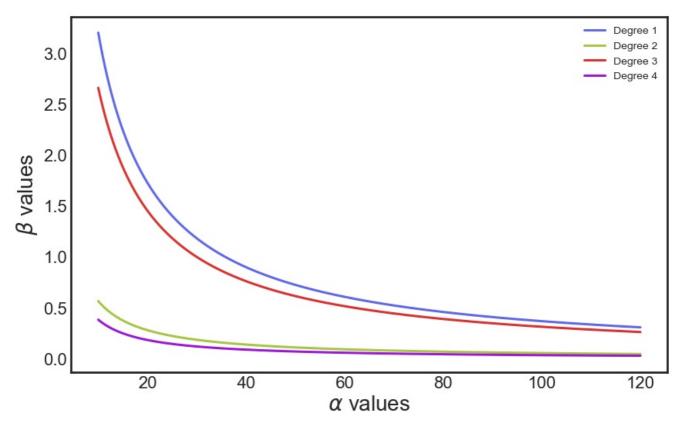
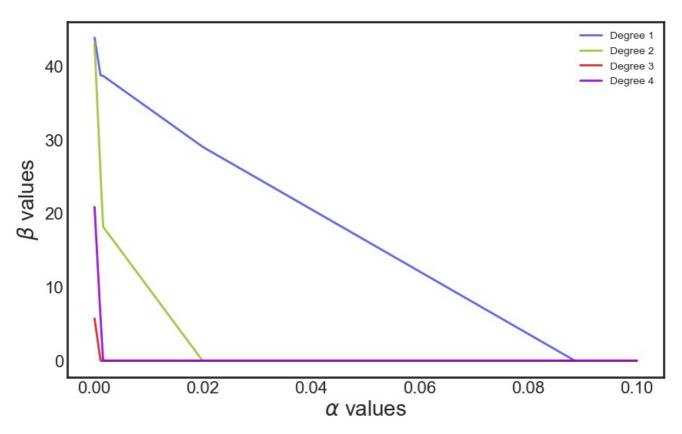
```
In [8]: %matplotlib inline
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          from sklearn.linear_model import Lasso
from sklearn.linear_model import Ridge
          from sklearn.preprocessing import PolynomialFeatures
          from sklearn.preprocessing import normalize
          import warnings
          warnings.filterwarnings('ignore')
 In [9]: large = 22; med = 16; small = 10
          params = {'axes.titlesize': large,
                      'legend.fontsize': med,
                      'figure.figsize': (16, 10),
                     'axes.labelsize': med,
                      'axes.titlesize': med,
                      'axes.linewidth': 2,
                      'xtick.labelsize': med,
                     'ytick.labelsize': med,
'figure.titlesize': large}
          plt.style.use('seaborn-white')
          plt.rcParams.update(params)
          %matplotlib inline
In [10]: df = pd.read_csv("bacteria_train.csv")
In [11]: df.head()
             Perc_population Spreading_factor
          0
                                   0.190708
                      1.535
          1
                      5.555
                                   0.326928
          2
                     -0.277
                                  -0.459699
          3
                      1.724
                                  -0.193013
          4
                      -0.550
                                  -0.835745
In [12]: x, y = df[['Spreading factor']], df['Perc population']
          maxdeg = 4
In [13]:
          x poly = PolynomialFeatures(degree= maxdeg, include bias=False).fit transform(x)
          x_poly = normalize(x_poly, axis=0)
In [14]: alpha_list = np.linspace(10,120,1000)
          len(alpha_list)
          1000
Out[14]:
In [15]: coeff_list = []
          for a in alpha list:
               ridge_reg = Ridge(alpha= a)
               ridge_reg.fit(x_poly, y)
               coeff_list.append(ridge_reg.coef_)
In [16]: ridge_trend = np.array(coeff_list).T
In [17]: colors = ['#5059E8','#9FC131FF','#D91C1C','#9400D3','#FF2F92','#336600','black']
          fig, ax = plt.subplots(figsize = (10,6))
          for i in range(maxdeg):
               ax.plot(alpha_list, np.abs(ridge_trend[i]), color=colors[i],
               alpha = 0.9, label = f'Degree {i+1}', lw=2.2)
ax.legend(loc='best',fontsize=10)
               ax.set_xlabel(r'$\alpha$ values', fontsize=20)
ax.set_ylabel(r'$\beta$ values', fontsize=20)
          fig.suptitle(r'Ridge ($L_2$) Regression');
```

Ridge (L_2) Regression

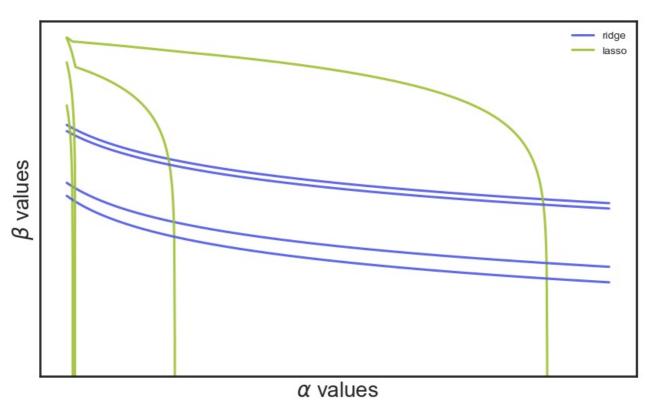


```
alpha_list = np.linspace(1e-4,1e-1,1000)
In [18]:
          len(alpha_list)
          1000
Out[18]:
In [19]:
          coeff_list = []
           for a in alpha_list:
               lasso_reg = Lasso(alpha=a,max_iter=250000)
               lasso_reg.fit(x_poly, y)
               coeff_list.append(lasso_reg.coef_)
In [20]: lasso_trend = np.array(coeff_list).T
In [21]: colors = ['#5059E8','#9FC131FF','#D91C1C','#9400D3','#FF2F92','#336600','black']
           fig, ax = plt.subplots(figsize = (10,6))
           for i in range(maxdeg):
               ax.plot(alpha_list, np.abs(lasso_trend[i]), color=colors[i],
               alpha = 0.9, label = f'Degree \{i+1\}', lw=2) ax.legend(loc='best', fontsize=10)
               ax.set_xlabel(r'$\alpha$ values', fontsize=20)
ax.set_ylabel(r'$\beta$ values', fontsize=20)
           fig.suptitle(r'Lasso ($L 1$) Regression');
```

Lasso (L₁) Regression



Ridge (L_2) vs Lasso (L_1) Regression



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