

COMP26120: Algorithms and Imperative Programming

Determinants and Permanents

Ian Pratt-Hartmann

Room KB2.38: email: ipratt@cs.man.ac.uk

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Outline

Introduction

Permutations

Determinants and Permanents

An efficient algorithm

Coda

- In this lecture, we are going to look at methods for computing $(n \times n)$ -**determinants** :

$$\begin{vmatrix} a_{1,1} & \dots & a_{1,n} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \dots & a_{n,n} \end{vmatrix}.$$

- We will begin with a simple method (which you have probably encountered in school).
- We will then describe a more sophisticated method (which you may also have encountered in school).
- We will compare their complexity.



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- A **permutation** is a 1–1 map of a set X onto itself.
- If X is a finite set, e.g. $[0, n - 1] = \{0, \dots, n - 1\}$, this means just jumbling X up.
- For example, if $n = 4$, we have the permutation

$$0 \mapsto 3 \quad 1 \mapsto 0 \quad 2 \mapsto 1 \quad 3 \mapsto 2,$$

which we might more easily depict as

$$[0, 1, 2, 3] \mapsto [3, 0, 1, 2].$$

- If σ and τ are permutations, their **product** , $\sigma \cdot \tau$, is their composition:

$$\sigma \cdot \tau(n) = \tau(\sigma(n)).$$



- The number of permutations on an n -element set is

$$n! = 1 \cdot 2 \cdot \dots \cdot (n - 1) \cdot n.$$

- A simple inductive proof shows that, for all $n \geq 4$:

$$2^n \leq n! \leq 2^{n^2}.$$

- That is: the function $n \mapsto n!$ is $\Omega(2^n)$ and $O(2^{n^2})$.

- A special kind of permutation is a **transposition** . If $X = [0, n - 1]$ the transposition $\sigma = (i, j)$ given by

$$\sigma(k) = \begin{cases} j & \text{if } k = i \\ i & \text{if } k = j \\ k & \text{ow.} \end{cases}$$

- It is easy to show that every permutation can be written as the product of transpositions. E.g.the permutation

$$[0, 1, 2, 3] \mapsto [1, 2, 3, 0].$$

can be written as

$$(0, 1) \cdot (0, 2) \cdot (0, 3) = (0, 3) \cdot (1, 3) \cdot (2, 3).$$

- Note that this is not unique! We define the **parity** of σ , written $\text{sgn}(\sigma)$, to be 1 if σ is the product of an even number of permutations, and -1 otherwise.

- For this to make sense, we must show that, if σ can be written as an even number of permutations, then it cannot be written as an odd number of permutations.
- To see this, first define

$$p(x_0, \dots, x_{n-1}) = \prod_{i < j} (x_j - x_i),$$

and then write

$$s(\sigma) = \frac{p(\sigma(x_0), \dots, \sigma(x_{n-1}))}{p(x_0, \dots, x_{n-1})}$$

The following are easy to check:

$$s(\sigma) = \pm 1$$

$$s(\sigma \cdot \tau) = s(\sigma)s(\tau)$$

$$s(t) = -1 \quad \text{for any transposition } t.$$

Thus, $\text{sgn}(\sigma) = s(\sigma)$, and so is well-defined.

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- Consider a pair of simultaneous linear equations

$$a_{1,1}x_1 + a_{1,2}x_2 = b_1$$

$$a_{2,1}x_1 + a_{2,2}x_2 = b_2,$$

which we can write in matrix form as

$$\begin{pmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}.$$

We can solve these by taking an appropriate $a_{2,1}/a_{1,1}$ time the first equation from the second:

$$(a_{2,2} - (a_{2,1}/a_{1,1})a_{1,2})x_2 = b_2,$$

as long as $(a_{2,2} - (a_{2,1}/a_{1,1})a_{1,2}) \neq 0$, or, equivalently,

$$a_{1,1}a_{2,2} - a_{1,2}a_{2,1} \neq 0.$$

- The quantity

$$a_{1,1}a_{2,2} - a_{1,2}a_{2,1} \neq 0.$$

is known as the **determinant** of the matrix

$$\begin{pmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{pmatrix},$$

and is usually denoted

$$\begin{vmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{vmatrix}.$$

The original system of equations has a unique solution if and only if this determinant is non-zero.

