

## Question 1

Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be a real symmetric  $n \times n$  matrix. Prove the following:

- The eigenvalues of  $\mathbf{A}$  are real
- The eigenvectors corresponding to distinct eigenvalues are orthogonal

Assume now that in the case of repeated eigenvalues the eigenvectors are *still* orthogonal (i.e., for  $\lambda_i = \lambda_j$  it is still the case that the corresponding eigenvectors  $\mathbf{v}_i$  and  $\mathbf{v}_j$  are orthogonal). Prove the following properties of the eigenvector matrix  $\mathbf{V} = [\mathbf{v}_1 \cdots \mathbf{v}_n]$ :

- $\mathbf{V}$  is an *orthoogonal* matrix, i.e.,  $\mathbf{V}^{-1} = \mathbf{V}^T$
- $\mathbf{V}^T \mathbf{A} \mathbf{V} = \mathbf{\Lambda}$  where  $\mathbf{\Lambda}$  is a diagonal matrix of eigenvalues

## Question 2

Let  $\mathbf{A} \in \mathbb{R}^{m \times n}$  be a real matrix. Prove the following results:

- The general solution of the linear system  $\mathbf{A}\mathbf{x} = \mathbf{b}$  is  $\mathbf{x} = \mathbf{x}_h + \mathbf{x}_p$  where  $\mathbf{x}_h$  is the *homogeneous solution* and satisfies  $\mathbf{A}\mathbf{x}_h = \mathbf{0}$  and  $\mathbf{x}_p$  is the *particular solution* and satisfies  $\mathbf{A}\mathbf{x}_p = \mathbf{b}$
- In the case where  $\mathbf{A}$  is *square*, the unique solution to the linear system  $\mathbf{A}\mathbf{x} = \mathbf{b}$  is  $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$  and this solution exists if and only if the matrix  $\mathbf{A}$  is nonsingular.

## Question 3

Let  $\mathbf{A}, \mathbf{B} \in \mathbb{R}^{n \times n}$  be constant  $n \times n$  matrices. Furthermore, let  $a(s)$  and  $b(s)$  each be arbitrary polynomials in  $s$  of degree  $m$ . Prove the following result:

$$a(\mathbf{A})b(\mathbf{B}) = b(\mathbf{B})a(\mathbf{A}) \iff \mathbf{AB} = \mathbf{BA}$$

## Question 4

Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be a constant  $n \times n$  matrix. Prove that the matrix exponential  $e^{\mathbf{A}}$  can be written as a linear combination of powers of  $\mathbf{A}$  of degree  $n - 1$  or less, i.e.,

$$e^{\mathbf{A}} = \sum_{k=0}^{n-1} \alpha_k \mathbf{A}^k$$

where  $\alpha_k$ , ( $k = 0, \dots, n - 1$ ) are real numbers.

## Question 5

Let  $\mathbf{B} \in \mathbb{R}^{n \times n}$  be an  $n \times n$  real matrix. Furthermore, let  $\mathbf{L}(\mathbf{A}) = \mathbf{A}\mathbf{B} + \mathbf{B}\mathbf{A}$  be a matrix operator whose input argument is  $\mathbf{A}$ . Show that  $\mathbf{L}$  is a linear operator.

## Question 6

Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be an  $n \times n$  real square matrix. Furthermore, let  $\mathbf{x}$  be an eigenvector of  $\mathbf{A}$  with eigenvalue  $\lambda$ . Prove that

$$\mathbf{A}^k \mathbf{x} = \lambda^k \mathbf{x}$$

for all  $k \geq 1$ .

## Question 7

Let  $p(s)$  be a polynomial of degree  $n$  in  $s \in \mathbb{C}$ . Furthermore, let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be an  $n \times n$  real square matrix. Finally, let  $\lambda$  be an eigenvalue of  $\mathbf{A}$  with eigenvector  $\mathbf{x}$ . Prove that  $p(\lambda)$  is an eigenvalue of  $p(\mathbf{A})$  with eigenvector  $\mathbf{x}$ .

## Question 8

Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be an  $n \times n$  real matrix. Suppose that the eigenvectors of  $\mathbf{A}$  form a complete set, i.e.,  $\text{span}(\mathbf{v}_1 \cdots \mathbf{v}_n) = \mathbb{R}^n$  (i.e., the set of eigenvectors  $(\mathbf{v}_1, \dots, \mathbf{v}_n)$  form a linearly independent set that spans  $\mathbb{R}^n$ ). Prove the following result:

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

where  $\mathbf{V} = [\mathbf{v}_1 \cdots \mathbf{v}_n]$  is the matrix of eigenvectors and  $\mathbf{D} = \text{diag}(\lambda_1, \dots, \lambda_n)$  is a diagonal matrix whose diagonal elements are the eigenvalues of  $\mathbf{A}$ .

## Question 9

Let  $p(s)$  be a polynomial in  $s \in \mathbb{C}$  of degree  $n$ . Furthermore, let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be an arbitrary square matrix. Finally, let  $\mathbf{T} \in \mathbb{R}^{n \times n}$  be a nonsingular square matrix. Prove that

$$p(\mathbf{T}^{-1} \mathbf{A} \mathbf{T}) = \mathbf{T}^{-1} p(\mathbf{A}) \mathbf{T}$$

## Question 10

Prove the following properties of the state transition matrix  $\Phi(t, \tau)$ :

(a)  $\Phi(t, \tau) = \Phi(t, t_1) \Phi(t_1, \tau)$

- (b)  $\Phi(t, \tau) = \Phi^{-1}(\tau, t)$
- (c)  $\frac{\partial \Phi(t, \tau)}{\partial t} = \mathbf{A}(t)\Phi(t, \tau)$
- (d)  $\frac{\partial \Phi(t, \tau)}{\partial \tau} = -\Phi(t, \tau)\mathbf{A}(\tau)$
- (e)  $\Phi(t, \tau)$  is nonsingular  $\forall t$  and  $\tau$

## Question 11

Let  $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x}$  be a linear time-invariant system with initial condition  $\mathbf{x}(0) = \mathbf{x}_0$ . The solution of this system can be written implicitly in integral form as

$$\mathbf{x}(t) = \mathbf{x}(0) + \int_0^t \mathbf{A}\mathbf{x}(\tau)d\tau$$

Suppose now that we use the following iterative procedure to solve the problem:

$$\mathbf{x}^{(k+1)}(t) = \mathbf{x}(0) + \int_0^t \mathbf{A}\mathbf{x}^{(k)}(\tau)d\tau$$

where  $\mathbf{x}^{(k)}(t)$  is the solution of the  $k^{\text{th}}$  iteration ( $k \geq 0$ ). Assuming that the zeroth iteration is  $\mathbf{x}^{(0)} = \mathbf{x}_0$ , prove the following result:

$$\lim_{k \rightarrow \infty} \mathbf{x}^{(k)}(t) = \mathbf{x}(t) = e^{\mathbf{A}t}\mathbf{x}_0$$

Repeat the procedure for  $\mathbf{x}^{(0)} = \bar{\mathbf{x}} \neq \mathbf{x}_0$ . How does your result differ in the second case as compared to the first case?