## NUMERICAL METHODS ECSE 543 - ASSIGNMENT 1

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### QUESTION 1

### Part a. The Choleski implementation is provided in Listing 2.

Code structure. To maintain portability and modularity of the code, object oriented principles were used for the software architecture. The choleski implementation is included in the CholeskiDecomposition() class. The solve(A, b) method solves the linear system of equations shown in equation (1) by performing choleski elimination.

$$Ax = b$$

The method accepts the matrix A, and the vector b (both of which will eventually be overwritten by the algorithm in order to conserve memory resources), and returns the vector x corresponding to the solution of equation (1). The algorithm works in two stages. The first stage performs a choleski factorization of A into  $LL^T$  (overwriting the lower triangular part of A by L), while simultaneously solving lower triangular system Ly = b using forward substitution (overwriting b with the solution y). At the end of this stage, the program state now contains L in the lower triangular half of the matrix A, and the solution to Ly = b in the vector b. In the second stage the program solves the system  $L^Tx = y$  using backwards substitution (overwriting y again with the solution x), where y is the solution to the system solved in the first stage. The program subsequently returns the vector x, which is the solution to equation (1).

Part b. For testing purposes, it was necessary to create a symmetric positive definite matrix. Such a matrix was created using the generate\_positive\_semidef(order, seed) method contained in the utils file in Listing 1. Given an order (the dimension of the desired matrix), and an integer valued seed (used to seed the random number generator with a standard normal distribution), the function creates a random matrix, multiplies it by its transpose, and returns the result. The mathematical proof for why such a matrix is symmetric positive definite is well established. Whether or not the matrix is singular in this semidefinite method is important, and this is being checked by comparing the rank of the matrix to its order. If the rank of the matrix is not equal to the order of the matrix, then the matrix is singular and a warning is printed to the console. Note that this check still does not prevent the matrix from having a poor condition number.

Date: October 16, 2016.

Part c. The testing of the choleski implementation was conducted using the code provided under the main() method in Listing 2 lines 90-111. The vector  $x^*$ , corresponding to the variable x in equation (1), is randomly generated with a standard normal distribution, and subsequently multiplied by the matrix A in order to generate a third vector b (i.e  $b = A \cdot x^*$ ). The matrix A and the vector b are subsequently supplied to the solver, and the result is compared with the vector  $x^*$  that was originally used to create b. A sample of the console output is provided in Figure 1 - the matrix A is of order 10 in this example. The error in the produced result is quantified using the 2-norm:

$$error = ||solve(A, b) - x^*||_2$$

As is seen in the console output, the error is only  $2 \cdot 10^{-13}$ , indeed the algorithm is producing the correct result. A possible reason for such a value of the error could be the roundoff error related to the condition of the randomly generated matrices.

Figure 1. Choleski Elimination Testing

**Part d.** A program used to solved tor the node voltages in a linear resistive network is provided in **Listing 3**. The *LinearResistiveNetworkSolver()* class is initialized with a filename from which to read the circuit description. The program, in the intializer, reads

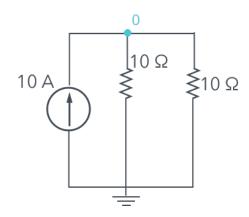
a list of network branches  $(J_k, R_k, E_k)$  and a reduced incidence matrix from a CSV file. The format of the file is as follows: a set of rows (corresponding to each branch in the network), containing the comma separated branch current, resistance, and voltage in that respective order. Then a period is printed on a new line, to signify the end of the network data. The subsequent comma separated rows denote the incidence matrix, where each row corresponds to a node, and each column to a branch. An entry of -1 is used to indicate current entering a branch, 1 is used to indicate current leaving a branch, and 0 is used to indicate that the branch does not interact directly with the given node. The program reads the data in the file sequentially (i.e first the rows of the branch data are read, and then the rows of the incidence matrix are read). Once the data is read, the program subsequently generates a linear system of equations using the aforementioned data, and solves the system via choleski elimination.

**Test Circuits.** Test circuit CSV descriptions (used to test the program), and their equivalent circuit diagrams and corresponding console outputs are shown below. In each case, the console output was consistent with the analytical results obtained by hand.

