021_Folien

November 23, 2018

```
In [5]: from matplotlib import pyplot as plt
    import numpy as np
    %matplotlib inline
```

1 Beschreibende Statistik

1.1 Übersicht

LagemaSS

Streuung

Graphiken

Mehrdimensionale Daten

- Diskrete Verteilung
- Kontinuierliche Verteilung

Abhängigkeit

1.2 Bivariate Stichproben

Beispiele:

- zwei Würfel
- Strom und Spannung
- Luftdruck und Höhe über Meer
- Regentropfen auf Blatt Papier (Koordinaten x und y)

2 Diskrete Verteilung

2.0.1 Kontingenztafel der absoluten Häufigkeiten

```
#1 \ #2 | head | tail |
         5 I
                  5 I
head
      | 4|
                  6 I
tail
2.0.2 ... mit Randverteilung
In [7]: '''calculate absolute frequencies (Python list)'''
       n11 = results.count([0, 0])
       n12 = results.count([0, 1])
       n21 = results.count([1, 0])
       n22 = results.count([1, 1])
       print('#1 \ #2 | head | tail | | sum ')
       print('----')
       print('head |{:5d} |{:5d} ||{:5d} '.format( n11,
                                                         n12,
                                                                n11+n12))
       print('tail |{:5d} |{:5d} ||{:5d} '.format( n21,
                                                                n21+n22))
                                                         n22,
       print('======')
       print('sum |{:5d} |{:5d} |.format(n11+n21, n12+n22, n11+n12+n21+n22))
#1 \ #2 | head | tail | | sum
          5 |
                 5 ||
                       10
head
tail
         4 |
                6 ||
                       10
______
          9 | 11 ||
      20
sum
In [8]: '''calculate absolute frequencies (numpy ndarray)'''
       npres = np.array(results)
                                          # convert data to ndarray
       n11 = (npres==[0, 0]).all(axis=1).sum() # comparing (2val)-array => boolean array
       n12 = (npres==[0, 1]).all(axis=1).sum() # ... only if both are True ("all")
       n21 = (npres==[1, 0]).all(axis=1).sum() # ... then count as "1" (else False="0")
       n22 = (npres==[1, 1]).all(axis=1).sum() # ... summing up these gives total number
       print('#1 \ #2| head | tail || sum ')
       print('----')
       print('head |{:5d} |{:5d} ||{:5d} '.format( n11,
                                                         n12,
                                                                n11+n12))
       print('tail |{:5d} |{:5d} ||{:5d} '.format( n21,
                                                         n22,
                                                                n21+n22))
       print('======')
       print('sum |{:5d} |{:5d} ||{:5d} '.format(n11+n21, n12+n22, n11+n12+n21+n22))
#1 \ #2 | head | tail | | sum
          5 I
                 5 ||
head
                       10
          4 l
                 6 11
tail
                11 ||
          9 |
                       20
SIIM
```

2.0.3 relative Häufigkeiten

two coins

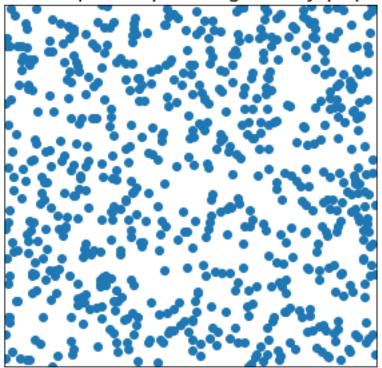
$$f_{ij} = \frac{h_{ij}}{n} \qquad f \in [0,1]$$

```
In [9]: '''calculate relative frequencies (numpy ndarray to list)'''
      nprel = np.array(results) # place holder for calculating relative frequency
      n = nprel.shape[0]
                              # number of value(pairs): 1st dimension of shape
      f11 = nprel.tolist().count([0, 0])/n # since a list again,
      f12 = nprel.tolist().count([0, 1])/n # ... we can count f21 = nprel.tolist().count([1, 0])/n # ... different el
                                       # ... different elements
      f22 = nprel.tolist().count([1, 1])/n
                                        # ... as before
      print('#1 \ #2| head | tail || sum ')
      print('----')
      print('head |{:5.2f} |{:5.2f} |.format( f11,
                                                         f12.
                                                               f11+f12))
      print('tail |{:5.2f} |{:5.2f} ||{:5.2f} '.format( f21,
                                                         f22,
                                                               f21+f22))
      print('======"')
      #1 \ #2 | head | tail | | sum
_____
head | 0.25 | 0.25 || 0.50
tail | 0.20 | 0.30 || 0.50
_____
     | 0.45 | 0.55 || 1.00
```

3 Kontinuierliche Verteilung

```
In [8]: '''Two dimensional random variable: raindrops on a paper'''
       f = plt.figure(figsize=(5, 5))
                                               # square
       np.random.seed(987654)
                                               # initialize random generator to same
       rain = np.random.random((2, 700))
                                              # draw 2x700 [0...1] random numbers
       plt.scatter(rain[0], rain[1])
                                              # let it rain
       plt.axis([0, 1, 0, 1])
                                               # (restrict frame)
       plt.title('rain drops keep falling on my paper') # (headline)
       ax = plt.gca()
                                               # (get instance of axes drawn)
       ax.axes.get_xaxis().set_visible(False) # (no tickmarks on x)
       ax.axes.get_yaxis().set_visible(False); # (
                                                             and y axis)
```

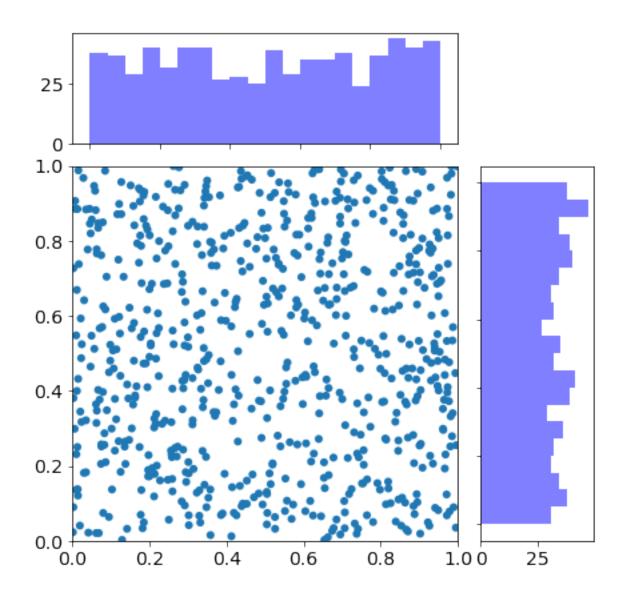
rain drops keep falling on my paper



4 Streudiagramm mit Randverteilung

der einzelnen Komponenten (Dimensionen)

```
In [10]: '''three sub-graph-axis-sets: axmx, axmy and axxy in one figure'''
        fig = plt.figure(figsize=(7, 7))
        bins = np.linspace(0.0, 1.0, 21)
                                                            # 20 bins from 0 to 1
         # define a 4x4 grid and distribute the virtual squares:
         # upper marginal 3x1, until 3rd column
         axmx = plt.subplot2grid((4, 4), (0, 0), colspan=3)
         # right marginal 1x3, start 2nd row, last col
         axmy = plt.subplot2grid((4, 4), (1, 3), rowspan=3)
         # main window, size 3x3, start 2nd row
         axxy = plt.subplot2grid((4, 4), (1, 0), colspan=3, rowspan=3)
         # x-marginal histogram
         axmx.hist(rain[0], color='b', bins=bins, label='x', alpha=.5)
         axmy.hist(rain[1], color='b', bins=bins, label='y', alpha=.5,
                                                   # y-marginal, rotated
                     orientation='horizontal')
        axxy.scatter(rain[0], rain[1], edgecolor='') # let it rain in big xy-panel
         axmx.xaxis.set_ticklabels([])
                                                     # no tickmarks
         axmy.yaxis.set_ticklabels([])
         axxy.axis([0, 1, 0, 1]);
                                                    # restrict area to full data range
```



5 KenngröSSen mehrdimensionaler Daten

5.1 Mittelwert

Randverteilung erlaubt Bestimmung des 2D-Mittelwerts

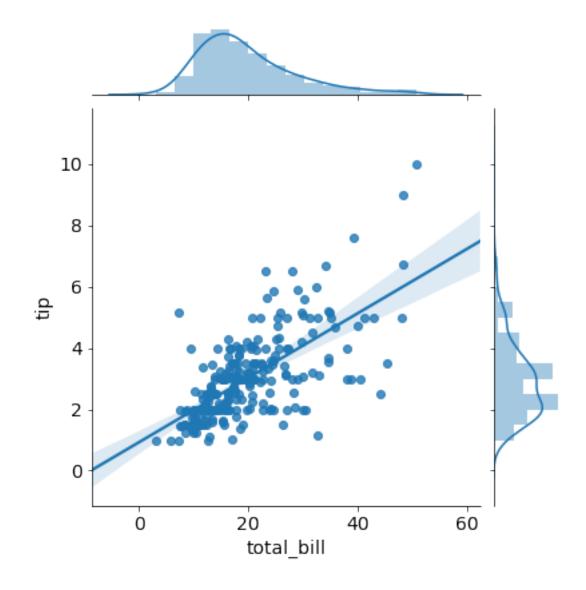
5.2 Varianz

Randverteilung erlaubt Bestimmung der 2D-Streuung(en)

es sollt auch in 2D

5.3 seaborn - weitere Bibliothek zur Datenexploration

```
In [5]: '''seaborn allows easy data exploration'''
    import seaborn as sns
    tips = sns.load_dataset("tips") # Load one of the data sets that come with seaborn
    sns.jointplot("total_bill", "tip", tips, kind='reg');
```

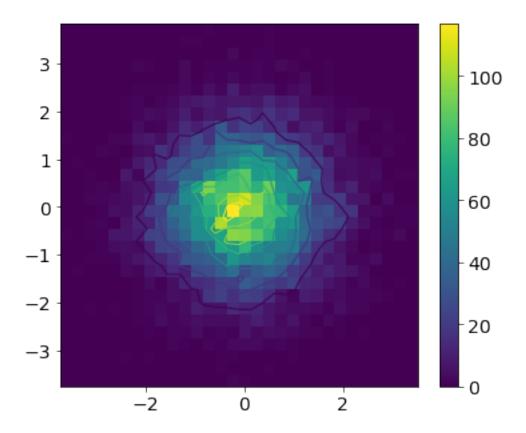


5.4 Dichte-Verteilung: 2D-Histogramm

Bei zu vielen Datenpunkten gibt es analog zum eindimensionalen Fall Histogramm die Möglichkeit in die dritte Dimension z die Häufigkeitsverteilung aufzutragen.

Matplotlib hat dazu hist2d farbkodiert.

```
In [6]: '''two dimensional density plot'''
    x1 = np.random.normal(size=10000)  # random x values
    x2 = np.random.normal(size=10000)  # random y values
    f = plt.figure(figsize=(6, 5))
    counts, xbins, ybins, image = plt.hist2d(x1, x2, bins=[30,30]) # counts-matrix
    plt.colorbar()  # visualize translation color-number
    plt.contour(xbins[:-1], ybins[:-1], counts.T, linewidths=1); # to use for contours
```



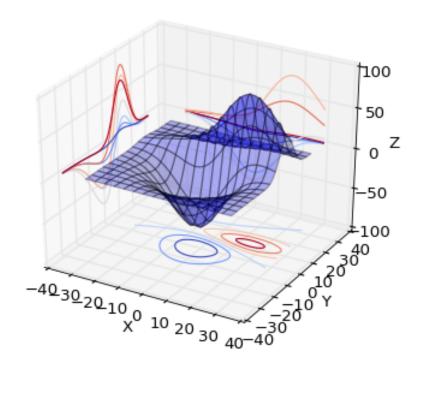
plt.hist2d?

5.5 3D-Darstellung

zweidimensionale Basis: $y = f(x_1, x_2)$ Graphische Darstellung - diskret: mit *Nadeln* - kontinuierlich: 3D-Graphik

In [13]:

Out[13]:

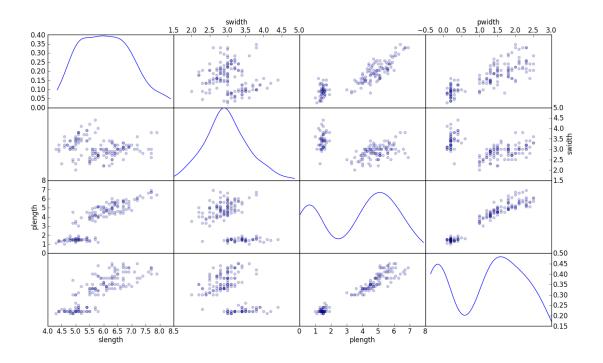


5.6 höherdimensional

• Streumatrix

In [14]:

Out[14]:



6 Zusammenfassung beschreibende mehrdimensionale Statistik

- Datenreduktion KenngröSSen
 - Mittelwerte
 - Standardabweichungen
- Struktur in den Daten erkennen
 - Form der Verteilung
 - Form der Randverteilung
- Anschauliche Darstellung
 - Kontingenztabelle, Vierfeldertafel
 - 2D-Histogramm

Ausblick:

- Abhängigkeit mehrerer Variabler
 - Korrelation
 - Regression

7 Links

- Matplotlib Graphikgalerie http://matplotlib.org/gallery.html
- Pandas Graphiken: http://pandas.pydata.org/pandas-docs/stable/visualization.html

8 Fragen?