Week 2

May 13, 2019

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*You are currently looking at **version 1.0** of this notebook. To download notebooks and datafiles, as well as get help on Jupyter notebooks in the Coursera platform, visit the Jupyter Notebook FAQ course resource.*

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1 The Series Data Structure

```
In [4]: import pandas as pd
        pd.Series

Out[4]: pandas.core.series.Series

In [5]: animals = ['Tiger', 'Bear', 'Moose']
        pd.Series(animals)

Out[5]: 0    Tiger
        1     Bear
        2    Moose
        dtype: object

In [6]: numbers = [1, 2, 3]
        pd.Series(numbers)

Out[6]: 0    1
        1    2
        2    3
        dtype: int64

In [7]: animals = ['Tiger', 'Bear', None]
        pd.Series(animals)

Out[7]: 0    Tiger
        1     Bear
        2     None
        dtype: object
```

```
In [8]: numbers = [1, 2, None]
        pd.Series(numbers)

Out[8]: 0    1.0
        1    2.0
        2    NaN
        dtype: float64

In [9]: import numpy as np
        np.nan == None

Out[9]: False

In [10]: np.nan == np.nan

Out[10]: False

In [11]: np.isnan(np.nan)

Out[11]: True

In [12]: sports = {'Archery': 'Bhutan',
                   'Golf': 'Scotland',
                   'Sumo': 'Japan',
                   'Taekwondo': 'South Korea'}
         s = pd.Series(sports)
         s

Out[12]: Archery           Bhutan
         Golf            Scotland
         Sumo               Japan
         Taekwondo    South Korea
         dtype: object

In [13]: s.index

Out[13]: Index(['Archery', 'Golf', 'Sumo', 'Taekwondo'], dtype='object')

In [14]: s = pd.Series(['Tiger', 'Bear', 'Moose'], index=['India', 'America', 'Canada'])
         s

Out[14]: India      Tiger
         America     Bear
         Canada     Moose
         dtype: object

In [15]: sports = {'Archery': 'Bhutan',
                   'Golf': 'Scotland',
                   'Sumo': 'Japan',
                   'Taekwondo': 'South Korea'}
         s = pd.Series(sports, index=['Golf', 'Sumo', 'Hockey'])
         s
```

```
Out[15]: Golf      Scotland
         Sumo         Japan
         Hockey         NaN
         dtype: object
```

2 Querying a Series

```
In [16]: sports = {'Archery': 'Bhutan',
                   'Golf': 'Scotland',
                   'Sumo': 'Japan',
                   'Taekwondo': 'South Korea'}
         s = pd.Series(sports)
         s

Out[16]: Archery           Bhutan
         Golf             Scotland
         Sumo                Japan
         Taekwondo    South Korea
         dtype: object

In [17]: s.iloc[3]

Out[17]: 'South Korea'

In [18]: s.loc['Golf']

Out[18]: 'Scotland'

In [19]: s[3]

Out[19]: 'South Korea'

In [20]: s['Golf']

Out[20]: 'Scotland'

In [21]: sports = {99: 'Bhutan',
                   100: 'Scotland',
                   101: 'Japan',
                   102: 'South Korea'}
         s = pd.Series(sports)

In [22]: s[0] #This won't call s.iloc[0] as one might expect, it generates an error instead

         ---------------------------------------------------------------------------

         KeyError                                  Traceback (most recent call last)
```

```
    <ipython-input-22-a5f43d492595> in <module>()
----> 1 s[0] #This won't call s.iloc[0] as one might expect, it generates an error instead


    /opt/conda/lib/python3.6/site-packages/pandas/core/series.py in __getitem__(self, key)
    601         key = com._apply_if_callable(key, self)
    602         try:
--> 603             result = self.index.get_value(self, key)
    604
    605             if not is_scalar(result):


    /opt/conda/lib/python3.6/site-packages/pandas/indexes/base.py in get_value(self, series,
   2167         try:
   2168             return self._engine.get_value(s, k,
-> 2169                                           tz=getattr(series.dtype, 'tz', None))
   2170         except KeyError as e1:
   2171             if len(self) > 0 and self.inferred_type in ['integer', 'boolean']:


    pandas/index.pyx in pandas.index.IndexEngine.get_value (pandas/index.c:3557)()


    pandas/index.pyx in pandas.index.IndexEngine.get_value (pandas/index.c:3240)()


    pandas/index.pyx in pandas.index.IndexEngine.get_loc (pandas/index.c:4279)()


    pandas/src/hashtable_class_helper.pxi in pandas.hashtable.Int64HashTable.get_item (panda


    pandas/src/hashtable_class_helper.pxi in pandas.hashtable.Int64HashTable.get_item (panda


    KeyError: 0
```

```
In [23]: s = pd.Series([100.00, 120.00, 101.00, 3.00])
         s

Out[23]: 0    100.0
         1    120.0
         2    101.0
         3      3.0
         dtype: float64

In [24]: total = 0
         for item in s:
```

```
            total+=item
        print(total)
```

```
324.0
```

In [25]:
```
import numpy as np

total = np.sum(s)
print(total)
```

```
324.0
```

In [26]:
```
#this creates a big series of random numbers
s = pd.Series(np.random.randint(0,1000,10000))
s.head()
```

Out[26]:
```
0    975
1    993
2    255
3    617
4    929
dtype: int64
```

In [27]:
```
len(s)
```

Out[27]:
```
10000
```

In [28]:
```
%%timeit -n 100
summary = 0
for item in s:
    summary+=item
```

```
1.74 ms ś 228 ţs per loop (mean ś std. dev. of 7 runs, 100 loops each)
```

In [29]:
```
%%timeit -n 100
summary = np.sum(s)
```

```
The slowest run took 9.59 times longer than the fastest. This could mean that an intermediate re
283 ţs ś 299 ţs per loop (mean ś std. dev. of 7 runs, 100 loops each)
```

