## ECSE 543 - ASSIGNMENT 1

MIDO ASSRAN - 260505216

QUESTION 1

**a.** Refer to listing

 $Date \hbox{: October 15, 2016.}$ 

```
# Cholesky Decomposition
  # Author: Mido Assran
5 # Date: 30, September, 2016
  # Description: CholeskyDecomposition solves the linear system of equations:
7 # Ax = b by decomposing matrix A using Cholesky 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
  class CholeskyDecomposition(object):
17
19
      def solve (self, A, b):
          :type A: np.array([float])
21
          :type b: np.array([float])
          :rtype: np.array([float])
23
           start_time = timeit.default_timer()
25
          # If the matrix, A, is not square, exit
27
           if A. shape [0] != A. shape [1]:
               return None
29
          n = A. shape [1]
31
33
          # Simultaneous cholesky factorization of A and chol-elimination
35
37
          # Cholesky factorization & forward substitution
```

```
39
          for j in range(n):
              # If the matrix A is not positive definite, exit
41
              if A[j,j] <= 0:
                  return None
43
                                      # Compute the j, j entry of chol(A)
              A[j,j] = A[j,j] ** 0.5
45
              b[j] /= A[j,j]
                                         # Compute the j entry of forward-sub
47
49
              for i in range (j+1, n):
                  A[i,j] /= A[j,j] # Compute the i, j entry of chol(A)
51
                  b[i] -= A[i,j] * b[j] # Look ahead modification of b
53
                   if A[i,j] = 0:
                                         # Optimization for matrix sparsity
                       continue
55
                  # Look ahead moidification of A
57
                   for k in range (j+1, i+1):
                      A[i,k] -= A[i,j] * A[k,j]
59
61
63
          # Now solve the upper traingular system
65
          # Transpose(A) is the upper-tiangular matrix of chol(A)
          A[:] = matrix\_transpose(A)
67
          # Backward substitution
69
          for j in range (n-1, -1, -1):
              b[j] /= A[j,j]
71
               for i in range(i):
73
                  b[i] -= A[i,j] * b[j]
75
          elapsed_time = timeit.default_timer() - start_time
77
```

```
print("Execution time: ", elapsed_time, end="\n\n")
79
           # The solution was overwritten in the vector b
            return b
81
83 if __name__ == "__main__":
       from utils import generate_positive_semidef, matrix_dot_vector
85
       order = 10
       seed = 5
87
       print("\n", end="\n")
89
       print ("# — TEST — #", end="\n")
print ("# — Cholesky Decomposition — #", end="\n")
print ("# — #", end = "\n\n")
91
       # Create a symmetric, real, positive definite matrix.
93
       A = generate_positive_semidef(order=order, seed=seed)
       x = np.random.randn(order)
95
       b = matrix_dot_vector(A=A, b=x)
       print("A: \n", A, end="\n\n")
97
       print("x:\n", x, end="\n\n")
99
       print("b (=Ax): \n", b, end="\n\n")
       chol_d = CholeskyDecomposition()
       v = chol_d.solve(A=A, b=b)
101
       print ("chol_d.solve(A, b):\n", v, end="\n")
       print ("2-norm error: \n", np. linalg.norm (v - x), end="\n")
103
       print ("# -----
                                                        - \#", end ="\n\n")
```

Listing 1. Cholesky.py

```
#
2 # Utils
  # -
4 # Author: Mido Assran
  # Date: 5, October, 2016
6 # Description: Utils provides a cornucopia of useful matrix
  # and vector helper functions.
  import random
10 import numpy as np
12 def matrix_transpose(A):
      :type A: np.array([float])
14
      :rtype: np.array([floats])
16
      # Initialize A_T(ranspose)
18
      A_T = \text{np.empty}([A. \text{shape}[1], A. \text{shape}[0]])
20
      # Set the rows of A to be the columns of A_T
      for i, row in enumerate(A):
22
           A_T[:, i] = row
24
      return A<sub>-</sub>T
26
28 def matrix_dot_matrix(A, B):
      :type A: np.array([float])
30
      :type B: np.array([float])
      :rtype: np.array([float])
32
34
      # If matrix shapes are not compatible return None
36
      if (A. shape [1] != B. shape [0]):
           return None
38
```

```
A_{dot_B} = np.empty([A.shape[0], B.shape[1]])
      A_{dot} = [:] = 0 # Initialize entries of the new matrix to zero
40
42
      B_T = matrix\_transpose(B)
      for i, row_A in enumerate(A):
44
           for j, column_B in enumerate(B_T):
               for k, v in enumerate(row_A):
46
                   A_{dot_B}[i, j] += v * column_B[k]
48
      return A_dot_B
50
52 def matrix_dot_vector(A, b):
      :type A: np.array([float])
54
      :type b: np.array([float])
      :rtype: np.array([float])
56
58
      # If matrix shapes are not compatible return None
      if (A. shape [1] != b. shape [0]):
60
           return None
62
      A_{dot_b} = np.empty([A.shape[0]])
      A_{-}dot_{-}b[:] = 0 # Initialize entries of the new vector to zero
64
      for i, row_A in enumerate(A):
66
           for j, val_b in enumerate(b):
               A_dot_b[i] += row_A[j] * val_b
68
      return A_dot_b
70
72
  def vector_to_diag(b):
74
      :type b: np.array([float])
      :rtype: np.array([float])
76
```

```
78
      diag_b = np.empty([b.shape[0], b.shape[0]])
      \operatorname{diag_b}[:] = 0
                       # Initialize the entries of the new diagonal matrix to zero
80
      for i, val in enumerate(b):
82
           diag_b[i, i] = val
84
      return diag_b
86
  def generate_positive_semidef(order, seed=0):
88
      :type order: int
      :type seed: int
90
      :rtype: np.array([float])
92
      np.random.seed(seed)
94
      A = np.random.randn(order, order)
      A = matrix_dot_matrix(A, matrix_transpose(A))
96
      return A
98
```

Listing 2. Utils.py