

Parameter estimation in ODE models: data interpolation, differential algebra, polynomial system solving

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Joint work with: Oren Bassik, Yosef Berman, Soo Go, Hoon Hong, Ilia Ilmer, Chris Rackauckas, Pedro Soto, and Chee Yap

Implementation:

<https://github.com/ilialmer/ParameterEstimation.jl>



Plan

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- Toy example to explain approach

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- Outline of approach

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- Future directions

Toy example

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$$y(t) = 1.2, 2.6, 1.2, 5.8$$

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Challenge: what are $y'(0)$ and $y''(0)$?

Example

$$\begin{cases} x'_1 = k_1 x_1 - k_2 x_1 x_2 \\ x'_2 = -k_3 x_2 + k_4 x_1 x_2 \end{cases}$$



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Which of the parameters can we find from $x_1(t)$?

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Can be checked using software:

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Which of the parameters can we find from $x_1(t)$?

Can be checked using software:

Yes: k_1, k_3, k_4

No: k_2

Example - modified

$$\begin{cases} x'_1 = k_1 x_1 - k_2 x_1 x_2 \\ x'_2 = -k_2 x_2 + k_3 x_1 x_2 \end{cases}$$



- x_1 - prey
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- x_1 - prey
- x_2 - predators

We can now in principle find all parameters

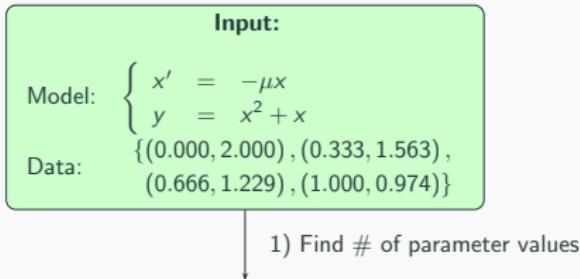
Flowchart

Input:

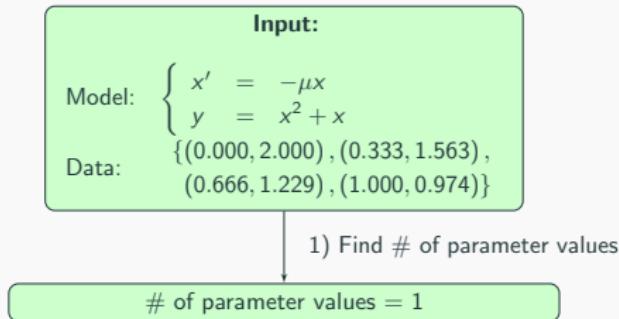
Model:
$$\begin{cases} x' = -\mu x \\ y = x^2 + x \end{cases}$$

