

BRYN MAWR COLLEGE
Department of Physics

Computational Module 0 – Introduction to Computational Methods for the Sciences

Prerequisite modules: None

Estimated completion time: ½ - 1 hour

Learning goals: Become familiar with basic issues in computing and key properties of computing languages



Image credit: xkcd.com

See http://www.explainxkcd.com/wiki/index.php/979:_Wisdom_of_the_Ancients for an explanation.

0.0 Computational Modules

This document is the first in a series of *computational modules* designed to allow an individual to build computational skills by working through the example problems in the modules, answering questions posed in the modules, and completing exercises at the end of each module. Some modules, including this one, may require that you consult other resources for additional information. Modules that should be worked through prior to the one being read will be indicated under *Prerequisite modules* at the top of the first page of the module. (**NOTE:** This module, and Module 1 – **A Brief Introduction to Python & Programming**, are prerequisites for all subsequent modules and may not be listed explicitly in all modules as prerequisites.) Each module also indicates an estimated time range for reading the module and completing the exercises. If you have used Python before or have other computing experience, the lower end of the range may apply to you; if you have not had such experience, the upper end may be more appropriate.

In later modules, in addition to Exercises you also will find **Breakpoint** questions sprinkled throughout. In debugging environments for programming languages, breakpoints are markers that can be put into code to pause the execution of a program, enabling the programmer to inspect the values of variables and otherwise confirm that the program is doing what it is supposed to be doing at that point. The module *Breakpoint* questions are meant to serve a similar purpose: usually, they get you to pause in your reading to think about what you've read, in order to confirm that your brain is doing what it should be with the material at that point. The answers to the *Breakpoint* questions are provided at the end of each module.

0.1 Why Learn Computer Programming?

For hundreds of years, progress was made in physics, chemistry, geology, astronomy, etc. without the use of computers. However, as those fields evolved, the problems tackled by scientists grew more complicated and more challenging to solve, due either to the complexity of the systems involved or to the vast amounts of data that needed to be analyzed. The advent of electronic computers made it feasible to attempt to solve some of these problems. Nowadays, the computer is indispensable for applications ranging from genomics (e.g., the Human Genome Project), to computational chemistry (used, for example, to predict the molecular structures of potential new pharmaceutical drugs), to climate modeling (needed to assess the impacts of climate change), to the search for planets beyond our solar system, to modeling the formation of large-scale structure during the evolution of the universe.

Besides enabling scientists to address the very important applications listed above (and countless others!), learning computer programming is very valuable for another reason: it teaches the programmer to think in an algorithmic way. (Until recently, computers could “think” *only* in that way. With the advent of large language models like ChatGPT, computers arguably now can think more like humans do.) By *algorithmic thinking* we simply mean thinking according to an ordered, logical sequence of steps: a recipe is a good everyday example of an algorithm. Algorithmic thinking arguably is the best way to solve most scientific problems (apart, perhaps, from those situations in which intuition might be needed), so becoming more proficient at such thinking can help one become a more effective scientist. Developing your algorithmic thinking skills is, in fact, the primary purpose of these modules. Learning the tools of programming is a secondary (but very important) focus.

0.2 Modern Computer Capabilities

Computer speeds often are characterized in terms of *flops* (or **FLOPS**) – floating-point operations per second – where a floating-point operation might be the addition of two floating-point numbers (numbers requiring a decimal representation, in contrast to integers). These days, a typical single-core processor in a laptop or desktop computer typically can perform maybe 10 billion floating-point operations per second; most computers have several cores and speeds in the range of several tens of gigaflops, or 10^{10} flops. Supercomputers operate in the many-petaflop range, with 1 petaflop = 10^6 gigaflops, or 10^{15} flops. As of November 2024, the fastest computer in the world was the El Capitan supercomputer, which reached 1.742 exaflops (10^{18} flops) using more than 11 *million* cores (<https://www.top500.org/>). Computing power in the 100-10000 teraflop (10^{14} - 10^{16} flops) range makes physically realistic simulations and analyses of structures and processes possible in reasonable time frames.

