

(1) Let V be a finite dimensional F -vector space. A linear transformation $T: V \rightarrow V$ is called idempotent if $T^2 = T$. Prove that if T is an idempotent linear transformation then there is a basis B of V such that the matrix of T with respect to B has the following form.

$$\begin{pmatrix} I_n & 0_{n \times m} \\ 0_{m \times n} & 0_{m \times m} \end{pmatrix}$$

60/60

where I_n is the $n \times n$ identity matrix and $0_{r \times s}$ denotes the $r \times s$ zero matrix.

Pf: First note that the only eigenvalues of T are 0 and 1 because: choose any basis for V . Consider $A = M_B(T)$. Clearly $A^2 = A$. But then: let $v \neq 0$ be an eigenvector with eigenvalue λ , i.e., $Av = \lambda v$. Is true that $Av = \lambda v \Rightarrow A^2 v = \lambda^2 v \Rightarrow \lambda v = \lambda^2 v \Rightarrow \lambda v - \lambda^2 v = 0$

$\Rightarrow (\lambda - \lambda^2)v = 0$ but since $v \neq 0 \Rightarrow \lambda - \lambda^2 = 0 \Rightarrow \lambda = \lambda^2$. The only solutions of this equation in any arbitrary field F are $\lambda = 0$ or $\lambda = 1$.

Now, we proved in class that if T has a basis consisting of eigenvectors, then there exists a basis B of V s.t. $M_B(T)$ is diagonal. Therefore, if $\lambda = 0$ and $\lambda = 1$, we can find a basis of eigenvectors for the eigenvalues $\lambda = 0$ or $\lambda = 1$, will prove what we wanted.

$V_0 = \{v \in V : T(v) = 0v = 0\} = \text{Ker}(T)$. Let $\{v_1, \dots, v_m\}$ be a basis for $\text{Ker}(T)$.
 $V_1 = \{v \in V : T(v) = 1 \cdot v = v\} = \text{Im}(T)$. Let $\{w_1, \dots, w_n\}$ be a basis for $\text{Im}(T)$.

By the dimension theorem, $\dim(V) = \dim(\text{Im}(T)) + \dim(\text{Ker}(T)) = n + m$. So the set $B = (w_1, \dots, w_n, v_1, \dots, v_m)$ is s.t. $\{v_1, \dots, v_m\} \cap \{w_1, \dots, w_n\} = \emptyset$ and any vector in V can be written as linear combinations of B . Hence, B is a basis. Moreover, the matrix of T with respect to B the desired form.

40

$$\begin{pmatrix} [T[w_1]]_B & [T[w_2]]_B & \dots & [T[w_n]]_B & [T[v_1]]_B & [T[v_2]]_B & \dots & [T[v_m]]_B \end{pmatrix} = \begin{pmatrix} I_n & 0_{n \times m} \\ 0_{m \times n} & 0_{m \times m} \end{pmatrix}$$

(2) Let V be a finite dimensional F -vector space. A linear transformation $T: V \rightarrow V$ is called nilpotent if $T^k = 0$, for some positive integer k .

(a) Prove that if T is a nilpotent linear transformation then there is a vector $v \neq 0$ in V such that $T(v) = 0$.

Pf: Let $T: V \rightarrow V$ be nilpotent. If $V = \{0\}$, then the result is trivial. Suppose then $V \neq \{0\}$. Pick $v_1 \in V$ s.t. $v_1 \neq 0$. Look at Tv_1 . there are two options: $Tv_1 = 0$, in which case we have found $v_1 \neq 0$ such that $Tv_1 = 0$.

Otherwise: $Tv_1 \neq 0$. Let $Tv_1 = v_2$, for some $v_2 \in V$, $v_2 \neq 0$. Now apply T again: $T(Tv_1) = Tv_2 \Leftrightarrow T^2v_1 = Tv_2$. Again, we have two possibilities: either $Tv_2 = 0$, in which case v_2 is the vector we wanted or $Tv_2 \neq 0$. Let $Tv_2 = v_3$.

(So that we apply T again: $T(T^2v_1) = T(Tv_2) \Leftrightarrow T^3v_1 = Tv_3$. Continue the process $k-1$ times. If at any step between 1 and $k-1$ we found v_i with $1 \leq i \leq k-1$ to be s.t. $T(v_i) = 0$, then we are done. Otherwise we have $T^{k-1}v_1 = v_{k-1}$; where $v_{k-1} \neq 0$. Apply T a final time: $T^k v_1 = Tv_{k-1}$; since T is nilpotent: $T^k v_1 = 0$ and v_1 is the vector we wanted, showing the result. ✓

(b) Prove that if W is a T -invariant subspace of V then both $T|_W$ and the induced linear transformation \bar{T} on V/W are nilpotent.

Pf: Let W be a T -invariant subspace of V , i.e., $T(W) \subseteq W$. Consider the basis $B_1 = (w_1, \dots, w_m)$ of W . We can complete this basis into a basis of V $(w_1, \dots, w_m, w_{m+1}, \dots, w_n)$. And so we may speak of the matrix of $T|_W$ in the basis B_1 : $M_{B_1}(T|_W)$, a square $m \times m$ matrix. Look at powers of $M_{B_1}(T|_W)$: $M_{B_1}(T|_W), M_{B_1}^2(T|_W), M_{B_1}^3(T|_W), \dots$. Since T is a nilpotent linear transformation, there exists a positive integer k such that $T^k = 0$. This means that for every $v \in V$: $T^k v = 0$. In particular, if we restrict T to W we get the result: $w \in W \Rightarrow w \in V$ and therefore, using the same argument as before $T^k w = 0 \Rightarrow T|_W^k = 0$ so $T|_W^k$ is nilpotent. A similar argument shows that the induced linear transformation is nilpotent: $T(v+W) = T(v) + W$ raised to k : $T(v+W)^k = T(v)^k + W = 0 + W = \bar{0}$.

(c) Prove that if T is a nilpotent linear transformation then there is a basis B of V such that the matrix of T with respect to B is strictly upper triangular (that is, all of the entries on the diagonal or below are 0).

Pf: Let $T: V \rightarrow V$ be nilpotent. By (a) there is a vector $v_1 \neq 0$ such that $Tv_1 = 0$. Let $B_1 = (v_1)$ be a basis for $\langle v_1 \rangle$. By (b) there is a basis B_2 for $\langle v_1, v_2 \rangle$ such that $T|_{\langle v_1, v_2 \rangle}$ is strictly upper triangular. Continue this process until we have a basis B for V such that $T|_B$ is strictly upper triangular.

M403 - Fall 2013 - HW 9 - Enrique Areyan

Pf: First note that $\lambda=0$ is the only eigenvalue of T because: $Tv = \lambda v$, $v \neq 0 \Rightarrow T^k v = T^{k-1}(Tv) = \lambda T^{k-1}(v) = 0 \Rightarrow \lambda T^{k-1}(v) = 0$ and $T^{k-1}(v) \neq 0 \Rightarrow \lambda = 0$.

Now, consider a basis for the eigenvalue 0. This is the same as a basis for $\text{Ker}(T)$. $B = (v_1, \dots, v_k)$ we can expand this basis to include a basis for $\text{Ker}(T^2)$. $B = (v_1, \dots, v_k, v_{k+1}, \dots, v_m)$ where (v_1, \dots, v_k) are a basis for $\text{Ker}(T)$ and v_{k+1}, \dots, v_m a basis for $\text{Ker}(T^2)$. Since T is nilpotent, eventually we get $\text{Ker}(T^k) = V$. Keep expanding the basis B until we get a basis for V : $B = (v_1, \dots, v_k, v_{k+1}, \dots, v_n)$. Consider $M_B(T)$. This matrix will be strictly upper triangular since $T(v_1), \dots, T(v_k) \in \text{Ker}(T)$ and $T(v_{k+1}), \dots, T(v_n)$ are in $\text{Ker}(T^p)$ for $k < p \leq n$ and thus are expressed using only vectors from previous v_{k+1}, \dots, v_n vectors (some, not all of them), the resulting matrix is strictly upper triangular.

(3). (a) Prove that the function $\text{Tr}: M_n(F) \rightarrow F$ given by sending A to $\text{Tr}(A)$ is a linear transformation.

Pf: We want to show:

(i) $\text{Tr}(A+B) \stackrel{?}{=} \text{Tr}(A) + \text{Tr}(B)$

(ii) $\text{Tr}(\alpha A) \stackrel{?}{=} \alpha \text{Tr}(A)$.

(i) $\text{Tr}(A+B) = \text{Tr}((a_{ij}) + (b_{ij})) = \text{Tr}((a_{ij} + b_{ij})) = \sum_{i=1}^n (a_{ii} + b_{ii})$
 $= \sum_{i=1}^n a_{ii} + \sum_{i=1}^n b_{ii} = \text{Tr}(A) + \text{Tr}(B)$.

(ii) $\text{Tr}(\alpha A) = \text{Tr}(\alpha(a_{ij})) = \text{Tr}(\alpha a_{ij}) = \sum_{i=1}^n \alpha a_{ii} = \alpha \sum_{i=1}^n a_{ii} = \alpha \text{Tr}(A)$.

\Rightarrow (i) & (ii) mean that Tr is linear.

(b) Prove that for all $A, B \in M_n(F)$, $\text{Tr}(AB) = \text{Tr}(BA)$.

Pf: $\text{Tr}(AB) = \text{Tr}((a_{ij})(b_{ij})) = \text{Tr}(c_{ij})$; where $C = AB$. we can write the elements of C explicitly: $(c_{ij}) = \sum_{k=1}^n a_{ik} b_{kj}$; hence,
 $\text{Tr}(AB) = \text{tr}((c_{ij})) = \text{tr}(\sum_{k=1}^n a_{ik} b_{kj}) = \sum_{i=1}^n \sum_{j=1}^n a_{ij} b_{ji} = \sum_{j=1}^n \sum_{i=1}^n b_{ji} a_{ij} = \text{Tr}(BA)$.

(c) Let $S: V \rightarrow V$ be a l.t. and let B, C be bases of V . Prove that $\text{Tr}(M_B(S)) = \text{Tr}(M_C(S))$. Give a definition of the trace of a l.t.

Pf: Let $M_C(S) = P M_B(S) P^{-1}$, where P is the change of basis matrix from the basis B to C . then:
 $\text{Tr}(M_C(S)) = \text{Tr}(P M_B(S) P^{-1})$ (grouping $M_B(S) P^{-1}$)
 $\text{Tr}(M_C(S)) = \text{Tr}((M_B(S) P^{-1}) P)$ (By part (b))
 $\text{Tr}(M_C(S)) = \text{Tr}(M_B(S) (P^{-1} P)) = \text{Tr}(M_B(S) I) = \text{Tr}(M_B(S)) \Rightarrow \boxed{\text{Tr}(M_C(S)) = \text{Tr}(M_B(S))}$

Definition: Let $T: V \rightarrow V$ be a linear transformation. Let B be an arbitrary basis of V . The trace of T is the trace of the matrix of T in the basis B , i.e., $\text{Tr}(T) = \text{tr}(M_B(T))$. This definition makes sense because the trace function is invariant under choice of basis which we proved before.

(4) From the book, page 126, problem 2.3.

Find all real 2×2 matrices that carry the line $y=x$ to the line $y=3x$.

Solution: We want to find a matrix $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ such that $a, b, c, d \in \mathbb{R}$.

$A \begin{bmatrix} x_1 \\ x_1 \end{bmatrix} = \begin{bmatrix} x_1 \\ 3x_1 \end{bmatrix}$

\downarrow

A vector on the line $y=x$ is of the form: (x_1, x_1) ; $x_1 \in \mathbb{R}$

A vector on the line $y=3x$ is of the form $(x_1, 3x_1)$ (or equivalently $(\frac{x_1}{3}, x_1)$; $x_1 \in \mathbb{R}$).

So, let us solve the linear system:

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x_1 \\ x_1 \end{bmatrix} = \begin{bmatrix} x_1 \\ 3x_1 \end{bmatrix} \Rightarrow \begin{cases} ax_1 + bx_1 = x_1 \\ cx_1 + dx_1 = 3x_1 \end{cases}$$

$\Rightarrow \begin{cases} (a+b-1)x_1 = 0 \\ (c+d-3)x_1 = 0 \end{cases}$

If $x_1 = 0$ then we are free to choose a, b, c, d . therefore, suppose $x_1 \neq 0$.

then $\begin{cases} a+b-1 = 0 \\ c+d-3 = 0 \end{cases} \Rightarrow \begin{cases} a+b = 1 \\ c+d = 3 \end{cases}$

therefore, any 2×2 real matrix $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$ that satisfies the conditions: (i) $a+b=1$ and (ii) $c+d=3$, will carry the line $y=x$ to the line $y=3x$.

(5) From the book, page 126, problem 3.4.

Let B be a complex $n \times n$ matrix. Prove or disprove: the operator T on the space of all $n \times n$ matrices defined by $T(A) = AB - BA$ is singular.

Solution: claim: the map T has a non-trivial kernel.

Pf: $T(I_n) = I_n B - B I_n = B - B = 0_{n \times n}$; where I_n is the identity $n \times n$.

So, $I_n \in \text{Ker}(T)$. Note that this is not the only matrix in $\text{Ker}(T)$.

For example; Let B be an idempotent matrix. Then

$$T(B) = BB - BB = B - B = 0. \text{ (this is trivial if } B=0\text{). (End of claim).}$$

Therefore, $\text{Ker}(T) \neq \{0\}$ and hence T is not an isomorphism, which means that T is singular.

(6) From the book, page 128, problem 0.4.

Let $A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$. Find a matrix P s.t. $P^{-1}AP$ is diagonal, and find a formula for the matrix A^{30} .

tlw

Solution: First let us find the eigenvalues of A : Eigenvalues of A satisfy

$$0 = \det(\lambda I - A) = \det \left(\begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} - \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \right) = \det \left(\begin{bmatrix} \lambda-2 & -1 \\ -1 & \lambda-2 \end{bmatrix} \right) = (\lambda-2)^2 - 1 = \lambda^2 - 4\lambda + 3$$

$$\Rightarrow \lambda^2 - 4\lambda + 3 = 0 \Rightarrow (\lambda-1)(\lambda-3) = 0 \Rightarrow \lambda_1 = 1 \text{ and } \lambda_2 = 3.$$

Let us find a basis for the eigenspaces:

$$V_1 = \{ v \in \mathbb{R}^2 : Av = v \} \Rightarrow \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \Rightarrow \begin{cases} 2x_1 + x_2 = x_1 \\ x_1 + 2x_2 = x_2 \end{cases}$$

$$\Rightarrow \begin{cases} x_1 + x_2 = 0 \\ x_1 + x_2 = 0 \end{cases} \Rightarrow x_1 = -x_2 \Rightarrow v_1 = \left\langle \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\rangle. \text{ Hence, an eigenvector is } \begin{pmatrix} 1 \\ -1 \end{pmatrix} \text{ for } \lambda = 1$$

$$V_2 = \{ v \in \mathbb{R}^2 : Av = 3v \} \Rightarrow \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 3 \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \Rightarrow \begin{cases} 2x_1 + x_2 = 3x_1 \\ x_1 + 2x_2 = 3x_2 \end{cases}$$

$$\Rightarrow \begin{cases} -x_1 + x_2 = 0 \\ x_1 - x_2 = 0 \end{cases} \Rightarrow x_1 = x_2 \Rightarrow v_2 = \left\langle \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\rangle. \text{ Hence, an eigenvector is } \begin{pmatrix} 1 \\ 1 \end{pmatrix} \text{ for } \lambda = 3.$$

the matrix P is: $P = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$. the inverse can be computed as follows:

$$\left[\begin{array}{cc|cc} 1 & 1 & 1 & 0 \\ -1 & 1 & 0 & 1 \end{array} \right] \xrightarrow{R_2 = R_2 + R_1} \left[\begin{array}{cc|cc} 1 & 1 & 1 & 0 \\ 0 & 2 & 1 & 1 \end{array} \right] \xrightarrow{R_2 = \frac{1}{2}R_2} \left[\begin{array}{cc|cc} 1 & 1 & 1 & 0 \\ 0 & 1 & \frac{1}{2} & \frac{1}{2} \end{array} \right] \xrightarrow{R_1 = R_1 - R_2} \left[\begin{array}{cc|cc} 1 & 0 & \frac{1}{2} & -\frac{1}{2} \\ 0 & 1 & \frac{1}{2} & \frac{1}{2} \end{array} \right]. \text{ check:}$$

$$P^{-1} = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \Rightarrow PP^{-1} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \frac{1}{2} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}. \text{ So we can diagonalize } A.$$

$$D = P^{-1}AP = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 & 3 \\ -1 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix}$$

Finally, let us find a formula for A^{30} . Note that $D = P^{-1}AP \Rightarrow$

$$D^{30} = (P^{-1}AP)^{30}; \text{ but } (P^{-1}AP)^{30} = P^{-1}A^{30}P \Rightarrow A^{30} = PD^{30}P^{-1}; \text{ where } D^{30} = \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix}^{30} = \begin{bmatrix} 1^{30} & 0 \\ 0 & 3^{30} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 3^{30} \end{bmatrix}$$

$$\Rightarrow D^{30} = P^{-1}A^{30}P \Rightarrow \boxed{A^{30} = PD^{30}P^{-1}}; \text{ concretely: } A^{30} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 3^{30} \end{bmatrix} \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 3^{30} + 1 & 3^{30} - 1 \\ 3^{30} - 1 & 3^{30} + 1 \end{bmatrix}$$