Data: $\{(0.000, 2.000), (0.333, 1.563), (0.666, 1.229), (1.000, 0.974)\}$

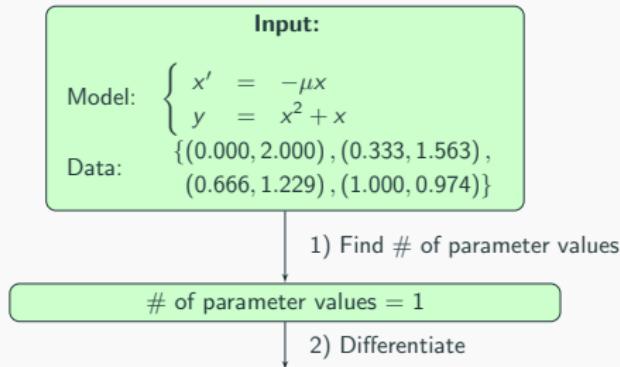
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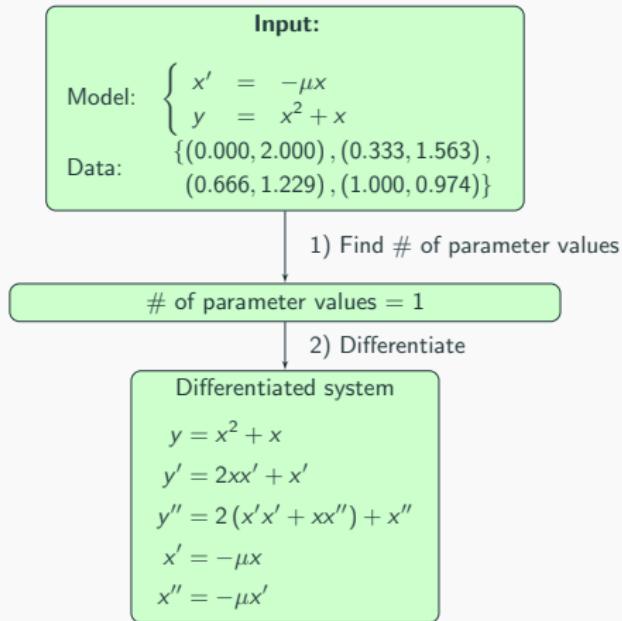
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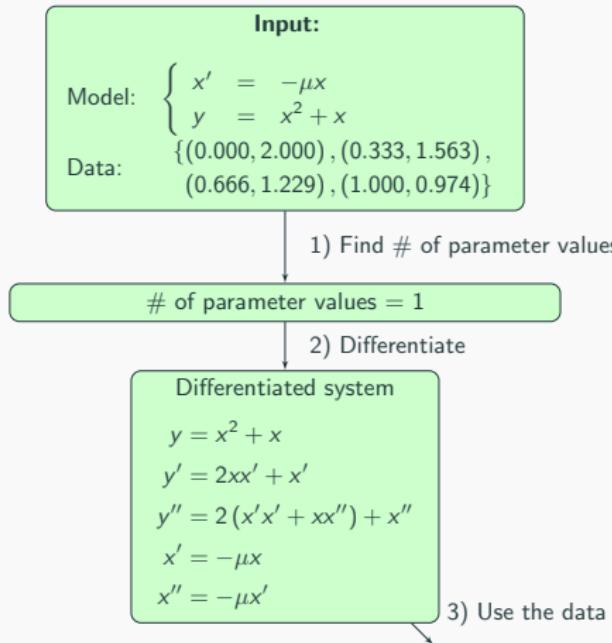
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