# $Test\ Circuit\ 1$

```
test_c1.csv
0, 10, 10
0, 10, 0
.
-1, +1
```

```
test_c2.csv
-10, 10, 0
0, 10, 0
.
-1, 1
```



```
midoassran@Midos-MacBook-Pro:~/documents/McGill/U(4)/ECSE 543/Assignment_1$ python lrn_solver.py

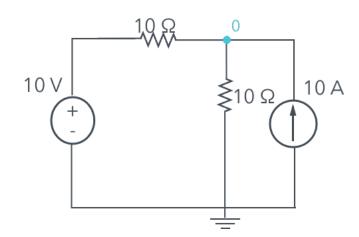
# ------ TEST ------ #
# ------ Manual CSV Data ------ #
# ------ Manual CSV Data ------ #
# ------ #
Cxecution time:
4.49499930255115e-05

Voltages:
Node 0: 50.0 Volts

midoassran@Midos-MacBook-Pro:~/documents/McGill/U(4)/ECSE 543/Assignment_1$ __
```

# $Test\ Circuit\ \mathcal{3}$

# test\_c3.csv 0, 10, 10 -10, 10, 0 . -1, -1



```
midoassran@Midos-MacBook-Pro:~/documents/McGill/U(4)/ECSE 543/Assignment_1$ python lrn_solver.py

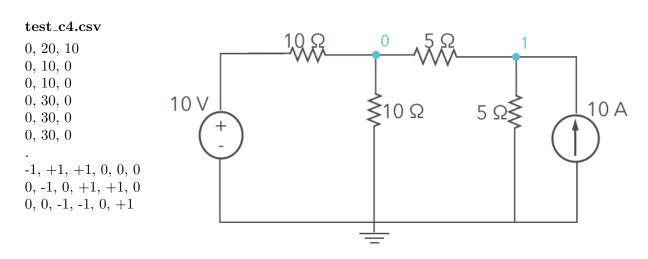
# ------ TEST ------ #
# ------ Manual CSV Data ------ #
# ------ Manual CSV Data ------ #
# ------ Mover.py

Execution time:
2.130295615643263e-05

Voltages:
Node 0: 55.0 Volts

midoassran@Midos-MacBook-Pro:~/documents/McGill/U(4)/ECSE 543/Assignment_1$ __
```

## Test Circuit 4



```
midoassran@Midos-MacBook-Pro:~/documents/McGill/U(4)/ECSE 543/Assignment_1$ python lrn_solver.py

# ------ TEST ----- #
# ------ Linear Resistive Network Solver ----- #
# ------ Manual CSV Data ----- #
# ------ #
Execution time:
    2.855702769011259e-05

Voltages:
    Node 0: 20.0 Volts
    Node 1: 35.0 Volts

midoassran@Midos-MacBook-Pro:~/documents/McGill/U(4)/ECSE 543/Assignment_1$ []
```

# $Test\ Circuit\ 5$

# 20 Ω 0 $test\_c5.csv$ 0, 20, 10 10 Ω 10 Ω 10 V 0, 10, 00, 10, 0 0, 30, 0 30 Ω 0, 30, 00, 30, 0 -1, +1, +1, 0, 0, 030 Ω 0, -1, 0, +1, +1, 00, 0, -1, -1, 0, +130 Ω

### QUESTION 2

Part a. To find the resistance across two diagonally opposing corners of a linear resistive N by N finite different mesh, the linear resistive network solver, provided in Listing 3, was used. This is the same program that was used in Question 1. The static method  $create\_lrn\_mesh\_data(N, fname)$  accepts an integer, N, denoting the size of the mesh, and a filename, to which a CSV description of the created mesh should be saved. It should be noted that this method also includes in the circuit description a test source placed across the diagonal of the mesh. This test source has a voltage of 1V, and an output resistance of  $1\Omega$ . The main() method in Listing 3 - lines 166-177 - calls the appropriate methods to create the resistive finite difference mesh, and subsequently solve for all the node voltages. Once all the node voltages are known, the voltage difference between the two corners of the mesh is used to construct a simple voltage division equation that is used to solver for the equivalent resistance of the mesh.

```
# Utils
  # Author: Mido Assran
5 # Date: 5, October, 2016
  # Description: Utils provides a cornucopia of useful matrix
7 # and vector helper functions.
9 import random
  import numpy as np
  def matrix_transpose(A):
13
      :type A: np.array([float])
       :rtype: np.array([floats])
15
17
      # Initialize A_T(ranspose)
      A_T = \text{np.empty}([A. \text{shape}[1], A. \text{shape}[0]])
19
      # Set the rows of A to be the columns of A_T
21
       for i, row in enumerate(A):
           A_T[:, i] = row
23
       return A<sub>-</sub>T
25
27
  def matrix_dot_matrix(A, B):
29
       :type A: np.array([float])
       :type B: np.array([float])
31
      :rtype: np.array([float])
33
      # If matrix shapes are not compatible return None
35
       if (A. shape [1] != B. shape [0]):
           return None
37
```

```
A_{dot_B} = np.empty([A.shape[0], B.shape[1]])
39
      A_{dot_B}[:] = 0 # Initialize entries of the new matrix to zero
41
      B_T = matrix_transpose(B)
43
      for i, row_A in enumerate(A):
           for j, column_B in enumerate(B_T):
45
               for k, v in enumerate(row_A):
                   A_{dot_B}[i, j] += v * column_B[k]
47
49
      return A_dot_B
51
  def matrix_dot_vector(A, b):
53
      :type A: np.array([float])
      :type b: np.array([float])
55
      :rtype: np.array([float])
57
      # If matrix shapes are not compatible return None
59
      if (A. shape [1] != b. shape [0]):
           return None
61
      A_{dot_b} = np.empty([A.shape[0]])
63
      A_{-}dot_{-}b[:] = 0 # Initialize entries of the new vector to zero
65
      for i, row_A in enumerate(A):
           for j, val_b in enumerate(b):
67
               A_dot_b[i] += row_A[j] * val_b
69
      return A_dot_b
71
73 def vector_to_diag(b):
75
      :type b: np.array([float])
      :rtype: np.array([float])
77
```

```
diag_b = np.empty([b.shape[0], b.shape[0]])
79
       \operatorname{diag_b}[:] = 0
                         # Initialize the entries to zero
81
       for i, val in enumerate(b):
           diag_b[i, i] = val
83
       return diag_b
85
87 def generate_positive_semidef(order, seed=0):
       :type order: int
89
       :type seed: int
       :rtype: np.array([float])
91
93
       np.random.seed(seed)
       A = np.random.randn(order, order)
95
       A = matrix_dot_matrix(A, matrix_transpose(A))
97
       # TODO: Replace matrix_rank with a custom function
       from numpy.linalg import matrix_rank
99
       if matrix_rank(A) != order:
           print("WARNING: Matrix is singular!", end="\n\n")
101
       return A
103
```

