

Fundamental Matrices

→ Read section 7.7

Consider the system of linear equations;

$$\mathbf{x}' = \mathbf{P}(t)\mathbf{x} \quad (1)$$

A fundamental matrix for (1) is any matrix $\mathbf{\Psi}(t)$ that satisfies

$$\mathbf{\Psi}'(t) = \mathbf{P}(t)\mathbf{\Psi}(t) \quad (2)$$

Note that (2) is a first order differential equation for an unknown matrix, and as such has a unique solution for every initial data $\mathbf{\Psi}(t_0) = \mathbf{B}$ (\mathbf{B} is any given matrix).

One way to form such a matrix is to use any fundamental set of solutions $\mathbf{x}^{(1)}(t), \dots, \mathbf{x}^{(n)}(t)$ of (1) as it's columns;

$$\mathbf{\Psi}(t) = \begin{pmatrix} x_1^{(1)}(t) & \cdots & x_1^{(n)}(t) \\ \vdots & & \vdots \\ x_n^{(1)}(t) & \cdots & x_n^{(n)}(t) \end{pmatrix}$$

Check that (2) holds for this matrix.

For this fundamental set of solutions, the general solution of (1) is

$$\begin{aligned} \mathbf{x}(t) &= c_1\mathbf{x}_1^{(1)}(t) + \cdots + c_n\mathbf{x}^{(n)}(t) \\ &= \mathbf{\Psi}(t)\mathbf{c}, \quad \mathbf{c} = (c_1, \dots, c_n) \end{aligned} \quad (3)$$

The same is true for *any* fundamental matrix $\mathbf{\Psi}(t)$; we can write the general solution of (1) as $\mathbf{\Psi}(t)\mathbf{c}$.

Let $\mathbf{x}(t) = \mathbf{\Psi}(t)\mathbf{c}$ be a solution of (1). If the initial condition is $\mathbf{x}(t_0) = \mathbf{x}_o$, then

$$\begin{aligned} \mathbf{\Psi}(t_0)\mathbf{c} &= \mathbf{x}_o \\ \rightarrow \mathbf{c} &= \mathbf{\Psi}^{-1}(t_0)\mathbf{x}_o \\ \rightarrow \mathbf{x}(t) &= \mathbf{\Psi}(t)\mathbf{\Psi}^{-1}(t_0)\mathbf{x}_o \end{aligned} \quad (4)$$

Now we look at a special type of fundamental matrix. We do this by choosing a special initial data matrix \mathbf{B} in (2). Let's look for one that satisfies $\mathbf{\Psi}(t_0) = \mathbf{I}$, the $n \times n$ identity matrix. Then, writing this fundamental matrix as $\mathbf{\Phi}(t)$, since $\mathbf{\Phi}(t_0) = \mathbf{I} = \mathbf{\Phi}^{-1}(t_0)$, $\mathbf{\Phi}(t_0)^{-1}\mathbf{x}_o = \mathbf{x}_o$, we can write (referring to (4));

$$\mathbf{x}(t) = \mathbf{\Phi}(t)\mathbf{x}_o \quad (5)$$

How to find $\mathbf{\Phi}(t)$? Here's one way. Take any fundamental set $\mathbf{x}^{(1)}(t), \dots, \mathbf{x}^{(n)}(t)$ with initial values $\mathbf{x}^{(1)}(t_0), \dots, \mathbf{x}^{(n)}(t_0) = \mathbf{x}_o^{(1)}, \dots, \mathbf{x}_o^{(n)}$. From this fundamental set we construct another $\mathbf{y}^{(1)}(t), \dots, \mathbf{y}^{(n)}(t)$ where $\mathbf{y}^{(j)}(t_0) = \mathbf{e}_j$, $1 \leq j \leq n$ and $\mathbf{e}_j = (0, 0, \dots, 0, 1, 0, \dots, 0)$ has all zeros except for a 1 in the j^{th} position. To do this we need to solve these systems of equations;

$$\mathbf{y}^{(j)}(t_0) = c_{j1}\mathbf{x}_o^{(1)} + c_{j2}\mathbf{x}_o^{(2)} + \cdots + c_{jn}\mathbf{x}_o^{(n)} = \mathbf{e}_j, \quad 1 \leq j \leq n$$

for the unknowns c_{jk} . Then we form $\mathbf{\Phi}(t)$ with columns made up of the $\mathbf{y}^{(j)}(t)$.

