

DataScience for Development and Social Change, 2015

Exploring Data with Python

Getting a first look at your
datasets...

Exploring Data

- Spend time with your dataset:
 - Understand where it came from - can you live with the assumptions the data collectors made?
 - Look at the data
 - Plot the data
 - Where are there missing values? Inconsistencies? Anomalies?
- Clean your data, find better datasets, get more data

The Pandas Library

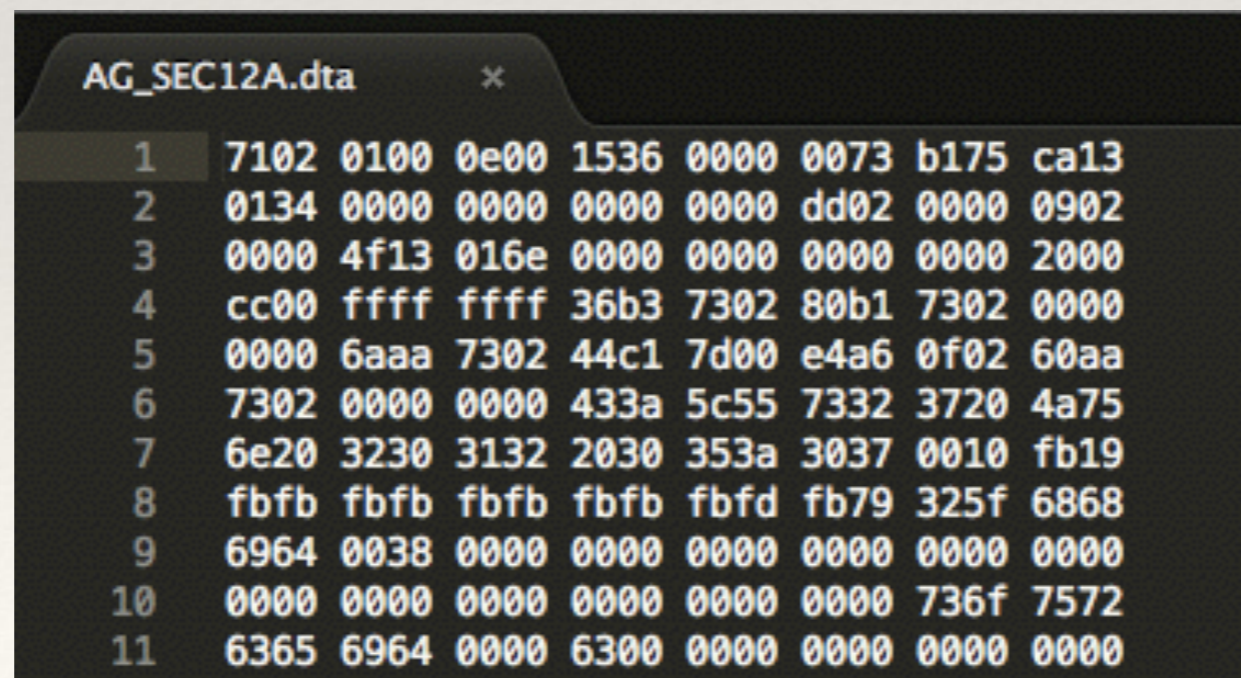
- ❖ Python has an R-like library for data analysis: Pandas
- ❖ <http://pandas.pydata.org/>
- ❖ “import pandas as pd”

Reading Files

- ❖ `read_csv`
- ❖ `read_excel`
- ❖ `read_sql` <- reads database files
- ❖ `read_json`
- ❖ `read_html` <- reads tables from HTML pages
- ❖ `read_stata` <- reads .dat files
- ❖ `read_clipboard` <- reads from your PC's clipboard

For example...

- ❖ `import pandas as pd`
- ❖ `df = pd.read_stata('LSMS Data / AG_SEC12A.dta')`
- ❖ Reads World Bank LSMS survey data into a “dataframe”



	AG_SEC12A.dta								
1	7102	0100	0e00	1536	0000	0073	b175	ca13	
2	0134	0000	0000	0000	0000	dd02	0000	0902	
3	0000	4f13	016e	0000	0000	0000	0000	2000	
4	cc00	ffff	ffff	36b3	7302	80b1	7302	0000	
5	0000	6aaa	7302	44c1	7d00	e4a6	0f02	60aa	
6	7302	0000	0000	433a	5c55	7332	3720	4a75	
7	6e20	3230	3132	2030	353a	3037	0010	fb19	
8	fbfb	fbfb	fbfb	fbfb	fbfd	fb79	325f	6868	
9	6964	0038	0000	0000	0000	0000	0000	0000	
10	0000	0000	0000	0000	0000	0000	736f	7572	
11	6365	6964	0000	6300	0000	0000	0000	0000	

What's in the DataFrame?

