# Eigenvalues and Eigenvectors

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#### 1 Preliminaries

#### 1.1 Subspaces

- $\cdot \mathbb{R}$ ,  $\mathbb{C}$  field of real and complex numbers.
- · Colspace of  $\mathbf{A}^{m \times n}$ :

$$C(\mathbf{A}) := \operatorname{span} \left\{ \operatorname{Cols}(\mathbf{A}) \right\} \subseteq \mathbb{R}^n$$

And  $Dim(\mathcal{C}(\mathbf{A})) = r$ . r is rank of  $\mathbf{A}$ .

· Rowspace of  $\mathbf{A}^{m \times n}$ :

$$\mathcal{R}(\mathbf{A}) := \operatorname{span} \{ \operatorname{Rows}(\mathbf{A}) \} = \mathcal{C}(\mathbf{A}^{\top}) \subseteq \mathbb{R}^{m}$$

And  $Dim(\mathcal{R}(\mathbf{A})) = r$ .

· Nullspace of  $\mathbf{A}^{m \times n}$ :

$$\mathcal{N}(\mathbf{A}) := \{ \mathbf{x} : \mathbf{A}\mathbf{x} = 0 \} \subset \mathbb{R}^n$$

 $Dim(\mathcal{N}(\mathbf{A})) = n - r.$ 

 $\cdot \mathcal{N}(\mathbf{A}^{\top}) := \{ \mathbf{x} : \mathbf{A}^{\top} \mathbf{x} = 0 \} \subset \mathbb{R}^{m}. \operatorname{Dim}(\mathcal{N}(\mathbf{A}^{\top})) = m - r.$ 

### 2 Properties

#### 2.1 Eigval

- · Eigval  $\lambda$ , eigvec  $\boldsymbol{v}$  such that:  $\boldsymbol{A}\boldsymbol{v} = \lambda \boldsymbol{v}$ .
- $\cdot$  cv is also an eigrec corresponding to  $\lambda$ , but only linearly indep. eigrecs are counted.

#### 2.2 Characteristic Polynomial

·  $\mathbf{A}$  is  $n \times n$  square, then the characteristic polynomial of  $\mathbf{A}$  is defined as.

$$P_{\boldsymbol{A}}(t) := \det(t\boldsymbol{I} - \boldsymbol{A})$$

- ·  $\lambda$  is eigval of  $\mathbf{A} \iff P_{\mathbf{A}}(\lambda) = 0 \iff \lambda$  is root of  $P_{\mathbf{A}}(t) = 0$ . Proof. ( $\Rightarrow$ )  $\lambda$  is eigval of  $\mathbf{A}$ : ( $\lambda \mathbf{I} - \mathbf{A})\mathbf{v} = 0$ .  $\mathbf{v}$  is an eigvec of  $\mathbf{A}$ , so  $\mathbf{v} \neq 0 \Rightarrow \mathbf{v} \in \mathcal{N}(\mathbf{A})$ . Hence  $\operatorname{Dim}(\mathcal{N}(\mathbf{A})) > 0 \Rightarrow r < n$ .  $\square$
- ·  $P_{\mathbf{A}}(t)$  can be represented as, sence we know the roots,

$$P_{\mathbf{A}}(t) = \prod_{j=1}^{n} (t - \lambda_j)$$

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· **D** being diagonal matrix, then  $P_{\mathbf{D}} = \prod_{j=1}^{n} (t - d_j)$ , which implies that  $\lambda_j = d_j$ , diagonal entries. Moreover,  $\mathbf{D}\mathbf{e}_j = d_j\mathbf{e}_j$ , so the *j*-th eigence is  $\mathbf{e}_j$  unit vec.

·  $\boldsymbol{L}$  being lower triangular, then  $P_{\boldsymbol{L}} = \prod_{j=1}^{n} (t - L_{jj})$ , which implies that  $\lambda_j = d_j$ . The last col of lower tri is  $L_{nn}\boldsymbol{e}_n$ , therefore  $\boldsymbol{L}\boldsymbol{e}_n = L_{nn}\boldsymbol{e}_n$ , i.e. the last eigence is  $\boldsymbol{e}_n$  unit vector. We can not tell about other eigences. Similar for  $\boldsymbol{U}$ .  $\boldsymbol{U}\boldsymbol{e}_1 = U_{11}\boldsymbol{e}_1$ , i.e. the first eigence of  $\boldsymbol{U}$  is  $\boldsymbol{e}_1$  unit vector.

