HW5

February 20, 2019

1 HDS 5230 High Performance Computing - HW5

Author: Miao Cai

```
In [ ]: import pandas as pd
        import os
        print("My working directory:\n" + os.getcwd())
        # os.chdir(r"C:\Users\evancarey\Dropbox\Work\SLU\Courses")
```

a. import only the first 100 rows of the patient.csv dataset using Python/pandas

```
In [22]: pt = pd.read_csv('healthcare2/Patient.csv', nrows = 100)
         pt.loc[0:10,:]
```

Out[22]:

	PatientID	FirstName	LastName	State	ZipCode	DateOfBirth	Gender	Race	Income
0	1	Diana	Huddleston	WI	53186	1962-02-27	female	NaN	1076.167979
1	2	Marion	Poston	IL	60527	1859-09-11	male	white	475.781094
2	3	Sandra	Hamby	IL	60126	1946-02-15	female	white	30.747987
3	4	Mildred	Krehbiel	ID	83702	1979-07-27	female	white	160.596425
4	5	Abigail	Flores	PA	19131	1983-02-19	female	?	NaN
5	6	Rusty	Thomas	AL	36107	NaN	male	black	171.378008
6	7	Robert	Alexander	CA	94539	1958-01-11	male	Missing	66.226314
7	8	Krista	Ward	WI	53219	1952-10-31	female	black	15.078950
8	9	Marti	Calabrese	MS	38801	1951-10-06	female	Missing	114.598911
9	10	Jeremy	Liu	CA	95526	1954-10-16	male	white	1081.877157
10	11	Catherine	Tatum	MI	48213	1983-07-12	female	Missing	35.058641

b. Examine the column names and the dtypes of the dataframe

```
In [23]: pt.columns

Out[23]: Index(['PatientID', 'FirstName', 'LastName', 'State', 'ZipCode', 'DateOfBirth',
                'Gender', 'Race', 'Income'],
               dtype='object')

In [24]: pt.dtypes

Out[24]: PatientID        int64
         FirstName       object
         LastName        object
         State           object
         ZipCode          int64
         DateOfBirth     object
         Gender          object
         Race            object
         Income         float64
         dtype: object
```

c. Create a dict of columns names and types using the to_dict() method

```
In [25]: col_types = pt.dtypes.to_dict()
         col_types

Out[25]: {'PatientID': dtype('int64'),
          'FirstName': dtype('O'),
          'LastName': dtype('O'),
          'State': dtype('O'),
          'ZipCode': dtype('int64'),
          'DateOfBirth': dtype('O'),
          'Gender': dtype('O'),
          'Race': dtype('O'),
          'Income': dtype('float64')}
```

d. Decide which columns you can compress by specifying a smaller dtype. For example, the default dtype of an integer is int64, but you may be able to fit hat integer data into the dtype int32, or int16, or uint16 (unsigned integer). It depends on the data! Consider turning the text data into categorical data. Try to make the dataframe as small as reasonably possible.

```
In [26]: pt.info(memory_usage='deep')
         col_types['PatientID']='uint16'
         col_types['FirstName']='category'
         col_types['LastName']='category'
         col_types['State']='category'
         col_types['ZipCode']='uint16'
         col_types['DateOfBirth']='category'
         col_types['Gender']='category'
         col_types['Race']='category'
         col_types['Income']='float32'
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 9 columns):
PatientID      100 non-null int64
FirstName      100 non-null object
LastName       100 non-null object
State          100 non-null object
ZipCode        100 non-null int64
DateOfBirth    92 non-null object
Gender         98 non-null object
Race           97 non-null object
Income         92 non-null float64
dtypes: float64(1), int64(2), object(6)
memory usage: 38.8 KB
```

e. Use the memory_usage(deep=True) dataframe method to calculate large your reduced file is.

```
In [27]: pt_reduced = pd.read_csv('healthcare2/Patient.csv', nrows = 100,
                                  dtype=col_types)
         pt_reduced.info(memory_usage='deep')
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 9 columns):
PatientID      100 non-null uint16
FirstName      100 non-null category
LastName       100 non-null category
State          100 non-null category
ZipCode        100 non-null uint16
DateOfBirth    92 non-null category
Gender         98 non-null category
Race           97 non-null category
Income         92 non-null float32
dtypes: category(6), float32(1), uint16(2)
memory usage: 30.7 KB
```

f. Import the patient.csv dataframe with default datatypes and calculate the memory_usage(deep=True). How much smaller is your reduced dataframe than the full dataframe?

```
In [28]: pt1 = pd.read_csv('healthcare2/Patient.csv')
         pt1.info(memory_usage='deep')
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20000 entries, 0 to 19999
Data columns (total 9 columns):
```

```
PatientID      20000 non-null int64
FirstName      20000 non-null object
LastName       20000 non-null object
State          20000 non-null object
ZipCode        20000 non-null int64
DateOfBirth    19000 non-null object
Gender         19431 non-null object
Race           19144 non-null object
Income         18600 non-null float64
dtypes: float64(1), int64(2), object(6)
memory usage: 7.6 MB
```

```
In [29]: pt1_reduced = pd.read_csv('healthcare2/Patient.csv',dtype=col_types)
         pt1_reduced.info(memory_usage='deep')
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20000 entries, 0 to 19999
Data columns (total 9 columns):
PatientID      20000 non-null uint16
FirstName      20000 non-null category
LastName       20000 non-null category
State          20000 non-null category
ZipCode        20000 non-null uint16
DateOfBirth    19000 non-null category
Gender         19431 non-null category
Race           19144 non-null category
Income         18600 non-null float32
dtypes: category(6), float32(1), uint16(2)
memory usage: 2.8 MB
```

g. Repeat a-f for the OutpatientVisit.csv file.

```
In [30]: op0 = pd.read_csv('healthcare2/OutpatientVisit.csv', nrows = 100)
         op0.loc[0:10,:]
```

