2, 5)
485 qs = [cos(t).n() for t in ts]
486 lp = Lagrangian_polynomial(ts, qs)
487 ts = np.linspace(0, pi / 2, 20)
488 Cos = [lp(x=t).n() for t in ts]
489 Sin = [lp.derivative(x)(x=t).n() for t in ts]
490 Zero = [abs(Cos[i] ^ 2 + Sin[i] ^ 2 - 1) for i in range(len(ts))]
491 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.

```

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

```

Here is the harmonic Lagrangian.

```

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

```

```

503 def parametric_path_action(Lagrangian, t0, q0, t1, q1):
504     def f(qs):
505         path = make_path(t0, q0, t1, q1, qs=qs)
506         return Lagrangian_action(Lagrangian, path, t0, t1)
507
508     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.

```

509 t0, t1 = 0, pi / 2
510 q0, q1 = cos(t0), cos(t1)
511 T = np.linspace(0, pi / 2, 5)
512 initial_qs = [cos(t).n() for t in T][1:-1]
513 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]`.)

```

514 def find_path(Lagrangian, t0, q0, t1, q1, n):
515     ts = np.linspace(t0, t1, n)
516     initial_qs = np.linspace(q0, q1, n)[1:-1]

```

```

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

```

0.000172462354236957

Great. All works!

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

```

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

```

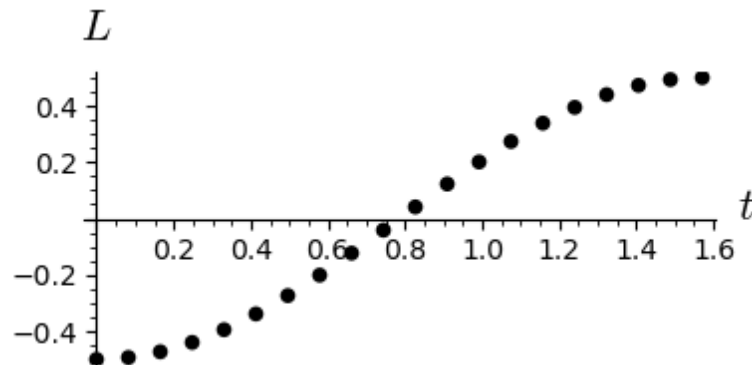


Figure 1.1: The harmonic Lagrangian as a function of the optimal path $q(t) = \cos t$, $t \in [0, \pi/2]$.

1.5 THE EULER-LAGRANGE EQUATIONS

1.5.1 Standard imports

```

535 load("utils1.4.sage")

```

```

536 load("utils1.5.sage")
537
538 t = var("t", domain="real")

```

```

539 load("show_expression.sage")

```

1.5.2 Derivation of the Lagrange equations

Harmonic oscillator

Here is a test on the harmonic oscillator.

```

540 load("utils1.4.sage")
541 k, m = var('k m', domain="positive")
542 q = column_path([literal_function("x")])

```

```

543 L = L_harmonic(m, k)
544 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 Γ . 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.)

```

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

```

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

```

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

```

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

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

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

```

547 show(compose(partial(L, 1), Gamma(q))(t))
548 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:

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

```

549 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

```

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

```

```

551 var("mu", domain="positive")
552
553 def L_orbital(m, mu):
554     def Lagrangian(local):
555         q = coordinate(local)
556         v = velocity(local)
557         return (1 / 2) * m * square(v) + mu / sqrt(square(q))
558
559     return Lagrangian

```

```

560 L = L_orbital(m, mu)
561 show(L(Gamma(q)(t)))

```

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

```

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

```

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

```

563 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 θ and $\dot{\theta}$.

```

564 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.

```

565 var("m g l", domain="positive")
566
567
568 def L_planar_pendulum(m, g, l):
569     def Lagrangian(local):
570         theta = coordinate(local).list()[0]
571         thetadot = velocity(local).list()[0]
572         T = (1 / 2) * m * l ^ 2 * square(thetadot)
573         V = m * g * l * (1 - cos(theta))
574         return T - V
575
576     return Lagrangian

```

```

577 L = L_planar_pendulum(m, g, l)
578 show(L(Gamma(q)(t)))

```

```

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

```

```

580  ..../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.

```

581  ..../sage/section1.5.sage
def L_Henon_Heiles(m):
582      def Lagrangian(local):
583          x, y = coordinate(local).list()
584          v = velocity(local)
585          T = (1 / 2) * square(v)
586          V = 1 / 2 * (square(x) + square(y)) + square(x) * y - y**3 / 3
587          return T - V
588
589      return Lagrangian