0.3 Programming Concepts

Computers are instructed to do what we want them to do using **computer code**, written in a **programming language**. The very earliest programmers used **machine code** to program their computers: it can be directly understood and executed by a computer's hardware, but is very difficult to program in. Modern “high level” programming languages readable by humans generally fall into two categories (although the distinctions between them are not precise). **Compiled languages** – e.g., some versions of Basic, C/C++, Fortran, and Java in some forms – are converted by **compilers** (other software) into machine code *before the user's software is run*. These languages tend to produce code that runs very fast – a boon to a programmer doing complicated calculations with large data sets. Their disadvantage is that one must “compile” the code before being able to run it. **Interpreted languages** – including Mathematica, MATLAB, and some other versions of Basic – use interpreters (other software) to translate the code into machine-executable form *when the code is run*: no pre-compilation is required. This approach allows the programmer to see the results of their program sooner but leads to code that may not run as fast as compiled code. (As already noted, the distinctions between these two types of languages are not clear-cut, and there are other ways of coding that fall in between these two categories.) Python – the language used in these modules – employs a hybrid approach, so it is easy to see results quickly, but the code also runs quickly.

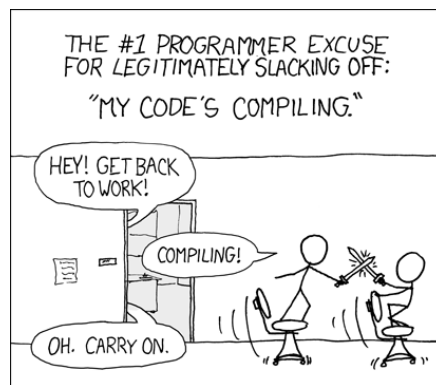


Image credit: xkcd.com

In addition to the compiled/interpreted language distinction, another important distinction associated with modern computing languages is the one between numeric and symbolic computing. As the name suggests, numeric computing involves the processing of actual numbers, e.g., the numerical derivative of $\sin(x)$ would involve computing the differences between its values for nearby points x and $x + \Delta x$, with x having particular values. Some modern languages also can perform symbolic computations, e.g. the symbolic derivative of $\sin(x)$ would be reported as $\cos(x)$, with no numbers involved. Python has a set of functions that one can call on (called “sympy,” for *symbolic Python*) that allow for symbolic computation.

Another feature possessed by some programming languages is their ability to be used for **object-oriented programming**. This programming style, which has become very common in the last few decades, enables the programmer to create reusable code modules that are easy to modify for new applications. Corporate programmers have tended to utilize this approach more than scientists have, but its use in the scientific realm is growing.

It's important to keep in mind that when running programs, computers are completely literal, no matter the programming language used. They operate based on precisely what you tell them, not on what they think you mean. If your code has an error in it, whether small or large, a computer will not understand and fix it. If you're lucky, it will produce an error message; if you're not, it simply will not do what you want it to, and you might not realize that. Therefore, in cases when incorrect output is not obvious, it's critical to test your code. A good way is to run it on data for which you know what the output should be. Don't assume the output is right just because you get a result!

The interested reader is encouraged to consult the relevant Wikipedia pages for more information on programming languages.

0.4 Starting with Python

For these modules, we will use Python as the main programming language. Its advantages are numerous: it is open-source (free) software, its syntax is easy to learn, it is an interpreted language that allows the user to see the code's output easily, it has extensions that enable one to do sophisticated interactive numerical analysis and visualization (similar to MATLAB and Mathematica), it is a widely-used language (including within the scientific/engineering community), and it is object-oriented. Moreover, a variant known as IPython has been developed which can be run in a browser-based format called a *notebook*. These notebooks can incorporate not only code, but also text, LaTeX, graphics, videos, and more.

These modules are not meant to teach you the full Python language. Module 1 presents a brief introduction to Python, but it won't provide all the information useful for the modules. Appendix A of this module lists additional resources for learning Python, and computer programming in general. A list of the basic topics you'll need to know about is presented in Appendix B. For instructions on how to download and install tools for running Python on a computer, see Appendix C. The remainder of this section will describe some aspects of programming, and the Python language, that you might not initially pick up in your study, but which will be important for these modules.

Currently, there are several versions of Python in common use: version 2.7, and versions 3.X, where "X" is a number from 1 to 13 (as of Jan. 2025). The 3.X versions differ from 2.7 (and earlier 2.X versions) in several ways, three of which are relevant to us: in the syntax of the `print` statement, which displays text or numeric output onscreen; in their treatment of *integer division* – the division of one integer by another; and in the behavior of the `input` statement, which receives input from a user. (Note: throughout the modules, Python code will be presented in Courier font, as in the word `print` shown above.) Python versions 2.X round the result of an integer division to an integer; for example, they would give the unexpected result $1/2 = 0$. Versions 3.X produce the expected result: $1/2 = 0.5$. In version 2.7 (and earlier), the syntax of the `print` statement was `print "Hello"`, while versions 3.X surround the content to be printed with parentheses: `print ("Hello")`. In the 2.X versions, the `input` statement retains the form of the user's input: if it's a number, `input` stores a number; if it's text, `input` stores that text. In the 3.X versions, the `input` statement converts all input to text, for consistency. *For these modules, we will assume use of version 3.X*, but beware that some Python packages may not have every associated extension fully implemented.

Here's the prototypical example of a first program – the traditional starting point when learning any programming language:

```
print("Hello World!")
```

Note that, by itself, this is a complete (if uninteresting) Python program! Nothing is required before the `print` statement to make it run. In some languages, every program, even one this simple, requires some prior instructions to the computer. Already, we see here some of the simplicity of Python.

Another way in which Python is easy to learn is that it treats variables very simply. In computer programming languages, *variables* can be one of several *types*, most commonly *integers*, *floating-point numbers*, *booleans*, and *characters/strings*. A string is just a set of characters, e.g. “time” or “x_init”; a floating-point number can be thought of simply as a real number; and a boolean is `True` or `False`. In some languages, such as C, each variable must have its type “declared” at the beginning of the program, before it is used (e.g., `int x` declares `x` as an integer); however, the Python language is not “statically typed,” which means that when you need a variable you simply define it, and Python assumes its type or figures out what kind of variable it is from the definition. For example, `x = 1` creates `x` as an integer, while `x = 1.0` or `x = float(1)` creates it as a floating-point number. (This could lead to a problem if the programmer does not keep in mind how a variable has been defined: remember that integer division works differently from normal division in 2.X versions!) Python also supports *complex numbers* as variables; they may appear in a later module.

Besides variables, other key programming structures are *operators* (multiplication, exponentiation, etc.) *loops* (including `for` and `while` loops; these allow a particular calculation to be repeated multiple times in sequence), *functions*, *lists*, and *branching statements* (usually called `if`, or `if-else` statements; some languages call these `switch` or `case` statements; they allow the code to follow different logical paths depending on the state of the code or on user input). Higher level programming languages typically also allow the programmer to include *comments*, which the computer ignores. In Python, these are identified by a hash sign, `#`: everything on a line following a hash is ignored. Multi-line comments are enclosed within triple sets of (single or double) quotes: `""" Here's nice text """`.

One other important thing to know about modern programming languages is that user-written programs often acquire additional capabilities by bringing in functions from pre-existing collections of programs. In Python, one speaks of this process as *importing packages*; in other languages, it's usually referred to as *linking to libraries* (a library being a collection of related functions, e.g., mathematical ones). Such sets of functions in the mathematical realm are designed for very efficient calculation. For example, performing more than simple mathematics in Python requires importing the `math` or `numpy` library.

Important libraries for our purposes are

- `math` (function list at <https://docs.python.org/2/library/math.html>)
- `numpy` (*numerical python*), with lots of mathematical functions in addition to those in `math` (see <http://docs.scipy.org/doc/numpy/reference/>)

- `sympy` (*symbolic python*), which has functions for doing symbolic mathematics (e.g., differentiating $\sin(x)$ to get $\cos(x)$; see <http://docs.sympy.org/latest/index.html>)
- `matplotlib`, which provides MATLAB-like plotting capability (see <http://matplotlib.org/py-modindex.html>)

With this very basic foundation established, the reader now is directed to read Module 1. It might also be helpful to work through one of the online tutorials listed in Appendix A, or consult a text, to gain some familiarity with Python and basic programming. (The book by Kinder & Nelson listed in the Appendix is a particularly good resource to supplement these modules.) An initial exposure to Python through these tutorials should include the programming topics listed in Appendix B. Appendix C describes how to install Python on your own computer and how to get started using those Python packages and IPython notebooks. Bryn Mawr & Haverford students might have access to an online Python platform; consult with your instructor or another physics faculty member.

Appendix A – Programming & Numerical Methods Resources

A.1 Programming Resources

There are many valuable resources that one can use to supplement these modules. A good general resource on all aspects of computing is stackoverflow.com. If you search online for an answer to a question – which is a good way to get an answer – this site is likely to come up at or near the top of the list. The site <https://www.python.org/> has lots of good material specifically on Python (and check out https://wiki.python.org/moin/BeginnersGuide/NonProgrammers#Graded_Introductory_Programming_Courses). Other useful resources for programming and Python include (as of this writing) those below. *In each category, the top one or two resources might be the best places to start.*

Tutorials:

<https://docs.python.org/2/tutorial/> – non-interactive.

www.codecademy.com – an extensive interactive introduction that will take a while to work through, but which is very suited to programming novices. Note that you choose Python as the language to learn at the bottom of the main page. (You do not have to create an account unless you want the website to keep track of your progress and remember where you leave off at the end of each session.)

http://www.python-course.eu/python3_course.php

www.learnpython.org – a concise but fairly comprehensive introduction to Python programming.

