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import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

df_euro_csv = pd.read_csv("Euro2012TEAM.csv")

df euro csv

	Team	Goals	 Subs off	Players Used
0	Croatia	4	 9	16
1	Czech Republic	4	 11	19
2	Denmark	4	 7	15
3	England	5	 11	16
4	France	3	 11	19
5	Germany	10	 15	17
6	Greece	5	 12	20
7	Italy	6	 18	19

```
8
            Netherlands
                                             7
                                                           15
9
                              2
                                             7
                                                           17
                 Poland
10
               Portugal
                              6
                                            14
                                                           16
11
    Republic of Ireland
                              1
                                            10
                                                           17
                                 . . .
12
                              5
                                                           16
                 Russia
                                             7
                                 . . .
13
                  Spain
                             12
                                            17
                                                           18
                              5
14
                                             9
                                                           18
                 Sweden
15
                              2
                                             9
                                                           18
                Ukraine
[16 rows x 35 columns]
df euro csv.to excel("Euro2012TeamXLS.xls")
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:1:
FutureWarning: As the xlwt package is no longer maintained, the xlwt
engine will be removed in a future version of pandas. This is the only
engine in pandas that supports writing in the xls format. Install
openpyxl and write to an xlsx file instead. You can set the option
io.excel.xls.writer to 'xlwt' to silence this warning. While this
option is deprecated and will also raise a warning, it can be globally
set and the warning suppressed.
  """Entry point for launching an IPvthon kernel.
df euro html = pd.read html("Euro2012TeamHTML.html")
df_euro_json = pd.read_json("Euro2012TeamJSON.json")
df euro xlsx = pd.read excel("Euro2012TeamXLSX.xlsx")
df euro csv.to excel("Euro2012TeamXLSX.xlsx")
df euro csv.to html("Euro2012TeamHTML.html")
df euro csv.to json("Euro2012TeamJSON.json")
#Create DataFrame
df = pd.DataFrame()
df
Empty DataFrame
Columns: []
Index: []
proglang = ['Python','Java','CSS','SQL']
df = pd.DataFrame(proglang)
df
        0
   Python
0
     Java
1
2
      CSS
3
      SQL
```

```
rating = [1,2,3,4]
df[1] = rating
df
        0
           1
   Python
           1
0
           2
1
     Java
           3
2
      CSS
3
      SQL 4
df.columns
Int64Index([0, 1], dtype='int64')
df.columns = ['Programming Language', 'Rating']
df
  Programming Language
                         Rating
0
                Python
                              1
                              2
1
                   Java
2
                    CSS
                              3
3
                              4
                    SQL
# DataFrame Using Dictionary
data = [{'a':'apple','b':'ball','c':'cat'},{'a':'doll','b':'egg'}]
df2 = pd.DataFrame(data)
df2
df3 = pd.DataFrame(data, index=['row1','row2'], columns=['a','b'])
df3
df4 = pd.DataFrame(data, index=['row1','row2'], columns=['a','b','c'])
df4
df5 = pd.DataFrame(data, index=['row1','row2'],
columns=['a','b','c','d'])
df5
                          d
          а
                b
                      С
      apple
             ball
                    cat NaN
row1
       doll
row2
              egg
                    NaN NaN
df0 = pd.DataFrame({'ID':[1,2,3,4,5],'Name':
['Prajakta', 'Priyanka', 'Sayali', 'Spruha', 'Sanika']})
df0
   ID
           Name
0
    1
       Prajakta
1
    2
       Priyanka
2
    3
         Sayali
3
    4
         Spruha
4
    5
         Sanika
```

```
# Create a DataFrame from Dictionary of Series
dict = {
    'A': pd.Series([1,2,3], index=['a','b','c']),
    'B': pd.Series([1,2,3,4,5], index=['a','b','c','d','e'])
}
df1 = pd.DataFrame(dict)
df1
     Α
        В
        1
  1.0
  2.0
        2
  3.0
        3
С
d NaN
       4
  NaN 5
# DataFrames of Random Numbers with Date Indices
df dates = pd.date range(start='2020-01-01', end='2022-02-12')
df_dates
'2022-02-03', '2022-02-04', '2022-02-05', '2022-02-06', '2022-02-07', '2022-02-08', '2022-02-09', '2022-02-10', '2022-02-11', '2022-02-12'],
               dtype='datetime64[ns]', length=774, freq='D')
df dates = pd.date range('today', periods=7)
df dates
DatetimeIndex(['2022-02-26 08:32:17.309393', '2022-02-27
08:32:17.309393'
                 2022-02-28 08:32:17.309393', '2022-03-01
08:32:17.309393'
                 2022-03-02 08:32:17.309393', '2022-03-03
08:32:17.309393'
                '2022-03-04 08:32:17.309393'],
               dtype='datetime64[ns]', freq='D')
df dates = pd.date range(start='2022-02-12', periods=7)
df dates
DatetimeIndex(['2022-02-12', '2022-02-13', '2022-02-14', '2022-02-15', '2022-02-16', '2022-02-17', '2022-02-18'],
               dtype='datetime64[ns]', freq='D')
m = np.random.random((7,7))
```

```
array([[0.29702402, 0.99340143, 0.68345537, 0.47936164, 0.66510966,
        0.87395624, 0.27447494],
       [0.39197339, 0.0680303 , 0.72509582, 0.24857744, 0.09357214,
        0.89464463, 0.16723556],
       [0.6468961, 0.71961162, 0.43241249, 0.74406802, 0.45960101,
        0.77317424, 0.58804368],
       [0.84525509, 0.2277863 , 0.4217118 , 0.18089405, 0.17060361,
        0.66518967, 0.34744506],
       [0.69801784, 0.94404295, 0.8365751 , 0.73892661, 0.9031584 ,
        0.09278367, 0.93831671],
       [0.81902038, 0.27757321, 0.73018993, 0.11121947, 0.03754569,
        0.43984833, 0.7681264 ],
       [0.53085847, 0.2496172, 0.95375302, 0.31225572, 0.22123399,
        0.65510711, 0.8135386411)
dframe = pd.DataFrame(m , index=df dates)
dframe
                   0
                                      2
                                                                5
                             1
                     0.993401
                               0.683455
2022-02-12
           0.297024
                                          . . .
                                              0.665110
                                                         0.873956
0.274475
2022-02-13
           0.391973
                     0.068030
                               0.725096
                                              0.093572
                                                         0.894645
                                          . . .
0.167236
2022-02-14
           0.646896
                     0.719612 0.432412
                                              0.459601 0.773174
                                          . . .
0.588044
2022-02-15
           0.845255 0.227786 0.421712
                                          . . .
                                              0.170604 0.665190
0.347445
2022-02-16 0.698018 0.944043 0.836575
                                          ... 0.903158 0.092784
0.938317
2022-02-17
           0.819020 0.277573 0.730190
                                          . . .
                                              0.037546 0.439848
0.768126
2022-02-18
           0.530858 0.249617
                               0.953753
                                              0.221234
                                                         0.655107
0.813539
[7 rows x 7 columns]
dframe.columns = [ 'C1' , 'C2' , 'C3' , 'C4' , 'C5' , 'C6' ,
'C7' ]
dframe
                  C1
                           C2
                                     C3
                                                     C5
                                                               C6
C7
2022-02-12
           0.297024
                     0.993401
                               0.683455
                                          . . .
                                              0.665110
                                                         0.873956
0.274475
2022-02-13
           0.391973
                     0.068030 0.725096
                                         . . .
                                              0.093572 0.894645
0.167236
2022-02-14
           0.646896
                     0.719612
                               0.432412
                                              0.459601
                                                         0.773174
                                          . . .
0.588044
           0.845255 0.227786 0.421712 ... 0.170604 0.665190
2022-02-15
0.347445
```

```
2022-02-16
            0.698018 0.944043 0.836575
                                                0.903158 0.092784
                                           . . .
0.938317
                                                0.037546 0.439848
2022-02-17
            0.819020 0.277573 0.730190
                                           . . .
0.768126
            0.530858 0.249617 0.953753 ...
                                                0.221234 0.655107
2022-02-18
0.813539
[7 rows x 7 columns]
dframe.index
DatetimeIndex(['2022-02-12', '2022-02-13', '2022-02-14', '2022-02-15', '2022-02-16', '2022-02-17', '2022-02-18'],
              dtype='datetime64[ns]', freq='D')
dframe.columns
Index(['C1', 'C2', 'C3', 'C4', 'C5', 'C6', 'C7'], dtype='object')
dframe.dtypes
      float64
C1
C2
      float64
C3
      float64
C4
      float64
C5
      float64
C6
      float64
C7
      float64
dtype: object
dframe.sort values(by='C1')
                  C1
                             C2
                                       C3
                                                       C5
                                                                 C6
C7
2022-02-12
            0.297024 0.993401 0.683455
                                                0.665110
                                           . . .
                                                           0.873956
0.274475
2022-02-13
            0.391973 0.068030 0.725096
                                                0.093572 0.894645
                                           . . .
0.167236
2022-02-18
            0.530858 0.249617 0.953753
                                           . . .
                                                0.221234 0.655107
0.813539
2022-02-14
            0.646896 0.719612 0.432412
                                                0.459601 0.773174
0.588044
2022-02-16
            0.698018 0.944043
                                 0.836575
                                                0.903158 0.092784
                                           . . .
0.938317
2022-02-17
            0.819020 0.277573 0.730190
                                           . . .
                                                0.037546 0.439848
0.768126
            0.845255 0.227786 0.421712
2022-02-15
                                                0.170604 0.665190
                                          . . .
0.347445
[7 rows x 7 columns]
dframe.sort_values(by='C1' , ascending=False)
```

```
C1
                             C2
                                        С3
                                                        C5
                                                                   C6
C7
            0.845255
2022-02-15
                       0.227786
                                  0.421712
                                                  0.170604
                                                            0.665190
0.347445
2022-02-17
            0.819020
                       0.277573
                                  0.730190
                                                  0.037546
                                                            0.439848
0.768126
2022-02-16
            0.698018
                       0.944043
                                  0.836575
                                                  0.903158
                                                            0.092784
0.938317
2022-02-14
            0.646896
                       0.719612
                                  0.432412
                                                  0.459601
                                                            0.773174
0.588044
2022-02-18
            0.530858
                       0.249617
                                  0.953753
                                                  0.221234
                                                            0.655107
0.813539
2022-02-13
            0.391973
                                  0.725096
                                                  0.093572
                       0.068030
                                                            0.894645
0.167236
2022-02-12
            0.297024
                       0.993401
                                  0.683455
                                                  0.665110
                                                            0.873956
0.274475
```

[7 rows x 7 columns]

df1

```
Α
          В
а
   1.0
          1
   2.0
          2
b
   3.0
          3
C
d
          4
   NaN
          5
   NaN
```



#Delete Column in DataFrame

del df1['B']

df1

A a 1.0 b 2.0 c 3.0

```
d NaN
e NaN
df5
                 b
                       С
                           d
      apple ball cat NaN
row1
row2
       doll
                     NaN NaN
               egg
df5.pop('c')
df5
                      d
           а
                 b
      apple ball NaN
row1
               egg NaN
row2
       doll
df
  Programming Language
                          Rating
0
                 Python
                                2
1
                    Java
2
                                3
                     CSS
3
                     S<sub>0</sub>L
                                4
df.index = [21,22,23,24]
df
   Programming Language
                           Rating
21
                  Python
                                 1
22
                                 2
                     Java
23
                      CSS
                                 3
24
                      SQL.
                                 4
df.loc[21]
Programming Language
                          Python
Rating
                                1
Name: 21, dtype: object
```



#Data Selection in DataFrame

```
df
                          Rating
   Programming Language
21
                  Python
22
                               2
                    Java
23
                     CSS
                               3
24
                               4
                     SQL.
df.index
Int64Index([21, 22, 23, 24], dtype='int64')
df.index = [1,2,3,4]
df
  Programming Language
                         Rating
1
                Python
                              1
2
                   Java
                              2
3
                              3
                    CSS
4
                    SQL
                              4
df.loc[1], df.loc[2], df.loc[3], df.loc[4] # accessing column value
(Programming Language
                          Pvthon
Rating
 Name: 1, dtype: object, Programming Language
                                                    Java
 Rating
 Name: 2, dtype: object, Programming Language
                                                    CSS
 Rating
 Name: 3, dtype: object, Programming Language
                                                    SQL.
 Rating
 Name: 4, dtype: object)
df.iloc[1] # accessing index value
Programming Language
                         Java
Rating
                            2
Name: 2, dtype: object
df.iloc[:]
  Programming Language
                         Rating
1
                Python
                              1
2
                              2
                   Java
3
                              3
                    CSS
4
                              4
                    SQL
df.iloc[0:4]
                         Rating
  Programming Language
1
                Python
                              1
2
                              2
                   Java
```

```
CSS
3
                               3
4
                    SQL
                               4
df.iloc[0:1], df.iloc[1:2], df.iloc[2:3], df.iloc[3:4]
   Programming Language
                           Rating
                  Python
                                     Programming Language
1
                                1,
                                                             Rating
2
                    Java
                                2,
                                     Programming Language
                                                             Rating
                     CSS
                                3,
                                     Programming Language
                                                             Rating
 4
                     S<sub>0</sub>L
                                4)
df.iloc[0:4], df.iloc[1:4], df.iloc[2:4], df.iloc[3:4]
   Programming Language
                           Rating
 1
                  Python
2
                                2
                    Java
3
                     CSS
                                3
 4
                                4,
                     SQL
                                     Programming Language
                                                             Rating
2
                                2
                    Java
 3
                     CSS
                                3
 4
                     SQL.
                                4,
                                     Programming Language
                                                             Rating
 3
                     CSS
                                3
 4
                     SQL
                                4,
                                     Programming Language
                                                             Rating
 4
                     SQL
                                4)
df.loc[df.Rating>2]
  Programming Language
                          Rating
3
                    CSS
                    SQL
                               4
4
df1
     Α
   1.0
а
  2.0
b
С
  3.0
d
  NaN
  NaN
df1.loc['a']
     1.0
Name: a, dtype: float64
#df1.iloc['a']
df1.iloc[0]
     1.0
Name: a, dtype: float64
dframe
```

```
C1
                              C2
                                        C3
                                                        C5
                                                                   C6
C7
2022-02-12
                                                  0.665110
                                                             0.873956
            0.297024
                       0.993401
                                  0.683455
0.274475
2022-02-13
            0.391973
                       0.068030
                                  0.725096
                                                  0.093572
                                                             0.894645
                                             . . .
0.167236
2022-02-14
            0.646896
                       0.719612
                                  0.432412
                                                  0.459601
                                                             0.773174
0.588044
2022-02-15
            0.845255
                       0.227786
                                  0.421712
                                                  0.170604
                                                             0.665190
                                             . . .
0.347445
2022-02-16
            0.698018
                       0.944043
                                  0.836575
                                                  0.903158
                                                             0.092784
                                             . . .
0.938317
2022-02-17
            0.819020
                       0.277573
                                  0.730190
                                             . . .
                                                  0.037546
                                                             0.439848
0.768126
2022-02-18
            0.530858
                       0.249617
                                  0.953753
                                                  0.221234
                                                             0.655107
0.813539
[7 rows x 7 columns]
dframe['2022-02-13':'2022-02-15']
                   C1
                              C2
                                                        C5
                                                                   C6
                                        C3
                                             . . .
C7
2022-02-13
            0.391973
                       0.068030
                                  0.725096
                                                  0.093572
                                                             0.894645
0.167236
2022-02-14
            0.646896
                       0.719612
                                  0.432412
                                                  0.459601
                                                             0.773174
                                             . . .
0.588044
2022-02-15
            0.845255
                       0.227786
                                 0.421712
                                                  0.170604
                                                             0.665190
                                             . . .
0.347445
[3 rows x 7 columns]
dframe.loc[:,['C2','C5']]
                   C2
                              C5
2022-02-12
            0.993401
                       0.665110
2022-02-13
            0.068030
                       0.093572
2022-02-14
            0.719612
                       0.459601
2022-02-15
            0.227786
                       0.170604
2022-02-16
            0.944043
                       0.903158
2022-02-17
            0.277573
                       0.037546
2022-02-18
            0.249617
                       0.221234
dframe.loc[:,:]
                   C1
                              C2
                                        C3
                                                                   C6
                                                        C5
                                             . . .
C7
2022-02-12
            0.297024
                       0.993401
                                 0.683455
                                                  0.665110
                                                             0.873956
                                             . . .
0.274475
2022-02-13
            0.391973
                       0.068030
                                  0.725096
                                                  0.093572
                                             . . .
                                                             0.894645
0.167236
```

```
2022-02-14
            0.646896 0.719612
                                0.432412
                                               0.459601
                                                         0.773174
                                          . . .
0.588044
                                0.421712
2022-02-15
            0.845255
                      0.227786
                                               0.170604
                                                          0.665190
                                           . . .
0.347445
            0.698018
                      0.944043
                                0.836575
                                               0.903158 0.092784
2022-02-16
                                          . . .
0.938317
            0.819020 0.277573
                                0.730190
2022-02-17
                                               0.037546 0.439848
                                           . . .
0.768126
2022-02-18
            0.530858 0.249617 0.953753
                                               0.221234 0.655107
                                          . . .
0.813539
[7 rows x 7 columns]
dframe.loc['2022-02-12':'2022-02-15',['C1','C3','C5','C7']]
                  C1
                            C3
                                      C5
2022-02-12
            0.297024
                      0.683455
                                0.665110
                                          0.274475
            0.391973
                      0.725096
2022-02-13
                                0.093572
                                          0.167236
2022-02-14
            0.646896
                      0.432412
                                0.459601
                                          0.588044
                      0.421712
2022-02-15
            0.845255
                                0.170604
                                          0.347445
dframe[dframe['C1']>0.5]
                            C2
                  C1
                                      C3
                                                      C5
                                                                C6
C7
2022-02-14
                     0.719612 0.432412
            0.646896
                                               0.459601
                                                          0.773174
                                          . . .
0.588044
2022-02-15
            0.845255 0.227786
                               0.421712
                                               0.170604
                                                         0.665190
                                          . . .
0.347445
2022-02-16
            0.698018 0.944043
                                0.836575
                                               0.903158 0.092784
                                           . . .
0.938317
2022-02-17
            0.819020 0.277573 0.730190
                                               0.037546 0.439848
                                          . . .
0.768126
2022-02-18
                      0.249617
                                               0.221234
            0.530858
                                0.953753
                                          . . .
                                                         0.655107
0.813539
[5 rows x 7 columns]
dframe[(dframe['C3']>0.2) \& (dframe['C5']<0.3)]
                                      C3 ...
                  C1
                            C2
                                                      C5
                                                                C6
C7
            0.391973 0.068030 0.725096
2022-02-13
                                               0.093572
                                                          0.894645
                                          . . .
0.167236
2022-02-15
                      0.227786 0.421712
            0.845255
                                               0.170604
                                                          0.665190
                                          . . .
0.347445
2022-02-17
            0.819020 0.277573 0.730190
                                               0.037546 0.439848
                                          . . .
0.768126
                      0.249617 0.953753
2022-02-18
            0.530858
                                          . . .
                                               0.221234
                                                         0.655107
0.813539
```

```
[4 rows x 7 columns]
dframe[(dframe['C1']>0.1) & (dframe['C2']<0.88) | (dframe['C3']<0.79)]
                   C1
                             C2
                                        C3
                                                        C5
                                                                  C6
                                            . . .
C7
2022-02-12
            0.297024
                       0.993401
                                 0.683455
                                                 0.665110
                                                            0.873956
                                            . . .
0.274475
2022-02-13
                                 0.725096
            0.391973
                       0.068030
                                                 0.093572
                                                            0.894645
0.167236
2022-02-14
            0.646896
                       0.719612
                                 0.432412
                                                 0.459601
                                                            0.773174
                                            . . .
0.588044
2022-02-15
            0.845255
                       0.227786
                                 0.421712
                                                 0.170604
                                                            0.665190
                                            . . .
0.347445
2022-02-17
            0.819020
                       0.277573
                                 0.730190
                                                 0.037546
                                                            0.439848
                                            . . .
0.768126
2022-02-18
                       0.249617
            0.530858
                                 0.953753
                                                 0.221234
                                                            0.655107
                                            . . .
0.813539
[6 rows x 7 columns]
dframe
                   C1
                             C2
                                        C3
                                                        C5
                                                                  C6
C7
2022-02-12
            0.297024
                       0.993401
                                 0.683455
                                                 0.665110
                                            . . .
                                                            0.873956
0.274475
2022-02-13
            0.391973
                       0.068030
                                 0.725096
                                                 0.093572
                                                            0.894645
                                            . . .
0.167236
2022-02-14
            0.646896
                       0.719612
                                 0.432412
                                                 0.459601 0.773174
                                            . . .
0.588044
2022-02-15
            0.845255
                       0.227786
                                 0.421712
                                                 0.170604
                                                            0.665190
                                            . . .
0.347445
2022-02-16
            0.698018
                       0.944043
                                 0.836575
                                                 0.903158
                                                            0.092784
0.938317
2022-02-17
            0.819020
                       0.277573
                                 0.730190
                                                 0.037546
                                                            0.439848
                                            . . .
0.768126
2022-02-18
                       0.249617
                                 0.953753
            0.530858
                                                 0.221234
                                                            0.655107
0.813539
[7 rows x 7 columns]
dframe[(dframe['C1']>0.05) | (dframe['C2']>0.14) | (dframe['C3']>0.40)
|(dframe['C4']>0.19)|(dframe['C5']>0.07)|(dframe['C6']<0.13)|
(dframe['C7']>0.18)]
                             C2
                                        C3
                                                        C5
                                                                  C6
                   C1
C7
2022-02-12
            0.297024 0.993401 0.683455
                                                 0.665110
                                                            0.873956
                                            . . .
0.274475
```

```
2022-02-13
           0.391973
                     0.068030
                               0.725096
                                              0.093572
                                                        0.894645
0.167236
                     0.719612
                                              0.459601 0.773174
2022-02-14
           0.646896
                               0.432412
0.588044
           0.845255
                     0.227786
                               0.421712
2022-02-15
                                              0.170604
                                                        0.665190
                                         . . .
0.347445
2022-02-16
           0.698018
                     0.944043
                               0.836575
                                              0.903158
                                                        0.092784
0.938317
2022-02-17
           0.819020
                     0.277573
                               0.730190
                                              0.037546 0.439848
0.768126
2022-02-18
           0.530858 0.249617
                               0.953753 ...
                                              0.221234 0.655107
0.813539
```

[7 rows x 7 columns]

dframe.iloc[0][0]

