

Structure and Interpretation of Classical Mechanics with Python and Sagemath

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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 Scheme code of the book to understand the Python code. In other words, I 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 use `re` to modify LaTeX strings. 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 this too.

```

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

```

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, products 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 by means of 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")

```

I load `show_expression` to control the \LaTeX output in this org file.

don't tangle

```

157 load("show_expression.sage")
158
159 def show(s, simplify=True):
160     return show_expression(s, simplify)

```

0.4.2 The Function class

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

../sage/functions.sage

```

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

```

The next function will be used as decorator on a function *f*, say, that returns another function *inner_f*, say, so that *inner_f* becomes a Function.

../sage/functions.sage

```

190 def Func(f):
191     def wrapper(*args, **kwargs):
192         return Function(f(*args, **kwargs))
193
194     return wrapper

```

Composition is just a recursive call of functions.

```

195 @Func
196 def compose(*funcs):
197     if len(funcs) == 1:
198         return lambda x: funcs[0](x)
199     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.

```

200 def f(x):
201     return 5 * x
202
203
204 F = Function(lambda x: f(x))

```

We will use quadratic functions often, so let's make a function for this. When x is a vector, Sagemath interprets x^2 as $x^t \cdot x$, which we want. The identity is just interesting. Perhaps we'll use it later.

```

205 identity = Function(lambda x: x)
206
207 def _square(x):
208     if isinstance(
209         x,
210         (
211             int,
212             float,
213             sage.symbolic.expression.Expression,
214             sage.rings.integer.Integer,
215         ),
216     ):
217         return x ^ 2
218     elif isinstance(x, (Vector, list, tuple)):
219         v = vector(x)
220         return v.dot_product(v)
221     elif isinstance(x, Matrix) and x.ncols() == 1:
222         return (x.transpose() * x)[0, 0]
223     else:
224         raise TypeError(f"Unsupported type: {type(x)}")
225
226
227 square = Function(lambda x: _square(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.

```

228 sin = Function(lambda x: sage.functions.trig.sin(x))
229 cos = Function(lambda x: sage.functions.trig.cos(x))

```

To use Sagemath functions we make an abbreviation.

```

230 function = sage.symbolic.function_factory.function

```

Now we can make symbolic functions like so.

```

231 V = Function(lambda x: function("V") (x))

```

0.4.4 Examples

```

232 x, y = var("x y", domain = RR)
233
234 show((square)(x + y).expand())

```

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

```

235 show((square + square)(x + y))

```

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

```

236 show((square * square)(x))

```

$$x^4$$

```

237 show((sin + cos)(x))

```

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

```

238 show((square + V)(x))

```

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

```

239 hh = compose(square, sin)
240 show((hh + hh)(x))

```

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

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

```

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

```

$$0$$

Next, we test differentiation and integration.

```

242 show(diff(-compose(square, cos)(x), x))
243 show(integrate((2 * sin * cos)(x), x))

```

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

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

Arithmetic with symbolic functions works too.

```

244 U = Function(lambda x: function("U")(x))
245 V = Function(lambda x: function("V")(x))

```

```

246 show((U + V)(x))
247 show((V + V)(x))
248 show((V(U(x))))
249 show((compose(V, U)(x)))

```

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

$$2V(x)$$

$$V(U(x))$$

$$V(U(x))$$

```

250 def f(x):
251     def g(y):
252         return x * y ^ 2
253
254     return g

```

```

255 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'`.

```

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

```

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

```

257 @Func
258 def f(x):
259     def g(y):
260         return x * y ^ 2
261
262     return g

```

```

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

```

80

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

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

```

264 def f(x):
265     def g(y):
266         return x * y ^ 2
267
268     return Function(lambda y: g(y))

```

```

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

```

80

0.5 DIFFERENTIATION

0.5.1 *Standard imports*

```

_____/sage/differentiation.sage_____
270 load(
271     "functions.sage",
272     "tuples.sage",
273 )
_____/sage/differentiation_tests.sage_____
274 load("differentiation.sage")
275
276 var("t", domain="real")
_____/sage/differentiation_tests.sage_____
277 load("show_expression.sage")
_____don't tangle_____

```

0.5.2 *Examples with matrices, functions and tuples*

```

_____/sage/differentiation_tests.sage_____
278 _ = var("a b c x y", domain=RR)
279 M = matrix([[a, b], [b, c]])
280 b = vector([a, b])
281 v = vector([x, y])
282 F = 1 / 2 * v * M * v + b * v + c
_____/sage/differentiation_tests.sage_____
283 show(F)
_____

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

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

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

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

```

$$ax + by + a$$

Repeated differentiation works nicely.

```

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

```

$$b$$

This is the Jacobian.

```

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

```

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

```

288 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.

```

289 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.

```

290 def F(v):
291     return 1 / 2 * v * M * v + b * v + c

```

```

292 show(diff(F(v), x)) # add the arguments to F
293 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.

```

294 jacobian(F, v) # F has no arguments
295 jacobian(F(v), v) # v is not a list

```

The Tuple class supports differentiation.

```

296 T = up(t, t ^ 2, t ^ 3, sin(3 * t))
297 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:

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

```

298 @Func
299 def D(f):
300     return lambda t: diff(f(t), t)
301     #return derivative(expr, t)

```

Here is an example.

```

302 q = Function(lambda t: function("q")(t))
303
304 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:

```

305 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.

```

306 F = 5 * q + 3 * t
307
308 show(diff(F, r)) # does not work
309 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.

```

310 def Jacobian(F):
311     def f(args, vrs):
312         if isinstance(args, (list, tuple)):
313             args = vector(args)
314         if isinstance(vrs, (list, tuple)):
315             vrs = vector(vrs)
316         subs = {
317             v: var(f"v{id(v)}", domain=RR)
318             for v in args.list()
319             if not v.is_symbol()
320         }
321         result = jacobian(F(args.subs(subs)), vrs.subs(subs).list())
322         inverse_subs = {v: k for k, v in subs.items()}
323         return result.subs(inverse_subs)
324
325     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.

```

326 v = var("v", domain=RR)
327
328
329 def F(v):
330     r, t = v.list()
331     return 5 * r ^ 3 + 3 * t ^ 2 * r
332
333
334 show(Jacobian(F)([v, t], [t]))
335 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.

