General linear ordinary differential equation

A linear ordinary differential equation has the form $(d\vec{v}/dt) = A \cdot \vec{v}$ where A is an $n \times n$ matrix and \vec{v} is an column *n*-vector, where A is *independent* of \vec{v} . So far we have studied the case where A does not vary with t. We now want to study the more general case where A = A(t) may vary with t.

Before we begin, here is a clarificatory remark. In the introduction to this course, we had pointed out that a "time-dependent" equation $(d\vec{v}/dt) = \vec{f}(t, \vec{v})$ can be converted into a time-independent one by introducing \vec{w} as a column vector with (n + 1) entries; the first n entries are the same as \vec{v} and the last entry is t. On the right hand side we can then substitute w_{n+1} in place of t to write $(d\vec{w}/dt) = \vec{g}(\vec{w})$ where the first n entries of \vec{g} are $f(w_{n+1}, \vec{w})$ and the last entry is $1 (= (dw_{n+1}/dt))$. If we attempt to use this in the above case, we see that the new matrix A becomes dependent on w_{n+1} and hence the equation is no longer linear.

To summarise, it is *important* that A remains *independent* of \vec{v} for us to call it a *linear* equation. However, we *can* allow A to depend on t as we will now do.

Linearity of the solution

As a consequence of the general statement of existence and uniqueness of solutions of ordinary differential equation, we will see that under some *very mild* hypothesis on A (such as continuity), there is a unique function $\vec{v}(t)$ which satisfies

$$\frac{d\vec{v}}{dt} = A(t) \cdot \vec{v} \text{ and } \vec{v}(0) = \vec{v_0}$$

for any given *initial* vector $\vec{v_0}$. This solution exists for small enough interval around 0 (for t) which is *independent* of the chosen $\vec{v_0}$.

In fact, if we start at a different initial time t_0 and put the condition $\vec{v}(t_0) = \vec{v_0}$ instead of the above initial condition, then there is a unique solution $\vec{v}(t)$ for a small enough interval around t_0 .

Let $\vec{u_0} = a\vec{v_0} + \vec{w_0}$ for some *constant* a and some chosen fixed vectors $\vec{v_0}$ and $\vec{w_0}$. We use the notation \vec{u} to denote the solution of

$$\frac{d\vec{u}}{dt} = A(t) \cdot \vec{u} \text{ and } \vec{u}(0) = \vec{u_0}$$

and we use the notation \vec{w} to denote the solution of

$$\frac{d\vec{w}}{dt} = A(t) \cdot \vec{w} \text{ and } \vec{w}(0) = \vec{w_0}$$

Consider the vector valued function $\vec{u'} = a\vec{v} + \vec{w}$. We note that

$$\vec{u'}(0) = a\vec{v}(0) + \vec{w}(0) = a\vec{v_0} + \vec{w_0} = \vec{u_0}$$

Moreover, since a is a constant

$$\frac{d\vec{u'}}{dt} = a\frac{d\vec{v}}{dt} + \frac{d\vec{w}}{dt} = aA(t)\cdot\vec{v} + A(t)\cdot\vec{w}$$

Now, matrix multiplication *commutes* with scalar multiplication (by a) and *preserves addition*. So

$$aA(t)\cdot\vec{v} + A(t)\cdot\vec{w} = A(t)\cdot a\vec{v} + A(t)\cdot\vec{w} = A(t)\cdot(a\vec{v} + \cdot\vec{w}) = A\vec{u'}$$

In other words, we see that $\vec{u'}$ also satisfies

$$\frac{d\vec{u'}}{dt} = A(t) \cdot \vec{u'} \text{ and } \vec{u'}(0) = \vec{u_0}$$

By the uniqueness of the solution, we see that $\vec{u'} = \vec{u}$.

We have shown the following. If \vec{v} is the solution of the Linear ODE with initial value $\vec{v_0}$ and \vec{w} is the solution of the same Linear ODE with (the other) initial value $\vec{w_0}$, then $a\vec{v} + \vec{w}$ is the solution of the same Linear ODE with the combined initial value $a\vec{v_0} + \vec{w_0}$.

Before proceeding further, let us see how linearity is important. Consider a general ODE $(d\vec{v}/dt) = \vec{f}(t, \vec{v})$ and let \vec{v} be a solution of this with initial value $\vec{v_0}$ and \vec{w} be a solution of this with initial value $\vec{w_0}$. We can form $\vec{u'} = a\vec{v} + \vec{w}$ as we did above. What does it satisfy? It is clear that, as above, we have

$$\vec{u'(0)} = a\vec{v}(0) + \vec{w}(0) = a\vec{v_0} + \vec{w_0}$$

However, when we examine the derivative we see

$$\frac{d\vec{u'}}{dt} = a\frac{d\vec{v}}{dt} + \frac{d\vec{w}}{dt} = a\vec{f}(t, \cdot \vec{v}) + \vec{f}(t, \cdot \vec{w})$$

The expression on the right-hand side *does not*, in general, simplify to $f(t, a\vec{v} + \vec{w})$ since we do not know that $\vec{f}(t, \vec{v})$ is linear with respect to the second (vectorial) argument \vec{v} . Thus, we appear to have a solution of an entirely different ordinary differential equation (if at all!).

Linear independence

Given that $\vec{v_i}(t)$, for $i = 1, \ldots, k$ are solutions of the linear ordinary differential equation $(d\vec{v}/dt) = A \cdot \vec{v}$. Further, let us assume that $\vec{v_1}(0), \vec{v_2}(0), \ldots, \vec{v_k}(0)$ are *linearly independent*. We now claim that, for every t, the vectors $\vec{v_1}(t), \vec{v_2}(t), \ldots, \vec{v_k}(t)$ are linearly independent. Suppose that a_1, a_2, \ldots, a_k are constants so that

$$a_1 \vec{v_1}(t_0) + a_2 \vec{v_2}(t_0) + \dots + a_k \vec{v_k}(t_0) = 0$$
 for some t_0

By uniqueness of the solution with initial time t_0 , it follows that

$$a_1 \vec{v_1}(t) + a_2 \vec{v_2}(t) + \dots + a_k \vec{v_k}(t) = 0$$
 for all t

In particular, it follows that

$$a_1 \vec{v_1}(0) + a_2 \vec{v_2}(0) + \dots + a_k \vec{v_k}(0) = 0$$

However, these vectors are given to be linearly independent and so we conclude that $a_i = 0$ for i = 1, ..., k as required.

This has a practical consequence. For each *i* from 1 to *n*, we let $\vec{v_i}$ denote the solution of the Linear ODE as above with the initial condition $\vec{v_i}(0) = \vec{e_i}$ where

$$\vec{e_i} = \begin{pmatrix} 0\\ \vdots\\ 1\\ \vdots\\ 0 \end{pmatrix}$$

is the standard basis vector of \mathbb{R}^n with 1 in the *i*-th row and 0 everywhere else. We now define the matrix $G = [v_1, \ldots, v_n]$ whose columns are $\vec{v_i}$; this is a matrix valued function of t. Then, the solution of the initial value problem with initial condition $\vec{v}(0) = \vec{v_0}$ is given by $\vec{v} = G \cdot \vec{v_0}$.

In other words, G plays the same role as $\exp(tA)$ did in the case when A has constant coefficients!

Another way to look at this is to say that we only need to find n solutions to the Linear ODE, corresponding to the initial positions $\vec{e_i}$. All other solutions are obtained as linear combinations of these solutions.

Determinant

Given an $n \times n$ matrix G, recall that its determinant is defined inductively as

$$\det(G) = G_{1,1} \det H(1,1) - G_{1,2} \det H(1,2) + \dots + (-1)^n G_{1,n} \det H(1,n)$$

where $G_{i,j}$ is the (i, j)-th entry of G and H(i, j) is the $(n-1) \times (n-1)$ matrix obtained from G by deleting the *i*-th row and *j*-th column. By an application of Leibniz rule and induction we can show that

$$\frac{d \det(G)}{dt} = \det G^{(1)} + \det G^{(2)} + \dots + \det G^{(n)}$$

where $G^{(k)}$ is the matrix obtained from G by differentiating the k-th column of G.

Now, let us assume that G is the matrix solution to a linear ODE as constructed above. Then each column $\vec{v_i}$ of G satisfies $(d\vec{v_i}/dt) = A \cdot \vec{v_i}$. So the right-hand side in the above expression becomes

$$\det[A \cdot \vec{v_1}, \vec{v_2}, \dots, \vec{v_n}] + \det[\vec{v_1}, A \cdot \vec{v_2}, \dots, \vec{v_n}] + \det[\vec{v_1}, \vec{v_2}, \dots, A \cdot \vec{v_n}]$$

An elementary but instructive exercise can be used to show that this expression is Tr(A) det(G) where Tr(A) denotes the trace of the matrix A. In other words, we have obtained the identity

$$\frac{d\det(G)}{dt} = \operatorname{Tr}(A)\det(G)$$

So, det(G) satisfies *this* ordinary differential equation. As we saw in the introduction, the solution of this differential equation is

$$\det(G) = \exp\left(\int_0^t \operatorname{Tr}(A)(t)dt\right)$$

where we have used det(G)(0) = 1 since we started with the standard basis at t = 0.

We note, in particular, that if the trace of A is *identically* 0, then det(G) is identically 1; in other words, the flow is "volume preserving".

There are other examples of polynomial identities that can be deduced for G based on certain linear properties of A. This is the starting point of the theory of Lie Algebras. We only mention one important property. If A is a skew-symmetric matrix, then G is orthogonal, i.e. $G \cdot G^t = \mathbf{1}_n$. In other words, if A is skew-symmetric then the flow preserves lengths and angles.

Wronskian

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A particular case of the above analysis is when we solve a linear equation of order n with one indeterminate:

$$\frac{d^n x}{dt^n} + a_1(t)\frac{d^{n-1}x}{dt^{n-1}} + \dots + a_{n-1}(t)\frac{dx}{dt} + a_n x = 0$$

As mentioned in the introduction, we can introduce new variables $x_1 = x$, $x_2 = (dx/dt)$, and so on till $x_n = (d^{n-1}x/dt^{n-1})$. This equation becomes

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -a_n & -a_{n-1} & -a_{n-3} & \dots & -a_1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{pmatrix}$$

It is clear that the trace of the matrix is $-a_1$ in this case. Suppose that $x_i(t)$ denotes the solution with initial condition

$$\frac{d^k x_i}{dt^k} \mid_{t=0} = \begin{cases} 0 & k \neq i \\ 1 & k = i \end{cases}$$

for $k = 0, \ldots, (N - 1)$. Then the Wronskian is defined as

$$W(t) = \det \begin{pmatrix} x_1 & x_2 & \dots & x_n \\ \frac{dx_1}{dt} & \frac{dx_2}{dt} & \dots & \frac{dx_n}{dt} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{d^{n-1}x_1}{dt^{n-1}} & \frac{d^{n-1}x_2}{dt^{n-1}} & \dots & \frac{d^{n-1}x_n}{dt^{n-1}} \end{pmatrix}$$

This is non-zero for all t and is given by $W(t) = \exp\left(\int -a_1(t)dt\right)$. In fact, one can show that for *any* collection of functions x_1, x_1, \ldots, x_n which are linearly independent, the Wronskian as defined above is non-zero.