LAA Assignment

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Driver Code

```
In [ ]: from helper_functions import *
        import pandas as pd
In [ ]: tol = 10**(-10)
        results = []
        for i in range(2,11):
            A = hilbertMatrix(i)
            b = 20*np.random.random_sample(size=i)-20
             results.append([i, \ luSolveNorm(A,b, \ 'GEPP', \ tol,i), \ luSolveNorm(A,b, \ 'GERP', \ tol,i),
                              luSolveNorm(A,b, 'GECP', tol,i),scipyLUNorm(A,b),
                             luSolveNorm(A,b, \verb"Cholesky", tol, i), scipyCholNorm(A,b)])
        df = pd.DataFrame(results, columns = ['Order', 'GEPP', 'GERP', 'GECP', 'ScipyLU', 'Cholesky', 'ScipyChol'])
In [ ]: # LU Norms Comparison
        df
Out[ ]:
           Order
                          GEPP
                                        GERP
                                                      GECP
                                                                 ScipyLU Cholesky
                                                                                      ScipyChol
        0
                   0.000000e+00
                                 0.000000e+00
                                               0.000000e+00 0.000000e+00
                                                                          3.061750 3.552714e-15
         1
               3
                   3.746768e-14
                                               3.746768e-14
         2
                   3.330136e-13
                                 3.330136e-13
                                               3.330136e-13 8.608002e-13 6.765263 7.429356e-13
         3
               5
                   4.482813e-11
                                 6.290807e-11
                                               7.787435e-11 8.103213e-11 10.394742 5.818810e-11
         4
               6
                   6.692461e-10
                                 1.009176e-09
                                               5.754911e-10 8.898829e-10 17.967841 7.384851e-10
         5
                   2.373236e-08
                                 2.234370e-08
                                               5.053934e-08
                                                             2.917992e-08 13.547332 3.123814e-08
         6
                   1.622581e-06
                                 1.019623e-06
                                               2.446209e-06
                                                            1.375434e-06 27.780372 1.612921e-06
         7
               9 -1.000000e+00 -1.000000e+00
                                               4.028373e-05
                                                             8.150863e-05 23.939432 6.917792e-05
               10 -1.000000e+00 -1.000000e+00 -1.000000e+00 4.743223e-04 27.324559 4.252664e-04
         8
In [ ]: np.random.seed(21)
        results = []
        result2 = []
        for i in [2,5,10,20,50,100,200,500,700,1000]:
             b=np.random.rand(i,1)*1000
            b_delta=np.random.rand(i,1)*(1e-12)
            A = hilbertMatrix(i)
             results.append([i,condNum(A,b,b_delta,'GEPP',i),condNum(A,b,b_delta,'GERP',i),
                             condNum(A,b,b_delta,'GECP',i)])
        for j in range(1,11):
             b=np.random.rand(j,1)*1000
            b_delta=np.random.rand(j,1)*(1e-12)
            A = hilbertMatrix(j)
            result2.append(condNum(A,b,b_delta,'Cholesky',j))
        df = pd.DataFrame(results, columns = ['Order', 'GEPP', 'GERP', 'GECP'])
        df['Cholesky'] = result2
In [ ]: # Bound on conditional number
```

Out[]:		Order	GEPP	GERP	GECP	Cholesky
	0	2	5.663295e-01	0.566330	0.566330	1.011436
	1	5	1.347086e+00	1.163510	1.202622	1.012774
	2	10	0.000000e+00	0.000000	0.547353	0.992819
	3	20	1.654403e+03	7.474145	1.976511	2.753290
	4	50	9.522863e+06	737.757323	5.622611	0.191970
	5	100	2.582011e+08	2465.640401	2.263894	0.255180
	6	200	1.581160e+10	78840.547729	3.224121	0.502048
	7	500	5.000193e+11	51065.531522	7.157211	2.326155
	8	700	1.196713e+12	90753.598613	11.012589	8.702383
	9	1000	2.900402e+12	778796.452293	18.476805	1.566430

Helper functions

```
In [ ]: import numpy as np
        from scipy.linalg import lu_factor, lu_solve, cholesky
        def hilbertMatrix(n):
            Generates an n x n Hilbert matrix
            H = np.zeros((n, n))
            for i in range(1, n + 1):
                for j in range(1, n + 1):
                    H[i - 1, j - 1] = 1 / (i + j - 1)
            return H
        def luGEPP(A,tol):
            LU decomposition with partial pivoting
            n = len(A)
            A = A.astype(float)
            P = np.identity(n, dtype=float)
            L = np.zeros((n,n),dtype=float)
            U = np.copy(A)
            for i in range(n):
                max_pivot = abs(U[i:,i])
                max_row = np.argmax(max_pivot) + i
                if abs(U[max_row, i]) < tol:</pre>
                     P[0,0] = -1
                     return (P,L,U)
                U[[i, max_row]] = U[[max_row, i]]
                P[[i, max_row]] = P[[max_row, i]]
                L[[i, max_row]] = L[[max_row, i]]
                for j in range(i+1, n):
                     L[j,i] = U[j,i] / U[i,i]
                     U[j,i:] = L[j,i] * U[i,i:]
            np.fill_diagonal(L, 1.0)
            return P, L, U
        def luGERP(A,tol):
            LU decomposition with rook pivoting
            n = len(A)
            A = A.astype(float)
            Pl = np.identity(n, dtype=float)
            Pr = np.identity(n, dtype=float)
            L = np.zeros((n,n),dtype=float)
            U = np \cdot copy(A)
            for i in range(n):
                 pivotColInd = np.argmax(abs(U[i:,i]))+i
                 pivotRowInd = np.argmax(abs(U[i:,i]))+i
                 if abs(U[pivotColInd, i]) < abs(U[i, pivotRowInd]):</pre>
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```
pivotInd = pivotRowInd
            if abs(U[i, pivotRowInd]) < tol:</pre>
                P1[0,0] = -1
                return (P1,Pr,L,U)
            Pr[:, [pivotInd, i]] = Pr[:, [i, pivotInd]]
            U[:, [pivotInd,i]] = U[:, [i, pivotInd]]
            if abs(U[pivotColInd, i]) < tol:</pre>
                P1[0,0] = -1
                return (P1,Pr,L,U)
            pivotInd = pivotColInd
            Pl[[pivotInd, i]] = Pl[[i, pivotInd]]
            L[[pivotInd, i], :i] = L[[i, pivotInd], :i]
            U[[pivotInd, i]] = U[[i, pivotInd]]
        for j in range(i+1, n):
            L[j, i] = U[j, i]/U[i, i]
            U[j, i:] -= L[j, i] * U[i, i:]
    return (Pl,Pr,L,U)
def luGECP(A,tol):
    LU decomposition with complete pivoting
    n = len(A)
    A = A.astype(float)
    Pl = np.identity(n, dtype=float)
    Pr = np.identity(n, dtype=float)
    L = np.identity(n,dtype=float)
    U = np.copy(A)
    for i in range(n-1):
        max = np.unravel_index(np.argmax(abs(U[i:,i:])), U[i:,i:].shape)
        rowInd = max[0] + i
        colInd = max[1] + i
        if abs(U[rowInd,colInd]) < tol:</pre>
            P1[0,0] = -1
            return (Pl,Pr,L,U)
        Pl[[rowInd, i]] = Pl[[i, rowInd]]
        Pr[:, [colInd, i]] = Pr[:, [i, colInd]]
        L[[rowInd, i], :i] = L[[i, rowInd], :i]
        U[:, [colInd, i]] = U[:, [i, colInd]]
        U[[rowInd, i]] = U[[i, rowInd]]
        for j in range(i+1, n):
            L[j, i] = U[j, i]/U[i, i]
            U[j, i:] -= L[j, i] * U[i, i:]
    return (P1,Pr,L,U)
def cholesky(A, tol):
    Cholesky decomposition
    n = len(A)
    mid = A.copy().astype(np.float64)
    B = np.eye(n, dtype=np.float64)
    P = np.eye(n, dtype=np.float64)
    for i in range(n):
        pivot = np.argmax(np.diag(mid[i:,i:])) + i
        mid[[i,pivot],:] = mid[[pivot,i],:]
        mid[:,[i,pivot]] = mid[:,[pivot,i]]
        P[[i,pivot]] = P[[pivot,i]]
        B[[i,pivot],:i] = B[[pivot,i],:i]
        alpha = np.sqrt(mid[i,i])
        if abs(alpha)<tol:</pre>
            P[0,0] = -1
            return P,B
        wAlpha = mid[i,i+1:]/alpha
        B[i,i] = alpha
        B[i+1:,i] = wAlpha
        mid[i+1:,i+1:] -= [[i*j for i in wAlpha] for j in wAlpha ]
    return P,B
def forward_sub(P, L, b):
    Forward substitution
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```
Pb = np.dot(P, b)
    n = len(L)
    y = np.zeros(n, dtype=float)
    for i in range(n):
       y[i] = Pb[i] - sum(L[i,:i] * y[:i])
    return y
def backward_sub(U, Q, y):
    Backward substitution
    n = len(y)
    x = np.zeros(n)
    x[n - 1] = y[n - 1] / U[n - 1, n - 1]
    for i in range(n - 2, -1, -1):
        x[i] = y[i]
        for j in range(i + 1, n):
           x[i] = U[i, j] * x[j]
        x[i] /= U[i, i]
    return Q@x
def luSolve(Pl, Pr, L, U, b):
    LU Solver using forward and backward substitution
   y = forward_sub(P1, L, b)
    x = backward_sub(U, Pr, y)
    return x
def luSolveNorm(A, b, solver, tol, i):
    Norm Ax-b using various solvers
    if solver == 'GEPP':
       P, L, U = luGEPP(A, tol)
        if P[0,0] == -1:
            return -1
        else:
           I = np.identity(i, dtype=float)
            x = luSolve(P, I, L, U, b)
            return np.linalg.norm(A@x-b)
    elif solver == 'Cholesky':
        P, L = cholesky(A, tol)
        if P[0,0] == -1:
            return -1
        else:
            x = luSolve(P, P.T, L, L.T, b)
            return np.linalg.norm(A@x-b)
    elif solver == 'GERP':
        Pl, Pr, L, U = luGERP(A, tol)
        if Pl[0,0] == -1:
            return -1
        else:
            x = luSolve(Pl, Pr, L, U, b)
            return np.linalg.norm(A@x-b)
    elif solver == 'GECP':
        Pl, Pr, L, U = luGECP(A, tol)
        if Pl[0,0] == -1:
            return -1
        else:
            x = luSolve(Pl, Pr, L, U, b)
            return np.linalg.norm(A@x-b)
def condNum(A, b, b_delta, solver, i):
    Condition number using various solvers
    if solver == 'GEPP':
       P, L, U = luGEPP(A, 0)
       Q = np.eye(i,dtype="float")
    elif solver == 'Cholesky':
       P, L = cholesky(A, 0)
        Q = P.T
        U = L.T
```

```
elif solver == 'GERP':
                               P,Q,L,U = luGERP(A,0)
                elif solver == 'GECP':
                               P,Q,L,U = luGECP(A,0)
                x = luSolve(P,Q,L,U,b)
                x_delta = luSolve(P,Q,L,U,b+b_delta)
                \textbf{return np.linalg.norm} (x\_delta-x,ord=2)*np.linalg.norm (b\_ord=2)/(np.linalg.norm (b\_delta,ord=2)*(np.linalg.norm (x\_ord=2))/(np.linalg.norm (b\_delta,ord=2))/(np.linalg.norm (b\_delta,ord=2))/(np
def scipyLUNorm(A,b):
                Evaluates the norm of the residual of the LU factorization
                lu, piv = lu_factor(A)
                exp_x = lu_solve((lu,piv),b)
                return np.linalg.norm(A@exp_x-b)
def scipyCholNorm(A, b):
                Evaluates the norm of the residual of the Cholesky factorization
                L = np.linalg.cholesky(A)
                y = np.linalg.solve(L, b)
                x = np.linalg.solve(L.T, y)
                return np.linalg.norm(A@x-b)
```