

Ordinary differential equations

Jasper Ty

What is this?

These are notes I am taking for the class MATH-623, *Ordinary Differential Equations*, at Drexel University, taught by Yixin Guo.

Some notation is changed from her notes, and I try to add as many missing details from proofs as possible.

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I Basic theory

I.1 Definitions

Definition 1.1.1. Let $J \subseteq \mathbb{R}$, $U \subseteq \mathbb{R}^n$, $\Lambda \subseteq \mathbb{R}^k$ be open sets, and let $\mathbf{f} : J \times U \times \Lambda \rightarrow \mathbb{R}^n$ is a smooth function. An **ordinary differential equation** is an equation of the form

$$\dot{\mathbf{x}} = \mathbf{f}(t, \mathbf{x}, \underline{\lambda}) \quad (\text{ODE})$$

where the dot denotes differentiation with respect to the independent variable t .

Morally, the individual parts of an ODE have the following meaning:

$t \in J$: an independent variable, typically time,

$\mathbf{x} \in U$: a dependent variable,

$\underline{\lambda} \in \Lambda$: a vector of parameters,

\mathbf{f} : a continuously differentiable function that encodes the (time) evolution of \mathbf{x} .

Definition 1.1.2. A **solution** of (ODE) is a function $\mathbf{F} : J_0 \rightarrow U$, where $J_0 \subseteq J \subseteq \mathbb{R}$, such that

$$\frac{d}{dt}\mathbf{F}(t) = \mathbf{f}(t, \mathbf{F}(t), \underline{\lambda}), \quad \forall t \in J_0$$

i.e, a function for which we can put $\mathbf{x} = \mathbf{F}(t)$ in (ODE) for all $t \in J_0$.

The **orbit** of the solution \mathbf{F} is the set

$$\{\mathbf{F}(t) \in U : t \in J_0\} \subseteq \mathbb{R}^n.$$

This is also called the **trajectory**, **integral curve**, or **solution curve**.

Evidently, an ODE can have many solutions. For example, if $J = U = \Lambda = \mathbb{R}$ and $f(t, x, \lambda) = \lambda x$, then it's well known that $F(t) = Ce^{\lambda t}$ is a solution to this ODE for all $C \in \mathbb{R}$.

Fortunately, both real-life and abstract experience tells us that in many cases, the following is a natural way to restrict solutions to an ODE.

Definition 1.1.3. An **initial-value problem** (IVP) is the system of equations

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{f}(t, \mathbf{x}, \underline{\lambda}), \\ \mathbf{x}(t_0) = \mathbf{x}_0. \end{cases} \quad (\text{IVP})$$

Namely, it is an (ODE) with additional data $(t_0, \mathbf{x}_0) \in J \times U$, encoding an **initial value** constraint.

A **solution** of (IVP) is, again a function $\mathbf{F} : J_0 \rightarrow U$ that solves the underlying ODE, but now subject to the condition that $\mathbf{F}(t_0) = \mathbf{x}_0$.

Example 1.1.4. The **forced Van der Pol** equation is defined to be

$$\begin{cases} \dot{x}_1 = x_2, \\ \dot{x}_2 = b(1 - x_1^2)x_2 - \omega^2 x_1 + a \cos \Omega t. \end{cases} \quad (\text{fVdP})$$

1.2 Existence and uniqueness

Let $J, U \subseteq \mathbb{R}$, and consider the IVP

$$\frac{dy}{dt} = f(t, y), \quad y(t_0) = y_0 \quad (\text{IVP-1D})$$

where $f : J \rightarrow U$ is continuously differentiable in t on some open interval containing t_0 .

Supposing we can integrate (IVP-1D) from t_0 to a given point t ,

$$\begin{aligned} \int_{t_0}^t \frac{dy}{d\tau} d\tau &= \int_{t_0}^t f(\tau, y) d\tau, \\ y(t) - y(t_0) &= \int_{t_0}^t f(\tau, y) d\tau, \\ y(t) &= y(t_0) + \int_{t_0}^t f(\tau, y) d\tau, \\ y(t) &= y_0 + \int_{t_0}^t f(\tau, y) d\tau. \end{aligned} \quad (\int \text{IVP-1D})$$

If F satisfies (\int IVP-1D), it satisfies (IVP-1D) and vice versa, and this is easily seen using the fundamental theorem of calculus.

Definition 1.2.1. The **Picard iterates** y_i , given the data for (IVP-1D), are defined recursively as follows:

$$\begin{cases} y_0(t) := y_0, \\ y_{i+1}(t) := y_0 + \int_{t_0}^t f(\tau, y_i(\tau)) d\tau. \end{cases} \quad (\text{Picard})$$

We have a weak but straightforward estimate on the y_i .

Lemma 1.2.2. For all $a > 0, b > 0$, define R to be the rectangle $[t_0, t_0 + a] \times [y_0 - b, y_0 + b]$. Then if we define

$$M := \max_{(t,y) \in R} |f(t,y)|, \quad \alpha := \min \left\{ a, \frac{b}{M} \right\}.$$

Then

$$|y_n(t) - y_0| \leq M(t - t_0) \text{ for all } t_0 \leq t \leq t_0 + \alpha \quad (1)$$

for all n .

Proof. We prove this by induction.

We note that if (1) holds for n , then for all $t_0 \leq t \leq \alpha$,

$$\begin{aligned} |y_n(t) - y_0| &\leq M(t - t_0) \\ &\leq M\alpha \\ &= M \cdot \min \left\{ a, \frac{b}{M} \right\} \\ &= \min \left\{ \frac{a}{M}, b \right\} \\ &\leq b. \end{aligned}$$

Hence

$$|f(t, y_n(t))| \leq M. \quad (2)$$

$n = 0$: We have that $|y_0(t) - y_0| = 0 \leq M(t - t_0)$ trivially for all $t \geq t_0$.

$n > 0$: Suppose that $|y_n(t) - y_0| \leq M(t - t_0)$ for all $t_0 \leq t \leq t_0 + \alpha$. Then

$$\begin{aligned} |y_{n+1}(t) - y_0| &= \left| \int_{t_0}^t f(\tau, y_n(\tau)) d\tau \right| \\ &\leq \int_{t_0}^t \underbrace{|f(\tau, y_n(\tau))|}_{\text{use (2)}} d\tau \\ &\leq \int_{t_0}^t M d\tau \\ &= M(t - t_0) \end{aligned}$$

for all $t_0 \leq t \leq t_0 + \alpha$. This completes the proof.

□

Next we show that the the Picard iterates y_n converge.

Theorem 1.2.3. Let $f(t, y)$ be continuously differentiable in both t and y . Then, on $[t_0, t_0 + \alpha]$, the Picard iterates y_n converge pointwise to a function y that satisfies (IVP-1D).

Proof. We have that

$$\begin{aligned} y_n(t) &= y_0(t) + [y_1(t) - y_0] + \cdots + [y_n(t) - y_{n-1}(t)] \\ &= y_0(t) + \sum_{i=1}^n y_i(t) - y_{i-1}(t). \end{aligned}$$

So, $y_n(t)$ converges if and only if $\sum_{i=1}^n y_i(t) - y_{i-1}(t)$ converges.

It suffices to prove that $\sum_{i=1}^n |y_i(t) - y_{i-1}(t)|$ converges.

We compute that

$$\begin{aligned} |y_i(t) - y_{i-1}(t)| &= \left| \left[y_0 + \int_{t_0}^t f(\tau, y_{i-1}(\tau)) d\tau \right] - \left[y_0 + \int_{t_0}^t f(\tau, y_{i-2}(\tau)) d\tau \right] \right| \\ &= \left| \int_{t_0}^t f(\tau, y_{i-1}(\tau)) - f(\tau, y_{i-2}(\tau)) d\tau \right| \\ &\leq \int_{t_0}^t \underbrace{|f(\tau, y_{i-1}(\tau)) - f(\tau, y_{i-2}(\tau))|}_{\text{apply mean value theorem}} d\tau \\ &= \int_{t_0}^t \left| \frac{\partial f(\tau, c(\tau))}{\partial y} \right| |y_{i-1}(\tau) - y_{i-2}(\tau)| d\tau \\ &\leq P \int_{t_0}^t |y_{i-1}(\tau) - y_{i-2}(\tau)| d\tau. \end{aligned}$$

Now, we can show inductively that

$$|y_i(t) - y_{i-1}(t)| \leq M P^{i-1} \frac{(t - t_0)^i}{i!} \quad (3)$$

for all $i \geq 1$.

The base case $i = 1$ ends up being (1)

$$|y_1(t) - y_0(t)| = |y_1(t) - y_0| \leq M(t - t_0).$$

For the induction step, suppose it were true for i , then

$$\begin{aligned}
 |y_{i+1}(t) - y_i(t)| &\leq P \int_{t_0}^t \underbrace{|y_{i-1}(\tau) - y_i(\tau)|}_{\text{apply induction hypothesis}} d\tau \\
 &\leq P \int_{t_0}^t \left[M P^{i-1} \frac{(\tau - t_0)^i}{i!} \right] d\tau \\
 &= M P^i \int_{t_0}^t \frac{(t - t_0)^i}{i!} d\tau \\
 &= M P^i \frac{(t - t_0)^{i+1}}{(i+1)!}.
 \end{aligned}$$

Hence, (3) holds for all i .

So

$$\begin{aligned}
 \sum_{i=1}^n |y_i(t) - y_{i-1}(t)| &\leq \sum_{i=1}^n M P^{i-1} \frac{(t - t_0)^i}{i!} \\
 &= M P^{-1} \left(\sum_{i=1}^n P^i \frac{(t - t_0)^i}{i!} \right) \\
 &= M P^{-1} \left(-1 + \sum_{i=0}^n P^i \frac{(t - t_0)^i}{i!} \right) \\
 &= M \frac{e^{P(t-t_0)} - 1}{P},
 \end{aligned}$$

which is a finite number.

So, we have shown that $y_n(t) \rightarrow y(t)$ for some function y .

Next, we show that y satisfies (IVP-1D).

□