

Data Loading, Storage and File Formats

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Reference

- **Chapter 6**
- Wes McKinney, **Python for Data Analysis**: Data Wrangling with Pandas, NumPy, and IPython, O'Reilly Media, 2nd Edition, 2018.
 - Material: <https://github.com/wesm/pydata-book>

Outline

6.1 Reading and Writing Data in Text Format

6.2 Binary Data Formats

6.3 Interacting with Web APIs

6.4 Interacting with Databases

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6.1 Reading and Writing Data in Text Format

6.2 Binary Data Formats

6.3 Interacting with Web APIs

6.4 Interacting with Databases

- Parsing Functions
- Reading Text Files in Pieces
- Writing Data to Text Format
- JSON Data
- XML and HTML: Web Scraping

Parsing Functions in pandas

Function	Description
<u>read_csv</u>	Load delimited data from a file, URL, or file-like object; use comma as default delimiter
<u>read_table</u>	Load delimited data from a file, URL, or file-like object; use tab (' \t ') as default delimiter
read_fwf	Read data in fixed-width column format (i.e., no delimiters)
read_clipboard	Version of read_table that reads data from the clipboard; useful for converting tables from web pages
read_excel	Read tabular data from an Excel XLS or XLSX file
read_hdf	Read HDF5 files written by pandas
read_html	Read all tables found in the given HTML document
read_json	Read data from a JSON (JavaScript Object Notation) string representation
read_msgpack	Read pandas data encoded using the MessagePack binary format
read_pickle	Read an arbitrary object stored in Python pickle format

Most used

Parsing Functions in pandas – cont.

Function	Description
<code>read_sas</code>	Read a SAS dataset stored in one of the SAS system's custom storage formats
<code>read_sql</code>	Read the results of a SQL query (using SQLAlchemy) as a pandas DataFrame
<code>read_stata</code>	Read a dataset from Stata file format
<code>read_feather</code>	Read the Feather binary file format

6.1 Reading and Writing Data in Text Format

- These parsing functions take **optional arguments** that fall into the following categories:
 - **Indexing**
 - **Type inference and data conversion**
 - **Datetime parsing**
 - **Iterating**
 - **Unclean data issues**
- **Many options**, so refer to the **online documentation** for **complex cases**.

Comma-Separated (CSV) Text Files

```
ex1.csv  
a,b,c,d,message  
1,2,3,4,hello  
5,6,7,8,world  
9,10,11,12,foo
```

```
  0  1  2  3  4  
0  1  2  3  4 hello  
...
```

- For files with **no headers**, ask pandas to **assign default** column names, or **specify names**.

```
df = pd.read_csv('ex1.csv')
```

```
df  
   a  b  c  d message  
0  1  2  3  4  hello  
1  5  6  7  8  world  
2  9 10 11 12   foo
```

```
pd.read_csv('ex2.csv',  
            header=None)
```

```
pd.read_csv('ex2.csv',  
            names=['a', 'b', 'c', 'd',  
                  'message'])
```

Alternative:

```
pd.read_table('ex1.csv', sep=',')
```


Comma-Separated (CSV) Text Files

- You can use **one of the file columns** as **index**.
- How to handle fields separated by a **variable amount of whitespace**?

`ex3.csv`

	A	B	C
aaa	-0.264438	-1.026059	-0.619500
bbb	0.927272	0.302904	-0.032399

Must be index, why?

```
names = ['a', 'b', 'c', 'd',  
         'message']
```

```
pd.read_csv('ex2.csv', names=names,  
           index_col='message')
```

	a	b	c	d
hello	1	2	3	4
world	5	6	7	8
foo	9	10	11	12

Regular expression
for one or more
white space

```
pd.read_table('ex3.txt', sep='\s+')
```

Comma-Separated (CSV) Text Files

- **Missing data** is usually either **not present** (empty string) or marked by some **sentinel value**.
- Can Specify the sentinel values.

```
ex5.csv
something,a,b,c,d,msg
one,1,2,3,4,NA
two,5,6,,8,foo
```

```
pd.read_csv('ex5.csv')
```

```
  something  a  b  c  d  msg
0         one  1  2  3.0  4  NaN
1         two  5  6  NaN  8  foo
```

```
sentinels = {'msg': ['foo', 'NA'],
             'something': ['two']}
```

```
pd.read_csv('examples/ex5.csv',
            na_values=sentinels)
```

```
  something  a  b  c  d  msg
0         one  1  2  3.0  4  NaN
1         NaN  5  6  NaN  8  NaN
```

