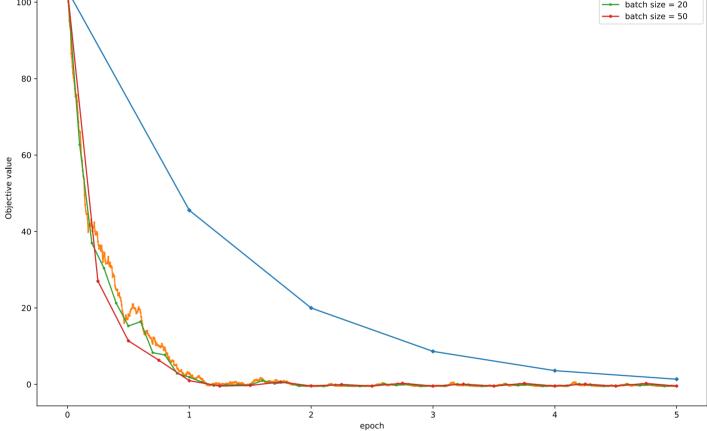
```
In [319...
          import csv
          import numpy as np
          import math
          import matplotlib.pyplot as plt
In [320...
          # read file and store to numpy array
          A = []
          B = []
          C = []
          with open('data_minibatch.csv', newline='') as csvfile:
              reader = csv.reader(csvfile)
              rows = []
              for row in reader:
                 rows.append([float(x) for x in row])
              # shuffle data
              shuffled_indices = np.random.permutation(len(rows))
              for i in range(len(rows)):
                  A.append(rows[i][0])
                  B.append(rows[i][1])
                  C.append(rows[i][2])
          A = np.array(A)
          B = np.array(B)
          C = np.array(C)
In [321...
          def f(x):
              return np.sum(A + (B * x) + (1/2 * (C * math.pow(x,2))))
In [322...
          def f_gradient(x):
              return np.sum(B + (C * x))
In [323...
          def f hessian():
              return np.sum(C)
In [324...
          # 2.1
          # f(x) is Lipchitz continiously differentiable if L > 0
          L = np.sum(np.abs(C))
          print('L: {:.3f}'.format(L))
          if L > 0:
             print('f(x) is Lipchitz continiously differentiable')
          print('Range of step size: 0 < t < {:.3f}\n'.format(2/L))</pre>
          # find x*, p*
          x_{opt} = -np.sum(B)/np.sum(C)
          print('x* is {:.3f} (hessian of f(x*) = {:.3f})'.format(x_opt, f_hessian()))
          p_{opt} = f(x_{opt})
          print('p* is {:.3f}'.format(p_opt))
         L: 166.644
         f(x) is Lipchitz continiously differentiable
         Range of step size: 0 < t < 0.012
         x* is -0.115 (hessian of f(x*) = 166.644)
         p* is -0.441
In [325...
          def search_direction(x, k, i):
              b = B[i:i+k]
              c = C[i:i+k]
              return - np.sum(b + (c * x))
In [326...
          def mini_batch(N,k,t):
              epoch = 5
              # initial x_previous
              x_prev = 1
              f_hist = [f(x_prev)]
              for e in range(epoch):
                  i = 0
                  for n in range(1, N//k + 1):
                      x = x_prev + (t * (search_direction(x_prev, k, i)))
                      f_hist.append(f(x))
                      x_prev = x
                      i += k
              return f_hist
In [327...
          # 2.2
          def result(N,t):
              print('step size:', t)
              plt.figure(figsize=(15, 10))
```

```
k = N #batch size
f hist = mini batch(N,k,t)
plt.plot([x/(N//k) for x in range(len(f_hist))], f_hist,marker='D', markersize=3, label='batch optimization')
k = 1 #batch_size
f_hist_1 = mini_batch(N,k,t)
plt.plot([x/(N//k) for x in range(len(f hist 1))], f hist 1, marker='.', markersize=3, label='batch size = 1')
k = 20 #batch size
f_hist_20 = mini_batch(N,k,t)
 plt.plot([x/(N//k) \  \, \textbf{for} \  \, \textbf{x} \  \, \textbf{in} \  \, \textbf{range}(len(f\_hist\_20))], \  \, f\_hist\_20, marker='*', \  \, \textbf{markersize=3}, \  \, label='batch \  \, \textbf{size} = 20') 
k = 50 #batch size
f hist 50 = \min  batch(N,k,t)
plt.plot([x/(N//k) for x in range(len(f_hist_50))], f_hist_50,marker='P', markersize=3, label='batch size = 50')
plt.xlabel("epoch")
plt.ylabel("Objective value")
plt.legend(loc='upper right')
plt.show()
```

In [328...

```
N = A.shape[0]
t = 0.01
result(N, t)
```

step size: 0.01 -- batch optimization batch size = 1 100 - batch size = 20 → batch size = 50



```
In [329...
          N = A.shape[0]
          t = 0.005
          result(N, t)
```

step size: 0.005

