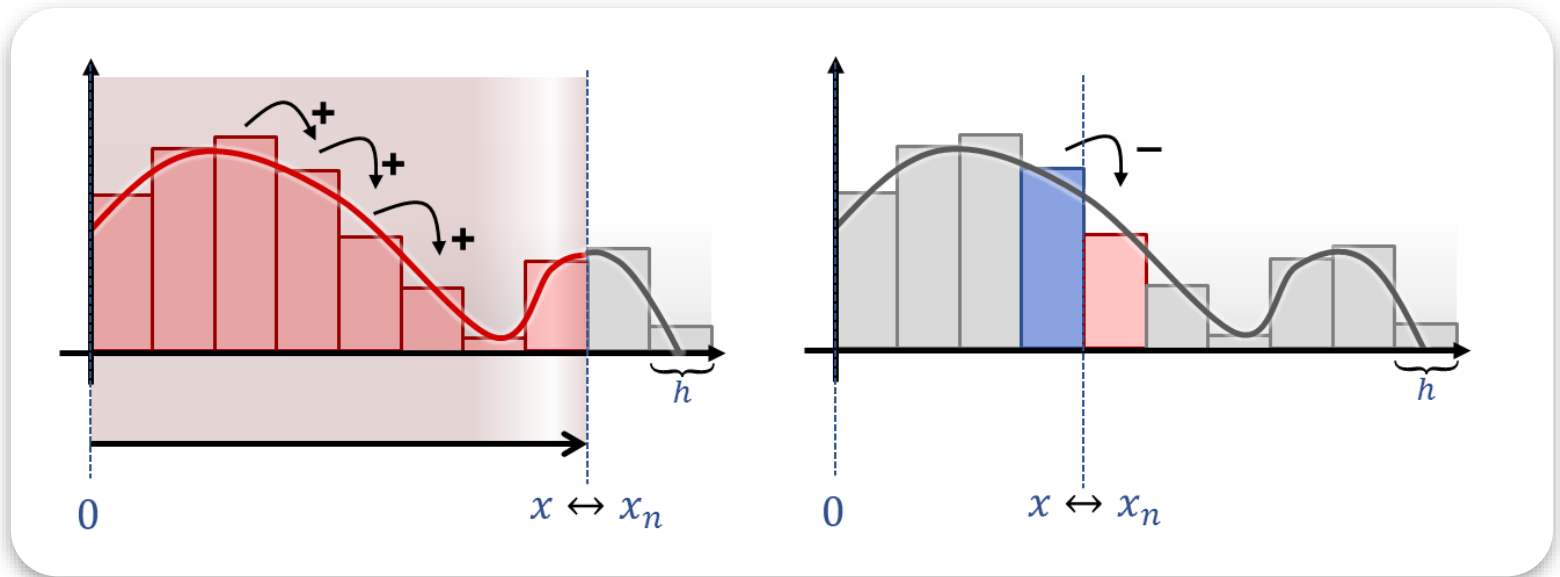


# Modelling 1

SUMMER TERM 2020



## LECTURE 20

# Differential & Functional Equations

# Functional Equations

(usually: Differential & Integral Equations)

# Functional Equations

## Searching for functions

- Implicit definition of a function

$$F(f) = 0$$

- Function  $f$ : unknown
- Function  $F$ : constraint on  $f$

# Functional Equations

## Linear functional equations

- $F$  is linear (“linear operator”).
  - We use  $L$  in that case.
  - We drop the  $(\cdot)$  in analogy to matrix multiplication

- Homogeneous:

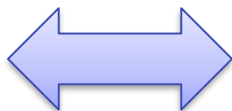
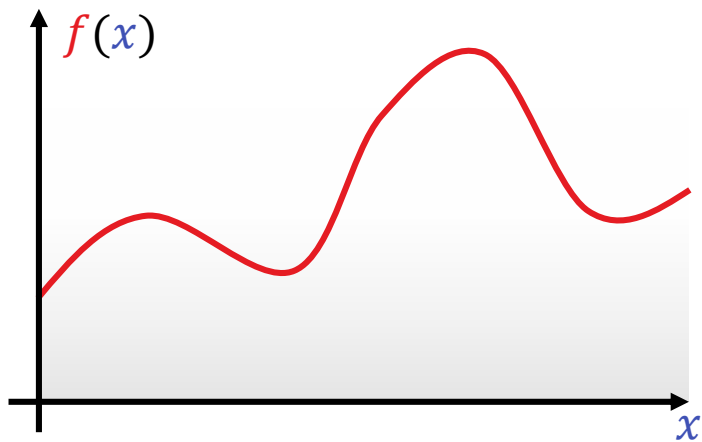
$$Lf = 0$$

- Inhomogeneous:

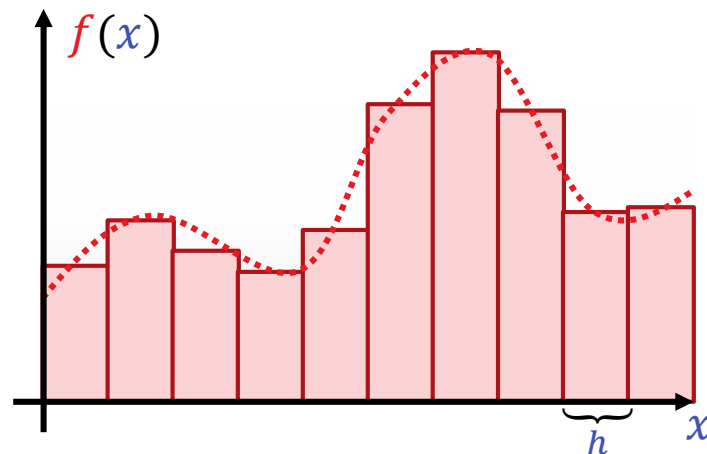
$$Lf = g$$

# Discrete Analogy

Function  $f$



Think of this:



# We Know the Structure...

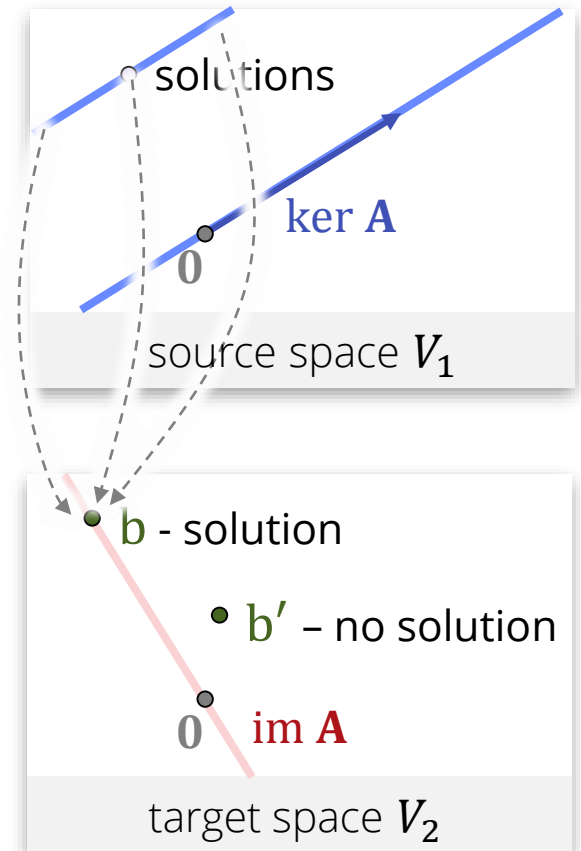
## **Structure of a functional equation**

- Linear solution space
- Add inhomogeneous solution to any homogeneous

# Reminder (Linear Maps)

## Solutions to linear system

- $\mathbf{Ax} = \mathbf{0}$ 
  - Solution space =  $\ker \mathbf{A}$
- $\mathbf{Ax} = \mathbf{b}$ 
  - Solution if and only if  $\mathbf{b} \in \text{im } \mathbf{A}$
- Set of all solutions:
  - One  $\mathbf{y}$  with  $\mathbf{Ay} = \mathbf{b}$
  - Add any solution of  $\mathbf{Ax} = \mathbf{0}$
  - Solution set:  $\mathbf{y} + \ker \mathbf{A}$



# Differential Equations



*Ordinary*

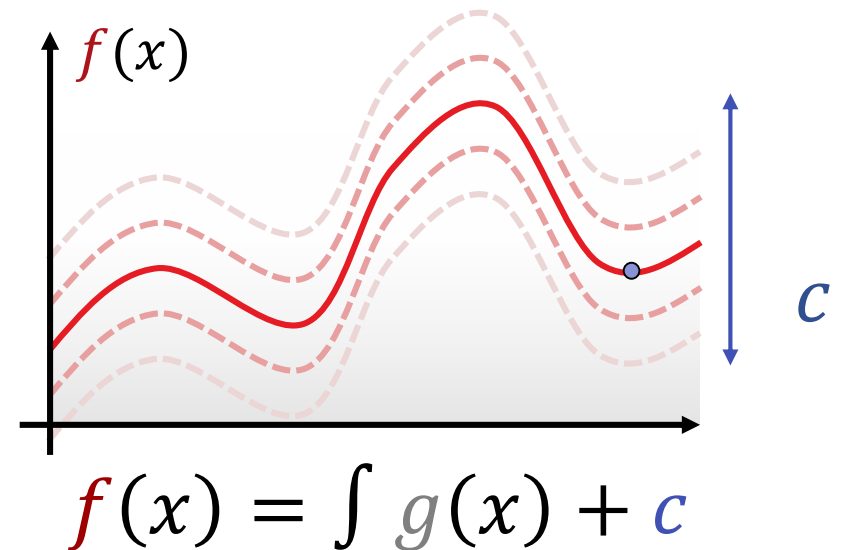
Differential Equations  
(ODEs)

# Examples

# Boundary Conditions

## Solution space

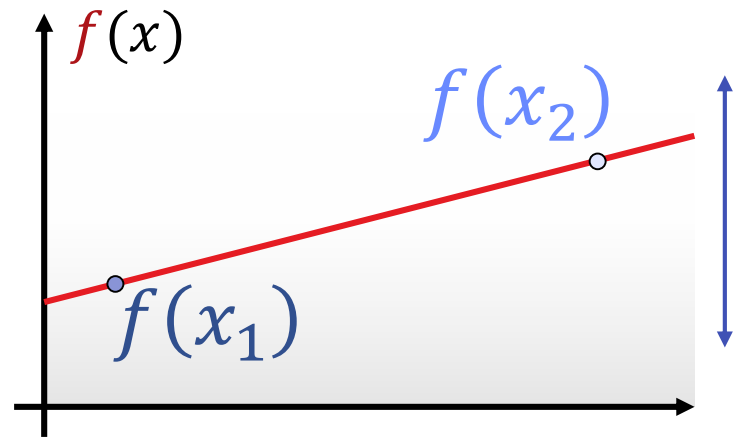
- $\frac{d}{dx} f(x) = g(x)$
- $\dim \ker \frac{d}{dx} = 1$
- Solution space is 1-dimensional
- One degree of freedom



# Boundary Conditions

## Solution space

- $\frac{d^2}{dx^2} f(x) = 0$
- $\dim \ker \frac{d^2}{dx^2} = 2$
- Solution space is 2-dimensional
- Two degrees of freedom



# Discretized

## Continuous Equation

$$\frac{d}{dx} f(x) + 1 \cdot f(x) = 0$$

## Matrix Approximation

$$\left[ \frac{1}{h} \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 & 0 \\ 0 & 0 & -1 & 1 & 0 \\ 0 & 0 & 0 & -1 & 1 \end{pmatrix} + \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \right] \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{pmatrix} = 0$$

# Discretized

## Continuous Equation

$$\frac{d^2}{dx^2} f(x) = 0$$

## Matrix Approximation

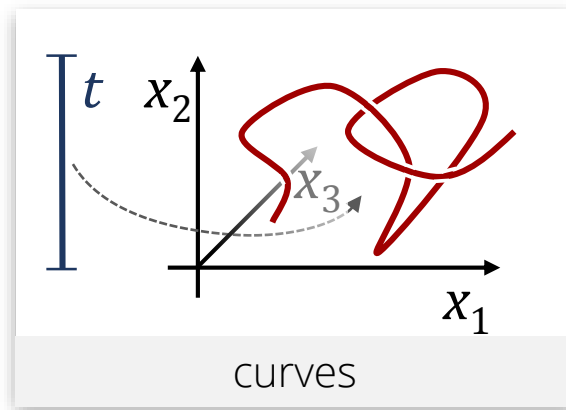
$$\frac{1}{h^2} \begin{pmatrix} -1 & 1 & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 \\ 0 & 1 & -2 & 1 & 0 \\ 0 & 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & 1 & -1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{pmatrix} = 0$$

What is it about?

# Designing Curves

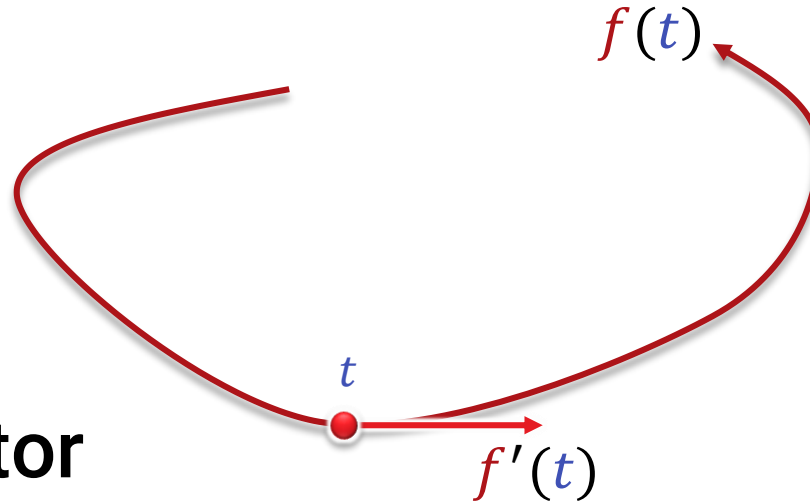
## One Parameter functions

- Functions  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  (“scalar field”)
- Functions  $f: \mathbb{R} \rightarrow \mathbb{R}^n$  (“curves”)
- Functions  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  (general case)





# Geometric Meaning



## Tangent Vector

- $f'$  is the tangent vector
  - Higher order derivatives: also vectors
- Physical particle
  - First derivative  $\dot{f} \cong$  velocity.
  - Second derivative  $\ddot{f} \cong$  acceleration

# Example

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

$$a \frac{d^2}{dt^2} f(t) + b \frac{d}{dt} f(t) + c f(t) = g(t)$$

## Example

- Linear
- 1dim ODE
- 2nd degree

# First Order Derivatives Suffice

## Higher order ODE

$$\frac{d^2}{dt^2} f(t) = g(t)$$

Convert to system (multi-dim.) of lower order DEs

### Substitution

$$v(t) := \frac{d}{dt} f(t)$$
$$\frac{d^2}{dt^2} f(t) = \frac{d}{dt} v(t)$$

### System

$$v(t) = \frac{d}{dt} f(t)$$
$$\frac{d}{dt} v(t) = g(t)$$

# General Form of an ODE

**Unknown**

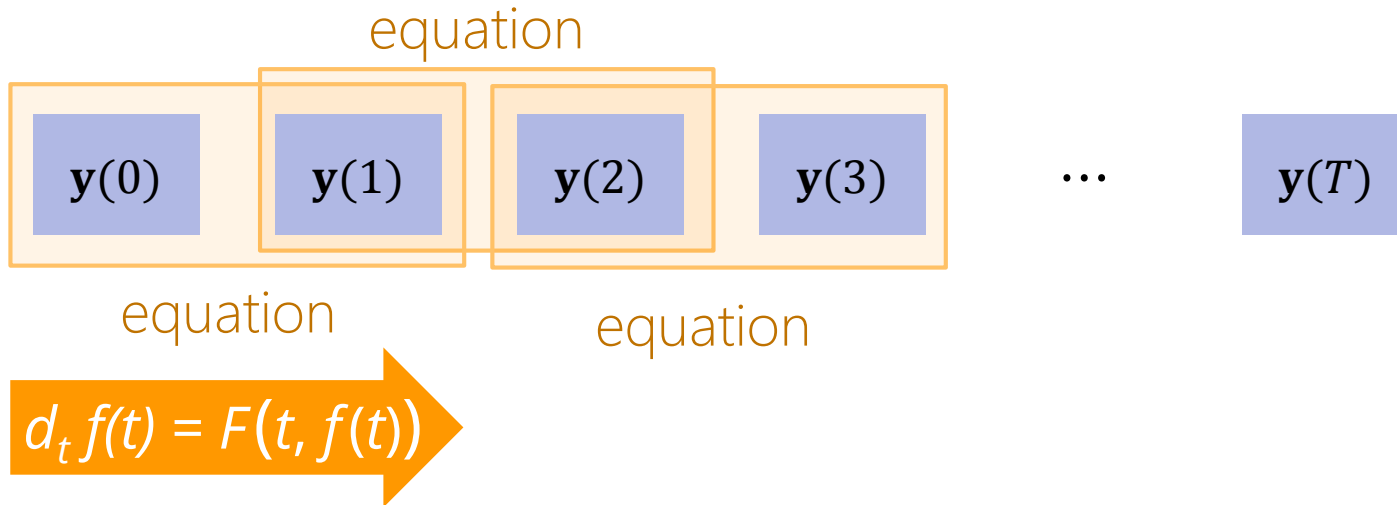
$$f: \mathbb{R} \rightarrow \mathbb{R}^n$$

