

## Another perspective on the matrix exponential

**Review.** We achieved the milestone to introduce a **matrix exponential** in such a way that we can treat a system of DEs, say  $\mathbf{y}' = M\mathbf{y}$  with  $\mathbf{y}(0) = \mathbf{c}$ , just as if the matrix  $M$  was a number: namely, the unique solution is simply  $\mathbf{y} = e^{Mx}\mathbf{c}$ .

The price to pay is that the matrix  $e^{Mx}$  requires some work to actually compute (and proceeds by first determining a different matrix solution  $\Phi(x)$  using eigenvectors and eigenvalues). We offer below another way to think about  $e^{Mx}$  (using Taylor series).

**(exponential function)**  $e^x$  is the unique solution to  $\mathbf{y}' = \mathbf{y}$ ,  $\mathbf{y}(0) = 1$ .

From here, it follows that  $e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots$

The latter is the Taylor series for  $e^x$  at  $x = 0$  that we have seen in Calculus II.

**Important note.** We can actually construct this infinite sum directly from  $\mathbf{y}' = \mathbf{y}$  and  $\mathbf{y}(0) = 1$ .

Indeed, observe how each term, when differentiated, produces the term before it. For instance,  $\frac{d}{dx} \frac{x^3}{3!} = \frac{x^2}{2!}$ .

**Review.** We defined the **matrix exponential**  $e^{Mx}$  as the unique matrix solution to the IVP

$$\mathbf{y}' = M\mathbf{y}, \quad \mathbf{y}(0) = I.$$

We next observe that we can also make sense of the matrix exponential  $e^{Mx}$  as a power series.

**Theorem 70.** Let  $M$  be  $n \times n$ . Then the **matrix exponential** satisfies

$$e^M = I + M + \frac{1}{2!}M^2 + \frac{1}{3!}M^3 + \dots$$

**Proof.** Define  $\Phi(x) = I + Mx + \frac{1}{2!}M^2x^2 + \frac{1}{3!}M^3x^3 + \dots$

$$\begin{aligned} \Phi'(x) &= \frac{d}{dx} \left[ I + Mx + \frac{1}{2!}M^2x^2 + \frac{1}{3!}M^3x^3 + \dots \right] \\ &= 0 + M + M^2x + \frac{1}{2!}M^3x^2 + \dots = M\Phi(x). \end{aligned}$$

Clearly,  $\Phi(0) = I$ . Therefore,  $\Phi(x)$  is the fundamental matrix solution to  $\mathbf{y}' = M\mathbf{y}$ ,  $\mathbf{y}(0) = I$ .

But that's precisely how we defined  $e^{Mx}$  earlier. It follows that  $\Phi(x) = e^{Mx}$ . Now set  $x = 1$ . □

**Example 71.** If  $A = \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$ , then  $A^{100} = \begin{bmatrix} 2^{100} & 0 \\ 0 & 5^{100} \end{bmatrix}$ .

**Example 72.** If  $A = \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$ , then  $e^A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix} + \frac{1}{2!} \begin{bmatrix} 2^2 & 0 \\ 0 & 5^2 \end{bmatrix} + \dots = \begin{bmatrix} e^2 & 0 \\ 0 & e^5 \end{bmatrix}$ .

Clearly, this works to obtain  $e^D$  for every diagonal matrix  $D$ .

In particular, for  $Ax = \begin{bmatrix} 2x & 0 \\ 0 & 5x \end{bmatrix}$ ,  $e^{Ax} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 2x & 0 \\ 0 & 5x \end{bmatrix} + \frac{1}{2!} \begin{bmatrix} (2x)^2 & 0 \\ 0 & (5x)^2 \end{bmatrix} + \dots = \begin{bmatrix} e^{2x} & 0 \\ 0 & e^{5x} \end{bmatrix}$ .

The following is a preview of how the matrix exponential deals with repeated characteristic roots.

**Example 73.** Determine  $e^{Ax}$  for  $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ .

**Solution.** If we compute eigenvalues, we find that we get  $\lambda = 0, 0$  (multiplicity 2) but there is only one 0-eigenvector (up to multiples). This means we are stuck with this approach—however, see next extra section.

The key here is to observe that  $A^2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ . It follows that  $e^{Ax} = I + Ax = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 0 & x \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & x \\ 0 & 1 \end{bmatrix}$ .