<http://interactivepython.org/runestone/static/thinkcspy/index.html> – an "interactive textbook."

<http://www.afterhoursprogramming.com/tutorial/Python/Introduction/> – an interactive tutorial

<http://software-carpentry.org/v5/novice/python/index.html> – Contrary to the word in the URL, this is not for novices, but might be a good way to test your skills after you have done some Python programming.

<http://quant-econ.net/py/index.html> – Part 1 is a nice summary of Python basics.

<http://www.tutorialspoint.com/python/> – a somewhat-interactive online tutorial that also provides a downloadable PDF tutorial. This site assumes that you have access to Python installed on a computer. It also has a nice online Quick Reference resource.

<http://learnpythonthehardway.org/book/> – non-interactive, but very user-friendly.

<http://www.cs.hmc.edu/csforall/> – non-interactive, but the basis for the introductory computer science course at Harvey Mudd College. (Section 2 in Chapter 1 is not useful for our purposes.)

Courses:

<http://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-189-a-gentle-introduction-to-programming-using-python-january-iap-2011/>

<https://www.coursera.org/course/pythonlearn> – requires (free) registration.

<https://www.udemy.com/introduction-to-computer-science-and-programming-python/?dtcode=cKFpSGu1kzK8>

Books (there are, of course, many more than those listed here):

A Student's Guide to Python for Physical Modeling: 2nd Edition, Jesse Kinder and Philip Nelson. An excellent brief survey of Python for scientific computing applications.

Think Python, Allen Downey, Jeffrey Elkin, and Chris Meyers, freely downloadable from this site:
<http://www.greenteapress.com/thinkpython/>

Python for Informatics, Charles Severance (freely downloadable from the coursera site above).

A.2 Numerical Methods Resources

Computational Physics, Mark Newman, CreateSpace Independent Publishing Platform, 2013. Uses Python. Easy to read, very clear explanations and derivations, good examples and choices of topic coverage. A great first choice of text.

Numerical Recipes, 3rd Ed., William Press, Saul Teukolsky, William Vetterling, Brian Flannery, Cambridge University Press, 2007. Uses C++. (The 2nd edition uses C, so its code examples may be easier to understand.) This is the classic text in the field, but it is more advanced than the text above. Explanations tend to be concise, but explicit code is given for many methods and the writing is engaging. Any important method not discussed in detail in the text is likely to have a reference cited.

A Survey of Computational Physics, Rubin Landau, Manuel José Páez, and Cristian Bordeianu, Princeton University Press, 2008. Uses Java. Introduces object-oriented programming early. Includes a number of topics and applications not in Newman's book.

Computational Physics, 2nd Ed., Nicholas Giordano and Hisao Nakanishi, Pearson Prentice Hall, 2006. Presents language-independent code in outline form only. Has lots of nice physics-related applications.

A Primer on Scientific Programming with Python, 2nd Ed., Hands Petter Langtangen, Springer, 2011. Clear writing, lots of code samples provided. Focuses more on programming than on scientific applications.

Appendix B – Basic Python topics to know

- Types of variables: integers, floating point numbers ("floats"), strings, booleans
- Basic mathematical operations: multiplication, division, addition, subtraction, power ($x**y$), modulo ($x\%y$ gives the remainder after dividing x by y an integer number of times; e.g., $16\%7 = 2$)
- Lists, arrays, and their differences: `[7, "howdy", 42.00]`, `[6, 12, 18, 24]`
- The simple print statement: `print("text")`

- The formatted print statement:

```
print("You typed the number %f %d times." %(num, times))
```

Here, `%f` is a placeholder indicating a floating-point number will be printed, `%d` indicates an integer to be printed. `%(num, times)` shows the values to be substituted for the `%f` and `%d`, in order.

- The input statement: `reply = input("Enter the value of x: ")`
- if and if-else statements:

```
if x > 100:
    print("x is pretty big")
```

```
if x > 100:
    print("x is pretty big")
else:
    print("x is not so big")
```

- The for loop:

```
for i in [1, 2, 3]:
    print(i+1)
```

- The while statement:

```
n = 0
while n < 4:
    print(n+1)
    n = n + 1
```

- Constructing simple functions:

```
def func():
    <function statements here...>
```

- Importing functions from modules (`from module import [function]` or `from module import *`) and importing entire modules (`import module`).

Appendix C – How to install Python on your computer or use it online

To install and run Python on your computer, you can use an open source product such as Anaconda, which includes many useful Python packages. You also can install Python and desired add-on packages directly.

To download and install Continuum's Anaconda, use this link:

<https://www.anaconda.com/download>

As an alternative to installing Anaconda, you can download Python itself from the following websites:

<https://www.python.org/download>

<https://www.python.org/download/windows/>

<https://www.python.org/download/mac>

If you install the Python package instead of Anaconda, you can edit and run Python scripts in IDLE, the Integrated DeveLopment Environment for Python and the default environment for most Python installations.

Using Anaconda

If you chose to use Anaconda, then after downloading and installing the package open the Launcher. In its window, you can choose to install the Spyder app – the Scientific PYthon DeveLopment EnviRonment, an *interactive development environment* (IDE) included with Anaconda – and/or the IPython notebook environment, which runs in your own browser.

To open Spyder, click on the Spyder icon or simply type `spyder` in the command prompt or terminal (opened using the **CMD.exe Prompt** launcher). In Spyder, you can open, view, and edit existing files, and create new files in the Editor (the left-hand panel) using the icons in the toolbar at the top of the window, or via the respective options in the File menu. You can execute the files by clicking the green triangular button in the toolbar or choosing the **Run** option in the **Run** menu. You can run a portion of a file by highlighting that portion and selecting **Run selection** under the **Run** menu. If you attempt to execute code that contains errors, a message will appear in the Console window at lower-right. Small orange triangles also will appear in the Editor window in front of the lines containing the errors; you can see descriptions of the errors if you hover your cursor over the triangles.

You can use the in-browser IPython notebook by launching the **jupyter notebook** environment from the Launcher. This will open a new tab in your browser; the page will present a list of the directories from which you can choose to run a pre-existing notebook, or you can start a new one using the **New** dropdown menu in the upper-right. More information on using notebooks is provided below.

Note: the packages incorporated in the Anaconda distribution are updated regularly, so you will want to check for updates periodically. To update Anaconda itself, open a command window, type `conda update conda`, and hit Enter. The computer will check for available packages and ask you to confirm their installation. To update the packages within Anaconda, type `conda update --all` (with two dashes in front of `all`) and hit Enter. Again, you will be asked to confirm any installations.

IPython

An IPython notebook provides a browser-based interactive computing environment. Besides being able to run executable code, the notebook also supports commentary text and other rich media such as LaTeX, HTML, and images.

The Anaconda distribution already includes IPython, so you can simply use the IPython notebook in Anaconda, although you might want to check <http://ipython.org/install.html> to make sure you have the latest version of IPython. If you have just basic Python installed, then go to the website below and follow the instructions to download and install the IPython notebook environment:

<http://ipython.org/ipython-doc/stable/install/install.html>

After installation, you can open and use an IPython notebook by typing `ipython notebook` in a terminal or command prompt window on your computer, or by using the appropriate launcher.

You also can run IPython on a remote server if you have been given an account on one. You connect to it via a browser, just as to any other web page, and all input and output is done through the browser.

Notebooks have an input/output format like that of Mathematica, with each input in a “cell.” Each cell is like a single file – it can include one or more lines of code, commentary text, mathematical equations in LaTeX format, etc. Cells can be modified both before and after execution.

The cell format provides the advantage of breaking down code into smaller sections for easier reading and debugging. Most if not all implementations will allow you to run either one cell at a time or all cells in order; you may even be able to run just a part of a cell. See the instructions or help system for your particular implementation.

To write commentary text in a notebook, select the **Markdown** option from the **Cell** menu. Simultaneously press the Shift and Enter keys to complete Markdown input (or to execute a code cell). Double-click a Markdown cell to enable modification or rewriting. You can also conveniently transform any cell between Markdown or Code formats by highlighting the cell and using the scrollable menu at the right end of the toolbar.

For more information on the topics discussed in this Appendix, see:

<http://ipython.org/install.html>

<https://docs.anaconda.com/>

<http://code.google.com/p/spyderlib/wiki/Features>

<http://ipython.org/ipython-doc/dev/notebook/>