Chapter1_async

March 14, 2022

1 Chapter1

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

[2]:
```
scores = {"name":['Ray','Japhy','Zosa'],
          "city":['San Francisco','San Francisco','Denver'],
          "score":[75,92,94]
         }
```

[3]:
```
df = pd.DataFrame(scores)
```

[4]:
```
df
```

[4]:

	name	city	score
0	Ray	San Francisco	75
1	Japhy	San Francisco	92
2	Zosa	Denver	94

[5]:
```
df['score']
```

[5]:
```
0    75
1    92
2    94
Name: score, dtype: int64
```

[6]:
```
df['name_city'] = df['name'] + '_' + df['city']
```

[7]:
```
df[df['score']>90]
```

[7]:

	name	city	score	name_city
1	Japhy	San Francisco	92	Japhy_San Francisco
2	Zosa	Denver	94	Zosa_Denver

[8]:
```
iris = pd.read_csv('./iris.csv')
```

[9]:
```
iris.shape
```

[9]:
```
(150, 5)
```

[10]:
```
iris.head(3)
```

[10]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa

[11]:
```
iris.tail(3)
```

[11]:

	sepal_length	sepal_width	petal_length	petal_width	species
147	6.5	3.0	5.2	2.0	virginica
148	6.2	3.4	5.4	2.3	virginica
149	5.9	3.0	5.1	1.8	virginica

[12]:
```
iris.dtypes
```

[12]:
```
sepal_length    float64
sepal_width     float64
petal_length    float64
petal_width     float64
species          object
dtype: object
```

[13]:
```
iris.loc[3:5]
```

[13]:

	sepal_length	sepal_width	petal_length	petal_width	species
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
5	5.4	3.9	1.7	0.4	setosa

[14]:
```
iris.loc[3,'sepal_length']
```

[14]:
```
4.6
```

[15]:
```
iris.iloc[3,0]
```

[15]:
```
4.6
```

[16]:
```
# iris.to_csv('iris-output.csv',index=False)
```

[17]:
```
emissions = pd.DataFrame({"country":['China','United States','India'],
          "year":['2018','2018','2018'],
          "co2_emissions":[10060000000.0,5410000000.0,2650000000.0]})
```

[18]:
```
emissions
```

[18]:

	country	year	co2_emissions
0	China	2018	1.006000e+10

```
1  United States  2018   5.410000e+09
2          India  2018   2.650000e+09
```

[19]:
```
# pd.set_option('display.max_rows', 2)
pd.reset_option('^display.', silent=True)

emissions
```

[19]:
```
         country  year  co2_emissions
0          China  2018   1.006000e+10
1  United States  2018   5.410000e+09
2          India  2018   2.650000e+09
```

[20]:
```
# pd.set_option('display.max_columns', 2)
pd.reset_option('^display.', silent=True)
emissions
```

[20]:
```
         country  year  co2_emissions
0          China  2018   1.006000e+10
1  United States  2018   5.410000e+09
2          India  2018   2.650000e+09
```

[21]:
```
pd.options.display.float_format = '{:,.2f}'.format
```

[22]:
```
emissions
```

[22]:
```
         country  year     co2_emissions
0          China  2018 10,060,000,000.00
1  United States  2018  5,410,000,000.00
2          India  2018  2,650,000,000.00
```

2 Chapter2

[23]:
```
planets = pd.read_csv('planets.csv')
```

[24]:
```
planets.head(3)
```

[24]:
```
            method  number  orbital_period  mass  distance  year
0  Radial Velocity       1          269.30  7.10     77.40  2006
1  Radial Velocity       1          874.77  2.21     56.95  2008
2  Radial Velocity       1          763.00  2.60     19.84  2011
```

[25]:
```
planets.dtypes
```

[25]:
```
method             object
number              int64
orbital_period    float64
```

```
mass              float64
distance          float64
year                int64
dtype: object
```

[26]:
```
planets.mean()
```

```
C:\Users\aadar\AppData\Local\Temp\ipykernel_11968\656747818.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.
  planets.mean()
```

[26]:
```
number               1.79
orbital_period   2,002.92
mass                 2.64
distance           264.07
year             2,009.07
dtype: float64
```

[27]:
```
planets['number'][0]/planets['mass'][0]
```

[27]:
```
0.14084507042253522
```

[28]:
```
planets['number'][0].astype(float)
```

[28]:
```
1.0
```

[29]:
```
planets['mass'][0].astype(int)
# note not a rounded value
```

[29]:
```
7
```

[30]:
```
planets['year'][0].astype(str)
```

[30]:
```
'2006'
```

[31]:
```
planets['year_dt'] = pd.to_datetime(planets['year'], format='%Y')
planets['year_dt']
```

[31]:
```
0      2006-01-01
1      2008-01-01
2      2011-01-01
3      2007-01-01
4      2009-01-01
          …
1030   2006-01-01
1031   2007-01-01
```

```
1032   2007-01-01
1033   2008-01-01
1034   2008-01-01
Name: year_dt, Length: 1035, dtype: datetime64[ns]
```

[32]:
```
names = pd.Series(['Pomeray, CODY ',' Wagner; Jarry','smith, Ray'])
```

[33]:
```
names = names.str.replace(';',',')
names
```

[33]:
```
0    Pomeray, CODY
1     Wagner, Jarry
2        smith, Ray
dtype: object
```

[34]:
```
names.str.len()
```

[34]:
```
0    14
1    14
2    10
dtype: int64
```

[35]:
```
names = names.str.strip()
names.str.len()
```

[35]:
```
0    13
1    13
2    10
dtype: int64
```

[36]:
```
names = names.str.lower()
names
```

[36]:
```
0    pomeray, cody
1    wagner, jarry
2       smith, ray
dtype: object
```

[37]:
```
names = names.str.split(', ')
names
```

[37]:
```
0    [pomeray, cody]
1    [wagner, jarry]
2       [smith, ray]
dtype: object
```

[38]:
```
names = pd.Series([i[::-1] for i in names])
names
```

[38]:
```
0    [cody, pomeray]
1    [jarry, wagner]
2       [ray, smith]
dtype: object
```

[39]:
```
names = [' '.join(i) for i in names]
names
```

[39]:
```
['cody pomeray', 'jarry wagner', 'ray smith']
```

3 Chapter3

[40]:
```
iris = pd.read_csv('./iris.csv')
iris.head(5)
```

[40]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.10	3.50	1.40	0.20	setosa
1	4.90	3.00	1.40	0.20	setosa
2	4.70	3.20	1.30	0.20	setosa
3	4.60	3.10	1.50	0.20	setosa
4	5.00	3.60	1.40	0.20	setosa

[41]:
```
# can flatten hierarchical index with reset_index()
iris.groupby(['species']).max()
```

[41]:

species	sepal_length	sepal_width	petal_length	petal_width
setosa	5.80	4.40	1.90	0.60
versicolor	7.00	3.40	5.10	1.80
virginica	7.90	3.80	6.90	2.50

[42]:
```
df = iris.groupby(['species']).agg({'sepal_length':
 ↪['mean','min','max'],'sepal_width':'count'})
df
```

[42]:

	sepal_length			sepal_width
	mean	min	max	count
species				
setosa	5.01	4.30	5.80	50
versicolor	5.94	4.90	7.00	50
virginica	6.59	4.90	7.90	50

[43]:
```
df['sepal_length']
```

[43]:

species	mean	min	max
setosa	5.01	4.30	5.80

```
versicolor  5.94 4.90 7.00
virginica   6.59 4.90 7.90
```

[44]:
```
df.columns = ['_'.join(col).strip() for col in df.columns.values]
df.reset_index()
df
```

[44]:
```
            sepal_length_mean  sepal_length_min  sepal_length_max  \
species
setosa                   5.01              4.30              5.80
versicolor               5.94              4.90              7.00
virginica                6.59              4.90              7.90

            sepal_width_count
species
setosa                     50
versicolor                 50
virginica                  50
```

[45]:
```
groupings = iris.groupby(['species'])
```

[46]:
```
groupings.get_group('setosa').head()
```

[46]:
```
   sepal_length  sepal_width  petal_length  petal_width species
0          5.10         3.50          1.40         0.20  setosa
1          4.90         3.00          1.40         0.20  setosa
2          4.70         3.20          1.30         0.20  setosa
3          4.60         3.10          1.50         0.20  setosa
4          5.00         3.60          1.40         0.20  setosa
```

[47]:
```
groupings.max()
```

[47]:
```
            sepal_length  sepal_width  petal_length  petal_width
species
setosa              5.80         4.40          1.90         0.60
versicolor          7.00         3.40          5.10         1.80
virginica           7.90         3.80          6.90         2.50
```

[48]:
```
groupings.apply(lambda x: x.max())
```

[48]:
```
            sepal_length  sepal_width  petal_length  petal_width     species
species
setosa              5.80         4.40          1.90         0.60      setosa
versicolor          7.00         3.40          5.10         1.80  versicolor
virginica           7.90         3.80          6.90         2.50   virginica
```

[49]:
```
groupings.filter(lambda x: x['petal_length'].max() <5)
```

[49]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.10	3.50	1.40	0.20	setosa
1	4.90	3.00	1.40	0.20	setosa
2	4.70	3.20	1.30	0.20	setosa
3	4.60	3.10	1.50	0.20	setosa
4	5.00	3.60	1.40	0.20	setosa
5	5.40	3.90	1.70	0.40	setosa
6	4.60	3.40	1.40	0.30	setosa
7	5.00	3.40	1.50	0.20	setosa
8	4.40	2.90	1.40	0.20	setosa
9	4.90	3.10	1.50	0.10	setosa
10	5.40	3.70	1.50	0.20	setosa
11	4.80	3.40	1.60	0.20	setosa
12	4.80	3.00	1.40	0.10	setosa
13	4.30	3.00	1.10	0.10	setosa
14	5.80	4.00	1.20	0.20	setosa
15	5.70	4.40	1.50	0.40	setosa
16	5.40	3.90	1.30	0.40	setosa
17	5.10	3.50	1.40	0.30	setosa
18	5.70	3.80	1.70	0.30	setosa
19	5.10	3.80	1.50	0.30	setosa
20	5.40	3.40	1.70	0.20	setosa
21	5.10	3.70	1.50	0.40	setosa
22	4.60	3.60	1.00	0.20	setosa
23	5.10	3.30	1.70	0.50	setosa
24	4.80	3.40	1.90	0.20	setosa
25	5.00	3.00	1.60	0.20	setosa
26	5.00	3.40	1.60	0.40	setosa
27	5.20	3.50	1.50	0.20	setosa
28	5.20	3.40	1.40	0.20	setosa
29	4.70	3.20	1.60	0.20	setosa
30	4.80	3.10	1.60	0.20	setosa
31	5.40	3.40	1.50	0.40	setosa
32	5.20	4.10	1.50	0.10	setosa
33	5.50	4.20	1.40	0.20	setosa
34	4.90	3.10	1.50	0.20	setosa
35	5.00	3.20	1.20	0.20	setosa
36	5.50	3.50	1.30	0.20	setosa
37	4.90	3.60	1.40	0.10	setosa
38	4.40	3.00	1.30	0.20	setosa
39	5.10	3.40	1.50	0.20	setosa
40	5.00	3.50	1.30	0.30	setosa
41	4.50	2.30	1.30	0.30	setosa
42	4.40	3.20	1.30	0.20	setosa
43	5.00	3.50	1.60	0.60	setosa
44	5.10	3.80	1.90	0.40	setosa
45	4.80	3.00	1.40	0.30	setosa

```
46          5.10         3.80          1.60         0.20  setosa
47          4.60         3.20          1.40         0.20  setosa
48          5.30         3.70          1.50         0.20  setosa
49          5.00         3.30          1.40         0.20  setosa
```

