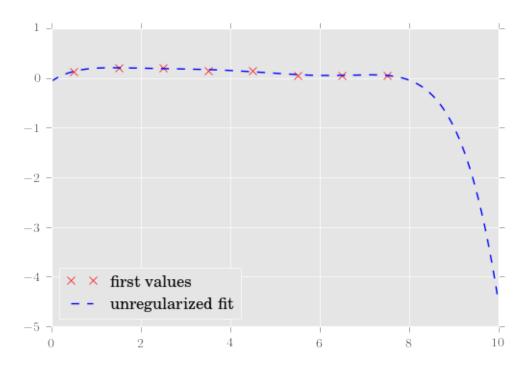
# Aufgabe3

# January 17, 2017

```
In [1]: import numpy as np
        import matplotlib.pyplot as plt
        import pandas as pd
        from scipy.optimize import curve_fit
       plt.style.use('ggplot')
       plt.rcParams['text.usetex'] = True
       plt.rcParams['text.latex.unicode'] = True
       plt.rcParams['font.family'] = 'lmodern'
        %matplotlib inline
In [2]: df = pd.read_csv('aufg_a.csv')
       print(df)
       print(df.columns)
           "y_0"
    X
0 0.5 0.132939
  1.5 0.204351
1
  2.5 0.197394
3 3.5 0.157457
  4.5 0.139232
5 5.5 0.060324
6
  6.5 0.056360
7 7.5 0.051944
Index(['x', ' "y_0"'], dtype='object')
```

## Aufgabenteil a) least squares ohne Regularisierung

#### Out[5]: <matplotlib.legend.Legend at 0x7f06a6072e80>



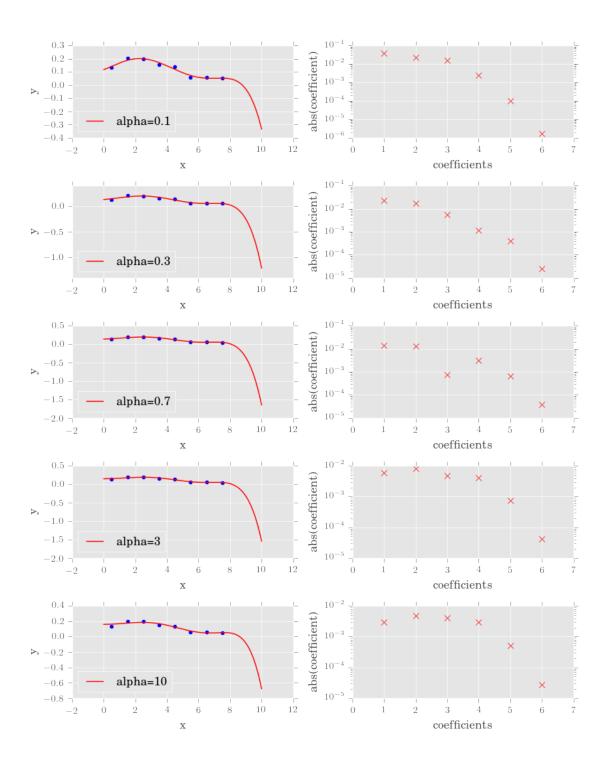
## Aufgabenteil b) least squares mit Regularisierung (Ridge)

```
In [6]: X = df['x'].values
    y = df[' "y_0"'].values
    alphas = [0.1,0.3,0.7,3,10]
    xx = np.linspace(0,10,1000)

In [9]: from sklearn.linear_model import LinearRegression
    from sklearn.linear_model import Ridge
    from sklearn.pipeline import make_pipeline
    from sklearn.preprocessing import PolynomialFeatures

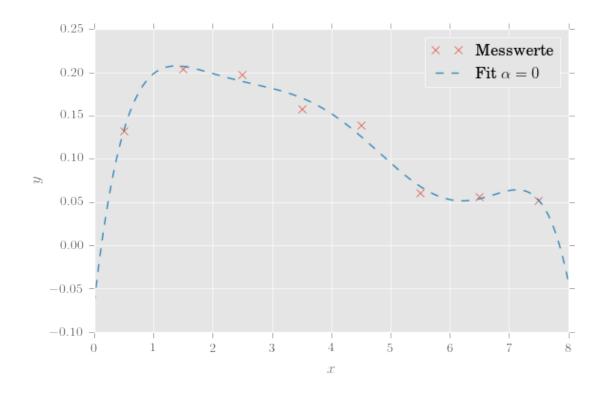
def plot_approximation(est, ax, label=None):
    """Plot the approximation of ``est`` on axis ``ax``. """
    ax.scatter(X, y)
    ax.plot(xx, est.predict(xx[:, np.newaxis]), color='red', label=label)
    ax.set_ylabel('y')
    ax.set_xlabel('x')
    ax.legend(loc='lower left') #, fontsize='small')
```

```
In [10]: fig, ax_rows = plt.subplots(5, 2, figsize=(8, 10))
        coeffs = pd.DataFrame()
        def plot_coefficients(est, ax, label=None, yscale='log'):
            coef = est.steps[-1][1].coef_.ravel()
            print(coef)
            if yscale == 'log':
                ax.semilogy(np.abs(coef), 'rx', label=label)
            else:
                ax.plot(np.abs(coef),'rx', label=label)
            ax.set_ylabel('abs(coefficient)')
            ax.set_xlabel('coefficients')
            ax.set_xlim((0, 7))
        degree = 6
        for alpha, ax_row in zip(alphas, ax_rows):
            ax_left, ax_right = ax_row
            est = make_pipeline(PolynomialFeatures(degree), Ridge(alpha=alpha))
            est.fit (X.reshape(-1,1), y)
            plot_approximation(est, ax_left, label='alpha=%r' % alpha)
            plot_coefficients(est, ax_right, label='Ridge(alpha=%r) coefficients'
        plt.tight_layout()
[ 0.0000000e+00
                  3.98408576e-02
                                  2.37780622e-02 -1.60774510e-02
  2.45036421e-03 -1.00768394e-04 -1.57905362e-06
[ 0.0000000e+00
                  2.29969149e-02
                                   1.77004081e-02 -5.56852913e-03
 -1.16203148e-03 3.90150061e-04 -2.51701597e-05
                  1.37608196e-02 1.31475191e-02
[ 0.0000000e+00
                                                    7.69817553e-04
 -3.22846727e-03 6.62187333e-04 -3.79484647e-05]
[ 0.00000000e+00 5.84671455e-03 8.15234448e-03
                                                    4.95334977e-03
 -4.17870714e-03 7.47236291e-04 -4.04580699e-05]
[0.000000000e+00 2.95787192e-03 4.96839389e-03]
                                                    4.06305987e-03
 -3.06483047e-03 5.20224266e-04 -2.68041849e-05]
```



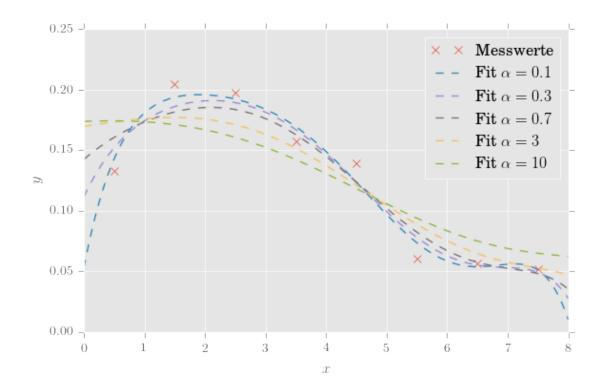
# Selbstimplementierte least squares

```
Gamma = np.dot(C,A)
              for alpha in alphas:
                  a = np.dot(np.dot(np.linalg.inv(np.dot(A.T,A) + alpha* np.dot(Gamma
                  y_plot = func(xx, *a)
                  plt.plot(xx,y_plot,'--',label=r'Fit $\alpha = {}$'.format(alpha))
              plt.xlabel(r'$x$')
             plt.ylabel(r'$y$')
             plt.legend(loc='best')
             plt.tight_layout()
             plt.savefig('{}.pdf'.format(name))
             plt.show()
             plt.clf()
In [12]: def get_err(y):
              return np.std(y)/np.sqrt(len(y))
         def getInvErr(y):
              return np.sqrt(len(y)) / np.std(y)
In [13]: def f_0(x):
             return x**0
         def f_{1}(x):
              return x
         def f_2(x):
              return x**2
         def f 3(x):
              return x**3
         def f_4(x):
             return x**4
         def f_5(x):
              return x**5
         def f_6(x):
             return x**6
         x = df['x']
         A = \text{np.array}([f_{-6}(x), f_{-5}(x), f_{-4}(x), f_{-3}(x), f_{-2}(x), f_{-1}(x), f_{-0}(x)]).T
In [14]: plotter([0], '3a')
```



