



# Numerical Linear Algebra

## SEAS Matlab Tutorial 2

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## Linear System of Equations

Linear system of equations.

- Given  $n$  linear equations in  $n$  unknowns.
- Matrix notation: find  $x$  such that  $Ax = b$ .

$$\begin{aligned} 0x_1 + 1x_2 + 1x_3 &= 4 \\ 2x_1 + 4x_2 - 2x_3 &= 2 \\ 0x_1 + 3x_2 + 15x_3 &= 36 \end{aligned}$$

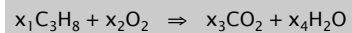
$$A = \begin{bmatrix} 0 & 1 & 1 \\ 2 & 4 & -2 \\ 0 & 3 & 15 \end{bmatrix}, b = \begin{bmatrix} 4 \\ 2 \\ 36 \end{bmatrix}$$

Among most fundamental problems in science and engineering.

- Chemical equilibrium. ↖ see Lab 2
- Google's PageRank algorithm.
- Linear and nonlinear optimization.
- Kirchoff's current and voltage laws.
- Hooke's law for finite element methods.
- Numerical solutions to differential equations.

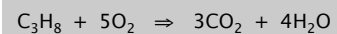
## Chemical Equilibrium

Ex: combustion of propane.



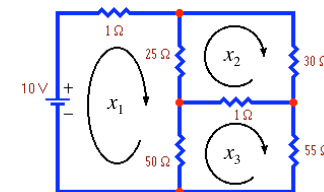
Stoichiometric constraints.

- Carbon:  $3x_1 = x_3$
  - Hydrogen:  $8x_1 = 2x_4$
  - Oxygen:  $2x_2 = 2x_3 + x_4$
  - Normalize:  $x_1 = 1$
- } conservation of mass



## Circuit Analysis

Ex: find current flowing in each branch of a circuit.



Kirchoff's current law.

- $10 = 1x_1 + 25(x_1 - x_2) + 50(x_1 - x_3)$
  - $0 = 25(x_2 - x_1) + 30x_2 + 1(x_2 - x_3)$
  - $0 = 50(x_3 - x_1) + 1(x_3 - x_2) + 55x_3$
- } conservation of electrical charge

Solution:  $x_1 = 0.2449, x_2 = 0.1114, x_3 = 0.1166$ .

## Gaussian Elimination

### Gaussian elimination.

- Among oldest and most widely used solutions.
- Repeatedly apply **row operations** until system is **upper triangular**.
- Solve "trivial" upper triangular system via **back substitution**.

$$\begin{bmatrix} 0 & 1 & 1 \\ 2 & 4 & -2 \\ 0 & 3 & 15 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \\ 36 \end{bmatrix}$$

```
>> A = [0 1 1; 2 4 -2; 0 3 15];
>> b = [4; 2; 36];
>> x = solve(A, b)
x =
-1
 2
 2
```

we are going to implement this

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## Gaussian Elimination

```
>> A = [0 1 1; 2 4 -2; 0 3 15]
A =
 0  1  1
 2  4 -2
 0  3 15
declare a matrix

>> A([1 2], :) = A([2 1], :)
A =
 2  4 -2
 0  1  1
 0  3 15
swap rows 1 and 2

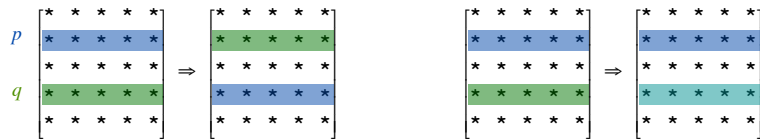
>> A(3, :) = A(3, :) - 3 * A(2, :)
A =
 2  4 -2
 0  1  1
 0  0 12
subtract 3 times row 2
from row 3
```

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## Elementary Row Operations

### Elementary row operations.

- Exchange row  $p$  and row  $q$ .
- Add a multiple  $\alpha$  of row  $p$  to row  $q$ .



```
A([p q], :) = A([q p], :);
b([p q], :) = b([q p], :);
```

```
A(q, :) = A(q, :) - alpha * A(p, :);
b(q, :) = b(q, :) - alpha * b(p, :);
```

**Key invariant.** Row operations preserve solutions.

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## Gaussian Elimination: Row Operations

$$\begin{bmatrix} 0 & 1 & 1 \\ 2 & 4 & -2 \\ 0 & 3 & 15 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \\ 36 \end{bmatrix}$$

↓ (interchange row 1 and 2)

$$\begin{bmatrix} 2 & 4 & -2 \\ 0 & 1 & 1 \\ 0 & 3 & 15 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \\ 36 \end{bmatrix}$$

↓ (subtract 3x row 2 from row 3)

$$\begin{bmatrix} 2 & 4 & -2 \\ 0 & 1 & 1 \\ 0 & 0 & 12 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \\ 24 \end{bmatrix}$$

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## Gaussian Elimination: Back Substitution

**Back substitution.** Upper triangular systems are easy to solve by examining equations in reverse order.

$$\begin{bmatrix} 2 & 4 & -2 \\ 0 & 1 & 1 \\ 0 & 0 & 12 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \\ 24 \end{bmatrix}$$

Eq 3.  $x_3 = 24/12 = 2$

Eq 2.  $x_2 = 4 - x_3 = 2$

Eq 1.  $x_1 = (2 - 4x_2 + 2x_3) / 2 = -1$

$$x_i = \frac{1}{a_{ii}} \left[ b_i - \sum_{j=i+1}^n a_{ij} x_j \right]$$

```
[m n] = size(A);
x = zeros(n, 1);
for i = n : -1 : 1
    total = 0.0;
    for j = i+1 : n
        total = total + A(i, j) * x(j);
    end
    x(i) = (b(i) - total) / A(i, i);
end
```

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## Gaussian Elimination: Back Substitution

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$$\begin{bmatrix} 2 & 4 & -2 \\ 0 & 1 & 1 \\ 0 & 0 & 12 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \\ 24 \end{bmatrix}$$

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$$x_i = \frac{1}{a_{ii}} \left[ b_i - \sum_{j=i+1}^n a_{ij} x_j \right]$$

vectorized version

```
[m n] = size(A);
x = zeros(size(b));
for i = n : -1 : 1
    j = i+1 : n
    x(i, :) = ((b(i, :) - A(i, j) * x(j, :)) / A(i, i);
end
```

vector inner product

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## Gaussian Elimination: Forward Elimination

**Forward elimination.** Apply row operations to make upper triangular.

**Pivot.** Zero out entries below pivot  $a_{pp}$ .

$$\alpha = a_{ip} / a_{pp}$$

$$a_{ij} = a_{ij} - \alpha a_{pj}$$

$$b_i = b_i - \alpha b_p$$

$$\begin{matrix} & & p & & \\ \begin{bmatrix} * & * & * & * & * \\ 0 & * & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & * & * & * \end{bmatrix} & \Rightarrow & \begin{bmatrix} * & * & * & * & * \\ 0 & * & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & 0 & * & * \\ 0 & 0 & 0 & * & * \\ 0 & 0 & 0 & * & * \end{bmatrix} \end{matrix}$$

```
for i = p+1 : n
    alpha = A(i, p) / A(p, p);
    b(i, :) = b(i, :) - alpha * b(p, :);
    A(i, :) = A(i, :) - alpha * A(p, :);
end
```

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## Gaussian Elimination: Forward Elimination

**Forward elimination.** Apply row operations to make upper triangular.

**Pivot.** Zero out entries below pivot  $a_{pp}$ .

$$\begin{matrix} \begin{bmatrix} * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \end{bmatrix} & \Rightarrow & \begin{bmatrix} * & * & * & * & * \\ 0 & * & * & * & * \\ 0 & * & * & * & * \\ 0 & * & * & * & * \\ 0 & * & * & * & * \end{bmatrix} & \Rightarrow & \begin{bmatrix} * & * & * & * & * \\ 0 & * & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & * & * & * \end{bmatrix} & \Rightarrow & \begin{bmatrix} * & * & * & * & * \\ 0 & * & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & 0 & * & * \\ 0 & 0 & 0 & * & * \end{bmatrix} & \Rightarrow & \begin{bmatrix} * & * & * & * & * \\ 0 & * & * & * & * \\ 0 & 0 & * & * & * \\ 0 & 0 & 0 & * & * \\ 0 & 0 & 0 & 0 & * \end{bmatrix} \end{matrix}$$

```
for p = 1 : n
    for i = p+1 : n
        alpha = A(i, p) / A(p, p);
        b(i, :) = b(i, :) - alpha * b(p, :);
        A(i, :) = A(i, :) - alpha * A(p, :);
    end
end
```

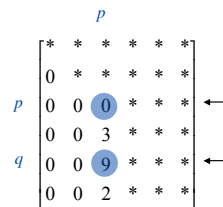
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## Gaussian Elimination: Partial Pivoting

**Remark.** Code on previous slide fails spectacularly if pivot  $a_{pp} = 0$ .

**Partial pivoting.** Swap row  $p$  with the row  $q$  that has **largest** entry in column  $p$  among rows below the diagonal.

```
q = p;
for i = p+1 : n
    if (abs(A(i, p)) > abs(A(q, p)))
        q = i;
    end
end
A([p q], :) = A([q p], :);
b([p q], :) = b([q p], :);
```



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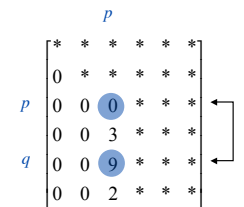
## Gaussian Elimination: Partial Pivoting

**Remark.** Code on previous slide fails spectacularly if pivot  $a_{pp} = 0$ .

**Partial pivoting.** Swap row  $p$  with the row  $q$  that has **largest** entry in column  $p$  among rows below the diagonal.

```
[val q] = max(abs(A(p:n, p)));
q = q + p - 1;
A([p q], :) = A([q p], :);
b([p q], :) = b([q p], :);
```

vectorized version



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## Gaussian Elimination with Partial Pivoting

```
function x = lsolve(A, b)
% LSOLVE Linear system of equation solver, bare bones version
% x = lsolve(A, b) returns the solution to the equation Ax = b,
% where A is an n-by-n nonsingular matrix, and b is a column
% vector of length n (or a matrix with several such columns).

[m n] = size(A);

% Gaussian elimination with partial pivoting
for p = 1 : n

    % find index q of largest element below diagonal in column p
    [val q] = max(abs(A(p:n, p)));
    q = q + p - 1;

    % swap with row p
    A([p q], :) = A([q p], :);
    b([p q], :) = b([q p], :);

    % zero out entries of A and b using pivot A(p, p)
    for i = p+1 : n
        alpha = A(i, p) / A(p, p);
        b(i, :) = b(i, :) - alpha * b(p, :);
        A(i, :) = A(i, :) - alpha * A(p, :);
    end
end

% back substitution
x = zeros(size(b));
for i = n : -1 : 1
    j = i+1 : n;
    x(i, :) = (b(i, :) - A(i, j) * x(j, :)) / A(i, i);
end
```

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$$x = A \setminus b;$$

## Singular Value Decomposition

**Singular value decomposition.** Given a real, square matrix  $A$ , the SVD is  $A = USV^T$ , where  $U$  and  $V$  are orthogonal, and  $S$  is diagonal.

$U U^T = I$       singular values in descending order

- Among most important concepts in matrix computation.
- Applications: statistics, signal processing, acoustics, vibrations, ....

$$\begin{bmatrix} 4 & -1 & 1 \\ -2 & -2 & 2 \\ 0 & 5 & 5 \end{bmatrix} = \begin{bmatrix} 0 & \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \\ 0 & \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} \sqrt{50} & 0 & 0 \\ 0 & \sqrt{20} & 0 \\ 0 & 0 & \sqrt{10} \end{bmatrix} \begin{bmatrix} 0 & -1 & 0 \\ \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \end{bmatrix}^T$$

$A$                        $U$                        $S$                        $V$

## Principal Component Analysis

**Principal component analysis (PCA).** Truncated SVD is  $A_r = U_r S_r V_r^T$ , where  $U_r$  and  $V_r$  are the first  $r$  columns of  $U$  and  $V$ , and  $S_r$  is the first  $r$  rows and columns of  $S$ .

**Fact.**  $A_r$  is the "best" rank  $r$  approximation to  $A$ .

$$\begin{bmatrix} 4 & -1 & 1 \\ -2 & -2 & 2 \\ 0 & 5 & 5 \end{bmatrix} = \begin{bmatrix} 0 & \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \\ 0 & \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} \sqrt{50} & 0 & 0 \\ 0 & \sqrt{20} & 0 \\ 0 & 0 & \sqrt{10} \end{bmatrix} \begin{bmatrix} 0 & -1 & 0 \\ \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \end{bmatrix}^T$$

$A$                        $U$                        $S$                        $V$

$$r=2 \quad \begin{bmatrix} 4 & 0 & 0 \\ -2 & 0 & 0 \\ 0 & 5 & 5 \end{bmatrix} = \begin{bmatrix} 0 & \frac{2}{\sqrt{5}} \\ 0 & \frac{1}{\sqrt{5}} \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \sqrt{50} & 0 \\ 0 & \sqrt{20} \end{bmatrix} \begin{bmatrix} 0 & -1 \\ \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & 0 \end{bmatrix}^T$$

$A_r$                        $U_r$                        $S_r$                        $V_r$                        $\|A_r - A\|_2 = \sqrt{10}$

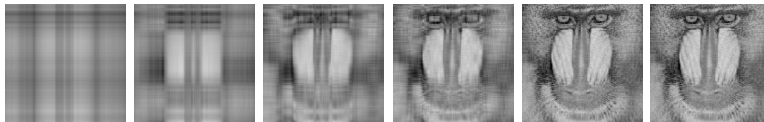
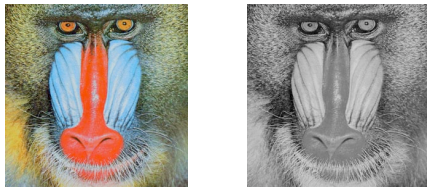
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## Image Processing: PCA

### Image processing.

- Read in color image.
- Convert to grayscale.
- Create  $n$ -by- $n$  matrix of grayscale values.
- Compute best rank  $\{ 1, 2, 5, 10, 25, 50 \}$  approximation.



## baboon.m

```
% MATLAB script that reads in the image baboon.jpg,
% converts it to grayscale, and forms a matrix of its
% grayscale values.
%
% Then it computes and plots the best rank r approximate
% to the matrix using the SVD. It saves each approximation
% as a JPEG image.

A = imread('baboon.jpg'); % read image from a file
A = rgb2gray(A);         % convert from color to grayscale
A = im2double(A);        % convert to double precision matrix
imshow(A);               % display the image in a window

[U S V] = svd(A);
for r = [1 2 5 10 25 50 100 298]
    Ar = U(:, 1:r) * S(1:r, 1:r) * V(:, 1:r)';
    imshow(Ar);
    pause;
    imwrite(Ar, sprintf('baboon-%d.jpg', r));
end
```

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## Faces



average

## 7 Principal Faces



Reference: Diego Nehab, COS 496