

Modelling and Simulating Complex Systems

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legato-team.eu

Speak up

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< Best Recent EN

DataAndComputa



Number: 73334
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14:19

2 messages,
0 comments,
2 votes

23/01/2019 14:21 by me [comment](#) ▾

This room will remain open during the whole course and allows you to ask questions without being shy, or interrupting the flow. You can also vote for a question if you find it relevant. All this can be done online. Enjoy! Stéphane BORDAS



+1
1 vote



23/01/2019 14:20 by me [comment](#) ▾

Dear course participants, welcome! I am opening this room so that you can start asking questions you're interested in hearing about tomorrow for the "upskilling" course you will be participating in.



+1
1 vote



Post a message



Speak up

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SpeakUp

EN

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Joined rooms

No rooms joined yet

My rooms

DataAndComputationalScienceUpskilling



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Acknowledgements

The University of Luxembourg

Hussein Rappel, PhD

Post-doctoral fellow, university of Luxembourg

Christian Vincenot, PhD

Biosphere Informatics Laboratory
Department of Social Informatics

Susanne Claus, PhD

University of Copenhagen, Denmark
Department of Computer Science

Pierre Kerfriden, PhD

Mines Paris-Tech, Paris, France

The legato-team
legato-team.eu



COMPLEXITY
EXPLORER
SANTA FE INSTITUTE

What is Truth?

- Sciences are plural, they have validity domains, hence truth has multiple faces;
- Knowledge is never complete and Research never finished;
- Knowledge itself is less important than the method used to acquire it;
- What we discover through research is the BEST we can discover at a given instant in time;
- Investing in Research indicates trust in the scientific method.

Data

- 90% of the data available today was generated in the last two years;
- Developing new scientific methodologies to optimise the acquisition of new understanding from data;
- Develop methodological core based on the language of mathematics and common to various application areas;
- These new methodologies will fuel science and technology by creating multi-disciplinary interactions in two fields: personalised medicine and advanced materials.

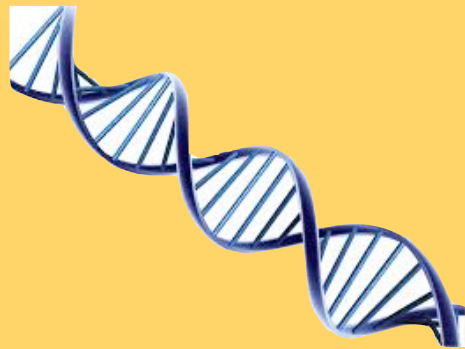
Access to TRUTH(S)

**Access to
Reality(ies)**

**Trust in
Institutions,
Methods,
Human
discourse**

$$\frac{\partial V}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} - rV = 0$$

MATHEMATICS



$$\hat{\mathcal{H}}\Psi = E\Psi$$

$$\psi(x) = \sum_{I=1}^m N_I(x) u_I$$

Outline

- Scientific method, experiments, analytical methods and models
- What is a complex system?
- What models are available to understand and predict?
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The scientific method and the 3 pillars of science

(or should we say 4?)

How do we think
about the world?



Aristotle

384-322 BC

Geocentrism

Copernicus

1473-1543

Heliocentrism

Galileo Galilei

1564-1642

innovative combination of experiments and mathematics

Isaac Newton

1643-1727

Universality of gravitation...

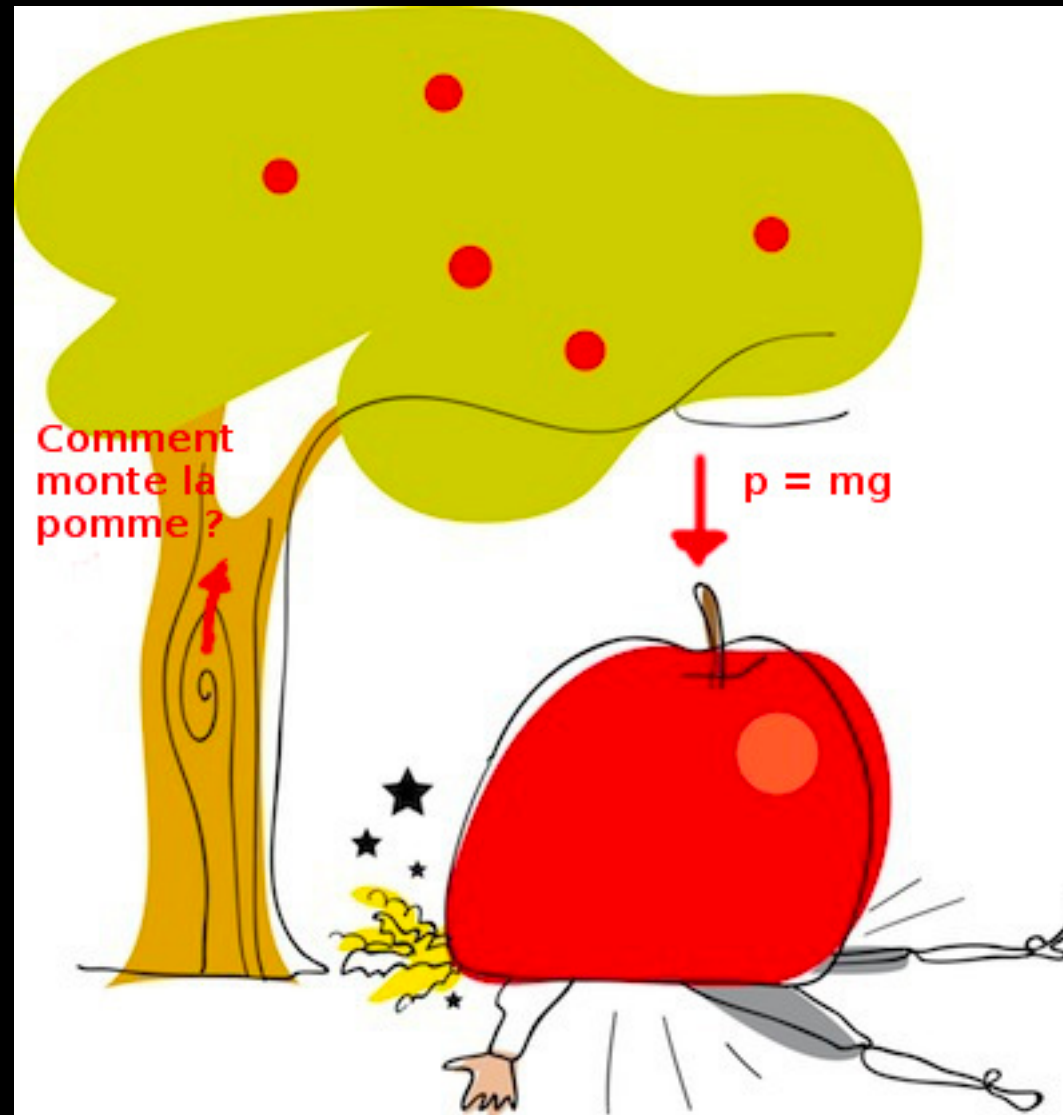
Pierre Simon Laplace

1749-1827

Our lack of knowledge is uncertainty

How do we think
about the world?





Inductive thinking
Observations -> conclusions
on governing laws



$$u^h(x) = \sum_{I=1}^m N_I(x) u_I$$

Write mathematical models from observations



Experiments to

- Formulate hypotheses and laws about the world
- Identify parameters of mathematical models



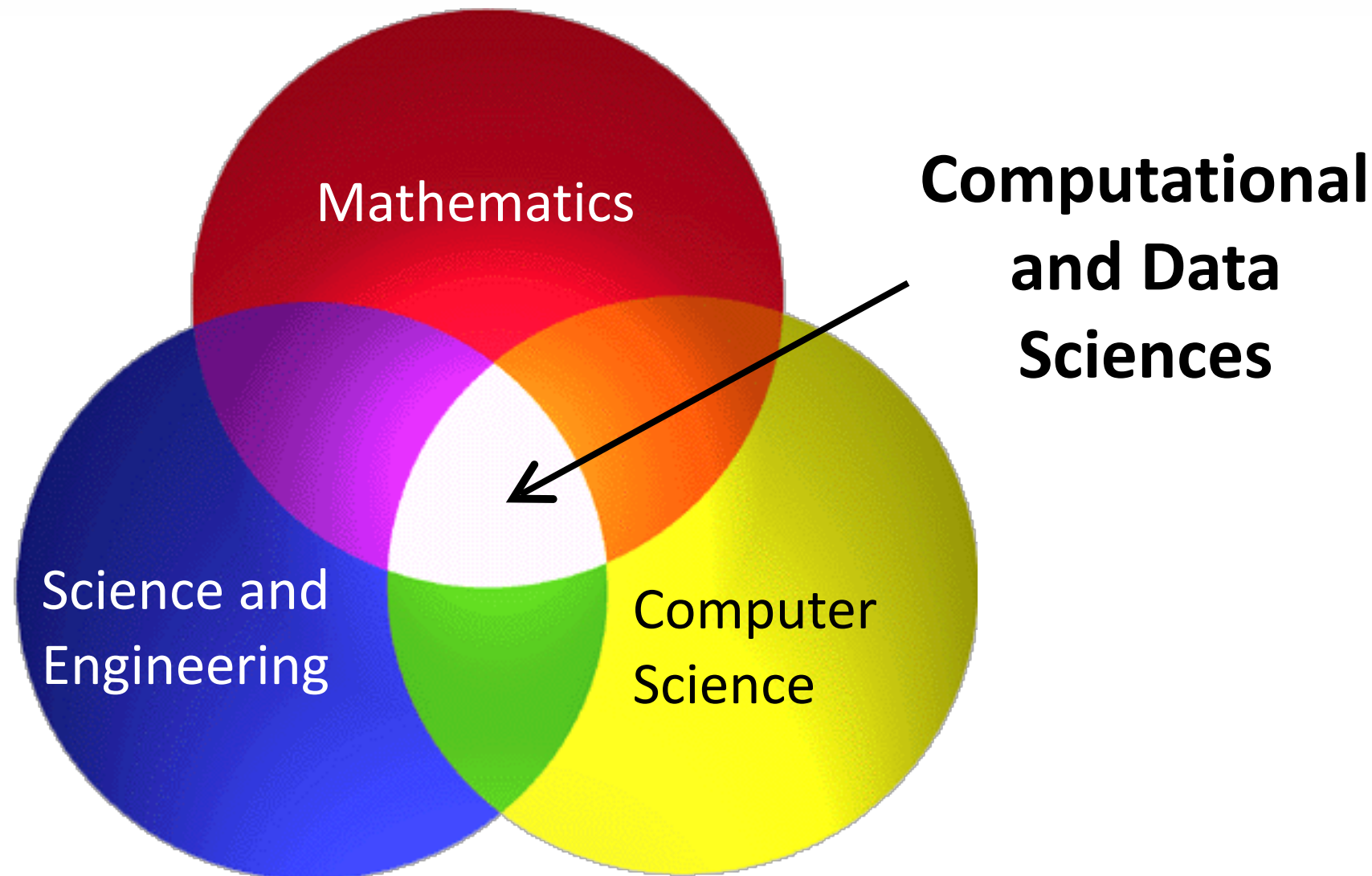
Conduct numerical experiments to

- Formulate hypotheses and laws about the world
- Identify parameters of mathematical models



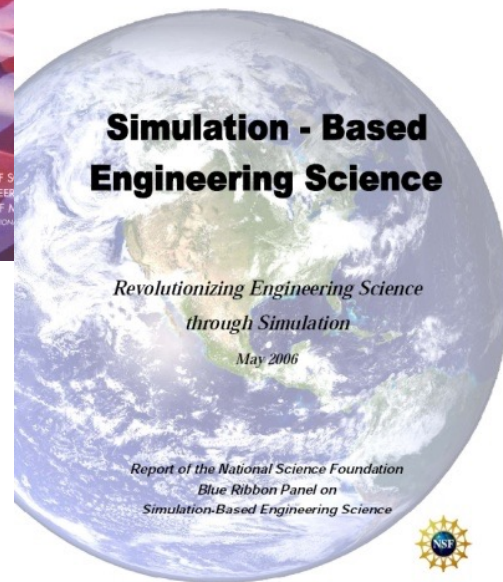
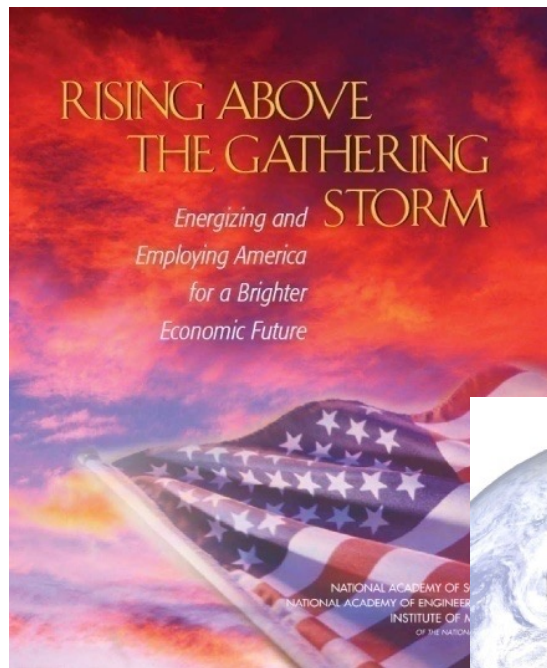
The third pillar of
scientific discovery

- Conduct numerical experiments to
- Formulate hypotheses and laws about the world
 - Identify parameters of mathematical models



Computational Sciences: the discipline concerned with the use of computational methods and devices to enable scientific discovery and engineering applications of science.

Simulation based science



RISING ABOVE THE GATHERING STORM

*Energizing and
Employing America
for a Brighter
Economic Future*

NATIONAL ACADEMY OF SCIENCES,
NATIONAL ACADEMY OF ENGINEERING, AND
INSTITUTE OF MEDICINE
OF THE NATIONAL ACADEMIES

**National Academy of Sciences
National Academy of Engineering
Institute of Medicine**

REPORT TO THE PRESIDENT

Computational Science: Ensuring America's Competitiveness

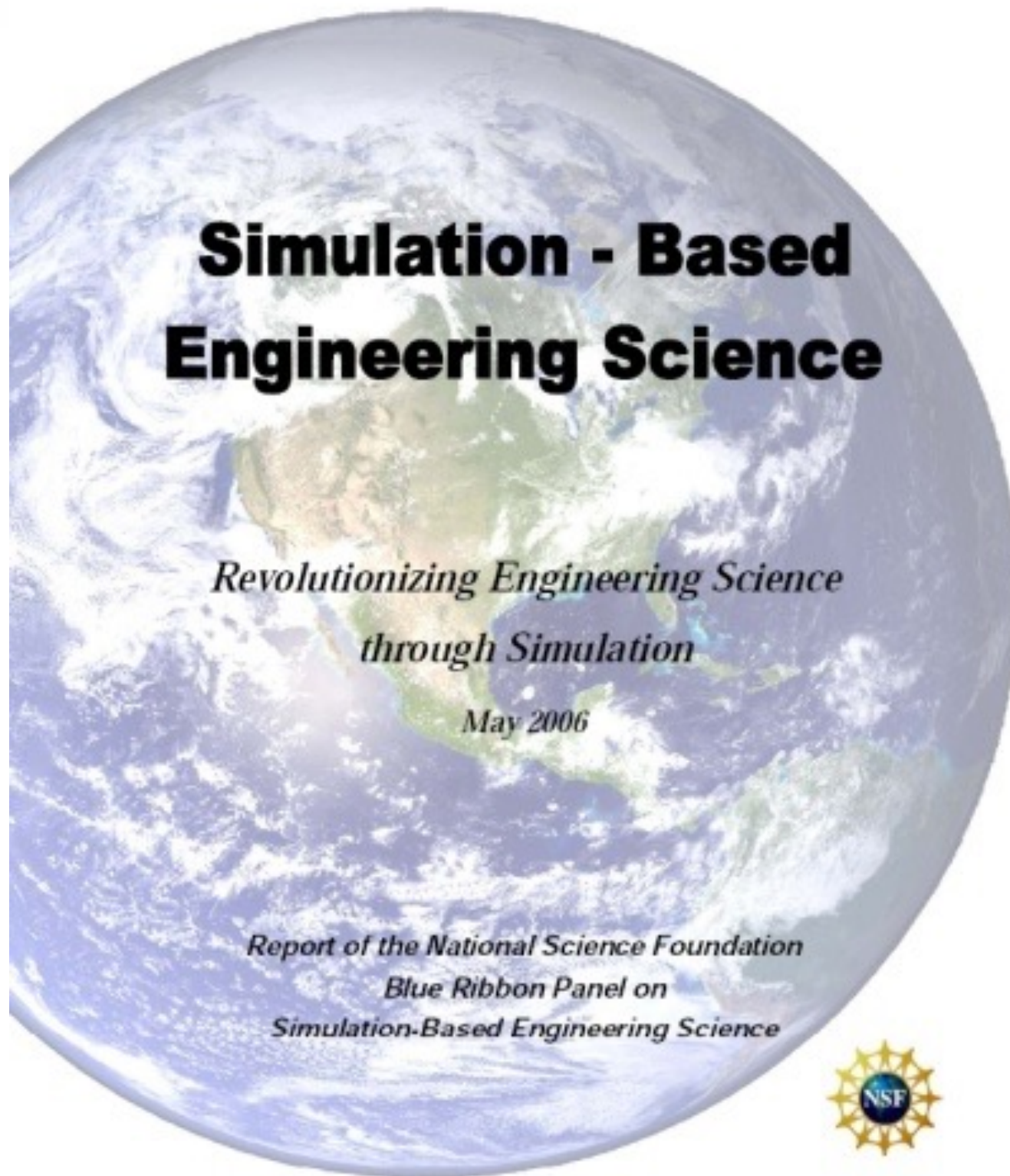
President's Information Technology
Advisory Committee

June 2005



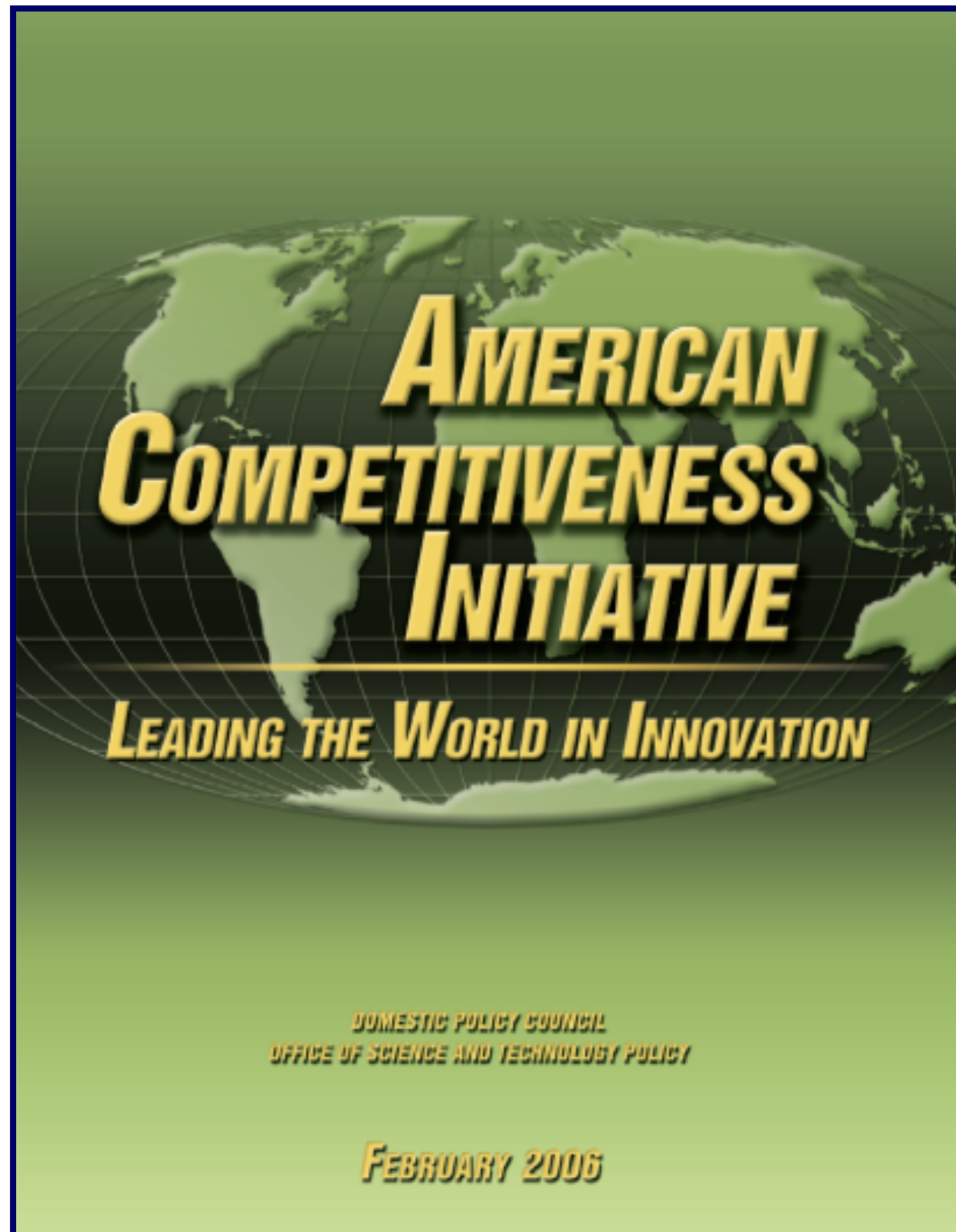
President's Information Technology
Advisory Committee

2005



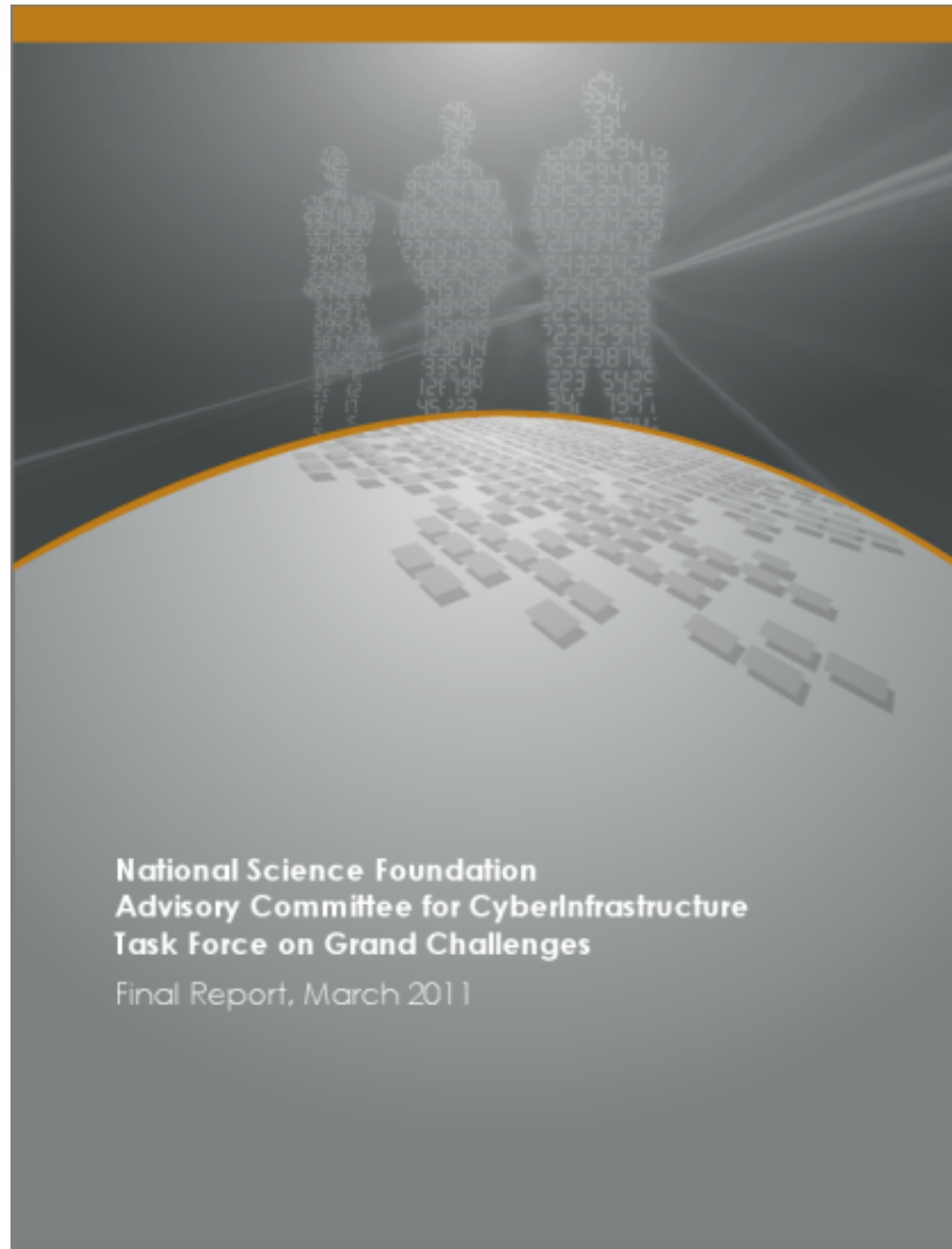
US National Science Foundation (NSF)

2006



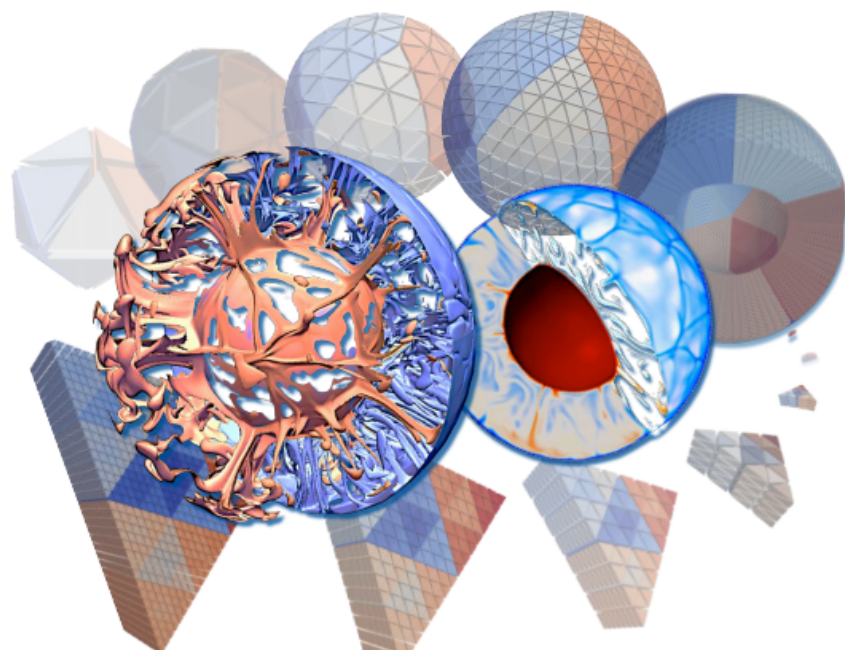
Office of Science and Technology Policy
2006

Simulation based science



Research and Education in Computational Science and Engineering

September 2016



arXiv:1610.02608v3 [cs.CE] 17 Oct 2016

Report from a workshop sponsored by the Society for Industrial and Applied Mathematics (SIAM) and the European Exascale Software Initiative (EESI-2), August 4-6, 2014, Breckenridge, Colorado

Workshop Organizers:

Officers of the SIAM Activity Group on Computational Science and Engineering (SIAG/CSE), 2013-2014:

Ulrich Rüde, Universität Erlangen-Nürnberg, Chair

Karen Willcox, Massachusetts Institute of Technology, Vice Chair

Lois Curfman McInnes, Argonne National Laboratory, Program Director

Hans De Sterck, Monash University, Secretary

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Max Gunzburger, Florida State University

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Michael Heroux, Sandia National Laboratories

Jan Hesthaven, École Polytechnique Fédérale de Lausanne

Peter Jimack, University of Leeds

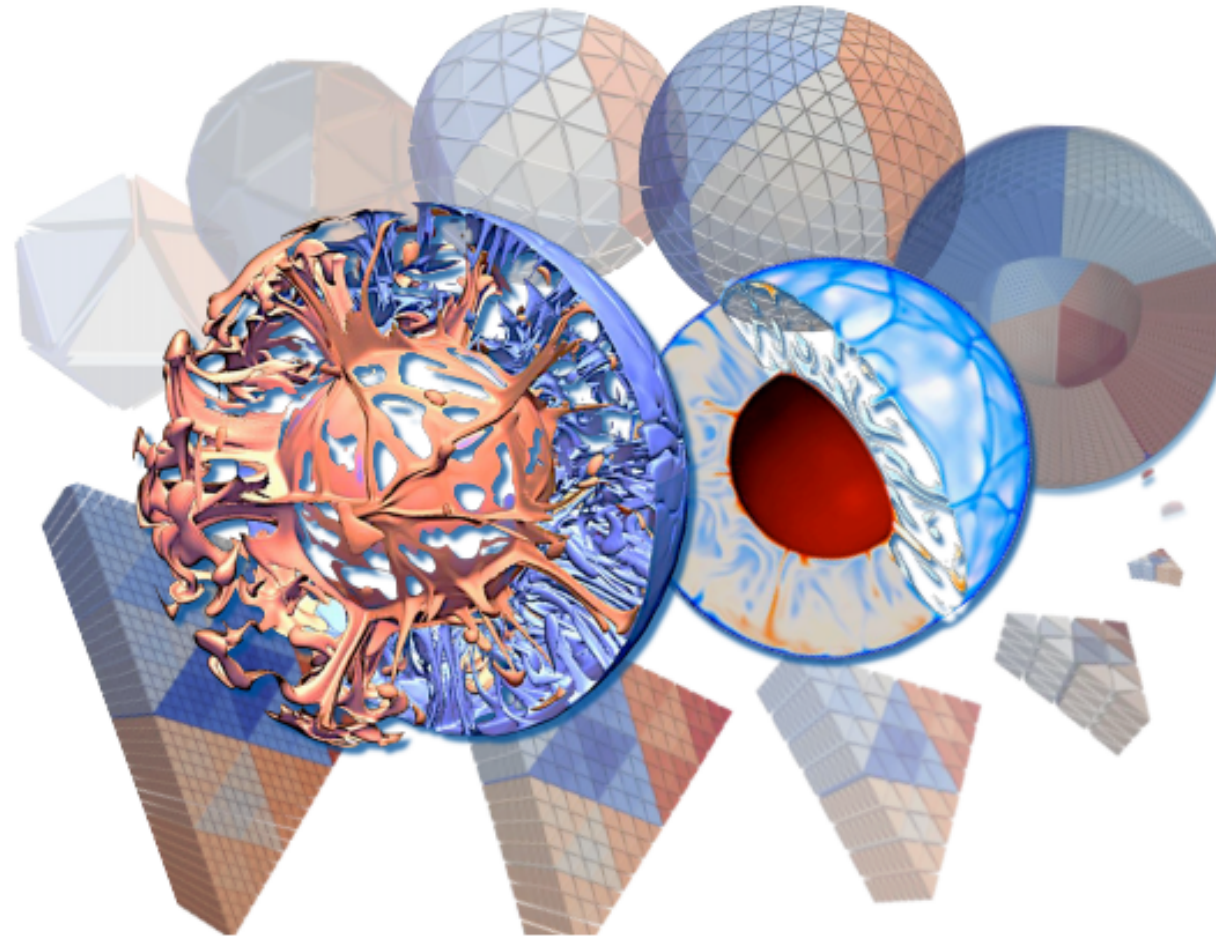
Chris Johnson, University of Utah

Kirk E. Jordan, IBM

David E. Keyes, KAUST

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Additional Contributors:

<https://arxiv.org/pdf/1610.02608.pdf>

1610.02608v3 [cs.CE] 17 Oct 2016

What computational & data sciences enable that traditional science doesn't

LOOK INTO THE PAST

- Earthquakes, climate, oil discovery, archeology, seismology, law, economics, finance

PROBE THE FUTURE

- Explore the effects of thousands of scenarios
- Drug design, space exploration, climate change, natural disasters, ...

