

What the decomposition gives you

Once $A = QDQ^T$ is in hand, every entry below is read off D alone. The orthogonal factor never changes — which is the whole reason functions of the matrix become functions of its eigenvalues.

FUNCTIONS OF THE MATRIX

4

1

Powers
 $k \geq 0$ integer; any k if all $\lambda > 0$

$A^k = QD^kQ^T$

The middle factors collapse: $QDQ^T QDQ^T = QD^2Q^T$, since $Q^TQ = I$. Raising a diagonal matrix to a power means raising each entry, so A^k costs no more than A — the reason **diagonalization** is worth the effort at all.

100 2

2

Inverse
all $\lambda \neq 0$

$A^{-1} = QD^{-1}Q^T$

Inverting a diagonal matrix means reciprocating its entries, so the whole inversion costs nothing once the factorization exists. A zero eigenvalue is precisely what makes it fail — the same condition as everywhere else, read off the diagonal.

3

Square root
all $\lambda \geq 0$ — **positive semi-definite**

$A^{1/2} = QD^{1/2}Q^T$

A symmetric square root exists exactly when no eigenvalue is negative, and it is unique among positive semi-definite matrices. This is the construction behind **Cholesky** and behind whitening a covariance matrix.

4

Exponential
always defined

$e^{At} = Qe^{Dt}Q^T$

No condition at all, since e^x is defined for every real x . This is what solves the linear system of differential equations $\dot{\mathbf{x}} = A\mathbf{x}$ — the solution is $e^{At}\mathbf{x}$, and the decomposition makes it computable.

$A \neq 0$

SCALARS, STRAIGHT OFF THE DIAGONAL

3

5

Trace
always

$\text{tr}(A) = \lambda_1 + \dots + \lambda_n$

The trace is **similarity invariant**, so $\text{tr}(A) = \text{tr}(D)$ and D is diagonal — the sum is immediate.

6

Determinant
always

$\det(A) = \lambda_1 \lambda_2 \dots \lambda_n$

Same argument with the multiplicative property: $\det Q \cdot \det D \cdot \det Q^T = \det A$, since $\det Q = \det Q^T = \pm 1$ and the two cancel.

7

Rank
always

count of $\lambda \neq 0$

Multiplying by an invertible matrix preserves rank, so $\text{rank}(A) = \text{rank}(D)$, which is the number of nonzero diagonal entries. For a symmetric matrix, algebraic and geometric multiplicity always agree — so this count is unambiguous in a way it is not for a general matrix.

The pattern is one rule: $f(A) = Qf(D)Q^T$, and $f(D)$ applies f to each diagonal entry. Powers, inverses, roots and exponentials are all the same statement with a different f , which is what makes a symmetric matrix genuinely easier to work with rather than merely nicer to look at.