[50]:
```
df = pd.DataFrame({"Region":
 ↪['North','West','East','South','North','West','East','South'],
          "Team":['One','One','One','One','Two','Two','Two','Two'],
          "Revenue":[7500,5500,2750,6400,2300,3750,1900,575],
            "Cost":[5200,5100,4400,5300,1250,1300,2100,50]})
df
```

[50]:
```
  Region Team  Revenue  Cost
0  North  One     7500  5200
1   West  One     5500  5100
2   East  One     2750  4400
3  South  One     6400  5300
4  North  Two     2300  1250
5   West  Two     3750  1300
6   East  Two     1900  2100
7  South  Two      575    50
```

[51]:
```
df.pivot(index='Region',columns='Team',values='Revenue')
```

[51]:
```
Team     One   Two
Region
East    2750  1900
North   7500  2300
South   6400   575
West    5500  3750
```

[52]:
```
df2 = df.set_index(['Region','Team'])
```

[53]:
```
stacked = pd.DataFrame(df2.stack())
stacked
```

[53]:
```
                       0
Region Team
North  One  Revenue  7500
            Cost     5200
West   One  Revenue  5500
            Cost     5100
East   One  Revenue  2750
            Cost     4400
South  One  Revenue  6400
            Cost     5300
North  Two  Revenue  2300
```

```
             Cost     1250
West   Two  Revenue  3750
             Cost     1300
East   Two  Revenue  1900
             Cost     2100
South  Two  Revenue   575
             Cost       50
```

[54]:
```
stacked.unstack('Region')
```

[54]:
```
                 0
Region        East North South  West
Team
One  Revenue  2750  7500  6400  5500
     Cost     4400  5200  5300  5100
Two  Revenue  1900  2300   575  3750
     Cost     2100  1250    50  1300
```

[55]:
```
df.head(3)
```

[55]:
```
  Region Team  Revenue  Cost
0  North  One     7500  5200
1   West  One     5500  5100
2   East  One     2750  4400
```

[56]:
```
df.melt(id_vars=['Region','Team'], var_name='value type')
```

[56]:
```
    Region Team value type  value
0   North  One    Revenue   7500
1    West  One    Revenue   5500
2    East  One    Revenue   2750
3   South  One    Revenue   6400
4   North  Two    Revenue   2300
5    West  Two    Revenue   3750
6    East  Two    Revenue   1900
7   South  Two    Revenue    575
8   North  One       Cost   5200
9    West  One       Cost   5100
10   East  One       Cost   4400
11  South  One       Cost   5300
12  North  Two       Cost   1250
13   West  Two       Cost   1300
14   East  Two       Cost   2100
15  South  Two       Cost     50
```

[57]:
```
# mean by default
df.pivot_table(index='Team',columns='Region',values='Revenue')
```

[57]:
```
Region  East  North  South  West
Team
One     2750   7500   6400  5500
Two     1900   2300    575  3750
```

[58]:
```
df1 = pd.DataFrame({'letter': ['A', 'B', 'C', 'D'],
                    'number': [1, 2, 3, 4]})
df2 = pd.DataFrame({'letter': ['C', 'D', 'E', 'F'],
                    'number': [3, 4, 5, 6]})
```

[59]:
```
df1.merge(df2,how='left',on='number')
```

[59]:
```
  letter_x  number letter_y
0        A       1      NaN
1        B       2      NaN
2        C       3        C
3        D       4        D
```

[60]:
```
df1.merge(df2,how='inner',left_on='number',right_on='number')
```

[60]:
```
  letter_x  number letter_y
0        C       3        C
1        D       4        D
```

[61]:
```
df1.merge(df2,how='right',on='number',suffixes=('','_right'))
```

[61]:
```
  letter  number letter_right
0      C       3            C
1      D       4            D
2    NaN       5            E
3    NaN       6            F
```

[62]:
```
# drop duplicates with .drop_duplicates()
df3 = pd.concat([df1,df2]).drop_duplicates().reset_index(drop=True)
df3
```

[62]:
```
  letter  number
0      A       1
1      B       2
2      C       3
3      D       4
4      E       5
5      F       6
```

