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Pandas Python
(from corey schafer)

used to read tabular data.

Introduction

* import pandas as pd — General Convention of importing
* df = pd.read_csv('path/to/file.txt') — read into the file
* df (short for Dataframe). — Prints the file.
* df.shape — columns x rows of the file (or) Shape of the file.
* df.info() — information of all the datatypes of the columns along with the shape of the dataframe

datatypes → (some of them) int64, float64, object
└ string type. (generally)

```
pd.set_option('display.max_columns', 36)
pd.set_option('display.max_rows', 20)
```

a max of 20 rows is displayed.
a max of '36' columns is displayed.

* df.head() ——→ get first 5 rows.
* df.head(7) ——→ set first 7 rows.
* df.tail() ——→ set last 5 rows.
* df.tail(7) ——→ set the last 7 rows.

Dataframe & Series Datatypes

Dataframe → 2-D Datastructure.

Dataframe is like to a Dictionary.

eg

```
people = {
    "name" : ["x", "y", "z"]
    "role" : ["A", "B", "C"]
}
```

x → A
y → B
z → C

people["name"] ⟶ ["x", "y", "z"]

to convert this Dictionary into Dataframe we can do

~~pd.DataFrame(people)~~ df = pd.DataFrame(people) *

print(df) ↓

	name	role
0	x	A
1	y	B
2	z	C

* type(df) = pandas.core.frame.DataFrame

indexes * INDEXES

print(df["name"]) | type(df["name"])

0	x
1	y
2	z

pandas.core.series.Series

Series → 1-D Datastructure like to Lists but a lot more functionality. (or) Array.

We can Also write df.name (or) df.role but

* df["name"] (✓) (or) df["role"] (✓) is more preferred.

printing out Multiple columns

df[ ]

↑ pass in 'l' column (or) a list of columns.

* df[["name", "role"]]

Now, we've seen how to access columns.

How to Access rows?

→ for that we use iloc & loc

df.iloc[,]

→ Pass in a list (or) single index of indexes (row indexes) — necessary argument *

→ Pass in the column index (or) list of column indexes — optional argument *

→ Used for slicing the Dataframe with integer-indexes.

* df.iloc[[0,1], 1]

0	A
1	B

* df.iloc[0]

→ prints out all the columns of row index 0.

i.e.

	name	role
0	X	A

* df.iloc[[1,2], [0,1]]

	name	role
1	Y	B
2	Z	C

But we can use custom-named indexes

	0	1	2
0			
1			
2			

← regular

df.loc → 111er to df.iloc

	name	role
0		
1		
2		

← custom-named

* df.iloc[[1,2], ["role", "name"]]

	role	name
1	B	Y
2	C	Z

* df.columns → shows (or) returns a list of all the columns in the DataFrame df.

["name", "role"]

We can Also use slicing

df.loc[[0,1,2], ["name", "role"]]

can be written as → df.loc[0:2, "name":"role"]

df.iloc[0:2, 0:1]

* While slicing in pandas 0:2 → Both are inclusive.

(✓) Pandas → 0:2 → 0, 1, 2

generally → 0:2 → 0, 1

Indexes

We can set a column to act as a index (Also indexes need not be unique according to pandas) we can do it by

* df.set_index("column_name")
  df

But this does not actually change df to set index as 'column_name' we can do that by

df = df.set_index("column_name")

(or)

df.set_index("column_name", inplace=True).

df.index → shows or displays the index column of the DataFrame df.

When we do this we are setting custom indexes

the use of this feature is

Imagine we have fields like "name" & "email" we have unique emails then that can act as a field

⇒ we do `df = df.set_index("email")`

& we can search based on email using 'loc'

* `df.loc("johndoe@gmail.com")`
  ↳ gets all info relating to the email "johndoe@gmail.com" the name & stuff.

* Note
  We cannot use ~~loc~~ `loc[0]` once we change the indexes since indexes 0, 1, ... do not exist since we changed it

  But, we can still use `iloc[0]` & stuff with iloc it works.
  (still)

We can reset indexes by doing

* `df = df.reset_index()`
  (or)
  `df.reset_index(inplace = True)`

  Now, all the integer indexes are restored.

We can use * `schema_df.sort_index()`
↳ sorts all the index

(in pandas) Can be used in many places where the changes are temporary

ascending = False.
inplace = True

*

`df.shape, df.index, df.columns.`

Note* df.loc & df.iloc return the smaller Dataframes wrt to the values we put in them. The returned Dataframes are of the type 'pandas.core.frame.DataFrame'

Same as all the Dataframe.

(A new sub-DataFrame is returned, of the original 'df'.

Filtering

Consider Dataframe

	Name	last	email
0	John	Doe	johndoe@gmail.com
1	Jane	Doe	Janedoe@gmail.com.
2	James	Brad	JamesB@gmail.com.
3	Jennifer	Anniston	JenniferAnniston@gmail.com.

* We use conditionals such as '==', '4', '1' to filter our DataFrame.

→ (df["last"] == "Doe") → filt

↓ → this will return a series of True/False values. (Array)

[True, True, False, False] → df[[True, True, False, False]]

→ df[filt] will give indexes 0, 1

→ it is a sub-DataFrame to the original DF (it is filtered.)

We can do df.loc[filt] instead → this has ↑ Advantage

* df.loc processes the Array of T/F values

We can do

→ df.loc[filt, ["Name", "email"]]

* to get only Name & email fields.

Using Conditionals. '&' and '|'

('and')

```
filt = (df["last"] == "Doe") & (df["name"] == "John")
df.loc[filt]
```

→ returns index '0'. Dataframe with

Negation → '~' (Not)

```
df.loc[~filt]
```

→ Dataframe with indexes '1', '2', '3'

In Pandas we use '|' for (or).

We can also use '>', '<', '>=', '<=' for filtering.

Examples

```
high_salary = (df["salary"] > 70000)
df[high_salary]
df.loc[high_salary, ["country", "prog_lang", "salary"]]
```

Now let's say we want to see if the "country" is in a set of our preferred countries

```
countries = ["USA", "UK", "India", "Germany", "Canada"]
```

we use (Python-Pandas func()) ".isin()" function to check if "country" is in the array

```
filt1 = df["country"].isin(countries)
```

Now let's say we want to see how many use "python" in their work.

consider "prog_lang" format as (HTML, CSS, Java, python ...)

We use a (Python-Pandas func()) ".str.contains()" to check if substring is present inside a main string.

we need to put a additional parameter 'na = False' to deal with NaN values in the DataFrame.

(Null or None) | NaN = None = NULL

Basically NULL values in Pandas DataFrame.

[redacted]

```
filt2 = df["prog_lang"].str.contains("Python", na=False)
df.loc[filt2]
```

filtering is one of the first things we do with Data