- Similar remarks apply to the trio of simultaneous linear equations

$$a_{1,1}x_1 + a_{1,2}x_2 + a_{1,3}x_3 = b_1$$

$$a_{2,1}x_1 + a_{2,2}x_2 + a_{2,3}x_3 = b_2$$

$$a_{3,1}x_1 + a_{3,2}x_2 + a_{3,3}x_3 = b_3,$$

which we can write in matrix form as

$$\begin{pmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}.$$

These equations have a unique solution if and only if

$$a_{1,1} \cdot \begin{vmatrix} a_{2,2} & a_{2,3} \\ a_{3,2} & a_{3,3} \end{vmatrix} - a_{1,2} \cdot \begin{vmatrix} a_{2,1} & a_{2,3} \\ a_{3,1} & a_{3,3} \end{vmatrix} + a_{1,3} \cdot \begin{vmatrix} a_{2,1} & a_{2,2} \\ a_{3,1} & a_{3,2} \end{vmatrix} \neq 0.$$



- More generally, the n -tuple of simultaneous linear equations

$$\begin{pmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \cdots & a_{n,n} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix}.$$

has a unique solution if and only if the **determinant**,

$$\begin{vmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \cdots & a_{n,n} \end{vmatrix} = a_{1,1} \begin{vmatrix} a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \vdots \\ a_{n,2} & \cdots & a_{n,n} \end{vmatrix} - \cdots \pm a_{1,n} \begin{vmatrix} a_{2,1} & \cdots & a_{2,n-1} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \cdots & a_{n,n-1} \end{vmatrix},$$

is non-zero.

- It's worth writing this out for $n = 3$:

$$\begin{vmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \end{vmatrix} = \\
 a_{1,1} \cdot \begin{vmatrix} a_{2,2} & a_{2,3} \\ a_{3,2} & a_{3,3} \end{vmatrix} - a_{1,2} \cdot \begin{vmatrix} a_{2,1} & a_{2,3} \\ a_{3,1} & a_{3,3} \end{vmatrix} + a_{1,3} \cdot \begin{vmatrix} a_{2,1} & a_{2,2} \\ a_{3,1} & a_{3,2} \end{vmatrix} = \\
 a_{1,1} \cdot (a_{2,2}a_{3,3} - a_{2,3}a_{3,2}) \\
 - a_{1,2} \cdot (a_{2,1}a_{3,3} - a_{2,3}a_{3,1}) \\
 + a_{1,3} \cdot (a_{2,1}a_{3,2} - a_{2,2}a_{3,1}) = \\
 a_{1,1}a_{2,2}a_{3,3} - a_{1,1}a_{2,3}a_{3,2} \\
 - a_{1,2}a_{2,1}a_{3,3} + a_{1,2}a_{2,3}a_{3,1} \\
 + a_{1,3}a_{2,1}a_{3,2} - a_{1,3}a_{2,2}a_{3,1}.
 \end{vmatrix}$$

- Notice how, in each term, the left indices are 1, 2, 3 and the right indices are a permutation of this sequence.

- More generally, we have

$$\begin{vmatrix} a_{1,1} & \dots & a_{1,n} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \dots & a_{n,n} \end{vmatrix} = \sum_{\sigma \in \text{perm}\{1, \dots, n\}} \text{sgn}(\sigma) \prod_{i=1}^n a_{i, \sigma(i)}.$$

- Obvious question: how do we compute this quantity?

- The definition

$$\begin{vmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \cdots & a_{n,n} \end{vmatrix} = a_{1,1} \begin{vmatrix} a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \vdots \\ a_{n,2} & \cdots & a_{n,n} \end{vmatrix} - \cdots \pm \begin{vmatrix} a_{2,1} & \cdots & a_{2,n-1} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \cdots & a_{n,n-1} \end{vmatrix}$$

gives us an easy recursive solution.

- Alternatively, we could step through the permutations of $\{1, \dots, n\}$ and add up the terms of

$$\sum_{\sigma \in \text{perm}\{1, \dots, n\}} \text{sgn}(\sigma) \prod_{i=1}^n a_{i, \sigma(i)}.$$

- But both of these methods take exponential time.

- While we are on the subject of determinants, we contrast the **determinant** of the matrix

$$\begin{pmatrix} a_{1,1} & \dots & a_{1,n} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \dots & a_{n,n} \end{pmatrix}$$

which we defined as

$$\sum_{\sigma \in \text{perm}\{1, \dots, n\}} \text{sgn}(\sigma) \prod_{i=1}^n a_{i, \sigma(i)},$$

with the **permanent** of the same matrix, defined as

$$\sum_{\sigma \in \text{perm}\{1, \dots, n\}} \prod_{i=1}^n a_{i, \sigma(i)},$$

- Obvious question: how do we compute this quantity?
- We will see below that the answers to these questions are very different.

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- An $n \times n$ matrix A is called **upper triangular** if all the entries below the diagonal are zero:

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n-1} & a_{1,n} \\ 0 & a_{2,2} & \cdots & a_{2,n-1} & a_{2,n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & a_{n-1,n-1} & a_{n-1,n} \\ 0 & 0 & \cdots & 0 & a_{n,n} \end{pmatrix}$$

- The determinant of an upper triangular matrix is simply the product of its diagonal elements:

$$\prod_{i=1}^n a_{i,i} = a_{1,1} \cdot a_{2,2} \cdot \cdots \cdot a_{n-1,n-1} \cdot a_{n,n}.$$

- Furthermore, if A is any $n \times n$ matrix, its determinant is unaffected by the following operations on:
 - transposing two rows;
 - transposing two columns;
 - adding a multiple of one row of to another;
 - adding a multiple of one column of to another/
- And if any row (or column) of a matrix is zero, then the determinant is zero.
- This gives us a was of computing determinants mroe efficiently:

- First, we may transpose columns (if necessary) so that $a_{1,1} \neq 0$. (If this is impossible, the determinant is 0.)

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n-1}, & a_{1,n} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n-1} & a_{2,n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n-1,1} & a_{n-1,2} & \cdots & a_{n-1,n-1} & a_{n-1,n} \\ a_{n,1} & a_{n,2} & \cdots & a_{n,n-1} & a_{n,n} \end{pmatrix}$$

- Now take suitable multiples of the first row off the subsequent rows so as to zero the first column:

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n-1}, & a_{1,n} \\ 0 & a_{2,2} & \cdots & a_{2,n-1} & a_{2,n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & a_{n-1,2} & \cdots & a_{n-1,n-1} & a_{n-1,n} \\ 0 & a_{n,2} & \cdots & a_{n,n-1} & a_{n,n} \end{pmatrix}$$

- Next, we may transpose columns (if necessary) so that $a_{2,2} \neq 0$. (If this is impossible, the determinant is 0.)

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \dots & a_{1,n-1} & a_{1,n} \\ 0 & a_{2,2} & \dots & a_{2,n-1} & a_{2,n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & a_{n-1,2} & \dots & a_{n-1,n-1} & a_{n-1,n} \\ 0 & a_{n,2} & \dots & a_{n,n-1} & a_{n,n} \end{pmatrix}$$

- Now take suitable multiples of the second row off the subsequent rows so as to zero the second column:

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \dots & a_{1,n-1} & a_{1,n} \\ 0 & a_{2,2} & \dots & a_{2,n-1} & a_{2,n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & a_{n-1,n-1} & a_{n-1,n} \\ 0 & 0 & \dots & a_{n,n-1} & a_{n,n} \end{pmatrix}$$

- Continue zeroing columns up to the $(n - 1)$ th:

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n-1}, & a_{1,n} \\ 0 & a_{2,2} & \cdots & a_{2,n-1} & a_{2,n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & a_{n-1,n-1} & a_{n-1,n} \\ 0 & 0 & \cdots & 0 & a_{n,n} \end{pmatrix}$$

- Now compute the determinant as

$$\prod_{i=1}^n a_{i,i}.$$

- Warning: the $a_{i,j}$'s above are not the same as they appear in different matrices!

- Let us now calculate the complexity of this procedure:
 - To zero the i column below the diagonal, we need to do a transposition of columns $O(n)$, and, for each of $(n - i + 1) \leq n$ rows below i th row, subtract a multiple of the i th row from that row, which contains $n - i + 1 \leq n$ non-zero elements: so, $O(n)$ operations per row.
 - So cost of zeroing i th column below diagonal is $O(n^2)$.
 - There are $n - 1 \leq n$ columns to zero below the diagonal, so cost of converting to upper-triangular form is $O(n^3)$.
 - Cost of multiplying diagonal elements is $O(n)$.
- So cost of whole procedure is $O(n^3 + n) = O(n^3)$.
- Although this looks complicated, this is much better than the $\Omega(2^n)$ methods outlined above.



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- We have seen how to compute the **determinant** of the matrix

$$\begin{pmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \vdots & \vdots \\ a_{n,1} & \cdots & a_{n,n} \end{pmatrix}.$$

- But what about the **permanent** ? Remember, this is defined as

$$\sum_{\sigma \in \text{perm}\{1, \dots, n\}} \prod_{i=1}^n a_{i, \sigma(i)}.$$

- It seems that there is no efficient method to calculate this quantity: the best-known algorithms run in exponential time.
- This fact (if it is a fact), has important consequences in the theory of combinatorial optimization and computational complexity.