```

```

590  ..../sage/section1.5.sage
L = L_Henon_Heiles(m)
591  q = column_path([literal_function("x"), literal_function("y")])
592  show(L(Gamma(q)(t)))

```

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

```

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

```

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

```

594  ..../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

```

595 var('R', domain="positive")
596
597
598 def L_sphere(m, R):
599     def Lagrangian(local):
600         theta, phi = coordinate(local).list()
601         alpha, beta = velocity(local).list()
602         L = m * R * (square(alpha) + square(beta * sin(theta))) / 2
603         return L
604
605     return Lagrangian

```

```

606 q = column_path([literal_function("phi"), literal_function("theta")])
607 L = L_sphere(m, R)
608
609 show(L(Gamma(q)(t)))

```

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

```

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

```

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

```

611 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.

```

612 q = column_path(
613     [
614         literal_function("x"),
615         literal_function("y"),
616     ]
617 )
618
619 L = L_free_particle(m)
620 show(compose(partial(L, 1), Gamma(q))(t))
621 show(compose(partial(L, 2), Gamma(q))(t))
622 show(D(compose(partial(L, 2), Gamma(q)))(t))
623 show(
624     (D(compose(partial(L, 2), Gamma(q))) - compose(partial(L, 1), Gamma(q)))(t)
625 )

```

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

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

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

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

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

```

626 def Lagrange_equations(L):
627     def f(q):
628         return D(compose(partial(L, 2), Gamma(q))) - compose(
629             partial(L, 1), Gamma(q)
630         )
631
632     return f

```

The free particle

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

```

633 var("a b c a0 b0 c0", domain="real")
634 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.

```

635 l_eq = Lagrange_equations(L_free_particle(m))(test_path)
636 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.

```

637 q = column_path([literal_function("x")])
638 l_eq = Lagrange_equations(L_free_particle(m))(q)
639 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

```

640 var("A phi omega", domain="real")
641 assume(A > 0)
642
643 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 .

```

644 l_eq = Lagrange_equations(L_harmonic(m, k))(proposed_path)(t)
645 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 1×1 matrix, we take the element `[0][0]`.

```

646 show(l_eq[0][0])

```

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

Let's factor out the cosine.

```

647 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.

```

648 var("G m m1 m2", domain="positive")
649
650
651 def L_central_polar(m, V):
652     def Lagrangian(local):
653         r, phi = coordinate(local).list()
654         rdot, phidot = velocity(local).list()
655         T = 1 / 2 * m * (square(rdot) + square(r * phidot))
656         return T - V(r)
657
658     return Lagrangian
659
660
661 def gravitational_energy(G, m1, m2):
662     def f(r):
663         return -G * m1 * m2 / r
664
665     return f

```

```

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

```

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

```

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

```

```

671 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?

```

672 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*

```

673  _____ ../sage/utills1.6.sage _____
load("utills1.5.sage")

674  _____ ../sage/section1.6.sage _____
load("utills1.6.sage")

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

```

1.6.2 *Constant acceleration*

We start with a point in a uniform gravitational field.

```

676  _____ ../sage/utills1.6.sage _____
var("t", domain="real")
677  var("g m", domain="positive")
678
679
680  def L_uniform_acceleration(m, g):
681      def Lagrangian(local):
682          x, y = coordinate(local).list()
683          v = velocity(local)
684          T = 1 / 2 * m * square(v)
685          V = m * g * y
686          return T - V
687
688      return Lagrangian

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

```

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

1.6.3 Central force field

```

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

```

Let us first try this on a concrete potential function.

```

700 def U(r):
701     return 1 / r

```

```

702 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.

```

703 U = Function(lambda x: function("U")(x))
704 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.

```

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

```

1.6.5 polar coordinates

```

714 def p_to_r(local):
715     r, phi = coordinate(local).list()
716     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 ϕ .

```

717 r = literal_function("r")
718 phi = literal_function("phi")
719 q = column_path([r, phi])
720 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.

```

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

```

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

Next is the derivative wrt r and ϕ .

```

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

```

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

```

723 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.