In [30]:
```
s+=2 #adds two to each item in s using broadcasting
s.head()
```

Out[30]:
```
0    977
1    995
2    257
3    619
4    931
dtype: int64
```

```
In [31]: for label, value in s.iteritems():
             s.set_value(label, value+2)
         s.head()

Out[31]: 0    979
         1    997
         2    259
         3    621
         4    933
         dtype: int64

In [32]: %%timeit -n 10
         s = pd.Series(np.random.randint(0,1000,10000))
         for label, value in s.iteritems():
             s.loc[label]= value+2


         ---------------------------------------------------------------------------

         KeyboardInterrupt                         Traceback (most recent call last)

         <ipython-input-32-28c01e28f9f7> in <module>()
     ----> 1 get_ipython().run_cell_magic('timeit', '-n 10', 's = pd.Series(np.random.randint(0,1


         /opt/conda/lib/python3.6/site-packages/IPython/core/interactiveshell.py in run_cell_magi
        2101             magic_arg_s = self.var_expand(line, stack_depth)
        2102             with self.builtin_trap:
     -> 2103                 result = fn(magic_arg_s, cell)
        2104             return result
        2105


         <decorator-gen-61> in timeit(self, line, cell)


         /opt/conda/lib/python3.6/site-packages/IPython/core/magic.py in <lambda>(f, *a, **k)
         185     # but it's overkill for just that one bit of state.
         186     def magic_deco(arg):
     --> 187         call = lambda f, *a, **k: f(*a, **k)
         188
         189         if callable(arg):


         /opt/conda/lib/python3.6/site-packages/IPython/core/magics/execution.py in timeit(self,
        1082                     break
        1083
     -> 1084         all_runs = timer.repeat(repeat, number)
```

```
   1085         best = min(all_runs) / number
   1086         worst = max(all_runs) / number


    /opt/conda/lib/python3.6/timeit.py in repeat(self, repeat, number)
    204         r = []
    205         for i in range(repeat):
--> 206             t = self.timeit(number)
    207             r.append(t)
    208         return r


    /opt/conda/lib/python3.6/site-packages/IPython/core/magics/execution.py in timeit(self,
    158         gc.disable()
    159         try:
--> 160             timing = self.inner(it, self.timer)
    161         finally:
    162             if gcold:


    <magic-timeit> in inner(_it, _timer)


    /opt/conda/lib/python3.6/site-packages/pandas/core/indexing.py in __setitem__(self, key,
    139             key = com._apply_if_callable(key, self.obj)
    140         indexer = self._get_setitem_indexer(key)
--> 141         self._setitem_with_indexer(indexer, value)
    142
    143     def _has_valid_type(self, k, axis):


    /opt/conda/lib/python3.6/site-packages/pandas/core/indexing.py in _setitem_with_indexer(
    578             self.obj._data = self.obj._data.setitem(indexer=indexer,
    579                                                     value=value)
--> 580             self.obj._maybe_update_cacher(clear=True)
    581
    582     def _align_series(self, indexer, ser, multiindex_indexer=False):


    /opt/conda/lib/python3.6/site-packages/pandas/core/generic.py in _maybe_update_cacher(se
   1455         """
   1456
-> 1457         cacher = getattr(self, '_cacher', None)
   1458         if cacher is not None:
   1459             ref = cacher[1]()


    KeyboardInterrupt:
```

```
In [ ]: %%timeit -n 10
        s = pd.Series(np.random.randint(0,1000,10000))
        s+=2
```

```
In [ ]: s = pd.Series([1, 2, 3])
        s.loc['Animal'] = 'Bears'
        s
```

```
In [34]: original_sports = pd.Series({'Archery': 'Bhutan',
                                      'Golf': 'Scotland',
                                      'Sumo': 'Japan',
                                      'Taekwondo': 'South Korea'})
         cricket_loving_countries = pd.Series(['Australia',
                                               'Barbados',
                                               'Pakistan',
                                               'England'],
                                              index=['Cricket',
                                                     'Cricket',
                                                     'Cricket',
                                                     'Cricket'])
         all_countries = original_sports.append(cricket_loving_countries)
```

```
In [35]: original_sports
```

```
Out[35]: Archery           Bhutan
         Golf            Scotland
         Sumo               Japan
         Taekwondo    South Korea
         dtype: object
```

```
In [36]: cricket_loving_countries
```

```
Out[36]: Cricket    Australia
         Cricket     Barbados
         Cricket     Pakistan
         Cricket      England
         dtype: object
```

```
In [37]: all_countries
```

```
Out[37]: Archery           Bhutan
         Golf            Scotland
         Sumo               Japan
         Taekwondo    South Korea
         Cricket        Australia
         Cricket         Barbados
         Cricket         Pakistan
         Cricket          England
         dtype: object
```

```
In [38]: all_countries.loc['Cricket']
```

```
Out[38]: Cricket    Australia
         Cricket     Barbados
         Cricket     Pakistan
         Cricket      England
         dtype: object
```

3 The DataFrame Data Structure

```
In [39]: import pandas as pd
         purchase_1 = pd.Series({'Name': 'Chris',
                                 'Item Purchased': 'Dog Food',
                                 'Cost': 22.50})
         purchase_2 = pd.Series({'Name': 'Kevyn',
                                 'Item Purchased': 'Kitty Litter',
                                 'Cost': 2.50})
         purchase_3 = pd.Series({'Name': 'Vinod',
                                 'Item Purchased': 'Bird Seed',
                                 'Cost': 5.00})
         df = pd.DataFrame([purchase_1, purchase_2, purchase_3], index=['Store 1', 'Store 1', 'S
         df.head()
```

```
Out[39]:          Cost Item Purchased   Name
         Store 1  22.5       Dog Food  Chris
         Store 1   2.5   Kitty Litter  Kevyn
         Store 2   5.0      Bird Seed  Vinod
```

```
In [40]: df.loc['Store 2']
```

```
Out[40]: Cost                      5
         Item Purchased    Bird Seed
         Name                  Vinod
         Name: Store 2, dtype: object
```

```
In [41]: type(df.loc['Store 2'])
```

```
Out[41]: pandas.core.series.Series
```

```
In [42]: df.loc['Store 1']
```

```
Out[42]:          Cost Item Purchased   Name
         Store 1  22.5       Dog Food  Chris
         Store 1   2.5   Kitty Litter  Kevyn
```

```
In [43]: df.loc['Store 1', 'Cost']
```

```
Out[43]: Store 1    22.5
         Store 1     2.5
         Name: Cost, dtype: float64
```

```
In [44]: df.T

Out[44]:                  Store 1       Store 1    Store 2
         Cost                22.5           2.5          5
         Item Purchased  Dog Food  Kitty Litter  Bird Seed
         Name               Chris         Kevyn      Vinod

In [45]: df.T.loc['Cost']

Out[45]: Store 1    22.5
         Store 1     2.5
         Store 2       5
         Name: Cost, dtype: object

In [46]: df['Cost']

Out[46]: Store 1    22.5
         Store 1     2.5
         Store 2     5.0
         Name: Cost, dtype: float64

In [47]: df.loc['Store 1']['Cost']

Out[47]: Store 1    22.5
         Store 1     2.5
         Name: Cost, dtype: float64

In [48]: df.loc[:,['Name', 'Cost']]

Out[48]:           Name  Cost
         Store 1  Chris  22.5
         Store 1  Kevyn   2.5
         Store 2  Vinod   5.0

In [49]: df.drop('Store 1')

Out[49]:          Cost Item Purchased   Name
         Store 2   5.0      Bird Seed  Vinod

In [50]: df

Out[50]:          Cost Item Purchased   Name
         Store 1  22.5       Dog Food  Chris
         Store 1   2.5   Kitty Litter  Kevyn
         Store 2   5.0      Bird Seed  Vinod

In [51]: copy_df = df.copy()
         copy_df = copy_df.drop('Store 1')
         copy_df
```

```
Out[51]:          Cost Item Purchased   Name
         Store 2   5.0      Bird Seed  Vinod

In [53]: copy_df.drop

Out[53]: <bound method NDFrame.drop of          Cost Item Purchased   Name
         Store 2   5.0      Bird Seed  Vinod>

In [54]: del copy_df['Name']
         copy_df

Out[54]:          Cost Item Purchased
         Store 2   5.0      Bird Seed

In [55]: df['Location'] = None
         df

Out[55]:          Cost Item Purchased   Name Location
         Store 1  22.5       Dog Food  Chris     None
         Store 1   2.5   Kitty Litter  Kevyn     None
         Store 2   5.0      Bird Seed  Vinod     None
```

