Performance Pandas

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StrataNYC 2015

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https://github.com/jreback/StrataNYC2015/tree/master/performance

Jeff Reback

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- former quant
- currently working on projects at Continuum
- core commiter to pandas for last 3 years
- manage pandas since 2013

Why Pandas?

- Vectorization for the masses
- ETL
- Fast and Efficient DataFrame
- Interoperability with ecosystem
- Database-like
- User friendly API
- Munging & data prep is a big part of the pipeline

http://pandas.pydata.org/

What do we care about when writing code?

Objectives

- feature set
- readability counts
- maintenance is a virtue
- tests & docs

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Objectives

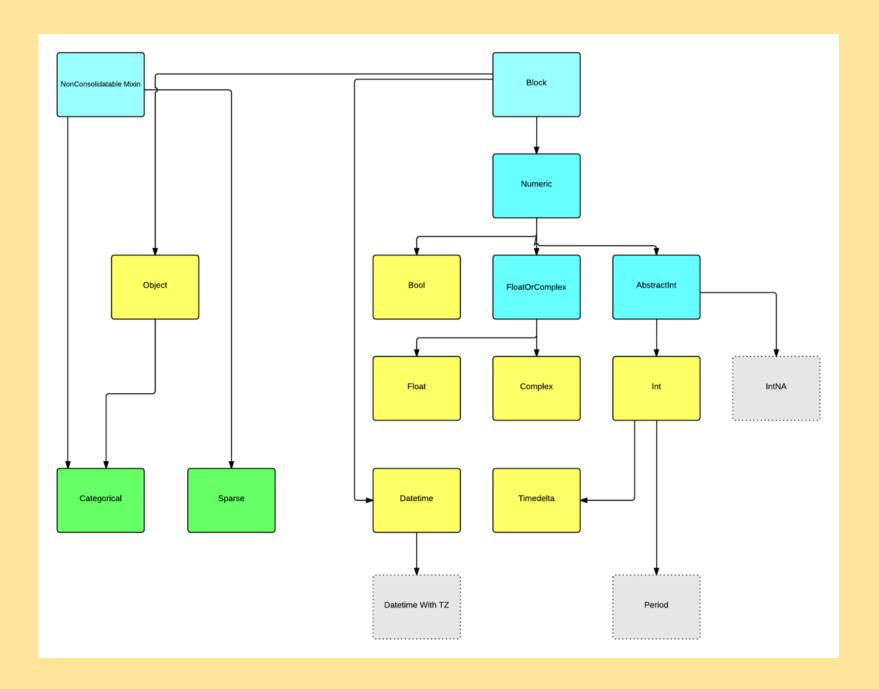
- feature set
- readability counts
- maintenance is a virtue
- tests & docs

Constraints

- implementation time
- runtime
- resource utilization

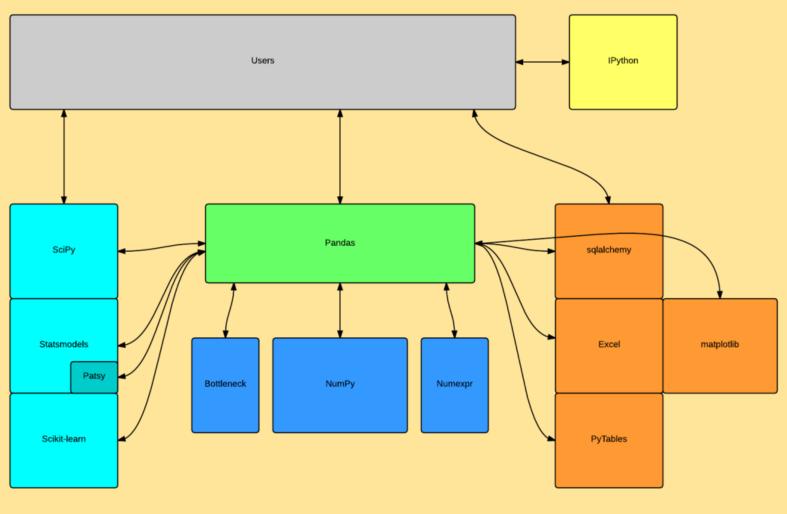
What drives pandas?

