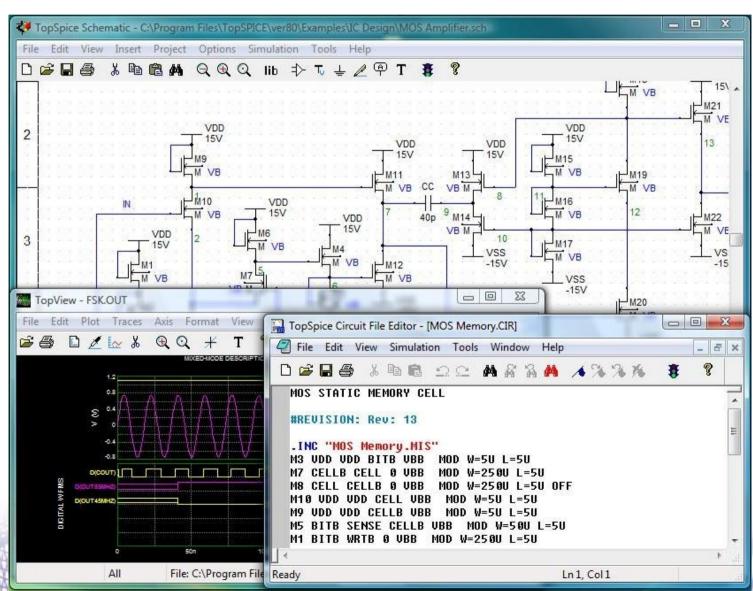
SPICE

 Simulation Program with Integrated Circuit Emphasis



SPICE under the covers

In a Nutshell: How SPICE Works

By Dr. Colin Warwick, Agilent Technologies, Inc.

Intended audience: This article won't help software engineers who have to implement circuit simulators: they invest time with the classic textbooks¹. But I believe that textbooks are over kill for SPICE users and that a shorter, simpler explanation is a better investment of time, hence this article.

Let's start simply with a onetime step (i.e. DC) solution of a circuit that consists of two unknown node voltages, V_1 , V_2 , a ground node V_0 , three known ohmic conductances, $G_{xy} = 1/R_{xy}$ (where $I_{xy} = G_{xy} \left(V_y - V_x \right)$ and x and y are the node indices), and three known current sources (Fig. 1).

You can solve a circuit using either Kirchhoff's current law or voltage law or both. These laws are named after the German physicist Gustav Robert Kirchhoff (1824–87)². SPICE is a modified nodal solver and uses the current law: the sum of the currents into each node is zero. We'll talk about what the 'modified' bit means in a future article on 'super nodes.' We'll also postpone a discussion about when Kirchhoff's laws break down for a future article (hint: Faraday's law trumps Kirchhoff's law).

The nodes are joined by branches, so the other ingredients are the branch constitutive equations of the components that join them; for example V = IR if it's an ohmic resistor, $V = L \, dI/dt$ for an inductor, etc. In this simple example, we have three simultaneous equations, one each from node 0, 1, and 2:

$$G_{01}(V_1 - V_0) - G_{20}(V_0 - V_2) + I_{20} - I_{01} = 0$$

$$G_{12}(V_2 - V_1) - G_{01}(V_1 - V_0) + I_{01} - I_{12} = 0$$

$$G_{20}(V_0 - V_2) - G_{12}(V_2 - V_1) + I_{12} - I_{20} = 0$$

... with three unknowns, V_0 , V_1 , and V_2 .

The same equations can be rearranged into matrix form; in this case the augmented (or indefinite) node conductance matrix relates the voltage and current vectors:

$$\begin{bmatrix} (G_{01} + G_{20}) & -G_{01} & -G_{20} \\ -G_{01} & (G_{12} + G_{01}) & -G_{12} \\ -G_{20} & -G_{12} & (G_{20} + G_{12}) \end{bmatrix} \begin{bmatrix} V_0 \\ V_1 \\ V_2 \end{bmatrix}$$

$$= \begin{bmatrix} I_{20} - I_{01} \\ I_{01} - I_{12} \\ I_{12} - I_{20} \end{bmatrix}$$

Note the 'pattern of four' that each conductance (e.g. G_{01} highlighted below) impresses into the conductance matrix (Table 1).

In SPICE parlance, making this 'pattern of four' impression is called 'stamping the matrix.' Conveniently, this 'stamping' generaliz-

es for any number of nodes and two termi-

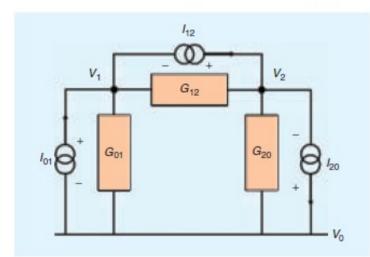


Fig. 1.

