Series A Series is very similar to a NumPy array (in fact it is built on top of the NumPy array object). What differentiates the NumPy array from a Series, is that Series is a **one-dimensional labeled array** capable of holding any data type (integers, strings, floating point numbers, Python objects, etc.). A Series can have axis labels, meaning it can be indexed by a label, instead of just a number location. Importing the required libraries pd is the abbreviation for pandas used by the data science community. In [149... import numpy as np import pandas as pd 1. Creating a Series pd.Series() You can create a series by calling pd.Series() • Syntax: pd.Series(data=None, index=None, dtype=None, name=None, copy=False, fastpath=False) data: data can be many different things: a Python dict an ndarray a scalar value (like 5) index: index is a list of axis labels. 1.1 Using List Note: If no index is given, an index with the values [0,..., len(data) - 1] is generated. # Create a pandas series using list In [150... $my_list = [10, 20, 30]$ pd.Series(data=my list) 10 Out[150]: 20 2 30 dtype: int64 In [151... # A pandas series with custom index my list = [10, 20, 30]labels = ['a', 'b', 'c'] # index must be the same length as data pd.Series(data=my list,index=labels) Out[151]: b 20 30 dtype: int64 In [152... # If we do not pass data and index with keyword arguments, # the first argument will be data and the second will be index. pd.Series(labels, my list) 10 а Out[152]: 20 b 30 С dtype: object 1.2 Using NumPy Array # Create a pandas series using numpy array In [153... arr = np.array([10,20,30])pd.Series(arr) 10 Out[153]: 20 30 dtype: int32 In [154... # Example pd.Series(arr, labels) 10 а Out[154]: 20 30 dtype: int32 1.3 Using Dictionary **Note:** All the keys in the dictionary will become the indices of the Series object. In [155... # Create a pandas series using dictionary $d = \{ 'a':10, 'b':20, 'c':30 \}$ pd.Series(d) 10 Out[155]: 20 30 dtype: int64 • If an index is passed, the values in data corresponding to the labels in the index will be pulled out and the remaining index values will NaN (not a number) is the standard missing data marker used in pandas. pd.Series(d,index=["a","b","d","e","f"]) In [156... 10.0 Out[156]: 20.0 b NaN d NaN NaN dtype: float64 1.4 From scalar value • If data is a scalar value, an index must be provided. The value will be repeated to match the length of index. In [157... pd.Series(5.0,index=["a", "b", "c", "d", "e"]) Out[157]: 5.0 С 5.0 5.0 d 5.0 dtype: float64 A pandas Series can hold a variety of object types: Even functions can be a part of the Series pd.Series([sum,print,len]) In [158... <built-in function sum> Out[158]: <built-in function print> <built-in function len> dtype: object 2) Selecting Data 2.1 Using Index The key to using a Series is understanding its index. Pandas makes use of these index names or numbers by allowing for fast look-ups of information (works like a hash table or dictionary). ser1 = pd.Series([1,2,3,4],index = ['USA', 'Germany','USSR', 'Japan']) In [159... print(ser1) USA 1 Germany 2 USSR 3 Japan 4 dtype: int64 Use indexing operators [] for quick and easy access # We can use pandas' explicit index (0,1..) as well as our own label to access the value. In [160... print(ser1["USA"]) print(ser1[0]) 1 print(ser1[-1]) # negative indexing In [161... 2.2 Slicing ser1 = pd.Series([1,2,3,4],index = ['USA', 'Germany','USSR', 'Japan']) In [162... print(ser1) USA 1 2 Germany USSR 3 4 Japan dtype: int64 ser1["USA":"USSR"] # **both** the start and the stop are included with custom indexing In [163... 1 USA Out[163]: Germany 2 USSR 3 dtype: int64 ser1[0:2] # but with default indexing start(included) and stop(excluded) In [164... USA 1 Out[164]: Germany 2 dtype: int64 2.3 fancy indexing ser1 = pd.Series([1,2,3,4],index = ['USA', 'Germany','USSR', 'Japan']) In [165... print(ser1) USA Germany USSR Japan dtype: int64 In [166... ser1[[0,3]] # we can pass multiple indexes in a list Out[166]: Japan dtype: int64 In [167... ser1[["USA","Japan"]] Out[167]: Japan dtype: int64 **Using Boolean Mask** # mask In [168... mask = ser1%2 == 0# Condition: even values # It generates a series of Boolean values. USA False Out[168]: True Germany USSR False True Japan dtype: bool In [169... extraxt_from_ser1 = ser1[mask] print(extraxt_from_ser1) # Even values only Germany Japan dtype: int64 3. Assign a value ser1 = pd.Series([1,2,3,4],index = ['USA', 'Germany','USSR', 'Japan']) In [170... print(ser1) USA 1 Germany 2 USSR 3 Japan dtype: int64 In [171... ser1[0]=232 print(ser1) 232 Germany Japan dtype: int64 In [172... | ser1["Japan"]=345 print(ser1) USA 232 Germany 2 USSR Japan dtype: int64 4. Series operations You can perform arithmetic operations like addition, subtraction, division and multiplication on two Series objects. The operations are performed only on the matching indexes. For all non-matching indexes, NaN (Not a Number) will be returned. In [173... ser1 = pd.Series([1,2,3,4],index = ['USA', 'Germany','USSR', 'Japan']) ser2 = pd.Series([1,2,5,4],index = ['USA', 'Germany','Italy', 'Japan']) # addition of two series In [174... ser1 + ser2 Germany Out[174]: Italy 8.0 Japan USA 2.0 dtype: float64 ser1*ser2 In [175... 4.0 Germany Out[175]: Italy NaN 16.0 Japan 1.0 dtype: float64 In [176... # An arithmetic operation on the array with a scalar value ser1*2 # This code multiplies each element of the array by 2 Out[176]: Germany USSR Japan dtype: int64 # Membership operator (checking index) In [177... "USA" in ser1 Out[177]: In [178... # relationship operator USA False Out[178]: Germany False USSR True Japan True dtype: bool In [179... # min() ser1.min() Out[179]: In [180... # max() ser1.max() Out[180]: In [181... # sum() ser1.sum() # It will give the sum of all the values. Out[181]: # mean() In [182... ser1.mean() Out[182]: In [183... # count ser1.count() Out[183]: 4.1 Series Attributes In [184... ser1 = pd.Series([1,2,3,4],index = ['USA', 'Germany','USSR', 'Japan']) USA Out[184]: Germany USSR Japan dtype: int64 In [185... # size ser1.size Out[185]: # shape In [186... ser1.shape Out[186]: In [187... # dtype ser1.dtype dtype('int64') Out[187]: In [188... # index ser1.index Index(['USA', 'Germany', 'USSR', 'Japan'], dtype='object') Out[188]: In [189... # values ser1.values array([1, 2, 3, 4], dtype=int64) Out[189]: 4.2 Series methods In [190... sr = pd.Series([33,22,22,11,44,22,55,66,77,33,55,22,22,88]) 33 Out[190]: 22 22 11 4 44 5 22 6 55 7 66 8 77 9 33 10 55 11 22 12 22 13 88 dtype: int64

value_counts()

In [191... sr.value_counts()

1

1

1

1

dtype: int64

sort_values()

Sort by the values.

11

22 22

22

33 33

55

8 77
13 88
dtype: int64

77 66

55

55

22

Great Job!

dtype: int64

6

10

11

22

335511

44

66

77

88

11

Out[191]:

In [192...

Out[192]:

In [193...

Out[193]:

Return a Series containing counts of unique values.

Sort a Series in ascending or descending order by some criterion.

sr.sort_values(ascending=False) # descending to ascending

Let's stop here for now and move on to DataFrames, which will expand on the concept of Series!

You can read the documentation for further details: click the link below. https://pandas.pydata.org/docs/reference/api/pandas.Series.html

sr.sort_values() # default is ascending

• Syntax: Series.value_counts(normalize=False, sort=True, ascending=False, bins=None, dropna=True)

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