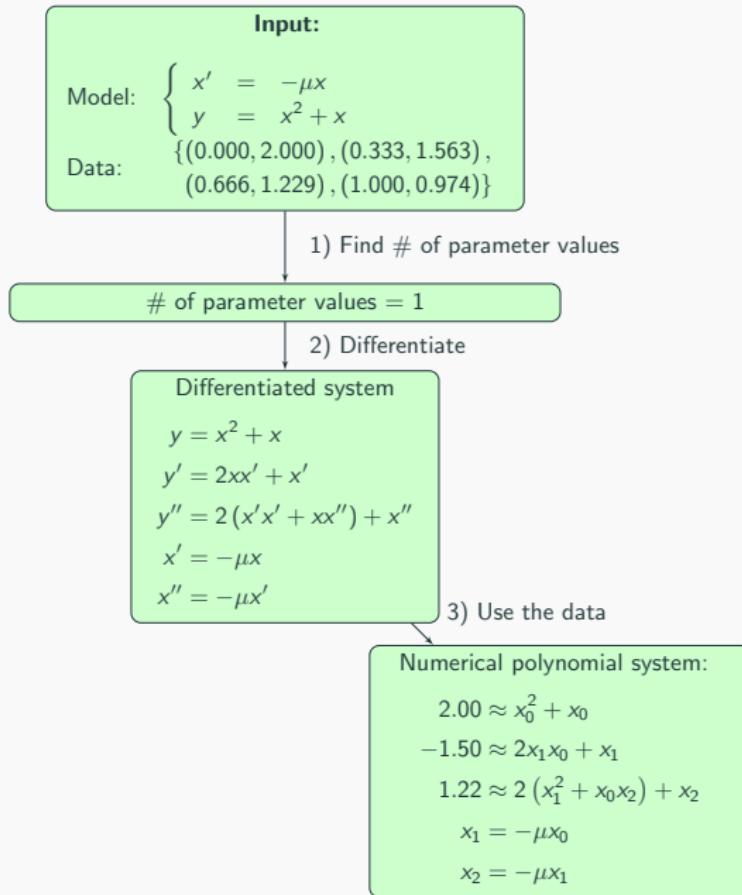
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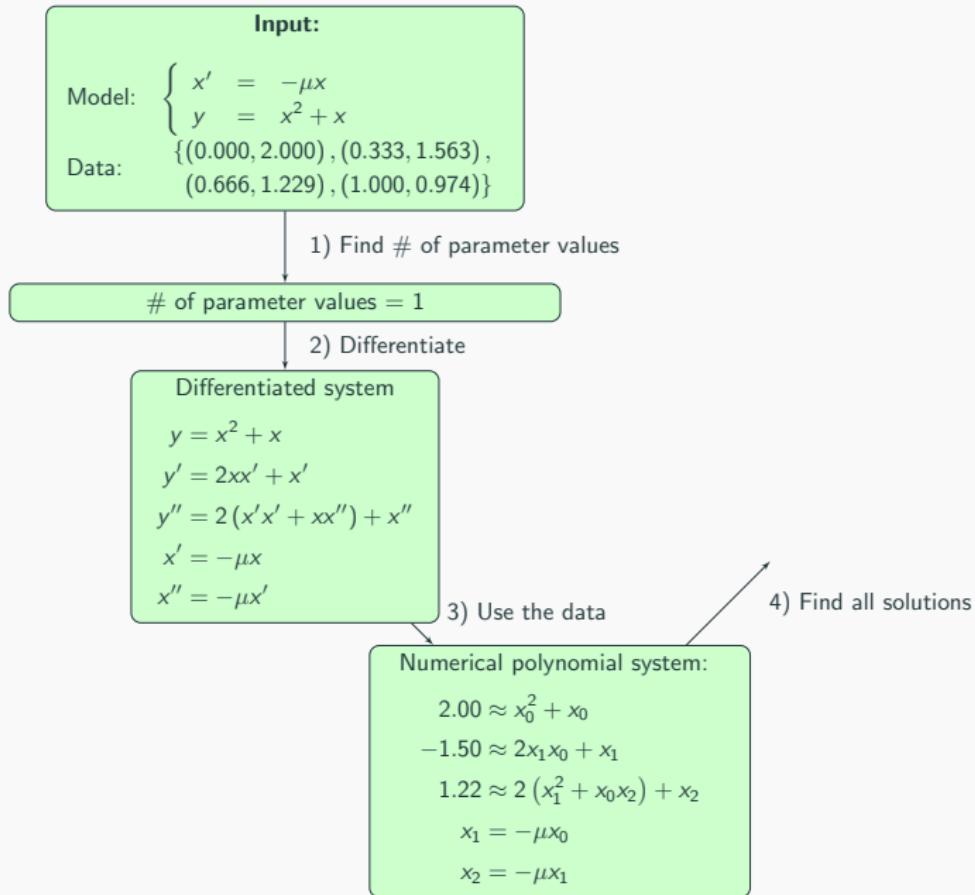
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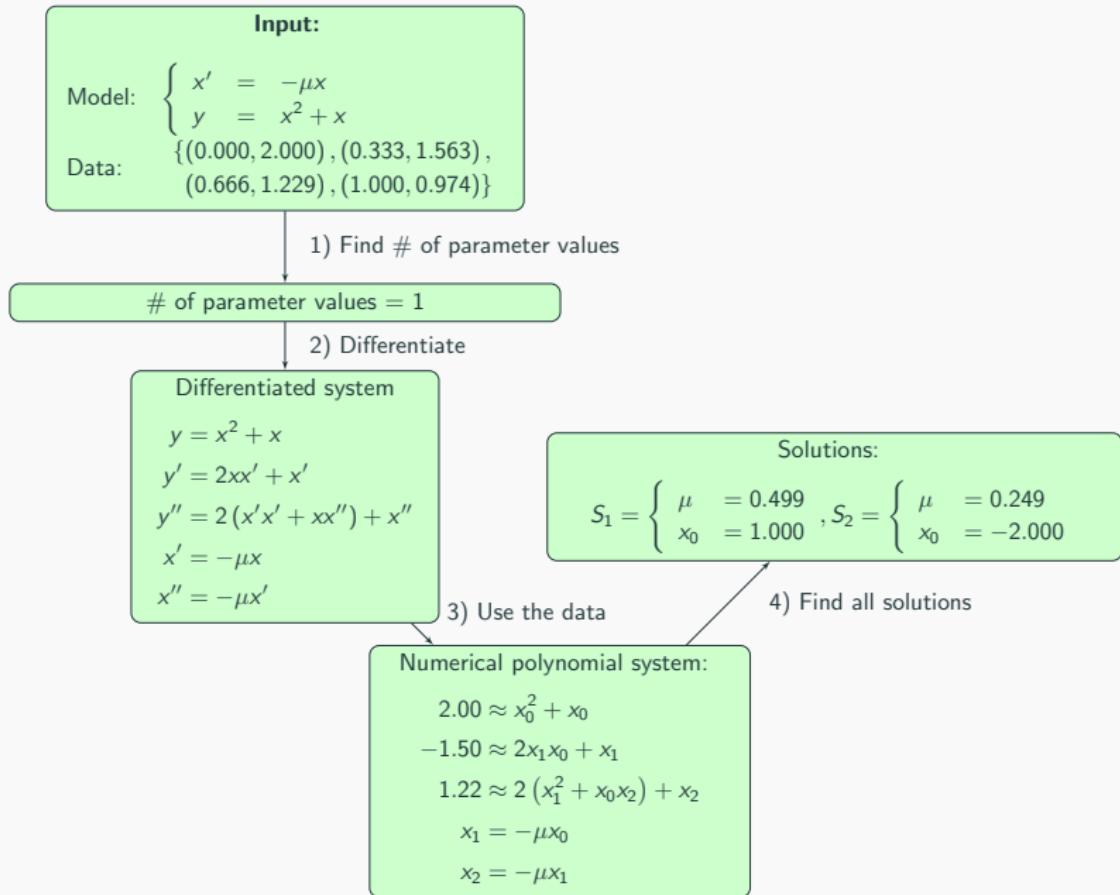
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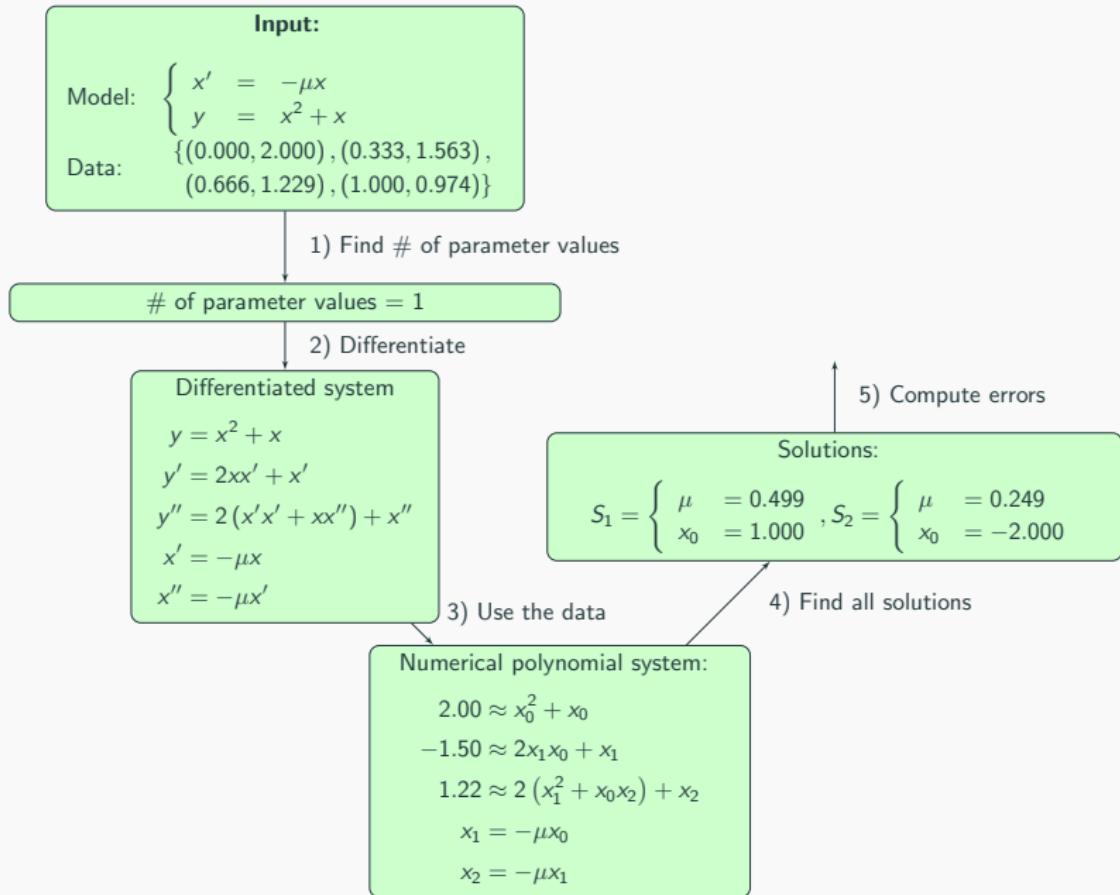
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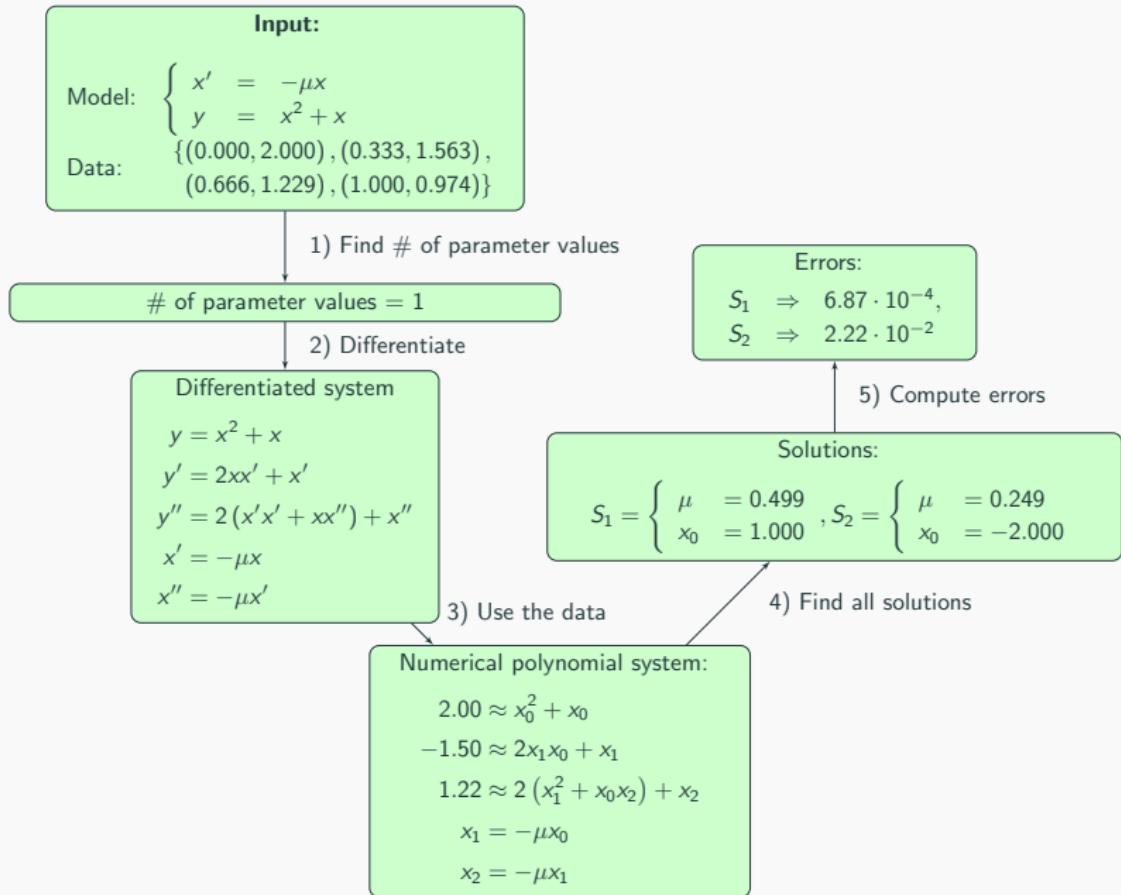
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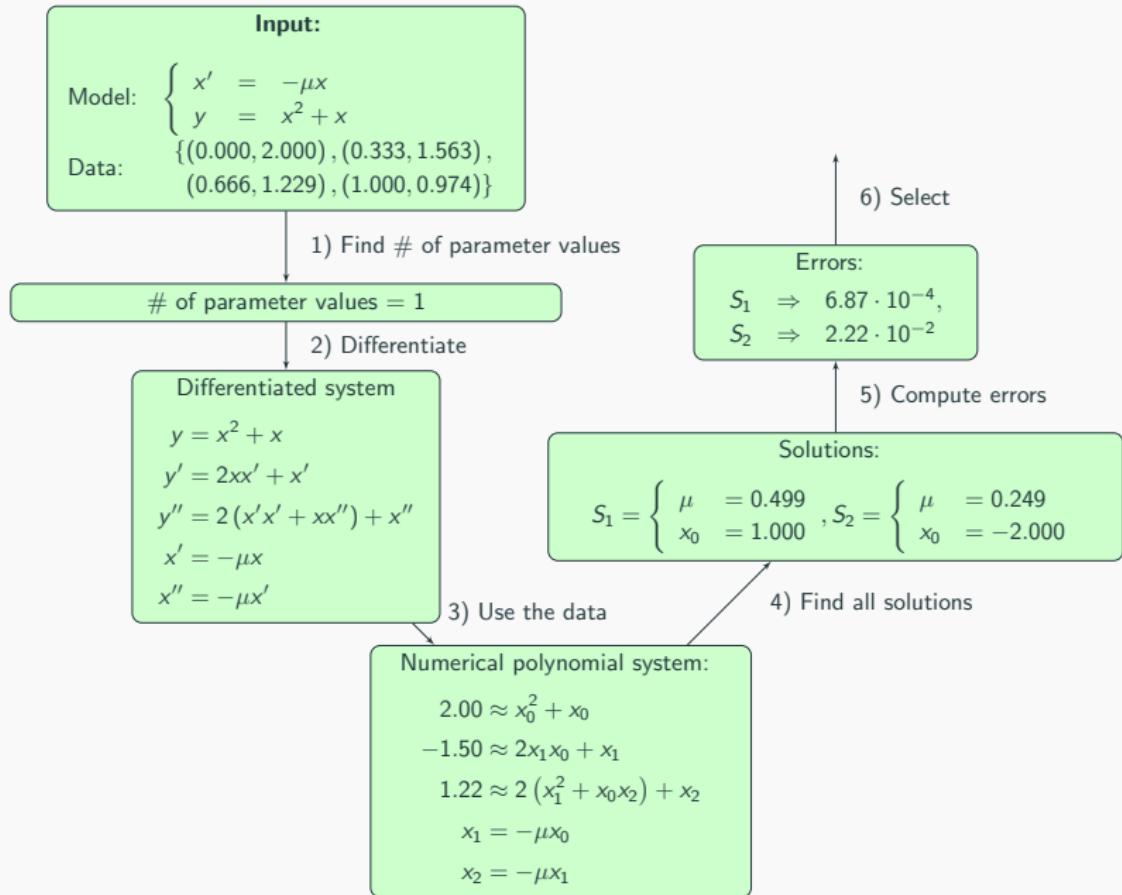
Flowchart



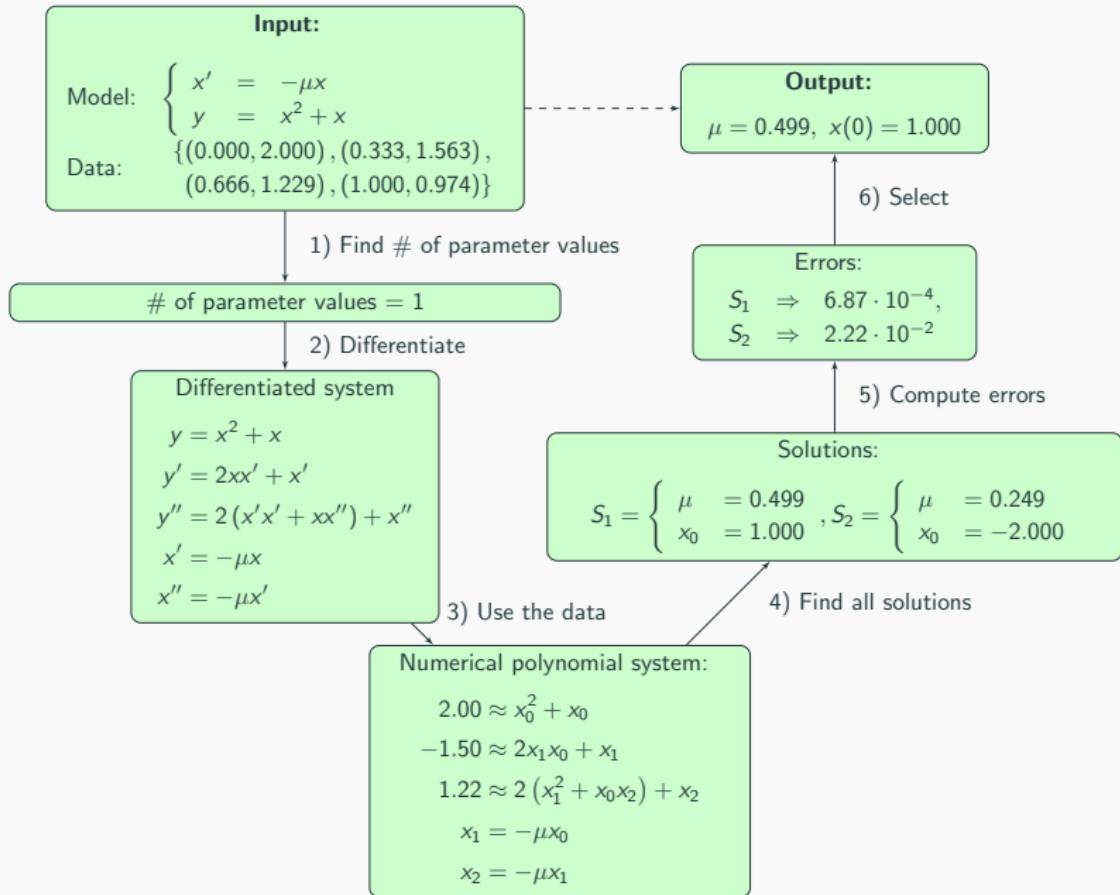
Flowchart



Flowchart



Flowchart



Using software in Julia

Using software in Julia

```
using ParameterEstimation
using ModelingToolkit

# Input:
# -- Differential model
@parameters mu
@variables t x(t) y(t)
D = Differential(t)
@named Sigma = ODESSystem([D(x) ~ -mu * x],
                           t, [x], [mu])
outs = [y ~ x^2 + x]

# -- Data
data = Dict(
    "t"      => [0.000, 0.333, 0.666, 1.000],
    x^2 + x => [2.000, 1.563, 1.229, 0.974])

# Run
res = estimate(Sigma, outs, data);
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The software returns this result:

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```

The software returns this result:

```
# Output:
Parameter(s)      : mu    = 0.499
Initial Condition(s): x(t) = 1.000
```

Performance

Mean of Relative Errors in %

Software		IQM			SciML			AMIGO2			Parameter Estimation.jl
Search Range		[0,1]	[0,2]	[0,3]	[0,1]	[0,2]	[0,3]	[0,1]	[0,2]	[0,3]	Any
Models	Harmonic, eq. (2)	66.3	68.7	101.9	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	Van der Pol, eq. (4)	7.5	0.0	9.6	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	FitzHugh-Nagumo, eq. (6)	82.4	131.1	210.3	8.3	15.9	20.4	0.0	0.0	0.2	0.0
	HIV, eq. (8)	85.6	104.2	144.7	14.8	36.9	102.9	18.5	43.2	28.4	0.0
	Mammillary 3, eq. (10)	76.1	104.3	128.2	11.9	21.5	24.7	0.0	0.0	0.0	0.0
	Lotka-Volterra, eq. (12)	72.2	75.4	75.4	54.7	76.9	111.7	13.6	112.6	29.7	0.0
	Crauste, eq. (14)	99.6	125.2	179.7	47.3	72.9	226.4	2.6	105.7	51.7	0.0
	Biohydrogenation, eq. (16)	93.7	184.0	130.3	77.7	151.1	306.6	17.8	13.5	28.3	0.0
	Mammillary 4, eq. (18)	94.1	109.6	105.3	66.2	67.7	118.8	29.8	56.7	59.0	0.1
	SEIR, eq. (20)	132.3	230.8	338.9	18.4	60.6	118.1	25.9	40.7	50.8	0.0

HIV model

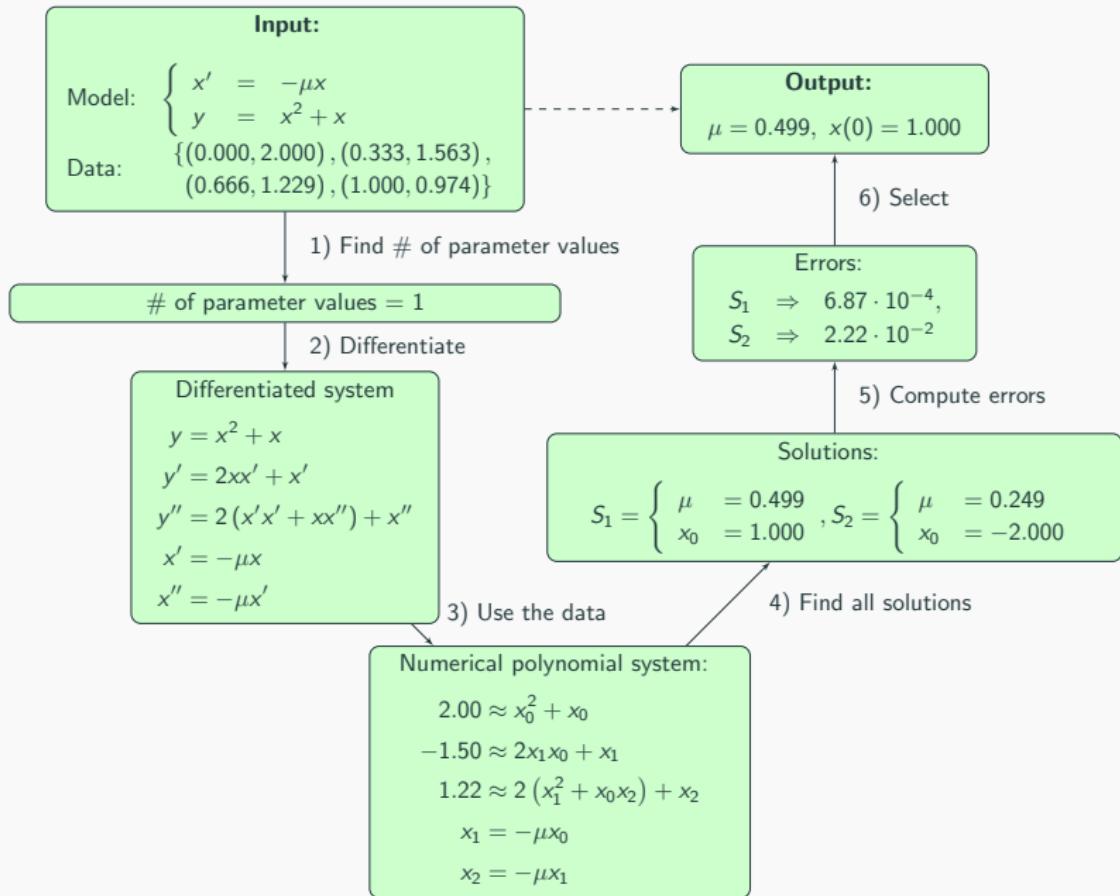
HIV infection dynamics during interaction with immune system
in various treatments.

HIV model

HIV infection dynamics during interaction with immune system
in various treatments.

$$\begin{cases} \dot{x} = \lambda - d x - \beta x v \\ \dot{y} = \beta x v - a y \\ \dot{v} = k y - u v \\ \dot{w} = c x y w - c q y w - b w \\ \dot{z} = c q y w - h z \end{cases}$$
$$\begin{cases} y_1 = w, \quad y_2 = z \\ y_3 = x, \quad y_4 = y + v \end{cases}$$

What can be proven?



Challenges and Future Steps

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- Proofs

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- Improve quality of derivative estimates

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- Noise