```

336 q = Function(lambda t: function("q")(t))
337 v = [q(t), t]
338 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.

```

339 def gradient(F, v):
340     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.

```

341 def gradient(F):
342     return lambda v: Jacobian(F)(v, v).T

```

```

343  ..../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.

```

344  ..../sage/differentiation_tests.sage
U = Function(lambda x: function("U")(square(x)))
345  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.

```

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

```

```

348  ..../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.

```

349  ..../sage/differentiation.sage
@Func
350  def partial(f, slot):
351      def wrapper(local):
352          if slot == 0:
353              selection = [time(local)]
354          elif slot == 1:
355              selection = coordinate(local)
356          elif slot == 2:
357              selection = velocity(local)
358          return Jacobian(f)(local, selection)
359
360      return wrapper

```

The main text contains many examples.

CHAPTER 1

1.4 COMPUTING ACTIONS

1.4.1 *Standard setup*

I make 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 gets tangled to `utils1.4.sage`, 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.

```

361 import numpy as np
362
363 load("functions.sage", "differentiation.sage", "tuples.sage")

```

BTW, don't load from `sage.all` `import *` after loading `utils.sage`, because that will lead to name space conflict with the Gamma function, which we define below.

```

364 load("utils1.4.sage")
365
366 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.

```

367 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.

```

368 def L_free_particle(mass):
369     def Lagrangian(local):
370         v = velocity(local)
371         return 1 / 2 * mass * square(v)
372
373     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`.

```

374 @Func
375 def literal_function(name):
376     return lambda t: function(name)(t)

```

It's a function.

```

377 x = literal_function("x")
378 print(x)

```

<__main__.Function object at 0x71122066e470>

Here are some operations on x .

```

379 show(x(t))
380 show((x+x)(t))
381 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, `path_function` 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`.

```

382 @Func
383 def path_function(lst):
384     #return lambda t: vector([l(t) for l in lst])
385     return lambda t: column_matrix([l(t) for l in lst])

```

```

386 q = path_function(
387     [
388         literal_function("x"),
389         literal_function("y"),
390     ]
391 )

```

Here is an example to see how to use q.

```

392 show(q(t))

```

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

```

393 show((q + q)(t))

```

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

```

394 show((2 * q)(t))

```

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

```

395 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.

```

396 def Gamma(q, n=3):
397     # if isinstance(q, np.ndarray):
398     #     q = vector(q.tolist()) # todo, is this still needed?
399
400     if n < 2:
401         raise ValueError("n must be > 1")
402     Dq = [q]
403     for k in range(2, n):
404         Dq.append(D(Dq[-1]))
405     return lambda t: up(t, *[v(t) for v in Dq])

```

When applying Gamma to a path, we get this.

```

406 local = Gamma(q)(t)
407 show(local)

```

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

We can include the acceleration too.

```

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

```

$$\begin{matrix} t \\ \begin{bmatrix} x \\ y \\ \dot{x} \\ \dot{y} \\ \ddot{x} \\ \ddot{y} \end{bmatrix} \end{matrix}$$

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

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

```

409 time = Function(lambda local: local[0])
410 coordinate = Function(lambda local: local[1])
411 velocity = Function(lambda local: local[2])

```

```

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

```

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

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.

```

413 q = path_function(
414     [
415         literal_function("x"),
416         literal_function("y"),
417         literal_function("z"),
418     ]
419 )

```

```

420 show(q(t))

```

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

```

421 show(D(q)(t))

```

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

```

422 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 .

```

423 load("functions.sage")
424 m = var('m', domain='positive')
425 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)`.

```
426  ..../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.

```
427  ..../sage/utis1.4.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`.

```
428  ..../sage/utis1.4.sage
def integral_latex_format(*args):
429      expr, var, a, b = args
430      return (
431          fr"\int_{{a}}^{{b}} "
432          + latex(expr)
433          + r"\, \text{trm}{d}\,",
434          + latex(var)
435      )
436
437
438  definite_integral._print_latex_ = integral_latex_format
```

Here is the action along a generic path q .

```
439  ..../sage/section1.4.sage
T = var("T", domain="positive")
440
441  def Lagrangian_action(L, q, t1, t2):
442      return definite_integral(compose(L, Gamma(q))(t), t, t1, t2)
443
444  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.

```

445 test_path = Function(lambda t: vector([4 * t + 7, 3 * t + 5, 2 * t + 1]))
446 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.

```

447 hard_path = lambda t: vector([4 * t + 7, 3 * t + 5, 2 * exp(-t) + 1])
448
449 result = Lagrangian_action(L_free_particle(mass=3), hard_path, 0, 10)
450 show(result)
451 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.

```

452 @Func
453 def make_eta(nu, t1, t2):
454     return lambda t: (t - t1) * (t - t2) * nu(t)
455
456
457 nu = Function(lambda t: vector([sin(t), cos(t), t ^ 2]))
458
459 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.)

