Effective Programming Practices for Economists

Data management with pandas

Creating variables

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Using numpy math functions

assume that `df` is the gapminder example

```
>>> import numpy as np
>>> df["log_life_exp"] = np.log(df["life_exp"])
>>> df
```

country	continent	year	life_exp	log_life_exp
0 Cuba	Americas	2002	77.16	4.35
1 Cuba	Americas	2007	78.27	4.36
2 Spain	Europe	2002	79.78	4.38
3 Spain	Europe	2007	80.94	4.39

• All functions you'll ever need are implemented:

```
np.log`
```

- **...**
- See docs for details
- Index is preserved
- Very fast, vectorized implementations

Arithmetic with Series

```
>>> df["gdp_billion"] = df["gdp_per_cap"] * df["pop"] / 1e9
>>> df
```

country	year	gdp_per_cap	рор	gdp_billion
0 Cuba	2002	6340.65	11226999	71.19
1 Cuba	2007	8948.10	11416987	102.16
2 Spain	2002	24835.47	40152517	997.21
3 Spain	2007	28821.06	40448191	1165.76

- **, **, *-*, *-*, */*, ... work as expected
- All calculations are aligned by index
- Not all Series have to come from the same DataFrame or be assigned to a DataFrame

Recoding values

```
>>> df["country_code"] = df["country"].replace(
... {"Cuba": "CUB", "Spain": "ESP"}
...)
>>> df

country continent year life_exp country_code
```

- country **0** Cuba Americas 2002 77.16 **CUB** 1 Cuba Americas 2007 78.27 **CUB 2** Spain 2002 79.78 ESP Europe **3** Spain Europe 2007 80.94 **ESP**
- Can be useful to create new variable or fix typos in string variables
- Not super fast, but faster than any looping approach

Vectorized if conditions

country	year	gdp_per_cap	рор	income_status
0 Cuba	2002	6340.65	11226999	not rich
1 Cuba	2007	8948.10	11416987	not rich
2 Spain	2002	24835.47	40152517	rich
3 Spain	2007	28821.06	40448191	rich

- `pd.Series.where` takes two Series as arguments:
 - 1. `cond`: Boolean Series determining where values are kept
 - 2. 'other': Series with **values** to be used where 'cond' is 'False'
- Can express general if conditions using nested where
- Vectorized and fast

When is it okay to loop?

Over columns: 🔽



```
clean = pd.DataFrame()
for var in varlist:
    clean[var] = clean_variable(df[var])
```

- Such a loop is not just ok, it is often the fastest and most readable option
- Accessing and inserting columns is fast
- Even if `clean_variable` is vectorized, it's runtime will completely dominate any loop overhead

Over rows: X

- Code example intentionally left blank
- Use the vectorized functions from above instead of loops
- List comprehensions, `df.apply`, map, etc. are just python loops in disguise and not faster in this case