

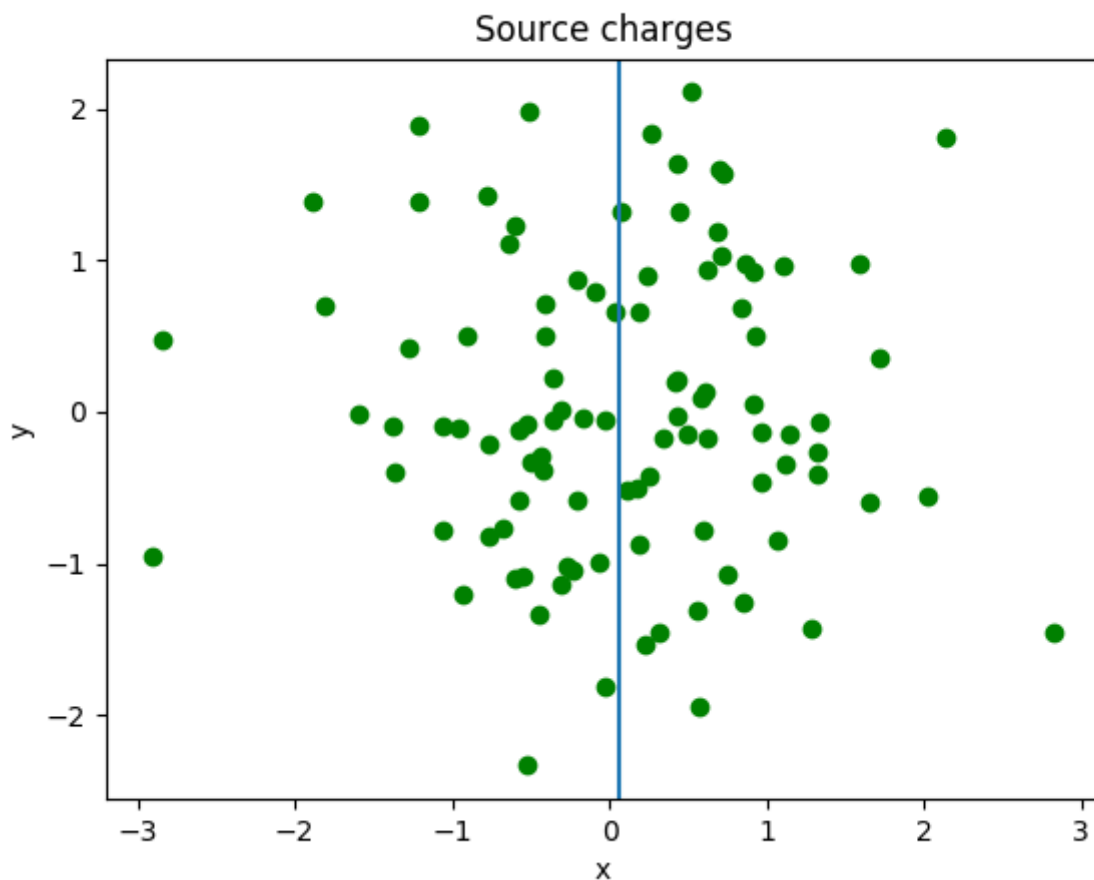
# Semiseparable Matrices

10 points

The electric potential from a point charge,  $Q$ , at a distance  $r$  from a charge is given as

$$V = \frac{1}{4\pi\epsilon_0} \frac{Q}{r}.$$

Now, consider computing the potential from a collection of point charges,  $q_1, q_2, \dots, q_n$  at the locations of each of the point charges. Suppose that the point charges are numbered in such a way that particles  $q_1, q_2, \dots, q_j$  are left of the midpoint, and that  $q_{j+1}, \dots, q_n$  are to the right of the midpoint of the particles.



The interaction matrix comprised that computes the potential at  $q_1, \dots, q_n$  given a vector of charges  $Q_1, \dots, Q_n$  has a special structure known as a *semiseparable* (or SS for short). SS matrices show up in a variety of different applications from computing potentials, to PDE solvers, to multivariate statistics in the form of the covariance matrix.