**Explicit Form**

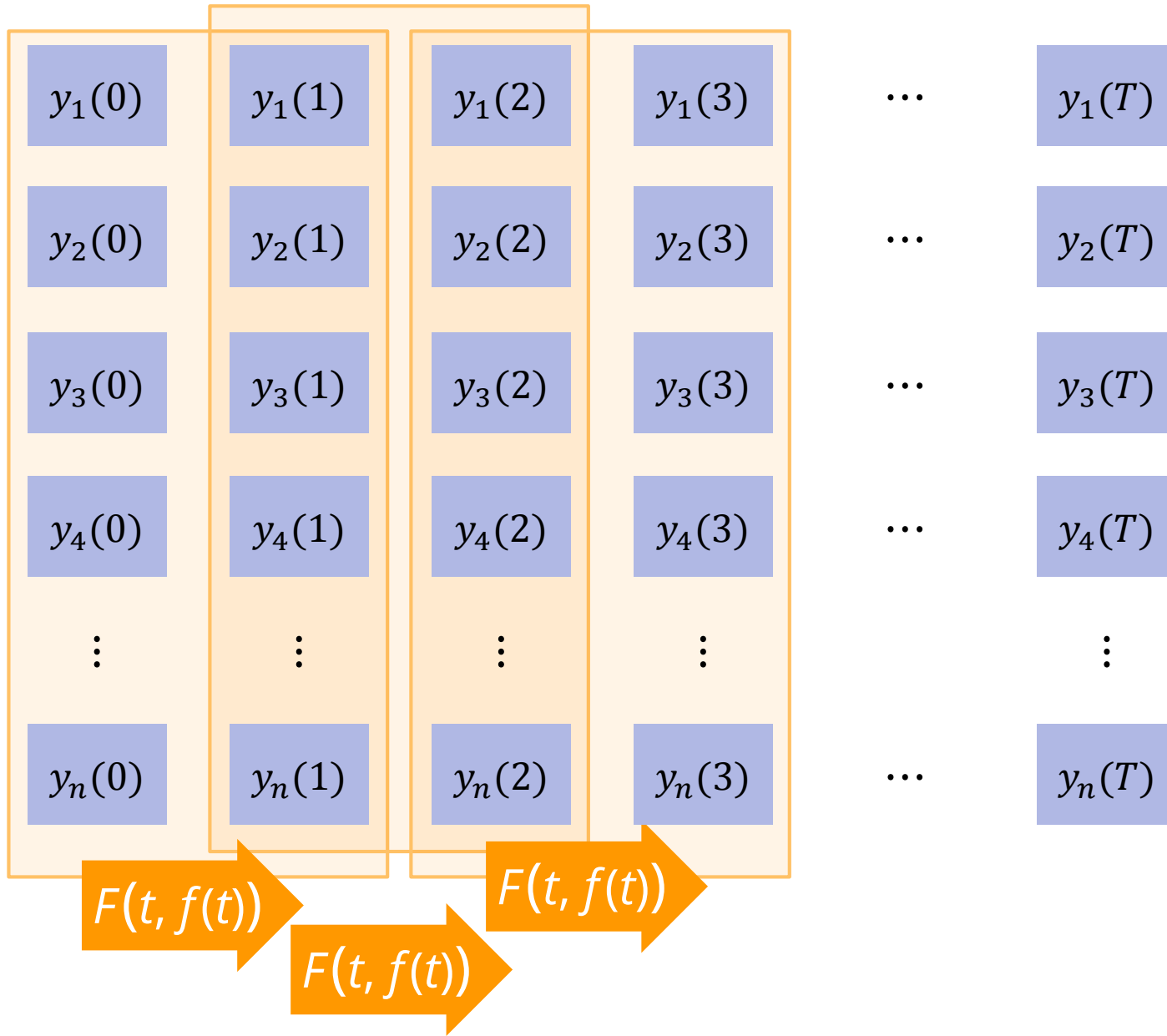
$$\frac{d}{dt} f(t) = F(t, f(t))$$

# Chain-Structure

## Causal Chain

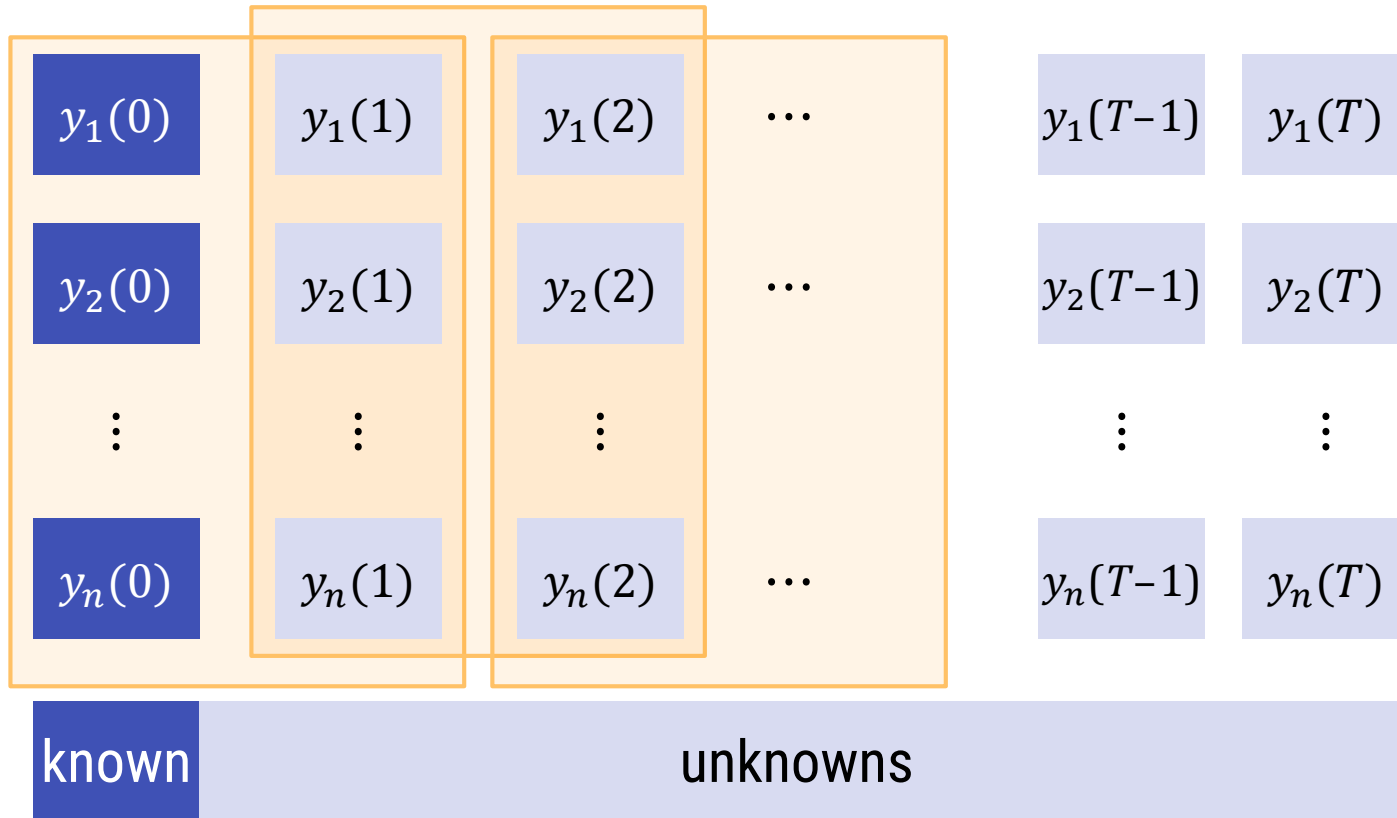


# Structure of ODE



# Initial Value Problems

# Initial Value Problems



## Solution

- Solve step-by-step
- Propagate information forward

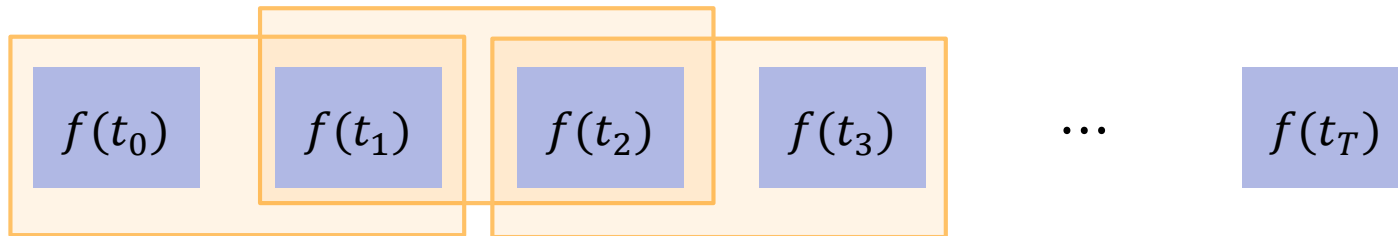


# Numerical Solvers

## ODE

$$f: \mathbb{R} \rightarrow \mathbb{R}^n, \quad \frac{d}{dt} f(t) = F(t, f(t))$$

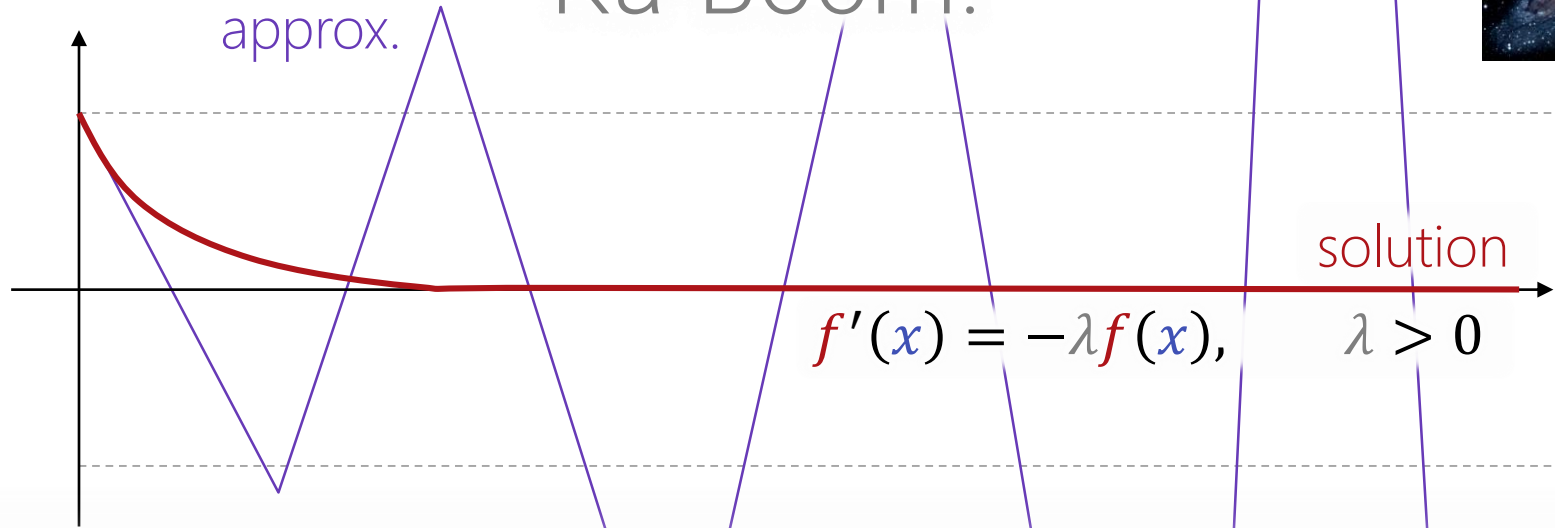
## Explicit Euler integrator



$$f(t_{i+1}) = f(t_i) + F(t_i, f(t_i)) \cdot (t_{i+1} - t_i)$$



Ka-Boom!

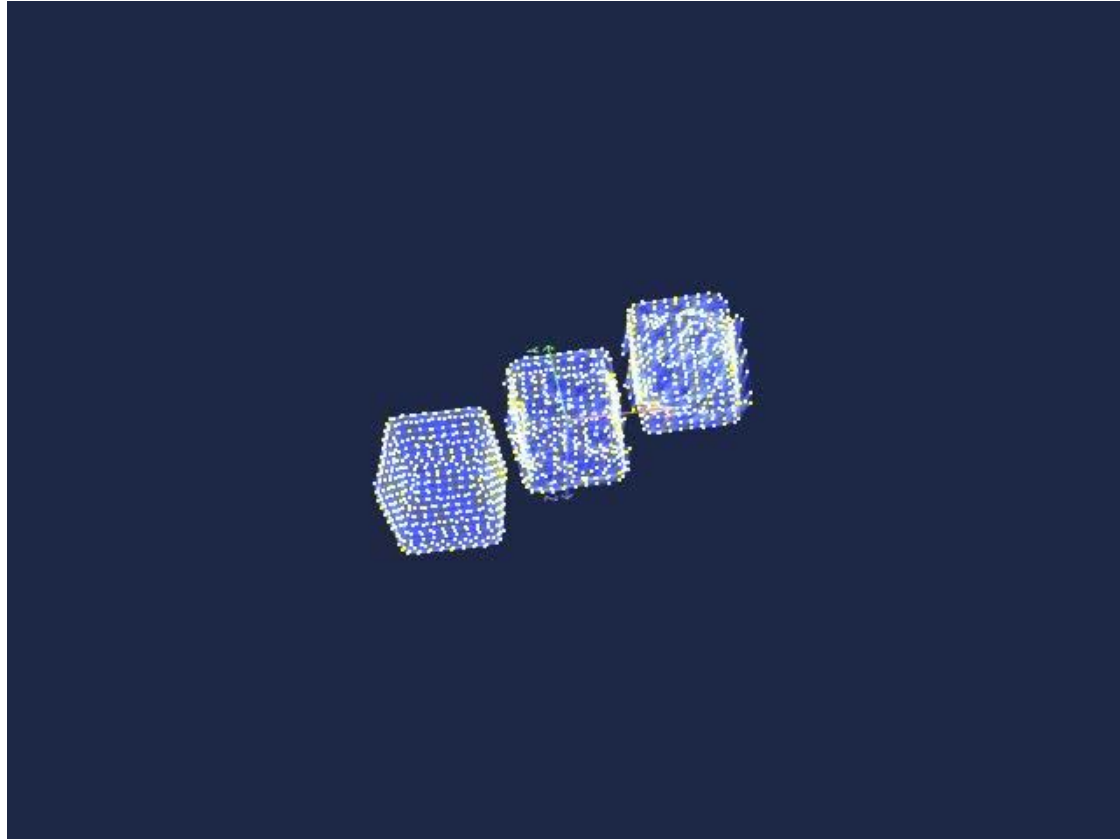


## Explicit Euler integrator

$$f(t_{i+1}) = f(t_i) + F(t_i, f(t_i)) \cdot (t_{i+1} - t_i)$$

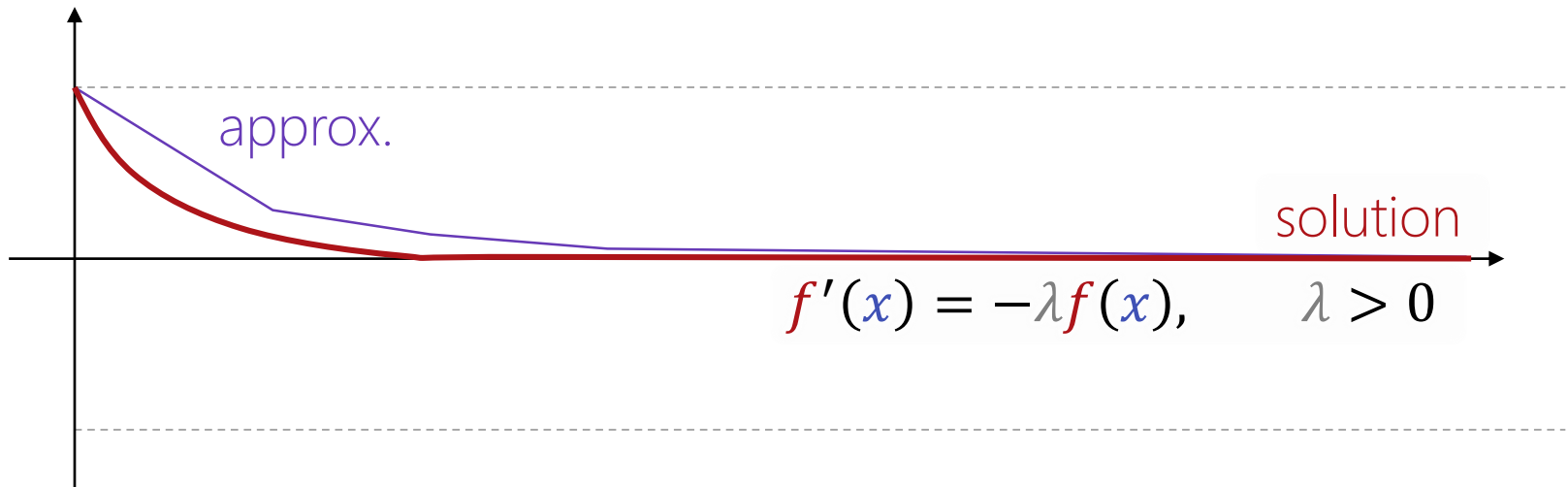
time step must be  
smaller than  $|2\lambda|$ ,  
 $\lambda$  being the most negative  
eigenvalue of  $F$

# An Actual Example...



**Stiff elasticity model** (that I screwed up myself :-)

# Ka-Boom!



## Implicit Euler integrator

$$f(t_{i+1}) = f(t_i) + F(t_{i+1}, f(t_{i+1})) \cdot (t_{i+1} - t_i)$$

*unconditionally stable  
(does not mean accurate)*

*downside:  
need to solve system of equations*

# Integrators - Variants

## Higher consistency order

- Local polynomial approximation

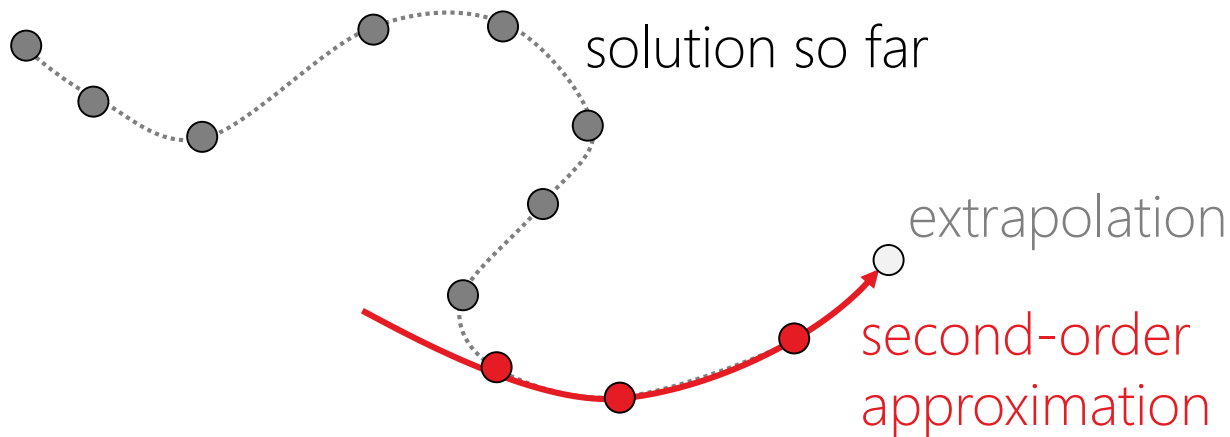
## Basic Idea

- Local polynomial approximation
- Fitted to multiple evaluations of  $F(t, y)$

## Two Variants

- Runge-Kutta Methods: single time step
  - RK4 most popular
- (Linear) Multi-Step Methods: incl. previous time steps
  - BDF-2 popular for stiff problems (i.e., huge  $|\lambda_{min}|$ )

# Integrators - Variants



## Multi-Step Methods

- Linear MSM: Fit polynomial to last  $k$  steps
  - Explicit: predict next value
  - Implicit: optimize for next value
  - Specific implicit MSM of degree 2 (BDF-2 method) is very stable and accuracy is ok

# Analytical Solutions

## ODE

$$f: \mathbb{R} \rightarrow \mathbb{R}^n, \quad \frac{d}{dt} f(t) = F(t, f(t))$$

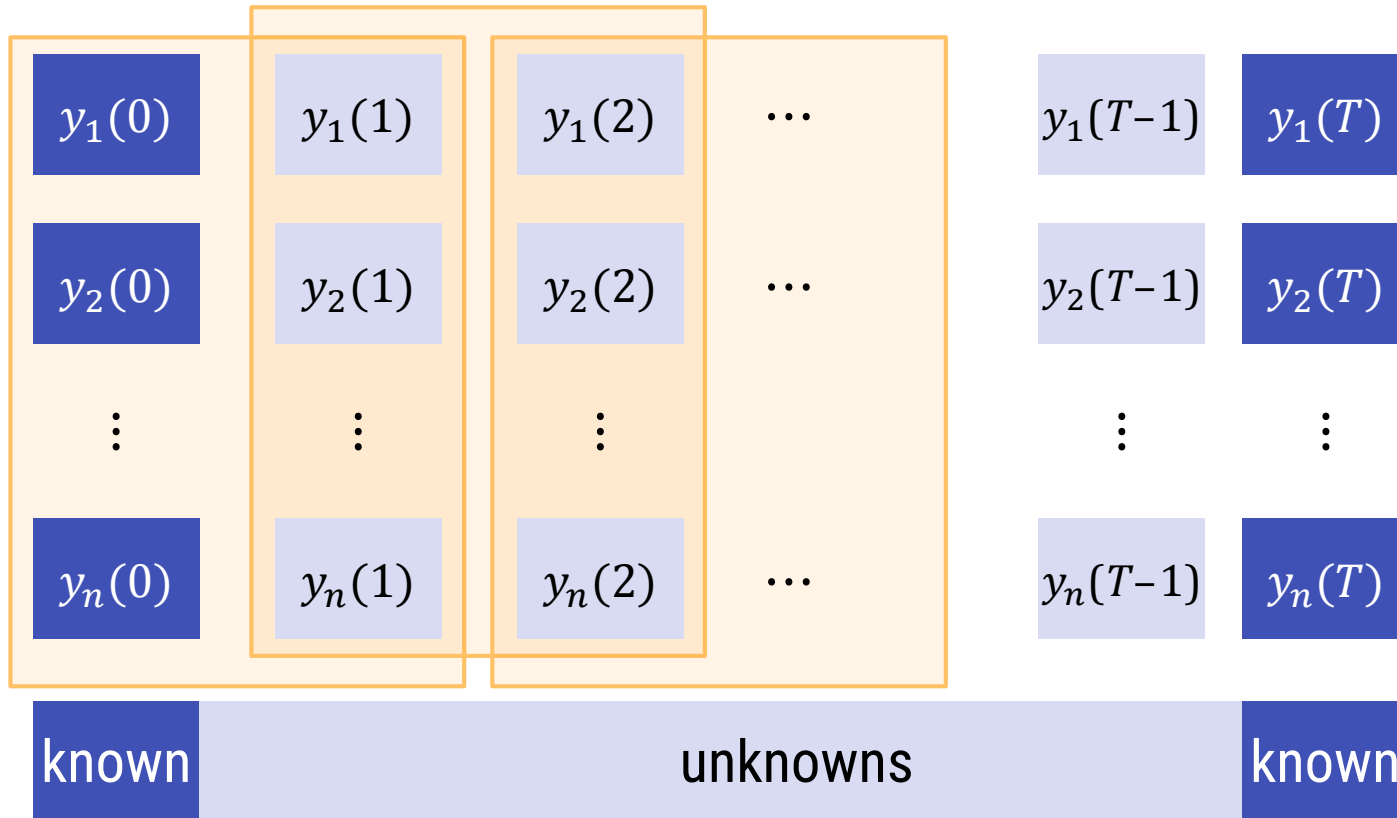
## Linear ODEs

- $F$  linear in variables
- Analytical solution possible
  - Via matrix factorization
  - Jordan-normal-form (in  $\mathbb{C}$ )
  - Solution based on complex exponentials
- Shift-invariant: Fourier transforms

# Boundary Value Problems



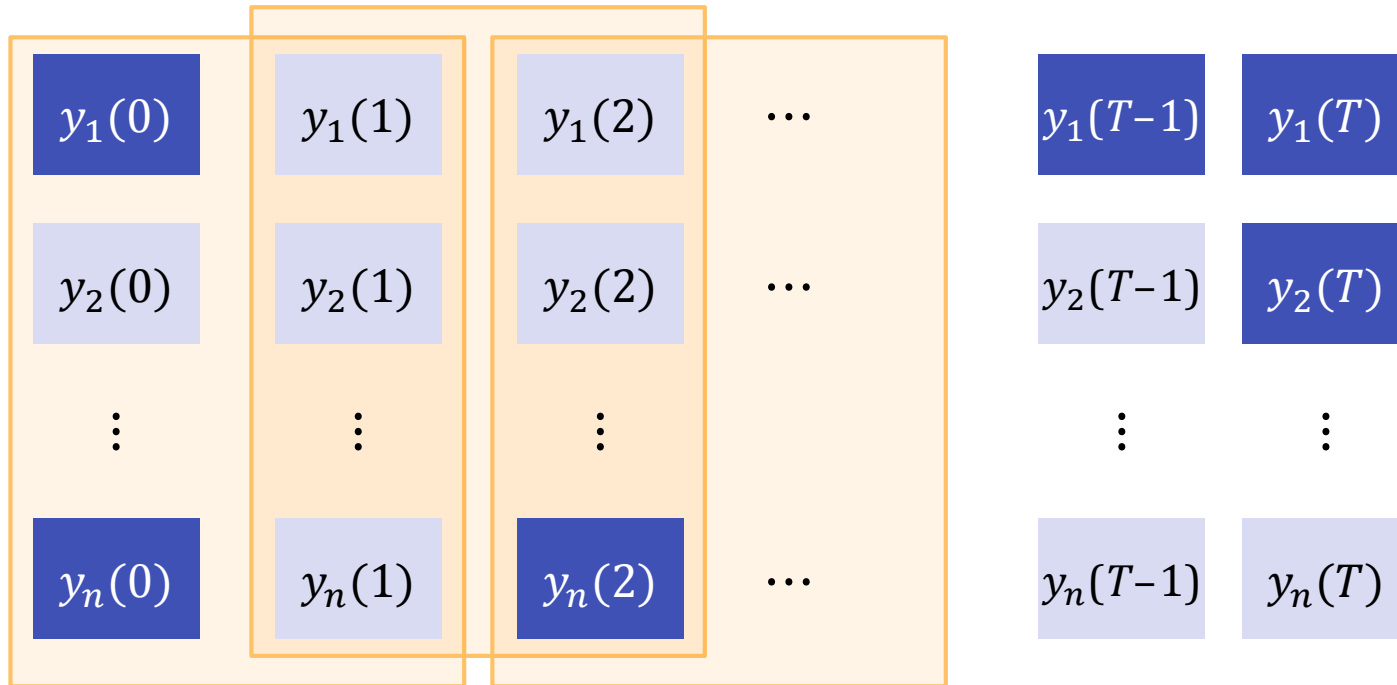
# Boundary Value Problems



## Solution

- Step-by-step does not work
- Need to solve global system of equations

# General constraints



## Variant

- Impose constraints all over the place
  - Analog to diffusion Images (link in script)
- Careful with degrees of freedom
  - We rather go for least-squares (more later)

ODE Example:  
Newtonian Physics

# Example

## Newtonian Physics

$$“F = m \cdot a”$$

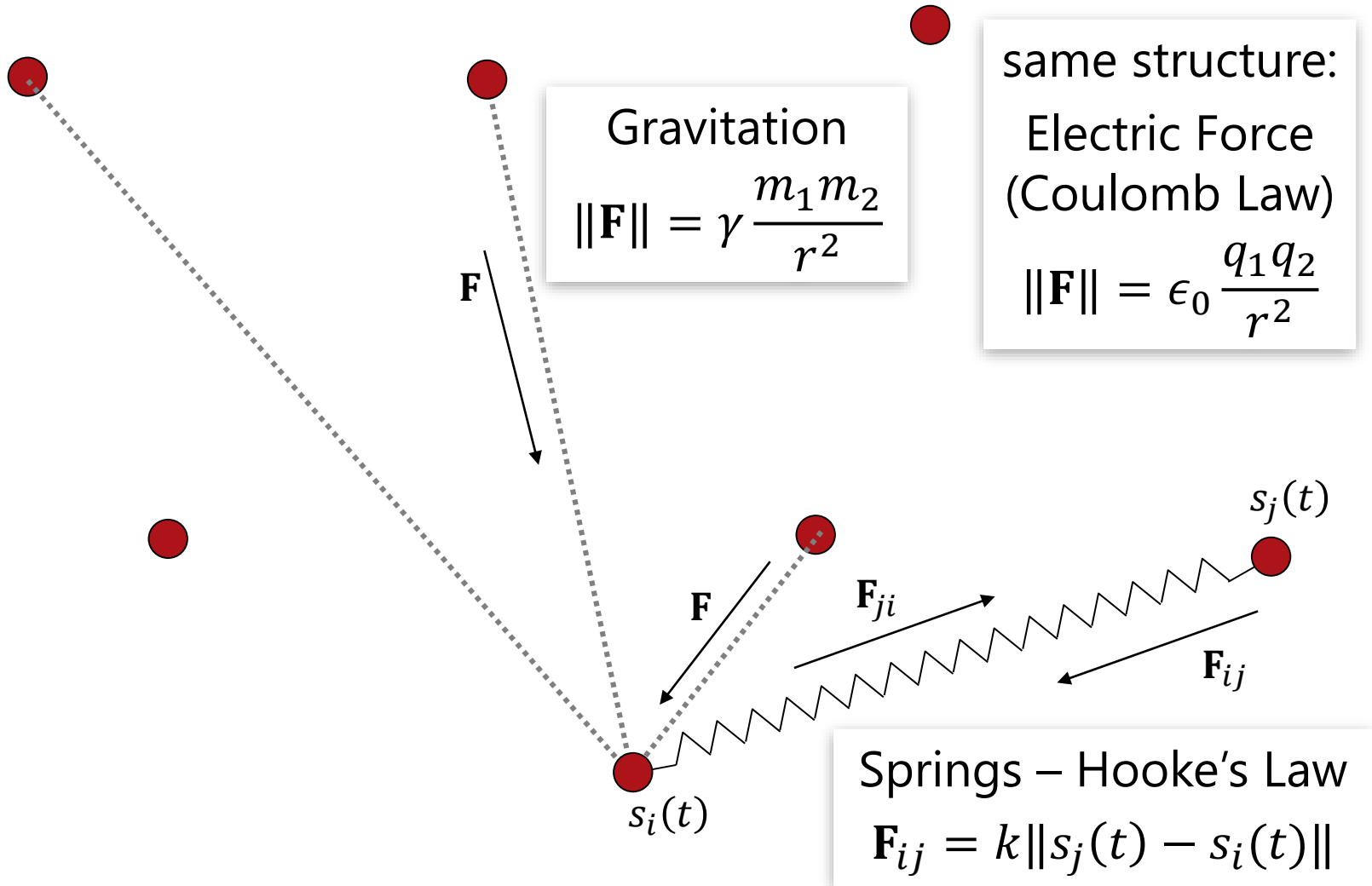
Which means

$$F(t, s(t)) = m \cdot a(t) = m \cdot \ddot{s}(t)$$

In other words...

$$\frac{d^2}{dt^2} s(t) = \frac{1}{m} F(t, s(t))$$

# Particle Systems



*Partial*

Differential Equations  
(PDEs)

# Multi-dimensional Inputs

## One Parameter functions

- Functions  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  (“scalar field”)
- Functions  $f: \mathbb{R} \rightarrow \mathbb{R}^n$  (“curves”)
- Functions  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  (general case)



*We'll stick to that case for simplicity.*

*No fundamental difference.*

# Partial Differential Equations

## Unknown

$$f: \mathbb{R}^n \rightarrow \mathbb{R}$$

## Explicit Form

$$Df(\mathbf{x}) = F(\mathbf{x}, f(\mathbf{x}))$$

$D$ : differential operator, including partials

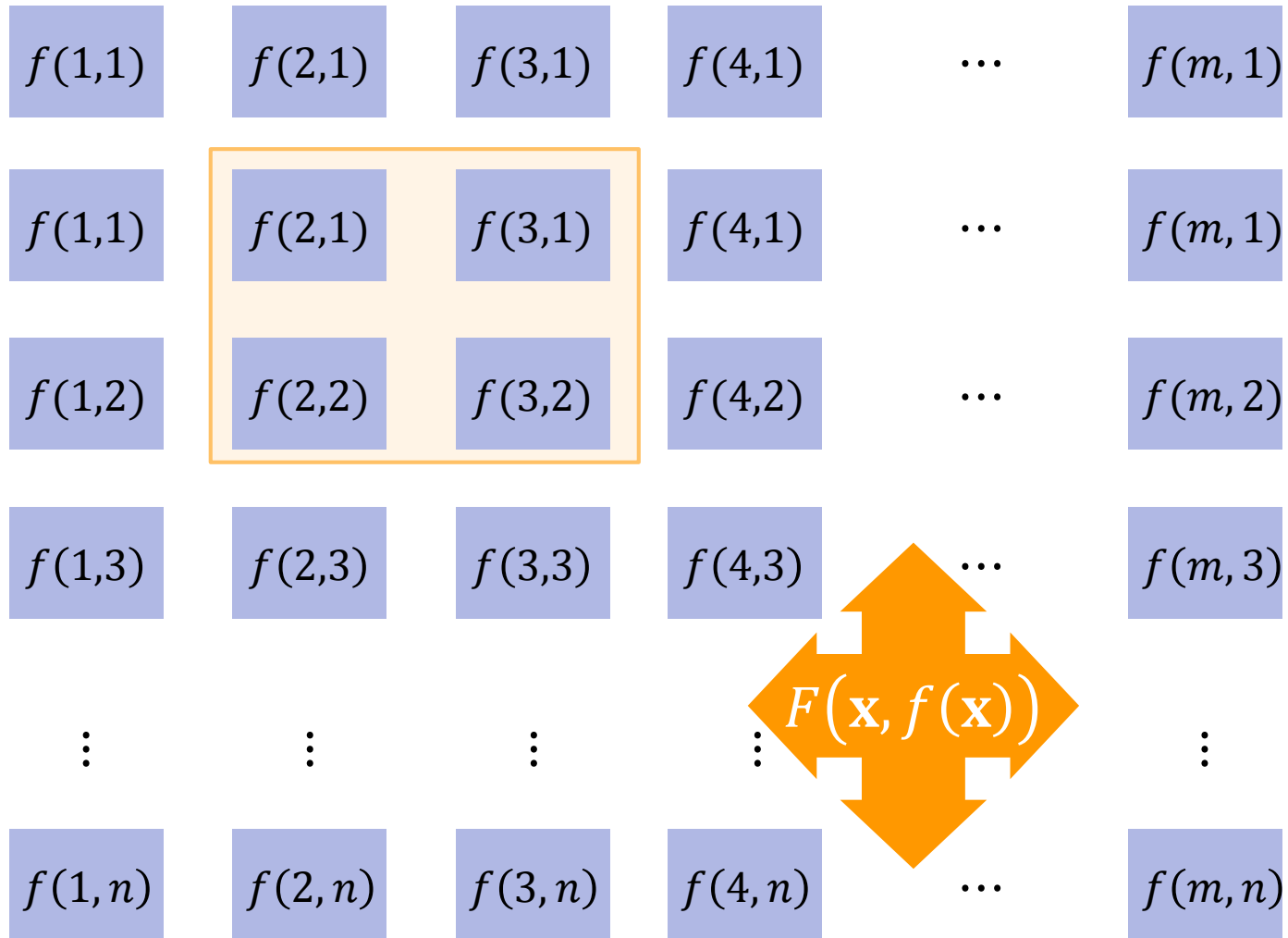
## Example

$$\frac{\partial^2}{\partial x_1^2} f(\mathbf{x}) + \frac{\partial^2}{\partial x_2^2} f(\mathbf{x}) = g(\mathbf{x})$$

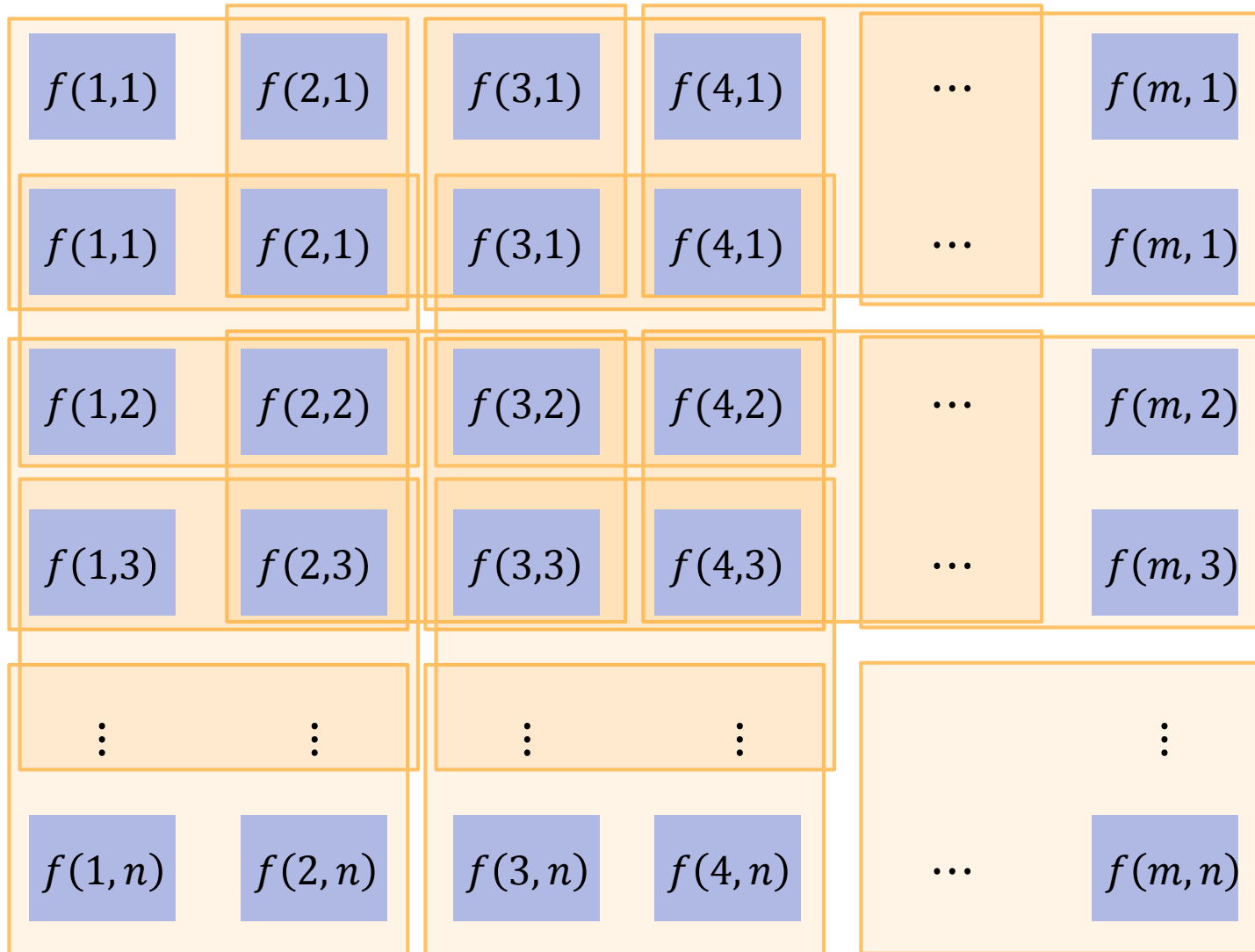
Laplacian  $\Delta f$



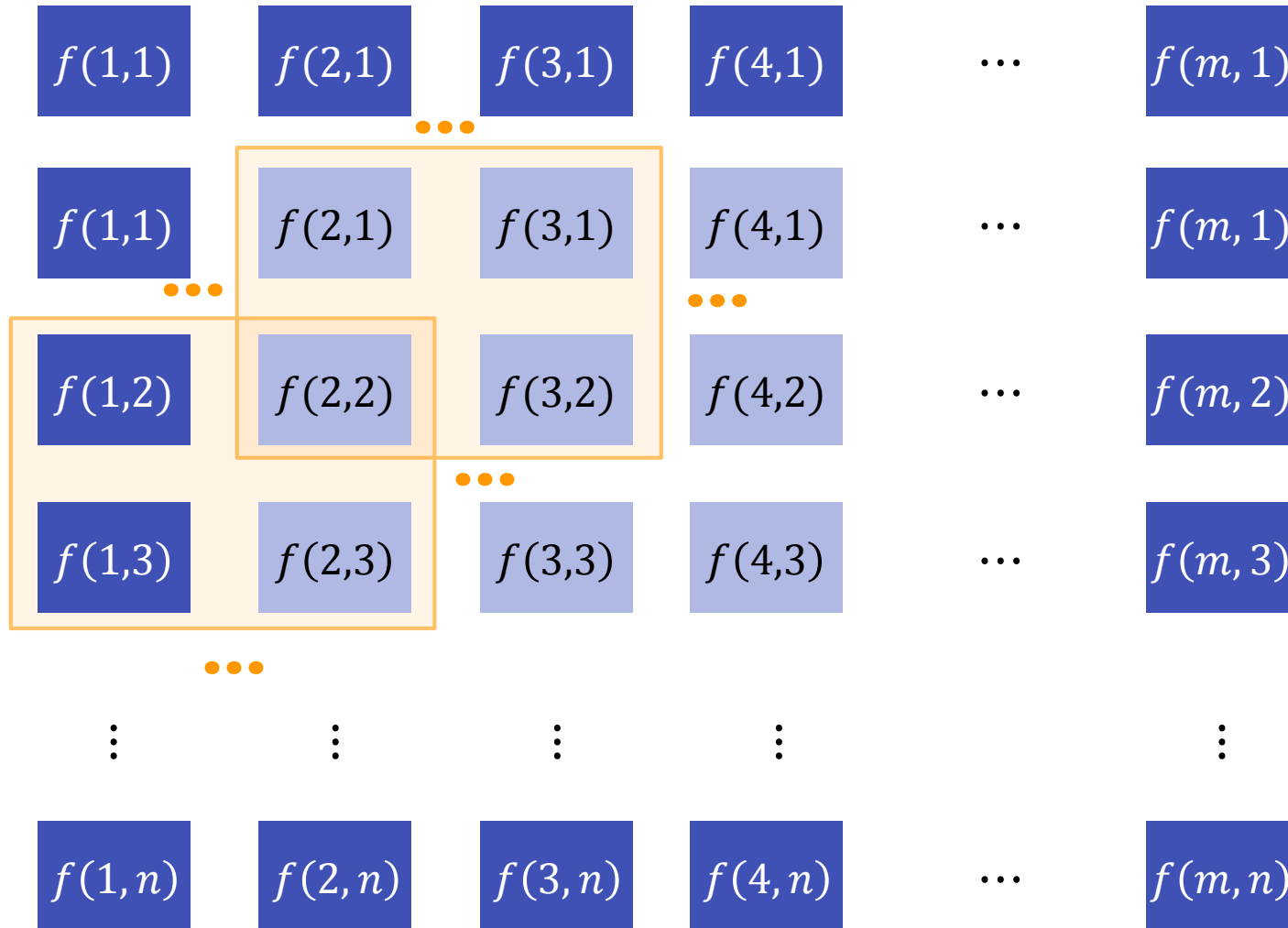
# Structure of PDE



# Structure of PDE



# Boundary Value Problem



# Solving PDEs



[Orzan et al. 2008]



[Orzan et al. 2008]

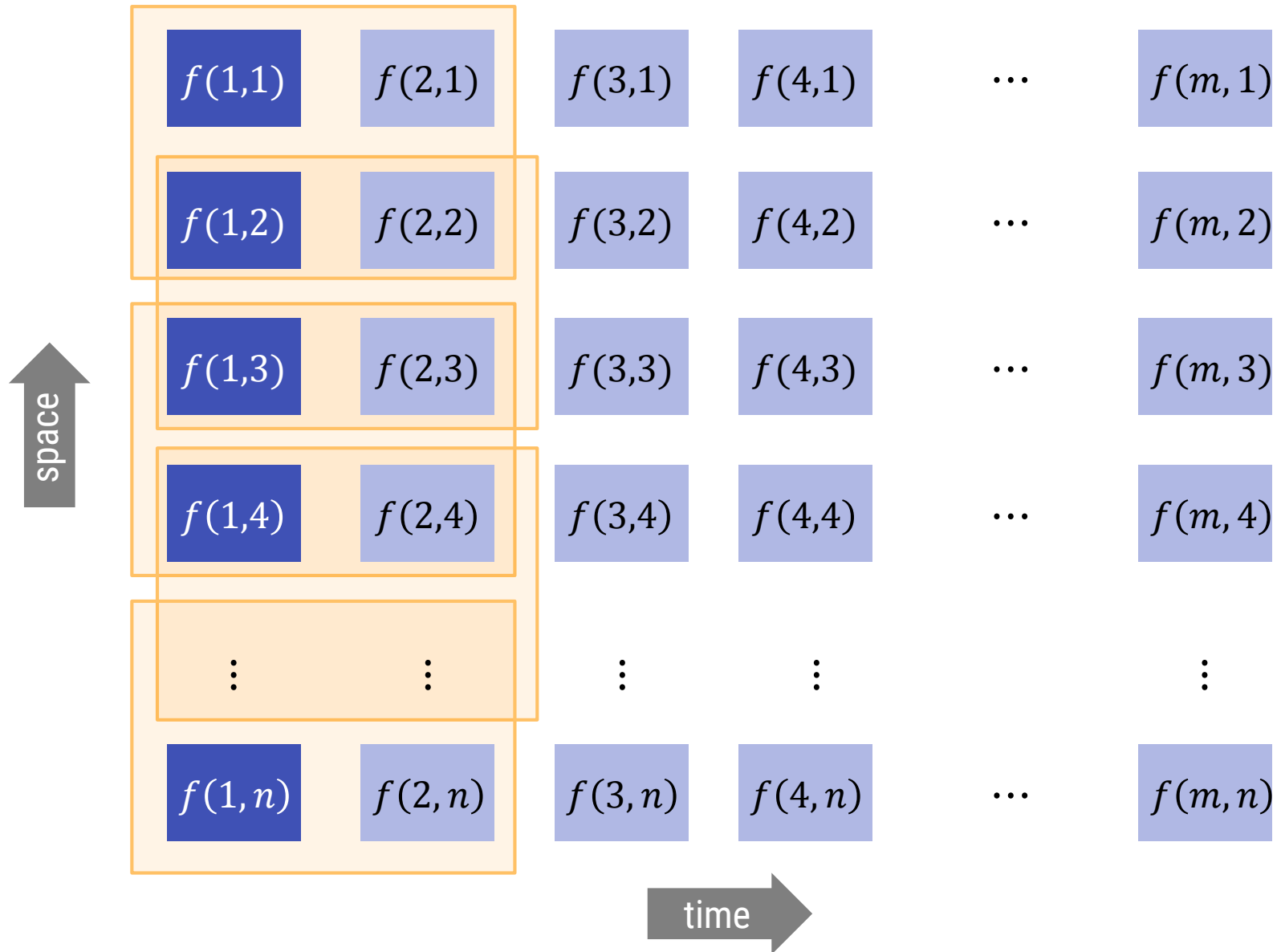
$\Delta f(x) = 0$   
- and -  
 $f(x) = g(x)$  for  $x \in \text{curve}$

## Usually boundary value problems

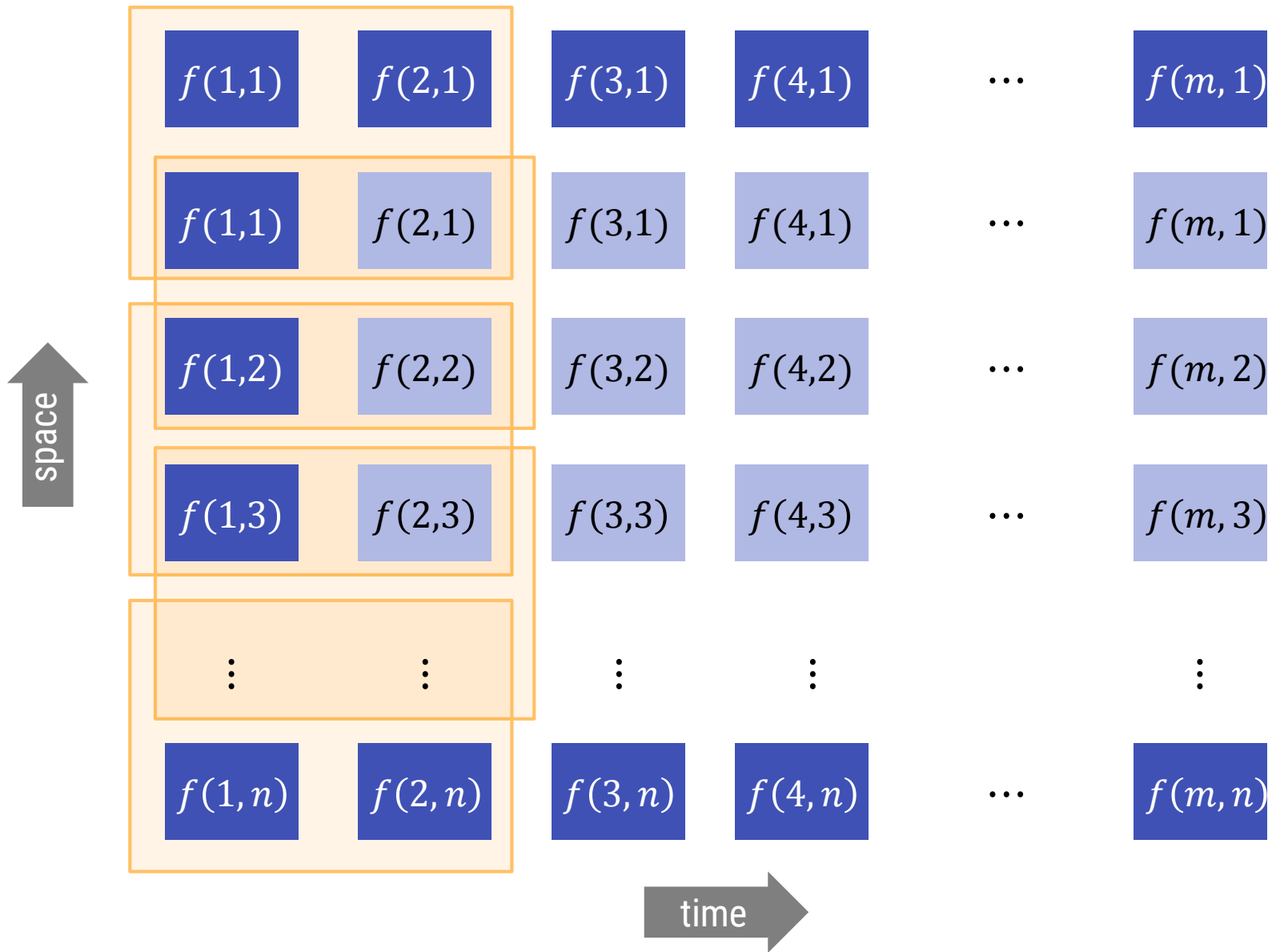
- Need global solver anyways
- No time stepping
- Linear PDEs: linear system of equations
  - We are looking at that case

**A. Orzan, A. Bousseau, H. Winnemöller, P. Barla, J. Thollot, D. Sales:**  
Diffusion Curves: A Vector Representation for Smooth-Shaded Images.  
In: *ACM Transactions on Graphics, SIGGRAPH 2008.*

# Initial Value Problem



# Fixed Spatial Boundaries



# Examples

## Heat Diffusion

$$f: \mathbb{R}^3 \supset (\Omega \times \mathbb{R}) \rightarrow \mathbb{R}$$

$$\partial_t f = -\lambda \underbrace{(\partial_x^2 + \partial_y^2)}_{\Delta f} f$$

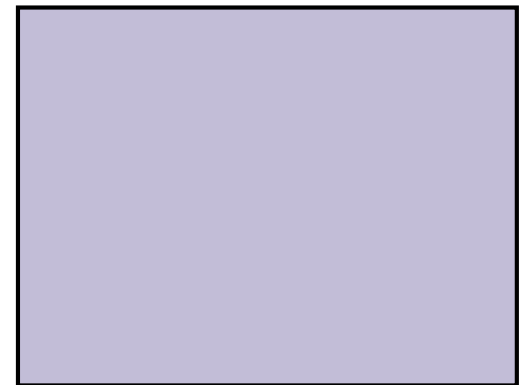
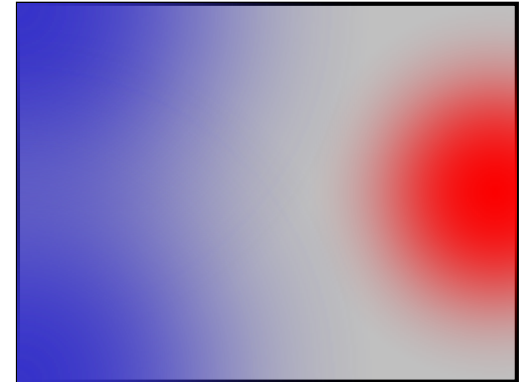
## Problem

- Initial heat distribution given
- Compute progression over time

## Class

- Second order, “parabolic”
- Smooths out details over time

$$f(\mathbf{x}, 0) = g(\mathbf{x})$$



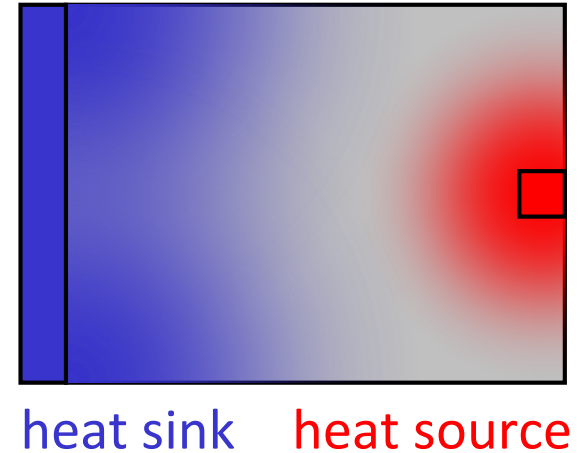
# Examples

## Diffusion Equation

$$f: \mathbb{R}^2 \supset \Omega \rightarrow \mathbb{R}$$

$$\underbrace{(\partial_x^2 + \partial_y^2)}_{\Delta f} f = 0$$

$$\forall \mathbf{x} \in \text{Boundary}: f(\mathbf{x}) = g(\mathbf{x})$$



## Problem

- Boundary conditions for heat given
- Compute steady state  $\partial_t = 0$



# Examples

## Wave Equation

$$f: \mathbb{R}^3 \supset (\Omega \times \mathbb{R}) \rightarrow \mathbb{R}$$

$$\partial_t^2 f = \lambda \underbrace{(\partial_x^2 + \partial_y^2)}_{\Delta f} f$$

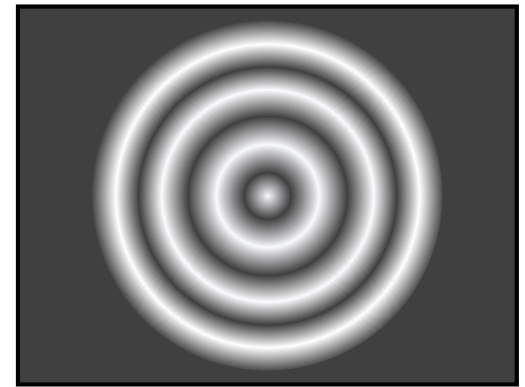
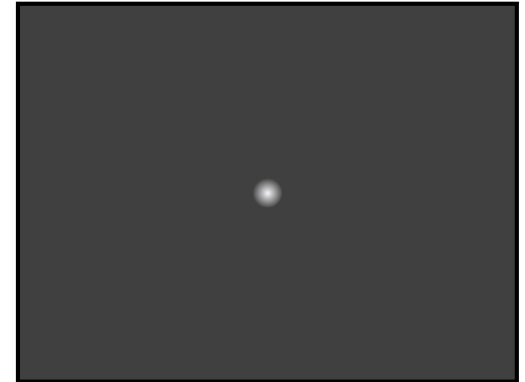
## Problem

- Driver function given (time/space)
- Compute progression over time

## Class

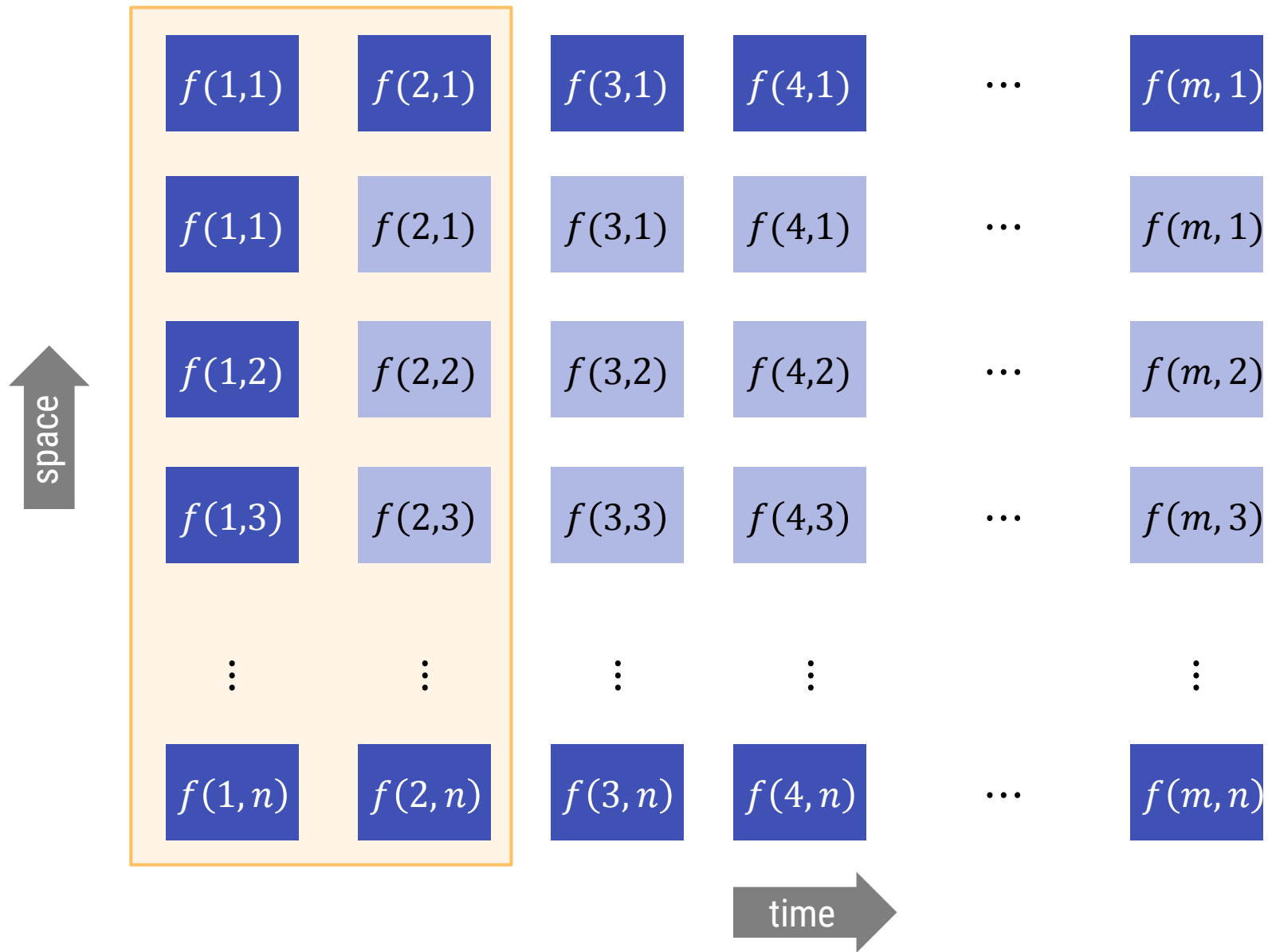
- Second order, “hyperbolic”
- Transports information through space (no information loss)

$$f(\mathbf{0}, t) = g(t)$$



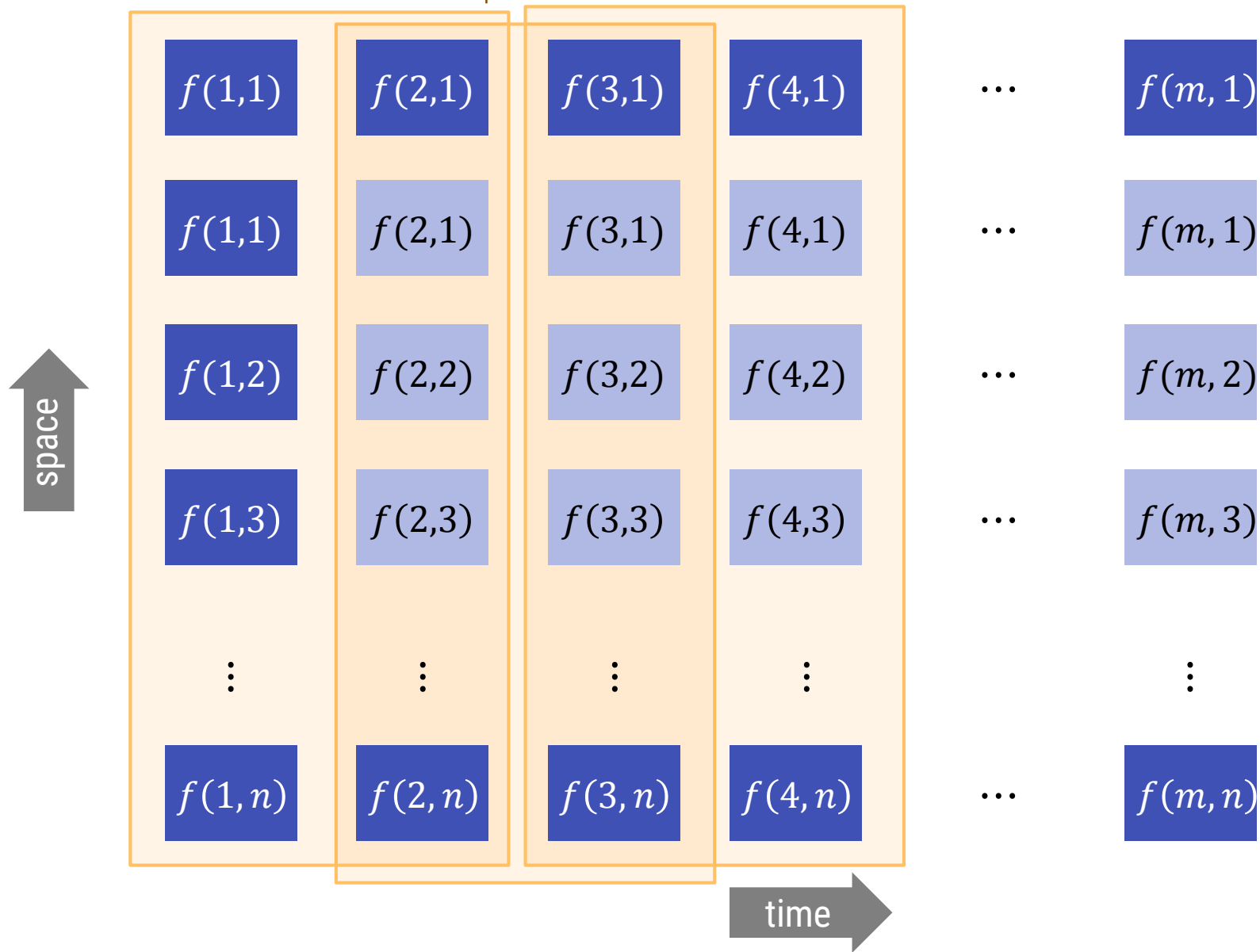
Solve system of equations  
for each time step

# Solver



Solve system of equations  
for each time step

# Solver



# Integral Equations

# Integral Equations

## Use integrals to construct $L$ :

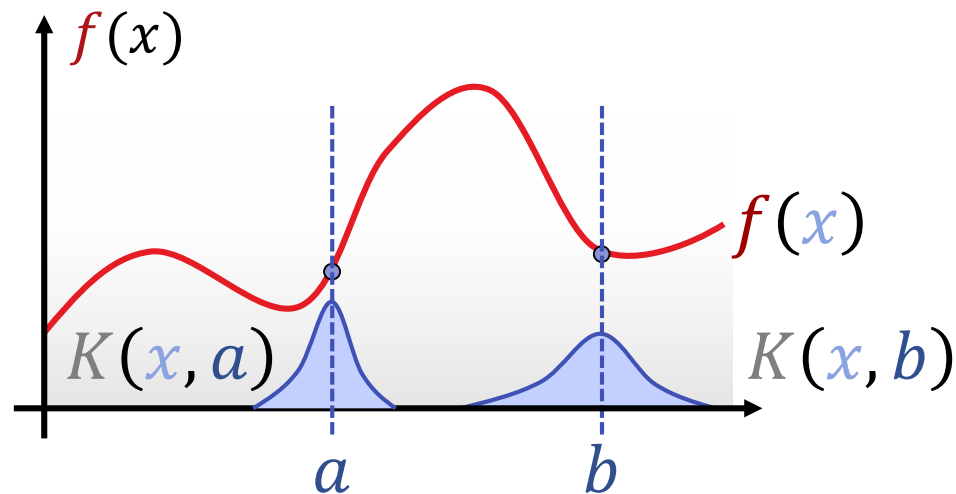
- For example:

$$f(x) = g(x) + \int_0^1 K(x, y) \cdot f(y) dy$$

- Given (known): functions  $K, g$
- Unknown: function  $f$
- As operator equation  $Lf = g$

$$L(f(x)) = \left[ f(x) - \int_0^1 K(x, y) \cdot f(y) dy \right]$$
$$g = g(x)$$

What does it do?

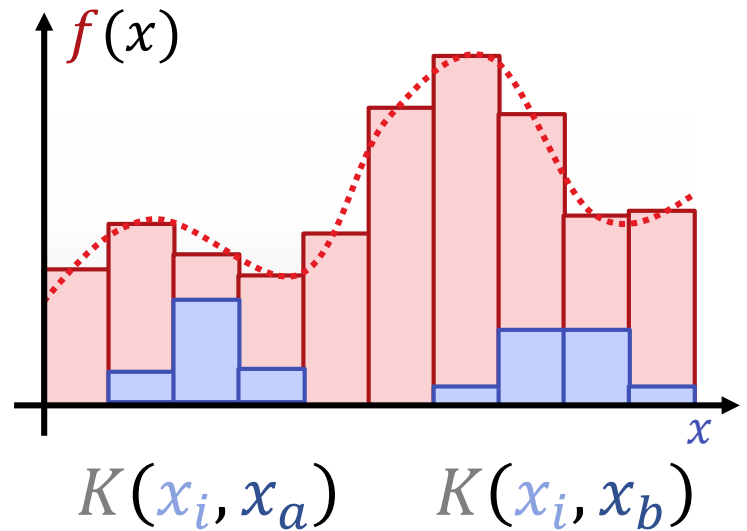
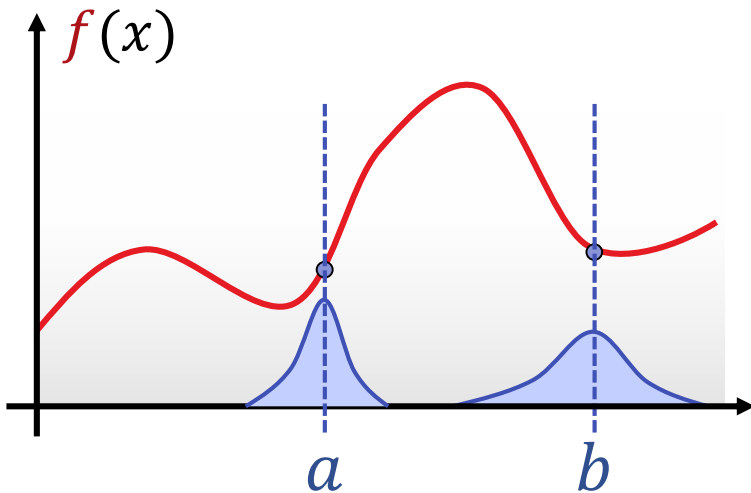


$$f(x) = g(x) + \int_0^1 K(x, y) \cdot f(y) dy$$

## Fredholm integral equation (2<sup>nd</sup> kind):

- Prescribe weighted averages of function values
- Add constant function

# Discrete Analogy

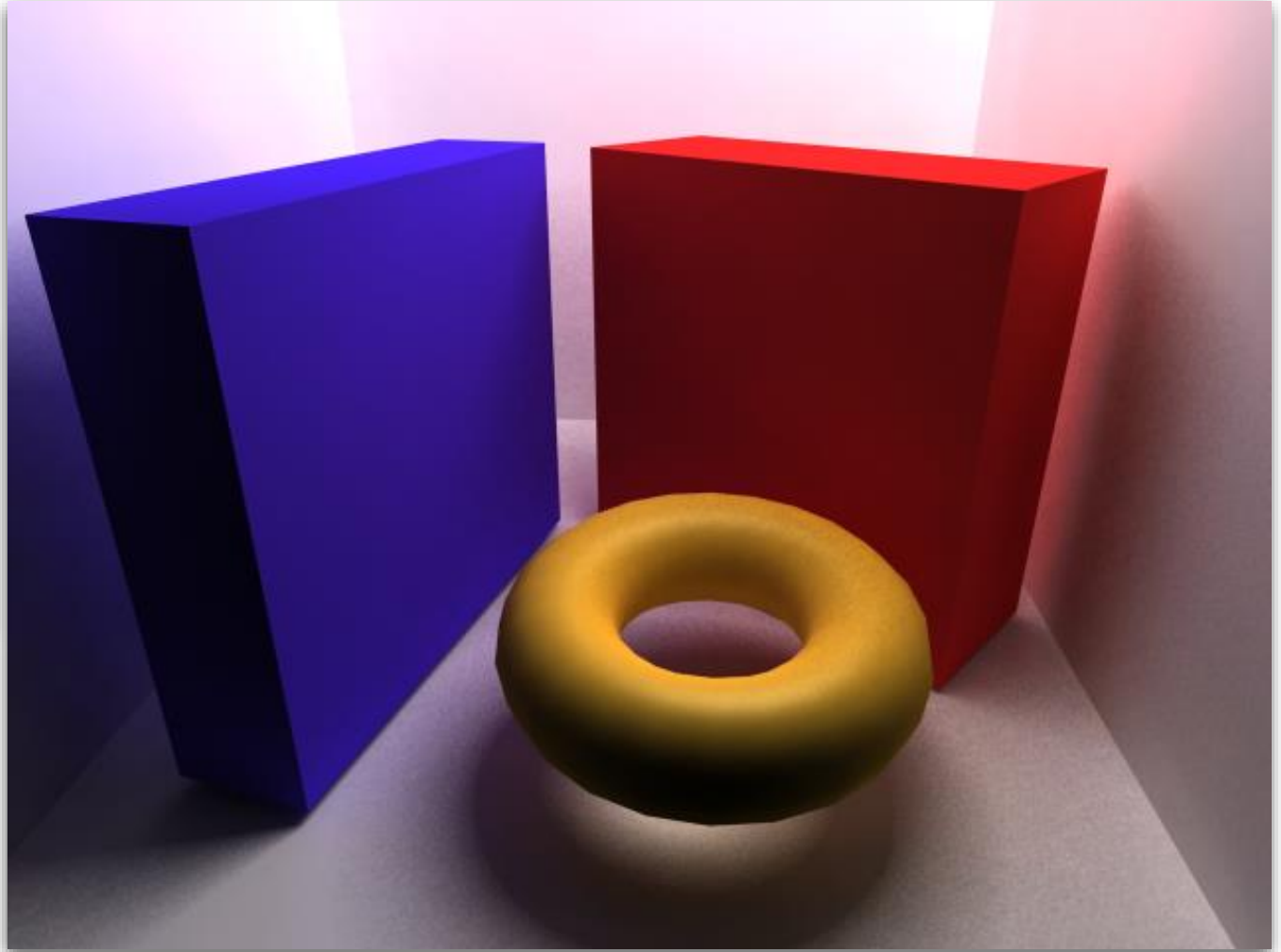
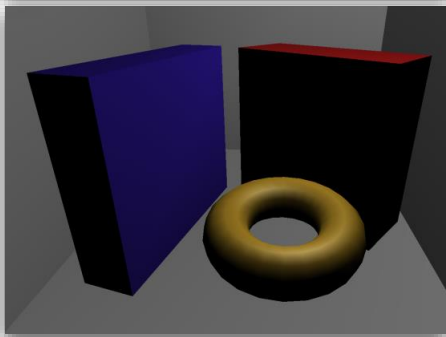
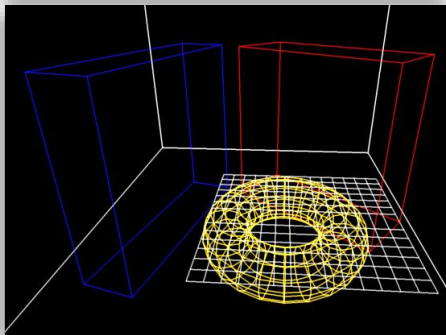
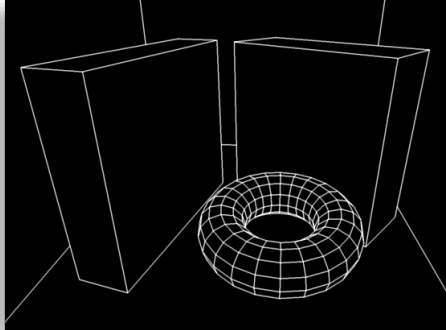


$$f(x) = g(x) + \int_0^1 K(x, y) \cdot f(y) dy$$

$$\mathbf{x} = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 2 \\ 3 \end{pmatrix} + \begin{pmatrix} 1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 3 & 3 & 1 \\ 0 & 0 & 1 & 2 & 0.5 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \mathbf{x}, \text{ i.e.,}$$

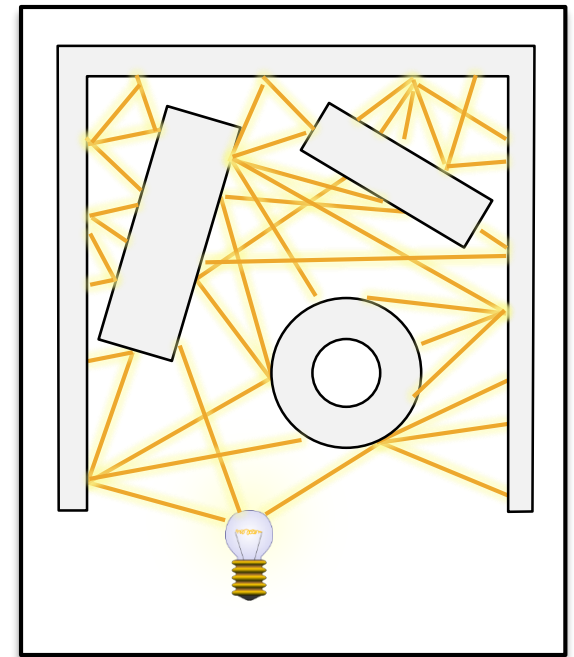
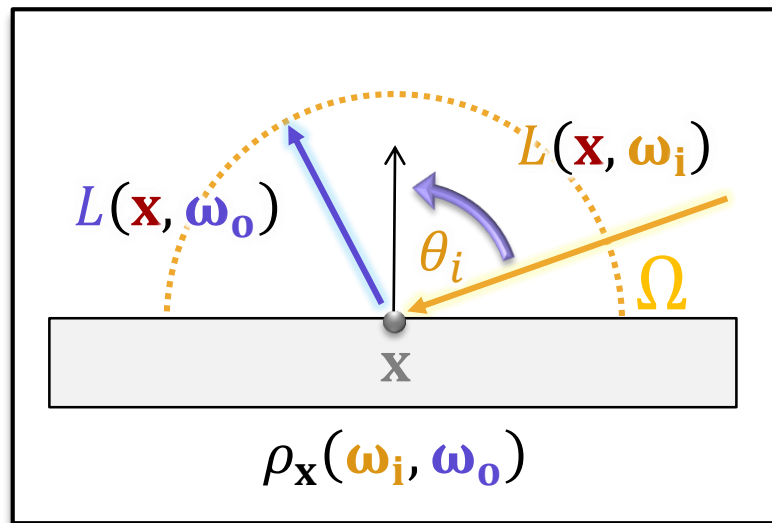
$$\left[ \mathbf{I} - \begin{pmatrix} 1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 3 & 3 & 1 \\ 0 & 0 & 1 & 2 & 0.5 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \right] \mathbf{x} = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 2 \\ 3 \end{pmatrix}$$

# “Global Illumination”





# Example: Rendering Equation



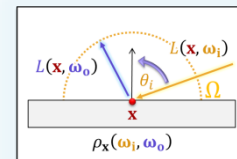
## Rendering Equation

$$L(\mathbf{x}, \omega_o) = \underbrace{E(\mathbf{x}, \omega_o)}_{\text{emission}} + \underbrace{\int_{\omega_i \in \Omega} [L(\mathbf{x}, \omega_i) \cdot \rho_{\mathbf{x}}(\omega_i, \omega_o) \cdot \cos \theta_i] d\omega_i}_{\text{reflection}}$$

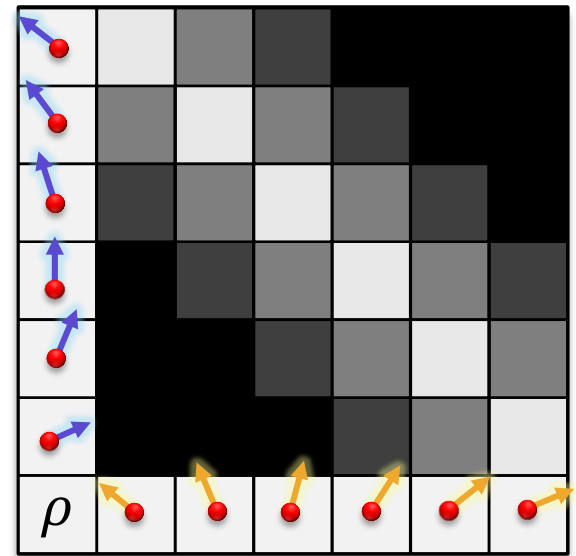
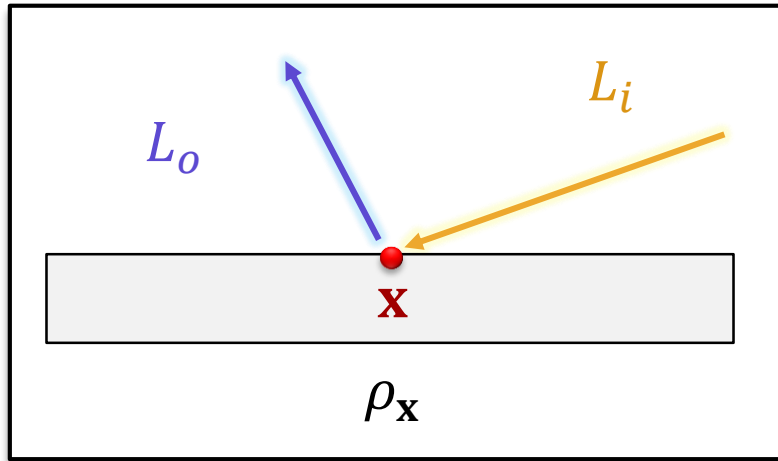
emission



reflection



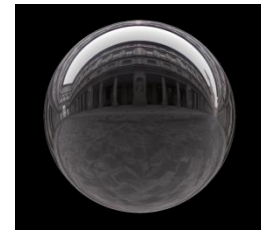
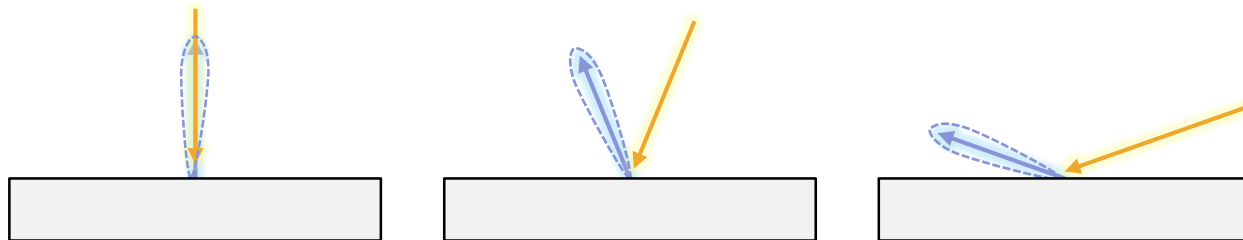
# Interaction with Surfaces



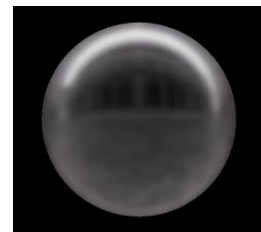
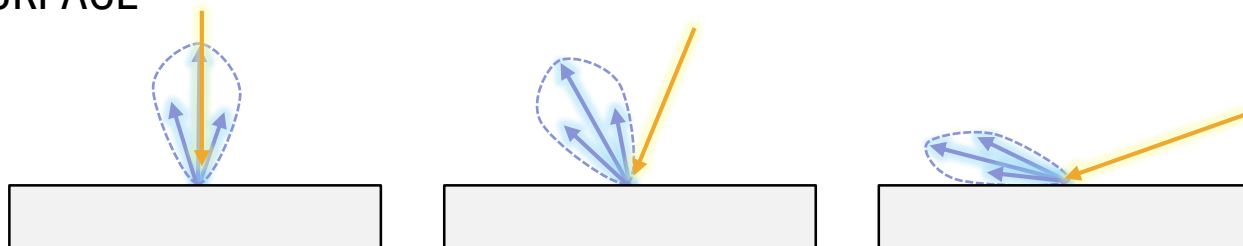
**Bi-direction Reflectance Distribution Function (BRDF)**

# Bidirectional Reflectance Distribution Function (BRDF)

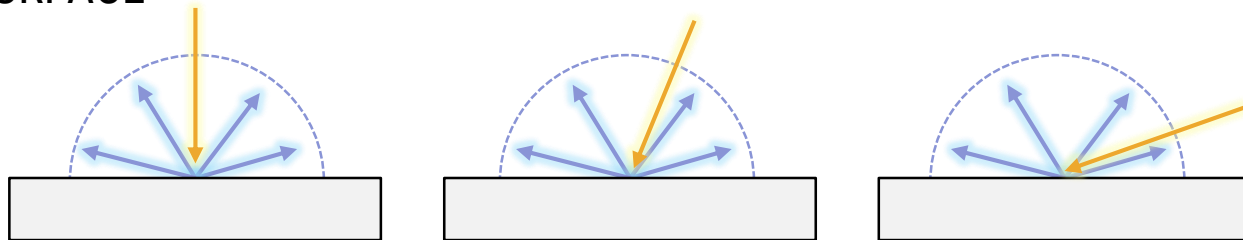
## MIRROR



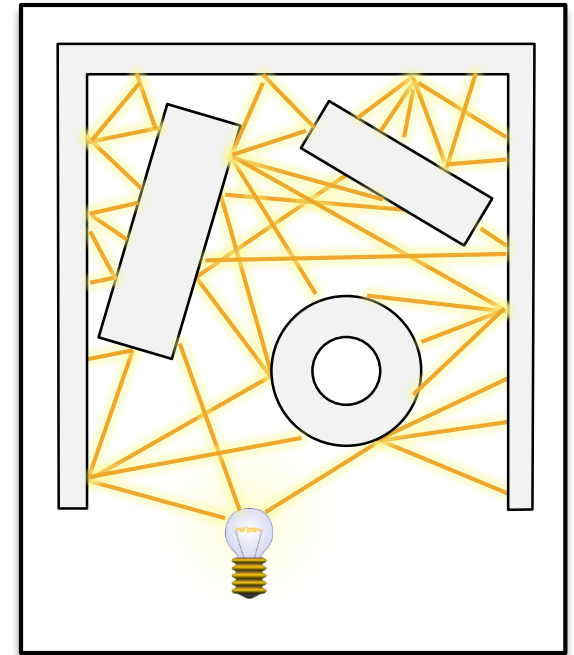
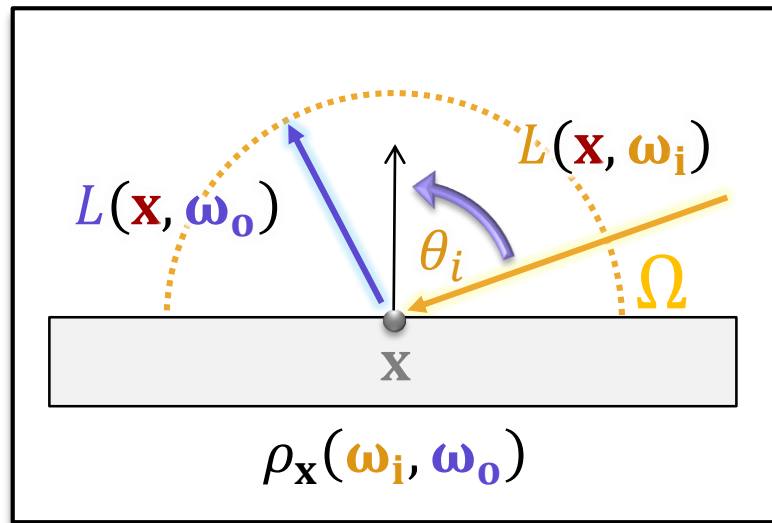
## GLOSSY SURFACE



## DIFFUSE SURFACE



# Example: Rendering Equation



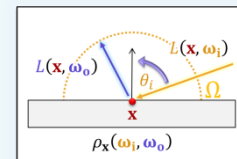
## Rendering Equation

$$L(\mathbf{x}, \omega_o) = \underbrace{E(\mathbf{x}, \omega_o)}_{\text{emission}} + \underbrace{\int_{\omega_i \in \Omega} [L(\mathbf{x}, \omega_i) \cdot \rho_{\mathbf{x}}(\omega_i, \omega_o) \cdot \cos \theta_i] d\omega_i}_{\text{reflection}}$$

emission



reflection



# Classification

# Linear Functional Equations

## Linear Functional Equation

- $Lf = g$  – solving a linear system
  - Discretization with “array”
    - “Finite differences” for differential equations
    - Replace  $f'(x)$  with  $[f(x_i) - f(x_{i-1})]/(x_i - x_{i-1})$
  - Discretization with linear ansatz: “finite elements”
  - Analytical solution?
    - If  $L$  is diagonalizable and we know the eigenfunctions: Diagonal system
    - Scaling of projections of  $g$  on eigenfunctions

# Linear Functional Equations

## Linear Time Evolution

- Linear ODE ( $\mathbf{x} \in \mathbb{R}^d$ )

- $\frac{d}{dt} \mathbf{x}(t) = \mathbf{A} \mathbf{x}$

- Solution

$$\mathbf{x}(t) = \exp(t \cdot \mathbf{A}) \mathbf{x}(0)$$

- $\mathbf{A}$  diagonalizable?

$$\begin{aligned} & \mathbf{U} \exp(t \cdot \mathbf{D}) \mathbf{U}^T \mathbf{x}(0) \\ &= \mathbf{U} \begin{pmatrix} \exp(\lambda_1)^t & & \\ & \ddots & \\ & & \exp(\lambda_d)^t \end{pmatrix} \mathbf{U}^T \mathbf{x}(0) \end{aligned}$$

- $\mathbf{A}$  not diagonalizable? – Jordan Normal Form
- Inhomogeneous case similar\*

\*) see e.g. <https://de.wikipedia.org/wiki/Matrixexponential>

# Linear Functional Equations

## Linear Time Evolution

- Linear PDE with simple (Markovian) time evolution

- $\frac{d}{dt} f = Lf$

- Solution

$$\mathbf{x}(t) = \exp(t \cdot \mathbf{L}) \mathbf{x}(0)$$

- $L$  diagonalizable? (“self-adjoint” = symmetric?)

- Eigenfunctions  $u_i$

$$f(t) = \sum_i u_i(t) \cdot \exp(t \cdot \lambda_1) \langle f(0), u_i \rangle$$



# Linear Functional Equations

## Shift invariant Operators

- Linear Operator shift invariant?
  - (Time invariant) ordinary differential equations
  - (Spatially/temporally uniform) partial differential equations
- We know the eigenbasis already!
  - Fourier-basis
  - For example in time evolution: We can write down the solution by exponential scaling of Fourier-coefficients
  - See tutorials (heat equation)