0.2970240205535962



```
for i in range(0,7):
  for j in range(0,7):
    print(dframe.iloc[i][j],end=" ")
  print('\r')
0.2970240205535962 0.9934014300214532 0.6834553697127193
0.47936163525774567 0.665109661037421 0.8739562368944179
0.27447493860838135
0.3919733929241598 0.06803030139282162 0.725095818211181
0.24857743823734124 0.09357214446558892 0.8946446333680803
0.167235555695172
0.6468960971307358 0.7196116239174362 0.43241249401959014
0.7440680169017216 0.45960101209093596 0.7731742355137042
0.5880436772892668
0.8452550912258495 0.22778629542411688 0.421711803234877
0.18089404763642591 0.17060360858690626 0.6651896667567393
0.3474450561889899
0.6980178395383771 0.9440429504645023 0.8365751033456689
0.7389266094644521 0.9031584022330947 0.09278367103618113
```

```
0.9383167071519896
0.8190203839065364 0.27757321017074355 0.7301899320621287
0.11121946783937142 0.03754569443727962 0.4398483285510715
0.7681264020131515
0.5308584704527858 0.24961720213081773 0.953753021545734
0.3122557217360349 0.2212339939525797 0.6551071068942877
0.8135386375759671
dframe.iloc[:,:]
                            C2
                                      C3
                  C1
                                                      C5
                                                                C6
                                          . . .
C7
            0.297024 0.993401
2022-02-12
                                0.683455
                                               0.665110
                                                          0.873956
0.274475
            0.391973
                      0.068030 0.725096
2022-02-13
                                               0.093572 0.894645
0.167236
2022-02-14
            0.646896 0.719612 0.432412
                                               0.459601 0.773174
                                           . . .
0.588044
                                               0.170604 0.665190
            0.845255 0.227786 0.421712
2022-02-15
                                           . . .
0.347445
2022-02-16
            0.698018 0.944043 0.836575
                                               0.903158 0.092784
                                           . . .
0.938317
2022-02-17
            0.819020 0.277573 0.730190
                                               0.037546 0.439848
                                           . . .
0.768126
2022-02-18
                      0.249617 0.953753
                                         ... 0.221234 0.655107
            0.530858
0.813539
[7 rows x 7 columns]
dframe.iloc[2:5, 2:5]
                  C3
                            C4
                                      C5
2022-02-14
            0.432412
                      0.744068
                                0.459601
2022-02-15
            0.421712
                      0.180894
                                0.170604
2022-02-16
            0.836575
                      0.738927
                                0.903158
dframe.iloc[:, 2:5]
                  C3
                            C4
                                      C5
2022-02-12
            0.683455
                      0.479362
                                0.665110
            0.725096
                      0.248577
                                0.093572
2022-02-13
2022-02-14
            0.432412
                      0.744068
                                0.459601
2022-02-15
            0.421712
                      0.180894
                                0.170604
2022-02-16
            0.836575
                      0.738927
                                0.903158
2022-02-17
            0.730190
                      0.111219
                                0.037546
2022-02-18
            0.953753
                      0.312256
                                0.221234
dframe.iloc[0][0]=10
dframe
                   C1
                             C2
                                       C3
                                                       C5
                                                                 C6
                                           . . .
C7
```

```
2022-02-12
            10.000000
                        0.993401
                                  0.683455
                                                   0.665110
                                                             0.873956
0.274475
                        0.068030
2022-02-13
             0.391973
                                   0.725096
                                                   0.093572
                                                             0.894645
0.167236
             0.646896
                        0.719612
                                   0.432412
                                                   0.459601
2022-02-14
                                                             0.773174
                                              . . .
0.588044
             0.845255
                                   0.421712
2022-02-15
                        0.227786
                                                   0.170604
                                                             0.665190
0.347445
2022-02-16
             0.698018
                        0.944043
                                  0.836575
                                                   0.903158
                                                             0.092784
                                             . . .
0.938317
2022-02-17
             0.819020
                        0.277573
                                   0.730190
                                                   0.037546
                                                             0.439848
0.768126
2022-02-18
                        0.249617
                                                   0.221234
             0.530858
                                  0.953753
                                                             0.655107
0.813539
```

[7 rows x 7 columns]



```
for i in range(0,7):
  for j in range(0,7):
    dframe.iloc[i][j]=((i+1)*(j+1))
dframe
              C1
                     C2
                           C3
                                  C4
                                        C5
                                               C6
                                                      C7
2022-02-12
                          3.0
                                 4.0
                                       5.0
                                              6.0
                                                     7.0
             1.0
                    2.0
2022-02-13
                   4.0
                                 8.0
                                      10.0
                                             12.0
                                                    14.0
             2.0
                          6.0
2022-02-14
                                      15.0
                                                    21.0
                   6.0
                          9.0
                                12.0
                                             18.0
             3.0
2022-02-15
             4.0
                   8.0
                         12.0
                                16.0
                                      20.0
                                             24.0
                                                    28.0
                                      25.0
                                                    35.0
2022-02-16
             5.0
                  10.0
                         15.0
                                20.0
                                             30.0
2022-02-17
             6.0
                  12.0
                         18.0
                                24.0
                                      30.0
                                             36.0
                                                    42.0
2022-02-18
                  14.0
                                                    49.0
             7.0
                         21.0
                                28.0
                                      35.0
                                             42.0
dframe[dframe['C1'].isin([1,3,5,7])]
                    C2
                                  C4
                                               C6
              C1
                           C3
                                        C5
                                                      C7
                                       5.0
2022-02-12
             1.0
                   2.0
                          3.0
                                 4.0
                                              6.0
                                                     7.0
2022-02-14
                   6.0
                          9.0
                                      15.0
                                                   21.0
             3.0
                                12.0
                                             18.0
2022-02-16
                  10.0
                         15.0
                                20.0
                                      25.0
                                                    35.0
             5.0
                                             30.0
2022-02-18
                  14.0
                         21.0
                                      35.0
                                                    49.0
             7.0
                                28.0
                                             42.0
```

```
dframe['C1']=123
dframe['C2']=234
dframe['C3']=345
dframe['C4']=456
dframe['C5']=567
dframe['C6']=678
dframe['C7']=789
dframe
              C1
                   C2
                         C3
                              C4
                                   C5
                                         C6
                                              C7
                  234
                                  567
                                             789
2022-02-12
             123
                        345
                             456
                                        678
             123
                  234
                        345
                                             789
2022-02-13
                             456
                                  567
                                        678
2022-02-14
            123
                  234
                        345
                             456
                                             789
                                  567
                                        678
2022-02-15
            123
                  234
                        345
                                  567
                                             789
                             456
                                        678
2022-02-16
            123
                  234
                        345
                             456
                                  567
                                        678
                                             789
2022-02-17
                       345
            123
                  234
                             456
                                  567
                                        678
                                             789
2022-02-18
             123
                  234
                        345
                             456
                                  567
                                        678
                                             789
```

dframe.at Access a single value for a row/column label pair.

Similar to loc, in that both provide label-based lookups. Use at if you only need to get or set a single value in a DataFrame or Series.

Raises

KeyError If 'label' does not exist in DataFrame.

See Also

DataFrame.iat: Access a single value for a row/column pair by integer position. DataFrame.loc: Access a group of rows and columns by label(s). Series.at: Access a single value using a label.

Examples

```
df = pd.DataFrame([[0, 2, 3], [0, 4, 1], [10, 20, 30]], ... index=[4, 5, 6], columns=['A', 'B', 'C']) df A B C 4 0 2 3 5 0 4 1 6 10 20 30
```

Get value at specified row/column pair

```
df.at[4, 'B'] 2
```

Set value at specified row/column pair

$$df.at[4, 'B'] = 10 df.at[4, 'B'] 10$$

Get value within a Series

```
df.loc[5].at['B'] 4
dframe.at[1:4,'C4']=777
dframe
```

/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:719: FutureWarning: Slicing a positional slice with .loc is not supported, and will raise TypeError in a future version. Use .loc with labels or .iloc with positions instead.

indexer = self. get setitem indexer(key)

	C1	C2	С3	C4	C5	C6	C7
2022-02-12	123	234	345	456	567	678	789
2022-02-13	123	234	345	777	567	678	789
2022-02-14	123	234	345	777	567	678	789
2022-02-15	123	234	345	777	567	678	789
2022-02-16	123	234	345	456	567	678	789
2022-02-17	123	234	345	456	567	678	789
2022-02-18	123	234	345	456	567	678	789

dframe.iat Access a single value for a row/column pair by integer position.

Similar to iloc, in that both provide integer-based lookups. Use iat if you only need to get or set a single value in a DataFrame or Series.

Raises

IndexError When integer position is out of bounds.

See Also

DataFrame.at: Access a single value for a row/column label pair. DataFrame.loc: Access a group of rows and columns by label(s). DataFrame.iloc: Access a group of rows and columns by integer position(s).

Examples

```
df = pd.DataFrame([[0, 2, 3], [0, 4, 1], [10, 20, 30]], ... columns=['A', 'B', 'C']) df A B C 0 0 2 3 1 0 4 1 2 10 20 30
```

Get value at specified row/column pair

```
df.iat[1, 2] 1
```

Set value at specified row/column pair

$$df.iat[1, 2] = 10 df.iat[1, 2] 10$$

Get value within a series

	C1	C2	С3	C4	C5	C6	C7
2022-02-12	123	234	345	456	567	678	789
2022-02-13	123	234	345	777	567	678	789

2022 02 14	122	224	245	777	F67	670	700
2022-02-14	123	234	345	///	207	0/8	789
2022-02-15	123	234	345	333	567	678	789
2022-02-16	123	234	345	456	567	678	789
2022-02-17	123	234	345	456	567	678	789
2022-02-18	123	234	345	456	567	678	789



dframe.iloc[5,5]=505
dframe

	C1	C2	С3	C4	C5	C6	C7
2022-02-12	123	234	345	456	567	678	789
2022-02-13	123	234	345	777	567	678	789
2022-02-14	123	234	345	777	567	678	789
2022-02-15	123	234	345	333	567	678	789
2022-02-16	123	234	345	456	567	678	789
2022-02-17	123	234	345	456	567	505	789
2022-02-18	123	234	345	456	567	678	789

(deep: bool_t = True) -> DataFrame Make a copy of this object's indices and data.

When deep=True (default), a new object will be created with a copy of the calling object's data and indices. Modifications to the data or indices of the copy will not be reflected in the original object (see notes below).

When deep=False, a new object will be created without copying the calling object's data or index (only references to the data and index are copied). Any changes to the data of the original will be reflected in the shallow copy (and vice versa).

Parameters

deep: bool, default True Make a deep copy, including a copy of the data and the indices. With deep=False neither the indices nor the data are copied.

Returns

copy : Series or DataFrame Object type matches caller.

Notes

When deep=True, data is copied but actual Python objects will not be copied recursively, only the reference to the object. This is in contrast to copy.deepcopy in the Standard Library, which recursively copies object data (see examples below).

While Index objects are copied when deep=True, the underlying numpy array is not copied for performance reasons. Since Index is immutable, the underlying data can be safely shared and a copy is not needed.

Examples

```
s = pd.Series([1, 2], index=["a", "b"]) s a 1 b 2 dtype: int64 s_copy = s.copy() s_copy a 1 b 2 dtype: int64 Shallow copy versus default (deep) copy:
```

s = pd.Series([1, 2], index=["a", "b"]) deep = s.copy() shallow = s.copy(deep=False) Shallow copy shares data and index with original.

s is shallow False s.values is shallow.values and s.index is shallow.index True Deep copy has own copy of data and index.

s is deep False s.values is deep.values or s.index is deep.index False Updates to the data shared by shallow copy and original is reflected in both; deep copy remains unchanged.

s[0] = 3 shallow[1] = 4 s a 3 b 4 dtype: int64 shallow a 3 b 4 dtype: int64 deep a 1 b 2 dtype: int64 Note that when copying an object containing Python objects, a deep copy will copy the data, but will not do so recursively. Updating a nested data object will be reflected in the deep copy.

s = pd.Series([[1, 2], [3, 4]]) deep = s.copy() s[0][0] = 10 s 0 [10, 2] 1 [3, 4] dtype: object deep 0 [10, 2] 1 [3, 4] dtype: object

```
dframe1 = dframe.copy(deep=True)
dframe1.copy()
```

```
C2
                         C3
                              C4
                                    C5
                                              C7
              C1
                                         C6
                  234
                             456
                                              789
2022-02-12
             123
                        345
                                   567
                                        678
2022-02-13
             123
                  234
                        345
                             777
                                   567
                                        678
                                              789
2022-02-14
             123
                  234
                        345
                             777
                                   567
                                        678
                                             789
2022-02-15
             123
                  234
                        345
                             333
                                   567
                                        678
                                              789
2022-02-16
             123
                  234
                        345
                             456
                                   567
                                        678
                                              789
             123
                        345
2022-02-17
                  234
                             456
                                              789
                                   567
                                        505
                  234
2022-02-18
             123
                       345
                             456
                                   567
                                        678
                                             789
```

	C1	C2	С3	C4	C5	C6	C7
2022-02-12	0	0	0	0	0	0	0
2022-02-13	0	0	0	0	0	0	0
2022-02-14	0	0	0	0	0	0	0
2022-02-15	0	0	0	0	0	0	0
2022-02-16	0	0	0	0	0	0	0

```
2022-02-17
              0
                  0
                      0
                           0
                               0
                                    0
                                        0
                                    0
2022-02-18
              0
                  0
                       0
                           0
                               0
                                        0
dframe1[dframe1['C1'] == 0]
                 C2
                     C3
                              C5
                                       C7
             C1
                          C4
                                   C6
2022-02-12
                  0
              0
                      0
                           0
                               0
                                    0
                                        0
2022-02-13
              0
                  0
                      0
                           0
                               0
                                    0
                                        0
2022-02-14
              0
                  0
                      0
                           0
                               0
                                    0
                                        0
2022-02-15
              0
                  0
                      0
                           0
                               0
                                    0
                                        0
2022-02-16
                  0
                      0
                           0
                               0
                                    0
                                        0
              0
2022-02-17
              0
                  0
                      0
                           0
                               0
                                    0
                                        0
2022-02-18
                  0
                      0
                           0
                               0
                                    0
                                        0
              0
dframe1[dframe1['C1'].isin([0])] = 99
dframe1
             C1
                 C2
                     C3
                          C4
                              C5
                                   C6
                                       C7
2022-02-12
                 99
                     99
                          99
                              99
                                   99
                                       99
             99
2022-02-13
             99
                 99
                     99
                          99
                              99
                                   99
                                       99
                 99
                     99
                          99
                              99
                                   99
                                       99
2022-02-14
             99
2022-02-15
             99
                 99
                     99
                          99
                              99
                                   99
                                       99
2022-02-16
             99
                 99
                     99
                          99
                              99
                                   99
                                       99
                     99
2022-02-17
             99
                 99
                          99
                              99
                                   99
                                       99
2022-02-18
                 99
                     99
                          99
                              99
                                   99
                                       99
             99
dframe1[dframe1['C1'] == 99]
             C1
                 C2
                     C3
                          C4
                              C5
                                   C6
                                       C7
2022-02-12
                 99
                     99
                          99
                              99
                                   99
             99
                                       99
                 99
                     99
                          99
                              99
                                   99
                                       99
2022-02-13
             99
2022-02-14
             99
                 99
                     99
                          99
                              99
                                   99
                                       99
                 99
                     99
                          99
                              99
                                   99
                                       99
2022-02-15
             99
2022-02-16
             99
                 99
                     99
                          99
                              99
                                   99
                                       99
2022-02-17
             99
                 99
                     99
                          99
                              99
                                   99
                                       99
2022-02-18
             99
                 99
                     99
                          99
                              99
                                   99
                                       99
#Dealing with NULL Values
dframe.at[0:8, 'C7'] = np.NaN
dframe.at[0:2, 'C6'] = np.NaN
dframe.at[5:6, 'C5'] = np.NaN
dframe
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:719:
FutureWarning: Slicing a positional slice with .loc is not supported,
and will raise TypeError in a future version. Use .loc with labels or
.iloc with positions instead.
  indexer = self. get setitem indexer(key)
                   C2
                         C3
              C1
                              C4
                                      C5
                                              C6 C7
                                   567.0
2022-02-12
             123
                  234
                        345
                             456
                                            NaN NaN
```

```
2022-02-13
                 234
                                567.0
           123
                      345
                           777
                                         NaN NaN
2022-02-14 123
                 234
                      345
                           777
                                567.0
                                      678.0 NaN
2022-02-15
           123
                 234
                     345
                          333
                                567.0
                                      678.0 NaN
2022-02-16
           123
                234
                     345
                          456
                                567.0
                                      678.0 NaN
2022-02-17
           123
                234
                     345
                           456
                                  NaN
                                      505.0 NaN
2022-02-18
           123
                 234
                     345
                           456 567.0 678.0 NaN
```

dframe.notna()

	C1	C2	С3	C4	C5	C6	C7
2022-02-12	True	True	True	True	True	False	False
2022-02-13	True	True	True	True	True	False	False
2022-02-14	True	True	True	True	True	True	False
2022-02-15	True	True	True	True	True	True	False
2022-02-16	True	True	True	True	True	True	False
2022-02-17	True	True	True	True	False	True	False
2022-02-18	True	True	True	True	True	True	False

dframe.isna()

	C1	C2	С3	C4	C5	C6	C7
2022-02-12	False	False	False	False	False	True	True
2022-02-13	False	False	False	False	False	True	True
2022-02-14	False	False	False	False	False	False	True
2022-02-15	False	False	False	False	False	False	True
2022-02-16	False	False	False	False	False	False	True
2022-02-17	False	False	False	False	True	False	True
2022-02-18	False	False	False	False	False	False	True



