## **4B**

import numpy as np import scipy.spatial

```
def generate\_sensors(k = 7, d = 2):
     Generate sensor locations.
     Input:
     k: The number of sensors.
     d: The spatial dimension.
     Output:
     sensor_loc: k * d numpy array.
     sensor_loc = 100*np.random.randn(k,d)
     return sensor loc
def generate data(sensor loc, k = 7, d = 2,
                       n = 1, original_dist = True):
     Generate the locations of n points.
 Input:
     sensor loc: k * d numpy array. Location of sensor.
     k: The number of sensors.
     d: The spatial dimension.
     n: The number of points.
     original_dist: Whether the data are generated from the original
     distribution.
     Output:
     obj_loc: n * d numpy array. The location of the n objects.
     distance: n * k numpy array. The distance between object and
     the k sensors.
     assert k, d == sensor loc.shape
     obj_loc = 100*np.random.randn(n, d)
     if not original dist:
       obj_loc += 1000
     distance = scipy.spatial.distance.cdist(obj loc,
                                                          sensor_loc,
```

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metric='euclidean')
                  distance += np.random.randn(n, k)
                  return obj. loc, distance
# Starter code for Part (b)
import math
def compute gradient(distance, loc, sensor loc):
      This function computes the gradient at a particular point
      total_x = 0
      total y = 0
      x1 = loc[0]
      y1 = loc[1]
      for i in range(7):
            ai = sensor_loc[i][0]
            bi = sensor loc[i][1]
            di = distance[i]
            total_x += 2 * (math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)/math.sqrt((ai - x1))/math.sqrt((ai - x1))/
(bi - y1)**2)
      for i in range(7):
            ai = sensor_loc[i][0]
            bi = sensor loc[i][1]
            di = distance[i]
            total_y += 2 * (math.sqrt((ai - x1)**2 + (bi -y1)**2) - di) * (bi - y1)/math.sqrt((ai - x1)**2 +
(bi - y1)**2)
      gradient = np.array([total x,total y]) * -1
      return gradient
def compute update(distance, current loc, sensor loc, step count, step size):
      """Computes the new point after the update at x."""
      return current_loc - step_size(step_count) * compute_gradient(distance, current_loc,
sensor loc)
def compute_updates(distance, p, sensor_loc, total_step_count, step_size):
      """Computes several updates towards the minimum of ||Ax-b|| from p.
      Params:
            b: in the equation ||Ax-b||
            p: initialization point
```

```
total_step_count: number of iterations to calculate
     step_size: function A, b,for determining the step size at step i
  positions = [np.array(p)]
  for k in range(total_step_count):
     positions.append(compute_update(distance, positions[-1], sensor_loc, k, step_size))
  return np.array(positions)
np.random.seed(0)
sensor_loc = generate_sensors()
obj loc, distance = generate data(sensor loc)
single_distance = distance[0]
total_step_count = 1000
step_size = lambda i: .05
initial position = np.array([0, 0])
positions = compute_updates(single_distance, initial_position, sensor_loc, total_step_count,
step_size)
#compute gradient(single distance, np.array([0, 0]), sensor loc)
print(positions)
initial_position = np.random.rand(2)*100
positions = compute_updates(single_distance, initial_position, sensor_loc, total_step_count,
step_size)
#compute_gradient(single_distance, np.array([0, 0]), sensor_loc)
print(positions)
```

```
import numpy as np import scipy.spatial
```

```
def generate sensors(k = 7, d = 2):
    Generate sensor locations.
    Input:
    k: The number of sensors.
    d: The spatial dimension.
    Output:
    sensor_loc: k * d numpy array.
    sensor loc = 100*np.random.randn(k,d)
    return sensor loc
def generate_data_given_location(sensor_loc, obj_loc, k = 7, d = 2):
    Generate the distance measurements given location of a single object and sensor.
    Input:
    obj loc: 1 * d numpy array. Location of object
    sensor loc: k * d numpy array. Location of sensor.
    k: The number of sensors.
    d: The spatial dimension.
    Output:
    distance: 1 * k numpy array. The distance between object and
    the k sensors.
    assert k, d == sensor loc.shape
    distance = scipy.spatial.distance.cdist(obj_loc,
                         sensor loc.
                         metric='euclidean')
    distance += np.random.randn(1, k)
    return obj loc, distance
# Starter code for Part (b)
```

```
import math
def compute_gradient(distance, loc, sensor_loc):
      This function computes the gradient at a particular point
      total x = 0
      total y = 0
      x1 = loc[0]
      y1 = loc[1]
      for i in range(7):
            ai = sensor_loc[i][0]
            bi = sensor_loc[i][1]
            di = distance[i]
            total_x += 2 * (math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2 + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)**2) - di) * (ai - x1)/math.sqrt((ai - x1)**2) + (bi - y1)/math.sqrt((ai - x1)**2) + (bi - y1)/math.sqrt((ai - x1))/math.sqrt((ai - x1))/math.
(bi -y1)**2)
      for i in range(7):
            ai = sensor_loc[i][0]
            bi = sensor_loc[i][1]
            di = distance[i]
            total_y += 2 * (math.sqrt((ai - x1)**2 + (bi -y1)**2) - di) * (bi - y1)/math.sqrt((ai - x1)**2 +
(bi -y1)**2)
      gradient = np.array([total x,total y]) * -1
      return gradient
def compute_update(distance, current_loc, sensor_loc, step_count, step_size):
      """Computes the new point after the update at x."""
      return current_loc - step_size(step_count) * compute_gradient(distance, current_loc,
sensor_loc)
def compute_updates(distance, p, sensor_loc, total_step_count, step_size):
      """Computes several updates towards the minimum of ||Ax-b|| from p.
      Params:
            b: in the equation ||Ax-b||
            p: initialization point
            total_step_count: number of iterations to calculate
            step_size: function A, b,for determining the step size at step i
      positions = [np.array(p)]
      for k in range(total_step_count):
            positions.append(compute_update(distance, positions[-1], sensor_loc, k, step_size))
      return np.array(positions)
```

```
# Sensor locations.
num_gd_replicates = 100
obj_locs = [[[i,i]] for i in np.arange(0,1000,100)]
np.random.seed(100)
total_step_count = 1000
step_size = lambda i: .05
est_pos = [list()] * 10
for k, obj_loc in enumerate(obj_locs):
  for j in range(num_gd_replicates):
     sensor_loc = generate_sensors()
     obj_loc, distance = generate_data_given_location(sensor_loc, obj_loc, k = 7, d = 2)
     single_distance = distance[0]
     initial_position = (obj_loc[0][0]+1)*np.random.randn(2)
     positions = compute_updates(single_distance, initial_position, sensor_loc,
total_step_count, step_size)
     print(positions[-1])
#for pos in est_pos:
# print(pos)
```