```

724 def L_central_polar(m, U):
725     def Lagrangian(local):
726         return compose(L_central_rectangular(m, U), F_to_C(p_to_r))(local)
727
728     return Lagrangian

```

```

729 # show(L_central_polar(m, U)(Gamma(q)(t)))
730 show(L_central_polar(m, U)(Gamma(q)(t)).simplify_full())

```

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

```

731 expr = Lagrange_equations(L_central_polar(m, U))(q)(t)
732 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

```

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

```

```

758     def Lagrangian(local):
759         return L_free_polar(m)(F_to_C(F(0mega))(local))
760
761     return Lagrangian
762
763
764
765     def r_to_p(local):
766         x, y = coordinate(local).list()
767         return column_matrix([sqrt(x * x + y * y), atan(y / x)])
768
769
770     def L_rotating_rectangular(m, 0mega):
771         def Lagrangian(local):
772             return L_rotating_polar(m, 0mega)(F_to_C(r_to_p)(local))
773
774         return Lagrangian

```

```

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

```

```

778     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.

```

779     expr = Lagrange_equations(L_rotating_rectangular(m, 0mega))(q)(t)
780     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 θ . So, we need a transformation from

θ to x and y . Note that the function `coordinate` returns a (1×1) column matrix which just contains θ . 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.

```

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

```

```

789 def L_pend(m, l, g, ys):
790     def Lagrangian(local):
791         return L_uniform_acceleration(m, g)(
792             F_to_C(dp_coordinates(l, ys))(local)
793         )
794
795     return Lagrangian

```

```

796 _ = var("l", domain="positive")
797
798 theta = column_path([literal_function("theta")])
799 ys = literal_function("y")
800
801 expr = L_pend(m, l, g, ys)(Gamma(theta)(t)).simplify_full()
802 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

```

803 load("utils1.6.sage")

```

```

804 load("utils1.7.sage")
805
806 var("t", domain=RR)

```

```

807 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.

```

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

```

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

```

813 F = compose(transpose, partial(L, 1))
814 show(F(local))
815 P = partial(L, 2)
816 show((F - partial(P, 0))(local))

```

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

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

```

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

```

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

Convert to vector.

```

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

```

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

```

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

```

We apply this to the harmonic oscillator.

```

827  ..../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:

```

828  def de_rhs(x, y):
829      return [y, -x]
830
831
832  sol = desolve_odeint(de_rhs(x, y), [1, 0], srange(0, 100, 0.05), [x, y])
833  pp = list(zip(sol[:, 0], sol[:, 1]))
834  p = points(pp, color='blue', size=3)
835  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.

```

836  def de_rhs(x, y):
837      return [1, -1]

```

The solution is to replace the numbers by expressions.

```

838  def convert_to_expr(n):
839      return SR(n)

```

And then define the function of differentials like this.

```

840  def de_rhs(x, y):
841      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)`.

```

842 def Lagrangian_to_state_derivative(L):
843     acceleration = Lagrangian_to_acceleration(L)
844     return lambda state: up(
845         convert_to_expr(1), velocity(state), acceleration(state)
846     )

```

```

847 show(Lagrangian_to_state_derivative(L)(local))

```

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

```

848 def harmonic_state_derivative(m, k):
849     return Lagrangian_to_state_derivative(L_harmonic(m, k))

```

```

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

```

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

```

851 def qv_to_state_path(q, v):
852     return lambda t: up(t, q(t), v(t))

```

```

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

```

```

861 res = Lagrange_equations_first_order(L_harmonic(m, k))(
862     column_path([literal_function("x"), literal_function("y")]),
863     column_path([literal_function("v_x"), literal_function("v_y")]),
864 )
865 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.

```

866 def make_dummy_vector(name, dim):
867     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.

```

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

```

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

```

881 def state_advancer(state_derivative, ics, T):
882     init_time = time(ics)
883     times = [init_time, init_time + T]
884     soln = evolve(state_derivative, ics, times)
885     return soln[-1]

```

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

```

886 state_advancer(
887     harmonic_state_derivative(m=2, k=1),
888     ics=up(0, column_matrix([1, 2]), column_matrix([3, 4])),
889     T=10,
890 )

```

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.

```

891 def periodic_drive(amplitude, frequency, phase):
892     def f(t):
893         return amplitude * cos(frequency * t + phase)
894
895     return f

```

With this we make the Lagrangian.