dframe = dframe.fillna(1020)
dframe

	C1	C2	С3	C4	C5	C6	C7
2022-02-12	123	234	345	456	567.0	1020.0	1020.0
2022-02-13	123	234	345	777	567.0	1020.0	1020.0
2022-02-14	123	234	345	777	567.0	678.0	1020.0
2022-02-15	123	234	345	333	567.0	678.0	1020.0
2022-02-16	123	234	345	456	567 0	678 O	1020 0

```
2022-02-17
             123
                  234
                        345
                             456
                                   1020.0
                                            505.0
                                                    1020.0
2022-02-18
             123
                  234
                        345
                             456
                                    567.0
                                            678.0
                                                    1020.0
dframe.at[0:5 ,
                 'C7'] = np.NaN
                 'C6'] = np.NaN
dframe.at[0:2 ,
dframe.at[5:6, 'C5'] = np.NaN
dframe
/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:719:
FutureWarning: Slicing a positional slice with .loc is not supported,
and will raise TypeError in a future version. Use .loc with labels or
.iloc with positions instead.
  indexer = self. get setitem indexer(key)
                   C2
                                                      C7
              C1
                         C3
                              C4
                                      C5
                                             C6
                                   567.0
2022-02-12
             123
                  234
                        345
                             456
                                            NaN
                                                     NaN
2022-02-13
             123
                  234
                        345
                                   567.0
                             777
                                            NaN
                                                     NaN
2022-02-14
             123
                  234
                        345
                             777
                                   567.0
                                          678.0
                                                     NaN
2022-02-15
             123
                  234
                        345
                             333
                                   567.0
                                          678.0
                                                     NaN
2022-02-16
             123
                  234
                        345
                             456
                                   567.0
                                          678.0
                                                     NaN
                  234
2022-02-17
             123
                        345
                             456
                                     NaN
                                          505.0
                                                  1020.0
2022-02-18
                  234
             123
                        345
                             456
                                   567.0
                                          678.0
                                                  1020.0
dframe.fillna(value={'C5':123, 'c6':789})
dframe
                   C2
                         C3
                              C4
                                             C6
                                                      C7
              C1
                                      C5
                                   567.0
2022-02-12
             123
                  234
                        345
                             456
                                            NaN
                                                     NaN
2022-02-13
             123
                  234
                        345
                             777
                                   567.0
                                            NaN
                                                     NaN
2022-02-14
                  234
                        345
                             777
                                   567.0
                                          678.0
             123
                                                     NaN
2022-02-15
             123
                  234
                        345
                             333
                                   567.0
                                          678.0
                                                     NaN
2022-02-16
             123
                  234
                        345
                             456
                                   567.0
                                          678.0
                                                     NaN
2022-02-17
             123
                  234
                        345
                             456
                                     NaN
                                          505.0
                                                  1020.0
2022-02-18
             123
                  234
                        345
                             456
                                   567.0
                                          678.0
                                                  1020.0
dframe.fillna(value={'C7':789}, limit=1)
dframe
                                                      C7
              C1
                   C2
                         C3
                              C4
                                      C5
                                             C6
                                   567.0
2022-02-12
             123
                  234
                        345
                             456
                                            NaN
                                                     NaN
2022-02-13
             123
                  234
                        345
                             777
                                   567.0
                                            NaN
                                                     NaN
2022-02-14
                  234
                        345
                             777
                                   567.0
                                          678.0
             123
                                                     NaN
2022-02-15
             123
                  234
                        345
                             333
                                   567.0
                                          678.0
                                                     NaN
2022-02-16
             123
                  234
                        345
                             456
                                   567.0
                                          678.0
                                                     NaN
2022-02-17
             123
                  234
                        345
                                          505.0
                             456
                                     NaN
                                                  1020.0
2022-02-18
             123
                  234
                        345
                             456
                                   567.0
                                          678.0
                                                  1020.0
dframe.dropna()
dframe
                   C2
                         C3
                              C4
                                      C5
                                              C6
                                                      C7
              C1
             123
                  234
                        345
2022-02-12
                             456
                                  567.0
                                            NaN
                                                     NaN
```

```
2022-02-13
             123
                   234
                         345
                                    567.0
                              777
                                              NaN
                                                       NaN
2022-02-14
             123
                   234
                         345
                              777
                                    567.0
                                            678.0
                                                       NaN
2022-02-15
             123
                   234
                         345
                              333
                                    567.0
                                            678.0
                                                       NaN
2022-02-16
             123
                   234
                         345
                              456
                                    567.0
                                            678.0
                                                       NaN
2022-02-17
             123
                   234
                         345
                              456
                                      NaN
                                            505.0
                                                    1020.0
2022-02-18
             123
                   234
                         345
                              456
                                    567.0
                                            678.0
                                                    1020.0
dframe.dropna(axis='columns')
dframe
                    C2
                          C3
                                                        C7
              C1
                               C4
                                       C5
                                               C6
2022-02-12
             123
                   234
                         345
                              456
                                    567.0
                                              NaN
                                                       NaN
2022-02-13
             123
                   234
                                    567.0
                         345
                              777
                                              NaN
                                                       NaN
2022-02-14
             123
                   234
                         345
                              777
                                    567.0
                                            678.0
                                                       NaN
2022-02-15
                   234
                                    567.0
             123
                         345
                              333
                                            678.0
                                                       NaN
2022-02-16
             123
                   234
                         345
                              456
                                    567.0
                                            678.0
                                                       NaN
2022-02-17
             123
                   234
                         345
                              456
                                      NaN
                                            505.0
                                                    1020.0
                                            678.0
2022-02-18
             123
                   234
                         345
                              456
                                    567.0
                                                    1020.0
dframe.dropna(subset=['C5','C6'])
              C1
                    C2
                          C3
                               C4
                                       C5
                                               C6
                                                        C7
                                    567.0
2022-02-14
             123
                   234
                         345
                              777
                                            678.0
                                                       NaN
                              333
2022-02-15
             123
                   234
                         345
                                    567.0
                                            678.0
                                                       NaN
2022-02-16
             123
                   234
                         345
                              456
                                    567.0
                                            678.0
                                                       NaN
2022-02-18
             123
                   234
                         345
                              456
                                    567.0
                                            678.0
                                                    1020.0
#Descriptive Statistics
dframe.fillna(55, inplace=True)
dframe
              C1
                    C2
                          C3
                               C4
                                       C5
                                               C6
                                                        C7
2022-02-12
             123
                   234
                         345
                              456
                                    567.0
                                             55.0
                                                      55.0
2022-02-13
             123
                   234
                         345
                              777
                                    567.0
                                             55.0
                                                      55.0
2022-02-14
             123
                   234
                         345
                              777
                                    567.0
                                            678.0
                                                      55.0
2022-02-15
                   234
                         345
                              333
                                    567.0
                                            678.0
                                                      55.0
             123
2022-02-16
             123
                   234
                         345
                              456
                                    567.0
                                            678.0
                                                      55.0
2022 - 02 - 17
                   234
                                     55.0
             123
                         345
                              456
                                            505.0
                                                    1020.0
                        345
2022-02-18
             123
                   234
                              456
                                    567.0
                                            678.0
                                                    1020.0
dframe.mean()
C1
      123.000000
C2
      234.000000
C3
      345.000000
C4
      530.142857
C5
      493.857143
C6
      475.285714
      330.714286
C7
dtype: float64
```

```
dframe.max()
C1
       123.0
C2
       234.0
       345.0
С3
C4
       777.0
C5
       567.0
C6
       678.0
C7
      1020.0
dtype: float64
dframe.min()
      123.0
C1
C2
      234.0
С3
      345.0
C4
      333.0
C5
       55.0
C6
       55.0
C7
       55.0
dtype: float64
dframe.median()
C1
      123.0
C2
      234.0
С3
      345.0
C4
      456.0
C5
      567.0
      678.0
C6
C7
       55.0
dtype: float64
dframe.std()
C1
        0.000000
C2
        0.000000
С3
        0.000000
C4
      174.514019
C5
      193.517810
C6
      293.977161
C7
      470.871785
dtype: float64
dframe.var()
C1
            0.00000
C2
            0.00000
С3
            0.00000
C4
       30455.142857
C5
       37449.142857
C6
       86422.571429
```

```
C7
      221720.238095
dtype: float64
dframe.quantile()
C1
      123.0
C2
      234.0
С3
      345.0
C4
      456.0
C5
      567.0
C6
      678.0
C7
       55.0
Name: 0.5, dtype: float64
dframe.quantile(0.25)
C1
      123.0
C2
      234.0
С3
      345.0
C4
      456.0
C5
      567.0
C6
      280.0
       55.0
C7
Name: 0.25, dtype: float64
dframe.quantile(0.50)
C1
      123.0
C2
      234.0
С3
      345.0
C4
      456.0
C5
      567.0
C6
      678.0
C7
       55.0
Name: 0.5, dtype: float64
dframe.quantile(0.75)
C1
      123.0
C2
      234.0
С3
      345.0
C4
      616.5
C5
      567.0
C6
      678.0
C7
      537.5
Name: 0.75, dtype: float64
dframe.quantile(0.75)-dframe.quantile(0.25)
C1
        0.0
C2
        0.0
С3
        0.0
```

```
C4
      160.5
C5
        0.0
      398.0
C6
C7
      482.5
dtype: float64
dframe.sum()
C1
       861.0
C2
      1638.0
C3
      2415.0
C4
      3711.0
C5
      3457.0
C6
      3327.0
C7
      2315.0
dtype: float64
dframe.describe()
          C1
                  C2
                         С3
                                      C4
                                                   C5
                                                                C6
C7
                 7.0
                        7.0
                                7.000000
                                             7.000000
                                                          7.000000
count
         7.0
7.000000
                      345.0
               234.0
                              530.142857
                                           493.857143
                                                        475.285714
mean
       123.0
330.714286
                 0.0
                        0.0
                              174.514019
                                           193.517810
                                                        293.977161
std
         0.0
470.871785
                      345.0
min
       123.0
               234.0
                              333.000000
                                            55.000000
                                                         55.000000
55.000000
                              456.000000
                                           567.000000
25%
       123.0
               234.0
                      345.0
                                                        280.000000
55.000000
50%
       123.0
               234.0
                      345.0
                             456.000000
                                           567.000000
                                                       678.000000
55.000000
       123.0
               234.0
                      345.0
                              616.500000
                                           567.000000
                                                       678.000000
75%
537.500000
       123.0
              234.0
                      345.0
                              777.000000
                                           567.000000
                                                       678.000000
max
1020.000000
dframe.skew()
C1
      0.000000
C2
      0.000000
C3
      0.000000
C4
      0.912895
C5
     -2.645751
C6
     -1.057380
C7
      1.229634
dtype: float64
dframe.kurt()
```

```
C1
      0.000000
C2
      0.000000
C3
      0.000000
C4
     -0.842404
C5
      7.000000
C6
     -1.078846
C7
     -0.840000
dtype: float64
dframe.corr()
    C1 C2
            C3
                      C4
                                 C5
                                           C6
                                                      C7
C1 NaN NaN NaN
                     NaN
                                NaN
                                          NaN
                                                     NaN
C2 NaN NaN NaN
                     NaN
                                NaN
                                          NaN
                                                     NaN
C3 NaN NaN NaN
                                          NaN
                     NaN
                                NaN
                                                     NaN
C4 NaN NaN NaN
                1.000000
                           0.187343 -0.307890 -0.290230
C5 NaN NaN NaN
                0.187343
                           1.000000 -0.044571 -0.645497
C6 NaN NaN NaN -0.307890 -0.044571
                                     1.000000
                                                0.270053
C7 NaN NaN NaN -0.290230 -0.645497
                                     0.270053
                                                1.000000
dframe.cov()
     C1
          C2
               C3
                              C4
                                            C5
                                                           C6
C7
C1 0.0
                        0.000000
                                      0.000000
                                                     0.00000
         0.0
              0.0
0.000000
                        0.000000
                                      0.000000
                                                     0.000000
C2 0.0
         0.0
              0.0
0.000000
              0.0
                        0.000000
                                      0.000000
                                                     0.000000
C3
    0.0
         0.0
0.000000
C4
    0.0
              0.0
                   30455.142857
                                   6326.857143 -15795.714286
         0.0
23849.285714
C5
    0.0
              0.0
                    6326.857143
                                  37449.142857
                                                -2535.619048
         0.0
58819.047619
C6
    0.0
         0.0
              0.0 -15795.714286
                                  -2535.619048
                                                86422.571429
37382,261905
    0.0
              0.0 -23849.285714 -58819.047619
C7
         0.0
                                                37382.261905
221720.238095
```

#import statistics as st



import statistics as st
dframe.at[3:6,'C1']=22
dframe

/usr/local/lib/python3.7/dist-packages/pandas/core/indexing.py:719: FutureWarning: Slicing a positional slice with .loc is not supported, and will raise TypeError in a future version. Use .loc with labels or .iloc with positions instead.

indexer = self._get_setitem_indexer(key)

	C1	C2	C3	C4	C5	C6	C7
2022-02-12	123	234	345	456	567.0	55.0	55.0
2022-02-13	123	234	345	777	567.0	55.0	55.0
2022-02-14	123	234	345	777	567.0	678.0	55.0
2022-02-15	22	234	345	333	567.0	678.0	55.0
2022-02-16	22	234	345	456	567.0	678.0	55.0
2022-02-17	22	234	345	456	55.0	505.0	1020.0
2022-02-18	123	234	345	456	567.0	678.0	1020.0

st.mean(dframe['C1'])

79.71428571428571

st.harmonic_mean(dframe['C1'])

41.44857768052516



```
arr = np. array([1, 2,3,4,5, 6,7,8])
st.median(arr)
4.5
st.median_low(arr)
4
st.median_high(arr)
5
st.mode(dframe['C7'])
55.0
st.variance(dframe['C1'])
2914.5714285714284
st.pvariance(dframe['C1'])
2498.204081632653
st.stdev(dframe['C1'])
53.986770866309726
st.pstdev(dframe['C1'])
49.982037589844744
dframe
             C1
                   C2
                        C3
                             C4
                                     C5
                                             C6
                                                     C7
2022-02-12
            123
                  234
                       345
                             456
                                  567.0
                                          55.0
                                                   55.0
2022-02-13
            123
                  234
                       345
                             777
                                  567.0
                                          55.0
                                                   55.0
2022-02-14
            123
                  234
                       345
                            777
                                  567.0
                                         678.0
                                                   55.0
2022-02-15
                  234
             22
                       345
                             333
                                  567.0
                                         678.0
                                                   55.0
2022-02-16
             22
                  234
                       345
                             456
                                  567.0
                                         678.0
                                                   55.0
2022-02-17
             22
                  234
                       345
                             456
                                   55.0
                                         505.0
                                                 1020.0
2022-02-18
            123
                  234
                       345
                             456
                                  567.0
                                         678.0
                                                 1020.0
dframe.apply(max)
C1
       123.0
C2
       234.0
С3
       345.0
C4
       777.0
C5
       567.0
C6
       678.0
C7
      1020.0
dtype: float64
```

```
dframe.apply(min)
C1
       22.0
C2
      234.0
С3
      345.0
C4
      333.0
C5
       55.0
C6
       55.0
C7
       55.0
dtype: float64
```



```
dframe.apply(sum)
C1
       558.0
C2
      1638.0
С3
      2415.0
C4
      3711.0
C5
      3457.0
C6
      3327.0
C7
      2315.0
dtype: float64
dframe.apply(np.sum)
C1
       558.0
C2
      1638.0
С3
      2415.0
C4
      3711.0
C5
      3457.0
C6
      3327.0
C7
      2315.0
dtype: float64
dframe.apply(np.sum ,axis=1)
2022-02-12
              1835.0
2022-02-13
              2156.0
2022-02-14
              2779.0
2022-02-15
              2234.0
```

2022-02-16 2357.0 2022-02-17 2637.0 2022-02-18 3423.0 Freq: D, dtype: float64											
<pre>dframe.applymap(np.sqrt)</pre>											
67	C1	C2	C3		C5	C6					
C7 2022-02-12	11.090537	15.297059	18.574176		23.811762	7.416198					
7.416198 2022-02-13	11.090537	15.297059	18.574176		23.811762	7.416198					
7.416198 2022-02-14	11.090537	15.297059	18.574176		23.811762	26.038433					
7.416198 2022-02-15	4.690416	15.297059	18.574176		23.811762	26.038433					
7.416198 2022-02-16	4.690416	15.297059	18.574176		23.811762	26.038433					
7.416198 2022-02-17	4.690416	15.297059	18.574176		7.416198	22.472205					
31.937439 2022-02-18 31.937439	11.090537	15.297059	18.574176		23.811762	26.038433					
[7 rows x 7 columns]											
import math dframe.appl		qrt)									
	C1	C2	C3		C5	C6					
C7 2022-02-12	11.090537	15.297059	18.574176		23.811762	7.416198					
7.416198											
2022-02-13 7.416198	11.090537	15.297059	18.574176	• • •	23.811762	7.416198					
2022-02-14 7.416198	11.090537	15.297059	18.574176		23.811762	26.038433					
2022-02-15	4.690416	15.297059	18.574176		23.811762	26.038433					
7.416198 2022-02-16	4.690416	15.297059	18.574176		23.811762	26.038433					
7.416198 2022-02-17	4.690416	15.297059	18.574176		7.416198	22.472205					
31.937439 2022-02-18 31.937439	11.090537	15.297059	18.574176		23.811762	26.038433					
[7 rows x 7	columns]										
dframe.appl	ymap(float)										

```
C1
                       C2
                              C3
                                      C4
                                             C5
                                                     C6
                                                             C7
            123.0
                    234.0
                           345.0
2022-02-12
                                   456.0
                                          567.0
                                                   55.0
                                                           55.0
                    234.0
                                   777.0
                                                           55.0
2022-02-13
            123.0
                           345.0
                                          567.0
                                                   55.0
2022-02-14
            123.0
                    234.0
                           345.0
                                   777.0
                                          567.0
                                                 678.0
                                                           55.0
             22.0
                    234.0
                           345.0
                                   333.0
2022-02-15
                                          567.0
                                                  678.0
                                                           55.0
2022-02-16
             22.0
                    234.0
                           345.0
                                   456.0
                                          567.0
                                                  678.0
                                                           55.0
                                                         1020.0
2022-02-17
             22.0
                    234.0
                           345.0
                                   456.0
                                           55.0
                                                 505.0
            123.0
                           345.0
                                   456.0
                                          567.0
                                                 678.0
                                                         1020.0
2022-02-18
                    234.0
dframe.apply(lambda x: min(x))
C1
       22.0
C2
      234.0
C3
      345.0
C4
      333.0
C5
       55.0
C6
       55.0
C7
       55.0
dtype: float64
dframe.apply(lambda x: x*x)
                C1
                       C2
                                        C4
                                                   C5
                                С3
                                                             C6
C7
                    54756
                           119025
2022-02-12
            15129
                                    207936
                                            321489.0
                                                         3025.0
3025.0
2022-02-13
            15129
                    54756
                           119025
                                    603729
                                            321489.0
                                                         3025.0
3025.0
                    54756
                           119025
                                    603729
                                            321489.0
2022-02-14
            15129
                                                       459684.0
3025.0
2022-02-15
              484
                    54756
                           119025
                                    110889
                                            321489.0
                                                       459684.0
3025.0
2022-02-16
               484
                    54756
                           119025
                                    207936
                                            321489.0
                                                       459684.0
3025.0
2022-02-17
               484
                    54756
                           119025
                                    207936
                                              3025.0
                                                       255025.0
1040400.0
2022-02-18
            15129
                    54756
                           119025
                                    207936
                                            321489.0
                                                       459684.0
1040400.0
dafl = pd.DataFrame ({'id' : ['1', '2', '3', '4', '5'], 'Name' :
['Prajakta', 'Prerna', 'Prema', 'Pari', 'Palkia']})
dafl
  id
          Name
      Prajakta
0
  1
1
  2
        Prerna
   3
2
         Prema
3
   4
          Pari
4
   5
        Palkia
```

```
daf2 = pd.DataFrame ({'id': ['1', '2', '6', '7', '8'], 'Score': [40 ,
60 , 80 , 90 , 70]})
daf2
  id
      Score
0
         40
   1
  2
         60
1
2
  6
         80
3
   7
         90
4
   8
         70
# Inner Join
pd.merge(dafl, daf2, on='id', how='inner')
                Score
          Name
  id
  1
      Prajakta
                    40
   2
        Prerna
                    60
# Full Outer Join
pd.merge(dafl, daf2, on='id', how='outer')
  id
          Name Score
      Prajakta
                  40.0
   1
   2
        Prerna
                  60.0
1
2
   3
         Prema
                   NaN
3
   4
          Pari
                   NaN
   5
4
        Palkia
                   NaN
5
  6
           NaN
                  80.0
6
   7
           NaN
                  90.0
7
   8
           NaN
                  70.0
# Left Outer Soon
pd.merge(dafl, daf2, on='id', how='left')
          Name
  id
                 Score
   1
      Prajakta
                  40.0
  2
        Prerna
                  60.0
1
2
  3
         Prema
                   NaN
3
   4
          Pari
                   NaN
   5
        Palkia
                   NaN
#Right Outer Join
pd.merge(dafl, daf2, on='id', how='right')
  id
          Name
                 Score
0
  1
      Prajakta
                    40
        Prerna
                    60
1
  2
2
                    80
  6
           NaN
3
  7
           NaN
                    90
4
   8
           NaN
                    70
```

```
import glob
path = r"C:\Users\User\Downloads\COVID-19-master\COVID-19-master\
csse_covid_19_data\csse_covid_19_daily_reports_us"
filenames = glob.glob(path + "\*.csv")
print(path, filenames)
covid=pd.DataFrame()
for i in filenames:
  df = pd.read csv(i)
  print(i)
  covid = covid.append(df, ignore index=True,sort=True)
C:\Users\User\Downloads\COVID-19-master\COVID-19-master\
csse covid 19 data\csse covid 19 daily reports us []
from google.colab import data table
data table.enable dataframe formatter()
df pokemon =
pd.read csv("https://raw.githubusercontent.com/lgreski/pokemonData/
master/Pokemon.csv")
df pokemon
       ID
                  Name
                                        ... Sp. Def Speed
                                                            Generation
0
        1
            Bulbasaur
                                                 65
                                                        45
                                                                      1
        2
1
                                                 80
                                                                      1
               Ivysaur
                                                        60
2
        3
                                                                      1
              Venusaur
                                                100
                                                        80
3
        4
           Charmander
                                                 50
                                                        65
                                                                      1
4
        5
           Charmeleon
                                                 65
                                                        80
                                                                      1
                                                       . . .
                                                . . .
1040
      896
            Glastrier
                                                110
                                                       30
                                                                      8
            Spectrier
                                                                      8
1041
      897
                                                 80
                                                       130
                                                        80
                                                                      8
1042
      898
               Calyrex
                                                 80
1043
               Calyrex
                                                        50
                                                                      8
      898
                            Ice Rider
                                                130
1044
      898
               Calyrex
                        Shadow Rider
                                                       150
                                                                      8
                                                100
[1045 \text{ rows x } 13 \text{ columns}]
df pokemon.head(), df pokemon.tail()
    ID
               Name Form
                          Type1
                                 ... Sp. Atk Sp. Def
Generation
0
     1
         Bulbasaur
                          Grass
                                            65
                                                      65
                                                             45
1
 1
     2
                                                      80
                                                             60
           Ivysaur
                          Grass
                                            80
                                 . . .
1
 2
     3
                          Grass
                                                     100
                                                             80
          Venusaur
                                           100
                                 . . .
1
3
     4 Charmander
                            Fire
                                            60
                                                      50
                                                             65
1
4
     5 Charmeleon
                            Fire
                                 . . . .
                                            80
                                                      65
                                                             80
1
```

```
[5 rows x 13 columns],
                                                               Generation
         ID
                   Name
                                   Form
                                          ... Sp. Def Speed
 1040
       896
             Glastrier
                                                          30
                                                                         8
                                                   110
                                                                         8
 1041
       897
             Spectrier
                                                    80
                                                          130
 1042
       898
               Calyrex
                                                    80
                                                          80
                                                                         8
               Calyrex
                                                          50
                                                                         8
 1043
                             Ice Rider
       898
                                                   130
 1044
                         Shadow Rider
                                                                         8
       898
               Calyrex
                                                   100
                                                         150
 [5 rows x 13 columns])
df_pokemon['Total'] = df_pokemon['HP'] + df_pokemon['Attack']
df pokemon
        ΙD
                                          ... Sp. Def Speed
                                                               Generation
                   Name
                                   Form
0
         1
             Bulbasaur
                                                    65
                                                           45
                                                                         1
         2
                                                                         1
1
               Ivysaur
                                                    80
                                                           60
2
                                                                         1
         3
              Venusaur
                                                   100
                                                           80
3
         4
                                                                         1
            Charmander
                                                    50
                                                          65
4
         5
                                                                         1
            Charmeleon
                                                          80
                                                    65
1040
      896
             Glastrier
                                                   110
                                                          30
                                                                         8
1041
      897
             Spectrier
                                                    80
                                                          130
                                                                         8
                                                                         8
1042
      898
               Calyrex
                                                    80
                                                          80
1043
                                                          50
                                                                         8
      898
               Calyrex
                             Ice Rider
                                                   130
                                                                         8
1044
               Calyrex
                          Shadow Rider
      898
                                                   100
                                                          150
[1045 rows x 13 columns]
df pokemon['Total'] = df pokemon['HP'] + df pokemon['Attack']
df pokemon
        ΙD
                                              Sp. Def Speed
                                                               Generation
                   Name
                                   Form
             Bulbasaur
0
         1
                                                    65
                                                           45
                                                                         1
1
         2
                                                                         1
               Ivysaur
                                                    80
                                                           60
2
         3
              Venusaur
                                                   100
                                                          80
                                                                         1
3
         4
            Charmander
                                                    50
                                                          65
                                                                         1
         5
4
                                                                         1
            Charmeleon
                                                    65
                                                          80
                                                          . . .
                                                                        . .
             Glastrier
                                                                         8
1040
      896
                                                   110
                                                          30
             Spectrier
                                                                         8
1041
      897
                                                    80
                                                          130
1042
      898
                                                          80
                                                                         8
               Calyrex
                                                    80
1043
      898
               Calyrex
                             Ice Rider
                                                   130
                                                          50
                                                                         8
1044
      898
               Calyrex
                          Shadow Rider
                                                   100
                                                         150
                                                                         8
[1045 \text{ rows x } 13 \text{ columns}]
df_pokemon['Total'] = df_pokemon.iloc[:,4:10].sum(axis=1)
df_pokemon.head(5)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.

"""Entry point for launching an IPython kernel.

	ID	Name Form	Type1	 Sp. Atk	Sp. Def	Speed	Generation
0	1	Bulbasaur	Grass	 65	65	45	1
1	2	Ivysaur	Grass	 80	80	60	1
2	3	Venusaur	Grass	 100	100	80	1
3	4	Charmander	Fire	 60	50	65	1
4	5	Charmeleon	Fire	 80	65	80	1

[5 rows x 13 columns]

```
cols = list(df_pokemon.columns)
df_pokemon = df_pokemon[cols[0:10] + [cols[-1]] + cols[10:12]]
df_pokemon.head(5)
```

	ID	Name Form	Type1	 Sp. Atk	Generation	Sp. Def	Speed
0	1	Bulbasaur	Grass	 65	1	65	45
1	2	Ivysaur	Grass	 80	1	80	60
2	3	Venusaur	Grass	 100	1	100	80
3	4	Charmander	Fire	 60	1	50	65
4	5	Charmeleon	Fire	 80	1	65	80

[5 rows x 13 columns]

```
cols = list(df_pokemon.columns)
df_pokemon = df_pokemon[cols[0:10] + [cols[-1]] + cols[10:12]]
df pokemon.head(5)
```

	ID	Name Form	Type1	 Sp. Atk	Speed	Generation	Sp. Def
0	1	Bulbasaur	Grass	 65	45	1	65
1	2	Ivysaur	Grass	 80	60	1	80
2	3	Venusaur	Grass	 100	80	1	100
3	4	Charmander	Fire	 60	65	1	50
4	5	Charmeleon	Fire	 80	80	1	65

[5 rows x 13 columns]

```
cols = list(df_pokemon.columns)
df_pokemon = df_pokemon[cols[0:10] + [cols[12]] + cols[10:12]]
df_pokemon.head(5)
```

	ID	Name Form	Type1	Sı	o. Atk	Sp. Def	Speed	Generation
0	1	Bulbasaur	Grass		65	65	45	1
1	2	Ivysaur	Grass		80	80	60	1
2	3	Venusaur	Grass		100	100	80	1
3	4	Charmander	Fire		60	50	65	1

