

Math 350 Fall 2011
Notes about Diagonalization and Invariant subspaces

1. DIAGONALIZATION OF A MATRIX

In this section, let A be an $n \times n$ matrix with entries in a field F .

Definition 1. The characteristic polynomial of A is by definition the polynomial

$$f(t) = \det(A - tI),$$

where I is the $n \times n$ identity matrix.

It is thus a polynomial of degree n in a single variable t with coefficients in F .

Question 1. Consider the following 2×2 real matrix

$$\begin{pmatrix} 1 & -4 \\ 0 & -1 \end{pmatrix}.$$

Compute its characteristic polynomial.

Definition 2. λ is said to be an eigenvalue of A if it is a zero of the characteristic polynomial of A in F .

So $\lambda \in F$ is an eigenvalue of A , if and only if $\det(A - \lambda I) = 0$. It then follows that the nullspace of $A - \lambda I$ is non-zero; so there exists a non-zero vector $v \in F^n$ such that $(A - \lambda I)v = 0$, i.e. $Av = \lambda v$.

Definition 3. A vector $v \in F^n$ is said to be an eigenvector of A , if $v \neq 0$ and $Av = \lambda v$ for some $\lambda \in F$.

In fact if $v \neq 0$ and $Av = \lambda v$, then v is said to be an eigenvector of A corresponding to the eigenvalue λ . Equivalently, if $v \neq 0$, then v is an eigenvector of A corresponding to λ , if and only if v is in the nullspace of $A - \lambda I$.

Question 2. Find all eigenvalues and eigenvectors of the matrix in Question 1.

Definition 4. A is said to be diagonalizable (over F) if there exists an invertible matrix P , and a diagonal matrix D , both of which are $n \times n$ and have entries in F , such that

$$A = PDP^{-1}.$$

If A is diagonalizable, a representation of A in the form PDP^{-1} , where P is invertible and D is diagonal, is called a diagonalization of A .

In other words, A is diagonalizable if and only if it is similar to a diagonal matrix (over F).

Question 3. Show that the matrix in Question 1 over \mathbb{R} is diagonalizable over \mathbb{R} , and diagonalize it.

Theorem 1. A is diagonalizable over F , if and only if A has n linearly independent eigenvectors in F^n .

Proof. Suppose $A = PDP^{-1}$ for some invertible matrix P and a diagonal matrix D . Then $AP = PD$, so if v_1, \dots, v_n are the columns of P , and $\lambda_1, \dots, \lambda_n$ are the entries on the diagonal of D , then $Av_i = \lambda_i v_i$ for all $i = 1, \dots, n$. In particular, v_1, \dots, v_n are eigenvectors of A . They are linearly independent since P is invertible.

Conversely, suppose v_1, \dots, v_n are linearly independent eigenvectors of A , say corresponding to eigenvalues $\lambda_1, \dots, \lambda_n \in F$. Let P be the matrix whose i -th row is v_i , and D be the diagonal matrix whose i -th entry on the diagonal is λ_i . Then P is invertible, since its columns are linearly independent, and we have $AP = PD$, since $Av_i = \lambda_i v_i$ for $i = 1, 2, \dots, n$. It follows that $A = PDP^{-1}$, which implies that A is diagonalizable over F . \square

Definition 5. A set of n linearly independent eigenvectors of A in F^n is said to be an eigenbasis of F^n associated to A .

The above theorem can now be reformulated as

Theorem 2. A is diagonalizable over F , if and only if there exists an eigenbasis of F^n associated to A .

Definition 6. The span of all eigenvectors of A corresponding to an eigenvalue λ of A is called the eigenspace of A corresponding to λ .

Thus any eigenspace of A is in particular a subspace of F^n . In general one should think about eigenspaces instead of eigenvectors as much as possible, since eigenspaces are *canonical* objects associated to a matrix, while eigenvectors are not.

2. DIAGONALIZATION OF A LINEAR MAP

In this and subsequent sections, V is a vector space over F of dimension $n < \infty$, and $T: V \rightarrow V$ is a linear map from V into itself.

Recall that given any ordered basis α of V , one can find a matrix representation $[T]_\alpha$ of T with respect to α . This matrix is an $n \times n$ matrix with entries in F , and if β is a different ordered basis of V , then $[T]_\beta$ and $[T]_\alpha$ are similar to each other. Since the determinant of any two similar matrices are the same, we can make the following definition:

Definition 7. The determinant of a linear map $T: V \rightarrow V$ is defined to be the determinant of $[T]_\alpha$, where α is any ordered basis of V . We denote this by $\det(T)$.

This definition is independent of the choice of α . With this we define:

Definition 8. The characteristic polynomial of a linear map $T: V \rightarrow V$ is defined to be the polynomial $f(t) = \det(T - tI)$, where $I: V \rightarrow V$ is the identity map.

In other words, the characteristic polynomial of T is given by

$$f(t) = \det([T]_\alpha - tI),$$

where α is any ordered basis of V and I is the identity matrix. This is independent of the choice of α .

Question 4. Let V be the space of real polynomials of degree ≤ 2 . It is a vector space over \mathbb{R} . Let $T: V \rightarrow V$ be the linear map

$$T(p(x)) = p'(x).$$

Find the characteristic polynomial of T .

Definition 9. λ is said to be an eigenvalue of T if it is a zero of the characteristic polynomial of T in F .

So $\lambda \in F$ is an eigenvalue of T , if and only if $\det(T - \lambda I) = 0$. It then follows that the nullspace of $T - \lambda I$ is non-zero; so there exists a non-zero vector $v \in V$ such that¹ $(T - \lambda I)v = 0$, i.e. $Tv = \lambda v$. Conversely, if there is a non-zero vector $v \in V$ such that $Tv = \lambda v$, then λ is an eigenvalue of T .

Definition 10. A vector $v \in V$ is said to be an eigenvector of T , if $v \neq 0$ and $Tv = \lambda v$ for some $\lambda \in F$.

In fact if $v \neq 0$ and $Tv = \lambda v$, then v is said to be an eigenvector of T corresponding to the eigenvalue λ . Equivalently, if $v \neq 0$, then v is an eigenvector of T corresponding to λ , if and only if v is in the nullspace of $T - \lambda I$.

Question 5. Find all the eigenvalues and eigenvectors of T where T is the linear map in Question 4.

Definition 11. A linear map $T: V \rightarrow V$ is said to be diagonalizable, if there exists an ordered basis α of V such that the matrix representation $[T]_\alpha$ of T is a diagonal matrix.

It follows that such an ordered basis of V consists only of eigenvectors of T ; in fact $\alpha = \{v_1, \dots, v_n\}$, and $[T]_\alpha$ is a diagonal matrix with entries $\lambda_1, \dots, \lambda_n$ on the diagonal, then $Tv_i = \lambda_i v_i$ for all $i = 1, \dots, n$, and thus α consists of eigenvectors of T .

Conversely, if α is an ordered basis of T consisting only of eigenvectors of T , say $\alpha = \{v_1, \dots, v_n\}$, and $Tv_i = \lambda_i v_i$ for all $i = 1, \dots, n$ where $\lambda_1, \dots, \lambda_n \in F$, then the matrix representation $[T]_\alpha$ of T is a diagonal matrix with entries $\lambda_1, \dots, \lambda_n$, and T is diagonalizable. Thus we have proved:

Theorem 3. A linear map $T: V \rightarrow V$ is diagonalizable, if and only if there exists a basis of V that consists only of eigenvectors of T .

Such a basis is usually called an *eigenbasis* of V associated to T . Since in a vector space of dimension n , n vectors form a basis if and only if they are linearly independent, we have

Theorem 4. A linear map $T: V \rightarrow V$ is diagonalizable, if and only if there exists n linearly independent eigenvectors of T , where $n = \dim(V)$.

¹From now on we write Tv for $T(v)$ for the sake of brevity.

Question 6. Show that the linear map T in Question 4 is not diagonalizable.