Mini-Projects

- Some ideas for mini-projects are on Blackboard now. Please use them for inspiration.
- Please choose one (or make up your own) and fill out a project proposal form by next week's class.
 - Proposal forms are mandatory, and they are also helpful to you.
- After I approve your mini-project (perhaps after modifications), then do your project and create your presentation.
- Presentations are on Feb 27 and March 2.
 - We'll use Matlab to determine order of presentations.

Mini-project Proposal Form

Your name:

Project description.

Please describe your project. Include a description of the problem to be solved, and describe any realworld applications involving this problem.

Inputs and outputs

Please describe any input data your program will use, i.e. is it text, numeric, or what? If the data comes from a third-party source (e.g. off the internet), please tell where it comes from.

Please describe what outputs your program will produce, for example, graphs, text output, etc.

Solution algorithm.

Please describe the algorithm you will use to solve the problem. In best case, this should be a numbered list of steps, similar to what is presented in class lectures.

Corner cases

Please describe any "corner cases" or particular inputs which require special treatment by your program.

Testing

Please describe how you will test your program to verify that it products correct results.

Available on Blackboard

A word about your mini-project presentations

- Give presentation from your laptop using PowerPoint or equivalent.
- Rule of thumb: 1 slide every 2 minutes.
- Count on 12 14 minutes.
 - 6 or 7 slides max.
 - Allow time for a question.
- Maximum 5 main points (main bullets) per slide.
- Pictures are better than words. Draw a picture to convey your idea if you can.

Mini-projects....

- Please rehearse your presentation before you give it in class.
- Make sure you know how to project using your laptop.
 - Suggestion: try to project using the math department's projector prior to class.
- Put a copy of your presentation (.pdf format) onto a flash drive in case of emergency.

Ideal presentation

- 3 slides introducing your general topic area, and talking about the specific problem you want to solve.
- 2 slides reviewing your computation and algorithm
- 1 slide presenting your results.
- Maybe one more slide to expand on one of the above. Maybe.....

Numerical Analysis 1 – Class 5

Solvers for dense systems of linear equations

Ax = b

Gauss elimination

- Algorithm to solve linear system A x = b
- Named after Gauss (early 19th c. German mathematician)
- Known to Chinese mathematicians as early as 179 AD.
- Example on white board



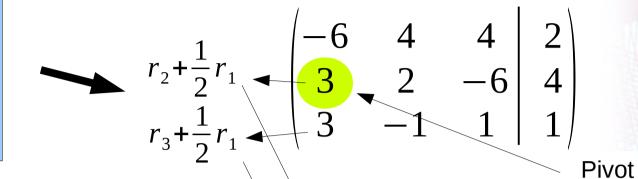
Johann Carl Friedrich Gauss, 1777-1855

Example

Start with

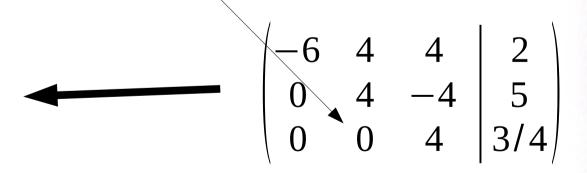
$$-6x+4y+4z=2$$

 $3x+2y-6z=4$
 $3x-y+z=1$



After elimination

$$-6x+4y+4z=2$$
 $4y-4z=5$
 $4z=3/4$

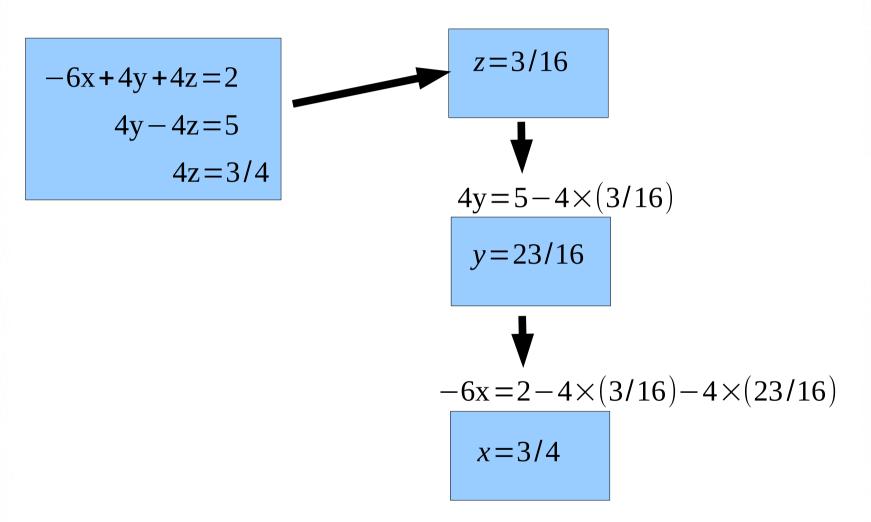


elements

in green

Use forward elimination to get triangular system

Backsubstitution



 Use backsubstitution on triangular system to get x, y, z values.

Remarks about Gauss elimination

- Use pivots to zero out elements underneath them
- Useful for dense matrices.
 - Use other solvers for sparse.
- Two stages to Gauss elimination:
 - Forward elimination
 - Back substitution
- Important: Gauss elimination is O(N³).
 - Demonstration on whiteboard.

Gaussian elimination code

```
function x = naive gauss(A,b);
  % Get length of input vector b.
                                                    Code is on Blackboard in
 n = length(b);
                                                    Class 5 folder.
  % preallocate output x vector.
 x = zeros(n,1);
  % Perform forward elimination to create triangular matrix.
 fprintf('----- Forward elimination -----\n')
 for k=1:n-1 % Iterate over pivots
   for i=k+1:n % Iterate over rows
     xmult = A(i,k)/A(k,k); % Divide by pivot
     for j=1:n % Third inner loop over cols -- algorithm is O(N^3)
       A(i,j) = A(i,j)-xmult*A(k,j);
     end
     b(i) = b(i) - xmult * b(k);
                                                     Elimination step: Three
   end
                                                     nested loops
  end
  % Now do back substitution to get x.
 fprintf('----- Backward substitution -----\n')
 x(n) = b(n)/A(n,n);
 for i=n-1:-1:1
   sum = b(i);
   for j=i+1:n
     sum = sum-A(i,j)*x(j);
                                           Back substitution step:
   end
                                           Two nested loops.
   x(i) = sum/A(i,i);
 end
```

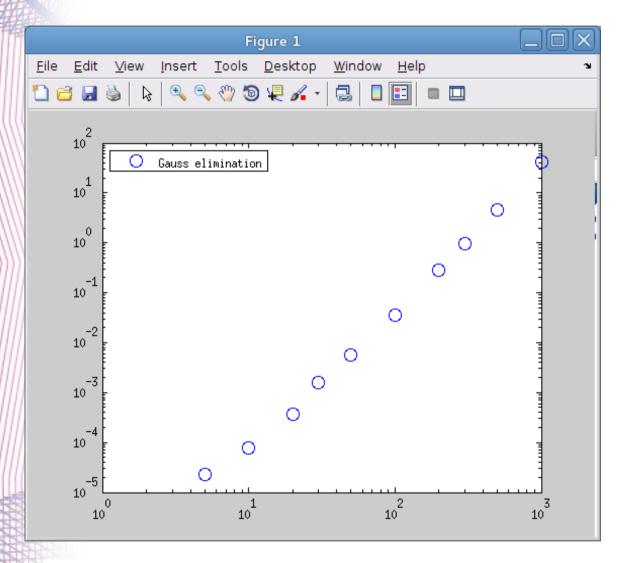
end

```
>> naive gauss(A, b)
----- Forward elimination
After iteration 1,
A =
            -0.6336
    0.1952
                       0.0297
                                -1.9038
           -0.1459
                      -1.0608
                               -3.4057
           -0.0281
                    0.0671
                               -0.3557
           6.2837
                    0.0578
                               17.7575
b =
    0.6970
   2.0223
    0.0976
   -6.5570
After iteration 2,
A =
            -0.6336
                    0.0297
                               -1.9038
    0.1952
            -0.1459
                      -1.0608
                              -3.4057
                    0.2717
                              0.3013
                  0
                    -45.6376 -128.9532
b =
    0.6970
   2.0223
   -0.2925
  80.5599
After iteration 3,
A =
    0.1952
           -0.6336
                      0.0297
                               -1.9038
            -0.1459
                      -1.0608
                              -3.4057
        0
                       0.2717 0.3013
                       0.0000 -78.3423
        0
                  0
b =
    0.6970
    2.0223
   -0.2925
   31.4191
```

Demo

~/Class4/NaiveGauss

Timing of naive_gauss



Gauss elimination: O(N³)

```
>> time_naive_gauss
Testing [3, 3] matrix
Testing [5, 5] matrix
Testing [10, 10] matrix
Testing [20, 20] matrix
Testing [30, 30] matrix
Testing [50, 50] matrix
Testing [100, 100] matrix
Testing [200, 200] matrix
Testing [300, 300] matrix
Testing [500, 500] matrix
Testing [500, 500] matrix
Testing [1000, 1000] matrix
Gauss elimination is O(2.990)
```

Pivoting

 Consider trying to solve this system using Gaussian elimination:

$$\begin{pmatrix} 0 & 3 & 4 \\ 2 & -6 & 1 \\ -1 & 7 & -3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -1 \\ 3 \\ 2 \end{pmatrix}$$

- System is non-singular, well conditioned.
- But we can't do Gaussian elimination due to 0 in upper left position.
- However, we can solve this system:

$$\begin{pmatrix} 2 & -6 & 1 \\ 0 & 3 & 4 \\ -1 & 7 & -3 \end{pmatrix} \begin{pmatrix} x_2 \\ x_1 \\ x_3 \end{pmatrix} = \begin{pmatrix} 3 \\ -1 \\ 2 \end{pmatrix}$$

Same as above, except with rows 1, 2 swapped.

Pivoting and stability

Consider simple system:

$$\begin{pmatrix} 1e-20 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

By observation, solution is close to

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

 But what is the result of Gaussian elimination on this system (on white board)?

Pivoting

- Pivoting: Means putting largest possible value into "pivot element" before doing elimination.
 - Partial pivoting: Row permutations only
 - Complete pivoting: Both row and column permutations
- Pivoting is important for two reasons:
 - Removing zeros
 - Numerical stability
- If you can't eliminate zeros using pivoting, your matrix is probably singular.

Next: LU decomposition

- Derived from Gauss elimination
- L = non-zeros on lower triangle

U = non-zeros on

upper triangle

Recall LU decomposition

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix}.$$

 Why is it good? Decompose matrix once, use for many different b vectors

$$Ax = b$$
 Hard to solve

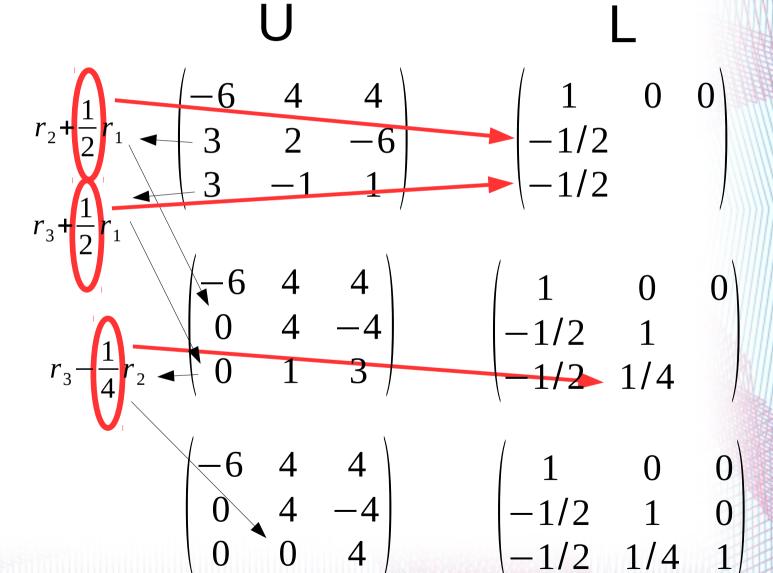
 $LUx = b$ \Rightarrow $Ly = b$ Easy to solve

 $Ux = y$

LU decomposition walk-through

 Put 1 on diagonal of L.

 Put neg of Gauss elimin coeffs into off-diags of L.



Check LU decomposition

$$LU = \begin{pmatrix} 1 & 0 & 0 \\ -1/2 & 1 & 0 \\ -1/2 & 1/4 & 1 \end{pmatrix} \begin{pmatrix} -6 & 4 & 4 \\ 0 & 4 & -4 \\ 0 & 0 & 4 \end{pmatrix}$$

$$= \begin{pmatrix} -6 & 4 & 4 \\ 3 & 2 & -6 \\ 3 & -1 & 1 \end{pmatrix}$$
We get original matrix back.

matrix back.

LU – the main points

- We want to solve Ax = b
- Forward iteration to get A = LU. $O(N^3)$
- Now we have LUx = b. Use back substitution to get x = U(Lb). $O(N^2)$
- This is useful for problems where A remains the same, but b changes.
- Generally used for dense matrices.
- Matlab command: [L, U] = lu(A)
- Default Matlab solver for Ax=b when A is square.

LU demo

-6

$$A = \begin{pmatrix} -6 & 4 & 4 \\ 3 & 2 & -6 \\ 3 & -1 & 1 \end{pmatrix}$$

$$L = \begin{pmatrix} 1 & 0 & 0 \\ -1/2 & 1 & 0 \\ -1/2 & 1/4 & 1 \end{pmatrix}$$

$$U = \begin{pmatrix} -6 & 4 & 4 \\ 0 & 4 & -4 \\ 0 & 0 & 4 \end{pmatrix}$$

Another demo

 Note: to get LU form in general, you need to pay attention to the permutation matrix returned along with decomposition.

What happened here?!?

```
>> A = randn(4,4)
A =
   -0.1953
             -1.2125
                         0.5279
                                    1.0228
    0.8734
              0.4541
                         0.6959
                                    0.5313
   -0.4659
              1.0691
                        -1.1500
                                   -1.5655
   -0.8709
              0.2928
                        -0.6465
                                   -0.9258
>> [L,U] = lu(A)
   -0.2236
             -0.8472
                         0.0483
                                    1.0000
    1.0000
               1.0000
   -0.5335
   -0.9971
              0.5686
                         1.0000
[]
    0.8734
              0.4541
                       0.6959
                                    0.5313
               1.3114
                        -0.7788
                                   -1.2821
                                    0.3329
                         0.4902
                                    0.0393
```

How to pivot? Permutation matrices

 Upon multiplication a permutation matrix will swap rows/cols. Examples:

$$P = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

$$PA = \begin{vmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{vmatrix} \begin{vmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \\ 9 & 10 & 11 & 12 \\ 13 & 14 & 15 & 16 \end{vmatrix} = \begin{vmatrix} 1 & 2 & 3 & 4 \\ 9 & 10 & 11 & 12 \\ 5 & 6 & 7 & 8 \\ 13 & 14 & 15 & 16 \end{vmatrix}$$

- PA swap rows
- AP swap columns.

Make a permutation matrix in Matlab

```
>> P = eye(5)
>> P = P(:, [1 4 3 2 5])
```

Row swap

>> A				
A =				
1 1 1 1 1	1 2 3 4 5	1 3 6 10 15	1 4 10 20 35	1 5 15 35 70
>> P*A				
ans =				
1 1 1 1 1	1 4 3 2 5	1 10 6 3 15	1 20 10 4 35	1 35 15 5 70

Column swap

>> A				
A =				
1 1 1 1 1	1 2 3 4 5	1 3 6 10 15	1 4 10 20 35	1 5 15 35 70
>> A*P				
ans =				
1 1 1 1 1	1 4 10 20 35	1 3 6 10 15	1 2 3 4 5	1 5 15 35 70

Permutation matrices

- When we do
 Gaussian
 elimination, we
 do pivoting by
 hand.
- When Matlab does pivoting, it can return a permutation matrix along with the matrix decomposition.

```
>> A = randn(4,4)
A =
                                  1.0228
   -0.1953
             -1.2125
                        0.5279
    0.8734
            0.4541
                        0.6959
                                  0.5313
   -0.4659 1.0691
                       -1.1500
                                 -1.5655
   -0.8709
            0.2928
                       -0.6465
                                 -0.9258
>> [L,U] = lu(A)
             -0.8472
   -0.2236
                        0.0483
                                  1.0000
    1.0000
              1.0000
   -0.5335
   -0.9971
              0.5686
                        1.0000
    0.8734
              0.4541
                        0.6959
                                  0.5313
              1.3114
                       -0.7788
                                 -1.2821
                        0.4902
                                  0.3329
```

0.0393

```
>> A
A =
   -0.1953
              -1.2125
                         0.5279
                                    1.0228
    0.8734
             0.4541
                       0.6959
                                    0.5313
   -0.4659
           1.0691
                       -1.1500
                                   -1.5655
              0.2928
   -0.8709
                       -0.6465
                                   -0.9258
>> [L,U,P] = lu(A)
                                                    Ask for
                                                    permutation matrix
L =
                                                    as part of return
    1.0000
                    0
                               0
                                         0
   -0.5335
              1.0000
                                         0
   -0.9971
             0.5686
                         1.0000
   -0.2236
              -0.8472
                         0.0483
                                    1.0000
U =
    0.8734
               0.4541
                         0.6959
                                    0.5313
               1.3114
                        -0.7788
                                   -1.2821 ◀
                                    0.3329
         0
                    0
                         0.4902
         0
                    0
                               0
                                    0.0393
P =
     0
                        0
                  0
                                    Permutation matrix
     0
            0
                  1
```

1

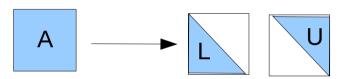
0

0

0

Next: Cholesky decomposition

- The story so far:
 - Gauss elim is O(N³)
 - LU is O(N³) for decomposition, O(N²) to re-use the decomposition for different b vectors.



- For symmetric, positive-definite matrices: Cholesky decomposition
 - Still O(N³), but 1/2 time required by LU

Derivation

Note symmetric

$$A = L L^{T}$$

Cholesky decomposition – SPD matrix only

$$\begin{vmatrix} a_{11} & a_{21} & a_{31} \\ a_{21} & a_{22} & a_{32} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{vmatrix} \begin{vmatrix} l_{11} & l_{21} & l_{31} \\ 0 & l_{22} & l_{32} \\ 0 & 0 & l_{33} \end{vmatrix}$$

$$= \begin{vmatrix} l_{11}^2 & same & same \\ l_{21}l_{11} & l_{21}^2 + l_{22}^2 & same \\ l_{31}l_{11} & l_{32}l_{21} + l_{32}l_{22} & l_{31}^2 + l_{32}^2 + l_{33}^2 \end{vmatrix}$$

Walk through on white board

Working backwards....

$$\begin{vmatrix} a_{11} & a_{21} & a_{31} \\ a_{21} & a_{22} & a_{32} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} l_{11}^2 & same \\ l_{21}l_{11} & l_{21}^2 + l_{22}^2 & same \\ l_{31}l_{11} & l_{32}l_{21} + l_{32}l_{22} & l_{31}^2 + l_{32}^2 + l_{33}^2 \end{vmatrix}$$

$$l_{11} = \sqrt{a_{11}}$$

$$l_{21} = a_{21}/l_{11} \longrightarrow l_{22} = \sqrt{a_{22} - l_{21}^2}$$

$$l_{31} = a_{31}/l_{11} \longrightarrow l_{32} = (a_{32} - l_{31}l_{21})/l_{22} \longrightarrow l_{33} = \sqrt{a_{33}^2 - l_{31}^2 - l_{32}^2}$$

 Each element depends only upon previously computed values.

Cholesky algorithm

- **1.** Initialize by computing --- $l_{11} = \sqrt{a_{11}}$
- 2. Loop on i = 2:N
- 3. Loop on j = 1:i-1 (lower triangle)
- 4. Compute off diagonals --- $l_{ij} = \left(\frac{1}{l_{ii}}\right) \left(a_{ij}^2 \sum_{k=1}^{j-1} l_{ik} l_{jk}\right)$
- 5. End i
- 6. Compute diagonal entry $\longrightarrow l_{ii} = \sqrt{a_{ii}^2 \sum_{i=1}^{i-1} l_{ij}^2}$
- 7. End i
- 7. End I

 8. Return lower triangular matrix $\begin{vmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{vmatrix}$

```
function L = mychol(A)
 % My simple implementation of the basic
 % Cholesly factorization algorithm.
 % Written for Numerical Analysis 1, Spring 2016.
 % Note: This version is not vectorized. Performance
 % may stink.
 N = size(A,1);
 L = zeros(size(A));
 fprintf('Computing [%d, %d]\n', 1, 1)
 L(1,1) = sqrt(A(1,1));
 for i=2:N
   % Do sum giving lower off-diagonals.
   for j=1:(i-1)
     s = 0;
     for k=1:(j-1)
        s = s + L(i,k)*L(i,k);
      end
      fprintf('Computing [%d, %d]\n', i, j)
     L(i,j) = (A(i,j) - s)/L(j,j);
   end
   % Do sum giving diagonal piece.
   s = 0;
   for j=1:(i-1)
     s = s + L(i,j)*L(i,j);
    end
   fprintf('Computing [%d, %d]\n', i, i)
   L(i,i) = sqrt(A(i,i) - s);
 end
```

Simple Matlab implementation

end

```
>> B = 100*randn(4,4)
                                              >> L = mychol(A)
                                              Computing [1, 1]
B =
                                              Computing [2, 1]
                                              Computing [2, 2]
 -135.5593
            -71.7969
                       -4.3032
                                 52.2341
                                              Computing [3, 1]
   45.9567
           85.5801
                      219.6252
                                 -55.9906
                                              Computing [3, 2]
   44.2279
           155.6688
                      -75.9326
                                 -63.3317
                                              Computing [3, 3]
  -25.6306 -38.9228
                      134.5036
                                -46.2980
                                              Computing [4, 1]
                                              Computing [4, 2]
>> A = B'*B
                                              Computing [4, 3]
                        This creates SPD
                                              Computing [4, 4]
A =
                        matrix
                                              T. =
   1.0e+04 *
                                                151.9914
                                                                            0
                                                                                      0
    2.3101
              2.1548
                        0.3871
                                  -1.1268
                                                141.7728 134.6364
                                                                                      0
              3.8226
                       0.2049
    2.1548
                                  -1.6599
                                                 25.4675
                                                          -11.5992
                                                                     267.0723
             0.2049
                       7.2111
    0.3871
                                  -1.3940
                                                -74.1381
                                                          -45.2172
                                                                     -47.0899
                                                                                47.5316
   -1.1268
             -1.6599
                       -1.3940
                                   1.2018
                                              >> L*L' - A
                                              ans =
                                                 1.0e-10 *
                                                       0
                                                                 0
                                                                            0
                                                                                      0
                                                       0
                                                                 0
                                                                                      0
                                                                 0
                                                                      -0.1455
                                                       0
```

```
function test mychol()
 basetol = 2e-11;
                                                                         Test
 pass = 0;
  fail = 0;
  % Generate representative sample of different sized matrices for test.
  for n = 3:13:200
   tol = n*basetol;
    % Generate random SPD matrix of size nxn.
   B = 100*randn(n,n);
   A = B'*B;
    % Now send to mychol
   L = mychol(A);
    % Now check decomposition
    diff = norm(L*L' - A);
    fprintf('Testing [%d, %d] matrix... diff = %e, tol = %e ', n, n, diff, tol)
    if (diff < tol)</pre>
      fprintf('Test passed!\n')
     pass = pass+1;
    else
      fprintf('Test failed!\n')
     fail = fail+1;
    end
  end
  if (fail > 0)
    fprintf('At end, at least one test failed -- general failure\n')
  else
    fprintf('All tests pass!\n')
  end
```

end

Testing different matrix sizes

```
>> test mychol
Testing [3, 3] matrix... diff = 2.943187e-12, tol = 6.000000e-11 Test passed!
Testing [16, 16] matrix...
                            diff = 3.382943e-11, tol = 3.200000e-10 Test passed!
Testing [29, 29] matrix...
                            diff = 8.032349e-11, tol = 5.800000e-10 Test passed!
Testing [42, 42] matrix...
                            diff = 1.963549e-10, tol = 8.400000e-10 Test passed!
Testing [55, 55] matrix...
                            diff = 2.601481e-10, tol = 1.100000e-09 Test passed!
                            diff = 3.717185e-10, tol = 1.360000e-09 Test passed!
Testing [68, 68] matrix...
                            diff = 3.559407e-10, tol = 1.620000e-09 Test passed!
Testing [81, 81] matrix...
                            diff = 5.008336e-10, tol = 1.880000e-09 Test passed!
Testing [94, 94] matrix...
Testing [107, 107] matrix...
                              diff = 5.222256e-10, tol = 2.140000e-09 Test passed!
                              diff = 7.884064e-10, tol = 2.400000e-09 Test passed!
Testing [120, 120] matrix...
Testing [133, 133] matrix...
                              diff = 9.766660e-10, tol = 2.660000e-09 Test passed!
                              diff = 1.187644e-09, tol = 2.920000e-09 Test passed!
Testing [146, 146] matrix...
Testing [159, 159] matrix...
                              diff = 1.042439e-09, tol = 3.180000e-09 Test passed!
Testing [172, 172] matrix...
                              diff = 1.229054e-09, tol = 3.440000e-09 Test passed!
Testing [185, 185] matrix...
                              diff = 1.499971e-09, tol = 3.700000e-09 Test passed!
Testing [198, 198] matrix...
                              diff = 2.454740e-09, tol = 3.960000e-09 Test passed!
All tests pass!
```

```
>> B = 100*randn(4,4)
B =
  -43.7288 -177.1440 -15.6100 -84.0567
  -58.3422 -68.3055 -131.0864 -140.8051
    5.8492 -85.7203
                     44.9550 155.4949
 -121.6094 59.1517 -179.1292 -162.2765
>> L = mychol(B)
                                            What if input
Computing [1, 1]
Computing [2, 1]
Computing [2, 2]
                                     matrix is not SPD?
Computing [3, 1]
Computing [3, 2]
Computing [3, 3]
Computing [4, 1]
Computing [4, 2]
Computing [4, 3]
Computing [4, 4]
L =
   0.0000 + 6.6128i 0.0000 + 0.0000i
                                      0.0000 + 0.0000i
                                                        0.0000 + 0.0000i
  0.0000 + 8.8227i 3.0877 + 0.0000i
                                      0.0000 + 0.0000i
                                                        0.0000 + 0.0000i
   0.0000 - 0.8845i - 30.2894 + 0.0000i
                                      0.0000 + 29.5247i
                                                        0.0000 + 0.0000i
   0.0000 + 18.3901i 71.7046 + 0.0000i
                                      0.0000 - 66.9436i
                                                        0.0000 + 22.0041i
>> T.*T.' - B
                                                Returned L does not
ans =
                                                satisfy B = L L^{T}
   1.0e+04 *
   0.0087
             0.0235
                      0.0010
                               0.0206
   0.0117
             0.0156
                      0.0030
                               0.0524
   -0.0012
            -0.0016
                      0.1745
                               -0.4320
    0.0243
             0.0324
                     -0.3986
                               1.0608
```

Another decomposition: QR

 Decompose matrix into product of orthogonal and upper triangular matrices.

$$A = QR$$
• Upper triangular R:
$$R = \begin{bmatrix} r_{11} & r_{12} & r_{13} & \cdots & r_{1N} \\ 0 & r_{22} & r_{23} & \cdots & r_{2N} \\ 0 & 0 & r_{33} & \cdots & r_{3N} \\ \vdots & \vdots & & & \\ 0 & 0 & 0 & \cdots & r_{NN} \end{bmatrix}$$

Orthogonal matrix Q

We will use the QR decomposition later....

Orthogonal matrix

Orthogonal matrix: columns are vectors orthogonal to each other.

$$Q = \begin{pmatrix} \vdots & \vdots & \vdots & \vdots \\ e_1 & e_2 & e_3 & \cdots & e_N \\ \vdots & \vdots & \vdots & \vdots \end{pmatrix}$$

Nice property of orthogonal matrices:

$$Q^T Q = I \Leftrightarrow Q^{-1} = Q^T$$

QR algorithm – Gram-Schmidt

$$A = (a_1 | a_2 | a_3 | \cdots | a_N)$$

Start by considering A as collection of column vectors.

$$=(e_1|e_2|e_3|\cdots|e_N)$$

Use Gram-Schmidt to compute orthonormal basis set e_i from a_i vectors. This forms orthogonal matrix Q.

$$=(e_1|e_2|e_3|\cdots|e_N)\begin{vmatrix}a_1\cdot e_1&a_2\cdot e_1&a_2\cdot e_1&\cdots&a_N\cdot e_1\\0&a_2\cdot e_2&a_2\cdot e_2&\cdots&a_N\cdot e_2\\0&0&a_3\cdot e_3&\cdots&a_N\cdot e_3\\\vdots&&&&\vdots\\0&0&0&\cdots&a_N\cdot e_N\end{vmatrix}$$
 se Gram-Schmidt to

Use e_i and a_i to compute elements of R matrix.

Gram-Schmidt procedure

- Details in handout on blackboard
- Goal: Given matrix A, create set of orthonormal basis vectors e, spanning space of A.
- The idea is to take each column of input matrix A, and shave off non-orthogonal components of each column.
- Picture on white board.

Gram-Schmidt procedure

Start with:
$$A = (a_1 | a_2 | a_3 | \cdots | a_N)$$

First vector:
$$u_1 = a_1$$

$$e_1 = \frac{u_1}{\|u_1\|}$$

$$u_2 = a_2 - (a_2 \cdot e_1) e_1$$

$$e_2 = \frac{u_2}{\|u_2\|}$$

$$u_3 = a_3 - (a_3 \cdot e_2) e_2 - (a_3 \cdot e_1) e_1$$

$$e_3 = \frac{u_3}{\|u_3\|}$$

Gram-Schmidt code

```
function e = gram schmidt(A)
  % This fcn returns the orthogonalization of the
  % matrix A computed using Gram-Schmidt
  N = size(A, 2);
  e = zeros(size(A));
  % Initialize computation by computing u and e for first col.
  un = A(:, 1);
  e(:, 1) = un/norm(un);
  % Iterate over remaining columns of A
  for k = 2:N
    an = A(:, k);
    un = an;
    % Iterate over previous e values and subtract off
    % component parallel to each ei
    for i = k-1:-1:1
      un = un - dot(an,e(:, i))*e(:, i);
    end
    % Compute next col of e
    e(:, k) = un/norm(un);
  end
end
```

Code implementing QR decomposition

```
function [Q, R] = my qr(A)
  % This fcn implements the QR algorithm using Gram-Schmidt
 e = gram schmidt(A);
 N = size(A, 1);
 0 = e;
 R = zeros(size(A));
  for r = 1:N
    for c = r:N
      R(r, c) = dot(A(:, c), e(:, r));
    end
  end
end
```

Remarks

- You can do a QR decomposition on a nonsquare matrix.
 - Important in linear regression.
- Gram-Schmidt is numerically unstable. Better methods use:
 - Householder transformation
 - Givens Rotations

Topics covered in this session

- Gauss elimination
 - Works for every type of matrix
 - $O(N^3)$
- LU decomposition
 - O(N³) decomposition, O(N²) re-use.
 - Pivoting and permutation matrices
- Cholesky
 - O(N³) but faster... for SPD matrices only.
- QR decomposition
 - Can do for non-square

Ax = b