```
In [2]:
               np.random.seed(1)
 In [3]:
               A_{,y} = datagen()
               n,p,n_train,n_test,A,y,A_test,y_test = datatreat(A_,y_,0.6)
              Number of obs: 173
              n train: 103
              n test: 70
              Number of explicative variables: 9
 In [4]:
               def mse ridge(x,lbd): return 0.5*np.linalg.norm(A@x-y)**2/n train+0.5*lbd*np.linalg.norm(x)**2
 In [5]:
               lbd1=0.1
               1bd2=5
               x0,trash=solve_ridge(A,y,0)
               x1,trash=solve_ridge(A,y,lbd1)
               x2,trash=solve_ridge(A,y,lbd2)
 In [6]:
               print("Norm of x with lambda=%.1f: "%0, np.round(np.linalg.norm(x0),3))
               print(x0)
               print("Norm of x with lambda=%.2f: "%lbd1, np.round(np.linalg.norm(x1),3))
               print("Norm of x with lambda=%.1f: "%1bd2, np.round(np.linalg.norm(x2),3))
               print(x2)
              Norm of x with lambda=0.0: 5.295
              [1.55005263 2.26343997 0.64557462 2.82098553 0.64259705 1.62561205
               1.11312589 1.2918067 2.48540551]
              Norm of x with lambda=0.10: 4.937
              [1.51234584 1.81786824 1.05966036 2.58615889 0.81948639 1.50306634
               1.17136137 1.33581472 2.21005061]
              Norm of x with lambda=5.0: 2.26
              [0.88544134 \ 0.83496142 \ 0.81579881 \ 0.67714247 \ 0.76494586 \ 0.56591435]
               0.6436692 0.74187305 0.79543403]
 In [7]:
               plt.scatter(np.arange(1,10),x0,color='blue',marker='+',label='lambda=%.1f'%0,s=50)
               \verb|plt.scatter(np.arange(1,10),x1,color='green',marker='+',label='lambda=%.1f'%lbd1,s=50)|
               plt.scatter(np.arange(1,10),x2,color='red',marker='+',label='lambda=%.1f'%1bd2,s=50)
               plt.xlabel("Id of Variable")
               plt.ylabel("Value of Variable")
               plt.legend(loc='upper left');
               plt.subplots_adjust(left=None, bottom=None, right=2, top=1, wspace=None, hspace=None)
                             lambda=0.0
                             lambda=0.1
                                                                                  +
                             lambda=5.0
                 2.5
              Value of Variable
                 2.0
                 1.5
                                                                                                                                        #
                 1.0
                                                                                                    #
                                                                                                                                                                             +
                                                                                               ld of Variable
 In [8]:
               \textbf{def} \ \texttt{mse\_elastic\_net(x,lbd,mu):} \ \textbf{return} \ 0.5*\texttt{np.linalg.norm(A@x-y)**2/n\_train+0.5*lbd*np.linalg.norm(x)**2+\texttt{mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+mu*np.linalg.norm(x)**2+m
               def dmse ridge(x,lbd): return A.T@(A@x-y)/n train+lbd*x
               def ISTA(tau,lbd,mu,x 0,n iter):
                     x=np.zeros((n_iter,x_0.size))
                     x[0]=x 0
                     fx=np.full(n_iter,mse_elastic_net(x[0],lbd,mu))
                     tau=tau*np.ones(n_iter)
                     for i in range(1, n_iter):
                           grad_ridge=dmse_ridge(x[i-1],lbd)
                           iter_GD=x[i-1]-tau[i]*grad_ridge
                           ind_left=np.argwhere(iter_GD<-tau[i] *mu)</pre>
                           ind_right=np.argwhere(iter_GD> tau[i]*mu)
                           x[i,ind_left] = iter_GD[ind_left] + tau[i] * mu
                           x[i,ind_right] = iter_GD[ind_right] -tau[i] *mu
                            fx[i]=mse_elastic_net(x[i],lbd,mu)
                     return x, fx
 In [9]:
               lbd=0.022857142857142857
               x 0=np.ones(A.shape[1])
               tau=0.1
               n iter=100
               mu=np.array([0.,1.,2.,3.,4.,5.,6.])
               N mu=mu.size
               x=np.zeros((N mu, n iter, A.shape[1]))
               Nb_of_zeros=np.ones(N_mu).astype(int)
               for i in range(N_mu):
                     x[i],trash=ISTA(tau,lbd,mu[i],x 0,n iter)
                     Nb of zeros[i]=np.count nonzero(x[i,-1]==0)
                     print("mu=%.0f: Number of zeros="%mu[i], Nb of zeros[i])
              mu=0: Number of zeros= 0
              mu=1: Number of zeros= 0
              mu=2: Number of zeros= 0
              mu=3: Number of zeros= 1
              mu=4: Number of zeros= 3
              mu=5: Number of zeros= 5
              mu=6: Number of zeros= 7
In [10]:
               fig, ((ax0,ax1,ax2,ax3),(ax4,ax5,ax6,ax7)) = plt.subplots(nrows=2,ncols=4)
               ax0.scatter(mu, Nb of zeros, color='black', marker='+', s=100)
               ax0.set title("Number of zeros in function of $\mu$")
               ax=[ax1,ax2,ax3,ax4,ax5,ax6,ax7]
               for i in range(0,N mu):
                     ax[i].plot(x[i])
                     ax[i].axhline(0,color="grey")
                     ax[i].set title("Convergence of coordinate values with \mu=%.0f"%mu[i])
                     ax[i].set xlabel("Number of iterations")
                     ax[i].set_ylabel("$(x_i)_{1\leq i\leq 9}$ values")
               fig.tight_layout() # Or equivalently, "plt.tight layout()"
               plt.subplots adjust(left=None, bottom=None, right=3, top=1.5, wspace=0.2, hspace=None)
                                                               Convergence of coordinate values with \mu=0
                        Number of zeros in function of \mu
                                                                                                         Convergence of coordinate values with \mu=1
                                                                                                                                                     Convergence of coordinate values with \mu=2
                                                           2.5
                                                                                                                                                 2.0
                                                         2.0
                                                                                                    1.5 values
1.0
                                                                                                                                                 1.5
