CSE 275 HW0

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TOTAL POINTS

13.5 / 20

QUESTION 1

1 Report 13.5 / 20

- \checkmark 2 pts -1: two cases for 2.1.1 (t/(1-b) can be greater then 1)
- -1: wrong answer for 2.1.2
 - 0.5 pts Wrong Answers for 1.4
 - 0 pts All correct.
- **20 pts** Please review your submission and provide a high-resolution version. You can request a regrade then.
- √ 1 pts Final Segmentation Quality
 - 1 pts 2.3: Wrong Answer
 - 1.5 pts 2.1: Wrong Answer
 - **1 pts** 1.2.1: Miss Case
 - 1 pts 2.2: pdf > 0
 - 1 pts Please give the mathematical proof.
- 3.5 Point adjustment

1 Report 13.5 / 20

- $\sqrt{-2 \text{ pts}}$ -1: two cases for 2.1.1 (t/(1-b) can be greater then 1)
- -1: wrong answer for 2.1.2
 - **0.5 pts** Wrong Answers for 1.4
 - 0 pts All correct.
- **20 pts** Please review your submission and provide a high-resolution version. You can request a regrade then.
- √ 1 pts Final Segmentation Quality
 - 1 pts 2.3: Wrong Answer
 - **1.5 pts** 2.1: Wrong Answer
 - 1 pts 1.2.1: Miss Case
 - 1 pts 2.2: pdf > 0
 - 1 pts Please give the mathematical proof.
- **3.5** Point adjustment

HW0

October 6, 2023

1 Homework 0

print(err)

1.1 Problem 1

```
1. Gradient of Lagrangian
     \nabla xL = AT(Ax-b) + 2 \times \lambda
        2. Unconstrained least square
     x = (ATA)-1ATb
     3.a
     x = (ATA+2 \lambda I)-1(ATb)
     3.b
     (your work here)
        4. Implement
 [1]: import numpy as np
      npz = np.load('HWO_P1.npz')
      A = npz['A']
      b = npz['b']
      eps = npz['eps']
      A.shape, A.dtype, b.shape, b.dtype, eps
 [1]: ((100, 30), dtype('float64'), (100,), dtype('float64'), array(0.5))
[19]: def solve(A, b, eps):
           # your implementation here
          errLimit, err = 0.00001, 1
          lambda1, lambda2 = 0, 1
          while (abs(err) > errLimit):
               lambdaMid = (lambda2 + lambda1)/2
               x = (np.linalg.inv(A.T @ A + 2*lambdaMid*np.identity(A.shape[1])) @ A.T_{\sqcup}
       →@ b)
               err = x@x-eps
               print("err:")
```

```
if (err < 0):
    lambda2 = lambdaMid
    else:
        lambda1 = lambdaMid
    # prod = prod @ prod
    return x

[20]: # Evaluation code, you need to run it, but do not modify
    x = solve(A,b,eps)
    print('x norm square', x@x) # x@x should be close to or less then eps
    print('optimal value', ((A@x - b)**2).sum())

err:
    0.040151344139552236
err:
    0.009837342359711232</pre>
```

-0.004181909633978131

0.002738827985615422

-0.0007433488978196845

0.0009922365769245989

0.00012307451577187045

-0.00031047872407591326

-9.378758695816058e-05

1.4622081245474483e-05

-3.958809708864974e-05

-1.248434417422395e-05

1.0685344478122971e-06

x norm square 0.5000010685344478 optimal value 17.220125343130423

err:

err:

err:

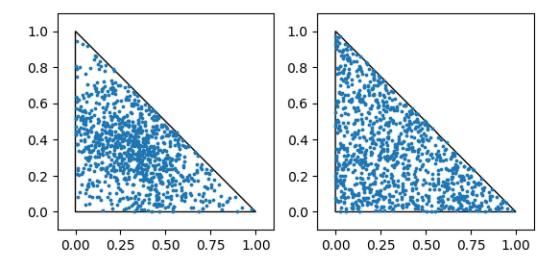
1.2 Problem 2

(2.2) Your proof here

```
[35]: import matplotlib.pyplot as plt
      from matplotlib.patches import Polygon
      import numpy as np
      pts = np.array([[0,0], [0,1], [1,0]])
      def draw_background(index):
          # DRAW THE TRIANGLE AS BACKGROUND
          p = Polygon(pts, closed=True, facecolor=(1,1,1,0), edgecolor=(0, 0, 0))
          plt.subplot(1, 2, index + 1)
          ax = plt.gca()
          ax.set_aspect('equal')
          ax.add_patch(p)
          ax.set xlim(-0.1,1.1)
          ax.set_ylim(-0.1,1.1)
      def incorr_pts():
          # incorrect method:
          alpha, beta, gamma = np.random.random(1000), np.random.random(1000), np.
       →random.random(1000) # sample in uniform distribution
          alpha, beta, gamma = alpha / (alpha + beta + gamma), beta / (alpha + beta +_{\sqcup}
       →gamma), gamma / (alpha + beta + gamma)
          # print(alpha.shape, alpha.T.shape, pts[0].shape)
```

```
# print(beta.shape, pts[1].shape)
    # print(qamma.shape, pts[2].shape)
    # print(beta)
   return np.outer(pts[0],alpha) + np.outer(pts[1],beta) + np.
 →outer(pts[2],gamma)
def corr_pts():
   # correct method:
   alpha, beta = np.random.random(1000), np.random.random(1000) # sample in_
 →uniform distribution
    # print(alpha.shape, alpha.T.shape, pts[0].shape)
    # print(beta.shape, pts[1].shape)
   pprime = np.outer(pts[0], np.ones(1000)) + np.outer((pts[1]-pts[0]),alpha)
 h np.outer((pts[2]-pts[0]),beta)
   # print(pprime.shape)
   pprimeSum = np.sum(pprime, axis=0)
    # print(pprimeSum)
   pprimeOut = pprime[:, pprimeSum > 1]
   print("The number of points outside the triangle:", pprimeOut.shape[1])
   pprime[:, pprimeSum > 1] = np.outer(pts[1], np.ones(pprimeOut.shape[1])) +__
 anp.outer(pts[2], np.ones(pprimeOut.shape[1])) - pprime[:, pprimeSum > 1]
   return pprime
# YOUR CODE HERE
incorrPts = incorr_pts()
corrPts = corr_pts()
draw_background(0)
# REPLACE THE FOLLOWING LINE USING YOUR DATA (incorrect method)
plt.scatter(incorrPts[0, :], incorrPts[1, :], s=3)
draw_background(1)
# REPLACE THE FOLLOWING LINE USING YOUR DATA (correct method)
plt.scatter(corrPts[0, :], corrPts[1, :], s=3)
plt.show()
```

The number of points outside the triangle: 511

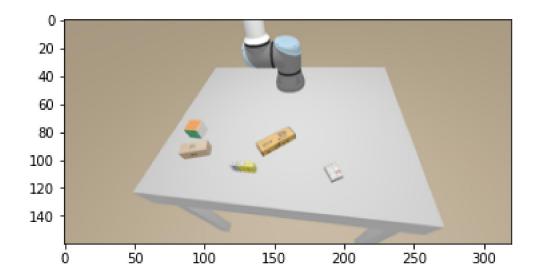


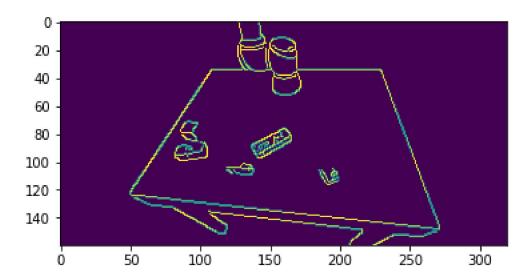
[]:

1.3 Problem 3

```
[]: import numpy as np
    npz = np.load("train.npz")
    images = npz["images"] # array with shape (N, Width, Height, 3)
    edges = npz["edges"] # array with shape (N, Width, Height)
[]: plt.figure()
    plt.imshow(images[0])
    plt.figure()
    plt.figure()
    plt.imshow(edges[0])
```

[]: <matplotlib.image.AxesImage at 0x7fd1d7d9ac10>





```
[]: images.shape, edges.shape, images.max(), np.unique(edges)

[]: ((1000, 160, 320, 3), (1000, 160, 320), 255, array([ 0, 255], dtype=uint8))

[]: # Build and train your neural network here, optionally save the weights

[]: # Test on the testing set import numpy as np import numpy as np import matplotlib.pyplot as plt npz = np.load("test.npz") test_images = npz["images"]
```

```
plt.figure(figsize=(10, 10))
for i, img in enumerate(test_images[:4]):
    plt.subplot(4, 2, i * 2 + 1)
    plt.imshow(img)

plt.subplot(4, 2, i * 2 + 2)
    # edge = evaluate your model on the test set, replace the following line
    edge = np.zeros(img.shape[:2])
    plt.imshow(edge)
```

