

# Math 302: Existence and uniqueness of solutions

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- When, in generality, can we say that a differential equation has solutions? When are solutions uniquely identified?

I won't go into every gritty detail. Rather, the goal is to give you a view of the broad ideas, and some techniques and tricks that are used to reason abstractly about differential equations.

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We've turned our problem into a new problem. That's kinda like progress!

# Solving systems of equations

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- $n$  equations
- about  $n$  unknown functions  $y_1, \dots, y_n$
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If we want to think about how to solve this, it makes sense to start with the simplest case, where we have only one unknown function.



# Solving $y' = F(x, y)$

- Suppose for now that  $F$  is continuous in a rectangle  $R$ :  $a_0 \leq x \leq a_1$  and  $b_0 \leq y \leq b_1$ .
- (We'll see later that we actually have to assume a little bit more about  $F$ .)
- Fix a point  $(x_0, y_0)$  in this rectangle, and suppose we have the initial condition  $y(x_0) = y_0$ .

# Solving $y' = F(x, y)$

**A trick:** Turn the differential equation into an integral equation.

Integrate both sides, starting at  $x_0$ , get:

$$y(x) - y(x_0) = \int_{x_0}^x F(t, y(t)) dt.$$

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- Derivatives are easier to compute than integrals.
- But for theoretical uses, integrals are better behaved.
- Since we want a theoretical result rather than calculations, this makes things easier.

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**Spoiler:** This stupid idea will work out (with an extra assumption on  $F$ ). The error gets smaller and smaller, so these **Picard approximations** converge to the true solution.

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- We turned the first-order differential equation  $y' = F(x, y)$  into an integral equation.
- We want to use [Picard's method](#) to find a solution to this integral equation.

# Picard's method

$$y(x) = y_0 + \int_{x_0}^x F(t, y(t)) dt$$

We want to solve this equation. The idea is to get the solution from a sequence of better and better approximations:

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Explaining just why this works will take some time.

Let's start by talking about convergence of sequences of functions.

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For example:

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  - The functions  $1, x^2, x^3, \dots$  are continuous on  $[0, 1]$  and converge (nonuniformly) to the discontinuous function which is 0 when  $x < 1$  and 1 when  $x = 1$ .

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A lot of things you might hope convergence is enough for actually need uniform convergence.

# Uniform convergence and integration

Suppose the functions  $y_0, y_1, \dots$  converge uniformly to  $y$  on a domain which includes the interval  $[a, b]$ .

- Then, you can swap the order of integration and taking a limit:

$$\lim_{n \rightarrow \infty} \int_a^b y_n(x) \, dx = \int_a^b \underbrace{\lim_{n \rightarrow \infty} y_n(x)}_{=y(x)} \, dx.$$

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We want to apply this to our sequence of Picard approximations:

$$\begin{aligned} \lim_{n \rightarrow \infty} y_n(x) &= \lim_{n \rightarrow \infty} \left( y_0 + \int_{x_0}^x F(t, y_{n-1}(t)) dt \right) \\ &= y_0 + \int_{x_0}^x \lim_{n \rightarrow \infty} F(t, y_{n-1}(t)) dt \end{aligned}$$

# Uniform convergence and integration

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To be able to bring the limit inside the integral we need that the sequence

$$F(t, y_0(t)), F(t, y_1(t)), \dots, F(t, y_n(t)), \dots$$

converges uniformly in the interval  $[x_0, x]$ .



# One way we might get convergence

Maybe we can show that the differences  $y_{n+1} - y_n$  get uniformly smaller, so that the telescoping series

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Let's look at this and think about what we might need.

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That's a start. We got the first step. But later ones are gonna be harder...

# Why is it harder?

We want to compute a bound on  $|y_{n+1} - y_n|$ .

Let's start

$$\begin{aligned} |y_{n+1} - y_n| &= \left| \int_{x_0}^x F(t, y_{n+1}(t)) - F(t, y_n(t)) dt \right| \\ &\leq \int_{x_0}^x |F(t, y_{n+1}(t)) - F(t, y_n(t))| dt \end{aligned}$$

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**The trouble:** If  $F$  changes values really quickly, we can't get a better bound on the distance between  $F(x_0, y_0)$  and  $F(x_1, y_1)$  for two points. So we need to assume more about  $F$ .

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A condition on  $F$  that makes this work is to assume that

$$|F(t, y_a) - F(t, y_b)| \leq L |y_a - y_b|$$

for some fixed bound  $L$  which works for all points  $y_a, y_b$ . This condition says that  $F$  can change, but it cannot change too quickly.

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So the sequence  $y_0, y_1, \dots$  of functions converges uniformly, call the limit  $y$ .

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We want to check that the limit  $y$  of the Picard approximations satisfies the integral equation

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Knowing that  $y_n \rightarrow y$  uniformly, we can check:

$$\begin{aligned} \lim_{n \rightarrow \infty} y_n(x) &= \lim_{n \rightarrow \infty} \left( y_0 + \int_{x_0}^x F(t, y_{n-1}(t)) dt \right) \\ &= y_0 + \int_{x_0}^x \lim_{n \rightarrow \infty} F(t, y_{n-1}(t)) dt \\ &= y_0 + \int_{x_0}^x F(t, \lim_{n \rightarrow \infty} y_{n-1}(t)) dt \end{aligned}$$

Using the Lipschitz condition on  $F$ , we can say that this last step is valid, we can uniformly bring the limit inside  $F$ .

# Summing things up

## Theorem (Picard–Lindelöf theorem)

Consider the differential equation

$$y' = F(x, y),$$

where  $F$  satisfies a certain Lipschitz condition on a rectangle  $R$ , and consider a point  $(x_0, y_0)$  in  $R$ . Then there is a unique solution to this equation, valid on the a subdomain of the rectangle, satisfying the initial condition  $y(x_0) = y_0$ , where  $(x_0, y_0)$  is a point in  $R$ .

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What about with  $n$  equations?

$$y'_1 = F_1(x, y_1, \dots, y_n)$$

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So we get solutions to linear differential equations of order  $> 1$ .

## Coda: Another existence theorem

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### Theorem (Peano existence theorem)

*Consider the differential equation  $y' = F(x, y)$  where  $F$  is continuous on a rectangle  $R$ , and consider a point  $(x_0, y_0)$  in  $R$ . Then there is a solution—not necessarily unique—to this equation, valid on a subdomain of the rectangle, satisfying  $y(x_0) = y_0$ .*

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That is, you can drop the assumption, but it comes at the cost of uniqueness.

For example, the equation

$$y' = \sqrt{|y|}$$

has multiple different solutions satisfying  $y(0) = 0$ :

$$y_1 = 0$$

$$y_2 = \frac{x^2}{4}$$