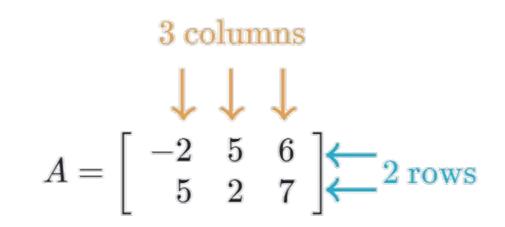
Matrix Multiplication

Randomized Algorithms: Week 3 Summer HSSP 2023 Emily Liu

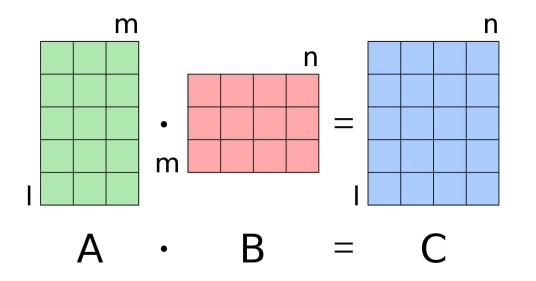
Review: Matrices

Matrix: 2D array of numbers (n rows, m columns \rightarrow **n x m** matrix)



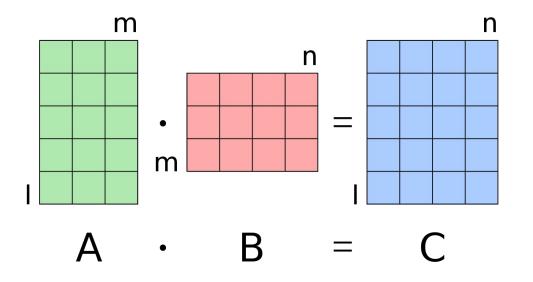
Matrix multiplication

To multiply matrices A and B, the second dimension of A must be the first dimension of B.



Matrix multiplication

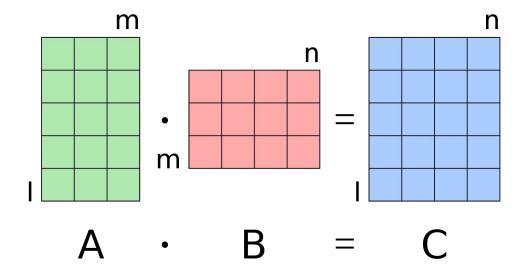
Usages: Many modern data-intensive processes, such as machine learning and dataset management, rely on matrix multiplication.



Matrix multiplication

Let $C = A \times B$. The entry $C[i][j] = sum(A[i][k] * B[k][j] \text{ for } k = 0 \dots m-1)$.

- dot product of ith row of A, jth column of B



Matrix multiplication: Example

$$\left[\begin{array}{rrr}1&8\\3&6\end{array}\right]\times\left[\begin{array}{rrr}4\\3\end{array}\right]$$

Matrix multiplication: Example

$$\begin{bmatrix} 1 & 8 \\ 3 & 6 \end{bmatrix} \times \begin{bmatrix} 4 \\ 3 \end{bmatrix} = \begin{bmatrix} 1*4 + 8*3 \\ 3*4 + 6*3 \end{bmatrix}$$
$$= \begin{bmatrix} 28 \\ 30 \end{bmatrix}$$

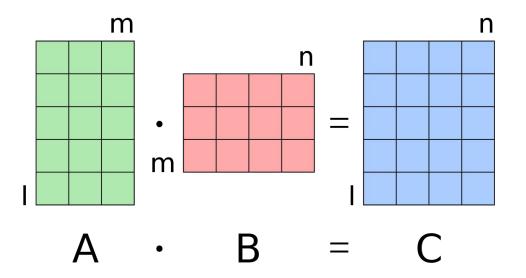
Time complexity of matrix multiplication

For each entry in C $(I \times n)$:

Need to compute the summation of m different product terms.

Therefore, overall complexity is **O(Imn)**.

For square matrices, is $O(n^3)$.



Matrix multiplication is slow...

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Naive method: O(n<sup>3</sup>)
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Strassen's method: $O(n^{\log 7}) \approx O(n^{2.8})$

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Williams et al (2020): O(n<sup>2.3728596</sup>)
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Duan et al (2022): O(n^{2.371866})

Williams et al (2023): O(n^{2.371552})

Changing the problem

Matrix multiplication

(given A and B, find A x B)

 \rightarrow

Matrix multiplication verification

(given A, B, and C, return True if A x B = C and False otherwise.)

Applications: error-checking, ID verification

Matrix multiplication verification

Task: Given A, B, and C, return True if $A \times B = C$ and False otherwise.

Observe:

- Matrix multiplication is a stronger problem than matrix multiplication verification; when we solve matrix multiplication (using any of the previous methods), we can solve the verification task in O(n³) O(n^{log 7}), etc...
- Randomized approach: can get to **O**(**n**²).

Frievalds' Algorithm

Assume A, B, and C are all n x n matrices.

Define a length n vector **r** consisting of 0's and 1's. Each entry in **r** has a $\frac{1}{2}$ probability of being 0 or 1.

Return True if the following expression is true, otherwise False.

A(Br) = Cr.

Time complexity of Frievalds

In general: Matrix-vector multiplication is faster than matrix multiplication.

RHS: Computation of C x r: $O(n * n * 1) = O(n^2)$

LHS: Compute A x B x r

- Compute $s = B \times r$ first in O(n²)
- Compute A x s in $O(n^2)$

Case 1: A x B = C (Correct answer: True)

- Frievalds will always return True, which is correct answer!

Case 2: A x B != C (Correct answer: False)

- Question: What is the probability Frievalds returns False?

Claim: When $AB \neq C$, $P(ABr \neq Cr) \geq \frac{1}{2}$.

What this means:

When AB \neq C, Frievalds has a $\leq \frac{1}{2}$ probability of being wrong.

Proof:

Let D = AB - C.

Now, we find $P(D r \neq 0)$.

Claim: Sine $D \neq 0$, there exist indices i, j such that $D[i][j] \neq 0$.

We can select a length n vector **v** such that v[j] = 1 and all other entries are 0. Observation: D v \neq 0.

Now consider a badly selected **r** such that D r = 0 (algorithm returns True).

Define: r' = r + v.

Observe: D r' = D r + D v = 0 + [something nonzero] \neq 0.

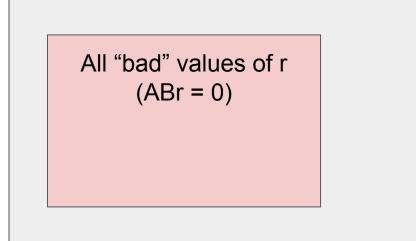
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Remember that v[i] = 0 for all i \neq j.
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This means that r' = r at all other indices that are not j.

All possible values of r

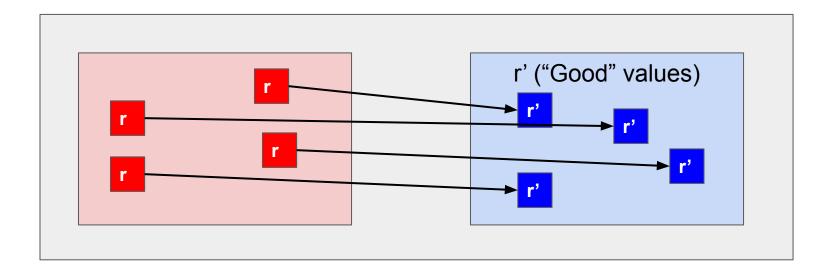
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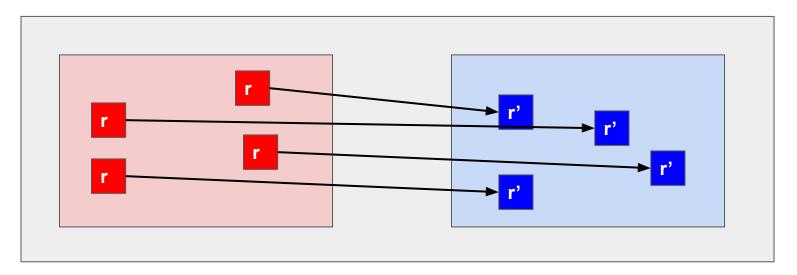
This means that r' = r at all other indices that are not j.



Remember that v[i] = 0 for all $i \neq j$.

This means that r' = r at all other indices that are not j.





For every "bad" r, such that ABr = 0, we can construct a "good" r' (that is also a valid vector we could select randomly) where ABr' \neq 0.

Thus, $P(ABr = 0) \le \frac{1}{2}$, and $P(correct) \ge \frac{1}{2}$.

Conclusion

Matrix multiplication (of n x n matrices) is slow; brute force is $O(n^3)$ and optimizations get to $O(n^{2.3})$ at best.

Frievalds' algorithm is a true-biased Monte Carlo algorithm that performs matrix multiplication **verification** in $O(n^2)$ time, using the fact that matrix-vector multiplication is faster than matrix-matrix multiplication.

Regardless of input, Frievalds' algorithm has at most a 50% chance of failure in the case where the algorithm should return False.

Exercises

Warmup: Matrix multiplication

- 1. Implement standard (brute-force) matrix multiplication in your programming language of choice.
- 2. (Optional) Implement Strassen's divide-and-conquer algorithm for matrix multiplication.
 - a. Resource: <u>https://www.geeksforgeeks.org/strassens-matrix-multiplication/</u>

Frievalds' Algorithm

- 3. Implement Frievalds' Algorithm in your programming language of choice for square matrices.
- 4. Try plotting the accuracy of Frievald's vs False cases versus the number of times you run the algorithm.
- 5. Extend Frievalds' Algorithm to matrix multiplication of any dimension.

Additional resources

Matrices:

https://www.khanacademy.org/math/precalculus/x9e81a4f98389efdf:matrices/x9e 81a4f98389efdf:mat-intro/a/intro-to-matrices

Matrix multiplication: <u>https://www.mathsisfun.com/algebra/matrix-multiplying.html</u>

Matmul in code: <u>https://www.geeksforgeeks.org/c-program-multiply-two-matrices/</u>

Frievalds implementation: <u>https://www.geeksforgeeks.org/freivalds-algorithm/</u>

Strassen's method: https://www.geeksforgeeks.org/strassens-matrix-multiplication/