

Day 2

BUSI 520: Python for Business Research

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Lambda functions

Convenient way to write short function definitions



```
In [9]: def f(x):  
        return x**2  
        print(f(3))  
  
        f = lambda x: x**2  
        print(f(3))
```

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Line continuations

- can use a backslash \ at the end of a line to continue it on the next
- or enclose in ()
- lists can be continued without a backslash



```
In [18]: def f(x):  
          return 2*x \  
              + 3  
          print(f(2))  
  
          def f(x):  
              return (  
                  2*x  
                  + 3  
              )  
          print(f(2))  
  
          lst = [  
              "a",  
              "b",  
              "c"  
          ]  
          print(lst)
```

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['a', 'b', 'c']



Reading and writing dataframes with pandas

- Can read csv, excel, sas, stata, sql, ...
- Can write also



Two ways to read stata dta files

1. `pd.read_stata`
2. `StataReader`

```
In [3]: import pandas as pd
df = pd.read_stata("WAGE1.DTA")
df
```

```
Out[3]:
```

	wage	educ	exper	tenure	nonwhite	female	married	numdep	smsa	no
0	3.10	11	2	0	0	1	0	2	1	
1	3.24	12	22	2	0	1	1	3	1	
2	3.00	11	2	0	0	0	0	2	0	
3	6.00	8	44	28	0	0	1	0	1	
4	5.30	12	7	2	0	0	1	1	0	
...
521	15.00	16	14	2	0	1	1	2	0	
522	2.27	10	2	0	0	1	0	3	0	
523	4.67	15	13	18	0	0	1	3	0	
524	11.56	16	5	1	0	0	1	0	0	
525	3.50	14	5	4	1	1	0	2	0	

526 rows × 24 columns




```
In [5]: from pandas.io.stata import StataReader

file = StataReader("WAGE1.dta")
variables = file.variable_labels()
df = file.read()
```

```
In [6]: variables
```

```
Out[6]: {'wage': 'average hourly earnings',  
        'educ': 'years of education',  
        'exper': 'years potential experience',  
        'tenure': 'years with current employer',  
        'nonwhite': '=1 if nonwhite',  
        'female': '=1 if female',  
        'married': '=1 if married',  
        'numdep': 'number of dependents',  
        'smsa': '=1 if live in SMSA',  
        'northcen': '=1 if live in north central U.S',  
        'south': '=1 if live in southern region',  
        'west': '=1 if live in western region',  
        'construc': '=1 if work in construc. indus.',  
        'ndurman': '=1 if in nondur. manuf. indus.',  
        'trcommpu': '=1 if in trans, commun, pub ut',  
        'trade': '=1 if in wholesale or retail',  
        'services': '=1 if in services indus.',  
        'profserv': '=1 if in prof. serv. indus.',  
        'profocc': '=1 if in profess. occupation',  
        'cleroocc': '=1 if in clerical occupation',  
        'servocc': '=1 if in service occupation',  
        'lwage': 'log(wage)',  
        'expersq': 'exper^2',  
        'tenursq': 'tenure^2'}
```



In [7]:

```
df
```

Out[7]:

	wage	educ	exper	tenure	nonwhite	female	married	numdep	smsa	no
0	3.10	11	2	0	0	1	0	2	1	
1	3.24	12	22	2	0	1	1	3	1	
2	3.00	11	2	0	0	0	0	2	0	
3	6.00	8	44	28	0	0	1	0	1	
4	5.30	12	7	2	0	0	1	1	0	
...	
521	15.00	16	14	2	0	1	1	2	0	
522	2.27	10	2	0	0	1	0	3	0	
523	4.67	15	13	18	0	0	1	3	0	
524	11.56	16	5	1	0	0	1	0	0	
525	3.50	14	5	4	1	1	0	2	0	

526 rows × 24 columns



What can we do with pandas?

- Explore
- Select
- Transform
- Aggregate
- Sort, rank, and cut
- Filter
- Aggregate by group
- Merge
- Plot



Explore

- `.info()`
- `.describe()`
- `.head()`
- `.tail()`
- `.columns`
- `.index`



Select

- `[]` for columns
- or `.column_name` for columns
- `.loc[]` to get rows using labels
- `.iloc[]` to get rows using index 0, 1, ...

Transform

- mathematical operations to individual columns or rows
- combine columns or combine rows in mathematical operations (add, subtract, ...)
- create new columns or rows
- `np.where(condition, if_true, if_false)`
- `.map(f)` or `.map(lambda x: ...)`
- exercises:
 1. change the female column to be "F" if 1 and "M" if 0
 2. create a new column that is 1 if experience ≤ 5 , is 2 if $5 < \text{experience} \leq 10$, 3 if $10 < \text{experience} \leq 20$, 4 otherwise
 3. create a new column that is "northcen" if northcen=1, is "south" if south=1, is "west" if west=1, and is "east" otherwise.



Aggregate

- `.sum()`, `.mean()`, `.median()`, `.quantile()`, `.std()`, `.var()`, `.min()`, `.max()`
- `.apply(lambda x: ...)`



Sort, rank, and cut

- `sort_values()` or `sort_index()`
- `.rank()`
- `pd.cut(data, bin_edges, right=True or False, labels=...)`
- `pd.qcut(data, number_of_groups, labels=...)`

Filter

- `df[df.wage<10]`
- `df[(df.wage<10) & (df.female==1)]`
- `df[(df.wage<10) | (df.female==1)]`
- `df.dropna()` or `df.dropna(subset=[...])`
- not filtering, but `.fillna` is also useful

Aggregate by group

- `df.groupby("female").wage.mean()`
- `df.groupby(["female", "married"]).wage.mean().unstack()`
- can use any aggregation function including custom functions with `.apply`
- `df.groupby("female").wage.apply(lambda x: x.quantile([0.2, 0.4, 0.6, 0.8]))`
- `df.groupby("female").wage.apply(lambda x: pd.qcut(x, 5, labels=range(1, 6)))`

```
In [20]: df["group"] = df.groupby(["female"], group_keys=False).wage.apply(  
        lambda x: pd.qcut(x, 5, labels=range(1, 6))  
        )  
df[["female", "wage", "group"]]
```

ask Julius

- to read WAGE1.DTA and describe it and then do some aggregation by groups with it
- to get the seaborn tips dataset and do some aggregation by groups with it
- to use pandas datareader to get the Fama-French factors from French's data library and merge them with MOM from French's data library
- to create boxplots for the Fama-French factors and MOM

