



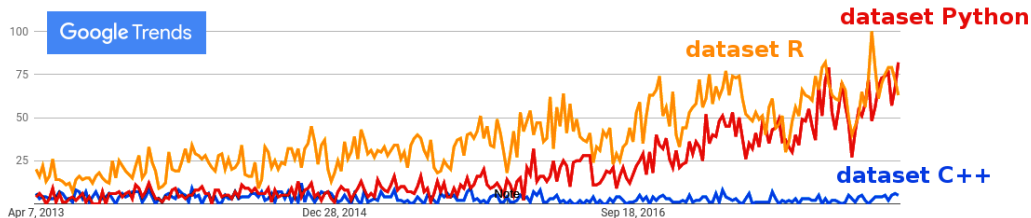
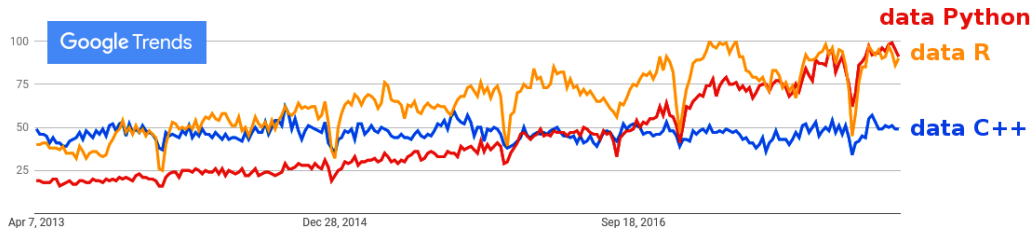
Scientific Python Ecosystem HATS

Jim Pivarski

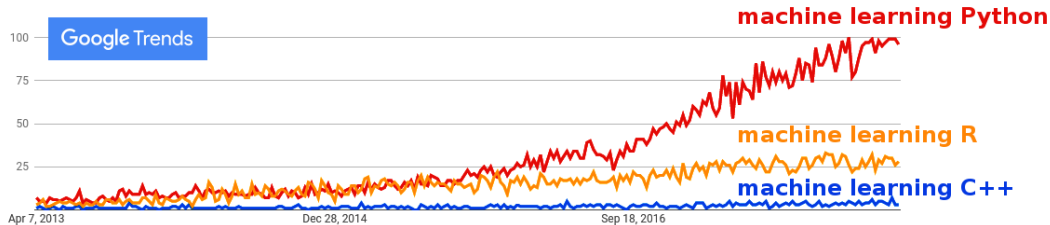
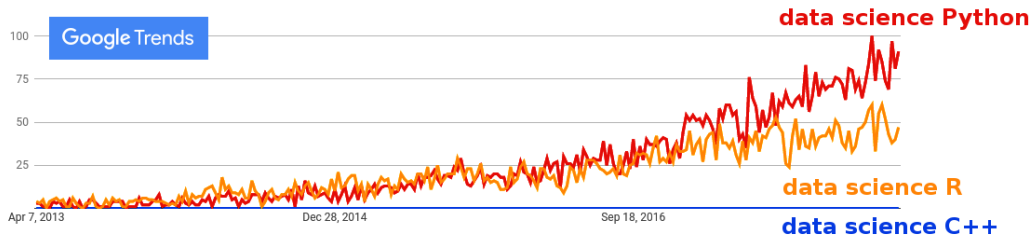
Princeton University – DIANA-HEP

June 4, 2018

A lot of data analysis is being in Python these days



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The “scientific Python ecosystem”



The word “ecosystem” is deliberately vague: no clear boundaries, but a few key components and many projects that share conventions.

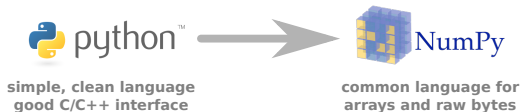


simple, clean language
good C/C++ interface

The “scientific Python ecosystem”