4 Dataframe Indexing and Loading

```
In [56]: costs = df['Cost']
         costs

Out[56]: Store 1    22.5
         Store 1     2.5
         Store 2     5.0
         Name: Cost, dtype: float64

In [57]: costs+=2
         costs

Out[57]: Store 1    24.5
         Store 1     4.5
         Store 2     7.0
         Name: Cost, dtype: float64

In [58]: df

Out[58]:          Cost Item Purchased   Name Location
         Store 1  24.5       Dog Food  Chris     None
         Store 1   4.5   Kitty Litter  Kevyn     None
         Store 2   7.0      Bird Seed  Vinod     None

In [59]: !cat olympics.csv
```

```
0,1,2,3,4,5,6,7,8,9,10,11,12,13,14,15
, Summer,01 !,02 !,03 !,Total, Winter,01 !,02 !,03 !,Total, Games,01 !,02 !,03 !,Combined total
Afghanistană(AFG),13,0,0,2,2,0,0,0,0,0,13,0,0,2,2
Algeriaă(ALG),12,5,2,8,15,3,0,0,0,0,15,5,2,8,15
Argentinaă(ARG),23,18,24,28,70,18,0,0,0,0,41,18,24,28,70
Armeniaă(ARM),5,1,2,9,12,6,0,0,0,0,11,1,2,9,12
Australasiaă(ANZ) [ANZ],2,3,4,5,12,0,0,0,0,0,2,3,4,5,12
Australiaă(AUS) [AUS] [Z],25,139,152,177,468,18,5,3,4,12,43,144,155,181,480
Austriaă(AUT),26,18,33,35,86,22,59,78,81,218,48,77,111,116,304
Azerbaijanăă(AZE),5,6,5,15,26,5,0,0,0,0,10,6,5,15,26
Bahamasă(BAH),15,5,2,5,12,0,0,0,0,0,15,5,2,5,12
Bahrainăă(BRN),8,0,0,1,1,0,0,0,0,0,8,0,0,1,1
Barbadosă(BAR) [BAR],11,0,0,1,1,0,0,0,0,0,11,0,0,1,1
Belarusă(BLR),5,12,24,39,75,6,6,4,5,15,11,18,28,44,90
Belgiumă(BEL),25,37,52,53,142,20,1,1,3,5,45,38,53,56,147
Bermudaă(BER),17,0,0,1,1,7,0,0,0,0,24,0,0,1,1
Bohemiaă(BOH) [BOH] [Z],3,0,1,3,4,0,0,0,0,0,3,0,1,3,4
Botswanaă(BOT),9,0,1,0,1,0,0,0,0,0,9,0,1,0,1
Braziliă(BRA),21,23,30,55,108,7,0,0,0,0,28,23,30,55,108
British West Indiesă(BWI) [BWI],1,0,0,2,2,0,0,0,0,0,1,0,0,2,2
Bulgariaă(BUL) [H],19,51,85,78,214,19,1,2,3,6,38,52,87,81,220
Burundiă(BDI),5,1,0,0,1,0,0,0,0,0,5,1,0,0,1
Cameroonăă(CMR),13,3,1,1,5,1,0,0,0,0,14,3,1,1,5
Canadaă(CAN),25,59,99,121,279,22,62,56,52,170,47,121,155,173,449
Chileă(CHI) [I],22,2,7,4,13,16,0,0,0,0,38,2,7,4,13
Chinaă(CHN) [CHN],9,201,146,126,473,10,12,22,19,53,19,213,168,145,526
Colombiaă(COL),18,2,6,11,19,1,0,0,0,0,19,2,6,11,19
Costa Ricaă(CRC),14,1,1,2,4,6,0,0,0,0,20,1,1,2,4
Ivory Coastăă(CIV) [CIV],12,0,1,0,1,0,0,0,0,0,12,0,1,0,1
Croatiaă(CRO),6,6,7,10,23,7,4,6,1,11,13,10,13,11,34
Cubaă(CUB) [Z],19,72,67,70,209,0,0,0,0,0,19,72,67,70,209
Cyprusă(CYP),9,0,1,0,1,10,0,0,0,0,19,0,1,0,1
Czech Republicăă(CZE) [CZE],5,14,15,15,44,6,7,9,8,24,11,21,24,23,68
Czechoslovakiaă(TCH) [TCH],16,49,49,45,143,16,2,8,15,25,32,51,57,60,168
Denmarkăă(DEN) [Z],26,43,68,68,179,13,0,1,0,1,39,43,69,68,180
Djiboutiăă(DJI) [B],7,0,0,1,1,0,0,0,0,0,7,0,0,1,1
Dominican Republicăă(DOM),13,3,2,1,6,0,0,0,0,0,13,3,2,1,6
Ecuadorăă(ECU),13,1,1,0,2,0,0,0,0,0,13,1,1,0,2
Egyptăă(EGY) [EGY] [Z],21,7,9,10,26,1,0,0,0,0,22,7,9,10,26
Eritreaă(ERI),4,0,0,1,1,0,0,0,0,0,4,0,0,1,1
Estoniaă(EST),11,9,9,15,33,9,4,2,1,7,20,13,11,16,40
Ethiopiaă(ETH),12,21,7,17,45,2,0,0,0,0,14,21,7,17,45
Finlandăă(FIN),24,101,84,117,302,22,42,62,57,161,46,143,146,174,463
Franceă(FRA) [O] [P] [Z],27,202,223,246,671,22,31,31,47,109,49,233,254,293,780
Gabonăă(GAB),9,0,1,0,1,0,0,0,0,0,9,0,1,0,1
Georgiaă(GEO),5,6,5,14,25,6,0,0,0,0,11,6,5,14,25
Germanyă(GER) [GER] [Z],15,174,182,217,573,11,78,78,53,209,26,252,260,270,782
United Team of Germanyă(EUA) [EUA],3,28,54,36,118,3,8,6,5,19,6,36,60,41,137
```

East Germanyă(GDR) [GDR],5,153,129,127,409,6,39,36,35,110,11,192,165,162,519
West Germanyă(FRG) [FRG],5,56,67,81,204,6,11,15,13,39,11,67,82,94,243
Ghanaă(GHA) [GHA],13,0,1,3,4,1,0,0,0,0,14,0,1,3,4
Great Britainăă(GBR) [GBR] [Z],27,236,272,272,780,22,10,4,12,26,49,246,276,284,806
Greeceă(GRE) [Z],27,30,42,39,111,18,0,0,0,0,45,30,42,39,111
Grenadaă(GRN),8,1,0,0,1,0,0,0,0,0,8,1,0,0,1
Guatemalaă(GUA),13,0,1,0,1,1,0,0,0,0,14,0,1,0,1
Guyanaă(GUY) [GUY],16,0,0,1,1,0,0,0,0,0,16,0,0,1,1
Haitiă(HAI) [J],14,0,1,1,2,0,0,0,0,0,14,0,1,1,2
Hong Kongă(HKG) [HKG],15,1,1,1,3,4,0,0,0,0,19,1,1,1,3
Hungaryă(HUN),25,167,144,165,476,22,0,2,4,6,47,167,146,169,482
Icelandă(ISL),19,0,2,2,4,17,0,0,0,0,36,0,2,2,4
Indiaă(IND) [F],23,9,6,11,26,9,0,0,0,0,32,9,6,11,26
Indonesiaă(INA),14,6,10,11,27,0,0,0,0,0,14,6,10,11,27
Iranăă(IRI) [K],15,15,20,25,60,10,0,0,0,0,25,15,20,25,60
Iraqă(IRQ),13,0,0,1,1,0,0,0,0,0,13,0,0,1,1
Irelandă(IRL),20,9,8,12,29,6,0,0,0,0,26,9,8,12,29
Israelă(ISR),15,1,1,5,7,6,0,0,0,0,21,1,1,5,7
Italyă(ITA) [M] [S],26,198,166,185,549,22,37,34,43,114,48,235,200,228,663
Jamaicaă(JAM) [JAM],16,17,30,20,67,7,0,0,0,0,23,17,30,20,67
Japanăă(JPN),21,130,126,142,398,20,10,17,18,45,41,140,143,160,443
Kazakhstanăă(KAZ),5,16,17,19,52,6,1,3,3,7,11,17,20,22,59
Kenyaă(KEN),13,25,32,29,86,3,0,0,0,0,16,25,32,29,86
North Koreaă(PRK),9,14,12,21,47,8,0,1,1,2,17,14,13,22,49
South Koreaă(KOR),16,81,82,80,243,17,26,17,10,53,33,107,99,90,296
Kuwaităă(KUW),12,0,0,2,2,0,0,0,0,0,12,0,0,2,2
Kyrgyzstanăă(KGZ),5,0,1,2,3,6,0,0,0,0,11,0,1,2,3
Latviaă(LAT),10,3,11,5,19,10,0,4,3,7,20,3,15,8,26
Lebanonăă(LIB),16,0,2,2,4,16,0,0,0,0,32,0,2,2,4
Liechtensteinăă(LIE),16,0,0,0,0,18,2,2,5,9,34,2,2,5,9
Lithuaniaă(LTU),8,6,5,10,21,8,0,0,0,0,16,6,5,10,21
Luxembourgăă(LUX) [O],22,1,1,0,2,8,0,2,0,2,30,1,3,0,4
Macedoniaă(MKD),5,0,0,1,1,5,0,0,0,0,10,0,0,1,1
Malaysiaă(MAS) [MAS],12,0,3,3,6,0,0,0,0,0,12,0,3,3,6
Mauritiusăă(MRI),8,0,0,1,1,0,0,0,0,0,8,0,0,1,1
Mexicoă(MEX),22,13,21,28,62,8,0,0,0,0,30,13,21,28,62
Moldovaă(MDA),5,0,2,5,7,6,0,0,0,0,11,0,2,5,7
Mongoliaă(MGL),12,2,9,13,24,13,0,0,0,0,25,2,9,13,24
Montenegroă(MNE),2,0,1,0,1,2,0,0,0,0,4,0,1,0,1
Moroccoă(MAR),13,6,5,11,22,6,0,0,0,0,19,6,5,11,22
Mozambiqueă(MOZ),9,1,0,1,2,0,0,0,0,0,9,1,0,1,2
Namibiaă(NAM),6,0,4,0,4,0,0,0,0,0,6,0,4,0,4
Netherlandsăă(NED) [Z],25,77,85,104,266,20,37,38,35,110,45,114,123,139,376
Netherlands Antillesăă(AHO) [AHO] [I],13,0,1,0,1,2,0,0,0,0,15,0,1,0,1
New Zealandăă(NZL) [NZL],22,42,18,39,99,15,0,1,0,1,37,42,19,39,100
Nigeră(NIG),11,0,0,1,1,0,0,0,0,0,11,0,0,1,1
Nigeriaă(NGR),15,3,8,12,23,0,0,0,0,0,15,3,8,12,23
Norwayă(NOR) [Q],24,56,49,43,148,22,118,111,100,329,46,174,160,143,477

