



10 Matrix decompositions

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Introduction

- In this topic, we will
 - Review prime and polynomial factorizations
 - Learn that the word decomposition is often synonymous with factorization
 - Introduce the concept of a matrix decomposition
 - Describe five matrix decompositions useful in engineering
 - Identify those decompositions we will see in this course





Decompositions and factorizations

- In secondary school, you have heard the word *factorization*
 - This is the process of breaking down a more complex structure into simpler structures
 - Another term for this process is *decomposing*
 - We will discuss *matrix decompositions*, but this process is also known as *matrix factorization*





Prime factorization

- In secondary school, you have already learned of factorizations or decompositions:

- Every positive integer has a unique prime factorization

$$91 = 7 \times 13$$

$$1970 = 2 \times 5 \times 197$$

$$1513083350813089415334$$

$$= 2 \times 3 \times 13^2 \times 137 \times 19249 \times 20411 \times 27722467$$

- Many public-key cryptosystems rely on the difficulty of factoring integers to send secure messages

$$103665097958289904342438570560669042336391651299123031549302027602242813630593634896235339$$

$$= 252851506086608242857479289768516098250631 \times 409984103170743628346516399525200434439687427869$$





Polynomial factorization

- In secondary school, you saw that all polynomials could be written as a factorization of either monic linear or quadratic polynomials multiplied by a scaling factor

$$\begin{aligned} 3x^7 + 33x^6 + 111x^5 + 276x^4 + 621x^3 + 363x^2 + 1533x - 2940 \\ = 3(x - 1)(x + 4)(x + 7)(x^2 - x + 5)(x^2 + 2x + 7) \end{aligned}$$

- In this course, we have taken this one step further:

$$\begin{aligned} 3x^7 + 33x^6 + 111x^5 + 276x^4 + 621x^3 + 363x^2 + 1533x - 2940 \\ = 3(x - 1)(x + 4)(x + 7)(x + 1 + \sqrt{6}j)(x + 1 - \sqrt{6}j) \left(x - \frac{1}{2} + \frac{\sqrt{19}}{2}j\right) \left(x - \frac{1}{2} - \frac{\sqrt{19}}{2}j\right) \end{aligned}$$





Matrix factorizations or decompositions

- A matrix is representation of a linear mapping or linear operator, and it is difficult to understand

$$A = \begin{pmatrix} 3 & 2 & 1 & -3 \\ 2 & -3 & 2 & -2 \\ -1 & 0 & 4 & 1 \\ 1 & 2 & -1 & -1 \end{pmatrix} \quad B = \begin{pmatrix} 3 & 2 & 1 & -3 \\ 2 & -3 & 2 & -2 \\ -2 & 0 & 4 & 1 \\ 1 & 2 & -1 & -1 \end{pmatrix}$$

- A factorization or decomposition of a matrix tries to rewrite the matrix as a product of matrices that provide information about the original matrix





Looking ahead

- In this course, we will see two, maybe three, matrix decompositions
 - In this chapter, we will see the PLU decomposition
 - This is usually just called an LU decomposition
 - Later, we will see:
 - The eigenvalue decomposition or *diagonalization*
 - The singular-value decomposition (SVD)
- Not all matrices may be so decomposed
 - We will see that diagonalization really only applies to those matrices where the matrix equals its transpose: $A = A^T$





Examples

- We will quickly describe the following:
 - LU decomposition
 - Cholesky decomposition
 - QR decomposition
 - Diagonalization (or eigenvalue decomposition)
 - Singular-value decomposition





PLU decomposition

- Every $m \times n$ matrix A can be written as the product $A = PLU$ where:
 - P is an $m \times m$ permutation matrix
 - L is a lower-triangular $m \times m$ matrix with all ones on the diagonal
 - U is an upper triangular $m \times n$ matrix

$$\begin{pmatrix} -1.5 & 2.1 & 6.4 & 1.7 \\ 2 & 3.9 & 3.1 & 11.2 \\ 5 & 3 & 2 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ -0.3 & 1 & 0 \\ 0.4 & 0.9 & 1 \end{pmatrix} \begin{pmatrix} 5 & 3 & 2 & 1 \\ 0 & 3 & 7 & 2 \\ 0 & 0 & -4 & 9 \end{pmatrix}$$





Cholesky decomposition

- Every real $n \times n$ matrix A where $A = A^T$ and the angle between \mathbf{u} and $A\mathbf{u}$ is less than 90° for all \mathbf{u} can be written as the product

$$A = LL^T$$

where:

- L is a lower-triangular $n \times n$ matrix with all non-zero positive real numbers on the diagonal

$$\begin{pmatrix} 0.09 & -0.06 & 0.03 & -0.18 \\ -0.06 & 0.2 & 0.18 & -0.6 \\ 0.03 & 0.18 & 0.9 & 0.54 \\ -0.18 & 0.6 & 0.54 & 7.09 \end{pmatrix} = \begin{pmatrix} 0.3 & 0 & 0 & 0 \\ -0.2 & 0.4 & 0 & 0 \\ 0.1 & 0.5 & 0.8 & 0 \\ -0.6 & 1.2 & 0 & 2.3 \end{pmatrix} \begin{pmatrix} 0.3 & -0.2 & 0.1 & -0.6 \\ 0 & 0.4 & 0.5 & 1.2 \\ 0 & 0 & 0.8 & 0 \\ 0 & 0 & 0 & 2.3 \end{pmatrix}$$





QR decomposition

- Every real $m \times n$ matrix A can be written as the product

$$A = QR$$

where:

- Q is an orthogonal $m \times m$ matrix
- R is an upper-triangular $m \times n$ matrix

$$\begin{pmatrix} 1.92 & 2.3 & 0.14 & 4.06 \\ -2.56 & 1.6 & 0.48 & -2.58 \end{pmatrix} = \begin{pmatrix} -0.6 & 0.8 \\ 0.8 & 0.6 \end{pmatrix} \begin{pmatrix} -3.2 & -0.1 & 0.3 & -4.5 \\ 0 & 2.8 & 0.4 & 1.7 \end{pmatrix}$$





Diagonalization

- Every real $n \times n$ matrix A where $A = A^T$ can be written as

$$A = UDU^T$$

where:

- U is an orthogonal $n \times n$ matrix
- D is a diagonal $n \times n$ matrix with all real entries on that diagonal

$$\begin{pmatrix} 1.569 & -0.405 & 0.595 & -0.817 \\ -0.405 & 0.633 & -1.519 & -1.595 \\ 0.595 & -1.519 & -0.183 & -1.155 \\ -0.817 & -1.595 & -1.155 & -1.119 \end{pmatrix}$$

$$= \begin{pmatrix} 0.1 & 0.5 & 0.5 & 0.7 \\ 0.5 & -0.1 & 0.7 & -0.5 \\ 0.5 & -0.7 & -0.1 & 0.5 \\ 0.7 & 0.5 & -0.5 & -0.1 \end{pmatrix} \begin{pmatrix} -3.2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1.7 & 0 \\ 0 & 0 & 0 & 2.4 \end{pmatrix} \begin{pmatrix} 0.1 & 0.5 & 0.5 & 0.7 \\ 0.5 & -0.1 & -0.7 & 0.5 \\ 0.5 & 0.7 & -0.1 & -0.5 \\ 0.7 & -0.5 & 0.5 & -0.1 \end{pmatrix}$$





Singular-value decomposition (SVD)

- Every real $m \times n$ matrix A can be written as the product

$$A = U\Sigma V^T$$

where:

- U is an orthogonal $m \times m$ matrix
- Σ is a diagonal $m \times n$ matrix with all real entries on that diagonal
- V is an orthogonal $n \times n$ matrix

$$\begin{pmatrix} -1.382 & -0.254 & 1.282 & -1.994 \\ 0.824 & -0.872 & -2.024 & 0.008 \end{pmatrix} = \begin{pmatrix} -0.6 & -0.8 \\ -0.8 & 0.6 \end{pmatrix} \begin{pmatrix} 1.7 & 0 & 0 & 0 \\ 0 & 3.2 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0.1 & 0.5 & 0.5 & 0.7 \\ 0.5 & -0.1 & -0.7 & 0.5 \\ 0.5 & 0.7 & -0.1 & -0.5 \\ 0.7 & -0.5 & 0.5 & -0.1 \end{pmatrix}$$





Looking ahead

- You do not have to know any of the matrix decompositions we described here but will not cover
 - This is just making you aware of the variety of different decompositions that are useful in different circumstances
- In this topic, we will only look at the LU decomposition
 - In future topics, we will look at the eigenvalue and singular-value decompositions





Summary

- Following this topic, you now
 - Have reviewed the decompositions or factorizations from secondary school
 - Prime and polynomial factorizations
 - Are aware of five useful matrix decompositions or matrix factorizations that you may come across in the future





References

- [1] https://en.wikipedia.org/wiki/Matrix_decomposition
- [2] https://en.wikipedia.org/wiki/LU_decomposition
- [3] https://en.wikipedia.org/wiki/Cholesky_decomposition
- [4] https://en.wikipedia.org/wiki/QR_decomposition
- [5] https://en.wikipedia.org/wiki/Diagonalizable_matrix
- [6] https://en.wikipedia.org/wiki/Singular_value_decomposition





Acknowledgments

None so far.





Colophon

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