The exponential of a matrix

In the special case of $\mathbf{x}' = \mathbf{A}\mathbf{x}$, where \mathbf{A} is a constant matrix (so the $\mathbf{P}(t)$ in (1) does not actually depend on time), another way to find $\Phi(t)$ is as follows. From (2) we see that any fundamental matrix $\Psi(t)$ of $\mathbf{x}' = \mathbf{A}\mathbf{x}$ satisfies

$$\Psi'(t) = \mathbf{A}\Psi(t) \quad (6)$$

And as before, we will call $\Phi(t)$ that particular fundamental matrix which satisfies $\Psi(t_o) = \mathbf{I}$. Recall the scalar case $x' = ax$, $x(t_o) = x_o$ whose solution is $x_o = e^{a(t-t_o)} = x_o \exp(a(t-t_o))$. This suggests we write the solution of (6) as $\Phi(t) = e^{\mathbf{A}(t-t_o)} = \exp(\mathbf{A}(t-t_o))$ (remember that $\Phi(t_o) = \mathbf{I}$). How should we define the exponential $\exp(\mathbf{B})$ of a matrix \mathbf{B} ? Recall the power series;

$$e^{at} = \exp(at) = 1 + at + \frac{(at)^2}{2!} + \frac{(at)^3}{3!} + \dots = \sum_{i=0}^{\infty} \frac{(at)^i}{i!}, \quad (7)$$

so we may be lead to define $\exp(\mathbf{A}(t-t_o))$ also by a power series;

$$e^{\mathbf{A}(t-t_o)} = \exp(\mathbf{A}(t-t_o)) = 1 + \mathbf{A}(t-t_o) + \frac{(\mathbf{A}(t-t_o))^2}{2!} + \frac{(\mathbf{A}(t-t_o))^3}{3!} + \dots = \sum_{i=0}^{\infty} \frac{(\mathbf{A}(t-t_o))^i}{i!} \quad (8)$$

It can be shown (along similar lines as done with (7)) that (8) converges for any $\mathbf{A}(t-t_o)$ (any matrix \mathbf{A} and any t, t_o), and so the expression $\exp(\mathbf{A}(t-t_o))$ makes sense as a matrix. Then, following the same lines as for (7), we can differentiate the series (8) and show that

$$\frac{d}{dt} \exp(\mathbf{A}(t-t_o)) = \mathbf{A} \exp(\mathbf{A}(t-t_o)) \quad (9)$$

Thus, if we define $\Phi(t)$ by $\Phi(t) = \exp(\mathbf{A}(t-t_o))$ (as given by formula (8)), then this is the solution of (6) with $\Phi(t_o) = \mathbf{I}$.

The next task is to actually compute the series (8). We will first look at a simple case when $\mathbf{A} = \mathbf{D}$ is a diagonal matrix with diagonal entries $(\lambda_1, \lambda_2, \dots, \lambda_n)$. Then it's easy to see that

$$\mathbf{D}^k = \begin{bmatrix} \lambda_1^k & & \\ & \ddots & \\ & & \lambda_n^k \end{bmatrix}$$

(all other entries 0). Therefore,

$$\Phi(t) = \exp(\mathbf{D}(t-t_o)) = \begin{bmatrix} e^{\lambda_1(t-t_o)} & & \\ & \ddots & \\ & & e^{\lambda_n(t-t_o)} \end{bmatrix} \quad (10)$$

(all other entries 0).

The next simplest case is when \mathbf{A} is *diagonalizable*. This means there is an invertible matrix \mathbf{T} such that

$$\mathbf{T}^{-1}\mathbf{A}\mathbf{T} = \mathbf{D} \quad (\iff \mathbf{A} = \mathbf{T}\mathbf{D}\mathbf{T}^{-1}) \quad (11)$$

where \mathbf{D} is a diagonal matrix. Recall that \mathbf{A} is diagonalizable if it has n linearly independent eigenvectors $\xi^{(1)}, \dots, \xi^{(n)}$. Let $\lambda_1, \dots, \lambda_n$ be the eigenvalues (counted with multiplicity, i.e., if λ occurs as a root of the characteristic equation 2 times, then it appears in the list of eigenvalues 2 times; so not necessarily are all the λ_i distinct); $\mathbf{A}\xi^{(i)} = \lambda_i\xi^{(i)}$.