Reading Text Files in Pieces

- If you want to **read a small number of rows**, use **nrows**.
- To **read a file in pieces**, specify a **chunksize** of rows.
- **Iterate** on the returned parser object **to aggregate** the value counts in the **'key'** column.
- There is also **chunker.get_chunk(n)**.

```
pd.read_csv('ex6.csv', nrows=5)

chunker = pd.read_csv('ex6.csv',
                      chunksize=1000)

chunker
<pandas.io.parsers.TextFileReader
at 0x7f6b1e2672e8>

tot = pd.Series([])
for c in chunker:
    tot = tot.add(
        c['key'].value_counts(),
        fill_value=0)

tot = tot.sort_values(
    ascending=False)
```

Writing Data to Text Format

- We can **write** the data out **to a comma-separated file**.
- Useful options: **sep**, **na_rep**, **index**, and **header**.
- You can also write only a **subset** of the columns, and in an **order** of your choosing.

```
data.to_csv('out.csv')
```

```
data.to_csv(sys.stdout,  
            sep='|',  
            na_rep='NULL')  
|something|a|b|c|d|message  
0|one|1|2|3.0|4|NULL
```

```
data.to_csv(sys.stdout,  
            index=False,  
            header=False)
```

```
data.to_csv(sys.stdout,  
            columns=['a', 'b', 'c'])
```

JSON Data

- **JSON** (short for **JavaScript Object Notation**) is a standard format for sending data.
- It is a **free-form** data format. Example:

```
{ "name": "Wes",  
  "places_lived": ["United States", "Spain"],  
  "pet": null,  
  "siblings": [{"name": "Scott", "age": 30},  
               {"name": "Katie", "age": 38}]  
}
```

Call it **j_str**

Nearly valid Python code. Exceptions: The null value is **null**. **Disallowing trailing commas** at the end of lists. All of the **keys** in an object must be **strings**.

JSON Data

- Python has built in JSON support.
- To convert a JSON string to Python form, use `json.loads`.
- `json.dumps` converts a Python object to JSON.
- Python `dict` to `DataFrame`.

```
result = json.loads(j_str)
```

```
asjson = json.dumps(result)
```

```
siblings = pd.DataFrame(  
    result['siblings'],  
    columns=['name', 'age'])
```

```
siblings  
   name  age  
0  Scott  30  
1  Katie  38
```

Also `json.load(fp)`

Also `json.dump(fp)`

JSON Data

- The default options for `pandas.read_json` assume that each object in the JSON array is a row in the table.
- To export data from pandas to JSON, use the `to_json`.

```
example.json
```

```
[{"a": 1, "b": 2, "c": 3},  
 {"a": 4, "b": 5, "c": 6},  
 {"a": 7, "b": 8, "c": 9}]
```

```
data = pd.read_json('example.json')
```

```
data
```

```
   a  b  c  
0  1  2  3  
1  4  5  6  
2  7  8  9
```

```
print(data.to_json(  
        orient='records'))
```

```
[{"a":1,"b":2,"c":3}, {"a":4,"b":5,"  
c":6}]
```

XML and HTML: Web Scraping

- Given the **html** document from the US FDIC [list](#) for bank failures, find the **years with most bank failures**.
- Pandas has **read_html** that returns a list of DataFrames.
- We need to extract years from the 'Closing Date' column.

```
conda install lxml
pip install beautifulsoup4 html5lib

tables = pd.read_html(
    'fdic_failed_bank_list.html')
failures = tables[0]

close_timestamps = pd.to_datetime(
    failures['Closing Date'])
close_timestamps.dt.year.
    value_counts()

2010 157
2009 140
2011 92
...
```


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- Pickle
- Using HDF5 Format
- Reading Microsoft Excel Files