[63]:
```
df4 = pd.concat([df1,df2],axis=1)
df4
```

[63]:

	letter	number	letter	number
0	A	1	C	3
1	B	2	D	4
2	C	3	E	5
3	D	4	F	6

[64]:
```
new_row = pd.Series(['Z',26],index=df3.columns)
df3.append(new_row,ignore_index=True)
```

```
C:\Users\aadar\AppData\Local\Temp\ipykernel_11968\696788980.py:2: FutureWarning:
The frame.append method is deprecated and will be removed from pandas in a
future version. Use pandas.concat instead.
  df3.append(new_row,ignore_index=True)
```

[64]:

	letter	number
0	A	1
1	B	2
2	C	3
3	D	4
4	E	5
5	F	6
6	Z	26

[65]:
```
join_df = pd.DataFrame({'letter': ['F','G', 'H', 'I'],
                        'number': [6, 7, 8, 9]})
```

[66]:
```
df2.join(join_df, rsuffix='_right')
```

[66]:

	letter	number	letter_right	number_right
0	C	3	F	6
1	D	4	G	7
2	E	5	H	8
3	F	6	I	9

[67]:
```
df = pd.DataFrame({"Species":['Chinook','Chum','Coho','Steelhead','Bull Trout'],
             "Population":['Skokomish','Lower␣
↪Skokomish','Skokomish','Skokomish','SF Skokomish'],
             "Count":[1208,2396,3220,6245,8216]})
df
```

[67]:

	Species	Population	Count
0	Chinook	Skokomish	1208
1	Chum	Lower Skokomish	2396
2	Coho	Skokomish	3220
3	Steelhead	Skokomish	6245
4	Bull Trout	SF Skokomish	8216

[68]:
```
import numpy as np
bins = [0, 2000, 4000, 6000, 8000, np.inf]
labels = ['Low Return', 'Below Avg Return', 'Avg Return', 'Above Avg Return',␣
 ↪'High Return']
```

[69]:
```
df['Count Category'] = pd.cut(df['Count'], bins, labels=labels)
df
```

[69]:
```
      Species       Population  Count    Count Category
0     Chinook        Skokomish   1208        Low Return
1        Chum  Lower Skokomish   2396  Below Avg Return
2        Coho        Skokomish   3220  Below Avg Return
3   Steelhead        Skokomish   6245  Above Avg Return
4  Bull Trout     SF Skokomish   8216       High Return
```

[70]:
```
fed_status ={"Chinook":"Threatened",
"Chum":"Not Warranted",
"Coho":"Not Warranted",
"Steelhead":"Threatened"}
```

[71]:
```
df['Federal Status'] = df['Species'].map(fed_status)
df
```

[71]:
```
      Species       Population  Count    Count Category Federal Status
0     Chinook        Skokomish   1208        Low Return     Threatened
1        Chum  Lower Skokomish   2396  Below Avg Return  Not Warranted
2        Coho        Skokomish   3220  Below Avg Return  Not Warranted
3   Steelhead        Skokomish   6245  Above Avg Return     Threatened
4  Bull Trout     SF Skokomish   8216       High Return            NaN
```

[72]:
```
df['Count Category'] = pd.Categorical(df['Count Category'],
                ordered=True,
                categories=labels)
df['Count Category']
```

[72]:
```
0          Low Return
1    Below Avg Return
2    Below Avg Return
3    Above Avg Return
4         High Return
Name: Count Category, dtype: category
Categories (5, object): ['Low Return' < 'Below Avg Return' < 'Avg Return' <
'Above Avg Return' < 'High Return']
```

[73]:
```
df.sort_values(by=['Count Category'],ascending=False)
```

[73]:
```
      Species       Population  Count    Count Category Federal Status
4  Bull Trout     SF Skokomish   8216       High Return            NaN
```

```
3   Steelhead         Skokomish   6245  Above Avg Return     Threatened
1        Chum  Lower Skokomish   2396  Below Avg Return  Not Warranted
2        Coho         Skokomish   3220  Below Avg Return  Not Warranted
0     Chinook         Skokomish   1208         Low Return     Threatened
```

[74]:
```
pd.get_dummies(df['Count Category'])
```

[74]:

	Low Return	Below Avg Return	Avg Return	Above Avg Return	High Return
0	1	0	0	0	0
1	0	1	0	0	0
2	0	1	0	0	0
3	0	0	0	1	0
4	0	0	0	0	1