```

460 def varied_free_particle_action(mass, q, nu, t1, t2):
461     eta = make_eta(nu, t1, t2)
462

```

```

463     def f(eps):
464         return Lagrangian_action(L_free_particle(mass), q + eps * eta, t1, t2).n()
465
466     return f
467
468 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_____
469 res = find_local_minimum(
470     varied_free_particle_action(3.0, test_path, nu, 0.0, 10.0), -2.0, 1.0
471 )
472 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_____
473 def Lagrangian_polynomial(ts, qs):
474     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_____
475 ts = np.linspace(0, pi / 2, 5)
476 qs = [cos(t).n() for t in ts]
477 lp = Lagrangian_polynomial(ts, qs)
478 ts = np.linspace(0, pi / 2, 20)
479 Cos = [lp(x=t).n() for t in ts]
480 Sin = [lp.derivative(x)(x=t).n() for t in ts]
481 Zero = [abs(Cos[i] ^ 2 + Sin[i] ^ 2 - 1) for i in range(len(ts))]
482 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.

```

483 def make_path(t0, q0, t1, q1, qs):
484     ts = np.linspace(t0, t1, len(qs) + 2)
485     qs = np.r_[q0, qs, q1]
486     return lambda t: vector([Lagrangian_polynomial(ts, qs)(t)])

```

Here is the harmonic Lagrangian.

```

487 def L_harmonic(m, k):
488     def Lagrangian(local):
489         q = coordinate(local)
490         v = velocity(local)
491         return (1 / 2) * m * square(v) - (1 / 2) * k * square(q)
492
493     return Lagrangian

```

```

494 def parametric_path_action(Lagrangian, t0, q0, t1, q1):
495     def f(qs):
496         path = make_path(t0, q0, t1, q1, qs=qs)
497         return Lagrangian_action(Lagrangian, path, t0, t1)
498
499     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.

```

500 t0, t1 = 0, pi / 2
501 q0, q1 = cos(t0), cos(t1)
502 T = np.linspace(0, pi / 2, 5)
503 initial_qs = [cos(t).n() for t in T][1:-1]
504 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]`.)

```

505 def find_path(Lagrangian, t0, q0, t1, q1, n):
506     ts = np.linspace(t0, t1, n)
507     initial_qs = np.linspace(q0, q1, n)[1:-1]

```

```

508     minimizing_qs = minimize(
509         parametric_path_action(Lagrangian, t0, q0, t1, q1),
510         initial_qs,
511     )
512     return make_path(t0, q0, t1, q1, minimizing_qs)
513
514 best_path = find_path(L_harmonic(m=1, k=1), t0=0, q0=1, t1=pi / 2, q1=0, n=5)
515 result = [
516     abs(best_path(t)[0].n() - cos(t).n()) for t in np.linspace(0, pi / 2, 10)
517 ]
518 show(max(result))

```

0.000172462354236957

Great. All works!

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

```

519 T = np.linspace(0, pi / 2, 20)
520 q = lambda t: vector([cos(t)])
521 lvalues = [L_harmonic(m=1, k=1)(Gamma(q)(t))(t=ti).n() for ti in T]
522 points = list(zip(ts, lvalues))
523 plot = list_plot(points, color="black", size=30)
524 plot.axes_labels(["$t$", "$L$"])
525 plot.save("../figures/Lagrangian.png", figsize=(4, 2))

```

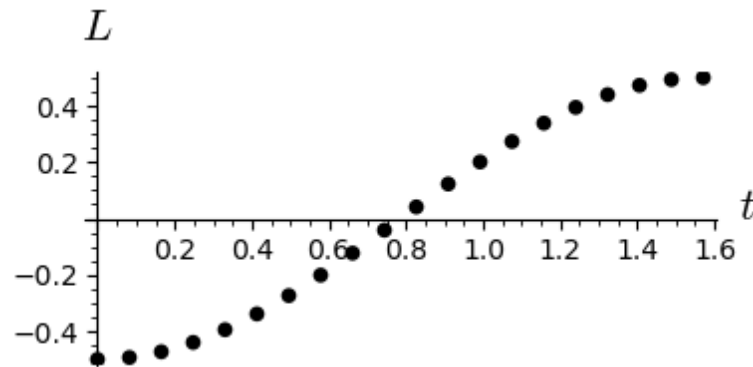


Figure 1.1: The harmonic Lagrangian as a function of the optimal path $q(t) = \cos t$, $t \in [0, \pi/2]$.

1.5 THE EULER-LAGRANGE EQUATIONS

1.5.1 Standard imports

```

526 load("utils1.4.sage")

```

```

527 load("utils1.5.sage")
528
529 t = var("t", domain="real")

```

```

530 load("show_expression.sage")

```

1.5.2 Derivation of the Lagrange equations

Harmonic oscillator

Here is a test on the harmonic oscillator.

```

531 load("utils1.4.sage")
532 k, m = var('k m', domain="positive")
533 q = path_function([literal_function("x")])

```

```

534 L = L_harmonic(m, k)
535 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.)

```

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

```

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

```

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

```

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

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

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

```

538 show(compose(partial(L, 1), Gamma(q))(t))
539 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:

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

```

540 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

```

541 q = path_function([literal_function("xi"), literal_function("eta")])

```

```

542 var("mu", domain="positive")
543
544 def L_orbital(m, mu):
545     def Lagrangian(local):
546         q = coordinate(local)
547         v = velocity(local)
548         return (1 / 2) * m * square(v) + mu / sqrt(square(q))
549
550     return Lagrangian

```

```

551 L = L_orbital(m, mu)
552 show(L(Gamma(q)(t)))

```

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

```

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

```

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

```

554 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 $\dot{\theta}$.

```

555 q = path_function([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.

```

556 var("m g l", domain="positive")
557
558
559 def L_planar_pendulum(m, g, l):
560     def Lagrangian(local):
561         theta = coordinate(local)[0]
562         thetadot = velocity(local)
563         T = (1 / 2) * m * l ^ 2 * square(thetadot)
564         V = m * g * l * (1 - cos(theta))
565         return T - V
566
567     return Lagrangian

```

```

568 L = L_planar_pendulum(m, g, l)
569 show(L(Gamma(q)(t)))

```

```

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

```

```

571 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.

```

572 def L_Henon_Heiles(m):
573     def Lagrangian(local):
574         x, y = coordinate(local).list()
575         v = velocity(local)
576         T = (1 / 2) * square(v)
577         V = 1 / 2 * (square(x) + square(y)) + square(x) * y - y**3 / 3
578         return T - V
579
580     return Lagrangian