6.2 Binary Data Formats

- Python's has built-in **pickle** serialization.
- **Serialization** is converting an object in memory to a byte stream that can be stored on disk or sent over a network.
- Python's pickle package has **dump** and **load**.
- Good for short-term storage.

```
import pickle
```

```
dogs_dict = { 'Ozzy': 3, 'Filou': 8,  
             'Luna': 5, 'Skippy': 10, 'Barco': 12,  
             'Balou': 9, 'Laika': 16 }
```

```
filename = 'dogs'  
outfile = open(filename, 'wb')  
pickle.dump(dogs_dict, outfile)  
outfile.close()
```

```
infile = open(filename, 'rb')  
new_dict = pickle.load(infile)  
infile.close()
```

6.2 Binary Data Formats

- All pandas objects have **to_pickle**.

```
frame = pd.DataFrame(  
    np.arange(9).reshape((3, 3)))  
frame.to_pickle('frame_pickle')
```

- The reverse is **read_pickle**.

```
pd.read_pickle('frame_pickle')  
   0  1  2  
0  0  1  2  
1  3  4  5  
2  6  7  8
```

Using HDF5 Format

- The **hierarchical data format** is efficient and cross platform.
- Pandas has built-in support for HDF5.
- Use **to_hdf** and **read_hdf** to access one or more pandas objects in an HDF5 file.

```
df = pd.DataFrame({'A': [1, 2, 3],  
                  'B': [4, 5, 6]})  
df.to_hdf('data.h5', key='df1',  
         mode='w')
```

```
s = pd.Series([1, 2, 3, 4])  
s.to_hdf('data.h5', key='s1')
```

```
pd.read_hdf('data.h5', 'df1')
```

```
   A  B  
0  1  4  
1  2  5  
2  3  6
```

Using HDF5 Format

- The **HDFStore** class works like a **dict** and handles the low-level details for writing and retrieving.

```
frame = pd.DataFrame({'a':  
                        np.random.randn(100)})
```

```
store = pd.HDFStore('mydata.h5')  
store['obj1'] = frame  
store['obj1_col1'] = frame['a']
```

```
frm2 = store['obj1']  
store.close()
```

Reading Microsoft Excel Files

- Pandas supports **reading** from **Excel 2003** (and higher) files using either the **ExcelFile** class or **pandas.read_excel** function.
- Need the packages **xlrd** and **openpyxl**.
- Writing is supported with **ExcelWriter**.

```
xlsx = pd.ExcelFile('ex1.xlsx')
pd.read_excel(xlsx, 'Sheet1')
```

```
   a  b  c  d  message
0  1  2  3  4    hello
1  5  6  7  8    world
2  9 10 11 12     foo
```

```
# Alternatively, for one sheet:
frame = pd.read_excel('ex1.xlsx',
                      'Sheet1')
```

```
writer = pd.ExcelWriter('ex2.xlsx')
frame.to_excel(writer, 'Sheet1')
writer.save() # Save and close
# Alternatively, for one sheet:
frame.to_excel('ex2.xlsx')
```

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6.3 Interacting with Web APIs

- Many **websites** have public APIs providing data feeds via **JSON**, e.g., [Weather Data](#).
- To find the last **30 GitHub issues** for pandas, we can make a **GET** HTTP request using the add-on **requests** library.

```
import requests
url = 'https://api.github.com/repos/pandas-dev/pandas/issues'

resp = requests.get(url)
data = resp.json() # List of dict
data[0]['title']
'Period does not round down for ...'
issues = pd.DataFrame(data,
                      columns=['number',
                               'title', 'labels',
                               'state'])
```


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6.4 Interacting with Databases

- The [SQLAlchemy project](#) is a popular **Python SQL toolkit** for interfacing with SQL databases.
- **Supports** SQLite, Postgresql, MySQL, Oracle, MS-SQL, Firebird, Sybase and others.
- pandas has **read_sql** that reads data easily from a SQLAlchemy connection.

```
import sqlalchemy as sqla
db = sqla.create_engine(
    'sqlite:///mydata.sqlite')
```

```
pd.read_sql('select * from test',
            db)
```

	a	b	c	d
0	Atlanta	Georgia	1.25	6
1	Tallahassee	Florida	2.60	3
2	Sacramento	California	1.70	5

Homework

- Solve the homework on **data loading and file formats.**

Summary

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