```
5 Charmeleon
                          Fire ...
                                                          80
4
                                         80
                                                   65
                                                                        1
[5 rows x 13 columns]
df pokemon.to csv('poke updated1.csv', index=False)
df pokemon.to html('poke updated1.html', index=False)
df pokemon.to json('poke updated1.json')
df pokemon.to excel('poke updated1.xlsx', index=False)
df pokemon.to excel('poke updated1 xls.xls', index=False)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:1:
FutureWarning: As the xlwt package is no longer maintained, the xlwt
engine will be removed in a future version of pandas. This is the only
engine in pandas that supports writing in the xls format. Install
openpyxl and write to an xlsx file instead. You can set the option
io.excel.xls.writer to 'xlwt' to silence this warning. While this
option is deprecated and will also raise a warning, it can be globally
set and the warning suppressed.
  """Entry point for launching an IPython kernel.
df pokemon.to csv('poke.txt' , sep='\t' , index=False)
df pokemon.loc[df pokemon['Type2'] == 'Dragon']
df pokemon
       ID
                 Name
                                Form
                                       ... Sp. Def Speed
                                                          Generation
0
        1
            Bulbasaur
                                                65
                                                      45
                                                                    1
1
        2
                                                                    1
              Ivysaur
                                                80
                                                      60
2
        3
                                                                   1
             Venusaur
                                               100
                                                      80
3
        4
                                                                   1
           Charmander
                                                50
                                                      65
        5
4
           Charmeleon
                                                      80
                                                                   1
                                                65
                                                     . . .
                                                                  . . .
                                               . . .
1040
      896
            Glastrier
                                               110
                                                      30
                                                                   8
            Spectrier
                                                                   8
1041
      897
                                                80
                                                     130
                                                                   8
1042
      898
              Calvrex
                                                80
                                                      80
1043
              Calvrex
                           Ice Rider
                                                      50
                                                                   8
      898
                                               130
1044
      898
              Calyrex
                       Shadow Rider
                                               100
                                                     150
                                                                   8
[1045 rows \times 13 columns]
df3 = df pokemon.loc[(df pokemon['Type2'] == 'Dragon') &
(df pokemon['Type1'] == 'Dark')]
df3
      ID
               Name Form Type1 ... Sp. Atk Sp. Def
Generation
649
    633
              Deino
                           Dark
                                . . .
                                          45
                                                    50
                                                           38
5
```

```
Zweilous
650
     634
                             Dark
                                              65
                                                         70
                                                                 58
5
651
           Hydreigon
     635
                             Dark
                                             125
                                                         90
                                                                 98
5
            Guzzlord
912
     799
                             Dark
                                              97
                                                         53
                                                                 43
                                   . . .
[4 rows x 13 columns]
df4
     = df3.reset index()
df4
   index
                                   ... Sp. Atk Sp. Def
                                                           Speed
            ID
                      Name Form
                                                                   Generation
0
     649
           633
                     Deino
                                             45
                                                      50
                                                              38
                                                                             5
1
           634
                  Zweilous
                                             65
                                                      70
                                                              58
     650
                                                                             5
2
                                                      90
     651
           635
                Hydreigon
                                            125
                                                              98
                                                                             7
3
     912
           799
                  Guzzlord
                                             97
                                                      53
                                                              43
[4 rows x 14 columns]
df3.reset index(drop=True , inplace=True)
df3
    ID
               Name Form Type1
                                  ... Sp. Atk
                                                 Sp. Def
                                                           Speed
                                                                   Generation
   633
             Deino
                           Dark
0
                                            45
                                                      50
                                                              38
                                                                             5
   634
          Zweilous
                                            65
                                                      70
                                                              58
1
                           Dark
                                                                             5
2
         Hydreigon
                           Dark
                                           125
                                                      90
                                                              98
   635
                                                                             7
3
   799
          Guzzlord
                           Dark
                                            97
                                                      53
                                                              43
                                  . . .
[4 rows x 13 columns]
df pokemon.head(10)
   ID
              Name Form
                           Type1
                                   ... Sp. Atk
                                                  Sp. Def
                                                            Speed
                                                                    Generation
0
    1
         Bulbasaur
                           Grass
                                                       65
                                                                45
                                             65
                                                                               1
                                   . . .
    2
1
           Ivysaur
                           Grass
                                             80
                                                       80
                                                                60
                                                                              1
                                   . . .
2
                                                      100
                                                                              1
    3
          Venusaur
                           Grass
                                            100
                                                                80
3
    4
        Charmander
                            Fire
                                             60
                                                       50
                                                                65
                                                                              1
                                   . . .
4
    5
        Charmeleon
                            Fire
                                                       65
                                                                80
                                                                              1
                                             80
                                   . . .
5
                                                              100
    6
         Charizard
                            Fire
                                                       85
                                                                              1
                                            109
6
    7
                                                       64
                                                                43
                                                                              1
          Squirtle
                           Water
                                             50
7
                                                       80
                                                                58
                                                                              1
    8
         Wartortle
                           Water
                                             65
8
         Blastoise
                                             85
                                                      105
                                                                78
                                                                              1
    9
                           Water
                                   . . .
9
   10
          Caterpie
                             Bug
                                             20
                                                       20
                                                                45
                                                                              1
                                   . . .
[10 rows x 13 columns]
df pokemon.Name.str.contains("rill").head(10)
0
     False
```

False

```
2
     False
     False
False
4
5
     False
     False
6
     False
7
8
     False
     False
Name: Name, dtype: bool
```

df_pokemon.loc[df_pokemon.Name.str.contains('rill')]

	ID	Name	Form	Type1	 Sp. Atk	Sp. Def
Spee	d Ge	neration				
14	15	Beedrill		Bug	 45	80
75		1				
182	183	Marill		Water	 20	50
40		2				
183	184	Azumarill		Water	 60	80
50		2				
297	298	Azurill		Normal	 20	40
20		3				
544	530	Excadrill		Ground	 50	65
88		5				
608	592	Frillish		Water	 65	85
40		5				
677	15	Beedrill	Mega Beedrill	Bug	 15	80
145		6	-	_		

[7 rows x 13 columns]

df_pokemon.loc[-df_pokemon.Name.str.contains('rill')].head(10)

	ID	Name Form	Type1	9	Sp. Atk	Sp. Def	Speed	Generation
0	1	Bulbasaur	Grass		65	65	45	1
1	2	Ivysaur	Grass		80	80	60	1
2	3	Venusaur	Grass		100	100	80	1
3	4	Charmander	Fire		60	50	65	1
4	5	Charmeleon	Fire		80	65	80	1
5	6	Charizard	Fire		109	85	100	1
6	7	Squirtle	Water		50	64	43	1
7	8	Wartortle	Water		65	80	58	1
8	9	Blastoise	Water		85	105	78	1
9	10	Caterpie	Bug		20	20	45	1

[10 rows x 13 columns]



 $\begin{array}{ll} df_pokemon.loc[df_pokemon['Type1'].str.contains('Grass') \ \& \\ df_pokemon['Type2'].str.contains("Poison")] \end{array}$

_	ID	Name		Form	Type1		Sp. Atk	Sp. Def
Spee 0	d Ge 1	neration Bulbasaur			Grass		65	65
45	1	1			01 033	• • • •	05	03
1	2	Ivysaur			Grass		80	80
60 2	3	1 Venusaur			Grass		100	100
80	5	1			01 055	• • • •	100	100
42	43	Oddish			Grass		75	65
30 43	44	1			Craca		OF	75
43 40	44	Gloom 1			Grass		85	75
44	45	- Vileplume			Grass		110	90
50	60	1			C		70	20
68 40	69	Bellsprout 1			Grass		70	30
69	70	Weepinbell			Grass		85	45
55		1						
70 70	71	Victreebel 1			Grass		100	70
314	315	Roselia			Grass		100	80
65		3						
411 55	406	Budew 4			Grass		50	70
412	407	4 Roserade			Grass		125	105
90		4			0.000			
606	590	Foongus			Grass		55	55
15 607	591	5 Amoonguss			Grass		85	80
30	JJ1	5			diass		05	00
673 80	3	Venusaur 6	Mega	Venusaur	Grass		122	120

[15 rows x 13 columns]

```
df pokemon.loc[df pokemon['Type1'].str.contains('grass |water',regex =
True)].head(10)
Empty DataFrame
Columns: [ID, Name, Form, Type1, Type2, Total, HP, Attack, Defense,
Sp. Atk, Sp. Def, Speed, Generation]
Index: []
df_pokemon.loc[df_pokemon['Type1'].str.contains('grass |water', case =
False , regex = True)].head(10)
    ID
              Name Form
                                 ... Sp. Atk Sp. Def
                          Type1
                                                         Speed
Generation
6
     7
          Squirtle
                          Water
                                            50
                                                     64
                                                             43
1
7
     8
         Wartortle
                          Water
                                            65
                                                     80
                                                             58
1
8
     9
                                                             78
         Blastoise
                          Water
                                            85
                                                    105
                                  . . .
1
53
    54
           Psyduck
                                                     50
                                                             55
                          Water
                                            65
1
54
    55
           Golduck
                                                     80
                                                             85
                          Water
                                            95
                                  . . .
1
59
            Poliwag
                                                     40
                                                             90
    60
                          Water
                                            40
                                  . . .
1
60
    61
         Poliwhirl
                          Water
                                            50
                                                     50
                                                             90
1
61
    62
         Poliwrath
                                            70
                                                     90
                                                             70
                          Water
1
71
         Tentacool
                                                             70
    72
                          Water
                                            50
                                                    100
1
        Tentacruel
72
   73
                          Water
                                            80
                                                    120
                                                            100
                                  . . .
[10 rows x 13 columns]
import re
df pokemon.loc[df pokemon['Type1'].str.contains('grass |water',flags =
re.I ,regex = True)].head(10)
    ID
               Name Form
                          Type1
                                  ... Sp. Atk Sp. Def
                                                          Speed
Generation
6
     7
          Squirtle
                          Water
                                            50
                                                     64
                                                             43
1
7
     8
         Wartortle
                          Water
                                            65
                                                     80
                                                             58
1
8
     9
         Blastoise
                                                             78
                                            85
                                                    105
                          Water
                                  . . .
1
53
    54
           Psyduck
                                            65
                                                     50
                                                             55
                          Water
1
54
    55
           Golduck
                          Water
                                            95
                                                     80
                                                             85
                                 . . .
```

1						
59 1	60	Poliwag	Water	 40	40	90
60 1	61	Poliwhirl	Water	 50	50	90
61 1	62	Poliwrath	Water	 70	90	70
71 1	72	Tentacool	Water	 50	100	70
72 1	73	Tentacruel	Water	 80	120	100

[10 rows x 13 columns]

 $df_pokemon.loc[df_pokemon.Name.str.contains('Wa',flags = re.I ,regex = True)].head(10)$

	ID	Name	Form	Type1	 Sp.	Atk	Sp. De	f Speed
Gene	ratio	n						
7 1	8	Wartortle		Water		65	80	58
59 1	60	Poliwag		Water		40	40	90
104 1	105	Marowak		Ground		50	80	9 45
259 3	260	Swampert		Water		85	90	60
316 3	317	Swalot		Poison		73	83	3 55
319 3	320	Wailmer		Water		70	3!	60
320 3	321	Wailord		Water		90	45	60
332 3	333	Swablu		Normal		40	75	5 50
367 3	365	Walrein		Ice		95	90	65
519 5	505	Watchog		Normal		60	69	77

[10 rows x 13 columns]

df_pokemon.loc[df_pokemon.Name.str.contains('^Wa[a-l]
+',flags=re.I ,regex=True)].head(10)

	ID	Name Form	Type1	 Sp. Atk	Sp. Def	Speed	Generation
319	320	Wailmer	Water	 70	35	60	3
320	321	Wailord	Water	 90	45	60	3
367	365	Walrein	Ice	 95	90	65	3

[3 rows x 13 columns]

df_pokemon.loc[df_pokemon.Name.str.contains('^[x-z]',flags =
re.I ,regex = True)]

Conon	ID	Name	Form	5	p. Def	Speed
40	ation 41	Zubat			40	55
1 144	145	Zapdos			90	100
1 177	178	Xatu			70	95
2 192	193	Yanma			45	95
2 262	263	Zigzagoon			41	60
3 334	335	Zangoose			60	90
3 476	469	Yanmega			56	95
4 537	523	Zebstrika			63	116
5 578	562	Yamask			65	30
5 586	570	Zorua			40	65
5 587 5	571	Zoroark			60	105
650 5	634	Zweilous			70	58
662 5	644	Zekrom			100	90
796	716	Xerneas			98	99
6 797	717	Yveltal			98	99
6 798	718	Zygarde50% Forme			95	95
6 825 7	718	Zygarde10% Forme			85	115
826 7	718	Zygarde	Complete Forme		95	85
839 7	734	Yungoos			30	45
909 7	796	Xurkitree			71	83
923 7	807	Zeraora			80	143
935 8	145	Zapdos	Galarian Zapdos		90	100
939	263	Zigzagoon	Galarian Zigzagoon		41	60

```
944
      562
                       Yamask
                                     Galarian Yamask
                                                                  65
                                                                         30
8
971
      835
                       Yamper
                                                                  50
                                                                         26
                                                         . . .
1028
      888
                       Zacian
                                        Crowned Sword
                                                                 115
                                                                        148
1029
                                Hero of Many Battles
      888
                       Zacian
                                                                 115
                                                                        138
8
1030
                    Zamazenta
                                      Crowned Shield
      889
                                                                 145
                                                                        128
1031
                    Zamazenta
                                Hero of Many Battles
      889
                                                                 115
                                                                        138
1037
      893
                       Zarude
                                                                  95
                                                                        105
                                                         . . .
[30 rows x 13 columns]
df pokemon['Name2'] = df pokemon.Name.str.extract(r'(^{w{3})')
df pokemon.head()
   ID
              Name Form
                          Type1
                                  ... Sp. Def
                                                Speed
                                                        Generation
                                                                      Name2
0
    1
        Bulbasaur
                          Grass
                                            65
                                                    45
                                                                        Bul
                                                                  1
1
    2
           Ivysaur
                          Grass
                                  . . .
                                            80
                                                    60
                                                                  1
                                                                        Ivy
2
    3
                                                                  1
          Venusaur
                          Grass
                                           100
                                                    80
                                                                        Ven
                                  . . .
                                                                  1
3
       Charmander
                           Fire
                                                                        Cha
                                  . . .
                                            50
                                                    65
       Charmeleon
                           Fire
                                                                  1
                                            65
                                                    80
                                                                        Cha
                                  . . .
[5 rows x 14 columns]
df_pokemon.loc[df_pokemon.Name.str.match(r'(^[B|b].*)')].head(5)
    ID
               Name Form
                           Type1
                                    ... Sp. Def
                                                  Speed
                                                         Generation
                                                                       Name2
0
     1
          Bulbasaur
                           Grass
                                   . . .
                                             65
                                                     45
                                                                   1
                                                                         Bul
8
                                                     78
                                                                   1
                                                                         Bla
     9
          Blastoise
                           Water
                                            105
                                    . . .
11
    12
        Butterfree
                                             80
                                                     70
                                                                   1
                                                                         But
                              Bug
14
    15
           Beedrill
                                             80
                                                     75
                                                                   1
                                                                         Bee
                              Bug
                                   . . .
68
    69
        Bellsprout
                                                     40
                                                                   1
                                                                         Bel
                           Grass
                                             30
```

#Replace values in DataFrame

[5 rows x 14 columns]



df_pokemon.head(10)

```
ID
              Name Form
                           Type1
                                   ... Sp. Def
                                                 Speed
                                                         Generation
                                                                       Name2
0
    1
         Bulbasaur
                           Grass
                                             65
                                                     45
                                                                    1
                                                                         Bul
    2
                                                                    1
1
           Ivysaur
                           Grass
                                             80
                                                     60
                                                                         Ivy
2
    3
                                                                    1
          Venusaur
                           Grass
                                            100
                                                     80
                                                                         Ven
3
                                                                    1
    4
       Charmander
                            Fire
                                             50
                                                     65
                                                                         Cha
4
       Charmeleon
                                                                    1
                            Fire
                                             65
                                                     80
                                                                         Cha
5
    6
        Charizard
                            Fire
                                             85
                                                    100
                                                                    1
                                                                         Cha
6
                                                                    1
    7
          Squirtle
                           Water
                                             64
                                                     43
                                                                         Squ
                                   . . .
7
    8
        Wartortle
                                             80
                                                     58
                                                                    1
                                                                         War
                           Water
8
    9
         Blastoise
                                                                    1
                                                                         Bla
                           Water
                                            105
                                                     78
   10
          Caterpie
                                             20
                                                     45
                                                                    1
                                                                         Cat
                             Bug
[10 rows x 14 columns]
```

```
df_pokemon['Type1'] = df_pokemon['Type1'].replace({'Grass' : 'Meadow'
, 'Fire' : 'Blaze'})
df pokemon.head(10)
```

	ID	Name Form	Type1	 Sp. Def	Speed	Generation	Name2
0	1	Bulbasaur	Meadow	 65	45	1	Bul
1	2	Ivysaur	Meadow	 80	60	1	Ivy
2	3	Venusaur	Meadow	 100	80	1	Ven
3	4	Charmander	Blaze	 50	65	1	Cha
4	5	Charmeleon	Blaze	 65	80	1	Cha
5	6	Charizard	Blaze	 85	100	1	Cha
6	7	Squirtle	Water	 64	43	1	Squ
7	8	Wartortle	Water	 80	58	1	War
8	9	Blastoise	Water	 105	78	1	Bla
9	10	Caterpie	Bug	 20	45	1	Cat

```
[10 rows x 14 columns]
```

```
df_pokemon['Type2'] = df_pokemon['Type2'].replace({'Poison' :
'Venom'})
df_pokemon.head()
```

```
ID
               Name Form
                            Type1
                                     ... Sp. Def
                                                   Speed
                                                           Generation
                                                                         Name2
0
    1
         Bulbasaur
                           Meadow
                                               65
                                                       45
                                                                            Bul
                                                                      1
1
    2
           Ivysaur
                           Meadow
                                               80
                                                       60
                                                                      1
                                                                            Ivy
                                     . . .
2
    3
          Venusaur
                           Meadow
                                              100
                                                       80
                                                                      1
                                                                            Ven
                                     . . .
3
    4
                                                       65
                                                                      1
        Charmander
                            Blaze
                                               50
                                                                            Cha
                                     . . .
4
    5
        Charmeleon
                            Blaze
                                               65
                                                       80
                                                                      1
                                                                            Cha
                                     . . .
[5 rows x 14 columns]
df pokemon['Type2'] = df pokemon['Type2'].replace(['Venom' ,
'Dragon'] , 'DANGER')
df pokemon.head(10)
   ID
               Name Form
                                                   Speed
                                                           Generation
                                                                         Name2
                            Type1
                                     ... Sp. Def
                           Meadow
0
    1
         Bulbasaur
                                                       45
                                                                            Bul
                                     . . .
                                               65
                                                                      1
    2
                                                                      1
1
           Ivysaur
                           Meadow
                                               80
                                                       60
                                                                            Ivy
                                     . . .
2
    3
          Venusaur
                           Meadow
                                                       80
                                                                      1
                                                                            Ven
                                              100
                                     . . .
3
    4
                                                                      1
        Charmander
                            Blaze
                                               50
                                                       65
                                                                            Cha
4
    5
        Charmeleon
                            Blaze
                                               65
                                                       80
                                                                      1
                                                                            Cha
                                     . . .
5
                            Blaze
    6
         Charizard
                                               85
                                                      100
                                                                      1
                                                                            Cha
                                     . . .
6
                                                                      1
    7
          Squirtle
                            Water
                                               64
                                                       43
                                                                            Squ
                                     . . .
7
         Wartortle
                                                       58
                                                                      1
    8
                            Water
                                               80
                                                                           War
                                     . . .
8
    9
         Blastoise
                            Water
                                              105
                                                       78
                                                                      1
                                                                            Bla
                                     . . .
9
   10
                                               20
                                                       45
                                                                      1
                                                                            Cat
          Caterpie
                               Bug
[10 rows x 14 columns]
df_pokemon.loc[df_pokemon['Type2'] == 'DANGER' , 'Name2'] = np.NaN
df pokemon.head(10)
   ID
               Name Form
                            Type1
                                     ... Sp. Def
                                                   Speed
                                                           Generation
                                                                         Name2
    1
                           Meadow
0
         Bulbasaur
                                               65
                                                       45
                                                                      1
                                                                            NaN
    2
1
           Ivysaur
                           Meadow
                                               80
                                                       60
                                                                      1
                                                                            NaN
                                     . . .
2
    3
                           Meadow
                                              100
                                                       80
                                                                      1
                                                                            NaN
          Venusaur
                                     . . .
3
    4
        Charmander
                            Blaze
                                               50
                                                       65
                                                                      1
                                                                            Cha
                                     . . .
4
    5
        Charmeleon
                            Blaze
                                               65
                                                       80
                                                                      1
                                                                            Cha
                                     . . .
5
                                                                      1
    6
         Charizard
                            Blaze
                                               85
                                                      100
                                                                            Cha
                                     . . .
6
    7
          Squirtle
                            Water
                                               64
                                                       43
                                                                      1
                                                                            Squ
                                     . . .
7
    8
         Wartortle
                                               80
                                                       58
                                                                      1
                                                                           War
                            Water
                                     . . .
8
    9
         Blastoise
                            Water
                                              105
                                                       78
                                                                      1
                                                                            Bla
9
   10
          Caterpie
                                               20
                                                       45
                                                                      1
                                                                            Cat
                               Bug
                                     . . .
[10 rows x 14 columns]
df pokemon.loc[df pokemon['Total']>400 , ['Name2' , 'Legendary']] =
'ALERT'
df pokemon.head(10)
   ID
               Name Form
                            Type1
                                     ... Speed Generation
                                                               Name2
Legendary
    1
         Bulbasaur
                           Meadow
                                             45
                                                            1
                                                                 NaN
                                     . . .
```

NaN 1 2	Ivysaur	Meadow		60	1	NaN	
NaN 2 3	Venusaur	Meadow		80	1	ALERT	
ALERT 3 4	Charmander	Blaze		65	1	Cha	
NaN 4 5 NaN	Charmeleon	Blaze		80	1	Cha	
5 6 ALERT	Charizard	Blaze		100	1	ALERT	
6 7 NaN	Squirtle	Water		43	1	Squ	
7 8 NaN	Wartortle	Water		58	1	War	
8 9 ALERT	Blastoise	Water		78	1	ALERT	
9 10 NaN	Caterpie	Bug		45	1	Cat	
[10 ro	ws x 15 columns]						
['ALER	emon.loc[df_poke T-1' , 'ALERT-2 emon.head(10)		l'] :	> 400 ,	['Legendary	' , 'Nam	ne2']] =
ID	Name Form	Type1		Speed	Generation	Name2	
Legend 0 1 NaN	Bulbasaur	Meadow		45	1	NaN	
1 2 NaN	Ivysaur	Meadow		60	1	NaN	
2 3 ALERT-	Venusaur 1	Meadow		80	1	ALERT-2	
3 4 NaN	Charmander	Blaze		65	1	Cha	
4 5 NaN	Charmeleon	Blaze		80	1	Cha	
5 6 ALERT-	Charizard 1	Blaze		100	1	ALERT - 2	
6 7 NaN	Squirtle	Water		43	1	Squ	
7 8 NaN	Wartortle	Water		58	1	War	
8 9 ALERT-	Blastoise 1	Water		78	1	ALERT - 2	
9 10 NaN	Caterpie	Bug		45	1	Cat	
[10 ro	ws x 15 columns]		a	nuurag.	A Edlabadkar edlabadkar@gi 9546 ŵ 98907		