Given an $\mathbf{x}(t)$ (which we think of as being the solution of $\mathbf{x}' = \mathbf{A} \mathbf{x}$, $\mathbf{x}(t_o) = \mathbf{x}_o$), let's define the vector $\mathbf{y}(t)$ by

$$\mathbf{x} = \mathbf{T} \mathbf{y} \iff \mathbf{T}^{-1} \mathbf{x} = \mathbf{y} \quad (12)$$

Then

$$\begin{aligned} \mathbf{x} = \mathbf{T} \mathbf{y} &\implies \mathbf{T} \mathbf{y}' = \mathbf{x}' \\ &\mathbf{T} \mathbf{y}' = \mathbf{A} \mathbf{x} \\ &= \mathbf{A} \mathbf{T} \mathbf{y} \\ &\implies \mathbf{y}' = \mathbf{T}^{-1} \mathbf{A} \mathbf{T} \mathbf{y} \\ &= \mathbf{D} \mathbf{y} \end{aligned} \quad (13)$$

So now we can write the solution of (13) as

$$\mathbf{y}(t) = \exp(\mathbf{D}(t - t_o)) \mathbf{y}_o, \quad \mathbf{y}_o = \mathbf{T}^{-1} \mathbf{x}_o \quad (14)$$

where $\exp(\mathbf{D}(t - t_o))$ is given by (10).

We now want to relate $\mathbf{y}(t)$ to the solution $\mathbf{x}(t)$. From (12) and (14),

$$\begin{aligned} \mathbf{x}(t) &= \mathbf{T} \mathbf{y}(t) \\ &= \mathbf{T} \exp(\mathbf{D}(t - t_o)) \mathbf{y}_o \\ &= \mathbf{T} \exp(\mathbf{D}(t - t_o)) \mathbf{T}^{-1} \mathbf{x}_o \\ &= \mathbf{\Omega}(t) \mathbf{x}_o, \end{aligned} \quad (15)$$

where we've written $\mathbf{\Omega}(t) = \mathbf{T} \exp(\mathbf{D}(t - t_o)) \mathbf{T}^{-1}$. Then since $\mathbf{x}'(t) = \mathbf{A} \mathbf{x}(t)$, comparing this with (15) implies that $\mathbf{\Omega}'(t) = \mathbf{A} \mathbf{\Omega}(t)$. Thus, $\mathbf{\Omega}(t)$ is a fundamental matrix of $\mathbf{x}' = \mathbf{A} \mathbf{x}$. Note that $\mathbf{\Omega}(t_o) = \mathbf{I}$. Thus,

$$\mathbf{\Phi}(t) = \mathbf{T} \exp(\mathbf{D}(t - t_o)) \mathbf{T}^{-1} = \mathbf{T} e^{\mathbf{D}(t - t_o)} \mathbf{T}^{-1} \quad (16)$$

is the special fundamental matrix of $\mathbf{x}' = \mathbf{A} \mathbf{x}$ with $\mathbf{\Phi}(t_o) = \mathbf{I}$.

Now let $\mathbf{Q}(t) = \exp(\mathbf{D}(t - t_o))$ and $\mathbf{\Psi}(t) = \mathbf{T} \mathbf{Q}(t)$. Then

$$\begin{aligned} \mathbf{\Psi}'(t) &= \mathbf{T} \mathbf{Q}'(t) \\ &= \mathbf{T} \mathbf{D} \mathbf{Q}(t) \\ &= \mathbf{T} \mathbf{D} (\mathbf{T}^{-1} \mathbf{T} \mathbf{Q}) \\ &= (\mathbf{T} \mathbf{D} \mathbf{T}^{-1}) \mathbf{T} \mathbf{Q}(t) \\ &= \mathbf{A} \mathbf{T} \mathbf{Q}(t) \\ &= \mathbf{A} \mathbf{\Psi}(t) \end{aligned}$$

and so $\mathbf{\Psi}(t)$ is a fundamental matrix of $\mathbf{x}' = \mathbf{A} \mathbf{x}$ (with $\mathbf{\Psi}(t_o) = \mathbf{T}$).

Note that $\mathbf{\Phi}(t) = \mathbf{\Psi}(t) \mathbf{\Psi}^{-1}(t_o)$;

$$\begin{aligned} \mathbf{\Psi}(t) \mathbf{\Psi}^{-1}(t_o) &= \mathbf{T} \mathbf{Q}(t) (\mathbf{T} \mathbf{Q}(t_o))^{-1} \\ &= \mathbf{T} \mathbf{Q}(t) \mathbf{Q}^{-1}(t_o) \mathbf{T}^{-1} \\ &= \mathbf{T} \mathbf{Q}(t) \mathbf{I} \mathbf{T}^{-1} \\ &= \mathbf{T} \mathbf{Q}(t) \mathbf{T}^{-1} \\ &= \mathbf{\Phi}(t) \end{aligned}$$

Repeated Eigenvalues

→ Read section 7.8 (and review section 7.3)

A is an $n \times n$ matrix. The characteristic equation $p_A(\lambda) = \det(\mathbf{A} - \lambda\mathbf{I})$ has roots λ_i which are eigenvalues of A . $p_A(\lambda)$ is a polynomial of order n and so has at most n real roots (but has exactly n complex roots). The associated eigenvectors $\xi^{(i)}$; $\mathbf{A}\xi^{(i)} = \lambda_i\xi^{(i)}$. If eigenvalue λ_i occurs m_i times as a root of $p_A(\lambda)$ we say that λ_i has algebraic multiplicity m_i . Each eigenvalue has at least one associated eigenvector, but it may happen that it has more than one. The geometric multiplicity q_i of the eigenvalue λ_i is the number of linearly independent eigenvectors associated with λ_i . Note that $1 \leq q_i \leq m_i$. We list the distinct eigenvalues as $\lambda_1, \dots, \lambda_k$, $k \leq n$, and their algebraic multiplicities m_1, \dots, m_k . Note that $m_1 + m_2 + \dots + m_k = n$.

We now switch to the notation r for the eigenvalues λ , to be consistent with the text.

We consider the system $\mathbf{x}' = \mathbf{A}\mathbf{x}$. For each (distinct) eigenvalue r with algebraic multiplicity m we need to construct m linearly independent solutions (then we will have a total of n linearly independent solutions and thus have a fundamental set of solutions).

Case I: Geometric multiplicity $q = m$. Let $\xi^{(1)}, \dots, \xi^{(m)}$ be m linearly independent eigenvectors for r . Then we have m linearly independent solutions

$$e^{rt}\xi^{(1)}, e^{rt}\xi^{(2)}, \dots, e^{rt}\xi^{(m)}$$

Case II: $q < m$. Here there are not enough linearly independent eigenvectors to construct independent solutions of the form $e^{rt}\xi$, so we have to look at a way to find more solutions for the eigenvalue r . We will only consider the case when $q = m - 1$, in particular, when $q = 2, m = 1$. Then r occurs twice as a root of $p_A(\lambda)$ but there is only one linearly independent eigenvector ξ . So we have one solution $\mathbf{x}(t) = e^{rt}\xi$. To find another we look for one of the form

$$\mathbf{x}(t) = \xi te^{rt} + \eta e^{rt} \quad (17)$$

for some unknown vector η . Substituting this into the equation $\mathbf{x}' = \mathbf{A}\mathbf{x}$ and comparing coefficients of te^{rt} and e^{rt} , we find that

$$\mathbf{A}\xi = r\xi, \quad \text{and} \quad (\mathbf{A} - r\mathbf{I})\eta = \xi \quad (18)$$

The first equation in (18) is just the eigenvector equation for r (which we know is true), so we need to show that the second equation has a solution, i.e., that there is an η which satisfies that equation. Now, we know that the matrix $\mathbf{A} - r\mathbf{I}$ is not invertible, so we can't solve the equation by $\eta = (\mathbf{A} - r\mathbf{I})^{-1}\xi$. However, the system $(\mathbf{A} - r\mathbf{I})\eta = \xi$ can always be solved for η (say, by Gaussian elimination, row reduction). Thus, we are able to find a second linearly independent solution. The vector η is called a *generalized eigenvector* of \mathbf{A} associated to the eigenvalue r .

It may turn out that when you solve (18) for η , you will get a sum of two or more vectors, one of which may be a multiple of ξ . Since ξ already occurs in the general solution, you can discard this part of η .

See the examples in the text.