```

896 _ = var("m l g A omega")
897
898
899 def L_periodically_driven_pendulum(m, l, g, A, omega):
900     ys = periodic_drive(A, omega, 0)
901
902     def Lagrangian(local):
903         return L_pend(m, l, g, ys)(local)
904
905     return Lagrangian

```

```

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

```

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

```

912 expr = Lagrange_equations(L_periodically_driven_pendulum(m, l, g, A, omega))(
913     q
914 )(t).simplify_full()
915 show(expr)

```

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

```

916 show(
917     Lagrangian_to_acceleration(
918         L_periodically_driven_pendulum(m, l, g, A, omega)
919     )(Gamma(q)(t)).simplify_full()
920 )

```

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

```

921 def pend_state_derivative(m, l, g, A, omega):
922     return Lagrangian_to_state_derivative(
923         L_periodically_driven_pendulum(m, l, g, A, omega)
924     )

```

```

925 expr = pend_state_derivative(m, l, g, A, omega)(Gamma(q)(t))
926 show(time(expr))
927 show(coordinate(expr).simplify_full())
928 show(velocity(expr).simplify_full())

```

$$1$$

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

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

```

929 def principal_value(cut_point):
930     def f(x):
931         return (x + cut_point) % (2 * np.pi) - cut_point
932
933     return f

```

```

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

So now we make the plot.

```

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

```

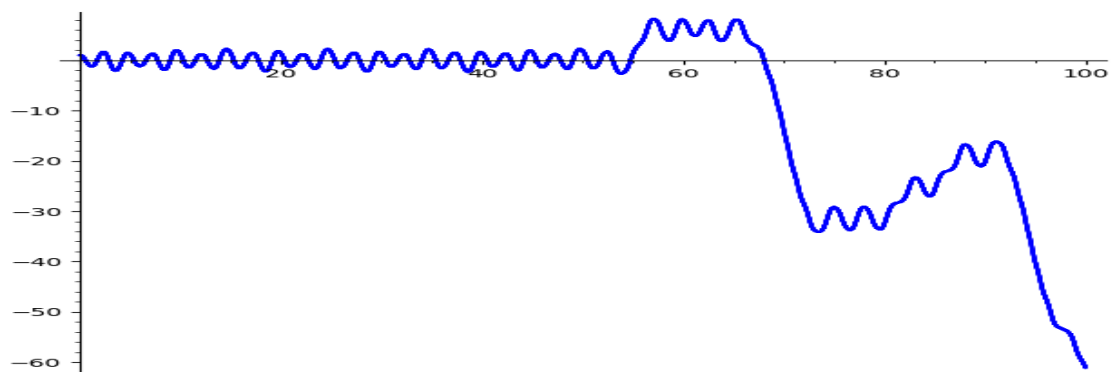
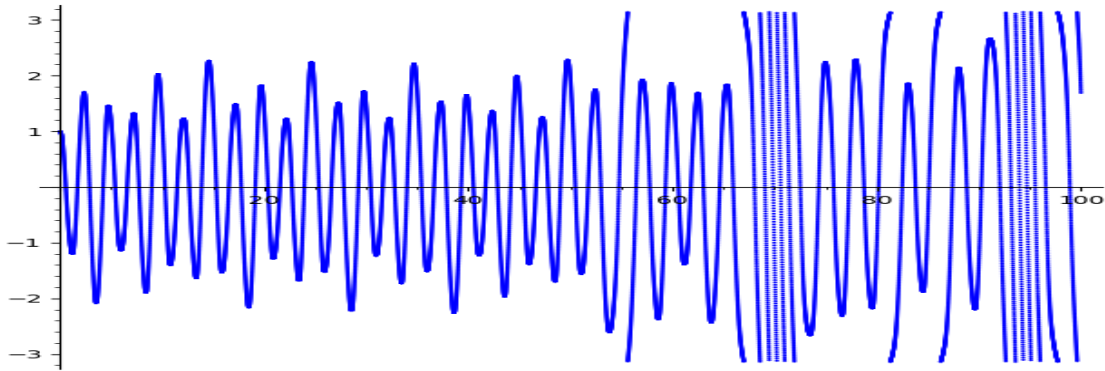
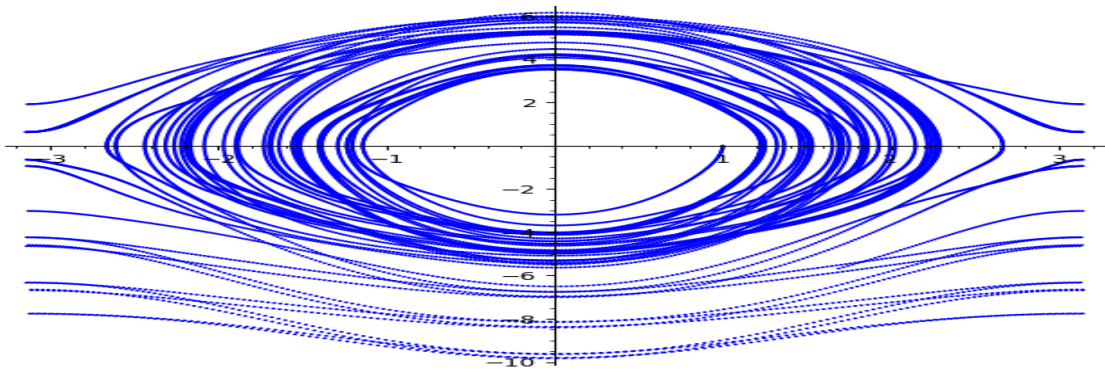