```

```

581 L = L_Henon_Heiles(m)
582 q = path_function([literal_function("x"), literal_function("y")])
583 show(L(Gamma(q)(t)))

```

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

```

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

```

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

```

585 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

```

586 var('R', domain="positive")
587
588
589 def L_sphere(m, R):
590     def Lagrangian(local):
591         theta, phi = coordinate(local).list()
592         alpha, beta = velocity(local).list()
593         L = m * R * (square(alpha) + square(beta * sin(theta))) / 2
594         return L
595
596     return Lagrangian

```

```

597 q = path_function([literal_function("phi"), literal_function("theta")])
598 L = L_sphere(m, R)
599
600 show(L(Gamma(q)(t)))

```

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

```

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

```

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

```

602 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.

```

603 q = path_function(
604     [
605         literal_function("x"),
606         literal_function("y"),
607     ]
608 )
609
610 L = L_free_particle(m)
611 show(compose(partial(L, 1), Gamma(q))(t))
612 show(compose(partial(L, 2), Gamma(q))(t))
613 show(D(compose(partial(L, 2), Gamma(q)))(t))
614 show(
615     (D(compose(partial(L, 2), Gamma(q))) - compose(partial(L, 1), Gamma(q)))(t)
616 )

```

$$\begin{bmatrix} 0 & 0 \\ m\dot{x} & m\dot{y} \\ m\ddot{x} & m\ddot{y} \\ m\ddot{x} & m\ddot{y} \end{bmatrix}$$

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

```

617 def Lagrange_equations(L):
618     def f(q):
619         return D(compose(partial(L, 2), Gamma(q))) - compose(
620             partial(L, 1), Gamma(q)
621         )
622
623     return f

```

The free particle

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

```

624 var("a b c a0 b0 c0", domain="real")
625 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 \mathcal{L}_{eq} we receive a function instead of vector.

```

626 l_eq = Lagrange_equations(L_free_particle(m))(test_path)
627 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.

```

628 q = path_function([literal_function("x")])
629 l_eq = Lagrange_equations(L_free_particle(m))(q)
630 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

```

631 var("A phi omega", domain="real")
632 assume(A > 0)
633
634 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 .

```

635 l_eq = Lagrange_equations(L_harmonic(m, k))(proposed_path)(t)
636 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 $[0][0]$.

```

637 show(l_eq[0][0])

```

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

Let's factor out the cosine.

```

638 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.

```

639 var("G m m1 m2", domain="positive")
640
641
642 def L_central_polar(m, V):
643     def Lagrangian(local):
644         r, phi = coordinate(local).list()
645         rdot, phidot = velocity(local).list()
646         T = 1 / 2 * m * (square(rdot) + square(r * phidot))
647         return T - V(r)
648
649     return Lagrangian
650
651

```

```

652 def gravitational_energy(G, m1, m2):
653     def f(r):
654         return -G * m1 * m2 / r
655
656     return f

```

```

        ../sage/section1.5.sage
657 q = path_function([literal_function("r"), literal_function("phi")])
658 V = gravitational_energy(G, m1, m2)
659 L = L_central_polar(m, V)
660 show(L(Gamma(q)(t)))

```

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

```

        ../sage/section1.5.sage
661 l_eq = Lagrange_equations(L)(q)(t)

```

```

        ../sage/section1.5.sage
662 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?

```

        ../sage/section1.5.sage
663 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

```

        ../sage/utils1.6.sage
664 load("utils1.5.sage")

```

```

        ../sage/section1.6.sage
665 load("utils1.6.sage")

```

```

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

```

1.6.2 Constant acceleration

We start with a point in a uniform gravitational field.

```

667 var("t", domain="real")
668 var("g m", domain="positive")
669
670
671 def L_uniform_acceleration(m, g):
672     def Lagrangian(local):
673         x, y = coordinate(local).list()
674         v = velocity(local)
675         T = 1 / 2 * m * square(v)
676         V = m * g * y
677         return T - V
678
679     return Lagrangian

```

```

680 q = path_function([literal_function("x"), literal_function("y")])
681 l_eq = Lagrange_equations(L_uniform_acceleration(m, g))(q)
682 show(l_eq(t))

```

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

1.6.3 Central force field

```

683 def L_central_rectangular(m, U):
684     def Lagrangian(local):
685         q = coordinate(local)
686         v = velocity(local)
687         T = 1 / 2 * m * square(v)
688         return T - U(sqrt(square(q)))
689
690     return Lagrangian

```

Let us first try this on a concrete potential function.

```

691 def U(r):
692     return 1 / r

```

```

693 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.

```

694 U = Function(lambda x: function("U")(x))
695 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.

```

696 def F_to_C(F):
697     def f(local):
698         return up(
699             time(local),
700             F(local),
701             partial(F, 0)(local) + partial(F, 1)(local) * velocity(local),
702         )
703
704     return f

```

1.6.5 polar coordinates

```

705 def p_to_r(local):
706     r, phi = coordinate(local).list()
707     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 .

```

708 r = literal_function("r")
709 phi = literal_function("phi")
710 q = path_function([r, phi])
711 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.

```
../sage/section1.6.sage
712 show((partial(p_to_r, 0)(Gamma(q)(t))))
```

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

Next is the derivative wrt r and ϕ .

```
../sage/section1.6.sage
713 show((partial(p_to_r, 1)(Gamma(q)(t))))
```

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

```
../sage/section1.6.sage
714 show(F_to_C(p_to_r)(Gamma(q)(t)))
```

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

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

```
../sage/section1.6.sage
715 def L_central_polar(m, U):
716     def Lagrangian(local):
717         return compose(L_central_rectangular(m, U), F_to_C(p_to_r))(local)
718
719     return Lagrangian
```

```
../sage/section1.6.sage
720 # show(L_central_polar(m, U)(Gamma(q)(t)))
721 show(L_central_polar(m, U)(Gamma(q)(t)).simplify_full())
```

