

# Diagonalization & Singular-Value Decomposition

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Use eigenvalues and eigenvectors to factor a matrix into a more useful product of matrices. (Diagonalization)  
Generalize this concept to non-diagonalizable matrices.

# Diagonalization

We know how to compute eigenvalues and eigenvectors, whether using MATLAB or by the characteristic polynomial. Fix a (diagonalizable) matrix  $A$ . Let  $D$  be a diagonal matrix whose diagonal entries are the eigenvalues  $\lambda_1, \dots, \lambda_n$  of  $A$ , and  $P$  be the matrix whose columns are the corresponding eigenvectors  $x_1, \dots, x_n$ , which form a basis of the vector space. Then:

$$A = PDP^{-1}.$$

Why? Consider how this decomposition acts on the eigenvectors.

# Non-diagonalizable matrices

Not every matrix is diagonalizable.

Consider the matrix

$$\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

This does not have a set of eigenvectors that span the space.  
How can we tell?

# Application: Matrix Powers

In many applications, we want to compute  $A^M$  without doing  $M - 1$  matrix multiplications. If we have a diagonalization  $A = PDP^{-1}$ , then

$$A^M = (PDP^{-1})(PDP^{-1})\cdots(PDP^{-1})$$

$$\implies A^M = PD(P^{-1}P)D(P^{-1}\cdots P)DP^{-1} = PD^M P^{-1}.$$

Power of a diagonal matrix  $D^M$  is the diagonal matrix with entries raised to the power.

# Singular Value Decomposition

Generalizes the notion of diagonalization for arbitrary  $m \times n$  matrices.

$A = U\Sigma V^*$  is a decomposition with  $U, V$  orthogonal matrices – corresponding to rotations – and  $\Sigma$  is a diagonal matrix ordered from largest to smallest entries – corresponding to stretching. The entries of  $\Sigma$  are called singular values of the matrix. They are eigenvalues of  $MM^*$  and  $M^*M$ .

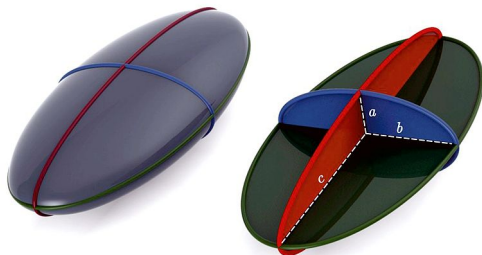
# SVD: Introduction

Strongly recommended piece on SVD:

<http://www.ams.org/samplings/feature-column/fcarc-svd>.

# Low rank approximation

If we throw away smaller singular values of the matrix, the matrix keeps doing its large-scaling operations, but instead of shrinking some vectors, it just collapses them to zero.



Imagine that instead of returning a vector in the larger ellipsoid at left, it returns a vector in the oval. (Image from Wikipedia)



# Applications: Image Processing

An image with an  $m \times n$  grid of grayscale pixels can be written as an  $m \times n$  matrix with entries between 0 and 255.

# Applications: Image Processing

For instance, consider this image:



This has  $700 \times 700$  pixels, almost all black or white, with some grayscale.