Chapter12

August 4, 2023

```
[274]: import pandas as pd
import matplotlib.pyplot as plt
import matplotlib.ticker as ticker
import seaborn as sns
import os
```

0.0.1 Scatter plots

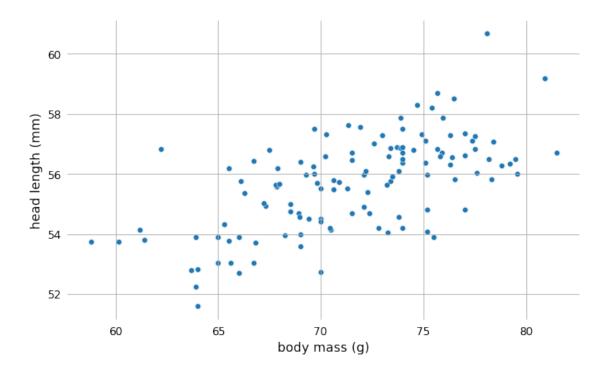
ax.grid()

```
[3]: blue_jays = pd.read_csv(os.path.join('data','blue_jays.csv'))

[4]: fig, ax = plt.subplots(1,1,figsize = (10,6))
    sns.scatterplot(
        data = blue_jays,
        x = 'Mass',
        y = 'Head'
        )
    ax.spines[:].set_visible(False)
    ax.xaxis.set_ticks_position('none')
    ax.yaxis.set_ticks_position("none")

ax.tick_params(axis='both', which='major', labelsize = 12)

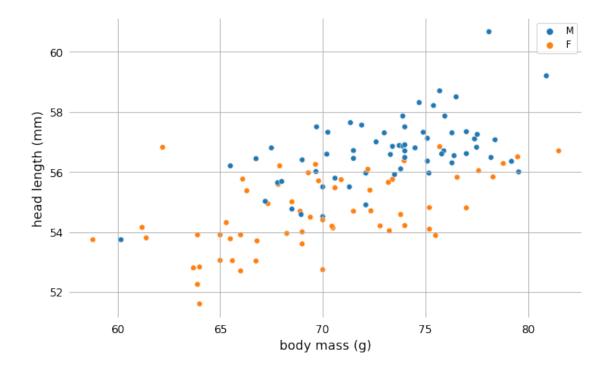
ax.set_xlabel('body mass (g)', fontsize=14)
    ax.set_ylabel('head length (mm)', fontsize=14)
```



```
[5]: fig, ax = plt.subplots(1,1,figsize = (10,6))
sns.scatterplot(
    data = blue_jays,
    x = 'Mass',
    y = 'Head',
    hue = "KnownSex"
    )
ax.spines[:].set_visible(False)
ax.xaxis.set_ticks_position('none')
ax.yaxis.set_ticks_position("none")

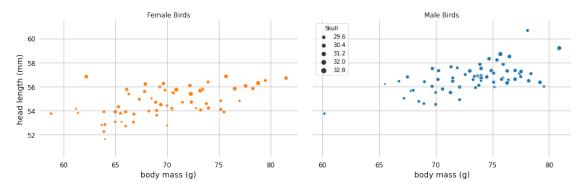
ax.set_xlabel('body mass (g)', fontsize=14)
ax.set_ylabel('head length (mm)', fontsize=14)

ax.tick_params(axis='both', which='major', labelsize = 12)
ax.legend(title='')
ax.grid()
```



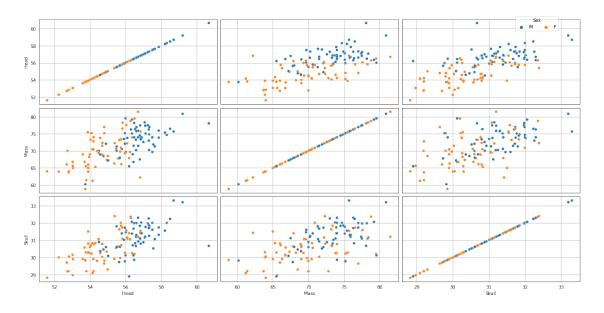
```
[8]: fig, (ax1, ax2) = plt.subplots(1,2,figsize = (14,4), sharey=True)
     plt.tight_layout()
     sns.scatterplot(
        data = blue_jays[blue_jays['KnownSex']=='F'],
         x = 'Mass',
         y = 'Head',
         color = '#ff7f0e',
         size = 'Skull',
         edgecolor = 'w',
         linewidth = 1,
         ax = ax1
         )
     sns.scatterplot(
         data = blue_jays[blue_jays['KnownSex']=='M'],
         x = 'Mass',
         y = 'Head',
         color = '#1f77b4',
         size = 'Skull',
         edgecolor = 'w',
         linewidth = 1,
         ax = ax2
         )
```

```
ax1.spines[:].set_visible(False)
ax1.xaxis.set_ticks_position('none')
ax1.yaxis.set_ticks_position("none")
ax1.set_xlabel('body mass (g)', fontsize=14)
ax1.set_ylabel('head length (mm)', fontsize=14)
ax1.set_ylim([50.1,61.5])
ax1.tick_params(axis='both', which='major', labelsize = 12)
ax1.get legend().remove()
ax1.set_title('Female Birds')
ax1.grid()
ax2.spines[:].set_visible(False)
ax2.xaxis.set_ticks_position('none')
ax2.yaxis.set_ticks_position("none")
ax2.set_ylim([50.1,61.5])
ax2.set_xlabel('body mass (g)', fontsize=14)
ax2.set_ylabel('head length (mm)', fontsize=14)
ax2.tick_params(axis='both', which='major', labelsize = 12)
ax2.set_title('Male Birds')
ax2.grid()
```



0.0.2 All-against-all Plot

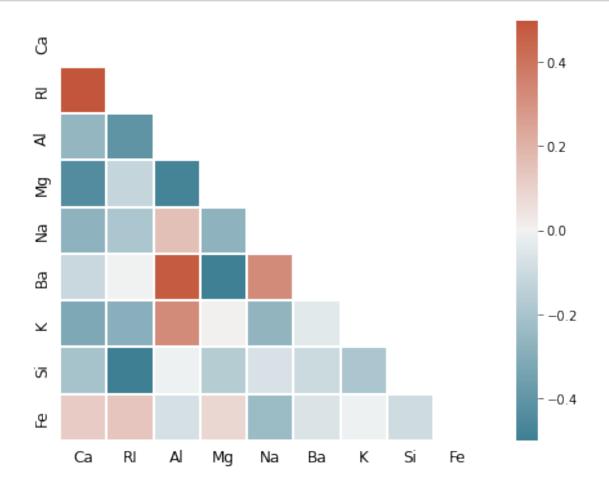
[66]: <matplotlib.legend.Legend at 0x7fa6cd05d160>



0.1 12.2 Correlograms

```
[174]: forensic_glass = pd.read_csv(os.path.join('data','forensic_glass.csv'))
corr_matrix = forensic_glass.iloc[:,:-1].corr()
```

```
[176]: cols = ['Ca', 'RI', 'Al', 'Mg', 'Na', 'Ba', 'K', 'Si', 'Fe']
       corr_matrix = corr_matrix[cols].reindex(cols)
[177]: fig, ax = plt.subplots(1,1,figsize=(10,6))
       mask = np.triu(np.ones_like(corr_matrix,dtype=bool))
       sns.heatmap(
           corr_matrix,
           mask = mask,
           # annot = True,
           linewidth = 2,
           square = True,
           center = 0,
           vmin = -.5,
           vmax = .5,
           cmap = sns.diverging_palette(220, 20, as_cmap=True)
       ax.xaxis.set_ticks_position('none')
       ax.yaxis.set_ticks_position('none')
       ax.tick_params(axis = 'both', which = 'major', labelsize = 12)
```



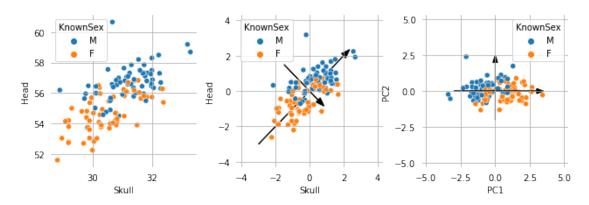
0.1.1 12.3 Dimension Reduction

```
[264]: from sklearn.decomposition import PCA
       from sklearn.preprocessing import StandardScaler
[335]: |blue_jays = pd.read_csv(os.path.join('data','blue_jays.csv'))
       fig, (ax1, ax2, ax3) = plt.subplots(1,3, figsize = (9,3))
       plt.tight_layout()
       sns.scatterplot(
           data = blue_jays,
           x = 'Skull',
           y = 'Head',
           hue = 'KnownSex',
           ax = ax1,
       ax1.spines[:].set_visible(False)
       ax1.xaxis.set_ticks_position('none')
       ax1.yaxis.set_ticks_position('none')
       ax1.xaxis.set_major_locator(ticker.MultipleLocator(2))
       ax1.yaxis.set_major_locator(ticker.MultipleLocator(2))
       ax1.grid()
       scaler = StandardScaler()
       blue_jays[['Head','Skull']] = scaler.fit_transform(blue_jays[['Head','Skull']])
       pca = PCA(n_components = 2).fit(blue_jays[['Head','Skull']])
       ax2.grid()
       comps = pca.components_
       exp_var = pca.explained_variance_
       comp0 = comps[0,:]*exp_var[0]
       comp1 = comps[1,:]*exp_var[1]
       x11 = -3
       y11 = comp0[1]/comp0[0]*x11
       dx12 = 5
       dy12 = comp0[1]/comp0[0]*x12
       x21 = -1.5
       y21 = comp1[1]/comp1[0]*x21
       dx22 = 2
       dy22 = comp1[1]/comp1[0]*x22
```

```
ax2.arrow(
   x11,
   y11,
   dx12,
    dy12,
   head_width = 0.35,
   head_length = 0.5,
    color = 'k'
)
ax2.arrow(
    x21,
    y21,
    dx22,
    dy22,
    head_width = 0.35,
    head_length = 0.5,
    color = 'k'
sns.scatterplot(
   data = blue_jays,
   x = 'Skull',
   y = 'Head',
   hue = 'KnownSex',
    ax = ax2,
)
ax2.spines[:].set_visible(False)
ax2.set_ylim([-4.3,4.3])
ax2.set_xlim([-4.3,4.3])
ax2.xaxis.set_major_locator(ticker.MultipleLocator(2))
ax2.yaxis.set_major_locator(ticker.MultipleLocator(2))
ax2.xaxis.set_ticks_position('none')
ax2.yaxis.set_ticks_position('none')
transf_features = pca.transform(blue_jays[['Head','Skull']])
blue_jays[['PC1','PC2']] = transf_features
sns.scatterplot(
   data = blue_jays,
    x = 'PC1',
    y = 'PC2',
    hue = 'KnownSex',
    ax = ax3
    )
```

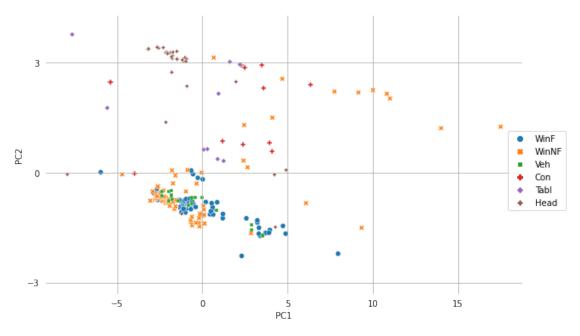
```
ax3.grid()
ax3.spines[:].set_visible(False)
ax3.set_ylim([-5.3,5.3])
ax3.set_xlim([-5.3,5.3])
ax3.xaxis.set_ticks_position('none')
ax3.yaxis.set_ticks_position('none')
ax3.xaxis.set_major_locator(ticker.MultipleLocator(2.5))
ax3.yaxis.set_major_locator(ticker.MultipleLocator(2.5))
ax3.arrow(-3.0,0.0,6.0,0.0,head_width = 0.35,head_length = 0.5, color = 'k')
ax3.arrow(0.0,-2.0,0.0,4.0,head_width = 0.35,head_length = 0.5, color = 'k')
```

[335]: <matplotlib.patches.FancyArrow at 0x7fa6bcdf75e0>



```
[350]: forensic_glass_data = forensic_glass.iloc[:,:-1]
       pca = PCA(n_components = 2).fit(forensic_glass_data)
       X = pca.transform(forensic_glass_data)
       transf_data = pd.DataFrame(X,columns = ['PC1','PC2'])
       transf_data['Type'] = forensic_glass['type']
[376]: fig, ax = plt.subplots(1,1,figsize = (10,6))
       sns.scatterplot(
           data = transf data,
           x = 'PC1',
           y = 'PC2',
           hue = 'Type',
           style = 'Type',
           # size = 20,
           ax = ax
       ax.spines[:].set_visible(False)
       ax.set_ylim([-3.3, 4.3])
       ax.grid()
       ax.yaxis.set_ticks_position('none')
```

```
ax.xaxis.set_ticks_position('none')
ax.legend(bbox_to_anchor=(1.1,0.6))
ax.yaxis.set_major_locator(ticker.MultipleLocator(3))
```

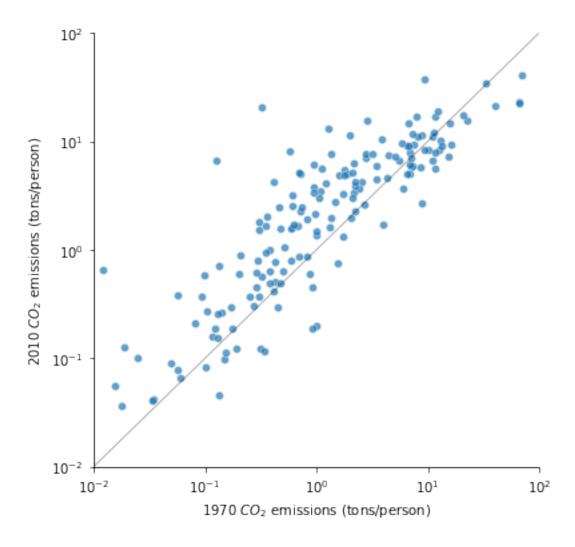


0.1.2 12.4 Paired Data

```
[380]:
     CO2_emissions = pd.read_csv(os.path.join('data','CO2_emissions.csv'))
[386]: CO2_emissions_2010= CO2_emissions[CO2_emissions['year']==2010]
      CO2 emissions 1970= CO2 emissions[CO2 emissions['year']==1970]
      CO2_emissions_merge = pd.merge(CO2_emissions_2010, CO2_emissions_1970, right_on_
       [422]: fig, ax = plt.subplots(1,1,figsize = (6,6))
      sns.scatterplot(
          data = CO2_emissions_merge,
          x = 'emissions_y',
          y = 'emissions_x',
          alpha = 0.7
      ax.set_ylim([0.0099,101])
      ax.set_xlim([0.0099,101])
      plt.xscale('log')
      plt.yscale('log')
      ax.set_xlabel('1970 $CO_2$ emissions (tons/person)')
      ax.set_ylabel('2010 $CO_2$ emissions (tons/person)')
```

```
ax.minorticks_off()
ax.spines["top"].set_visible(False)
ax.spines['right'].set_visible(False)
ax.plot([0,100],[0,100],color='k',alpha = 0.6, linewidth = .5)
```

[422]: [<matplotlib.lines.Line2D at 0x7fa6c3a61d90>]



```
[435]: import matplotlib.lines as mlines

# Import Data

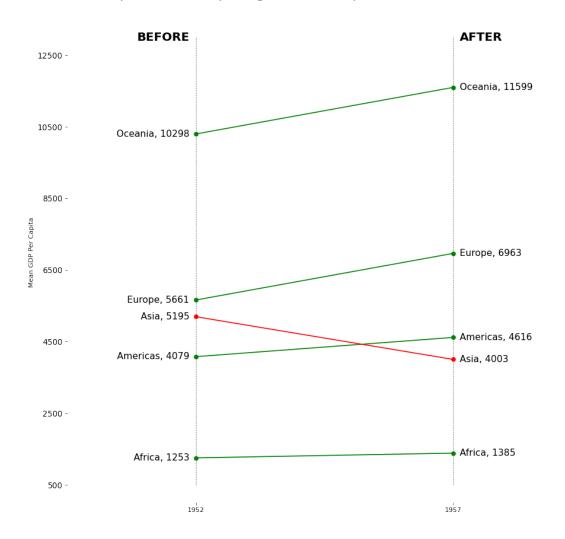
df = pd.read_csv("https://raw.githubusercontent.com/selva86/datasets/master/

Gdppercap.csv")
```

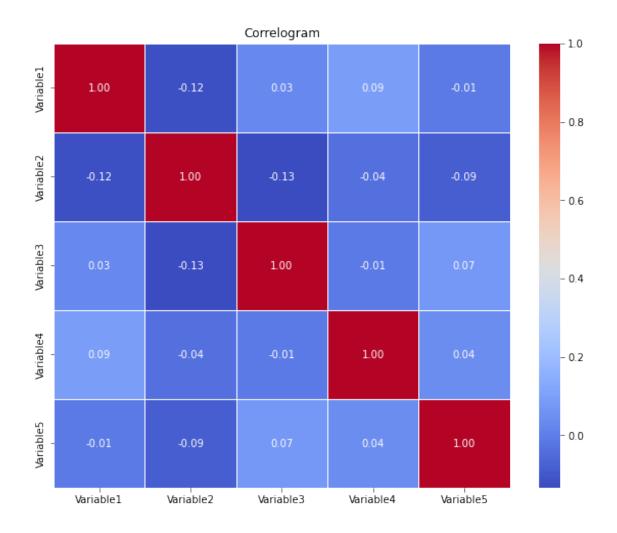
```
left_label = [str(c) + ', '+ str(round(y)) for c, y in zip(df.continent,__
 ⇔df['1952'])]
right_label = [str(c) + ', '+ str(round(y)) for c, y in zip(df.continent,_
⇔df['1957'])]
klass = ['red' if (y1-y2) < 0 else 'green' for y1, y2 in zip(df['1952'], _{\sqcup}

df['1957'])]
# draw line
# https://stackoverflow.com/questions/36470343/
→how-to-draw-a-line-with-matplotlib/36479941
def newline(p1, p2, color='black'):
    ax = plt.gca()
    1 = mlines.Line2D([p1[0],p2[0]], [p1[1],p2[1]], color='red' if p1[1]-p2[1]
 →> 0 else 'green', marker='o', markersize=6)
    ax.add_line(1)
    return 1
fig, ax = plt.subplots(1,1,figsize=(14,14), dpi= 80)
# Vertical Lines
ax.vlines(x=1, ymin=500, ymax=13000, color='black', alpha=0.7, linewidth=1,__
 ⇔linestyles='dotted')
ax.vlines(x=3, ymin=500, ymax=13000, color='black', alpha=0.7, linewidth=1,__
 ⇔linestyles='dotted')
# Points
ax.scatter(y=df['1952'], x=np.repeat(1, df.shape[0]), s=10, color='black', ___
 \rightarrowalpha=0.7)
ax.scatter(y=df['1957'], x=np.repeat(3, df.shape[0]), s=10, color='black',
 ⇒alpha=0.7)
# Line Segments and Annotation
for p1, p2, c in zip(df['1952'], df['1957'], df['continent']):
    newline([1,p1], [3,p2])
    ax.text(1-0.05, p1, c + ', ' + str(round(p1)), horizontalalignment='right',
 ⇔verticalalignment='center', fontdict={'size':14})
    ax.text(3+0.05, p2, c + ', ' + str(round(p2)), horizontalalignment='left', u
 ⇔verticalalignment='center', fontdict={'size':14})
# 'Before' and 'After' Annotations
ax.text(1-0.05, 13000, 'BEFORE', horizontalalignment='right', L
 ⇔verticalalignment='center', fontdict={'size':18, 'weight':700})
ax.text(3+0.05, 13000, 'AFTER', horizontalalignment='left',
 overticalalignment='center', fontdict={'size':18, 'weight':700})
# Decoration
```

Slopechart: Comparing GDP Per Capita between 1952 vs 1957



```
[]:
 []:
 []:
 []:
[72]: import numpy as np
      import pandas as pd
      import seaborn as sns
      import matplotlib.pyplot as plt
      # Sample data (Replace this with your own dataset)
      data = pd.DataFrame({
          'Variable1': np.random.rand(100),
          'Variable2': np.random.rand(100),
          'Variable3': np.random.rand(100),
          'Variable4': np.random.rand(100),
          'Variable5': np.random.rand(100)
      })
      # Calculate the correlation matrix
      correlation_matrix = data.corr()
      # Create a heatmap (correlogram) using seaborn
      plt.figure(figsize=(10, 8))
      sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f',__
       →linewidths=0.5)
      plt.title('Correlogram')
      plt.show()
```



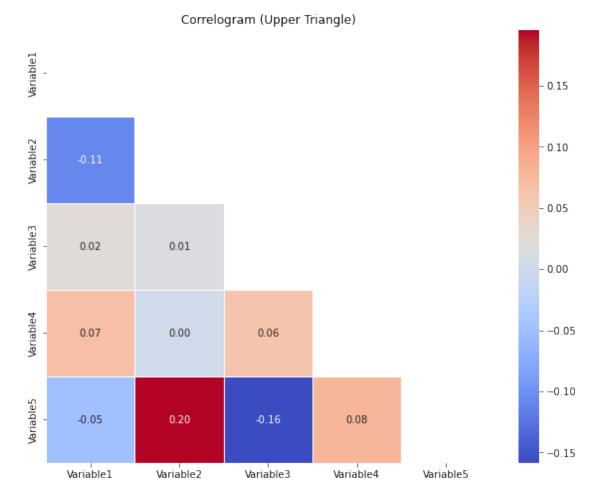
```
[73]: import numpy as np
  import pandas as pd
  import seaborn as sns
  import matplotlib.pyplot as plt

# Sample data (Replace this with your own dataset)
data = pd.DataFrame({
    'Variable1': np.random.rand(100),
    'Variable2': np.random.rand(100),
    'Variable3': np.random.rand(100),
    'Variable4': np.random.rand(100),
    'Variable5': np.random.rand(100)
})

# Calculate the correlation matrix
correlation_matrix = data.corr()
```

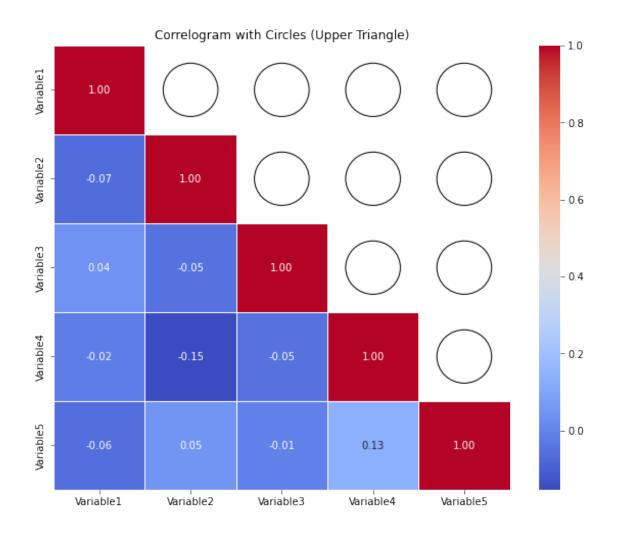
```
# Create a mask for the lower triangle
mask = np.triu(np.ones_like(correlation_matrix, dtype=bool))

# Create a heatmap (correlogram) using seaborn, masking the lower triangle
plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f', usinewidths=0.5, mask=mask)
plt.title('Correlogram (Upper Triangle)')
plt.show()
```



```
[75]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib.patches import Circle
```

```
# Sample data (Replace this with your own dataset)
data = pd.DataFrame({
    'Variable1': np.random.rand(100),
    'Variable2': np.random.rand(100),
    'Variable3': np.random.rand(100),
    'Variable4': np.random.rand(100),
    'Variable5': np.random.rand(100)
})
# Calculate the correlation matrix
correlation_matrix = data.corr()
# Create a circular mask for the upper triangle
mask = np.zeros_like(correlation_matrix, dtype=bool)
mask[np.triu_indices_from(mask, k=1)] = True
# Create a heatmap (correlogram) using seaborn, masking the upper triangle with
\hookrightarrow circles
plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f',__
 ⇒linewidths=0.5, mask=mask)
# Add circular cutouts to the heatmap
for i in range(correlation_matrix.shape[0]):
    for j in range(i + 1, correlation_matrix.shape[1]):
        if mask[i, j]:
            circle = Circle((j + 0.5, i + 0.5), 0.3, facecolor='white',
⇔edgecolor='black')
            plt.gca().add_patch(circle)
plt.title('Correlogram with Circles (Upper Triangle)')
plt.show()
```



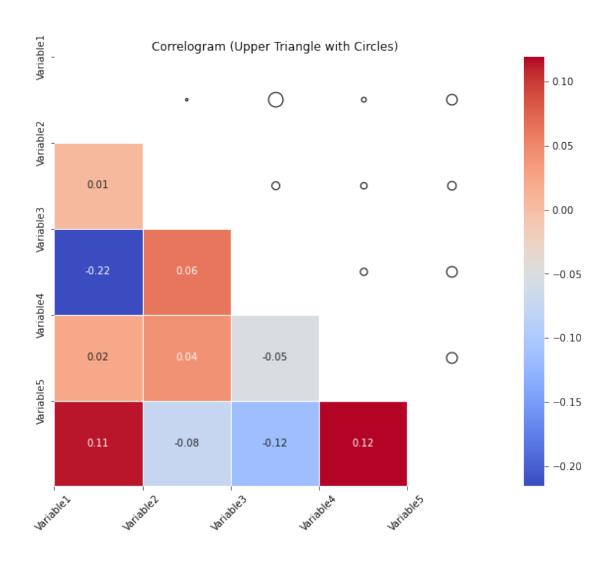
```
[78]: import numpy as np
  import pandas as pd
  import seaborn as sns
  import matplotlib.pyplot as plt

# Sample data (Replace this with your own dataset)
data = pd.DataFrame({
    'Variable1': np.random.rand(100),
    'Variable2': np.random.rand(100),
    'Variable3': np.random.rand(100),
    'Variable4': np.random.rand(100),
    'Variable5': np.random.rand(100)
})

# Calculate the correlation matrix
correlation_matrix = data.corr()
```

```
# Create a mask for the lower triangle
mask = np.triu(np.ones_like(correlation_matrix, dtype=bool))
# Create a heatmap using seaborn
plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f',__
 →linewidths=0.5, mask=mask)
# Plot circles over the heatmap
for i in range(len(correlation_matrix)):
   for j in range(i+1, len(correlation_matrix)):
       plt.scatter(j + 0.5, i + 0.5, s=1000 * abs(correlation_matrix.iloc[i,__

→j]), color='white', edgecolor='black')
plt.title('Correlogram (Upper Triangle with Circles)')
plt.xticks(range(len(correlation_matrix.columns)), correlation_matrix.columns,_
 →rotation=45)
plt.yticks(range(len(correlation_matrix.columns)), correlation_matrix.columns)
plt.show()
```



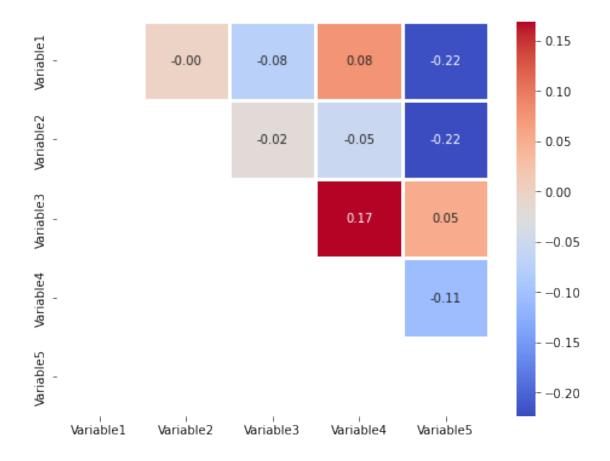
```
[90]: import numpy as np
  import pandas as pd
  import seaborn as sns
  import matplotlib.pyplot as plt

# Sample data (Replace this with your own dataset)
data = pd.DataFrame({
    'Variable1': np.random.rand(100),
    'Variable2': np.random.rand(100),
    'Variable3': np.random.rand(100),
    'Variable4': np.random.rand(100),
    'Variable5': np.random.rand(100)
})

# Calculate the correlation matrix
```

```
correlation_matrix = data.corr()
# Create a mask for the upper triangle
mask = np.tril(np.ones_like(correlation_matrix, dtype=bool))
# Create a heatmap using seaborn
plt.figure(figsize=(8, 6))
sns.heatmap(correlation_matrix, annot=True, square=False, cmap='coolwarm',_
 # Plot circles over the heatmap
# for i in range(len(correlation_matrix)):
    for j in range(i + 1, len(correlation_matrix)):
         plt.scatter(j + 0.5, i + 0.5, s=1000 * abs(correlation_matrix.iloc[i, local))
⇔j]), color='white', edgecolor='black')
# plt.title('Correlogram (Lower Triangle with Circles)')
# plt.xticks(range(len(correlation_matrix.columns)), correlation_matrix.
 ⇔columns, rotation=45)
# plt.yticks(range(len(correlation_matrix.columns)), correlation_matrix.columns)
# plt.show()
```

[90]: <AxesSubplot:>



```
[81]:
     correlation_matrix
[81]:
                Variable1 Variable2 Variable3
                                                 Variable4
                                                            Variable5
     Variable1
                 1.000000
                            0.133832
                                       0.022393
                                                 -0.063948
                                                            -0.014000
     Variable2
                 0.133832
                            1.000000
                                      -0.008763
                                                 -0.025814
                                                             0.044087
     Variable3
                 0.022393
                           -0.008763
                                       1.000000
                                                  0.201253
                                                            -0.108385
     Variable4 -0.063948
                           -0.025814
                                       0.201253
                                                  1.000000
                                                            -0.060605
     Variable5 -0.014000
                            0.044087 -0.108385
                                                 -0.060605
                                                             1.000000
[84]: np.tril(np.ones_like(correlation_matrix,dtype=bool))
[84]: array([[ True, False, False, False, False],
             [ True, True, False, False, False],
             [ True, True, True, False, False],
             [ True, True, True, False],
             [ True,
                     True, True, True, True]])
 []:
[74]: mask
```

```
[74]: array([[ True, True, True,
                                           True],
                                    True,
             [False, True, True,
                                    True,
                                          True],
             [False, False, True,
                                   True,
                                          True],
             [False, False, False, True,
                                           True],
             [False, False, False, False,
                                          Truell)
[67]: test = blue_jays[['Head','Mass','Skull']]
[69]: test.corr()
[69]:
                Head
                           Mass
                                    Skull
             1.000000 0.629445
                                0.667793
      Head
      Mass
             0.629445
                      1.000000
                                 0.552863
      Skull 0.667793 0.552863
                                1.000000
 []:
[59]: for ax in axes.axes.flatten():
         print(ax.xaxis.get_label())
     Text(0.5, 435.893333333333, 'Head')
     Text(0.5, 435.893333333333, 'Mass')
     Text(0.5, 435.893333333333, 'Skull')
     Text(0.5, 235.0466666666662, 'Head')
     Text(0.5, 235.0466666666662, 'Mass')
     Text(0.5, 235.0466666666662, 'Skull')
     Text(0.5, 17.20000000000003, 'Head')
     Text(0.5, 17.20000000000003, 'Mass')
     Text(0.5, 17.20000000000003, 'Skull')
 []:
 []:
 []:
[52]: labels.replace({'0':'Male','1':"Female"})
      AttributeError
                                                 Traceback (most recent call last)
      /Users/bopei/Library/CloudStorage/Box-Box/bp/USF/Courses/Data visualization/
        →Python Version/Chapter12.ipynb Cell 9 in <cell line: 1>()
       ----> <a href='vscode-notebook-cell:/Users/bopei/Library/CloudStorage/Box-Box/b
        →USF/Courses/Data%20visualization/Python%20Version/Chapter12.
        sipynb#X21sZmlsZQ%3D%3D?line=0'>1</a> labels.replace({'0':'Male','1':"Female"}
      AttributeError: 'dict_keys' object has no attribute 'replace'
```

```
[49]: for i in labels:
          print(i)
     0
     1
[51]: labels()
                                                 Traceback (most recent call last)
       TypeError
       /Users/bopei/Library/CloudStorage/Box-Box/bp/USF/Courses/Data visualization/
        →Python Version/Chapter12.ipynb Cell 11 in <cell line: 1>()
       ----> <a href='vscode-notebook-cell:/Users/bopei/Library/CloudStorage/Box-Box/b'/
        →USF/Courses/Data%20visualization/Python%20Version/Chapter12.
        →ipynb#X23sZmlsZQ%3D%3D?line=0'>1</a> labels()
       TypeError: 'dict_keys' object is not callable
[42]: axes._legend_data.values()
[42]: dict values([<matplotlib.collections.PathCollection object at 0x7fa6cbec60a0>,
      <matplotlib.collections.PathCollection object at 0x7fa6cbec66d0>])
[43]: axes._legend_data.keys()
[43]: dict_keys(['0', '1'])
[26]: for a in ax.axes.flatten():
          print(a)
     AxesSubplot(0.038953,0.685453;0.300983x0.296831)
     AxesSubplot(0.346287,0.685453;0.300983x0.296831)
     AxesSubplot(0.65362,0.685453;0.300983x0.296831)
     AxesSubplot(0.038953,0.375504;0.300983x0.296831)
     AxesSubplot(0.346287,0.375504;0.300983x0.296831)
     AxesSubplot(0.65362,0.375504;0.300983x0.296831)
     AxesSubplot(0.038953,0.0655556;0.300983x0.296831)
     AxesSubplot(0.346287,0.0655556;0.300983x0.296831)
     AxesSubplot(0.65362,0.0655556;0.300983x0.296831)
 []:
 [9]: blue_jays
 [9]:
                           BirdID KnownSex
                                            BillDepth BillWidth BillLength
           Unnamed: 0
                                                                                Head \
      0
                    1 0000-00000
                                         Μ
                                                 8.26
                                                             9.21
                                                                        25.92
                                                                               56.58
      1
                    2 1142-05901
                                                 8.54
                                                            8.76
                                                                        24.99
                                                                               56.36
                    3 1142-05905
                                                 8.39
                                                            8.78
                                                                        26.07 57.32
      2
                                         М
```

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3
           4 1142-05907
                                                  9.30
                                                            23.48 53.77
                                F
                                       7.78
4
           5 1142-05909
                                       8.71
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                                                            25.47 57.32
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                                       8.70
           119
                962-62176
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                                                            24.62 56.61
118
                                M
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                962-62181
                                M
                                       7.96
                                                  9.80
                                                            25.07 55.68
120
                962-62184
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                                       7.90
                                                  9.30
                                                            23.60 53.90
           121
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           122
                962-62185
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    Mass Skull Sex
0
    73.30 30.66
1 75.10 31.38
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2 70.25 31.25
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3 65.50 30.29
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   74.90 31.85
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           ... ...
118 77.00 31.99
119 68.00 30.61
                   1
120 63.90 30.30
                   0
121 70.45 29.90
                   0
122 66.00 29.70
                   0
[123 rows x 10 columns]
```

[]: