



Hervé Mignot EQUANCY

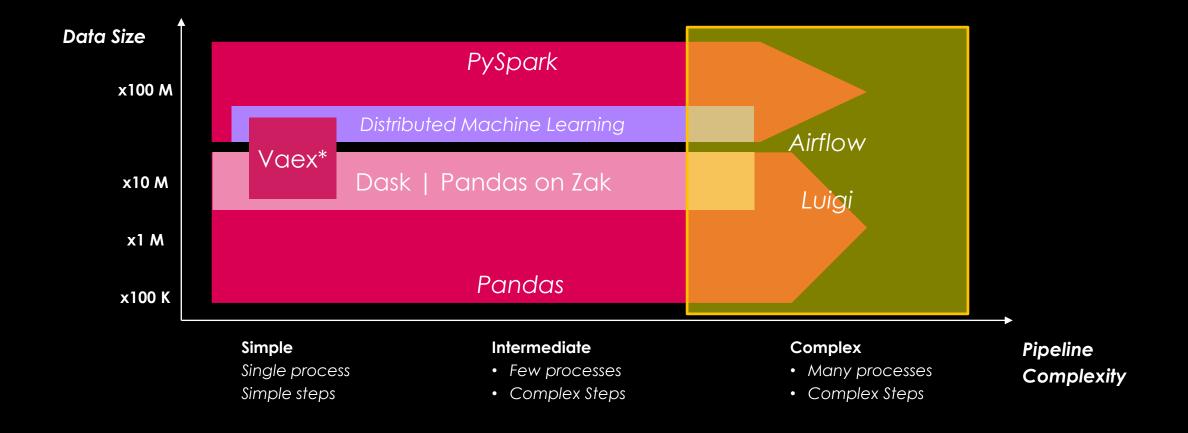


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Building Pipelines with Python

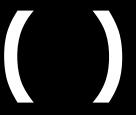




Our tools

Using pandas to build data transformation pipelines







Method Chaining

Brackets

lambda



Full credits to Tom Augspurger (@TomAugspurger)



https://tomaugspurger.github.io/

Effective Pandas
https://leanpub.com/effective-pandas

- ☐ Effective Pandas
- Method Chaining
- □ Indexes
- □ Fast Pandas

- □ Tidy Data
- □ Visualization
- □ Time Series



Modern Pandas – Method Chaining

Method chaining is composing functions application over an object.

Many data libraries API inspired from this functional programming pattern:

- dplyr (R)
- Apache Spark (Scala, Python, R)
- •

Example (reading a csv file, renaming a column, taking the first 6 rows into a pandas dataframe):

Modern Pandas – Functions

Method chaining is composing functions application over an object

What?	Method
Compute columns	<pre>.assign(col = val, col = val,)</pre>
Drop columns, rows	<pre>.drop('val', axis=[0 1]) .loc[condition for rows to be kept, list of columns]</pre>
Call user defined function	<pre>.pipe(fun, [args])</pre>
Rename columns or index	<pre>.rename(columns=mapper) .rename(mapper, axis=['columns' 'index'])</pre>
Copy or replace	.where(cond, other)
Filter rows on "where expr"	<pre>.query(where expr) .loc[dataframe expression using where expr]</pre>
Drop missing values	.dropna([subset= <i>list</i>])
Sort against values	<pre>sort_values([subset=list])</pre>



Hands-on!



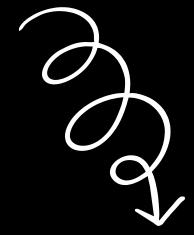






Our data set





https://www.prix-carburants.gouv.fr/rubrique/opendata/

0	1	2	3	4	5	6	7	8	9
0	1000001	1000	R	4620114.0	519791.0	2016-01-02T09:01:58	1.0	Gazole	1026.0
1	1000001	1000	R	4620114.0	519791.0	2016-01-04T10:01:35	1.0	Gazole	1026.0

Reading & preparing the data

```
df = (pd.read_csv('./Prix2017.zip',
                  sep=';',
                  header=None,
                  dtype={1: str},
                  parse_dates = [5],
     # Rename columns
     .rename(columns={0: 'station_id', 1:'zip_code',
                      3: 'latitude', 4: 'longitude', 5: 'date',
                      7: 'gas_type', 8: 'price'})
     # Recompute columns
     .assign(
         price = lambda x: x['price'] / 1000,
         latitude = lambda x: x['latitude'] / 100000,
         longitude = lambda x: x.longitude / 100000,
     # Drop columns
     .drop([2, 6,], axis=1)
```