Listing 1. utils.py

```
#
  # Choleski Decomposition
  # Author: Mido Assran
5 # Date: 30, September, 2016
  # Description: CholeskiDecomposition solves the linear system of equations:
7 # Ax = b by decomposing matrix A using Choleski factorization and using
  # forward and backward substitution to determine x. Matrix A must
9 # be symmetric, real, and positive definite.
11 import random
  import timeit
13 import numpy as np
  from utils import matrix_transpose
15
  DEBUG = True
17
  class CholeskiDecomposition (object):
19
      def __init__(self):
           if DEBUG:
21
               np.core.arrayprint._line_width = 200
23
      def solve (self, A, b):
25
          :type A: np.array([float])
          :type b: np.array([float])
27
           :rtype: np.array([float])
29
          start_time = timeit.default_timer()
31
          # If the matrix, A, is not square, exit
33
           if A. shape [0] != A. shape [1]:
               return None
35
37
          n = A. shape [1]
```

```
39
          # Simultaneous Choleski factorization of A and chol-elimination
41
          # Choleski factorization & forward substitution
43
          for j in range(n):
45
              # If the matrix A is not positive definite, exit
              if A[j,j] <= 0:
47
                  return None
49
                                      # Compute the j,j entry of chol(A)
              A[j,j] = A[j,j] ** 0.5
              b[j] /= A[j,j]
                                        # Compute the j entry of forward-sub
51
              for i in range (j+1, n):
55
                  A[i,j] /= A[j,j] # Compute the i, j entry of chol(A)
                  b[i] -= A[i,j] * b[j] # Look ahead modification of b
57
59
                   if A[i,j] == 0:
                                         # Optimization for matrix sparsity
                       continue
61
                  # Look ahead moidification of A
                   for k in range (j+1, i+1):
63
                      A[i,k] = A[i,j] * A[k,j]
65
67
          # Now solve the upper traingular system
69
71
          # Transpose(A) is the upper-tiangular matrix of chol(A)
          A[:] = matrix\_transpose(A)
73
          # Backward substitution
          for j in range (n-1, -1, -1):
75
              b[j] /= A[j,j]
77
```

```
for i in range(i):
                      b[i] -= A[i,j] * b[j]
79
81
             elapsed_time = timeit.default_timer() - start_time
83
             if DEBUG:
                  print("Execution time:\n", elapsed_time, end="\n\n")
85
             # The solution was overwritten in the vector b
87
             return b
89
   if __name__ == "__main__":
        from utils import generate_positive_semidef, matrix_dot_vector
        order = 10
93
        seed = 5
95
        print("\n", end="\n")

      print ("# — TEST — #", end="\n")

      print ("# — Choleski Decomposition — #", end="\n")

      print ("# — #", end="\n")

      #", end="\n")

      #", end="\n")

      #", end="\n\n")

97
99
        chol_d = CholeskiDecomposition()
101
        # Create a symmetric, real, positive definite matrix.
        A = generate_positive_semidef(order=order, seed=seed)
        x = np.random.randn(order)
103
        b = matrix_dot_vector(A=A, b=x)
        print ("A:\n", A, end="\n\n")
105
        print("x:\n", x, end="\n\n")
        print("b (=Ax): \n", b, end="\n\n")
107
        v = chol_d.solve(A=A, b=b)
        print("result = solve(A, b):\n", v, end="\n\n")
109
        print ("2-norm error:\n", np.linalg.norm(v - x), end="\n\n")
        print ("# -------#", end="\n\n")
111
```

Listing 2. choleski.py

```
# Linear Resistive Network Solver
  # Author: Mido Assran
5 # Date: 30, September, 2016
  # Description: LinearResistiveNetworkSolver reads a CSV description of
7 # a linear resistive network, and determines all the node voltages
  # of the circuit by constructing a linear system of equations,
9 # and solving the system using Choleski Decomposition.
11 import random
  import csv
13 import numpy as np
  from choleski import Choleski Decomposition
15 from utils import matrix_transpose, matrix_dot_matrix, matrix_dot_vector, vector_to_diag
17 DEBUG = False
19 class LinearResistiveNetworkSolver(object):
      #-----#
21
      \# A \rightarrow The matrix 'A' in the system of equations Ax = b
      \# _b \rightarrow The vector 'b' in the system of equations Ax = \_b
23
      def __init__(self, fname):
25
          :type fname: String
27
          :rtype: void
          " " "
29
          if DEBUG:
              np.core.arrayprint._line_width = 200
31
          #-----#
33
          # Program first reads branch data, then swtiches to reading the
          # incidence matrix. Flag goes high when the the program
35
          # swtiches to reading the incidence matrix.
37
          flag = False
          network_branches = []
```

```
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```

```
39
           incidence\_matrix = []
           reader = csv.reader(open(fname, 'r'))
           for row in reader:
41
               if len(row) = 1 and row[0] = ".":
43
                   flag = True
                   continue
               elif len(row) == 0:
45
                   continue
               if not flag:
47
                   network_branches += [list(row)]
49
               else:
                   incidence_matrix += [list(row)]
           network_branches = np.array(network_branches, dtype=np.float64)
51
           incidence_matrix = np.array(incidence_matrix, dtype=np.float64)
          J = network_branches [:, 0]
53
          Y = vector_to_diag(1 / network_branches[:, 1])
          E = network_branches [:, 2]
55
          A = matrix_dot_matrix (A=matrix_dot_matrix (A=incidence_matrix, B=Y),
                                  B=matrix_transpose(incidence_matrix))
57
          b = matrix_dot_vector(A=incidence_matrix,
                                 b=(J - matrix_dot_vector(A=Y, b=E)))
59
           self.A = A
           self._b = b
61
63
      def solve (self):
           :rtype: numpy.array([float64])
65
           chol_decomp = CholeskiDecomposition()
67
          # Choleski decomposition will overwrite A, and b
           return chol_decomp.solve(A=self._A, b=self._b)
69
71
      @staticmethod
      def create_lrn_mesh_data(N, fname):
73
75
          :type N: int
           :type fname: String
77
           :rtype: void
```

```
num\_nodes = (N + 1) ** 2
79
            num_branches = 2 * (N ** 2) + 2 * N + 1
            incidence_matrix = np.empty([num_nodes, num_branches])
81
            network_branches = np.empty([num_branches, 3])
            incidence_matrix[:] = 0
83
            network_branches[:] = 0
85
            for i, row in enumerate (network_branches):
                if i = (num\_branches - 1):
87
                    network_branches[i, :] = np.array([0, 1, 1])
                else:
89
                    network\_branches[i, :] = np.array([0, 1e3, 0])
91
           node_num = 0
93
           # Iterate through node rows of mesh
           for level in range (N + 1):
95
               # Iterate through node columns of mesh
97
                for column in range (N + 1):
99
                    # If the node has a left branch
                    if (node_num \% (N + 1) != 0):
101
                        left\_branch = node\_num + (level * N) - 1
                        incidence\_matrix[node\_num, left\_branch] = -1
                        if DEBUG:
                            print("L:", node_num, left_branch, end="\t")
105
                    # If the node has a right branch
107
                    if ((node_num + 1) \% (N + 1) != 0):
                        right_branch = node_num + (level * N)
109
                        incidence_matrix[node_num, right_branch] = 1
                        if DEBUG:
111
                            print("R:", node_num, right_branch, end="\t")
113
                    # If the node has a top branch
                    if (node_num < (num_nodes - (N + 1))):
115
                        top\_branch = node\_num + ((level + 1) * N)
```

```
117
                         incidence_matrix [node_num, top_branch] = 1
                         if DEBUG:
                             print("T:", node_num, top_branch, end="\t")
119
121
                    # If the node has a botom branch
                    if (node_num > N):
                        bottom\_branch = (node\_num - 1) + ((level - 1) * N)
123
                         incidence\_matrix[node\_num, bottom\_branch] = -1
                         if DEBUG:
125
                             print("B:", node_num, bottom_branch, end="\t")
127
                    if DEBUG:
                        print ("\n")
129
131
                    node_num += 1
           # Add the branch of the test source
133
            incidence_matrix[0, -1] = -1
            incidence_matrix[-1, -1] = 1
135
137
           # Write data to file fname.csv
            fwriter = csv.writer(open(fname, 'w'))
            for i, row in enumerate(network_branches):
139
                fwriter.writerow(row)
141
           # Write a period to separate network_branches from
           # the incidence_matrix
143
            fwriter.writerow(".")
145
            for i, row in enumerate (incidence_matrix):
                fwriter.writerow(row)
147
149
151
   if _-name_- == "_-main_-":
      # print("\n", end="\n")
      # print("# — #", end="\n")
# print("# — Linear Resistive Network Solver — #", end="\n")
```

```
# print("# — #", end="\n")
# print("# — #", end="\n\n")
         # lrn = LinearResistiveNetworkSolver("data/test_c1.csv")
        # voltages = lrn.solve()
159
        # print("Voltages:", end="\n")
         # for i, v in enumerate(voltages):
161
                 print(" Node", i, end=": ")
                print(v, "Volts", end="\n")
163
         \# print("\n", end="\n")
165

      print ("# — TEST — #", end="\n")

      print ("# — Linear Resistive Network Solver — #", end="\n")

      print ("# — Finite Difference Mesh — #", end="\n")

      print ("# — #", end="\n")

      #", end="\n")

      #", end="\n")

167
169
         new_fname = "data/test_save.csv"
        N = 15
171
         print ("Mesh size:\n", N, "x", N, end="\n")
         LinearResistiveNetworkSolver.create_lrn_mesh_data(N=N, fname=new_fname)
173
         lrn = LinearResistiveNetworkSolver(new_fname)
         voltages = lrn.solve()
175
         r_{eq} = (voltages[0] - voltages[-1]) / (1 - (voltages[0] - voltages[-1]))
         print("Resistance:\n", r_eq, "Ohms", end="\n\n")
```

Listing 3. lrn\_solver.py