## Math 313 Lecture #21§5.1: Eigenvectors and Eigenvalues

We have learned many things about solving the matrix equation  $A\vec{x} = \vec{b}$ .

Now we turn our attention to solving another problem in linear algebra, namely the equation

$$A\vec{x} = \lambda\vec{x}$$

for an  $n \times n$  matrix A and a scalar  $\lambda$ .

Example. For the matrix

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}$$

and the scalar  $\lambda = 3$ , the vector

$$\vec{x} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

satisfies  $A\vec{x} = \lambda \vec{x}$  because

$$A\vec{x} = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix}$$
 and  $\lambda\vec{x} = 3 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix}$ .

Are there other choices of  $\vec{x}$  and  $\lambda$  that work too?

Certainly the choice of  $\vec{x} = \vec{0}$  works for any scalar  $\lambda$ , but this does not depend on what A is, so we will exclude this *trivial* case.

Definitions. An eigenvector of an  $n \times n$  matrix A is a **NONZERO** vector  $\vec{x}$  such that  $A\vec{x} = \lambda \vec{x}$  for some scalar  $\lambda$ .

An **eigenvalue** of an  $n \times n$  matrix A is a scalar (could be zero) such that there is a **NONZERO** vector  $\vec{x}$  such that  $A\vec{x} = \lambda \vec{x}$ ; such a vector is called an eigenvector corresponding to  $\lambda$ .

What is the significance of the eigenvectors and eigenvalues of A?

Example (Continued). We know that  $\lambda = 3$  is an eigenvalue of the matrix

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix},$$

and a corresponding eigenvector is

$$\vec{x} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}.$$

The eigenvector satisfies the equation  $A\vec{x} = 3\vec{x}$ , or rewritten,

$$A\vec{x} - 3\vec{x} = \vec{0}.$$

We would like to factor out the common  $\vec{x}$ , but we might mistakenly write

$$(A-3)\vec{x} = \vec{0}.$$

What is incorrect with this equation? There is no way to subtract the scalar 3 from the  $2 \times 2$  matrix A.

Instead, we write  $3\vec{x} = 3I\vec{x}$  for the  $2 \times 2$  identity matrix I, so that

$$(A-3I)\vec{x} = \vec{0}.$$

This homogeneous equation has a nontrivial solution in  $\vec{x}$ .

By the Inverse Matrix Theorem the matrix A - 3I is not invertible, which we verify directly: the determinant of

$$A - 3I = \begin{bmatrix} -2 & 1\\ 4 & -2 \end{bmatrix}$$

is 0, and so A - 3I is not invertible.

The nonzero vector  $\vec{x}$  belongs to the null space of A - 3I; in particular,  $\vec{x}$  is a basis for Nul(A - 3I) because

$$A - 3I \sim \begin{bmatrix} 2 & 1 \\ 0 & 0 \end{bmatrix}$$

so that  $u_1$  is a basic variable,  $u_2 = t$  is a free variable, and the solution set of  $(A-3I)\vec{u} = \vec{0}$  consists of the vectors

$$\vec{u} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = t \begin{bmatrix} 1 \\ 2 \end{bmatrix} = t\vec{x}.$$

Are there other values of  $\lambda$  for which  $A - \lambda I$  is not invertible? Yes,  $\lambda = -1$ .

An eigenvector of A corresponding to  $\lambda = -1$  is a nontrivial solution of  $(A + I)\vec{u} = \vec{0}$ , namely

$$\vec{x} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}.$$

It so happens that this  $\vec{x}$  is a basis for Nul(A + I).

We see in this example that eigenvectors of A are basis vectors of null spaces of matrices of the form  $A - \lambda I$ .

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Thus  $\lambda$  is an eigenvalue of A if and only if the homogeneous equation  $(A - \lambda I)\vec{x} = \vec{0}$  has a nontrivial solution.

For an eigenvalue  $\lambda$  of A we call the null space of  $A - \lambda I$  the **eigenspace** of A corresponding to  $\lambda$ .

For a real eigenvalue  $\lambda$  of a matrix A, the matrix transformation  $\vec{x} \mapsto A\vec{x} = \lambda \vec{x}$  acts like dilation/contraction on the eigenspace Nul $(A - \lambda I)$ .

At this point, we only have the method of guess and check to find eigenvalues of a square matrix.

But for a certain type of square matrix, the eigenvalues are readily found.

Theorem 1. The eigenvalues of a triangular square matrix are its diagonal entries.

WARNING: this does not say that we can find eigenvalues by row reduction. A counterexample to this is the invertible matrix

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}$$

whose eigenvalues are 3 and -1, but the eigenvalues of the I (the reduced row echelon form of A) are 1 and 1.

Proof. Suppose A is in triangular form with diagonal entries  $a_{ii}$  for i = 1, 2, ..., n.

For a fixed i set  $\lambda = a_{ii}$ .

One of the diagonal entries of  $A - \lambda I$  is zero (namely the (i, i) entry), and so det $(A - \lambda I) = 0$ .

By the Inverse Matrix Theorem, there is a nontrivial solution  $\vec{x}$  of  $(A - \lambda I)\vec{x} = \vec{0}$ .

This means that  $A\vec{x} = \lambda \vec{x}$ , and so  $\lambda$  is an eigenvalue.

The eigenvectors of a square matrix associated to distinct eigenvalues enjoy a valuable property.

Theorem 2. If  $\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_r$  are eigenvectors corresponding to distinct eigenvalues of a square matrix, then the set  $\{\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_r\}$  is linearly independent.

Proof. We prove this by way of contradiction, by supposing that the set of eigenvectors is linearly dependent.

Since  $\vec{v}_1 \neq \vec{0}$ , we can apply the linear dependent Theorem: there is a least index p such that the eigenvector  $v_{p+1}$  is a linear combination of the preceding p eigenvectors.

The choice of the least index p means that the set of eigenvectors  $\{\vec{v}_1, \ldots, \vec{v}_p\}$  is linearly independent (for otherwise there would be a smaller choice of p).

There are weights  $c_1, c_2, \ldots, c_p$  such that

$$c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_p \vec{v}_p = \vec{v}_{p+1}.$$

To this we apply the matrix transformation  $\vec{x} \mapsto A\vec{x}$  to obtain

$$c_1 A \vec{v}_1 + c_2 A \vec{v}_2 + \dots + c_p A \vec{v}_p = A \vec{v}_{p+1}.$$

Since  $A\vec{v}_i = \lambda_i \vec{v}_i$  for each i = 1, 2, ..., p + 1, we obtain

$$c_1\lambda_1\vec{v}_1 + c_2\lambda_2\vec{v}_2 + \dots + c_p\lambda_p\vec{v}_p = \lambda_{p+1}\vec{v}_{p+1}.$$

Multiplying the equation  $c_1\vec{v}_1 + c_2\vec{v}_2 + \cdots + c_p\vec{v}_p = \vec{v}_{p+1}$  through by  $\lambda_{p+1}$  gives

$$c_1\lambda_{p+1}\vec{v}_1 + c_2\lambda_{p+1}\vec{v}_2 + \dots + c_p\lambda_{p+1}\vec{v}_p = \lambda_{p+1}\vec{v}_{p+1}.$$

Subtracting this equation from  $c_1\lambda_1\vec{v}_1 + c_2\lambda_2\vec{v}_2 + \cdots + c_p\lambda_p\vec{v}_p = \lambda_{p+1}\vec{v}_{p+1}$  gives

$$c_1(\lambda_1 - \lambda_{p+1})\vec{v}_1 + c_2(\lambda_2 - \lambda_{p+1})\vec{v}_2 + \dots + c_p(\lambda_p - \lambda_{p+1})\vec{v}_p = \vec{0}.$$

The linear independence of the set  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p\}$  now implies that  $c_i(\lambda_i - \lambda_{p+1}) = 0$  for each  $i = 1, 2, \dots, p$ .

Thus  $c_i = 0$  for each i = 1, 2, ..., p since the eigenvalues are distinct.

This implies that  $v_{p+1} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_p \vec{v}_p = \vec{0}.$ 

[What is the contradiction here?]

But  $\vec{v}_{p+1}$  is an eigenvector, meaning it is nonzero, a contradiction.