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In [ ]: # Import standard numeric and visualization libs  
import numpy as np  
import matplotlib.pyplot as plt  
  
# Import my householder QR factorization implementation  
# import sys  
# sys.path.insert(0, "/home/masvgp/dev/byu_num_lin_alg_2022/src/")  
# from factorizations.householder import house, formQ
```

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In [ ]: # My implementation of householder factorization and it's associated utility functi
def e(i, length):
    """Generate a unit coordinate vector, i.e., Given an index i and a length, retu

    Args:
        i (int): Integer index in which 1.0 will be assigned
        length (int): Length of the desired output vector

    Returns:
        arr: Unit coordinate vector with 1.0 in the ith position
    """
    e_i = np.zeros(length, dtype=np.float64)
    e_i[i] = 1.0
    return e_i

def reflector(v):
    """Given a unit vector v, return the householder reflector  $I - 2 * \text{np.outer}(v,$ 

    Args:
        v (arr): Vector
    Returns:
        reflector (arr): A  $\text{len}(v) \times \text{len}(v)$  array representing a householder reflect
    """
    I = np.eye(len(v))
    outer = np.outer(v, v)

    return (I - 2 * outer)

def house(A, reduced=False):
    """Compute the factor R of a QR factorization of an  $m \times n$  matrix A with  $m \geq n$ .

    Args:
        A (arr): A numpy array of shape  $m \times n$ 
    Returns:
        R (arr): The upper diagonal matrix R in a QR factorization
    """
    m = A.shape[0] # Get row-dim of A
    n = A.shape[1] # Get col-dim of A
    W = np.zeros((m, n))
    #Q = np.eye(m, m)
    # V = np.zeros((m, n))

    # Cast A as type np.float64
    A = A.astype(np.float64)

    for k in range(n):
        # Get the first column of the  $(m-k+1, n-k+1)$  submatrix of A
        x = A[k:m, k]

        # Compute e_1, sign function, and the norm of x
        e_1 = e(0, len(x))
        sign = np.sign(x[0])
        norm_x = np.linalg.norm(x)

        # Compute vk, the vector reflected across the Householder hyperplane

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vk = (sign * norm_x * e_1) + x
vk = vk / np.linalg.norm(vk)

# Store vk in W
W[k:m, k] = vk

# Apply the reflector to the k:m, k:n submatrix of A to put
# zeros below the diagonal of the kth column of A
A[k:m, k:n] = reflector(vk) @ A[k:m, k:n]

if reduced == True:
    # return np.around(A[:n, :n], decimals=8), W
    return A[:n, :n], W
else:
    # Return full R
    return A, W
    # return np.around(A, decimals=8), W

def formQ(W, reduced=False):
    """Form Q from the lower diagonal vk's formed in the householder algorithm.

    Args:
        W (arr): An m x n matrix containing the vk in the kth column

    Returns:
        arr: Returns an m x m orthonormal matrix Q such that QR = A
    """
    m = W.shape[0]
    n = W.shape[1]
    Q = np.eye(m, m)
    for k in range(n):
        Qk = np.eye(m, m)
        Qk[k:m, k:m] = reflector(W[k:m, k])
        Q = Q @ Qk

    if reduced == True:
        # Return reduced Q
        return Q[:m, :n]
        # return np.around(Q[:m, :n], decimals=8)
    else:
        # Return full Q
        return Q
        # return np.around(Q, decimals=6)

```

Problem 4

Use the result of Problem 3 to develop a Python function `lq` that takes as input a matrix A and returns Q and L where $A = QL$ is a QL -decomposition, and test it with random matrices of sizes 10×7 and 75×50 . Organize your code in a single Python notebook and include it in the PDF you turn in. By test your code, I mean that you compute the QL decomposition, that you verify Q has orthonormal columns, you verify that L is lower triangular, and that you compute $\|A - QL\|_2$, where A is a random matrix for which you

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compute the QL decomposition. I should be able to run your single file code and observe
In [ ]: def ql(A):
        """Compute the QL factorization of an m x n matrix A.

        Args:
            A (arr): An m x n matrix
        Return:
            Q (arr): a unitary matrix
            L (arr): A lower triangular matrix
        """
        # Get dims of A
        m = A.shape[0]
        n = A.shape[1]

        # Define the reversal matrix K_n
        Kn = np.flip(np.eye(n), axis=0)

        # Compute reduced QR(AK_n) - \hat{Q} is m x n, \hat{R} is n x n.
        Rhat, W = house(A @ Kn, reduced=True)
        Qhat = formQ(W, reduced=True)

        # Compute Q = \hat{Q}K_n
        Q = Qhat @ Kn

        # Compute L = K_n\hat{R}K_n
        L = Kn @ Rhat @ Kn

        # return Q, L
        return Q, L

```

Testing

1. Define matrices A and B of size 10×7 and 75×50 , respectively. Then do the following:
2. Compute $QL(A)$ and $QL(B)$
3. Verify that Q has orthonormal columns
4. Verify that L is lower triangular
5. Compute $\|A - QL\|_2$ and $\|B - QL\|_2$, where A and B are the random matrices defined in (1).

Testing for the 10×7 random matrix a

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In [ ]: # 1. Define random matrix of size 10 x 7
A = np.random.normal(loc=0.0, scale=1.0, size=(10, 7))

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In [ ]: # 2. Compute $QL(A)$
QA, LA = ql(A)

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In [ ]: # 3. Verify that $Q$ has orthonormal columns
# Compute $Q^T @ Q$
QATQA = QA.T @ QA

# Check the diagonals are 1.0
print(f"The dimensions of $Q^T @ Q$ are: {QATQA.shape[0]} x {QATQA.shape[1]}")
print(f"Check that the diagonal entries of $Q^T @ Q$ sum to {QATQA.shape[0]} by che

# Set diagonal entries to 0.0
for i in range(QATQA.shape[0]):
    QATQA[i, i] = 0.0

# Compute the max absolute value of the the array $Q^T @ Q$ with diagonal set to 0.0.
print("Check the max absolute value of the array after setting the diagonal entries
```

The dimensions of $Q^T @ Q$ are: 7 x 7
 Check that the diagonal entries of $Q^T @ Q$ sum to 7 by checking the sum of the diagonal entries.
 The sum is equal to 7: True
 Check the max absolute value of the array after setting the diagonal entries to 0.
 0.
 This effectively checks the max absolute value of the off-diagonal entries.
 The max absolute value of the resulting array is: 3.2963142379103426e-16

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In [ ]: # 4. Verify that $L$ is Lower triangular
# Extract upper diagonal part of $L$
LA_upper_diag = [LA[i, j] for i in range(LA.shape[0]) for j in range(LA.shape[1]) if i < j]
# Extract the lower diagonal part of $L$
LA_lower_diag = [LA[i, j] for i in range(LA.shape[0]) for j in range(LA.shape[1]) if i > j]

print("The max absolute value of the upper diagonal entries of $L$ is: " + str(np.max(LA_upper_diag)))
print("The max absolute value of the lower diagonal entries of $L$ is: \n" + str(np.max(LA_lower_diag)))
```

The max absolute value of the upper diagonal entries of L is: 4.2312988466972233e-16
 The max absolute value of the lower diagonal entries of L is:
 3.0734874283890266

```
In [ ]: # 5. Compute $\|A - QL\|_2$, where $A$ is the random matrix
print("The L2-norm of $A - QL$ is: " + str(np.linalg.norm(A - QA @ LA)))
```

The L2-norm of $A - QL$ is: 3.831465111543962e-15

Testing for the 75×50 random matrix B .

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In [ ]: # 1. Define random matrix of size 75 x 50
B = np.random.normal(loc=0.0, scale=1.0, size=(75, 50))
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In [ ]: # Compute $QL$ for the 75 x 50 matrix $B$
QB, LB = ql(B)
```

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In [ ]: # 3. Verify that $Q$ has orthonormal columns
# Compute $Q^T @ Q$
QBTQB = QB.T @ QB

# Check the diagonals are 1.0
print(f"The dimensions of $QB.T @ QB$ are: {QBTQB.shape[0]} x {QBTQB.shape[1]}")
print(f"Check that the diagonal entries of $QB.T @ QB$ sum to {QBTQB.shape[0]} by che

# Set diagonal entries to 0.0
for i in range(QBTQB.shape[0]):
    QBTQB[i, i] = 0.0

# Compute the max absolute value of the the array $Q^T Q$ with diagonal set to 0.0.
print("Check the max absolute value of the array after setting the diagonal entries
```

The dimensions of \$QB.T @ QB\$ are: 50 x 50

Check that the diagonal entries of \$QB.T @ QB\$ sum to 50 by checking the sum of the diagonal entries.

The sum is equal to 50: True

Check the max absolute value of the array after setting the diagonal entries to 0.0.

This effectively checks the max absolute value of the off-diagonal entries.

The max absolute value of the resulting array is: 8.721678135614107e-16

```
In [ ]: # 4. Verify that $L$ is Lower triangular
# Extract upper diagonal part of $L$
LB_upper_diag = [LB[i, j] for i in range(LB.shape[0]) for j in range(LB.shape[1]) if i < j]
# Extract the lower diagonal part of $L$
LB_lower_diag = [LB[i, j] for i in range(LB.shape[0]) for j in range(LB.shape[1]) if i > j]

print("The max absolute value of the upper diagonal entries of $L$ is: " + str(np.max(LB_upper_diag)))
print("The max absolute value of the lower diagonal entries of $L$ is: \n" + str(np.max(LB_lower_diag)))
```

The max absolute value of the upper diagonal entries of \$L\$ is: 2.9251450570491478e-15

The max absolute value of the lower diagonal entries of \$L\$ is:

9.757443792666914

```
In [ ]: # 5. Compute $\|B - QL\|_2$ where $B$ is the random matrix
print("The L2-norm of $B - QL$ is: " + str(np.linalg.norm(B - QB @ LB)))
```

The L2-norm of \$B - QL\$ is: 1.0843623836693697e-13