

# Eigen-Stuff

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We will be discussing all the eigen stuff - eigenvectors, eigenvalues, eigenspace, and etc.

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A square  $n \times n$  matrix  $A$  can be viewed as a linear transformation of an  $n$  dimensional vector space  $V$  to itself. A linear transformation is a function, so this square matrix  $A$  maps vector  $x \in V$  to another vector  $x' \in V$ . Some of the vectors of the vector space have their direction unchanged under the linear transformation, and their will at most be a change in scale  $\lambda$ . This, in terms of an equation, is  $Ax = \lambda x$  where  $x$  is nonzero. We will call such  $x$  as **eigenvector** and its corresponding  $\lambda$  as **eigenvalue**.

To find such eigenvectors computationally wise involve some steps. First, we notice that if we have  $Ax = \lambda x$ , we will have  $(A - \lambda I)x = 0$  where this  $I$  is the  $n \times n$  identity matrix. Since  $x$  is nonzero, the matrix  $A - \lambda I$  must be **singular**. Singular means not invertible, and the columns of a singular matrix should be linearly dependent in order for it to be not invertible. Now, we recall that a singular matrix will have determinate 0, so  $\det A - \lambda I = 0$ . Solving this equation should give us the eigenvalues.

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We should note that the set of eigenvectors can form a vector space, which we call the **eigenspace**. The dimension of that space may not be the same as the number of eigenvectors since they may not be linearly independent. If all the eigenvalues are distinct, then the eigenvectors will be linearly independent. When there are repeated eigenvalues, more checking will be needed. Now, if the eigenspace has the same dimension as the original vector space, we can perform a change-of-basis to the matrix and get something very helpful.

For the  $n \times n$  matrix  $A$  with eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$  and eigenvectors  $v_1, v_2, \dots, v_n$ , we will have the eigenmatrix  $Q = [v_1 \ v_2 \ \dots \ v_n]$ . Then,  $AQ = [\lambda_1 v_1 \ \lambda_2 v_2 \ \dots \ \lambda_n v_n]$ . Also, if we have  $\Lambda$  the diagonal matrix with the eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$  on the diagonal, we would have  $Q\Lambda = [\lambda_1 v_1 \ \lambda_2 v_2 \ \dots \ \lambda_n v_n]$ . So, we have gotten  $AQ = Q\Lambda$ , or  $A = Q\Lambda Q^{-1}$ . This rewrites the original matrix  $A$  using eigen stuff, so it is known as the **eigendecomposition** of the matrix  $A$ .

As mentioned earlier, not all matrices can be decomposed like this. Those that can be eigendecomposed will have the same dimension of eigenspace and vector space, and they are known as **diagonalisable**, since the process of eigendecomposition will turn into a diagonalisation with a small twist of  $Q^{-1}AQ = \Lambda$  where  $\Lambda$  is a diagonal matrix.