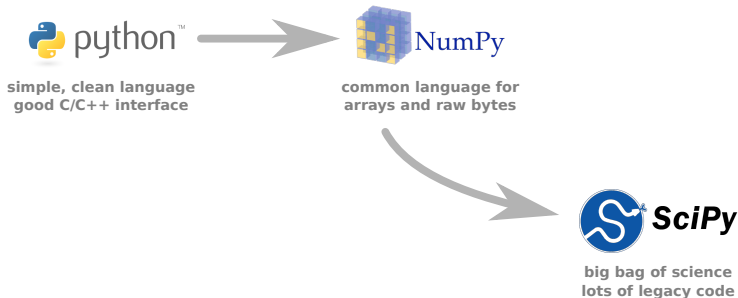
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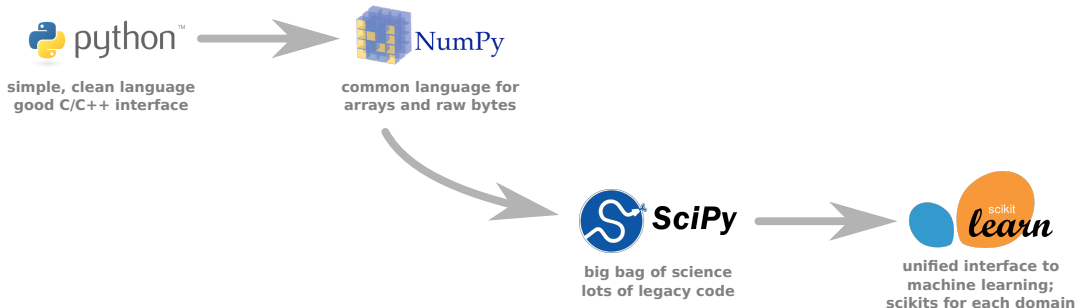
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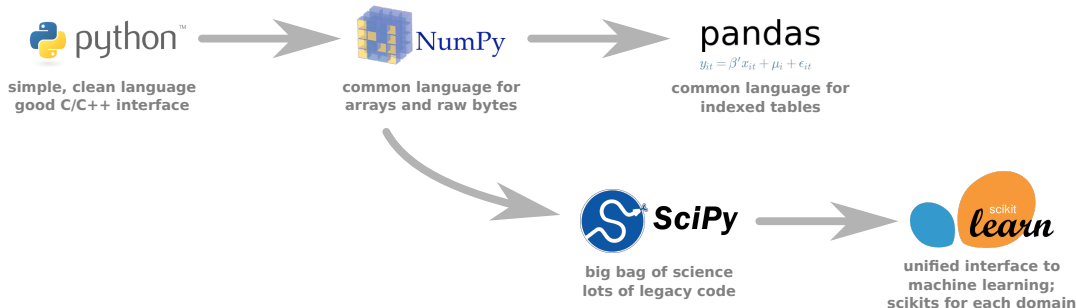
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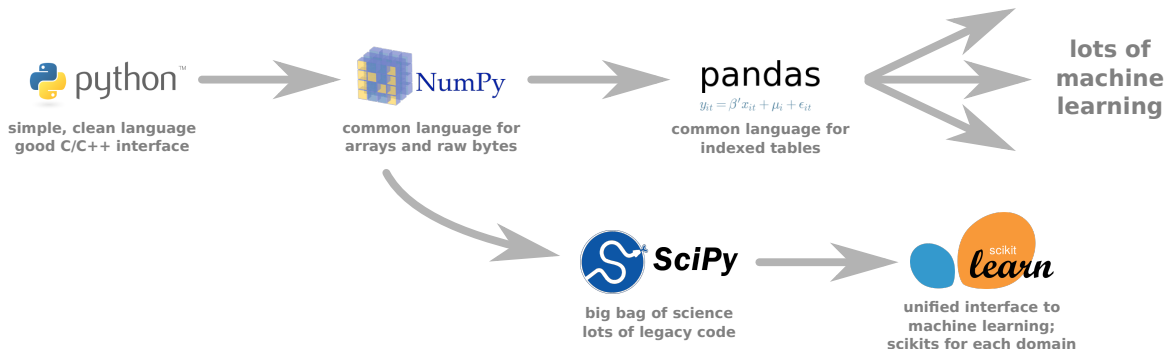
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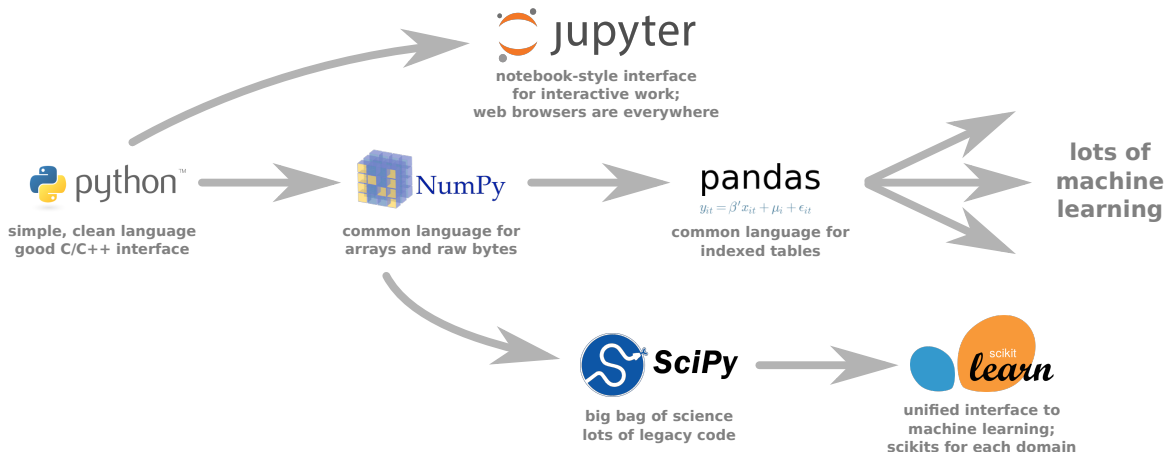
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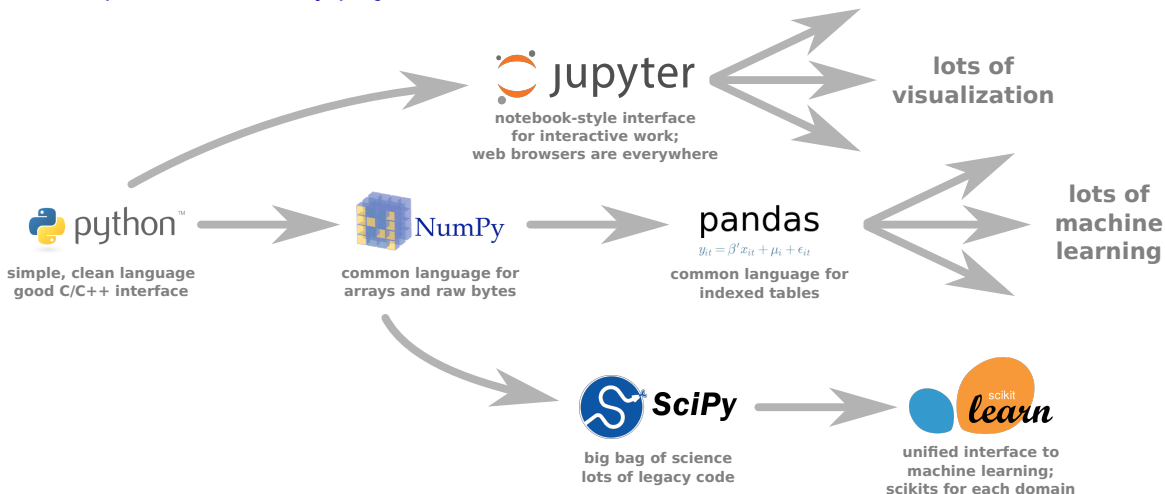
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The “scientific Python ecosystem”



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Some of what we need is readily available in this ecosystem



What we can find off-the-shelf

- fast calculations
- fast data sharing with C/C++
- libraries of special functions, matrix math
- fitting/minimization, integration, differentiation, interpolation
- symbolic algebra
- advanced statistics
- machine learning
- graphics, advanced plotting
- graphical interfaces, user workflows
- parallel and distributed processing

What we must develop in-house

- reading/writing ROOT files
- integration with collaboration frameworks
- advanced histogramming
- efficient variable-length lists (e.g. number of electrons per event)
- HEP functions (e.g. boost 4-vectors)
- HEP-style limit setting



1. **Numpy**: common language for bulk data exchange. Introduction to vectorization and fancy indexing.
2. **uproot***: implementation of ROOT I/O in Python and Numpy. Quick way to get your data into Numpy and everything else. Includes PyTorch example.
3. **histbook***: HEP-style histogramming in Numpy, Pandas, and Vega (web graphics).
4. **Numba and OAMap***: Numba is a just-in-time Python compiler for number crunching. OAMap (“object-array mapping”) extends it for HEP-style data.
5. **and more**: SymPy, iminuit, NumPythia, pyjet, PyPDT...
6. **LIGO**: discover a black hole with Numpy and Pandas.

***my own projects**