

COMBINE16, Newcastle

# **AMICI: An ODE simulation framework for sensitivity analysis of large-scale models**

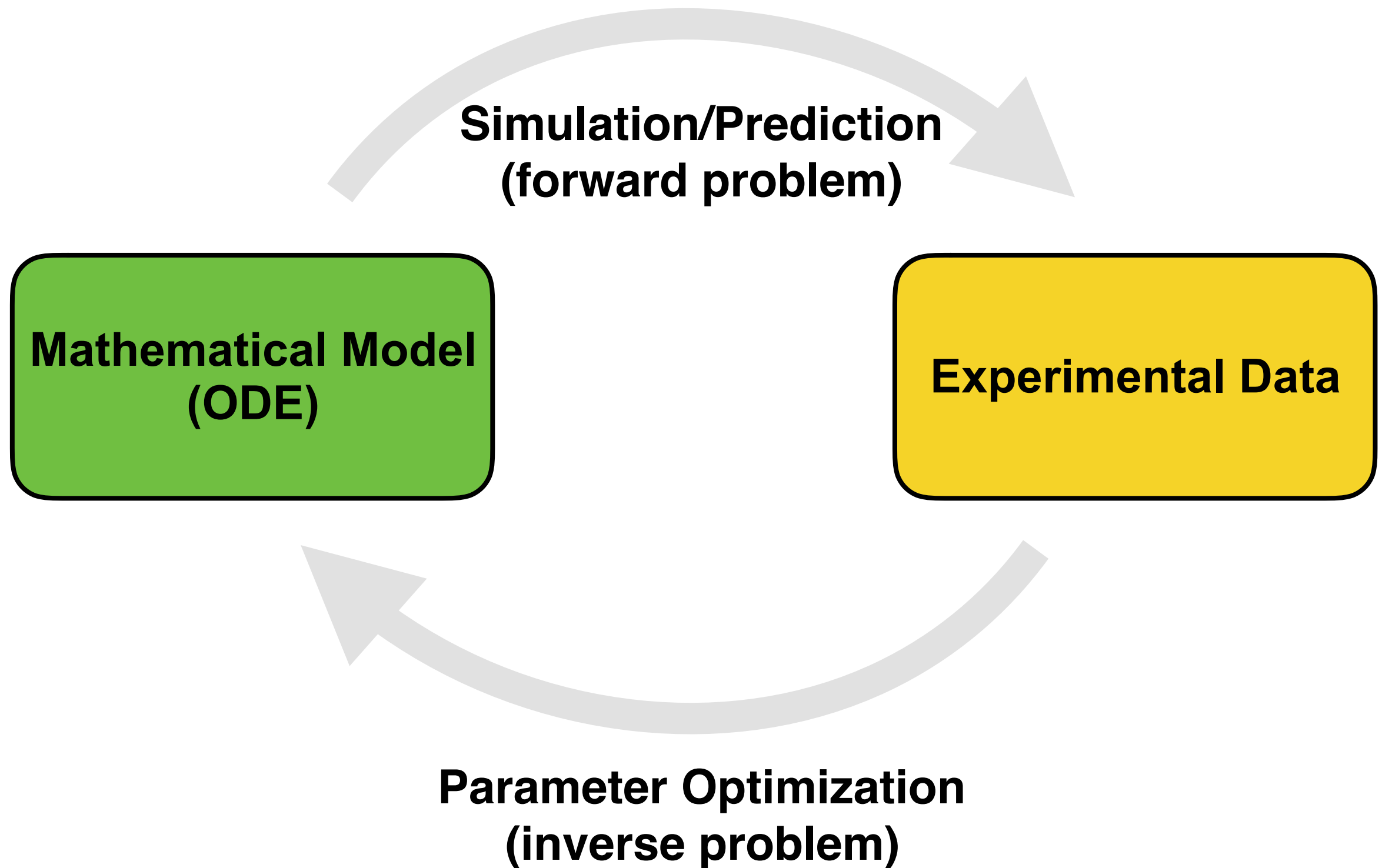
**Fabian Fröhlich**<sup>1,2</sup>, Jan Hasenauer<sup>1,2</sup>

<sup>1</sup>Helmholtz Zentrum München, Institute of Computational Biology

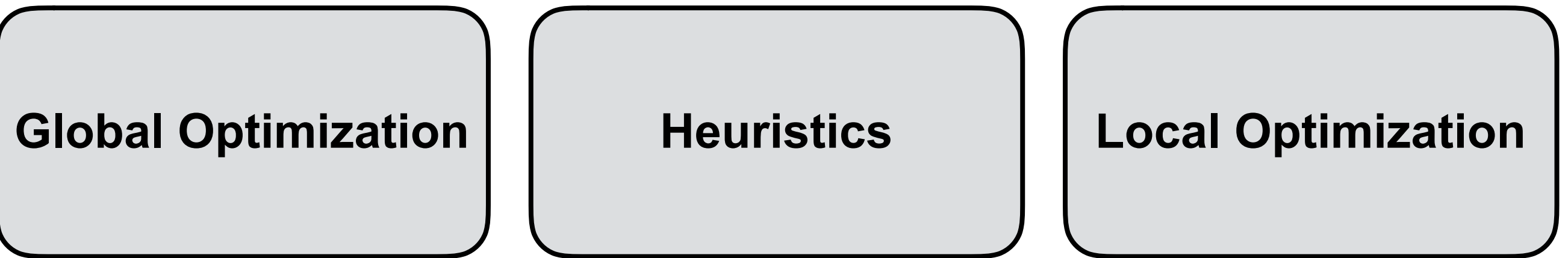
<sup>2</sup>Technische Universität München, Center for Mathematics



# Forward/Inverse Problem



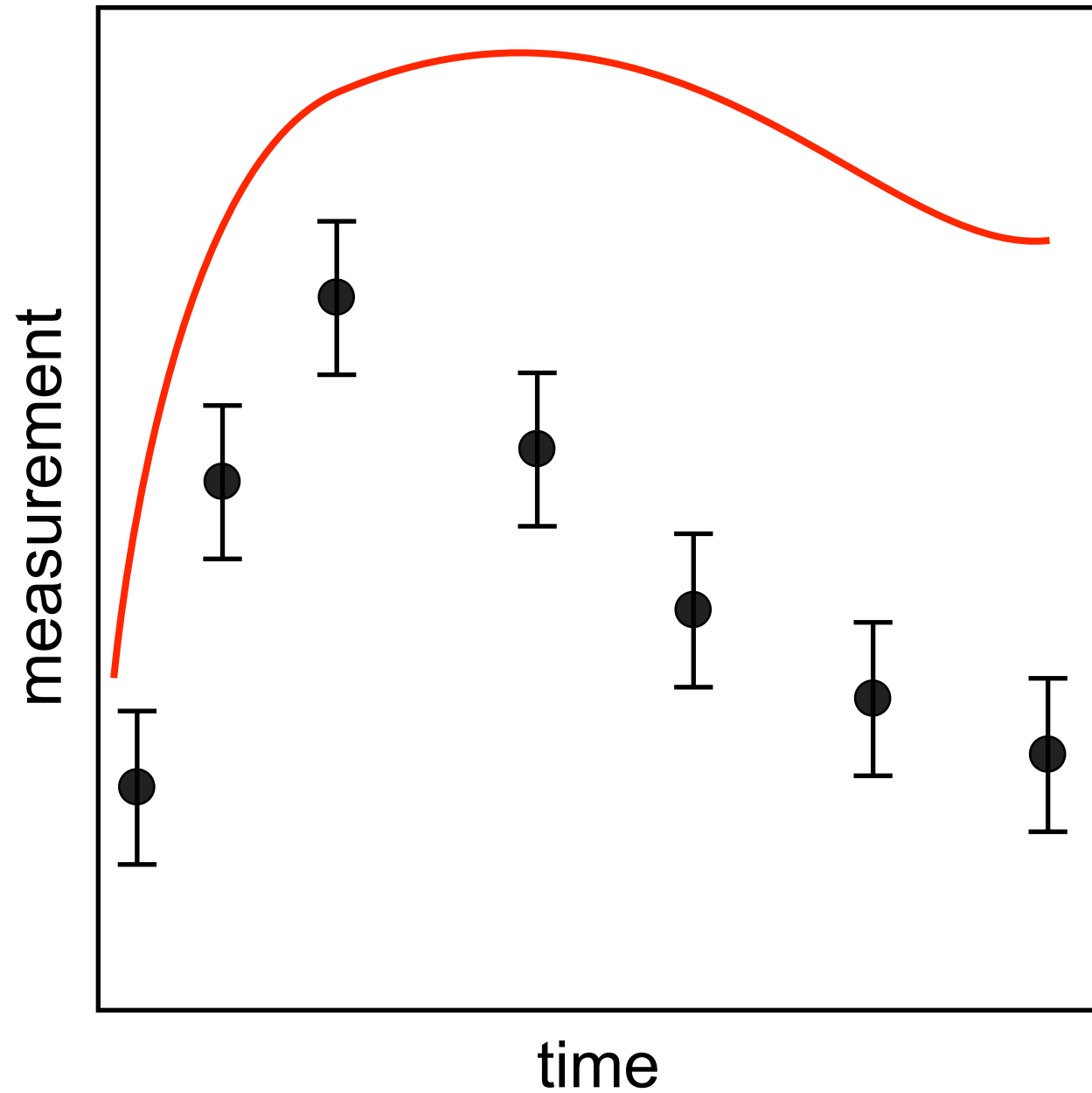
# Optimization Algorithms



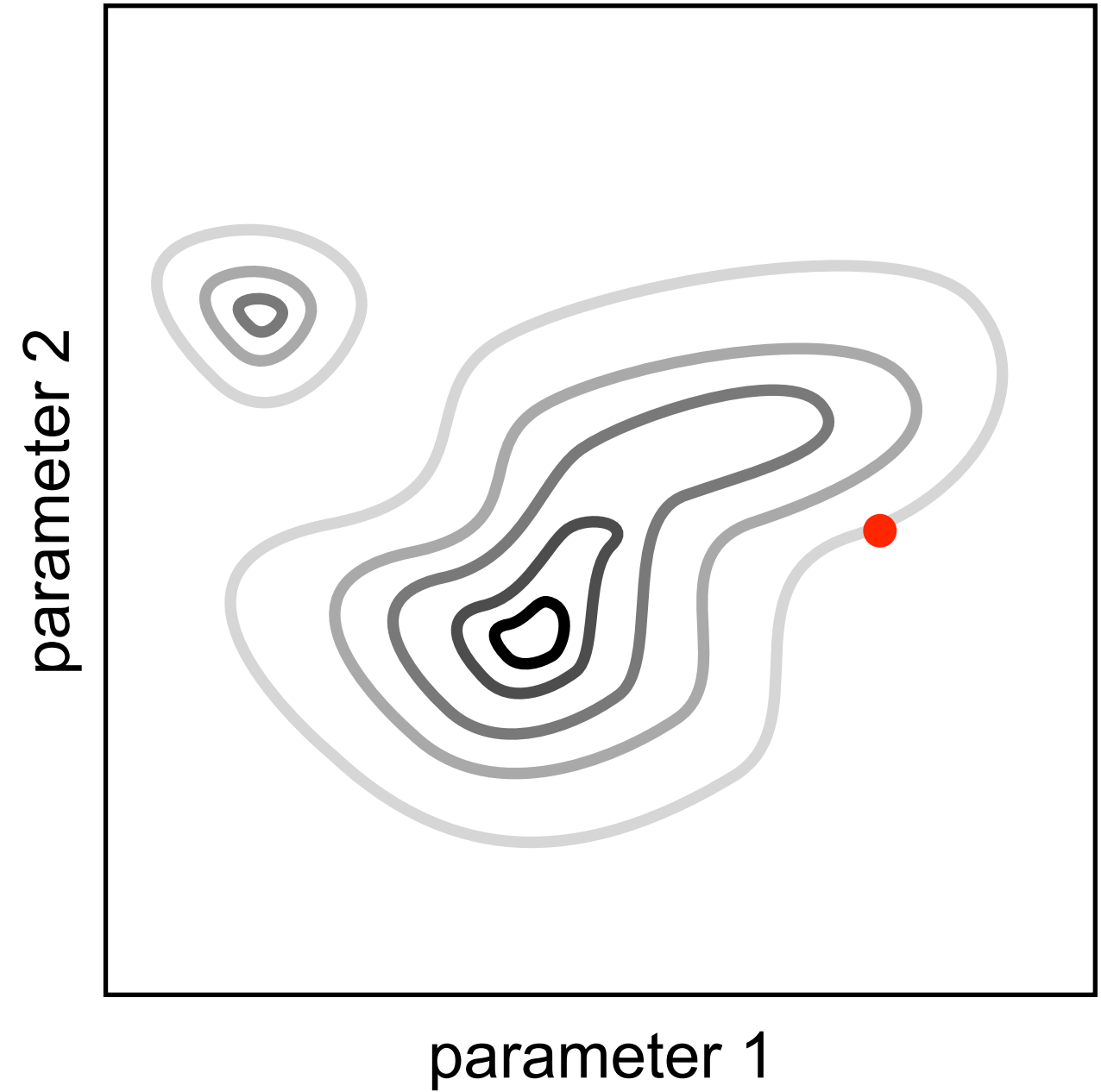
**in our group:** multi-start local  
gradient based optimization

# Parameter Estimation

## Model-data comparison

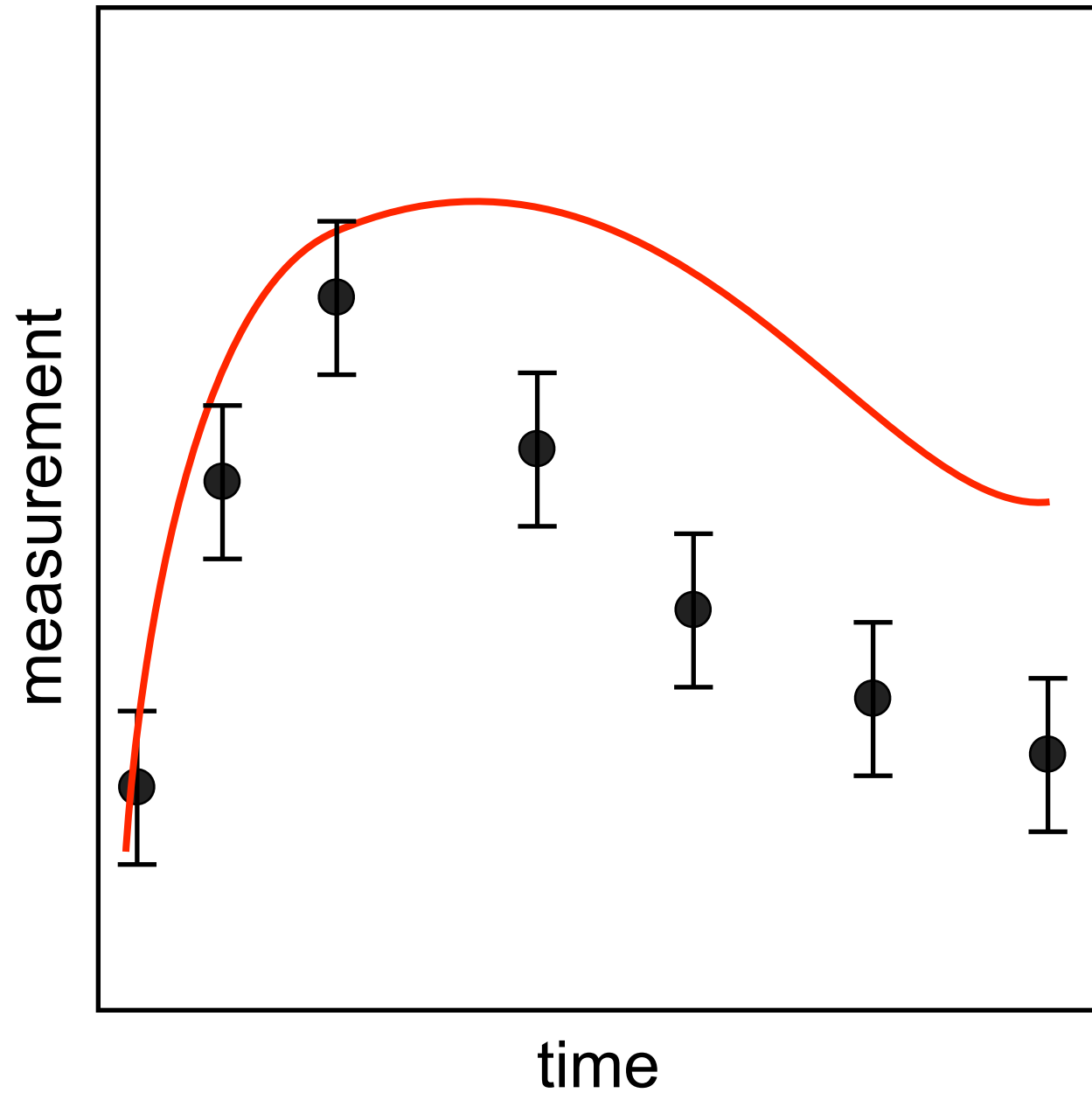


## Objective function landscape

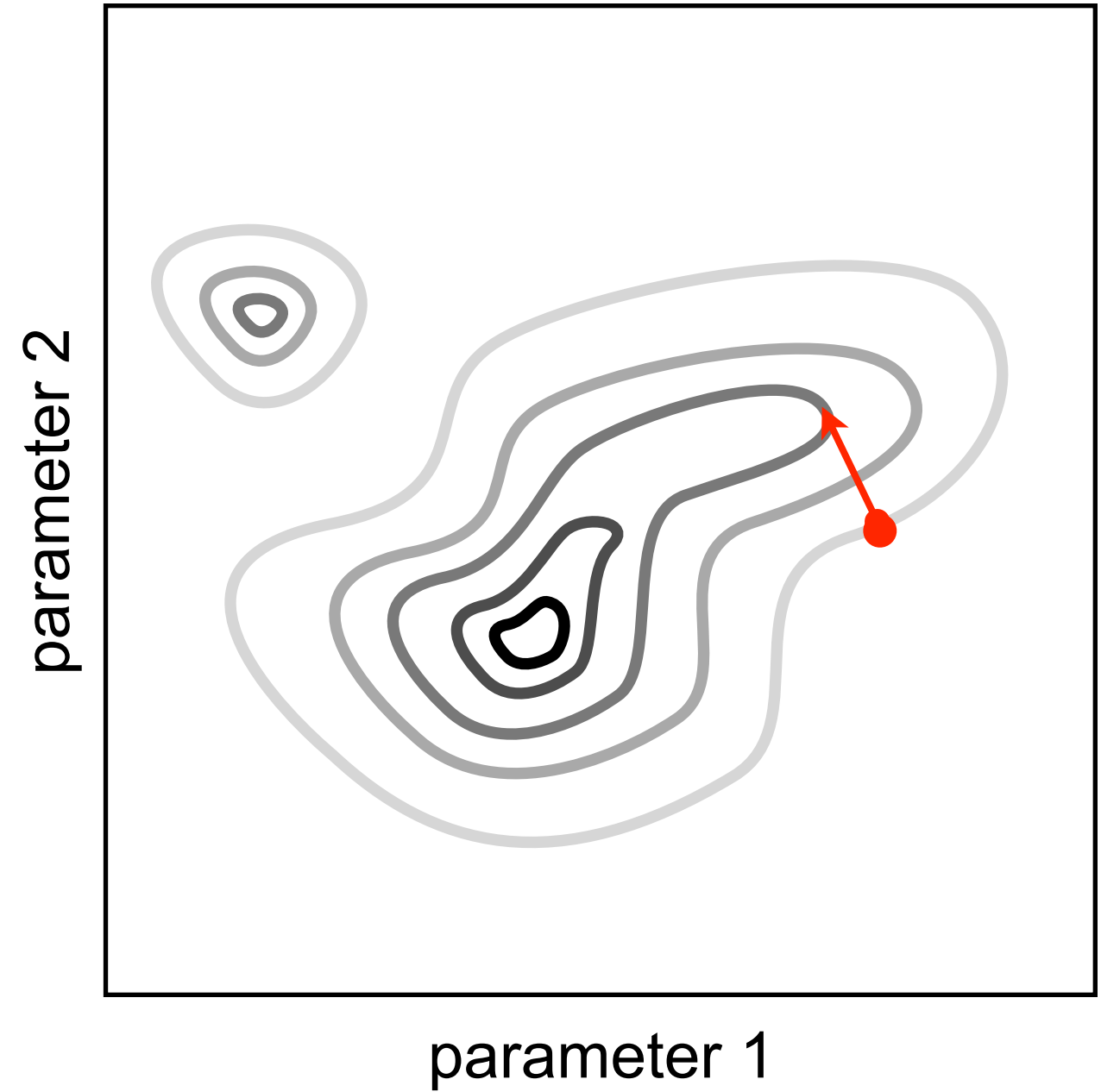


# Parameter Estimation

## Model-data comparison

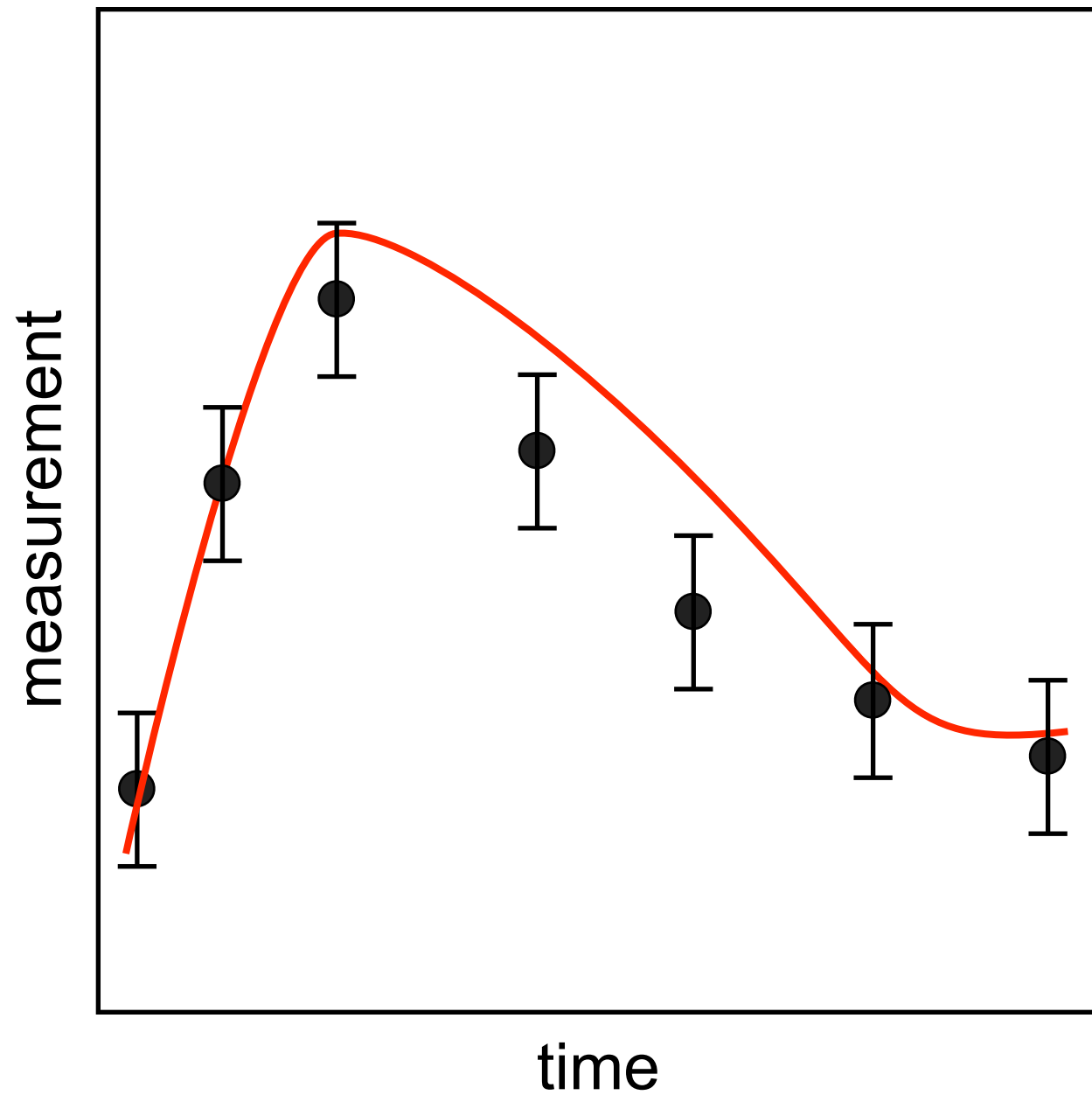


## Objective function landscape

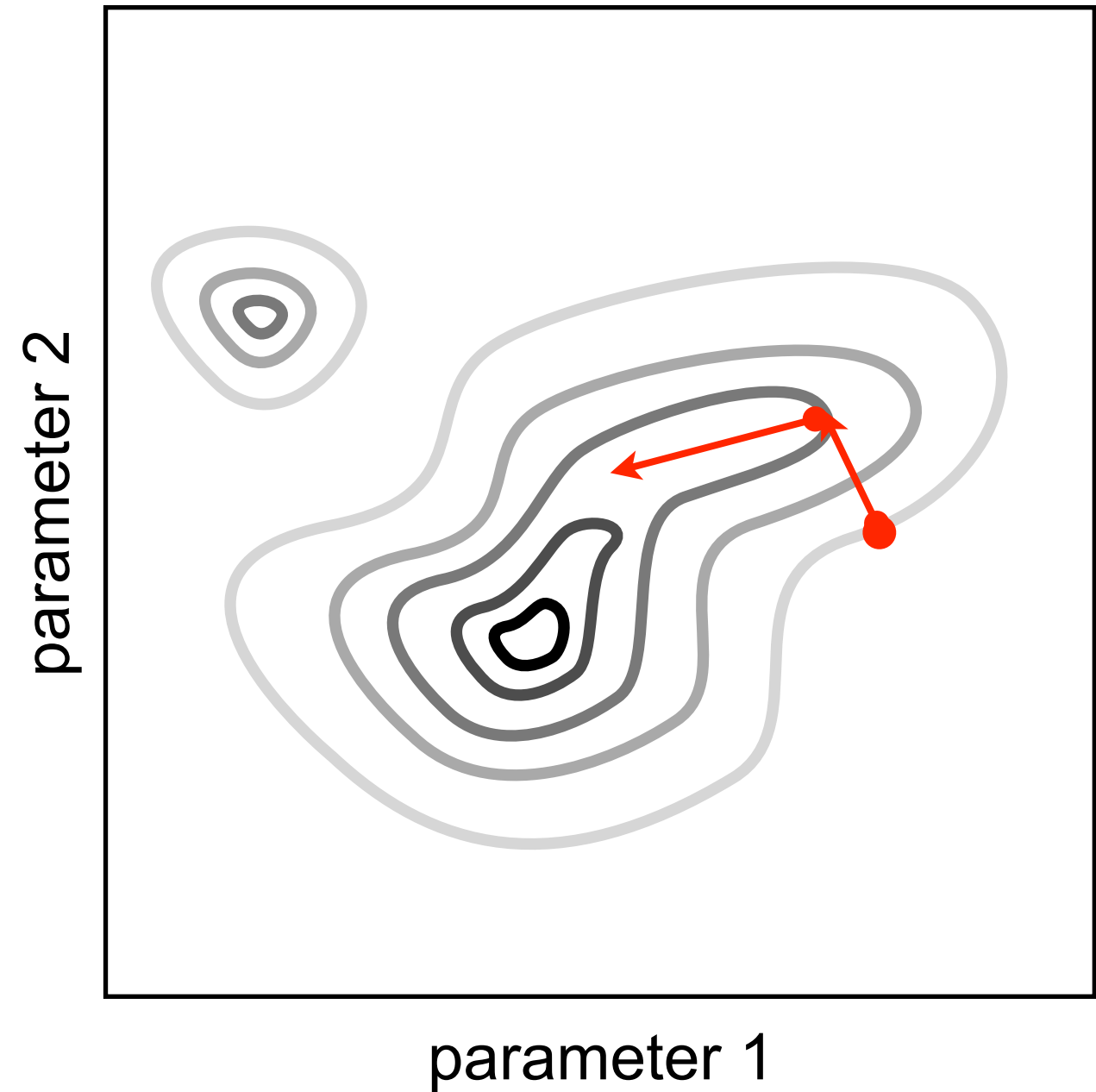


# Parameter Estimation

## Model-data comparison

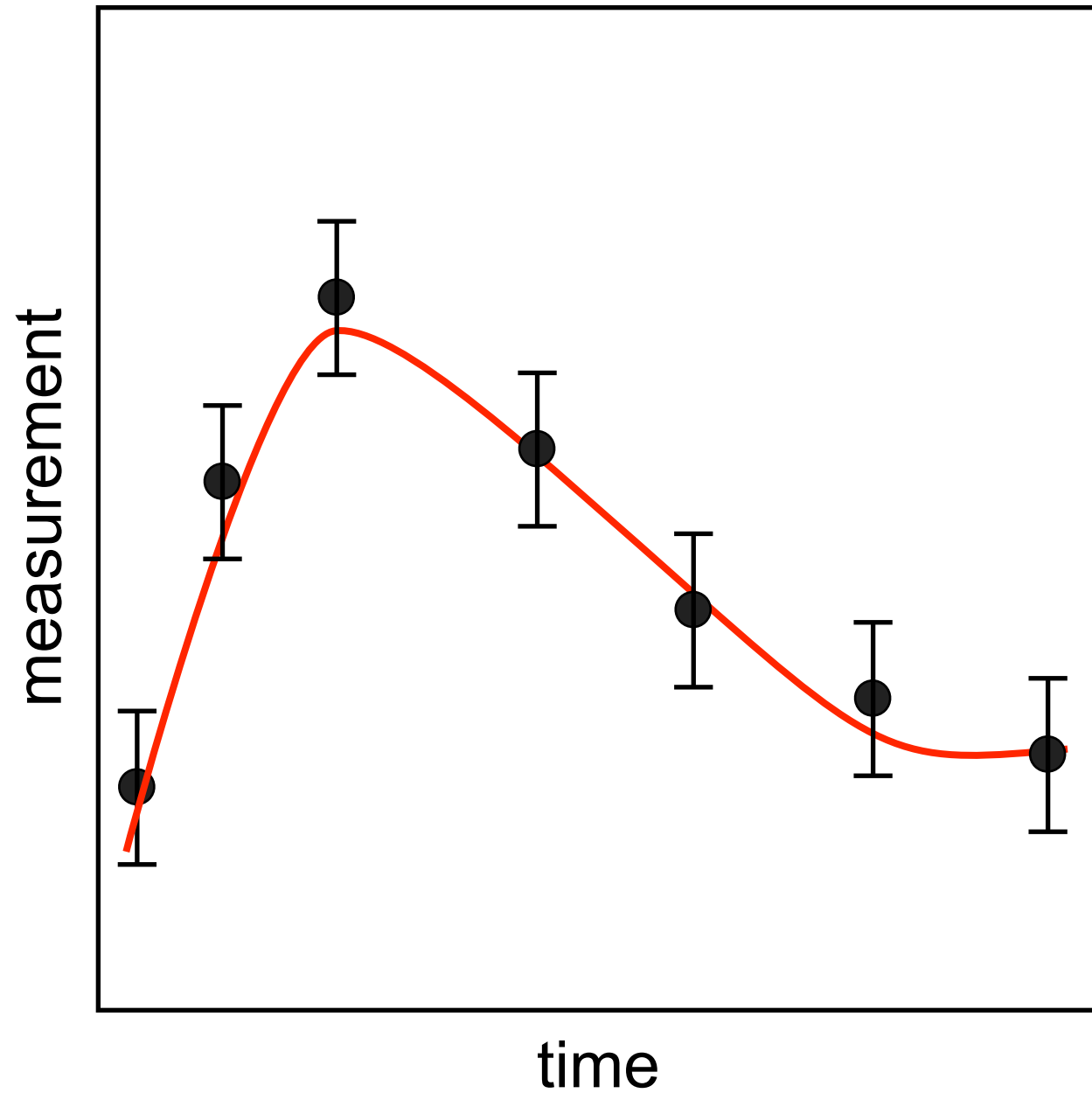


## Objective function landscape

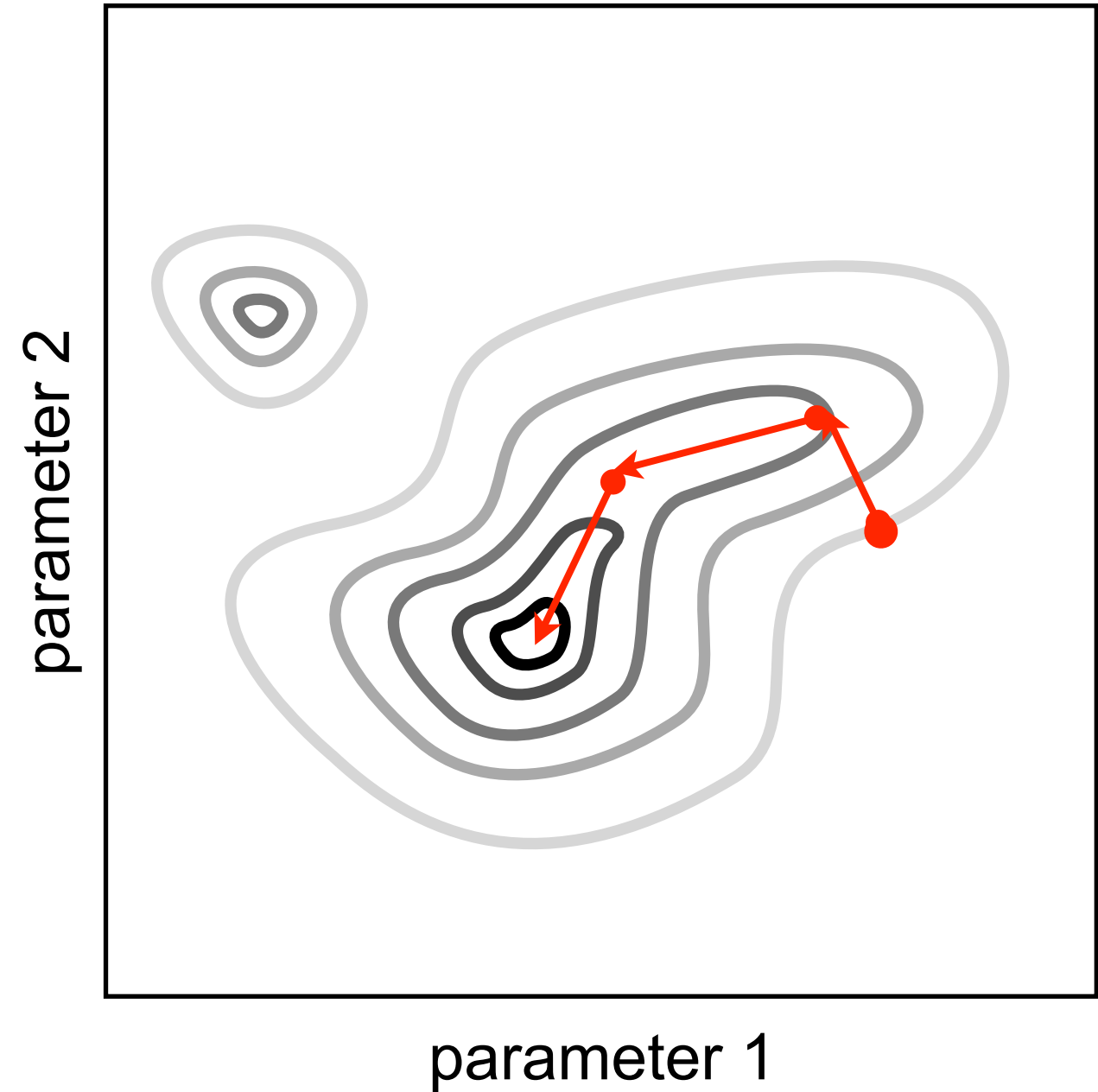


# Parameter Estimation

## Model-data comparison

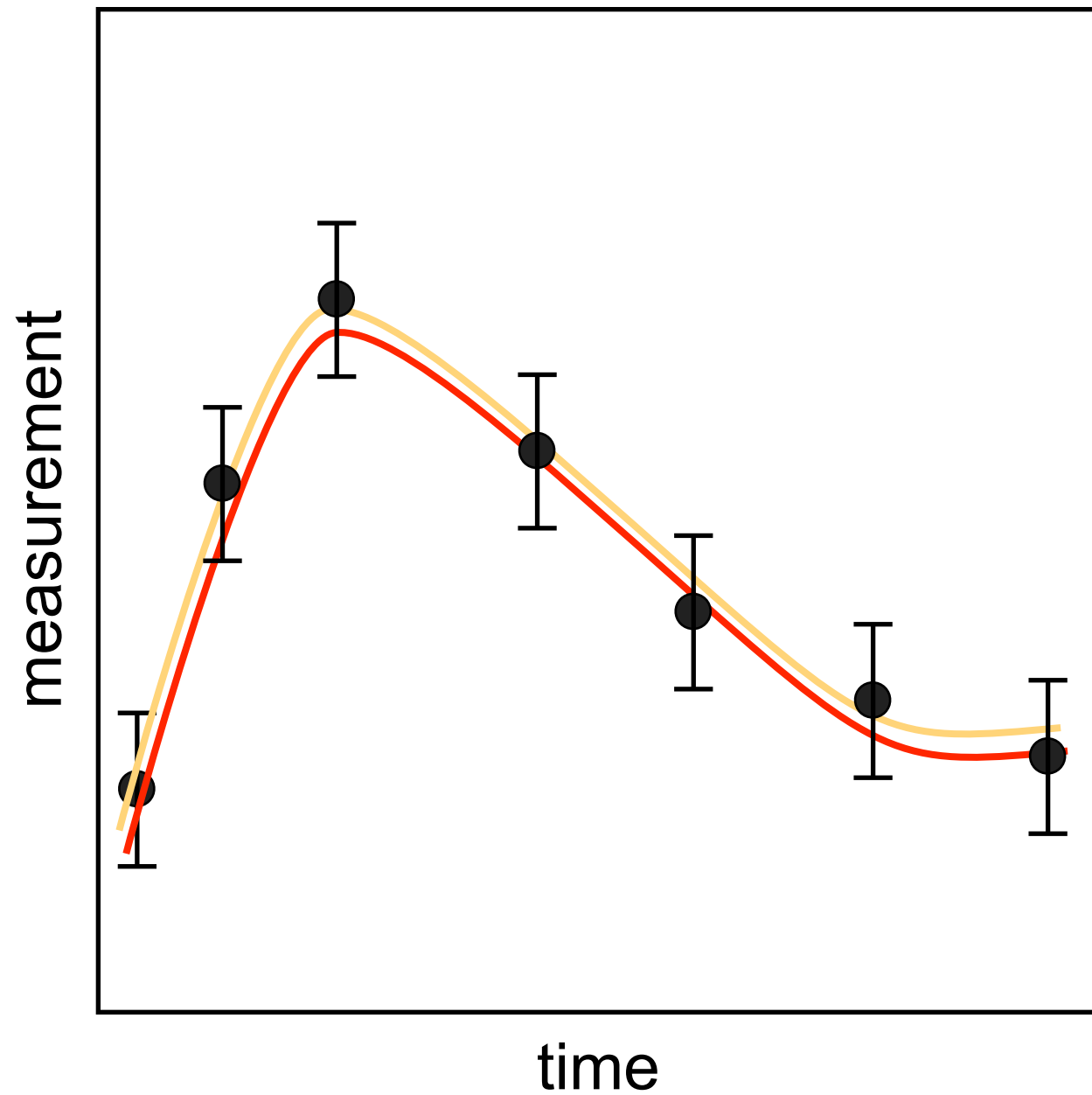


## Objective function landscape

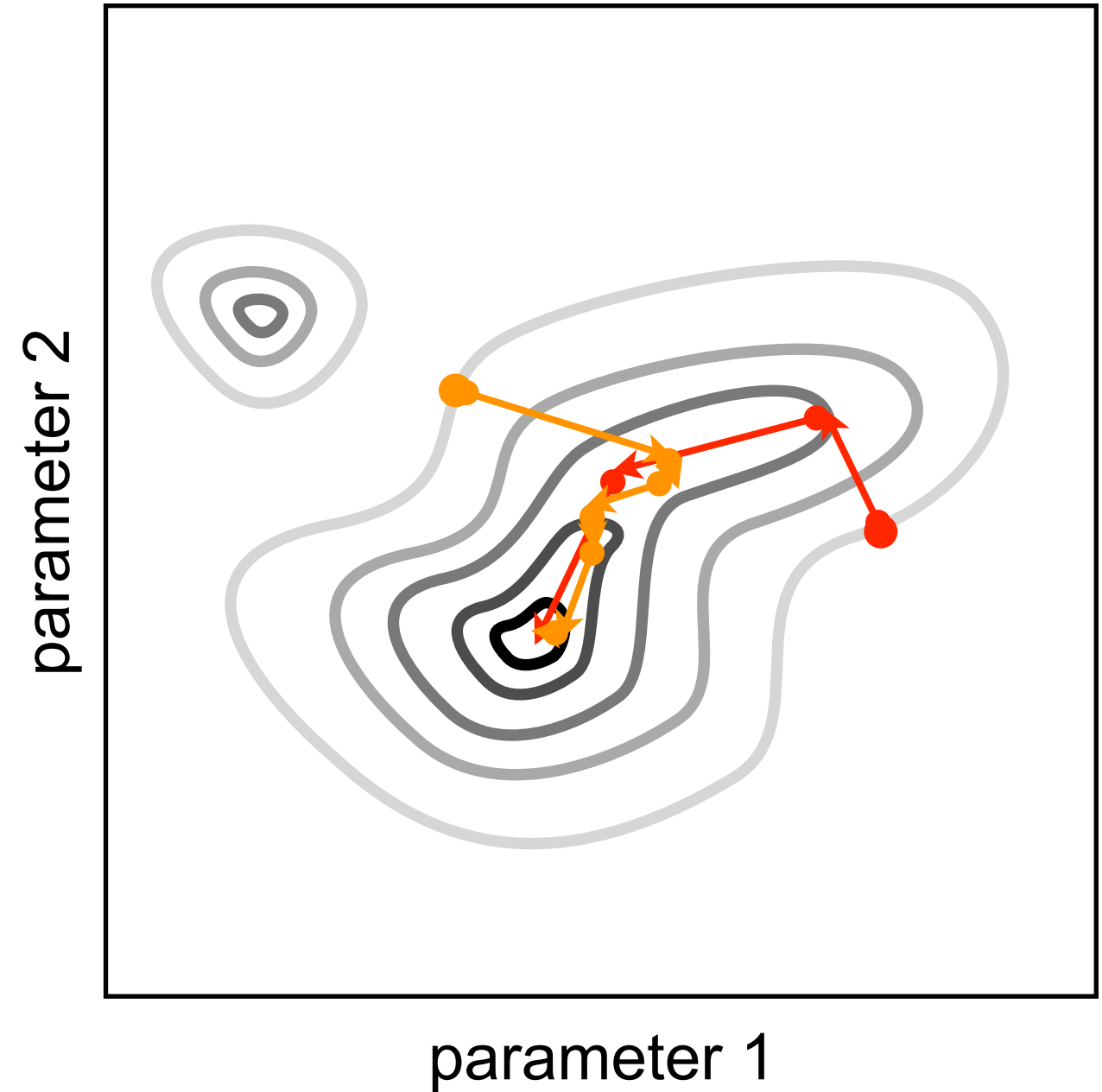


# Parameter Estimation

## Model-data comparison



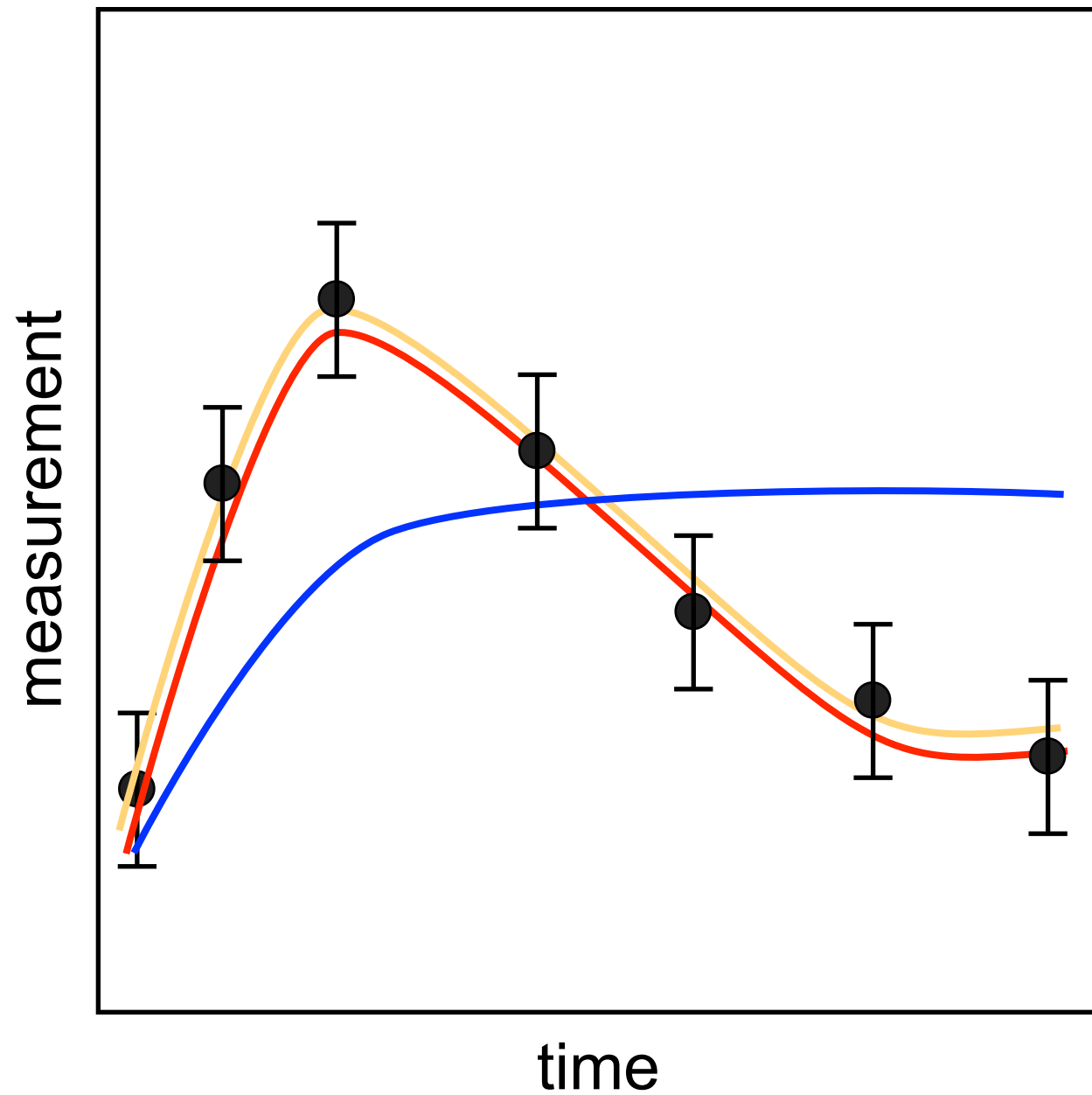
## Objective function landscape



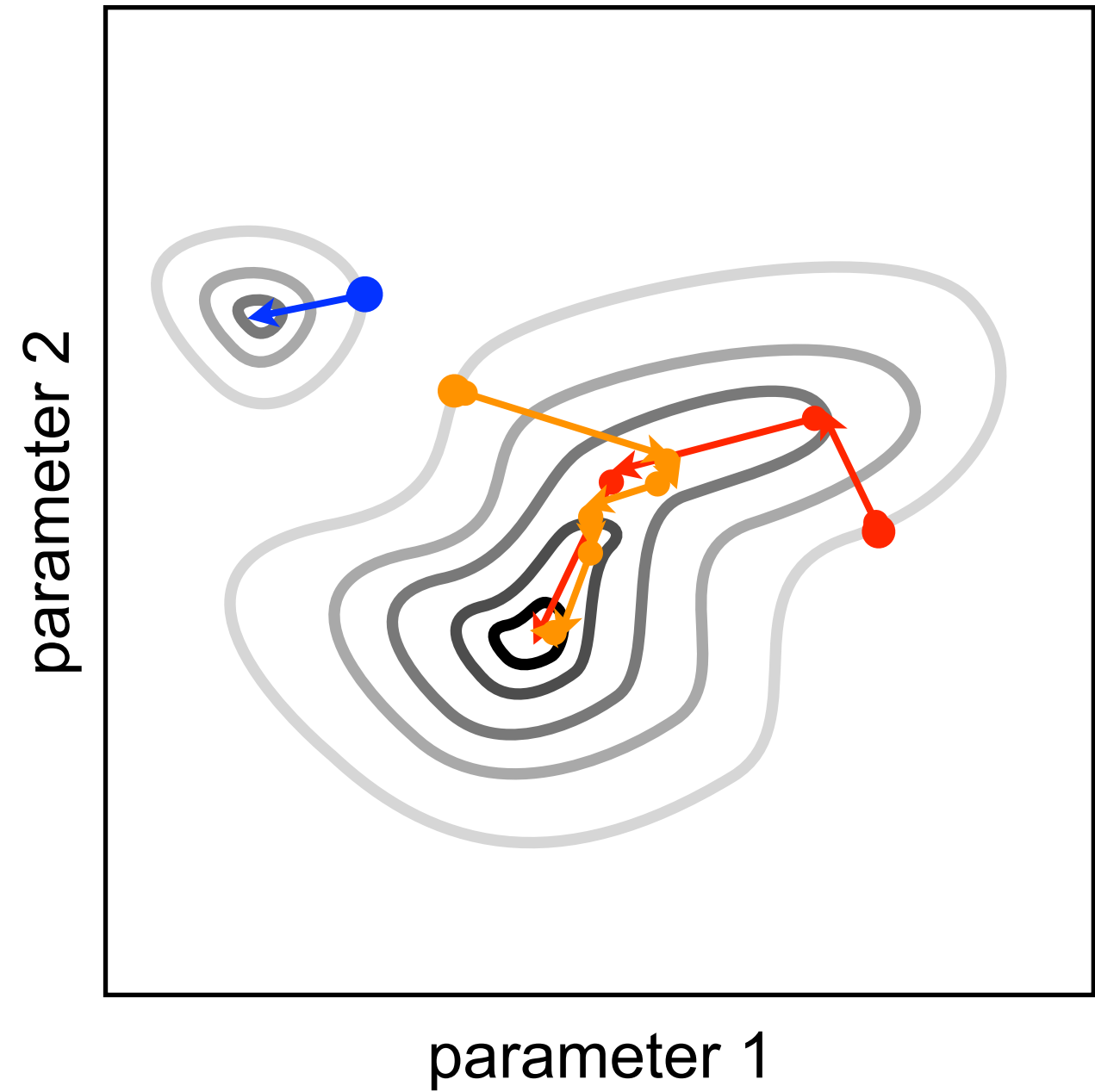


# Parameter Estimation

## Model-data comparison

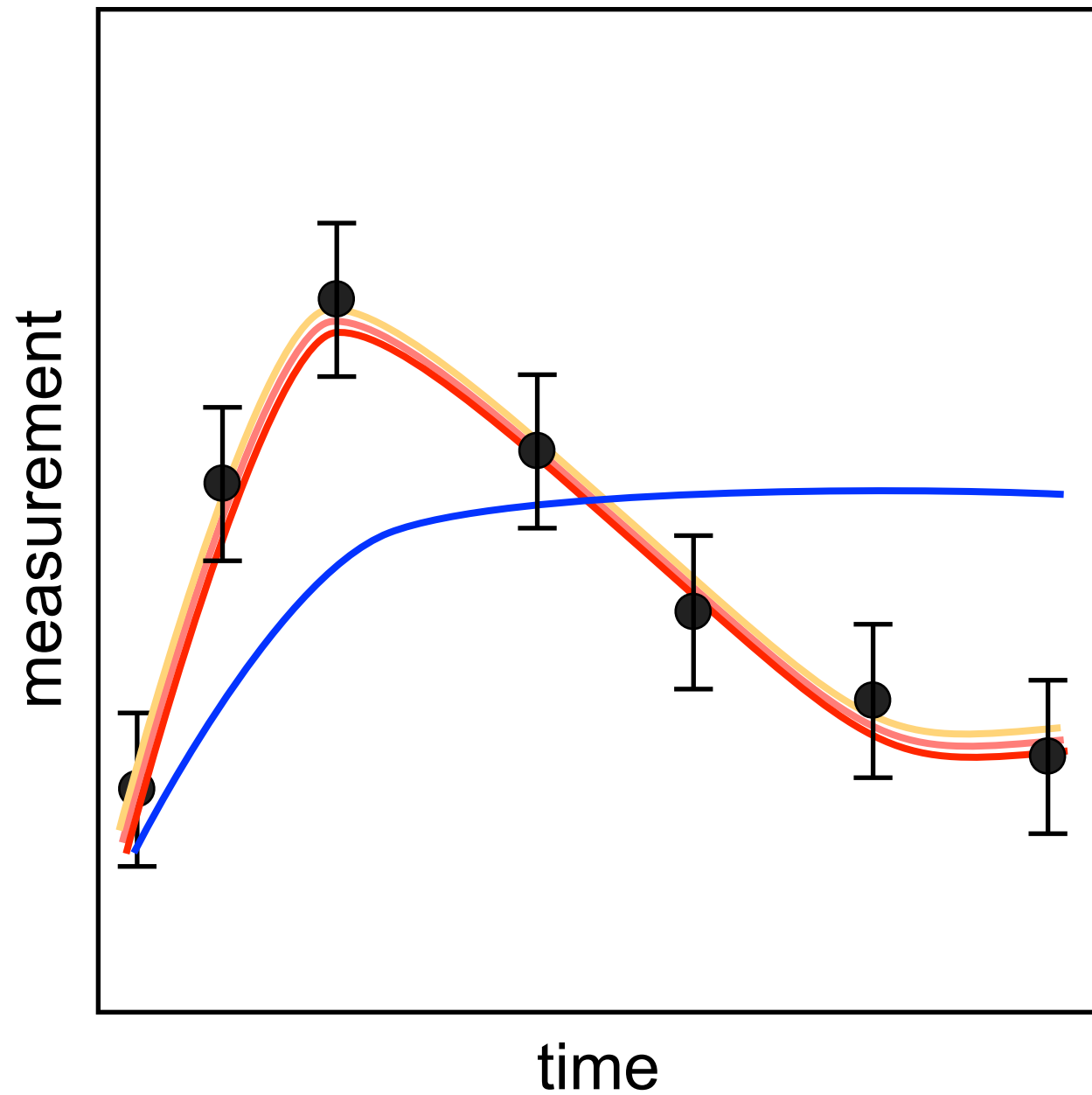


## Objective function landscape

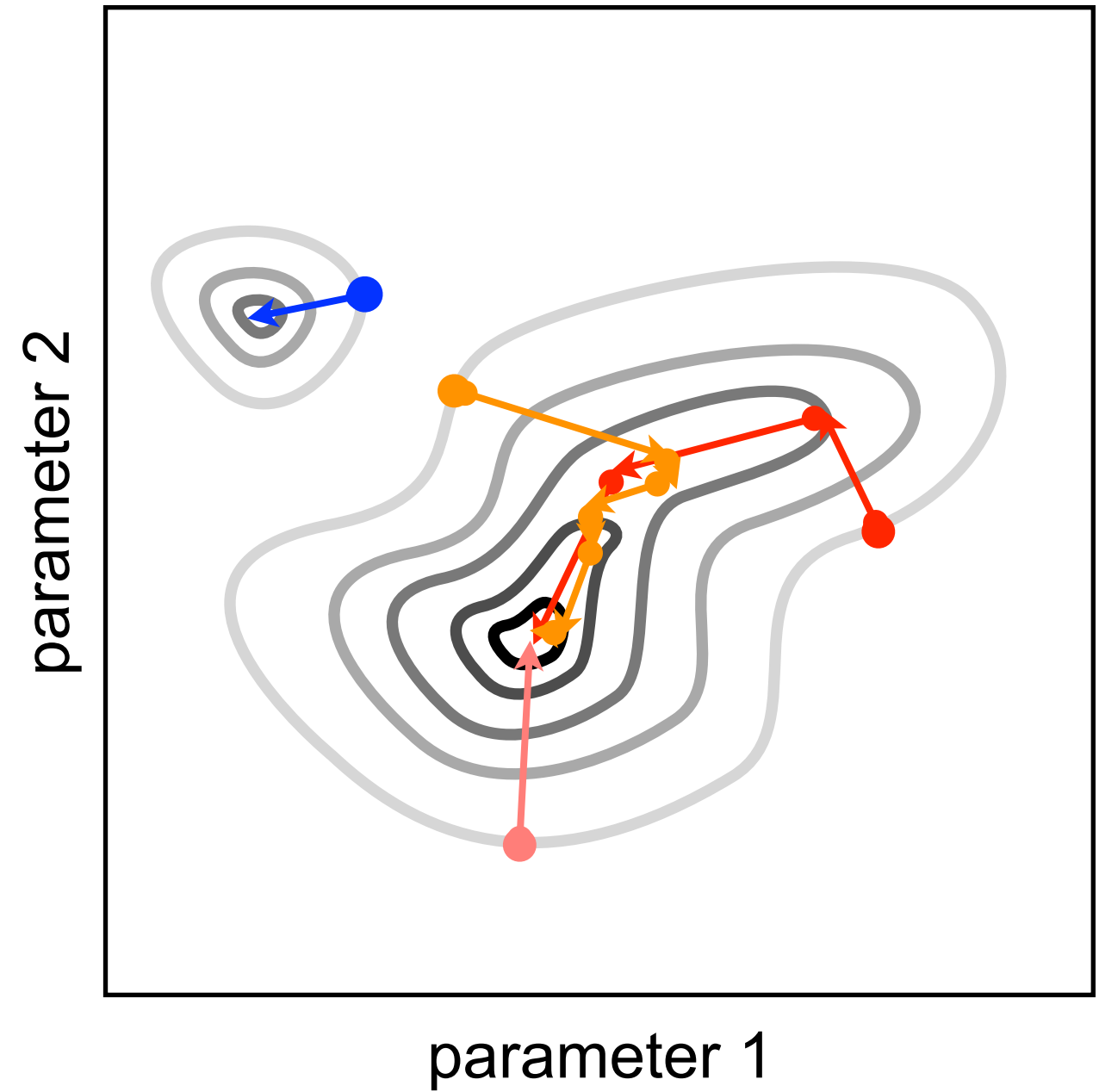


# Parameter Estimation

## Model-data comparison

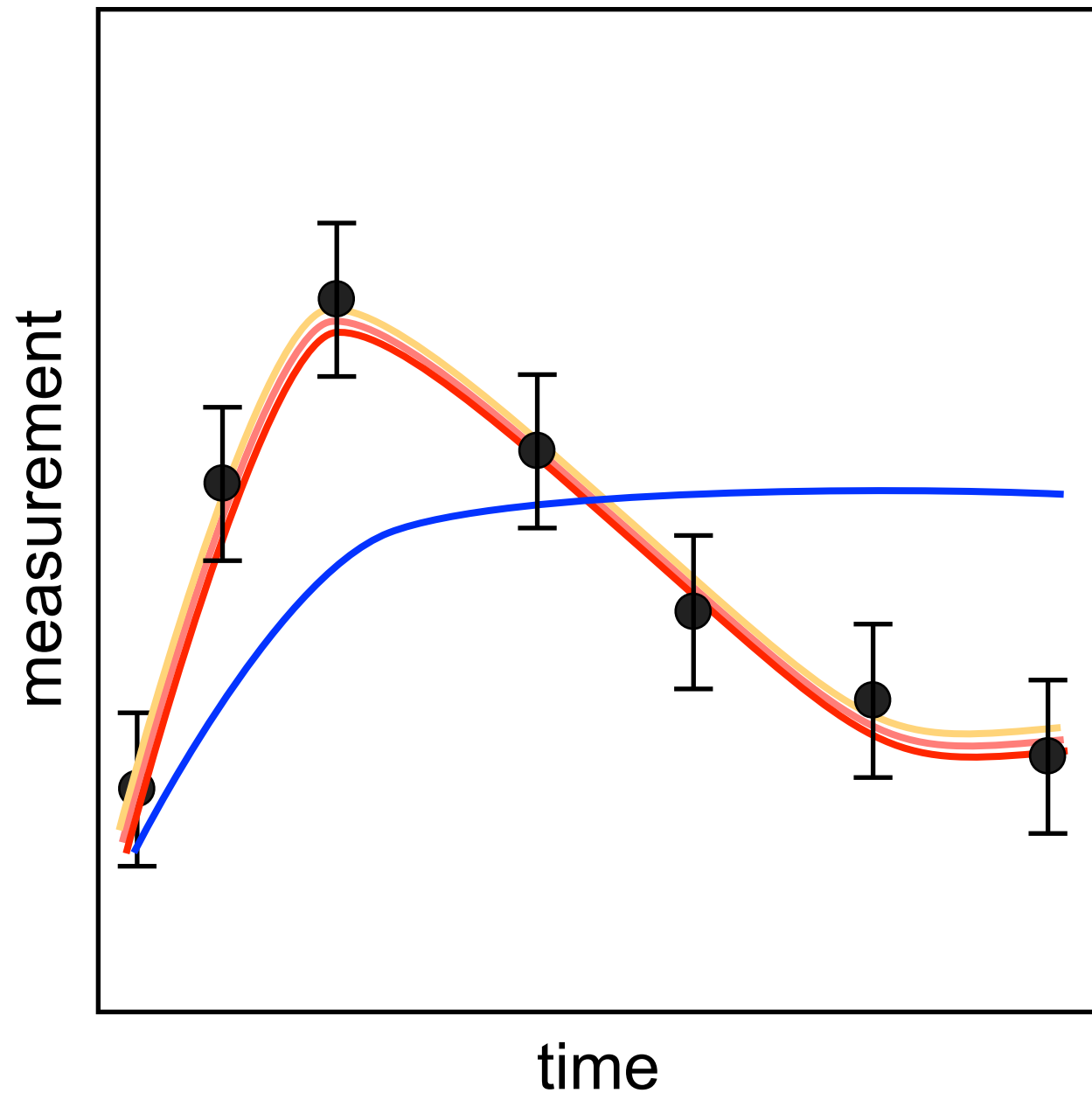


## Objective function landscape

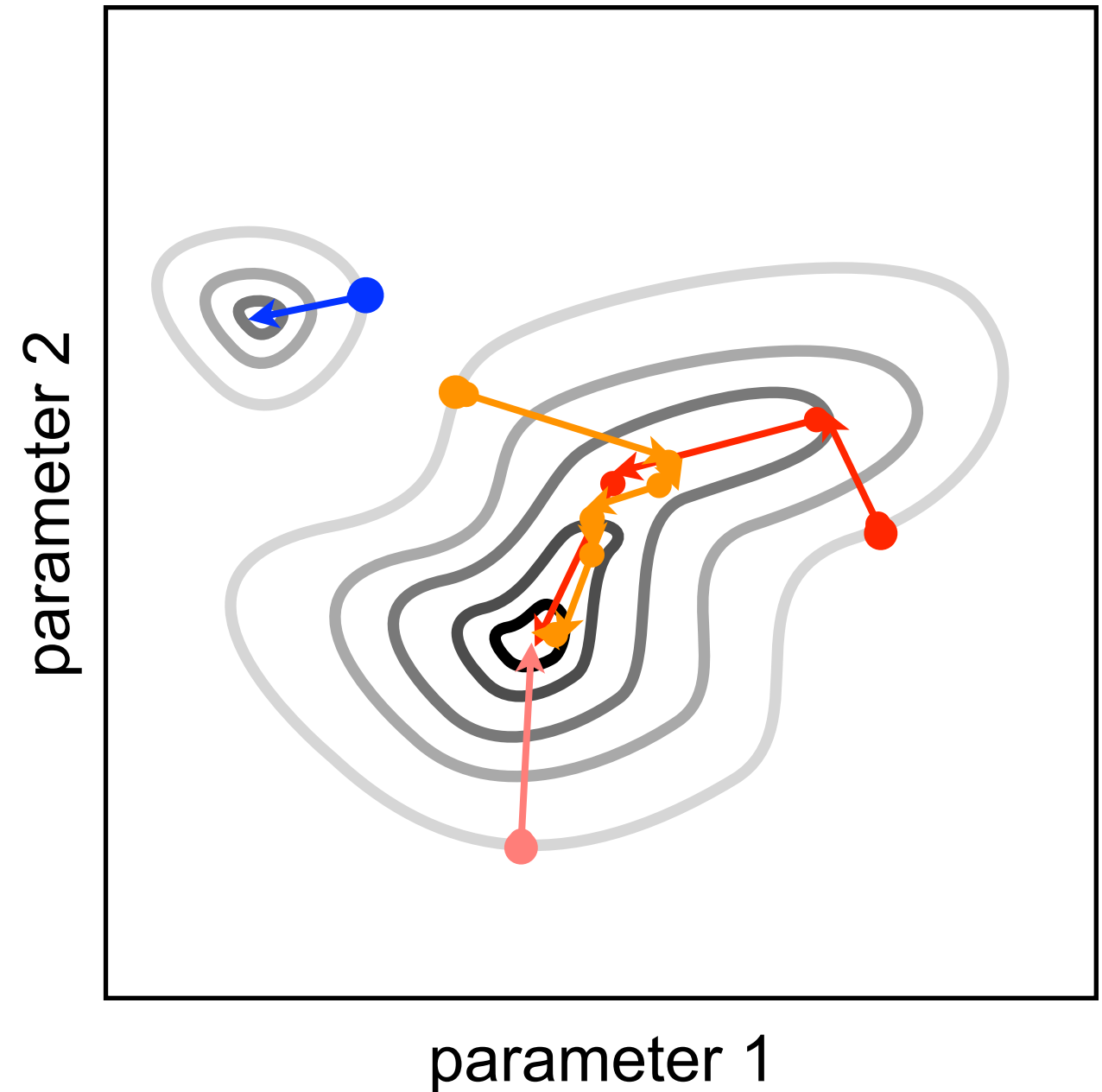


# Parameter Estimation

## Model-data comparison



## Objective function landscape

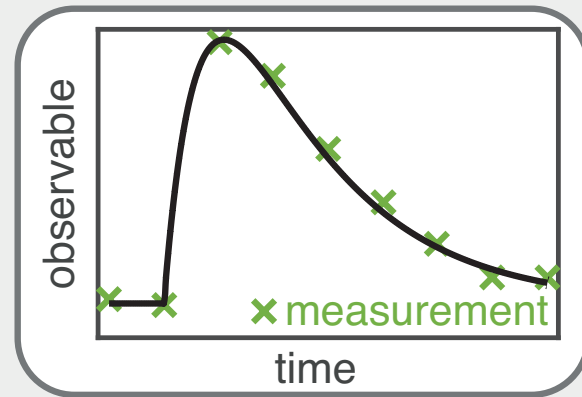


**Bottleneck: gradient computation**

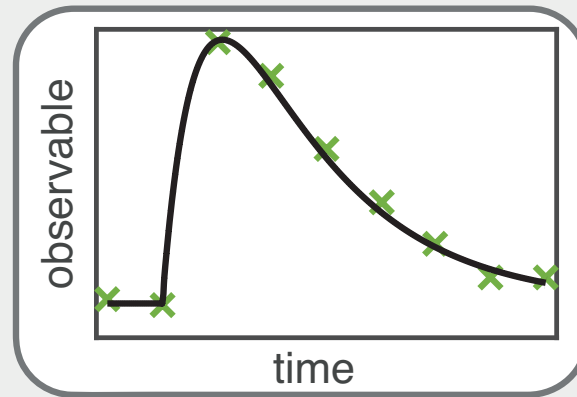
# Gradient Computation Schemes

Step 1

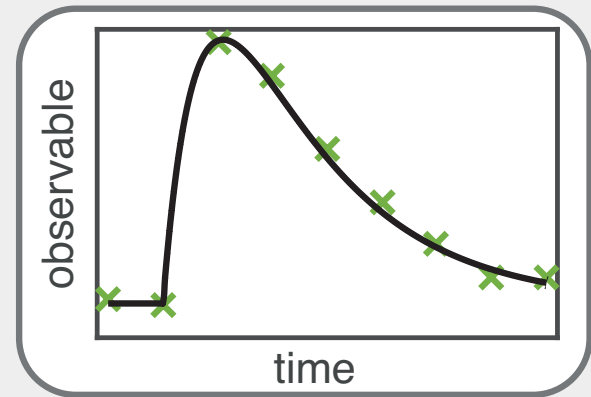
Finite Differences



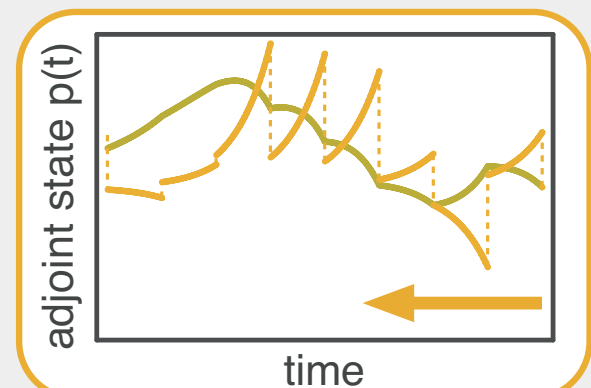
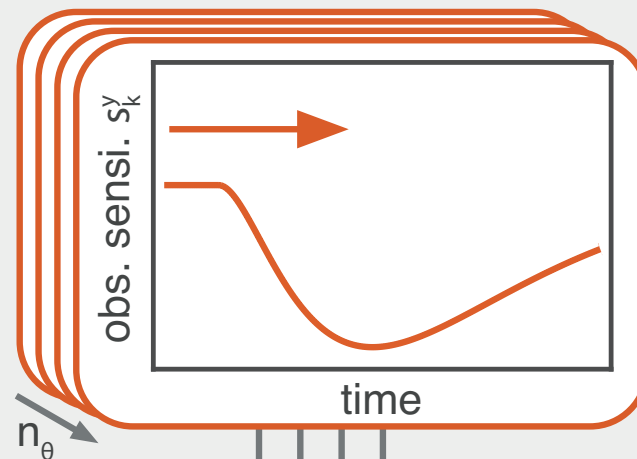
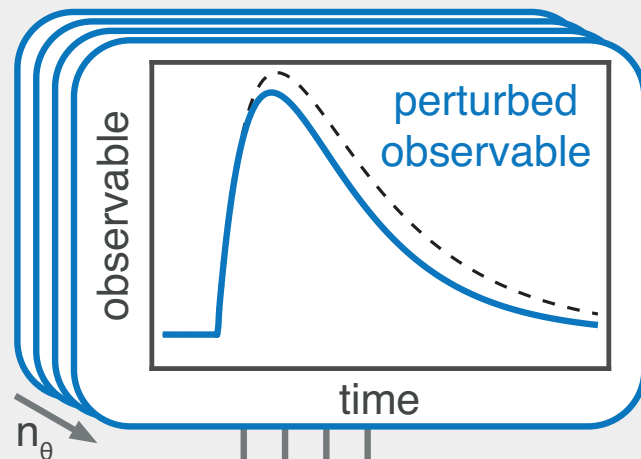
Forward Sensitivities



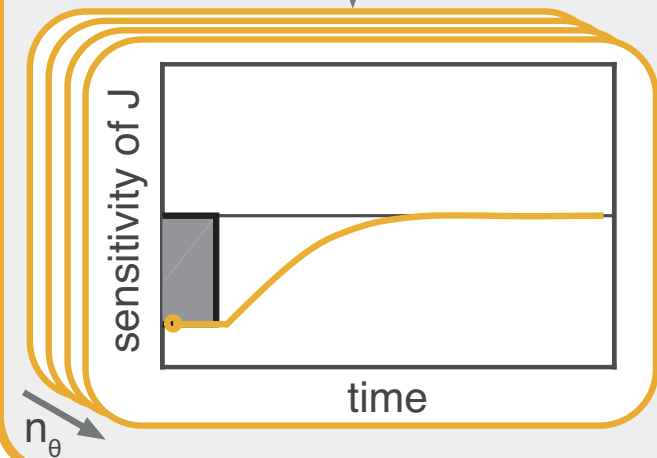
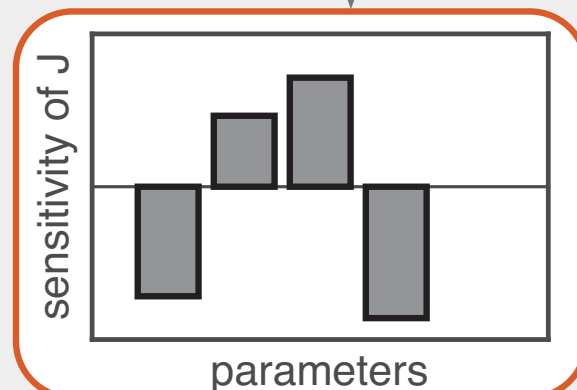
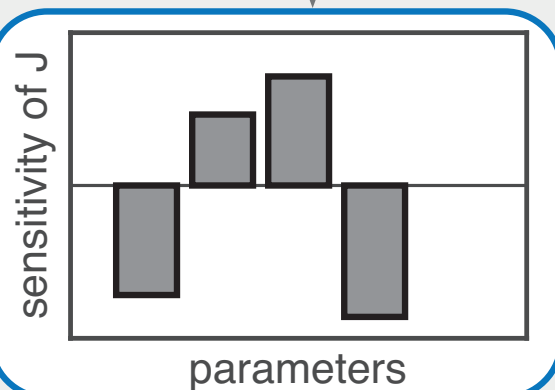
Adjoint sensitivities



Step 2



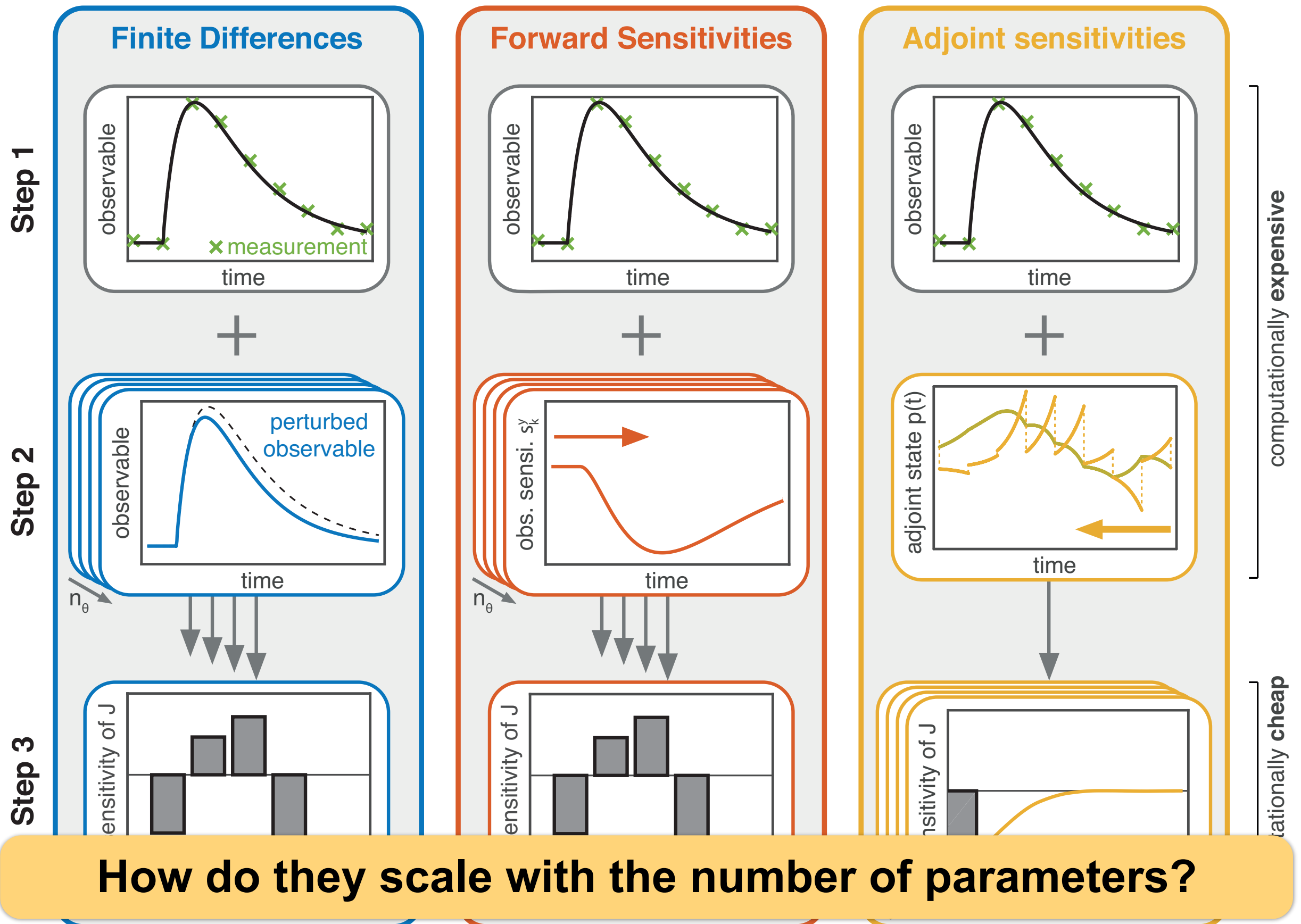
Step 3



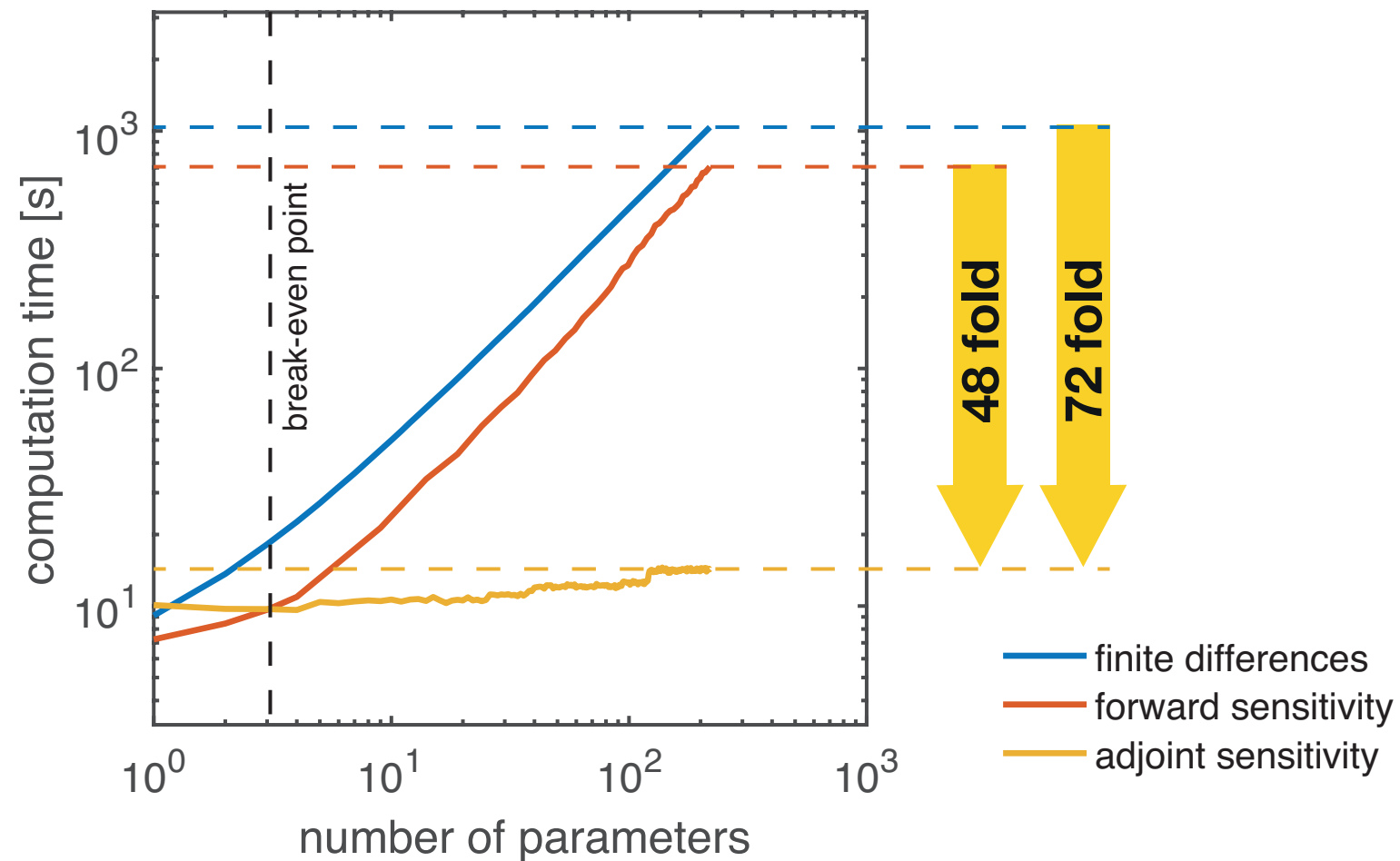
computationally **expensive**

computationally **cheap**

# Gradient Computation Schemes



# Speedup of Gradient Computation



## Model:

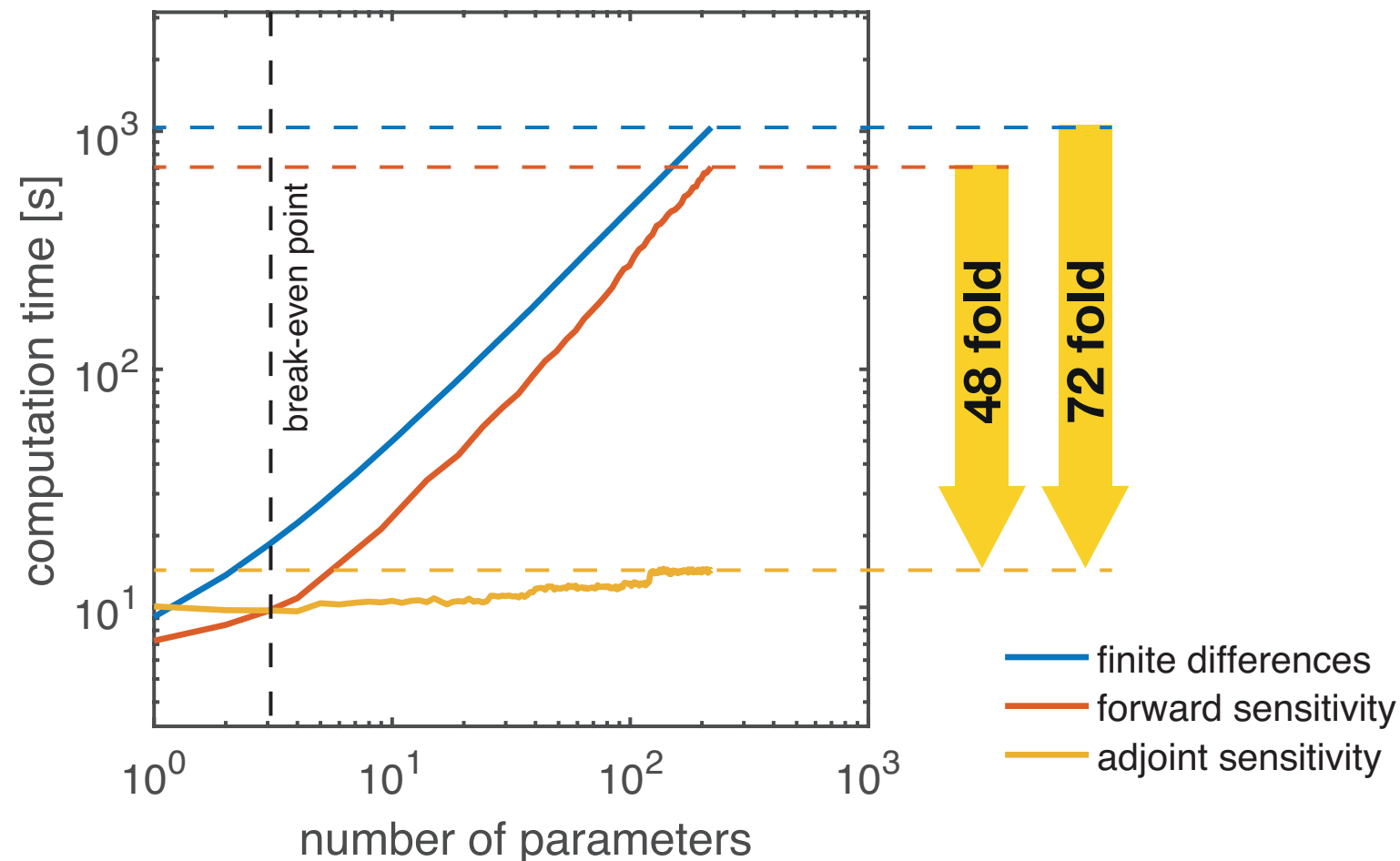
BIOMD0000000255

Chen2009 ErbB Signaling

219 parameters

500 state variables

# Speedup of Gradient Computation

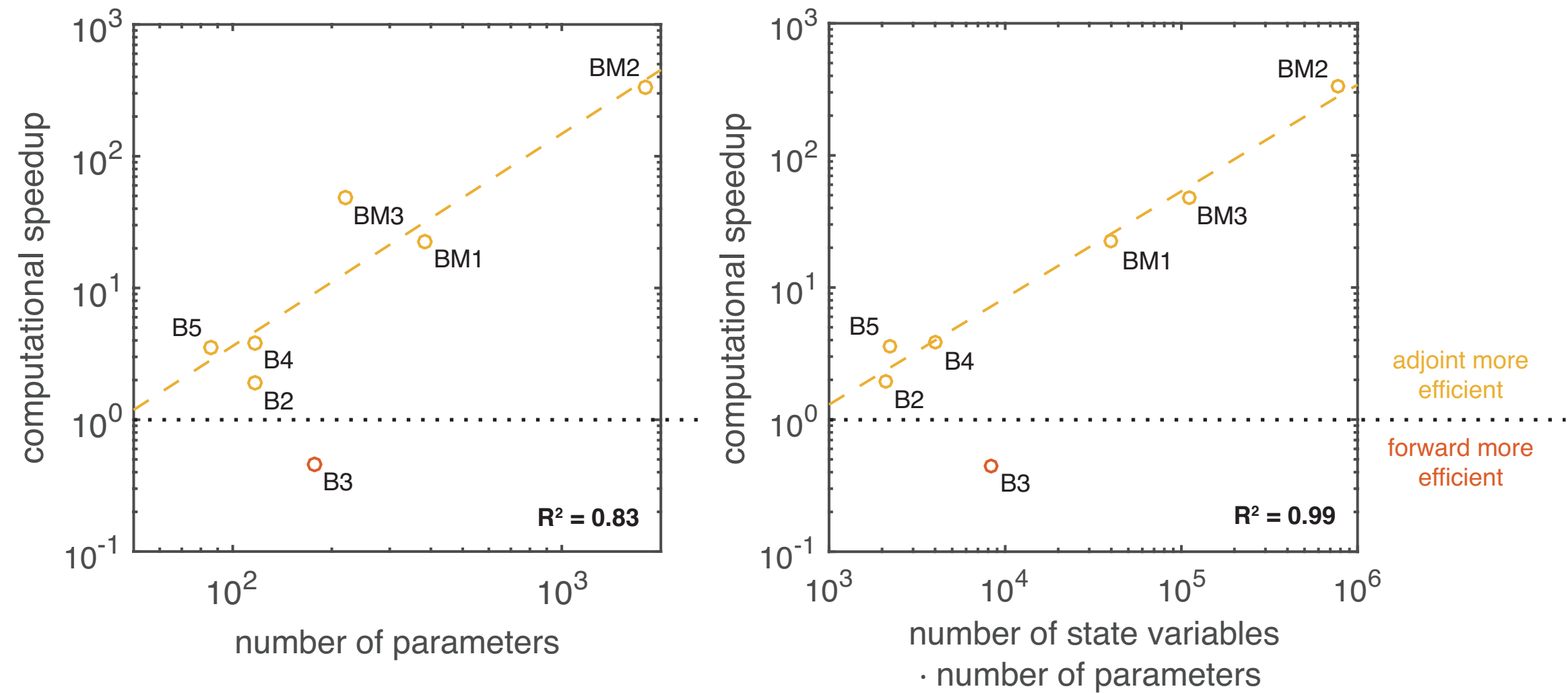


## Model:

BIOMD0000000255  
Chen2009 ErbB Signaling  
219 parameters  
500 state variables

**Does this translate to other models?**

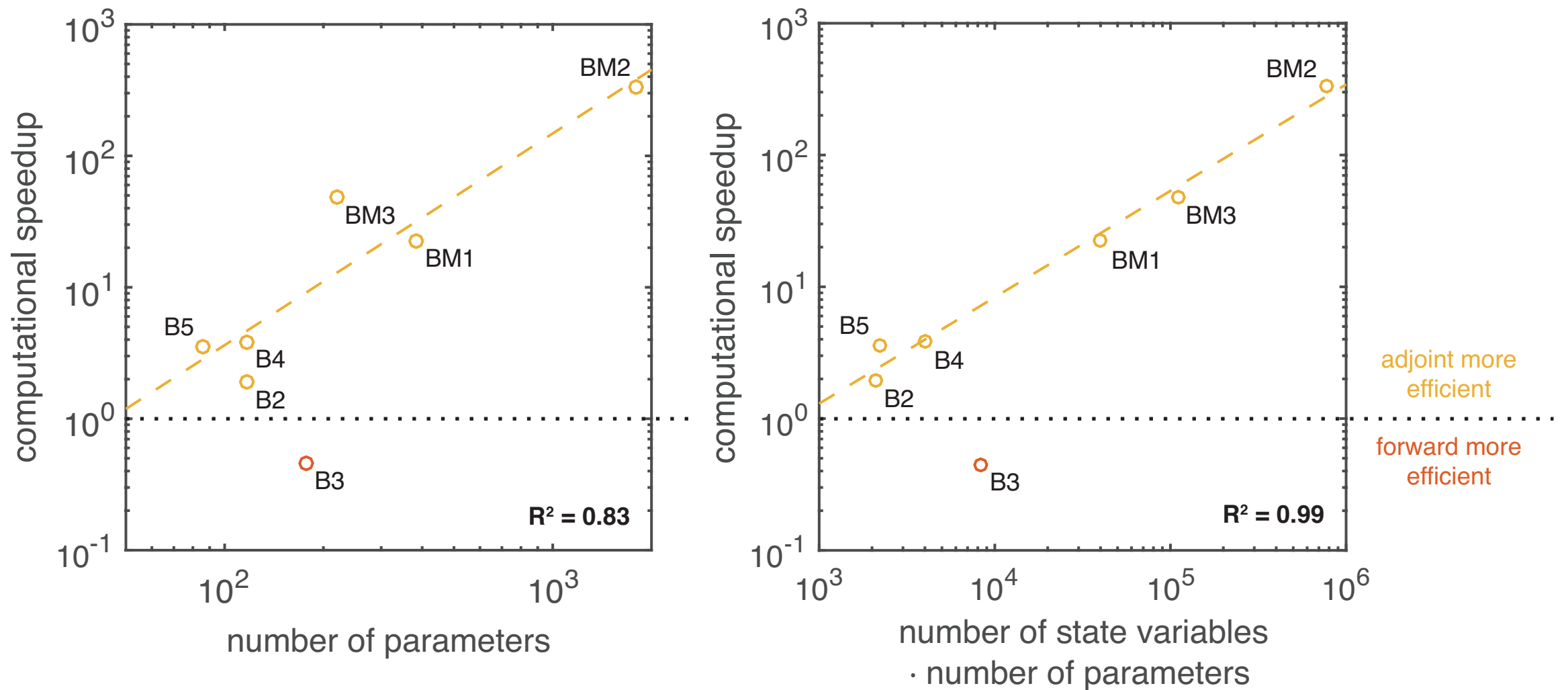
# Speedup of Gradient Computation



	B2	B3	B4	B5	BM1	BM2	BM3
#states	18	47	34	26	104	431	500
#parameters	116	178	117	86	383	1801	219
BIOM ID					474	235	255
	BioPreDynBench, Villaverde et. al 2015						



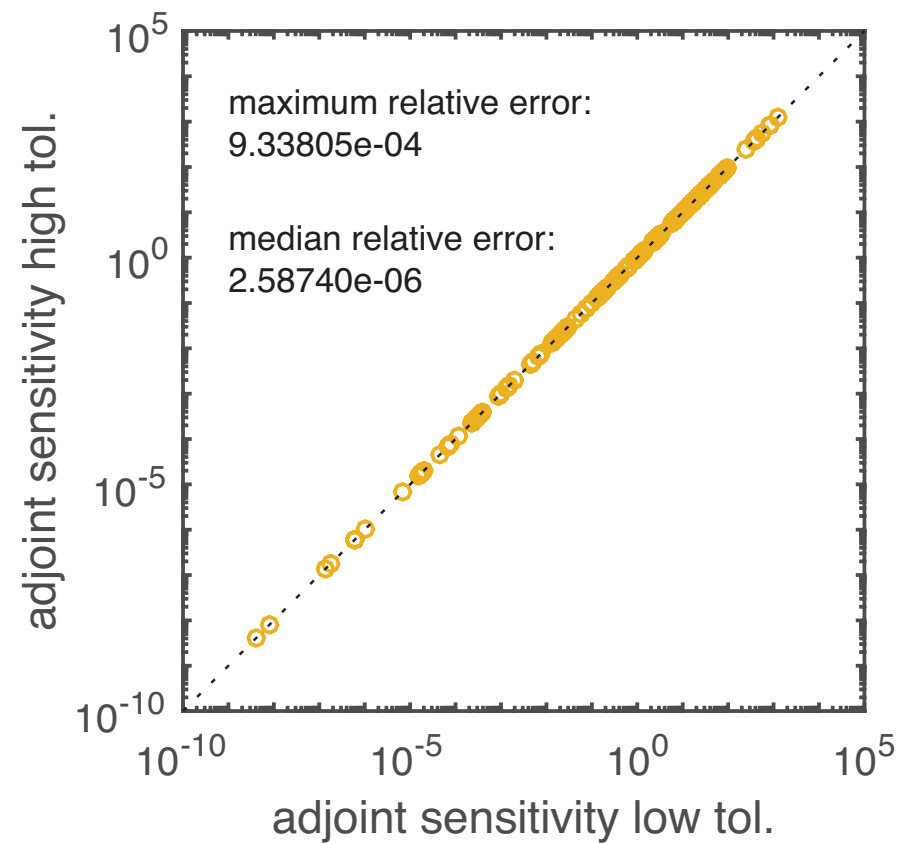
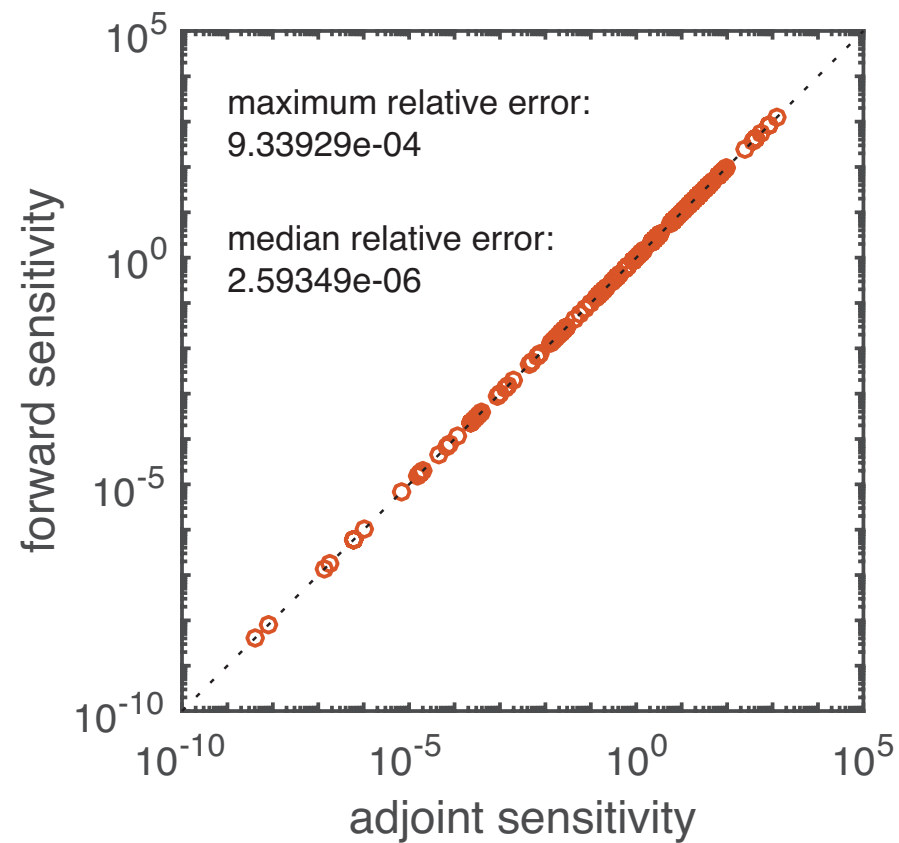
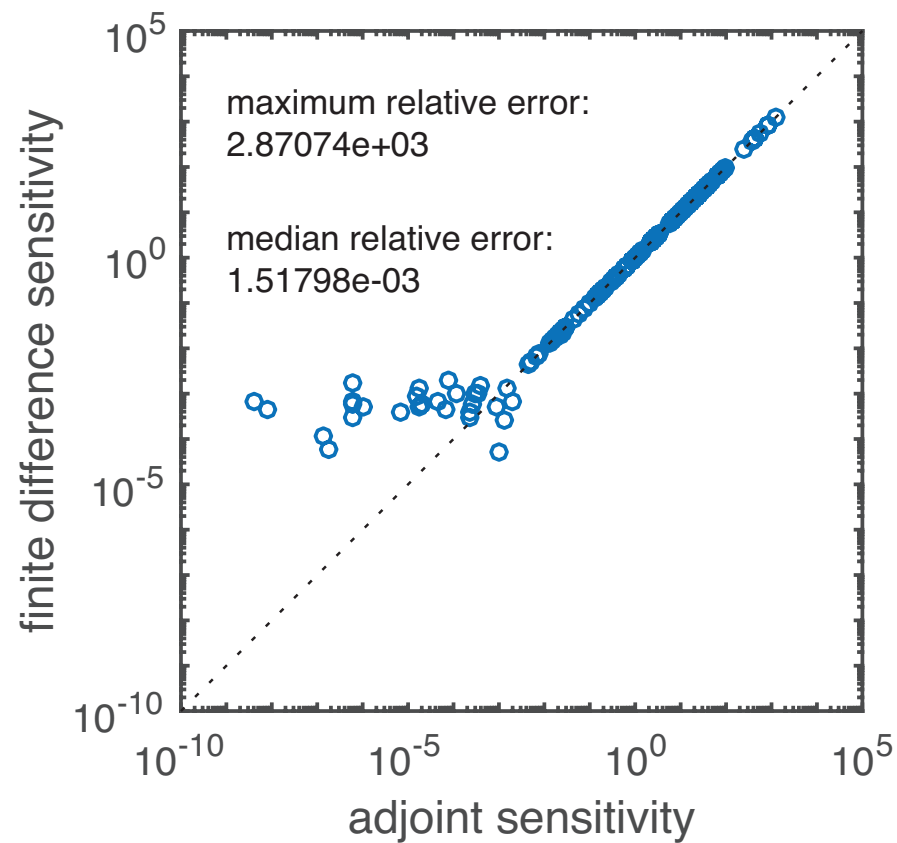
# Speedup of Gradient Computation



	B2	B3	B4	B5	BM1	BM2	BM3
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#parameters	116	178	117	86	383	1801	219
BIOM ID					474	235	255

**Speedup is almost universal.**

# Accuracy of Gradient Computation



absolute tolerance:  $10^{-16}$   
relative tolerance:  $10^{-8}$

## Model:

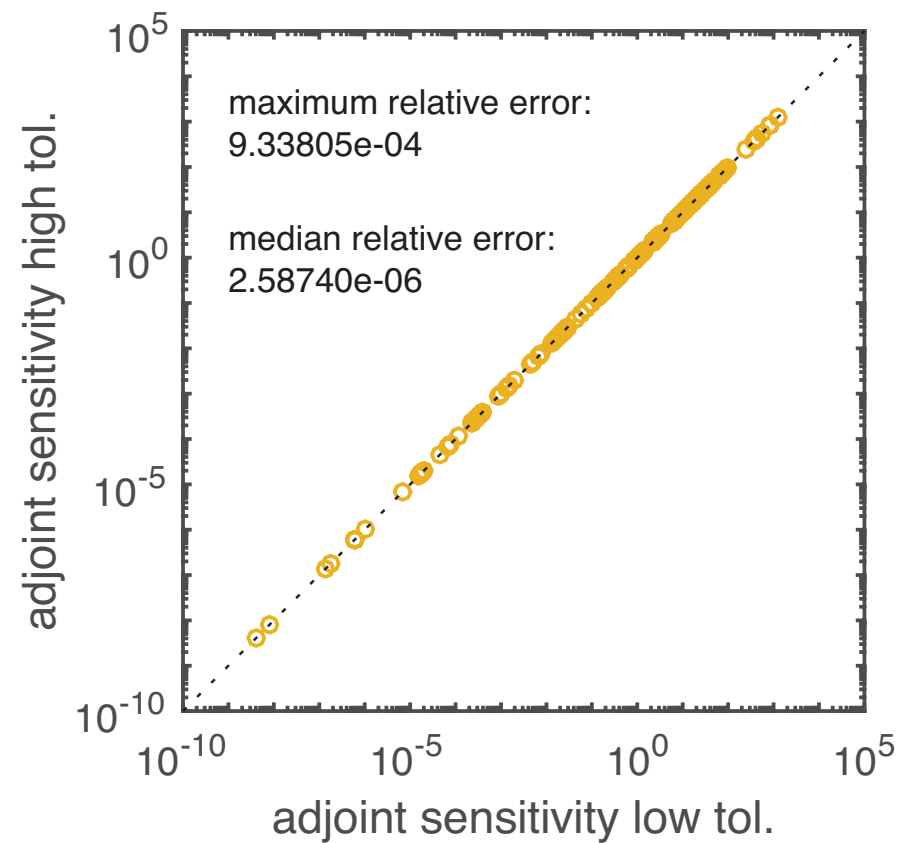
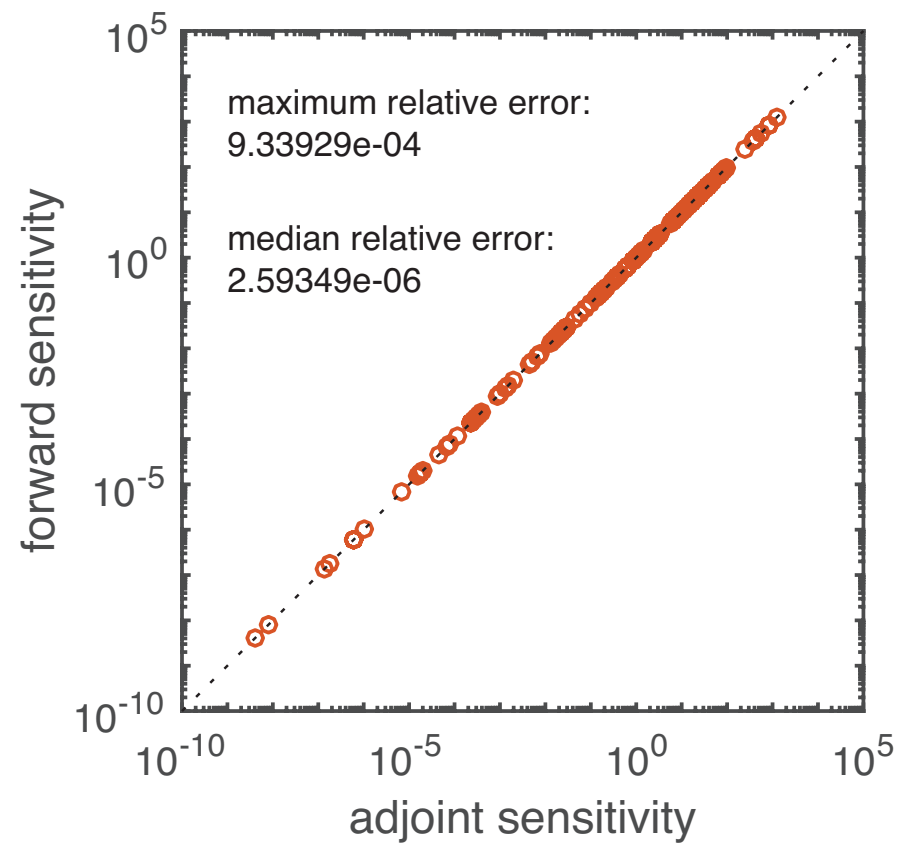
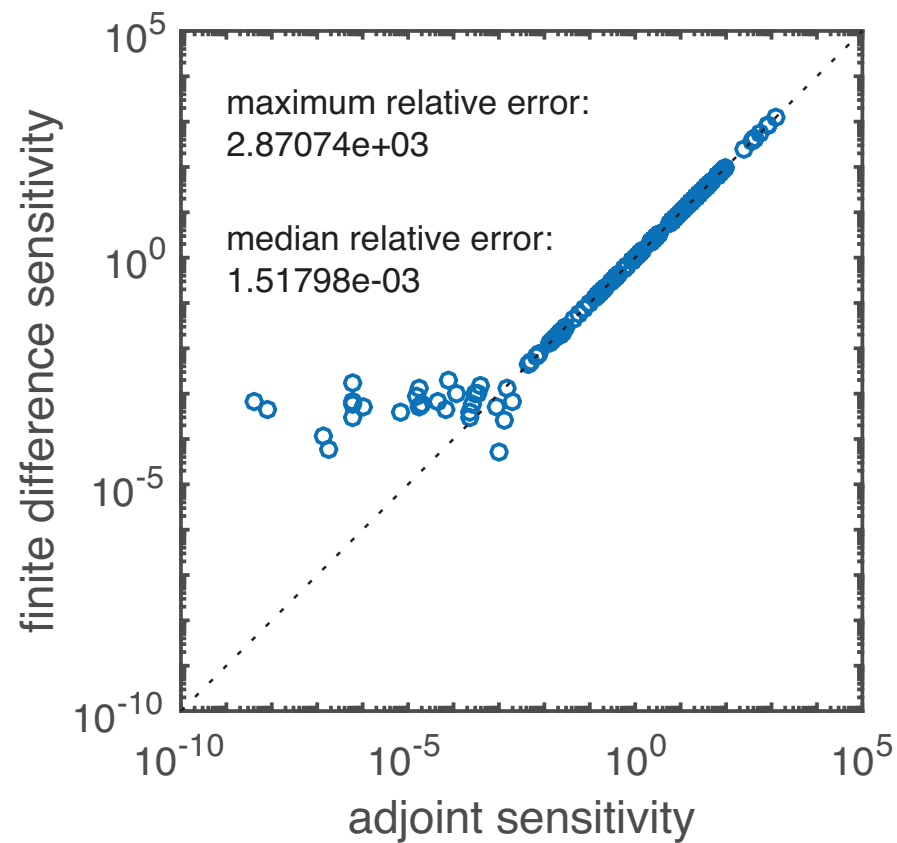
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# Accuracy of Gradient Computation



absolute tolerance:  $10^{-16}$   
relative tolerance:  $10^{-8}$

## Model:

BIOMD0000000255

Chen2009 ErbB Signaling

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**Forward and adjoint sensitivities have similar accuracy.**

# Application: Cancer Signalling

**ALACRIS**  
Theranostics GmbH

## Model properties

State variables: 1230

Parameters: 4256

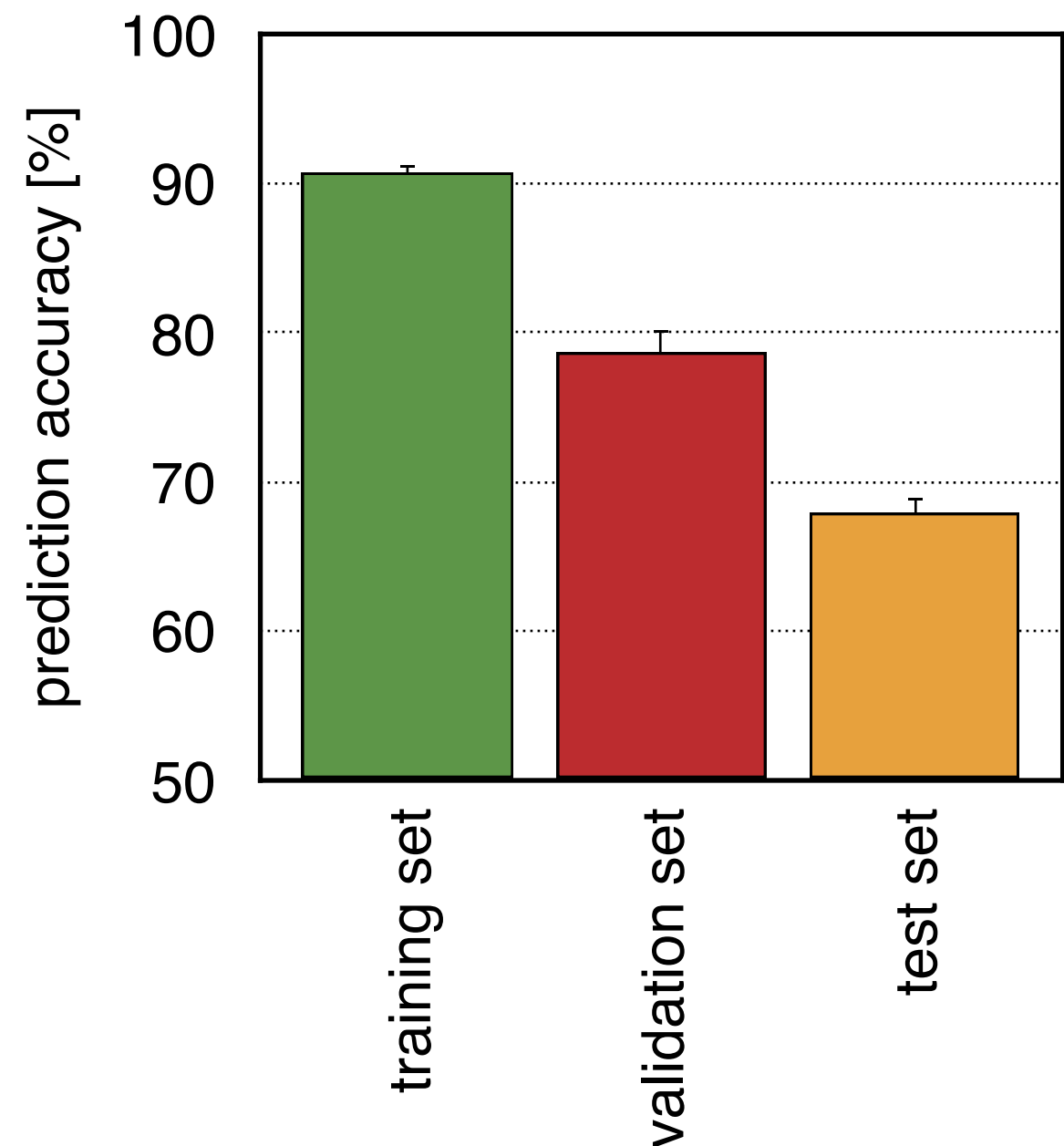
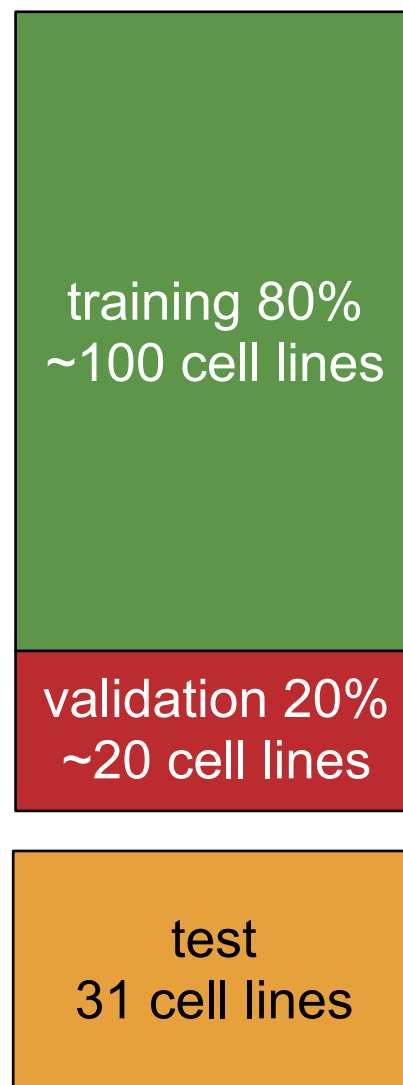
Experimental

Conditions: 5500



## Classification

Responder/Non-Responder  
to 7 different drugs



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## Model properties

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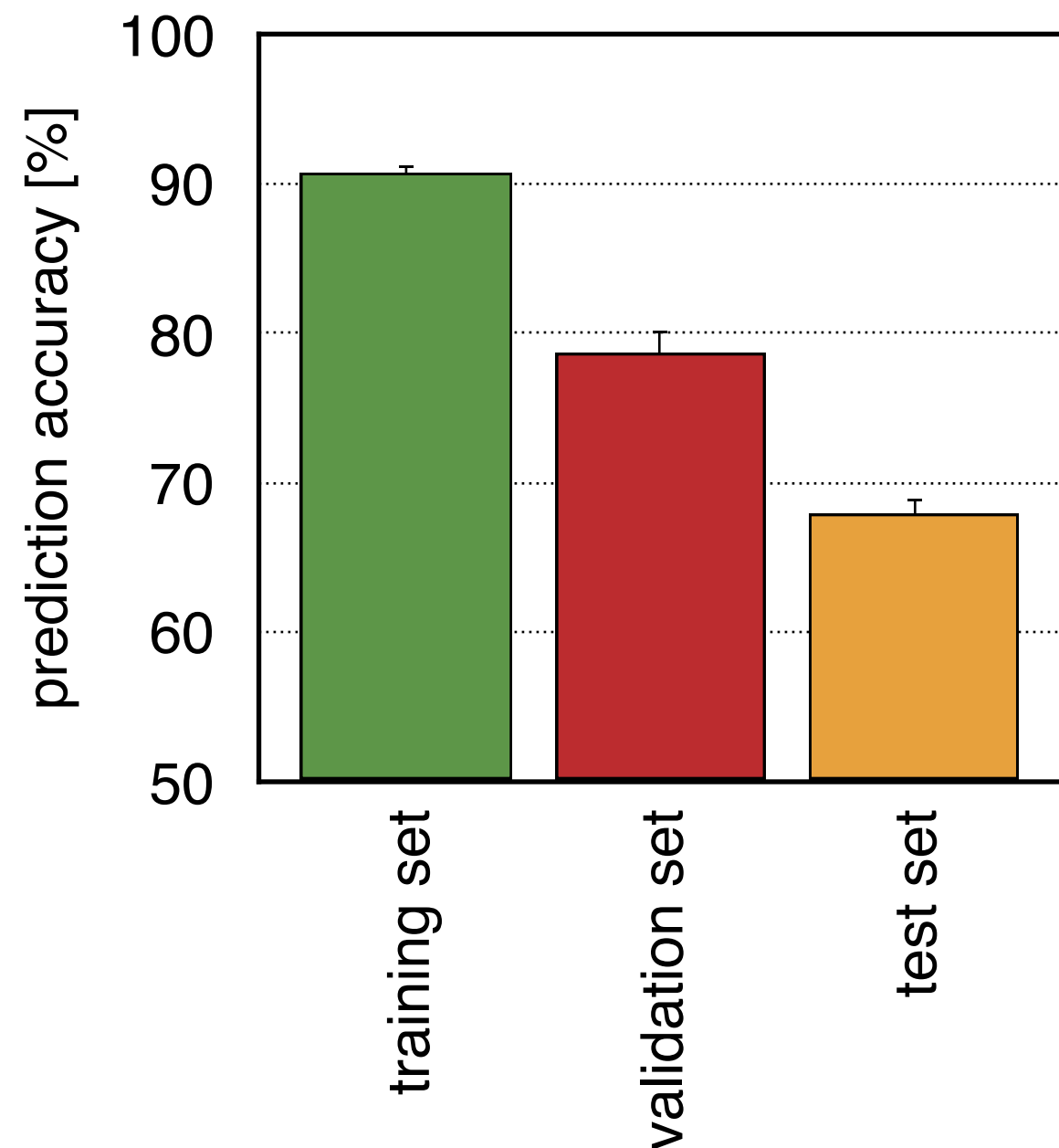
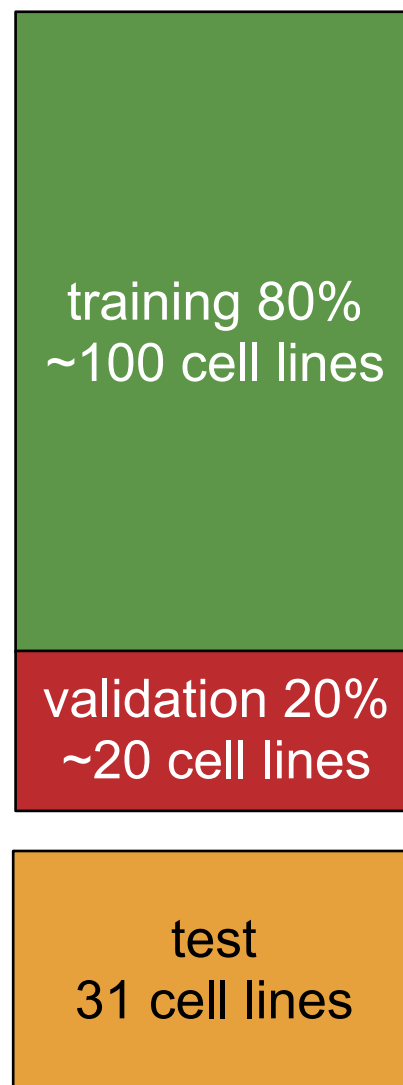
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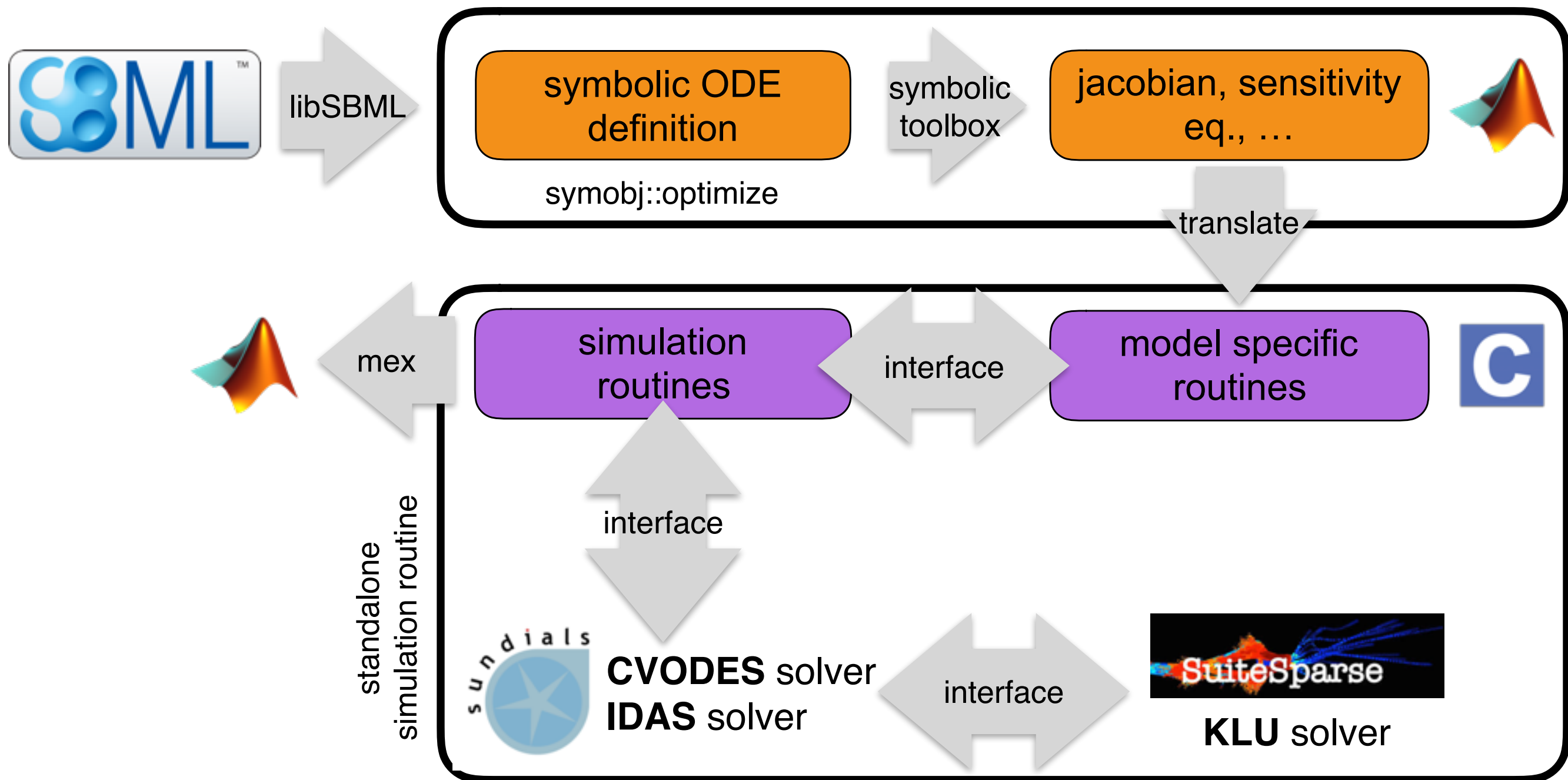
## Classification

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**Adjoint approach makes large-scale parametrization possible**

# Implementation: AMICI



# AMICI: Key Features

- Forward and Adjoint sensitivity analysis
- Second order sensitivities analysis
- Directional second order sensitivities
- Stead state sensitivities
- Support for models with events
- Supports models with thousands of states and parameters

**AMICI**



[https://github.com/  
ICB-DCM/AMICI](https://github.com/ICB-DCM/AMICI)

# Summary

- Adjoint sensitivity analysis is more efficient and as accurate as forward sensitivity analysis for large scale models
- Modular Implementation in AMICI which generates simulation routines as native C code

# Outlook

- HPC support
- Support for SED-ML
- Support for different ODE solvers?
- Rewrite of symbolic processing in C/python?
- Standard for model-data metric?



# Acknowledgment

*ICB, Helmholtz Zentrum München*

**Jan Hasenauer**

Paul Stapor

Daniel Weindl

Fabian Theis

*University of Klagenfurt*

**Barbara Kaltenbacher**



**Thank you for your attention!**