Hi, everyone. Today, we’re going to talk about linear transformations. So, we’ve seen linear transformations incognito all the time until now. We’ve played around with matrices. Matrices multiplying vectors, say, in \( \mathbb{R}^n \) and producing vectors in \( \mathbb{R}^m \).

So really, the language of linear transformations only provides a nicer framework when we want to analyze linear operations on more abstract vector spaces, like the one we have in this problem here. We’re going to work with the space of two by two matrices. And we’re going to analyze the operation, have the matrix \( A \), and we produce its transpose. OK. So please take a few minutes to try the problem on your own and come back.

Hi, again. OK. So the first question we need to ask ourselves is, indeed, why is \( T \) a linear operator? So what are the abstract properties that a linear operator satisfies? Well, what happens when \( T \) acts on the sum of two matrices, \( A \) and \( B \)?

So we’ve produced-- this is the matrix. The transpose of \( A + B \). But we know that this is \( A \) transpose, \( B \) transpose. And so, this is exactly \( T \) of \( A + B \). So the transformation that we’re analyzing takes the sum of two matrices into the sum of their transformations. OK. Similarly, it takes a multiple of a transformation into the multiple of the transformations. So it takes the matrix \( C \) times \( A \) to \( C \) times \( A \) transpose, which is \( C \ T \) of \( A \).

OK. So it is a linear operator. Now, can we figure out what its inverse is? Well, what does the transpose do? It takes a column and flips it into a row. So what happens if we apply the operation once again? Well, it's going to take the row and turn it back down to the column.

So applying the transformation twice, we come back to the original situation. So therefore, \( t \) squared is the identity. And from this, we infer that the inverse is the transformation itself.

Now, this was part one. Part two, we’ll compute the matrix of the linear transformation in the following two bases. So the first basis is, in fact-- it is the
standard basis for the space of two by two matrices. And the way we compute the matrix, we first compute what $T$ does to each of the basis elements. So $T$ of $v_1$. Let's go back. So here.

So $T$ takes the transpose of this matrix. And we see that the transpose of $1, 0, 0, 0$ is $1, 0, 0, 0$. So it's a symmetric matrix. So $T$ of $v_1$. Here's $v_1$. What about $T$ of $v_2$? Come back here. So this one comes here. $0$ comes here. And so we actually get $v_3$. So $T$ of $v_2$ is $v_3$. Similarly, $T$ of $v_3$ is $v_2$.

And finally, $T$ of $v_4$. Well, $v_4$ is a symmetric matrix as well. So the transpose doesn't change it. OK. Now, we encode this into a matrix in the following way. Essentially, the first column will tell us how $T$ of $v_1$ is expressed as a linear combination of the basis elements.

Or in this case, it's just $v_1$. So it's going to be $1$ times $v_1$ plus $0$, $v_2$ plus $0$, $v_3$ plus $0$, $v_4$. $T$ of $v_2$ is $v_3$. So we have $0, 0, 1, 0$. $T$ of $v_3$ is $0$ $v_1$, $1$ $v_2$, $0$ $v_3$, $0$ $v_4$. And $T$ of $v_4$ is $0$ $v_1$, $0$ $v_2$, $0$ $v_3$, $1$ $v_4$.

OK. So we've written down the matrix of the linear transformation, $T$, in the standard basis. And you can check that this is exactly what we want. The representation of some matrix, say, $1, 2, 3, 4$, in this standard basis is, it's the vector $1, 2, 3, 4$. $T$ takes this to its transpose $1, 3, 2, 4$.

So this in the basis is represented as $1, 3, 2, 4$. Right? And it's not hard to see that $MT$ multiplies this vector, we get exactly this vector. So we'll pause for a bit, so that I erase the board. And we're going to return with the representation of $T$ in the basis, $w_1, w_2, w_3,$ and $w_4$.

OK. So let's compute now the matrix $T$ in the basis $w_1, w_2, w_3,$ and $w_4$. We played the same game. We look at how $T$ acts on each of the basis vectors. So $T$ of $w_1$--well, $w_1$ is a symmetric matrix. So $T$ of $w_1$ is $w_1$. Similarly, with $w_2$ and $w_3$. They're all symmetric.

What about $w_4$? Well, we see that the $1$ comes down here. The negative one comes up here. I mean, in the end we just get the negative of the $w_4$. So, let me just
write this out. We had $T$ of $w_1$ equal to $w_1$, $T$ of $w_2$ equal to $w_2$, $T$ of $w_3$ equal to $w_3$, and $T$ of $w_4$, was negative on $w_4$.

So therefore, the matrix of the linear transformation, $T$, in this basis. I'm going to call the matrix $M$ prime. $T$ has a fairly simple expression. The only non-zero entries are on a diagonal. And they’re precisely 1, 1, 1, and negative 1.

And finally, let’s tackle the eigenvalues slash eigenvectors issue. Well, you’ve seen what an eigenvector for a matrix is. And the idea for an eigenvalue, eigenvector for a linear transformation is virtually the same.

And we are looking for the vectors $v$ and the scale is lambda such that $T$ of $v$ is lambda of $v$. But if you guys look back to what we just did with $w_1$, $w_2$, $w_3$, and $w_4$, you’ll see precisely that $w_1$, $w_2$, and $w_3$ are eigenvectors for $T$ with eigenvalue 1. And $w_4$ is an eigenvector for $T$ with eigenvalue negative 1.

So yeah, we essentially have solved the problem knowing a very, very nice basis in which we computed the linear transformation, $T$. So I'll leave it at that.