<matplotlib.figure.Figure at 0x7f06a1179a20>

In [15]: plotter([0.1,0.3,0.7,3,10],'3b')



<matplotlib.figure.Figure at 0x7f06a11612b0>

3

4

0.131928

0.148677

0.092533

0.110090

0.199207

0.082094

```
Out[16]:
                                  "y_1
                                             "y_2
                                                        "y_3
                                                                   "y_4
                                                                              "y_5
                       "y_0
               Х
         0
             0.5
                  0.103939
                             0.149753
                                        0.112739
                                                   0.184439
                                                              0.145831
                                                                         0.113267
                                                                                    0.1554
             1.5
                  0.219092
                             0.193544
                                        0.160651
                                                   0.170565
                                                              0.184050
                                                                         0.163337
                                                                                    0.2121
         1
                                                                                     0.1709
             2.5
                  0.177677
                             0.166459
                                        0.236241
                                                   0.205570
                                                              0.197066
                                                                         0.230060
             3.5
                  0.139580
                             0.163861
                                        0.167301
                                                   0.141728
                                                              0.143816
                                                                         0.161717
                                                                                     0.1538
             4.5
                  0.115663
                             0.112004
                                        0.111168
                                                   0.102507
                                                              0.108234
                                                                         0.085693
                                                                                     0.1136
         5
             5.5
                  0.094581
                             0.089400
                                        0.079080
                                                   0.066638
                                                              0.063737
                                                                         0.101826
                                                                                     0.0663
                             0.058229
                                        0.081565
                                                   0.072236
                                                              0.088597
                                                                         0.079524
                                                                                     0.0628
             6.5
                  0.086982
             7.5
                  0.062485
                             0.066751
                                        0.051256
                                                   0.056316
                                                              0.068670
                                                                         0.064577
                                                                                    0.0647
                                                 "y_40
                                                                                  "y_43
                 "y_7
                            "y_8
                                                            "y_41
                                                                       "y_42
         0
             0.155187
                        0.105173
                                              0.080580
                                                         0.111108
                                                                    0.122147
                                                                               0.146955
             0.166457
                        0.186264
                                                         0.215006
                                                                    0.147765
                                                                               0.164665
                                              0.206292
         2
             0.181053
                        0.186724
                                              0.182444
                                                         0.188019
                                                                    0.144067
                                                                               0.239231
```

0.179982

0.114844

0.103984

0.169037

0.126542

0.055161

0.213696

0.161769

0.073740

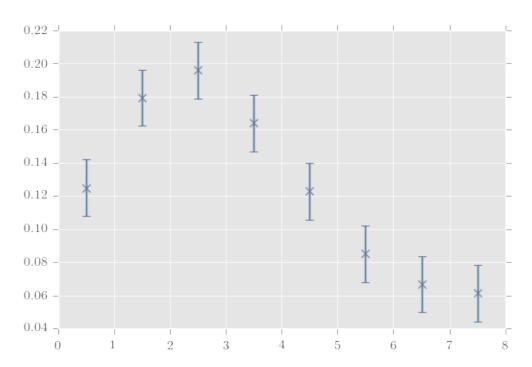
0.157921

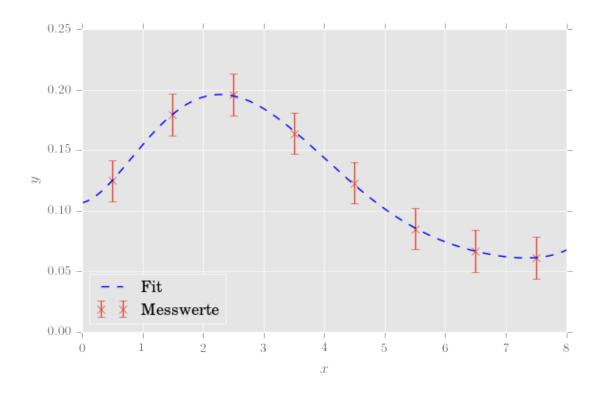
0.111842

0.091454

```
6 0.063917 0.060868
                                         0.065893 0.055365 0.074342 0.052247
                                . . .
         7 0.060249 0.069581
                                 . . .
                                         0.065981 0.079762 0.062473 0.035685
               "y_44
                                  "y_46
                                             "y_47
                                                       "y_48
                                                                "y_49
                         "y_45
         0.076564 \quad 0.114331 \quad 0.132773 \quad 0.167562 \quad 0.135968 \quad 0.131620
         1 \quad 0.238038 \quad 0.203325 \quad 0.203236 \quad 0.213183 \quad 0.223799 \quad 0.148379
         2 0.103315 0.198792 0.201374 0.197383 0.126766 0.193543
         3 0.229623 0.207216 0.125185 0.119271 0.212534 0.193191
         4 0.142766 0.097612 0.132306 0.101199 0.085060 0.094930
         5 0.069506 0.064214 0.104694 0.063192 0.094040 0.098975
         7 0.069191 0.059323 0.057535 0.052826 0.068084 0.070466
         [8 rows x 51 columns]
In [18]: def getLineArrays(line):
            y = []
             for i in range (50):
                y.append(Y[i][line])
             return np.array(y)
         #ekelhaft geht bestimmt auch so x, *y = ...
         x, y_0, y_1, y_2, y_3, y_4, y_5, y_6, y_7, y_8, y_9, y_10, y_11, y_12, y_1
             'aufg_c.csv', delimiter=', ', skip_header=1, unpack=True)
         # Zusammenfassen der y-Werte in einer Matrix
         Y = ([y_0, y_1, y_2, y_3, y_4, y_5, y_6, y_7, y_8, y_9, y_{10}, y_{11}, y_{12}]
              y_26, y_27, y_28, y_29, y_30, y_31, y_32, y_33, y_34, y_35, y_36, y_
         # Erzeugen von Arrays von y-Werten, die die Inhalte der Zeilen der
         # csv-Tabelle enthalten
        y_0 = getLineArrays(0)
        y_1 = getLineArrays(1)
        y_2 = getLineArrays(2)
        y_3 = getLineArrays(3)
        y_4 = getLineArrays(4)
        y_5 = getLineArrays(5)
        y_6 = getLineArrays(6)
        y_7 = getLineArrays(7)
         \# Berechnung des Vektors y als Mittelwertvektor von y_0 bis y_7
        y = np.array((np.mean(y_0), np.mean(y_1), np.mean(y_2), np.mean(
             y_3), np.mean(y_4), np.mean(y_5), np.mean(y_6), np.mean(y_7))
         # Berechnung der Gewichtungsmatrix
         invErr = (getInvErr(y_0), getInvErr(y_1), getInvErr(y_2), getInvErr(
             y_3), getInvErr(y_4), getInvErr(y_5), getInvErr(y_6), getInvErr(y_7))
        W = np.diag(invErr)
         # Plot der Messwerte
         err = qet_err(y)
        plt.errorbar(x, y, yerr=err, fmt='x', label='Messwerte')
```

```
plt.errorbar(x, y, yerr=err, fmt='x', label='Messwerte')
# Berechnung des Loesungsvektors a
a = np.dot(np.dot(np.dot(np.linalg.inv(np.dot(A.T, np.dot(W, A))), A.T), Westernamental actions and the second seco
```





In [ ]: