

General Techniques

1. Matrix Factorization and/or algorithm
2. Perturbation theory and condition numbers.
3. Effect of round off error (ie. flops).
4. Analysis of speed of algorithm
5. Utilize software.

Matrix Multiplication

By convention, a matrix will be indicated by a capital letter (A, B, C, \dots)
a vector will be indicated using lower case letters.

Matrix $A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{bmatrix}$

this matrix has m columns and n rows

The entries of the matrix may be real or complex

$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix}$ is an m -tuple of real numbers

we can multiply A by x . Let's say $b = Ax$
 b is an n -tuple $b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$

$$b_i = a_{i1}x_1 + a_{i2}x_2 + \dots + a_{im}x_m$$

$$= \sum_{j=1}^m a_{ij}x_j$$

$$b_i = a_{i1}x_1 + a_{i2}x_2 + \dots + a_{im}x_m$$

$$= \sum_{j=1}^n a_{ij}x_j$$

In other words, the i th component of b is the inner product (dot product) of the i th row of A with x .

Example is

Here $n=2$ and $m=3$

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix} = \begin{bmatrix} 50 \\ 122 \end{bmatrix}$$

Pseudo-code for this type of multiplication

row-oriented multiplication \rightarrow

```

b ← 0
for i = 1, 2, ..., n
  for j = 1, 2, ..., m
    b_i ← b_i + a_ij x_j
  
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Another way to view matrix-vector multiplication

$$A = \begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix}$$

$$b = Ax$$

$$b = \begin{bmatrix} a_{11} \\ \vdots \\ a_{n1} \end{bmatrix} x_1 + \begin{bmatrix} a_{12} \\ \vdots \\ a_{n2} \end{bmatrix} x_2 + \dots + \begin{bmatrix} a_{1m} \\ \vdots \\ a_{nm} \end{bmatrix} x_m$$

That b is a linear combination of the columns of A .

as in the example above

of A .

Back to the example above

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix} = \begin{bmatrix} 50 \\ 122 \end{bmatrix}$$

$$\begin{bmatrix} 50 \\ 122 \end{bmatrix} = \begin{bmatrix} 1 \\ 4 \end{bmatrix} 7 + \begin{bmatrix} 2 \\ 5 \end{bmatrix} 8 + \begin{bmatrix} 3 \\ 6 \end{bmatrix} 9$$

Proposition If $b = Ax$, then b is a linear combination of the columns of A .

If we let A_j describe the j th column of A we have

$$b = \sum_{j=1}^m A_j x_j$$

Pseudo code

$b \leftarrow 0$
for $j = 1, \dots, m$
 $b \leftarrow b + A_j x_j$

column-oriented
multiplication

$b \leftarrow 0$
for $j = 1, \dots, m$
 for $i = 1, \dots, n$
 $b_i \leftarrow b_i + a_{ij} x_j$

Flop Counts

Real numbers are stored in floating point format.

FLOPs \equiv Floating-point arithmetic operators.

$b_i \leftarrow b_i + a_{ij} x_j$ this involves
2 flops because there are 2 operations,

To determine/estimate "how long" a computer
may take to run we may count the number of flops

Let's say A is an n by m matrix

for $b \leftarrow 0$
 $i = 1, \dots, n$

for $j = 1, \dots, m$

$b_i \leftarrow b_i + A_{ij} x_j \leftarrow 2 \text{ flops}$

Since we repeat the line $b_i \leftarrow b_i + A_{ij} x_j$
 nm times there are $2nm$ flops

Let B be a 300 by 400 matrix

C be a 600 by 400 matrix

column vector x is the correct length

Bx $2nm = 2(300)(400)$

Big O

Cx $2(600)(400)$ \leftarrow twice as
much as Bx

A lot of calculations/algorithms involve square
matrices. That is if B is an n by n

matrix

$O(n^2)$

the process is order n^2
The notation emphasizes the dependence on n .

To get set up

1. Install Python and/or Anaconda

2. Install language interpreter

example: VS code

Jupyter

3. Install NumPy

if using VS code

In the command window type:

Python3 -m pip install numpy

Vandermonde Matrix

Let $\{x_1, \dots, x_m\}$ be a sequence of numbers. If p and q are polynomials of degree $< n$ and α is a scalar

then $p+q$ is of degree $< n$

αp is of degree $< n$

The values of the polynomials at x_i satisfy the linearity properties

$$(p+q)(x_i) = p(x_i) + q(x_i)$$

$$(\alpha p)(x_i) = \alpha (p(x_i))$$

Thus the map of vector coefficient p to $(p(x_1), p(x_2), \dots, p(x_m))$ is linear

$$A = \begin{bmatrix} 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & x_2 & x_2^2 & \dots & x_2^{n-1} \\ \vdots & x_m & x_m^2 & \dots & x_m^{n-1} \end{bmatrix}$$