

# Solving Linear Systems

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This is the key lecture about solving a system of  $n$  linear constant coefficient equations.

$$\frac{dy}{dt} = Ay$$

$y$  is a vector. How do we solve this system?

Eigenvectors are vectors that go in their own way. So when you have an eigenvector, it's like you have a one by one problem and the  $A$  becomes just a number  $\lambda$ . So for a general vector, everything is mixed together. But for an eigenvector, everything stays one dimensional. The  $A$  changes just to  $\lambda$  for that special direction. And of course, as always, we need  $n$  of those eigenvectors because we want to take the starting value.

We take our starting vector, which is probably not an eigenvector. We would make it a combination of eigenvectors. And we are OK because we are assuming that we have  $n$  independent eigenvectors.

$$y(0) = c_1x_1 + c_2x_2 + \dots + c_nx_n$$

And now we follow each starting value.

$$y(t) = c_1e^{\lambda_1 t}x_1 + \dots + c_ne^{\lambda_n t}x_n$$

You can see the stable time point. When  $\lambda_i < 0$  then the term is going to 0.

We will see an example. A Markov differential equation.

$$A = \begin{bmatrix} -1 & 2 \\ 1 & -2 \end{bmatrix}$$

The column will add up to 0. So there is the eigenvalue of 0 our powers is like the eigenvalue 0 for differential equations. Because  $e^{0t} = 1$ . So anyway, let's find the eigenvalues of that.

The first eigenvalue is 0 and the corresponding eigenvector is  $x_1 = [2, 1]^T$ . And my second eigenvalue is  $-3$  and the eigenvector is  $x_2 = [1, -1]^T$ . So we have done the preliminary work. Given this matrix, we have got the eigenvalues and eigenvectors. Now we take

$$y(0) = c_1x_1 + c_2x_2$$

And now the

$$y(t) = c_1e^{0t}x_1 + c_2e^{-3t}x_2$$

That's the evolution of a Markov process, a continuous Markov process. Compared to the powers of a matrix, this is a continuous evolving evolution of this vector. Now we are interested in steady state. Steady state is what happens as  $t \rightarrow \infty$ . Obviously, the second term goes quickly to 0. So we have  $c_1x_1$  is the steady state. We are thinking that no matter how you start, no matter what  $y_0$  is, as time goes on, the  $x_2$  part is going to disappear. And if you only have the  $x_1$  part in that ratio 2 : 1. So again, if we had movement between  $y_1$  and  $y_2$  or we have things evolving in time, the steady state is  $c_1x_1$ . This is the differential equations happen to have a Markov matrix. And we will show why 0 is an eigenvalue of the Markov matrix.

We have now two examples of the following fact. That if all columns add to  $s$ , then  $\lambda = s$  is an eigenvalue. That was the point from Markov matrices,  $s = 1$ . And if all rows of  $A$  add to  $s$ , then  $\lambda = s$  is an eigenvalue because  $A^T$  and  $A$  share the same eigenvalues. And maybe you would like to just see why that's true.

If we want the eigenvalues of a matrix, we look at  $\det(A - \lambda I)$ . That gives us eigenvalues of  $A$ . If we want the eigenvalues of  $A^T$ , we will look at  $\det(A^T - \lambda I)$ . But why are they the same? Because the determinant of a matrix and the determinant of its transpose are equal.

So for special matrices, we are able to identify important fact about their eigenvalue.