```
df = pd.read csv('poke updated1.csv')
df.head(5)
   ID
              Name Form
                         Type1
                                 ... Sp. Atk
                                               Sp. Def
                                                         Speed
                                                                Generation
0
    1
        Bulbasaur
                         Grass
                                                            45
                                           65
                                                     65
                                                                          1
    2
           Ivysaur
                         Grass
                                                     80
                                                            60
                                                                          1
1
                                           80
                                 . . .
2
    3
                                                    100
         Venusaur
                         Grass
                                          100
                                                            80
                                                                          1
                                 . . .
3
    4
       Charmander
                          Fire
                                           60
                                                     50
                                                            65
                                                                          1
                                 . . .
4
       Charmeleon
    5
                           Fire
                                           80
                                                     65
                                                            80
                                                                          1
[5 rows x 13 columns]
df.groupby(['Type1']).mean().head(10)
                   ID
                             Total
                                            HP
                                                        Sp. Def
                                                                      Speed
Generation
Type1
                                                 . . .
          399.691358
                       384.382716
                                    57.024691
                                                      65.074074
                                                                  63.259259
Bug
4.037037
          476.260870
                       441.478261
                                    70.065217
Dark
                                                      68.369565
                                                                  76.695652
                                                 . . .
5.434783
          549.219512
                       560.195122
                                    85.243902
                                                      84.097561
                                                                  84.341463
Dragon
                                                 . . .
5.073171
Electric
          474.967742
                       430.241935
                                    62.241935
                                                      71.661290
                                                                 86.758065
                                                 . . .
4.645161
Fairy
          541.772727
                       431.590909
                                    72.954545
                                                      89.181818
                                                                  59.545455
4.954545
          499.809524
Fighting
                       482.714286
                                    72.928571
                                                      68.214286
                                                                  71.333333
5.047619
Fire
          401.923077
                       464.184615
                                    69.430769
                                                      71.292308
                                                                  74.369231
4.230769
                                                      70.000000
Flying
          752.750000
                       427.750000
                                    69.625000
                                                                  87.000000
6.750000
Ghost
          549.976190
                       441.000000
                                    64.809524
                                                      78.809524
                                                                  63.047619
                                                 . . .
5.166667
          438.648352
                       429.967033
                                    66.736264
                                                      70.054945
                                                                  59.769231
Grass
4.395604
[10 rows x 9 columns]
df.groupby(['Type1']).mean().sort values('Attack',ascending=False).hea
d(10)
                   ΙD
                             Total
                                            HP
                                                        Sp. Def
                                                                      Speed
Generation
Type1
                                                 . . .
          549.219512
                       560.195122
                                    85.243902
                                                      84.097561
Dragon
                                                                 84.341463
5.073171
Fighting
          499.809524
                       482.714286
                                    72.928571
                                                      68.214286
                                                                 71.333333
                                                . . .
```

5.047619 Ground 4.195122	408.170732	470.000000	70.853659		64.317073	61.804878
Steel 5.138889	520.222222	513.638889	69.527778		76.416667	53.638889
Rock	462.700000	477.100000	67.100000		72.800000	57.666667
4.616667 Fire	401.923077	464.184615	69.430769		71.292308	74.369231
4.230769 Dark	476.260870	441.478261	70.065217		68.369565	76.695652
5.434783 Ice	451.378378	442.756757	72.270270		72.729730	66.135135
4.945946 Psychic	468.864198	470.765432	73.456790		87.864198	80.432099
4.876543 Poison 4.073171	341.780488	447.707317	74.975610		72.731707	65.926829
[10 rows	x 9 columns]					
df.groupb ad(10)	y(['Type1'])	.mean().sort	_values('De	fense	',ascending	=False).he
Conomotio	ID	Total	НР		Sp. Def	Speed
Generatio Type1	П					
Steel 5.138889	520.222222	513.638889				
2.130009		3131030003	69.527778		76.416667	53.638889
Rock	462.700000	477.100000	69.527778		76.416667 72.800000	53.63888957.666667
4.616667 Ground						
4.616667 Ground 4.195122 Dragon	462.700000	477.100000	67.100000		72.800000	57.666667
4.616667 Ground 4.195122 Dragon 5.073171 Ghost	462.700000 408.170732	477.100000 470.000000	67.100000 70.853659		72.800000 64.317073	57.666667 61.804878
4.616667 Ground 4.195122 Dragon 5.073171 Ghost 5.166667 Poison	462.700000 408.170732 549.219512	477.100000 470.000000 560.195122	67.100000 70.853659 85.243902		72.800000 64.317073 84.097561	57.666667 61.804878 84.341463
4.616667 Ground 4.195122 Dragon 5.073171 Ghost 5.166667 Poison 4.073171 Fighting	462.700000 408.170732 549.219512 549.976190	477.100000 470.000000 560.195122 441.000000	67.100000 70.853659 85.243902 64.809524		72.800000 64.317073 84.097561 78.809524	57.666667 61.804878 84.341463 63.047619
4.616667 Ground 4.195122 Dragon 5.073171 Ghost 5.166667 Poison 4.073171 Fighting 5.047619 Water	462.700000 408.170732 549.219512 549.976190 341.780488	477.100000 470.000000 560.195122 441.000000 447.707317	67.100000 70.853659 85.243902 64.809524 74.975610		72.800000 64.317073 84.097561 78.809524 72.731707	57.666667 61.804878 84.341463 63.047619 65.926829
4.616667 Ground 4.195122 Dragon 5.073171 Ghost 5.166667 Poison 4.073171 Fighting 5.047619 Water 3.738806 Ice	462.700000 408.170732 549.219512 549.976190 341.780488 499.809524	477.100000 470.000000 560.195122 441.000000 447.707317 482.714286	67.100000 70.853659 85.243902 64.809524 74.975610 72.928571		72.800000 64.317073 84.097561 78.809524 72.731707 68.214286	57.666667 61.804878 84.341463 63.047619 65.926829 71.333333
4.616667 Ground 4.195122 Dragon 5.073171 Ghost 5.166667 Poison 4.073171 Fighting 5.047619 Water 3.738806	462.700000 408.170732 549.219512 549.976190 341.780488 499.809524 377.619403	477.100000 470.000000 560.195122 441.000000 447.707317 482.714286 439.888060	67.100000 70.853659 85.243902 64.809524 74.975610 72.928571 70.888060		72.800000 64.317073 84.097561 78.809524 72.731707 68.214286 71.395522	57.666667 61.804878 84.341463 63.047619 65.926829 71.333333 66.335821

[10 rows x 9 columns]

```
df.groupby(['Type1']).mean().sort values('Speed',ascending=False).head
(10)
                  ID
                           Total
                                          HP
                                                     Sp. Def
                                                                   Speed
Generation
Type1
          752.750000
                      427.750000
                                  69.625000
                                                   70.000000
                                                              87.000000
Flying
6.750000
Electric
          474.967742
                      430.241935
                                  62.241935
                                                   71.661290
                                                              86.758065
                                              . . .
4.645161
Dragon
          549.219512
                      560.195122
                                  85.243902
                                                   84.097561
                                                               84.341463
                                              . . .
5.073171
                      470.765432
Psychic
          468.864198
                                  73.456790
                                              . . .
                                                   87.864198
                                                              80.432099
4.876543
          476.260870
Dark
                      441.478261
                                   70.065217
                                                   68.369565
                                                               76.695652
5.434783
Fire
          401.923077
                      464.184615
                                  69.430769
                                                   71.292308
                                                              74.369231
4.230769
Fighting
          499.809524
                      482.714286
                                  72.928571
                                                   68.214286
                                                              71.333333
                                              . . .
5.047619
Normal
          380.547826
                      423.017391
                                   77.052174
                                                   64.486957
                                                               69.678261
3.782609
Water
          377.619403
                      439.888060
                                  70.888060
                                                   71.395522
                                                              66.335821
                                              . . .
3.738806
                      442.756757
                                  72.270270
Ice
          451.378378
                                              . . .
                                                   72.729730
                                                              66.135135
4.945946
[10 rows x 9 columns]
df.sum()
ID
                                                          460604
              BulbasaurIvysaurVenusaurCharmanderCharmeleonCh...
Name
Form
                                           Female
                                                   Male
              GrassGrassGrassFireFireFireWaterWaterBugB...
Type1
Type2
              PoisonPoison Flying
                                              FlyingPoisonPoi...
Total
                                                          468947
HP
                                                           73221
Attack
                                                           84088
Defense
                                                           78021
Sp. Atk
                                                           76308
Sp. Def
                                                           75542
Speed
                                                           71904
Generation
                                                             4655
dtype: object
df.groupby(['Type2']).sum().head(5)
              ID
                   Total
                             HP
                                               Sp. Atk
                                                        Sp. Def
                                  Attack
Generation
```

32470

449

2065

2475

608

Type2								
2072	211000	207355	33198	37217		33130	33245	
2073 Bug	6095	3461	516	650		540	649	
56 Dark	12485	14196	2079	2961		2177	1884	
139 Dragon	19697	18223	2970	3004		3290	2701	
193 Electric 59	4494	5199	838	883		817	731	
[5 rows x	9 colum	ins]						
df.count()							
ID Name Form Type1 Type2 Total HP Attack Defense Sp. Atk Sp. Def Speed Generatio dtype: in	ID 1045 Name 1045 Form 1045 Typel 1045 Type2 1045 Total 1045 HP 1045 Attack 1045 Defense 1045 Sp. Atk 1045 Sp. Def 1045 Speed 1045 Generation 1045							
df['count df.groupb		2']).cou	ınt()['c	ount1']				
Type2 Bug Dark Dragon Electric Fairy Fighting Fire Flying Ghost Grass Ground Ice Normal Poison	492 9 27 32 11 41 32 17 112 26 28 39 20 11 38							

```
Psychic
              40
Rock
              15
Steel
              35
Water
              20
Name: count1, dtype: int64
df['count1'] = 0
df.groupby(['Type1']).count()['count1']
Type1
Bug
              81
Dark
              46
              41
Dragon
Electric
              62
Fairy
              22
Fighting
              42
Fire
              65
Flying
               8
              42
Ghost
Grass
              91
Ground
              41
Ice
              37
Normal
             115
Poison
              41
              81
Psychic
Rock
              60
Steel
              36
Water
             134
Name: count1, dtype: int64
df['count1'] = 0
df.groupby(['Type1','Type2']).count()['count1']
df.columns
Index(['ID', 'Name', 'Form', 'Type1', 'Type2', 'Total', 'HP',
'Attack',
        'Defense', 'Sp. Atk', 'Sp. Def', 'Speed', 'Generation',
'count1'l.
      dtype='object')
#Loading Data in Chunks
for df in pd.read csv('poke updated1.csv', chunksize=10):
  print(df)
   ID
              Name Form
                         Type1
                                 ... Sp. Atk
                                               Sp. Def
                                                         Speed
                                                                Generation
0
    1
        Bulbasaur
                          Grass
                                 . . .
                                           65
                                                     65
                                                            45
                                                                           1
                                                                           1
1
    2
           Ivysaur
                          Grass
                                           80
                                                     80
                                                            60
                                 . . .
2
         Venusaur
                                                                          1
    3
                                                    100
                                                            80
                          Grass
                                          100
                                 . . .
3
    4
       Charmander
                           Fire
                                           60
                                                     50
                                                            65
                                                                          1
                                 . . .
4
    5
                                                                          1
       Charmeleon
                           Fire
                                                     65
                                                            80
                                           80
                                 . . .
5
    6
        Charizard
                           Fire
                                                     85
                                                                           1
                                          109
                                                           100
                                 . . .
```

6 7 8 9	7 8 9 10	Squirtle Wartortle Blastoise Caterpie	Water Water Water Bug		6 8	0 5 5 0	1	64 80 05 20	43 58 78 45		1 1 1 1
_	ID		_	l	Sp.	Atk	Sp.	Def	Spe	ed	
10	erat 11	Metapod	Bug	j		25		25		30	
1 11	12	Butterfree	Bug	,		90		80		70	
1 12	13	Weedle	Bug	j		20		20		50	
1 13	14	Kakuna	Buç	,		25		25		35	
1 14	15	Beedrill	Bug	j		45		80		75	
1 15	16	Pidgey	Norma			35		35		56	
1 16	17	Pidgeotto	Norma			50		50		71	
1 17	18	Pidgeot	Norma			70		70	1	01	
1 18	19	Rattata	Norma			25		35		72	
1 19 1	20	Raticate	Norma			50		70		97	
_	ID	s x 13 colu Name		Type1		Sp.	Atk	Sp.	Def	Speed	
20	erat 21	spearow	No	rmal			31		31	70	
	22	Fearow	No	rmal			61		61	100	
1 22	23	Ekans	Po	oison			40		54	55	
1 23	24	Arbok	Po	oison			65		79	80	
1 24	25	Pikachu	Elec	ctric			50		50	90	
1 25	26	Raichu	Elec	ctric			90		80	110	
1 26	27	Sandshrew	Gı	ound			20		30	40	
1 27	28	Sandslash	Gı	ound			45		55	65	
1 28 1	29	Nidoran	Female Po	oison			40		40	41	

29 1	30	Nidorina	Po	ison		55	55	56
_	ID		m Type	1	Sp. Atk	Sp. Def	Speed	
30 1	erat: 31	Nidoqueen	Poiso	n	75	85	76	
31 1	32	Nidoran Mal	e Poiso	n	40	40	50	
32 1	33	Nidorino	Poiso	n	55	55	65	
33 1	34	Nidoking	Poiso	n	85	75	85	
34 1	35	Clefairy	Fair	y	60	65	35	
35 1	36	Clefable	Fair	y	95	90	60	
36 1	37	Vulpix	Fire	e	50	65	65	
37 1	38	Ninetales	Fire	e	81	100	100	
38 1	39	Jigglypuff	Norma	ι	45	25	20	
39 1	40	Wigglytuff	Norma	ι	85	50	45	
_	ID	s x 13 columns] Name Form	Type1	S _l	p. Atk S	p. Def S	Speed	
40 1	erat: 41	Zubat	Poison		30	40	55	
41 1	42	Golbat	Poison		65	75	90	
42 1	43	Oddish	Grass		75	65	30	
43 1	44	Gloom	Grass		85	75	40	
44 1	45	Vileplume	Grass		110	90	50	
45 1	46	Paras	Bug		45	55	25	
46 1	47	Parasect	Bug		60	80	30	
47 1	48	Venonat	Bug		40	55	45	
48 1	49	Venomoth	Bug		90	75	90	
49 1	50	Diglett	Ground		35	45	95	

[10	row ID	s x 13 columns] Name Form		Sn. Atk	Sp. Def	Speed
Gene	erat		1,7001	 Spi Nek	Spr BC	эрсси
50	51	Dugtrio	Ground	 50	70	120
1 51	52	Meowth	Normal	 40	40	90
1 52	53	Persian	Normal	 65	65	115
1 53 1	54	Psyduck	Water	 65	50	55
54 1	55	Golduck	Water	 95	80	85
55 1	56	Mankey	Fighting	 35	45	70
56	57	Primeape	Fighting	 60	70	95
1 57 1	58	Growlithe	Fire	 70	50	60
58	59	Arcanine	Fire	 100	80	95
1 59 1	60	Poliwag	Water	 40	40	90
	ID	s x 13 columns] Name For		 Sp. Atk	Sp. Def	Speed
Gene		Name For			Sp. Def	Speed 90
Gene 60 1 61	ID erat	Name For	m Type1	 50	-	•
Gene 60 1 61 1 62	ID erat 61	Name For ion Poliwhirl	m Typel Water	 50 70	50	90
Gene 60 1 61 1 62 1 63	ID erat 61 62	Name For ion Poliwhirl Poliwrath	m Typel Water Water	 50 70 105	50 90	90
Gene 60 1 61 1 62 1 63 1 64	ID erat 61 62 63	Name For ion Poliwhirl Poliwrath Abra	m Typel Water Water Psychic	 50 70 105 120	50 90 55	90 70 90
Gene 60 1 61 1 62 1 63 1 64 1 65	ID erat 61 62 63 64	Name For ion Poliwhirl Poliwrath Abra Kadabra	Typel Water Water Psychic Psychic	 50 70 105 120 135	50 90 55 70	90 70 90 105
Gene 60 1 61 1 62 1 63 1 64 1 65 1 66	ID erat 61 62 63 64 65	Name For ion Poliwhirl Poliwrath Abra Kadabra Alakazam	Water Water Psychic Psychic Psychic	 50 70 105 120 135 35	50 90 55 70 95	90 70 90 105 120
Gene 60 1 61 1 62 1 63 1 64 1 65 1 66 1	ID erat 61 62 63 64 65 66	Name For ion Poliwhirl Poliwrath Abra Kadabra Alakazam Machop	Typel Water Water Psychic Psychic Psychic Fighting	 50 70 105 120 135 35	50 90 55 70 95 35	90 70 90 105 120 35
Gene 60 1 61 1 62 1 63 1 64 1 65 1 66 1	ID erat 61 62 63 64 65 66 67	Name Forion Poliwhirl Poliwrath Abra Kadabra Alakazam Machop Machoke	Water Water Psychic Psychic Psychic Fighting Fighting	 50 70 105 120 135 35 50	50 90 55 70 95 35 60	90 70 90 105 120 35 45

[10 rows x 13 columns]

6	ID		Form	Type1		Sp.	Atk	S _I	p. De	ef S	peed	
70	erat 71	victreebel		Grass			100)	7	70	70	
1 71	72	Tentacool		Water			50)	16	00	70	
1 72	73	Tentacruel		Water			86)	12	20	100	
1 73	74	Geodude		Rock			36)	3	30	20	
1 74	75	Graveler		Rock			45	•	2	15	35	
1 75	76	Golem		Rock			55	•	6	55	45	
1 76	77	Ponyta		Fire			65	•	6	55	90	
1 77 1	78	Rapidash		Fire			86)	8	30	105	
78 1	79	Slowpoke		Water			40)	4	10	15	
79 1	80	Slowbro		Water			100)	8	30	30	
[10	ID	s x 13 colun Name	_	Туре	1		Sp.	Atk	Sp.	Def	Speed	d
80	erat 81	ion Magnemite		Electri	.C			95		55	4!	5
1 81	82	Magneton		Electri	.C			120		70	70	0
1 82	83	Farfetch'd		Norma	ıl			58		62	60	0
1 83	84	Doduo		Norma	ıl			35		35	7!	5
1 84	85	Dodrio		Norma	ıl			60		60	110	0
1 85	86	Seel		Wate	r			45		70	4!	5
1 86	87	Dewgong		Wate	r			70		95	70	0
1 87 1	88	Grimer		Poiso	n			40		50	2!	5
88 1	89	Muk		Poiso	n			65		100	50	0
89 1	90	Shellder		Wate	er			45		25	40	0
_	row ID erat		_	Type1		S	р. А	itk	Sp.	Def	Speed	

90	91	Cloyster	Water		85	45	70
1 91 1	92	Gastly	Ghost		100	35	80
92 1	93	Haunter	Ghost		115	55	95
93 1	94	Gengar	Ghost		130	75	110
94 1	95	0nix	Rock		30	45	70
95	96	Drowzee	Psychic		43	90	42
1 96	97	Hypno	Psychic		73	115	67
1 97 1	98	Krabby	Water		25	25	50
98 1	99	Kingler	Water		50	50	75
99 1	100	Voltorb	Electric		55	55	100
_	ID		_	el	Sp. Atk	Sp. Def	Speed
Gene 100 1	eration 101	on Electrode	Electr	ic	80	80	150
101 1	102	Exeggcute	Gra	ss	60	45	40
102 1	103	Exeggutor	Gra	ss	125	75	55
103 1	104	Cubone	Grou	nd	40	50	35
104 1	105	Marowak	Grou	nd	50	80	45
105 1	106	Hitmonlee	Fighti	ng	35	110	87
106 1	107	Hitmonchan	Fighti	ng	35	110	76
107 1	108	Lickitung	Norm	al	60	75	30
108 1	109	Koffing	Pois	on	60	45	35
109 1	110	Weezing	Pois	on	85	70	60
_	ID		s] Form Typel	S _l	p. Atk S	Sp. Def	Speed
Gene 110 1	eratio 111	on Rhyhorn	Ground		30	30	25

111	112	Rhydon	Ground		45	45	40
1 112	113	Chansey	Normal		35	105	50
1 113	114	Tangela	Grass		100	40	60
1 114	115	Kangaskhan	Normal		40	80	90
1 115	116	Horsea	Water		70	25	60
1 116	117	Seadra	Water		95	45	85
1 117	118	Goldeen	Water		35	50	63
1 118 1	119	Seaking	Water		65	80	68
119 1	120	Staryu	Water		70	55	85
[10	rows ID	x 13 columns] Name Form	Туре	1	Sp. Atk	Sp. Def	Speed
Gene 120 1	ratio 121	n Starmie	Wate	er	100	85	115
121 1	122	Mr. Mime	Psychi	.c	100	120	90
122 1	123	Scyther	Bu	ıg	55	80	105
123 1	124	Jynx	Ic	e	115	95	95
124 1	125	Electabuzz	Electri	.c	95	85	105
125 1	126	Magmar	Fir	e	100	85	93
126 1	127	Pinsir	Bu	ıg	55	70	85
127 1	128	Tauros	Norma	ıl	40	70	110
128 1	129	Magikarp	Wate	er	15	20	80
129 1	130	Gyarados	Wate	er	60	100	81
_	ID	x 13 columns] Name Form	Type1	S	p. Atk S	p. Def S	Speed
130	ratio 131	Lapras	Water		85	95	60
1 131 1	132	Ditto	Normal		48	48	48

132	133	Eevee	Normal		45	65	55
1 133 1	134	Vaporeon	Water		110	95	65
134 1	135	Jolteon	Electric		110	95	130
135 1	136	Flareon	Fire		95	110	65
136 1	137	Porygon	Normal		85	75	40
137 1	138	0manyte	Rock		90	55	35
138 1	139	0mastar	Rock		115	70	55
139 1	140	Kabuto	Rock		55	45	55
	ID	x 13 columns Name		1	Sp. Atk	Sp. Def	Speed
140 1	eratio 141	Kabutops	Roc	k	65	70	80
141 1	142	Aerodactyl	Roc	k	60	75	130
142 1	143	Snorlax	Norma	ι	65	110	30
143 1	144	Articuno	Ic	e	95	125	85
144 1	145	Zapdos	Electri	с	125	90	100
145 1	146	Moltres	Fir	e	125	85	90
146 1	147	Dratini	Drago	n	50	50	50
	148	Dragonair	Drago	n	70	70	70
148 1	149	Dragonite	Drago	n	100	100	80
149 1	150	Mewtwo	Psychi	с	154	90	130
_	ID	x 13 columns Name	_		Sp. Atk	Sp. Def	Speed
150	eratio 151	n Mew	Psychic		100	100	100
1 151	152	Chikorita	Grass		49	65	45
2 152 2	153	Bayleef	Grass		63	80	60

153 2	154	Meganium	Gra	SS			83		100	80
154 2	155	Cyndaquil	Fi	re			60		50	65
155 2	156	Quilava	Fi	re			80		65	80
156 2	157	Typhlosion	Fi	re			109		85	100
157 2	158	Totodile	Wat	er			44		48	43
158 2	159	Croconaw	Wat	er			59		63	58
159 2	160	Feraligatr	Wat	er			79		83	78
_	ID	x 13 columns] Name Form	Type1		. Sp	. Atk	Sp.	Def	: Sp	peed
160 2	ratio 161	Sentret	Normal			35		45)	20
161 2	162	Furret	Normal			45		55	<u>;</u>	90
162 2	163	Hoothoot	Normal			36		56	j	50
163 2	164	Noctowl	Normal			86		96	<u>)</u>	70
164 2	165	Ledyba	Bug			40		80)	55
165 2	166	Ledian	Bug			55		110)	85
166 2	167	Spinarak	Bug			40		40)	30
167 2	168	Ariados	Bug		•	60		70)	40
168 2	169	Crobat	Poison			70		80	i	130
169 2	170	Chinchou	Water			56		56	,	67
_	ID	x 13 columns] Name Form	Тур	e1		Sp.	Atk	Sp.	Def	Speed
170 2	ratio 171	Lanturn	Wat	er			76		76	67
171 2	172	Pichu	Electr	ic			35		35	60
172 2	173	Cleffa	Fai	ry			45		55	15
173 2	174	Igglybuff	Norm	al			40		20	15

174 2	175	Togepi	Fairy	 40	65	20
175 2	176	Togetic	Fairy	 80	105	40
176 2	177	Natu	Psychic	 70	45	70
177 2	178	Xatu	Psychic	 95	70	95
178 2	179	Mareep	Electric	 65	45	35
179 2	180	Flaaffy	Electric	 80	60	45
_	rows ID eratio	x 13 columns] Name For	m Typel	 Sp. Atk	Sp. Def	Speed
180	181	Ampharos	Electric	 115	90	55
181	182	Bellossom	Grass	 90	100	50
182	183	Marill	Water	 20	50	40
183 2	184	Azumarill	Water	 60	80	50
184 2	185	Sudowoodo	Rock	 30	65	30
185 2	186	Politoed	Water	 90	100	70
186 2	187	Hoppip	Grass	 35	55	50
187 2	188	Skiploom	Grass	 45	65	80
188 2	189	Jumpluff	Grass	 55	95	110
189 2	190	Aipom	Normal	 40	55	85
-	ID	x 13 columns] Name Fo	rm Typel	 Sp. Atk	Sp. Def	Speed
190	eratio 191	Sunkern	Grass	 30	30	30
2 191	192	Sunflora	Grass	 105	85	30
2 192	193	Yanma	Bug	 75	45	95
2 193 2	194	Wooper	Water	 25	25	15
194 2	195	Quagsire	Water	 65	65	35

105	106	Ганаан	Davabáa		120	0.5	110
195 2	196	Espeon	Psychic		130	95	110
196 2	197	Umbreon	Dark		60	130	65
197 2	198	Murkrow	Dark		85	42	91
198	199	Slowking	Water	·	100	110	30
2 199 2	200	Misdreavus	Ghost		85	85	85
-	rows ID eratio	x 13 columns] Name Form	Type1		Sp. Atk	Sp. Def	Speed
200	201	Unown	Psychic		72	48	48
201	202	Wobbuffet	Psychic		33	58	33
202	203	Girafarig	Normal		90	65	85
203	204	Pineco	Bug		35	35	15
204	205	Forretress	Bug		60	60	40
205 2	206	Dunsparce	Normal		65	65	45
206 2	207	Gligar	Ground		35	65	85
207	208	Steelix	Steel		55	65	30
208	209	Snubbull	Fairy		40	40	30
209	210	Granbull	Fairy		60	60	45
_	ID	x 13 columns] Name Form	Type1	S _I	o. Atk S	Sp. Def S	Speed
Gene 210	eration 211	on Qwilfish	Water		55	55	85
2 211	212	Scizor	Bug		55	80	65
2 212	213	Shuckle	Bug		10	230	5
2 213	214	Heracross	Bug		40	95	85
2 214	215	Sneasel	Dark		35	75	115
2 215 2	216	Teddiursa	Normal		50	50	40

216	217	Ursaring	Normal .		75	75	55
2 217 2	218	Slugma	Fire .		70	40	20
218 2	219	Magcargo	Fire .		90	80	30
219 2	220	Swinub	Ice .		30	30	50
_	rows ID eratio		Typel	. Sp	. Atk Sp	o. Def S	peed
220 2		Piloswine	Ice		60	60	50
2 221 2	222	Corsola	Water		65	95	35
222	223	Remoraid	Water		65	35	65
2 223 2	224	Octillery	Water		105	75	45
224	225	Delibird	Ice		65	45	75
225 2	226	Mantine	Water		80	140	70
226 2	227	Skarmory	Steel		40	70	70
227 2	228	Houndour	Dark		80	50	65
228 2	229	Houndoom	Dark		110	80	95
229 2	230	Kingdra	Water		95	95	85
	ID	x 13 columns] Name Form	Type1		Sp. Atk	Sp. Def	Speed
230	eratio 231	n Phanpy	Ground		40	40	40
2 231	232	Donphan	Ground		60	60	50
2 2 3 2	233	Porygon2	Normal		105	95	60
2 233	234	Stantler	Normal		85	65	85
2 2 3 4	235	Smeargle	Normal		20	45	75
2 235	236	Tyrogue	Fighting		35	35	35
2 236 2	237	Hitmontop	Fighting		35	110	70

237 2	238	Smoochum	Ice		85	65	65
238 2	239	Elekid	Electric		65	55	95
2 2 2 2	240	Magby	Fire		70	55	83
[10	rows ID	x 13 columns] Name Form	Type1		Sp. Atk	Sp. Def	Speed
Gene 240 2	eratio 241	on Miltank	Normal		40	70	100
241 2	242	Blissey	Normal		75	135	55
242 2	243	Raikou	Electric		115	100	115
243 2	244	Entei	Fire		90	75	100
244 2	245	Suicune	Water		90	115	85
245 2	246	Larvitar	Rock		45	50	41
246 2	247	Pupitar	Rock		65	70	51
247 2	248	Tyranitar	Rock		95	100	61
248 2	249	Lugia	Psychic		90	154	110
249 2	250	Ho-oh	Fire		110	154	90
_	ID		Type1	:	Sp. Atk	Sp. Def	Speed
250	eration 251	Celebi	Psychic		100	100	100
2 251	252	Treecko	Grass		65	55	70
3 252	253	Grovyle	Grass		85	65	95
3 253	254	Sceptile	Grass		105	85	120
3 254	255	Torchic	Fire		70	50	45
3 255	256	Combusken	Fire		85	60	55
3 256	257	Blaziken	Fire		110	70	80
3 257 3	258	Mudkip	Water		50	50	40

258	259	Marshtomp	Water	 . 60	70	50
3 259 3	260	Swampert	Water	 . 85	90	60
_	ID	x 13 columns] Name Form	Type1	 Sp. Atk	Sp. Def	Speed
Gene 260 3	eratio 261		Dark	 30	30	35
261 3	262	Mightyena	Dark	 60	60	70
262 3	263	Zigzagoon	Normal	 30	41	60
263 3	264	Linoone	Normal	 50	61	100
264 3	265	Wurmple	Bug	 20	30	20
265 3	266	Silcoon	Bug	 25	25	15
266 3	267	Beautifly	Bug	 100	50	65
267 3	268	Cascoon	Bug	 25	25	15
268 3	269	Dustox	Bug	 50	90	65
269 3	270	Lotad	Water	 40	50	30
	rows ID eratio	x 13 columns] Name Form	Type1	 Sp. Atk	Sp. Def	Speed
270 3	271	Lombre	Water	 60	70	50
271 3	272	Ludicolo	Water	 90	100	70
272 3	273	Seedot	Grass	 30	30	30
273 3	274	Nuzleaf	Grass	 60	40	60
274 3	275	Shiftry	Grass	 90	60	80
275 3	276	Taillow	Normal	 30	30	85
276 3	277	Swellow	Normal	 75	50	125
277 3	278	Wingull	Water	 55	30	85
278 3	279	Pelipper	Water	 95	70	65

279 3	280	Ralts	Psychic .	••	45	35	40
	ID	x 13 columns; Name N			Sp. Atk	Sp. Def	Speed
Gene 280 3	eratio 281	on Kirlia	Psychic		65	55	50
281 3	282	Gardevoir	Psychic		125	115	80
282	283	Surskit	Bug		50	52	65
283 3	284	Masquerain	Bug		100	82	80
284 3	285	Shroomish	Grass		40	60	35
285 3	286	Breloom	Grass		60	60	70
286 3	287	Slakoth	Normal		35	35	30
287 3	288	Vigoroth	Normal		55	55	90
288 3	289	Slaking	Normal		95	65	100
289 3	290	Nincada	Bug		30	30	40
_	rows ID eratio	x 13 columns; Name Fo			Sp. Atk	Sp. Def	Speed
290 3	291	Ninjask	Bug		50	50	160
291 3	292	Shedinja	Bug		30	30	40
292 3	293	Whismur	Normal		51	23	28
293 3	294	Loudred	Normal		71	43	48
294 3	295	Exploud	Normal		91	73	68
295 3	296	Makuhita	Fighting		20	30	25
296 3	297	Hariyama	Fighting		40	60	50
297 3	298	Azurill	Normal		20	40	20
298 3	299	Nosepass	Rock		45	90	30
299 3	300	Skitty	Normal		35	35	50

[10	rows ID	x 13 columns] Name Form	Type1	Sp	. Atk	Sp. Def	Speed
300	eratio 301		Normal		55	55	90
301	302	Sableye	Dark		65	65	50
3 302 3	303	Mawile	Steel		55	55	50
303 3	304	Aron	Steel		40	40	30
304 3	305	Lairon	Steel		50	50	40
305 3	306	Aggron	Steel		60	60	50
306 3	307	Meditite	Fighting		40	55	60
307 3	308	Medicham	Fighting		60	75	80
308 3	309	Electrike	Electric		65	40	65
309 3	310	Manectric	Electric		105	60	105
_	ID	x 13 columns] Name Form	Type1	Sp.	Atk	Sp. Def	Speed
Gene		Name Form	Type1 Electric	Sp.	Atk 85	Sp. Def	Speed 95
Gene 310 3 311	ID eratio	Name Form		-		-	-
Gene 310 3 311 3 312	ID eratio 311	Name Form n Plusle	Electric		85	75	95
Gene 310 3 311 3 312 3 313	ID eratio 311 312	Name Form n Plusle Minun	Electric Electric		85 75	75 85	95 95
Gene 310 3 311 3 312 3 313 3 314	ID eratio 311 312 313	Name Form n Plusle Minun Volbeat	Electric Electric Bug		85 75 47	75 85 85	95 95 85
Gene 310 3 311 3 312 3 314 3 315	ID eratio 311 312 313 314	Name Form n Plusle Minun Volbeat Illumise	Electric Electric Bug Bug		85 75 47 73	75 85 85 85	95 95 85 85
Gene 310 3 311 3 312 3 314 3 315 3 316	ID eratio 311 312 313 314 315	Name Form Plusle Minun Volbeat Illumise Roselia	Electric Electric Bug Bug Grass		85 75 47 73 100	75 85 85 85 80	95 95 85 85 65
Gene 310 3 311 3 312 3 314 3 315 3 316 3 317	ID eratio 311 312 313 314 315 316	Name Form Plusle Minun Volbeat Illumise Roselia Gulpin	Electric Electric Bug Bug Grass Poison		85 75 47 73 100 43	75 85 85 85 80 53	95 95 85 85 65 40
Gene 310 3 311 3 312 3 314 3 315 3 316 3	ID eratio 311 312 313 314 315 316 317	Name Form Plusle Minun Volbeat Illumise Roselia Gulpin Swalot	Electric Electric Bug Bug Grass Poison Poison		85 75 47 73 100 43 73	75 85 85 85 80 53 83	95 95 85 85 65 40

[10 rows x 13 columns]

Cana	ID		Form	Type1		Sp. Atk	Sp. Def	Speed
320 3	321	Wailord		Water		90	45	60
321 3	322	Numel		Fire		65	45	35
322 3	323	Camerupt		Fire		105	75	40
323 3	324	Torkoal		Fire		85	70	20
324 3	325	Spoink		Psychic		70	80	60
3 325 3	326	Grumpig		Psychic		90	110	80
326 3	327	Spinda		Normal		60	60	60
3 327 3	328	Trapinch		Ground		45	45	10
328 3	329	Vibrava		Ground		50	50	70
329 3	330	Flygon		Ground		80	80	100
_	ID		nns] Form	Type1	9	Sp. Atk	Sp. Def	Speed
330	eratio 331	on Cacnea		Grass		85	40	35
3 331	332	Cacturne		Grass		115	60	55
3 332	333	Swablu		Normal		40	75	50
333	334	Altaria		Dragon		70	105	80
3 334	335	Zangoose		Normal		60	60	90
335	336	Seviper		Poison		100	60	65
3 336	337	Lunatone		Rock		95	85	70
3 337 3	338	Solrock		Rock		55	65	70
338 3	339	Barboach		Water		46	41	60
	340	Whiscash		Water		76	71	60
	ID			Type1		Sp. Atk	Sp. Def	Speed
Gene	eratio	on						

340	341	Corphish		Water	• • • • •	50		35	35	
3 341	342	Crawdaunt		Water		90		55	55	
3 342	343	Baltoy		Ground	٠	40		70	55	
3 343 3	344	Claydol		Ground	٠	70	1	20	75	
344 3	345	Lileep		Rock		61		87	23	
345 3	346	Cradily		Rock		81	1	07	43	
346 3	347	Anorith		Rock		40		50	75	
347 3	348	Armaldo		Rock		70		80	45	
348 3	349	Feebas		Water	·	10		55	80	
349 3	350	Milotic		Water	·	100	1	25	81	
	rows	x 13 colum	nc 1							
	ID	Name	15]	Form	Type1	Sp	. Atk	Sp.	Def	Speed
350	ratio 351	on Castform			Normal		70		70	70
3 351	351	Castform	Sunny	Form	Fire		70		70	70
3 352	351	Castform	Rainy	Form	Water		70		70	70
3 353 3	351	Castform	Snowy	Form	Ice		70		70	70
354 3	352	Kecleon			Normal		60		120	40
355 3	353	Shuppet			Ghost		63		33	45
356 3	354	Banette			Ghost		83		63	65
357 3	355	Duskull			Ghost		30		90	25
358 3	356	Dusclops			Ghost		60		130	25
	357	Tropius			Grass		72		87	51
[10	rows ID	x 13 colum Name		Tvpe1	9	Sp. Atk	Sp. D	ef S	peed	
	ratio	n								
360 3	358	Chimecho	ŀ	Psychic		95		90	65	

361	359	Absol	Dark		75	60	75
3 362	360	Wynaut	Psychic		23	48	23
3 363	361	Snorunt	Ice		50	50	50
3 364 3	362	Glalie	Ice		80	80	80
365 3	363	Spheal	Ice		55	50	25
366 3	364	Sealeo	Ice		75	70	45
367 3	365	Walrein	Ice		95	90	65
368 3	366	Clamperl	Water		74	55	32
369 3	367	Huntail	Water		94	75	52
_	ID	x 13 columns] Name Form	Type1		Sp. Atk	Sp. Def	Speed
370	eration 368		Water		114	75	52
3 371	369	Relicanth	Water		45	65	55
3 372 3	370	Luvdisc	Water		40	65	97
373 3	371	Bagon	Dragon		40	30	50
374 3	372	Shelgon	Dragon		60	50	50
375 3	373	Salamence	Dragon		110	80	100
	374	Beldum	Steel		35	60	30
377 3	375	Metang	Steel		55	80	50
378 3	376	Metagross	Steel		95	90	70
379 3	377	Regirock	Rock		50	100	50
	ID	x 13 columns] Name	Form	٦	ypel	. Sp. Atk	Sp. Def
Spee 380 50	ed Ge 378	eneration Regice 3			Ice	. 100	200
381 50	379	Registeel 3		9	Steel	. 75	150

382 110	380	Latias 3			Dragon		13	LO	130
383 110	381	Latios 3			Dragon		13	30	110
384 90	382	Kyogre 3			Water		15	50	140
385 90	383	Groudon 3			Ground		10	90	90
386 95	384	Rayquaza 3			Dragon		15	50	90
387 100	385	Jirachi 3			Steel		10	90	100
388 150	386	Deoxys 3	Normal	Forme	Psychic		15	50	50
389 150	386	Deoxys 3	Attack	Forme	Psychic		18	30	20
[10 390	rows ID 386	x 13 columns Name Deoxys	Defense		Sp.	Def 160	Speed 90	Gen	eration 3
391 392	386 387	Deoxys Turtwig	Speed	d Forme		90 55	180 31		3 3 4
393	388	Grotle				65	36		4
394 395	389 390	Torterra Chimchar				85 44	56 61		4 4
396	391	Monferno				52	81		4
397	392	Infernape				71	108		4
398 399	393 394	Piplup Prinplup				56 76	40 50		4 4
[10	rows ID	x 13 columns	_	Type	ı Cn	۸ + ا د	C۳	Dof	Cnood
Gene	יטו eratio		Form	Type	l Sp	. ALK	Sp.	Def	Speed
400	395	Empoleon		Wate	r	111		101	60
401 4	396	Starly		Norma	l	30		30	60
402 4	397	Staravia		Norma	l	40		40	80
403 4	398	Staraptor		Norma	l	50		60	100
404 4	399	Bidoof		Norma	l	35		40	31
405 4	400	Bibarel		Norma	l	55		60	71
406 4	401	Kricketot		Bug	g	25		41	25
407 4	402	Kricketune		Buç	g	55		51	65
408	403	Shinx	E	Electri	c	40		34	45

4 409 4	404	Luxio		Electr	ric .		66)	4	9	60
[10	rows ID	x 13 column Name	s]	Form	т	ype1		Sn	Δ+k	Sn	Def
Spee		eneration		1 01 111	'	урст		Jp.	ACK	Jp.	DCT
410 70	405	Luxray 4			Elec	tric			95		79
411	406	Budew			G	rass			50		70
55 412	407	4 Roserade			G	rass			125		105
90 413	408	4 Cranidos				Rock			30		30
58 414	409	4 Rampardos				Rock			65		50
58 415	410	4 Shieldon				Rock			42		88
30 416	411	4 Bastiodon				Rock			47		138
30		4									
417 36	412	Burmy 4				Bug	• • •		29		45
418 36	413	Wormadam 4	Plant	Cloak		Bug			79		105
419 36	413	Wormadam 4	Sandy	Cloak		Bug			59		85
		·									
[10		x 13 column	s]								
_	ID	Name		Form	Т	ype1		Sp.	Atk	Sp.	Def
Spee		eneration	- .	61 1					60		0.5
420 36	413	Wormadam 4	Trash	Cloak		Bug			69		95
421 66	414	Mothim 4				Bug			94		50
422 70	415	Combee 4				Bug			30		42
423	416	Vespiquen				Bug			80		102
40 424	417	4 Pachirisu			Elec	tric			45		90
95 425	418	4 Buizel			W	ater			60		30
85 426	419	4 Floatzel			W	ater			85		50
115 427	420	4 Cherubi			G	rass			62		53
35		4			_				- -		
428 85	421	Cherrim 4			G	rass	• • •		87		78
429	422	Shellos			W	ater			57		62

34	4

[10	rows	x 13 columns]				
Cana	ID	Name Form	Type1	 Sp. Atk	Sp. Def	Speed
430	eratio 423	on Gastrodon	Water	 92	82	39
4	123		nacei	 32	02	33
431	424	Ambipom	Normal	 60	66	115
4 432 4	425	Drifloon	Ghost	 60	44	70
433 4	426	Drifblim	Ghost	 90	54	80
434 4	427	Buneary	Normal	 44	56	85
435 4	428	Lopunny	Normal	 54	96	105
436 4	429	Mismagius	Ghost	 105	105	105
437	430	Honchkrow	Dark	 105	52	71
4 438	431	Glameow	Normal	 42	37	85
4 439 4	432	Purugly	Normal	 64	59	112
[10	rows	x 13 columns]				
	ID	Name Form	Type1	 Sp. Atk	Sp. Def	Speed
Gene	ID eratio	Name Form		-	-	-
	ID	Name Form	Type1 Psychic	Sp. Atk	Sp. Def	Speed 45
Gene 440 4 441	ID eratio	Name Form		 -	-	-
Gene 440 4 441 4 442	ID eratio 433	Name Form on Chingling	Psychic	 65	50	45
Gene 440 4 441 4 442 4 443	ID eratio 433 434	Name Form on Chingling Stunky	Psychic Poison	 65 41	50 41	45 74
Gene 440 4 441 4 442 4 443 4 444	ID eratio 433 434 435	Name Form On Chingling Stunky Skuntank	Psychic Poison Poison	 65 41 71	50 41 61	45 74 84
Gene 440 4 441 4 442 4 443 4 444 4 445	ID 433 434 435 436	Name Form On Chingling Stunky Skuntank Bronzor	Psychic Poison Poison Steel	 65 41 71 24	50 41 61 86	45 74 84 23
Gene 440 4 441 4 442 4 443 4 444 4 445 4 446	ID eratio 433 434 435 436 437	Name Form On Chingling Stunky Skuntank Bronzor Bronzong	Psychic Poison Poison Steel Steel	 65 41 71 24 79	50 41 61 86 116	45 74 84 23 33
Gene 440 4 441 442 4 443 4 444 445 4 446 4	ID 433 434 435 436 437 438	Name Form On Chingling Stunky Skuntank Bronzor Bronzong Bonsly	Psychic Poison Poison Steel Steel Rock	 65 41 71 24 79 10	50 41 61 86 116 45	45 74 84 23 33 10
Gene 440 4 441 4 442 4 443 4 444 445 4 446 4 447 4	ID 433 434 435 436 437 438 439	Name Form Chingling Stunky Skuntank Bronzor Bronzong Bonsly Mime Jr.	Psychic Poison Poison Steel Steel Rock Psychic	 65 41 71 24 79 10 70	50 41 61 86 116 45 90	45 74 84 23 33 10 60
Gene 440 4 441 4 442 4 443 4 444 4 445 4 446 4 447 4	ID 433 434 435 436 437 438 439 440	Name Form Chingling Stunky Skuntank Bronzor Bronzong Bonsly Mime Jr. Happiny	Psychic Poison Poison Steel Steel Rock Psychic Normal	 65 41 71 24 79 10 70 15	50 41 61 86 116 45 90 65	45 74 84 23 33 10 60 30

[10	rows ID	x 13 columns] Name Form	Type1	S	Sp. Atk	Sp. Def	Speed
Gene	eratio		. , , , ,		γ	op. 50.	Specu
450	443	Gible	Dragon		40	45	42
4			_				
451	444	Gabite	Dragon		50	55	82
4 452	445	Garchomp	Dragon		80	85	102
432	443	dar Crionip	Diagon	• • •	00	65	102
453	446	Munchlax	Normal		40	85	5
4							
454	447	Riolu	Fighting		35	40	60
4							
455	448	Lucario	Fighting		115	70	90
4 456	449	Uinnonotos	Cround		38	42	22
450	449	Hippopotas	Ground	• • •	30	42	32
457	450	Hippowdon	Ground		68	72	47
4		pp	0.000				
458	451	Skorupi	Poison		30	55	65
4							
459	452	Drapion	Poison		60	75	95
4							
[10	rovic	x 13 columns]					
[10	ID	Name Form	Tyne1	Sn	Δ+k	Sp. Def	Sneed
Gene	eratio		турст	Jp	, ACK	Sp. DC1	Specu
460	453	Croagunk	Poison		61	40	50
4		J					
461	454	Toxicroak	Poison		86	65	85
4			_				
462	455	Carnivine	Grass		90	72	46
4 463	456	Finneon	Water		49	61	66
4	430	TIMEON	water		43	01	00
464	457	Lumineon	Water		69	86	91
4							
465	458	Mantyke	Water		60	120	50
4		_	_				
466	459	Snover	Grass		62	60	40
4 467	460	Ahomosnov	Crace		92	85	60
407	400	Abomasnow	Grass		92	63	60
468	461	Weavile	Dark		45	85	125
4							
469	462	Magnezone	Electric		130	90	60
4							
		10 1 1					
[10		x 13 columns]	T 1	_	-ا⊥۸ م`	C	Cnaad
	ID	Name Form	ryper	5	p. ATK	Sp. Def	Speed

Gene	eratio	on							
470 4	463	Lickilicky		Norma	al	80	9	95	50
471 4	464	Rhyperior		Grour	nd	5!	5	55	40
472 4	465	Tangrowth		Gras	SS	110	9 .	50	50
4 473 4	466	Electivire	I	Electri	ic	9!	5	85	95
474	467	Magmortar		Fir	^е	12!	5	95	83
4 475	468	Togekiss		Fair	^y	120	9 1	15	80
4 476 4	469	Yanmega		Ви	ıg	116	<u> </u>	56	95
4 477 4	470	Leafeon		Gras	SS	60	9	65	95
478	471	Glaceon		Id	e	130	9	95	65
4 479 4	472	Gliscor		Grour	nd	4!	5	75	95
	rows	x 13 columns	:1						
	ID	Name	, ,	Form	Туре	1	Sp. Atk	Sp. D	ef
Spee 480	ed G€ 473	eneration Mamoswine			Ic	0	70		60
80	4/3	4			10	e	70		00
481 90	474	Porygon-Z 4			Norma	l	135		75
482 80	475	Gallade 4			Psychi	с	65	1	.15
483 40	476	Probopass 4			Roc	k	75	1	.50
484 45	477	Dusknoir 4			Ghos	t	65	1	.35
485 110	478	Froslass 4			Ic	e	80		70
486 91	479	Rotom 4			Electri	с	95		77
487 86	479	Rotom 4	Heat I	Rotom	Electri	с	105	1	.07
488 86	479	Rotom 4	Wash F	Rotom	Electri	с	105	1	.07
489 86	479		Frost F	Rotom	Electri	c	105	1	.07
[10	rows	x 13 columns	;]						
400	ID	Name	_	Form		•	Speed	Generat	_
490 491	479 479	Rotom Rotom		n Roton w Roton		107 107	86 86		4

492 493 494 495 496 497 498 499	480 481 482 483 484 485 486 487	Uxie Mesprit Azelf Dialga Palkia Heatran Regigigas Giratina	Altered Forme		130 105 70 100 120 106 110 120	95 80 115 90 100 77 100 90	4 4 4 4 4 4 4
[10	rows ID	x 13 columr Name	ns] Form	Type1		Sn Atk	Sp. Def
Spee		eneration	101111	1 y pc 1		Sp. Ack	Spi Dei
500 90	487	Giratina 4	Origin Forme	Ghost		120	100
501 85	488	Cresselia 4		Psychic		75	130
502 80	489	Phione 4		Water	`	80	80
503	490	Manaphy		Water		100	100
100 504	491	4 Darkrai		Dark		135	90
125 505	492	4 Shaymin	Land Forme	Grass		100	100
100 506	492	4 Shaymin	Sky Forme	Grass		120	75
127 507	493	4 Arceus		Normal		120	120
120 508	494	4 Victini		Psychic		100	100
100 509	495	5 Snivy		Grass		45	55
63	493	5		01 055		43	55
	ID	x 13 columr Name	_	Sp.	Atk S	Sp. Def	Speed
Gene	eratio						
510 5	496	Servine	Grass	• • •	60	75	83
511 5	497	Serperior	Grass		75	95	113
512 5	498	Tepig	Fire		45	45	45
513	499	Pignite	Fire		70	55	55
5 514	500	Emboar	Fire		100	65	65
5 515	501	0shawott	Water		63	45	45
5 516 5	502	Dewott	Water		83	60	60

517	503	Samurott		Water			108	3	76)	70
5 518 5	504	Patrat		Normal			35	5	39)	42
5 5 5	505	Watchog		Normal			66)	69)	77
_	ID			Type1		Sp	. Atl	c Sp	. Def	- Sp	eed
Gene 520 5	eratio 506	on Lillipup		Normal			25	5	45	5	55
5 521 5	507	Herdier		Normal			35	5	65	5	60
522 5	508	Stoutland		Normal			45	5	96)	80
523 5	509	Purrloin		Dark			56)	37	7	66
524 5	510	Liepard		Dark			88	3	50)	106
525 5	511	Pansage		Grass			53	3	48	3	64
526 5	512	Simisage		Grass			98	3	63	3	101
527 5	513	Pansear		Fire			53	3	48	3	64
528 5	514	Simisear		Fire			98	3	63	3	101
	515	Panpour		Water			53	3	48	3	64
_	rows ID eratio		_	Тур	e1		Sp.	Atk	Sp.	Def	Speed
530	516	Simipour		Wat	er			98		63	101
5 531 5	517	Munna	ı	Psych	ic			67		55	24
532 5	518	Musharna	ı	Psych	ic			107		95	29
533 5	519	Pidove	!	Norm	al			36		30	43
534 5	520	Tranquill		Norm	al			50		42	65
535 5	521	Unfezant		Norm	al			65		55	93
536 5	522	Blitzle	!	Electr	ic			50		32	76
5 537 5	523	Zebstrika	ı	Electr	ic			80		63	116

538 5	524	Roggenrola	Rock	 25	25	15
539 5	525	Boldore	Rock	 50	40	20
_	ID	x 13 columns] Name F		 Sp. Atk	Sp. Def	Speed
540	eration 526	on Gigalith	Rock	 60	80	25
5 541 5	527	Woobat	Psychic	 55	43	72
542 5	528	Swoobat	Psychic	 77	55	114
543 5	529	Drilbur	Ground	 30	45	68
5 5 5	530	Excadrill	Ground	 50	65	88
545 5	531	Audino	Normal	 60	86	50
546 5	532	Timburr	Fighting	 25	35	35
5 5 5	533	Gurdurr	Fighting	 40	50	40
548 5	534	Conkeldurr	Fighting	 55	65	45
549 5	535	Tympole	Water	 50	40	64
_	rows ID eratio	x 13 columns] Name F		 Sp. Atk	Sp. Def	Speed
550 5	536	Palpitoad	Water	 65	55	69
	537	Seismitoad	Water	 85	75	74
552 5	538	Throh	Fighting	 30	85	45
553 5	539	Sawk	Fighting	 30	75	85
5 554 5	540	Sewaddle	Bug	 40	60	42
555 5	541	Swadloon	Bug	 50	80	42
5 556 5	542	Leavanny	Bug	 70	80	92
5 557 5	543	Venipede	Bug	 30	39	57
558 5	544	Whirlipede	Bug	 40	79	47

559 5	545	Scolipede	Bu	g	55	j (69 13	12
560 561 562 563 564 565 566 567 568 569	rows 1D 546 547 548 549 550 551 552 553 554	x 13 columns Name Cottonee Whimsicott Petilil Lilligant Basculin Basculin Sandile Krokorok Krookodile Darumaka	Red-Striped Blue-Striped		Sp.	Def Spe 50 75 50 75 55 55 45 70 45	eed Ger 66 116 30 90 98 98 65 74 92 50	neration 5 5 5 5 5 5 5 5
570 571 572 573 574 575 576 577 578 579	rows ID 555 555 556 557 558 559 560 561 562 563	x 13 columns Name Darmanitan Darmanitan Maractus Dwebble Crustle Scraggy Scrafty Sigilyph Yamask Cofagrigus	5] For Standard Mod Zen Mod	e	Sp. Def 55 105 67 35 75 76 115 86 65	95 55 60 55 645 448 65 65 73 97 63	Generat	5 5 5 5 5 5 5 5 5 5
[10	rows ID	x 13 columns Name	_	S	p. Atk	Sp. Def	Speed	
Gene	eratio	on				-	•	
580 5	564	Tirtouga	Water		53	45	22	
581 5	565	Carracosta	Water		83	65	32	
582 5	566	Archen	Rock		74	45	70	
583	567	Archeops	Rock		112	65	110	
5 584 5	568	Trubbish	Poison		40	62	65	
585 5	569	Garbodor	Poison		60	82	75	
586 5	570	Zorua	Dark		80	40	65	
587 5	571	Zoroark	Dark		120	60	105	
588 5	572	Minccino	Normal		40	40	75	

589 5	573	Cinccino	Normal .		65	60	115
_	ID	x 13 columns] Name Form	Type1	Sp.	Atk	Sp. Def	Speed
590	eration 574	on Gothita	Psychic		55	65	45
5 591 5	575	Gothorita	Psychic		75	85	55
592 5	576	Gothitelle	Psychic		95	110	65
593 5	577	Solosis	Psychic		105	50	20
594 5	578	Duosion	Psychic		125	60	30
595 5	579	Reuniclus	Psychic		125	85	30
596 5	580	Ducklett	Water		44	50	55
597 5	581	Swanna	Water		87	63	98
598 5	582	Vanillite	Ice		65	60	44
599 5	583	Vanillish	Ice		80	75	59
	ID	x 13 columns] Name Form	Type1	Sp	o. Atk	Sp. Def	Speed
600	eration 584	on Vanilluxe	Ice		110	95	79
5 601	585	Deerling	Normal		40	50	75
5 602	586	Sawsbuck	Normal		60	70	95
5 603	587	Emolga	Electric		75	60	103
5 604 5	588	Karrablast	Bug		40	45	60
605 5	589	Escavalier	Bug		60	105	20
606 5	590	Foongus	Grass		55	55	15
607 5	591	Amoonguss	Grass		85	80	30
608 5	592	Frillish	Water		65	85	40
609 5	593	Jellicent	Water		85	105	60

[10	rows ID	x 13 columns] Name Form	Type1	 Sn. Atk	Sp. Def	Speed
	eratio	on		•	-	•
610 5	594	Alomomola	Water	 40	45	65
611 5	595	Joltik	Bug	 57	50	65
612 5	596	Galvantula	Bug	 97	60	108
613 5	597	Ferroseed	Grass	 24	86	10
614 5	598	Ferrothorn	Grass	 54	116	20
615 5	599	Klink	Steel	 45	60	30
616 5	600	Klang	Steel	 70	85	50
617 5	601	Klinklang	Steel	 70	85	90
618 5	602	Tynamo	Electric	 45	40	60
619	603	Eelektrik	Electric	 75	70	40
5						
	rows	x 13 columns]				
[10	ID	x 13 columns] Name Form	Type1	 Sp. Atk	Sp. Def	Speed
[10 Gene 620	ID eratio	Name Form	Typel Electric	 Sp. Atk 105	Sp. Def	Speed 50
[10 Gene 620 5 621	ID eratio 604	Name Form on		 •	-	-
[10 Gene 620 5 621 5 622	ID eratio 604	Name Form on Eelektross	Electric	 105	80	50
[10 Gene 620 5 621 5 622 5 623	ID eratio 604 605	Name Form on Eelektross Elgyem	Electric Psychic	 105 85	80 55	50 30
[10 Gene 620 5 621 5 622 5 623 5 624	ID eratio 604 605 606	Name Form on Eelektross Elgyem Beheeyem	Electric Psychic Psychic	 105 85 125	80 55 95	50 30 40
[10 Gene 620 5 621 5 622 5 623 5 624 5 625	ID eration 604 605 606 607	Name Form on Eelektross Elgyem Beheeyem Litwick	Electric Psychic Psychic Ghost	 105 85 125 65	80 55 95 55	50 30 40 20
[10 Gene 620 5 621 5 622 5 623 5 624 5 625 5	ID eratio 604 605 606 607 608	Name Form on Eelektross Elgyem Beheeyem Litwick Lampent	Electric Psychic Psychic Ghost Ghost	 105 85 125 65 95	80 55 95 55 60	50 30 40 20 55
[10 Gene 620 5 621 5 622 5 623 5 624 5 625 5 626 5	ID eration 604 605 606 607 608 609	Name Form on Eelektross Elgyem Beheeyem Litwick Lampent Chandelure	Electric Psychic Psychic Ghost Ghost Ghost	 105 85 125 65 95 145	80 55 95 55 60 90	50 30 40 20 55 80
[10 Gene 620 5 621 5 622 5 623 5 624 5 625 5 626 5	ID eration 604 605 606 607 608 609 610	Name Form Definition Eelektross Elgyem Beheeyem Litwick Lampent Chandelure Axew	Electric Psychic Psychic Ghost Ghost Ghost Dragon	 105 85 125 65 95 145 30	80 55 95 55 60 90 40	50 30 40 20 55 80 57

[10 rows x 13 columns]

Conc	ID eratio	Name I	Form	Type1		. Sp. /	Atk	Sp.	Def	Speed	
630 5	614	Beartic		Ice			70		80	50	
631 5	615	Cryogonal		Ice			95		135	105	
632 5	616	Shelmet		Bug			40		65	25	
633 5	617	Accelgor		Bug			100		60	145	
634	618	Stunfisk		Ground			81		99	32	
5 635 5	619	Mienfoo		Fighting			55		50	65	
636 5	620	Mienshao		Fighting			95		60	105	
637 5	621	Druddigon		Dragon			60		90	48	
638 5	622	Golett		Ground			35		50	35	
639 5	623	Golurk		Ground			55		80	55	
_	ID		s] Form	Typel		Sp. A	tk	Sp.	Def	Speed	
640	eratio 624	n Pawniard		Dark		4	40		40	60	
5 641	625	Bisharp		Dark		(60		70	70	
5 642	626	Bouffalant		Normal		4	40		95	55	
5 643	627	Rufflet		Normal		3	37		50	60	
5 644 5	628	Braviary		Normal		į	57		75	80	
645 5	629	Vullaby		Dark		4	45		65	60	
646 5	630	Mandibuzz		Dark		į	55		95	80	
647 5	631	Heatmor		Fire		10	95		66	65	
648 5	632	Durant		Bug		4	48		48	109	
649 5	633	Deino		Dark		4	45		50	38	
[10 650	rows ID 634	x 13 columns Name Zweilous	5]	For	n n	Sp		ef Sp 70	eed 58	Generat	ion 5

651 652 653 654 655 656 657 658 659	635 636 637 638 639 640 641 641 642	Hydreigon Larvesta Volcarona Cobalion Terrakion Virizion Tornadus Tornadus Thundurus	Incarnate Forme Therian Forme Incarnate Forme			90 55 105 72 90 129 80 90 80	98 60 100 108 108 108 111 121	5 5 5 5 5 5 5 5
[10		x 13 columns	_		C	D - 4	C	Cananation
660	ID	Name	Form	• • •	Sp.		Speed	Generation
660 661	642 643	Thundurus Reshiram	Therian Forme	• • •		80 120	101 90	5
662	644	Zekrom				100	90	5
663	645	Landorus	Incarnate Forme			80	101	5 5 5 5 5 5 5
664	645	Landorus	Therian Forme	• • •		80	91	5
665	646	Kyurem	THEFTAIL FORME	• • •		90	95	5
666	646	Kyurem	Black Kyurem			90	95	5
667	646	Kyurem	White Kyurem			100	95	5
668	647	Keldeo	Ordinary Form			90	108	5
669	647	Keldeo	Resolute Form			90	108	5
[10	rows ID	x 13 columns	s] Form		. Sp	. Def	Speed	Generation
670	648	Meloetta	Aria Forme			128	90	5
671	648	Meloetta	Pirouette Forme			77		5
672	649	Genesect				95	99	5
673	3	Venusaur	Mega Venusaur			120	80	6
674	6	Charizard	Mega Charizard X			85	100	6
675	6	Charizard	Mega Charizard Y			115	100	6
676	9	Blastoise	Mega Blastoise			115	78	6
677	15	Beedrill	Mega Beedrill			80	145	6
678	18	Pidgeot	Mega Pidgeot			80	121	6
679	65	Alakazam	Mega Alakazam		•	105	150	6
[10	rows	x 13 columns	s 1					
	ID	Name	Form		. Sp	. Def	Speed	Generation
680	80	Slowbro	Mega Slowbro		. '	80	['] 30	6
681	94	Gengar	Mega Gengar			95	130	6
682	115	Kangaskhan	Mega Kangaskhan			100	100	6
683	127	Pinsir	Mega Pinsir			90	105	6
684	130	Gyarados	Mega Gyarados			130	81	6
685	142	Aerodactyl	Mega Aerodactyl			95	150	6
686	150	Mewtwo	Mega Mewtwo X			100	130	6
687	150	Mewtwo	Mega Mewtwo Y			120	140	6
688	181	Ampharos	Mega Ampharos			110	45	6
689	208	Steelix	Mega Steelix	• •	•	95	30	6

[10 rows x 13 columns]

690 691 692 693 694 695 696 697 698	ID 212 214 229 248 254 257 260 282 302 303	Name Scizor Heracross Houndoom Tyranitar Sceptile Blaziken Swampert Gardevoir Sableye Mawile	Form Mega Scizor Mega Heracross Mega Houndoom Mega Tyranitar Mega Sceptile Mega Blaziken Mega Swampert Mega Gardevoir Mega Sableye Mega Mawile	 Sp.	Def 100 105 90 120 85 80 110 135 115	Speed 75 75 115 71 145 100 70 100 20 50	Generation 6 6 6 6 6 6 6 6
[10 700 701 702 703 704 705 706 707 708 709	ID 306 308 319 323 334 354 359 362 373	x 13 column Name Aggron Medicham Manectric Sharpedo Camerupt Altaria Banette Absol Glalie Salamence	Form Mega Aggron Mega Medicham Mega Manectric Mega Sharpedo Mega Camerupt Mega Altaria Mega Banette Mega Absol Mega Glalie Mega Salamence	 Sp.	Def 80 85 80 65 105 105 83 60 80 90	Speed 50 100 135 105 20 80 75 115 100 120	Generation 6 6 6 6 6 6 6 6
710 711 712 713 714 715 716 717 718 719	ID 376 380 381 382 383 384 428 445 448	x 13 column Name Metagross Latias Latios Kyogre Groudon Rayquaza Lopunny Garchomp Lucario Abomasnow x 13 column	Form Mega Metagross Mega Latias Mega Latios Primal Kyogre Primal Groudon Mega Rayquaza Mega Lopunny Mega Garchomp Mega Lucario Mega Abomasnow	 Sp.	Def 110 150 120 160 90 100 95 70 105		Generation 6 6 6 6 6 6 6 6
720 721 722 723 724 725 726 727 728 729	1D 475 531 650 651 652 653 654 655 656	Name Gallade Audino Chespin Quilladin Chesnaught Fennekin Braixen Delphox Froakie Frogadier	Form Mega Gallade Mega Audino	 Sp.	Def 115 126 45 58 75 60 70 100 44 56	Speed 110 50 38 57 64 60 73 104 71	Generation 6 6 6 6 6 6 6 6 6 6 6

[10	rows ID	x 13 columns] Name Fo	orm Type1		Sp. Atl	< Sp. De	f Speed
Gene	eration 658		Water		103	•	1 122
6 731	659	Bunnelby	Normal		32		6 57
6 732	660	Diggersby	Normal		5(77 78
6 733	661	Fletchling	Normal		40		8 62
6 734	662	Fletchinder	Fire		56		2 84
6 735	663	Talonflame	Fire		74		9 126
6 736	664	Scatterbug	Bug		27		25 35
6 737	665	Spewpa	Bug		27		0 29
6 738	666	Vivillon	Bug		9(0 89
6 739	667	Litleo	Fire		73	3 5	4 72
6							
[10	rows ID	x 13 columns] Name Form	Type1	S _I	o. Atk	Sp. Def	Speed
		Name Form	Typel Fire	S _l	o. Atk 109	Sp. Def	Speed
Gene 740 6 741	ID eratio	Name Form on		·		•	·
Gene 740 6 741 6 742	ID eratio 668	Name Form on Pyroar	Fire		109	66	106
Gene 740 6 741 6 742 6 743	ID eratio 668 669	Name Form on Pyroar Flabébé	Fire Fairy		109 61	66	106 42
Gene 740 6 741 6 742 6 743 6 744	ID eration 668 669 670	Name Form on Pyroar Flabébé Floette	Fire Fairy Fairy		109 61 75	66 79 98	106 42 52
Gene 740 6 741 6 742 6 743 6 744 6 745	ID eratio 668 669 670 671	Name Form Pyroar Flabébé Floette Florges	Fire Fairy Fairy Fairy		109 61 75 112	66 79 98 154	106 42 52 75
Gene 740 6 741 6 742 6 743 6 744 6 745 6 746	ID eratio 668 669 670 671	Name Form Pyroar Flabébé Floette Florges Skiddo	Fire Fairy Fairy Fairy Grass		109 61 75 112 62	66 79 98 154 57	106 42 52 75 52
Gene 740 6 741 6 742 6 743 6 744 6 745 6 746 6 747	ID eratio 668 669 670 671 672	Name Form Pyroar Flabébé Floette Florges Skiddo Gogoat	Fire Fairy Fairy Grass Grass		109 61 75 112 62 97	66 79 98 154 57 81	106 42 52 75 52 68
Gene 740 6 741 6 742 6 743 6 744 6 745 6 746 6 747 6	ID eratio 668 669 670 671 672 673	Name Form Pyroar Flabébé Floette Florges Skiddo Gogoat Pancham	Fire Fairy Fairy Fairy Grass Grass Fighting		109 61 75 112 62 97 46	66 79 98 154 57 81 48	106425275526843
Gene 740 6 741 6 742 6 743 6 744 6 745 6 746 6 747	ID eratio 668 669 670 671 672 673 674	Name Form Pyroar Flabébé Floette Florges Skiddo Gogoat Pancham Pangoro	Fire Fairy Fairy Fairy Grass Grass Fighting Fighting		109 61 75 112 62 97 46 69	66 79 98 154 57 81 48 71	106 42 52 75 52 68 43 58

[10 rows x 13 columns]

750 751 752 753 754 755 756 757 758 759	ID 678 679 680 681 682 683 684 685	Name Meowstic Meowstic Honedge Doublade Aegislash Aegislash Spritzee Aromatisse Swirlix Slurpuff	Form Male Female Shield Forme Blade Forme	 Sp. Def 81 81 37 49 140 50 65 89 57	104 104 28 35 60 60 23 29 49	Ger	neration 6 6 6 6 6 6 6 6
[10	rows ID	x 13 columns Name	_	 Sp. At	k Sp.	Def	Speed
Gene	ratio 686	on Inkay	Dark	 3	7	46	45
6 761	687	Malamar	Dark	 6	8	75	73
6 762	688	Binacle	Rock	 3	9	56	50
6 763	689	Barbaracle	Rock	 5	4	86	68
6 764	690	Skrelp	Poison	 6	Θ	60	30
6 765	691	Dragalge	Poison	 9	7	123	44
6 766 6	692	Clauncher	Water	 5	8	63	44
767 6	693	Clawitzer	Water	 12	Θ	89	59
768 6	694	Helioptile	Electric	 6	1	43	70
769 6	695	Heliolisk	Electric	 10	9	94	109
_	ID	x 13 columns Name F	=	 Sp. Atk	Sp. [Def	Speed
770	eration 696	on Tyrunt	Rock	 45		45	48
6 771	697	Tyrantrum	Rock	 69		59	71
6 772	698	Amaura	Rock	 67		63	46
6 773 6	699	Aurorus	Rock	 99		92	58
774 6	700	Sylveon	Fairy	 110	:	130	60
775	701	Hawlucha	Fighting	 74		63	118

6 776	702	Dedenne	Electric		81	67	101	
6 777	703	Carbink	Rock		50	150	50	
6								
778 6	704	Goomy	Dragon	• • • •	55	75	40	
779	705	Sliggoo	Dragon		83	113	60	
6								
[10	rows	x 13 columns	5]					
Cana	ID	Name	Form	Type1	Sp.	Atk S	p. Def	Speed
780	eratio 706	Goodra		Dragon		110	150	80
6				_				
781 6	707	Klefki		Steel	• • •	80	87	75
782	708	Phantump		Ghost		50	60	38
6 783	709	Trovonant		Chast		65	ດລ	E.6
763 6	709	Trevenant		Ghost	• • •	65	82	56
784	710	Pumpkaboo	Average Size	Ghost		44	55	51
6 785	710	Pumpkaboo	Small Size	Ghost		44	55	56
6		·						
786 6	710	Pumpkaboo	Large Size	Ghost		44	55	46
787	710	Pumpkaboo	Super Size	Ghost		44	55	41
6	711	Couraciet	Avenage Cite	Chast		EO	75	0.4
788 6	711	Gourgeist	Average Size	Ghost		58	75	84
789	711	Gourgeist	Small Size	Ghost		58	75	99
6								
[10	rows	x 13 columns	5]					
700	ID	Cour		Form		f Speed	Gener	
790 791	711 711		rgeist Large S rgeist Super S		. 7! . 7!			6 6
792	712		rgmite		. 3!			6
793	713		/alugg		. 40			
794	714		Noibat		. 40			6 6 6
795	715		pivern		. 80			6
796	716		erneas		. 98			6
797	717		/eltal		. 98			6
798	718	Zygarde50%			. 9!			6
799	719		iancie	• •	. 150			6
[10	rows	x 13 columns	s 1					
	ID	Name	=	orm	Sp. Def	Speed	Genera	tion
800	719	Diancie	Mega Diano		110	110		6

801 802 803 804 805 806 807 808 809	720 720 721 19 20 25 26 27 28		Hoo Alola Alola Partn Alo Alolan	a Confined pa Unbound an Rattata n Raticate er Pikachu lan Raichu Sandshrew Sandslash		130 130 90 35 80 60 85 35 65	70 80 70 72 77 120 110 40 65		6 6 7 7 7 7 7
[10		x 13 columns]	Form	Cn	Dof (`nood	Cononati	ian
810 811 812 813 814 815 816 817 818	ID 37 38 50 51 52 53 74 75 76 88	Diglett Dugtrio Meowth Persian Geodude	lolan Alola Alola Alola Alola Alolan	Form an Vulpix Ninetales n Diglett n Dugtrio an Meowth n Persian n Geodude Graveler lan Golem an Grimer	Sp.	Def 5 65 100 45 70 40 65 30 45 65 50	90 110 90 115 20 35 45 25	Generat	7 7 7 7 7 7 7 7
[10	rows ID	x 13 columns] Name		Form	9	Sp. Def	Speed	
Gene 820 7	eratio 89	on	Muk	Alo	lan Muk		100	50	
821 7	103	Exeg	gutor	Alolan Exe	eggutor		75	45	
822 7	105	Ма	rowak	Alolan M	Marowak		80	45	
823 7	133		Eevee	Partne	r Eevee		85	75	
7 824 7	658	Gre	ninja	Ash-G	reninja		71	132	
825	718	Zygarde10%	Forme				85	115	
7 826 7	718	Zy	garde	Complete	e Forme		95	85	
7 827 7	722	R	owlet				50	42	
828	723	Da	rtrix				70	52	
7 829 7	724	Deci	dueye				100	70	
	rows ID eratio] Form	Typel	. Sp. At	k Sp.	Def	Speed	

830	725	Litten	Fire		60	4	0	70
7 831	726	Torracat	Fire		80	5	0	90
7 832	727	Incineroar	Fire		80	9	0	60
7 833	728	Popplio	Water		66	5	6	40
7 834 7	729	Brionne	Water		91	8	1	50
835 7	730	Primarina	Water		126	11	6	60
836 7	731	Pikipek	Normal		30	3	0	65
837 7	732	Trumbeak	Normal		40	5	0	75
838 7	733	Toucannon	Normal		75	7	5	60
839 7	734	Yungoos	Normal		30	3	0	45
[10 840 841 842 843 844 845 846 847 848 849	rows ID 735 736 737 738 739 740 741 741 741	x 13 columns] Name Gumshoos Grubbin Charjabug Vikavolt Crabrawler Crabominable Oricorio Oricorio Oricorio	Pom-Pom Sty Pa'u Sty	 le le	· · · · · · · · · · · · · · · · · · ·	45 75 75 47 67 70 70	ed Ge 45 46 36 43 63 43 93 93	eneration 7 7 7 7 7 7 7 7
_	rows ID eratio	x 13 columns] Name		Form	Sp	. Def	Speed	
850 7	742	Cutiefly				40	84	
851 7	743	Ribombee				70	124	
852 7	744	Rockruff				40	60	
853 7	744	Rockruff	Own Tempo Roc	kruff		40	60	
854 7	745	Lycanroc	Midday	Form		65	112	
855 7	745	Lycanroc	Midnight	Form		75	82	
, 856	745	Lycanroc	Dusk	Form		65	110	

7	7.46		6.1.5		0.5	40
857 7	/46	Wishiwashi	Solo Form		25	40
858 7	746	Wishiwashi	School Form		135	30
859 7	747	Mareanie			52	45
_	rows ID eratio		Typel Sp	. Atk S	Sp. Def S	Speed
860 7	748	Toxapex	Poison	53	142	35
861 7	749	Mudbray	Ground	45	55	45
862 7	750	Mudsdale	Ground	55	85	35
863 7	751	Dewpider	Water	40	72	27
864 7	752	Araquanid	Water	50	132	42
865 7	753	Fomantis	Grass	50	35	35
866 7	754	Lurantis	Grass	80	90	45
867 7	755	Morelull	Grass	65	75	15
868 7	756	Shiinotic	Grass	90	100	30
869 7	757	Salandit	Poison	71	40	77
_	rows ID eratio	x 13 columns] Name Form	Typel	Sp. Atk	Sp. Def	Speed
870	758	Salazzle	Poison	111	60	117
7 871	759	Stufful	Normal	45	50	50
7 872 7	760	Bewear	Normal	55	60	60
873 7	761	Bounsweet	Grass	30	38	32
874 7	762	Steenee	Grass	40	48	62
875 7	763	Tsareena	Grass	50	98	72
876 7	764	Comfey	Fairy	82	110	100
877	765	0ranguru	Normal	90	110	60

7 878	766	Passimian	Fighting		40	60	80	
7 879 7	767	Wimpod	Bug		20	30	80	
	ID	x 13 columns Name] Form	Type1	Sp	. Atk Sp	o. Def	Speed
Gene 880 7	ratio 768	on Golisopod		Bug		60	90	40
881 7	769	Sandygast		Ghost		70	45	15
882 7	770	Palossand		Ghost		100	75	35
883 7	771	Pyukumuku		Water		30	130	5
884 7	772	Type: Null		Normal		95	95	59
885 7	773	Silvally		Normal		95	95	95
886 7	774	Minior	Meteor Form	Rock		60	100	60
887 7	774	Minior	Core Form	Rock		100	60	120
888 7	775	Komala		Normal		75	95	65
889 7	776	Turtonator		Fire		91	85	36
_	ID	x 13 columns Name	=	1 9	Sp. Atk	Sp. Def	Speed	
890 7	ratio 777	Togedemaru	Electri	c	40	73	96	
891 7	778	Mimikyu	Ghos-	t	50	105	96	
892 7	779	Bruxish	Wate	r	70	70	92	
893 7	780	Drampa	Norma	ι	135	91	36	
894 7	781	Dhelmise	Ghos-	t	86	90	40	
895 7	782	Jangmo-o	Drago	n	45	45	45	
896 7	783	Hakamo-o	Dragoi	n	65	70	65	
7 897 7	784	Kommo-o	Dragoi	n	100	105	85	
898	785	Tapu Koko	Electri	c	95	75	130	

7 899 7	786	Tapu Lele	Psychic		130	115	95
	ID	x 13 columns Name Fo		Sp.	Atk	Sp. Def	Speed
900 7	ratio 787	n Tapu Bulu	Grass		85	95	75
901 7	788	Tapu Fini	Water		95	130	85
902 7	789	Cosmog	Psychic		29	31	37
903 7	790	Cosmoem	Psychic		29	131	37
904 7	791	Solgaleo	Psychic		113	89	97
905	792	Lunala	Psychic		137	107	97
7 906	793	Nihilego	Rock		127	131	103
7 907	794	Buzzwole	Bug		53	53	79
7 908	795	Pheromosa	Bug		137	37	151
7 909 7	796	Xurkitree	Electric	• • •	173	71	83
[10	rows ID	x 13 columns; Name	1	Form	Sp	. Def Sp	eed
Gene 910	ratio 797	n Celesteela				101	61
7 911	798	Kartana				31	109
7 912	799	Guzzlord				53	43
7 913	800	Necrozma				89	79
7 914	800	Necrozma	Dusk Mane Ned	crozma		109	77
7 915	800	Necrozma	Dawn Wings Nec	crozma		127	77
7 916	800	Necrozma	Ultra Ned	crozma		97	129
7 917	801	Magearna				115	65
7 918	802	Marshadow				90	125
7 919	803	Poipole				67	73

[10	rows ID	x 13 columns] Name	Form	Sp. Def	Sneed
920	ratio 804			 73	121
7 921 7	805	Stakataka		 101	13
922 7	806	Blacephalon		 79	107
923 7	807	Zeraora		 80	143
7 924 7	808	Meltan		 35	34
925 7	809	Melmetal		 65	34
926 8	52	Meowth	Galarian Meowth	 40	40
927 8	77	Ponyta	Galarian Ponyta	 65	90
928 8	78	Rapidash	Galarian Rapidash	 80	105
929 8	79	Slowpoke	Galarian Slowpoke	 40	15
[10	rows ID	x 13 columns]	Form	 Sp. Def	Speed
Gene	ID ratio	Name on	Form	Sp. Def	•
	ID	Name		 Sp. Def	
Gene 930 8 931	ID ratio	Name on	Form		30
Gene 930 8 931 8 932	ID ratio 80	Name on Slowbro	Form Galarian Slowbro	 70	30 55
Gene 930 8 931 8 932 8 933	ID eratio 80 83	Name on Slowbro Farfetch'd	Form Galarian Slowbro Galarian Farfetch'd	 70 62	30 55 60
Gene 930 8 931 8 932 8 933 8	ID eratio 80 83 110	Name on Slowbro Farfetch'd Weezing	Form Galarian Slowbro Galarian Farfetch'd Galarian Weezing	 70 62 70	30 55 60 100
Gene 930 8 931 8 932 8 933 8 934 8	ID 80 83 110 122	Name on Slowbro Farfetch'd Weezing Mr. Mime	Form Galarian Slowbro Galarian Farfetch'd Galarian Weezing Galarian Mr. Mime	 70 62 70 90	30 55 60 100 95
Gene 930 8 931 8 932 8 933 8 934 8 935 8	ID 80 83 110 122 144	Name Slowbro Farfetch'd Weezing Mr. Mime Articuno	Form Galarian Slowbro Galarian Farfetch'd Galarian Weezing Galarian Mr. Mime Galarian Articuno	 70 62 70 90 100	30 55 60 100 95 100
Gene 930 8 931 8 932 8 933 8 934 8 935 8 936 8	ID 80 83 110 122 144 145	Name Slowbro Farfetch'd Weezing Mr. Mime Articuno Zapdos	Form Galarian Slowbro Galarian Farfetch'd Galarian Weezing Galarian Mr. Mime Galarian Articuno Galarian Zapdos	 70 62 70 90 100	30 55 60 100 95 100 90
Gene 930 8 931 8 932 8 933 8 934 8 935 8	1D 80 83 110 122 144 145 146	Name Slowbro Farfetch'd Weezing Mr. Mime Articuno Zapdos Moltres	Form Galarian Slowbro Galarian Farfetch'd Galarian Weezing Galarian Mr. Mime Galarian Articuno Galarian Zapdos Galarian Moltres	 70 62 70 90 100 90 125	30 55 60 100 95 100 90 30

[10	rows ID	x 13 columns]			F	- orm		. Sn	. Def	Speed
Gene	eratio				•	0	• •	. 56		opeca
940 8	264	Linoone	Gal	arian	Lind	one			61	100
941 8	554	Darumaka	Gala	rian D	arun	naka			45	50
942 8	555	Darmanitan	Galarian	Standa	rd N	1ode		•	55	95
943 8	555	Darmanitan	Gala	rian Z	en N	1ode			55	135
944 8	562	Yamask	Ga	larian	Yan	nask			65	30
945 8	618	Stunfisk	Gala	rian S	tuni	fisk			84	32
946 8	810	Grookey							40	65
947 8	811	Thwackey							60	80
948 8	812	Rillaboom							70	85
949 8	813	Scorbunny						•	40	69
_	ID			pel .	9	Sp. /	Atk	Sp.	Def	Speed
	eratio									
950 8	814	Raboot	F	ire .	• •		55		60	94
951 8	815	Cinderace	F	ire .	• •		65		75	119
952 8	816	Sobble	Wa	ter .	• •		70		40	70
953 8	817	Drizzile	Wa	ter .	• •		95		55	90
954 8	818	Inteleon	Wa	ter .	• •		125		65	120
955 8	819	Skwovet	Nor	mal .			35		35	25
956 8	820	Greedent	Nor	mal .			55		75	20
957 8	821	Rookidee	Fly	ing .			33		35	57
958	822	Corvisquire	Fly	ing .			43		55	77
8 959 8	823	Corviknight	Fly	ing .			53		85	67
[10										

960	eratio 824	on Blipbug	Bug		25	45	45
8 961 8	825	Dottler	Bug		50	90	30
962 8	826	Orbeetle	Bug		80	120	90
963 8	827	Nickit	Dark		47	52	50
964 8	828	Thievul	Dark		87	92	90
965 8	829	Gossifleur	Grass		40	60	10
966 8	830	Eldegoss	Grass		80	120	60
967 8	831	Wooloo	Normal		40	45	48
968 8	832	Dubwool	Normal		60	90	88
969 8	833	Chewtle	Water		38	38	44
	rows ID	x 13 columns] Name Form	Tvpe1	9	Sp. Atk	Sp. Def	Speed
	eratio	on				•	-
970 8	834	Drednaw	Water	• • •	48	68	74
971 8	835	Yamper	Electric		40	50	26
972 8	836	Boltund	Electric		90	60	121
973 8	837	Rolycoly	Rock		40	50	30
974 8	838	Carkol	Rock		60	70	50
975 8	839	Coalossal	Rock		80	90	30
976 8	840	Applin	Grass		40	40	20
977 8	841	Flapple	Grass		95	60	70
978	842	Appletun	Grass		100	80	30
8 979 8	843	Silicobra	Ground		35	50	46
_	ID	x 13 columns] Name	Form		Sp. Def		Generation
980 981	844 845	Sandaconda Cramorant			70 95	71 85	8

982 983 984 985 986 987 988 989	846 847 848 849 849 850 851 852	Arrokuda Barraskewda Toxel Toxtricity Toxtricity Sizzlipede Centiskorch Clobbopus		Key Form Ded Form		30 50 35 70 70 50 90	66 136 40 75 75 45 65 32	8 8 8 8 8 8
[10	rows >	x 13 columns] Name		Type1		Sp. Atk	Sp. Def	Speed
Gene	ration 853			Fighting		70	80	42
8					• • • •			
991 8	854	Sinistea		Ghost	• • • •	74	54	50
992 8	855	Polteageist		Ghost	• • •	134	114	70
993 8	856	Hatenna		Psychic	• • •	56	53	39
994 8	857	Hattrem		Psychic		86	73	49
995	858	Hatterene		Psychic		136	103	29
8 996	859	Impidimp		Dark		55	40	50
8 997	860	Morgrem		Dark		75	55	70
8 998	861	Grimmsnarl		Dark		95	75	60
8 999 8	862	0bstagoon		Dark		60	81	95
-	ID	x 13 columns] Name	•	Type1		Sp. Atk	Sp. Def	Speed
Gene	ration 863	n Perrserker		Steel		50	60	50
8 1001	. 864	Cursola		Ghost		145	130	30
8 1002	865	Sirfetch'd		Fighting		68	82	65
8 1003	866	Mr. Rime		Psychic		110	100	70
8 1004	867	Runerigus		Ground		50	105	30
8 1005		_				50	61	34
8		Milcery		Fairy				
1006 8	869	Alcremie		Fairy	• • • •	110	121	64

1007	870	Falinks		Fightin	g		7	0	60	75	5
8 1008	871	Pincurchin		Electri	.c		9	1	85	15	,
8 1009 8	872	Snom		Ic	e		4	5	30	20)
[10 r 1010 1011 1012 1013 1014 1015 1016 1017 1018 1019	ows x ID 873 874 875 875 876 876 877 877 878	13 columns Nam Frosmot Stonjourne Eiscue Eiscue Indeede Indeede Morpek Morpek Cufan Copperaja	e r e e e o Full o	Ice Noice	Male male Mode		Sp.	Def 90 20 90 50 95 105 58 49 69	Speed 65 70 50 130 95 85 97 97 40 30	Gener	Ration 8 8 8 8 8 8 8 8 8 8 8 8 8
	ows x ID ation	13 columns Name]		Foi	∩m		Sp. D	ef Sp	eed	
1020	880	Dracozolt							70	75	
1021 8	881	Arctozolt							80	55	
1022 8	882	Dracovish							80	75	
1023 8	883	Arctovish							90	55	
1024 8	884	Duraludon							50	85	
1025 8	885	Dreepy							30	82	
1026 8	886	Drakloak							50	102	
1027 8	887	Dragapult							75	142	
1028 8	888	Zacian		Crowne	d Swo	∩d		1	.15	148	
1029 8	888	Zacian	Hero o	f Many	Battle	es		1	.15	138	
[10 r		13 columns]		Га			C D			
	ID ation	Name		Che	Foi			·	ef Sp		
1030 8	889	Zamazenta		Crowned						128	
1031	889	Zamazenta	него о	т Many	Battle	es		1	.15	138	

0										
8 1032 8	890	Eternatus						Ć	95 1	30
0 1033 8	890	Eternatus		Ete	ernan	nax		25	50 1	30
1034 8	891	Kubfu						ŗ	50	72
1035 8	892	Urshifu	Single	Strike	e Sty	/le		6	50	97
1036 8	892	Urshifu	Rapid	Strike	e Sty	/le		6	50	97
1037 8	893	Zarude						Ć	95 1	95
1038 8	894	Regieleki						Į.	50 2	90
1039 8	895	Regidrago							50	80
[10 r	ows x	13 columns]								
1040 1041 1042 1043 1044	ID 896	Name Glastrier Spectrier Calyrex Calyrex	Ice F Shadow F			Sp.	Def 110 80 80 130 100	Speed 30 130 80 50 150	Gene	ration 8 8 8 8 8
[5 ro	WS X	13 columns]								
df										
1040 1041 1042 1043 1044	ID 896 897 898 898	-		Form Rider Rider			Def 110 80 80 130 100	Speed 30 130 80 50 150	Gene	ration 8 8 8 8 8
[5 ro	WS X	13 columns]								
<pre>df1 = pd.DataFrame() for df in pd.read_csv('poke_updated1.csv', chunksize=10): df1 = pd.concat([df1 ,df]) df1.head(15)</pre>										
	D	Name For	m Type1	L	Sp.	Atk	Sp	. Def	Speed	
		ulbasaur	Grass	S		65		65	45	
	2	Ivysaur	Grass	S		80		80	60	
1 2	3	Venusaur	Grass	S		100		100	80	

1 3	4	Char	mar	nder		Fire		66	50	65
1 4	5	Char				Fire		86		80
1 5	6			zard		Fire		109		100
1 6	7			rtle		Water		50		43
1		•								
7	8			rtle		Water		65		58
8 1	9	Bla	sto	oise		Water		85	105	78
9 1	10	Ca	tei	rpie		Bug		20	20	45
10 1	11	M	leta	apod		Bug		25	5 25	30
11 1	12	Butt	ert	free		Bug		96	80	70
12	13		Wee	edle		Bug		26	20	50
1 13	14		Kal	kuna		Bug		25	5 25	35
1 14 1	15	Ве	edı	rill		Bug		45	80	75
[15	row	s x 1	.3 (colu	mns]					
#Sta	ack &	Unst	ack	in P	andas					
	=pd . P']]		Inc	dex.	from_pr	oduct([['20	910', '2	2015'], ['	Literacy'
dat	a=([[80,7	,88	3,6]	, [90,8	3,92,7]	, [89	9,7,91,8	3], [87,6,	93,8]])
df6	=pd.	DataF	ran	ne(d	ata,ind	lex=['I	ndia	','USA',	'Russia',	'China'])
df6										
0 1 2 3 India 80 7 88 6 USA 90 8 92 7 Russia 89 7 91 8 China 87 6 93 8										
st_ st_		df6.	sta	ack()					
Ind	ia	0 1 2	88	7						

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Russia
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dtype: int64
unst_df = st_df.unstack()
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unst_df = unst_df.unstack()
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dtype: int64
unst_df = unst_df.unstack()
unst_df
   India USA
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2	88	92	91	93
3	6	7	8	8