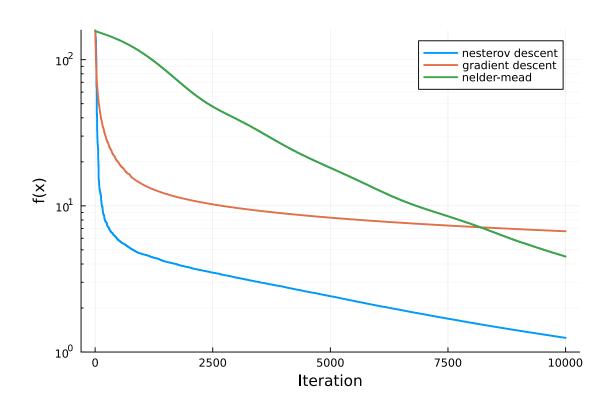
Problem 1

March 17, 2023

```
[1]: using Plots
     using AdvConvex.HW3
     using AdvConvex.HW4
    using Optim
    [ Info: Precompiling Plots
    [91a5bcdd-55d7-5caf-9e0b-520d859cae80]
    [ Info: Precompiling AdvConvex
    [a70558b1-94d0-46ca-a15d-76cbf33c1d08]
    [ Info: Precompiling Optim
    [429524aa-4258-5aef-a3af-852621145aeb]
[2]: mat = get_spam_data()
     X_train, Y_train, X_test, Y_test = train_test_split(mat, 0.05)
[2]: ([-2.3025850929940455 -1.7147984280919266 ... -2.3025850929940455
     -2.3025850929940455; -2.3025850929940455 -2.3025850929940455 ...
     -2.3025850929940455 -2.3025850929940455; ...; 2.4932054526026954 3.7864597824528
    ... 2.7788192719904172 2.4932054526026954; 4.883559211528279 6.499937405290376 ...
     4.11251186617755 4.160444363926624], [1.0, -1.0, -1.0, -1.0, -1.0, 1.0,
     1.0, 1.0, -1.0 ... -1.0, -1.0, -1.0, 1.0, 1.0, -1.0, 1.0, 1.0, 1.0]
     [-2.3025850929940455 \ -2.3025850929940455 \ \dots \ -0.030459207484708574
    -1.3862943611198906; -2.3025850929940455 -2.3025850929940455 ...
    -1.3093333199837622 -1.6094379124341003; ...; 2.4069451083182885
     1.9600947840472698 ... 4.763028270603671 3.893859034800475; 3.7864597824528
    2.7788192719904172 ... 8.159975242934362 6.933520486868163], [1.0, -1.0, -1.0,
     -1.0, -1.0, -1.0, -1.0, 1.0, -1.0, 1.0 ... -1.0, -1.0, 1.0, -1.0, 1.0,
    -1.0, -1.0, 1.0, -1.0
[4]: f = LogRegProblem(X_test,Y_test)
     f(w) = HW3. (f, w)
     prob = DifferentiableProblem(f, f)
     nest_solver = NesterovDescentSolver(
          = 1e-2
          = 0.0.
         max_iter = 10^4,
         linesearch = BackTrackingLineSearch(),
```

```
w_opt_nest, hist_nest = HW4.solve(nest_solver, prob, zeros(size(X_test, 1)));
[5]: gd_solver = GradientDescentSolver(
          = 1e-3,
          = 1e-10,
         max_iter = 10^4,
         linesearch = BackTrackingLineSearch(),
     w_opt_gd, hist_gd = HW3.solve(gd_solver, prob, zeros(size(X_test, 1)));
[6]: res = optimize(f, zeros(size(X_test, 1)), NelderMead(),
         Optim.Options(iterations=10_000, show_trace=false, store_trace=true)
     )
[6]: * Status: failure (reached maximum number of iterations)
      * Candidate solution
         Final objective value:
                                    4.497449e+00
      * Found with
         Algorithm:
                        Nelder-Mead
      * Convergence measures
         \sqrt{(\Sigma(y-\bar{y})^2)/n}
                        1.0e-08
      * Work counters
         Seconds run:
                        0 (vs limit Inf)
         Iterations:
                       10000
         f(x) calls:
                       13724
[7]: plot(
         hist_nest.f, yscale=:log10,
         label="nesterov descent", lw=2,
         xlabel="Iteration", ylabel="f(x)",
         ylims=(10^(floor(log10(last(hist_nest.f)))),Inf), yminorgrid=true)
     plot!(hist_gd.f, label="gradient descent", lw=2)
     plot!(getfield.(res.trace, :value), label="nelder-mead", lw=2)
[7]:
```

