

Importing Text Files I

```
open(file_name, 'r') # open the file
file.read() # read the file
file.close() # close the file
file.closed() # check if the file is closed
```

It is a good practice to close the file after reading it when using 'open'

Importing Text Files II

```
with open(file_name) as file:
    # open the file

file.read() # read the file
file.readline() # read line by line
```

When using the 'with' statement there is no need to close the file

Importing Flat Files with Numpy I

```
import numpy as np

np.loadtxt(file_name, delimiter=',')

skiprows=1

usecols=[0, 2]
```

dtype = str

loadtxt only works with numeric data

Importing Flat Files with Numpy II

```
import numpy as np

np.recfromcsv(file, delimiter=',', names=True, dtype=None)

np.genfromtxt(file, delimiter=',', names=True, dtype=None)
```

with the functions `recfromcsv()` and `genfromtxt()` we are able to import data with different types

Importing Stata Files

```
import pandas as pd

df = pd.read_stata('disarea.dta')
```

Importing Flat Files With Pandas

```
import pandas as pd

pd.read_csv(file)

nrows=5 # argument for the number of rows to load
```

```
import numpy

# argument for no header
# argument to set delimiter
# argument to skip a specific row
# argument to only show specific columns
# argument to recognize a string as a NaN Value
```

to import the data as string

Import pickled files

```
import pickle

import numpy

with open(file_name, 'r') as file:
    # the file as file

pickle.load(file) # open the file
```

Importing Spreadsheet Files

```
import pandas as pd

pd.ExcelFile(file) # opening the file

xl.sheet_names # exporting the sheet names
```

```
xl.parse(sheet_name/index) # loading sheet to dataframe
```

```
skiprows=[index] # skipping specific row
```

```
names=[List of Names] # naming the sheet's columns
```

```
usecols=[0,] # parse specific columns
```

skiprows, names and usecols are all arguments of the function `parse()`

Importing SAS Files

```
from sas7bdat import SAS7BDAT
```

```
import pandas as pd
```

```
with SAS7BDAT('file_name') as file:
```

```
file.to_dataframe()
```

Importing HDF5 files

```
import numpy as np      import numpy
import h5py              importing the
                        h5py library
h5py.File(file, 'r')    reading the
                        file
```

Importing MATLAB files

```
import scipy.io          importing
                        scipy.io
scipy.io.loadmat('file_ reading
name')                  the file
```

Relational databases I

```
import pandas as pd
from sqlalchemy import create_engine
```

```
engine = create_engine('database://username:password@hostname:port/database')
con = engine.connect()
```

```
rs = con.execute('SELECT * FROM Album')
```

```
df = pd.DataFrame(rs.fetchall())
```

```
df.columns = rs.keys
```

```
con.close()
```

The best practice is to close the connection

Relational databases II

```
engine = create_engine('database://username:password@hostname:port/database')
with engine.connect() as con:
```

```
rs = con.execute('sql code')
```

```
df = pd.DataFrame(rs.fetchmany(size=3))
```

With 'open' you don't have to close the connection at the end

Relational databases III

```
engine = create_engine('database://username:password@hostname:port/database')
df = pd.read_sql_query('sql code', engine)
```

Fastest way to connect to a database engine and perform query

importing pandas
importing the necessary library

creating an engine
performing query

connecting to the engine
performing query
save as a dataframe
set columns names
close the connection



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