Pakistană(PAK),16,3,3,4,10,2,0,0,0,0,18,3,3,4,10
Panamaă(PAN),16,1,0,2,3,0,0,0,0,0,16,1,0,2,3
Paraguayă(PAR),11,0,1,0,1,1,0,0,0,0,12,0,1,0,1
Peruă(PER) [L],17,1,3,0,4,2,0,0,0,0,19,1,3,0,4
Philippinesă(PHI),20,0,2,7,9,4,0,0,0,0,24,0,2,7,9
Polandă(POL),20,64,82,125,271,22,6,7,7,20,42,70,89,132,291
Portugală(POR),23,4,8,11,23,7,0,0,0,0,30,4,8,11,23
Puerto Ricoă(PUR),17,0,2,6,8,6,0,0,0,0,23,0,2,6,8
Qatară(QAT),8,0,0,4,4,0,0,0,0,0,8,0,0,4,4
Romaniaă(ROU),20,88,94,119,301,20,0,0,1,1,40,88,94,120,302
Russiaă(RUS) [RUS],5,132,121,142,395,6,49,40,35,124,11,181,161,177,519
Russian Empireă(RU1) [RU1],3,1,4,3,8,0,0,0,0,0,3,1,4,3,8
Soviet Unionă(URS) [URS],9,395,319,296,1010,9,78,57,59,194,18,473,376,355,1204
Unified Teamă(EUN) [EUN],1,45,38,29,112,1,9,6,8,23,2,54,44,37,135
Saudi Arabiaă(KSA),10,0,1,2,3,0,0,0,0,0,10,0,1,2,3
Senegală(SEN),13,0,1,0,1,5,0,0,0,0,18,0,1,0,1
Serbiaă(SRB) [SRB],3,1,2,4,7,2,0,0,0,0,5,1,2,4,7
Serbia and Montenegroă(SCG) [SCG],3,2,4,3,9,3,0,0,0,0,6,2,4,3,9
Singaporeă(SIN),15,0,2,2,4,0,0,0,0,0,15,0,2,2,4
Slovakiaă(SVK) [SVK],5,7,9,8,24,6,2,2,1,5,11,9,11,9,29
Sloveniaă(SLO),6,4,6,9,19,7,2,4,9,15,13,6,10,18,34
South Africaă(RSA),18,23,26,27,76,6,0,0,0,0,24,23,26,27,76
Spainăă(ESP) [Z],22,37,59,35,131,19,1,0,1,2,41,38,59,36,133
Sri Lankaă(SRI) [SRI],16,0,2,0,2,0,0,0,0,0,16,0,2,0,2
Sudană(SUD),11,0,1,0,1,0,0,0,0,0,11,0,1,0,1
Surinameă(SUR) [E],11,1,0,1,2,0,0,0,0,0,11,1,0,1,2
Swedenă(SWE) [Z],26,143,164,176,483,22,50,40,54,144,48,193,204,230,627
Switzerlandă(SUI),27,47,73,65,185,22,50,40,48,138,49,97,113,113,323
Syriaă(SYR),12,1,1,1,3,0,0,0,0,0,12,1,1,1,3
Chinese Taipeiă(TPE) [TPE] [TPE2],13,2,7,12,21,11,0,0,0,0,24,2,7,12,21
Tajikistană(TJK),5,0,1,2,3,4,0,0,0,0,9,0,1,2,3
Tanzaniaă(TAN) [TAN],12,0,2,0,2,0,0,0,0,0,12,0,2,0,2
Thailandă(THA),15,7,6,11,24,3,0,0,0,0,18,7,6,11,24
Togoă(TOG),9,0,0,1,1,1,0,0,0,0,10,0,0,1,1
Tongaă(TGA),8,0,1,0,1,1,0,0,0,0,9,0,1,0,1
Trinidad and Tobagoă(TRI) [TRI],16,2,5,11,18,3,0,0,0,0,19,2,5,11,18
Tunisiaă(TUN),13,3,3,4,10,0,0,0,0,0,13,3,3,4,10
Turkeyă(TUR),21,39,25,24,88,16,0,0,0,0,37,39,25,24,88
Ugandaă(UGA),14,2,3,2,7,0,0,0,0,0,14,2,3,2,7
Ukraineă(UKR),5,33,27,55,115,6,2,1,4,7,11,35,28,59,122
United Arab Emiratesă(UAE),8,1,0,0,1,0,0,0,0,0,8,1,0,0,1
United Statesă(USA) [P] [Q] [R] [Z],26,976,757,666,2399,22,96,102,84,282,48,1072,859,750,2681
Uruguayă(URU),20,2,2,6,10,1,0,0,0,0,21,2,2,6,10
Uzbekistană(UZB),5,5,5,10,20,6,1,0,0,1,11,6,5,10,21
Venezuelaă(VEN),17,2,2,8,12,4,0,0,0,0,21,2,2,8,12
Vietnamă(VIE),14,0,2,0,2,0,0,0,0,0,14,0,2,0,2
Virgin Islandsă(ISV),11,0,1,0,1,7,0,0,0,0,18,0,1,0,1
Yugoslaviaă(YUG) [YUG],16,26,29,28,83,14,0,3,1,4,30,26,32,29,87

```
Independent Olympic Participantsă(IOP) [IOP],1,0,1,2,3,0,0,0,0,0,1,0,1,2,3
Zambiaă(ZAM) [ZAM],12,0,1,1,2,0,0,0,0,0,12,0,1,1,2
Zimbabweă(ZIM) [ZIM],12,3,4,1,8,1,0,0,0,0,13,3,4,1,8
Mixed teamă(ZZX) [ZZX],3,8,5,4,17,0,0,0,0,0,3,8,5,4,17
Totals,27,4809,4775,5130,14714,22,959,958,948,2865,49,5768,5733,6078,17579
```

In [60]:
```
df = pd.read_csv('olympics.csv')
df.head()
```

Out[60]:

	0	1	2	3	4	5	6	7	8
0	NaN	Summer	01 !	02 !	03 !	Total	Winter	01 !	02 !
1	Afghanistană(AFG)	13	0	0	2	2	0	0	0
2	Algeriaă(ALG)	12	5	2	8	15	3	0	0
3	Argentinaă(ARG)	23	18	24	28	70	18	0	0
4	Armeniaă(ARM)	5	1	2	9	12	6	0	0