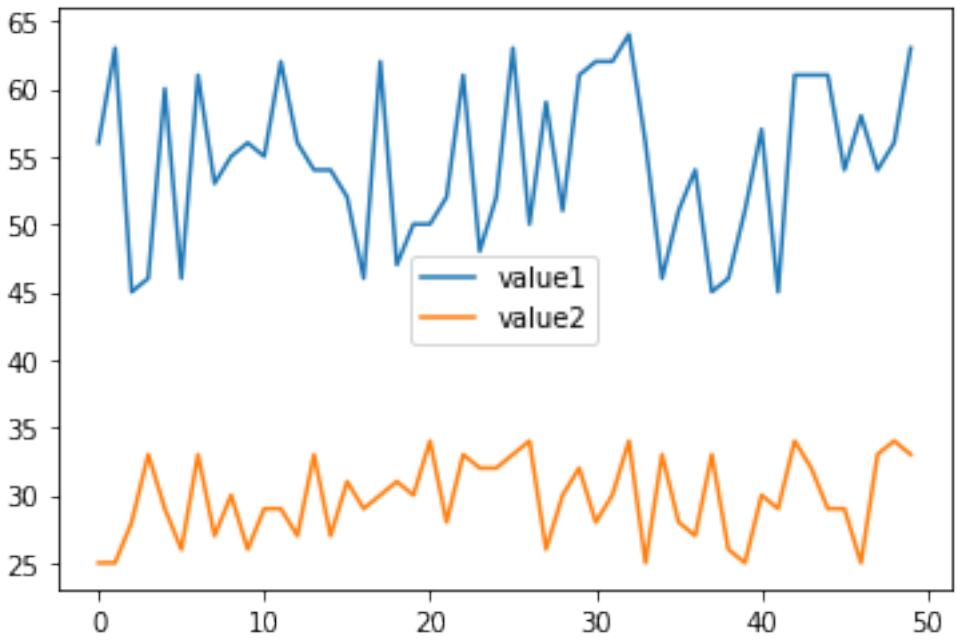
4 Chapter4

[75]:
```
import numpy as np
daterange = pd.period_range('1/1/1950', freq='1d', periods=50)
date_df = pd.DataFrame(data=daterange,columns=['day'])
date_df['value1'] = np.random.randint(45,65,size=(len(date_df)))
date_df['value2'] = np.random.randint(25,35,size=(len(date_df)))
date_df.head(3)
```

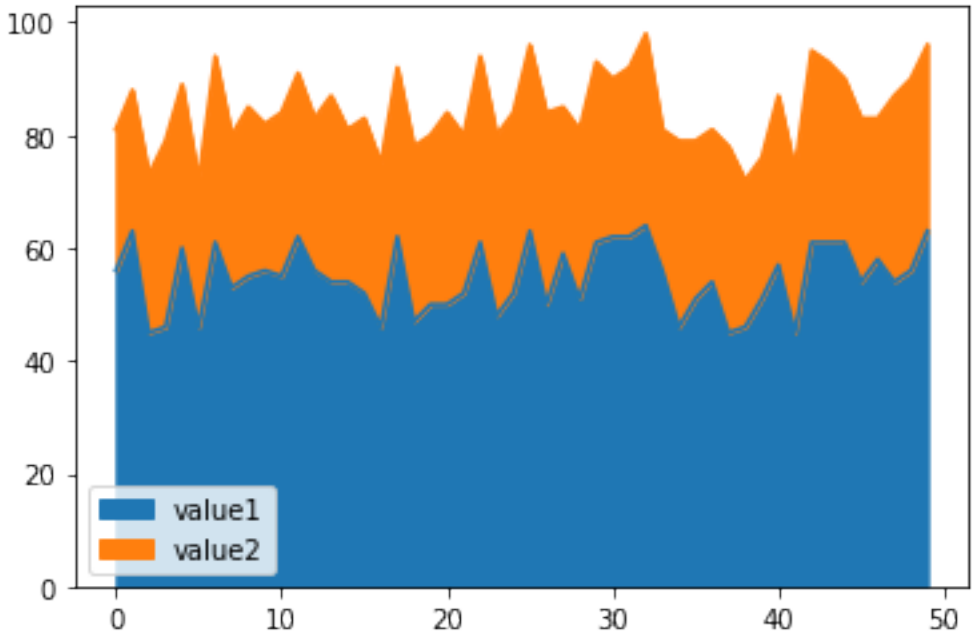
[75]:

	day	value1	value2
0	1950-01-01	56	25
1	1950-01-02	63	25
2	1950-01-03	45	28

[76]:
```
ax = date_df.plot();
```

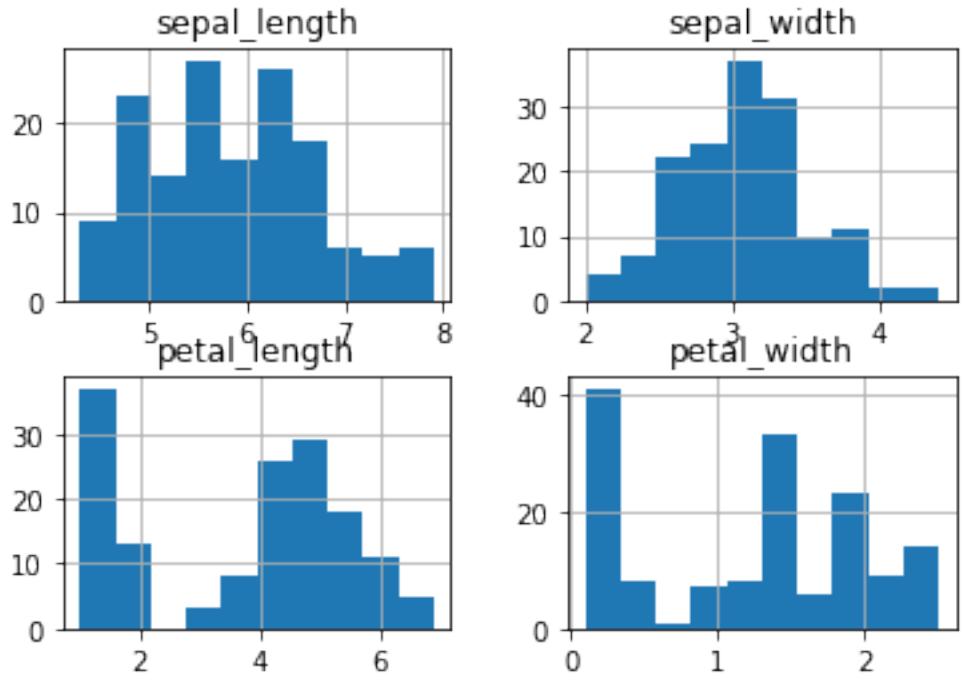


[77]:
```
date_df.plot.area(stacked=True);
```

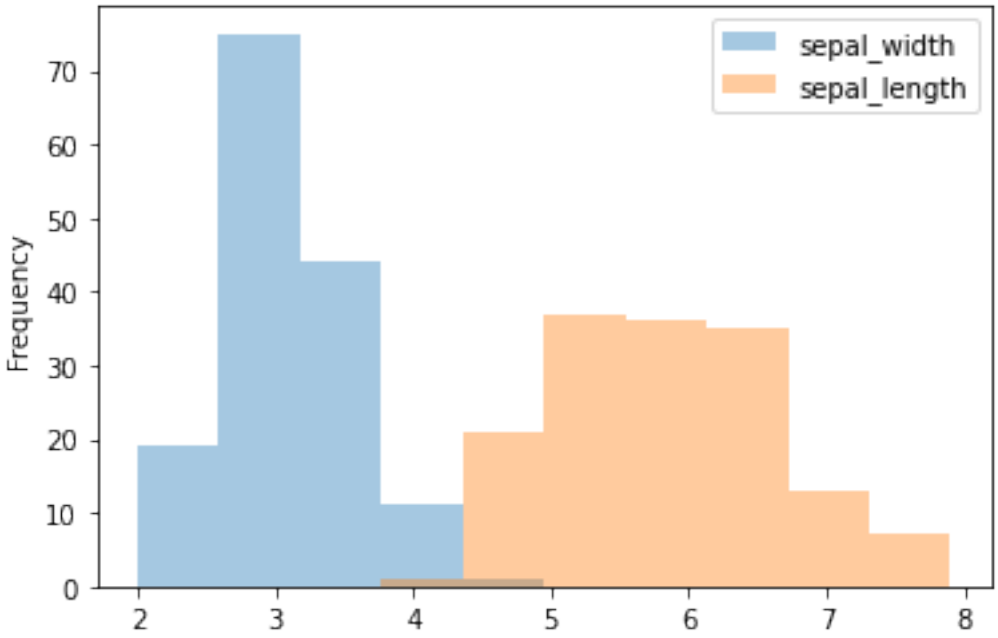


[78]:
```
iris = pd.read_csv('./iris.csv')
```

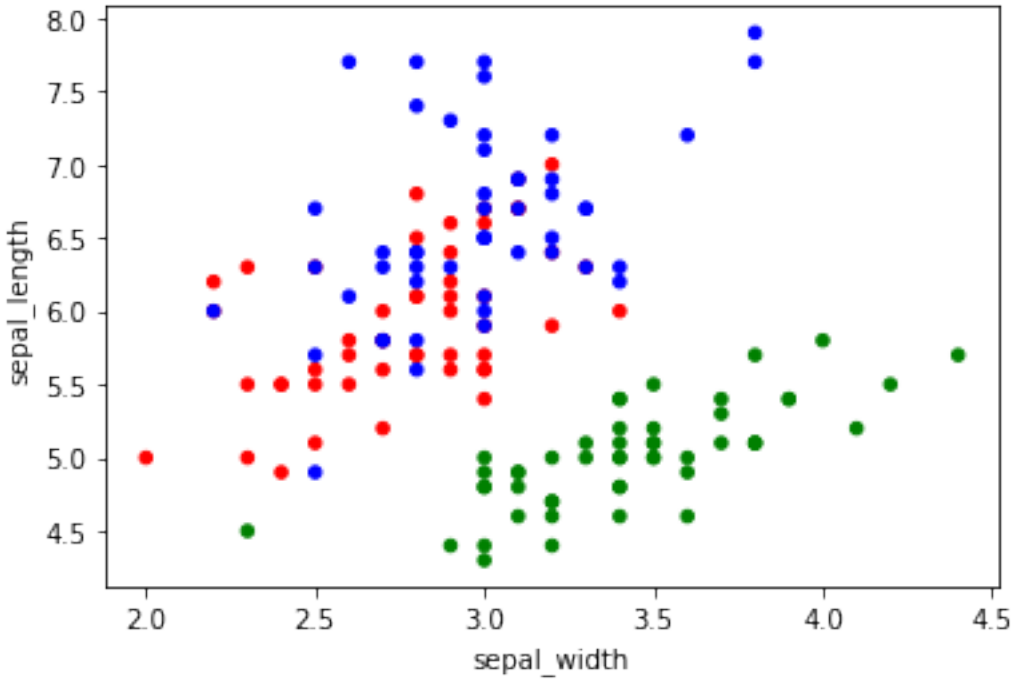
[79]:
```
iris.hist();
```



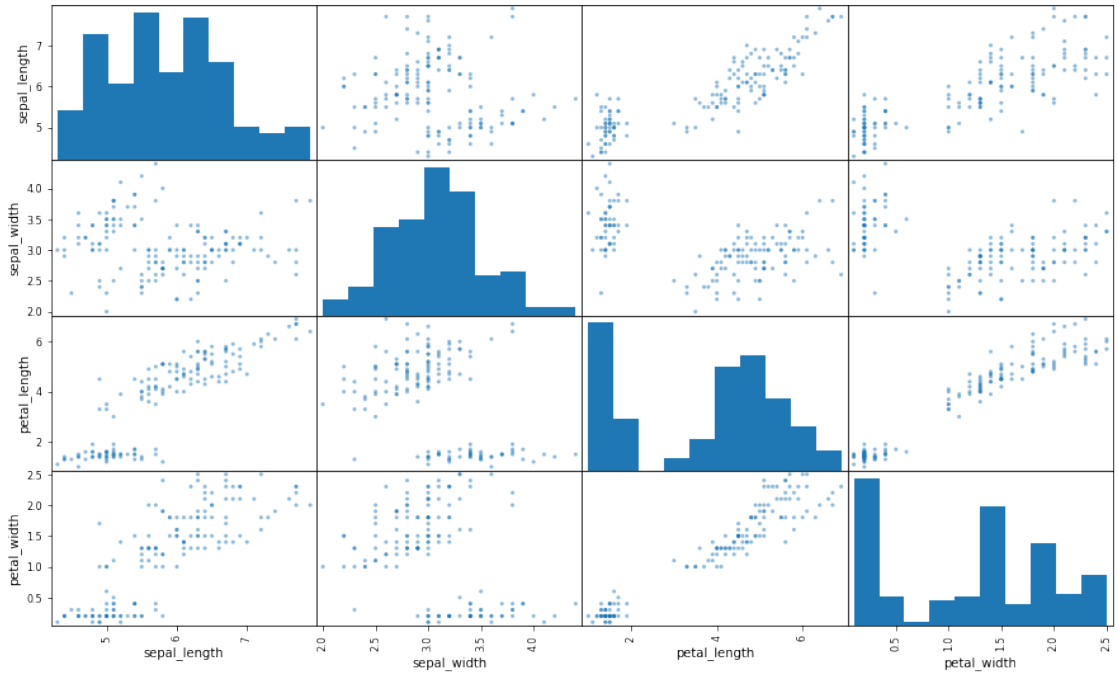
[80]:
```
iris[['sepal_width','sepal_length']].plot.hist(alpha=0.4);
```



[81]:
```
colors = {"versicolor":"red","setosa":"green","virginica":"blue"}
iris['colors'] = iris['species'].map(colors)
iris.plot.scatter(x='sepal_width', y='sepal_length', color=iris['colors']);
```



[82]:
```
from pandas.plotting import scatter_matrix
scatter_matrix(iris,figsize=(15, 9),);
```



[83]:
```
iris.mean()
```

```
C:\Users\aadar\AppData\Local\Temp\ipykernel_11968\935066809.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.
  iris.mean()
