In [1]:	<pre>import numpy as np import matplotlib.pyplot as plt</pre>
	<pre>from utils import datagen from utils import datatreat from utils import tau_GD from utils import solve_ridge</pre>
	<pre>from utils import eigen_val from utils import PGD from utils import FWGD</pre>
In [2]: In [3]:	<pre>np.random.seed(1) A ,y = datagen()</pre>
	<pre>n,p,n_train,n_test,A,y,A_test,y_test = datatreat(A_,y_,128) mu,L=eigen_val(A,0)</pre> Number of obs: 173
In [4]:	n train: 128 n test: 45 Number of explicative variables: 9
	<pre>def mse(x): return 0.5*np.linalg.norm(A@x-y)**2/n_train def dmse(x): return A.T@(A@x-y)/n_train</pre>
In [5]:	<pre>def dmse_ridge(x,lbd): return A.T@(A@x-y)/n_train+lbd*x def mse_elastic_net(x,lbd,mu): return 0.5*np.linalg.norm(A@x-y)**2/n_train+0.5*lbd*np.linalg.norm(x)**2+mu*np.l def ISTA(tau,lbd,mu,x_0,n_iter):</pre>
	<pre>x=np.zeros((n_iter,x_0.size)) x[0]=x_0 fx=np.full(n iter,mse elastic net(x[0],lbd,mu))</pre>
	<pre>tau=tau*np.ones(n_iter) for i in range(1,n_iter):</pre>
	<pre>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) ind_right=np.argwhere(iter_GD> tau[i]*mu)</pre>
	<pre>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)</pre>
In [6]:	return x, fx
III [O].	<pre>x,trash=solve_ridge(A,y,0) R_L2_star=np.linalg.norm(x) R_L2=5.25 print(R_L2_star,R_L2)</pre>
In [7]:	5.429668585968503 5.25 x_0=np.ones(p) n_iter=100
In [8]:	<pre>tau_max,tau_opt=tau_GD(A,0) ''' x star L2,trash=solve ridge(A,y,0.0292125258648815)</pre>
	<pre>print(np.linalg.norm(x_star_L2)) fx_star_L2=mse(x_star_L2) print(fx_star_L2) print(fx_star_L2)</pre>
	<pre>x,trash=ISTA(tau_opt,0.0292125258648815,0,x_0,2000) print(np.linalg.norm(x[-1])) print(mse(x[-1])) x_star_L2=x[-1]</pre>
	fx_star_L2=mse(x_star_L2) 5.25000000000017 2.412218672742907
In [9]:	<pre>def proj_L2(x,R): return R*x/(np.maximum(np.linalg.norm(x),R)) def iter_FW_L2(x,R): return -R*dmse(x)/np.linalg.norm(dmse(x))</pre>
	<pre>def theta_FW_fix(x,s,k): return 2/(k+1) def theta_FW_opt(x,s,k): N=np.linalg.norm(A@(s-x))**2 D=np.dot(A@(s-x),A@x-y)</pre>
	C=-D/N if(C<0): return 0 elif(C>1): return 1 else: return C
In [10]:	<pre>xPGD,fxPGD=PGD(lambda x: proj_L2(x,R_L2), tau_opt, x_0, mse, dmse, n_iter) errPGD=np.linalg.norm(xPGD-x_star_L2,axis=1) errfPGD=fxPGD-fx star L2</pre>
	<pre>print(np.max(np.linalg.norm(xPGD,axis=1))) print(fxPGD[-1]) print("-"*50) xFWGD_fix,fxFWGD_fix=FWGD(lambda x: iter_FW_L2(x,R_L2),theta_FW_fix, x_0, mse, dmse, n_iter)</pre>
	<pre>errFWGD_fix=np.linalg.norm(xFWGD_fix-x_star_L2,axis=1) errfFWGD_fix=fxFWGD_fix-fx_star_L2 print(np.max(np.linalg.norm(xFWGD_fix,axis=1))) print(fxFWGD_fix[-1]) print("-"*50)</pre>
	<pre>xFWGD_opt,fxFWGD_opt=FWGD(lambda x: iter_FW_L2(x,R_L2),theta_FW_opt, x_0, mse, dmse, n_iter) errFWGD_opt=np.linalg.norm(xFWGD_opt-x_star_L2,axis=1) errfFWGD_opt=fxFWGD_opt-fx_star_L2</pre>
	<pre>print(np.max(np.linalg.norm(xFWGD_opt,axis=1))) print(fxFWGD_opt[-1]) #FW_upperbound_L2=rho=2*L*4*R_L2**2/np.arange(2,n_iter+2)</pre>
	5.25000000000001 2.412219475445704
In [11]:	5.25 2.41221867472749
	<pre>fig, (ax0,ax1) = plt.subplots(nrows=1,ncols=2) ax0.plot(np.log10(np.arange(1,n_iter+1)),np.log10(fxPGD),label='PGD') ax0.plot(np.log10(np.arange(1,n_iter+1)),np.log10(fxFWGD_fix),label='FWGD(fix)') ax0.plot(np.log10(np.arange(1,n_iter+1)),np.log10(fxFWGD_opt),label='FWGD(opti)')</pre>
	<pre>ax0.set_xlabel('\$log_{10}\$k') ax0.set_ylabel('\$log_{10}f(x_k)\$') ax0.set_title('Convergence of the \$log_{10}\$ of the objective function') ax0.legend()</pre>
	<pre>ax1.plot(np.log10(errfPGD), label='PGD') ax1.plot(np.log10(errfFWGD_fix), label='FWGD(fix)') ax1.plot(np.log10(errfFWGD_opt), label='FWGD(opti)') ax1.set_xlabel('k') ax1.set_ylabel('\$log {10}(f(x k)-f^*)\$')</pre>