CHOOSE MODELS

- Explore consequences of breakdown of models and theory...

OPTIMISE

- Optimize procedures, designs, products, systems, etc.

Why is multi-disciplinary research important?

- Most Grand Challenge problems today involve complex phenomena and systems that lie on **disciplinary boundaries**.
- **Interdisciplinary research** moves beyond simple collaboration and teaming to integrate **data**, methodologies, perspectives, and concepts from multiple disciplines in order to advance fundamental understanding and to solve **real world problems**.
- Interdisciplinary research holds the promise of pushing fields forward and accelerating scientific **discovery**.
- Interdisciplinary study and training **prepares a workforce** that undertakes scientific challenges in new and innovative ways.

Mission

To provide the infrastructure and intellectual leadership for developing outstanding interdisciplinary research programs in computational and data sciences.

To enable world-leading education in computational and data sciences in Luxembourg.

To promote and facilitate digital literacy and provide the fundamental building blocks necessary to the development of Industry 4.0 in Luxembourg.

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Complex systems

- Large number of interacting components
- Evolving over time
- Decentralised decisions vs. Centralised control
- **Local** interactions -> **emergence** of **global** patterns

Examples

- Biological systems (brain, cancer, bacteria...)
- Policy and government
- Environment (weather, ice sheet, pollution...)
- Economy, stock market
- Ecosystems (bats, fish...)
- Functional/sensing materials (graphene...)

Emergence

- Micro (local) level leads to patterns at the macro level
 - Ant/bee colonies
 - Housing patterns, traffic jams
 - Populations in ecosystems
 - Pressure of gases
 - Pricing
 - Effect of individual behaviours in societies

Two questions about emergence

- You know the micro, you want to understand the macro
- You observe the macro, you want to deduce the micro rules

THIS IS HARD

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What is a model?



When we were kids ;-) or grownup kids...

What is a model?



Biologists

What is a model?

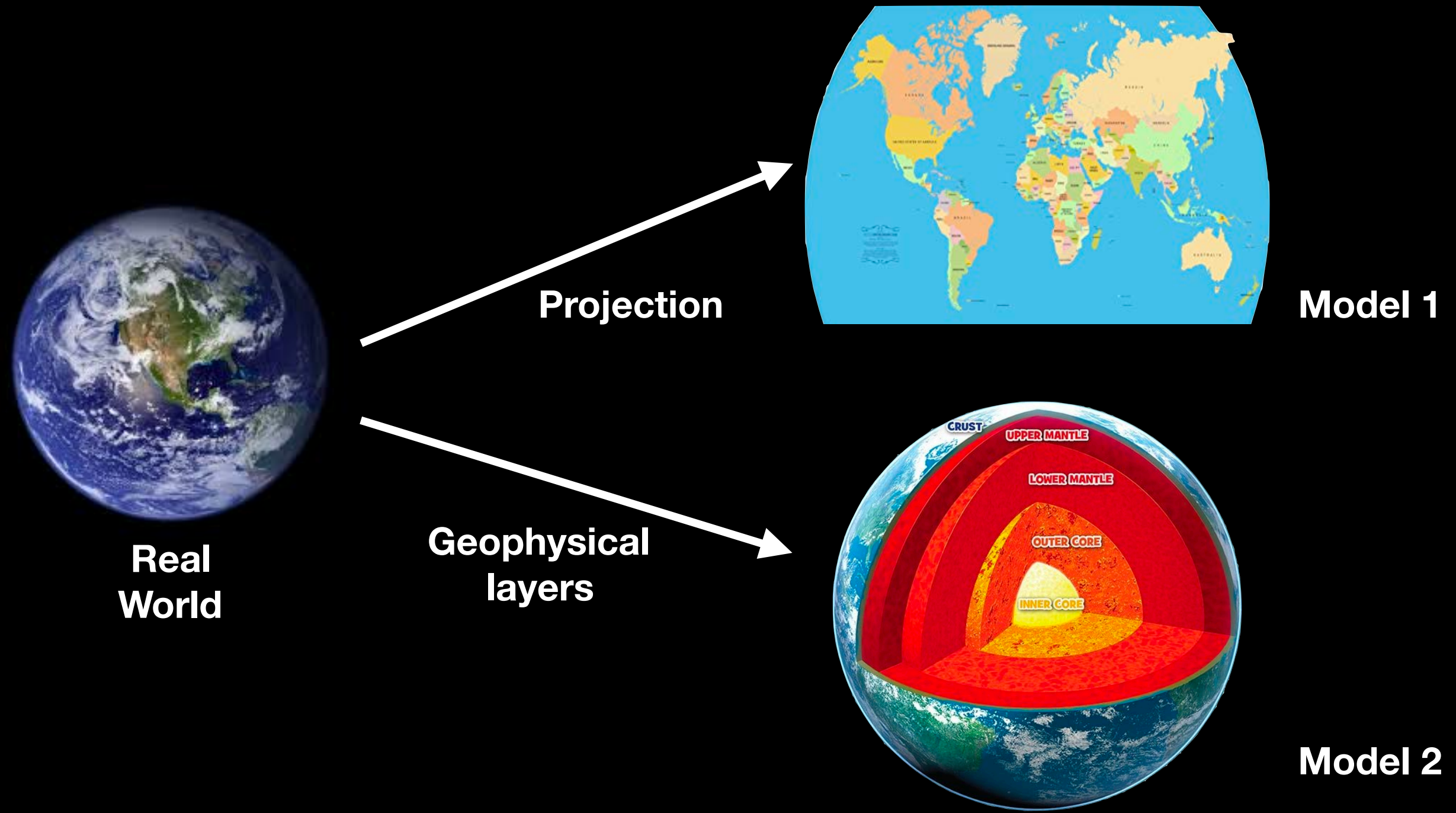


If you are in fashion...

What is a model?

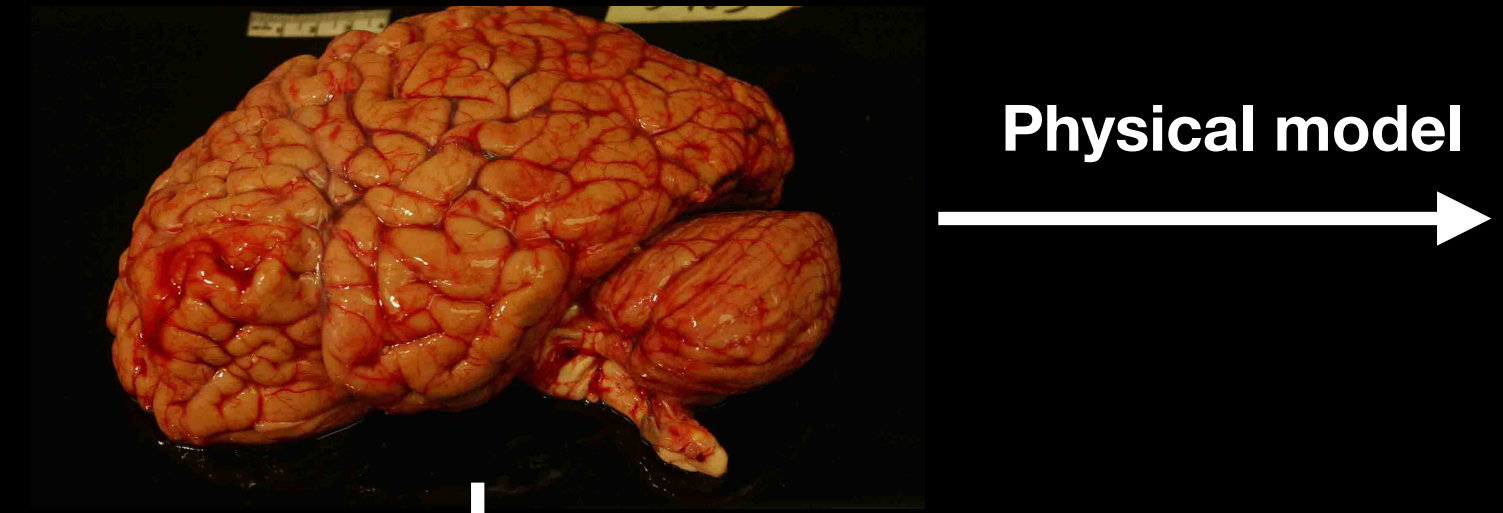
- An abstract description of a process, object, system, event which exaggerates certain aspects compared to others
- “Essentially, all models are wrong, but some are useful”
George Box, 1987
- The choice of the model depends on the quantities of interest (QoI)

What is a model?

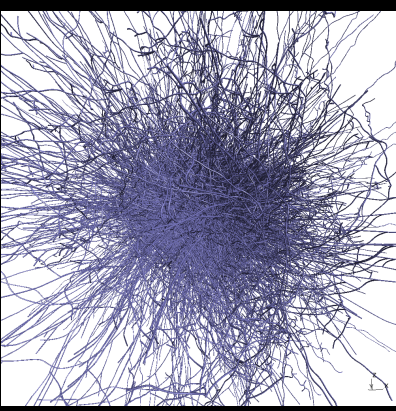
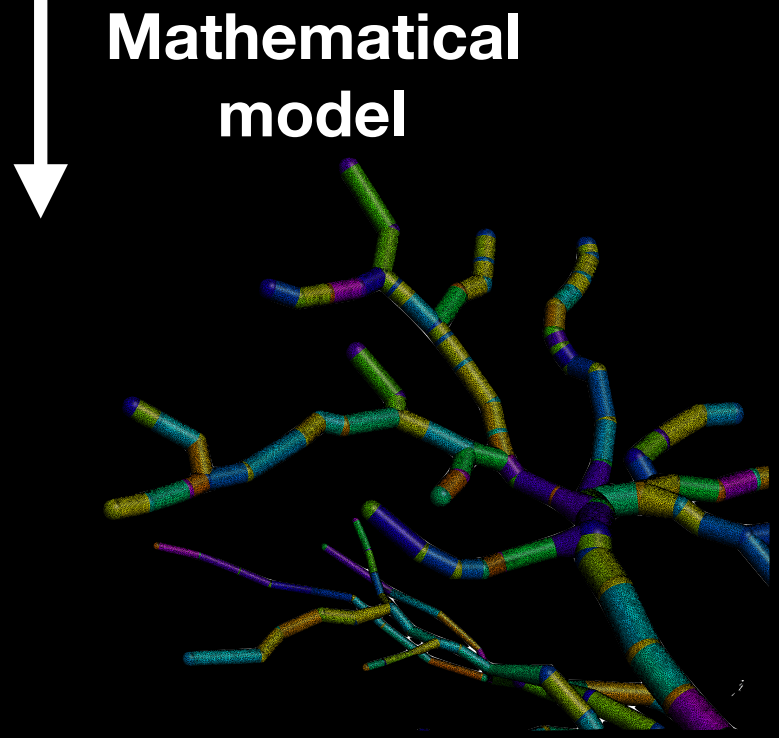
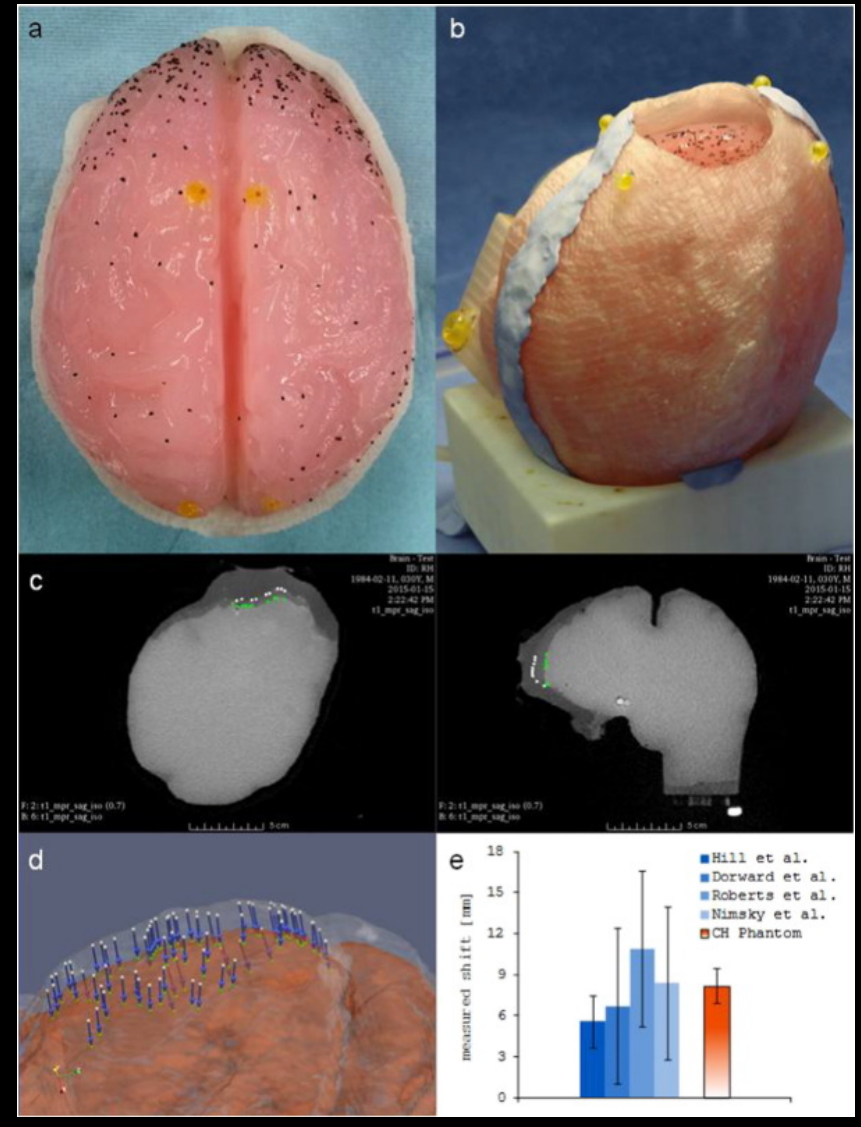


If you are into geography or geophysics

What is a model?



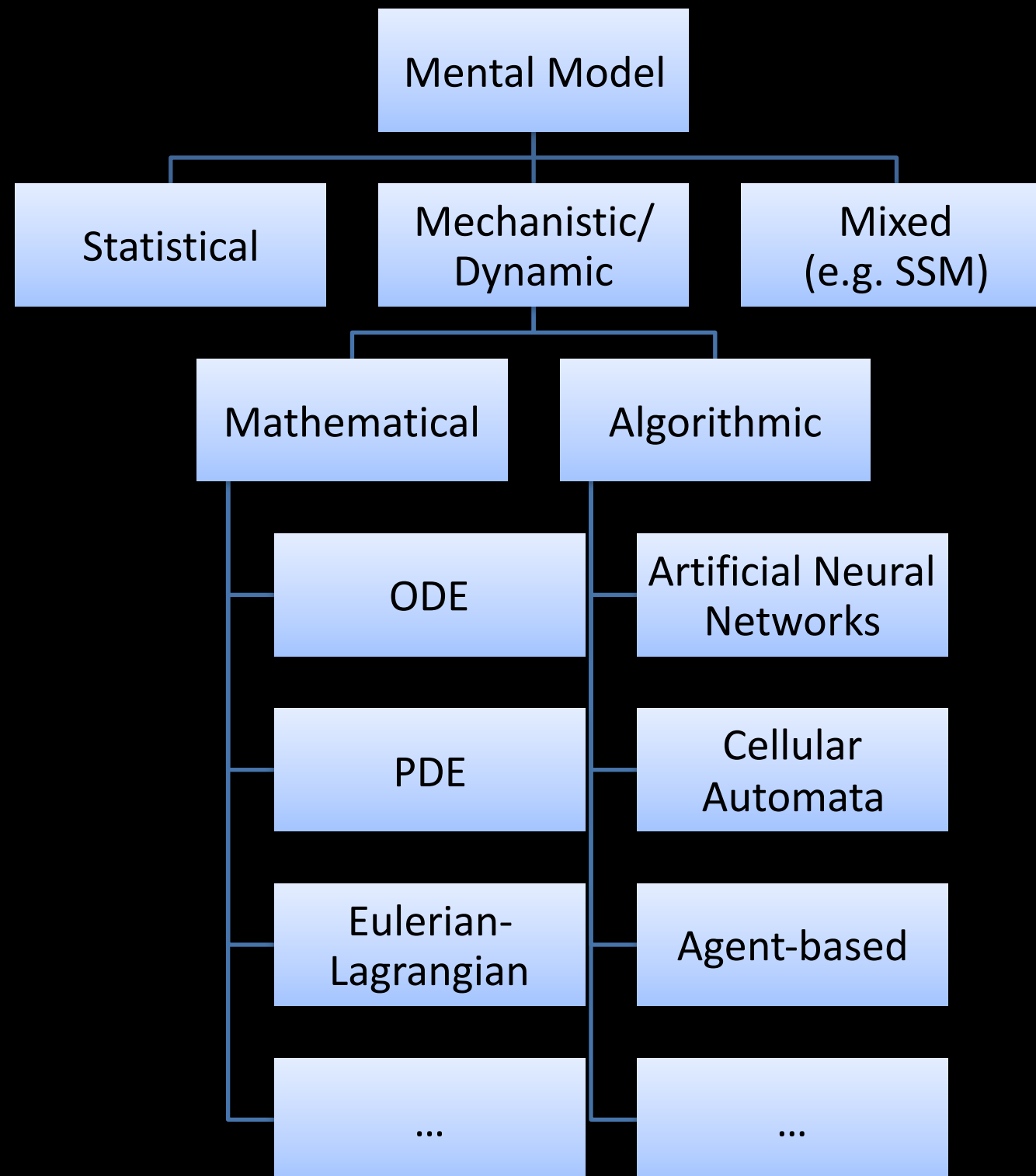
Physical model
→



J.F. Remacle
Quantity of Interest: neuro-transmission

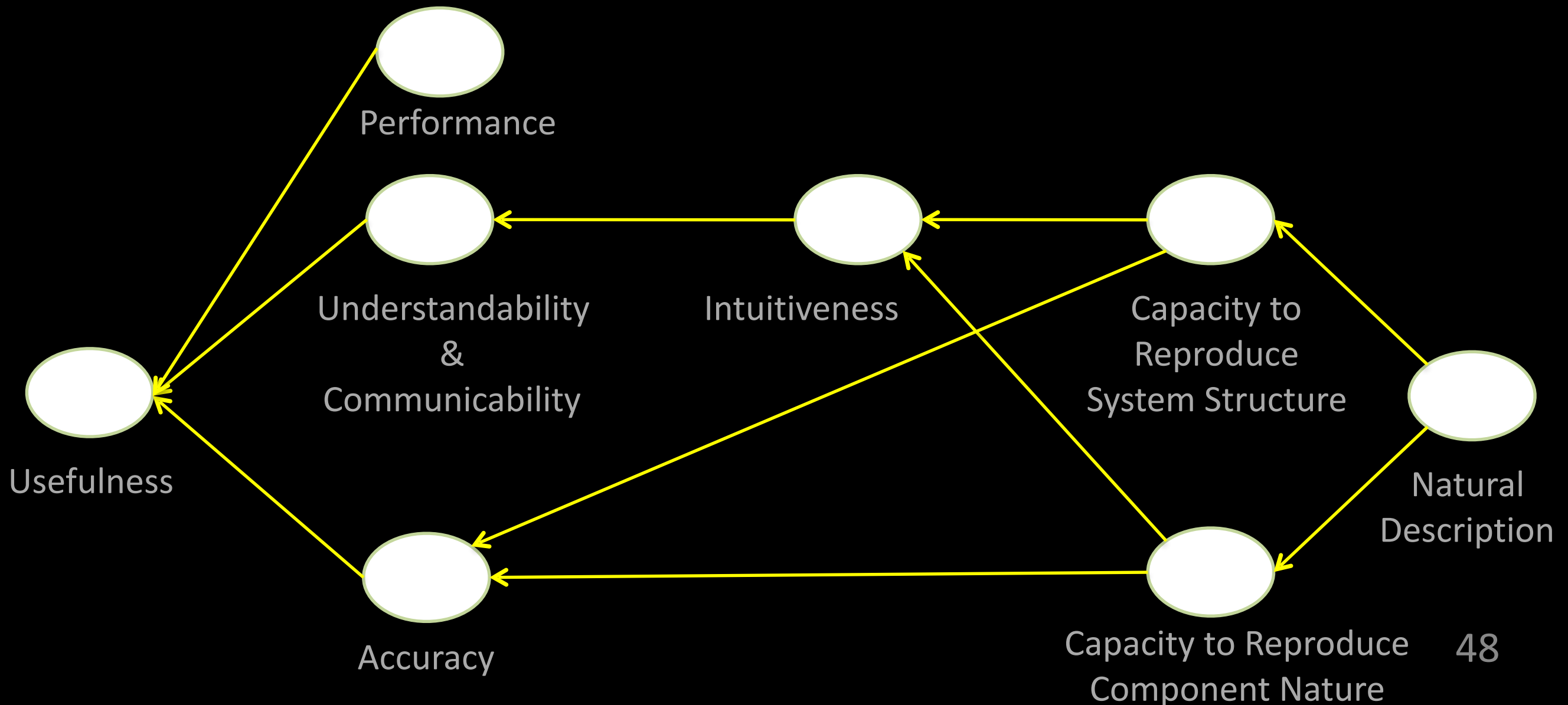
Tissue phantoms/mimics
(Dini, Imperial College London)
Quantity of Interest: stiffness

Types of models



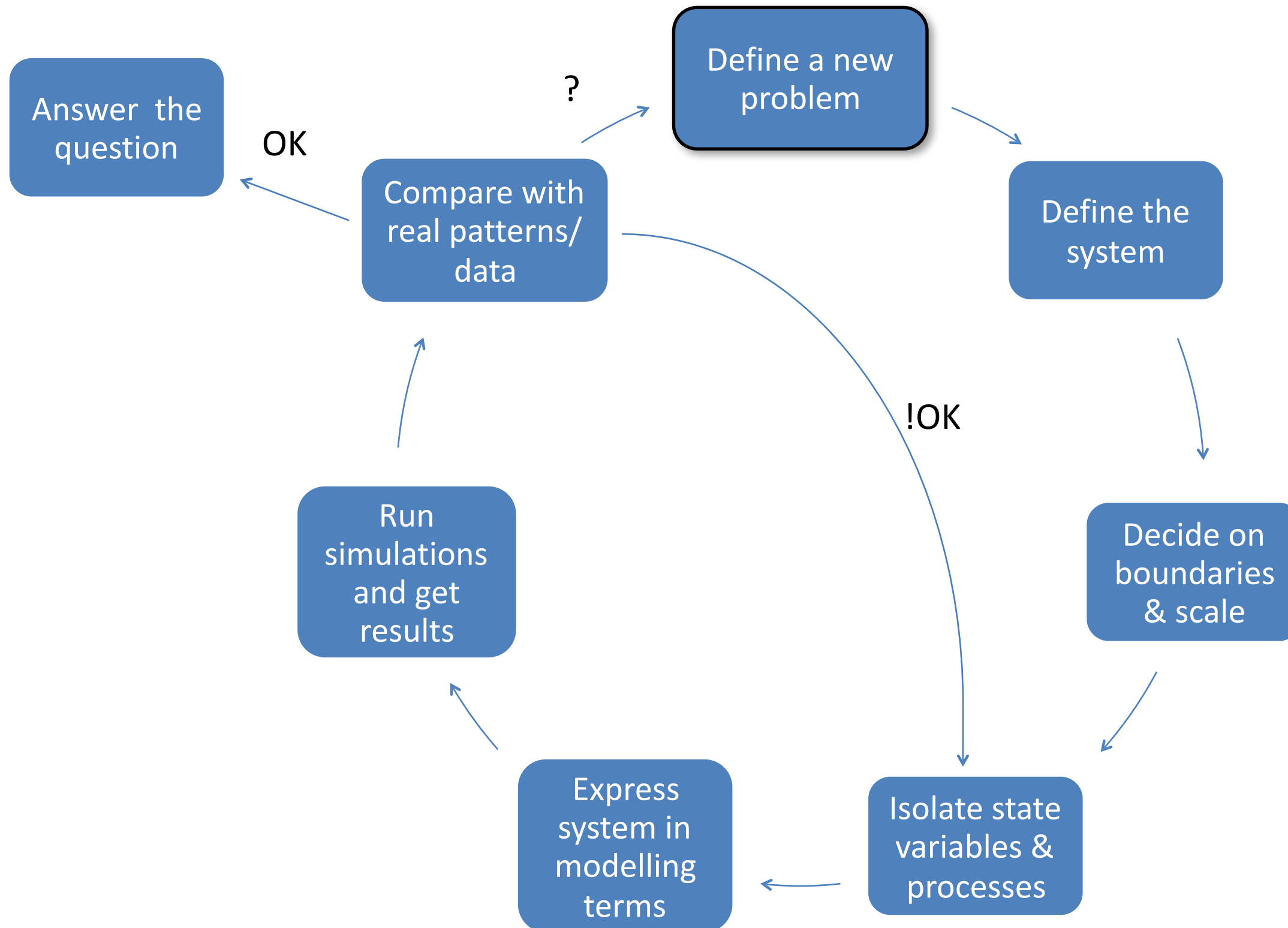
What is a “good” dynamic model?

- “All models are wrong. Some of them are useful.” -- George E.P. Box (1979)

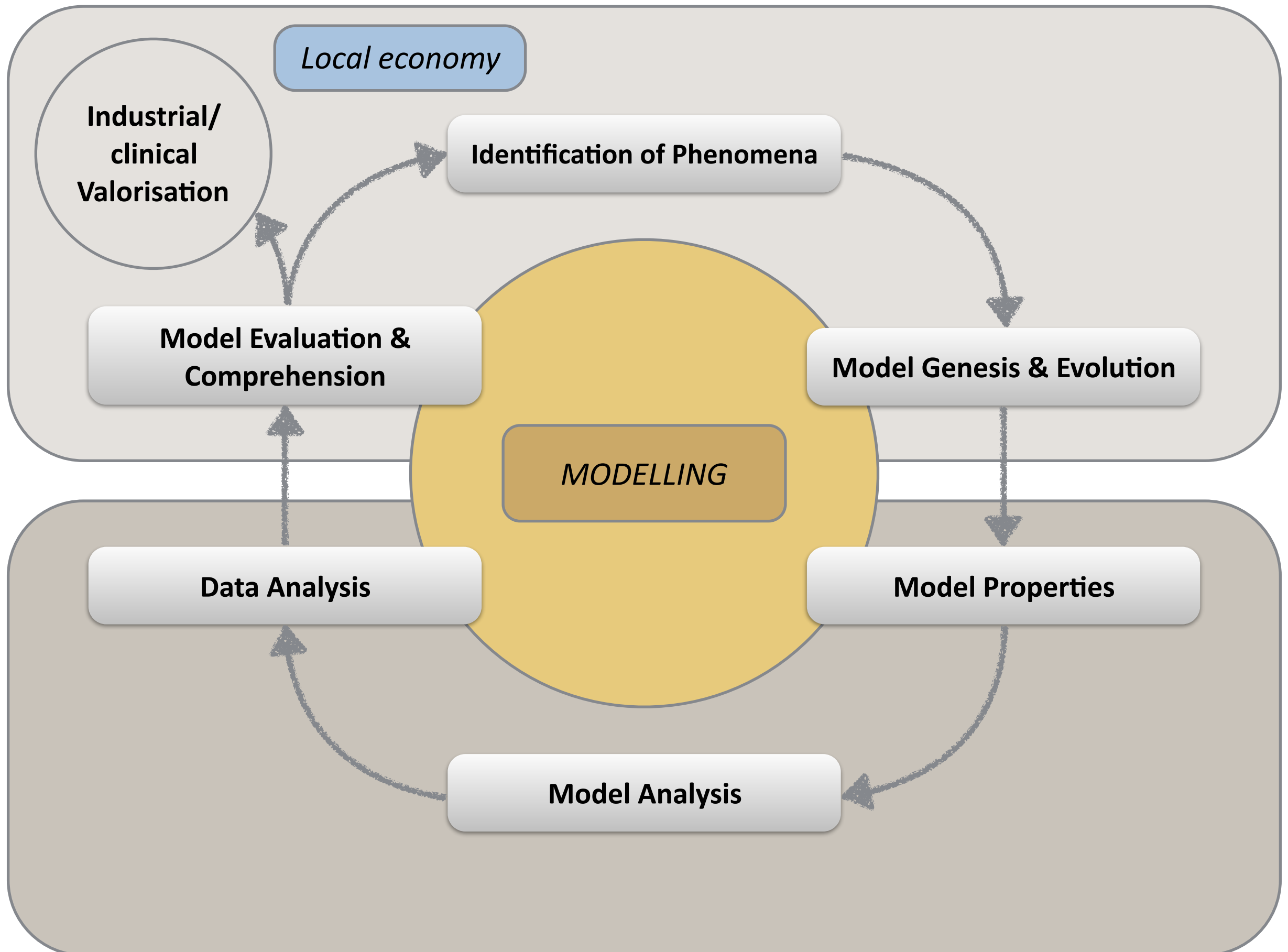


How to build dynamic models?

The modelling process



COMPUTATIONAL & DATA SCIENCES

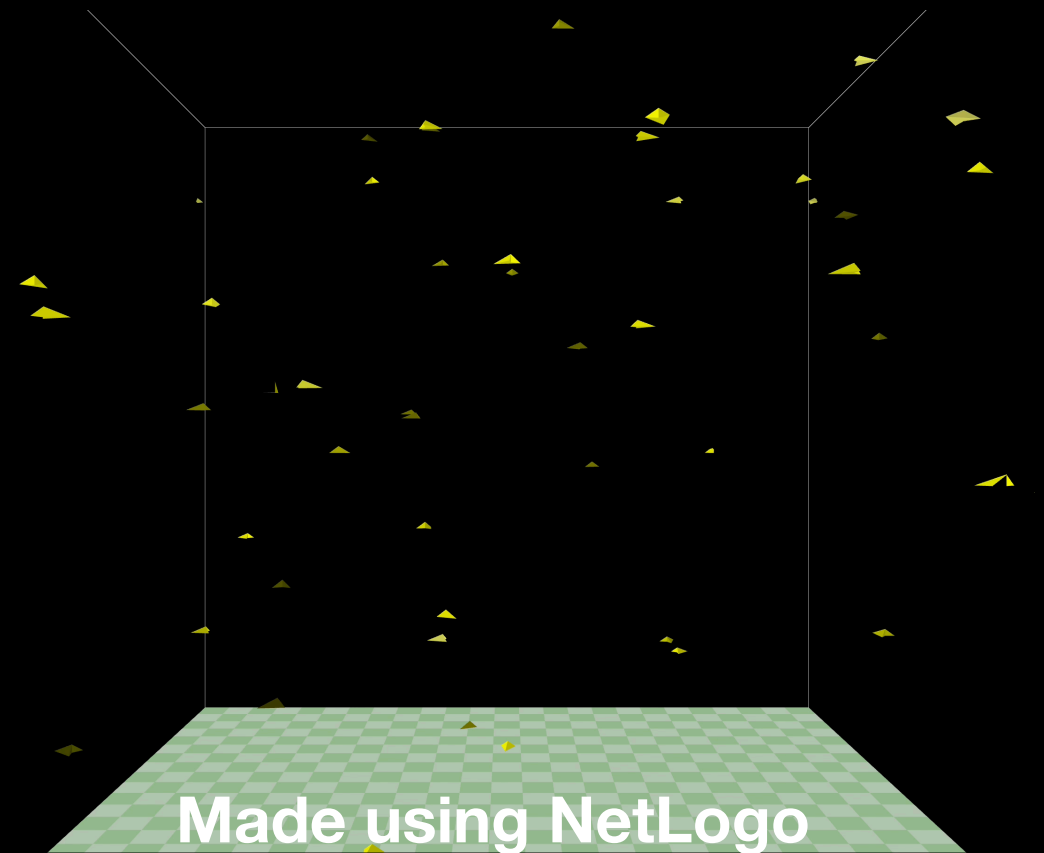


Stirling murmuration (Rome)



Flocking of Stirlings

- Is there a leader?
- Are the global patterns attributable to local rules? Can we derive global equations governing their behaviour? Is the process deterministic?
- What is their acceleration, maximum velocity, reaction time, minimum distance with other birds, line of sight, manoeuvrability?

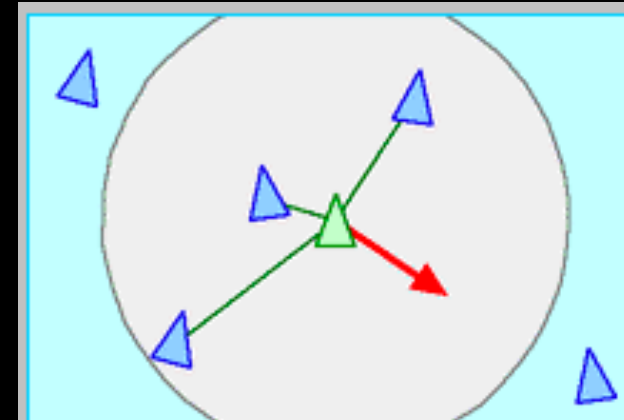
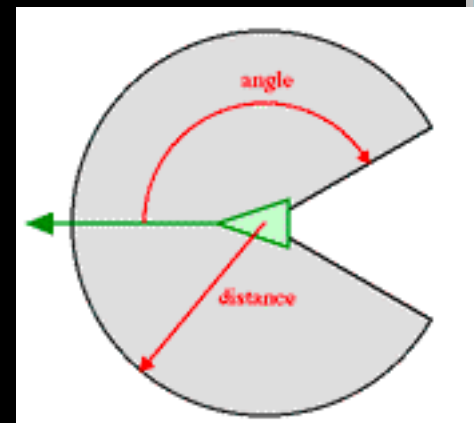
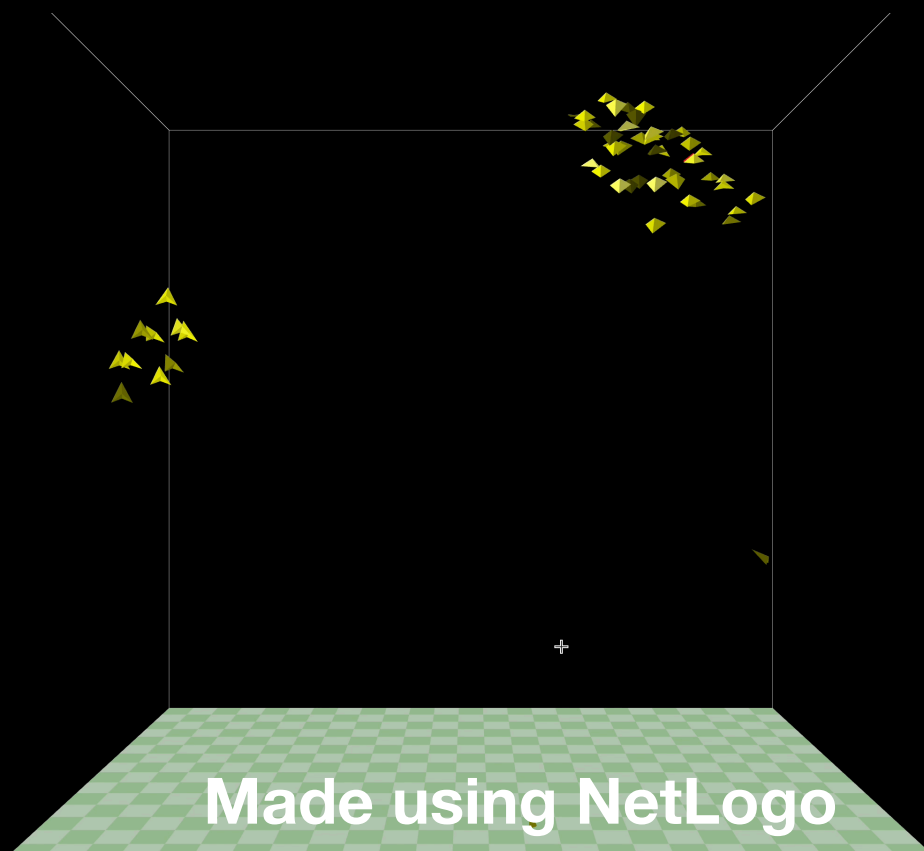


Made using NetLogo

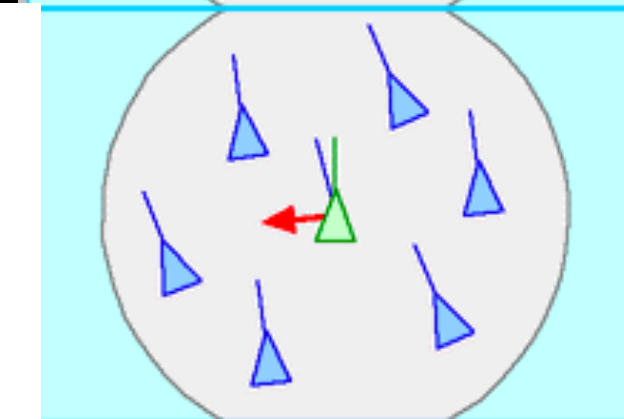
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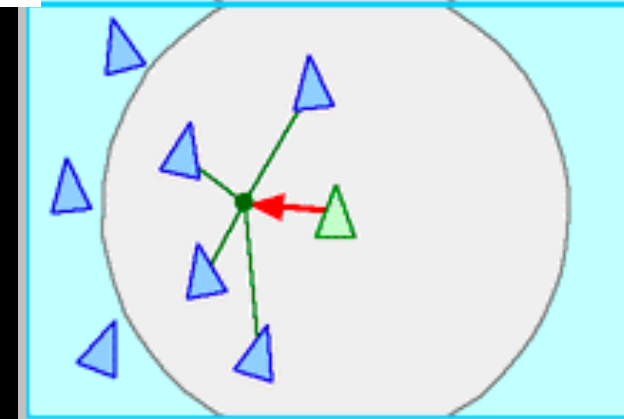
Boids flocking local rules for global behaviour



Separation: steer to avoid crowding local flockmates



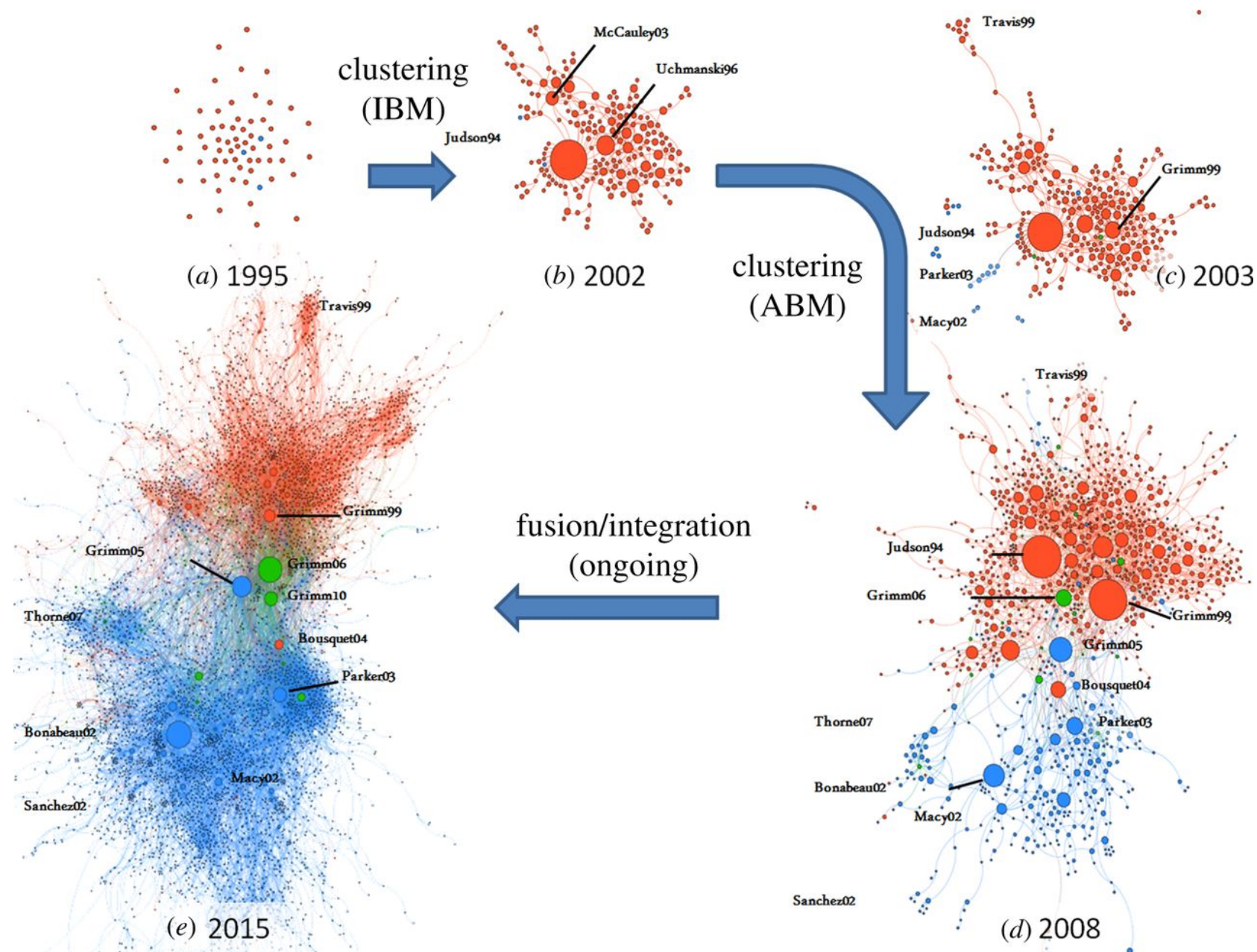
Alignment: steer towards the average heading of local flockmates



Cohesion: steer to move toward the average position of local flockmates

Applications of ABM/IBM

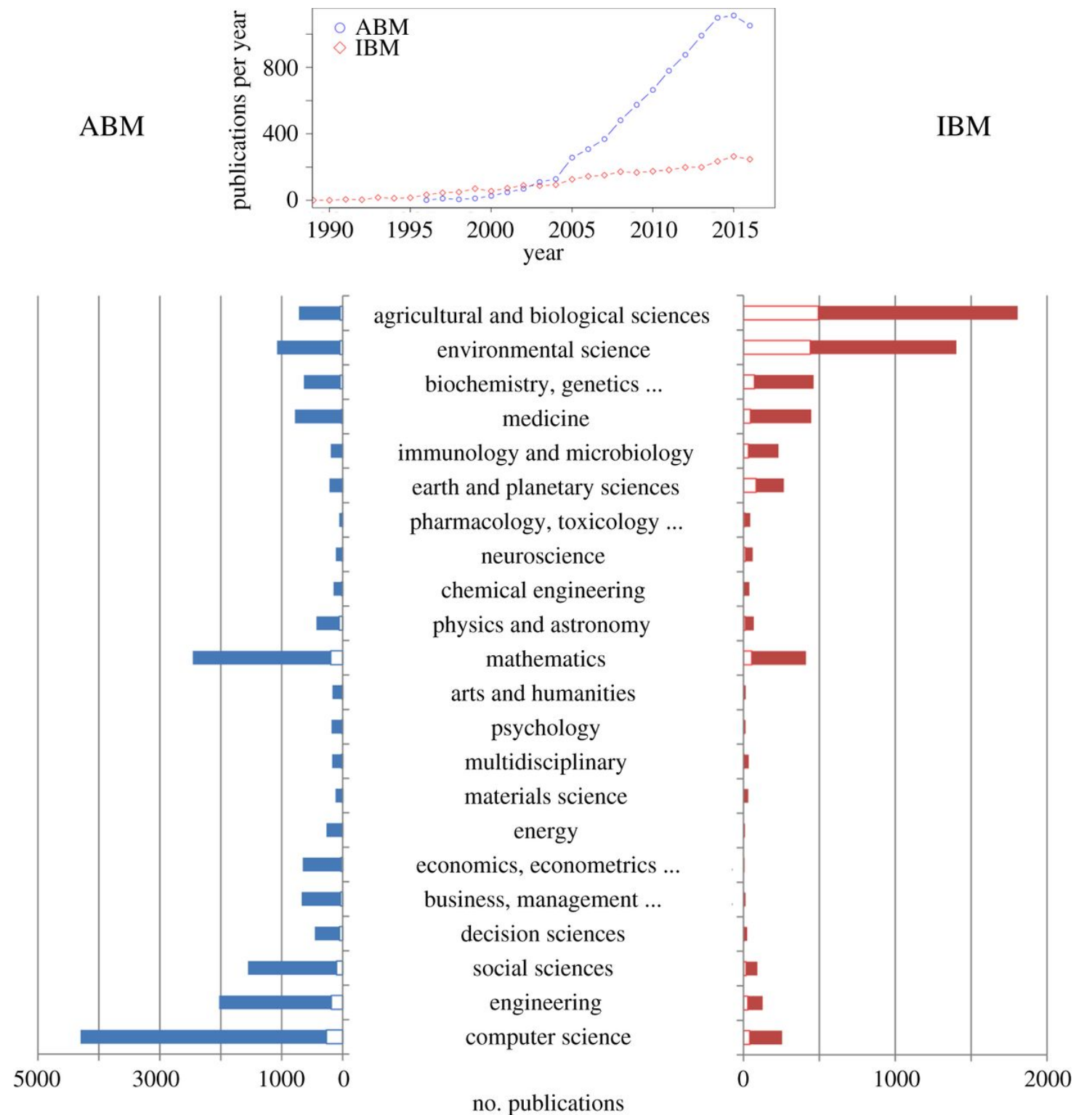
- A growing, unified community



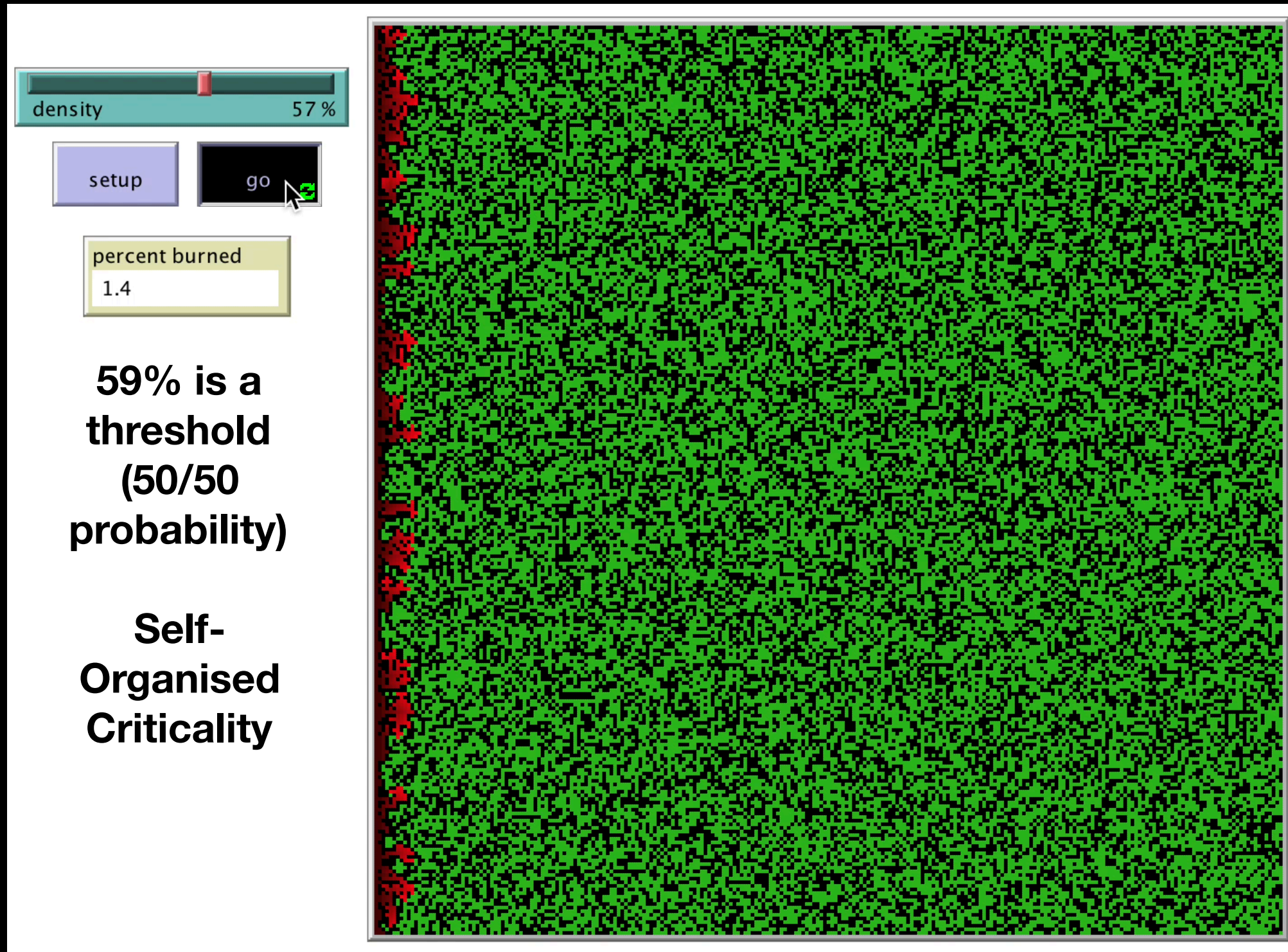
(Vincenot, 2018)

Applications of ABM/IBM

- Over 12,000 publications
- Everywhere in science!



Mathematical models



Fire spreading through a forest for various tree densities by NetLogo

Self-Organised Criticality

The Forest-Fire model belongs to the class of Self-Organized-Critical (SOC) systems, which are governed by a **slow driving energy input and burst (avalanches) of dissipative outputs** resulting often in **fractal** structures. These systems were introduced by P. Bak et al. [2] in 1987 using the example of a **sandpile** model. These SOC models can be applied to many different fields, famous applications are for instance: **earthquakes, solar flares, co-evolution, forest fires, hydraulic fracture and more. In addition they show scaling laws** and are related to critical phenomena.

http://guava.physics.uiuc.edu/~nigel/courses/563/Essays_2010/PDF/Funke.pdf

<https://pdfs.semanticscholar.org/ec58/3f6f99f1d15a1d1ae2de1d243b648efd2ba8.pdf>

<http://www.uvm.edu/pdodds/files/papers/others/1993/grassberger1993.pdf>

<https://www.sciencedirect.com/science/article/pii/B9780128001301000047>

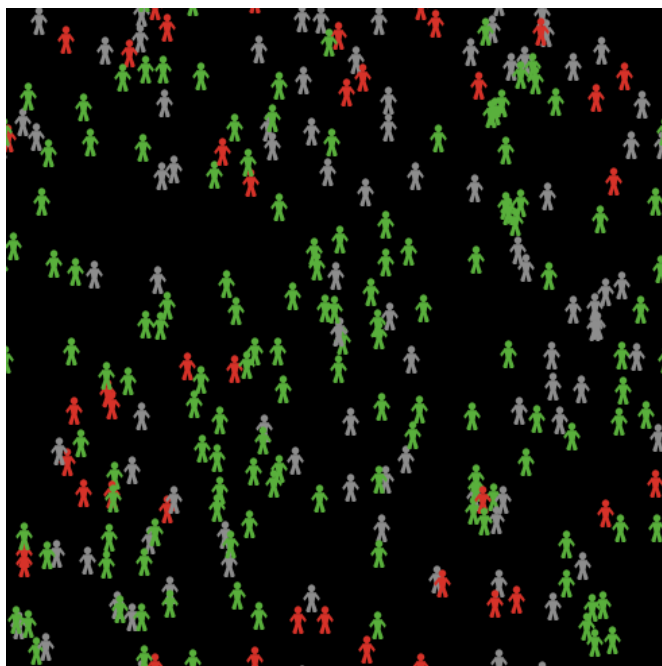
Algorithmic Languages

- ABMs are always coded as algorithms
- Most often, simple deductive behavior:
e.g. “if hungry, search food”
condition *action*
- Implementations require coding skills, but simplified languages exist

Netlogo



- Most used by ecologists
- Simplified language
- Slow, but good for relatively simple models
- Simple random movement model →



```
to setup
  clear-all
  create-turtles 10
  reset-ticks
end

to go
  ask turtles [
    fd 1 ;; forward 1 step
    rt random 10 ;; turn right
    lt random 10 ;; turn left
  ]
  tick
end
```


Netlogo

File Edit Tools Zoom Tabs Help

Interface Info Code

Edit Delete Add abc Button | normal speed | view updates | Settings... | continuous

Setup Start Virus

Disease-Variables
SARS

fatality-rate 11 %

avg.-reproductive-ratio 2.5

reproductive-ratio-range 0.5

avg.-incubation-period 6.4 days

incubation-period-range 5.0 days

symptom-length-lowest 14 days

symptom-length-highest 24 days


patient0-x 205 patient0-y -21

On Off get-again?

#-people-per-agent 5 million

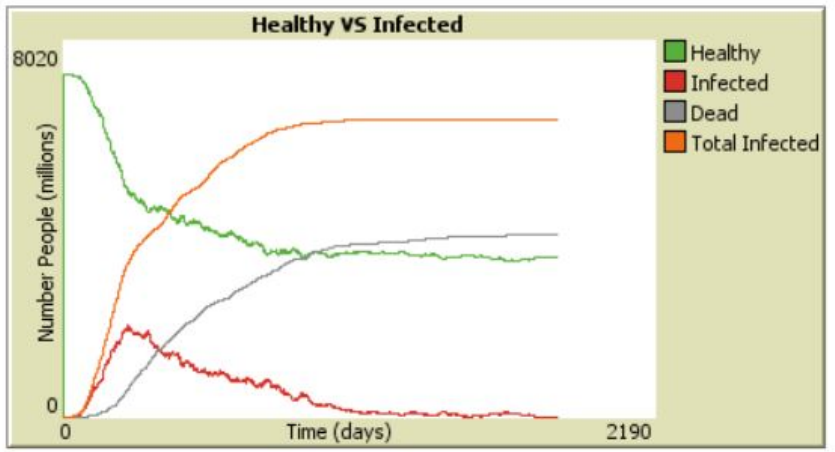
Total Turtles 1460

Days: 1825 3D



City Information:

Healthy VS Infected



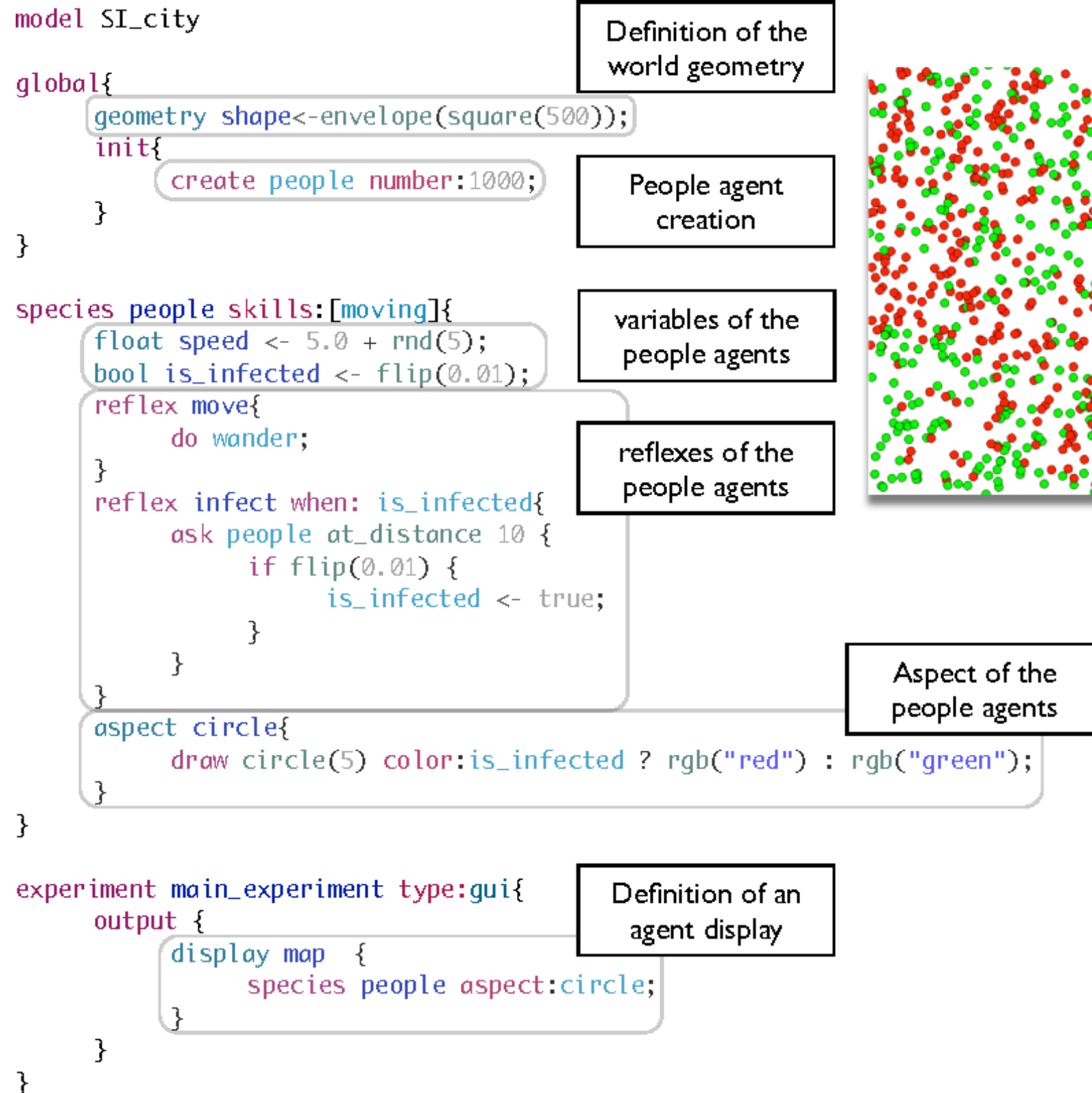
State	Value (millions)
Healthy	3400
Infected	0
Dead	3900
Total Infected	6355

Years 5

Gama



- Computer scientists
- More complex language
- Powerful, but difficult



Gama

schelling - /Users/patricktaillandier/Desktop/GAMA_15_releases/macosx.cocoa.x86/eclipse/plugins/msi.gama.models_1.0.0.201207122055/models/schelling/models/segregation_google_maps.gaml

cycle 13

Segregation

Parameters

- Environment
 - Width and height of the environment: 80
 - Name of image file to load: '../images/hanoi.png'
- Model segregation System parameters for experiment 'schelling'
 - Random number generator: 'mersenne'
 - Random seed Define: 0.0
- User interface

Charts

Unhappy = 0 (0%)
Happy = 2 035 (100%)

100
75
50
25
0

0 1 2 3 4 5 6 7 8 9 10 11 12 13 14

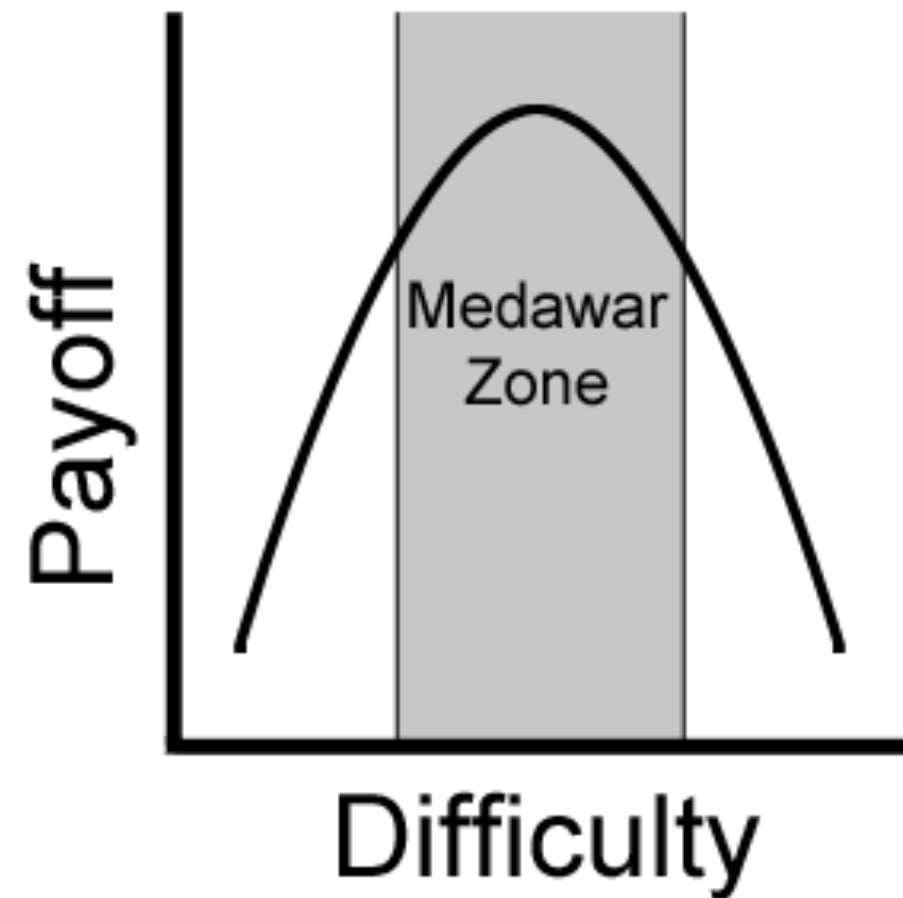
0.0

■ happy — similarity

250M of 347M

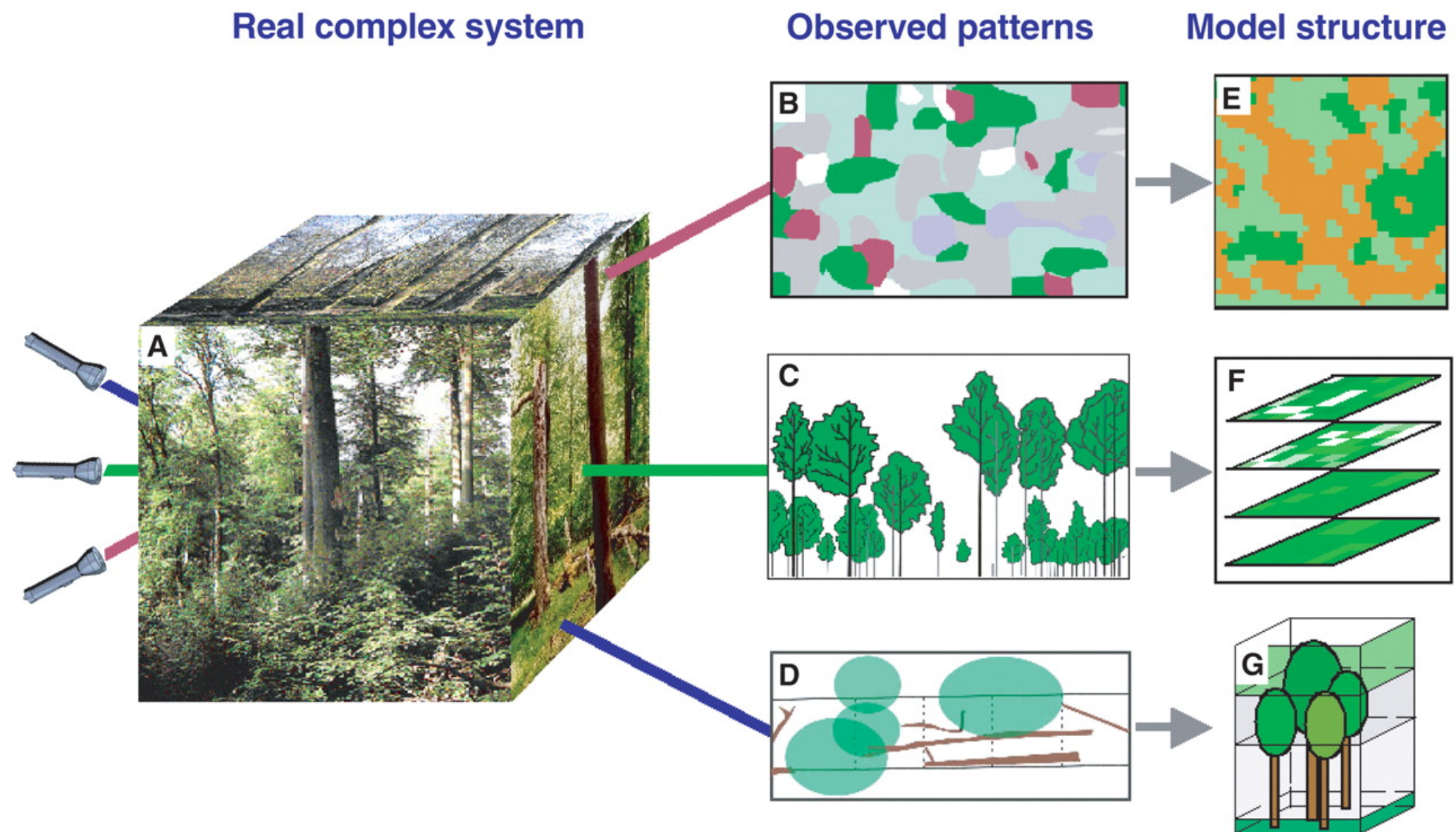
KISS

- “Keep it Simple, Stupid” (KISS) principle
- Try to make models as simple as possible
- Many processes/parameters does NOT mean better accuracy!!

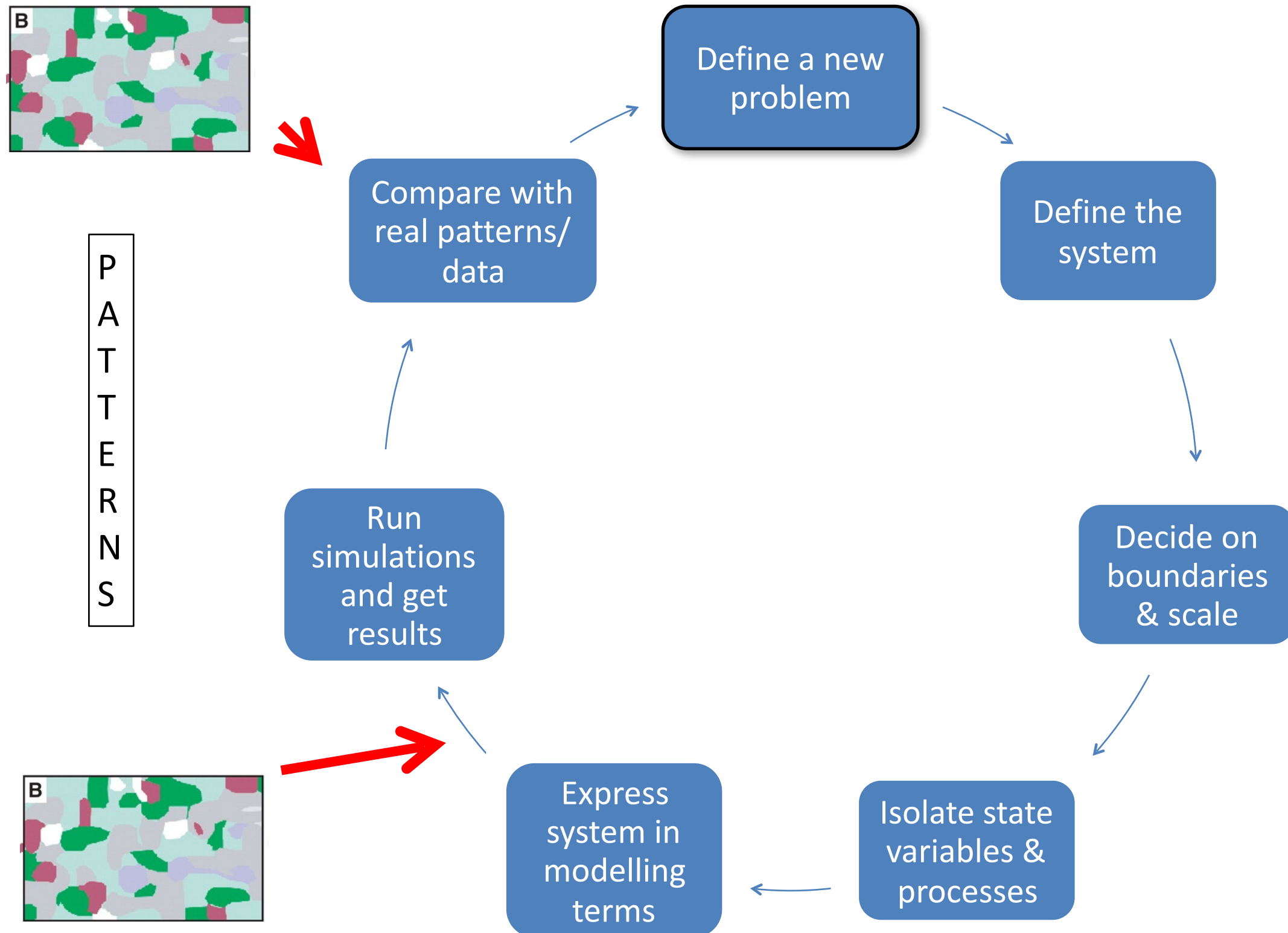


Pattern-oriented Modelling

- Use patterns instead of numerical fitting to validate model (Grimm et al. 2005)



Pattern-oriented Modelling



ODD: Describing ABM Models

- The ODD Protocol
 - Overview, Design concepts, and Details (ODD)
 - A strict set of guidelines to describe and publish ABMs
 - Guarantees the replicability of models and studies
 - More than 2000 citations
 - 2nd version already; 3rd version in preparation
 - See Grimm et al. (2006 and 2010)
- Please always use when developing ABMs!

Useful Reads

Agent-based Modelling

- Grimm V. and Railsback S.F. 2011. *Individual-based Modeling and Ecology*. **Princeton University Press**.
- Janssen M. 2017. *Introduction to Agent-based Modeling*. [Online book] <https://cbie.gitbook.io/introduction-to-agent-based-modeling/>

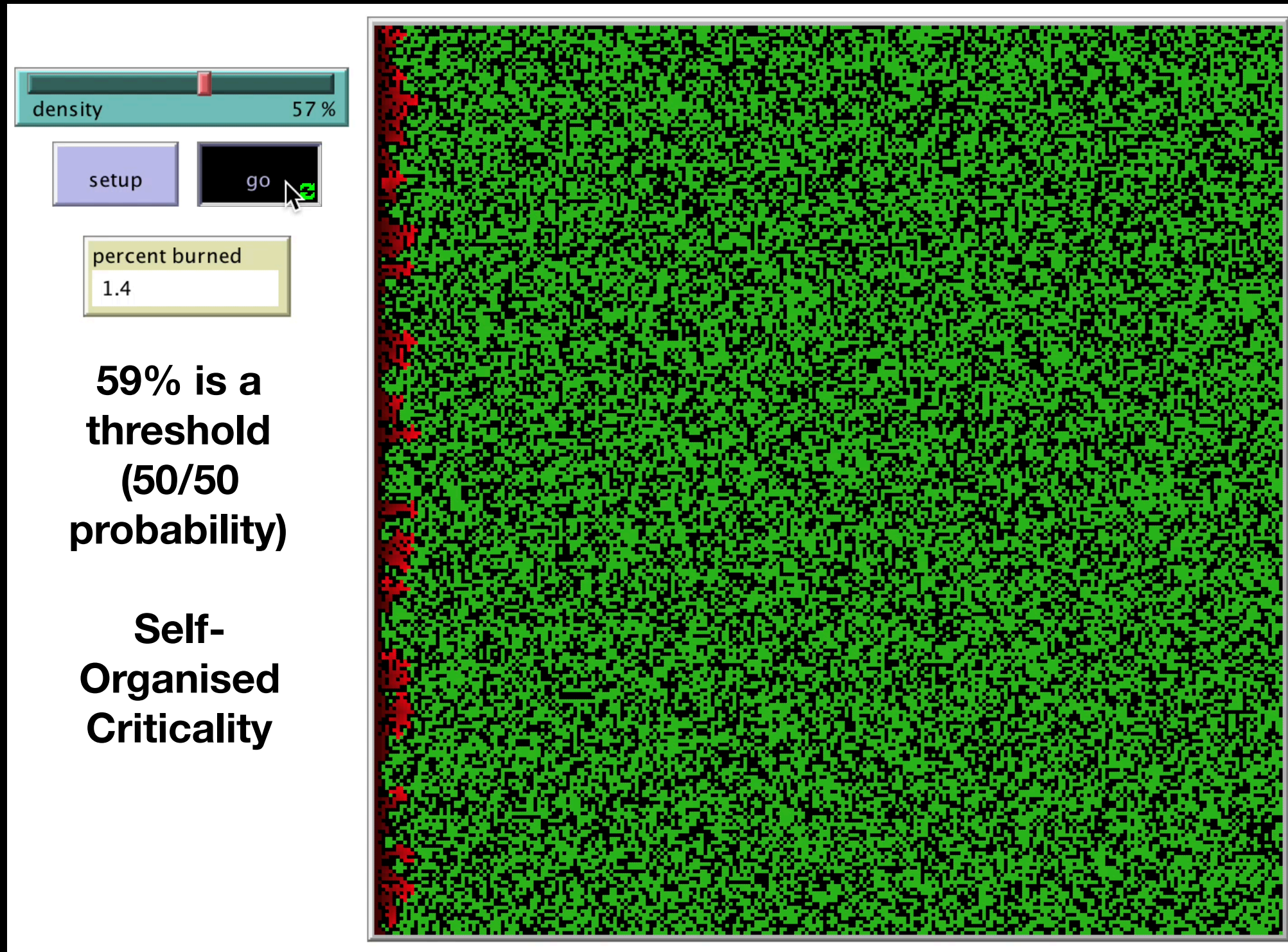
Vegetation model

- Vincenot C.E., Carteni F., Bonanomi G., Mazzoleni S., Giannino F. 2017. *Plant-soil negative feedback explains vegetation dynamics and patterns at multiple scales*. **Oikos**, 126:1319-1328.
- Vincenot C.E., Carteni F., Mazzoleni S., Rietkerk M., Giannino F. 2016. *Spatial Self-Organization of Vegetation Subject to Climatic Stress—Insights from a System Dynamics—Individual-Based Hybrid Model*. **Frontiers in Plant Science** fpls.2016.00636.

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Mathematical models



Fire spreading through a forest for various tree densities by NetLogo

Physics-based or mathematical models

Physics-based model

T (K) is the temperature of the fire layer,

$S \in [0, 1]$ is the fuel supply mass fraction (the relative amount of fuel remaining),

k ($m^2 s^{-1}$) is the thermal diffusivity,

A ($K s^{-1}$) is the temperature rise per second at the maximum burning rate with full initial fuel load and no cooling present,

B (K) is the proportionality coefficient in the modified Arrhenius law,

C (K^{-1}) is the scaled coefficient of the heat transfer to the environment,

C_S (s^{-1}) is the fuel relative disappearance rate,

T_a (K) is the ambient temperature, and

\vec{v} (ms^{-1}) is the wind speed given by atmospheric data or model.

The model is derived from the conservation of energy, balance of fuel supply, and the fuel reaction rate:

$$\frac{dT}{dt} = \nabla \cdot (k \nabla T) - \vec{v} \cdot \nabla T + A \left(S e^{-B/(T-T_a)} - C(T - T_a) \right), \quad (1)$$

$$\frac{dS}{dt} = -C_S S e^{-B/(T-T_a)}, \quad T > T_a, \quad (2)$$

$$\frac{dT}{dt} = \underbrace{\nabla \cdot (k \nabla T)}_{\text{Heat diffusion}} - \underbrace{\vec{v} \cdot \nabla T}_{\text{Advection}} + \underbrace{A \left(S e^{-B/(T-T_a)} - C (T - T_a) \right)}_{\text{Fuel consumption}}, \quad (1)$$

$$\frac{dS}{dt} = -C_S S e^{-B/(T-T_a)}, \quad T > T_a, \quad (2)$$

with the initial values

$$S(t_{\text{init}}) = 1 \text{ and } T(t_{\text{init}}) = T_{\text{init}}. \quad (3)$$

The diffusion term $\nabla \cdot (k \nabla T)$ models short-range heat transfer by radiation in a semi-permeable medium, $\vec{v} \cdot \nabla T$ models heat advected by the wind, $S e^{-B/(T-T_0)}$ is the rate fuel is consumed due to burning, and $AC(T - T_a)$ models the convective heat lost to the atmosphere. The reaction rate $e^{-B/(T-T_a)}$ is obtained by modifying the reaction rate $e^{-B/T}$ from the Arrhenius law by an offset to force zero reaction at ambient temperature, with the resulting reaction rate smoothly dependent on temperature.

Questions: Identification of parameters?

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- Scientific method, experiments, analytical methods and models
- What is a complex system?
- What models are available to understand and predict?
- What are agent-based models?
- Equation-based (mathematical) models
- How to choose the “best” model?

When use agent-based models

- Not too few, nor too many - medium numbers of entities
- Heterogeneous systems
- Local interactions
- Rich environments

Not too many, not too few

- Casti ,1996
 - Too few agents, the system is too simple: game theory and ethnography are sufficient
 - Too many agents, averages work well, statistical descriptions

Richness of the environments

- Social networks
- Geographical systems
- The environment can itself be an agent

Compare agent-based models (ABM), mathematical models (MM), statistical models (SM)

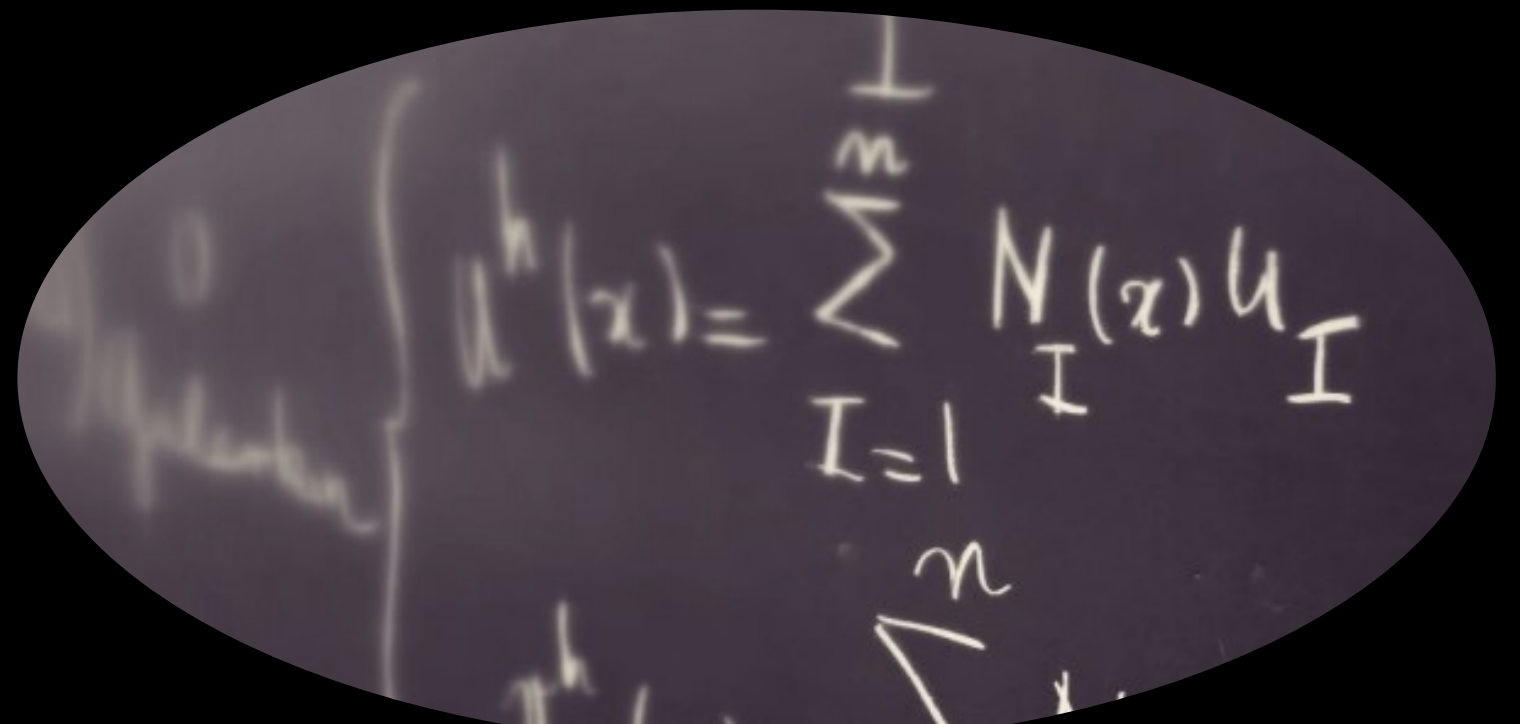
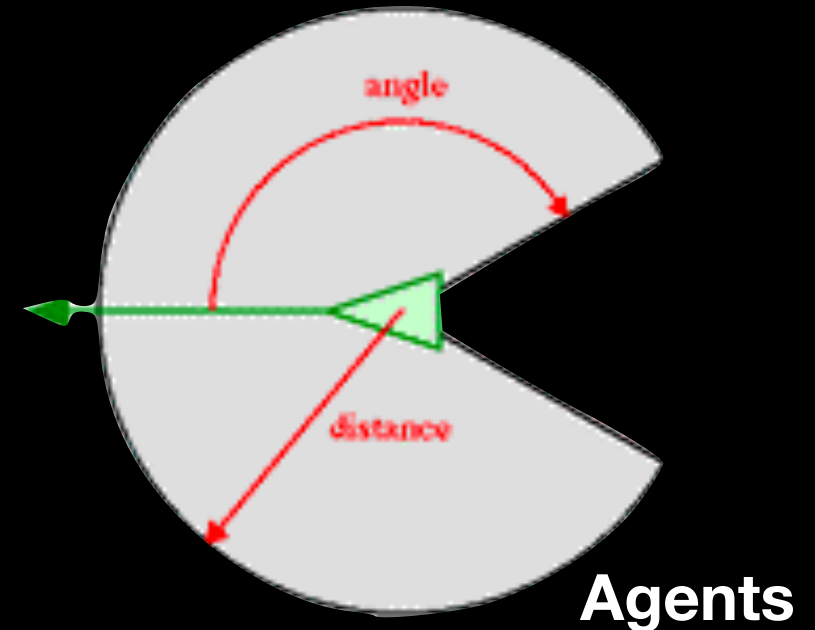
- MM are often continuous: Nano-wolf problem (Wilson 1998), but if you can write equations, do it
- ABM and MM approaches need parameters which are difficult to measure (agents need local parameters and rules)
- SM need large data sets of high quality
- ABM can be coupled to MM
- Both ABM and MM can learn from SM and Machine Learning and become adaptive

Some future prospects

- Lab experiments are costly, they are sometimes impossible, dangerous... they can help generate theories
- ABM or MM can be created from lab experiments
- Models can help scale up from experiments
- Experiments are done within a laboratory setting, far removed from real-life, where the models would be used
- Digital twins can avoid such issues, but are still illusory in practice
- Models can help provide insights into sensitivities and uncertainties

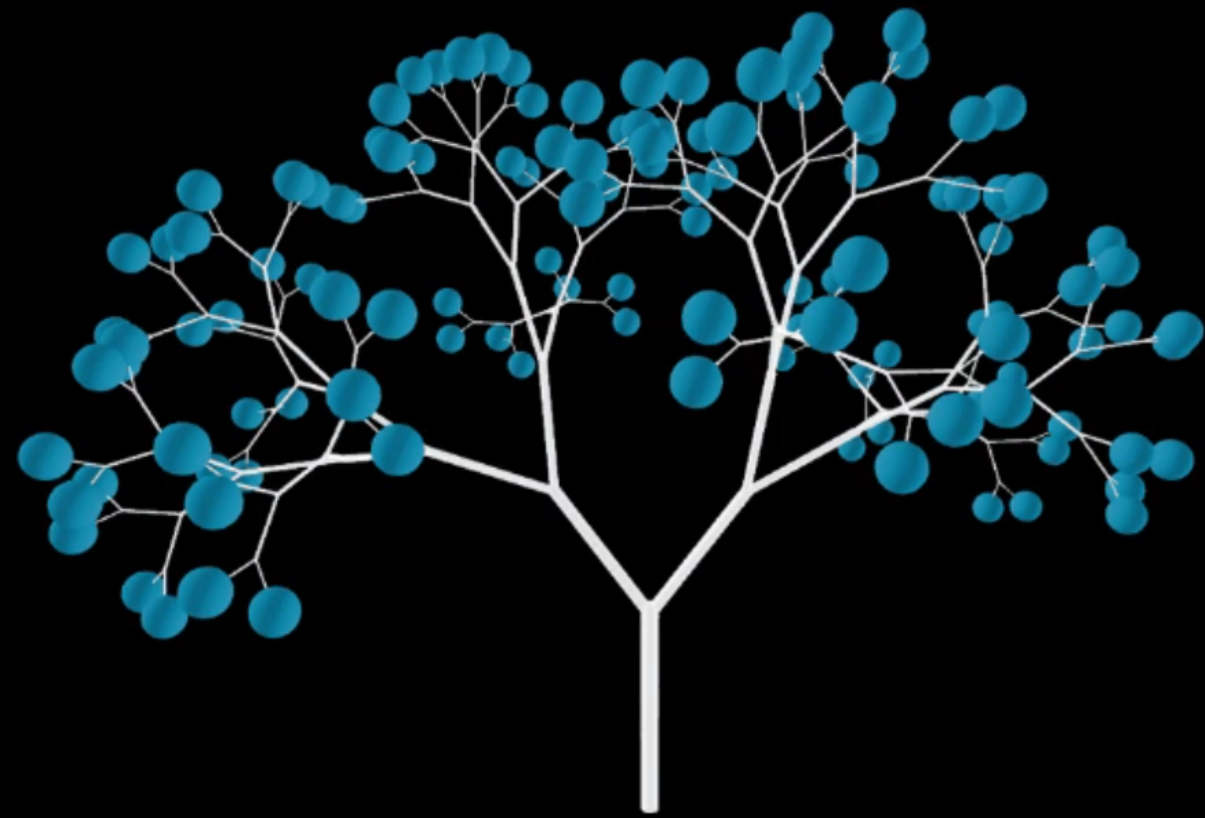
Use of models

- Describe systems
- Explain behaviour
- Experiment and test systems
- Measure sensitivities
- Create analogies
- Educate
- Predict



Mathematical models

Thank you to



COMPLEXITY
EXPLORER
SANTA FE INSTITUTE

Outline

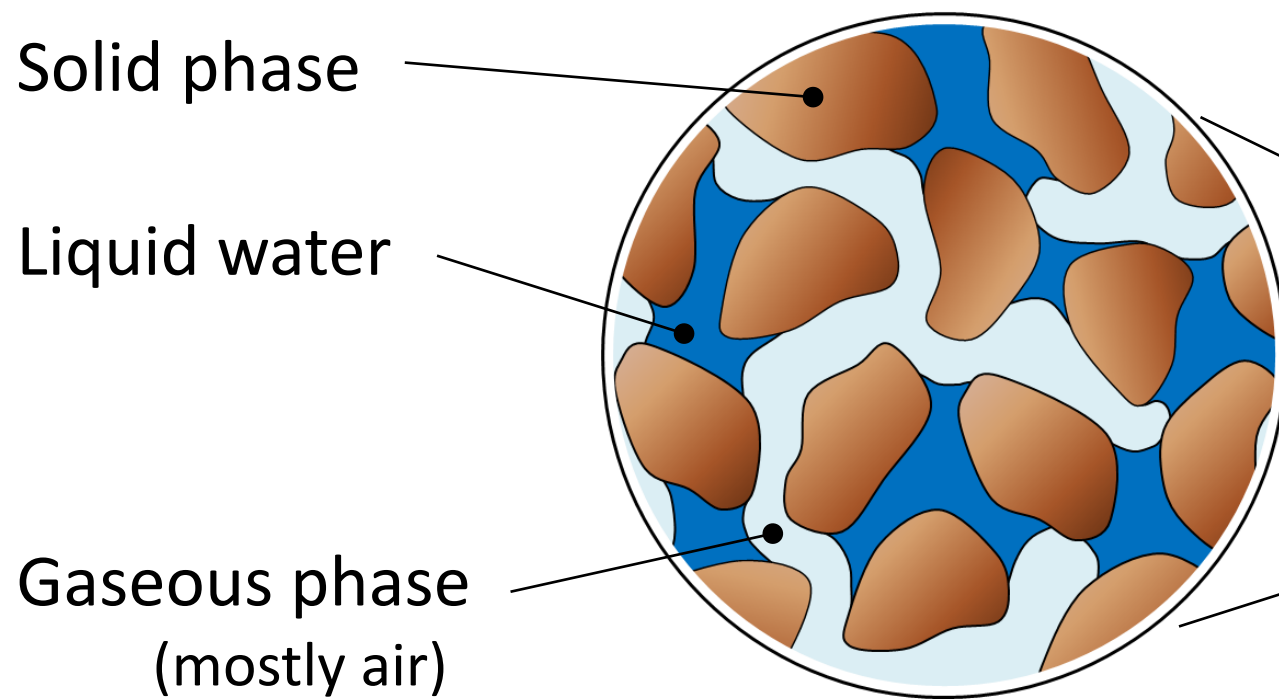
A focus on equation-based models

- How can we control the quality of simulations, verification and validation?
- Why are set-in-stone-models limited?
- How can we leverage statistical models to improve our models?

Porous media models

Microscale (pore scale)

Interfaces identifiable



RVE

Macroscale (Darcy scale)

At each point more phases coexist



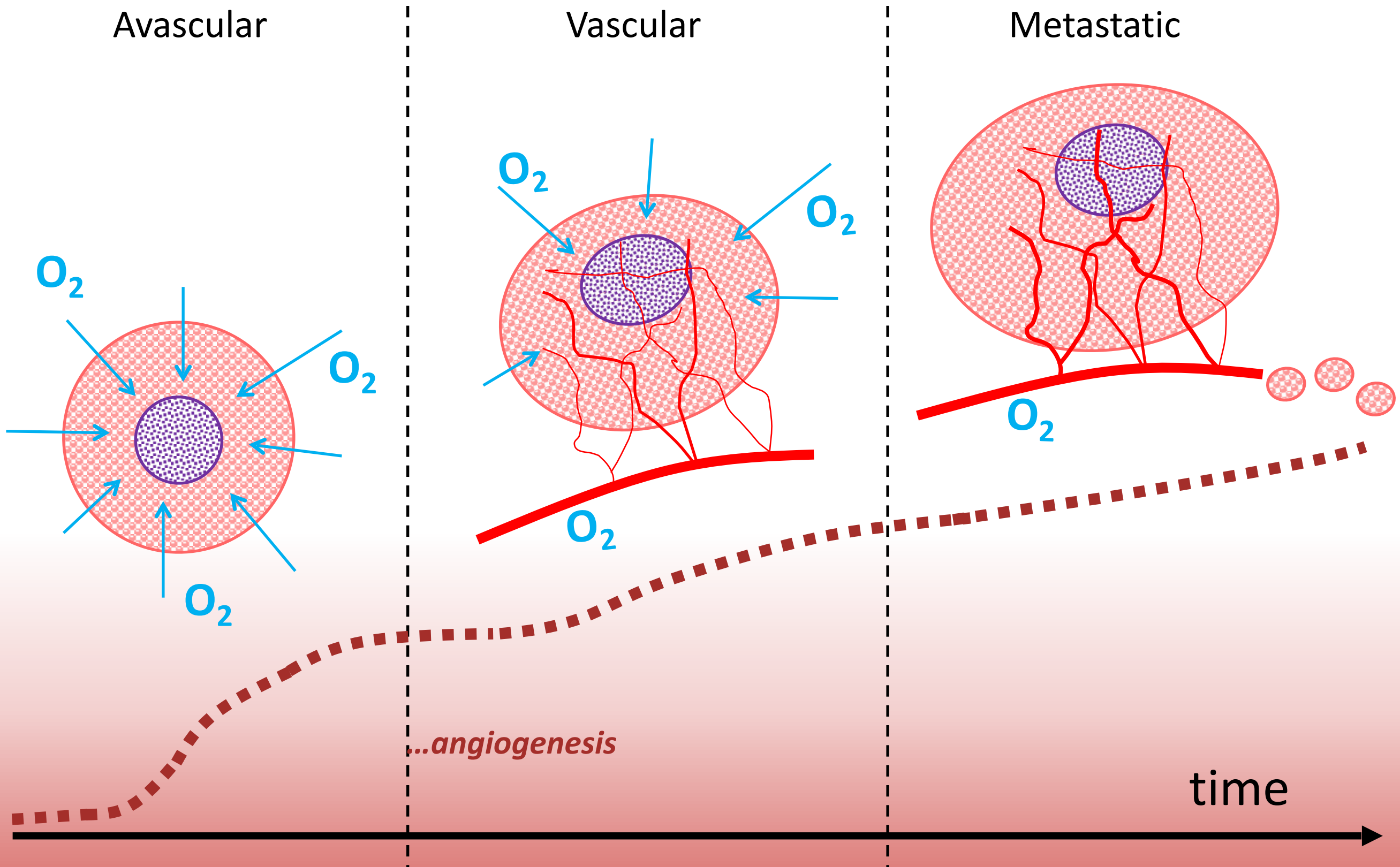
From Ms. Stanton – Science (<http://www.oconee.k12.sc.us/>)

Microscale
Conservation EQS

TCAT
Thermodynamically Constrained
Averaging Theory
(Gray & Miller, 2005)

Macroscale
Conservation EQS

Can also help model tumours





MANY PROBLEMS SHARE THE SAME FORMALISM

Option Pricing



Black, Fischer, and Myron Scholes. "The pricing of options and corporate liabilities." *The journal of political economy* (1973): 637-654.

$$\frac{\partial V}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} - rV = 0$$



MANY PROBLEMS HAVE THE SAME FORMALISM

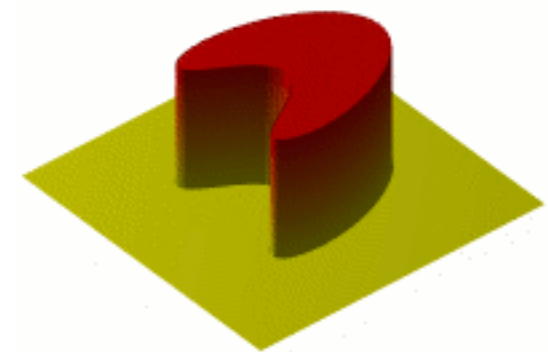
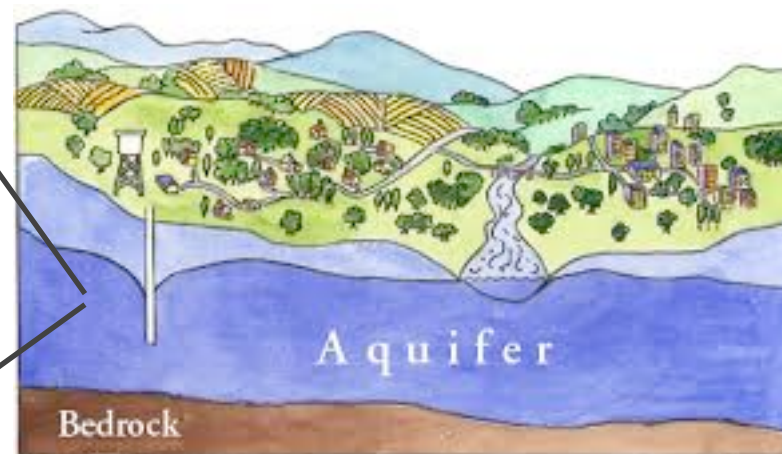
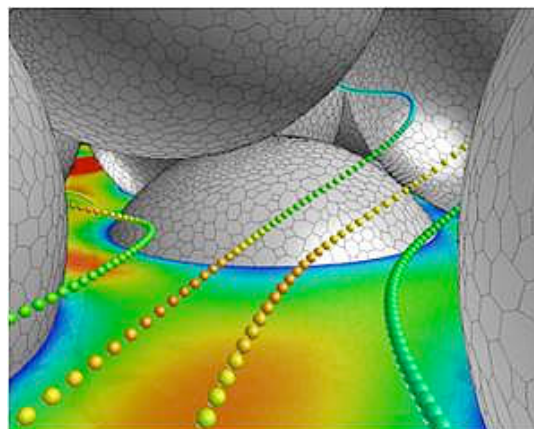
Option Pricing



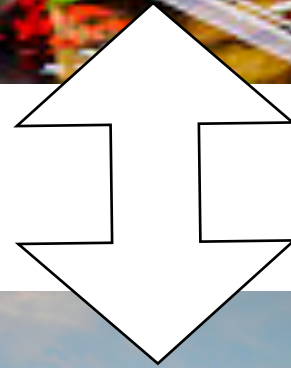
Black, Fischer, and Myron Scholes. "The pricing of options and corporate liabilities." *The journal of political economy* (1973): 637-654.

$$\frac{\partial V}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} - rV = 0$$

Transport in Porous Media



Barenblatt, Grigory Isaakovich. *Scaling, self-similarity, and intermediate asymptotics: dimensional analysis and intermediate asymptotics*. Vol. 14. Cambridge University Press, 1978.



Mathematical Modelling

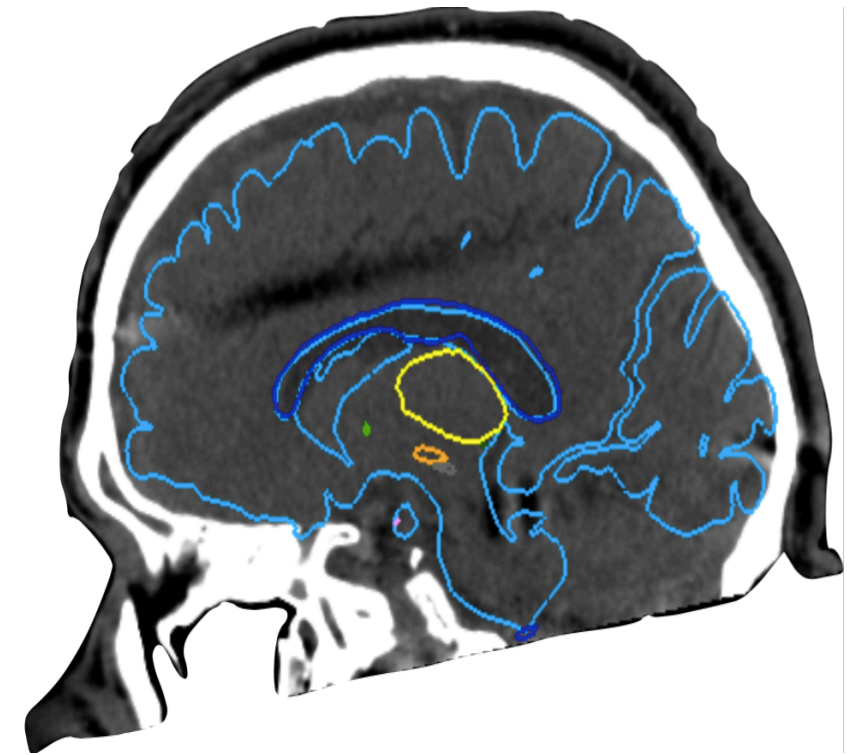
Continuous
Problem

Mathematical Modelling

Continuous
Problem

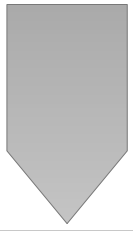


Bijar, Rohan, Perrier &
Payan 2015



Mathematical Modelling

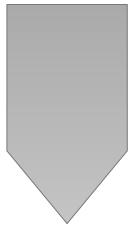
Continuous
Problem



Mathematical
Model

Mathematical Modelling

Continuous
Problem



Mathematical
Model

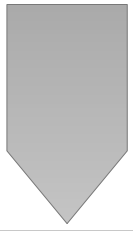
$$\min_{\mathbf{u} \in \mathbf{V}} \frac{1}{2} \int_{\Omega} \boldsymbol{\sigma}(\mathbf{u}, \beta) : \boldsymbol{\varepsilon}(\mathbf{u}) \, d\mathbf{x} - \int_{\Omega} \mathbf{g} \cdot \mathbf{u} \, d\mathbf{x}$$

with

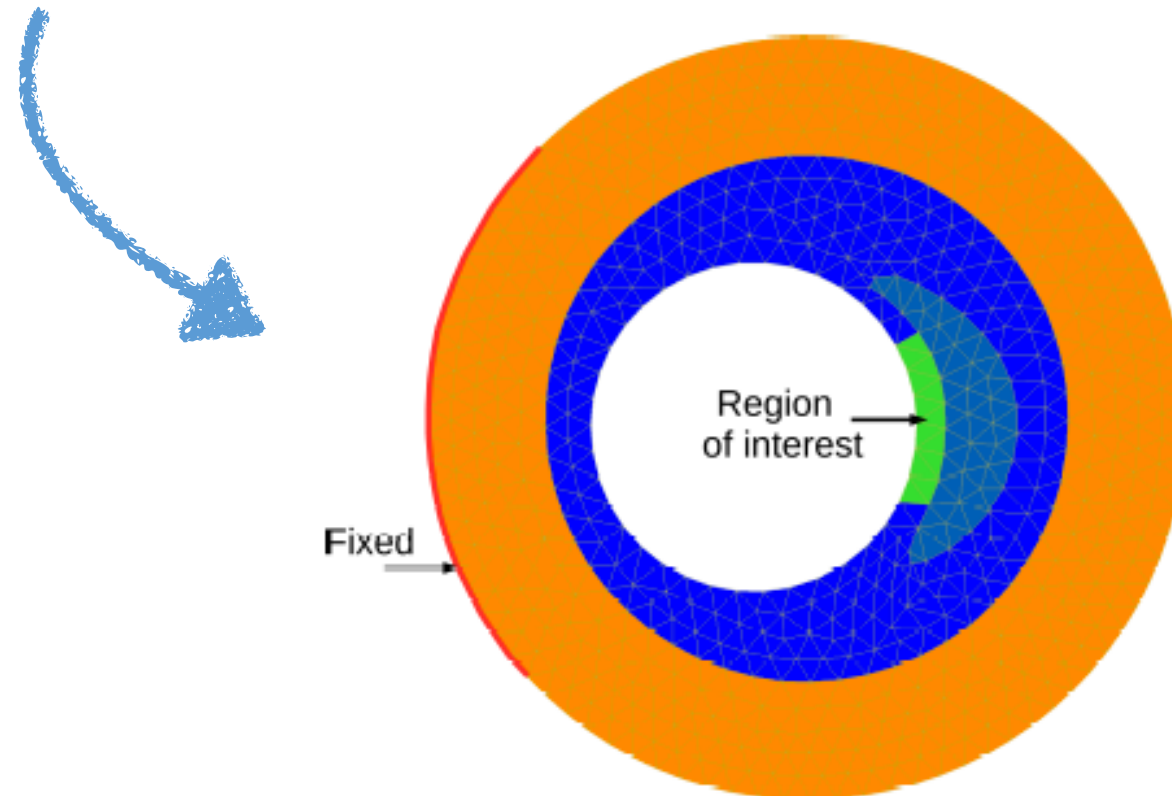
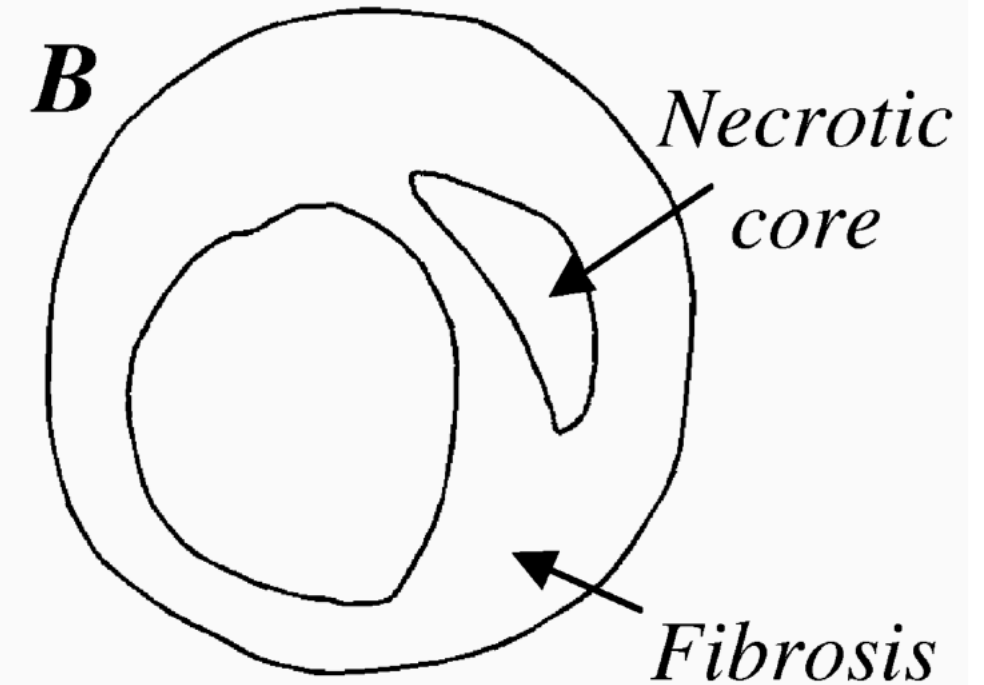
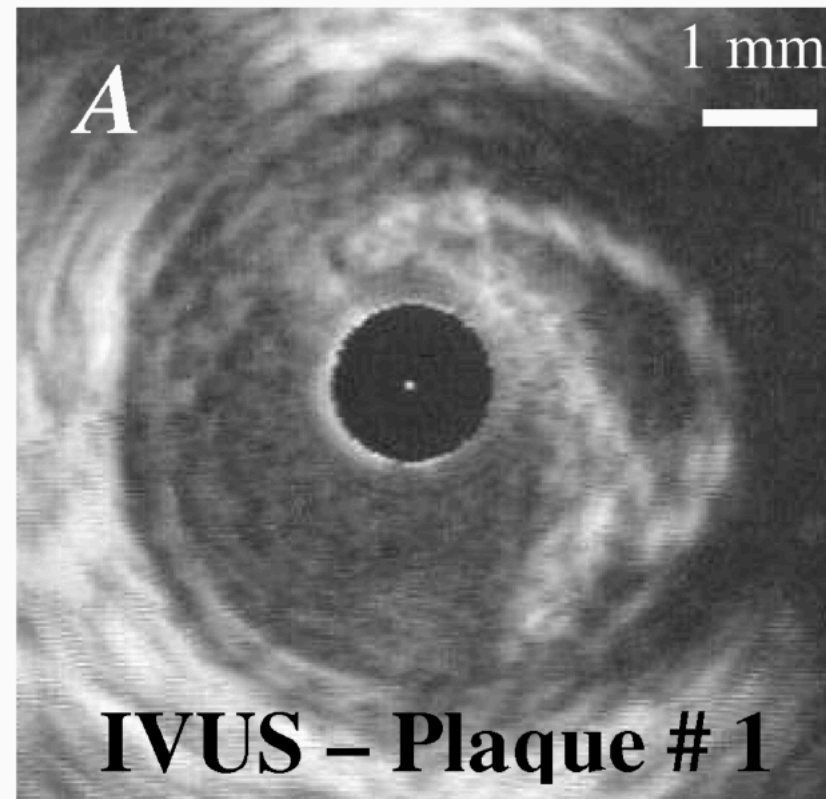
$$\boldsymbol{\sigma}(\mathbf{u}, \beta) = \underbrace{\boldsymbol{\sigma}_P(\mathbf{u})}_{\substack{\text{passive} \\ \text{material}}} + \underbrace{\boldsymbol{\sigma}_A(\beta)}_{\substack{\text{muscular} \\ \text{activation}}} \left\{ \begin{array}{l} \boldsymbol{\sigma}_A(\beta) = \beta T e_A \otimes e_A \\ e_A : \text{fiber direction} \\ T : \text{tension} \\ \beta : \text{activation} \end{array} \right.$$