In this problem, you will solve a linear system with an SS matrix of the following form using the Sherman-Morrison formula

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}.$$

In general, semiseparable matrices have diagonal blocks which have full rank and off-diagonal blocks which have (approximately) low rank. In this particular problem, we assume that the SS matrix has off-diagonal blocks with rank 1 (i.e. an outer product of two vectors). In this particular problem,  $A_{11}$  is a  $n \times n$  matrix, and  $A_{22}$  is a  $m \times m$  matrix.  $x_1$  and  $b_1$  are length  $n$  vectors.  $x_2$  and  $b_2$  are length  $m$  vectors.  $A_{12} = \mathbf{u}\mathbf{v}^T$  where  $\mathbf{u}$  is a length  $n$  vector and  $\mathbf{v}$  is a length  $m$  vector. Further,  $A_{21} = \mathbf{w}\mathbf{p}^T$ , where  $\mathbf{w}$  is a length  $m$  vector and  $\mathbf{p}$  is a length  $n$  vector.

### What do I need to do?

1. Using Gaussian-Elimination on the block-structure of the matrix and applying the Sherman-Morrison formula, solve the linear system for  $x_1$  and  $x_2$ .
2. Solving a semiseparable system in this fashion has some performance benefits. In a comment block, please describe the expected cost, in terms of  $n$  and  $m$ , for solving the entire system without taking advantage of the semiseparable structure of the matrix, such as by calling `la.solve`.
3. In another comment block, specify the expected cost, in terms of  $n$  and  $m$ , for solving the system by taking advantage of the semiseparable structure of the matrix.

**NOTE:** You may assume that the cost of solving a generic  $p \times p$  system is  $cost \approx Cp^3$ , where  $C$  is some constant that does not depend on the problem size. (You may also find a detailed operation count for LU in our textbook.) You need not specify  $C$ , you may simply use it as a known constant. For your cost analysis, please only consider the leading order terms, e.g., you may simplify  $2nm^2 + 3n^2m + n^3 + m^3 + 457nm + 1589n$  as  $2nm^2 + 3n^2m + n^3 + m^3$ .

INPUT:

- `A11` : A numpy array of the top left  $n \times n$  block of  $A$ .
- `A11_inv` : A numpy array of the inverse of the  $A_{11}$  block of  $A$ .
- `A22` : A numpy array of the bottom right  $m \times m$  block of  $A$ .
- `A22_inv` : A numpy array of the inverse of the  $A_{22}$  block of  $A$ .
- `u`, `p` : numpy arrays of size  $n$ .
- `v`, `w` : numpy arrays of size  $m$ .
- `b` : A numpy array of size  $n + m$  with the right-hand side of the linear system.

OUTPUT:

- `x1` : A numpy array of size  $n$  containing the first  $n$  entries of the solution.
- `x2` : A numpy array of size  $m$  containing the last  $m$  entries of the solution.

**Problem set-up code** (click to view)

**Starter code** (click to view)

**Answer\***

```
38 #2: B1 = b1 - A12A22-1*b2
39 B1 = b1 - np.dot(u,np.dot(v.T,np.dot(A22_inv,b2)))
40
41 #3: yy = A11-1 * B
42 yy = np.dot(A11_inv,B1)
```

```

43
44 #4: x1 = y + ((pT y)/(1 - pT z))z
45 denom = 1 - np.dot(p.T, zz)
46 pty = np.dot(p.T, yy)
47 x1 = yy + np.dot(pty/denom, zz)
48
49 '''
50 To solve x1 x2 directly we may need to get the inverse of A(n+m)*(n+m).
51 So that would be O((n+m)^3). Then we multiply A^-1 with b which give us O
52 The total leading term would be O(n^3+m^3+3nm^2+3n^2m)
53 '''
54 '''
55 If we take advantage of the semiseperarable matrix, the above four steps c
56 #1:m^2+m+n+n^2
57 #2:m^2+m+n
58 #3:n^2
59 #4:n+n+m+n

```

Press F9 to toggle full-screen mode. Set editor mode in user profile (/profile/).

#### Overall grade

**The overall grade is 100%.**

#### Autograder feedback

**The autograder assigned 9/9 points.**

**Your answer is correct.**

Here is some feedback on your code:

- 'x1' looks good
- 'x2' looks good
- Execution time: 0.3 s -- Time limit: 10.0 s

Your code ran on relate-02.cs.illinois.edu.

#### Human feedback

**The human grader assigned 1/1 points.**

The following code is a valid answer:

```
import numpy as np
#split b into b1 and b2
n = len(u)
m = len(v)
b1 = b[:n]
b2 = b[n:]

b2 -= np.inner(p,A11_inv@b1)*w
z=A22_inv@w
y=A22_inv@b2
my_v = np.inner(p,A11_inv@u)*v
x2 = y+(np.inner(my_v,y)/(1-np.inner(my_v,z)))*z
x1 = b1 - np.dot(v,x2) * u
x1 = A11_inv@x1
```