- dtype segregation
- block memory layout



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- dtype segregation
- block memory layout
- computation backends



Compute Stack

IO Stack

Graphics Stack

Computation Backends

- numpy
- bottleneck
- numexpr
- DyND?

http://slides.com/jeffreback/ds4ds-pandas#/

What drives pandas?

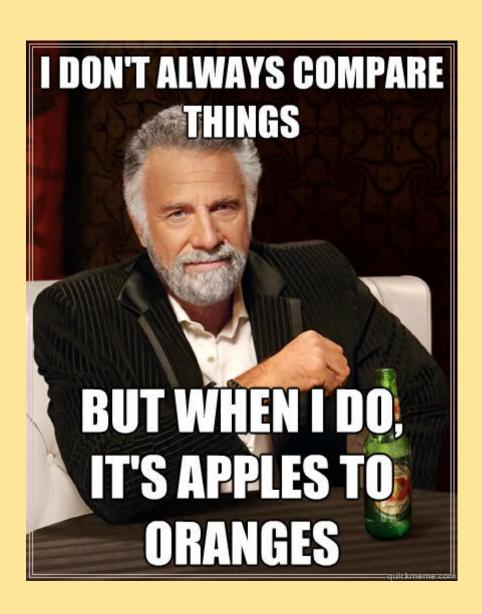
- dtype segregation
- block memory layout
- computation backends
- cython for critical parts
- hashtable for indexing

how to make pandas perform

- 1. Have Correct Code
- 2. Profile / Compare

how to make pandas perform

- 1. Have Correct Code
- 2. Profile / Compare
- 3. Refer to Rules #1 and #2



Programmers waste enormous amounts of time thinking about, or worrying about, the speed of noncritical parts of their programs, and these attempts at efficiency actually have a strong negative impact when debugging and maintenance are considered.

" premature optimization is the root of all evil (or at least most of it) in programming.

How to make pandas *fast*

- algo
- idioms
- built-in / vectorization
 - pandas/numpy
 - bottleneck/numexpr
 - cython
- ad-hoc cython/numba

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How to make pandas fast slow

apply across the rows

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- apply across the rows
- itertuples/iterrows

How to make pandas fast slow

- apply across the rows
- itertuples/iterrows
- iterative updating

.values, a double edged sword



Do's

- have the correct dtypes
- pd.concat
- Categoricals
- Use idioms & builtin
- .apply across columns

Don'ts

- repeated insertions
- micro optimize
- use loops / re-invent the wheel
- .apply across rows
- .applymap
- nest groupby.apply()
- inplace=True

Memory Considerations

conversions

Memory Considerations

- conversions
- categoricals

Memory Considerations

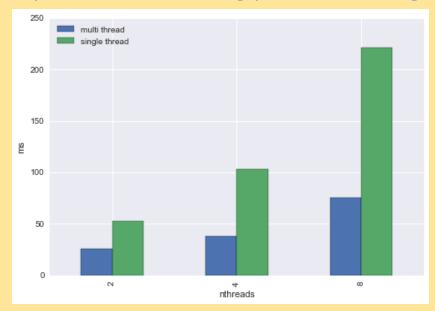
- conversions
- categoricals
- iterators

I/O & Serialization

- HDF5
- bcolz
- CSV
- SQL
- JSON
- pickle
- msgpack

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 - http://continuum.io/blog/pandas-releasing-the-gil

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- out-of-core
- dask
 - threading
 - multi-process
 - distributed

How to contribute

https://github.com/pydata/pandas/issues

This Talk

https://github.com/jreback/StrataNYC2015/tree/master/performance

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