	9	10	11	12	13	14	15
0	03 !	Total	Games	01 !	02 !	03 !	Combined total
1	0	0	13	0	0	2	2
2	0	0	15	5	2	8	15
3	0	0	41	18	24	28	70
4	0	0	11	1	2	9	12

In [61]:
```
df = pd.read_csv('olympics.csv', index_col = 0, skiprows=1)
df.head()
```

Out[61]:

	Summer	01 !	02 !	03 !	Total	Winter	01 !.1
Afghanistană(AFG)	13	0	0	2	2	0	0
Algeriaă(ALG)	12	5	2	8	15	3	0
Argentinaă(ARG)	23	18	24	28	70	18	0
Armeniaă(ARM)	5	1	2	9	12	6	0
Australasiaă(ANZ) [ANZ]	2	3	4	5	12	0	0

	02 !.1	03 !.1	Total.1	Games	01 !.2	02 !.2
Afghanistană(AFG)	0	0	0	13	0	0
Algeriaă(ALG)	0	0	0	15	5	2
Argentinaă(ARG)	0	0	0	41	18	24
Armeniaă(ARM)	0	0	0	11	1	2
Australasiaă(ANZ) [ANZ]	0	0	0	2	3	4

	03 !.2	Combined total
Afghanistană(AFG)	2	2
Algeriaă(ALG)	8	15
Argentinaă(ARG)	28	70
Armeniaă(ARM)	9	12
Australasiaă(ANZ) [ANZ]	5	12

In [62]:
```
df.columns
```

```
Out[62]: Index([' Summer', '01 !', '02 !', '03 !', 'Total', ' Winter', '01 !.1',
                '02 !.1', '03 !.1', 'Total.1', ' Games', '01 !.2', '02 !.2', '03 !.2',
                'Combined total'],
               dtype='object')
```

```
In [63]: for col in df.columns:
             if col[:2]=='01':
                 df.rename(columns={col:'Gold' + col[4:]}, inplace=True)
             if col[:2]=='02':
                 df.rename(columns={col:'Silver' + col[4:]}, inplace=True)
             if col[:2]=='03':
                 df.rename(columns={col:'Bronze' + col[4:]}, inplace=True)
             if col[:1]=='':
                 df.rename(columns={col:'#' + col[1:]}, inplace=True)

         df.head()
```

Out[63]:

	# Summer	Gold	Silver	Bronze	Total	# Winter	\
Afghanistană(AFG)	13	0	0	2	2	0	
Algeriaă(ALG)	12	5	2	8	15	3	
Argentinaă(ARG)	23	18	24	28	70	18	
Armeniaă(ARM)	5	1	2	9	12	6	
Australasiaă(ANZ) [ANZ]	2	3	4	5	12	0	

	Gold.1	Silver.1	Bronze.1	Total.1	# Games	Gold.2	\
Afghanistană(AFG)	0	0	0	0	13	0	
Algeriaă(ALG)	0	0	0	0	15	5	
Argentinaă(ARG)	0	0	0	0	41	18	
Armeniaă(ARM)	0	0	0	0	11	1	
Australasiaă(ANZ) [ANZ]	0	0	0	0	2	3	

	Silver.2	Bronze.2	Combined total
Afghanistană(AFG)	0	2	2
Algeriaă(ALG)	2	8	15
Argentinaă(ARG)	24	28	70
Armeniaă(ARM)	2	9	12
Australasiaă(ANZ) [ANZ]	4	5	12

5 Querying a DataFrame

```
In [64]: df['Gold'] > 0
```

```
Out[64]: Afghanistană(AFG)                    False
         Algeriaă(ALG)                         True
         Argentinaă(ARG)                       True
         Armeniaă(ARM)                         True
         Australasiaă(ANZ) [ANZ]               True
         Australiaă(AUS) [AUS] [Z]             True
```

Austriaă(AUT)	True
Azerbaijană(AZE)	True
Bahamasă(BAH)	True
Bahrainăă(BRN)	False
Barbadosă(BAR) [BAR]	False
Belarusă(BLR)	True
Belgiumă(BEL)	True
Bermudaă(BER)	False
Bohemiaă(BOH) [BOH] [Z]	False
Botswanaă(BOT)	False
Brazilă(BRA)	True
British West Indiesă(BWI) [BWI]	False
Bulgariaă(BUL) [H]	True
Burundiă(BDI)	True
Cameroonă(CMR)	True
Canadaă(CAN)	True
Chileă(CHI) [I]	True
Chinaă(CHN) [CHN]	True
Colombiaă(COL)	True
Costa Ricaă(CRC)	True
Ivory Coastă(CIV) [CIV]	False
Croatiaă(CRO)	True
Cubaă(CUB) [Z]	True
Cyprusă(CYP)	False
	...
Sri Lankaă(SRI) [SRI]	False
Sudanăă(SUD)	False
Surinameă(SUR) [E]	True
Swedenă(SWE) [Z]	True
Switzerlandă(SUI)	True
Syriaă(SYR)	True
Chinese Taipeiă(TPE) [TPE] [TPE2]	True
Tajikistanăă(TJK)	False
Tanzaniaă(TAN) [TAN]	False
Thailandă(THA)	True
Togoă(TOG)	False
Tongaă(TGA)	False
Trinidad and Tobagoă(TRI) [TRI]	True
Tunisiaă(TUN)	True
Turkeyă(TUR)	True
Ugandaă(UGA)	True
Ukraineă(UKR)	True
United Arab Emiratesă(UAE)	True
United Statesă(USA) [P] [Q] [R] [Z]	True
Uruguayă(URU)	True
Uzbekistanăă(UZB)	True
Venezuelaă(VEN)	True
Vietnamă(VIE)	False

```
Virgin Islandsă(ISV)                              False
Yugoslaviaă(YUG) [YUG]                             True
Independent Olympic Participantsă(IOP) [IOP]      False
Zambiaă(ZAM) [ZAM]                                False
Zimbabweă(ZIM) [ZIM]                               True
Mixed teamă(ZZX) [ZZX]                             True
Totals                                             True
Name: Gold, dtype: bool
```

In [65]:
```
only_gold = df.where(df['Gold'] > 0)
only_gold.head()
```

Out[65]:

	# Summer	Gold	Silver	Bronze	Total	# Winter	\
Afghanistană(AFG)	NaN	NaN	NaN	NaN	NaN	NaN	
Algeriaă(ALG)	12.0	5.0	2.0	8.0	15.0	3.0	
Argentinaă(ARG)	23.0	18.0	24.0	28.0	70.0	18.0	
Armeniaă(ARM)	5.0	1.0	2.0	9.0	12.0	6.0	
Australasiaă(ANZ) [ANZ]	2.0	3.0	4.0	5.0	12.0	0.0	

	Gold.1	Silver.1	Bronze.1	Total.1	# Games	Gold.2	\
Afghanistană(AFG)	NaN	NaN	NaN	NaN	NaN	NaN	
Algeriaă(ALG)	0.0	0.0	0.0	0.0	15.0	5.0	
Argentinaă(ARG)	0.0	0.0	0.0	0.0	41.0	18.0	
Armeniaă(ARM)	0.0	0.0	0.0	0.0	11.0	1.0	
Australasiaă(ANZ) [ANZ]	0.0	0.0	0.0	0.0	2.0	3.0	

	Silver.2	Bronze.2	Combined total
Afghanistană(AFG)	NaN	NaN	NaN
Algeriaă(ALG)	2.0	8.0	15.0
Argentinaă(ARG)	24.0	28.0	70.0
Armeniaă(ARM)	2.0	9.0	12.0
Australasiaă(ANZ) [ANZ]	4.0	5.0	12.0

In [66]:
```
only_gold['Gold'].count()
```

Out[66]: 100

In [67]:
```
df['Gold'].count()
```

Out[67]: 147

In [68]:
```
only_gold = only_gold.dropna()
only_gold.head()
```

Out[68]:

	# Summer	Gold	Silver	Bronze	Total	# Winter	\
Algeriaă(ALG)	12.0	5.0	2.0	8.0	15.0	3.0	
Argentinaă(ARG)	23.0	18.0	24.0	28.0	70.0	18.0	
Armeniaă(ARM)	5.0	1.0	2.0	9.0	12.0	6.0	
Australasiaă(ANZ) [ANZ]	2.0	3.0	4.0	5.0	12.0	0.0	

```
Australiaă(AUS) [AUS] [Z]      25.0  139.0   152.0   177.0  468.0      18.0

                            Gold.1  Silver.1  Bronze.1  Total.1  # Games  \
Algeriaă(ALG)                  0.0       0.0       0.0      0.0     15.0
Argentinaă(ARG)                0.0       0.0       0.0      0.0     41.0
Armeniaă(ARM)                  0.0       0.0       0.0      0.0     11.0
Australasiaă(ANZ) [ANZ]        0.0       0.0       0.0      0.0      2.0
Australiaă(AUS) [AUS] [Z]      5.0       3.0       4.0     12.0     43.0

                            Gold.2  Silver.2  Bronze.2  Combined total
Algeriaă(ALG)                  5.0       2.0       8.0            15.0
Argentinaă(ARG)               18.0      24.0      28.0            70.0
Armeniaă(ARM)                  1.0       2.0       9.0            12.0
Australasiaă(ANZ) [ANZ]        3.0       4.0       5.0            12.0
Australiaă(AUS) [AUS] [Z]    144.0     155.0     181.0           480.0
```

In [69]:
```
only_gold = df[df['Gold'] > 0]
only_gold.head()
```

Out[69]:
```
                            # Summer  Gold  Silver  Bronze  Total  # Winter  \
Algeriaă(ALG)                     12     5       2       8     15         3
Argentinaă(ARG)                   23    18      24      28     70        18
Armeniaă(ARM)                      5     1       2       9     12         6
Australasiaă(ANZ) [ANZ]            2     3       4       5     12         0
Australiaă(AUS) [AUS] [Z]         25   139     152     177    468        18

                            Gold.1  Silver.1  Bronze.1  Total.1  # Games  \
Algeriaă(ALG)                    0         0         0        0       15
Argentinaă(ARG)                  0         0         0        0       41
Armeniaă(ARM)                    0         0         0        0       11
Australasiaă(ANZ) [ANZ]          0         0         0        0        2
Australiaă(AUS) [AUS] [Z]        5         3         4       12       43

                            Gold.2  Silver.2  Bronze.2  Combined total
Algeriaă(ALG)                    5         2         8              15
Argentinaă(ARG)                 18        24        28              70
Armeniaă(ARM)                    1         2         9              12
Australasiaă(ANZ) [ANZ]          3         4         5              12
Australiaă(AUS) [AUS] [Z]      144       155       181             480
```

In [70]:
```
len(df[(df['Gold'] > 0) | (df['Gold.1'] > 0)])
```

Out[70]:
```
101
```

In [71]:
```
df[(df['Gold.1'] > 0) & (df['Gold'] == 0)]
```

Out[71]:
```
                          # Summer  Gold  Silver  Bronze  Total  # Winter  Gold.1  \
Liechtensteină(LIE)             16     0       0       0      0        18       2
```

```
                        Silver.1  Bronze.1  Total.1  # Games  Gold.2  Silver.2  \
Liechtensteină(LIE)            2         5        9       34       2         2

                        Bronze.2  Combined total
Liechtensteină(LIE)            5               9
```

6 Indexing Dataframes

In [72]:
```
df.head()
```

Out[72]:
```
                             # Summer  Gold  Silver  Bronze  Total  # Winter  \
Afghanistană(AFG)                  13     0       0       2      2         0
Algeriaă(ALG)                      12     5       2       8     15         3
Argentinaă(ARG)                    23    18      24      28     70        18
Armeniaă(ARM)                       5     1       2       9     12         6
Australasiaă(ANZ) [ANZ]             2     3       4       5     12         0

                             Gold.1  Silver.1  Bronze.1  Total.1  # Games  Gold.2  \
Afghanistană(AFG)                 0         0         0        0       13       0
Algeriaă(ALG)                     0         0         0        0       15       5
Argentinaă(ARG)                   0         0         0        0       41      18
Armeniaă(ARM)                     0         0         0        0       11       1
Australasiaă(ANZ) [ANZ]           0         0         0        0        2       3

                             Silver.2  Bronze.2  Combined total
Afghanistană(AFG)                   0         2               2
Algeriaă(ALG)                       2         8              15
Argentinaă(ARG)                    24        28              70
Armeniaă(ARM)                       2         9              12
Australasiaă(ANZ) [ANZ]             4         5              12
```

In [73]:
```
df['country'] = df.index
df = df.set_index('Gold')
df.head()
```

Out[73]:
```
      # Summer  Silver  Bronze  Total  # Winter  Gold.1  Silver.1  Bronze.1  \
Gold
0           13       0       2      2         0       0         0         0
5           12       2       8     15         3       0         0         0
18          23      24      28     70        18       0         0         0
1            5       2       9     12         6       0         0         0
3            2       4       5     12         0       0         0         0

      Total.1  # Games  Gold.2  Silver.2  Bronze.2  Combined total  \
Gold
0           0       13       0         0         2               2
5           0       15       5         2         8              15
18          0       41      18        24        28              70
```

```
1           0       11       1         2         9              12
3           0        2       3         4         5              12

                       country
Gold
0           Afghanistană(AFG)
5               Algeriaă(ALG)
18            Argentinaă(ARG)
1               Armeniaă(ARM)
3     Australasiaă(ANZ) [ANZ]
```

In [74]:
```
df = df.reset_index()
df.head()
```

Out[74]:

	Gold	# Summer	Silver	Bronze	Total	# Winter	Gold.1	Silver.1	\
0	0	13	0	2	2	0	0	0	
1	5	12	2	8	15	3	0	0	
2	18	23	24	28	70	18	0	0	
3	1	5	2	9	12	6	0	0	
4	3	2	4	5	12	0	0	0	

	Bronze.1	Total.1	# Games	Gold.2	Silver.2	Bronze.2	Combined total	\
0	0	0	13	0	0	2	2	
1	0	0	15	5	2	8	15	
2	0	0	41	18	24	28	70	
3	0	0	11	1	2	9	12	
4	0	0	2	3	4	5	12	

	country
0	Afghanistană(AFG)
1	Algeriaă(ALG)
2	Argentinaă(ARG)
3	Armeniaă(ARM)
4	Australasiaă(ANZ) [ANZ]

In [75]:
```
df = pd.read_csv('census.csv')
df.head()
```

Out[75]:

	SUMLEV	REGION	DIVISION	STATE	COUNTY	STNAME	CTYNAME	\
0	40	3	6	1	0	Alabama	Alabama	
1	50	3	6	1	1	Alabama	Autauga County	
2	50	3	6	1	3	Alabama	Baldwin County	
3	50	3	6	1	5	Alabama	Barbour County	
4	50	3	6	1	7	Alabama	Bibb County	

