# UNSW Mathematics Society Presents MATH2221/2121 Seminar Part 1



Presented by Steve Jang, Joanna Lin

# Overview I

Part I: Linear ODEs
 Linear Differential Operators
 Wronskian and Linear Independence
 Methods
 Power Series
 Bessel Equation
 Lagrange Equation (2221 only)

2. Dynamical Systems
Lipschitz (2221 only)
Linear Dynamical Systems
Matrices
Equilibrium Points
First Integrals (2221 only)

# 1. Part I: Linear ODEs

# Differential Operators

#### Definition 1: Linear differential operators

Define the linear differential operator L of order m to be

$$Lu(x) = \sum_{j=0}^{m} a_j(x) \cdot D^j u(x)$$
  
=  $a_m(x) D^m u(x) + a_{m-1}(x) D^{m-1} u(x) + \dots + a_0(x) u(x),$ 

where  $D^j u = \frac{d^j u}{du^j}$  and  $D^0 u = u$ .

### Definition 2: Singular ODEs

An ODE is said to be singular with respect to [a, b] if the leading coefficient vanishes for any  $x \in [a, b]$ . E.g. xu' - u = 0 is singular on [0, 1].

# Homogeneous and inhomogeneous ODEs

#### Definition 3: Homogeneous ODEs

An ODE is said to be homogeneous if the right hand side is 0. That is, we have a differential equation of the form

$$Lu = 0.$$

• Example: u'' + u' + u = 0.

### Definition 4: Inhomogeneous ODEs

An ODE is said to be inhomogeneous if the right hand side is not 0. Then we can write the differential equation as

$$Lu = f$$
.

• Example:  $u'' + u' + u = \cos(x)$ .

# Initial-value problems (IVP)

#### Definition 5: Initial-value problems

Consider an m-th order differential equation

$$Lu = f$$
, on  $[a, b]$  (1)

along with the values

$$u(a) = v_0, u'(a) = v_1, ..., u^{(m-1)}(a) = v_{m-1}.$$
 (2)

The problem (1) with (2) is called an initial-value problem.

- Example: u' + u = x, u(0) = 0.
- Solution:  $u(x) = x 1 + e^{-x}$ .

# Unique solutions

### Theorem 1: Unique solution for IVP

If f is continuous on [a, b] and the ODE Lu = f is non-singular on [a, b], then the IVP (1) and (2) has a unique solution.

### Theorem 2: Homogeneous equation solution

If L is a linear m-th order differential operator and non-singular on [a, b], then the set of all solutions to the homogenous equation Lu = 0 on [a, b] forms a vector space of dimension m.

• What does this mean?: The solution space has a basis of dimension m, with elements  $u_1, ..., u_m$ . And so every solution to the homogeneous equation can be written as a linear combination of this basis:

$$u(x) = c_1 u_1(x) + \dots + c_m u_m(x), \quad \forall x \in [a, b].$$

This is called the general solution.

# Unique solutions

We can do the same for an inhomogeneous equation Lu = f by fixing a particular solution  $u_P$ . Then for any solution u,  $L(u - u_p) = f - f = 0$  and so we can write  $u - u_p$  as a linear combination of the homogeneous equation basis:

$$u(x) - u_p(x) = c_1 u_1(x) + \dots + c_m u_m(x), \quad \forall x \in [a, b].$$

Rearranging, we have the general solution for an inhomogeneous differential equation:

$$u(x) = u_P(x) + c_1 u_1(x) + \dots + c_m u_m(x), \quad \forall x \in [a, b].$$

# Solving an inhomogeneous DE

### Example 1

Solve the second order ODE

$$2u'' + u' - u = 10\sin(x).$$

Characteristic equation  $2t^2 + t - 1 = 0$ . Solutions  $t = -1, \frac{1}{2}$  so a basis for the homogeneous solution space is  $\{e^{-x}, e^{x/2}\}$ . Particular solution guess  $u_P(x) = A\cos(x) + B\sin(x)$ . Substituting gives A = -1 and B = -3, so  $u_P(x) = -\cos(x) - 3\sin(x)$ . We can write the general solution as

$$u(x) = c_1 e^{-x} + c_2 e^{x/2} - \cos(x) - 3\sin(x).$$

# Linear Independence

#### Definition 6: Linear independence

Let  $u_1(x), ..., u_m(x)$  be functions on some interval  $I \subset \mathbb{R}$ . We say that  $u_1, ..., u_m$  are **linearly independent** if

$$a_1u_1(x) + \dots + a_mu_m(x) = 0, \quad \forall x \in I,$$

implies that the constants  $a_1, ..., a_m$  are all 0. Otherwise, we say that that  $u_1, ..., u_m$  are **linearly dependent**.

# Wronskian

#### Definition 7: Wronskian

The **Wronskian** of the functions  $u_1, ..., u_m$  is the  $m \times m$  determinant

$$W(x) = W(x; u_1, ..., u_m) = \det(D^{i-1}u_j).$$

For example, a  $3 \times 3$  Wronskian is

$$W(x) = W(x; u_1, ..., u_m) = \det \begin{pmatrix} u_1 & u_2 & u_3 \\ u'_1 & u'_2 & u'_3 \\ u''_1 & u''_2 & u''_3 \end{pmatrix}.$$

### Lemmas

#### Lemma 1

If  $u_1, ..., u_m$  are linearly dependent over an interval  $I \subset \mathbb{R}$  then  $W(x; u_1, ..., u_m) = 0$  for all  $x \in I$ .

#### Lemma 2

If  $u_1, ..., u_m$  are solutions to the ODE

$$a_m(x)u^{(m)}(x) + a_{m-1}(x)u^{(m-1)}(x) + \dots + a_0(x)u(x) = 0$$

on an interval  $I \subset \mathbb{R}$ , then the Wronskian satisfies

$$a_m(x)W'(x) + a_{m-1}(x)W(x) = 0 \quad \forall x \in I.$$

# Wronskian

### Example 2: MATH2221 2014 T2 2.iii).b

Given the functions  $u_1, u_2$ , prove that if they are solutions to a second-order, homogeneous linear differential equation

$$a_2(x)u'' + a_1(x)u' + a_0(x)u = 0,$$

then the Wronskian W satisfies

$$a_2(x)W' + a_1(x)W = 0.$$

$$W = u_1 u_2' - u_1' u_2, W' = u_1 u_2'' - u_1'' u_2.$$

$$a_2W' + a_1W = a_2(u_1u_2'' - u_1''u_2) + a_1(u_1u_2' - u_1'u_2)$$
  
=  $u_1(a_2u_2'' + a_1u_2') - u_2(a_2u_1'' + a_1u_1').$ 

# Wronskian

Add and subtract  $a_0u_1u_2$ :

$$a_2W' + a_1W = u_1(a_2u_2'' + a_1u_2' + a_0u_2) - u_2(a_2u_1'' + a_1u_1' + a_0u_1).$$

Since  $u_1$  and  $u_2$  are solutions to the ODE, the RHS is 0.

# Linear Independence of solutions

#### Theorem 3: Linear independence

Let  $u_1, ..., u_m$  be solutions to the non-singular, linear, homogeneous m-th order ODE Lu = 0 on the interval [a, b]. Then either

W(x) = 0 and the m solutions are linearly dependent,

or

 $W(x) \neq 0$  and the m solutions are linearly independent.

# Polynomial solution guess

#### Theorem 4

Let L = p(D) be a linear differential operator of order m with constant coefficients. Assume that  $p(0) \neq 0$ . Then for any integer  $r \geq 0$ , there exists a unique polynomial  $u_P$  of degree r such that  $Lu_P = x^r$ .

This means that if our linear ODE has a polynomial on the RHS, we should guess a polynomial of the same degree for our particular solution.

# Exponential solution guess

#### Theorem 5

Let L = p(D) and  $\mu \in \mathbb{C}$ . If  $p(\mu) \neq 0$ , then the function

$$u_P(x) = \frac{e^{\mu x}}{p(\mu)}$$

satisfies  $Lu_P = e^{\mu x}$ .

This means that we should guess a multiple of  $e^{\mu x}$  when the RHS of a linear ODE is  $e^{\mu x}$  if it is not already a solution of the homeogeneous solution.

# Polynomial + exponential

#### Theorem 6

Let L = p(D) and assume  $\mu \in \mathbb{C}$ . If  $p(\mu) \neq 0$  then for any integers  $r \geq 0$ , there exists a unique polynomial v of degree r such that

$$u_P(x) = v(x)e^{\mu x}$$

satisfies  $Lu_p = x^r e^{\mu x}$ .

So if the RHS of the inhomogeneous linear ODE is a polynomial times an exponential, and the exponential isn't in the homogeneous solution, then the guess for the particular solution should be a polynomial times exponential.

# General solutions

#### Example 3: MATH2221 2014 T2 2.ii)

Let 
$$p(z) = (z-1)(z+2)^2(z^2+1)$$
 and  $D = \frac{d}{dx}$ .

- Write down the general solution  $u_H$  of the 5-th order, linear homogeneous ODE p(D)u = 0.
- Write down the form of a particular solution  $u_P$  to the inhomogeneous ODE

$$p(D)u = e^{-2x} + x^2 + \cos(x).$$

The zeros of p(z) are 1, -2 (multiplicity 2), i and -i. So the homogeneous ODE has general solution

$$u_H(x) = c_1 e^x + c_2 e^{-2x} + c_3 x e^{-2x} + c_4 \cos(x) + c_5 \sin(x).$$

### General solutions

Now consider the inhomogeneous ODE. The solution form will need to contain a  $x^2e^{-2x}$  term (since all lower powers are in the homogeneous solution space), as well a second order polynomial, as well as  $x\cos(x)$  and  $x\sin(x)$  terms (since  $\cos(x)$  and  $\sin(x)$  are in the homogeneous solution space). Putting these together,

$$u_P(x) = a_1 x^2 e^{-2x} + (a_2 x^2 + a_3 x + a_4) + x(a_5 \cos(x) + a_6 \sin(x)).$$

### Reduction of order

#### Theorem 7: Reduction of order

If we know a solution  $u_1(x) \neq 0$  to the ODE

$$u'' + p(x)u' + q(x)u = 0$$

then we can find a second solution

$$u_2(x) = u_1(x) \int \frac{1}{u_1(x)^2 \exp(\int p(x)dx)} dx.$$

## Reduction of order

#### Example 4

Find the general solution to

$$x^2y'' + 2xy' - 2y = 0$$

given that  $y_1(x) = x$  is a solution.

We need to rewrite the ODE in the form from Theorem 3:

$$y'' + \frac{2}{x}y' - \frac{2}{x^2}y = 0.$$

Then  $\exp\left(\int p(x)dx\right) = \exp\left(\int \frac{2}{x}dx\right) = \exp\left(2\ln(x)\right) = x^2$ . Substituting into the reduction of order formula,

$$y_2(x) = x \int \frac{1}{x^4} dx = -\frac{1}{3x^2}.$$

# Annihilator method (2221 only)

This method gives us a way to find particular solutions to an inhomogeneous differential equation. Start with an ODE

$$Lu = f(x).$$

We find a differential operator A(D) which **annihilates** f(x), meaning A(D)f(x) = 0. For example  $D^3$  annihilates  $x^2$ . We apply this differential operator to both sides of our ODE:

$$A(D)Lu = 0.$$

A solution to this homogeneous differential equation will be a particular solution to the original differential equation.

# Annihilator method (2221 only)

#### Example 5

Find a particular solution to the ODE

$$y'' - 2y' + y = e^x + \sin(x).$$

The LHS differential operator is  $p(D) = D^2 - 2D + 1 = (D-1)^2$ . The function  $e^x$  is annihilated by (D-1). The function  $\sin(x)$  is annihilated by  $(D^2+1)$ . So the annihilator is  $A(D) = (D-1)(D^2+1)$ . Apply the annihilator to both sides:

$$A(D)p(D)y(x) = (D-1)^3(D^2+1)y(x) = 0.$$

Solutions to the characteristic equation here is 1 (multiplicity 3), i and -i. The particular solution is of the form

$$y_P(x) = c_1 e^x + c_2 x e^x + c_3 x^2 e^x + c_4 \sin(x) + c_5 \cos(x).$$

# Annihilator method (2221 only)

The first two terms in this particular solution are contained in the homogeneous solution of the original ODE, so we discard them:

$$y_P(x) = c_3 x^2 e^x + c_4 \sin(x) + c_5 \cos(x).$$

Substitute into the original ODE, coefficients are  $c_3 = \frac{1}{2}$ ,  $c_4 = 0$  and  $c_5 = \frac{1}{2}$ . So our particular solution is

$$y_P(x) = \frac{1}{2}x^2e^x + \frac{1}{2}\cos(x).$$

# Variation of parameters

If we have a linear, 2nd order, inhomogeneous ODE with leading coefficient 1,

$$Lu = u'' + p(x)u' + q(x)u = f(x).$$

Let  $u_1, u_2$  be a basis for the homogeneous solution space, and let W(x) be the Wronskian  $W(x; u_1, u_2)$ . Then a particular solution to the inhomogeneous equation is

$$u(x) = v_1(x)u_1(x) + v_2(x)u_2(x),$$

where

$$v_1'(x) = \frac{-u_2(x)f(x)}{W(x)}$$
, and  $v_2'(x) = \frac{u_1(x)f(x)}{W(x)}$ .

# Variation of parameters

### Example 6: MATH2121 2018 T2 1.i)

Use the variation of parameters method to solve

$$y'' - 3y' + 2y = e^x \sin(x).$$

Homogeneous solution has characteristic equation  $t^2 - 3t + 2 = 0$  so  $u_1(x) = e^x$  and  $u_2(x) = e^{2x}$ . The Wronskian is

$$W(x) = \det \begin{pmatrix} e^x & e^{2x} \\ e^x & 2e^{2x} \end{pmatrix} = e^{3x}.$$

Then

$$v_1'(x) = \frac{-e^{2x}e^x\sin(x)}{e^{3x}} = -\sin(x),$$

and

$$v_2'(x) = \frac{e^x e^x \sin(x)}{e^{3x}} = e^{-x} \sin(x).$$

# Variation of parameters

Integrating both  $v_1$  and  $v_2$  we have

$$v_1(x) = \cos(x)$$
, and  $v_2(x) = -\frac{1}{2}e^{-x}(\sin(x) + \cos(x))$ .

Then a particular solution is

$$u(x) = \cos(x)e^x - \frac{1}{2}e^{-x}(\sin(x) + \cos(x))e^{2x}$$
$$= \frac{1}{2}e^x(\cos(x) - \sin(x)).$$

consider a general second-order, linear, homogeneous ODE

$$Lu = a_2(x)u'' + a_1(x)u' + a_0(x)u = 0,$$

or equivalently

$$Lu = u'' + p(x)u' + q(x)u = 0$$

with  $p(x) = \frac{a_1(x)}{a_2(x)}$  and  $q(x) = \frac{a_0(x)}{a_2(x)}$ . Assume that  $a_0, a_1, a_2$  are analytic at 0 (ie. they have local convergent power series), and  $a_2(0) \neq 0$ . Then p and q are analytic at 0, so we can find power series expansions of both:

$$p(z) = \sum_{k=0}^{\infty} p_k z^k, \quad q(z) = \sum_{k=0}^{\infty} q_k z^k \text{ for } |z| < \rho,$$

where  $\rho > 0$ .

#### Theorem 8

If coefficients p(z) and q(z) are analytic for  $|z| < \rho$ , then the formal power series for the solution u(z) constructed in the previous slide, is also analytic for  $|z| < \rho$ .

This means that if we find where p(z) and q(z) are both analytic, the power series solution u(z) is also analytic in this region.

### Example 7: MATH2121 2018 T2 1.iii)

We aim to construct a series solution to the ODE about the ordinary point  $x_0 = 0$ :

$$(1 - x2)y'' - 2xy' + 20y = 0, \quad y(0) = 1, y'(0) = 0,$$

of the form

$$y(x) = \sum_{n=0}^{\infty} A_n x^n.$$

- Give the recurrence relation for the coefficients  $A_n$ .
- Explain from the recurrence relation that one of the series will terminate yielding a polynomial solution, and the other does not.
- Write down the polynomial solution.

Note that

$$y'(x) = \sum_{n=1}^{\infty} nA_n x^{n-1}, \quad y''(x) = \sum_{n=2}^{\infty} n(n-1)A_n x^{n-2}.$$

Then we substitute into the ODE:

$$Ly = y'' + (-x^2y'' - 2xy' + 20y)$$

$$= \sum_{n=2}^{\infty} n(n-1)A_n x^{n-2} + \sum_{n=2}^{\infty} -n(n-1)A_n x^n$$

$$- \sum_{n=1}^{\infty} 2nA_n x^n + \sum_{n=0}^{\infty} 20A_n x^n$$

$$= \sum_{n=2}^{\infty} n(n-1)A_n x^{n-2} + \sum_{n=0}^{\infty} -n(n-1)A_n x^n$$

$$- \sum_{n=2}^{\infty} 2nA_n x^n + \sum_{n=0}^{\infty} 20A_n x^n.$$

We combine the last three sums as follows.

$$Ly = \sum_{n=2}^{\infty} n(n-1)A_n x^{n-2} + \sum_{n=0}^{\infty} (-n^2 - n + 20)A_n x^n.$$

In the first sum, we change n to n+2.

$$Ly = \sum_{n=0}^{\infty} (n+1)(n+2)A_{n+2}x^n + \sum_{n=0}^{\infty} (-n^2 - n + 20)A_nx^n.$$

We can finally combine both sums.

$$Ly = \sum_{n=0}^{\infty} [(n+1)(n+2)A_{n+2} - (n+5)(n-4)A_n] x^n.$$

Since Ly = 0, we need  $(n+1)(n+2)A_{n+2} - (n+5)(n-4)A_n = 0$ . Rearranging,

$$A_{n+2} = \frac{(n+5)(n-4)}{(n+1)(n+2)} A_n.$$

We have  $A_0 = 1$  and  $A_1 = 0$ , so all odd terms are zero and the even terms terminate Since  $A_6 = 0$ . Hence the polynomial solution is

$$y(x) = 1 - 10x^2 + \frac{35}{3}x^4.$$

### Example 8: MATH2221 2015 T2 1.iii)

Consider the ODE

$$(1+z^2)u'' - zu' - 3u = 0.$$

• Find the recurrence relation satisfied by the coefficients  $A_k$  in any power series solution:

$$u = \sum_{k=0}^{\infty} A_k z^k.$$

- Show that  $A_5 = A_7 = A_9 = \dots = 0$ .
- Hence find the solution for which u(0) = 0, u'(0) = 6.

Note that

$$u'(z) = \sum_{k=1}^{\infty} k A_k z^{k-1}, \quad u''(z) = \sum_{k=2}^{\infty} k(k-1) A_k z^{k-2}.$$

Substitute into the ODE:

$$Lu = u'' + (z^{2}u'' - zu' - 3u)$$

$$= \sum_{k=2}^{\infty} k(k-1)A_{k}z^{k-2} + \sum_{k=2}^{\infty} k(k-1)A_{k}z^{k}$$

$$- \sum_{k=1}^{\infty} kA_{k}z^{k} - \sum_{k=0}^{\infty} A_{k}z^{k}$$

$$= \sum_{k=2}^{\infty} k(k-1)A_{k}z^{k-2} + \sum_{k=0}^{\infty} k(k-1)A_{k}z^{k}$$

$$- \sum_{k=2}^{\infty} kA_{k}z^{k} - \sum_{k=0}^{\infty} A_{k}z^{k}.$$

#### Power series

Combine the last three sums.

$$Lu = \sum_{k=2}^{\infty} k(k-1)A_k z^{k-2} + \sum_{k=0}^{\infty} (k^2 - 2k - 3)A_k z^k.$$

In the first sum, change k to k+2.

$$Lu = \sum_{k=0}^{\infty} (k+1)(k+2)A_{k+2}z^k + \sum_{k=0}^{\infty} (k^2 - 2k - 3)A_k z^k.$$

Finally, we can combine the sums.

$$Lu = \sum_{k=0}^{\infty} (k+1) \left[ (k+2)A_{k+2} + (k-3)A_k \right] z^k.$$

### Power series

Since Lu = 0, we need  $(k+2)A_{k+2} + (k-3)A_k = 0$ . Rearranging,

$$A_{k+2} = -\frac{k-3}{k+2} A_k.$$

Since  $A_5 = -\frac{3-3}{3+2}A_3 = 0$ , then all odd terms past  $A_5$  are zero. Also  $A_1 = 6$  and  $A_3 = 4$ .

The even terms start at  $A_0 = 0$  so all even terms past this are zero. Hence

$$u(z) = 6z + 4z^3.$$

# Singular/Cauchy-Euler ODEs

For singular ODEs, we only need to check the case when the leading coefficient vanishes at the origin.

A second-order Cauchy-Euler ODE has the form

$$Lu = ax^2u'' + bxu' + cu = f(x),$$

where a, b, c are constants with  $a \neq 0$ . This is singular at x = 0. Applying this differential operator L to  $x^r$ ,

$$Lx^r = \left[ar(r-1) + br + c\right]x^r,$$

we can see that  $x^r$  is a solution to the homogeneous equation Lu = 0 iff

$$ar(r-1) + br + c = 0.$$

# Singular/Cauchy-Euler ODEs

#### Lemma 3

Suppose there are distinct solutions  $r_1, r_2$  to the equation ar(r-1) + br + c = 0. That is,  $r_1 \neq r_2$ . Then the general solution of the homogeneous equation Lu = 0 is

$$u(x) = C_1 x^{r_1} + C_2 x^{r_2}, \quad x > 0.$$

#### Lemma 4

Suppose there is one solution  $r_1$  to ar(r-1) + br + c = 0. Then the general solution to the homogeneous equation Lu = 0 is

$$C_1 x^{r_1} + C_2 x^{r_1} \log(x), \quad x > 0.$$

## Cauchy-Euler ODEs

For a particular solution to the inhomogeneous Cauchy-Euler equation

$$ax^2u'' + bxu' + cu = x^r,$$

we can use the particular solution guess  $u(x) = \alpha x^r$ .

## Cauchy-Euler ODEs

#### Example 9: MATH2121 2016 T2 2.i)

Find the general solution of the Cauchy-Euler ODE

$$2x^2y'' + 7xy' + 3y = 13x^{1/4}, \quad x > 0.$$

We want to solve the equation ar(r-1) + br + c = 0, where a = 2, b = 7 and c = 3. That is,  $2r^2 + 5r + 3 = 0$ . Solutions are  $r_1 = -1$  and  $r_2 = -3/2$ . Using Lemma 3, we have general homogeneous solution

$$y_H(x) = C_1 x^{-1} + C_2 x^{-3/2}.$$

A particular solution can be found by applying variation of parameters, giving us

$$y_P(x) = -8x^{1/4}$$
.

So 
$$y(x) = C_1 x^{-1} + C_2 x^{-3/2} - 8x^{1/4}$$
.

## Cauchy-Euler ODEs - Complex Roots

#### Example

Find the general solution of the Cauchy-Euler ODE

$$x^2u'' - xu' + 2u = 0, \quad x > 0.$$

We want to solve the equation ar(r-1) + br + c = 0, where a = 1, b = -1 and c = 2. That is,  $r^2 - 2r + 2 = 0$ . Solutions are  $r_1 = 1 + i$  and  $r_2 = 1 - i$ . Using Lemma 3, we have general homogeneous solution

$$u_H(x) = C_1 x^{1+i} + C_2 x^{1-i}$$
.

But we want real solutions, and this is clearly not real! However, we can use the fact that

$$x^{1+i} = \exp(\ln(x^{1+i})) = \exp((1+i)\ln x)$$
  
=  $xe^{i\ln x}$   
=  $x(\cos(\ln x) + i\sin(\ln x)).$ 

# Cauchy-Euler ODEs - Complex Roots - Continued

Similarly, we have  $x^{1-i} = x(\cos(\ln x) - i\sin(\ln x))$ . Using these facts, we can get real solutions that form a basis for the solution space, by taking  $1/2(x^{1+i} + x^{i-1}) = x\cos(\ln x)$  and  $1/2(x^{1+i} - x^{i-1}) = x\sin(\ln x)$ . Thus our general solution is

$$u_H(x) = B_1 x \cos(\ln x) + B_2 x \sin(\ln x)$$

for any constants  $B_1, B_2$ .

### Frobenious normal form

A frequent form of ODE that appears in many applications can be written in **Frobenious normal form**:

$$z^{2}u'' + zP(z)u' + Q(z)u = 0,$$

where P(z) and Q(z) are analytic at 0. Let  $P_0 = P(0)$  and  $Q_0 = Q(0)$ , and define a series F as

$$F(z;r) = z^r \sum_{k=0}^{\infty} A_k(r) z^k.$$

Consider the equation  $r(r-1) + P_0r + Q_0 = 0$  with solutions  $r_1$  and  $r_2$ .

#### Lemma 5

If  $r_1 \neq r_2$ , then  $f(z; r_1)$  is a solution to the Frobenious normal form ODE. If  $r_1 - r_2$  is **not** a whole number, then a second linearly independent solution is  $F(z; r_2)$ .

The Bessel equation with parameter  $\nu$  is:

$$z^2u'' + zu' + (z^2 - \nu^2)u = 0.$$

This ODE is in Frobenious normal form, with indicial polynomial:

$$I(r) = (r + \nu)(r - \nu),$$

and we seek a series solution:

$$u(z) = \sum_{k=0}^{\infty} A_k z^{k+r}.$$

We assume  $\text{Re}(\nu) \geq 0$ , so  $r_1 = \nu$  and  $r_2 = -\nu$ .

With the normalisation:

$$A_0 = \frac{1}{2^{\nu} \Gamma(1+\nu)}$$

the series solution is called the **Bessel function of order**  $\nu$  and is denoted:

$$J_{\nu}(z) = \frac{(z/2)^{\nu}}{\Gamma(1+\nu)} \left[ 1 - \frac{(z/2)^{\nu}}{1+\nu} + \frac{(z/2)^4}{2!(1+\nu)(2+\nu)} - \dots \right].$$

And from the functional equation  $\Gamma(1+z)=z\Gamma(z)$ :

$$J_{\nu}(z) = \sum_{k=0}^{\infty} \frac{(-1)k(z/2)^{2k+\nu}}{k!\Gamma(k+1+\nu)}$$

If  $\nu$  is not an integer, then a second linearly independent, solution is:

$$J_{-\nu}(z) = \sum_{k=0}^{\infty} \frac{(-1)^k (z/2)^{2k-\nu}}{k!\Gamma(k+1-\nu)}.$$

For an integer  $\nu = n \in \mathbb{Z}$ , since  $\Gamma(n+1) = n!$ , we have:

$$J_n(z) = \sum_{k=0}^{\infty} \frac{(-1)^k (z/2)^{2k+n}}{k!(k+n)!}.$$

Also, since  $\frac{1}{\Gamma(z)} = 0$  for z = 0, -1, -2, ..., we find that  $J_n$  and  $J_{-n}$  are linearly dependent; in fact:

$$J_{-n}(z) = (-1)^n J_n(z).$$

#### Example 10: MATH2221 2015 T2 2.ii)

1. Use term-by-term differentiation to prove that for  $\nu \in \mathbb{R}$  and x > 0:

$$\frac{d}{dx}(x^{\nu}J_{\nu}(x)) = x^{\nu}J_{\nu-1}(x).$$

2. Hence evaluate the definite integral:

$$I = \int_{0}^{1} x^{\frac{7}{2}} J_{\frac{1}{2}}(x) dx.$$

Note that

$$x^{\nu}J_{\nu}(x) = \sum_{k=0}^{\infty} x^{\nu} \frac{(-1)^k (x/2)^{2k+\nu}}{k!\Gamma(k+1+\nu)}.$$

Moving the  $x^{\nu}$  into the fraction,

$$x^{\nu} J_{\nu}(x) = \sum_{k=0}^{\infty} \frac{(-1)^k (x)^{2k+2\nu}}{2^{2k+\nu} k! \Gamma(k+1+\nu)}.$$

Take one term in this sum and differentiate with respect to x:

$$\frac{d}{dx}\left(\frac{(-1)^k(x)^{2k+2\nu}}{2^{2k+\nu}k!\Gamma(k+1+\nu)}\right) = \frac{(-1)^k(2k+2\nu)x^{2k+2\nu-1}}{2^{2k+\nu}k!\Gamma(k+1+\nu)}.$$

Since  $\Gamma(k+1+\nu) = (k+\nu)\Gamma(k+1+\nu)$ , then

$$\frac{d}{dx}\left(\frac{(-1)^k(x)^{2k+2\nu}}{2^{2k+\nu}k!\Gamma(k+1+\nu)}\right) = \frac{(-1)^k(2k+2\nu)x^{2k+2\nu-1}}{2^{2k+\nu}k!(k+\nu)\Gamma(k+\nu)}.$$

However there is a factor of  $2(k + \nu)$  in both the numerator and denominator. So

$$\frac{d}{dx}\left(\frac{(-1)^k(x)^{2k+2\nu}}{2^{2k+\nu}k!\Gamma(k+1+\nu)}\right) = \frac{(-1)^k(x)^{2k+2\nu-1}}{2^{2k+\nu-1}k!\Gamma(k+\nu)}.$$

Finally, factor out a factor  $x^{\nu}$ :

$$\frac{d}{dx} \left( \frac{(-1)^k (x)^{2k+2\nu}}{2^{2k+\nu} k! \Gamma(k+1+\nu)} \right) = x^{\nu} \frac{(-1)^k (x/2)^{2k+\nu-1}}{k! \Gamma(k+\nu)}.$$

Since the derivative is linear,

$$\frac{d}{dx}(x^{\nu}J_{\nu}(x)) = \sum_{k=0}^{\infty} \frac{d}{dx} \left( \frac{(-1)^k (x)^{2k+2\nu}}{2^{2k+\nu}k!\Gamma(k+1+\nu)} \right).$$

Substituting in the derivative we found,

$$\frac{d}{dx}(x^{\nu}J_{\nu}(x)) = x^{\nu} \sum_{k=0}^{\infty} \frac{(-1)^k (x/2)^{2k+\nu-1}}{k!\Gamma(k+\nu)}.$$

The RHS here is just  $x^{\nu}J_{\nu-1}(x)$ , so we are done. To find the integral  $\int_{0}^{1} x^{\frac{7}{2}}J_{\frac{1}{2}}(x)dx$ , we separate and use integration by parts:

$$\int_{0}^{1} x^{2} \cdot x^{\frac{3}{2}} J_{\frac{1}{2}}(x) dx$$

where  $u' = x^{\frac{3}{2}} J_{\frac{1}{2}}(x)$  and  $v = x^2$ .

Applying integration by parts:

$$I = [x^{2}.x^{\frac{3}{2}}J_{\frac{3}{2}}(x)]_{0}^{1} - \int_{0}^{1} 2x \cdot x^{\frac{3}{2}}J_{\frac{3}{2}}(x)dx$$
$$= J_{\frac{3}{2}}(1) - 2\int_{0}^{1} x^{\frac{5}{2}}J_{\frac{3}{2}}(x)dx.$$

However by our previous result,  $\frac{d}{dx}(x^{\nu}J_{\nu}(x)) = x^{\nu}J_{\nu-1}(x)$ . So the antiderivative of  $x^{5/2}J_{3/2}$  is  $x^{5/2}J_{5/2}$ , hence

$$\begin{split} &=J_{\frac{3}{2}}(1)-2[x^{\frac{5}{2}}J_{\frac{5}{2}}(x)]_{0}^{1}\\ &=J_{\frac{3}{2}}(1)-2J_{\frac{5}{2}}(1). \end{split}$$

# Legendre equation (2221 only)

The **Legendre equation** with paramater  $\nu$  is:

$$(1 - z2)u'' - 2zu' + \nu(\nu + 1)u = 0.$$

This ODE is not singular at z = 0, so the solution has an ordinary Taylor series expansion:

$$u = \sum_{k=0}^{\infty} A_k z^k.$$

The  $A_k$  must satisfy:

$$(k+1)(k+2)A_{k+2} - [k(k+1) - \nu(\nu+1)]A_k = 0.$$

The recurrence relation is:

$$A_{k+2} = \frac{(k-\nu)(k+\nu+1)}{(k+1)(k+2)} A_k$$
 for  $k \ge 0$ .

# Legendre equation (2221 only)

We have:

$$u(z) = A_0 u_0(z) + A_1 u_1(z)$$

where:

$$u_0(z) = 1 - \frac{\nu(\nu+1)}{2!}z^2 + \frac{(\nu-2)\nu(\nu+1)(\nu+3)}{4!}z^-...$$

and:

$$u_1(z) = z - \frac{(\nu - 1)(\nu - 2)}{3!}z^3 + \frac{(\nu - 3)(\nu - 1)(\nu + 2)(\nu + 4)}{5!}z^5 - \dots$$

Suppose now that  $\nu = n$  is a non-negative integer. If n is even, then the series for  $u_0(z)$  terminates, whereas if n is odd, then the series for  $u_1(z)$  terminates. The terminating solution is then called the

**Legendre polynomial** of degree n and is denoted by  $P_n(z)$  with the normalisation:

$$P_n(1) = 1.$$

## 2. Dynamical Systems

## Dynamical Systems

**State variables** are natural variables which depend on a single independent variable. A **dynamical system** is a natural process described by these state variables. The state of a system at a given time is described by the values of the state variables at that instant.

Note that any nth order ODE can be writen as a system of **first order** ODEs (not vice versa):

$$\frac{d^n y}{dt^n} = g\left(y, \frac{dy}{dt}, ..., \frac{d^{n-1}y}{dt^{n-1}}\right)$$
$$\frac{dx}{dt} = f(x_1, x_2, ..., x_n)$$

### Non-autonomous ODEs

#### Definition 8

A system of ODEs of the form:

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x})$$

is said to be autonomous.

#### Definition 9

In a non-autonomous system,  $\mathbf{F}$  may depend explicitly on t:

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}, t).$$

# Lipschitz (2221 only)

#### Definition 10

The number  $L \in \mathbb{R}$  is a Lipschitz constant for a function  $f : [a, b] \to \mathbb{R}$  if

$$|f(x) - f(y)| \le L|x - y| \quad \forall x, y \in [a, b].$$

We say that the function f is Lipschitz if a Lipschitz constant for f exists.

#### Theorem 9

If f is Lipschitz, then f is uniformly continuous.

#### Lemma 6

If  $f: I \to \mathbb{R}$  is differentiable and f' is continuous on I, then f is Lipschitz.

# Lipschitz (2221 only)

#### Example 11

$$f(x) = 2\sqrt{x}, \quad x \ge 0$$

Where is f Lipschitz?

Since  $f'(x) = \frac{1}{\sqrt{x}}$  is continuous on x > 0, f is Lipschitz on any closed interval [a, b] such that 0 < a < b (Lemma 6).

Suppose a > 0 and consider some interval [0, a] containing 0. Let  $x \in (0, a]$ . Then,

$$\left| \frac{f(x) - f(0)}{x - 0} \right| = \frac{1}{\sqrt{x}} \to \infty \text{ as } x \to 0.$$

Hence, a Lipschitz constant does not exist for f on a non-empty closed interval containing 0.

# Lipschitz (2221 only)

#### Example 11

Find a Lipschitz constant for the function  $f:[0,\pi]\to\mathbb{R}$  defined by  $f(x)=x\cos x$ .

First note that

$$f'(x) = \cos x - x \sin x$$

$$\leq |\cos x - x \sin x|$$

$$\leq |\cos x| + |x \sin x| \qquad \text{(by triangle inequality)}$$

$$\leq 1 + |x| \leq 1 + \pi \quad \text{for } x \in [0, \pi].$$

For any  $x, y \in [0, \pi]$  where x < y, there exists  $c \in (x, y)$  such that

$$|f(y) - f(x)| = |f'(c)||y - x| \le (1 + \pi)|y - x|$$

by the Mean Value Theorem. Hence,  $1 + \pi$  is a Lipschitz constant.

## Lipschitz Vector Field (2221 only)

We extend the definition of Lipschitz to vector fields.

#### Definition 11

A vector field  $\mathbf{F}: S \subseteq \mathbb{R}^n \to \mathbb{R}^n$  is Lipschitz on S if

$$||\mathbf{F}(\mathbf{x}) - \mathbf{F}(\mathbf{y})|| \le L ||\mathbf{x} - \mathbf{y}|| \quad \forall \mathbf{x}, \mathbf{y} \in S.$$

Here,

$$||\mathbf{x}|| = \left(\sum_{j=1}^{m} x_j^2\right)^{1/2}$$

denotes the Euclidean norm of the vector  $\mathbf{x} \in \mathbb{R}^m$ .

We say that  $\mathbf{F}(\mathbf{x},t)$  is Lipschitz in  $\mathbf{x}$  if, for all t:

$$||\mathbf{F}(\mathbf{x},t) - \mathbf{F}(\mathbf{y},t)|| \le L ||\mathbf{x} - \mathbf{y}||, \quad \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^m.$$

# Existence and Uniqueness Theorem (2221 only)

We want to find solutions to a non-autonomous system. The following theorem guarantees a unique solution for a non-autonomous system, under certain conditions.

#### Theorem 10

Let  $r > 0, \tau > 0$  and suppose  $S = \{(\mathbf{x}, t) : ||\mathbf{x} - \mathbf{x}_0|| \le r, |t - t_0| \le \tau\}$ . If  $\mathbf{F}(\mathbf{x}, t)$  is Lipschitz and  $||\mathbf{F}(\mathbf{x}, t)|| \le M$  for  $(\mathbf{x}, t) \in S$  then the initial value problem defined by

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}, t), \quad \mathbf{x}(t_0) = \mathbf{x}_0$$

has a unique solution  $\mathbf{x}(t)$  over a time interval  $|t - t_0| < \min(r/M, \tau)$ .

In fact,  $\mathbf{x}(t)$  is continuous and differentiable.

### Notes

- The existence of solutions follow from continuity in x and t.
- The uniqueness of solutions follow from the Lipschitz condition in x.
- The theorem is a local existence theorem. It provides for the existence of solutions over a finite time interval.
- Outside of the time interval specified by the theorem, there may or may not be a unique solution.

## Linear systems of ODEs

#### Definition 12

We say that the  $n \times n$ , first-order system of ODEs:

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}, t)$$

is linear if the RHS has the form:

$$\mathbf{F}(\mathbf{x},t) = A(t)\mathbf{x} + \mathbf{b}(t)$$

for some  $n \times n$  matrix-valued function  $A(t) = [a_{i,j}(t)]$  and a vector-valued function  $\mathbf{b}(t) = [b_i(t)]$ .

The linear first-order system is autonomous when A and  $\mathbf{b}$  are constant.

## Global Existence and Uniqueness

We have a stronger existence result in the linear case:

#### Theorem 11

If A(t) and  $\mathbf{b}(t)$  are continuous for  $0 \le t \le T$ , then the linear initial-value problem

$$\frac{d\mathbf{x}}{dt} = A(t)\mathbf{x} + \mathbf{b}(t)$$
 with  $\mathbf{x}(0) = \mathbf{x}_0$ ,

has a unique solution  $\mathbf{x}(t)$  for  $0 \le t \le T$ .

## A special case

It is much easier to work with the special case when A(t) = A is a constant  $n \times n$  matrix and  $\mathbf{b}(t) = \mathbf{0}$ :

$$\frac{d\mathbf{x}}{dt} = A\mathbf{x}.$$

The general solution to this system is

$$\mathbf{x}(t) = \sum_{i=1}^{n} c_i e^{\lambda_i t} \mathbf{v}_i,$$

where  $\lambda_i$  is the *i*-th eigenvalue with corresponding eigenvector  $\mathbf{v}_i$  and  $c_1, ..., c_n$  are constants.

### Initial-valued system

Recall that for an  $n \times n$  complex matrix A, we define the exponential of a matrix as

$$e^A = \sum_{k=0}^{\infty} \frac{A^k}{k!} = I + A + \frac{1}{2!}A^2 + \frac{1}{3!}A^3 + \cdots$$

Consider the initial-valued system

$$\frac{d\mathbf{x}}{dt} = A\mathbf{x}, \quad \mathbf{x}(0) = \mathbf{x}_0.$$

Then we can write the solution as a matrix exponential,

$$\mathbf{x}(t) = e^{tA}\mathbf{x}_0.$$

## Example

#### Example 12: MATH2121 2016 T2 2.iii)

For an  $n \times n$  matrix A.

- 1. State the definition of  $e^A$ .
- 2. Show that if  $A\mathbf{v} = \lambda \mathbf{v}$ , then  $e^A \mathbf{v} = e^{\lambda} \mathbf{v}$ .

The definition of the matrix exponential is

$$e^A = \sum_{k=0}^{\infty} \frac{A^k}{k!} = I + A + \frac{A^2}{2!} + \frac{A^3}{3!} + \cdots$$

If  $A\mathbf{v} = \lambda \mathbf{v}$  then

$$e^{A}\mathbf{v} = \left(\sum_{k=0}^{\infty} \frac{1}{k!} A^{k}\right) \mathbf{v}$$
$$= \sum_{k=0}^{\infty} \frac{1}{k!} \left(A^{k} \mathbf{v}\right).$$

## Example

But we can find  $A^k \mathbf{v}$ :

$$A^k \mathbf{v} = A^{k-1} \lambda \mathbf{v} = \dots = \lambda^k \mathbf{v}.$$

Hence

$$e^{A}\mathbf{v} = \sum_{k=0}^{\infty} \frac{1}{k!} \left( A^{k} \mathbf{v} \right)$$
$$= \left( \sum_{k=0}^{\infty} \frac{\lambda^{k}}{k!} \right) \mathbf{v}.$$

This is the exponential function Taylor series at the point  $\lambda$ , so

$$e^A \mathbf{v} = e^{\lambda} \mathbf{v}.$$

## Sanity Check

#### Matrix Exponential and Linear Combination of Eigenvectors

Using a similar method as what we've done above, it can be shown that  $e^{tA}\mathbf{v} = e^{\lambda t}\mathbf{v}$  where  $\mathbf{v}$  is an eigenvector for eigenvalue  $\lambda$  of the matrix A. The above result confirms that the solution obtained by writing  $\mathbf{x}(t) = e^{tA}\mathbf{x_0}$  is the same as writing  $\mathbf{x}(t) = \sum_{i=1}^{n} c_i e^{\lambda_i t} \mathbf{v}_i$ , where  $\mathbf{v}_i$  is the eigenvector for eigenvalue  $\lambda_i$  of A because

$$e^{tA}\mathbf{x_0} = e^{tA}\mathbf{x}(0) = e^{tA}\sum_{i=1}^n c_i\mathbf{v}_i = \sum_{i=1}^n c_ie^{tA}\mathbf{v}_i = \sum_{i=1}^n c_ie^{\lambda_i t}\mathbf{v}_i.$$

# Calculating the Matrix Exponential

We now want to find a practical way of calculating  $e^{tA}$ , which requires finding  $A^k$  for  $k \in \mathbb{N}$ . To do this efficiently, we have a few methods and properties to look to.

- If A is a nilpotent matrix that is,  $A^n = \mathbf{0}$  for some  $n \in \mathbb{Z}^+$  we can sum all the terms in the Taylor series expansion up to the smallest such n.
- If A is diagonalisable, we can first diagonalise the matrix A. We explore the method of diagonalisation in more detail.
- If A can be written as a sum of two matrices B and C such that BC = CB, then we can write  $e^A = e^B e^C$ . We can use the above two methods to find  $e^B$  and  $e^C$ .

### Diagonalising a matrix

#### Definition 13

An  $n \times n$  complex matrix A is diagonalisable if there exists a non-singular matrix  $n \times n$  matrix M such that  $M^{-1}AM$  is diagonal.

#### Theorem 12

An  $n \times n$  complex matrix A is diagonalisable if and only if there exists a basis  $\{\mathbf{v}_1, ..., \mathbf{v}_n\}$  for  $\mathbb{C}^n$ , where  $\mathbf{v}_1, ..., \mathbf{v}_n$  are eigenvectors of A. In fact the columns of M are the eigenvectors of A,  $M = (\mathbf{v}_1 | \cdots | \mathbf{v}_n)$ , and  $M^{-1}AM = \Lambda$  where:

$$\Lambda = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$$

for eigenvalues  $\lambda_i$  corresponding to eigenvector  $\mathbf{v}_i$ .

#### Matrix Powers

In general since  $M^{-1}AM = \Lambda$ , we can efficiently calculate  $A^k$ :

$$A^{k} = \overbrace{M\Lambda M^{-1} \cdot M\Lambda M^{-1} \cdots M\Lambda M^{-1}}^{k \text{ times}} = M\Lambda^{k} M^{-1}.$$

This is better because

$$\Lambda^k = \begin{pmatrix} \lambda_1^k & & \\ & \ddots & \\ & & \lambda_n^k \end{pmatrix}.$$

In fact, due to the Taylor series of the exponential function, we can simplify the solution to the initial-valued system further.

## Exponential of a diagonalisable matrix

#### Theorem 13

If  $A = M\Lambda M^{-1}$  is diagonalisable, then:

$$e^A = Me^{\Lambda}M^{-1}$$
 and  $e^{\Lambda} = \begin{pmatrix} e^{\lambda_1} & & \\ & \ddots & \\ & & e^{\lambda_n} \end{pmatrix}$ .

We can now find the exponential of tA:

$$e^{tA} = Me^{t\Lambda}M^{-1}$$
, and  $e^{t\Lambda} = \begin{pmatrix} e^{t\lambda_1} & & \\ & \ddots & \\ & & e^{t\lambda_n} \end{pmatrix}$ .

### Equilibrium points

#### Definition 14

We say that  $\mathbf{a} \in \mathbb{R}^n$  is an equilibrium point for the dynamical system  $\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x})$  if

$$\mathbf{F}(\mathbf{a}) = \mathbf{0}.$$

Suppose **a** is an equilibrium point for the system  $\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x})$ . Consider the initial-valued system

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}), \quad \mathbf{x}(0) = \mathbf{a}.$$

Then the solution is the constant function  $\mathbf{x}(t) = \mathbf{a}$ .

### Stable Equilibrium

#### Definition 15

An equilibrium point **a** is **stable** if for every  $\epsilon > 0$ , there exists  $\delta > 0$  such that whenever  $||\mathbf{x}_0 - \mathbf{a}|| < \delta$ , the solution of

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}), \quad \mathbf{x}(0) = \mathbf{x}_0$$

satisfies

$$||\mathbf{x}(t) - \mathbf{a}|| < \epsilon \quad \forall t > 0.$$

Intuitively: if a solution starts close enough to the stable equilibrium point, then they will remain close to the stable equilibrium point.

## Asymptotic Stability

This is a stronger form of stability, on a particular subset of  $\mathbb{R}^n$ .

#### Definition 16

Let N be an open subset of  $\mathbb{R}^n$  that contains an equilibrium point  $\mathbf{a}$ . We say that  $\mathbf{a}$  is asymptotically stable in N if  $\mathbf{a}$  is stable, and whenever  $\mathbf{x}_0 \in N$  the solution of

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}) \quad \mathbf{x}(0) = \mathbf{x}_0$$

satisfies

$$\mathbf{x}(t) \to \mathbf{a}$$
 as  $t \to \infty$ .

We call N a domain of attraction for  $\mathbf{a}$ .

Intuitively: not only do the solutions stay close to the stable equilibrium point, but they also approach the equilibrium point as t goes to infinity.

#### Linear constant case

Consider the linear system

$$\frac{d\mathbf{x}}{dt} = A\mathbf{x} + \mathbf{b}, \quad \mathbf{x}(0) = \mathbf{x}_0,$$

where  $det(A) \neq 0$ . Then the unique equilibrium point of this system is

$$\mathbf{a} = -A^{-1}\mathbf{b},$$

and the solution to the system is

$$\mathbf{x}(t) = \mathbf{a} + e^{tA} \left( \mathbf{x}_0 - \mathbf{a} \right).$$

### Linear constant case

#### Theorem 14

Consider the previous linear constant coefficient system. Let  $\lambda_1, ..., \lambda_n$  be the eigenvalues of A. The equilibrium point  $\mathbf{a} = -A^{-1}\mathbf{b}$  is:

- 1. Stable if and only if  $Re(\lambda_i) \leq 0$  for all j.
- 2. asymptotically stable if and only if  $Re(\lambda_i) < 0$  for all j.

In the second case, the domain of attraction is the whole of  $\mathbb{R}^n$ .

# Classification of 2D Linear Systems

Туре	Eigenvalues	Eigenvectors	X(t)	Classification
1: $\mathbf{B} = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$	$\lambda_1 \neq \lambda_2 \in \mathbb{R}$	$v^{(1)}$ , $v^{(2)}$	$\left[\boldsymbol{v^{(1)}}e^{\lambda_1 t} \ \boldsymbol{v^{(2)}}e^{\lambda_2 t}\right]$	Improper Node
	$\lambda_1 < 0 < \lambda_2$	$v^{(1)}, v^{(2)}$	$\left[\boldsymbol{v^{(1)}}e^{\lambda_1 t} \ \boldsymbol{v^{(2)}}e^{\lambda_2 t}\right]$	Saddle Point
2: $\mathbf{B} = \begin{pmatrix} \lambda & \gamma \\ 0 & \lambda \end{pmatrix}$ $\lambda, \gamma \in \mathbb{R}$	$\lambda_1 = \lambda_2 = \lambda \in \mathbb{R}$ (multiplicity 2)	$v^{(1)}, v^{(2)}$ ( $v^{(2)}$ generalised eigenvector)	$\left[v^{(1)}e^{\lambda t} \left(v^{(2)}+tv^{(1)}\right)e^{\lambda t}\right]$	Deficient Node
	$\lambda_1 = \lambda_2 = \lambda$ (2D eigenspace)	$v^{(1)}, v^{(2)}$ any basis of $\mathbb{R}^2$	$\left[\boldsymbol{v^{(1)}}e^{\lambda_1 t} \;\; \boldsymbol{v^{(2)}}e^{\lambda_2 t}\right]$	Star (or proper) Node
3: $\mathbf{B} = \begin{pmatrix} \alpha & -\omega \\ \omega & \alpha \end{pmatrix}$ $\alpha, \omega \in \mathbb{R}$	$\lambda_1 = i\beta = \overline{\lambda_2}$ $(\beta \neq 0) \in \mathbb{R}$	$v^{(2)} = \overline{v^{(1)}}$	$\left[\mathbb{R}\mathrm{e}\!\left(\pmb{v}^{(1)}e^{i\beta t}\right)\mathbb{I}\mathrm{m}\!\left(\pmb{v}^{(2)}e^{i\beta t}\right)\right]$	Centre (or vortex)
	$\lambda_1 = \alpha + i\beta = \overline{\lambda_2}$ $(\alpha \neq 0, \beta \neq 0) \in \mathbb{R}$	$v^{(2)} = \overline{v^{(1)}}$	$\left[\mathbb{R}\mathrm{e}\!\left(\boldsymbol{v}^{(1)}e^{(\alpha+i\beta)t}\right)\mathbb{I}\mathrm{m}\!\left(\boldsymbol{v}^{(2)}e^{(\alpha+i\beta)t}\right)\right]$	Spiral point (or focus)

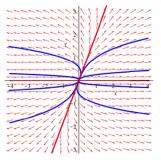
## Improper Node

In the case where  $0 < \lambda_1 < \lambda_2$ ,

- 1. Draw lines representing the eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$  of  $\lambda_1$  and  $\lambda_2$ .
- 2. Construct trajectories starting from the equilibrium point.
- 3. Initially, the trajectory is close to  $\mathbf{v}_1$ , but approaches the direction of  $\mathbf{v}_2$  quickly.
- 4. All trajectories are directed away from the equilibrium point.

When 
$$\lambda_1 = 1$$
,  $\mathbf{v}_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ ,  $\lambda_2 = 2$ ,  $\mathbf{v}_2 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ , solutions have the form

$$\mathbf{x} = Ae^t \begin{pmatrix} 1 \\ 2 \end{pmatrix} + Be^{2t} \begin{pmatrix} 1 \\ 0 \end{pmatrix}.$$



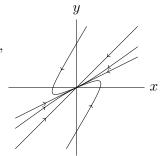
### Improper Node

In the case where  $\lambda_1 < \lambda_2 < 0$ ,

- 1. Draw lines representing the eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$  of  $\lambda_1$  and  $\lambda_2$ .
- 2. Construct trajectories starting far from the equilibrium point.
- 3. The trajectory starts close to the direction of  $\mathbf{v}_1$ , but approaches  $\mathbf{v}_2$  as it approaches the equilibrium point.

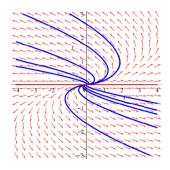
When 
$$\lambda_1 = -3$$
,  $\mathbf{v}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ ,  $\lambda_2 = -1$ ,  $\mathbf{v}_2 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ , solutions have the form

$$\mathbf{x} = Ae^{-3t} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + Be^{-t} \begin{pmatrix} 2 \\ 1 \end{pmatrix}.$$



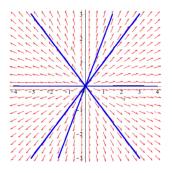
### Deficient Node

Here  $\lambda_1 = \lambda_2 \in \mathbb{R}$ , and the eigenspace is one dimensional (out of the scope of the course).



#### Star Node

Here  $\lambda_1 = \lambda_2 \neq 0$  and all nonzero vectors are eigenvectors. Therefore, all trajectories are either being attracted or repelled by the equilibrium points in a straight line.



### Saddle Point

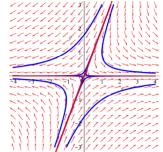
Here,  $\lambda_1 < 0 < \lambda_2$ .

- 1. Draw lines representing the eigenvectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$  of  $\lambda_1$  and  $\lambda_2$  respectively. Trajectories on  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are straight lines. Those on  $\mathbf{v}_1$  are attracted towards the equilibrium point, whereas those on  $\mathbf{v}_2$  are repelled.
- 2. Draw all other trajectories starting close to  $\mathbf{v}_1$ , initially approaching the equilibrium point. Such trajectories are eventually repelled from the equilibrium point, approaching  $\mathbf{v}_2$ .

When 
$$\lambda_1 = -1$$
,  $\mathbf{v}_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ ,  $\lambda_2 = 1$ ,

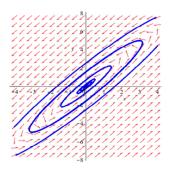
 $\mathbf{v}_2 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ , solutions have the form

$$\mathbf{x} = Ae^{-t} \begin{pmatrix} 1\\2 \end{pmatrix} + Be^t \begin{pmatrix} 1\\0 \end{pmatrix}$$



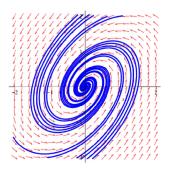
#### Centre

Here  $\lambda_1$ ,  $\lambda_2$  are purely imaginary. If eigenvalues are purely imaginary, the trajectory forms orbits around the equilibrium point as  $e^{it} = \cos(t) + i\sin(t)$ . We can determine whether the orbit is clockwise or anticlockwise by finding the value of  $d\mathbf{x}/dt$  at a point.



### Spiral

Here  $\lambda_1 = \alpha + i\beta = \bar{\lambda_2}$ . The trajectory will spiral inward when  $\alpha < 0$  or outward when  $\alpha > 0$ . We can determine whether the spiral is is clockwise or anticlockwise by finding the value of  $d\mathbf{x}/dt$  at a point.



## Equilibrium points

#### Example 13: MATH2121 2018 T2 2.iii)

Solve for x(t) and y(t) and determine the type and stability of the equilibrium point of the following system of differential equations:

$$\begin{aligned} \frac{dx}{dt} &= x + y; \\ \frac{dy}{dt} &= 2x. \end{aligned}$$

The system can be written as

$$\frac{d}{dt} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 2 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}.$$

## Equilibrium points

Let  $A = \begin{pmatrix} 1 & 1 \\ 2 & 0 \end{pmatrix}$ . By solving  $\det(A - \lambda I) = 0$ , we find that the eigenvalues of A are  $\lambda_1 = -1$  and  $\lambda_2 = 2$ . The eigenvectors are given by  $\mathbf{v}_1 = \begin{pmatrix} 1 \\ -2 \end{pmatrix}$  and  $\mathbf{v}_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .

Hence, the solutions to the differential equation can be written in the form

$$\begin{pmatrix} x(t) \\ y(t) \end{pmatrix} = c_1 e^{-t} \begin{pmatrix} 1 \\ -2 \end{pmatrix} + c_2 e^{2t} \begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

By equating the components, we get  $x(t) = c_1 e^{-t} + c_2 e^{2t}$  and  $y(t) = -2c_1 e^{-t} + c_2 e^{2t}$ .

To find the equilibrium point, we solve x + y = 0 and 2x = 0 simultaneously. The only solution is the point (0,0). Since  $\lambda_1 < 0 < \lambda_2$ , the equilibrium point (0,0) is a saddle point. Since  $\text{Re}(\lambda_2) > 0$  then (0,0) is an unstable equilibrium point. Hence (0,0) is an unstable saddle point.

# First integrals (2221 only)

#### Definition 17

A function  $G: \mathbb{R}^n \to \mathbb{R}$  is a **first integral** for a system of ODEs

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x})$$

if  $G(\mathbf{x}(t))$  is constant for every solution  $\mathbf{x}(t)$ .

Geometrically: G is a first integral iff

$$\nabla G(\mathbf{x}) \perp \mathbf{F}(\mathbf{x})$$
 for all  $\mathbf{x}$ .

# First integrals (2221 only)

#### Example 14

Consider the system of ODEs

$$\frac{dx_1}{dt} = x_1 x_2,$$
$$\frac{dx_2}{dt} = -x_1^2.$$

Prove that  $G(\mathbf{x}) = x_1^2 + x_2^2$  is a first integral.

Set the function  $\mathbf{F}(\mathbf{x})$  as the RHS of the system:

$$\mathbf{F}(\mathbf{x}) = \begin{pmatrix} x_1 x_2 \\ -x_1^2 \end{pmatrix}.$$

## First integrals (2221 only)

We want to find the gradient of G:

$$\nabla G(\mathbf{x}) = \begin{pmatrix} 2x_1 \\ 2x_2 \end{pmatrix}.$$

So

$$\nabla G(\mathbf{x}) \cdot \mathbf{F}(\mathbf{x}) = \begin{pmatrix} 2x_1 \\ 2x_2 \end{pmatrix} \cdot \begin{pmatrix} x_1 x_2 \\ -x_1^2 \end{pmatrix} = 2x_1^2 x_2 - 2x_2 x_1^2 = 0.$$

This means that all solutions to the system of ODEs must be mapped to a constant under G. That is, if  $(x_1, x_2)$  is a solution then  $x_1^2 + x_2^2 = C$  for some C. In other words, all the solutions to the system of ODEs lie on some circle around the origin.