	<pre>ax1.set_title('Convergence of the \$log_{10}\$ of the function error ') ax1.legend() fig.tight layout() # Or equivalently, "plt.tight layout()"</pre>
	plt.subplots_adjust(left=None, bottom=None, right=2, top=1, wspace=0.2, hspace=None) Convergence of the log10 of the objective function Convergence of the log10 of the function error
	18 - 16 - FWGD(fix) FWGD(opti) 0 - FWGD(fix) FWGD(opti)
	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
	0.8 0.6 0.4
In [12]:	0.00 0.25 0.50 0.75 1.00 1.25 1.50 1.75 2.00 0 20 40 60 80 100 k
[]	<pre>x,trash=solve_ridge(A,y,0) R_L1_star=np.linalg.norm(x,1) R_L1=np.floor(R_L1_star) print(R_L1_star,R_L1)</pre>
In [13]:	14.483338245059231 14.0 x,trash=ISTA(tau_opt,0,0.204557386591880,x_0,2000) print(np.linalg.norm(x[-1],1))
	<pre>print(mse(x[-1])) x_star_L1=x[-1] fx_star_L1=mse(x_star_L1) 14.00000000000000004</pre>
In [14]:	<pre>2.448790065633061 def proj_simplex(x,R): u=np.flip(np.sort(x))</pre>
	<pre>K=0 while((((np.sum(u[:K+1])-R)/(K+1)) < u[K])): K=K+1 if(K==x.size-1): break</pre>
	<pre>tau=(np.sum(u[:K+1])-R)/(K+1) return np.maximum(x-tau,0) def proj_L1(x,R): if(np.linalg.norm(x,1) <=R): return x</pre>
	<pre>else: return np.sign(x)*proj_simplex(x,R) def iter_FW_L1(x,R): k star=np.argmax(np.abs(dmse(x)))</pre>
	<pre>s=np.zeros(x.size) s[k_star]=-R*np.sign(dmse(x)[k_star]) return s</pre>
In [15]:	<pre>n_iter=200 xPGD,fxPGD=PGD(lambda x: proj_L1(x,R_L1), tau_opt, x_0, mse, dmse, n_iter) errPGD=np.linalg.norm(xPGD-x_star_L1,axis=1) errfPGD=fxPGD-fx star L1</pre>
	<pre>print(np.max(np.linalg.norm(xPGD,1,axis=1))) print(fxPGD[-1]) print("-"*50) xFWGD fix,fxFWGD fix=FWGD(lambda x: iter FW L1(x,R L1),theta FW fix, x 0, mse, dmse, n iter)</pre>
	<pre>errFWGD_fix=np.linalg.norm(xFWGD_fix-x_star_L1,axis=1) errfFWGD_fix=fxFWGD_fix-fx_star_L1 print(np.max(np.linalg.norm(xFWGD_fix,1,axis=1))) print(fxFWGD_fix[-1]) print("-"*50)</pre>
	<pre>xFWGD_opt,fxFWGD_opt=FWGD(lambda x: iter_FW_L1(x,R_L1),theta_FW_opt, x_0, mse, dmse, n_iter) errFWGD_opt=np.linalg.norm(xFWGD_opt-x_star_L1,axis=1) errfFWGD_opt=fxFWGD_opt-fx_star_L1</pre>
	<pre>print(np.max(np.linalg.norm(xFWGD_opt,1,axis=1))) print(fxFWGD_opt[-1]) #FW_upperbound_L1=rho=2*L*9*R_L1**2/np.arange(2,n_iter+2)</pre>
	14.00000000000004 2.448790153147365
In [16]:	13.611947270229187 2.5603952885808123 fig, (ax0,ax1) = plt.subplots(nrows=1,ncols=2)
	<pre>ax0.plot(np.log10(np.arange(1,n_iter+1)),np.log10(fxPGD),label='PGD') ax0.plot(np.log10(np.arange(1,n_iter+1)),np.log10(fxFWGD_fix),label='FWGD(fix)') ax0.plot(np.log10(np.arange(1,n_iter+1)),np.log10(fxFWGD_opt),label='FWGD(opti)') ax0.set xlabel('\$log {10}\$k')</pre>
	<pre>ax0.set_ylabel('\$log_{10}f(x_k)\$') ax0.set_title('Convergence of the \$log_{10}\$ of the objective function') ax0.legend()</pre>
	<pre>ax1.plot(np.log10(errfPGD),label='PGD') ax1.plot(np.log10(errfFWGD_fix),label='FWGD(fix)') ax1.plot(np.log10(errfFWGD_opt),label='FWGD(opti)') ax1.set_xlabel('k') ax1.set_ylabel('\$log_{10}(f(x_k)-f^*)\$')</pre>
	<pre>ax1.set_title('Convergence of the \$log_{10}\$ of the function error ') ax1.legend() fig.tight_layout() # Or equivalently, "plt.tight_layout()"</pre>
	plt.subplots_adjust(left=None, bottom=None, right=2, top=1, wspace=0.2, hspace=None) Convergence of the log10 of the objective function Convergence of the log10 of the function error
	2.00 - 1.75 - 1.50 - PGD FWGD(fix) FWGD(opti) 2 - 0 - FWGD(opti) PGD FWGD(fix) FWGD(opti)
	$ \begin{array}{c} 150 \\ $
	0.75 - 0.50
	0.0 0.5 1.0 1.5 2.0 0 25 50 75 100 125 150 175 200 k