	CENSUS2010POP	ESTIMATESBASE2010	POPESTIMATE2010	...	\
0	4779736	4780127	4785161	...	
1	54571	54571	54660	...	
2	182265	182265	183193	...	

```
3          27457             27457            27341      ...
4          22915             22919            22861      ...

   RDOMESTICMIG2011  RDOMESTICMIG2012  RDOMESTICMIG2013  RDOMESTICMIG2014  \
0          0.002295         -0.193196          0.381066          0.582002
1          7.242091         -2.915927         -3.012349          2.265971
2         14.832960         17.647293         21.845705         19.243287
3         -4.728132         -2.500690         -7.056824         -3.904217
4         -5.527043         -5.068871         -6.201001         -0.177537

   RDOMESTICMIG2015  RNETMIG2011  RNETMIG2012  RNETMIG2013  RNETMIG2014  \
0         -0.467369     1.030015     0.826644     1.383282     1.724718
1         -2.530799     7.606016    -2.626146    -2.722002     2.592270
2         17.197872    15.844176    18.559627    22.727626    20.317142
3        -10.543299    -4.874741    -2.758113    -7.167664    -3.978583
4          0.177258    -5.088389    -4.363636    -5.403729     0.754533

   RNETMIG2015
0     0.712594
1    -2.187333
2    18.293499
3   -10.543299
4     1.107861

[5 rows x 100 columns]
```

```
In [76]: df['SUMLEV'].unique()
```

```
Out[76]: array([40, 50])
```

```
In [77]: df=df[df['SUMLEV'] == 50]
         df.head()
```

```
Out[77]:    SUMLEV  REGION  DIVISION  STATE  COUNTY   STNAME         CTYNAME  \
         1      50       3         6      1       1  Alabama  Autauga County
         2      50       3         6      1       3  Alabama  Baldwin County
         3      50       3         6      1       5  Alabama  Barbour County
         4      50       3         6      1       7  Alabama     Bibb County
         5      50       3         6      1       9  Alabama   Blount County

            CENSUS2010POP  ESTIMATESBASE2010  POPESTIMATE2010     ...       \
         1          54571              54571            54660     ...
         2         182265             182265           183193     ...
         3          27457              27457            27341     ...
         4          22915              22919            22861     ...
         5          57322              57322            57373     ...

            RDOMESTICMIG2011  RDOMESTICMIG2012  RDOMESTICMIG2013  RDOMESTICMIG2014  \
         1          7.242091         -2.915927         -3.012349          2.265971
```

```
2         14.832960         17.647293         21.845705         19.243287
3         -4.728132         -2.500690         -7.056824         -3.904217
4         -5.527043         -5.068871         -6.201001         -0.177537
5          1.807375         -1.177622         -1.748766         -2.062535

   RDOMESTICMIG2015  RNETMIG2011  RNETMIG2012  RNETMIG2013  RNETMIG2014  \
1         -2.530799     7.606016    -2.626146    -2.722002     2.592270
2         17.197872    15.844176    18.559627    22.727626    20.317142
3        -10.543299    -4.874741    -2.758113    -7.167664    -3.978583
4          0.177258    -5.088389    -4.363636    -5.403729     0.754533
5         -1.369970     1.859511    -0.848580    -1.402476    -1.577232

   RNETMIG2015
1    -2.187333
2    18.293499
3   -10.543299
4     1.107861
5    -0.884411

[5 rows x 100 columns]
```

In [78]:
```
columns_to_keep = ['STNAME',
                   'CTYNAME',
                   'BIRTHS2010',
                   'BIRTHS2011',
                   'BIRTHS2012',
                   'BIRTHS2013',
                   'BIRTHS2014',
                   'BIRTHS2015',
                   'POPESTIMATE2010',
                   'POPESTIMATE2011',
                   'POPESTIMATE2012',
                   'POPESTIMATE2013',
                   'POPESTIMATE2014',
                   'POPESTIMATE2015']
df = df[columns_to_keep]
df.head()
```

Out[78]:
```
    STNAME         CTYNAME  BIRTHS2010  BIRTHS2011  BIRTHS2012  BIRTHS2013  \
1  Alabama  Autauga County         151         636         615         574
2  Alabama  Baldwin County         517        2187        2092        2160
3  Alabama  Barbour County          70         335         300         283
4  Alabama     Bibb County          44         266         245         259
5  Alabama   Blount County         183         744         710         646

   BIRTHS2014  BIRTHS2015  POPESTIMATE2010  POPESTIMATE2011  POPESTIMATE2012  \
1         623         600            54660            55253            55175
2        2186        2240           183193           186659           190396
```

```
3         260         269            27341            27226            27159
4         247         253            22861            22733            22642
5         618         603            57373            57711            57776

   POPESTIMATE2013  POPESTIMATE2014  POPESTIMATE2015
1            55038            55290            55347
2           195126           199713           203709
3            26973            26815            26489
4            22512            22549            22583
5            57734            57658            57673
```

In [79]:
```
df = df.set_index(['STNAME', 'CTYNAME'])
df.head()
```

Out[79]:

STNAME	CTYNAME	BIRTHS2010	BIRTHS2011	BIRTHS2012	BIRTHS2013
Alabama	Autauga County	151	636	615	574
	Baldwin County	517	2187	2092	2160
	Barbour County	70	335	300	283
	Bibb County	44	266	245	259
	Blount County	183	744	710	646

STNAME	CTYNAME	BIRTHS2014	BIRTHS2015	POPESTIMATE2010
Alabama	Autauga County	623	600	54660
	Baldwin County	2186	2240	183193
	Barbour County	260	269	27341
	Bibb County	247	253	22861
	Blount County	618	603	57373

STNAME	CTYNAME	POPESTIMATE2011	POPESTIMATE2012	POPESTIMATE2013
Alabama	Autauga County	55253	55175	55038
	Baldwin County	186659	190396	195126
	Barbour County	27226	27159	26973
	Bibb County	22733	22642	22512
	Blount County	57711	57776	57734

STNAME	CTYNAME	POPESTIMATE2014	POPESTIMATE2015
Alabama	Autauga County	55290	55347
	Baldwin County	199713	203709
	Barbour County	26815	26489
	Bibb County	22549	22583
	Blount County	57658	57673

In [80]:
```
df.loc['Michigan', 'Washtenaw County']
```

Out[80]:
```
BIRTHS2010     977
BIRTHS2011    3826
```

```
BIRTHS2012           3780
BIRTHS2013           3662
BIRTHS2014           3683
BIRTHS2015           3709
POPESTIMATE2010    345563
POPESTIMATE2011    349048
POPESTIMATE2012    351213
POPESTIMATE2013    354289
POPESTIMATE2014    357029
POPESTIMATE2015    358880
Name: (Michigan, Washtenaw County), dtype: int64
```

```
In [81]: df.loc[ [('Michigan', 'Washtenaw County'),
                  ('Michigan', 'Wayne County')] ]
```

Out[81]:

		BIRTHS2010	BIRTHS2011	BIRTHS2012	BIRTHS2013	\
STNAME	CTYNAME					
Michigan	Washtenaw County	977	3826	3780	3662	
	Wayne County	5918	23819	23270	23377	

		BIRTHS2014	BIRTHS2015	POPESTIMATE2010	\
STNAME	CTYNAME				
Michigan	Washtenaw County	3683	3709	345563	
	Wayne County	23607	23586	1815199	

		POPESTIMATE2011	POPESTIMATE2012	POPESTIMATE2013	\
STNAME	CTYNAME				
Michigan	Washtenaw County	349048	351213	354289	
	Wayne County	1801273	1792514	1775713	

		POPESTIMATE2014	POPESTIMATE2015
STNAME	CTYNAME		
Michigan	Washtenaw County	357029	358880
	Wayne County	1766008	1759335

7 Missing values

```
In [82]: df = pd.read_csv('log.csv')
         df
```

Out[82]:

	time	user	video	playback position	paused	volume
0	1469974424	cheryl	intro.html	5	False	10.0
1	1469974454	cheryl	intro.html	6	NaN	NaN
2	1469974544	cheryl	intro.html	9	NaN	NaN
3	1469974574	cheryl	intro.html	10	NaN	NaN
4	1469977514	bob	intro.html	1	NaN	NaN
5	1469977544	bob	intro.html	1	NaN	NaN
6	1469977574	bob	intro.html	1	NaN	NaN

```
7   1469977604     bob     intro.html                1    NaN     NaN
8   1469974604  cheryl     intro.html               11    NaN     NaN
9   1469974694  cheryl     intro.html               14    NaN     NaN
10  1469974724  cheryl     intro.html               15    NaN     NaN
11  1469974454     sue  advanced.html               24    NaN     NaN
12  1469974524     sue  advanced.html               25    NaN     NaN
13  1469974424     sue  advanced.html               23  False    10.0
14  1469974554     sue  advanced.html               26    NaN     NaN
15  1469974624     sue  advanced.html               27    NaN     NaN
16  1469974654     sue  advanced.html               28    NaN     5.0
17  1469974724     sue  advanced.html               29    NaN     NaN
18  1469974484  cheryl     intro.html                7    NaN     NaN
19  1469974514  cheryl     intro.html                8    NaN     NaN
20  1469974754     sue  advanced.html               30    NaN     NaN
21  1469974824     sue  advanced.html               31    NaN     NaN
22  1469974854     sue  advanced.html               32    NaN     NaN
23  1469974924     sue  advanced.html               33    NaN     NaN
24  1469977424     bob     intro.html                1   True    10.0
25  1469977454     bob     intro.html                1    NaN     NaN
26  1469977484     bob     intro.html                1    NaN     NaN
27  1469977634     bob     intro.html                1    NaN     NaN
28  1469977664     bob     intro.html                1    NaN     NaN
29  1469974634  cheryl     intro.html               12    NaN     NaN
30  1469974664  cheryl     intro.html               13    NaN     NaN
31  1469977694     bob     intro.html                1    NaN     NaN
32  1469977724     bob     intro.html                1    NaN     NaN
```

```
In [83]: df.fillna
```

```
Out[83]: <bound method DataFrame.fillna of          time    user          video  playback posit
0   1469974424  cheryl     intro.html                5  False    10.0
1   1469974454  cheryl     intro.html                6    NaN     NaN
2   1469974544  cheryl     intro.html                9    NaN     NaN
3   1469974574  cheryl     intro.html               10    NaN     NaN
4   1469977514     bob     intro.html                1    NaN     NaN
5   1469977544     bob     intro.html                1    NaN     NaN
6   1469977574     bob     intro.html                1    NaN     NaN
7   1469977604     bob     intro.html                1    NaN     NaN
8   1469974604  cheryl     intro.html               11    NaN     NaN
9   1469974694  cheryl     intro.html               14    NaN     NaN
10  1469974724  cheryl     intro.html               15    NaN     NaN
11  1469974454     sue  advanced.html               24    NaN     NaN
12  1469974524     sue  advanced.html               25    NaN     NaN
13  1469974424     sue  advanced.html               23  False    10.0
14  1469974554     sue  advanced.html               26    NaN     NaN
15  1469974624     sue  advanced.html               27    NaN     NaN
16  1469974654     sue  advanced.html               28    NaN     5.0
17  1469974724     sue  advanced.html               29    NaN     NaN
```

```
18  1469974484  cheryl     intro.html                 7    NaN     NaN
19  1469974514  cheryl     intro.html                 8    NaN     NaN
20  1469974754     sue  advanced.html                30    NaN     NaN
21  1469974824     sue  advanced.html                31    NaN     NaN
22  1469974854     sue  advanced.html                32    NaN     NaN
23  1469974924     sue  advanced.html                33    NaN     NaN
24  1469977424     bob     intro.html                 1   True    10.0
25  1469977454     bob     intro.html                 1    NaN     NaN
26  1469977484     bob     intro.html                 1    NaN     NaN
27  1469977634     bob     intro.html                 1    NaN     NaN
28  1469977664     bob     intro.html                 1    NaN     NaN
29  1469974634  cheryl     intro.html                12    NaN     NaN
30  1469974664  cheryl     intro.html                13    NaN     NaN
31  1469977694     bob     intro.html                 1    NaN     NaN
32  1469977724     bob     intro.html                 1    NaN     NaN>
```

In [84]:
```
df = df.set_index('time')
df = df.sort_index()
df
```

Out[84]:

```
                user          video  playback position paused  volume
time
1469974424  cheryl     intro.html                 5  False    10.0
1469974424     sue  advanced.html                23  False    10.0
1469974454  cheryl     intro.html                 6    NaN     NaN
1469974454     sue  advanced.html                24    NaN     NaN
1469974484  cheryl     intro.html                 7    NaN     NaN
1469974514  cheryl     intro.html                 8    NaN     NaN
1469974524     sue  advanced.html                25    NaN     NaN
1469974544  cheryl     intro.html                 9    NaN     NaN
1469974554     sue  advanced.html                26    NaN     NaN
1469974574  cheryl     intro.html                10    NaN     NaN
1469974604  cheryl     intro.html                11    NaN     NaN
1469974624     sue  advanced.html                27    NaN     NaN
1469974634  cheryl     intro.html                12    NaN     NaN
1469974654     sue  advanced.html                28    NaN     5.0
1469974664  cheryl     intro.html                13    NaN     NaN
1469974694  cheryl     intro.html                14    NaN     NaN
1469974724  cheryl     intro.html                15    NaN     NaN
1469974724     sue  advanced.html                29    NaN     NaN
1469974754     sue  advanced.html                30    NaN     NaN
1469974824     sue  advanced.html                31    NaN     NaN
1469974854     sue  advanced.html                32    NaN     NaN
1469974924     sue  advanced.html                33    NaN     NaN
1469977424     bob     intro.html                 1   True    10.0
1469977454     bob     intro.html                 1    NaN     NaN
1469977484     bob     intro.html                 1    NaN     NaN
1469977514     bob     intro.html                 1    NaN     NaN
```

```
1469977544     bob     intro.html                  1    NaN     NaN
1469977574     bob     intro.html                  1    NaN     NaN
1469977604     bob     intro.html                  1    NaN     NaN
1469977634     bob     intro.html                  1    NaN     NaN
1469977664     bob     intro.html                  1    NaN     NaN
1469977694     bob     intro.html                  1    NaN     NaN
1469977724     bob     intro.html                  1    NaN     NaN
```

In [85]:
```
df = df.reset_index()
df = df.set_index(['time', 'user'])
df
```

Out[85]:

time	user	video	playback position	paused	volume
1469974424	cheryl	intro.html	5	False	10.0
	sue	advanced.html	23	False	10.0
1469974454	cheryl	intro.html	6	NaN	NaN
	sue	advanced.html	24	NaN	NaN
1469974484	cheryl	intro.html	7	NaN	NaN
1469974514	cheryl	intro.html	8	NaN	NaN
1469974524	sue	advanced.html	25	NaN	NaN
1469974544	cheryl	intro.html	9	NaN	NaN
1469974554	sue	advanced.html	26	NaN	NaN
1469974574	cheryl	intro.html	10	NaN	NaN
1469974604	cheryl	intro.html	11	NaN	NaN
1469974624	sue	advanced.html	27	NaN	NaN
1469974634	cheryl	intro.html	12	NaN	NaN
1469974654	sue	advanced.html	28	NaN	5.0
1469974664	cheryl	intro.html	13	NaN	NaN
1469974694	cheryl	intro.html	14	NaN	NaN
1469974724	cheryl	intro.html	15	NaN	NaN
	sue	advanced.html	29	NaN	NaN
1469974754	sue	advanced.html	30	NaN	NaN
1469974824	sue	advanced.html	31	NaN	NaN
1469974854	sue	advanced.html	32	NaN	NaN
1469974924	sue	advanced.html	33	NaN	NaN
1469977424	bob	intro.html	1	True	10.0
1469977454	bob	intro.html	1	NaN	NaN
1469977484	bob	intro.html	1	NaN	NaN
1469977514	bob	intro.html	1	NaN	NaN
1469977544	bob	intro.html	1	NaN	NaN
1469977574	bob	intro.html	1	NaN	NaN
1469977604	bob	intro.html	1	NaN	NaN
1469977634	bob	intro.html	1	NaN	NaN
1469977664	bob	intro.html	1	NaN	NaN
1469977694	bob	intro.html	1	NaN	NaN
1469977724	bob	intro.html	1	NaN	NaN

In [86]:
```
df = df.fillna(method='ffill')
```

```
df.head()
```

Out[86]:

time	user	video	playback position	paused	volume
1469974424	cheryl	intro.html	5	False	10.0
	sue	advanced.html	23	False	10.0
1469974454	cheryl	intro.html	6	False	10.0
	sue	advanced.html	24	False	10.0
1469974484	cheryl	intro.html	7	False	10.0

In [ ]: