Differential Equations

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Introduction

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1 2/3/16: Background on \mathbb{R} ; Basic Existence Question of ODE's

1.1 Romeo and Juliet

$$\begin{cases} R' = aR + bJ \\ J' = cR + dJ \end{cases}$$

These equations model the rate of change of Romeo's and Juliet's feelings. We call this a linear system of two coupled differential equations of first order in two unknowns.

- What makes it linear is that the functions and variables appear in a linear fashion.
- What makes it coupled is that both equations have both R and J in them.
- An **uncoupled system** would look like:

$$\begin{cases} R' = aR \\ J' = bJ \end{cases}$$

• First-order refers to the fact that all the derivatives are the first derivatives.

"Identically cautious lovers":

$$R' = aR + bJ \quad a < 0, b > 0$$
$$J' = bR + aJ \quad |a| > |b|$$

We may have initial conditions, R(0) and J(0), and plot them on a **phase plane** with R against J. In this case, no matter where the starting point is, the trajectory will go towards a **stable node**.

In the case of |a| < |b|, points will move asymptotically towards R = -J and R = J. In the case of |a| = |b|, points will cycle around the origin infinitely.



1.2 Supremum and Infimum of a Set $A \subseteq \mathbb{R}$

• If $A \in (-\infty, b]$ for some $b \in \mathbb{R}$, we say A is bounded above, and that b is an **upper** bound for A.

Theorem 1.1 (Supremum Theorem). If $A \in \mathbb{R}$, $A \neq \emptyset$, and $A \subseteq (-\infty, b]$ for some $b \in \mathbb{R}$, then there exists $a \in \mathbb{R}$ such that $A \subseteq (-\infty, a]$ but if x < a, then $A \not\subseteq (-\infty, x]$. We write $a = \sup A$, call it the **supremum** of A.

Why is this necessary? Consider the set $\mathcal{A} = \{-\frac{1}{n} | n \in \mathbb{N}\}$. It does not have a maximum persay, but it has a supremum sup $\mathcal{A} = 0$.

Consider this example: What is $\sup (-\mathbb{N})$? It is -1, which also happens to be the maximum of the set. e

Theorem 1.2. If max A exists as a real number, then $\sup A = \max A$.

But to answer all these questions, we need to figure out: what exactly are the real numbers?

1.3 What is \mathbb{R} ?

Let $x = (s, N, d_1, d_2, d_3, \dots, d_k, \dots)$, where:

- $s \in \{+1, -1\}$
- $N \in \mathbb{Z}$
- $d_k \in \mathbb{D} = \{0, 1, 2, 3, 4, 5, 6, 7, 8, 9\}$
- $\neg(\exists k: d_{k+1} = d_{k+2} = \cdots = 0)$, this is to prevent multiple sequences from being the same number

In this case, "2.49" is shorthand for (+1, 2, 4, 8, 9, 9, 9, ...)

2 2/4/16: Background in \mathbb{R} ; Fundamental Existence/U-niqueness Question

2.1 Supremums and Infimums in Integrals

Theorem 2.1 (Supremum/Infimum Theorem).

- 1. If \mathcal{A} is a non-empty set of \mathbb{R} , and is bounded above (i.e. $\mathcal{A} \subseteq (-\infty, b]$ for some $b \in \mathbb{R}$), then there is a least upper bound for \mathcal{A} , namely $a \in \mathbb{R}$ such that
 - (a) $A \subseteq (-\infty, a]$
 - (b) if x < a, then $\mathcal{A} \nsubseteq (-\infty, x]$

This a is called the called the **supremum** of A, written sup A.

2. inf A. This is the <u>greatest lower bound</u> for A, or the **infimum**, provided $A \neq \emptyset$ and A has a lower bound at all.

Recall that the Riemann integral is taking the limit of a partition over an interval [a, b]. But when we take the limit, we make the mesh of the partition, $||\mathcal{P}||$, approach zero, where

$$\mathcal{P} = \max_{1 \le i \le n} \Delta x_i$$

To fix this, we can define:

$$\underline{\int}_{a}^{b} f(x) \ dx = \sup \left\{ \sum_{i=1}^{n} [\inf \{ f(x) \mid x_{i-1} \le x \le x_i \} \Delta x_i] \ \middle| \ a = x_0 < x_1 < \dots < x_n = b \right\}$$

This is a "down-and-up" procedure. The sum of the rectangle areas is a down approximation since we use the minimum possible height to find the area. Then, we take the supremum of that, since for any lower approximation there will always be a higher approximation. Turns out there will never be a maximum; that's why we take the supremum. This is a **lower Riemann sum**.

We can also define the same thing for an **upper Riemann sum**:

$$\int_{a}^{b} f(x) \ dx = \inf \left\{ \sum_{i=1}^{n} \left[\sup \{ f(x) \mid x_{i-1} \le x \le x_i \} \Delta x_i \right] \mid a = x_0 < x_1 < \dots < x_n = b \right\}$$

Therefore, the following inequality is true:

$$\int_{a}^{b} f \le \int_{a}^{\overline{b}} f$$

If these two are equal, then we say that f is **Riemann integrable**.

Here's an example of a function that is NOT Riemann integrable:

$$f(x) = \begin{cases} 0 \text{ if } x \in \mathbb{Q} \cap [0, 1] \\ 1 \text{ if } x \in [0, 1] \setminus \mathbb{Q} \end{cases}$$

Note that $\underline{\underline{f}}_0^1 f = 0$ and $\overline{\underline{f}}_0^1 f = 1$, so this is not Riemann integrable.

2.2 Real Numbers, Again

We have shorthand for our previous definition of the real numbers.

$$\mathbb{R} = \{0\} \cup \{(s, N, d_1, d_2, \dots, d_k, \dots \mid s \in \{-1, +1\}, N \in \mathbb{Z}^+, d_k \in \mathbb{D}, \text{no 0-tail}\}\$$

and the positive reals:

$$\mathbb{R}^+ = \{(s, N, d_1, d_2, \dots) \mid s = +1\}$$

Let us write $x = \underline{N}.d_1d_2d_3...$ and $y = \underline{M}.e_1e_2e_3...$

We also define negation as:

$$-(s, N, d_1, d_2, \dots) := (-s, N, d_1, d_2, \dots)$$

Then we can define the "less than" operation as follows:

- If $x, y \in \mathbb{R}^+$, then x < y if either N < M or N = M and $d_1 < e_1$ or N = M, $d_1 = e_1$ and $d_2 < e_2$, or...
- $0 < x \text{ if } x \in \mathbb{R}^+$
- x < 0 if $x \in \mathbb{R}^+$
- x < y if $x \in \mathbb{R}^-, y \in \mathbb{R}^+$.
- $x, y \in \mathbb{R}^-$, and x < y if -y < -x

3 2/5/16: Fundamental Existence of Uniqueness Theorem

3.1 Terminology

A differential equation is a relation between one or more unknown functions and at least some (but finitely many) of their derivatives, plus the independent variables.

Examples:

$$y' + 2xy - x^{2} = 3$$
$$y''' + 2x^{2}y'' - 3x^{3}y' + xy - x^{5} + 1 = 0$$
$$(y')^{y''} - e^{y'''} + x = 0$$

Or,

$$\vec{y}' = \mathbf{A}(x)\vec{y}$$

where

$$\vec{y} = \vec{y}(x) = \begin{bmatrix} y_1(x) \\ y_2(x) \\ \vdots \\ y_n(x) \end{bmatrix}$$

$$\mathbf{A}(x) = \begin{bmatrix} a_{11}(x) & a_{12}(x) & \cdots & a_{1n}(x) \\ a_{21}(x) & a_{22}(x) & \cdots & a_{2n}(x) \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1}(x) & a_{n2}(x) & \cdots & a_{nn}(x) \end{bmatrix}$$

3.2 A Treatise on PDE's

There are two different types of differential equations: ODE's (ordinary, where all unknown functions depend on a single, same independent variable) and PDE's (partial, anything else).

Wave equation:
$$\frac{\partial^2 u}{\partial x^2} = c^2 \frac{\partial^2 u}{\partial t^2}$$

 $u = g(x-t) + h(x+t)$

4 2/9/16: Basic Existence and Uniqueness Theorem

Theorem 4.1 (Flow Theorem). Let $\vec{F}(\vec{x}) = (F_1(\vec{x}), F_2(\vec{x}), \dots, F_n(\vec{x}))$ be a vector field defined on some closed and bounded region $\mathcal{D} \subseteq \mathbb{R}^n$. Also assume \vec{F} is C^1 ; namely, $\frac{\partial F_i}{\partial x_j}$ is continuous everywhere interior to D, for any i and j.

Let \vec{p} be a specific point interior to \mathcal{D} . Then \exists a function $\vec{\sigma}(t)$ from some "time" interval $(-\varepsilon, \varepsilon)$ with $\varepsilon > 0$ into \mathcal{D} , such that $\vec{\sigma}(0) = \vec{p}$ and $\vec{\sigma}'(t) = \vec{F}(\vec{\sigma}(t))$ for any $t \in (-\varepsilon, \varepsilon)$.

This theorem basically says that in a vector field, we can use the vector field to get the velocity of a curve. We call $\vec{\sigma}(t)$ a **flow** for \vec{F} , starting at \vec{p} . This flow is, in fact, unique, in the sense that any two flows for the same \vec{F} starting at the same point must agree whenever they are both defined.

This is meaningful in that we can treat it as a differential equation:

$$\begin{cases}
\sigma'_1 &= F_1(\sigma_1, \sigma_2, \dots, \sigma_n) \\
\sigma'_2 &= F_2(\sigma_1, \sigma_2, \dots, \sigma_n) \\
\vdots &\vdots \\
\sigma'_n &= F_n(\sigma_1, \sigma_2, \dots, \sigma_n)
\end{cases}$$

$$\begin{cases}
\sigma_1(0) &= p_1 \\
\sigma_2(0) &= p_2 \\
\vdots &\vdots \\
\sigma_n(0) &= p_n
\end{cases}$$

4.1 Second-Order

$$mx'' = -kx$$
, $x(0) = x_0$, $x'(0) = v_0$
 $x = x(t)$, $v = v(t) = x'(t)$, $a = a(t) = x''(t)$

where k > 0 is the spring constant. We can rewrite this as:

$$\begin{cases} x' = v = F_1(x, v) \\ v' = -\frac{k}{m}x = F_2(x, v) \end{cases} \text{ and } \begin{cases} x(0) = x_0 \\ v(0) = v_0 \end{cases}$$

The Flow Theorem will tell us there is a unique solution, for some time interval.

5 2/10/16: The Flow Theorem

5.1 Application: n^{th} order initial value problem (IVP)

$$\begin{cases} x &= x(t) \\ x^{(n)} &= F(t, x, x', x'', \dots, x^{(n-1)}) \\ x(t_0) &= x_{00} \\ x'(t_0) &= x_{10} \\ x''(t_0) &= x_{20} \\ \vdots \\ x^{(n-1)}(t_0) &= x_{(n-1)0} \end{cases}$$

 $f(t)x^{(n)} = F(t, x, x', x'', \dots, x^{(n-1)})$ is an n^{th} order ODE in standard form.

A **singularity** (or singular point) of this equation is a value t_0 where $f(t_0) = 0$. At this point, the equation ceases to be of n^{th} order. If f(t) is of constant sign in the time interval on which we'd like to solve the equation, we just divide through by f(t) to get our desired form (which is the regular case, as opposed to the singular case).

Here, the Flow Theorem says that there is a unique solution x = x(t) defined in some time interval $(t_0 - \varepsilon, t_0 + \varepsilon)$ where $\varepsilon > 0$.

To apply this:

$$\begin{cases} x_0(t) &= t \\ x_1 &= x_1(t) = x(t) \\ x_2 &= x_2(t) = x'(t) \\ x_3 &= x_3(t) = x''(t) \\ \vdots &&\\ x_n &= x_n(t) = x^{(n-1)}(t) \end{cases}$$

becomes

$$\begin{cases} x'_0 &= 1 = F_0(x_0, x_1, x_2, \dots, x_n) \\ x'_1 &= x_2 = F_1(x_0, x_1, x_2, \dots, x_n) \\ x'_2 &= x_3 = F_2(x_0, x_1, x_2, \dots, x_n) \\ x'_3 &= x_4 = F_3(x_0, x_1, x_2, \dots, x_n) \\ \vdots &\vdots \\ x'_{n-1} &= x_n = F_{n-1}(x_0, x_1, x_2, \dots, x_n) \\ x'_n &= F(t, x_1, x_2, \dots, x_n) = F_n(\dots) \end{cases}$$
 and
$$\begin{cases} x_0(t_0) &= t_0 \\ x_1(t_0) &= x_{00} \\ x_2(t_0) &= x_{10} \\ \vdots &\vdots \\ x_n(t_0) &= x_{(n-1)0} \end{cases}$$

This shows that we can recast an n^{th} order IVP into an n+1 order system.

However, for the Flow Theorem to apply, \vec{F} needs to be C^1 . Therefore, our hypothesis in the IVP is that F is C^1 , meaning that $\frac{\partial F}{\partial t}, \frac{\partial F}{\partial x}, \frac{\partial F}{\partial x'}, \dots$ are continuous.

6 2/11/16: Proof of the Flow Theorem

We will prove the Flow Theorem for two dimensions only; the proof can be extended to greater than two dimensions.

Proof. Let $\vec{F}(x,y) = (A(x,y), B(x,y))$ be a vector field. By hypothesis, A and B are defined on a closed, bounded region \mathcal{D} , and they are C^1 on \mathcal{D} . Then we need to solve the following equation:

$$\vec{x}' = \vec{F}(\vec{x})$$

 $\vec{x}(0) = \vec{p} = \langle p, q \rangle$

We need to see how fast A(x, y) is changing.

1. $|A(x_1, y_1) - A(x_2, y_2)| = |A(x_1, y_1) - A(x_1, y_2) + A(x_1, y_2) - A(x_2, y_2)|$ (Triangle Inequality) $\leq |A(x_1, y_1) - A(x_1, y_2)| + |A(x_1, y_2) - A(x_2, y_2)|$ $(MVT) \leq \left| \frac{\partial A}{\partial y}(x_1, y^*)(y_1 - y_2) \right| + \left| \frac{\partial A}{\partial x}(x^*, y_2)(x_1 - x_2) \right|$

Take K to be some upper bound for all the partial derivatives of A and B on \mathcal{D} .

$$\left| \frac{\partial A}{\partial y}(x_1, y^*)(y_1 - y_2) \right| + \left| \frac{\partial A}{\partial x}(x^*, y_2)(x_1 - x_2) \right| \le K(|x_1 - x_2| + |y_1 - y_2|)$$

Similarly:

$$|B(x_1, y_2) - B(x_2, y_2)| \le K(|x_1 - x_2| + |y_1 - y_2|)$$

This is called the **Lipschitz Condition**.

2. Also, note that A and B are continuous in \mathcal{D} and so by the Extreme Value Theorem, we can find an upper bound M for |A| and |B| on \mathcal{D} , i.e.

$$M = \max\left(\max_{(x,y)\in\mathcal{D}} |A(x,y)|, \max_{(x,y)\in\mathcal{D}} |B(x,y)|\right)$$

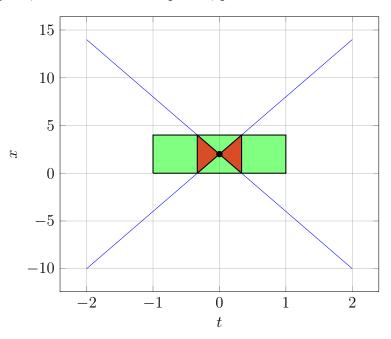
3. The point (p,q) is assumed to be interior to \mathcal{D} (not on the boundary).

We can therefore encase the point (0,p) in a rectangle in the tx-plane defined by $R: [-r,r] \times [p-s,p+s] \subseteq \operatorname{proj}_x \mathcal{D}, r,s>0$. Draw two lines with slopes M and -M through the point. We will consider the "bowtie" region formed by the intersections of the lines with the rectangle, joining them oppositely, and the lines themselves. Call the x-intersections -h and h.

Define $h := \min(r, \frac{s}{M}) > 0$. This is to formally define the bowtie region and consider the two possible pictures depending on the size of M.

Now, we want to construct the solution of the differential equation within the bowtie region.

Whoopsies, there was a screwup here, proof to be fixed in the future.



7 2/12/16: Separable and First-Order Linear Equations

7.1 Multiplicatively Separable Functions

$$F(t,x) = f(t)g(x)$$

A non-example of a separable function is $F(t,x) = t^2 + x^2$. An example is $F(t,x) = t^2x^3$.

For our purposes, we will work with first-order ODE's with scalar functions.

7.2 Separable ODE

$$x' = f(t)g(x)$$

There are other ways we can write this equation:

- General Form: G(t, x, x') = 0
- Standard Form: $\phi(t)x' = F(t,x)$
 - Regular Case: x' = F(t, x), F is the "slope function"
 - Singular Case: This is when we solve in an interval $(t_0 \delta, t_0 + \delta)$ where $\delta > 0$ and $\phi(t_0) = 0$.

To solve this type of equation:

$$x'(t) \equiv f(t)g(x(t))$$
$$\frac{x'(t)}{g(x(t))} = f(t)$$
$$\int_{a}^{t} \frac{x'(\tau)}{g(x(\tau))} d\tau = \int_{a}^{t} f(\tau) d\tau$$

Letting $u = x(\tau)$ and $du = x'(\tau) d\tau$:

$$\underbrace{\int_{x(a)}^{x(t)} \frac{du}{g(u)}}_{G(x(t))} = \underbrace{\int_{a}^{t} f(\tau) d\tau}_{F(t)}$$

$$\boxed{G(x(t)) = F(t) + C}$$

7.3 Example

$$x' = t^{2}x^{3}$$

$$\frac{x'}{x^{3}} = t^{2}$$

$$\int \frac{dx}{x^{3}} = \int t^{2} dt + C$$

$$\frac{x^{-2}}{-2} = \frac{t^{3}}{3} + C$$

$$x^{-2} = C - \frac{2}{3}t^{3}$$

$$x = \pm \frac{1}{\sqrt{C - \frac{2}{3}t^{3}}}$$