Out[30]:

	VisitID	StaffID	PatientID	VisitDate	ICD10_1	ICD10_2	ICD10_3	\
0	1	46	1	2013-08-10	E10621	K269	NaN	
1	2	50	1	2013-12-02	K269	E10621	NaN	
2	3	13	1	2014-06-29	E10621	K269	NaN	
3	4	23	1	2014-09-19	K269	E10621	NaN	
4	5	9	1	2015-05-29	K269	E10621	NaN	
5	6	46	1	2016-05-07	E10621	K269	NaN	
6	7	7	1	2016-10-07	E10621	K269	NaN	
7	8	18	1	2016-11-07	K269	E10621	NaN	
8	9	23	1	2017-01-14	K269	E10621	NaN	
9	10	5	1	2017-01-29	E10621	K269	NaN	
10	11	2	1	2017-06-29	K269	E10621	NaN	

```
    ClinicCode
0           15
1           55
2            1
3            3
4            5
5           15
6           41
7           31
8            3
9           14
10          55
```

```
In [31]: op0.columns
```

```
Out[31]: Index(['VisitID', 'StaffID', 'PatientID', 'VisitDate', 'ICD10_1', 'ICD10_2',
                'ICD10_3', 'ClinicCode'],
               dtype='object')
```

```
In [32]: op0.dtypes
```

```
Out[32]: VisitID         int64
         StaffID         int64
         PatientID       int64
         VisitDate      object
         ICD10_1        object
         ICD10_2        object
         ICD10_3       float64
         ClinicCode      int64
         dtype: object
```

```
In [33]: col_types = op0.dtypes.to_dict()
         col_types
```

```
Out[33]: {'VisitID': dtype('int64'),
          'StaffID': dtype('int64'),
          'PatientID': dtype('int64'),
          'VisitDate': dtype('O'),
          'ICD10_1': dtype('O'),
          'ICD10_2': dtype('O'),
          'ICD10_3': dtype('float64'),
          'ClinicCode': dtype('int64')}
```

```
In [34]: pt.info(memory_usage='deep')
         col_types['VisitID']='uint16'
         col_types['StaffID']='uint16'
         col_types['PatientID']='uint16'
         col_types['VisitDate']='category'
```

```
        col_types['ICD10_1']='category'
        col_types['ICD10_2']='category'
        col_types['ICD10_3']='category'
        col_types['ClinicCode']='uint16'
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 9 columns):
PatientID      100 non-null int64
FirstName      100 non-null object
LastName       100 non-null object
State          100 non-null object
ZipCode        100 non-null int64
DateOfBirth    92 non-null object
Gender         98 non-null object
Race           97 non-null object
Income         92 non-null float64
dtypes: float64(1), int64(2), object(6)
memory usage: 38.8 KB
```

```
In [35]: op0_reduced = pd.read_csv('healthcare2/OutpatientVisit.csv', nrows = 100,
                                   dtype=col_types)
         op0_reduced.info(memory_usage='deep')
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 8 columns):
VisitID       100 non-null uint16
StaffID       100 non-null uint16
PatientID     100 non-null uint16
VisitDate     100 non-null category
ICD10_1       100 non-null category
ICD10_2       63 non-null category
ICD10_3       0 non-null category
ClinicCode    100 non-null uint16
dtypes: category(4), uint16(4)
memory usage: 12.7 KB
```

```
In [36]: op = pd.read_csv('healthcare2/OutpatientVisit.csv')
         op.info(memory_usage='deep')
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 174690 entries, 0 to 174689
Data columns (total 8 columns):
VisitID       174690 non-null int64
StaffID       174690 non-null int64
PatientID     174690 non-null int64
```

```
VisitDate     173252 non-null object
ICD10_1       174690 non-null object
ICD10_2       59785 non-null object
ICD10_3       19362 non-null object
ClinicCode    174690 non-null int64
dtypes: int64(4), object(4)
memory usage: 39.7 MB
```

```
In [37]: op_reduced = pd.read_csv('healthcare2/OutpatientVisit.csv',dtype=col_types)
         op_reduced.info(memory_usage='deep')
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 174690 entries, 0 to 174689
Data columns (total 8 columns):
VisitID       174690 non-null uint16
StaffID       174690 non-null uint16
PatientID     174690 non-null uint16
VisitDate     173252 non-null category
ICD10_1       174690 non-null category
ICD10_2       59785 non-null category
ICD10_3       19362 non-null category
ClinicCode    174690 non-null uint16
dtypes: category(4), uint16(4)
memory usage: 3.7 MB
```