

- Differential Equations
 - Why we end up with these a lot in physics
 - Ordinary differential equations

ODEs

- $\frac{dy}{dt} = f(t)$

Ex: $\frac{dx}{dt} = -gt$
 $\frac{d^2x}{dt^2} = a$

- $\frac{dy}{dt} = f(t, y)$

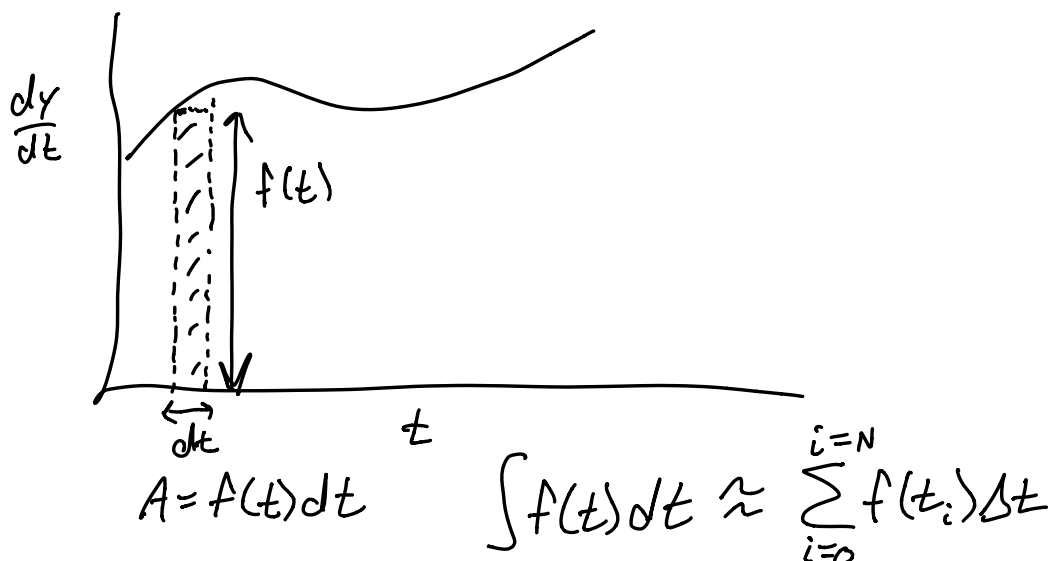
Ex: $\frac{dN}{dt} = -\frac{1}{\tau} N$

$$\frac{d^2x}{dt^2} = \frac{GM}{x^2}$$

- We want to be able to predict and analyze the motion and behavior of these sorts of systems
 - Not just their derivatives
- How do we solve these?
 - The first type is often very simple
 - Can integrate directly
 - Even if $f(t)$ is very complicated, we can calculate it everywhere and numerically integrate

$$\begin{aligned} \frac{dx}{dt} &= -gt \longrightarrow dx = -gt dt \\ \int dx &= \int -gt dt \\ x(t) &= -\frac{1}{2}gt^2 + C \end{aligned}$$

- If $f(t)$ is complicated, we can just do a Reimann sum or some better method (jupyter notebook)



- But what if $dy/dt = f(t,y)$?
 - We can't just plot the derivative and do a Reimann sum because we don't *know* the derivative everywhere
 - The derivative depends on the function itself, which is changing according to the derivative!
 - Usually, we know the value of y at some starting time t , usually $t=0$
 - Initial value problem

$$\frac{dy}{dt} = f(t, y)$$

$$\frac{dy}{dt} \longrightarrow y \longrightarrow \frac{dy}{dt}$$

Sometimes can still integrate directly ...

What if we know $y(t_0) = y_0$?

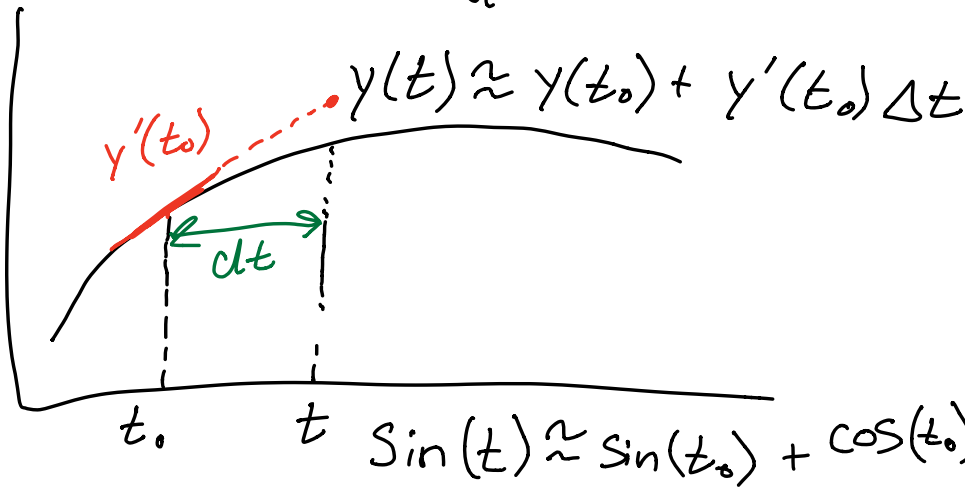
$$\text{Then } \frac{dy}{dt} = f(t_0, y_0)$$

Taylor's Thm:

$$y(t) = y(t_0) + \frac{dy(t_0, y_0)}{dt} \cdot (t - t_0)$$

$$+ \frac{1}{2} \frac{d^2 y(t_0, y_0)}{dt^2} \cdot (t - t_0)^2$$

$$+ \frac{1}{6} \frac{d^3 y(t_0, y_0)}{dt^3} \cdot (t - t_0)^3 + \dots$$



- Now "know" $y(t) + f(t, y(t)) \approx \frac{1}{2} \sin(t) \Delta t^2$

- can find $y(t + \Delta t)$

- Starting at a known initial value, we can repeat Taylor's approximation indefinitely until we reach the desired stop time
- Often, we don't know the second derivative. If dt is small enough, we can ignore it

Code

- Discretize time in small steps Δt

$$t_n = t_0 + n\Delta t$$

$$y_n = y(t_n)$$

$$\text{Start: } y(t_0) = y_0$$

$$y_1 \approx y_0 + \frac{dy_0}{dt} \Delta t$$

$$y_2 \approx y_1 + \frac{dy_1}{dt} \Delta t$$

$$y_n \approx y_{n-1} + \frac{dy_{n-1}}{dt} \Delta t$$

$$\text{Ex: } \frac{dy}{dt} = -\frac{1}{\tau} y$$

$$y(0) = 5$$

$$\frac{dy}{dt}(0) = -\frac{1}{\tau} 5$$

$$y_1 = y_0 + \frac{dy_0}{dt} \Delta t$$

$$= 5 - \frac{1}{\tau} 5 \Delta t = \# = y_1$$

$$y_2 = y_1 + \frac{dy_1}{dt} \Delta t, \dots$$

- This is only an approximation!
 - What are some errors associated with it?
 - Truncation error
 - Round off error

$$y_n = y_{n-1} + \frac{dy_{n-1}}{dt} \Delta t + \cancel{\frac{1}{2} \frac{d^2 y_{n-1}}{dt^2} \Delta t^2} + \dots$$

Approximation will be off
by a term $\sim \Delta t^2$

- truncation error = $\mathcal{O}(\Delta t^2)$

$$y_n - y_{n-1} \sim \Delta t^2$$

- Error at n^{th} pt

- Each step we incur an error of $\sim \Delta t^2$

- At point t , we have taken $\sim \frac{1}{\Delta t}$ steps
 $n = \frac{t - t_0}{\Delta t}$

Each step: $E \sim \Delta t^2$

$$E_n \sim \Delta t^2 \cdot n = \Delta t^2 \cdot \frac{1}{\Delta t} = \Delta t$$

$$E_n \sim \Delta t \quad \text{Want } E_n \sim 10^{-6}$$

$$\text{need } \Delta t \sim 10^{-6}$$

- Show Jupyter lab global error demo
- So do we want to make Δt as small as possible?

- Smaller $\Delta t \rightarrow$ smaller
truncation
error

BUT

more computing power
required

- on a computer, Δt
cannot be arbitrarily
small

- smallest pos. #
(machine eps)

$$1 + \epsilon = 1$$

Any float operation
is only accurate to $\mathcal{O}(\epsilon)$

Floating Pt Error $\sim \epsilon$

$$E_n \sim \frac{\epsilon}{\Delta t}$$

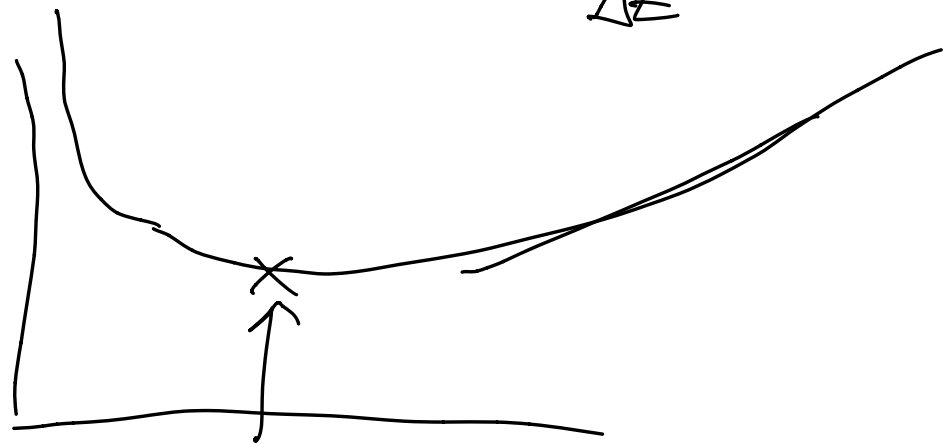
$$E_n = E_n^{\text{trunc}} + E_n^{\text{FP}}$$

$$\uparrow \quad = \frac{\epsilon}{\Delta t} + \Delta t$$

total

$$\Delta t \gg \epsilon, E_n \sim \Delta t$$

$$\Delta t \sim \epsilon, E_n \sim \frac{\epsilon}{\Delta t}$$



$$\Delta t \sim \epsilon^{\frac{1}{2}}$$

$$\sim 10^{-8}$$

Must balance this w/
available resources

Small in comparison to
what?