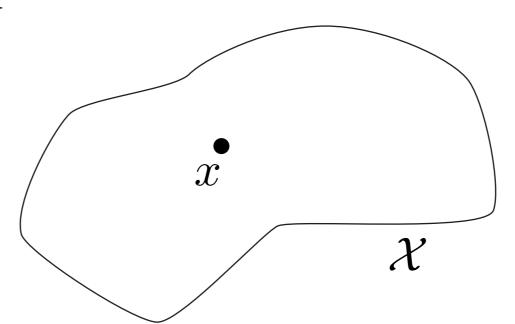
Numerical Integration of Ordinary Differential Equations

Reading:

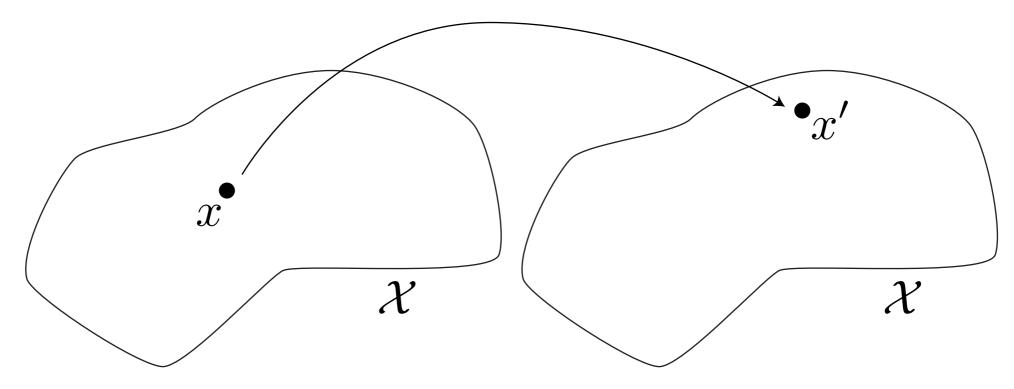
NDAC Secs. 2.8 and 6.1

State Space: \mathcal{X}

State: $x \in \mathcal{X}$



Dynamic: $T: \mathcal{X} \rightarrow \mathcal{X}$



Dynamical System ...
For example, continuous time ...

Ordinary differential equation: $\dot{\vec{x}} = \vec{F}(\vec{x})$ ($\dot{} = \frac{d}{dt}$)

State: $\vec{x}(t) \in \mathbf{R}^n$ $\vec{x} = (x_1, x_2, ..., x_n)$

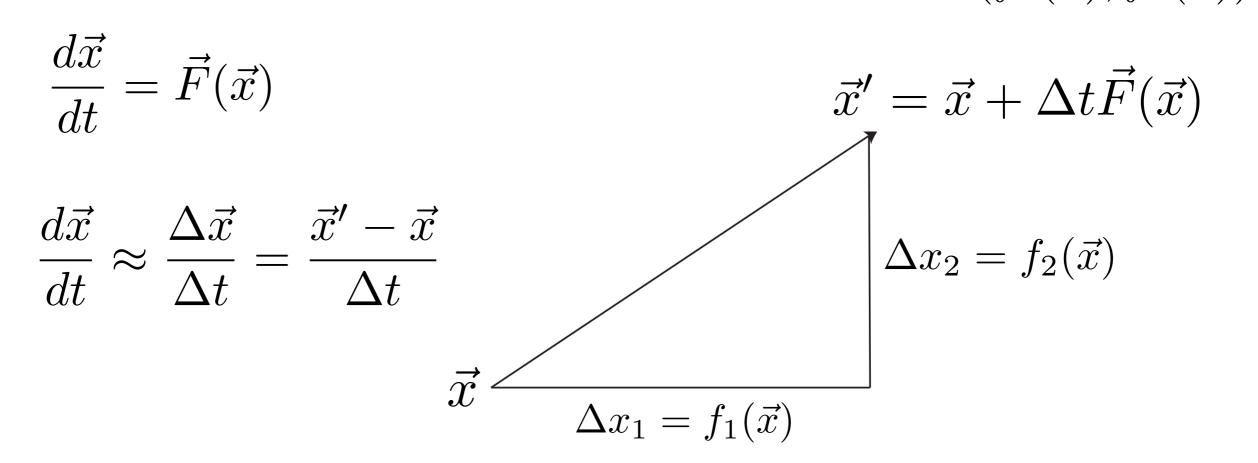
Initial condition: $\vec{x}(0)$

Dynamic: $\vec{F}: \mathbf{R}^n \to \mathbf{R}^n$ $\vec{F} = (f_1, f_2, \dots, f_n)$

Dimension: n

 $\mathcal{X} = \mathbf{R}^2 \quad \vec{x} = (x_1, x_2)$ $\vec{F} = (f_1(\vec{x}), f_2(\vec{x}))$

Geometric view of an ODE:



Each state \vec{x} has a vector attached $\vec{F}(\vec{x})$

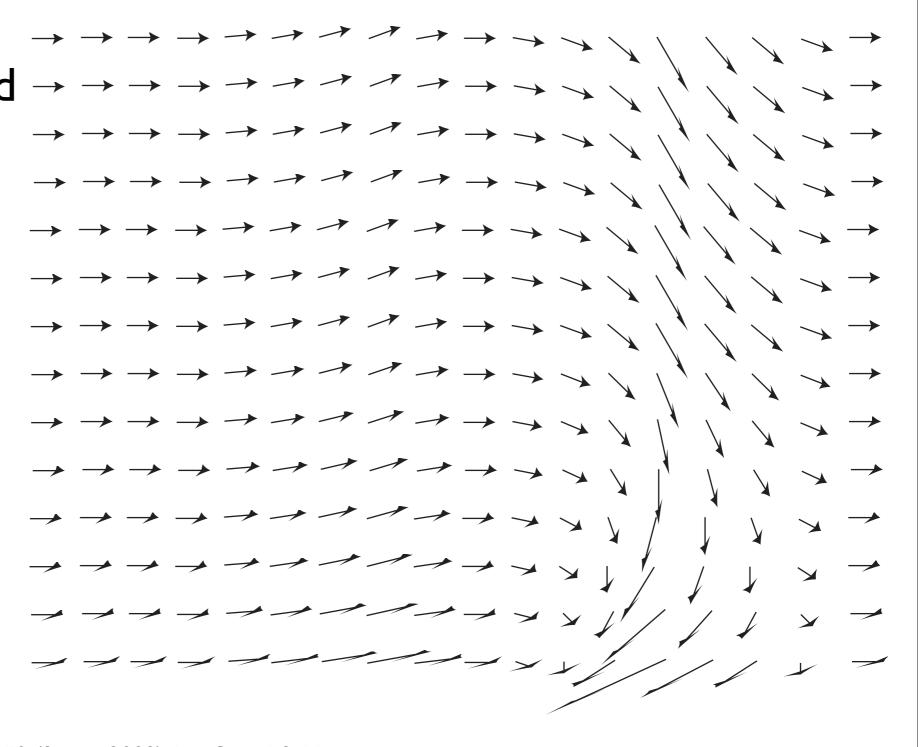
that says to what next state to go: $\vec{x}' = \vec{x} + \Delta t \cdot \vec{F}(\vec{x})$.

Vector field for an ODE (aka Phase Portrait)

 $\mathcal{X} = \mathbf{R}^2$

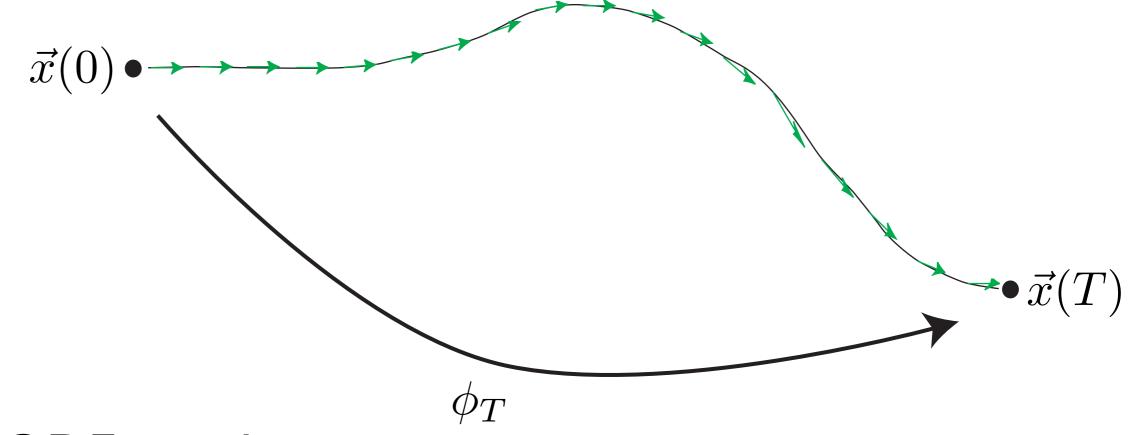
A set of rules:

Each state has a vector attached That says to what next state to go



Time-T Flow:
$$\vec{x}(T) = \phi_T(\vec{x}(0)) = \int_0^T dt \ \dot{\vec{x}} = \int_0^T dt \ \vec{F}(\vec{x}(t))$$

The solution of the ODE, starting from some IC Simply follow the arrows



Point: ODE is only instantaneous, flow gives state for *any* time t.

Euler Method in ID:

$$\dot{x} = f(x)$$

$$\frac{dx}{dt} \approx \frac{\Delta x}{\Delta t} = \frac{x_{n+1} - x_n}{\Delta t}$$

$$x(t_0 + \Delta t) \approx x_1 = x_0 + f(x_0)\Delta t$$

$$x_{n+1} = x_n + f(x_n)\Delta t$$

A discrete-time map!

Improved Euler Method in ID:

$$\dot{x} = f(x)$$

A trial (Euler) step:

$$\widehat{x}_n = x_n + f(x_n)\Delta t$$

The resulting better estimate (averaged at t_n and t_{n+1}):

$$x_{n+1} = x_n + \frac{1}{2} \left[f(x_n) + f(\widehat{x}_n) \right] \Delta t$$

Fourth-order Runge-Kutta Method in 1D:

Intermediate estimates:

$$k_1 = f(x_n)\Delta t$$

$$k_2 = f(x_n + \frac{1}{2}k_1)\Delta t$$

$$k_3 = f(x_n + \frac{1}{2}k_2)\Delta t$$

$$k_4 = f(x_n + k_3)\Delta t$$

Final estimate:

$$x_{n+1} = x_n + \frac{1}{6} \left[k_1 + 2k_2 + 2k_3 + k_4 \right]$$

Good trade-off between accuracy and time-step size.

Runge-Kutta in nD:

$$\dot{\vec{x}} = \vec{f}(\vec{x}), \ x \in \mathbb{R}^n \qquad \vec{f}: \mathbb{R}^n \to \mathbb{R}^n$$

Intermediate estimates:

$$\vec{k}_1 = \vec{f}(\vec{x}_n)\Delta t$$

$$\vec{k}_2 = \vec{f}(\vec{x}_n + \frac{1}{2}\vec{k}_1)\Delta t$$

$$\vec{k}_3 = \vec{f}(\vec{x}_n + \frac{1}{2}\vec{k}_2)\Delta t$$

$$\vec{k}_4 = \vec{f}(\vec{x}_n + \vec{k}_3)\Delta t$$

Final estimate:

$$\vec{x}_{n+1} = \vec{x}_n + \frac{1}{6} \left[\vec{k}_1 + 2\vec{k}_2 + 2\vec{k}_3 + \vec{k}_4 \right]$$