Hands-on Activity 9.2 Customized Visualizations using Seaborn - Colaboratory

#### Instructions:

Create a Python notebook to answer all shown procedures, exercises and analysis in this section.

#### Resources:

Download the following datasets:

- fb\_stock\_prices\_2018.csv
- earthquakes-1.csv

### Procedures:

sns.stripplot(

4.0

mb

mww

- 9.4 Introduction to Seaborn
- 9.5 Formatting Plots
- 9.6 Customizing Visualizations

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
import pandas as pd
fb = pd.read_csv('/content/fb_stock_prices_2018.csv', index_col='date', parse_dates=True
)
quakes = pd.read_csv('/content/earthquakes.csv')

quakes.assign(
time=lambda x: pd.to_datetime(x.time, unit='ms')
).set_index('time').loc('2018-09-28'].query(
"parsed_place == 'Indonesia' and tsunami == 1 and mag == 7.5"
)
```

time mag magType place tsunami parsed\_place

2018-09-28 10:02:43.480 7.5 mww 78km N of Palu, Indonesia 1 Indonesia

```
x='magType',
y='mag',
hue='tsunami',
data=quakes.query('parsed_place == "Indonesia"')
)

<Axes: xlabel='magType', ylabel='mag'>

7.5

7.0

6.5

6.0

9

5.5

5.0

4.5
```

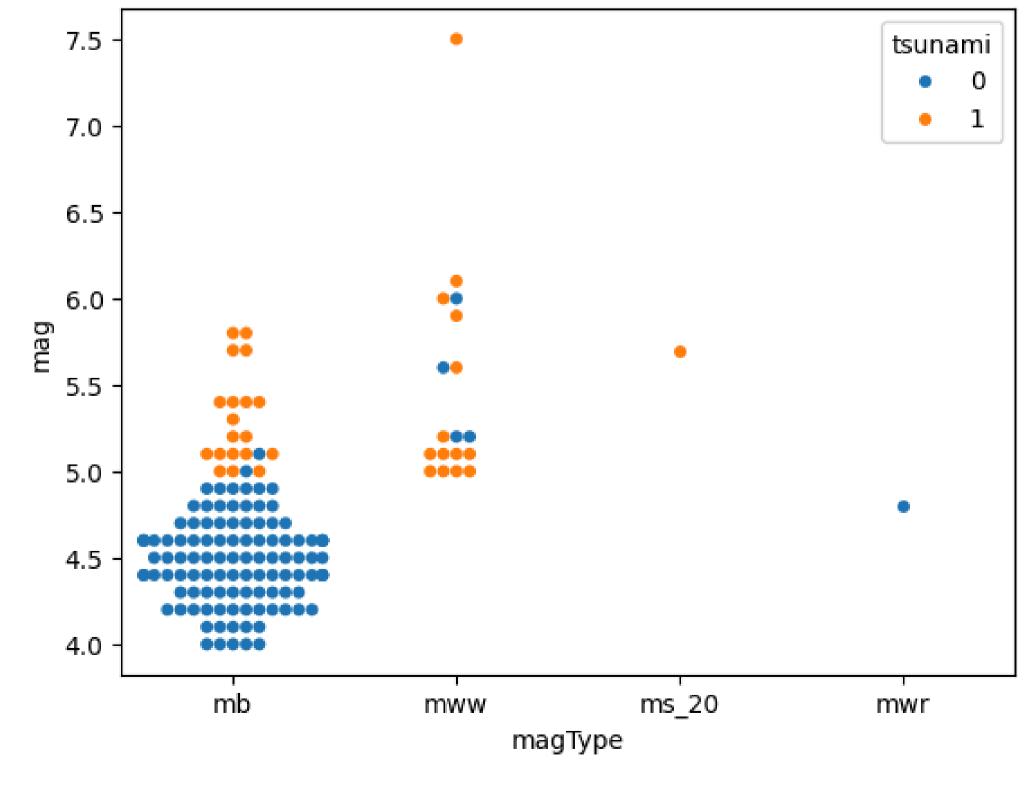
ms\_20

magType

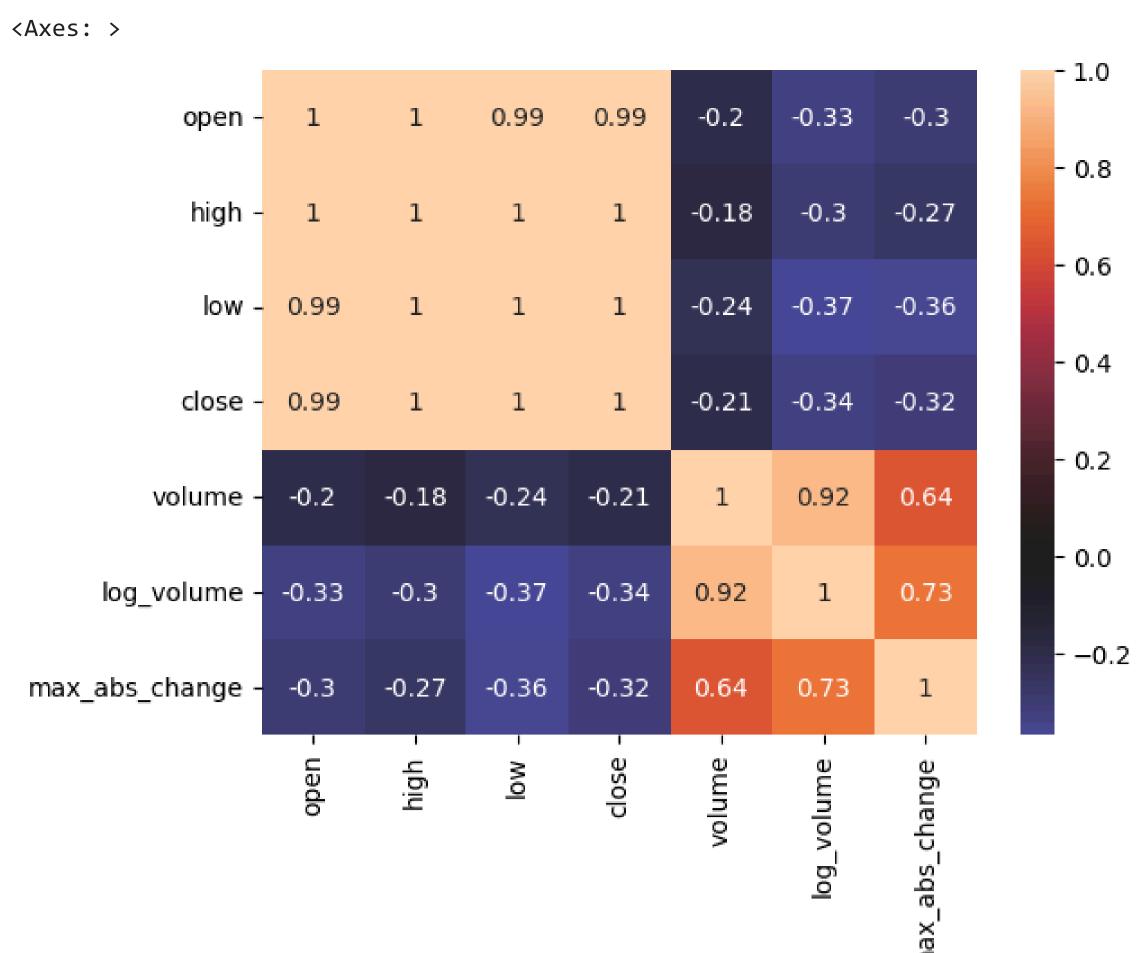
mwr

```
sns.swarmplot(
x='magType',
y='mag',
hue='tsunami',
data=quakes.query('parsed_place == "Indonesia"')
)
```

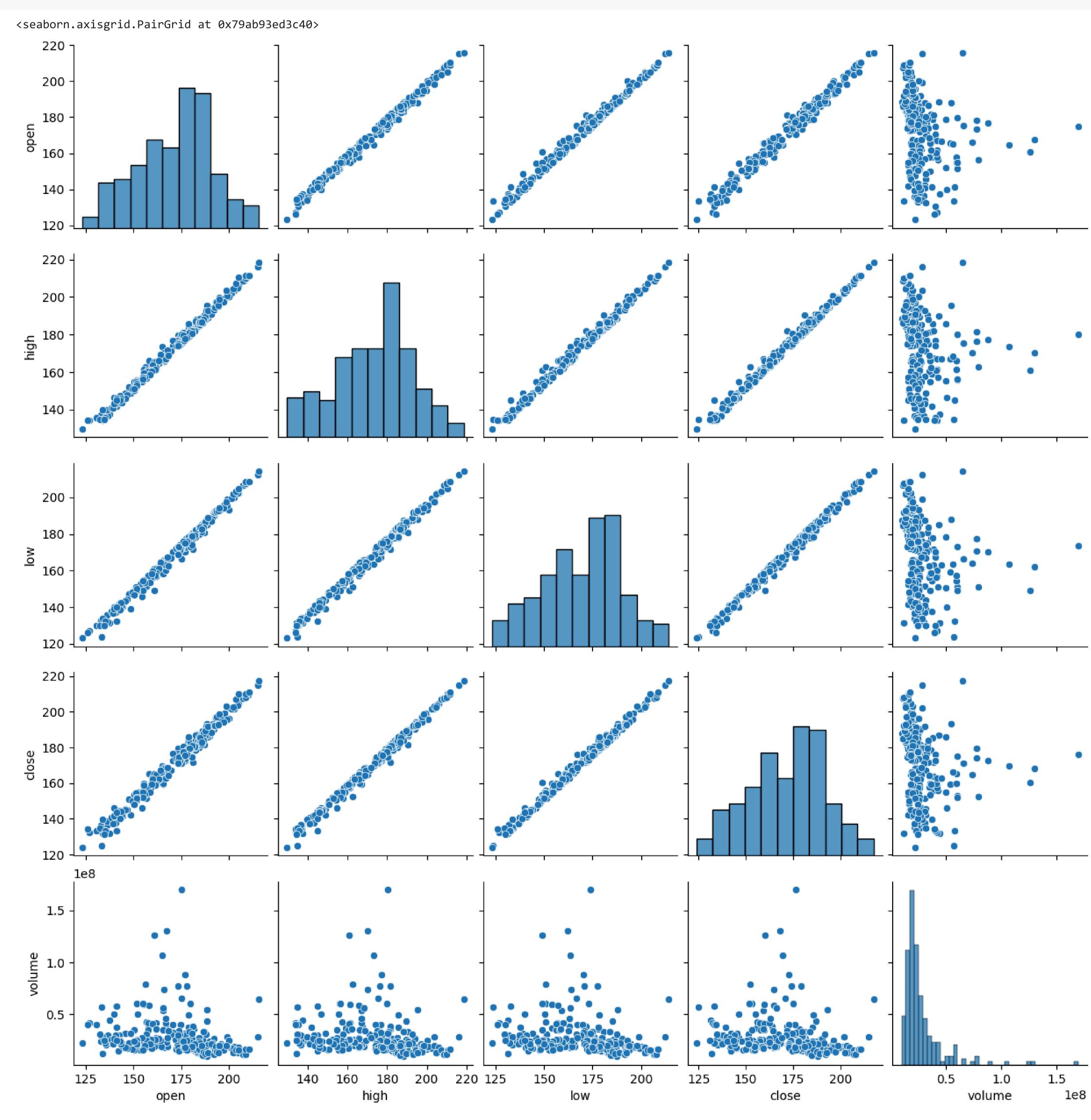
<Axes: xlabel='magType', ylabel='mag'>
/usr/local/lib/python3.10/dist-packages/seaborn/categorical.py:3398: UserWarning: 10.2% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.
warnings.warn(msg, UserWarning)



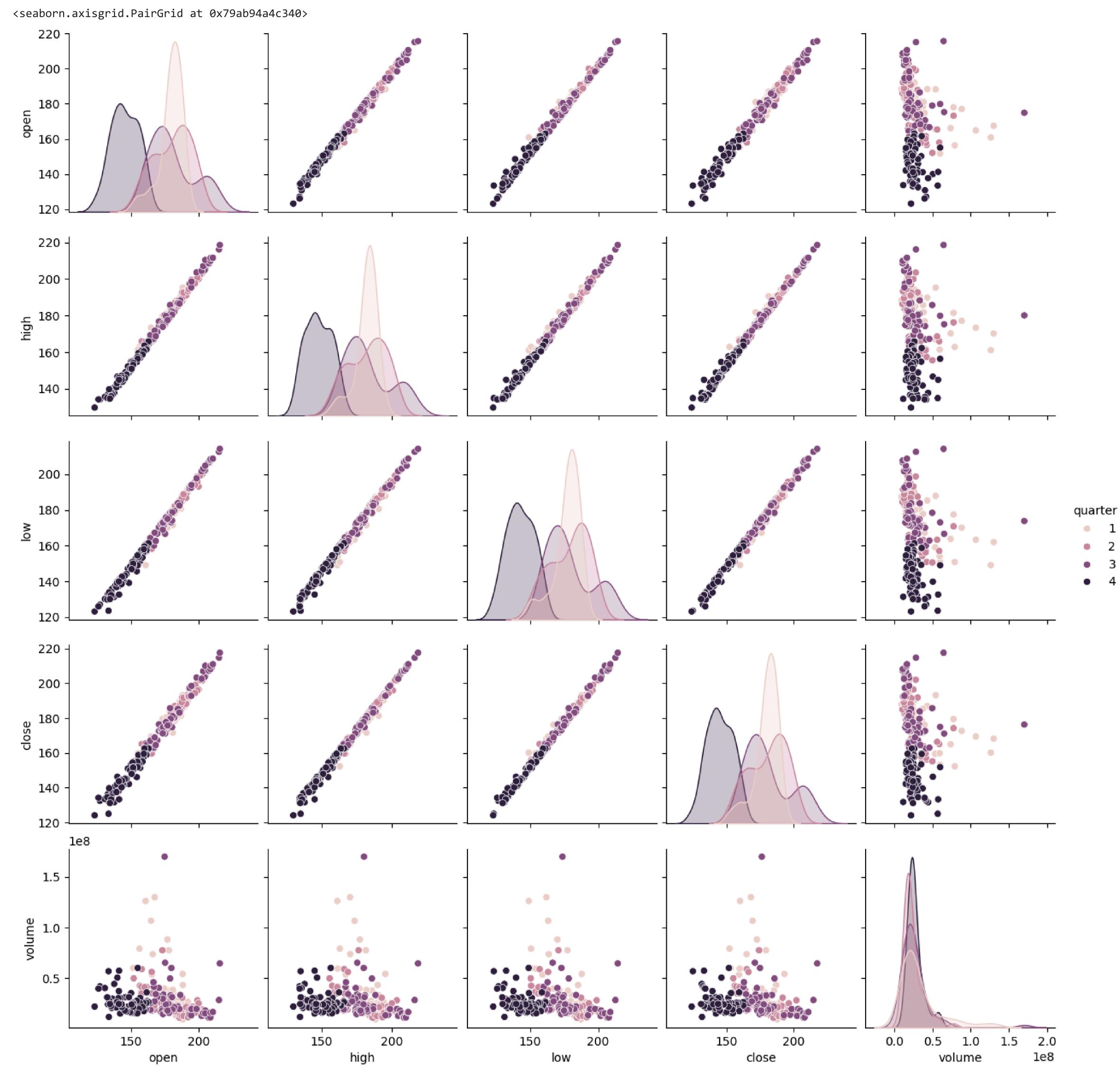
```
sns.heatmap(
fb.sort_index().assign(
log_volume=np.log(fb.volume),
max_abs_change=fb.high - fb.low
).corr(),
annot=True, center=0
```



#### sns.pairplot(fb)

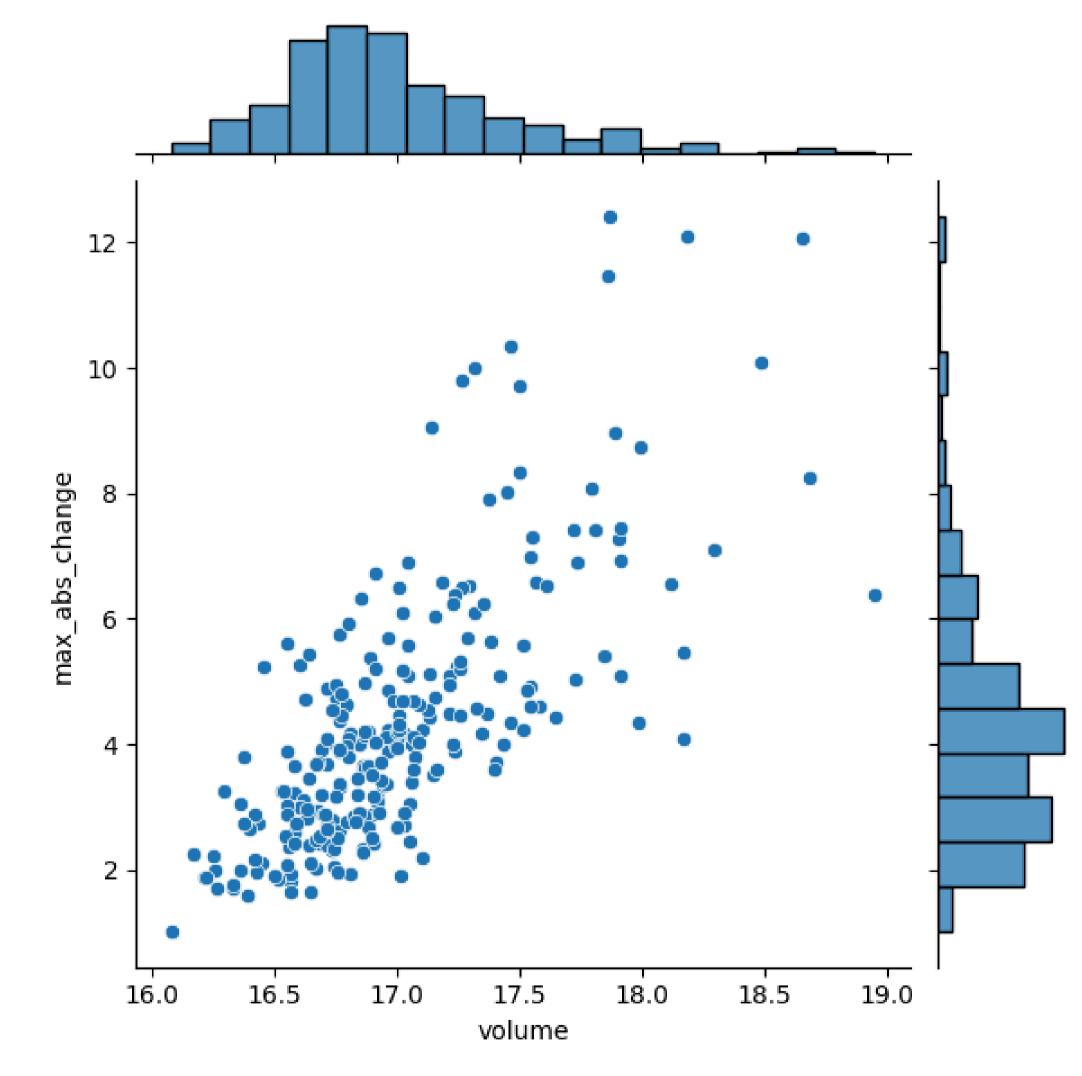


sns.pairplot(
fb.assign(quarter=lambda x: x.index.quarter),
diag\_kind='kde',
hue='quarter'



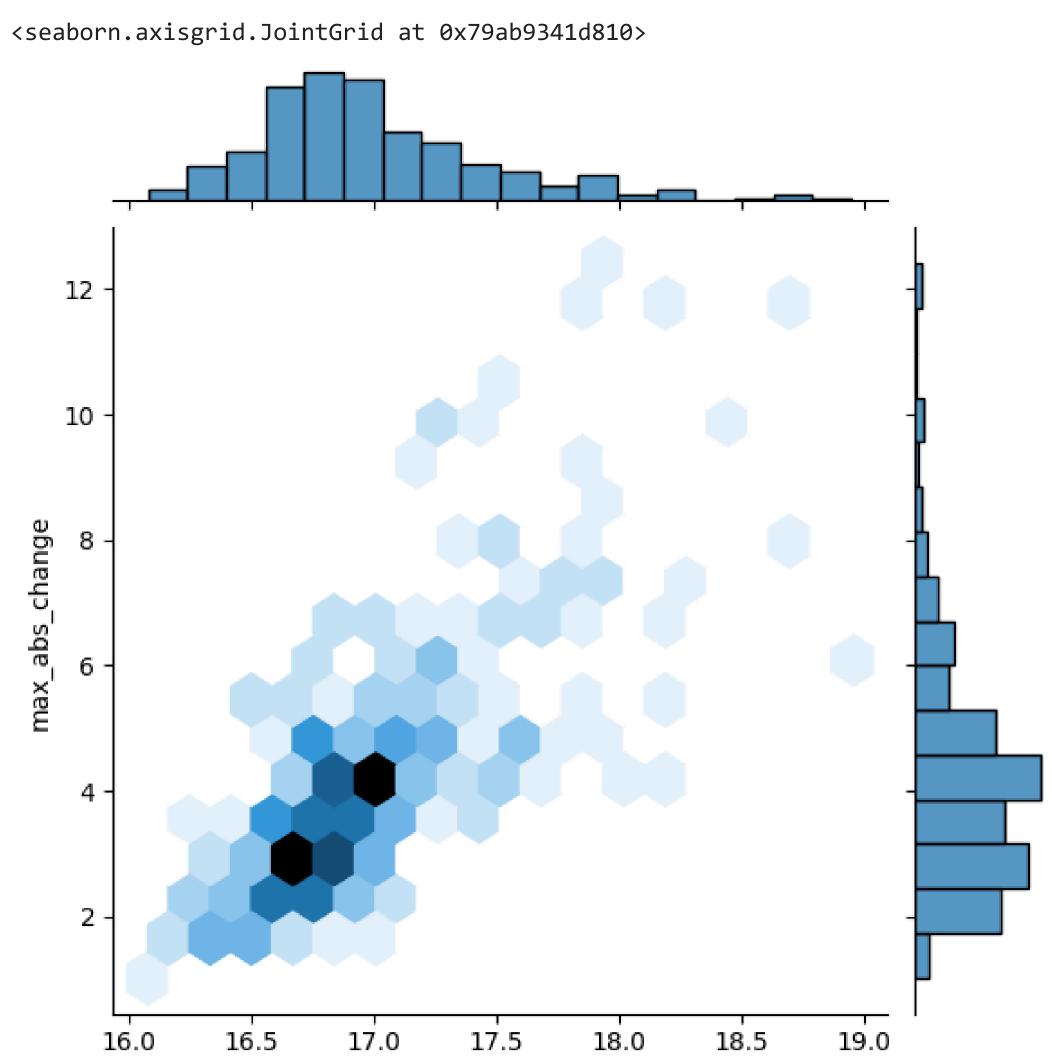
sns.jointplot(
x='volume',
y='max\_abs\_change',
data=fb.assign(
volume=np.log(fb.volume),
max\_abs\_change=fb.high - fb.low
)

<seaborn.axisgrid.JointGrid at 0x79ab92f2ded0>



sns.jointplot(
x='volume',
y='max\_abs\_change',
kind='hex',
data=fb.assign(
volume=np.log(fb.volume),
max\_abs\_change=fb.high - fb.low
)

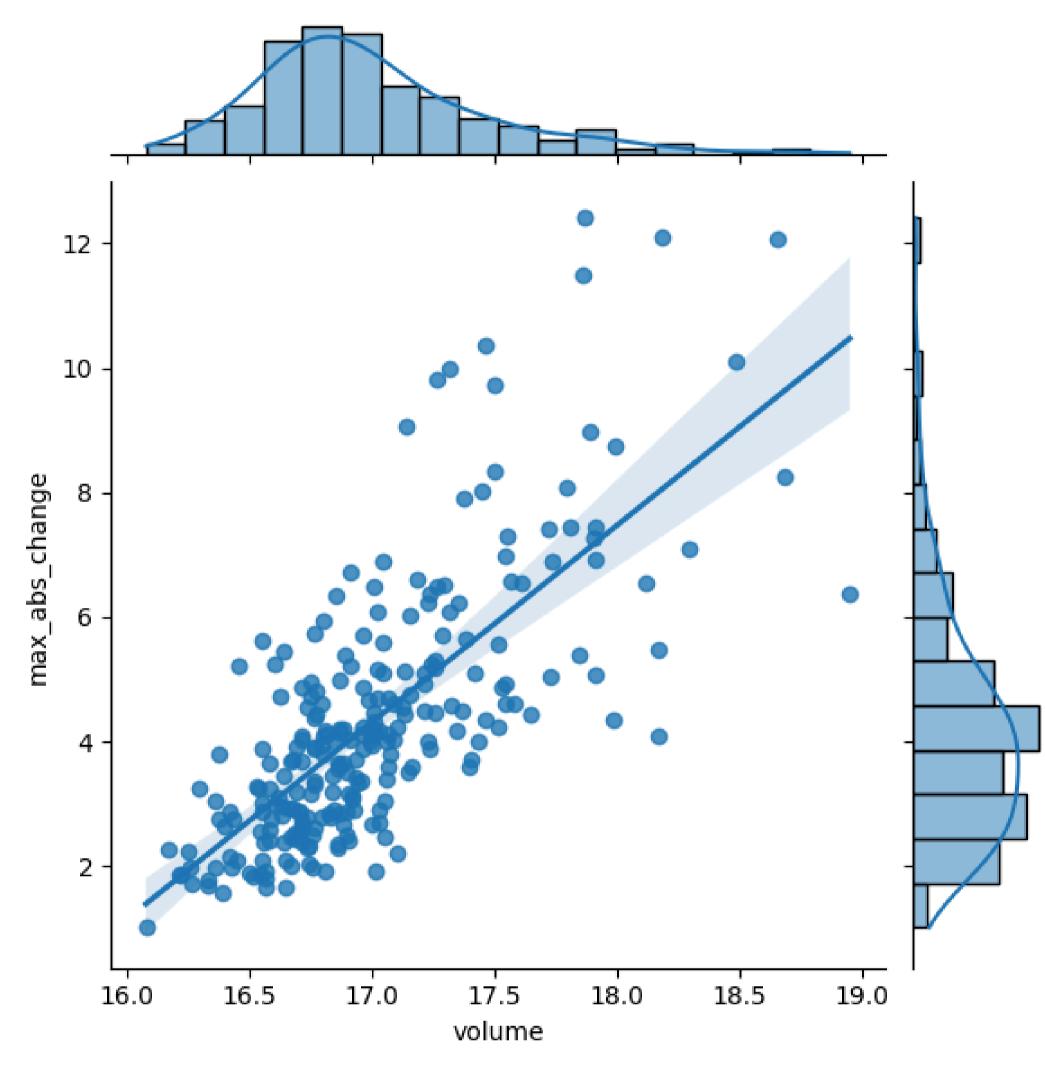
Hands-on Activity 9.2 Customized Visualizations using Seaborn - Colaboratory



volume

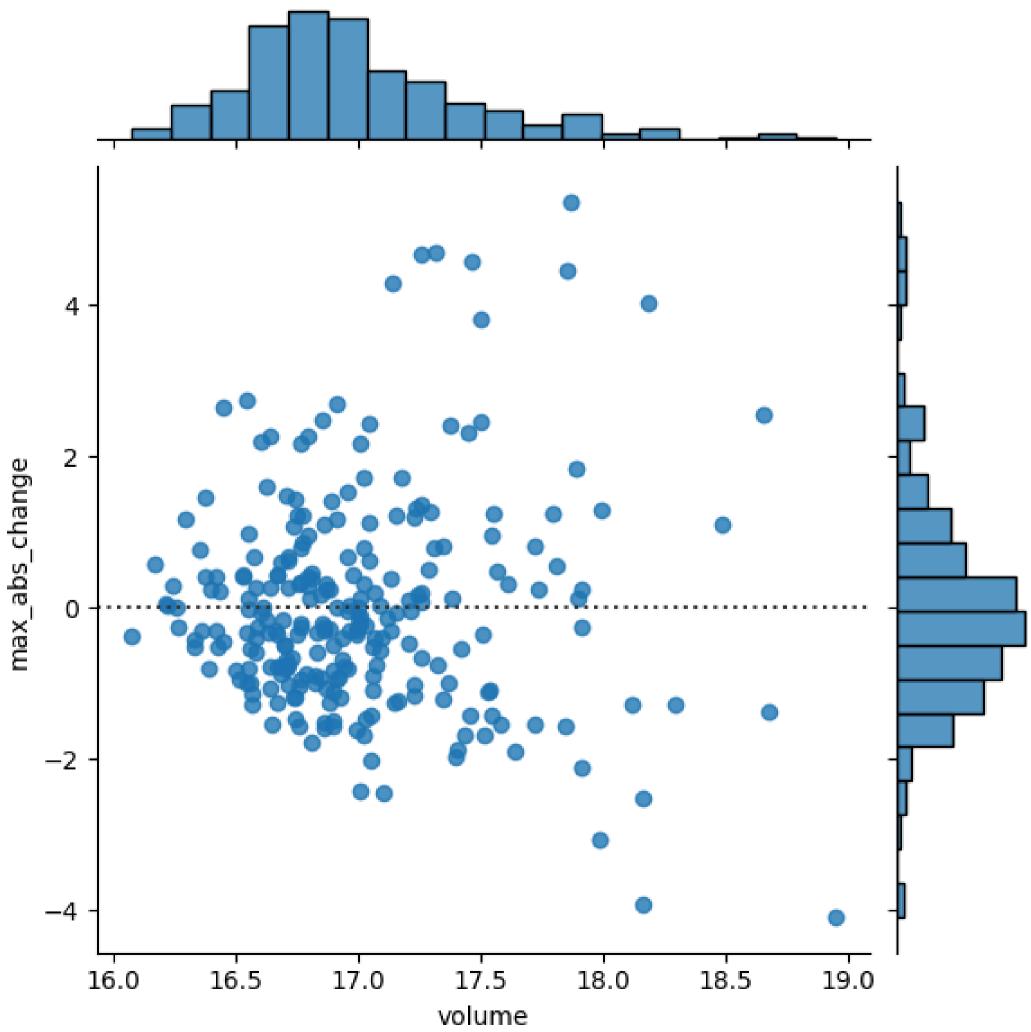
sns.jointplot(
x='volume',
y='max\_abs\_change',
kind='reg',
data=fb.assign(
volume=np.log(fb.volume),
max\_abs\_change=fb.high - fb.low
)

<seaborn.axisgrid.JointGrid at 0x79ab93d30c40>



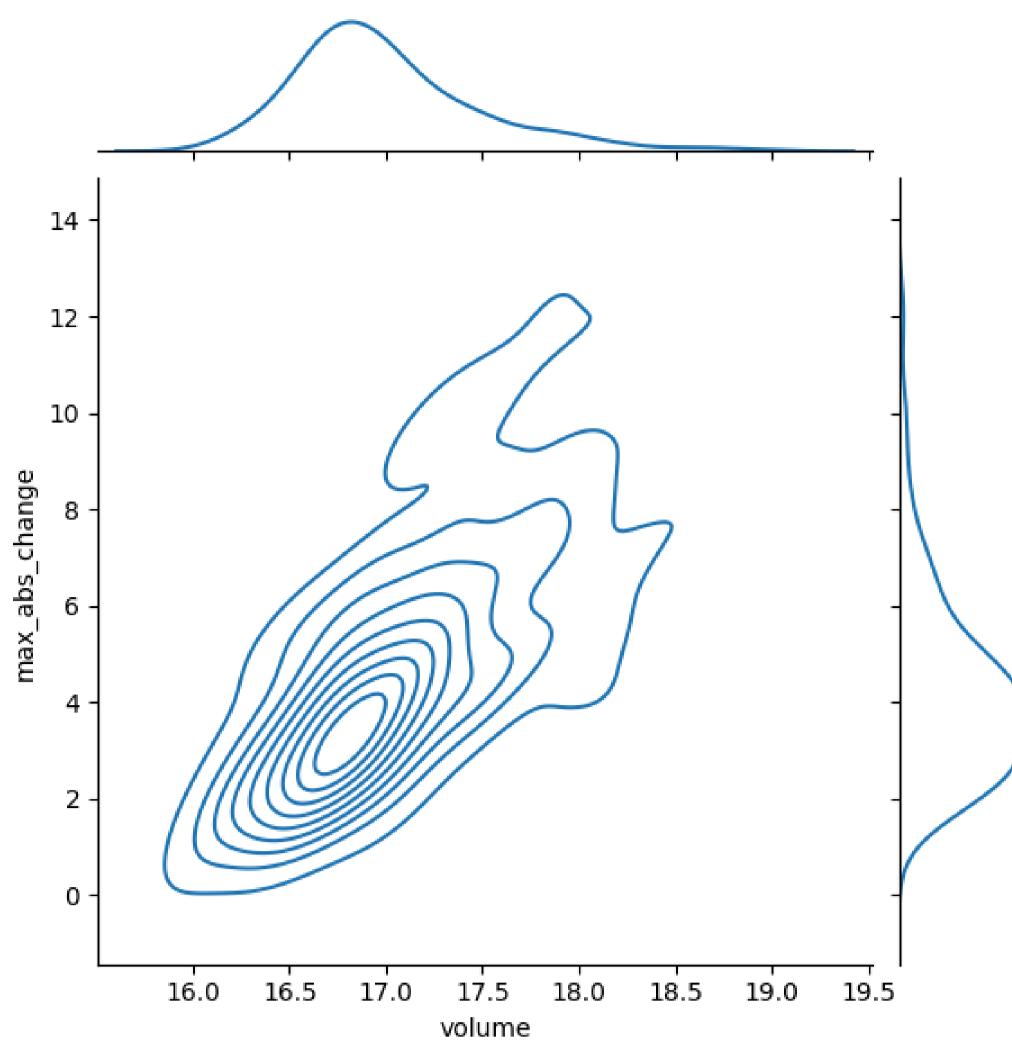
sns.jointplot(
x='volume',
y='max\_abs\_change',
kind='resid',
data=fb.assign(
volume=np.log(fb.volume),
max\_abs\_change=fb.high - fb.low
)

<seaborn.axisgrid.JointGrid at 0x79ab92800970>



```
sns.jointplot(
x='volume',
y='max_abs_change',
kind='kde',
data=fb.assign(
volume=np.log(fb.volume),
max_abs_change=fb.high - fb.low
)
)
)
```

<seaborn.axisgrid.JointGrid at 0x79ab92802830>



```
fb_reg_data = fb.assign(
    volume=np.log(fb.volume),
    max_abs_change=fb.high - fb.low
).iloc[:,-2:]

import itertools

iterator = itertools.repeat("I'm an iterator", 1)
for i in iterator:
    print(f'-->{i}')
print('This printed once because the iterator has been exhausted')
```

```
print(T -->{1} )
print('This printed once because the iterator has been exhausted')

# Reset the iterator
iterator = itertools.repeat("I'm an iterator", 1)
for i in iterator:
    print(f'-->{i}')
```

-->I'm an iterator
This printed once because the iterator has been exhausted
-->I'm an iterator

```
import itertools
iterable = list(itertools.repeat("I'm an iterable", 1))
for i in iterable:
    print(f'-->{i}')
print('This prints again because it\'s an iterable:')
for i in iterable:
    print(f'-->{i}')
```

-->I'm an iterable
This prints again because it's an iterable:
-->I'm an iterable

import itertools

```
import matplotlib.pyplot as plt
import seaborn as sns
def reg_resid_plots(data):
   Using seaborn, plot the regression and residuals
   plots side-by-side for every permutation of 2 columns
   in the data.
   Parameters:
        - data: A pandas DataFrame
    Returns:
        A matplotlib Figure object.
   num_cols = data.shape[1]
   permutation_count = num_cols * (num_cols - 1)
   fig, ax = plt.subplots(permutation_count, 2, figsize=(15, 8))
   for (x, y), axes, color in zip(
        itertools.permutations(data.columns, 2),
        ax,
        itertools.cycle(['royalblue', 'darkorange'])
```

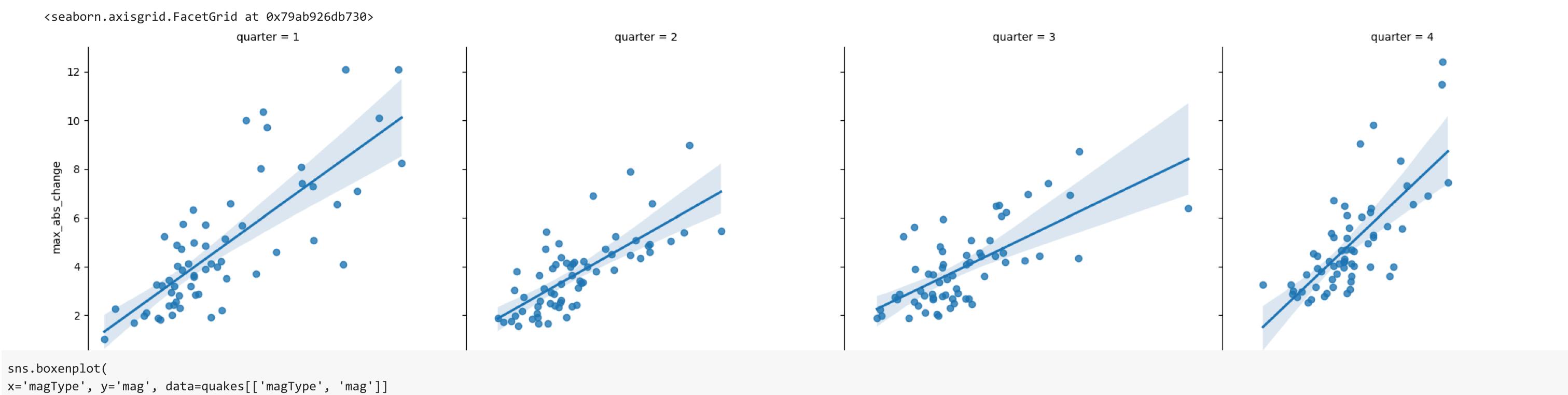
for subplot, func in zip(axes, (sns.regplot, sns.residplot)):

func(x=x, y=y, data=data, ax=subplot, color=color)

```
sns.lmplot(
x='volume',
y='max_abs_change',
data=fb.assign(
volume=np.log(fb.volume),
max_abs_change=fb.high - fb.low,
quarter=lambda x: x.index.quarter
),
col='quarter'
)
```

plt.close()

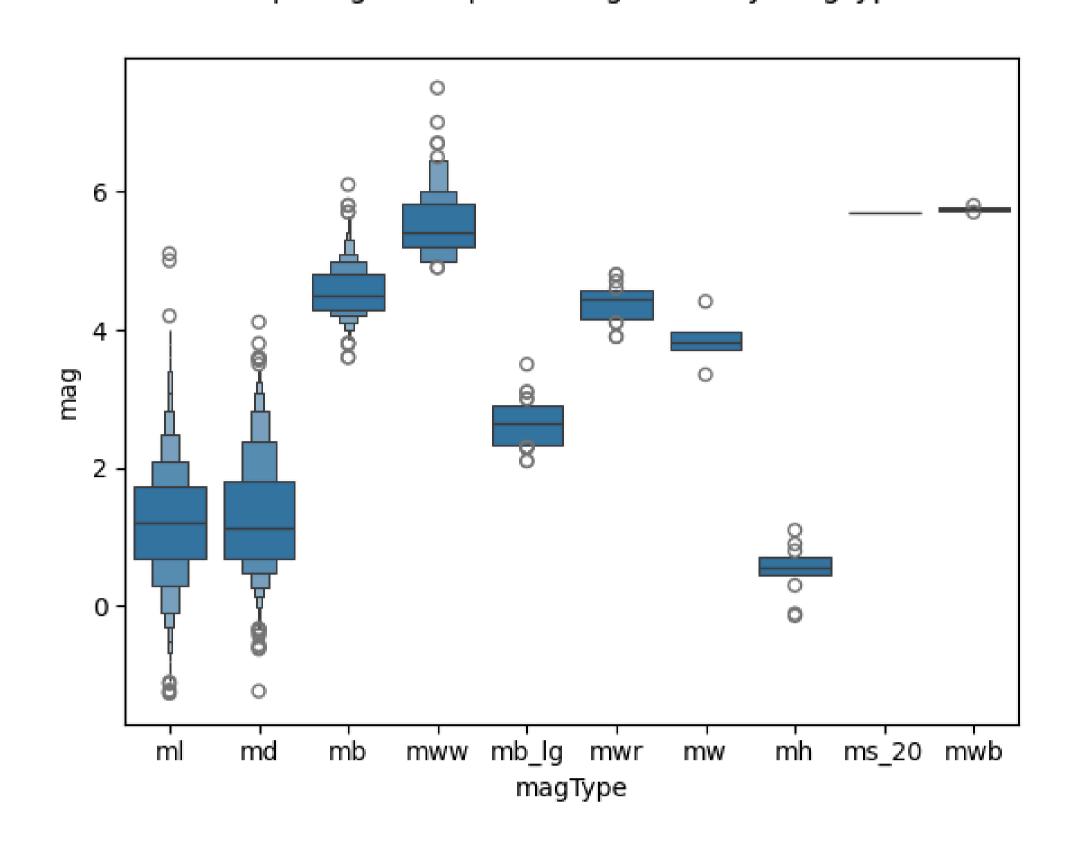
return fig



Text(0.5, 0.98, 'Comparing earthquake magnitude by magType')

Comparing earthquake magnitude by magType

plt.suptitle('Comparing earthquake magnitude by magType')



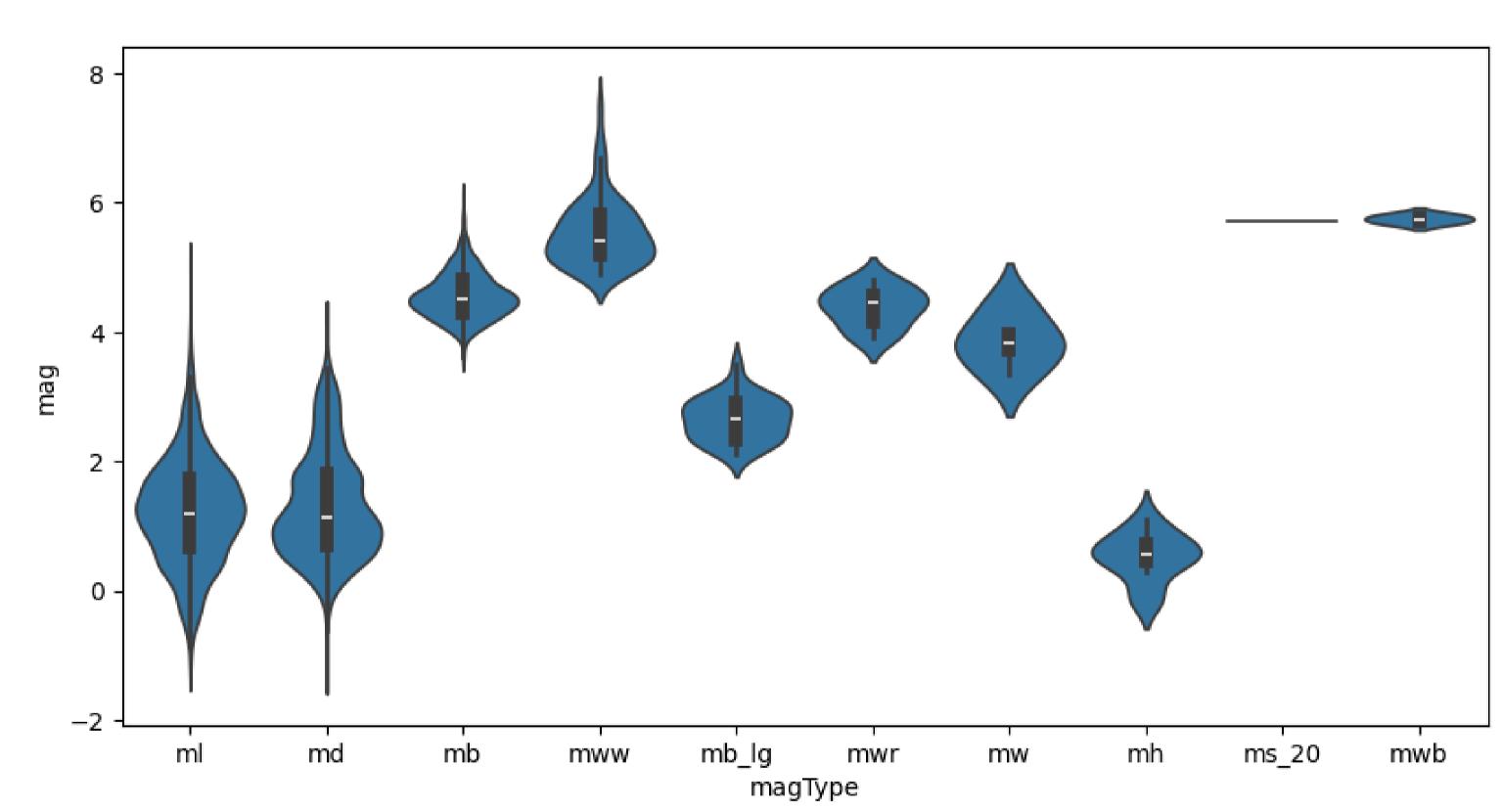
```
fig, axes = plt.subplots(figsize=(10, 5))
sns.violinplot(
x='magType', y='mag', data=quakes[['magType', 'mag']],
ax=axes, scale='width' # all violins have same width
)
plt.suptitle('Comparing earthquake magnitude by magType')
```

<ipython-input-95-4950d579cebb>:2: FutureWarning:

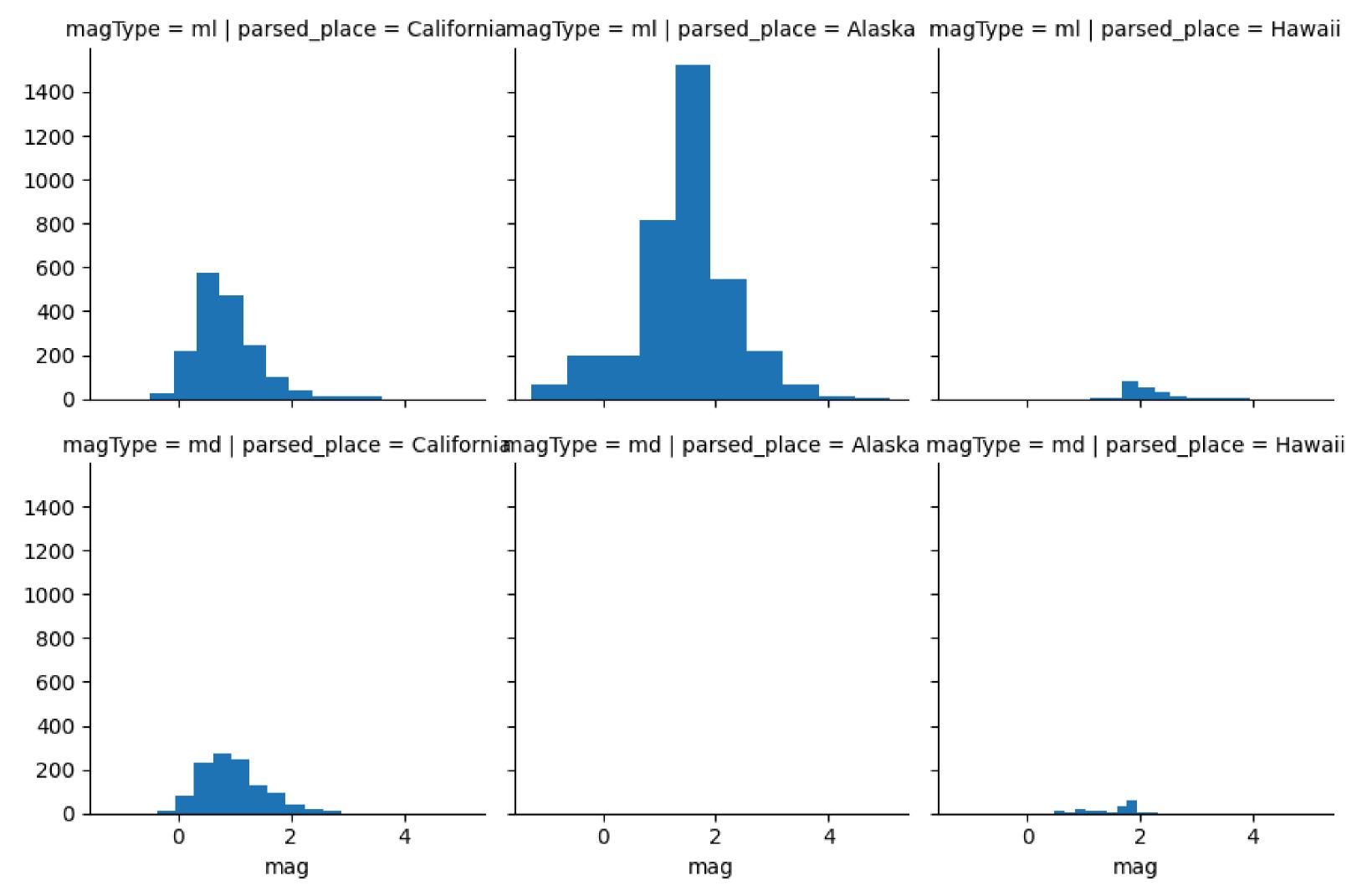
The `scale` parameter has been renamed and will be removed in v0.15.0. Pass `density\_norm='width'` for the same effect. sns.violinplot(

Text(0.5, 0.98, 'Comparing earthquake magnitude by magType')

## Comparing earthquake magnitude by magType



```
g = sns.FacetGrid(
quakes[
  (quakes.parsed_place.isin([
  'California', 'Alaska', 'Hawaii'
]))\
& (quakes.magType.isin(['ml', 'md']))
],
  row= 'magType',
  col='parsed_place'
)
g = g.map(plt.hist, 'mag')
```

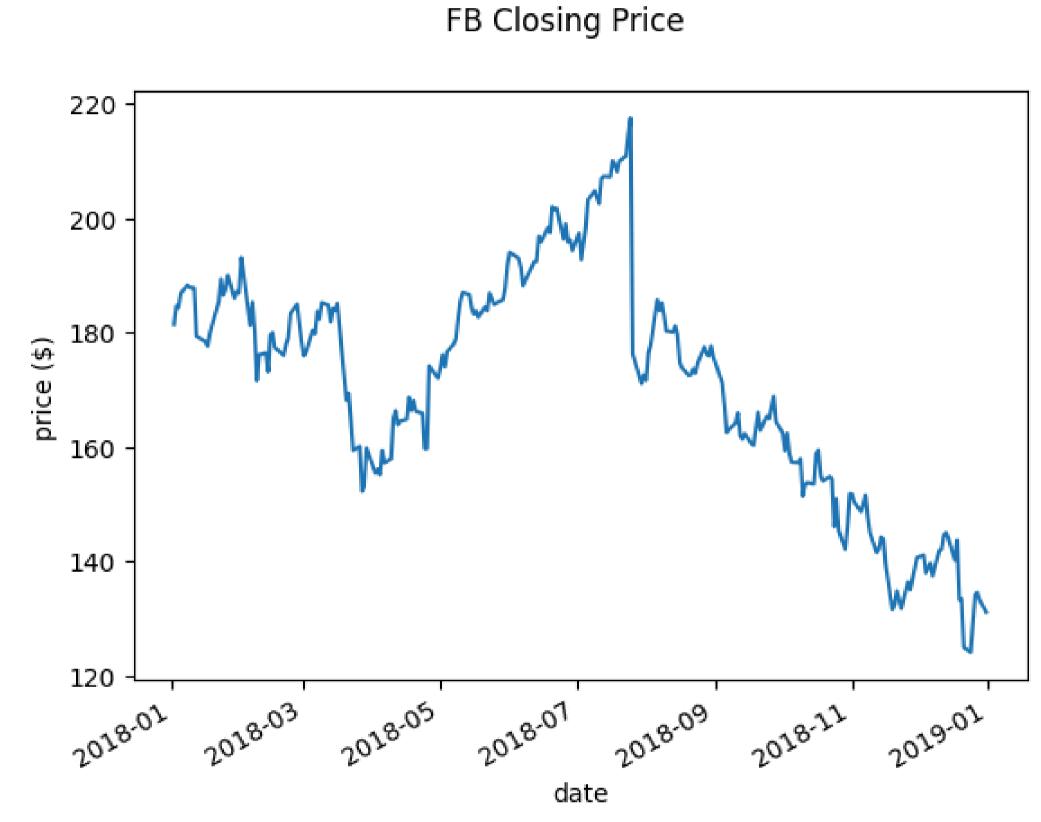


%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sns
fb = pd.read\_csv(
 '/content/fb\_stock\_prices\_2018.csv', index\_col='date', parse\_dates=True

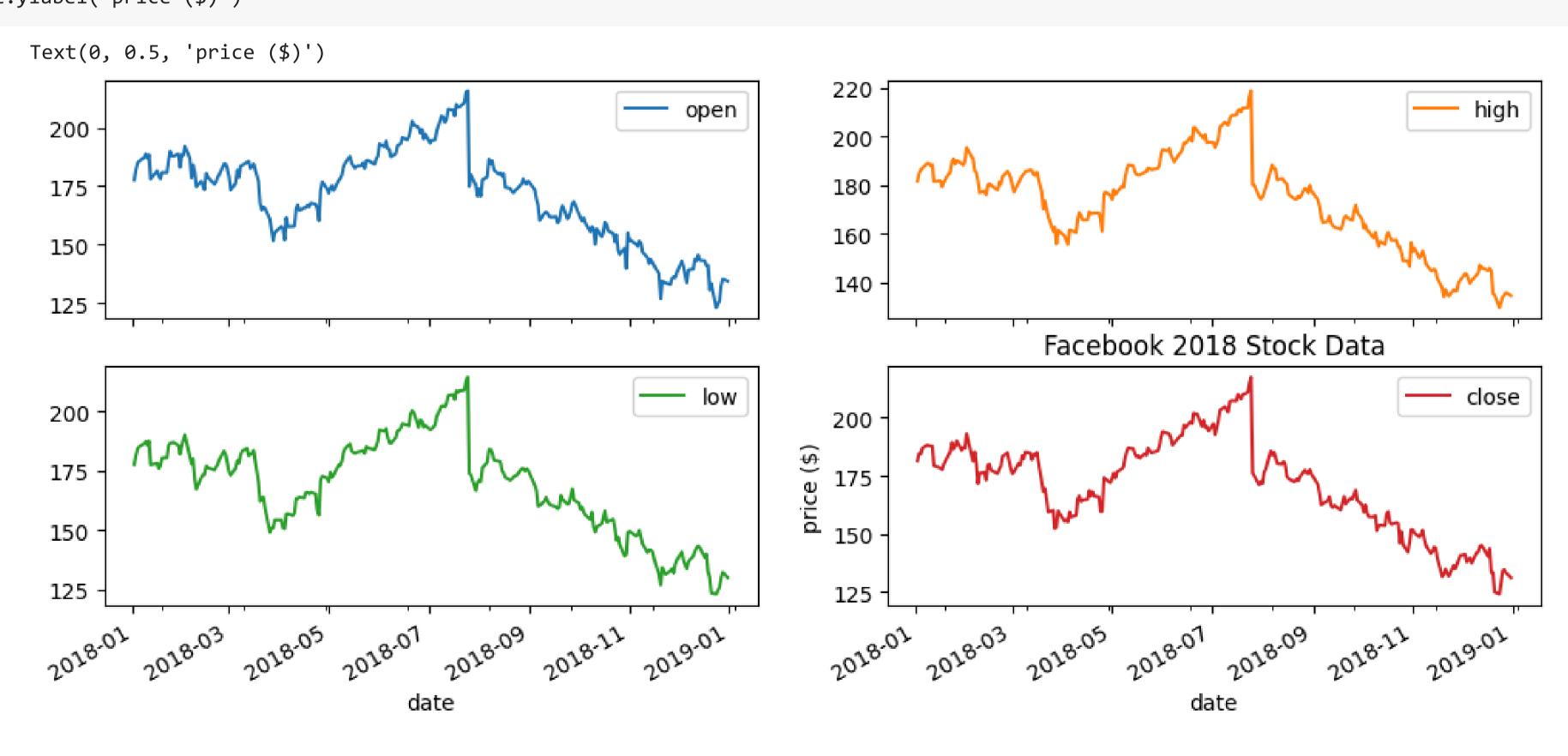
fb.close.plot()
plt.suptitle('FB Closing Price')
plt.xlabel('date')
plt.ylabel('price (\$)')

Text(0, 0.5, 'price (\$)')

#### ER Clasin



fb.iloc[:,:4].plot(subplots=True, layout=(2, 2), figsize=(12, 5))
plt.title('Facebook 2018 Stock Data')
plt.xlabel('date')
plt.ylabel('price (\$)')



fb.iloc[:,:4].plot(subplots=True, layout=(2, 2), figsize=(12, 5))
plt.suptitle('Facebook 2018 Stock Data')
plt.xlabel('date')
plt.ylabel('price (\$)')

```
Text(0, 0.5, 'price ($)')
                                               Facebook 2018 Stock Data
                                                                                                                  high
 200
                                                               200
 175
                                                               180
                                                               160
 150
                                                               140 -
 125
                                                                                                                 close
                                                    low
 200 -
                                                            buce
175 -
 175 -
 150
 125
                                                               125
```

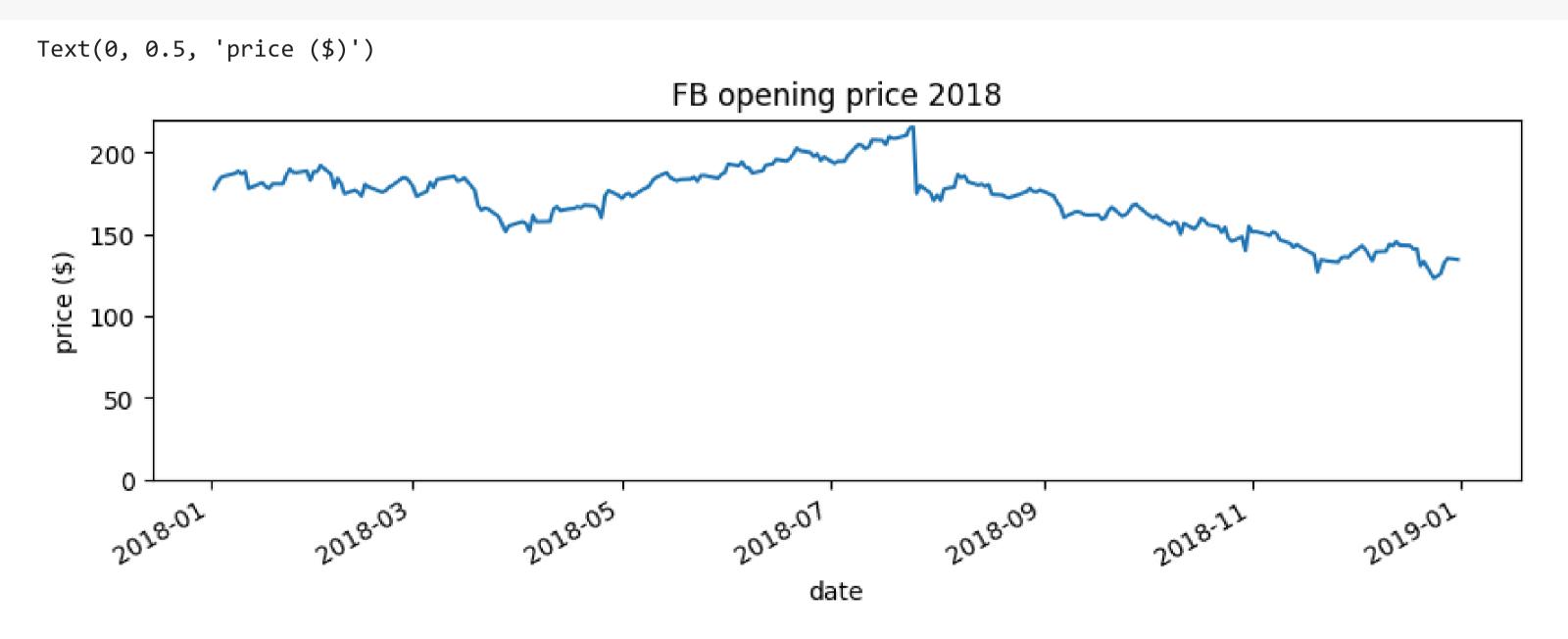
```
fb.assign(
ma=lambda x: x.close.rolling(20).mean()
).plot(
y=['close', 'ma'],
title='FB closing price in 2018',
label=['closing price', '20D moving average']
)
plt.legend(loc='lower left')
plt.ylabel('price ($)')
```

fb.open.plot(figsize=(10, 3), title='FB opening price 2018')
plt.ylim(0, None)
plt.ylabel('price (\$)')

import matplotlib.pyplot as plt

https://colab.research.google.com/drive/1Wrc2Y11FDXGVkyafnKy96mUDjzirl5NS#printMode=true

import calendar



import pandas as pd # Assuming you're using pandas for your data

# Plot the opening price with the initial setup
fb.open.plot(figsize=(10, 3), rot=0, title='FB opening price 2018')

# Get the current locations and labels (this will be modified)
locs, labels = plt.xticks()

month\_labels = [calendar.month\_name[i] for i in range(1, 13, 2)]

# Then, calculate new locations. Assume you want them evenly spaced, based on the plotted data range
new\_locs = np.linspace(start=locs[0], stop=locs[-1], num=len(month\_labels))

# Set the new locations and labels
plt.xticks(ticks=new\_locs, labels=month\_labels)

plt.ylabel('Price (\$)')
plt.show()

FB opening price 2018

200 - 180 - 140 - 120 - 1

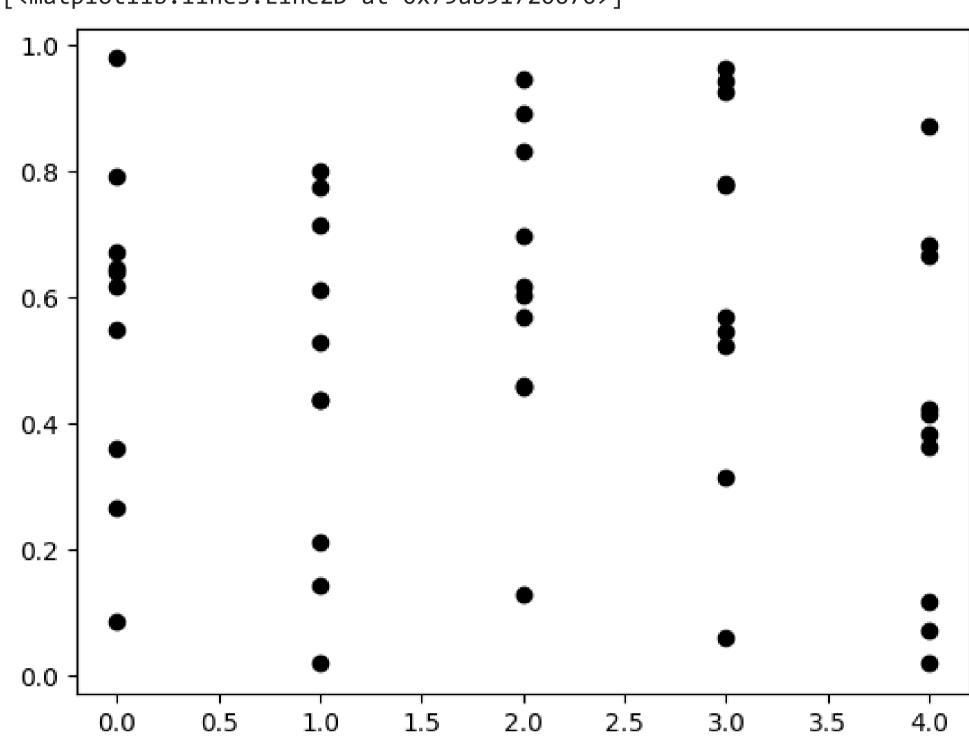
```
Hands-on Activity 9.2 Customized Visualizations using Seaborn - Colaboratory
import matplotlib.ticker as ticker
ax = fb.close.plot(
figsize=(10, 4),
title='Facebook Closing Price as Percentage of Highest Price in Time Range'
ax.yaxis.set_major_formatter(
ticker.PercentFormatter(xmax=fb.high.max())
ax.set_yticks([
fb.high.max()*pct for pct in np.linspace(0.6, 1, num=5)
]) # show round percentages only (60%, 80%, etc.)
ax.set_ylabel(f'percent of highest price (${fb.high.max()})')
```

Text(0, 0.5, 'percent of highest price (\$218.62)')

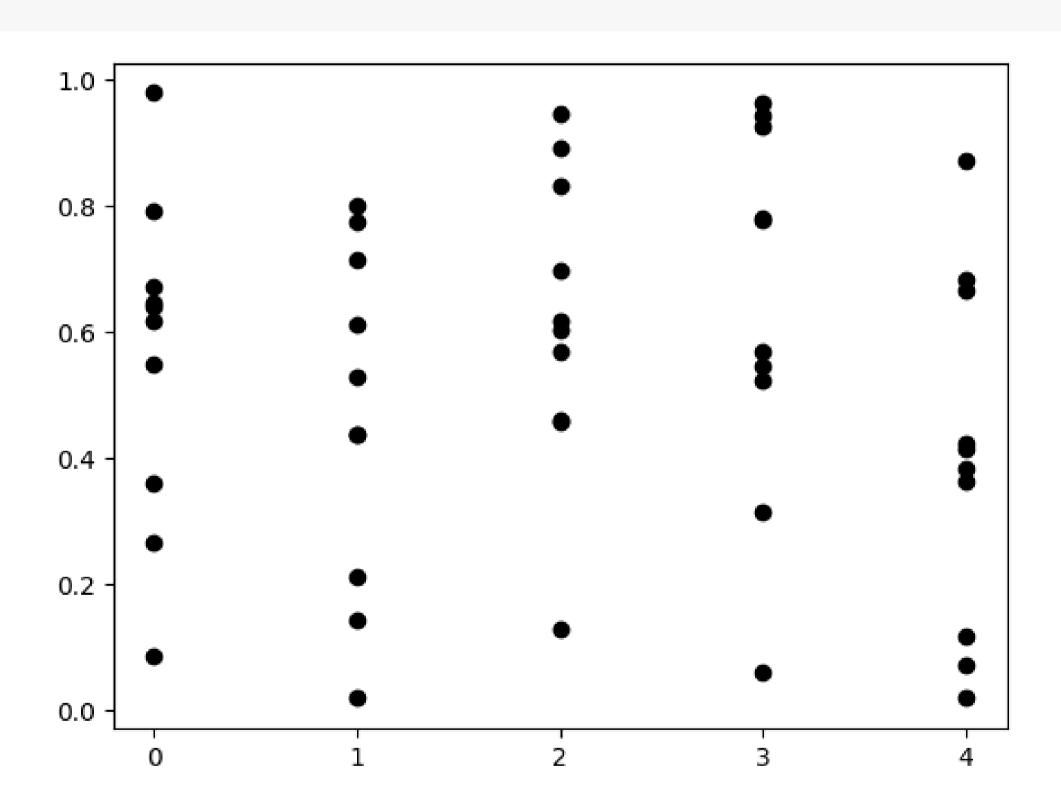


```
fig, ax = plt.subplots(1, 1)
np.random.seed(0)
ax.plot(np.tile(np.arange(0, 5), 10), np.random.rand(50), 'ko')
```

[<matplotlib.lines.Line2D at 0x79ab91720670>]



```
fig, ax = plt.subplots(1, 1)
np.random.seed(0)
ax.plot(np.tile(np.arange(0, 5), 10), np.random.rand(50), 'ko')
ax.get_xaxis().set_major_locator(
ticker.MultipleLocator(base=1)
```



```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
fb = pd.read_csv(
'/content/fb_stock_prices_2018.csv', index_col='date', parse_dates=True
```

from pandas.plotting import scatter\_matrix scatter\_matrix(fb, figsize=(10, 10))

```
<Axes: xlabel='high', ylabel='open'>,
      <Axes: xlabel='low', ylabel='open'>,
      <Axes: xlabel='close', ylabel='open'>,
      <Axes: xlabel='volume', ylabel='open'>],
      [<Axes: xlabel='open', ylabel='high'>,
      <Axes: xlabel='high', ylabel='high'>,
      <Axes: xlabel='low', ylabel='high'>,
      <Axes: xlabel='close', ylabel='high'>,
      <Axes: xlabel='volume', ylabel='high'>],
      [<Axes: xlabel='open', ylabel='low'>,
      <Axes: xlabel='high', ylabel='low'>,
      <Axes: xlabel='low', ylabel='low'>,
      <Axes: xlabel='close', ylabel='low'>,
      <Axes: xlabel='volume', ylabel='low'>],
      [<Axes: xlabel='open', ylabel='close'>,
      <Axes: xlabel='high', ylabel='close'>,
      <Axes: xlabel='low', ylabel='close'>,
      <Axes: xlabel='close', ylabel='close'>,
      <Axes: xlabel='volume', ylabel='close'>],
      [<Axes: xlabel='open', ylabel='volume'>,
      <Axes: xlabel='high', ylabel='volume'>,
      <Axes: xlabel='low', ylabel='volume'>,
      <Axes: xlabel='close', ylabel='volume'>,
      <Axes: xlabel='volume', ylabel='volume'>]], dtype=object)
  200
o 160
  140 -
  220 -
  200 -
high 180
  140
  200 -
  180
<u></u> 8 160
  140
  200
9 180
160
volume
1.0
                                                                         150
                                                                           close
```

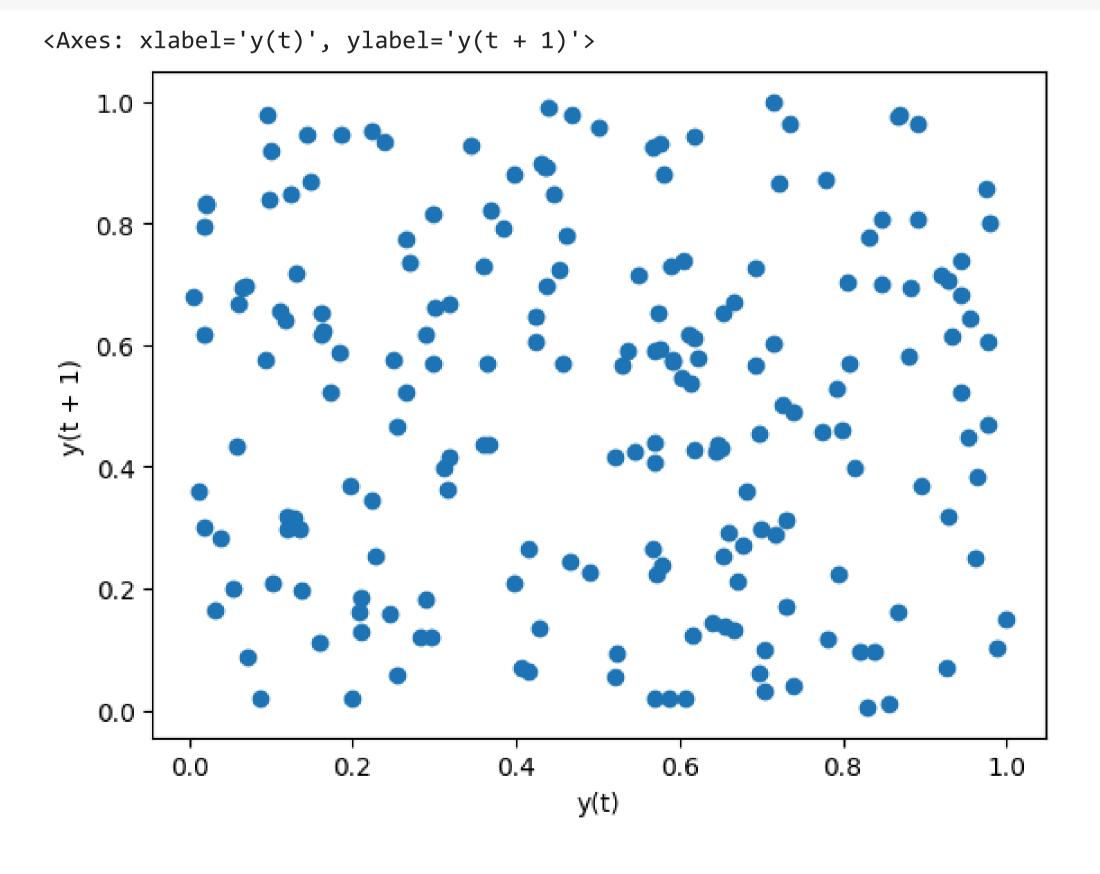
scatter\_matrix(fb, figsize=(10, 10), diagonal='kde')

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array([[<Axes: xlabel='open', ylabel='open'>,

```
<Axes: xlabel='low', ylabel='open'>,
      <Axes: xlabel='close', ylabel='open'>,
      <Axes: xlabel='volume', ylabel='open'>],
      [<Axes: xlabel='open', ylabel='high'>,
      <Axes: xlabel='high', ylabel='high'>,
      <Axes: xlabel='low', ylabel='high'>,
      <Axes: xlabel='close', ylabel='high'>,
      <Axes: xlabel='volume', ylabel='high'>],
      [<Axes: xlabel='open', ylabel='low'>,
      <Axes: xlabel='high', ylabel='low'>,
      <Axes: xlabel='low', ylabel='low'>,
      <Axes: xlabel='close', ylabel='low'>,
      <Axes: xlabel='volume', ylabel='low'>],
      [<Axes: xlabel='open', ylabel='close'>,
      <Axes: xlabel='high', ylabel='close'>,
      <Axes: xlabel='low', ylabel='close'>,
      <Axes: xlabel='close', ylabel='close'>,
      <Axes: xlabel='volume', ylabel='close'>],
      [<Axes: xlabel='open', ylabel='volume'>,
      <Axes: xlabel='high', ylabel='volume'>,
      <Axes: xlabel='low', ylabel='volume'>,
      <Axes: xlabel='close', ylabel='volume'>,
      <Axes: xlabel='volume', ylabel='volume'>]], dtype=object)
  200 -
e lao 1
  140 -
  220
  200 -
high 180
  140
  200
  180
  160
  200
95 180
160
volume
                                                                           close
```

from pandas.plotting import lag\_plot
np.random.seed(0) # make this repeatable
lag\_plot(pd.Series(np.random.random(size=200)))

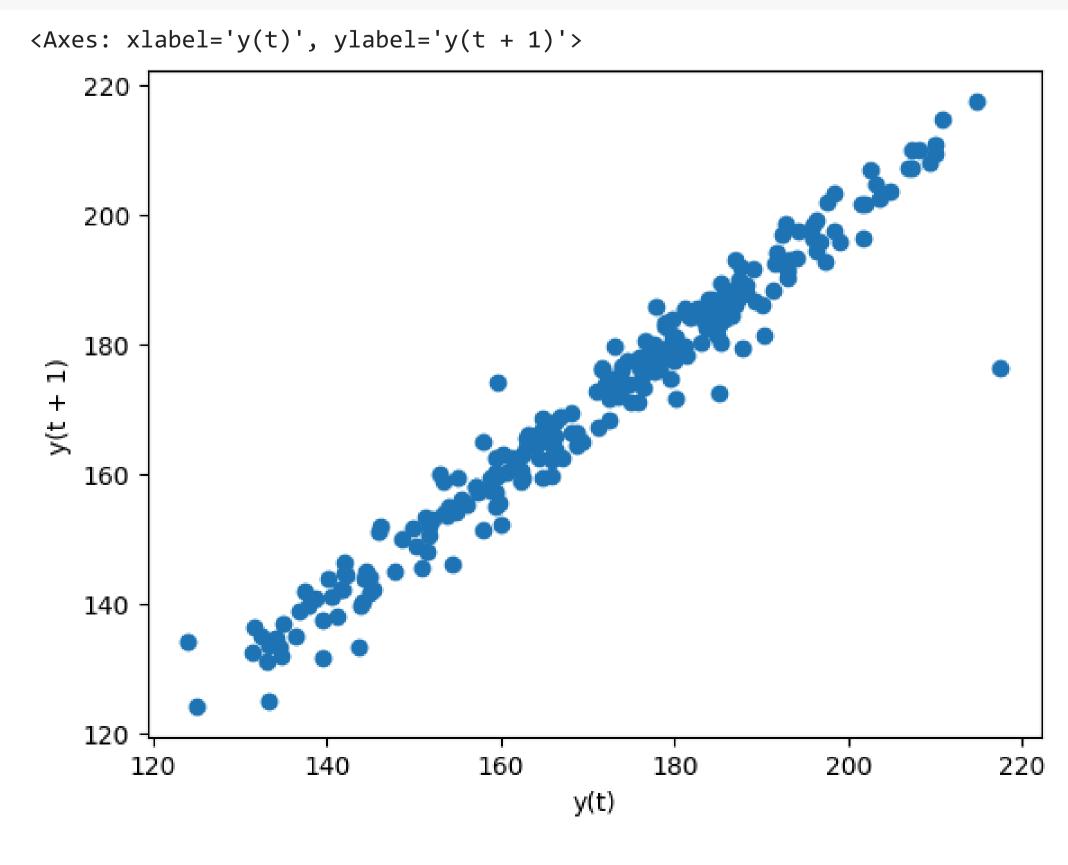


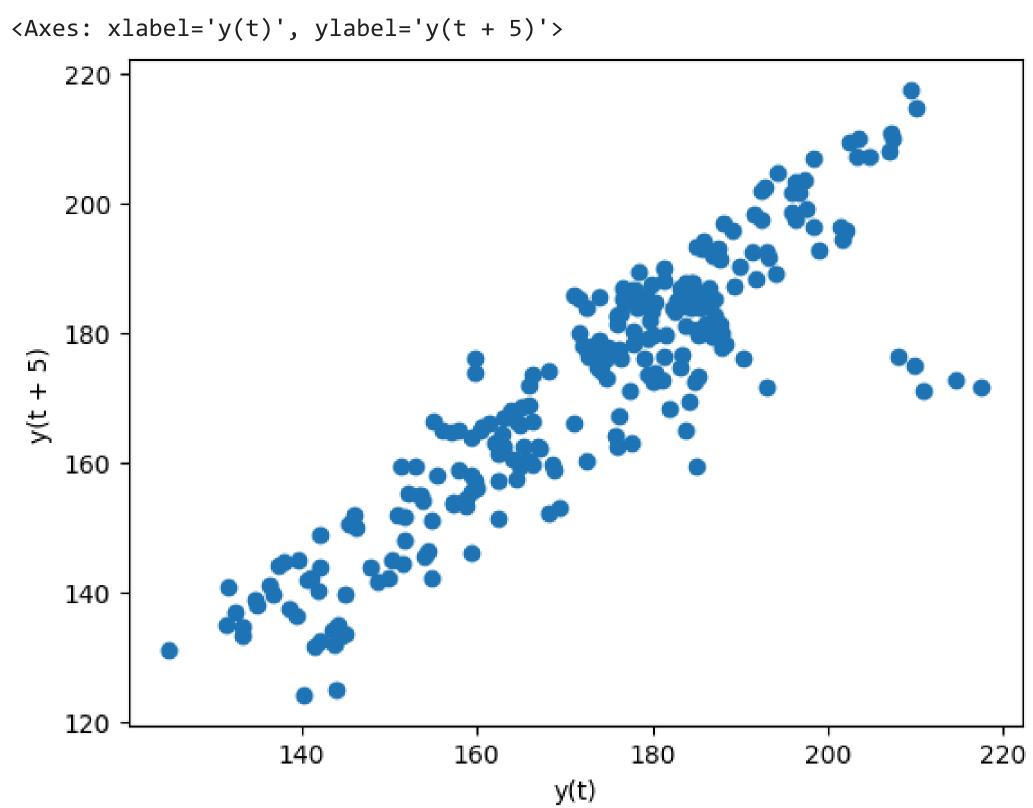
## lag\_plot(fb.close)

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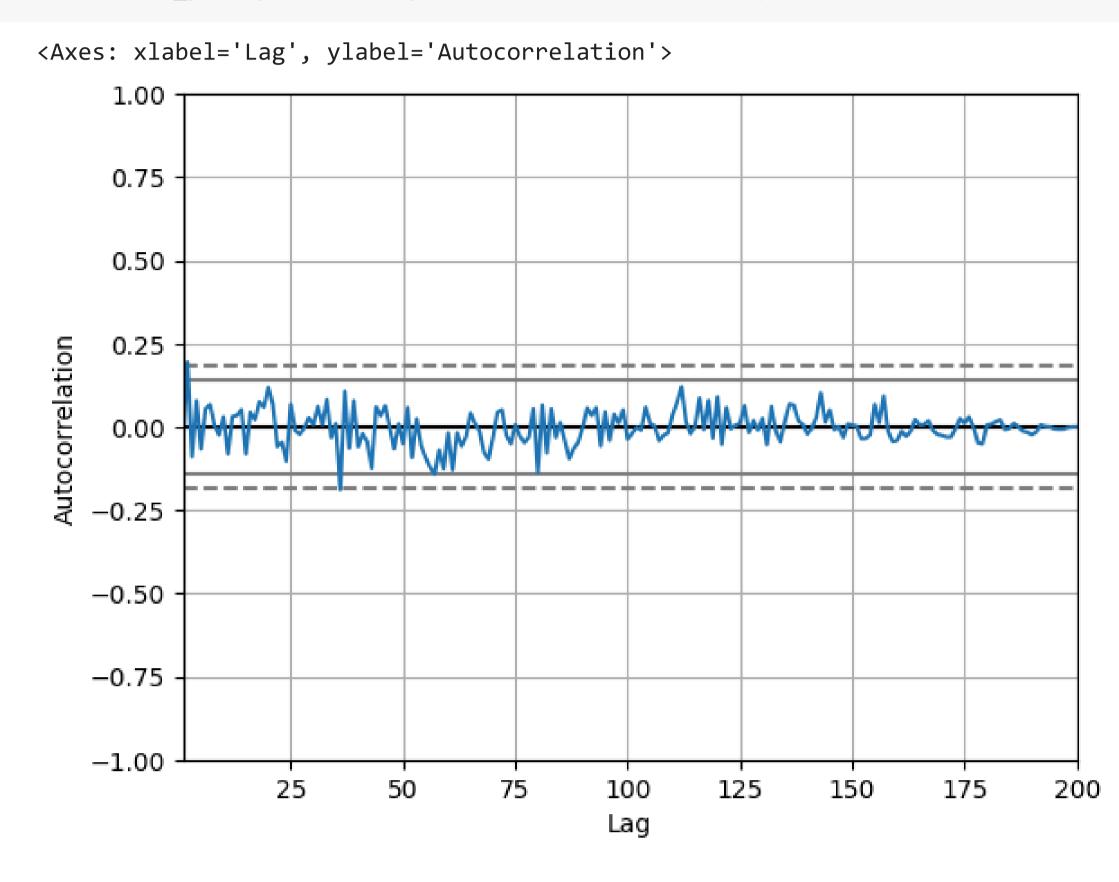
array([[<Axes: xlabel='open', ylabel='open'>,

<Axes: xlabel='high', ylabel='open'>,

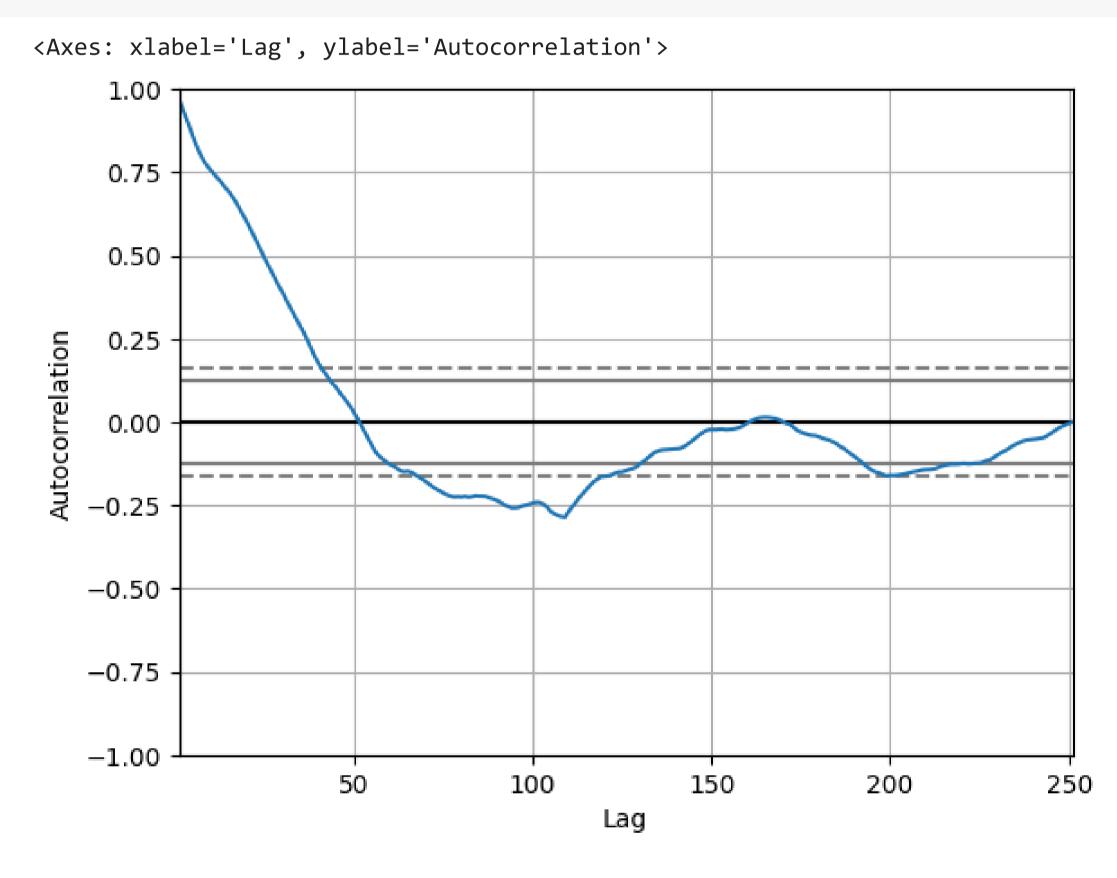




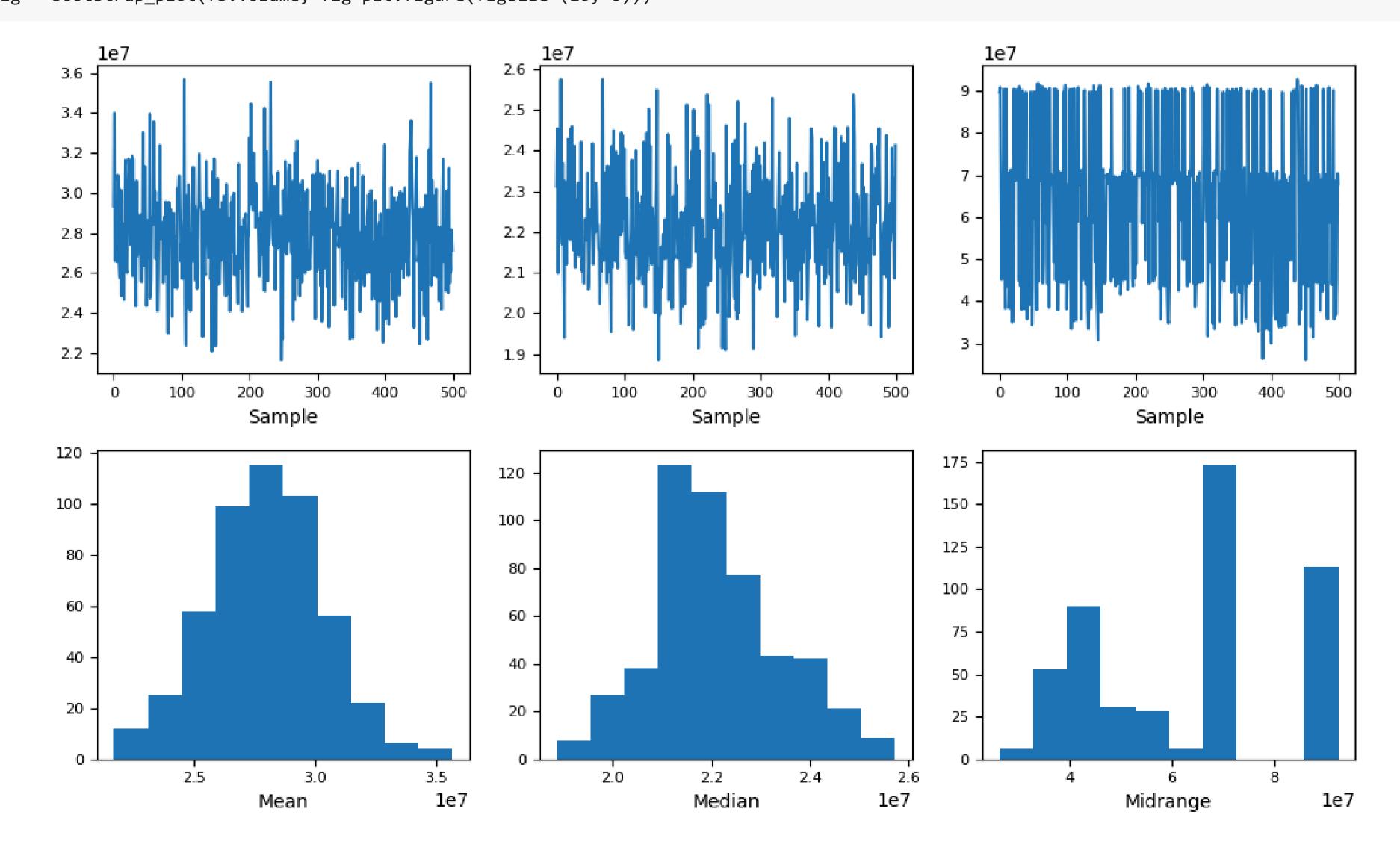
from pandas.plotting import autocorrelation\_plot
np.random.seed(0) # make this repeatable
autocorrelation\_plot(pd.Series(np.random.random(size=200)))



#### autocorrelation\_plot(fb.close)



# from pandas.plotting import bootstrap\_plot fig = bootstrap\_plot(fb.volume, fig=plt.figure(figsize=(10, 6)))



## Data Analysis:

The heatmap analysis revealing the correlation between earthquake magnitudes (with a specific focus on those categorized under magType 'mb') and the occurrences of tsunamis likely showcases a nuanced relationship between these natural phenomena. A demonstrated low to moderate positive correlation not only aligns with intuitive expectations—that larger earthquakes have a higher propensity to trigger tsunamis—but also highlights the complex interplay of factors beyond mere magnitude, such as geological conditions and proximity to water bodies, in tsunami generation. In parallel, a box plot detailing Facebook's traded volume and closing prices, supplemented by Tukey fences, provides a

granular look into the stock's trading behavior. This visualization method is particularly adept at spotlighting the central tendencies and variabilities within the data, with the identification of outliers potentially signaling periods of heightened volatility or market sensitivity to external events. Such insights are invaluable for investors and researchers alike, offering a clearer understanding of both natural disaster dynamics and financial market reactions.

## Supplementary Activity:

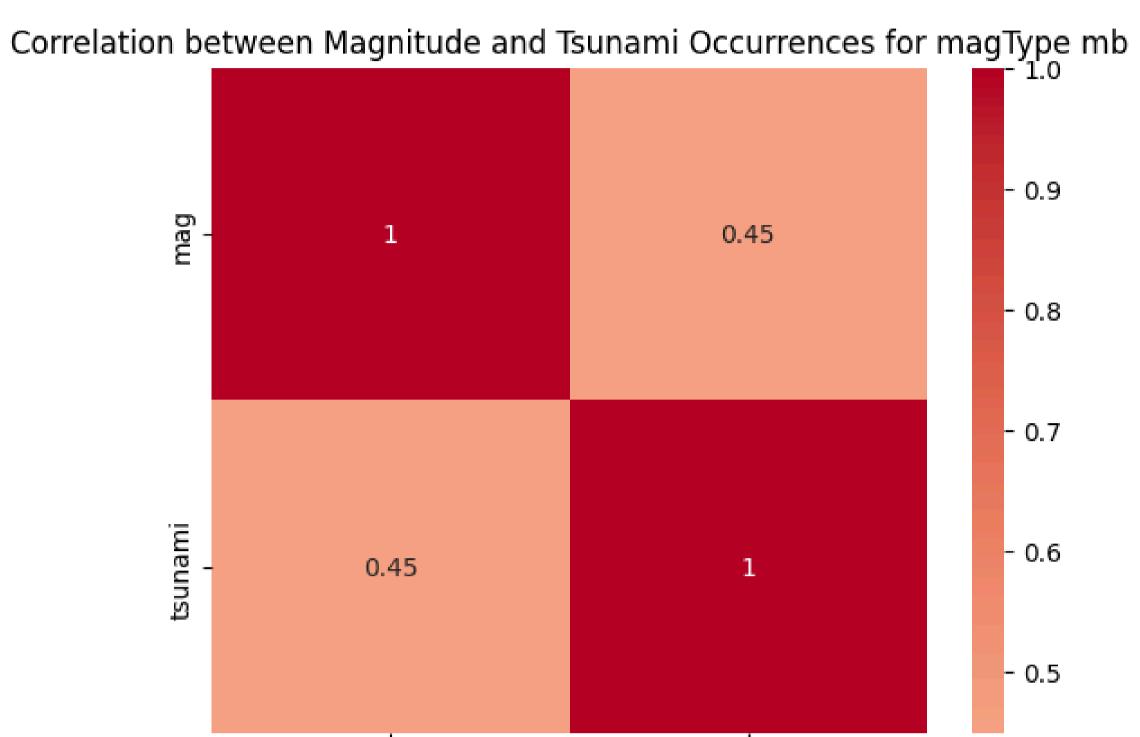
Using the CSV files provided and what we have learned so far in this module complete the following exercises:

```
import seaborn as sns
import matplotlib.pyplot as plt

# Filter earthquakes with magType 'mb'
mb_quakes = quakes[quakes['magType'] == 'mb']

# Calculate correlation matrix
corr_matrix = mb_quakes[['mag', 'tsunami']].corr()

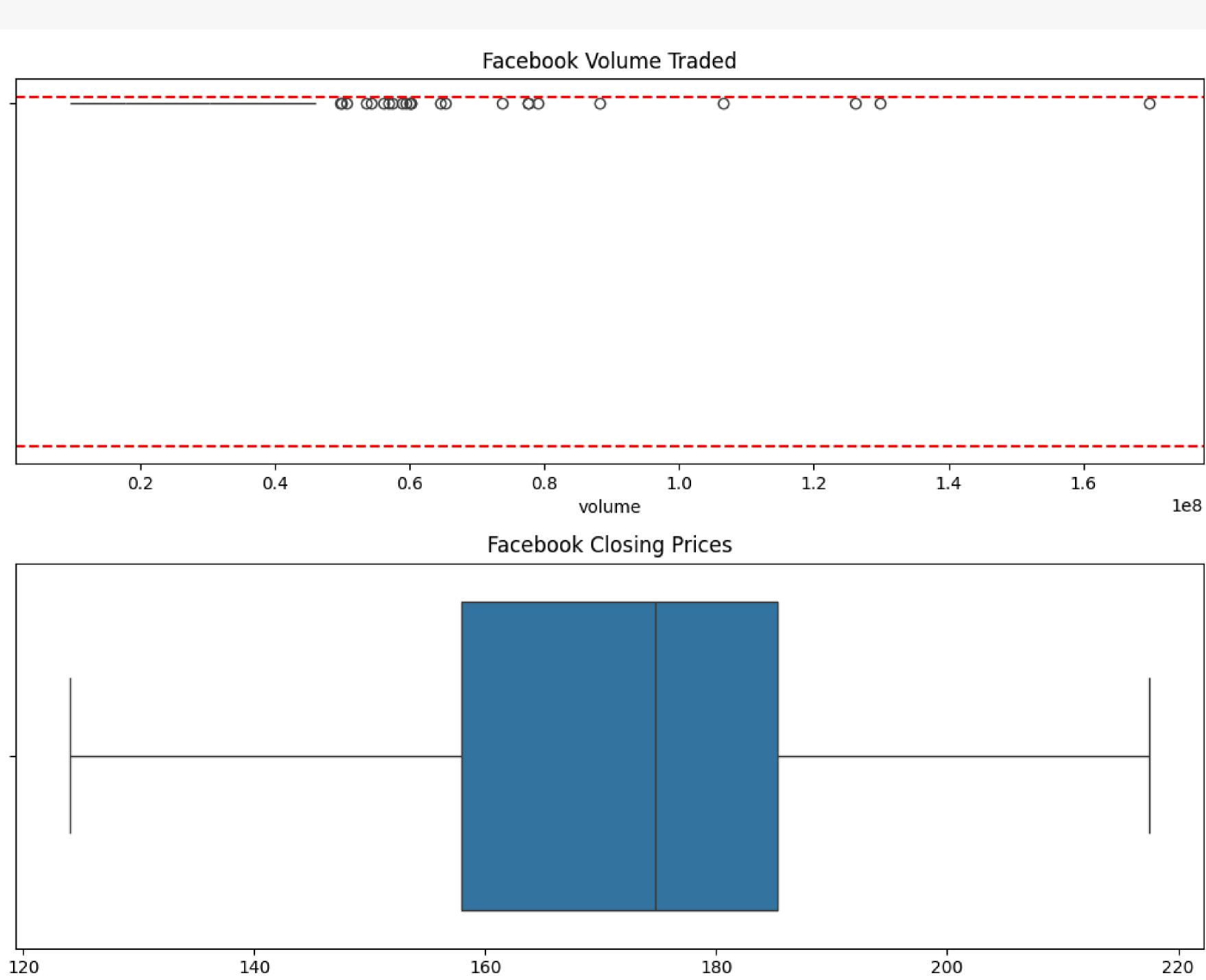
# Create heatmap
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', center=0)
plt.title('Correlation between Magnitude and Tsunami Occurrences for magType mb')
plt.show()
```



mag

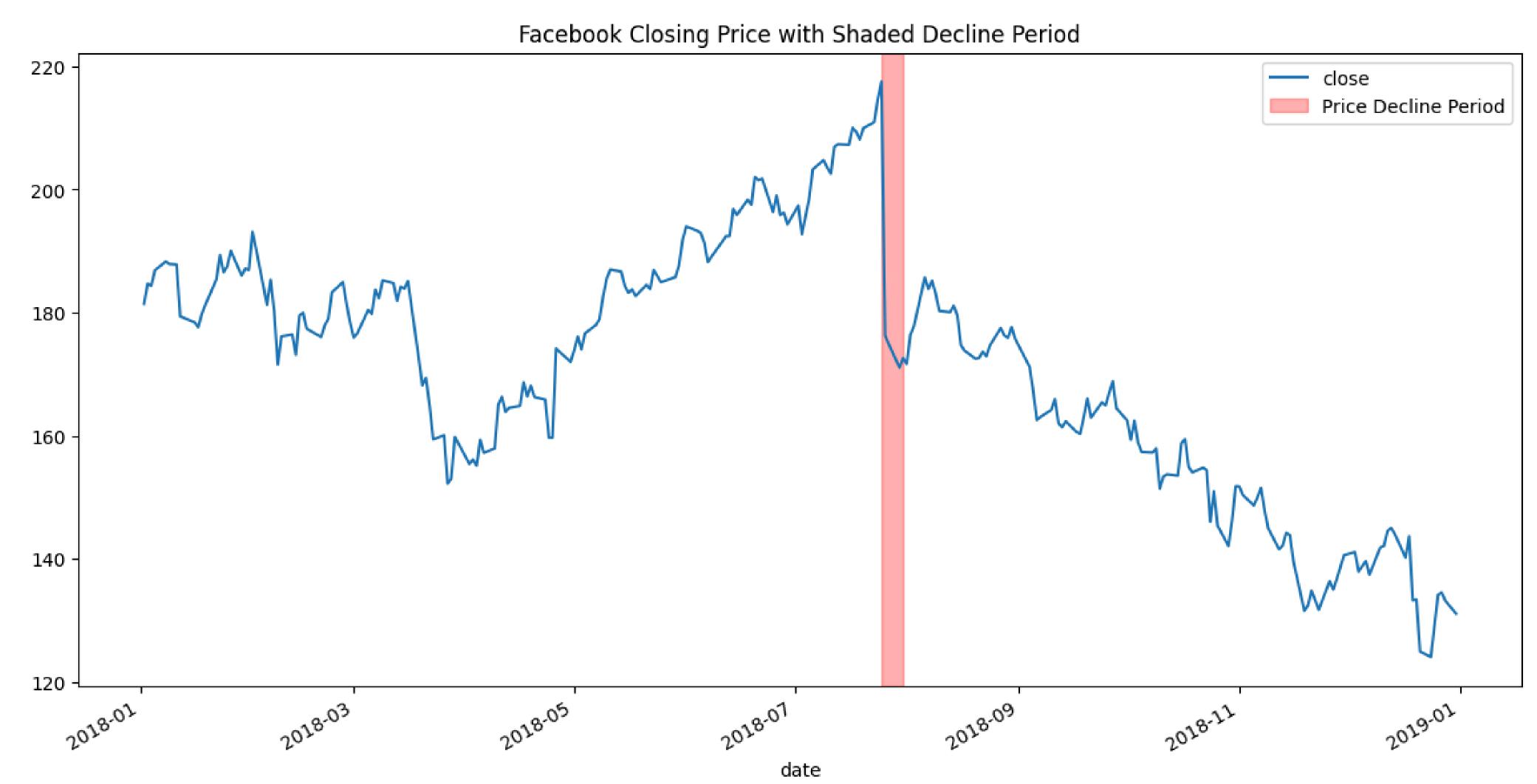
tsunami

```
fig, ax = plt.subplots(2, 1, figsize=(10, 8))
# Box plot for volume traded
sns.boxplot(x=fb['volume'], ax=ax[0])
ax[0].set_title('Facebook Volume Traded')
# Calculate IQR for volume traded
Q1_volume = fb['volume'].quantile(0.25)
Q3_volume = fb['volume'].quantile(0.75)
IQR_volume = Q3_volume - Q1_volume
lower_bound_volume = Q1_volume - 1.5 * IQR_volume
upper_bound_volume = Q3_volume + 1.5 * IQR_volume
# Drawing reference lines for volume
ax[0].axhline(y=lower_bound_volume, color='red', linestyle='--')
ax[0].axhline(y=upper_bound_volume, color='red', linestyle='--')
# Box plot for closing prices
sns.boxplot(x=fb['close'], ax=ax[1])
ax[1].set_title('Facebook Closing Prices')
# Similar calculations and reference lines for closing prices are omitted for brevity
plt.tight_layout()
plt.show()
```

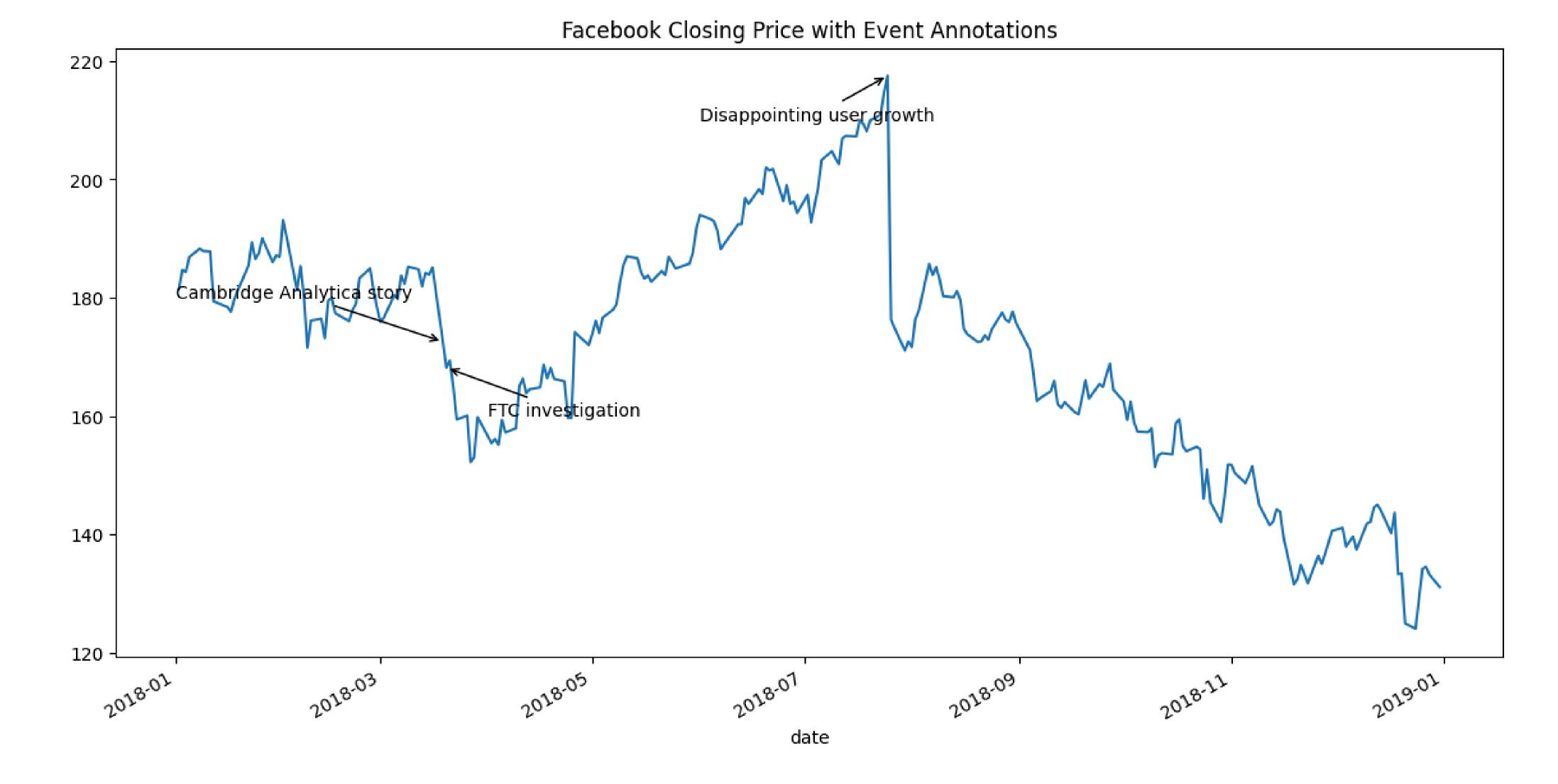


close

```
fb['close'].plot(figsize=(14, 7))
plt.axvspan('2018-07-25', '2018-07-31', color='red', alpha=0.3, label='Price Decline Period')
plt.legend()
plt.title('Facebook Closing Price with Shaded Decline Period')
plt.show()
```



```
fb['close'].plot(figsize=(14, 7))
plt.annotate('Disappointing user growth', xy=('2018-07-25', fb.loc['2018-07-25', 'close']), xytext=('2018-06', 210), arrowprops=dict(arrowstyle='->'))
plt.annotate('Cambridge Analytica story', xy=('2018-03-19', fb.loc['2018-03-19', 'close']), xytext=('2018-01', 180), arrowprops=dict(arrowstyle='->'))
plt.annotate('FTC investigation', xy=('2018-03-20', fb.loc['2018-03-20', 'close']), xytext=('2018-04', 160), arrowprops=dict(arrowstyle='->'))
plt.title('Facebook Closing Price with Event Annotations')
plt.show()
```



color - cman(i)

```
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
import pandas as pd
def reg_resid_plots(data, x_vars, y_var, colormap='tab10'):
   Plots regression and residual plots for pairs of columns in a DataFrame.
   Parameters:
    - data: DataFrame containing the data
    - x_vars: list of column names to be used as independent variables
    - y_var: column name to be used as dependent variable
    - colormap: name of the matplotlib colormap to use
    # Get the colormap
   cmap = plt.cm.get_cmap(colormap, len(x_vars))
   # Setup the matplotlib figure
   fig, axs = plt.subplots(len(x_vars), 2, figsize=(10, 5 * len(x_vars)))
   for i, x_var in enumerate(x_vars):
        # Set the color for current plot
```