Definition 12. If $T: V \rightarrow V$ is diagonalizable, then a diagonalization of T is by definition an ordered eigenbasis of V associated to T , together with the matrix representation $[T]_\alpha$.

In other words, it is a set of n linearly independent eigenvectors of T , and the diagonal matrix that consists of their corresponding eigenvalues.

Question 7. Let V be the space of all real polynomials of degree ≤ 1 . It is a vector space over \mathbb{R} . Let $S: V \rightarrow V$ be defined by

$$S(p(x)) = 2(x-1)p'(x)$$

for all $p(x) \in V$. Show that S is diagonalizable, and find a diagonalization of S .

Finally, comparing what we had in these two sections, we have:

Theorem 5. Suppose A is an $n \times n$ matrix with entries in F . Then A is diagonalizable, if and only if the linear map $T: F^n \rightarrow F^n$ associated to A is diagonalizable.

We also have:

Definition 13. The span of all eigenvectors of T corresponding to an eigenvalue λ of T is called the eigenspace of T corresponding to λ .

Thus any eigenspace of T is in particular a subspace of V . Again in general one should think about eigenspaces instead of eigenvectors as much as possible when one considers linear maps.

3. INVARIANT SUBSPACES

Again, in this section, $T: V \rightarrow V$ is a linear map from V into itself, where V is a vector space over a field F , and $\dim(V) = n < \infty$.

Definition 14. W is said to be an invariant subspace associated to T if it is a subspace of V , and $T(w) \in W$ for all $w \in W$.

Sometimes we also say that W is a T -invariant subspace of V .

Question 8. For example, let $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be given by

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2 & -4 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}.$$

Show that the x -axis is a T -invariant subspace of \mathbb{R}^2 , while the y -axis is not.

If $T: V \rightarrow V$ is a linear map, then the following are all trivial T -invariant subspaces of V : $\{0\}$, V , the nullspace of T , and the range of T . Also, any eigenspace of T , i.e. the nullspace of $T - \lambda I$ for any λ , is a T -invariant subspace of V ; so is the range of $T - \lambda I$ for any λ .

Definition 15. If $T: V \rightarrow V$ is linear and W is a T -invariant subspace of V , one can define the restriction of T to W by

$$\begin{aligned} T|_W &: W \rightarrow W \\ T|_W(w) &= T(w) \quad \text{for all } w \in W. \end{aligned}$$

The restriction of T to W is a linear map from W to W . One important fact is:

Theorem 6. If W is a T -invariant subspace of V , then the characteristic polynomial of $T|_W$ divides the characteristic polynomial of T ; more precisely, if $f(t)$ and $g(t)$ are the characteristic polynomials of T and $T|_W$ respectively, then there is a polynomial $q(t)$ with coefficients in F such that

$$f(t) = g(t)q(t).$$

Proof. Suppose $\alpha = \{v_1, \dots, v_m\}$ is an ordered basis of W . Extend it to an ordered basis $\beta = \{v_1, \dots, v_n\}$ of V . Then the matrix representation $[T]_\beta$ of T is of the form

$$\begin{pmatrix} A & B \\ 0 & C \end{pmatrix}$$

where A is an $m \times m$ matrix, B is an $m \times (n - m)$ matrix, C is an $(n - m) \times (n - m)$ matrix and 0 is the $(n - m) \times m$ zero matrix. In fact A is then the matrix representation $[T|_W]_\alpha$. It follows that if $f(t)$ and $g(t)$ are the characteristic polynomials of T and $T|_W$ respectively, then

$$f(t) = \det \left(\begin{pmatrix} A & B \\ 0 & C \end{pmatrix} - tI \right) = \det \begin{pmatrix} A - tI & B \\ 0 & C - tI \end{pmatrix},$$

so

$$f(t) = \det(A - tI) \det(C - tI) = g(t) \det(C - tI),$$

and letting $q(t) = \det(C - tI)$ we have the claim. \square

Question 9. For example, if W is the x -axis, and T is the linear map in Question 8, find the characteristic polynomial of $T|_W$, and show that it divides the characteristic polynomial of T .

One important example of construction of invariant subspaces is the following. Suppose $T: V \rightarrow V$ is a linear map, where V is a finite dimensional vector space over a field F . Suppose $v \in V$, $v \neq 0$. Let k be the largest positive integer such that $\{v, Tv, T^2v, \dots, T^{k-1}v\}$ are linearly independent². Such k exists since V is finite dimensional. Let W be the span of $\{v, Tv, \dots, T^{k-1}v\}$. Then W is a subspace of V , and we claim that W is T -invariant: in fact if $w \in W$, then

$$w = a_0v + a_1Tv + \dots + a_{k-1}T^{k-1}v$$

for some scalars $a_0, \dots, a_{k-1} \in F$, so

$$Tw = a_0Tv + a_1T^2v + \dots + a_{k-1}T^k v.$$

²From now on, we write T^i for the composition of i copies of T . In other words, T^2v is a shorthand of $T(Tv)$, T^3v is a shorthand for $T(T(Tv))$, etc. More generally, if $f(t)$ is a polynomial in a single variable t , say $f(t) = a_0 + a_1t + \dots + a_mt^m$, and $T: V \rightarrow V$ is a linear map, then $f(T)$ is by definition a linear map from V to V such that $f(T)v = a_0v + a_1Tv + \dots + a_mT^mv$ for all $v \in V$.

But $T^k v$ is a linear combination of $v, Tv, \dots, T^{k-1}v$, since $\{v, Tv, \dots, T^{k-1}v, T^k v\}$ is linearly dependent. Thus we have $T^k v \in W$, which proves that W is T -invariant.

Now $\{v, Tv, \dots, T^{k-1}v\}$ is a basis of W (why?), so $\dim(W) = k$. If

$$T^k v = -b_0 v - b_1 Tv - \dots - b_{k-1} T^{k-1} v$$

for some scalars b_0, \dots, b_{k-1} , then the characteristic polynomial of $T|_W$ can be shown to be

$$g(t) = (-1)^k (b_0 + b_1 t + \dots + b_{k-1} t^{k-1} + t^k)$$

(check!), so in particular $g(T)(v) = (-1)^k (b_0 v + b_1 Tv + \dots + b_{k-1} T^{k-1} v + T^k v) = 0$.

Now if $f(t)$ is the characteristic polynomial of T , then $g(t)$ divides $f(t)$, i.e. there exists a polynomial $q(t)$ with coefficients in F such that $f(t) = g(t)q(t)$. It follows that $f(T)(v) = q(T)g(T)v = q(T)(0) = 0$. Since this is true for all $v \in V$, we see that $f(T)$ is the zero map from V to V . This proves:

Theorem 7 (Cayley-Hamilton). *If V is a finite dimensional vector space over a field F , $T: V \rightarrow V$ is a linear map, and $f(t)$ is the characteristic polynomial of T , then the linear map $f(T): V \rightarrow V$ is the zero map, i.e. T ‘satisfies’ the characteristic polynomial of T .*

Corollary 8. *If A is an $n \times n$ matrix with entries in a field F , and $f(t)$ is the characteristic polynomial of A , then $f(A)$ is the zero matrix.*

4. CRITERIA ABOUT DIAGONALIZABILITY

Again in this section, V is a vector space over a field F of dimension $n < \infty$, and $T: V \rightarrow V$ is a linear map from V into itself. We state some important criteria about the diagonalizability of T .

Theorem 9. *If $v_1, \dots, v_k \in V$ are eigenvectors of T corresponding to the eigenvalues $\lambda_1, \dots, \lambda_k \in F$, and the eigenvalues $\lambda_1, \dots, \lambda_k$ are pairwise distinct, then the set of vectors $\{v_1, \dots, v_k\}$ is linearly independent.*

Proof. The proof is by induction on k . The claim is trivial if $k = 1$. Now when $k = 2$, i.e. if v_1, v_2 are eigenvectors of T corresponding to eigenvalues λ_1 and λ_2 , with $\lambda_1 \neq \lambda_2$, then suppose

$$a_1 v_1 + a_2 v_2 = 0$$

for some scalars a_1, a_2 . Applying $T - \lambda_2 I$ to both sides, we get

$$a_1 (\lambda_1 - \lambda_2) v_1 = 0,$$

which implies $a_1 = 0$. Hence $a_2 = 0$. So the case $k = 2$ is proved.

More generally, suppose the claim is true for a certain value of k . Suppose one has $k + 1$ eigenvectors of T , namely v_1, \dots, v_{k+1} , corresponding to eigenvalues $\lambda_1, \dots, \lambda_{k+1}$, such that no two of the λ_i 's are equal. Then if a_1, \dots, a_{k+1} are scalars such that

$$a_1 v_1 + \dots + a_{k+1} v_{k+1} = 0,$$

by applying $T - \lambda_{k+1}I$ to both sides, we get

$$a_1(\lambda_1 - \lambda_{k+1})v_1 + \cdots + a_k(\lambda_k - \lambda_{k+1})v_k = 0.$$

By induction hypothesis, $\{v_1, \dots, v_k\}$ are linearly independent, so

$$a_1(\lambda_1 - \lambda_{k+1}) = a_2(\lambda_2 - \lambda_{k+1}) = \cdots = a_k(\lambda_k - \lambda_{k+1}) = 0.$$

Since $\lambda_1, \dots, \lambda_k$ are all different from λ_{k+1} , this implies

$$a_1 = \cdots = a_k = 0,$$

so $a_{k+1} = 0$ as well, and that completes the induction. \square

Corollary 10. *If $T: V \rightarrow V$ has n distinct eigenvalues over F , where $n = \dim(V)$, then T is diagonalizable.*

Corollary 11. *Suppose V_1, \dots, V_k are eigenspaces of T corresponding to eigenvalues $\lambda_1, \dots, \lambda_k$, and the eigenvalues $\lambda_1, \dots, \lambda_k$ are pairwise distinct.*

(a) *If $v_1 \in V_1, \dots, v_k \in V_k$ satisfies*

$$v_1 + \cdots + v_k = 0,$$

then

$$v_1 = \cdots = v_k = 0.$$

(b) *If β_1, \dots, β_k are linearly independent subsets of V_1, \dots, V_k respectively, then $\beta_1 \cup \beta_2 \cup \cdots \cup \beta_k$ is a linearly independent subset of V .*

These are easy corollaries of the previous theorem. Their proofs are left as exercises.

Next is a little piece of terminology from algebra:

Definition 16. *If $p(t)$ is a polynomial of one variable with coefficients in a field F and that has degree n , then we say that $p(t)$ splits over F if there are scalars $c, a_1, \dots, a_n \in F$ such that*

$$p(t) = c(t - a_1)(t - a_2) \cdots (t - a_n).$$

In other words, $p(t)$ is said to split over F if and only if $p(t)$ can be completely factorized into a product of linear factors with coefficients in F .

Question 10. *Show that the polynomial $t^2 + 1$ splits over \mathbb{C} , but not over \mathbb{R} .*

Theorem 12. *Suppose V is a finite dimensional vector space over F , and a linear map $T: V \rightarrow V$ is diagonalizable. Then the characteristic polynomial of T splits over F .*

Definition 17. *If λ is an eigenvalue of a linear map $T: V \rightarrow V$, its algebraic multiplicity is the number of times the factor $(t - \lambda)$ appears in the factorization of the characteristic polynomial of T , and its geometric multiplicity is the dimension of the eigenspace corresponding to λ .*

Question 11. Suppose V is the vector space of real polynomials of degree ≤ 2 on \mathbb{R} . Let $T: V \rightarrow V$ be the linear map defined by

$$T(p(x)) = 2(x-1)p'(x) + (x-1)^2p''(x).$$

Find the eigenvalues of T , and their algebraic and geometric multiplicities.

From Theorem 12, we have:

Lemma 1. Suppose $T: V \rightarrow V$ is a linear map and $\dim(V) = n$. If T is diagonalizable, then the sum of the algebraic multiplicities of the eigenvalues of T is equal to n .

This is because if T is diagonalizable over F , then by Theorem 12, the characteristic polynomial of T splits, meaning that the characteristic polynomial of T has exactly n zeroes (counting multiplicities) in F . Thus the sum of the algebraic multiplicities of the eigenvalues of T is n , as desired.

The main result of this section is:

Theorem 13. The algebraic multiplicity of any eigenvalue of T is bigger than or equal to its geometric multiplicity. Furthermore, T is diagonalizable, if and only if equality holds for every eigenvalue of T , i.e. if and only if the algebraic multiplicity of every eigenvalue of T is equal to its geometric multiplicity.

Proof. If λ is an eigenvalue of T and M, m are its algebraic multiplicity and geometric multiplicities respectively, we first show $m \leq M$. To do so, let W be the eigenspace of T corresponding to λ , and $T|_W$ be the restriction of T to W . Let also $f(t)$ be the characteristic polynomial of T , and $g(t)$ be the characteristic polynomial of $T|_W$. Then by Theorem 6, $g(t)$ divides $f(t)$. But $\dim(W) = m$, and by computing explicitly the characteristic polynomial of $T|_W$, we get

$$g(t) = (\lambda - t)^m.$$

Now M is the number of factors of $(\lambda - t)$ that arises in the factorization of $f(t)$. So from $(\lambda - t)^m$ divides $f(t)$, we conclude that $m \leq M$, which is the desired claim.

Next, suppose the algebraic multiplicity of every eigenvalue of T is equal to its geometric multiplicity. Then let $\lambda_1, \dots, \lambda_k$ be a listing of the distinct eigenvalues of T , and M_1, \dots, M_k be their corresponding algebraic or geometric multiplicities. By Lemma 1, we get

$$M_1 + M_2 + \dots + M_k = n.$$

Now let V_1, \dots, V_k be the eigenspaces of T corresponding to $\lambda_1, \dots, \lambda_k$. Then their dimensions are M_1, \dots, M_k respectively. Let β_1, \dots, β_k be a basis of V_1, \dots, V_k respectively. Then $\beta := \beta_1 \cup \dots \cup \beta_k$ is linearly independent by Corollary 11(b). Furthermore, β has $M_1 + \dots + M_k = n$ vectors, and all elements of β are all eigenvectors of T . Thus T has n linearly independent eigenvectors, which shows that T is diagonalizable.

Now suppose T is diagonalizable. Then there exists an eigenbasis β of V associated to T . Let $\lambda_1, \dots, \lambda_k$ be a listing of the distinct eigenvalues of T , and V_1, \dots, V_k

be the eigenspaces of T corresponding to $\lambda_1, \dots, \lambda_k$ respectively. For $i = 1, \dots, k$, let $\beta_i = \beta \cap V_i$, and n_i be the number of elements of β_i . Then β is the disjoint union of β_1, \dots, β_k , since every vector in β is an eigenvector, and thus belongs to exactly one of the β_i 's. It follows that

$$(1) \quad n = \sum_{i=1}^k n_i.$$

If now m_i is the geometric multiplicity of λ_i , and M_i be the algebraic multiplicity of λ_i , then

$$(2) \quad n_i \leq m_i \leq M_i \quad \text{for all } i = 1, \dots, k.$$

The first inequality holds since β_i is a set of linearly independent vectors in V_i , where $m_i = \dim(V_i)$ and $n_i =$ number of elements of β_i ; the second inequality holds by the first part of our theorem. On the other hand,

$$(3) \quad \sum_{i=1}^k M_i = n$$

by Lemma 1. From (1), (2) and (3), we conclude that $m_i = M_i$ for all $i = 1, \dots, k$. This proves that the algebraic and geometric multiplicities are the same for each eigenvalue of T , and we conclude the proof of the theorem. \square

Question 12. Show that the linear map T in Question 11 is diagonalizable using Theorem 13.