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

```
../sage/section1.6.sage
722 expr = Lagrange_equations(L_central_polar(m, U))(q)(t)
723 show(expr.simplify_full().expand())
```

$$\begin{bmatrix} -m r \dot{\phi}^2 + m \ddot{r} + \frac{r D_0(U)(\sqrt{r^2})}{\sqrt{r^2}} & m r^2 \ddot{\phi} + 2 m r \dot{\phi} \dot{r} \end{bmatrix}$$

1.6.6 Coriolis and centrifugal forces

```

724 def L_free_rectangular(m):
725     def Lagrangian(local):
726         v = velocity(local)
727         return 1 / 2 * m * square(v)
728
729     return Lagrangian
730
731
732 def L_free_polar(m):
733     def Lagrangian(local):
734         return L_free_rectangular(m)(F_to_C(p_to_r)(local))
735
736     return Lagrangian
737
738
739 def F(Omega):
740     def f(local):
741         t = time(local)
742         r, theta = coordinate(local).list()
743         return vector([r, theta + Omega * t])
744
745     return f
746
747
748 def L_rotating_polar(m, Omega):
749     def Lagrangian(local):
750         return L_free_polar(m)(F_to_C(F(Omega))(local))
751
752     return Lagrangian
753
754
755
756 def r_to_p(local):
757     x, y = coordinate(local).list()
758     return column_matrix([sqrt(x * x + y * y), atan(y / x)])
759
760
761 def L_rotating_rectangular(m, Omega):
762     def Lagrangian(local):
763         return L_rotating_polar(m, Omega)(F_to_C(r_to_p)(local))
764
765     return Lagrangian

```

```

766 _ = var("Omega", domain="positive")
767 q_xy = path_function([literal_function("x"), literal_function("y")])
768 expr = L_rotating_rectangular(m, Omega)(Gamma(q_xy)(t)).simplify_full()

```

```

769 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.

```

770 expr = Lagrange_equations(L_rotating_rectangular(m, Omega))(q)(t)
771 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.

```

772 def dp_coordinates(l, ys):
773     "From theta to x, y coordinates."
774     def f(local):
775         t = time(local)
776         theta = coordinate(local)[0, 0]
777         return column_matrix([l * sin(theta), ys(t) - l * cos(theta)])
778
779     return f

```

```

780 def L_pend(m, l, g, ys):
781     def Lagrangian(local):
782         return L_uniform_acceleration(m, g)(
783             F_to_C(dp_coordinates(l, ys))(local)
784         )
785
786     return Lagrangian

```

```

787 _ = var("l", domain="positive")
788
789 theta = path_function([literal_function("theta")])
790 ys = literal_function("y")
791
792 expr = L_pend(m, l, g, ys)(Gamma(theta)(t)).simplify_full()
793 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

```

794 load("utils1.6.sage")

```

```

795 load("utils1.7.sage")
796
797 var("t", domain=RR)

```

```

798 load("show_expression.sage")

```

1.7.2 Acceleration and state derivative

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

```

799 q = path_function([literal_function("x"), literal_function("y")])
800 local = Gamma(q)(t)
801 m, k = var("m k", domain="positive")
802 L = L_harmonic(m, k)
803 show(L(local))

```

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

```

804 F = compose(transpose, partial(L, 1))
805 show(F(local))
806 P = partial(L, 2)
807 show((F - partial(P, 0))(local))

```

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

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

```

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

```

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

Convert to vector.

```

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

```

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

```

810 def Lagrangian_to_acceleration(L):
811     def f(local):
812         P = partial(L, 2)
813         F = compose(transpose, partial(L, 1))
814         M = (F - partial(P, 0)) - partial(P, 1) * velocity
815         return partial(P, 2)(local).solve_right(M(local))
816
817     return f

```

We apply this to the harmonic oscillator.

```

818 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:

don't tangle

```

819 def de_rhs(x, y):
820     return [y, -x]
821
822
823 sol = desolve_odeint(de_rhs(x, y), [1, 0], srange(0, 100, 0.05), [x, y])
824 pp = list(zip(sol[:, 0], sol[:, 1]))
825 p = points(pp, color='blue', size=3)
826 p.save(f'circle.png')

```

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

don't tangle

```

827 def de_rhs(x, y):
828     return [1, -1]

```

The solution is to replace the numbers by expressions.

../sage/utlils1.7.sage

```

829 def convert_to_expr(n):
830     return SR(n)

```

And then define the function of differentials like this.

don't tangle

```

831 def de_rhs(x, y):
832     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)`.

../sage/utlils1.7.sage

```

833 def Lagrangian_to_state_derivative(L):
834     acceleration = Lagrangian_to_acceleration(L)
835     return lambda state: up(
836         convert_to_expr(1), velocity(state), acceleration(state)
837     )

```

../sage/section1.7.sage

```

838 show(Lagrangian_to_state_derivative(L)(local))

```

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

```

839 def harmonic_state_derivative(m, k):
840     return Lagrangian_to_state_derivative(L_harmonic(m, k))

```

```

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

```

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

```

842 def qv_to_state_path(q, v):
843     return lambda t: up(t, q(t), v(t))

```

```

844 def Lagrange_equations_first_order(L):
845     def f(q, v):
846         state_path = qv_to_state_path(q, v)
847         res = D(state_path)
848         res -= compose(Lagrangian_to_state_derivative(L), state_path)
849         return res
850
851     return f

```

```

852 res = Lagrange_equations_first_order(L_harmonic(m, k))(
853     path_function([literal_function("x"), literal_function("y")]),
854     path_function([literal_function("v_x"), literal_function("v_y")]),
855 )
856 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.

```

857 def make_dummy_vector(name, dim):
858     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.

```

859 def evolve(state_derivative, ics, times):
860     dim = coordinate(ics).nrows()
861     coordinates = make_dummy_vector("q", dim)
862     velocities = make_dummy_vector("v", dim)
863     space = up(t, coordinates, velocities)
864     soln = desolve_odeint(
865         des=state_derivative(space).list(),
866         ics=ics.list(),
867         times=times,
868         dvars=space.list(),
869         atol=1e-13,
870     )
871     return soln

```

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

```

872 def state_advancer(state_derivative, ics, T):
873     init_time = time(ics)
874     times = [init_time, init_time + T]
875     soln = evolve(state_derivative, ics, times)
876     return soln[-1]

```

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

```

877 state_advancer(
878     harmonic_state_derivative(m=2, k=1),
879     ics=up(0, column_matrix([1, 2]), column_matrix([3, 4])),
880     T=10,
881 )

```

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.

```

882 def periodic_drive(amplitude, frequency, phase):
883     def f(t):
884         return amplitude * cos(frequency * t + phase)
885
886     return f

```

With this we make the Lagrangian.

```

887 _ = var("m l g A omega")
888
889
890 def L_periodically_driven_pendulum(m, l, g, A, omega):
891     ys = periodic_drive(A, omega, 0)
892
893     def Lagrangian(local):
894         return L_pend(m, l, g, ys)(local)
895
896     return Lagrangian

```

```

897 q = path_function([literal_function("theta")])
898 show(
899     L_periodically_driven_pendulum(m, l, g, A, omega)(
900         Gamma(q)(t)
901     ).simplify_full()
902 )

```

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

```

903 expr = Lagrange_equations(L_periodically_driven_pendulum(m, l, g, A, omega))(
904     q
905 )(t).simplify_full()
906 show(expr)

```

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

```

907 show(
908     Lagrangian_to_acceleration(
909         L_periodically_driven_pendulum(m, l, g, A, omega)
910     )(Gamma(q)(t)).simplify_full()
911 )

```

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

```

912 def pend_state_derivative(m, l, g, A, omega):
913     return Lagrangian_to_state_derivative(
914         L_periodically_driven_pendulum(m, l, g, A, omega)
915     )

```

```

916 expr = pend_state_derivative(m, l, g, A, omega)(Gamma(q)(t))
917 show(time(expr))
918 show(coordinate(expr).simplify_full())
919 show(velocity(expr).simplify_full())

```

$$1$$

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

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

```

920 def principal_value(cut_point):
921     def f(x):
922         return (x + cut_point) % (2 * np.pi) - cut_point
923
924     return f

```

```

925 def plot_driven_pendulum(A, T, step_size=0.01):
926     times = srange(0, T, step_size, include_endpoint=True)
927     soln = evolve(
928         pend_state_derivative(m=1, l=1, g=9.8, A=A, omega=2 * sqrt(9.8)),
929         ics=up(0, column_matrix([1]), column_matrix([0])),
930         times=times,
931     )
932     thetas = soln[:, 1]
933     pp = list(zip(times, thetas))
934     p = points(pp, color='blue', size=3)
935     p.save(f'../figures/driven_pendulum_{A:.2f}.png')
936
937     thetas = principal_value(np.pi)(thetas)
938     pp = list(zip(times, thetas))
939     p = points(pp, color='blue', size=3)
940     p.save(f'../figures/driven_pendulum_{A:.2f}_principal_value.png')
941
942     thetadots = soln[:, 2]
943     pp = list(zip(thetas, thetadots))
944     p = points(pp, color='blue', size=3)
945     p.save(f'../figures/driven_pendulum_{A:.2f}_trajectory.png')
946

```

So now we make the plot.

```

947  ..../sage/section1.7.sage
plot_driven_pendulum(A=0.1, T=100, step_size=0.005)

```

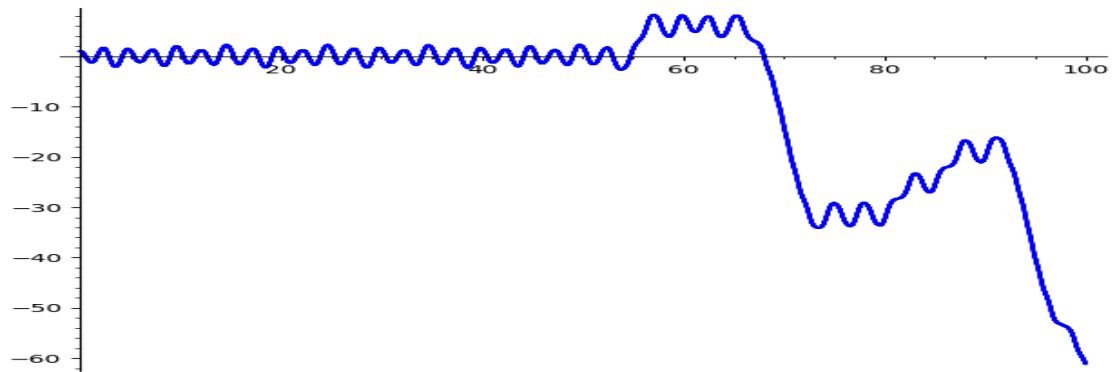


Figure 1.2: The angle of the vertically driven pendulum as a function of time. Obviously, around $t = 80$, the pendulum makes a few revolutions, and then starts to wobble again.

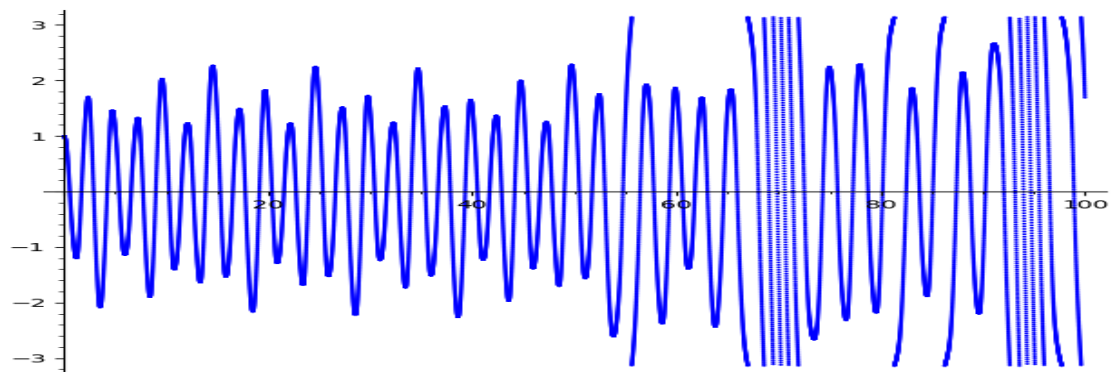


Figure 1.3: The angle on $(-\pi, \pi]$.

1.8 CONSERVED QUANTITIES

1.8.1 *Standard imports*

```

948  ..../sage/utis1.8.sage
load("utis1.6.sage")

```

```

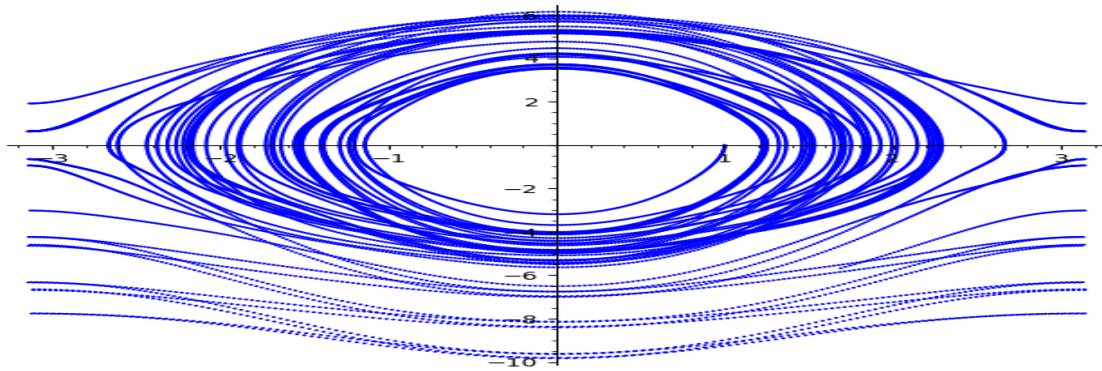
949  ..../sage/section1.8.sage
load("utis1.8.sage")
950
951  var("t", domain=RR)

```

```

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

```

Figure 1.4: The trajectory of θ and $\dot{\theta}$.

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.

```

953 def Lagrangian_to_energy(L):
954     P = partial(L, 2)
955     LL = Function(lambda local: L(local))
956     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.

```

957 q = path_function(
958     [
959         literal_function("r"),
960         literal_function("theta"),
961         literal_function("phi"),
962     ]
963 )

```

Next the transformation from spherical to 3D rectangular coordinates.

```

964 def s_to_r(sperical_state):
965     r, theta, phi = coordinate(sperical_state).list()

```

```

966     return vector(
967         [r * sin(theta) * cos(phi), r * sin(theta) * sin(phi), r * cos(theta)]
968     )

```

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

```

          ../sage/section1.8.sage
969 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.

```

          ../sage/section1.8.sage
970 V = Function(lambda r: function("V")(r))
971
972 def L_3D_central(m, V):
973     def Lagrangian(local):
974         return L_central_rectangular(m, V)(F_to_C(s_to_r)(local))
975
976     return Lagrangian

```

```

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

```

$$\left[-\frac{r D_0(V)(\sqrt{r^2}) - (mr \sin(\theta)^2 \dot{\phi}^2 + mr \dot{\theta}^2) \sqrt{r^2}}{\sqrt{r^2}} \quad m \cos(\theta) r^2 \sin(\theta) \dot{\phi}^2 \quad 0 \right]$$

```

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

```

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

```

          ../sage/section1.8.sage
979 def ang_mom_z(m):
980     def f(rectangular_state):
981         xyx = vector(coordinate(rectangular_state))
982         v = vector(velocity(rectangular_state))
983         return xyx.cross_product(m * v)[2]
984
985     return f
986
987
988 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 $E = T + V$.

```

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

```

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

1.8.4 The Restricted Three-Body Problem

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

```

990  ..../sage/section1.8.sage
991  var("G M0 M1 a", domain="positive")
992
993  def distance(x, y):
994      return sqrt(square(x - y))
995
996
997  def angular_freq(M0, M1, a):
998      return sqrt(G * (M0 + M1) / a ^ 3)
999
1000
1001  def V(a, M0, M1, m):
1002      Omega = angular_freq(M0, M1, a)
1003      a0, a1 = M1 / (M0 + M1) * a, M0 / (M0 + M1) * a
1004
1005      def f(t, origin):
1006          pos0 = -a0 * column_matrix([cos(Omega * t), sin(Omega * t)])
1007          pos1 = a1 * column_matrix([cos(Omega * t), sin(Omega * t)])
1008          r0 = distance(origin, pos0)
1009          r1 = distance(origin, pos1)
1010          return -G * m * (M0 / r0 + M1 / r1)
1011
1012      return f
1013
1014  def L0(m, V):
1015      def f(local):
1016          t, q, v = time(local), coordinate(local), velocity(local)
1017          return 1 / 2 * m * square(v) - V(t, q)
1018
1019      return f

```

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

```

1020 q = path_function([literal_function("x"), literal_function("y")])
1021 expr = (sqrt(G*M0 + G*M1)*t) / a^(3/2)
1022 A = var('A')
1023
1024 show(
1025     Lagrangian_to_energy(L0(m, V(a, M0, M1, m)))(Gamma(q)(t))
1026     .simplify_full()
1027     .expand()
1028     .subs({expr: A})
1029 )

```

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

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.

```

1030 def rotation_matrix(axis, theta):
1031     """
1032     Return the 3x3 rotation matrix for a rotation of angle theta (in radians)
1033     about the given axis. The axis is specified as an iterable of 3 numbers.
1034     """
1035     # Convert the axis to a normalized vector
1036     axis = vector(axis).normalized()
1037     x, y, z = axis
1038     c = cos(theta)
1039     s = sin(theta)
1040     t = 1 - c # common factor
1041
1042     # Construct the rotation matrix using Rodrigues' formula
1043     R = matrix(
1044         [
1045             [c + x**2 * t, x * y * t - z * s, x * z * t + y * s],
1046             [y * x * t + z * s, c + y**2 * t, y * z * t - x * s],
1047             [z * x * t - y * s, z * y * t + x * s, c + z**2 * t],
1048         ]
1049     )
1050     return R

```

```

1051 def F_tilde(angle_x, angle_y, angle_z):
1052     def f(local):

```

```

1053     return (
1054         rotation_matrix([1, 0, 0], angle_x)
1055         * rotation_matrix([0, 1, 0], angle_y)
1056         * rotation_matrix([0, 0, 1], angle_z)
1057         * coordinate(local)
1058     )
1059
1060     return f

```

```

1061 q = path_function(
1062     [literal_function("x"), literal_function("y"), literal_function("z")]
1063 )

```

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

```

1064 def Rx(s):
1065     return lambda local: F_tilde(s, 0, 0)(local)
1066
1067
1068 s, u, v = var("s u v")
1069 latex.matrix_delimiters(left='[', right=']')
1070 latex.matrix_column_alignment("c")
1071 show(Rx(s)(Gamma(q)(t)))
1072 show(diff(Rx(s)(Gamma(q)(t)), s)(s=0))

```

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

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

And now we check the result of the book. The computation of $D F_{\text{tilde}}$ 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 F_{tilde} (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 F_{tilde} . Therefore we unpack the argument x of the lambda function to convert the list $[s, u, v]$ into three separate arguments. The last step is to fill in $s = u = v = 0$.

Note that we differentiate wrt s, u, v and not wrt t . In itself, using t would not be a problem, but since we pass $\text{Gamma}(q)(t)$ to 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.

```

1073 U = Function(lambda r: function("U")(r))
1074
1075
1076 def the_Noether_integral(local):
1077     L = L_central_rectangular(m, U)
1078     Ftilde = lambda x: F_tilde(*x)(local)
1079     DF0 = Jacobian(Ftilde)([s, u, v], [s, u, v])(s=0, u=0, v=0)
1080     return partial(L, 2)(local) * DF0

```

```

1081 show(the_Noether_integral(Gamma(q)(t)).simplify_full())

```

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

CONTENTS

1.9 ABSTRACTION OF PATH FUNCTIONS

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

1.9.1 *Standard imports*

```
_____ ../sage/utils1.9.sage _____  
1082 load("utils1.6.sage")  
_____  
_____ ../sage/section1.9.sage _____  
1083 load("utils1.9.sage")  
1084  
1085 var("t", domain=RR)  
_____  
_____ don't tangle _____  
1086 load("show_expression.sage")  
_____
```

1.9.2 *Understanding F_{to_C}*

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

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

```
_____ ../sage/section1.9.sage _____  
1087 r, theta = literal_function("r"), literal_function("theta")  
1088 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 _____  
1089 show((r(t), theta(t)), simplify=False)  
_____
```

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

A `path_function` 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.)

```

1090 q_prime = path_function([r, theta])
1091 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 :

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

```

1092 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.

```

1093 show(Gamma(q_prime)(t), simplify=False)

```

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

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 .

```

1094 F = p_to_r
1095 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.

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

```

1096 Q = lambda t: compose(p_to_r, Gamma(q_prime))(t)
1097 show(Gamma(Q)(t), simplify=False)

```

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

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

```

1098 def f_bar(q_prime):
1099     q = lambda t: compose(F, Gamma(q_prime))(t)
1100     return lambda t: Gamma(q)(t)

```

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

```

1101 show(f_bar(q_prime)(t))

```

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

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

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

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

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

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

```

1102 def osculating_path(local):
1103     t = time(local)
1104     q = coordinate(local)
1105
1106     def wrapper(t_prime):
1107         res = q
1108         dt = 1
1109         factorial = 1
1110         for k in range(2, len(local)):
1111             factorial *= k
1112             dt *= t_prime - t
1113             res += local[k] * dt / factorial
1114         return res
1115
1116     return wrapper

```

Here is an example.

```

1117 t_prime = var("tt", domain="positive", latex_name="t'")
1118 q = path_function([literal_function("r"), literal_function("theta")])
1119 local = Gamma(q)(t)
1120 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`.

```

1121 def Gamma_bar(f_bar):
1122     def wrapped(local):
1123         t = time(local)
1124         q_prime = osculating_path(local)
1125         return f_bar(q_prime)(t)
1126
1127     return wrapped

```

```

1128 show(Gamma_bar(f_bar)(local))

```

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

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

```

1129 def F_to_C(F):
1130     def C(local):
1131         n = len(local)
1132
1133         def f_bar(q_prime):
1134             q = lambda t: compose(F, Gamma(q_prime))(t)
1135             return lambda t: Gamma(q, n)(t)
1136
1137         return Gamma_bar(f_bar)(local)
1138
1139     return C

```

```

1140 show(F_to_C(p_to_r)(local))

```

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

Here is the total time derivative.

```

1141 @Func
1142 def Dt(F):
1143     def DtF(local):
1144         n = len(local)
1145
1146         def DF_on_path(q):
1147             return D(lambda t: F(Gamma(q, n - 1)(t)))
1148
1149         return Gamma_bar(DF_on_path)(local)
1150
1151     return lambda state: DtF(local)

```

1.9.3 Lagrange equations at a moment

```

1152 def Euler_Lagrange_operator(L):
1153     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.

```

1154 q = path_function([literal_function("x")])
1155 local = Gamma(q, 4)(t)
1156 show(local)

```

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

```

1157 m, k = var("m k", domain="positive")
1158 L = L_harmonic(m, k)
1159 show(Euler_Lagrange_operator(L)(local))

```

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