Mathematical Modelling

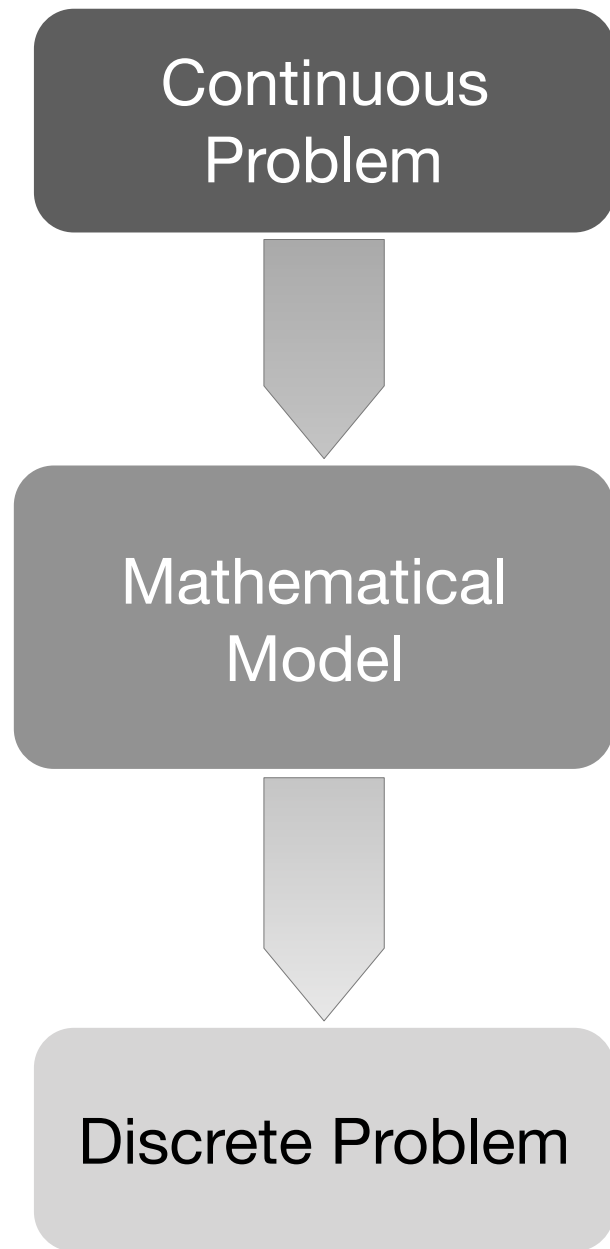
Continuous Problem



Mathematical Model

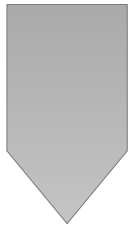


Mathematical Modelling



Mathematical Modelling

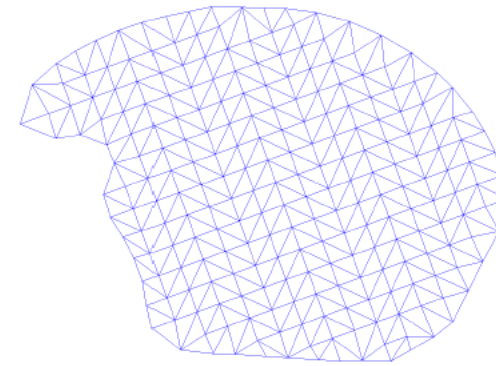
Continuous Problem



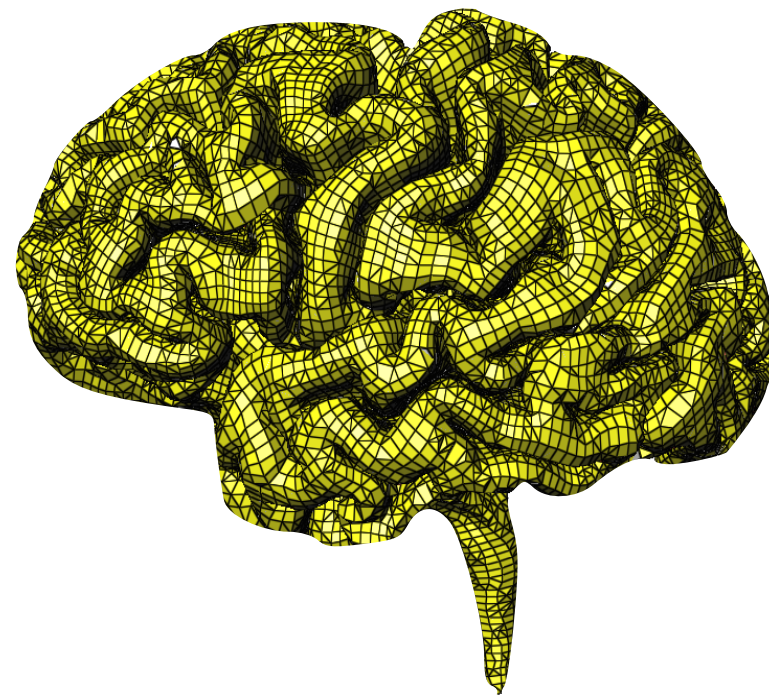
Mathematical Model



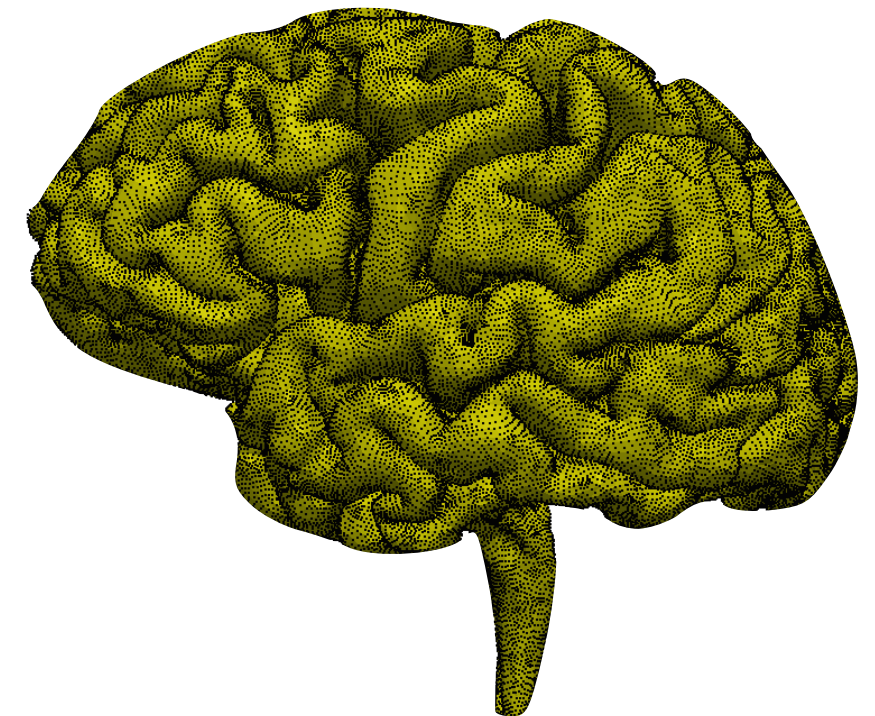
Discrete Problem



Finite element mesh of a tongue with F. Chouly et al.

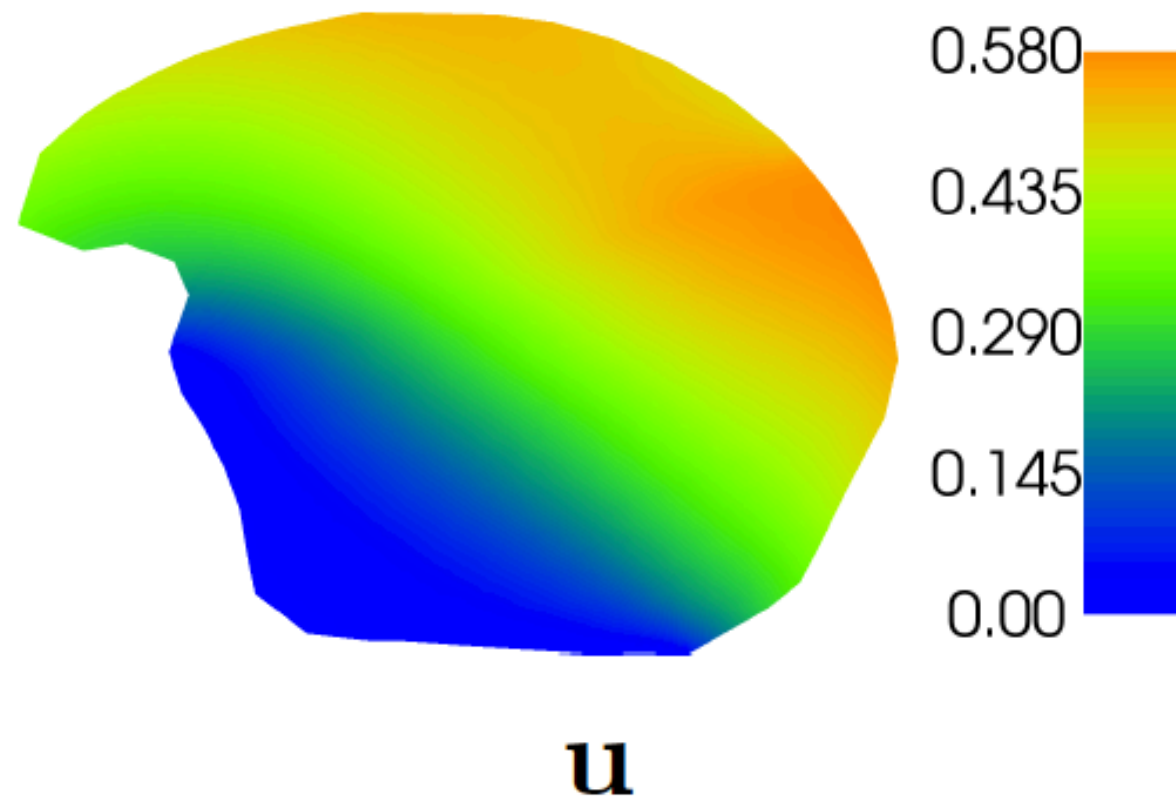
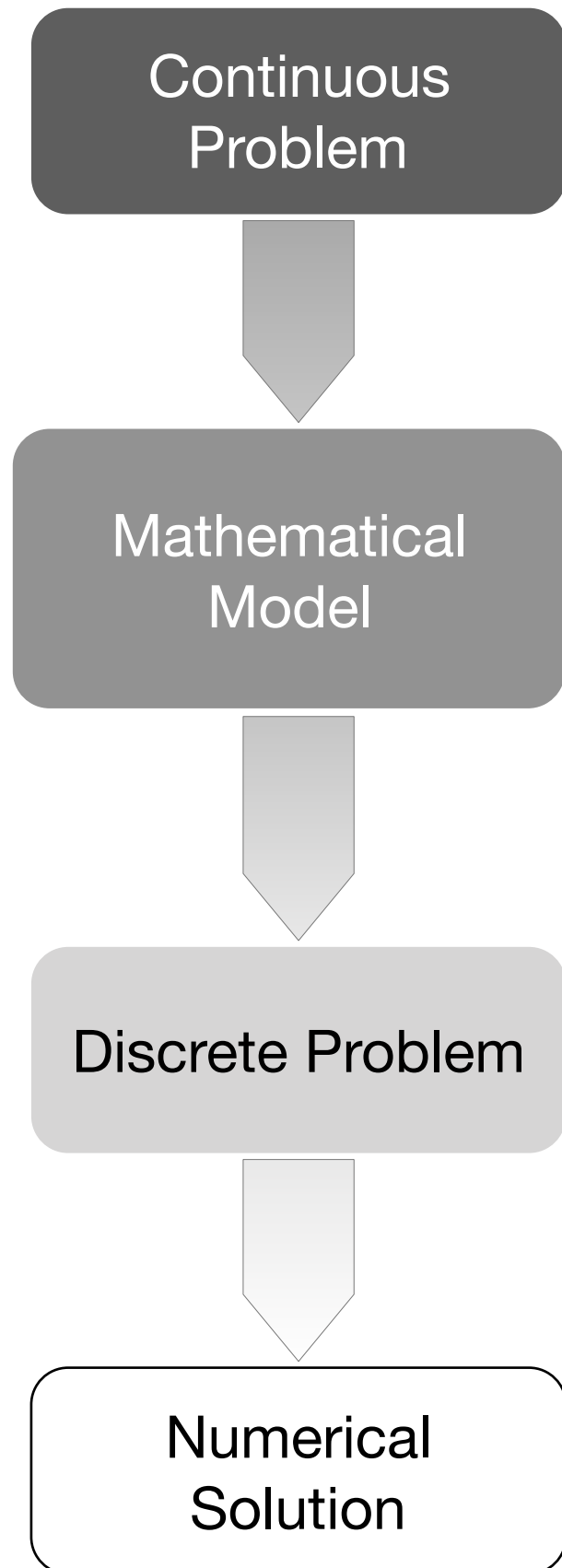


Hexahedral mesh of a brain with Bruno Lévy, Inria

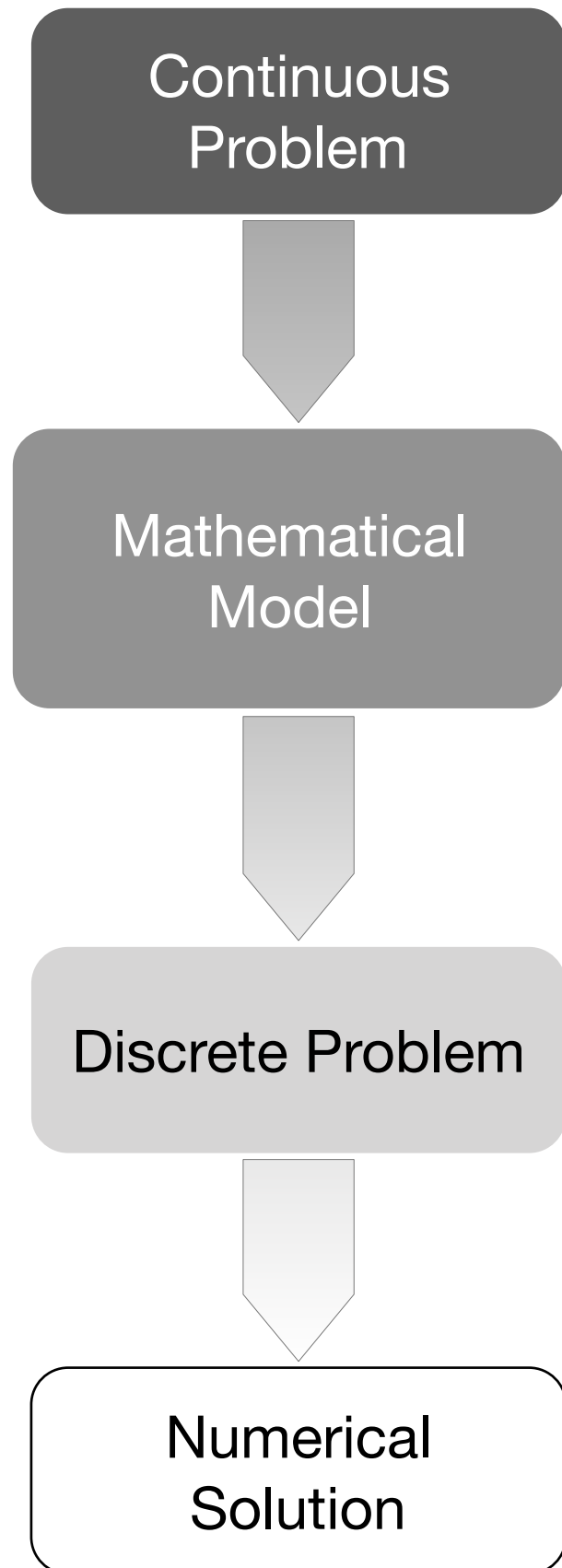


Meshless brain discretization with Bruno Lévy, Inria

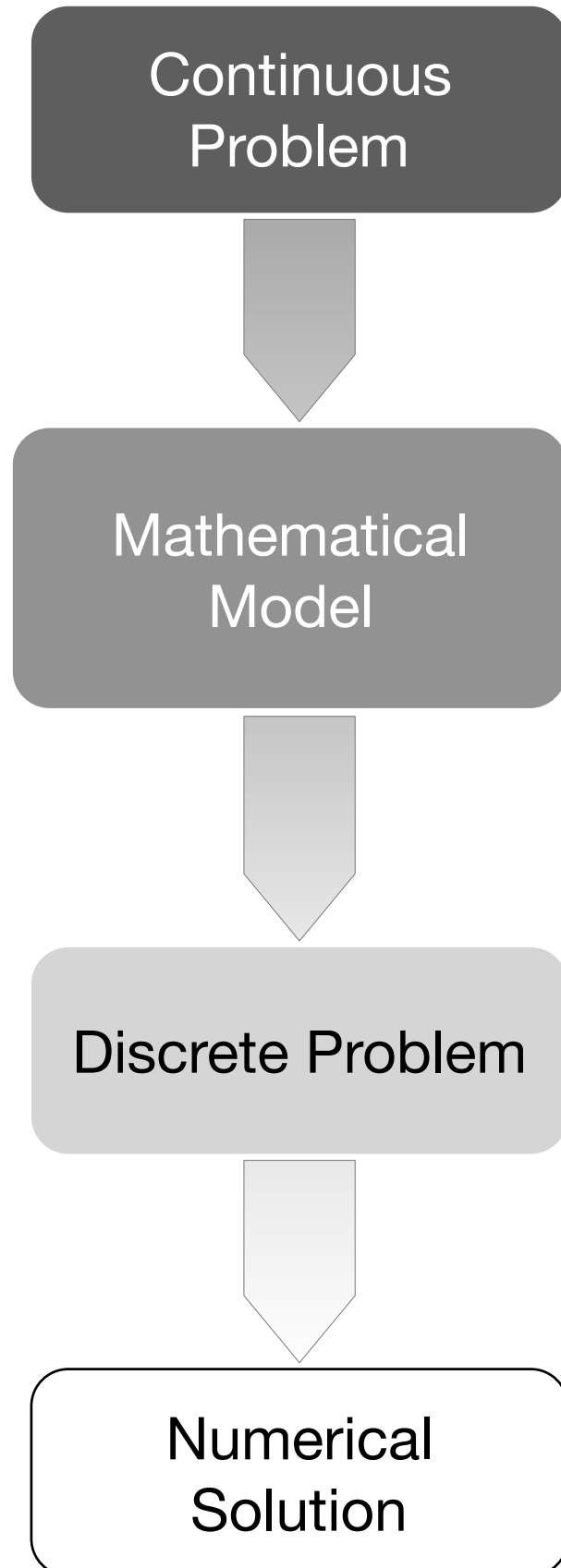
Mathematical Modelling



Mathematical Modelling



Mathematical Modelling

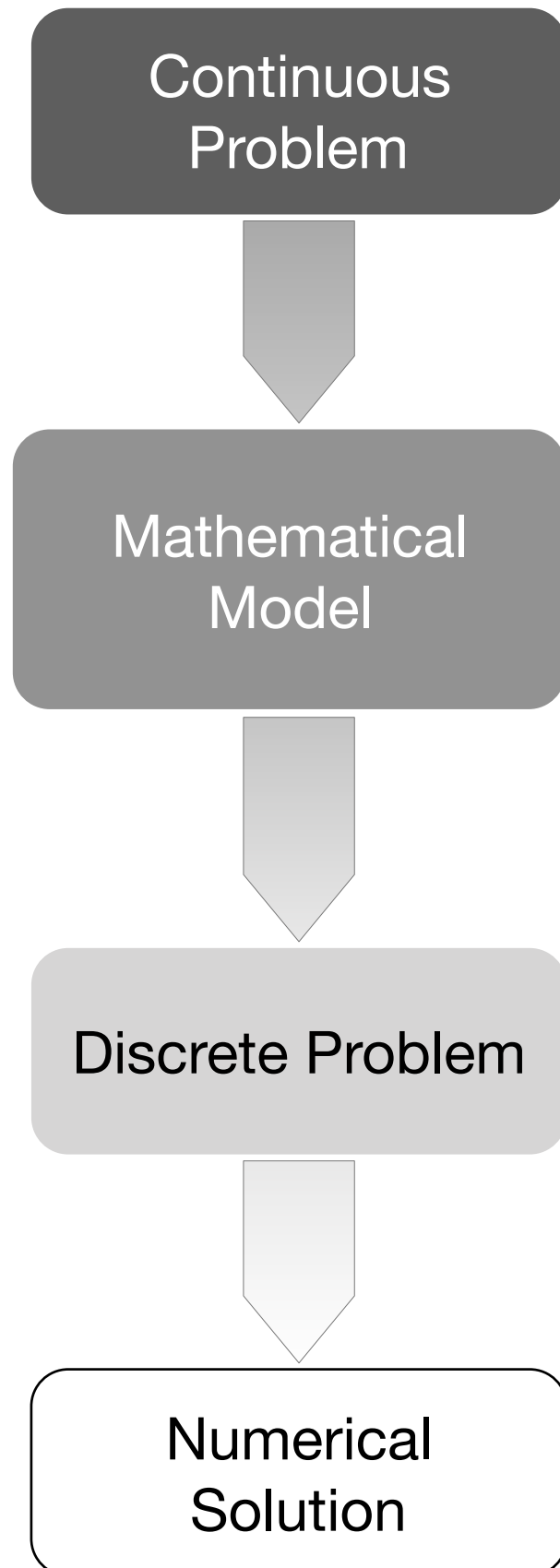


Bijar, Rohan, Perrier & Payan 2015

\neq

$$\min_{\mathbf{u} \in \mathbf{V}} \frac{1}{2} \int_{\Omega} \boldsymbol{\sigma}(\mathbf{u}, \beta) : \boldsymbol{\varepsilon}(\mathbf{u}) \, d\mathbf{x} - \int_{\Omega} \mathbf{g} \cdot \mathbf{u} \, d\mathbf{x}$$

Mathematical Modelling



Model Error



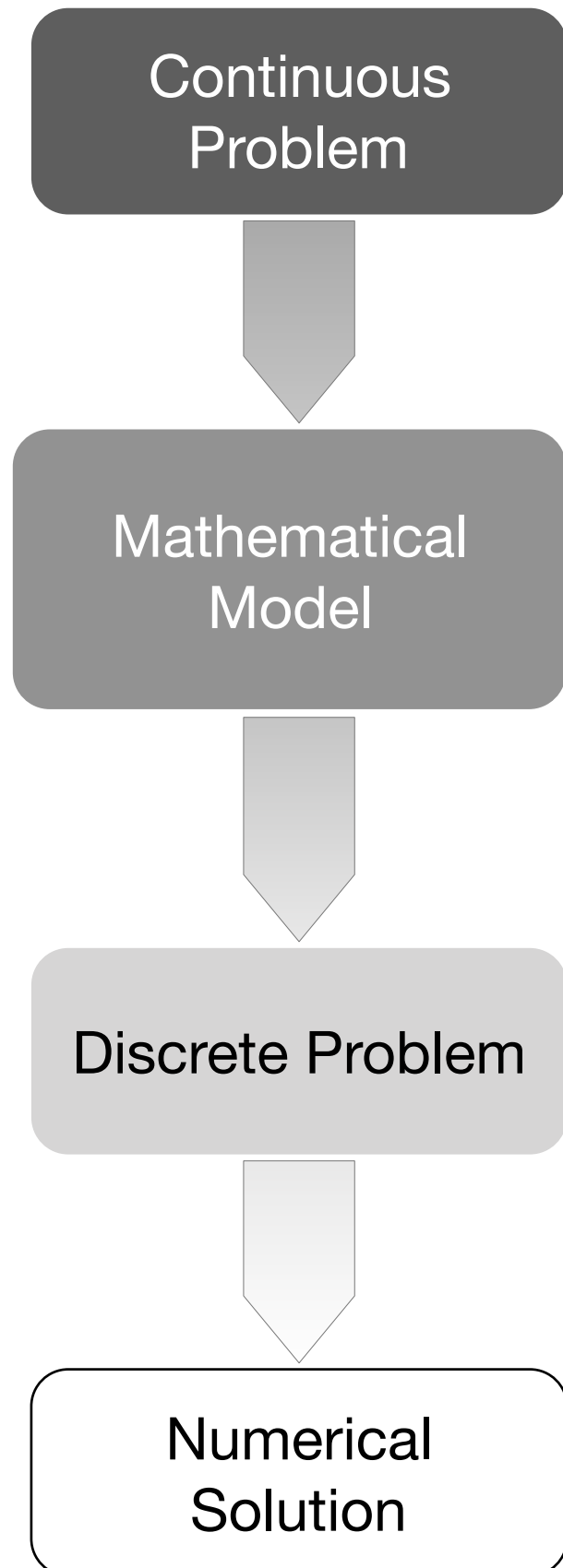
Bijar, Rohan, Perrier & Payan 2015

\neq

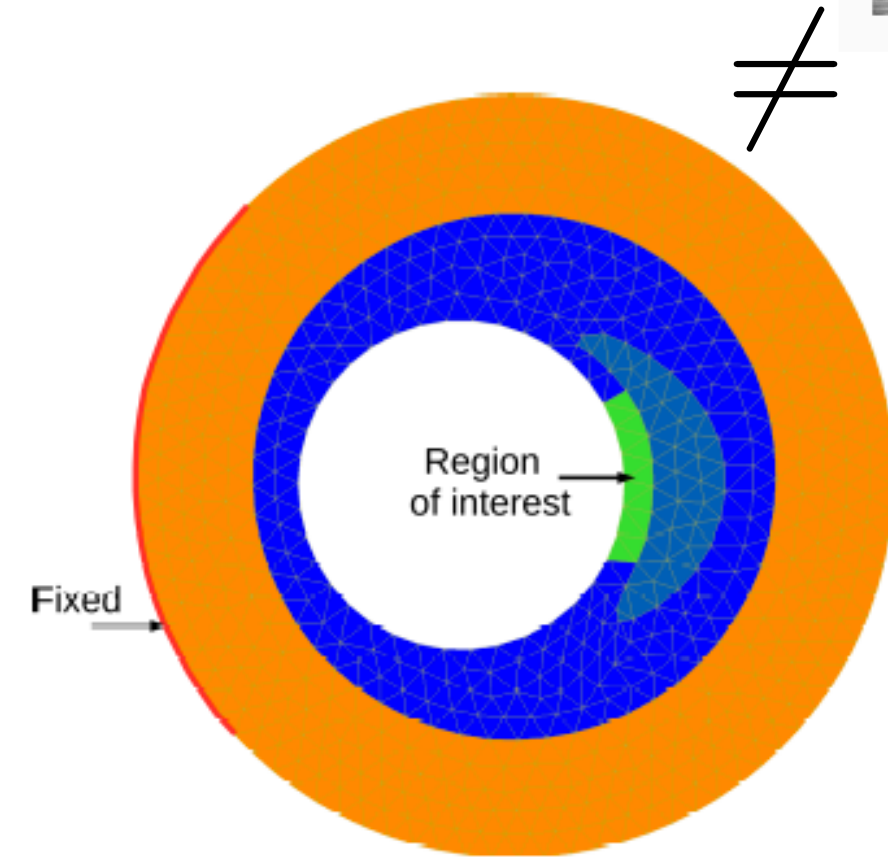
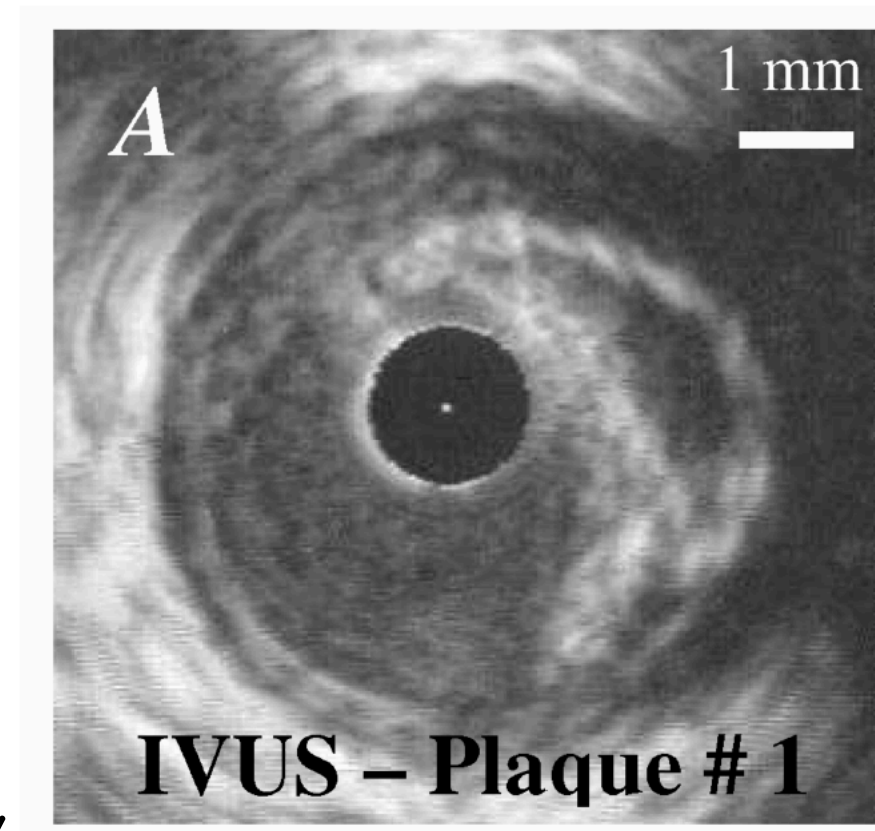
$$\min_{\mathbf{u} \in \mathbf{V}} \frac{1}{2} \int_{\Omega} \boldsymbol{\sigma}(\mathbf{u}, \boldsymbol{\beta}) : \boldsymbol{\varepsilon}(\mathbf{u}) \, d\mathbf{x} - \int_{\Omega} \mathbf{g} \cdot \mathbf{u} \, d\mathbf{x}$$

Physical Problem
Constitutive Model
Material Parameters

Mathematical Modelling

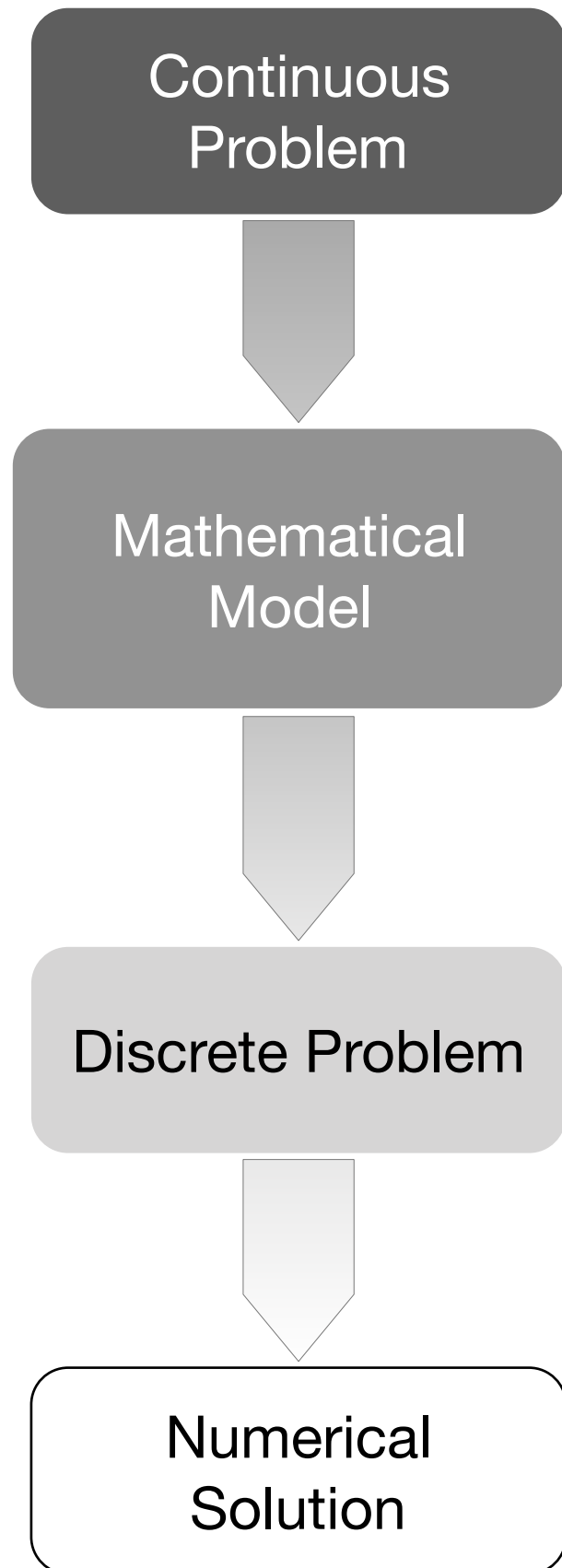


Model Error



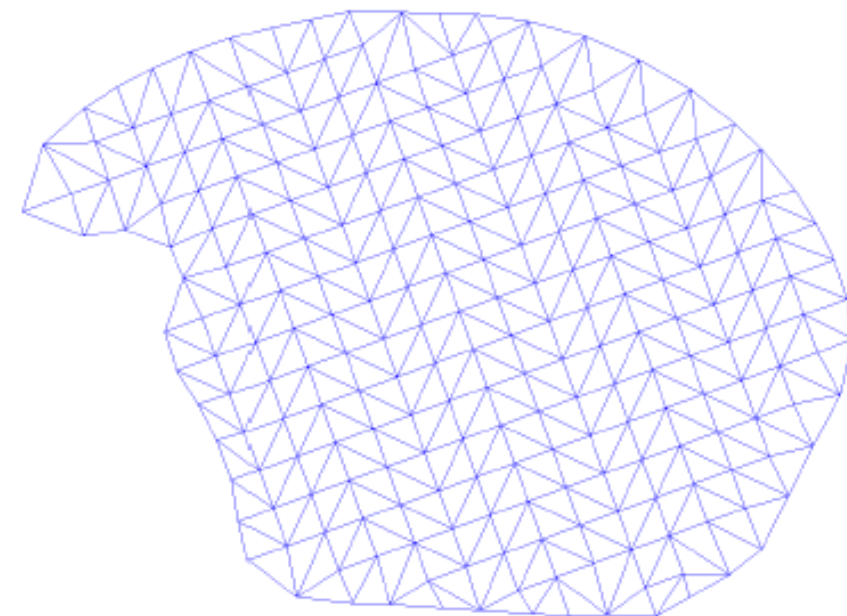
Geometry
Boundary conditions

Mathematical Modelling

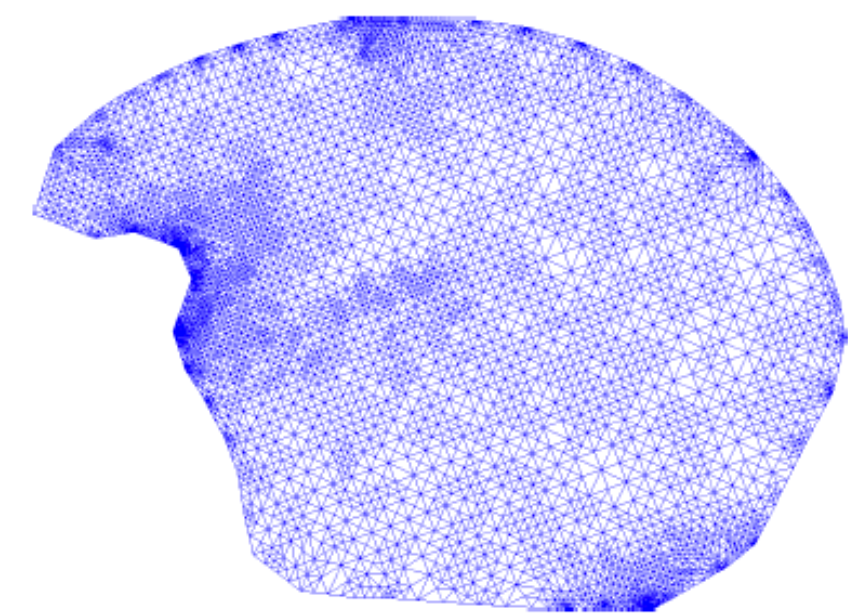


Model Error

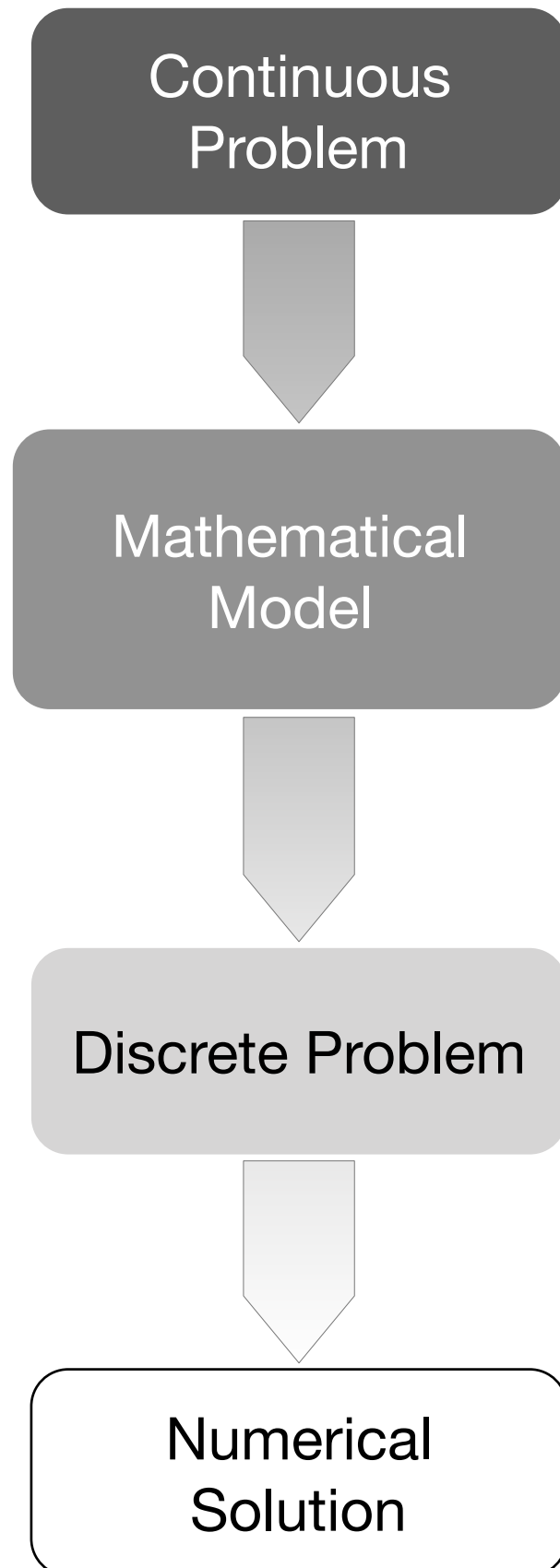
Discretization Error



vs.

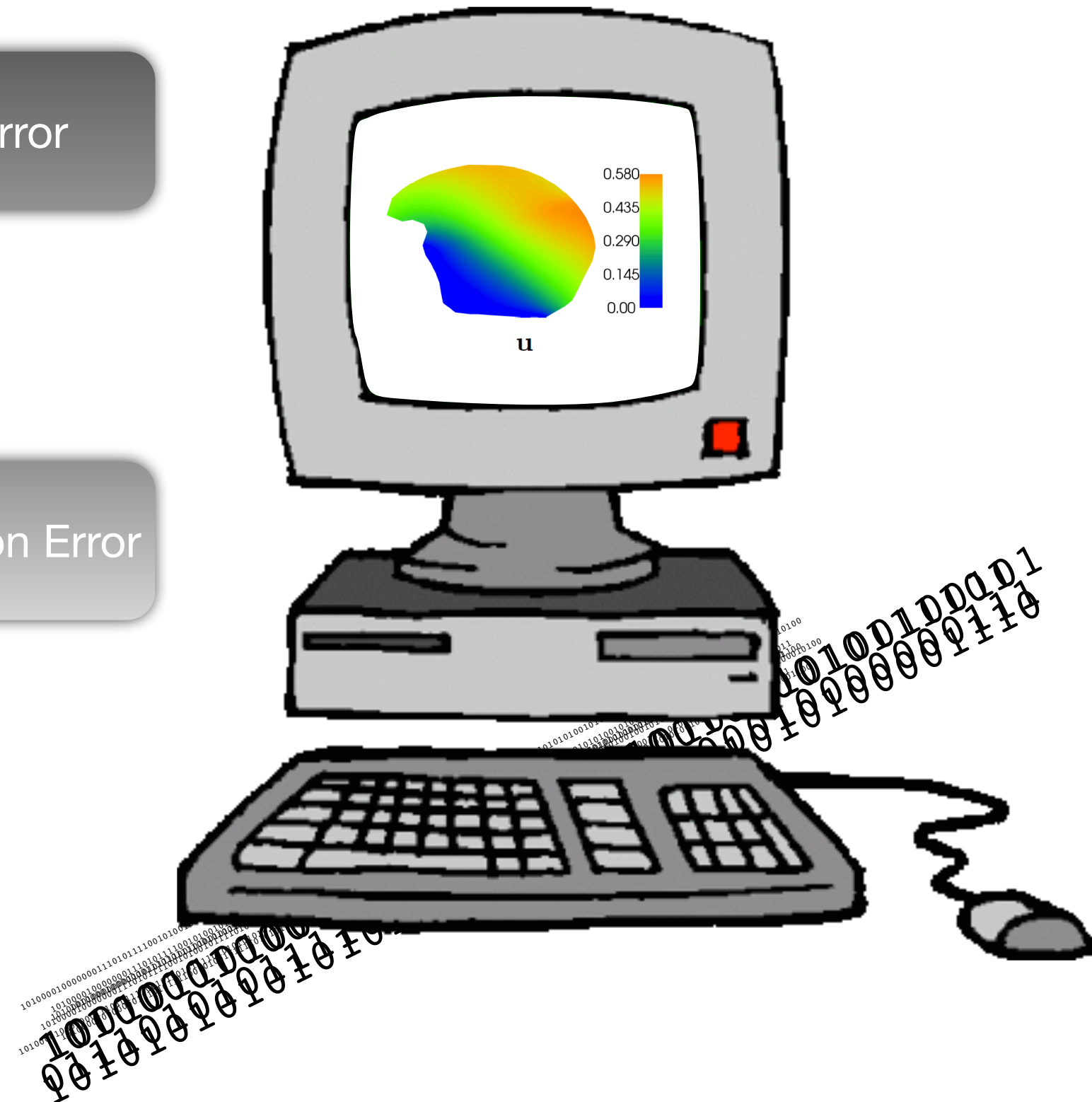


Mathematical Modelling

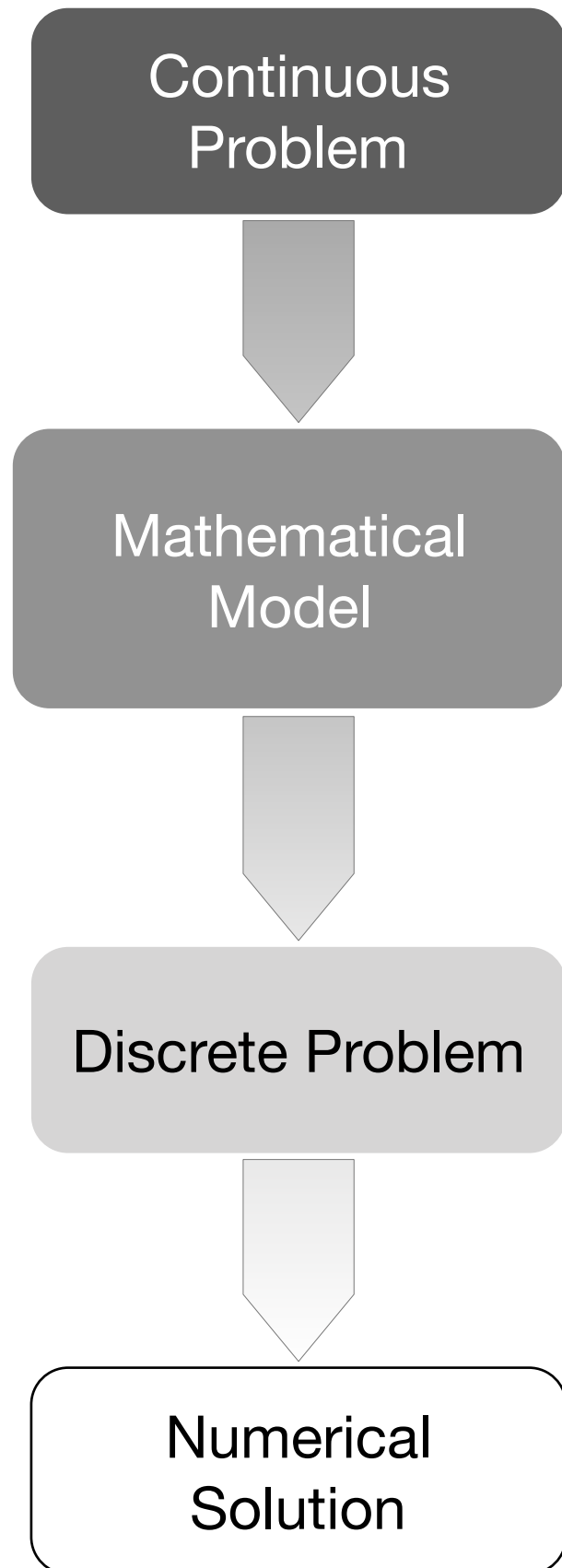


Model Error

Discretization Error



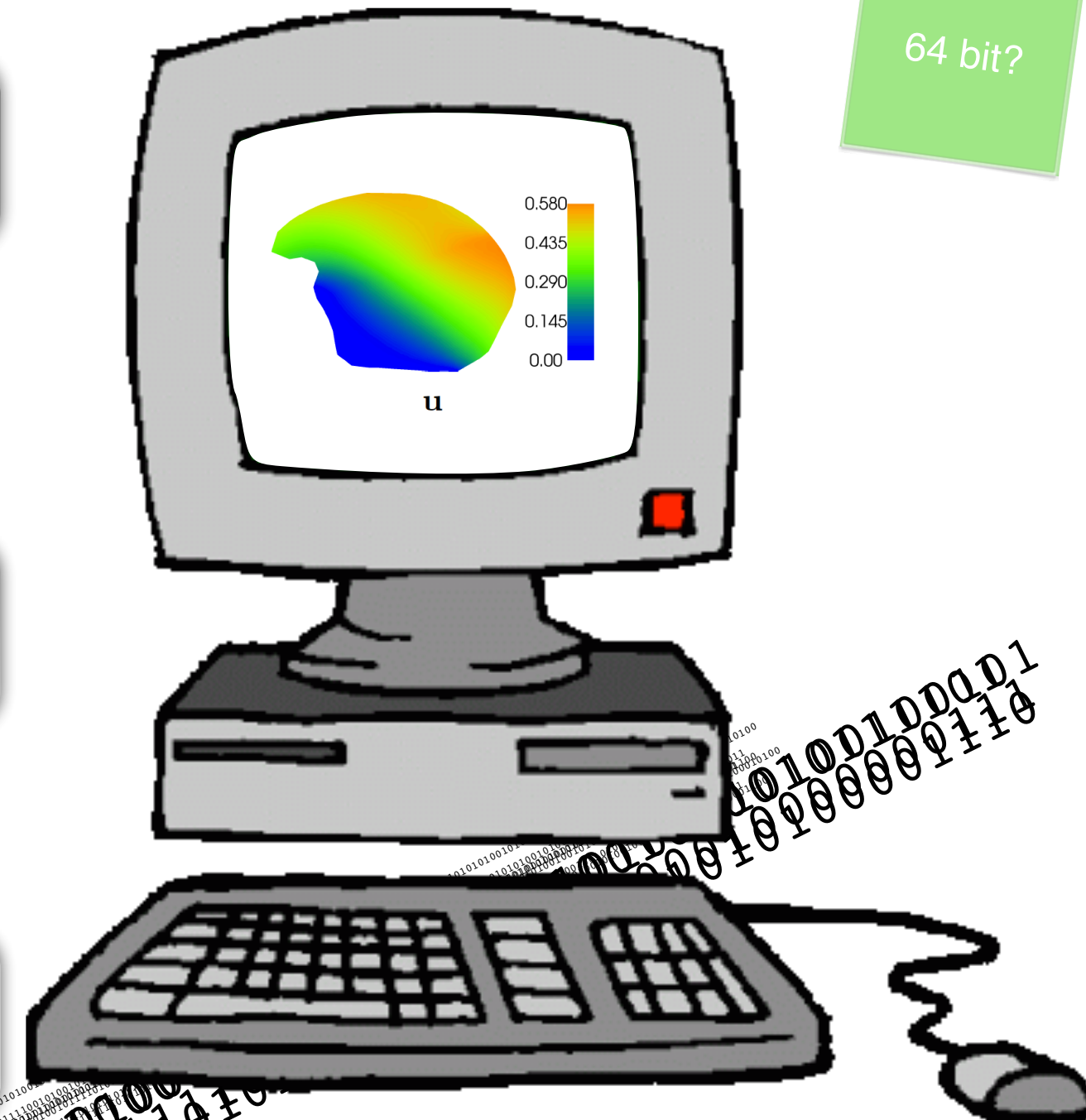
Mathematical Modelling



Model Error

Discretization Error

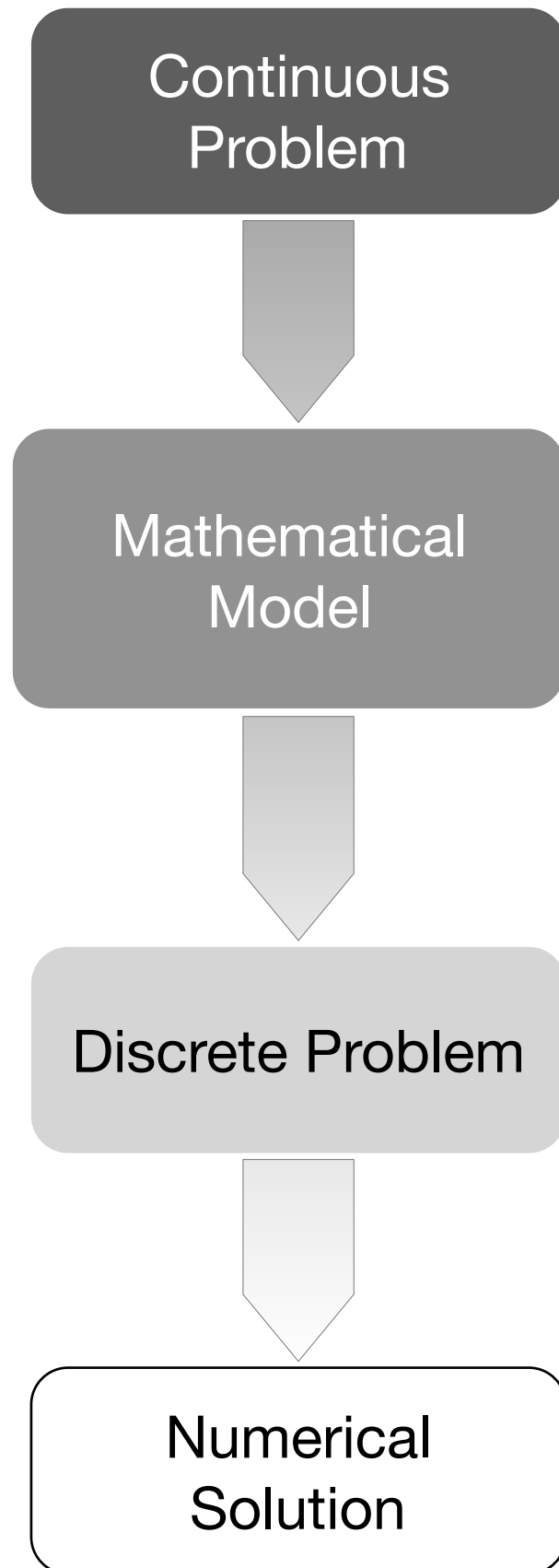
Numerical Error



32 bit?

64 bit?

