## Performance Pandas

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StrataNYC 2015

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

### Jeff Reback

@jreback

- 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
- 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

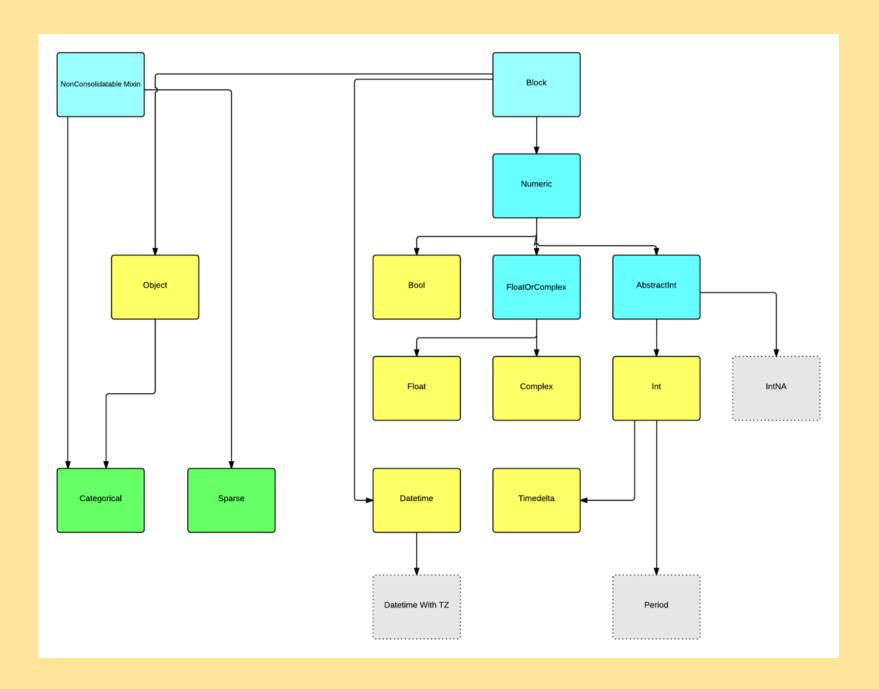
- feature set
- readability counts
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#### **Constraints**

- implementation time
- runtime
- resource utilization

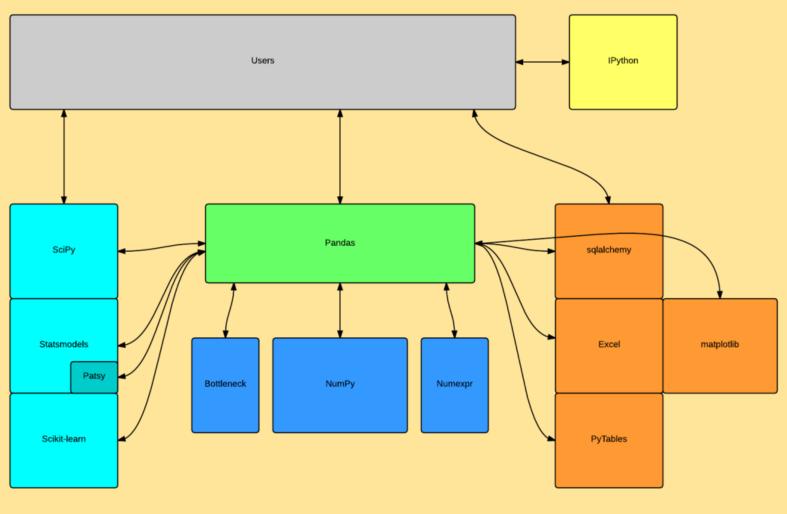
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- block memory layout



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- computation backends



Compute Stack

IO Stack

Graphics Stack

## **Computation Backends**

- numpy
- bottleneck
- numexpr
- DyND?

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

## What drives pandas?

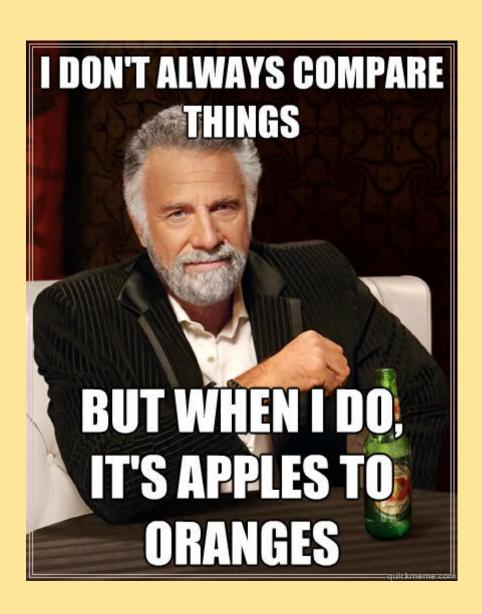
- dtype segregation
- block memory layout
- computation backends
- cython for critical parts
- hashtable for indexing

## how to make pandas perform

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

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- 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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- 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

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- conversions
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## **Memory Considerations**

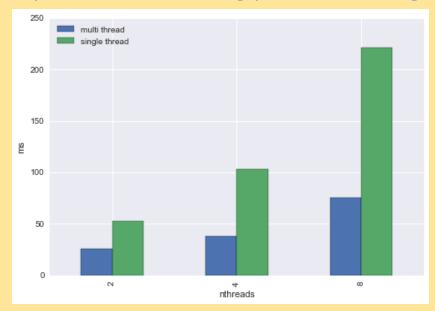
- conversions
- categoricals
- iterators

## I/O & Serialization

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

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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

@jreback