#### 2.3 Multiplicity

- $\lambda(A)$  is the set of all eigvals of A, it is acturally the spectrum of A.
- · If  $\lambda \in \lambda(\mathbf{A})$  is a root of multiplicity  $m_{\lambda}$  of  $P_{\mathbf{A}}(t) = 0$ , define  $m_{\lambda}$  as multiplicity of eigval  $\lambda$ .
- · Square matrix **A** has exactly n eigvals  $(\lambda \in \mathbb{C})$ , counted with multiplicity, i.e.

$$\sum_{\lambda \in \lambda(\mathbf{A})} m_{\lambda} = n$$

This is because, due to fundamental principal of algebra,  $P_{\mathbf{A}}(t) = 0$  has exactly n roots, counted with multiplicity.

#### 2.4 Eigspace

· Eigenspace (eigspace) of  $\lambda$ :  $V_{\lambda} := \{ \boldsymbol{v} : \boldsymbol{A}\boldsymbol{v} = \lambda \boldsymbol{v} \}$ , i.e. the set of all eigenspace corresponding to  $\lambda$ . Dim $(V_{\lambda})$  is the number of linearly indep. eigenspace corresponding to  $\lambda$ . We have

$$1 \leq \text{Dim}(V_{\lambda}) \leq m_{\lambda}$$

Thm. eigvecs corresponding to different eigvals of  $\mathbf{A}$  are linearly indep.

*Proof.* Let  $v_1, ..., v_p$  correspond to different eigvals of  $A: \lambda_1, ..., \lambda_p \ (p \le n)$ . It suffices to show

$$c_1 v_1 + ... + c_n v_n = 0 \Rightarrow c_1 = ... = c_n = 0$$

Show by contradiction: suppose otherwise, i.e.

$$(\dagger): c_1 \mathbf{v}_1 + ... + c_{n-1} \mathbf{v}_{n-1} = \mathbf{v}_n$$

Apply  $\boldsymbol{A}$  both sides:

$$m{A}(c_1m{v}_1 + ... + c_{p-1}m{v}_{p-1}) = m{A}m{v}_p$$
  
 $c_1\lambda_1m{v}_1 + ... + c_{p-1}\lambda_{p-1}m{v}_{p-1} = \lambda_pm{v}_p$ 

 $(\dagger) \times \lambda_p$ , substracted from last equation:

$$\sum_{j=1}^{p-1} c_j (\lambda_1 - \lambda_p) \boldsymbol{v}_j = \lambda_p \boldsymbol{v}_p - \lambda_p \boldsymbol{v}_p = 0$$

Which is not possible since  $\exists c_j \neq 0$ , and  $(\lambda_j - \lambda_p) \neq 0 \ \forall j$ . Contradiction.  $\Box$ 

#### 2.5 Miscellaneous

- ·  $\mathbf{A}$  singular  $\iff 0 \in \lambda(\mathbf{A})$ . Proof.  $\det(0\mathbf{I} - \mathbf{A}) = 0 \iff \mathbf{A}$  singular.  $\square$
- ·  $\boldsymbol{A}$  invertible.  $\lambda \in \lambda(\boldsymbol{A}), \boldsymbol{v} \in V_{\lambda}(\boldsymbol{A})$ . Then  $\frac{1}{\lambda} \in \lambda(\boldsymbol{A}^{-1}), \boldsymbol{v} \in V_{\frac{1}{\lambda}}(\boldsymbol{A}^{-1})$ . Proof.  $\boldsymbol{A}\boldsymbol{v} = \lambda \boldsymbol{v} \Rightarrow \boldsymbol{A}^{-1}\boldsymbol{A}\boldsymbol{v} = \lambda \boldsymbol{A}^{-1}\boldsymbol{v} \Rightarrow \frac{1}{\lambda}\boldsymbol{v} = \boldsymbol{A}^{-1}\boldsymbol{v}$ .

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·  $\boldsymbol{A}$  has eigtuple  $(\lambda, \boldsymbol{v})$ , then  $\boldsymbol{A}^k$  with  $(\lambda^k, v)$ .  $P(\boldsymbol{A})$  is a polynomial of  $\boldsymbol{A}$ , has eigtuple  $(P(\lambda), \boldsymbol{v})$ .

·  $\boldsymbol{A}$  and  $\boldsymbol{A}^{\top}$  have same eigvals. Proof.  $P_{\boldsymbol{A}}(t) = \det(t\boldsymbol{I} - \boldsymbol{A}) = \det((t\boldsymbol{I} - \boldsymbol{A})^{\top}) = \det(t\boldsymbol{I} - \boldsymbol{A}^{\top}) = P_{\boldsymbol{A}^{\top}}(t)$ . Thus have same roots.

· First, second and the last term of  $P_{\mathbf{A}}(t)$ :

$$P_{\mathbf{A}}(t) = t^n - \operatorname{tr}(\mathbf{A})t^{n-1} + \dots + (-1)^n \det(\mathbf{A})$$

Moreover,  $\sum_{j=1}^{n} \lambda_j = \operatorname{tr}(\boldsymbol{A})$  and  $\prod_{j=1}^{n} \lambda_j = \operatorname{det}(\boldsymbol{A})$ .

Proof.  $P_{\mathbf{A}}(t) = \det(t\mathbf{I} - \mathbf{A}).$ 

 $P_{\mathbf{A}}(0) = \det(-\mathbf{A}) = (-1)^n \det(\mathbf{A}).$ 

 $t^n, t^{n-1}$  arises with prod term of diagonal entries:

$$\prod_{j=1}^{n} (t - A_{jj}) = t^{n} - t^{n-1} \sum_{j=1}^{n} A_{jj} + \dots$$

which have coefficients 1 and  $-tr(\mathbf{A})$ . Moreover

$$P_{\mathbf{A}}(t) = \prod_{j=1}^{n} (t - \lambda_j) = t^n - t^{n-1} \sum_{j=1}^{n} \lambda_j + \dots + (-1)^n \prod_{j=1}^{n} \lambda_j$$

Finished the proof.  $\Box$ 

## 3 Diagonal Form

· A square, A is diagonalizable iff  $\exists$  diagonal  $\Omega$ , invertible V, s.t.

$$\mathbf{A} = \mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1}$$

Entries of  $\Lambda$  are eigvals of A, and cols of V are corresponding eigvecs.

*Proof.*  $AV = V\Lambda V^{-1}V = V\Lambda$ .

 $AV = (Av_1|...|Av_n); V\Lambda = (V\lambda_1e_1|...|V\lambda_ne_n) = (v_1\lambda_1|...|v_n\lambda_n). \square$ 

Since V is nonsingular,  $\{v_k\}$  must be linearly indep.

- $\cdot$  **A** is diagonalizable iff it has *n* linearly indep. eigences.
- · If  $\mathbf{A} = \mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1}$ , then  $\mathbf{A}^p = \mathbf{V} \mathbf{\Lambda}^p \mathbf{V}^{-1}$ . Proof.  $\mathbf{A}^p = \mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1} \mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1} ... \mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1} = \mathbf{V} \mathbf{\Lambda}^p \mathbf{V}^{-1}$ .  $\square$ Similarly,  $\mathbf{A}^{-1} = \mathbf{V} \mathbf{\Lambda}^{-1} \mathbf{V}^{-1}$ , and  $\mathbf{A}^{-p} = \mathbf{V} \mathbf{\Lambda}^{-p} \mathbf{V}^{-1}$ .

## 4 Diagonally Dominant Matrices

·  $R_j$  defined as sum of absolute value of entries on j-th row except for the main diagonal one  $A_{jj}$ .

$$R_j := \sum_{k=1, k \neq j}^n |A_{jk}|$$

- · **A** is a weakly diagonal dominant matrix iff  $|A_{jj}| \ge R_j$  for all j = 1, ..., n.
- · **A** is a strictly diagonal dominant matrix iff  $|A_{jj}| > R_j$  for all j = 1, ..., n.

Thm. (Gershgorin)  $A^{n\times n}$ , for any eigval  $\lambda$ , there exists index j, s.t.

$$|\lambda - A_{jj}| \le R_j$$

Alternative statement:

$$\lambda(\mathbf{A}) \subseteq \bigcup_{j=1}^{n} D(A_{jj}, R_{j})$$

Where  $D(A_{jj}, R_j) := \{z \in \mathbb{C} : |z - A_{jj}| \leq R_j\}$  being disc in complex field, centerred at  $A_{jj}$ , with radius  $R_j$ . Let  $\boldsymbol{v}$  be eigval corresponding to  $\lambda$ , index j is acturally the index of entry in  $\boldsymbol{v}$  who has biggest absolute value.

Proof.  $\mathbf{A}\mathbf{v} = \lambda \mathbf{v}, \mathbf{v} = (v_1, ..., v_n)^{\top}. \ j = \operatorname{argmax} |v_j|.$ 

 $\langle \boldsymbol{a}_{j}, \boldsymbol{v} \rangle = \lambda v_{j}$ , angle stands for inner product,  $\boldsymbol{a}_{j}$  is j-th row of  $\boldsymbol{A}$ . I.e.

$$\lambda v_{j} = \sum_{i=1}^{n} A_{ji} v_{i} = A_{jj} v_{j} + \sum_{i=1, i \neq j}^{n} A_{ji} v_{i}$$

$$\lambda - A_{jj} = \frac{1}{v_{j}} \sum_{i=1, i \neq j}^{n} A_{ji} v_{i}$$

$$|\lambda - A_{jj}| \le \frac{1}{|v_{j}|} \sum_{i=1, i \neq j}^{n} |A_{ji}| |v_{i}| = \sum_{i=1, i \neq j}^{n} |A_{ji}| \frac{|v_{i}|}{|v_{j}|}$$

$$\le \sum_{i=1, i \neq j}^{n} |A_{ji}| = R_{j}$$

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