Figure 1.2: The angle of the vertically driven pendulum as a function of time. Obviously, around $t = 80$, the pendulum makes a few revolutions, and then starts to wobble again.

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

1.8 CONSERVED QUANTITIES

1.8.1 Standard imports

```

..... ../sage/Utils1.8.sage .....
957 load("utils1.6.sage")
.....

..... ../sage/section1.8.sage .....
958 load("utils1.8.sage")
959
960 var("t", domain=RR)
.....

..... don't tangle .....
961 load("show_expression.sage")
.....

```

1.8.2 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 1×1 matrix. As we use the Lagrangian in functional arithmetic, we convert L into a Function.

```

962 def Lagrangian_to_energy(L):
963     P = partial(L, 2)
964     LL = Function(lambda local: L(local))
965     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.

```

966 q = column_path(
967     [
968         literal_function("r"),
969         literal_function("theta"),
970         literal_function("phi"),
971     ]
972 )

```

Next the transformation from spherical to 3D rectangular coordinates.

```

973 def s_to_r(spherical_state):
974     r, theta, phi = coordinate(spherical_state).list()
975     return vector(
976         [r * sin(theta) * cos(phi), r * sin(theta) * sin(phi), r * cos(theta)]
977     )

```

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

```

978 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.

```

979 V = Function(lambda r: function("V")(r))
980
981 def L_3D_central(m, V):
982     def Lagrangian(local):
983         return L_central_rectangular(m, V)(F_to_C(s_to_r)(local))
984
985     return Lagrangian

```

```

986 show(partial(L_3D_central(m, V), 1)(Gamma(q)(t)).simplify_full())

```

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

```

987 show(partial(L_3D_central(m, V), 2)(Gamma(q)(t)).simplify_full())

```

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

```

988 def ang_mom_z(m):
989     def f(rectangular_state):
990         xyx = vector(coordinate(rectangular_state))
991         v = vector(velocity(rectangular_state))
992         return xyx.cross_product(m * v)[2]
993
994     return f
995
996
997 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 $E = T + V$.

```

998 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.

```

999 var("G M0 M1 a", domain="positive")
1000
1001
1002 def distance(x, y):
1003     return sqrt(square(x - y))
1004
1005
1006 def angular_freq(M0, M1, a):

```

```

1007     return sqrt(G * (M0 + M1) / a ^ 3)
1008
1009
1010 def V(a, M0, M1, m):
1011     Omega = angular_freq(M0, M1, a)
1012     a0, a1 = M1 / (M0 + M1) * a, M0 / (M0 + M1) * a
1013
1014     def f(t, origin):
1015         pos0 = -a0 * column_matrix([cos(Omega * t), sin(Omega * t)])
1016         pos1 = a1 * column_matrix([cos(Omega * t), sin(Omega * t)])
1017         r0 = distance(origin, pos0)
1018         r1 = distance(origin, pos1)
1019         return -G * m * (M0 / r0 + M1 / r1)
1020
1021     return f
1022
1023 def L0(m, V):
1024     def f(local):
1025         t, q, v = time(local), coordinate(local), velocity(local)
1026         return 1 / 2 * m * square(v) - V(t, q)
1027
1028     return f

```

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

```

..... ../sage/section1.8.sage .....
1029 q = column_path([literal_function("x"), literal_function("y")])
1030 expr = (sqrt(G*M0 + G*M1)*t) / a^(3/2)
1031 A = var('A')
1032
1033 show(
1034     Lagrangian_to_energy(L0(m, V(a, M0, M1, m)))(Gamma(q)(t))
1035     .simplify_full()
1036     .expand()
1037     .subs({expr: A})
1038 )

```

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

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 .....
1039 def rotation_matrix(axis, theta):

```