Reading & preparing the data – 1/4

```
df = (pd.read_csv('./Prix2017.zip',
                  sep=';',
                  header=None,
                  dtype={1: str},
                  parse_dates = [5],
     # Rename columns
     .rename(columns={0: 'station_id', 1:'zip_code',
                      3: 'latitude', 4: 'longitude', 5: 'date',
                      7: 'gas type', 8: 'price'})
     # Recompute columns
     .assign(
         price = lambda x: x['price'] / 1000,
         latitude = lambda x: x['latitude'] / 100000,
         longitude = lambda x: x.longitude / 100000,
     # Drop columns
     .drop([2, 6,], axis=1)
```

Reading & preparing the data -2/4

```
df = (pd.read_csv('./Prix2017.zip',
                  sep=';',
                  header=None,
                  dtype={1: str},
                  parse_dates = [5],
     # Rename columns
     .rename(columns={0: 'station_id', 1:'zip_code',
                      3: 'latitude', 4: 'longitude', 5: 'date',
                      7: 'gas_type', 8: 'price'})
     # Recompute columns
     .assign(
         price = lambda x: x['price'] / 1000,
         latitude = lambda x: x['latitude'] / 100000,
         longitude = lambda x: x.longitude / 100000,
     # Drop columns
     .drop([2, 6,], axis=1)
```

Reading & preparing the data – 3/4

```
df = (pd.read_csv('./Prix2017.zip',
                  sep=';',
                  header=None,
                  dtype={1: str},
                  parse_dates = [5],
     # Rename columns
     .rename(columns={0: 'station id', 1:'zip code',
                      3: 'latitude', 4: 'longitude', 5: 'date',
                      7: 'gas_type', 8: 'price'})
     # Recompute columns
     .assign(
         price = lambda x: x['price'] / 1000,
         latitude = lambda x: x['latitude'] / 100000,
         longitude = lambda x: x.longitude / 100000,
     # Drop columns
     .drop([2, 6,], axis=1)
```

Reading & preparing the data -4/4

```
df = (pd.read_csv('./Prix2017.zip',
                  sep=';',
                  header=None,
                  dtype={1: str},
                  parse_dates = [5],
     # Rename columns
     .rename(columns={0: 'station id', 1:'zip code',
                      3: 'latitude', 4: 'longitude', 5: 'date',
                      7: 'gas_type', 8: 'price'})
     # Recompute columns
     .assign(
         price = lambda x: x['price'] / 1000,
         latitude = lambda x: x['latitude'] / 100000,
         longitude = lambda x: x.longitude / 100000,
     # Drop columns
     .drop([2, 6,], axis=1)
```

Result

	station_id	zip_code	latitude	longitude	date	gas_type	price
0	1000001	01000	46.20114	5.19791	2016-01-02 09:01:58	Gazole	1.026
1	1000001	01000	46.20114	5.19791	2016-01-04 10:01:35	Gazole	1.026
2	1000001	01000	46.20114	5.19791	2016-01-04 12:01:15	Gazole	1.026
3	1000001	01000	46.20114	5.19791	2016-01-05 09:01:12	Gazole	1.026
4	1000001	01000	46.20114	5.19791	2016-01-07 08:01:13	Gazole	1.026



Charting prices evolutions

```
(df
.dropna(subset=['date'])
 .groupby(['gas_type', pd.Grouper(key='date', freq='1W')])
['price']
.mean()
 .unstack(0)
.rename_axis('Gas price changes', axis=1)
 .plot()
```

Charting prices evolutions

```
(df
 .dropna(subset=['date'])
 .loc[df['gas_type'].isin(df['gas_type'].value_counts().index[:4])]
 .groupby(['gas_type', pd.Grouper(key='date', freq='1W')])
['price']
 .mean()
.unstack(0)
# .rename_axis('Gas price changes', axis=1)
 .plot()
                     Jan
2016
```

Data Quality – Chained Assertions

Use assertions for testing constraints against data frames

engarde is a module defining a set of functions & decorators to check these

Defining methods (monkey patching) on pd.DataFrame allows chained assertions

is_shape
none_missing
unique_index
within_range
within_set
has_dtypes

Logging & debugging

Encapsulate logging calls within pandas DataFrame methods

No module known, could be an addition to engarde (Tom Augspurger discussed logging)

```
import logging
def log shape(df):
    logging.info('%s' % df.shape)
    return df
pd.DataFrame.log shape = log shape
stations_df = (pd.read_csv('./Stations2017.zip', sep='|', header=None, dtype={1: str},
                           names=['station_id', 'zip_code', 'type',
                                  'latitude', 'longitude', 'address', 'city'],
               # Verify data frame structure
               .log shape()
                assign(latitude = lambda x: x.latitude / 100000, longitude = lambda x: x.longitude / 100000)
```

equancy

Merci



See you soon on...

modernpandas.io



Banskv



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