                                                           1.5
                                                         10 0.5
                                                                                                                                               i 1.0
                                                                                                                                               X)
                                                                                                      0.5
                                                                                          80
                                                                                                                                          100
                                                                                                                                                                  40
                                                                                                                                                     Convergence of coordinate values with \mu = 6
                    Convergence of coordinate values with µ=3
                                                               Convergence of coordinate values with µ=4
                                                                                                         Convergence of coordinate values with µ=5
                                                                                                                                                 3.0
                                                           3.0
                                                                                                      3.0
                2.5
                                                                                                                                                 2.5
                                                           2.5
                                                                                                      2.5
              2.0
1.5
                                                         values
                                                                                                                                                 2.0
                                                           2.0
                                                                                                      2.0
                1.5
                                                                                                                                                 1.5
                                                         0.5 1.5 (x) 0.5
                                                                                                    υ<sub>0</sub> 1.5
              10 (X)
                                                                                                                                                 1.0
                                                                                                      1.0
                                                                                                    (X<sub>1</sub>)
                                                           0.5
                                                                                                      0.5
                           20
                                                                                                                    Number of iterations
                                                                                                                                                               Number of iterations
                               Number of iterations
                                                                          Number of iterations
In [11]:
               def mse test(x): return np.linalg.norm(A test@x-y test)**2/n test
               x0, trash0=solve ridge(A, y, 0)
               f_test0=mse_test(x0)
               f test=np.zeros(N mu)
               for i in range(N_mu):
                     f test[i]=mse_test(x[i,-1])
               print("Solution by minimizing mse vanilla: ",f test0)
               print("Solution by minimizing elastic net: ",f test)
              Solution by minimizing mse vanilla: 5.543818661163919
              Solution by minimizing elastic net: [ 5.56298257 7.42404873 13.45203449 23.43782296 36.57344807 46.95951165
               57.96436243]
In [12]:
               n iter=1000
               mu=np.linspace(0,.1,20)
               N mu=mu.size
               f_search=np.zeros(N_mu)
               x_ridge_lambda,trash = solve_ridge(A,y,lbd)
               f_lambda = mse_test(x_ridge_lambda)
               for i in range (N_mu):
                     x,trash=ISTA(tau,lbd,mu[i],x_0,n_iter)
                     f_{search[i]} = mse_{test(x[-1])}
               i opti = f search.argmin()
               mu_opti = mu[i_opti]
               plt.title("Mse Test with elastic net values")
               plt.plot(mu,f_search,label='MSE Test')
               plt.plot( [mu_opti, mu_opti], [f_search.min(), f_search.max()], 'r--', label="$\mu^*=$%f"%mu_opti)
               plt.xlabel('$\mu$');
               plt.ylabel('MSE test');
               plt.axhline(f lambda,label='$\lambda=\lambda^*, \mu=0$',color='grey')
               plt.axhline(f_search[i_opti],label='$\lambda=\lambda^*, \mu=\mu^*$',color='red')
               plt.legend(loc="upper right");
                                      Mse Test with elastic net values
                                                                          MSE Test
                 5.534
                                                                         \mu^* = 0.047368
                 5.533
                                                                          \lambda = \lambda^*, \mu \neq 0
                                                                          \lambda = \lambda^*, \mu = \mu^*
                 5.532
              ₹ 5.531
                 5.530
                 5.529
                         0.00
                                     0.02
                                                 0.04
                                                             0.06
                                                                         0.08
                                                                                     0.10
```

In [1]:

import numpy as np

import matplotlib.pyplot as plt

from utils import datagen
from utils import datatreat
from utils import solve ridge