

## Part 4 - Useful packages for Librarians

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### importing packages

Often times you need to import some packages that contain functions to help you do things more quickly, and often times you give that package a nickname like so:

```
import package_name as package_nickname
```

therefore you can use `package_nickname` later in your code to access the packages functionality.

### Pandas, a package for working with tabular data

Pandas is one of the most popular packages in python, as it works with spreadsheet style data. You can quickly import a csv with its import function:

```
import pandas as pd # everyone uses pd as the nickname for pandas  
  
df = pd.read_csv('data/dataset.csv')
```

When reading or writing files in Python, by default, the jupyter notebook will look in the folder where the notebook is saved, therefore in the example above, we are using a relative subfolder called `data` with a `dataset.csv` saved in it.

### Accessing data with pandas;

The beauty of Python is that all packages try to work in the same way as the base kinds of objects in python, so a Pandas dataframe, which is the way Pandas represents a table, can be thought of as a dictionary full of lists, where the dictionary key is the name of the column, and the list index is the row number. Here is an example:

```
df['column 1'][0] # will output the value of the first row in column 1
```

### Handy things to do with a dataframe:

```
df.dtypes #an attribute that contains the type of data in each column  
df.head() # a method that prints the top five rows of the table  
df.shape # an attribute tells you the number of rows and columns  
df.columns # an attribute that tells you the names of all the columns
```

**Note:** Dataframes also have attributes, which are kind of like methods but without the `()`. You can think of them as metadata variables attached to the dataframe, and accessed using the full stop and the name of the attribute.

## Another super handy package, dcxml

This package was created to make dublin core xml files in Python, all we need to do is first import the package:

```
from dcxml import simpledc
```

Here we are using a slightly different code, where we are importing one thing (`simpledc`) from the package (`dcxml`). `simpledc` is a special python object that carries several methods to create dublin core xml.

To create a string xml file, we will use the `simpledc.tostring()` method, which takes in data in a dictionary format:

```
data = {
    'contributors' : ['CERN'],
    'coverage' : ['Geneva'],
    'creators' : ['CERN'],
    'dates' : ['2002'],
    'descriptions' : ['Simple Dublin Core generation'],
    'formats' : ['application/xml'],
    'identifiers' : ['dublin-core'],
    'languages' : ['en'],
    'publishers' : ['CERN'],
    'relations' : ['Invenio Software'],
    'rights' : ['MIT'],
    'sources' : ['Python'],
    'subject' : ['XML'],
    'titles' : ['Dublin Core XML'],
    'types' : ['Software'],
    'extra': ['extra']
}
```

All we need to do is put that dictionary into the `simpledc.tostring()` method and we have an xml!

```
xml = simpledc.tostring(data) # takes in a dictionary
print(xml) # print it out and see!
```

## Magic doodad, convert\_df\_row\_to\_dictionary

This doodad was created for this workshop, and will help us convert one of our metadata records, as a row of a dataframe, into a dictionary full of lists, to prepare it for the `simpledc.tostring()` method:

this function takes in a dataframe as the first argument, and a row index number as the second, and outputs a dictionary formatted for the `simplifiedc` methods:

```
from jgarber_respitch.workshop_tools import convert_df_row_to_dictionary  
convert_df_row_to_dictionary(df, 1)
```