- ❖ `len(df)`
- ❖ `df.dtypes`
- ❖ `df.describe()`

```
>>> df.dtypes
y2_hhid      object
sourceid    category
ag12a_0b     object
ag12a_01     category
ag12a_02_1   category
ag12a_02_2   category
ag12a_02_3   category
ag12a_02_4   category
ag12a_02_5   category
ag12a_02_6   category
ag12a_03     category
ag12a_04     category
ag12a_05     float64
ag12a_06     float64
dtype: object
>>> df.describe()

```

	ag12a_05	ag12a_06
count	37.000000	398.000000
mean	11559.459459	2.113065
std	32714.883289	6.167664
min	300.000000	0.000000
25%	2000.000000	0.000000
50%	3500.000000	1.000000
75%	10000.000000	2.000000
max	200000.000000	99.000000

View the DataFrame's head/tail rows

- ❖ `df.head()`
- ❖ `df.tail()`
- ❖ `df.head(3)`
- ❖ `df[10:20]`

```
>>> df.head(3)
```

	y2_hhid	sourceid	\
0	0101014002017101	GOVERNMENT EXTENSION	
1	0101014002017101	NGO	
2	0101014002017101	COOPERATIVE/FARMER'S ASSOCIATION	

	ag12a_0b	ag12a_01	ag12a_02_1	ag12a_02_2	ag12a_02_3	\
0	SERIKALI	YES	YES	NO	YES	
1	NGO	NO	NaN	NaN	NaN	
2	USHIRIKA / CHAMA CHA WAKU	NO	NaN	NaN	NaN	

	ag12a_02_4	ag12a_02_5	ag12a_02_6	ag12a_03	ag12a_04	ag12a_05	ag12a_06
0	NO	YES	YES	AVERAGE	NO	NaN	1
1	NaN	NaN	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN

Access individual columns

```
df['sourceid']
```

```
df[['sourceid','ag12a_01','ag12a_02_2']]
```

```
df[df.ag12a_01 == 'YES']
```

```
df[(df.ag12a_01 == 'YES') & (df.ag12a_02_1 == 'NO')]
```


Check Unique Column Values

```
>>> for k in df.keys():
...     print("{}: {}".format(k, df[k].unique()))
...
y2_hhid: ['0101014002017101' '0101014002028401' '0101014002029701' ...,
'5502018021005902' '5502018021005905' '5502018021006801']
sourceid: ['GOVERNMENT EXTENSION' 'NGO' "COOPERATIVE/FARMER'S ASSOCIATION"
'LARGE SCALE FARMER' 'OTHER']
ag12a_0b: ['SERIKALI' 'NGO' 'USHIRIKA / CHAMA CHA WAKU' 'MKULIMA/MFUGAJI MKUBWA'
'NYINGINE, TAJA' 'GOVERNMENT EXTENSION' "COOPERATIVE/FARMER'S ASSO"
'LARGE SCALE FARMER' 'OTHER']
ag12a_01: ['YES' 'NO']
ag12a_02_1: ['YES' 'NO' nan]
ag12a_02_2: ['YES' 'NO' nan]
ag12a_02_3: ['YES' 'NO' nan]
ag12a_02_4: ['YES' 'NO' nan]
ag12a_02_5: ['YES' 'NO' nan]
ag12a_02_6: ['YES' 'NO' nan]
ag12a_03: ['GOOD' 'AVERAGE' 'BAD' nan]
ag12a_04: ['YES' 'NO' nan]
ag12a_05: [
    nan    1000.    3500.    7500.    11200.    16000.    200000.    1500.
    5000.    1200.    20000.    3000.    10000.    2500.    8000.    12000.
    2000.    40000.    17000.    6000.    300.    4500.    500.]
ag12a_06: [ 1. nan  2.  0.  24.  4.  12.  3.  6.  40.  99.  5.  10.  30.  25.]
```