## Extra: The case of repeated eigenvalues with too few eigenvectors

**Review.** To construct a fundamental matrix solution  $\Phi(x)$  to  $\mathbf{y}' = M\mathbf{y}$ , we compute eigenvectors:

Given a  $\lambda$ -eigenvector  $\mathbf{v}$ , we have the corresponding solution  $\mathbf{y}(x) = \mathbf{v}e^{\lambda x}$ .

If there are enough eigenvectors, we can collect these as columns to obtain  $\Phi(x)$ .

The next example illustrates how to proceed if there are not enough eigenvectors.

In that case, instead of looking only for solutions of the type  $\mathbf{y}(x) = \mathbf{v}e^{\lambda x}$ , we also need to look for solutions of the type  $\mathbf{y}(x) = (\mathbf{v}x + \mathbf{w})e^{\lambda x}$ . This can only happen if an eigenvalue is a repeated root of the characteristic polynomial.

**Example 74.** Let  $M = \begin{bmatrix} 8 & 4 \\ -1 & 4 \end{bmatrix}$ .

- Determine the general solution to  $\mathbf{y}' = M\mathbf{y}$ .
- Determine a fundamental matrix solution to  $\mathbf{y}' = M\mathbf{y}$ .
- Compute  $e^{Mx}$ .
- Solve the initial value problem  $\mathbf{y}' = M\mathbf{y}$  with  $\mathbf{y}(0) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ .

**Solution.**

- We determine the eigenvectors of  $M$ . The characteristic polynomial is:

$$\det(M - \lambda I) = \det\left(\begin{bmatrix} 8 - \lambda & 4 \\ -1 & 4 - \lambda \end{bmatrix}\right) = (8 - \lambda)(4 - \lambda) + 4 = \lambda^2 - 12\lambda + 36 = (\lambda - 6)(\lambda - 6)$$

Hence, the eigenvalues are  $\lambda = 6, 6$  (meaning that 6 has multiplicity 2).

- To find eigenvectors  $\mathbf{v}$  for  $\lambda = 6$ , we need to solve  $\begin{bmatrix} 2 & 4 \\ -1 & -2 \end{bmatrix}\mathbf{v} = \mathbf{0}$ .  
Hence,  $\mathbf{v} = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$  is an eigenvector for  $\lambda = 6$ . There is no independent second eigenvector.

- We therefore search for a solution of the form  $\mathbf{y}(x) = (\mathbf{v}x + \mathbf{w})e^{\lambda x}$  with  $\lambda = 6$ .

$$\mathbf{y}'(x) = (\lambda\mathbf{v}x + \lambda\mathbf{w} + \mathbf{v})e^{\lambda x} = M\mathbf{y} = (M\mathbf{v}x + M\mathbf{w})e^{\lambda x}$$

Equating coefficients of  $x$ , we need  $\lambda\mathbf{v} = M\mathbf{v}$  and  $\lambda\mathbf{w} + \mathbf{v} = M\mathbf{w}$ .

Hence,  $\mathbf{v}$  must be an eigenvector (which we already computed); we choose  $\mathbf{v} = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$ .

[Note that any multiple of  $\mathbf{y}(x)$  will be another solution, so it doesn't matter which multiple of  $\begin{bmatrix} -2 \\ 1 \end{bmatrix}$  we choose.]

$\lambda\mathbf{w} + \mathbf{v} = M\mathbf{w}$  or  $(M - \lambda)\mathbf{w} = \mathbf{v}$  then becomes  $\begin{bmatrix} 2 & 4 \\ -1 & -2 \end{bmatrix}\mathbf{w} = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$ .

One solution is  $\mathbf{w} = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$ . [We only need one.]

Hence, the general solution is  $C_1 \begin{bmatrix} -2 \\ 1 \end{bmatrix} e^{6x} + C_2 \left( \begin{bmatrix} -2 \\ 1 \end{bmatrix} x + \begin{bmatrix} -1 \\ 0 \end{bmatrix} \right) e^{6x}$ .

- The corresponding fundamental matrix solution is  $\Phi = \begin{bmatrix} -2e^{6x} & -(2x+1)e^{6x} \\ e^{6x} & xe^{6x} \end{bmatrix}$ .

- Note that  $\Phi(0) = \begin{bmatrix} -2 & -1 \\ 1 & 0 \end{bmatrix}$ , so that  $\Phi(0)^{-1} = \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix}$ . It follows that

$$e^{Mx} = \Phi(x)\Phi(0)^{-1} = \begin{bmatrix} -2e^{6x} & -(2x+1)e^{6x} \\ e^{6x} & xe^{6x} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix} = \begin{bmatrix} (2x+1)e^{6x} & 4xe^{6x} \\ -xe^{6x} & -(2x-1)e^{6x} \end{bmatrix}.$$

- The solution to the IVP is  $\mathbf{y}(x) = e^{Mx} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} (2x+1)e^{6x} & 4xe^{6x} \\ -xe^{6x} & -(2x-1)e^{6x} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} (2x+1)e^{6x} \\ -xe^{6x} \end{bmatrix}$ .

## Phase portraits and phase plane analysis

Our goal is to visualize the solutions to systems of equations. This works particularly well in the case of systems of two differential equations. A system that can be written as

$$\begin{aligned}\frac{dx}{dt} &= f(x, y) \\ \frac{dy}{dt} &= g(x, y)\end{aligned}$$

is called **autonomous** because it doesn't depend on the independent variable  $t$ .

**Comment.** Can you show that if  $x(t)$  and  $y(t)$  are a pair of solutions, then so is the pair  $x(t + t_0)$  and  $y(t + t_0)$ ?

We can visualize solutions to such a system by plotting the points  $(x(t), y(t))$  for increasing values of  $t$  so that we get a curve (and we can attach an arrow to indicate the direction we're flowing along that curve). Each such curve is called the **trajectory** of a solution.

Even better, we can do such a **phase portrait** without solving to get a formula for  $(x(t), y(t))$ ! That's because we can combine the two equations to get  $\frac{dy}{dx} = \frac{g(x, y)}{f(x, y)}$ , which allows us to make a slope field! If a trajectory passes through a point  $(x, y)$ , then we know that the slope at that point must be  $\frac{dy}{dx} = \frac{g(x, y)}{f(x, y)}$ .

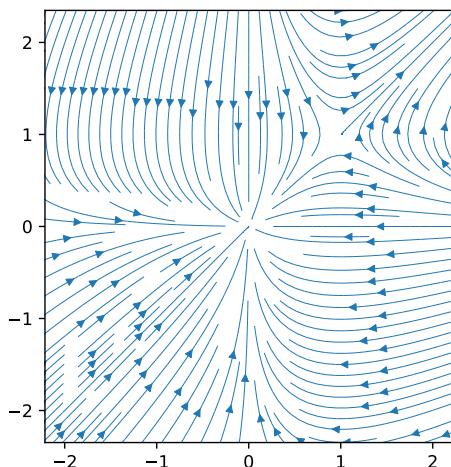
This allows us to sketch trajectories. However, it does not tell us everything about the corresponding solution  $(x(t), y(t))$  because we don't know at which times  $t$  the solution passes through the points on the curve.

However, we can visualize the speed with which a solution passes through the trajectory by attaching to a point  $(x, y)$  not only the slope  $\frac{g(x, y)}{f(x, y)}$  but the vector  $\begin{bmatrix} f(x, y) \\ g(x, y) \end{bmatrix}$ . That vector has the same direction as the slope but it also tells us in which direction we are moving and how fast (by its magnitude).

**Example 75.** Sketch some trajectories for the system  $\frac{dx}{dt} = x \cdot (y - 1)$ ,  $\frac{dy}{dt} = y \cdot (x - 1)$ .

**Solution.** Let's look at the point  $(x, y) = (2, -1)$ , for instance. Then the DEs tell us that  $\frac{dx}{dt} = x \cdot (y - 1) = -4$  and  $\frac{dy}{dt} = y \cdot (x - 1) = -1$ . We therefore attach the vector  $\left( \frac{dx}{dt}, \frac{dy}{dt} \right) = (-4, -1)$  to  $(x, y) = (2, -1)$ .

Note that if we use  $\frac{dy}{dx} = \frac{y \cdot (x - 1)}{x \cdot (y - 1)}$  directly, we find the slope  $\frac{dy}{dx} = \frac{-1}{-4} = \frac{1}{4}$ . This is slightly less information because it doesn't tell us that we are moving "left and down" as the arrows in the following plot indicate:



**Comment.** In this example, we can solve the slope-field equation  $\frac{dy}{dx} = \frac{y(x-1)}{x(y-1)}$  using separation of variables.

Do it! We end up with the implicit solutions  $y - \ln|y| = x - \ln|x| + C$ .

If we plot these curves for various values of  $C$ , we get trajectories in the plot above. However, note that none of this solving is needed for plotting by itself.

**Sage.** We can make Sage create such phase portraits for us!

```
>>> x,y = var('x y')
>>> streamline_plot((x*(y-1),y*(x-1)), (x,-3,3), (y,-3,3))
```