Pandas_and_Alternatives

July 19, 2024

1 Pandas and Alternatives

1. Import pandas aliased as 'pd' and numpy aliased as 'np'

```
[1]: import pandas as pd
```

2. Create a DataFrame named 'df' by reading the file 'USCG.Search.Rescue.Stats.csv' using the pandas read_csv method.

```
[2]: df = pd.read_csv('USCG.Search.Rescue.Stats.csv')
```

3. View the top 5 rows of data using the DataFrame head() method.

```
[3]: df.head()
```

[3]:	Fiscal Year	Cases	Responses	Sorties	Lives Saved	\
1964	NaN	41525	NaN	2932	NaN	
1965	NaN	38586	NaN	1984	NaN	
1966	NaN	43366	NaN	2629	NaN	
1967	NaN	42225	NaN	3028	NaN	
1968	NaN	46922	NaN	2434	NaN	

	Lives Los	t Aiter CG	Notification	Lives Los	t Before	CG	Notification	\
1964			NaN				NaN	
1965			NaN				NaN	
1966			NaN				NaN	
1967			NaN				NaN	
1968			NaN				NaN	

	Total	Lives	Unaccounted	For
1964	NaN			NaN
1965	NaN			NaN
1966	NaN			NaN
1967	NaN			NaN
1968	NaN			NaN

4. View the last 5 rows by using the DataFrame tail() method.

```
[4]: df.tail()
 [4]:
            Fiscal Year
                         Cases
                                 Responses
                                             Sorties
                                                      Lives Saved \
      2011
                 20512.0
                          43954
                                    21566.0
                                                              259.0
                                                 3793
      2012
                 19787.0
                          43940
                                    21609.0
                                                 4037
                                                              284.0
      2013
                 17803.0
                          38272
                                    19420.0
                                                 3753
                                                              226.0
      2014
                 17508.0
                          38282
                                    19032.0
                                                 3443
                                                              170.0
      2015
                 16456.0
                          37215
                                    18781.0
                                                 3536
                                                              169.0
            Lives Lost After CG Notification
                                                Lives Lost Before CG Notification
      2011
                                         476.0
                                                                               735.0
      2012
                                         429.0
                                                                               713.0
      2013
                                         425.0
                                                                               651.0
      2014
                                         425.0
                                                                               595.0
      2015
                                         434.0
                                                                               603.0
                   Lives Unaccounted For
            Total
            392.0
      2011
                                        NaN
      2012
            440.0
                                        NaN
      2013
            252.0
                                        NaN
      2014
            308.0
                                        NaN
      2015
            330.0
                                        NaN
        5. View the values in the 'Cases' column using dot syntax, bracket syntax, loc[] or iloc[].
[11]: df.loc[:,'Cases'].count()
[11]: 52
        6. Use describe() to view the summary statistics for the DataFrame.
[15]: df.describe()
[15]:
              Fiscal Year
                                    Cases
                                                Responses
                                                                Sorties
                                                                         Lives Saved
                                52.000000
      count
                 46.000000
                                                46.000000
                                                              52.000000
                                                                            46.000000
              46296.608696
                            58013.769231
                                             67666.586957
                                                            4339.230769
                                                                           670.956522
      mean
                            13480.714228
      std
              17438.646933
                                            29300.537271
                                                            1334.134847
                                                                           499.839128
                                                            1984.000000
      min
              16456.000000
                            37215.000000
                                             18781.000000
                                                                           169.000000
      25%
             31676.250000
                            46632.750000
                                            33202.750000
                                                            3348.500000
                                                                           281.750000
      50%
              50621.500000
                            55945.500000
                                             81711.500000
                                                            4221.000000
                                                                           383.500000
      75%
             57072.750000
                            69049.750000
                                            88433.750000
                                                            5484.500000
                                                                          1118.750000
      max
             77954.000000
                            86222.000000
                                            110267.000000
                                                            7889.000000
                                                                          1783.000000
             Lives Lost After CG Notification
                                                  Lives Lost Before CG Notification
                                      37.000000
      count
                                                                            46.000000
                                     508.486486
                                                                          1079.956522
      mean
      std
                                     134.761028
                                                                           394.869765
                                     180.000000
                                                                           533.000000
      min
```

25% 50% 75% max		425.000000 492.000000 593.000000 800.000000	751.000000 998.000000 1440.750000 1821.000000
	Total	Lives Unaccounted For	
count	16.000000	0.0	
mean	468.000000	NaN	
std	149.916866	NaN	
min	252.000000	NaN	
25%	336.750000	NaN	
50%	437.500000	NaN	
75%	584.250000	NaN	
max	732.000000	NaN	

7. You can filter for particular values by comparing a colum to a value within the square bracket syntax. This creates a mask on the fly. Lets look at all of the rows whose case count is higher than the mean. You can get this number from the summary statistics above.