```

1040 """
1041 Return the 3x3 rotation matrix for a rotation of angle theta (in radians)
1042 about the given axis. The axis is specified as an iterable of 3 numbers.
1043 """
1044 # Convert the axis to a normalized vector
1045 axis = vector(axis).normalized()
1046 x, y, z = axis
1047 c = cos(theta)
1048 s = sin(theta)
1049 t = 1 - c # common factor
1050
1051 # Construct the rotation matrix using Rodrigues' formula
1052 R = matrix(
1053     [
1054         [c + x**2 * t, x * y * t - z * s, x * z * t + y * s],
1055         [y * x * t + z * s, c + y**2 * t, y * z * t - x * s],
1056         [z * x * t - y * s, z * y * t + x * s, c + z**2 * t],
1057     ]
1058 )
1059 return R

```

```

1060 def F_tilde(angle_x, angle_y, angle_z):
1061     def f(local):
1062         return (
1063             rotation_matrix([1, 0, 0], angle_x)
1064             * rotation_matrix([0, 1, 0], angle_y)
1065             * rotation_matrix([0, 0, 1], angle_z)
1066             * coordinate(local)
1067         )
1068
1069     return f

```

```

1070 q = column_path(
1071     [literal_function("x"), literal_function("y"), literal_function("z")]
1072 )

```

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

```

1073 def Rx(s):
1074     return lambda local: F_tilde(s, 0, 0)(local)
1075
1076 s, u, v = var("s u v")
1077 latex.matrix_delimiters(left='[', right=']')
1078 latex.matrix_column_alignment("c")
1079 show(Rx(s)(Gamma(q)(t)))
1080 show(diff(Rx(s)(Gamma(q)(t)), s)(s=0))
1081

```

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

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

And now we check the result of the book. The computation of $D F_tilde$ is somewhat complicated. Observe that F_tilde is a function of the rotation angles, and returns a function that takes `local` as argument. Now we want to differentiate 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 F_tilde to `local`, and use a lambda function to provide the angles as the variables. This gives us $Ftilde$ (note that I drop the underscore in this name). There is one further point: F_tilde expects three angles, while the Jacobian provides the list $[s, u, v]$ as the argument to $Ftilde$. 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 $\Gamma(q)(t)$ to 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.

```

1082 U = Function(lambda r: function("U")(r))
1083
1084
1085 def the_Noether_integral(local):
1086     L = L_central_rectangular(m, U)
1087     Ftilde = lambda x: F_tilde(*x)(local)
1088     DF0 = Jacobian(Ftilde)([s,u, v], [s, u, v])(s=0, u=0, v=0)
1089     return partial(L, 2)(local) * DF0

```

```

1090 show(the_Noether_integral(Gamma(q)(t)).simplify_full())

```

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

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/utis1.9.sage_____
1091 load("utis1.6.sage")
_____/sage/section1.9.sage_____
1092 load("utis1.9.sage")
1093
1094 var("t", domain=RR)
_____/sage/section1.9.sage_____
1095 load("show_expression.sage")
_____don't tangle_____

```

1.9.2 Understanding F_to_C

The Scheme code starts with defining Γ_{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_____
1096 r, theta = literal_function("r"), literal_function("theta")
1097 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_____
1098 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.)

```

_____/sage/section1.9.sage_____
1099 q_prime = column_path([r, theta])
1100 show(q_prime(t), simplify=False)
_____

```

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

The function Γ 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 :

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

```
1101  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.

```
1102  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 .

```
1103  F = p_to_r
1104  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 Γ to the composite function $F \circ \Gamma[q]$ to obtain a function that maps t to a local tuple in the transformed space.

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

```

1105 Q = lambda t: compose(p_to_r, Gamma(q_prime))(t)
1106 show(Gamma(Q)(t), simplify=False)

```

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

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

```

1107 def f_bar(q_prime):
1108     q = lambda t: compose(F, Gamma(q_prime))(t)
1109     return lambda t: Gamma(q)(t)

```

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

```

1110 show(f_bar(q_prime)(t))

```

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

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.

```

1111 def osculating_path(local):
1112     t = time(local)
1113     q = coordinate(local)

```

```

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

```

Here is an example.

```

----- ../sage/section1.9.sage -----
1126 t_prime = var("tt", domain="positive", latex_name="t'")
1127 q = column_path([literal_function("r"), literal_function("theta")])
1128 local = Gamma(q)(t)
1129 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`.

```

----- ../sage/section1.9.sage -----
1130 def Gamma_bar(f_bar):
1131     def wrapped(local):
1132         t = time(local)
1133         q_prime = osculating_path(local)
1134         return f_bar(q_prime)(t)
1135
1136     return wrapped

```

```

----- ../sage/section1.9.sage -----
1137 show(Gamma_bar(f_bar)(local))

```

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

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

```

1138 def F_to_C(F):
1139     def C(local):
1140         n = len(local)
1141
1142         def f_bar(q_prime):
1143             q = lambda t: compose(F, Gamma(q_prime))(t)
1144             return lambda t: Gamma(q, n)(t)
1145
1146         return Gamma_bar(f_bar)(local)
1147
1148     return C

```

```

1149 show(F_to_C(p_to_r)(local))

```

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

Here is the total time derivative.

```

1150 @Func
1151 def Dt(F):
1152     def DtF(local):
1153         n = len(local)
1154
1155         def DF_on_path(q):
1156             return D(lambda t: F(Gamma(q, n - 1)(t)))
1157
1158         return Gamma_bar(DF_on_path)(local)
1159
1160     return lambda state: DtF(local)

```

1.9.3 Lagrange equations at a moment

```

1161 def Euler_Lagrange_operator(L):
1162     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.

```

1163 q = column_path([literal_function("x")])
1164 local = Gamma(q, 4)(t)
1165 show(local)

```

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

```

1166 m, k = var("m k", domain="positive")
1167 L = L_harmonic(m, k)
1168 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*

```

1169  _____ ../sage/utils3.1.sage _____
load("utils1.6.sage")
_____

1170  _____ ../sage/section3.1.sage _____
load("utils3.1.sage")
1171
1172  t = var("t", domain="real")
_____

_____ don't tangle _____
1173  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.

```

_____ ../sage/utils3.1.sage _____
1174  def Hamilton_equations(Hamiltonian):
1175      def f(q, p):
1176          state_path = qp_to_H_state_path(q, p)
1177          return D(state_path) - compose(
1178              Hamiltonian_to_state_derivative(Hamiltonian), state_path
1179          )
1180
1181      return f
_____

```

This needs the next function.

```

_____ ../sage/utils3.1.sage _____
1182  def qp_to_H_state_path(q, p):
1183      def f(t):
1184          return up(t, q(t), p(t))
1185
1186      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.

```

1187 def transpose(M):
1188     return M.T
1189
1190
1191 def row_path(lst):
1192     return lambda t: transpose(column_path(lst)(t))
1193
1194
1195 def row_matrix(lst):
1196     return transpose(column_matrix(lst))

```

Let's try what we built.

```

1197 q = column_path([literal_function("q_x"), literal_function("q_y")])
1198 p = row_path([literal_function("p_x"), literal_function("p_y")])

```

```

1199 show(p(t))

```

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

```

1200 H_state = qp_to_H_state_path(q, p)(t)
1201 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 $\partial_2 H$.

```

1202 def Hamiltonian_to_state_derivative(Hamiltonian):
1203     def f(H_state):
1204         return up(
1205             SR(1),
1206             partial(Hamiltonian, 2)(H_state).T,
1207             -partial(Hamiltonian, 1)(H_state),
1208         )
1209
1210     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.

```

1211 var("m")
1212
1213 def H_rectangular(m, V):
1214     def f(state):
1215         q, p = coordinate(state), momentum(state)
1216         return square(p) / 2 / m + V(q)
1217
1218     return f

```

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

```

1219 momentum = Function(lambda H_state: H_state[2])

```

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

```

1220 V = Function(lambda x: function("V")(*x.list()))

```

This is the Hamiltonian.

```

1221 H = H_rectangular
1222 show(H(m, V)(H_state))

```

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

Partial derivatives work.

```

1223 show(partial(H(m, V), 1)(H_state))

```

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

```

1224 show(Hamiltonian_to_state_derivative(H(m, V))(H_state))

```

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

```

1225  ..../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.

```

1226  ..../sage/utils3.1.sage
def Legendre_transform(F):
1227      def G(w):
1228          zeros = {var(f"v_{i}") : 0 for i in range(w.ncols())}
1229          b = gradient(F)(list(zeros.keys())).subs(zeros)
1230          M = Hessian(F)(list(zeros.keys())).subs(zeros)
1231          v = M.solve_right(w.T - b)
1232          return w * v - F(v)
1233
1234      return G

```

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

```

1235  ..../sage/utils3.1.sage
def Lagrangian_to_Hamiltonian(Lagrangian):
1236      def f(H_state):
1237          t = time(H_state)

```

```

1238     q = coordinate(H_state)
1239     p = momentum(H_state)
1240
1241     def L(qdot):
1242         return Lagrangian(up(t, q, qdot))
1243
1244     return Legendre_transform(L)(p)
1245
1246     return f

```

```

1247 res = Lagrangian_to_Hamiltonian(L_central_rectangular(m, V))(H_state)
1248 show(res)

```

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

```

1249 show(res.simplify_full())

```

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

```

1250 var("m g l")
1251 q = column_path([literal_function("theta")])
1252 p = row_path([literal_function("p")])

```

Here is exercise 3.1.

```

1253 # space = make_named_space(["\\theta"])
1254 H_state = qp_to_H_state_path(q, p)(t)
1255 show(Lagrangian_to_Hamiltonian(L_planar_pendulum(m, g, l))(H_state))

```

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

```

1256 q = column_path([literal_function("q_x"), literal_function("q_y")])
1257 p = row_path([literal_function("p_x"), literal_function("p_y")])
1258 H_state = qp_to_H_state_path(q, p)(t)
1259 show(Lagrangian_to_Hamiltonian(L_Henon_Heiles(m))(H_state))

```

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

```

1260 def L_sphere(m, R):
1261     def Lagrangian(local):
1262         theta, phi = coordinate(local).list()
1263         thetadot, phidot = velocity(local).list()
1264         return 1 / 2 * m * R ^ 2 * (
1265             square(thetadot) + square(phidot * sin(theta))
1266         )
1267
1268     return Lagrangian
1269
1270
1271 var("R", domain="positive")

```

```

1272 q = column_path([literal_function("theta"), literal_function("phi")])
1273 p = row_path([literal_function("p_x"), literal_function("p_y")])
1274 H_state = qp_to_H_state_path(q, p)(t)
1275 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

```

1276 load("utils3.1.sage")

```

```

1277 load("utils3.2.sage")
1278
1279 t = var("t", domain="real")

```

```

1280 load("show_expression.sage")

```

3.2.3 The Poisson Bracket

This is the Poisson bracket.

```

1281 @Func

```

```

1282 def Poisson_bracket(F, G):
1283     def f(state):
1284         left = (partial(F, 1) * compose(transpose, partial(G, 2)))(state)
1285         right = (partial(F, 2) * compose(transpose, partial(G, 1)))(state)
1286         return (left - right).simplify_full()
1287
1288     return f

```

We can make general state functions like so.

```

1289 @Func
1290 def state_function(name):
1291     return lambda H_state: function(name)(
1292         time(H_state), *coordinate(H_state).list(), *momentum(H_state).list()
1293     )

```

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.

```

1294 q = column_matrix([var("q-x"), var("q-y")])
1295 p = row_matrix([var("p-x"), var("p-y")])
1296 sigma = up(t, q, p)
1297 H = state_function("H")
1298
1299 show(Poisson_bracket(coordinate, H)(sigma))
1300 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.

```

1301 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.

```

1302 F = state_function("F")
1303 G = state_function("G")
1304
1305 show((Poisson_bracket(F, G) + Poisson_bracket(G, F))(sigma))

```

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

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

```

1306 show(
1307     (
1308         Poisson_bracket(F, G + H)
1309         - Poisson_bracket(F, G)
1310         - Poisson_bracket(F, H)
1311     )(sigma)
1312 )

```

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

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.

```

1313 constant = Function(lambda H_state: function("c")())

```

Is it indeed constant?

```

1314 show(Jacobian(constant)(sigma, sigma))

```

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

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

```

1315 show(
1316     (Poisson_bracket(F, constant * G) - constant * Poisson_bracket(F, G))(
1317         sigma
1318     ).simplify_full()
1319 )

```

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

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

```

1320 jacobi = (
1321     Poisson_bracket(F, Poisson_bracket(G, H))
1322     + Poisson_bracket(G, Poisson_bracket(H, F))
1323     + Poisson_bracket(H, Poisson_bracket(F, G))
1324 )
1325
1326 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.

```

1327 def f(H_state):
1328     return function("f")(
1329         *coordinate(H_state).list(), *momentum(H_state).list()
1330     )

```

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

```

1331 show(diff(f(sigma), time(sigma)))

```

$$0$$

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

```

1332 V = Function(lambda q: function("V")(*q.list()))
1333
1334 var(m, domain="positive")
1335
1336 H = H_rectangular(m, V)

```

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

```

1337 q = column_matrix([var("q")])
1338 p = row_matrix([var("p")])
1339 sigma = up(t, q, p)
1340
1341 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.