```

[83]:
```
sepal_length    5.84
sepal_width     3.06
petal_length    3.76
petal_width     1.20
dtype: float64
```

[84]:
```
iris.median()
```

```
C:\Users\aadar\AppData\Local\Temp\ipykernel_11968\1297003277.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.
  iris.median()
```

[84]:
```
sepal_length    5.80
sepal_width     3.00
petal_length    4.35
petal_width     1.30
```

```
dtype: float64
```

[85]:
```
iris.mode()
```

[85]:

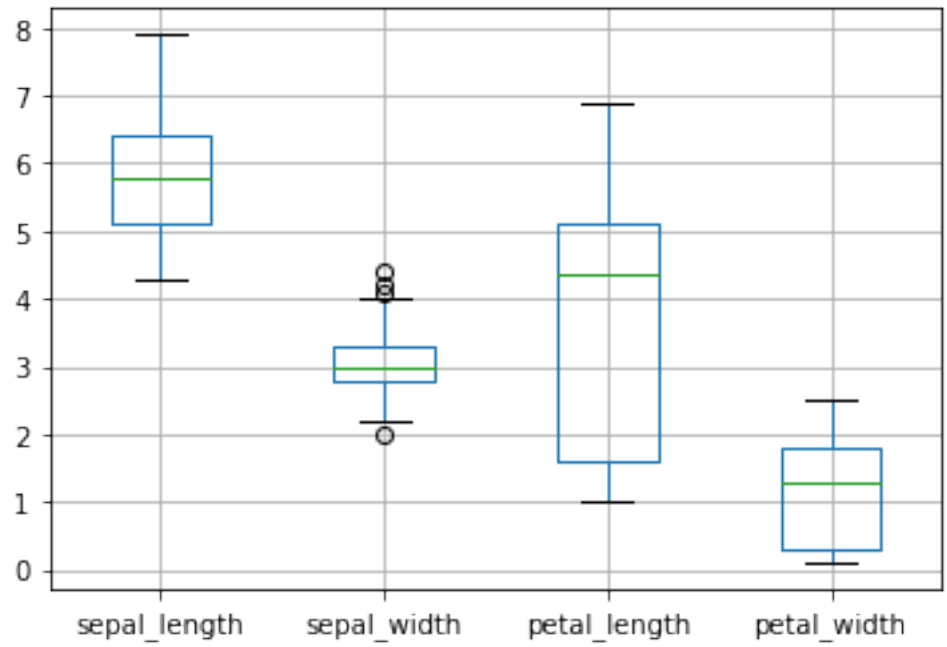
	sepal_length	sepal_width	petal_length	petal_width	species	colors
0	5.00	3.00	1.40	0.20	setosa	blue
1	NaN	NaN	1.50	NaN	versicolor	green
2	NaN	NaN	NaN	NaN	virginica	red

[86]:
```
iris.std()
```

```
C:\Users\aadar\AppData\Local\Temp\ipykernel_11968\3849825860.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.
  iris.std()
```

[86]:
```
sepal_length    0.83
sepal_width     0.44
petal_length    1.77
petal_width     0.76
dtype: float64
```

[87]:
```
iris.boxplot();
```



[88]:
```
iris.describe()
```

[88]:

	sepal_length	sepal_width	petal_length	petal_width
count	150.00	150.00	150.00	150.00
mean	5.84	3.06	3.76	1.20
std	0.83	0.44	1.77	0.76
min	4.30	2.00	1.00	0.10
25%	5.10	2.80	1.60	0.30
50%	5.80	3.00	4.35	1.30
75%	6.40	3.30	5.10	1.80
max	7.90	4.40	6.90	2.50

[89]:
```
iris.corr()
```

[89]:

	sepal_length	sepal_width	petal_length	petal_width
sepal_length	1.00	-0.12	0.87	0.82
sepal_width	-0.12	1.00	-0.43	-0.37
petal_length	0.87	-0.43	1.00	0.96
petal_width	0.82	-0.37	0.96	1.00

[90]:
```
iris.corr().style.background_gradient(cmap='RdYlGn', axis=None)
```

[90]: <pandas.io.formats.style.Styler at 0x2a02f402880>

5 Chapter 5

[91]:
```
# !pip install pandas_profiling
from pandas_profiling import ProfileReport
```

[92]:
```
iris = pd.read_csv('./iris.csv')
```

[93]:
```
profile = ProfileReport(iris,title="Iris Data Profile")
```

[94]:
```
profile.to_notebook_iframe()

#profile.to_widgets() if using Jupyter
```

```
Summarize dataset:   0%|          | 0/5 [00:00<?, ?it/s]

Generate report structure:   0%|          | 0/1 [00:00<?, ?it/s]

Render HTML:   0%|          | 0/1 [00:00<?, ?it/s]

<IPython.core.display.HTML object>
```

[95]:
```
# profile.to_file("iris-profile.html")
```

[96]:
```
import pandas as pd
import geopandas
```

```
PROJ: proj_create_from_database: SQLite error on SELECT name, type,
coordinate_system_auth_name, coordinate_system_code, datum_auth_name,
```

```
datum_code, area_of_use_auth_name, area_of_use_code, text_definition, deprecated
FROM geodetic_crs WHERE auth_name = ? AND code = ?: no such column:
area_of_use_auth_name
```

[97]:
```
peaks = pd.DataFrame(
    {'Peak Name': ['Green Mtn.', 'So. Boulder Peak', 'Bear Peak', 'Flagstaff␣
↪Mtn.', 'Mt. Sanitas'],
     'Latitude': [39.9821, 39.9539, 39.9603, 40.0017, 40.0360968],
     'Longitude': [-105.3016, -105.2992, -105.2952, -105.3075, -105.3061024]})
```

[98]:
```
gdf = geopandas.GeoDataFrame(
    peaks, geometry=geopandas.points_from_xy(peaks.Longitude, peaks.Latitude))
gdf
```

[98]:

	Peak Name	Latitude	Longitude	geometry
0	Green Mtn.	39.98	-105.30	POINT (-105.30160 39.98210)
1	So. Boulder Peak	39.95	-105.30	POINT (-105.29920 39.95390)
2	Bear Peak	39.96	-105.30	POINT (-105.29520 39.96030)
3	Flagstaff Mtn.	40.00	-105.31	POINT (-105.30750 40.00170)
4	Mt. Sanitas	40.04	-105.31	POINT (-105.30610 40.03610)

[99]:
```
token ='your token'
```

[100]:
```
import plotly.express as px
px.set_mapbox_access_token(token)
gdf['size'] = 65

fig = px.scatter_mapbox(gdf,
                        lat=gdf.geometry.y,
                        lon=gdf.geometry.x,
                        color="Peak Name",
                        hover_name="Peak Name",
                        mapbox_style='outdoors',
                        size='size',
                        zoom=10)

fig.show()
```

[101]:
```
# !pip install Dask
```

[102]:
```
import dask.dataframe as dd
df = dd.read_csv('iris.csv')
df.head()
```

[102]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.10	3.50	1.40	0.20	setosa
1	4.90	3.00	1.40	0.20	setosa
2	4.70	3.20	1.30	0.20	setosa

```
3          4.60         3.10          1.50         0.20  setosa
4          5.00         3.60          1.40         0.20  setosa
```

[103]:
```
# !pip install databricks
```

[104]:
```
# !pip install pyspark
```

[105]:
```
import pandas as pd
import numpy as np
import databricks.koalas as ks
from pyspark.sql import SparkSession
```

```
WARNING:root:Found pyspark version "3.2.1" installed. The pyspark version 3.2
and above has a built-in "pandas APIs on Spark" module ported from Koalas. Try
`import pyspark.pandas as ps` instead.
WARNING:root:'PYARROW_IGNORE_TIMEZONE' environment variable was not set. It is
required to set this environment variable to '1' in both driver and executor
sides if you use pyarrow>=2.0.0. Koalas will set it for you but it does not work
if there is a Spark context already launched.
```

[106]:
```
pdf = pd.DataFrame(np.random.randn(6, 4),  columns=list('ABCD'))
pdf.head()
```

[106]:

	A	B	C	D
0	0.28	0.31	0.87	-0.17
1	0.12	1.25	-0.34	1.33
2	0.85	0.25	-0.63	-0.88
3	0.00	1.56	0.09	0.77
4	-0.50	-0.42	-0.06	-1.40

[ ]:
```
kdf = ks.from_pandas(pdf)
kdf.head()
```