[20]: df[df.Cases > df.Cases.mean()]

[20]:	Fiscal Year	Cases	Responses	Sorties	Lives Saved	\
1972	51539.0	60328	72306.0	2633	1389.0	
1973	55107.0	64182	77209.0	2918	1474.0	
1974	59335.0	67692	79950.0	2751	1509.0	
1975	62334.0	70551	81561.0	3024	1254.0	
1976	67179.0	75069	87807.0	2995	1112.0	
1977	74637.0	82601	96021.0	4121	1458.0	
1978	77954.0	86222	100262.0	4386	1556.0	
1979	72517.0	79858	92117.0	5747	949.0	
1980	73345.0	81476	93726.0	6868	1235.0	
1981	71781.0	78951	91432.0	6339	1080.0	
1982	68552.0	75717	87715.0	5675	1359.0	
1983	63980.0	72585	85796.0	5946	1121.0	
1984	57431.0	66073	80698.0	5645	1148.0	
1985	60775.0	70237	88449.0	6497	1076.0	
1986	51765.0	68805	89318.0	4307	475.0	
1987	55998.0	66656	87211.0	5785	1015.0	
1988	54199.0	63446	83616.0	4307	583.0	
1989	52776.0	64027	81862.0	3981	461.0	
1990	53097.0	64971	84033.0	4407	463.0	
1991	52782.0	66409	84872.0	5465	368.0	
1992	53294.0	69856	88388.0	5543	399.0	
1993	53026.0	69784	88147.0	5826	415.0	
1994	53899.0	70337	108758.0	7889	338.0	
1995	49704.0	63679	110267.0	4453	304.0	
2004	32418.0	59995	33460.0	5557	281.0	

	Lives	Lost After C	G Notification	Lives L	ost	Before	CG	Notification	\
1972			NaN					1389.0	•
1973			NaN					1474.0	
1974			NaN					1509.0	
1975			NaN					1254.0	
1976			NaN					1112.0	
1977			NaN					1458.0	
1978			NaN					1556.0	
1979			672.0					1621.0	
1980			586.0					1821.0	
1981			637.0					1717.0	
1982			446.0					1805.0	
1983			640.0					1761.0	
1984			319.0					1467.0	
1985			259.0					1335.0	
1986			180.0					655.0	
1987			576.0					1591.0	
1988			449.0					1032.0	
1989			646.0					1107.0	
1990			622.0					1085.0	
1991			748.0					1116.0	
1992			540.0					939.0	
1993			800.0					1215.0	
1994			593.0					931.0	
1995			468.0					772.0	
2004			502.0					783.0	
	Total	Lives Unacc	ounted For						
1972	NaN	Lives onacc	NaN						
1973	NaN		NaN						
1974	NaN		NaN						
1975	NaN		NaN						
1976	NaN		NaN						
1977	NaN		NaN						
1978	NaN		NaN						
1979	NaN		NaN						
1980	NaN		NaN						
1981	NaN		NaN						
1982	NaN		NaN						
1983	NaN		NaN						
1984	NaN		NaN						
1985	NaN		NaN						
1986	NaN		NaN						
1987	NaN		NaN						
1988	NaN		NaN						
1989	NaN		NaN						

1990	NaN	NaN
1991	NaN	NaN
1992	NaN	NaN
1993	NaN	NaN
1994	NaN	NaN
1995	NaN	NaN
2004	691.0	NaN

9. Now lets create a NumPy array with the same data. Pandas DataFrames have a to_numpy() method. Use this method to create an array named 'np_array'.

```
[29]: import numpy as np
np_array = df.to_numpy()
```

10. Call the shape attribute on the array.

```
[32]: np_array.shape
```

- [32]: (52, 9)
 - 11. Use the array reshape() method to return a 4 x 13 x 9 array (the arguments to the method will be these numbers).

```
[35]: np_array = np_array.reshape(4,13,9)
np_array.shape
```

- [35]: (4, 13, 9)
 - 12. Import the dask.dataframe module aliased as 'dd'

```
[37]: import dask.dataframe as dd
```

13. the dask.dataframe module has a read_csv() method which works in a similar fasion to the Pandas one. Use this method to read the file 'USCG.Search.Rescue.Stats.csv' into a dask DataFrame named 'ddf'

```
[39]: ddf = dd.read_csv('USCG.Search.Rescue.Stats.csv')
```

14. Call the DataFrames std() method.

```
[40]: ddf.std()
```

[40]: Dask Series Structure:

npartitions=1
Cases float64
Total ...
dtype: float64

Dask Name: dataframe-std, 9 tasks

15. Notice that this did not calculate the standard deviation due to dask's use of lazy evaluation. add a .compute() after the std() to compute the result.

[41]: ddf.std().compute()

[41]:	Fiscal Year	17438.646933
	Cases	13480.714228
	Responses	29300.537271
	Sorties	1334.134847
	Lives Saved	499.839128
	Lives Lost After CG Notification	134.761028
	Lives Lost Before CG Notification	394.869765
	Total	149.916866
	Lives Unaccounted For	NaN
	Lives Lost After CG Notification Lives Lost Before CG Notification Total	134.761028 394.869765 149.916866

dtype: float64

[]: