Software Manual for Math 5610

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1 Introduction

The purpose of this document is twofold. The primary purpose is educational. Creating functional and well-structured code is a difficult task. In most computational course work, once an assignment is finished, the code quickly becomes lost and forgotten, never to be opened or utilized again. Knowing that I would eventually create this document drove me to write better code, so that I wouldn't need to remind myself of how it works before adding it to the document.

Its second reason is professional, or rather, pre-professional. It is an attempt to internalize the principles of well-documented and well-maintained code, so that in my professional career I will develop code which can be picked up and understood by co-workers easily.

I was once told that the key to success in Software Development is to make yourself dispensable. Creating code which is difficult to understand and maintain will make employers depend on your knowledge of the code, and create a poor product. Chances are you'll get stuck at the same job working on the same project for many years, making your knowledge specific and stale. Writing code which is readable and maintainable makes it easy for others to pick up where you left off, so you can move on to bigger and better things.

The main body of this document will be the documentation of code which has been developed over the course of the semester. The sections of said body will rely heavily upon the structures contained in the Appendix. For example, matrix operations such as Gaussian Elimination and Back Substitution will be explored in the main body, while the Matrix structure itself is contained in the Appendix.

The first section of this document contains routines dealing with error and precision in regards to machine computing. Topics such as relative and absolute error, vector and matrix norms, and the rounding unit will be discussed. Basic routines relating to all these topics will be given.

Section two will explore nonlinear equations of one variable. In particular, it contains various root-finding methods, such as Bisection and the Newton Method.

The next section discusses methods for solving linear systems of equations. It is broken into subsections, such as Direct Methods (Gaussian Elimination, LU Factorization) and Iterative Methods (Jacobi Iteration, Conjugate Gradient Method).

TODO: Summary of the last section TODO: Intro Conclusion

2 Precision and Error

- 2.1 Error
- 2.1.1 Absolute Error
- 2.1.2 Relative Error
- 2.1.3 Absolute Error in Complex Numbers
- 2.1.4 Relative Error in Complex Numbers

2.2 Rounding

- 2.3 Machine Precision
- 2.3.1 Single Precision
- 2.3.2 Double Precision

- 2.4 Vector Norms
- 2.4.1 One Norm/Manhattan Norm
- 2.4.2 Infinity Norm
- 2.4.3 L2 Norm
- 2.5 Matrix Norms
- 2.5.1 One Norm
- 2.5.2 Infinity Norm

3 Nonlinear Equations In One Variable

TODO: Section Header
3.1 Bisection
Description:
Input:
1.
2.
3.
4.
Output: Code Written:
Usage Sample:
//Written by Andrew Sheridan in C++
Console Output

3.2 Fixed Iteration
Description:
Input:
1.
2.
3.
4.
Output:
Code Written:
Usage Sample:
//Written by Andrew Sheridan in C++
Console Output

3.3 Newton Method
Description:
Input:
1.
2.
3.
4.
Output:
Code Written:
Usage Sample:
//Written by Andrew Sheridan in C++
Console Output

3.4 Secant Method
Description:
Input:
1.
2.
3.
4.
Output:
Code Written:
Usage Sample:
//Written by Andrew Sheridan in C++
Console Output

3.5 Hybrid Method
Description:
Input:
1.
2.
3.
4.
Output:
Code Written:
Usage Sample:
1 //Written by Andrew Sheridan in C++
Console Output

4 Matrix Operations

4.1 Direct Methods

Todo: Write up description for this section

4.1.1 Back Substitution

Description: Solves an upper triangular system of equations using back substitution.

Input:

- 1. Matrix A (Upper-triangular matrix. Note: Will not be reduced.)
- 2. Vector b (Right-Hand-Side vector.)

Output: Vector (the solution to the system of equations)

Returns NULL if the Matrix and Vector are of incorrect size.

Code Written:

```
/// Solves a set of linear equations using back substitution
  /// Does not reduce matrix A
3 //A: The matrix to be reduced
4 // b: Right-Hand-Side
5 Vector BackSubstitution(Matrix A, Vector b) {
     if (A.GetRows() != b.GetSize()) return NULL;
     Vector x(b.GetSize());
     for(int i = b.GetSize() - 1; i >= 0; i--)
10
11
       x[i] = b[i];
12
       for (int j = i + 1; j < b.GetSize(); j++) {
   x[i] -= A[i][j] * x[j];</pre>
13
14
       {\tt x\,[\,i\,]} \ /{=} \ {\tt A\,[\,i\,]\,[\,i\,]}\,;
16
     }
17
     return x;
18
  }
19
```

```
//Written by Andrew Sheridan in C++
int size = 4;
Matrix matrix = MatrixFactory::Instance()->UpperTriangular(size, size)

Vector vector = Vector(size);
vector.InitializeAllOnes();
vector = matrix * vector;
Vector resultVector = BackSubstitution(matrix, vector);

std::cout << "Upper triangular matrix" << std::endl;
matrix.Print();
std::cout << "Test vector" << std::endl;
vector.Print();</pre>
```

```
14 std::cout << "Result of back substition " << std::endl;
15 resultVector.Print();</pre>
```

consore bubble					
Upper trian	gular matrix				
40.713	0.428471	0.690885	0.71915		
0	40.4911	0.780028	0.410924		
0	0	40.5797	0.139951		
0	0	0	40.401		
Test vector					
42.5515	41.6821	40.7196	40.401		
Result of back substition					
1	1	1	1		

4.1.2 Forward Substitution

Description: Solves a lower-triangular set of linear equations using forward substitution.

Input:

- 1. Matrix A (lower triangular system of equations)
- 2. Vector b (right-hand-side)

Output: Vector (solution to the system of equations)

Returns NULL if the Matrix and Vector are of incorrect size.

Code Written:

```
/// Solves a set of linear equations using forward substitution
  /// Does not reduce matrix A
3 //A: The lower-triangular matrix
4 //b: right-hand-side
  Vector ForwardSubstitution(Matrix A, Vector b) {
    if (A.GetRows() != b.GetSize()) return NULL;
     Vector x(b.GetSize());
     x[0] = b[0];
10
     for (unsigned i = 0; i < A.GetRows(); i++) {</pre>
11
12
       x[i] = b[i];
       for (unsigned j = 0; j < i; j++) { x[i] = x[i] - (A[i][j] * x[j]);
13
14
15
       x[i] = x[i] / A[i][i];
16
17
    return x;
19
20 }
```

```
//Written by Andrew Sheridan in C++

int size = 4;

Matrix matrix = MatrixFactory::Instance()->LowerTriangular(size, size)

Vector vector = Vector(size);

vector.InitializeAllOnes();

vector = matrix * vector;

Vector resultVector = ForwardSubstitution(matrix, vector);

std::cout << "Upper triangular matrix" << std::endl;
matrix.Print();
std::cout << "Test vector" << std::endl;
vector.Print();
std::cout << "Result of back substition" << std::endl;
resultVector.Print();
```

Upper triangular matrix					
40.713	0	0	0		
0.428471	40.6909	0	0		
0.71915	0.491119	40.78	0		
0.410924	0.579694	0.139951	40.401		
Test vector					
40.713	41.1194	41.9903	41.5316		
Result of back substition					
1	1	1	1		

4.1.3 Gaussian Elimination

Description: Reduces a matrix to its upper-triangular form using Gaussian Elimination.

Input:

- 1. Matrix A (matrix to be reduced to upper-triangular form)
- 2. Vector b (right-hand-side, will be reduced along with the matrix)

Output: void

Code Written:

```
_{1} /// Reduces a matrix right-hand-side b to upper-triangular form using \leftrightarrow
        Gaussian Elimination
   //A: The matrix to be reduced
  // b: Right-Hand-Side
  void GaussianElimination(Matrix& A, Vector& b) {
     for (unsigned k = 0; k < A.GetRows(); k++) {
        for (unsigned i = k + 1; i < A.GetRows(); i++) {
  double factor = A[i][k] / A[k][k];</pre>
6
          {\tt A\,[\,i\,]\,[\,j\,]} \; = \; {\tt A\,[\,i\,]\,[\,j\,]} \; - \; {\tt factor} \! * \! {\tt A\,[\,k\,]\,[\,j\,]} \, ;
9
10
          A[i][k] = 0;
11
          b[i] = b[i] - factor*b[k];
12
13
     }
14
  }
15
```

Usage Sample:

```
1 //Written by Andrew Sheridan in C++
_2 int size = 4;
     \texttt{Matrix matrix} = \texttt{MatrixFactory} :: \texttt{Instance}() \ -\!\!\!> \ \texttt{DiagonallyDominant}(\texttt{size} \! \leftarrow\!\!\!\!-
 4
      Vector vector(size);
     vector.InitializeAllOnes();
      vector = matrix * vector;
     Matrix reducedMatrix = matrix;
     {\tt Vector\ resultVector}\ =\ {\tt vector}\ ;
10
     GaussianElimination(reducedMatrix, resultVector);
11
     std::cout << "Gaussian Elimination" << std::endl;
std::cout << "Test Matrix" << std::endl;</pre>
13
     matrix.Print();
std::cout << "Test vector" << std::endl;</pre>
14
     vector.Print();
std::cout << "Result of gaussian elimination" << std::endl;</pre>
16
17
     reducedMatrix.PrintAugmented(resultVector);
```

Console Output

Gaussian Elimination

Test Matrix

40.713	0.428471	0.690885	0.71915	
0.491119	40.78	0.410924	0.579694	
0.139951	0.401018	40.6273	0.324151	
0.244759	0.694755	0.593902	40.6318	
Test vector				
42.5515	42.2618	41.4924	42.1652	
Result of gaussian elimination				
40.713	0.428471	0.690885	0.71915	42.5515
0	40.7749	0.40259	0.571019	41.7485
0	0	40.621	0.316084	40.9371
0	0	0	40.6132	40.6132

4.1.4 LU Factorization

Description: Finds the LU Factorization of a system of linear equations. L is lower-triangular, U, upper-triangular.

Input:

- 1. Matrix A (Coefficient Matrix)
- 2. Vector b (Right-Hand-Side)

Output: Matrix*

The first entry in this array is L, the second entry, U.

Code Written:

```
/// Finds the LU factorization of matrix A.
   /// RHS b will be modified
3 // A: The nxn coefficient matrix
4 // b: Right-Hand-Side
5 Matrix* LUFactorization(Matrix A, Vector& b) {
      if (A.GetRows() != b.GetSize()) return NULL;
      {\tt Matrix} \ \ L\left({\tt A.GetRows}\left(\right)\,,\ {\tt A.GetColumns}\left(\right)\right);
      L. InitializeIdentityMatrix();
9
10
      \quad \text{for (int } k = 0; \ k < \texttt{A.GetRows()}; \ k++) \ \{
11
          12
             \begin{array}{lll} \mbox{double factor} \, = \, \mbox{A[i][k]} \, / \, \, \mbox{A[k][k]}; \end{array}
13
            L[i][k] = factor;
14
              \begin{array}{lll}  & \mbox{for} \ (\,\mbox{int}\ j \,=\, 0\,;\ j \,<\, \mbox{A.GetColumns}\,(\,)\,;\ j++) \ \{ \end{array} 
15
                {\tt A\,[\,i\,]\,[\,j\,]} \; = \; {\tt A\,[\,i\,]\,[\,j\,]} \; - \; {\tt factor} \! * \! {\tt A\,[\,k\,]\,[\,j\,]} \, ;
16
17
            b[i] = b[i] - factor*b[k];
19
20
      Matrix* LU = new Matrix[2]{L, A};
22
23
      return LU;
```

```
1 //Written by Andrew Sheridan in C++
2 int size = 4;
3 Matrix matrix = MatrixFactory::Instance()->DiagonallyDominant(size, \( \to \) size);
4 Vector vector(size);
5 vector = matrix * vector;
6 Matrix* LU = LUFactorization(matrix, vector);
7 Matrix L = LU[0];
8 Matrix U = LU[1];
9
10 std::cout << "LU Factorization" << std::endl;
11 std::cout << "Starting system: " << std::endl;
12 matrix.Print();
13 std::cout << "Result of Scaled LU Factorization" << std::endl;
14 U.Print();</pre>
```

LU Factorization

Starting system:

40.713	0.428471	0.690885	0.71915
0.491119	40.78	0.410924	0.579694
0.139951	0.401018	40.6273	0.324151
0.244759	0.694755	0.593902	40.6318

Result of Scaled LU Factorization 40.713 0.428471 0.690885

Result of	r Scaled LU Factor	rization	
40.713	0.428471	0.690885	0.71915
0	40.7749	0.40259	0.571019
0	-5.55112e-17	40.621	0.316084
0	7.9659e-19	0	40.6132

1	0	0	0
0.012063	1	0	0
0.0034375	0.0097988	1	0
0.00601183	0.0169756	0.0143501	1

4.1.5 Cholesky Decomposition

Description: Computes the Cholesky Decomposition of a symmetric-positive-definite matrix.

Input:

1. Matrix A (Symmetric Positive Definite Matrix)

Output: Matrix

Returns NULL if the matrix is not SPD.

Code Written:

```
1 ///Computes the Cholesky Decomposition of an n by n matrix A
  /// Returns NULL if the matrix is not SPD
_3 Matrix CholeskyDecomposition(Matrix& A) {
    if (A.IsSymmetric() = false)
4
       return NULL;
    Matrix L(A.GetRows(), A.GetColumns()); //Initialize the new matrix
    for (unsigned i = 0; i < A.GetRows(); i++) {</pre>
9
       for (unsigned j = 0; j < (i + 1); j++) {
10
11
         double entry = 0;
         for (unsigned k = 0; k < j; k++) {
12
13
           entry += L[i][k] * L[j][k];
14
         double sqrtValue = A[i][i] - entry;
15
         if (sqrtValue < 0)
16
           return NULL;
17
18
         // Conditional assignment. If the entry is diagonal, assign to \hookleftarrow
             the squre root of the previous value.
           Otherwise, Do computation for a nondiagonal entry
20
         L[i][j] = i = j ? std::sqrt(sqrtValue) : (1.0 / L[j][j] * (A[i \leftarrow i))
21
             ][j] - entry));
22
    }
23
24
25
    return L;
  }
26
```

```
//Written by Andrew Sheridan in C++

int size1 = 4;

Matrix matrix1 = MatrixFactory::Instance()->SPD(size1);

Matrix cholesky1 = CholeskyDecomposition(matrix1);

if (cholesky1 != NULL) {

std::cout << "Our diagonally dominant test matrix and the result of ←

computing the Cholesky Decomposition of the matrix." << std::←

endl;

matrix1.Print();

cholesky1.Print();

Matrix transpose = cholesky1.Transpose();

std::cout << "L Transpose" << std::endl;

transpose.Print();
```

•	•	test matrix an		of	computing	the	Cholesky	Decomposition	ı of	the	mati
40.713	0.428471	0.690885	0.71915								
0.428471	40.4911	0.780028	0.410924								
0.690885	0.780028	40.5797	0.139951								
0.71915	0.410924	0.139951	40.401								
6.38067	0	0	0								
0.0671514	6.36291	0	0								
0.108278	0.121447	6.36814	0								
0.112708	0.0633917	0.0188514	6.35484								
L Transpose											
6.38067	0.0671514	0.108278	0.112708								
0	6.36291	0.121447	0.0633917								
0	0	6.36814	0.0188514								
0	0	0	6.35484								
Multiplying	the Cholesky	Decomposition	n by its trans	spos	se.						
40.713	0.428471	0.690885	0.71915								
0.428471	40.4911	0.780028	0.410924								
0.690885	0.780028	40.5797	0.139951								
0.71915	0.410924	0.139951	40.401								

4.2 Pivoting Strategies

There will be systems of equations whose pivots will cause the system to yield poor results. To deal with this we can use pivoting strategies to use rows or columns whose pivots are well conditioned for solving the system. In this subsection I have modified some existing direct methods to use pivoting.

4.2.1 Scaled Gaussian Elimination

Description: This is Gaussian Elimination with Scaled Partial Pivoting. When pivots cause the normal Gaussian Elimination routine to give poor results, this algorithm should be used.

Input:

- 1. Matrix A
- 2. Vector b

Output: Matrix

This is the upper triangular matrix which has been reduced via Gaussian Elimination. Will return null if the matrix and vector are of incorrect size.

Code Written:

```
1 /// Reduces an matrix A and right-hand-side b to upper-triangular form←
        using Gaussian Elimination
  // A: The coefficient matrix
   // b: Right-Hand-Side
4 Matrix GaussianEliminationWithScaledPivoting(Matrix A, Vector& b) {
     if (A.GetRows() != b.GetSize()) return NULL;
6
     int n = b.GetSize();
     for (int k = 0; k < n; k++) {
9
       double* ratios = new double [n - k]; // New vector of size n - k to \leftarrow
10
             store the ratios
       for (int i = k; i < n; i++) {
11
12
         double rowMax = A[i].FindMaxMagnitudeStartingAt(k);
         ratios[i - k] = rowMax / A[i][k];
13
14
       int newPivot = FindMaxIndex(ratios, n - k) + k; //Find the best \leftarrow
15
            row for this iteration
16
        \begin{array}{ll} {\tt Vector} & {\tt temp} \, = \, {\tt A} \, [\, {\tt k} \, ] \, ; & // \\ {\tt A} \, [\, {\tt k} \, ] \, = \, {\tt A} \, [\, {\tt newPivot} \, ] \, ; & / \end{array} 
17
                                // Switch the current row with the best row ←
18
            for this iteration
       A[newPivot] = temp; //
19
20
       double tempEntry = b[k];
       b[k] = b[newPivot];
22
       b[newPivot] = tempEntry;
23
24
       for (int i = k + 1; i < n; i++)
25
          for (int j = 0; j < n; j++) {
27
            A[i][j] = A[i][j] - factor*A[k][j];
28
         b[i] = b[i] - (factor*b[k]);
30
31
```

Usage Sample:

```
//Written by Andrew Sheridan in C++
int size = 4;

Matrix matrix = MatrixFactory::Instance() -> Random(size, size);

Vector vector(size);
vector.InitializeAllOnes();
vector = matrix * vector;
Vector resultVector = vector;
Matrix reducedMatrix = GaussianEliminationWithScaledPivoting(
reducedMatrix, resultVector);

std::cout << "Gaussian Elimination With Scaled Pivoting" << std::endl;
std::cout << "Test Matrix" << std::endl;
matrix.Print();
std::cout << "Test vector" << std::endl;
vector.Print();
std::cout << "Result of gaussian elimination" << std::endl;
reducedMatrix.PrintAugmented(resultVector);</pre>
```

Console Output

```
Gaussian Elimination With Scaled Pivoting
```

Test Matrix

0.712955	0.428471	0.690885	0.71915
0.491119	0.780028	0.410924	0.579694
0.139951	0.401018	0.627317	0.324151
0.244759	0.694755	0.593902	0.631792

Test vector

2.55146 2.26177 1.49244 2.16521

Result of gaussian elimination

0.139951	0.401018	0.627317	0.324151	1
0	-0.00658268	-0.50321	0.0648859	1
0	0	120.911	-16.8459	1
0	0	0	-0.309529	1

4.2.2 Scaled LU Factorization

Description: This is a modified version of the LU Factorization method, which uses pivoting to ensure that poor pivots are not chosen.

Input:

- 1. Matrix A
- 2. Vector b

Output: Matrix*

An array of matrices. The first entry is the lower-diagonal matrix L, the second entry, the upper-diagonal matrix U.

Code Written:

```
1 /// Finds the LU Factorization of matrix A with RHS b. Returns a pair \leftrightarrow
        of matrices. The first is L, the second, U.
 2 // A: The nxn coefficient matrix
3 // b: Right-Hand-Side
4 // n: The size of the matrices
5 Matrix* ScaledLUFactorization(Matrix A, Vector b) {
      if (A.GetRows() != b.GetSize()) return NULL;
     {\tt Matrix} \ \ L\left({\tt A.GetRows}\left(\right)\,,\ {\tt A.GetColumns}\left(\right)\right);
     L.InitializeIdentityMatrix();
      \quad \text{for (unsigned } \texttt{k} = \texttt{0}; \texttt{ k} < \texttt{A.GetRows()}; \texttt{ k++}) \ \{
10
         \texttt{Vector ratios}(\texttt{A}.\texttt{GetRows}() - \texttt{k}); \text{ } / \text{New vector of size } \texttt{A}.\texttt{GetRows}() \ \leftarrow \\
11
             - k to store the ratios
         for (unsigned i = k; i < A.GetRows(); i++) {</pre>
12
           double rowMax = A[i].FindMaxMagnitudeStartingAt(k);
13
           \mathtt{ratios}\,[\,\mathtt{i}\,-\,\mathtt{k}\,]\,=\,\mathtt{rowMax}\,\,/\,\,\mathtt{A}\,[\,\mathtt{i}\,]\,[\,\mathtt{k}\,]\,;
14
15
        unsigned newPivot = ratios.FindMaxIndex() + k; //Find the best row↔
16
               for this iteration
17
         Vector temp = A[k]; //
18
        A[k] = A[newPivot]; // Switch the current row with the best row \leftarrow
19
              for this iteration
        A[newPivot] = temp; //
20
21
22
        double tempEntry = b[k];
        b[k] = b[newPivot];
23
        b[newPivot] = tempEntry;
24
25
         for (unsigned i = k + 1; i < A.GetRows(); i++) {
26
           double factor = A[i][k] / A[k][k];
           \texttt{L[i][k]} = \texttt{factor};
28
            for \ (unsigned \ j = k + 1; \ j < \texttt{A.GetColumns}(); \ j++) \ \{
29
             A[i][j] = A[i][j] - factor*A[k][j];
30
31
           A[i][k] = 0;
           \texttt{b[i]} = \texttt{b[i]} - \texttt{factor*b[k]};
33
34
35
36
      return new Matrix[2]{ L, A };
38
```

```
1 //Written by Andrew Sheridan in C++
_2 int size = 4;
4 Matrix matrix = MatrixFactory::Instance() -> Random(size, size);
5 Vector vector(size);
6 vector.InitializeAllOnes();
7 Vector resultVector = matrix * vector;
s \ \mathtt{Matrix*} \ \mathtt{LU} = \mathtt{ScaledLUFactorization} \, (\mathtt{matrix} \, , \ \mathtt{resultVector}) \, ;
9 Matrix L = LU[0];
10 Matrix U = LU[1];
12 std::cout << "LU Factorization with pivoting" << std::endl;
13 std::cout << "Test Matrix" << std::endl;
14 matrix.Print();
15 std::cout << "Test vector" << std::endl;</pre>
_{16} vector.Print();
17 std::cout << "Lower-Diagonal Matrix L" << std::endl;
18 L.Print();
{\tt 19} \ \mathtt{std}::\mathtt{cout} <<\ "Upper-Diagonal\ Matrix\ U" <<\ \mathtt{std}::\mathtt{end1};
20 U.Print();
  Console Output
  Test Matrix
                                                0.71915
  0.712955
                 0.428471
                                 0.690885
  0.491119
                  0.780028
                                 0.410924
                                                0.579694
  0.139951
                  0.401018
                                 0.627317
                                                0.324151
  0.244759
                 0.694755
                                 0.593902
                                                0.631792
  Test vector
                  1
                                 1
                                                1
  Lower-Diagonal Matrix L
                 0
                                 0
                                                0
  3.50923
                  1
                                 0
                                                0
  5.09433
                  245.257
                                 1
                                                0
  1.7489
                  95.2855
                                 0.381754
                                                1
  Upper-Diagonal Matrix U
  0.139951
                  0.401018
                                 0.627317
                                                0.324151
                  -0.00658268 -0.50321
  0
                                                0.0648859
                                                -16.8459
  0
                  0
                                 120.911
  0
                  0
                                 0
                                                -0.309529
```

4.3 Linear Least Squares Problems

This section deals with the algorithms used to solve Least Squares Problems. That is, solutions for the algebraic problem

$$min_x||b - Ax||_2$$

This is particularly useful in problems dealing with data fitting.

4.3.1 Least Squares

Description: This algorithm solves the equation

$$min_x||b - Ax||_2$$

for x using Cholesky Decomposition followed by forward and back substitution. Note that within this method we call three other methods, which are CholeskyDecomposition, ForwardSubstitution, and Back-Substitution. The details for these methods can be found earlier in this section.

Input:

- 1. Matrix A (system of equations)
- 2. Vector B (right-hand-side)

Output: Vector (solution to the system of equations)

Code Written:

```
/// A Least Squares algorithm via Normal Equations
  /// Requires a matrix A and a vector b
3 Vector LeastSquares(Matrix A, Vector b) {
     {\tt Matrix}\ {\tt AT}\ =\ {\tt A.Transpose}\,(\,)\;;
     \mathtt{Matrix}\ \mathtt{B}\ =\ \mathtt{AT}\ *\ \mathtt{A}\,;
     Vector y = AT * b;
6
     std::cout << "B:" << std::endl;
     B.Print();
9
     std::cout << "Y: " << std::endl;
10
     y.Print();
11
12
     Matrix G = CholeskyDecomposition(B);
13
     Vector z = ForwardSubstitution(G, y);
14
     {\tt Vector} \  \, {\tt x} \, = \, {\tt BackSubstitution} \, ({\tt G.Transpose} \, (\,) \; , \; \, {\tt z} \, ) \; ; \\
15
16
17
      return x;
```

```
//Written by Andrew Sheridan in C++

matrix[0][0] = 1;

matrix[0][1] = 0;

matrix[0][2] = 1;

matrix[1][0] = 2;

matrix[1][1] = 3;
```

```
7 matrix[1][2] = 5;
8 matrix[2][0] = 5;
9 matrix[2][1] = 3;
10 matrix[3][0] = 3;
11 matrix[3][1] = 5;
12 matrix[3][2] = 4;
13 matrix[4][0] = -1;
14 matrix[4][0] = -1;
15 matrix[4][1] = 6;
16 matrix[4][2] = 3;
17 Vector vector = Vector(5);
18 vector[0] = 4;
19 vector[1] = -2;
20 vector[2] = 5;
21 vector[3] = -2;
22 vector[4] = 1;
23
24 Vector result = LeastSquares(matrix, vector);
25 result.Print();
```

B:					
40	30	10			
30	79	47			
10	47	55			
Υ:					
18	5	-21			
0.347226	0.399004	-0.785917			

4.3.2 Gram Schmidt

Description: Computes the QR factorization of a matrix.

Input:

1. Matrix A

Output: Matrix* This is an array of Matrices with two entries. The first is Q, the second, R.

Code Written:

```
1 ///Computes the QR factorization of Matrix A
  ///Returns a pair of matrices in an array. The first is Q, the second, ←
3 Matrix* GramSchmidt(Matrix A) {
     if (A.GetRows() != A.GetColumns()) return NULL;
4
5
     {\tt Matrix} \ {\tt r(A.GetRows()} \ , \ {\tt A.GetColumns())} \ ;
     {\tt Matrix} \ \ q({\tt A.GetRows()} \ , \ \ {\tt A.GetColumns()} \ ) \ ;
     for (int k = 0; k < A.GetRows(); k++) {
       r[k][k] = 0;
10
       \quad \text{for (int i = 0; i < A.GetRows(); i++)} \\
11
         r[k][k] = r[k][k] + A[i][k] * A[i][k];
12
13
       r[k][k] = sqrt(r[k][k]);
14
15
       \quad \text{for (int i = 0; i < A.GetRows(); i++)} \\
16
17
         q[i][k] = A[i][k] / r[k][k];
18
        for (int j = k + 1; j < A.GetColumns(); j++) {
19
         r[k][j] = 0;
20
          for (int i = 0; i < A.GetRows(); i++)</pre>
21
22
            r[k][j] += q[i][k] * A[i][j];
23
          for (int i = 0; i < A.GetRows(); i++)
24
            A[i][j] = A[i][j] - r[k][j] * q[i][k];
25
26
27
     Matrix* QR = new Matrix[2];
28
     \mathtt{QR}\,[\,0\,] \ = \ \mathtt{q}\,;
29
30
     QR[1] = r;
     return QR;
31
32 }
```

```
//Written by Andrew Sheridan in C++
2 Matrix matrix = MatrixFactory::Instance()->DiagonallyDominant(5, 5);
3 Matrix* QR = GramSchmidt(matrix);
4 Matrix Q = QR[0];
5 Matrix R = QR[1];
6
7 std::cout << "Starting Matrix: " << std::endl;
8 matrix.Print();
9 std::cout << "Q: " << std::endl;
10 Q.Print();
11 std::cout << "R: " << std::endl;
```

```
12 R. Print();
14 std::cout << "Q * R" << std::endl;
_{15} Matrix TestResult = Q * R;
16 TestResult.Print();
17
_{18} Matrix Norm = matrix - TestResult;
\begin{array}{lll} & \texttt{double} & \texttt{normValue} = \texttt{Norm.OneNorm()}; \\ & \texttt{20} & \texttt{std::cout} << " \mid \mid A - (Q * R) \mid \mid & : " << \texttt{normValue} << \texttt{std::endl}; \\ \end{array}
  Console Output
   _____
  Starting Matrix:
  50.713
                 0.428471
                                 0.690885
                                                0.71915
                                                               0.491119
  0.780028
                 50.4109
                                 0.579694
                                                0.139951
                                                               0.401018
                                 50.2448
  0.627317
                 0.324151
                                                0.694755
                                                               0.593902
  0.631792
                 0.440257
                                 0.0837265
                                                50.7123
                                                               0.427863
  0.29778
                 0.492085
                                 0.740296
                                                0.357729
                                                               50.4172
  Q:
  0.99971
                 -0.0156152
                                 -0.0123673
                                                -0.0123431
                                                               -0.00545983
  0.0153768
                 0.999777
                                 -0.00647483 -0.00868187
                                                              -0.00956081
  0.0123664
                 0.00613281
                                 0.999798
                                                -0.00149318 -0.0145344
  0.0124546
                 0.00843422
                               0.00118936
                                              0.999863
                                                               -0.00674982
  0.00587018
                0.00962131
                                 0.0144128
                                                0.00657804
                                                               0.999811
  R.:
  50.7276
                 1.21588
                                 1.32633
                                                1.36339
                                                               0.805775
  0
                 50.4034
                                 0.884747
                                                0.564112
                                                               0.88559
                 0
                                 50.2331
                                                0.750286
                                                               1.31227
  0
  0
                 0
                                                50.6966
                                                               0.749021
                                 0
                 0
  0
                                 0
                                                0
                                                               50.3896
  Q * R
                 0.428471
                                 0.690885
                                                               0.491119
  50.713
                                                0.71915
                 50.4109
  0.780028
                                 0.579694
                                                0.139951
                                                               0.401018
  0.627317
                 0.324151
                                 50.2448
                                                0.694755
                                                               0.593902
  0.631792
                 0.440257
                                 0.0837265
                                                50.7123
                                                               0.427863
```

0.740296

||A - (Q * R)|| : 7.21645e-15

0.492085

0.29778

0.357729

50.4172

4.4 Iterative Methods

This section will approach the problem of solving linear systems of equation via iterative methods. This is beneficial because iterative methods can be less expensive, particularly when dealing with sparse systems of equations.

4.4.1 Jacobi Iteration

Description: Solves a system of equations using Jacobi Iteration

Input:

- 1. Matrix A
- 2. Vector x0 (initial guess)
- 3. Vector b (Right-Hand-Side)
- 4. int maxIterations
- 5. double tolerance

Output: Vector

Code Written:

```
///Solves the system of equations using Jacobi Iteration
  ///A: The Matrix
3//x0: The initial guess
   ///b: The Right-Hand-Side
5 Vector JacobiIteration(Matrix A, Vector xO, Vector b, int ←
       maxIterations, double tolerance) {
     int iterations = 0;
     int n = A.GetRows();
     {\tt Vector\ newX\,(x0)}\,;
     double error = 10 * tolerance;
     while(iterations < maxIterations && tolerance < error){</pre>
10
       Vector oldX(newX);
11
        for (int i = 0; i < n; i++) {
12
          \mathtt{newX}\,[\,\mathtt{i}\,] \;=\; \mathtt{b}\,[\,\mathtt{i}\,]\,;
13
          for (int j = 0; j < i; j++) {
  newX[i] = newX[i] - A[i][j] * oldX[j];</pre>
15
16
          17
18
19
          newX[i] = newX[i] / A[i][i];
20
          \mathtt{error} \ = \ (\mathtt{oldX} \ - \ \mathtt{newX}) \, . \, \mathtt{L2Norm} \, (\,) \; ;
21
          iterations++;
23
24
     return newX;
26
```

```
1 //Written by Andrew Sheridan in C++
2 Matrix m1(3);
3 m1 [0][0] = 7;

4 m1 [0][1] = 3;

5 m1 [0][2] = 1;
6 m1[1][0] = -3;
7 m1[1][1] = 10;
8 m1[1][2] = 2;
^{13}~\texttt{Vector}~\texttt{b1}\,(3)\,;
16 \ b1[2] = 2;
_{18} Vector x1(3);
19
_{20} int maxIter = 10000;
\begin{array}{lll} {\tt 21} & {\tt double} & {\tt tolerance} = 0.00001; \\ \end{array}
22
{\tt 24} \ \ {\tt Vector} \ \ {\tt result1} = {\tt JacobiIteration(m1, x1, b1, maxIter, tolerance);}
25 result1.Print();
```

The results of Jacobi Iteration with max iterations of 10000 and tolerance of 1e-05 0.223242 0.448775 0.0909775

4.4.2 Conjugate Gradient Method

Description: Uses the iterative method of the Conjugate Gradient Method to solve system of linear equations starting with an initial guess.

Input:

- 1. Matrix A
- 2. Vector b (Right-Hand-Side)
- 3. Vector x0 (initial guess)
- 4. double tolerance

Output: Vector (the solution to the system of equations)

Code Written:

```
///Solves a matrix and RHS using Conjugate Gradient Method
  ///A : The matrix to be solved
3 ///b : The RHS vector 4 ///x0 : The initial guess vector
   ///tol : The tolerance of our method
_{6} Vector ConjugateGradient(Matrix A, Vector b, Vector x0, double tol) {
     \texttt{Vector rk} = \texttt{b} - (\texttt{A} * \texttt{x0});
      double dk = rk * rk;
      double bd = b * b;
     int k = 0;
10
      Vector pk = rk;
11
12
      {\tt Vector} \ {\tt xk} \, = \, {\tt x0} \, ;
      while (dk > tol * tol * bd) {
13
        \label{eq:Vector_sk} \hat{\mbox{Vector}} \ \ \hat{\mbox{Sk}} \ = \ \mbox{A} \ \ * \ \mbox{pk} \, ;
14
        double ak = dk / (pk * sk);
        \texttt{Vector xkp1} = \texttt{xk} + (\texttt{ak} * \texttt{pk});
16
        Vector rkp1 = rk - (ak * sk);
17
18
        double dkp1 = rkp1 * rkp1;
        Vector pkp1 = rkp1 + ((dkp1 / dk) * pk);
19
20
        k++;
        //All values have been computed, set all kth values to equal the k \leftarrow
             +1 value in preparation for next iteration
        xk = xkp1;
        rk = rkp1;
23
        {\tt pk} \; = \; {\tt pkp1} \, ; \\
24
25
        dk = dkp1;
26
27
      return xk;
```

```
//Written by Andrew Sheridan in C++
2 Matrix m5(3); //Initialize 3x3 matrix
3 m5[0][0] = 7; //Assign values
4 m5[0][1] = 3;
5 m5[0][2] = 1;
6 m5[1][0] = 3;
7 m5[1][1] = 10;
```

```
8 m5[1][2] = 2;
9 m5[2][0] = 1;
10 m5[2][1] = 2;
11 m5[2][2] = 15;

12
13 Vector b5(3);
14 b5[0] = 28;
15 b5[1] = 31;
16 b5[2] = 22;
17
18 Vector x5(3);
19
20 Vector result5 = ConjugateGradient(m5, b5, x5, 0.000001);
21 std::cout << "Result of conjugate gradient method: " << std::endl;
22 result5.Print();</pre>
Console Output
-----
Result of conjugate gradient method:------
3 2 1
```

4.5 Eigenvalues and Singular Values

This section contains methods which aid in the approximation of Eigenvalues and Singular values.

4.5.1 PowerMethod

Description: Finds an approximation of the largest eigenvalue of a matrix.

Input:

- 1. Matrix A (the Matrix whose eigenvalue will be approximated)
- 2. Vector x0 (the initial guess)
- 3. double tol (the tolerance of the algorithm)
- 4. int maxIter (the maximum number of iterations)

Output: double (the approximation of the largest eigenvalue)

Code Written:

```
///Finds an approximation of the largest Eigenvalue of a matrix
2 ///A : The square matrix
3 ///x0 : The initial guess
4 ///tol : The tolerance of the algorithm
_{5} ///maxIter : The maximum number of iterations to be executed by the \hookleftarrow
6 double PowerMethod(Matrix A, Vector x0, double tol, int maxIter) {
     double error = 10 * tol;
     int k = 0;
     Vector y = A * x0;
9
10
     Vector xk = x0;
     double lambda_k = 0;
11
12
     while (error > tol && k < maxIter) {
       {\tt Vector\ xkp1\ =\ y\ /\ y.L2Norm\,(\,)\ ;}
14
       y = A * xkp1;
15
       double lambda_kp1 = xkp1 * y;
16
       {\tt error} \, = \, {\tt abs} \, (\, {\tt lambda\_kp1} \, - \, {\tt lambda\_k} \, ) \, ;
17
18
       lambda_k = lambda_kp1;
19
20
       k++;
21
22
23
     return lambda_k;
24
```

```
//Written by Andrew Sheridan in C++

Matrix A(2); //New 2x2 matrix

A[0][0] = 2; //Initialize entries

A[0][1] = -12;

A[1][0] = 1;

A[1][1] = -5;
```

```
8
9    Vector x_0(2); //New vector size 2
10    x_0[0] = 1; //Initialize entries
11    x_0[1] = 1;
12
13    //Find the largest eigenvalue of A
14    double lambda = PowerMethod(A, x_0, 0.0001, 1000);
15    std::cout << "Lambda: " << lambda << std::endl;</pre>
```

Console Output

Lambda: -2.00005

4.5.2 InversePowerMethod

Description: Approximates the smallest eigenvalue of a matrix.

Input: Finds an approximation of the smallest Eigenvalue of a matrix

- 1. Matrix A (square matrix whose smallest eigenvalue will be approximated)
- 2. Vector x0 (initial guess vector)
- 3. double tol (error tolerance)
- 4. int maxIter (maximum number of iterations)

Output: double (approximation of the smallest eigenvalue)

Code Written:

```
1 ///Finds an approximation of the smallest Eigenvalue of a matrix
  ///A : The square matrix
3 ///x0: The initial guess
4 ///tol: The tolerance of the algorithm
_{5} ///maxIter : The maximum number of iterations to be executed by the \leftrightarrow
_{6} double InversePowerMethod(Matrix A, Vector x0, double tol, int maxIter\hookleftarrow
       ) {
     int k = 0:
8
     Matrix U = A;
    Matrix L = LUFactorization(U, x0); // L is returned, U is modified to \leftarrow
10
          be upper triangular
     Vector y = BackSubstitution(U, x0); // Solve for y by doing back <math>\leftarrow
11
          substitution.
     double lambda_x = 0;
     while (error > tol && k < maxIter) {</pre>
13
14
       {\tt Vector} \ {\tt x} \ = \ {\tt y} \ / \ {\tt y.L2Norm} \, (\,) \; ;
       y = SolveSystem(A, x); //Does Gaussian Elimination then Back <math>\leftarrow
15
            Substitution
16
       double lambda_xp1 = x * y;
       error = abs(lambda_xp1 - lambda_x);
17
18
19
       lambda_x = lambda_xp1;
       k++;
20
     }
21
     output.close();
23
^{24}
     return lambda_x;
25 }
```

Usage Sample:

```
//Written by Andrew Sheridan in C++

Matrix A(2);
A[0][0] = 2;
A[0][1] = -12;
A[1][0] = 1;
A[1][1] = -5;
```

Console Output

Lambda: -0.999999

5 Citations

6 Appendices

6.1 Appendix B: Vector Class

```
1 #pragma once
2 //Andrew Sheridan
3 //Math 5610
4 //Written in C++
5 // Vector.h
7 class Vector {
s public:
    //Initialization and Destruction
10
     Vector();
     Vector(unsigned n);
11
     {\tt Vector}\left( \begin{array}{c} {\tt const} \end{array} \right. \, {\tt Vector} \, \left. \& {\tt v} \right);
12
     Vector(double* v, unsigned size);
13
     Vector operator=(const Vector& v);
14
     ~Vector();
16
17
     void InitializeRandomEntries();
     void InitializeAllOnes();
18
19
     //Overloaded Operators
20
     double& operator[] (unsigned x) { return entries[x]; }
21
     friend double operator*( Vector& a, Vector& b);
22
     friend Vector operator*(Vector& a, double constant);
     friend Vector operator*(double constant, Vector& a);
^{24}
     friend Vector operator+(Vector& a, Vector& b);
25
     friend Vector operator/(Vector& a, double constant);
friend Vector operator+(Vector& a, Vector& b);
27
28
29
     //Basic Algorithms
30
     double FindMaxMagnitudeStartingAt(unsigned start);
31
    unsigned FindMaxIndex();
32
    double L2Norm();
33
    //Accessing Data
35
    void Print();
    unsigned GetSize() { return size; }
void SetSize(unsigned newSize) { size = newSize; }
37
38
40 protected:
    unsigned size;
41
43 private:
    double* entries; //The stored values of the vector
44
45 };
```

```
1 //Andrew Sheridan
2 //Math 5610
3 //Written in C++
4 // Vector.cpp
6 #pragma once
7 #include "Vector.h"
* #include <random>
9 #include <iostream>
10 #include <iomanip>
12 ///Default constructor
13 Vector::Vector():size(0){}
15 /// Initializes the entries to an empty array of size n
16 Vector::Vector(unsigned n) : size(n){
    entries = new double[size];
     for (unsigned i = 0; i < n; i++) {
18
19
      entries[i] = 0;
20
21 }
22
23 ///Copy Constructor
24 Vector:: Vector (const Vector &v) : size(v.size) {
    entries = new double[size];
     for (unsigned i = 0; i < size; i++) {
26
       entries[i] = v.entries[i];
27
29 }
30
31 ///Copy Constructor
_{32} ///v: an array of doubles
33 ///n: the size of v
34 Vector::Vector(double* v, unsigned n) :size(n) {
     \mathtt{entries} \, = \, \underset{}{\mathrm{new}} \  \, \underset{}{\mathrm{double}} \, [\, \mathtt{size} \, ] \, ;
35
     for (unsigned i = 0; i < size; i++) {
       \mathtt{entries}\,[\,\mathtt{i}\,] \;=\; \mathtt{v}\,[\,\mathtt{i}\,]\,;
37
38
39 }
40
41 /// Copy Assignment Operator
42 Vector Vector::operator= (const Vector& v) {
   \mathtt{size} = \mathtt{v.size};
43
     entries = new double[v.size];
     for (unsigned i = 0; i < v.size; i++) {
45
       entries[i] = v.entries[i];
46
47
     return *this;
48
49 }
50
51 ///Destructor
52 Vector: ~ Vector()
53 {
54 }
55
56 ///Inner Product
57 double operator* (Vector& a, Vector& b) {
    if (a.size != b.size)
58
       return NULL;
59
     double sum = 0;
61
     for (int i = 0; i < a.size; i++) {
      \mathtt{sum} \mathrel{+}= \mathtt{a[i]} * \mathtt{b[i]};
62
63
```

```
64
     return sum;
66
   ///Multiply each entry by a constant
67
 68 Vector operator* (Vector& a, double constant) {
     Vector newVector(a.size);
69
 70
     for (int i = 0; i < a.size; i++) {
       newVector[i] = a[i] * constant;
71
 72
     return newVector;
 73
74 }
75
    ///Multiply each entry by a constant
 76
77 Vector operator* (double constant, Vector& a) {
     Vector newVector(a.GetSize());
     for (int i = 0; i < a.GetSize(); i++) {</pre>
 79
        \label{eq:newVector} \mbox{\tt newVector}\left[\,\mbox{\tt i}\,\right] \; = \; \mbox{\tt a}\left[\,\mbox{\tt i}\,\right] \; * \; \mbox{\tt constant} \; ;
80
      return newVector;
 82
 83 }
s_5 ///Subtracts the elements of vector a from the elements of vector b s_6 Vector operator- (Vector& a, Vector& b) {
     if (a.size != b.size) return a;
87
      Vector newVector(a.size);
 88
 89
     for (int i = 0; i < a.size; i++) {
       {\tt newVector[i]} \, = \, {\tt a[i]} \, - \, {\tt b[i]};
90
91
 92
     return newVector;
93 }
94
    ///Divides the entries of the vector by a constant
95
96 Vector operator/ (Vector& a, double constant) {
     Vector newVector(a.GetSize());
      for (int i = 0; i < a.GetSize(); i++) {</pre>
98
        newVector[i] = a[i] / constant;
99
100
     return newVector:
101
102 }
103
104 ///Returns a new vector where its entries are the sum of corresponding ←
         entries in a and b
105 Vector operator+ (Vector& a, Vector& b) {
     if (a.size != b.size) {
106
107
        std::cout << "Vectors are not same size. Returning first vector." <math>\leftarrow
           << std::endl;
108
        return a;
109
      Vector newVector(a.size);
110
     for (int i = 0; i < a.size; i++) {
112
       newVector[i] = a[i] + b[i];
113
     return newVector;
114
115 }
116
117 /// Initializes the entries to values between 0 and 1
118 void Vector::InitializeRandomEntries() {
      std::mt19937 generator(123); //Random number generator
      std::uniform_real_distribution<double> dis(0.0, 1.0); //Desired \leftarrow
120
          distribution
      for (unsigned i = 0; i < size; i++) {
122
        entries[i] = dis(generator); //Assign each entry in entries to ←
123
             random number between 0 and 1
```

```
124
125 }
126
127 /// Initializes the entries to 1
128 void Vector::InitializeAllOnes() {
    for (unsigned i = 0; i < size; i++) {
  entries[i] = 1; //Assign each entry in entries to 1</pre>
129
130
131
132 }
133
134 ///Finds the entry with the largest magnitude, starting with entry "←
        start".
135 double Vector::FindMaxMagnitudeStartingAt(unsigned start) {
     double max = 0;
136
137
      for (unsigned i = start; i < size; i++) {</pre>
        double value = std::abs(entries[i]);
138
        if (value > max) max = value;
139
140
141
      return max;
142 }
143
144 ///Finds the index of the value with the largest magnitude
145 unsigned Vector::FindMaxIndex() {
     double max = 0;
146
      unsigned index = -1;
147
      for (unsigned i = 0; i < size; i++) {
148
       double value = std::abs(entries[i]);
149
        \quad \  \  \text{if} \ \ (\, \texttt{value} \, > \, \texttt{max}\,) \ \ \{ \\
150
151
           {\tt max} = {\tt value};
           index = i;
152
        }
153
154
      return index;
155
157
_{158} ///Computes the L2 norm of the vector
159 double Vector::L2Norm() {
      double sum = 0;
160
      for (int i = 0; i < size; i++) {
161
       sum += (entries[i] * entries[i]);
162
163
164
      return std::sqrt(sum);
165 }
166
167 ///Outputs the vector's entries to the console
168 void Vector::Print() {
     for (unsigned i = 0; i < size; i++) {
169
        std::cout << std::setw(13) << std::left << entries[i];
170
171
172
      \mathtt{std}::\mathtt{cout} <\!< \mathtt{std}::\mathtt{endl} <\!< \mathtt{std}::\mathtt{endl};
173 }
```

6.2 Appendix C: Matrix Class

```
1 //Andrew Sheridan
2 //Math 5610
3 //Written in C++
4 // Matrix.h
6 #pragma once
7 #include "Vector.h"
9 class Matrix{
10 public:
     //Initialization and Deconstruction
    Matrix() = default;
12
    Matrix(unsigned size);
13
    Matrix(unsigned rowCount, unsigned columnCount);
14
    Matrix (const Matrix &m);
15
    Matrix operator = (const Matrix& m);
16
     ~Matrix();
17
18
    void InitializeIdentityMatrix();
    void InitializeRandom();
20
    void InitializeRange(double minValue, double maxValue);
21
    void InitializeDiagonallyDominant();
22
23
     //Overloaded Operators
24
    Vector & operator[] (unsigned row) { return entries[row]; }
25
    friend bool operator == (const Matrix& A, const Matrix& B);
26
    friend bool operator != (const Matrix& A, const Matrix& B);
27
    friend Vector operator * (const Matrix& A, Vector& x);
28
    friend Vector operator / (const Matrix& A, Vector& x);
29
    friend Matrix operator * (Matrix A, Matrix B);
friend Matrix operator - (Matrix A, Matrix B);
30
31
32
33
     //Basic Algorithms
    bool IsSymmetric();
34
    Matrix Transpose();
    double OneNorm();
36
    double InfinityNorm();
37
    //Getters and Setters
39
    unsigned GetRows() { return rows; }
40
    unsigned GetColumns() { return columns; }
41
    void SetRows(unsigned r) { rows = r; }
42
43
    void SetColumns(unsigned c) { columns = c; }
44
    //Output
45
46
    void Print();
    void PrintAugmented(Vector v);
47
48
49 private:
    Vector* entries; //The entries of the matrix
50
52
    unsigned rows;
    unsigned columns;
53
```

```
1 //Andrew Sheridan
2 //Math 5610
3 //Written in C++
4 // Matrix.cpp
6 #include "Matrix.h"
7 #include "Vector.h"
s #include <random>
9 #include <iostream>
10 #include <iomanip>
11
_{13} #pragma region Constructors and Initialization
14 ///Initializes an nxn matrix of size "size". All entries are zeroes.
15 Matrix::Matrix(unsigned size) : rows(size), columns(size)
16 {
17
     entries = new Vector[size];
     for (unsigned i = 0; i < size; i++) {
18
19
        entries[i] = Vector(size);
        for (unsigned j = 0; j < size; j++) {
20
          entries[i][j] = 0;
21
22
23
     }
24 }
26 ///Initializes an nxm matrix. All entries are zeroes.
27 Matrix::Matrix(unsigned rowCount, unsigned columnCount) : rows(\leftarrow
       rowCount), columns(columnCount) {
     \mathtt{entries} \ = \ \overset{\cdot}{\mathrm{new}} \ \mathtt{Vector} \left[ \ \mathtt{rowCount} \ \right];
28
     for (unsigned i = 0; i < rowCount; i++) {</pre>
29
       entries[i] = Vector(columnCount);
30
        for (unsigned j = 0; j < columnCount; j++) {
31
32
          entries[i][j] = 0;
       }
33
     }
34
  }
35
36
37 ///Copy Constructor
38 Matrix::Matrix(const Matrix &m): rows(m.rows), columns(m.columns) {
     \mathtt{entries} \ = \ \underset{}{\mathrm{new}} \ \mathtt{Vector} \left[ \ \mathtt{rows} \ \right];
39
     for (unsigned int i = 0; i < rows; i++) {
        entries[i] = Vector(columns);
41
        for (unsigned int j = 0; j < columns; j++) {
42
          entries[i][j] = m.entries[i][j];
44
       }
     }
45
46 }
47
  //Deconstructor
49 Matrix: Matrix() {
50 }
52 ///Assignment operator overload
53 Matrix Matrix::operator=(const Matrix& m) {
     this \rightarrow rows = m.rows;
54
     \begin{array}{lll} t\,h\,i\,s\,-\!\!>\!\!columns\,=\,m\,.\,columns\,; \end{array}
55
     this -> entries = new Vector[m.rows];
     for (int i = 0; i < m.rows; i++) {
57
       this->entries[i] = m.entries[i];
58
59
60
61
     return *this;
62 }
```

```
63
   ///Initializes the matrix to have ones on the main diagonal
65
   void Matrix::InitializeIdentityMatrix() {
     for (unsigned i = 0; i < rows; i++) {
67
       for (unsigned j = 0; j < i; j++) {
68
69
         entries[i][j] = 0;
70
       entries [i][i] = 1;
71
       for (unsigned j = i + 1; j < rows; j++) {
72
         entries[i][j] = 0;
73
74
75
     }
76 }
   ///Sets the values of all entries between 0 and 1
78
   void Matrix::InitializeRandom() {
79
     std::mt19937 generator(123); //Random number generator
     std::uniform_real_distribution<double> dis(0.0, 1.0); //Desired \leftrightarrow
81
         distribution
82
     \quad \text{for (unsigned i} = 0; \ i < \text{columns}; \ i++) \ \{
83
        for (unsigned j = 0; j < rows; j++) {
84
          85
               random number between 0 and 1
86
     }
87
88 }
89
90 ///Sets the values of all entries between minValue and maxValue
_{91} void Matrix::InitializeRange(double minValue, double maxValue) {
     std::mt19937 generator(123); //Random number generator
92
     std::uniform_real_distribution < double > dis(minValue, maxValue); //
93
         Desired distribution
94
     for (unsigned i = 0; i < columns; i++) {
95
        for (unsigned j = 0; j < rows; j++) {
96
          \texttt{entries[i][j]} = \texttt{dis(generator)}; \text{ //Assign each entry in matrix } to \leftarrow
97
               random number between 0 and 1
       }
98
     }
99
100
101
   /// Sets the values of the entries to be diagonally dominant
102
103
   void Matrix::InitializeDiagonallyDominant() {
     std::mt19937 generator(123); //Random number generator
104
     std::uniform\_real\_distribution < double > dis(0.0, 1.0); //Desired \leftarrow
105
          distribution
106
     for (unsigned i = 0; i < rows; i++) {
107
       for (unsigned j = 0; j < columns; j++) { entries[i][j] = dis(generator); //Assign each entry in matrix to\leftarrow
108
109
               random number between 0 and 1
110
     }
111
112
     for (unsigned k = 0; k < rows; k++) {
113
114
       entries[k][k] += 10 * rows; //Add 10*n to all diagonal entries
115
116 }
117
118 #pragma endregion
120 #pragma region Printing
```

```
121 ///Outputs matrix to the console
122 void Matrix::Print() {
      for (unsigned i = 0; i < rows; i++) {
123
        for (unsigned j = 0; j < columns; j++) {
124
          std::cout << std::setw(13) << std::left << entries[i][j];
125
126
127
        std::cout << std::endl;</pre>
128
      \mathtt{std}::\mathtt{cout} <\!< \mathtt{std}::\mathtt{endl}\;;
129
130
131
132 ///Outputs an augmented coefficient matrix to the console
   void Matrix::PrintAugmented(Vector v) {
133
     if (v.GetSize() != rows) {
  std::cout << "Vector and Matrix do not match sizes" << std::endl;</pre>
134
135
        return;
136
137
138
      for (unsigned i = 0; i < rows; i++) {
139
        for (unsigned j = 0; j < columns; j++) {
140
          std::cout << std::setw(13) << std::left << entries[i][j];
141
142
        std::cout << " | " << v[i] << std::endl;
143
144
      std::cout << std::endl;</pre>
145
146
147 #pragma endregion
148
149 #pragma region Comparison Operations
   ///Compares two matrices. Returns true if all corresponding entries \hookleftarrow
150
        are equal, and false if any are not equal.
   bool operator == (const Matrix& A, const Matrix& B) {
151
      if (A.columns != B.columns) return false;
152
      if (A.rows != B.rows) return false;
154
      \quad \text{for (unsigned i = 0; i < A.rows; i++) } \{
155
        for (unsigned j = 0; j < A.columns; j++) {
  if (A.entries[i][j] != B.entries[i][j]) {</pre>
156
157
158
             return false;
159
160
        }
161
      return true;
162
163 }
164
   ///Compares two matrices. Returns true if any corresponding entries ←
165
        are not equal, and false if all are equal.
   bool operator != (const Matrix& A, const Matrix& B) {
166
      if (A.columns != B.columns) return true;
167
      if (A.rows != B.rows) return true;
168
169
170
      for (unsigned i = 0; i < A.rows; i++) {
        for (unsigned j = 0; j < A.columns; j++) {
  if (A.entries[i][j] != B.entries[i][j]) {</pre>
171
172
173
             return true;
174
        }
175
      }
176
177
      return false;
178 }
180 ///Checks to see if matrix is symmetric
181 bool Matrix::IsSymmetric() {
     for (unsigned int i = 0; i < rows; i++) {
```

```
for (unsigned int j = 0; j \ll i; j++) {
183
          if (entries[i][j] != entries[j][i])
184
            return false;
185
186
187
     return true;
188
189
190
191 #pragma endregion
192
193
194 #pragma region operations
   /// Returns the transpose of the n by n matrix A
196 Matrix Matrix::Transpose() {
197
     Matrix matrix(columns, rows);
198
     for (unsigned i = 0; i < rows; i++) {
199
        for (unsigned j = 0; j < columns; j++) {
200
          matrix[j][i] = entries[i][j];
201
202
203
204
     return matrix;
205 }
206
   ///Multiplies an nxn matrix A by the vector x
207
   Vector operator *(const Matrix& A, Vector& x) {
    if (A.columns != x.GetSize()) return NULL;
209
210
211
     Vector result(A.rows);
     for (unsigned i = 0; i < A.rows; i++) {
212
213
        result[i] = 0;
        \label{eq:formula} \mbox{for (unsigned j = 0; j < A.columns; j++) } \{
214
          result[i] += A.entries[i][j] * x[j];
215
216
217
     }
218
     return result;
219 }
220
   ///Divides an nxn matrix A by the vector x
221
222 Vector operator /(const Matrix& A, Vector& x) {
     if (A.columns != x.GetSize()) return NULL;
223
224
     Vector result(A.rows);
225
     for (unsigned i = 0; i < A.rows; i++) {
226
        result[i] = 0;
        for (unsigned j = 0; j < A.columns; j++) {
228
          \texttt{result[i]} \; +\!\!\!= \; \texttt{A.entries[i][j]} \; / \; \texttt{x[j]};
220
230
231
232
     {\color{red} \textbf{return}} \ {\color{red} \textbf{result}};
233 }
234
   ///Multiplies an matrix A by matrix B
236
237 Matrix operator* (Matrix A, Matrix B) {
     if (A.columns != B.rows) throw "Incompatible sizes";
238
239
240
     Matrix matrix(A.rows, B.columns);
     Matrix bTranspose = B.Transpose();
241
242
     //std::cout << "Starting multiplication. A:" << std::endl;
243
     /*A. Print();
244
     std::cout << "B: " << std::endl;
245
     B. Print();
```

```
std::cout << "Bt: " << std::endl;
247
      bTranspose.Print();*/
^{248}
249
      for (unsigned i = 0; i < A.rows; i++) {
250
        for (unsigned j = 0; j < B.columns; j++) {
251
          // \, matrix \, [\,i\,] \, [\,j\,] \, = \, Dot Product \, (A[\,i\,] \,, \, \, bTranspose \, [\,j\,] \,, \, \, A. \, columns \, ) \, ;
252
253
          matrix[i][j] = A[i] * bTranspose[j];
254
255
256
      return matrix;
257 }
258
259
    //Multiplies an matrix A by matrix B
260 Matrix operator - (Matrix A, Matrix B) {
      if (A.columns != B.columns && A.rows != B.rows) throw "Incompatible ←
          sizes";
      Matrix matrix(A.rows, B.columns);
262
263
      for (unsigned i = 0; i < A.rows; i++) {
        for (unsigned j = 0; j < A.columns; j++) {
264
          matrix[i][j] = A[i][j] - B[i][j];
265
266
267
     }
268
      {\color{red} \textbf{return}} \ \ \textbf{matrix} \; ;
269 }
270
    ///Computes the 1-norm of the matrix
271
272 double Matrix::OneNorm() {
      double columnMax = 0;
273
      for (unsigned i = 0; i < rows; i++) {
        double columnSum = 0;
275
        for (unsigned j = 0; j < columns; j++) {
276
          columnSum += std::abs(entries[i][j]);
277
278
279
        if (columnSum > columnMax)
          columnMax = columnSum;
280
281
      return columnMax;
282
283 }
284
   ///Computes the inifinity norm of an n by n matrix A
285
286 double Matrix::InfinityNorm() {
287
      double rowMax = 0;
      for (unsigned j = 0; j < columns; j++) {
288
        double rowSum = 0;
289
        for (unsigned i = 0; i < rows; i++) {
290
          rowSum += std::abs(entries[i][j]);
291
292
        if (rowSum > rowMax)
293
          rowMax = rowSum:
294
295
296
      return rowMax;
297 }
298
299 #pragma endregion
```

6.3 Appendix D: Matrix Factory

This component is used to create matrices of various types, such as symmetric or diagonally dominant matrices. It was created in order to simplify the matrix.cpp file, and to keep the initialization logic separate from the operation logic.

6.3.1 Usage:

Because this component is a singleton, the syntax to use it is more complex, but useful because a new matrix can be created in a single line of your main code. If we wanted to create a new identity matrix, the code would look like this.

```
Matrix A = MatrixFactory::Instance() → Identity(rowCount, columnCount → );
```

This way, there need not be any declaration or construction of an instance of the Matrix Factory. We can just ask the singleton for an instance, and use that instance to product the specified matrix, all in one line.

6.3.2 Header File

```
//Andrew Sheridan
   //Math 5610
  //Written in C++
4 //MatrixFactory.h
5 #pragma once
6 #include "Matrix.h"
  ///A singleton which creates new matrices
9 class MatrixFactory {
10 public:
     static MatrixFactory* Instance();
11
    {\tt Matrix\ Identity} \, (\, {\tt unsigned\ rows} \, , \, \, {\tt unsigned\ columns} \, ) \, ;
12
    Matrix Ones (unsigned rows, unsigned columns);
    {\tt Matrix\ UpperTriangular(unsigned\ rows\,,\ unsigned\ columns)}\,;
14
    Matrix LowerTriangular(unsigned rows, unsigned columns);
15
     Matrix Random(unsigned rows, unsigned columns);
16
    {\tt Matrix\ RandomRange(unsigned\ rows\,,\ unsigned\ columns\,,\ double\ min\,,\ \hookleftarrow}
17
         double max);
     Matrix DiagonallyDominant(unsigned rows, unsigned columns);
    Matrix Symmetric(unsigned size);
19
    Matrix SPD(unsigned size);
20
21
22 private:
23
    MatrixFactory() {};
    MatrixFactory(MatrixFactory const&) {};
24
    MatrixFactory& operator=(MatrixFactory const&) {};
     static MatrixFactory* m_instance;
26
27 };
```

6.3.3 Code Written

```
1 //Andrew Sheridan
2 //Math 5610
3 //Written in C++
4 //MatrixFactory.cpp
6 #include "MatrixFactory.h"
7 #include <random>
9 MatrixFactory* MatrixFactory::m_instance = nullptr;
10
11 ///Returns the instance of the MatrixFactory singleton
12 MatrixFactory* MatrixFactory::Instance() {
if (!m_instance)
14
      m_instance = new MatrixFactory();
15
    return m_instance;
17 }
18
19 ///Creates an identity matrix
20 Matrix MatrixFactory:: Identity(unsigned rows, unsigned columns) {
21
  Matrix m(rows, columns);
   m.InitializeIdentityMatrix();
    return m;
23
24 }
_{26} ///Creates a matrix where every entry is a 1
27 Matrix MatrixFactory::Ones(unsigned rows, unsigned columns) {
   Matrix m(rows, columns);
28
29
    for (int i = 0; i < rows; i++) {
      for (int j = 0; j < columns; j++) {
  m[i][j] = 1;</pre>
30
31
      }
32
33
    }
34
    return m;
35 }
36
37 ///Creates a matrix where every entry is a value between 0 and 1
38 Matrix MatrixFactory::Random(unsigned rows, unsigned columns) {
   Matrix m(rows, columns);
39
40
    m.InitializeRandom();
41
    return m;
42 }
44 ///Creates a matrix where every entry is a value between minValue and \hookleftarrow
      maxValue
45 Matrix MatrixFactory::RandomRange(unsigned rows, unsigned columns, \leftarrow
      double minValue, double maxValue) {
    Matrix m(rows, columns);
    m.InitializeRange(minValue, maxValue);
47
    return m;
48
49 }
50
51 ///Creates an upper-triangular matrix
52 Matrix MatrixFactory::UpperTriangular(unsigned rows, unsigned columns)\hookleftarrow
53
    Matrix m(rows, columns);
    std::mt19937 generator(123); //Random number generator
    std::uniform_real_distribution<double> dis(0.0, 1.0); //Desired \leftarrow
55
        distribution
56
```

```
for (unsigned i = 0; i < rows; i++) {
57
         for (unsigned j = i; j < columns; j++) {
 58
           m[i][j] = dis(generator); //Assign entry in matrix to random \leftarrow
59
                number between 0 and 1
 60
        m[i][i] += 10 * rows; //Add 10*n to all diagonal entries
61
 62
63
64
      return m;
 65 }
66
 67 ///Creates an upper-triangular matrix
 68 Matrix MatrixFactory::LowerTriangular(unsigned rows, unsigned columns)\leftarrow
 69
      Matrix m(rows, columns);
      std::mt19937 generator(123); //Random number generator
 70
      \texttt{std}:: \texttt{uniform\_real\_distribution} < \overset{\frown}{double} > \ \texttt{dis}\left(0.0, \ 1.0\right); \ // \ \texttt{Desired} \ \leftarrow
 71
           distribution
72
      for (unsigned i = 0; i < rows; i++) {
 73
         for (unsigned j = 0; j \le i; j++) {
 74
           \texttt{m[i][j]} = \texttt{dis(generator)}; \; // \text{Assign entry in matrix to random} \; \leftarrow
75
                number between 0 and 1
 76
        m[i][i] += 10 * rows; //Add 10*n to all diagonal entries
77
 78
79
 80
      return m;
81 }
82
 83 ///Creates a Diagonally Dominant matrix
84 Matrix MatrixFactory::DiagonallyDominant(unsigned rows, unsigned \leftarrow
        columns) {
      Matrix m(rows, columns);
      m.InitializeDiagonallyDominant();
 86
 87
      return m;
88 }
89
90 ///Creates a symmetric matrix, where all entries have values between 0 \leftrightarrow
         and 1
91 Matrix MatrixFactory::Symmetric(unsigned size) {
      Matrix m(size);
 92
      std::mt19937 generator(123); //Random number generator
93
      \mathtt{std}:: \mathtt{uniform\_real\_distribution} < \mathtt{double} > \ \mathtt{dis} \, \big( \, 0.0 \, , \  \, 1.0 \big) \, ; \  \, \big/ / \, \mathtt{Desired} \, \, \hookleftarrow \,
94
           distribution
95
96
      for (unsigned i = 0; i < size; i++) {
         for (unsigned j = i; j < size; j++) {
97
           double value = dis(generator); //Assign each entry in matrix to \leftarrow
98
                random\ number\ between\ 0\ and\ 1
           \,\mathtt{m}\,[\,\mathtt{i}\,]\,[\,\mathtt{j}\,]\,\,=\,\,\mathtt{value}\,;
99
           m[j][i] = value;
100
101
102
103
      return m;
104
105 }
106
   ///Creates a symmetric positive definite matrix
107
108 Matrix MatrixFactory::SPD(unsigned size) {
      std::mt19937 generator(123); //Random number generator
109
      std::uniform_real_distribution<double> dis(0.0, 1.0); //Desired \leftrightarrow
110
           distribution
111
```

```
Matrix matrix(size);
112
113
         for (unsigned i = 0; i < size; i++) {
  for (unsigned j = i; j < size; j++) {
    double value = dis(generator); //Assign each entry in matrix to ←
    random number between 0 and 1</pre>
114
115
116
                 \mathtt{matrix}\,[\,\mathtt{i}\,]\,[\,\mathtt{j}\,] \;=\; \mathtt{value}\,;
117
                matrix[j][i] = value;
118
119
120
121
         for (unsigned k = 0; k < size; k++) { matrix[k][k] += 10 * size; //Add 10*n to all diagonal entries
122
123
124
125
         return matrix;
126
127 }
```

6.4 Appendix E: Complex Numbers

TODO

6.4.1 Usage:

Because this component is a singleton, the syntax to use it is more complex, but useful because a new matrix can be created in a single line of your main code. If we wanted to create a new identity matrix, the code would look like this.

```
\label{eq:additive} \begin{array}{ll} {\tt 1} \  \, {\tt Matrix} \  \, {\tt A} \, = \, {\tt MatrixFactory} :: {\tt Instance} \, () \, \longrightarrow \, {\tt Identity} ({\tt rowCount} \, , \, \, {\tt columnCount} \longleftrightarrow \, ) \, ; \end{array}
```

This way, there need not be any declaration or construction of an instance of the Matrix Factory. We can just ask the singleton for an instance, and use that instance to product the specified matrix, all in one line.

6.4.2 Header File

```
#pragma once
2 #include < cmath >
4 class Complex
5
6 public:
     double real;
7
     double imaginary;
     Complex (double newReal, double newImaginary)
10
11
       real = newReal;
       imaginary = newImaginary;
13
14
15
     Complex operator+(Complex c) {
16
17
       double realResult = real + c.real;
       double imaginaryResult = imaginary + c.imaginary;
18
19
       Complex result(realResult, imaginaryResult);
       return result;
20
     }
21
22
23
     Complex operator - (Complex c) {
       double realResult = real - c.real;
^{24}
25
       double imaginaryResult = imaginary - c.imaginary;
       Complex result(realResult, imaginaryResult);
26
       return result:
27
     }
29
30
     Complex operator*(Complex c) {
       double realResult = real * c.real - (imaginary * c.imaginary);
31
       \label{eq:condition} \textbf{double} \  \, \textbf{imaginaryResult} \, = \, \textbf{real} \, * \, \textbf{c.imaginary} \, + \, \textbf{c.real} \, * \, \textbf{imaginary};
32
33
       Complex result(realResult, imaginaryResult);
       return result;
34
     }
35
36
     Complex operator/(Complex c) {
37
38
       double realNumerator = real*c.real - imaginary*c.imaginary;
       double imaginaryNumerator = imaginary*c.real - real*c.imaginary;
39
       double denominator = (c.real*c.real) - (c.imaginary*c.imaginary);
40
```

```
return Complex(realNumerator / denominator, imaginaryNumerator / ← denominator);

denominator);

denominator);

denominator / ← denominator, imaginaryNumerator / ← denominator);

denominator);

denominator / denominator, imaginaryNumerator / ← denominator);

denomin
```