

Outline

- Matrix ODEs
- Classification of 2D linear systems with constant coefficients
- Inhomogeneous Systems

Last time:

Given $\dot{x}(t) = Ax(t)$, $x(0) = x_0$, $A \in M_{n \times n}(\mathbb{R})$, $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$,

$$x(t) = \exp(tA)x_0.$$

This time:

Suppose we can find a matrix P s.t. $A = P\Lambda P^{-1}$, for Λ diagonal and the columns of P eigenvectors of A .

Then $x(t) = \exp(tA)x_0 = \exp(tP\Lambda P^{-1})x_0 = P\exp(t\Lambda)P^{-1}x_0$.

Let $P = [v_1 \ v_2 \ \dots \ v_n]$, so v_1, \dots, v_n are eigenvectors,
and $\Lambda = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$, where $\lambda_1, \dots, \lambda_n$ are eigenvalues

Then $P\exp(t\Lambda) = [v_1 \ v_2 \ \dots \ v_n] \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} = [v_1 e^{\lambda_1 t} \ \dots \ v_n e^{\lambda_n t}]$

And let $P^{-1}x_0 = y_0 = \begin{bmatrix} y_{0,1} \\ \vdots \\ y_{0,n} \end{bmatrix}$

Then $x(t) = [v_1 e^{\lambda_1 t} \ \dots \ v_n e^{\lambda_n t}] \begin{bmatrix} y_{0,1} \\ \vdots \\ y_{0,n} \end{bmatrix} = y_{0,1} v_1 e^{\lambda_1 t} + \dots + y_{0,n} v_n e^{\lambda_n t}$

Thus, all solutions are linear combinations of the eigenvectors of A times $e^{\lambda_i t}$.

Ex. $\dot{x}(t) = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} x(t)$, $x(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

We want to find the eigenvalues + eigenvectors.

Trick from linear algebra

$$\left\{ \begin{array}{l} \text{Given a } 2 \times 2 \text{ matrix } A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \\ \text{Tr}(A) = a + d, \quad \det(A) = ad - bc. \\ \text{Then } \lambda^2 - \text{Tr}(A) \cdot \lambda + \det(A) = 0, \end{array} \right. \quad \left(\begin{array}{l} \left| \begin{array}{cc} \lambda - a & b \\ c & \lambda - d \end{array} \right| = \lambda^2 - (a+d)\lambda + ad - bc \\ = \lambda^2 - \text{Tr}(A) + \det(A) \end{array} \right)$$

$$\Rightarrow \begin{array}{l} \lambda^2 - \lambda - 2 = 0 \\ (\lambda - 2)(\lambda + 1) = 0 \\ \lambda = 2, -1. \end{array} \quad \begin{array}{l} \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2x_1 \\ 2x_2 \end{bmatrix} \Rightarrow \begin{array}{l} x_1 + 2x_2 = 2x_1 \\ x_1 = 2x_2 \end{array} \left. \vphantom{\begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix}} \right\} v_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \lambda_1 = 2 \\ \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -x_1 \\ -x_2 \end{bmatrix} \Rightarrow \begin{array}{l} x_1 + 2x_2 = -x_1 \\ x_1 = -x_2 \end{array} \left. \vphantom{\begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix}} \right\} v_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \lambda_2 = -1 \end{array}$$

$$\begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}^{-1} = \frac{1}{3} \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix}$$

Method 1:

$$\begin{aligned} \Rightarrow \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} &= \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix} \cdot \frac{1}{3} \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix} \\ x(t) &= \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} e^{2t} & 0 \\ 0 & e^{-t} \end{bmatrix} \cdot \frac{1}{3} \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} e^{2t} & 0 \\ 0 & e^{-t} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} e^{2t} \\ e^{-t} \end{bmatrix} \\ &= \begin{bmatrix} 2e^{2t} - e^{-t} \\ e^{2t} + e^{-t} \end{bmatrix} \end{aligned}$$

Method 2: Ansatz: $x(t) = c_1 \cdot \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{2t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t}$

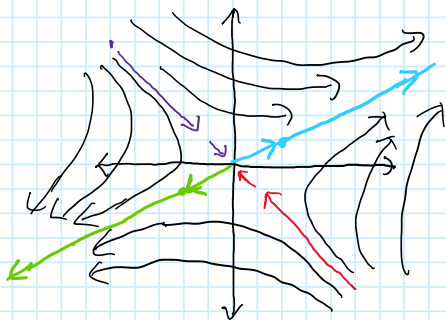
$$x(0) = \begin{bmatrix} 2c_1 - c_2 \\ c_1 + c_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \Rightarrow c_1 = 1, c_2 = 1$$

$$\Rightarrow x(t) = \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{2t} + \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t} = \begin{bmatrix} 2e^{2t} - e^{-t} \\ e^{2t} + e^{-t} \end{bmatrix}$$

Phase portrait: $A = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix}$, $\dot{x} = Ax$.

Then $\lambda_1 = 2$, $\lambda_2 = -1$

$$v_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \quad v_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \quad \text{So } x(t) = c_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{2t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t}$$



Similar to slope field.

$$\text{Suppose } x(0) = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \Rightarrow x(t) = \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{2t}$$

$$x(0) = \begin{bmatrix} -2 \\ -1 \end{bmatrix} \Rightarrow x(t) = - \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{2t}$$

$$x(0) = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \Rightarrow x(t) = \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t}$$

$$x(0) = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \Rightarrow x(t) = - \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t}$$

$$x(0) = \begin{bmatrix} -5 \\ 5 \end{bmatrix} \Rightarrow x(t) = 5 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t}$$

The qualitative behavior of the ODE given different starting points is captured by the phase portrait and completely determined by the eigenvalues & eigenvectors.

Cases:

2 real eigenvalues

↳ both positive

↳ both negative

↳ mix of positive & negative

1 real eigenvalue of multiplicity 2

↳ 2 linearly ind. eigenvectors

↳ 1 linearly ind. eigenvector

2 complex conjugate eigenvalues

↳ real part negative

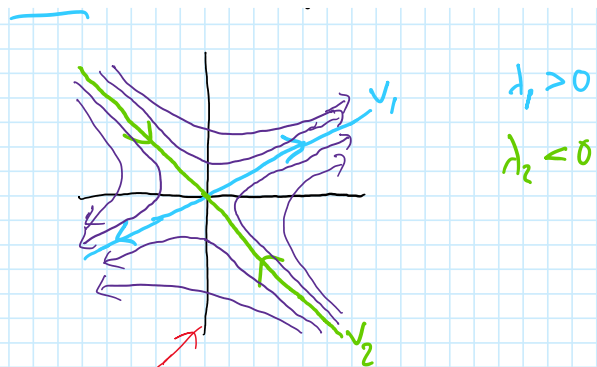
↳ real part = 0

↳ real part positive

Case: 2 real eigenvalues

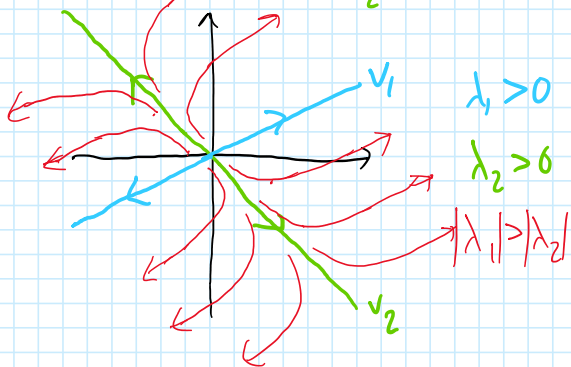
$\begin{bmatrix} \text{---} \\ \text{---} \\ \text{---} \end{bmatrix} \quad | \quad \begin{bmatrix} \text{---} \\ \text{---} \end{bmatrix} \quad \rightarrow \quad v_1 \quad \lambda_1 > 0$

$$x(t) = c_1 v_1 e^{\lambda_1 t} + c_2 v_2 e^{\lambda_2 t}$$

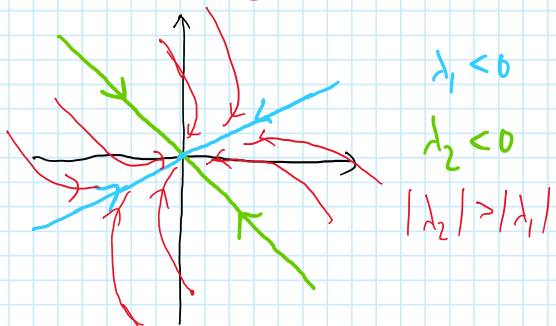


$$x(t) = c_1 v_1 \cdot e^{\lambda_1 t} + c_2 v_2 \cdot e^{\lambda_2 t}$$

Saddle point



Source. Everything goes away from the origin, but the component along v_1 gets magnified more quickly



Sink. Everything goes to the origin, but the component along v_2 decays more quickly

Case: 1 real eigenvalue

Suppose we have 2 linearly ind. eigenvectors v_1, v_2 .

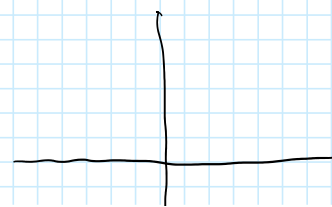
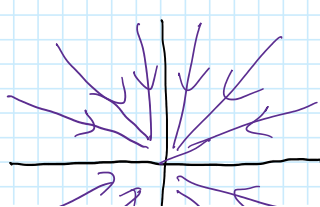
Then $A = P \cdot \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} P^{-1}$, where $P = [v_1 \ v_2]$.

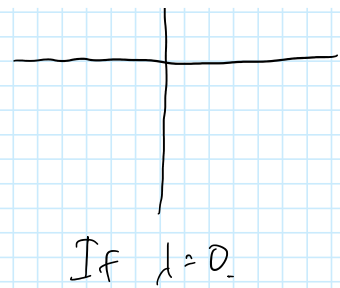
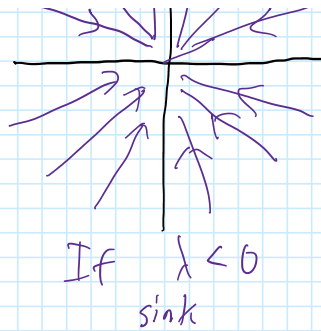
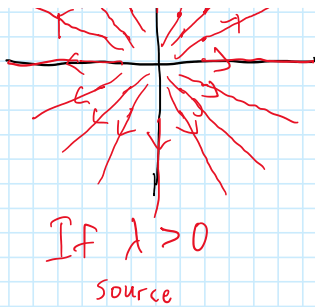
But diagonal matrices commute with all other matrices

So $A = P P^{-1} \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix}$.

Then for $\dot{x} = Ax$, we get $x = \exp(\lambda A) x_0 = \begin{bmatrix} e^{\lambda t} & 0 \\ 0 & e^{\lambda t} \end{bmatrix} x_0 = e^{\lambda t} x_0$

So $x(t) = c_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} e^{\lambda t} + c_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix} e^{\lambda t} = e^{\lambda t} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$





What if we only have 1 eigenvector?

e.g. $A = \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix}$ eigenvalue $\lambda = 3$

$$\left. \begin{aligned} \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3x_1 \\ 3x_2 \end{bmatrix} \\ \left. \begin{aligned} 3x_1 + x_2 = 3x_1 \\ x_2 = 3x_2 \end{aligned} \right\} x_2 = 0 \Rightarrow v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}. \end{aligned}$$

$$\dot{x} = Ax \quad x(0) = x_0$$

$$\Rightarrow x = \exp(tA)x_0$$

$$\exp(tA) = \exp\left(t\left(\begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} + \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}\right)\right)$$

Note $\begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 3 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 3 \\ 0 & 0 \end{bmatrix}$, so the two matrices commute.

$$\begin{aligned} \text{Thus, } \exp(tA) &= \exp\left(t\begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix}\right) \cdot \exp\left(t\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}\right) \\ &= \begin{bmatrix} e^{3t} & 0 \\ 0 & e^{3t} \end{bmatrix} \cdot \exp\left(\begin{bmatrix} 0 & t \\ 0 & 0 \end{bmatrix}\right). \end{aligned}$$

$$\text{Recall } \exp(X) = I + X + \frac{X^2}{2!} + \frac{X^3}{3!} + \frac{X^4}{4!} + \dots$$

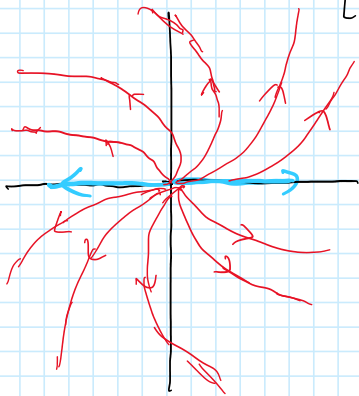
$$\text{Note } \begin{bmatrix} 0 & t \\ 0 & 0 \end{bmatrix}^2 = \begin{bmatrix} 0 & t \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & t \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}.$$

$$\begin{aligned} \text{So } \exp\left(\begin{bmatrix} 0 & t \\ 0 & 0 \end{bmatrix}\right) &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 0 & t \\ 0 & 0 \end{bmatrix} + \cancel{\frac{\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}}{2!}} + \cancel{\frac{\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}}{3!}} + \dots \\ &= \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix}. \end{aligned}$$

$$\Rightarrow \exp(tA) = \begin{bmatrix} e^{3t} & 0 \\ 0 & e^{3t} \end{bmatrix} \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} e^{3t} & te^{3t} \\ 0 & e^{3t} \end{bmatrix}$$

$$\Rightarrow \exp(tA) = \begin{bmatrix} e^{3t} & 0 \\ 0 & e^{3t} \end{bmatrix} \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} e^{3t} & te^{3t} \\ 0 & e^{3t} \end{bmatrix}$$

$$\text{Thus, } x(t) = c_1 \begin{bmatrix} e^{3t} \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} te^{3t} \\ e^{3t} \end{bmatrix} = c_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} e^{3t} + c_2 \begin{bmatrix} te^{3t} \\ e^{3t} \end{bmatrix}$$



Try using slope field

$$\dot{x} = \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix} x$$

$$\dot{x} = \begin{bmatrix} 3 \\ 0 \end{bmatrix} x_1 + \begin{bmatrix} 1 \\ 3 \end{bmatrix} x_2$$

If we start on the eigenvector, we stay on the eigenvector. Otherwise, the solution has weird behavior due to the extra polynomial t term.

Complicated, and I won't ask you to draw this case.

Case: Two complex conjugate eigenvalues
Then we don't have real eigenvectors, but we do have complex ones.

Let $\lambda = \alpha \pm \beta i$ be the two complex eigenvalues of A , ($\lambda_1^* = \lambda_2$)
and let v_1, v_2 be the corresponding complex eigenvectors.

Because v_1, v_2 are eigenvectors to conjugate eigenvalues, they are also conjugate WLOG
i.e. $v_1^* = v_2$.

$$\text{Let } P = [v_1, v_2].$$

$$\text{Then } A = P \Lambda P^{-1}, \text{ where } \Lambda = \begin{bmatrix} \alpha + \beta i & 0 \\ 0 & \alpha - \beta i \end{bmatrix}.$$

$$\text{Let } \dot{x} = Ax$$

$$\text{Then } x = \exp(tA) x_0, \text{ where } x_0 = x(0).$$

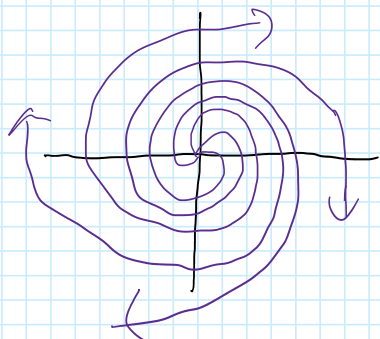
$$x(t) = P \exp(t\Lambda) P^{-1} x_0. \text{ Say } y_0 = P^{-1} x_0, \text{ the coordinates in this other basis}$$

$$\text{Then } x(t) = [v_1, v_2] \begin{bmatrix} e^{(\alpha + \beta i)t} & 0 \\ 0 & e^{(\alpha - \beta i)t} \end{bmatrix} \begin{bmatrix} y_{0,1} \\ y_{0,2} \end{bmatrix}$$

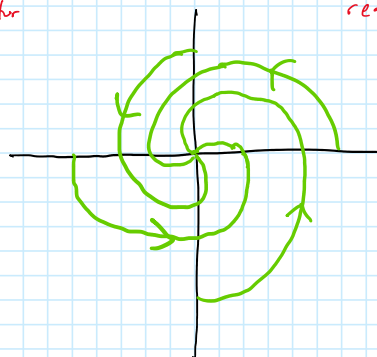
$$x(t) = v_1 e^{(\alpha + \beta i)t} \cdot y_{0,1} + v_2 e^{(\alpha - \beta i)t} y_{0,2}$$

In order for the final solution to be real for all t , $y_{0,1}^* = y_{0,2}$.

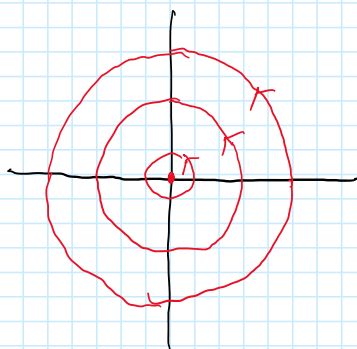
$$\begin{aligned} \text{Thus } x(t) &= 2 \operatorname{Re} \left(v_1 e^{(\alpha + \beta i)t} y_{0,1} \right) \\ &= 2 \operatorname{Re} \left(v_1 y_{0,1} e^{\alpha t} \cos(\beta t) + v_1 y_{0,1} e^{\alpha t} i \sin(\beta t) \right) \\ &= 2 \cos(\beta t) e^{\alpha t} \underbrace{\operatorname{Re}(v_1 y_{0,1})}_{\text{real vector}} - 2 \sin(\beta t) e^{\alpha t} \underbrace{\operatorname{Im}(v_1 y_{0,1})}_{\text{real vector}} \end{aligned}$$



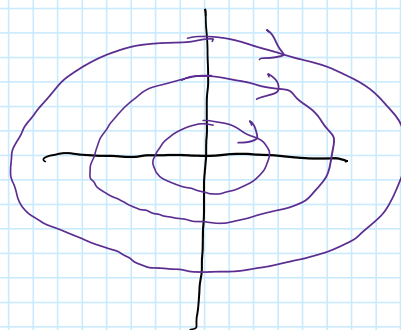
If $\alpha > 0$
spiral source



If $\alpha < 0$
spiral sink



If $\alpha = 0$
center



Terminology: A linear system is **stable** if all solutions are bounded, and **asymptotically stable** if all solutions converge to 0 as $t \rightarrow \infty$.

Inhomogeneous systems:

Given $\dot{x}(t) = Ax(t) + g(t)$, $x(0) = x_0$,

$$x(t) = \exp(tA)x_0 + \int_0^t \exp((t-s)A)g(s)ds. \quad (\text{Reschl 3.48})$$

Recall given $x^{(n)} + c_{n-1}x^{(n-1)} + \dots + c_1\dot{x} + c_0x = g(t)$,

$$x(t) = x_h(t) + \int_0^t u(t-s)g(s)ds$$

HW, Teschl 3.57

So you can explicitly solve the inhomogeneous eqn. using this formula.

So you can explicitly solve the inhomogeneous eqn. using this formula.
 But remember, there are other ways to solve linear ODEs with constant coefficients, such as **undetermined coefficients**, **variation of parameter**, **series methods**, etc. These all still work here.

Ex.

$$\dot{x}(t) = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} x(t) + \begin{bmatrix} 1 \\ t \end{bmatrix}$$

From above, $\lambda_1 = 2$ $v_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$

$\lambda_2 = -1$ $v_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

So $x_c = c_1 \cdot \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{2t} + c_2 \cdot \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t}$

Guess $x_p = \begin{bmatrix} k_1 t + k_2 \\ k_3 t + k_4 \end{bmatrix}$ (method of undetermined coefficients)

Then $\dot{x}_p = \begin{bmatrix} k_1 \\ k_3 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} k_1 t + k_2 \\ k_3 t + k_4 \end{bmatrix} + \begin{bmatrix} 1 \\ t \end{bmatrix}$

$$\begin{bmatrix} k_1 \\ k_3 \end{bmatrix} = \begin{bmatrix} k_1 t + k_2 + 2k_3 t + 2k_4 + 1 \\ k_1 t + k_2 + t \end{bmatrix}$$

$$\begin{cases} k_1 = t(k_1 + 2k_3) + (k_2 + 2k_4 + 1) \\ k_3 = t(k_1 + 1) + k_2 \end{cases}$$

$$\begin{cases} k_1 + 2k_3 = 0 \\ k_2 + 2k_4 + 1 = k_1 \\ k_2 = k_3 \\ k_1 + 1 = 0 \end{cases} \Rightarrow \begin{cases} k_1 = -1 \\ k_2 = \frac{1}{2} \\ k_3 = \frac{1}{2} \\ k_4 = -\frac{5}{4} \end{cases} \Rightarrow x_p = \begin{bmatrix} -t + \frac{1}{2} \\ \frac{1}{2}t - \frac{5}{4} \end{bmatrix}$$

$$\Rightarrow x = x_c + x_p = c_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} e^{2t} + c_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} e^{-t} + \begin{bmatrix} -t + \frac{1}{2} \\ \frac{1}{2}t - \frac{5}{4} \end{bmatrix}$$