

Eigenvectors

We consider again the eigenvalue equation

$$f(\mathbf{v}) = \lambda \mathbf{v},$$

where $f \in \mathcal{L}(V)$, $\lambda \in \mathbb{F}$. Let us recall also its matrix formulation in the two equivalent forms considered previously:

$$A\mathbf{x} = \lambda\mathbf{x} \iff (\lambda I - A)\mathbf{x} = \mathbf{0}.$$

16.1 Eigenspaces

The first observation is that \mathbf{v} belongs to a certain subspace of V .

Proposition 16.1 Let (λ, \mathbf{v}) be an eigenpair of f . Then $\mathbf{v} \in E_\lambda := \ker(f - \lambda id_V)$. Moreover, $E_\lambda \leq V$.

Proof. Since $\lambda \mathbf{v} = (\lambda id_V)\mathbf{v}$, we find that

$$f(\mathbf{v}) - (\lambda id_V)(\mathbf{v}) = \mathbf{0} \iff (f - \lambda id_V)(\mathbf{v}) = \mathbf{0} \iff \mathbf{v} \in \ker(f - \lambda id_V).$$

Finally, since $f, id_V \in \mathcal{L}(V)$, the map $f - \lambda id_V \in \mathcal{L}(V)$, by closure in $\mathcal{L}(V)$. The result then follows, as the kernel of an endomorphism on V is a subspace of V . ■

The subspace property in the previous result suggests the next definition.

Definition 16.1 — Eigenspace. The subspace E_λ is the eigenspace of f associated with eigenvalue λ .

Note that for any λ , E_λ is non-trivial, since it contains at least one non-zero vector: an eigenvector associated with λ . This means that $\dim E_\lambda \geq 1$. Let us derive further properties of eigenspaces.

Proposition 16.2 Let $(\lambda, \mathbf{v}), (\lambda', \mathbf{v}')$ denote two distinct eigenpairs. Then $E_\lambda \cap E_{\lambda'} = \{\mathbf{0}_V\}$.

Proof. Assume, for a contradiction, that there exists a nonzero \mathbf{u} such that $\mathbf{u} \in E_\lambda \cap E_{\lambda'}$. Then

$$f(\mathbf{u}) = \lambda \mathbf{u} = \lambda' \mathbf{u} \implies (\lambda - \lambda')\mathbf{u} = \mathbf{0}_V \implies \lambda = \lambda',$$

which is the contradiction we sought. ■

We immediately obtain the following corollary.

Corollary 16.3 Eigenvectors corresponding to different eigenvalues (i.e., from different eigenspaces) are linearly independent.

Another corollary is included below.

Corollary 16.4 Let $\dim E_\lambda = 1$ for all $\lambda \in \text{spf}$. Then V is a direct sum of eigenspaces:

$$V = \bigoplus_{\lambda \in \text{spf}} E_\lambda.$$

In particular, the eigenvectors form a basis for V .

Can we actually have $\dim E_\lambda > 1$? The answer is provided by the following example:

Example 16.1 Let $V(\mathbb{F})$ be an n -dimensional vector space and let $f = id_V$. Then its matrix representation is I_n , which has a single eigenvalue $\lambda = 1$, with algebraic multiplicity n . Moreover, each canonical vector $\mathbf{e}_i \in \mathbb{R}^n$ is an eigenvector for λ , so that $E_\lambda = \mathbb{R}^n$ and $\dim E_\lambda = n$.

It is clear from this example that the algebraic multiplicity of λ is related to the dimension of E_λ . Let us look at this more closely.

16.2 Geometric multiplicity

Definition 16.2 The **geometric multiplicity** of λ is denoted by $\gamma(\lambda)$ and is defined to be the dimension of its associated eigenspace: $\gamma(\lambda) := \dim E_\lambda$.

Let $B_\lambda = \{\mathbf{v}_1, \dots, \mathbf{v}_r\}$ be a basis for E_λ . Then each element \mathbf{v}_j of B_λ is an eigenvector of f . Therefore, the geometric multiplicity of λ can be viewed as the number of linearly independent eigenvectors associated with λ .

The following result provides some initial insight into the existence of eigenspaces of dimension greater than one.

Proposition 16.5 Let $f \in \mathcal{L}(V(\mathbb{F}))$, where $V(\mathbb{F})$ is an n -dimensional vector space. Then every eigenvalue λ of f has geometric multiplicity no greater than the algebraic multiplicity: $\gamma(\lambda) \leq \alpha(\lambda)$.

Proof. Let $1 \leq r = \gamma(\lambda)$ and $B_\lambda = \{\mathbf{v}_1, \dots, \mathbf{v}_r\}$ be a basis for E_λ , i.e., $f(\mathbf{v}_k) = \lambda \mathbf{v}_k$ for $k = 1, 2, \dots, r$. Let us choose now a basis B for V containing B_λ . Then the matrix representation of f takes the form

$$A = \begin{bmatrix} \lambda I_r & B \\ O & C \end{bmatrix}.$$

Hence, using the properties of determinants,

$$p_A(t) = \det(tI - A) = \det(tI_r - \lambda I_r) \cdot \det(tI_{n-r} - C) = (t - \lambda)^r p_C(t),$$

where $p_C \in \mathcal{P}_{n-r}(\mathbb{F})$ is the characteristic polynomial of C . Therefore, $\alpha(\lambda) \geq r$, since there are at least r factors $t - \lambda$ of $p_A(t)$. Thus, $\gamma(\lambda) = |B_\lambda| = r \leq \alpha(\lambda)$. ■

An important consequence of the above result is that the sum of geometric multiplicities is no greater than n :

$$\sum_{k=1}^r \gamma(\lambda_k) \leq \sum_{k=1}^r \alpha(\lambda_k) = n.$$

Since the geometric multiplicity $\gamma(\lambda)$ can be viewed as the number of linearly independent eigenvectors associated with λ , we deduce that the total number of linearly independent eigenvectors of a linear map

can be less than n .

Can we have indeed $\gamma(\lambda) \leq \alpha(\lambda)$ for some λ ? The answer is yes; here is an example we encountered when we discussed invariant subspaces and which we recall below.

Example 16.2 Let $V = \mathbb{R}^2$ and let

$$f(\mathbf{v}) = A\mathbf{v}, \quad A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}.$$

Then the eigenvalues are $\lambda_1 = \lambda_2 = \lambda = 1$, but we find that $\gamma(\lambda) = 1 < \alpha(\lambda) = 2$, since

$$(\lambda I - A)\mathbf{v} = \mathbf{0} \iff \begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix} \mathbf{v} = \mathbf{0} \iff \begin{cases} v_1 \in \mathbb{R} \\ v_2 = 0 \end{cases} \iff \mathbf{v} \in \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\} =: E_\lambda.$$

Note. This is an example of so-called **Jordan block** (of size 2, with eigenvalue $\lambda = 1$).

There are plenty examples where the geometric multiplicity can take any value between 1 and n .

Exercise 16.1 The following matrices have a single eigenvalue $\lambda = 1$. In each case, find the geometric multiplicity $\gamma(\lambda)$.

$$\text{i. } A = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}, \quad \text{ii. } B = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$