2

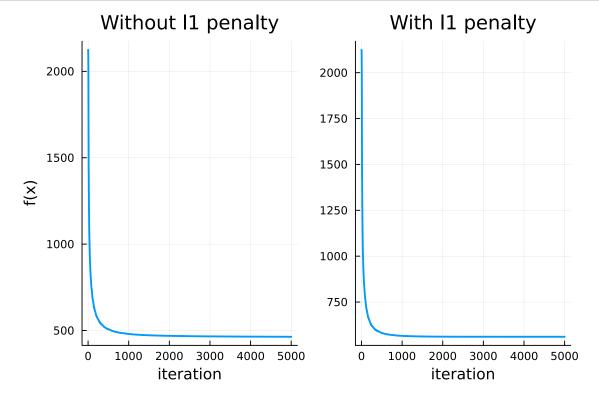


Problem 2

March 17, 2023

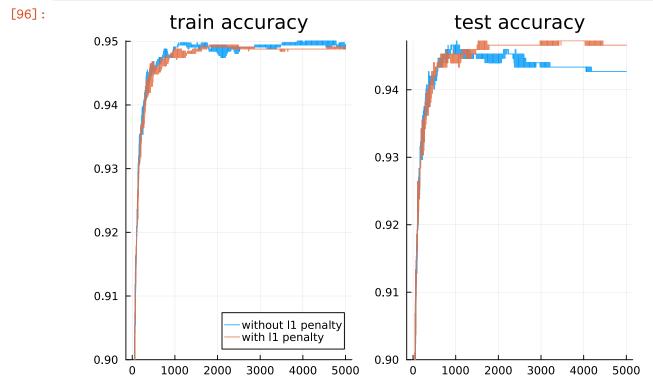
```
[73]: using Plots
      using AdvConvex.HW3
      using AdvConvex.HW4
      using Optim
      using LinearAlgebra
      \# NOTE: nbconvert doesn't render convenient unicode stuff like lambdas and \sqcup
       \neg nablas
[74]: mat = get_spam_data()
      X_train, Y_train, X_test, Y_test = train_test_split(mat, 0.334);
[90]: f = LogRegProblem(X_train,Y_train)
      f(x) = HW3. (f,x)
      prob = DifferentiableProblem(f, f)
      solver = GradientDescentSolver(
           = 1e-4,
           = 0.0,
          max_iter=5_000,
          linesearch = BackTrackingLineSearch()
      w_opt1, hist1 = solve(solver, prob, zeros(size(X_train, 1)));
[91]: = 5.0
      1 = PenaltyLogRegProblem(f, )
      g(1::PenaltyLogRegProblem,w) = 1.logreg(w)
      g(1,w) = HW3. (1.logreg, w)
      h(1::PenaltyLogRegProblem,w) = 1. * norm(w, 1)
      loss(1::PenaltyLogRegProblem, w) = g(1,w) + h(1,w)
      prox_th(1::PenaltyLogRegProblem, t, y) = sign(y)*max(abs(y) - t*1., 0.0)
      p = ProximalProblem(
          w -> loss(1, w),
          w \to g(1, w),
          (y,t) -> HW4.prox_th(1,t,y)
```

[92]:



```
train_acc1 = map(hist1.x) do x
    HW3.accuracy(x, X_train, Y_train)
end

train_acc2 = map(hist2.x) do x
    HW3.accuracy(x, X_train, Y_train)
end;
```



Problem 3

March 17, 2023

```
[1]: import cvxpy as cvx
     import numpy as np
     import imageio.v3 as iio
     import scipy
     import random
     import matplotlib.pyplot as plt
[2]: Y = iio.imread('SheppLogan_150x150.png')
     Y = Y / Y.max()
    n1, n2 = Y.shape
[3]: orig_shape = Y.shape
     flat = Y.flatten()
     n_mutated = len(flat) // 10
     rand_idxs = random.sample(range(0,len(flat)-1), n_mutated)
     flat[rand_idxs] += np.random.rand(n_mutated).astype(np.float32)
     Y_noisy = flat.reshape(orig_shape)
[4]: D = np.zeros(Y.shape)
     for i in range(n1-1):
         D[i,i] = -1
         D[i, i+1] = 1
     D[n1-1, n2-1] = -1
     Lh_tilde = scipy.sparse.kron(D, np.identity(n1))
     Lv_tilde = scipy.sparse.kron(np.identity(n2), D)
[5]: def TV(X):
        X = X.flatten()
         y_h = Lh_tilde @ X
         y_v = Lv_tilde @ X
         y = np.vstack((y_h, y_v))
         return np.sum(np.linalg.norm(y, 2, axis=0))
     def TV_cvx(X):
        X = X.flatten()
         y_h = Lh_tilde @ X
```

```
y_v = Lv_tilde @ X
y = cvx.vstack((y_h, y_v))
return cvx.sum(cvx.norm(y, 2, axis=0))

tau = 0.25*TV(Y_noisy)
```

```
[6]: X = cvx.Variable((n1,n2))
  objective = cvx.Minimize(0.5*cvx.norm(Y_noisy-X,'fro'))
  constraints = [0 <= X, X <= 1, TV_cvx(X) <= tau]
  prob = cvx.Problem(objective, constraints)
  result = prob.solve()
  Y_pred = X.value</pre>
```

```
fig, axes = plt.subplots(1,3,figsize=(15,7))
axes[0].imshow(Y_noisy, cmap='gray')
axes[0].axis('off')
axes[0].set_title("Noisy")
axes[1].imshow(Y_pred, cmap='gray')
axes[1].axis('off')
axes[1].set_title("De-noised")
axes[2].imshow(Y, cmap='gray')
axes[2].axis('off')
axes[2].set_title("True")
plt.show()
```