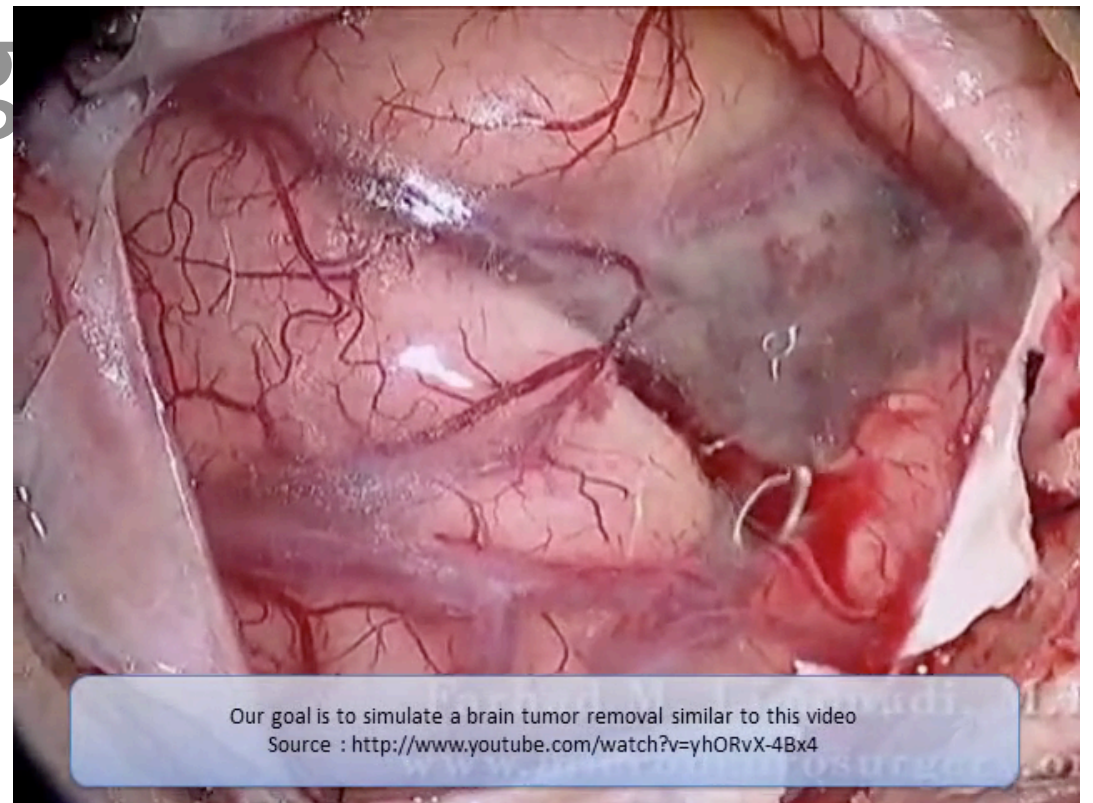
Mathematical Modelling



Model Error

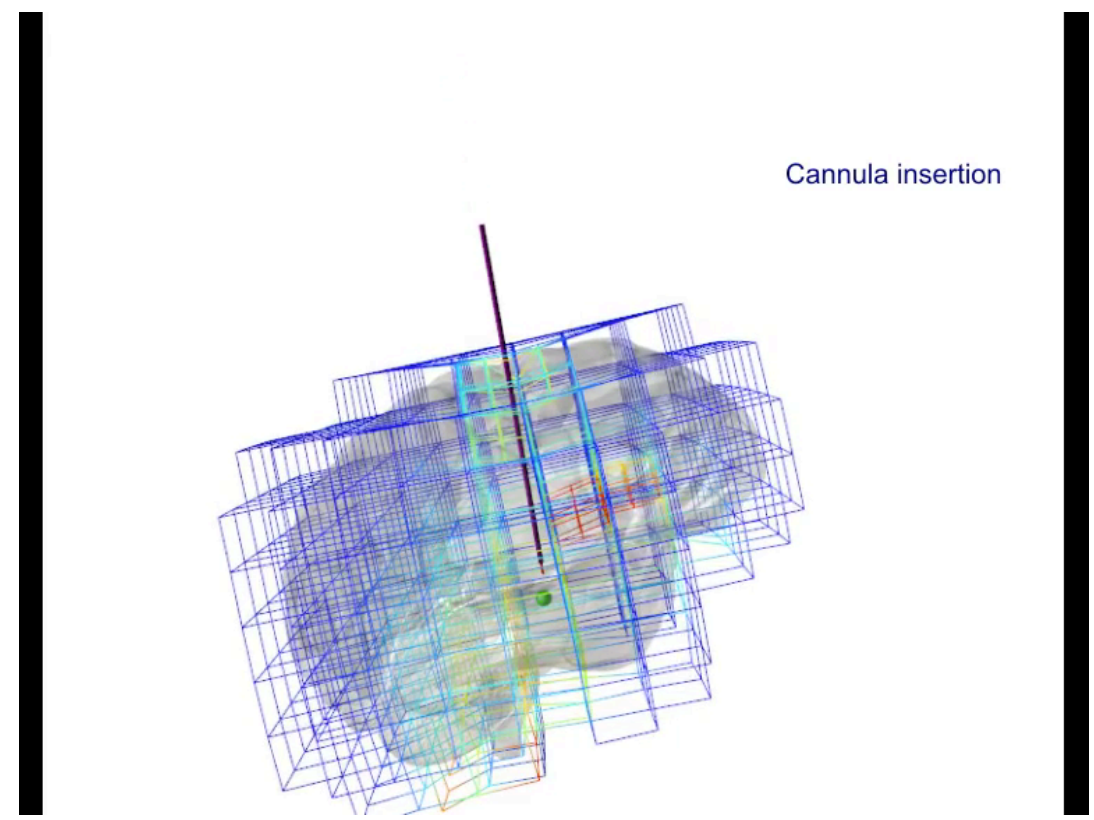
Discretization Error

Numerical Error

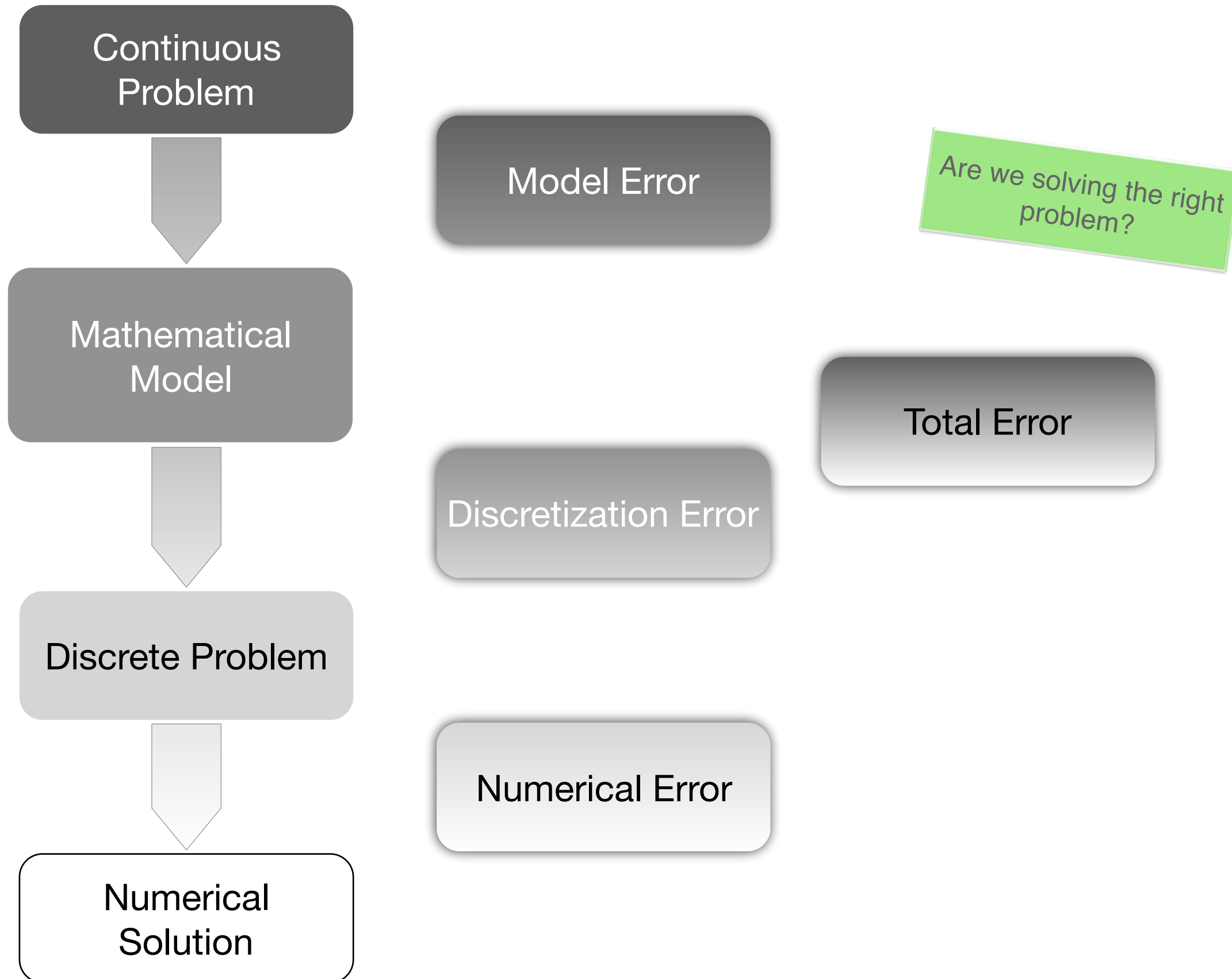


Reality
vs.
Simulation

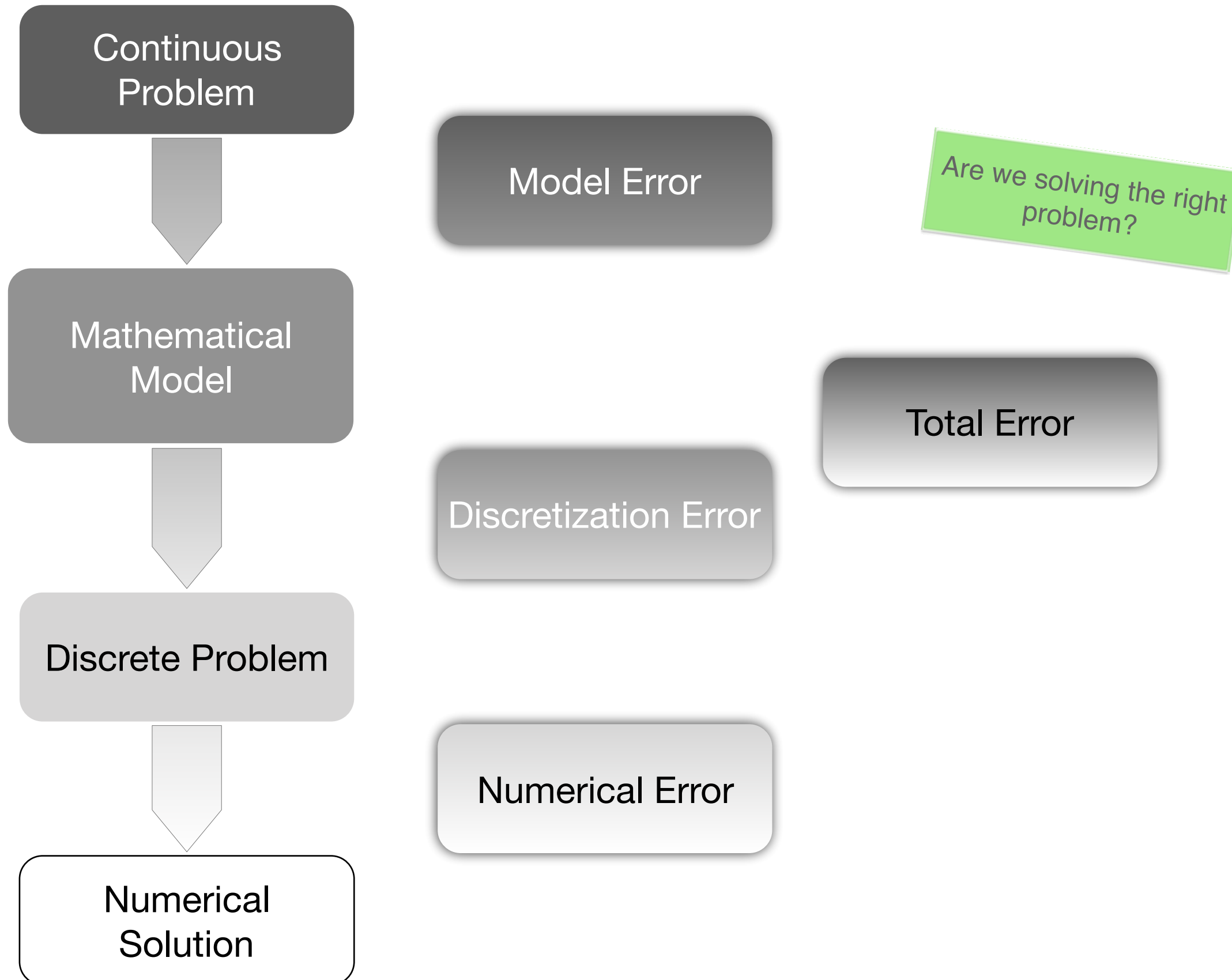
Total Error



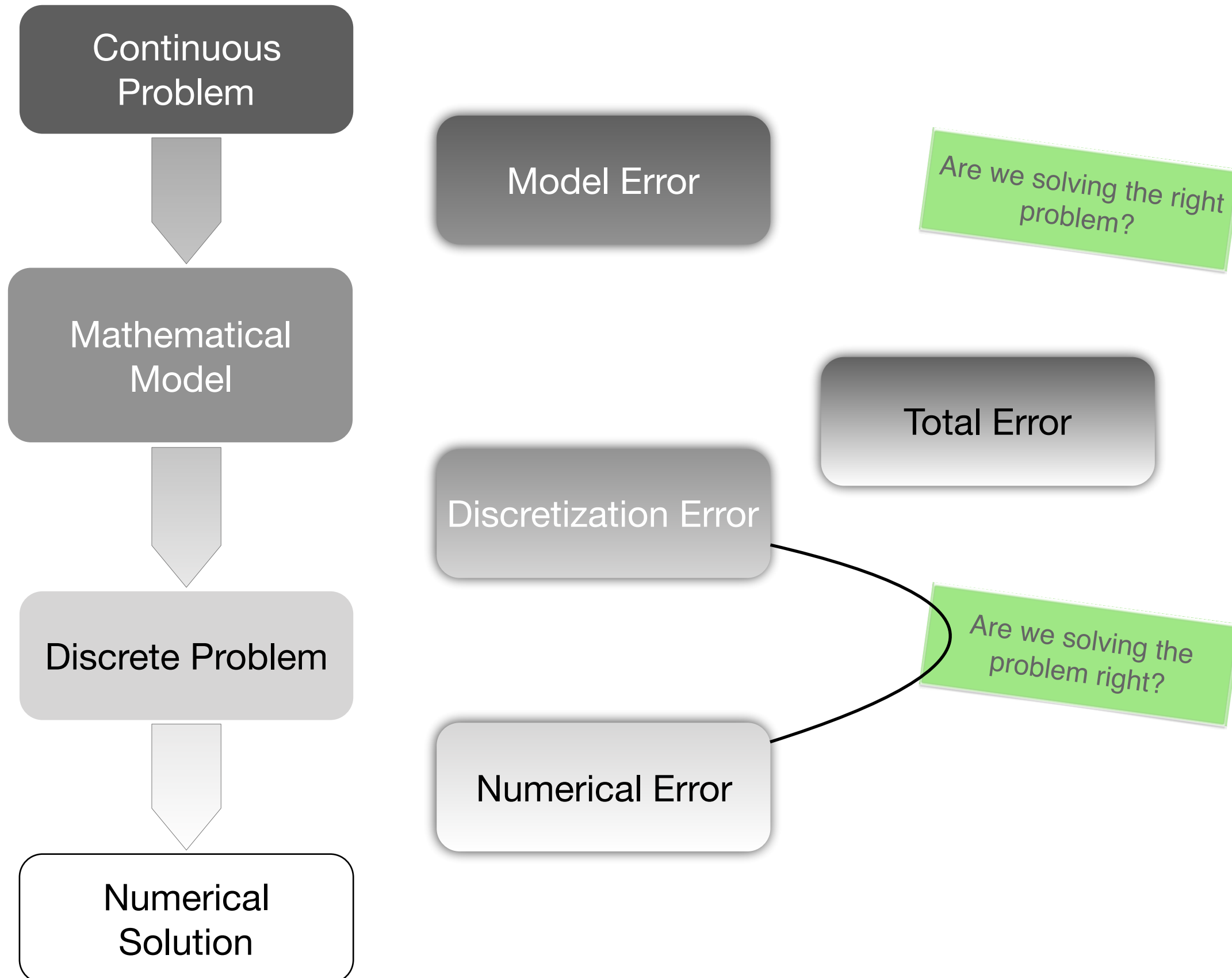
Mathematical Modelling



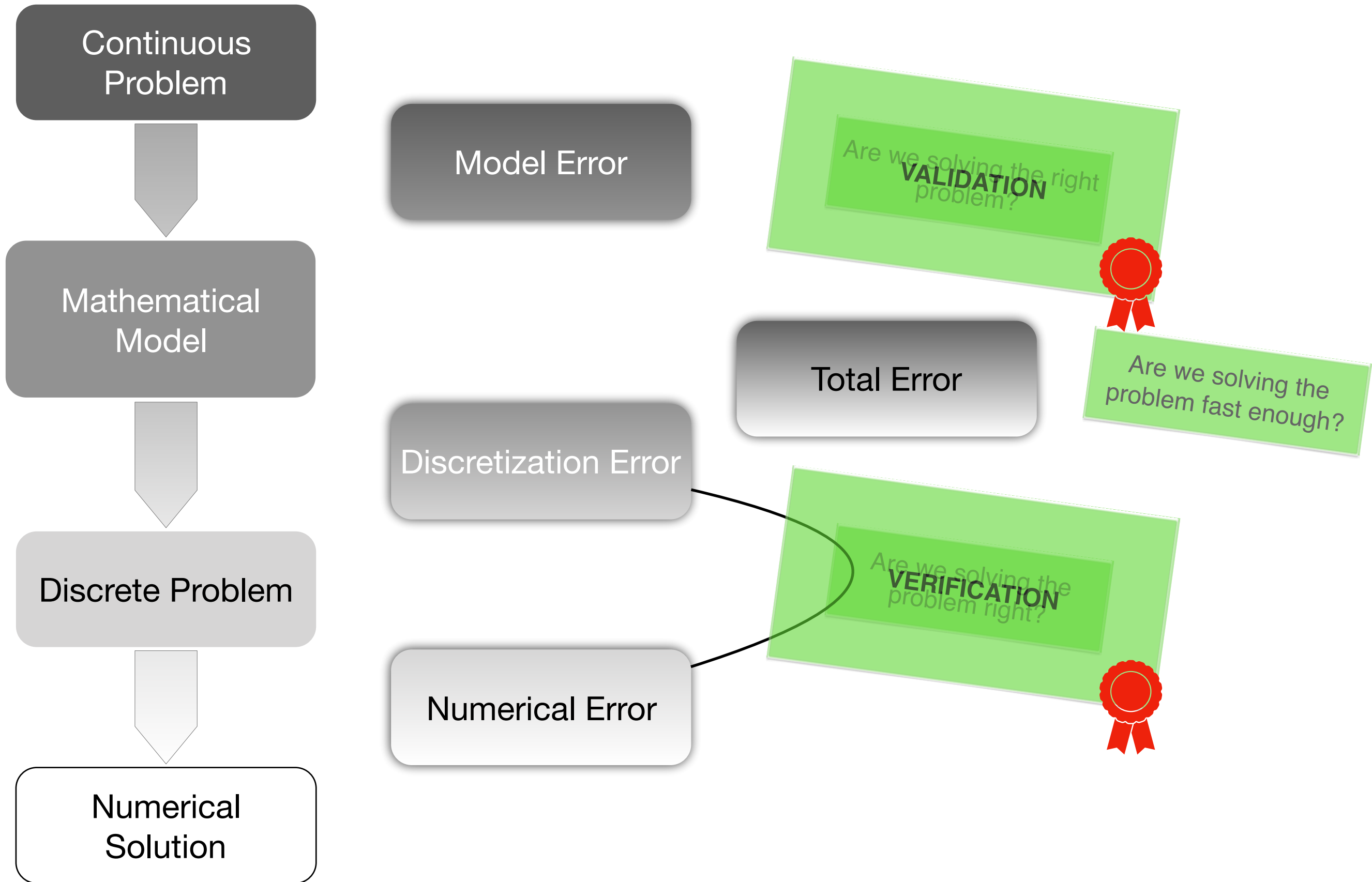
Mathematical Modelling



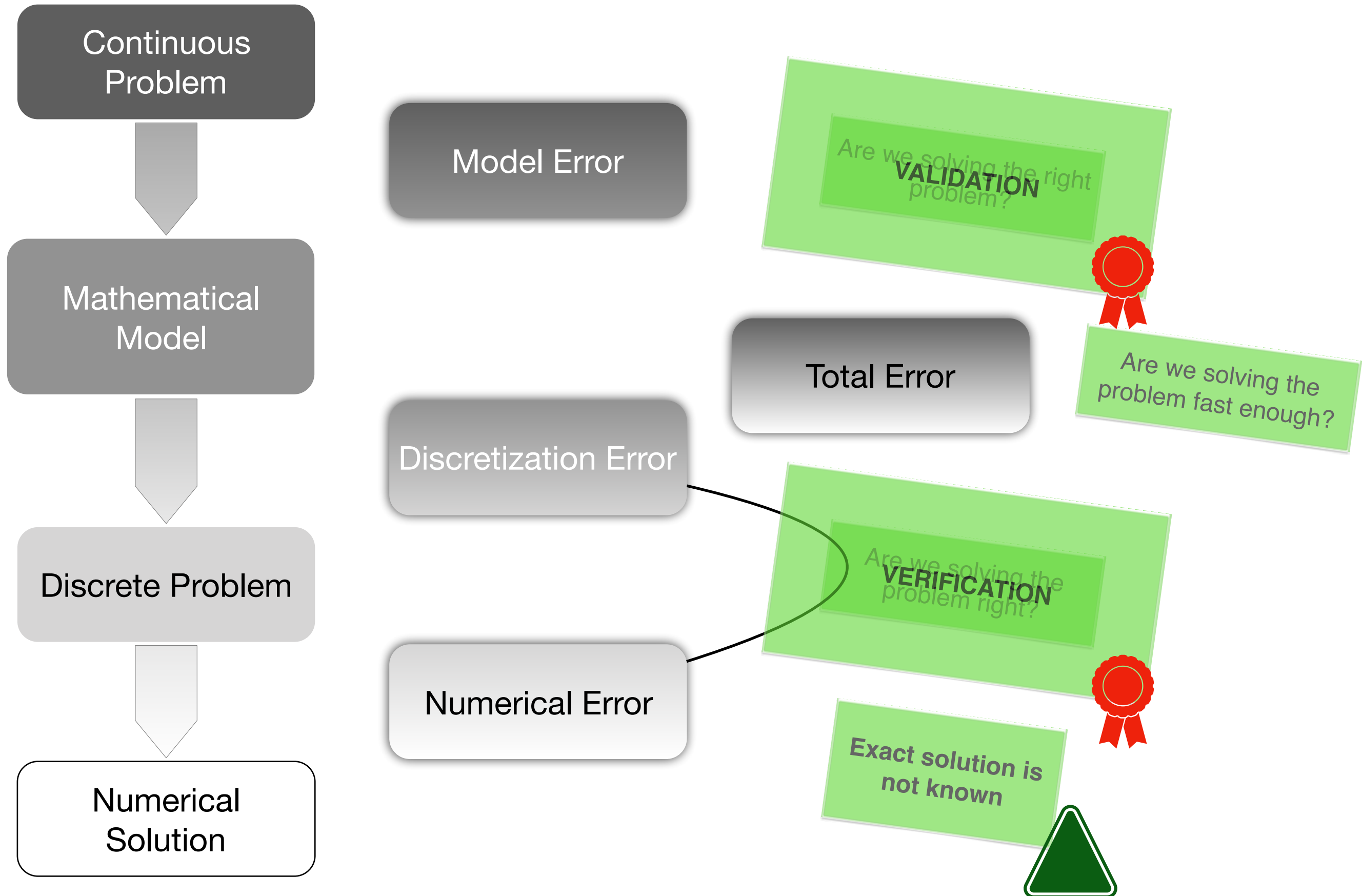
Mathematical Modelling



Mathematical Modelling



Mathematical Modelling



Outline

A focus on equation-based models

- How can we control the quality of simulations, verification and validation?
- Why are set-in-stone-models limited?
- How can we leverage statistical models to improve our models?

Outline

- **Data-driven modelling: Beyond setting models in stone**

- Data assimilation

- How can we learn from observations “on-the-fly”.

- The power of digital twins.

- **Future challenges**

Introduction to data assimilation

- Bayesian inference
- Kalman filtering

Model and parameter
identification through Bayesian
inference in solid mechanics

Hussein Rappel

h.rappel@gmail.com

September 07, 2018



Bayesian inference

Primer



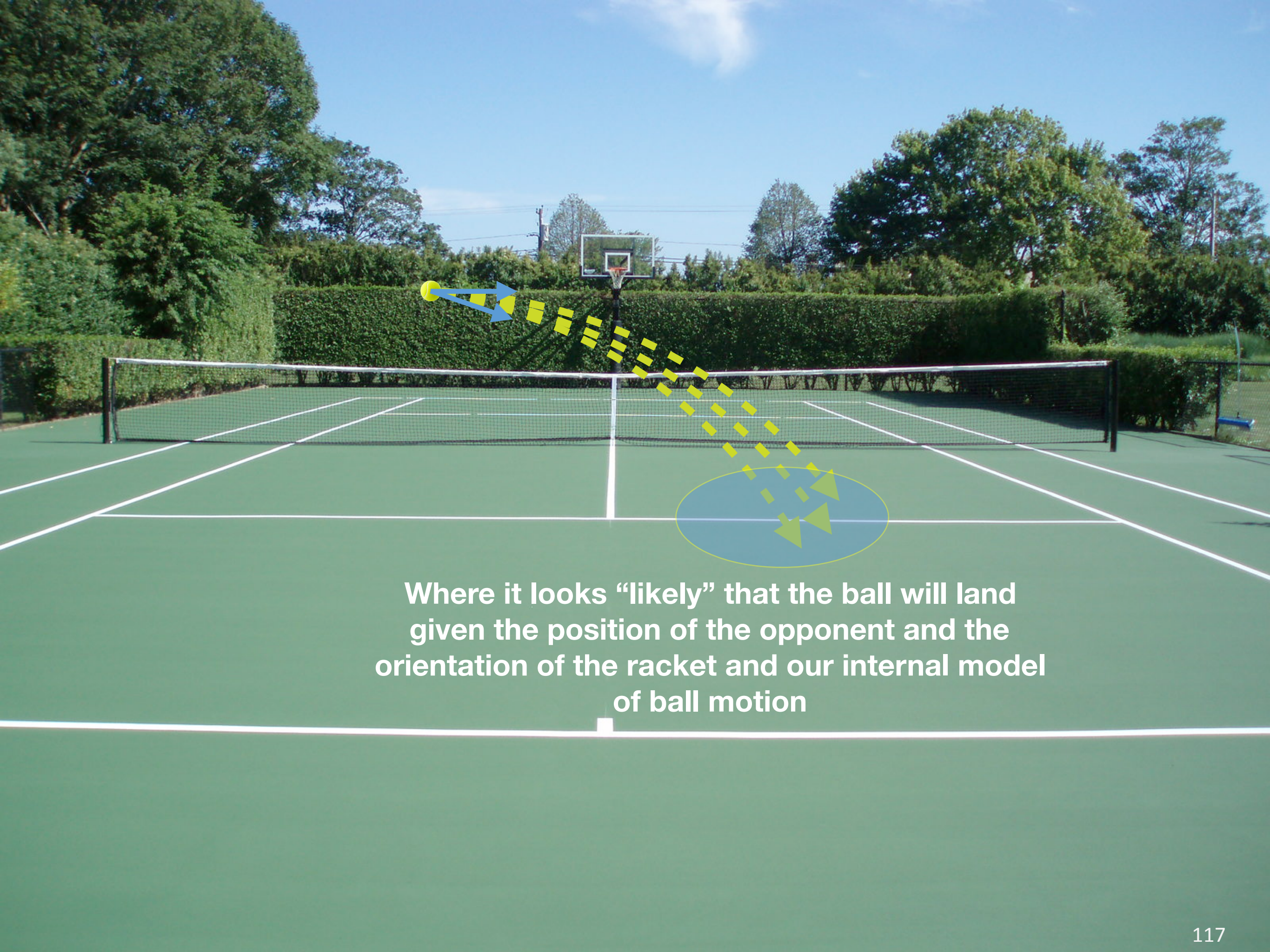




**Where the ball “usually” lands
based on experience**

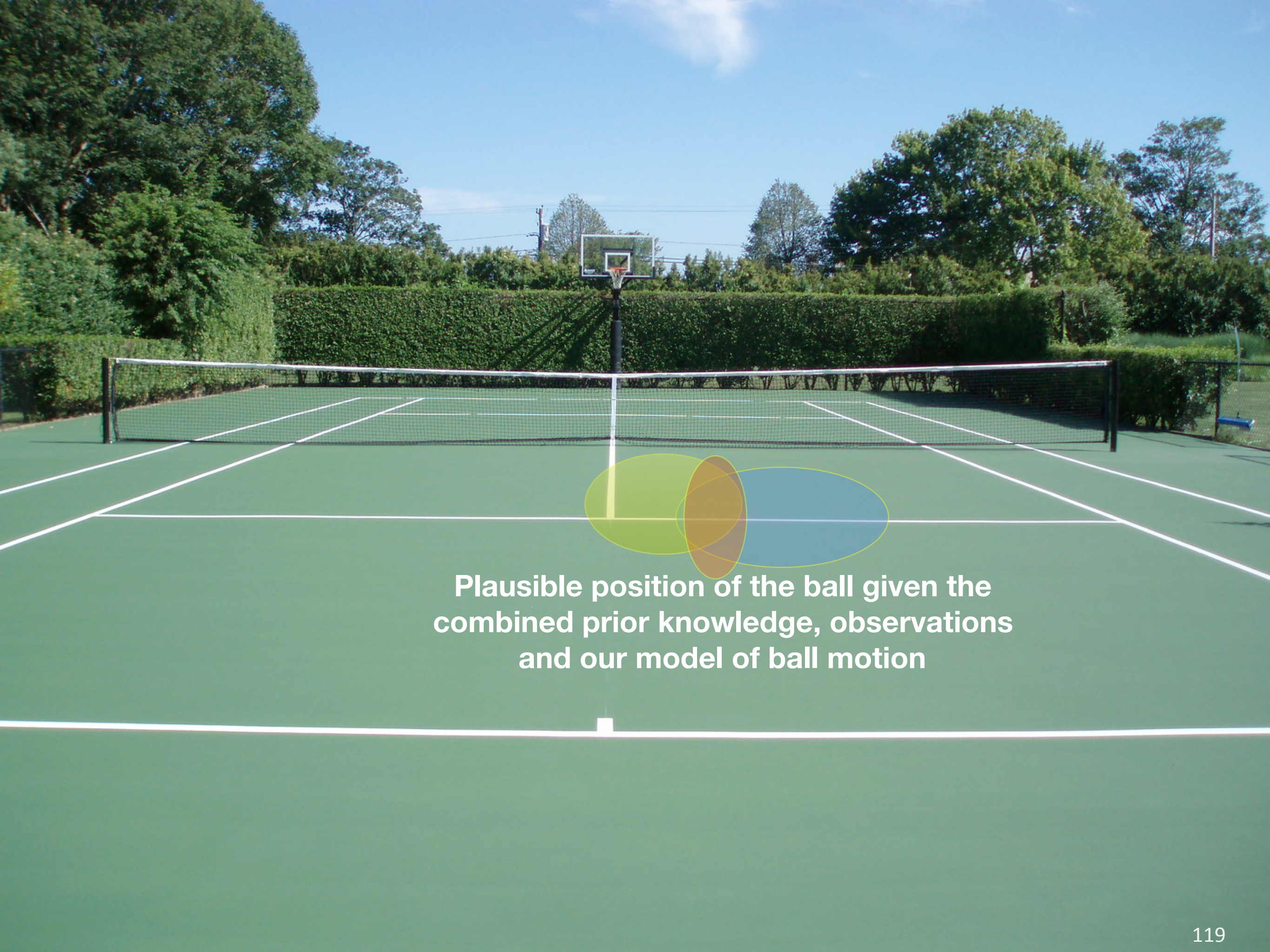


**Where the ball “usually” lands
based on experience**

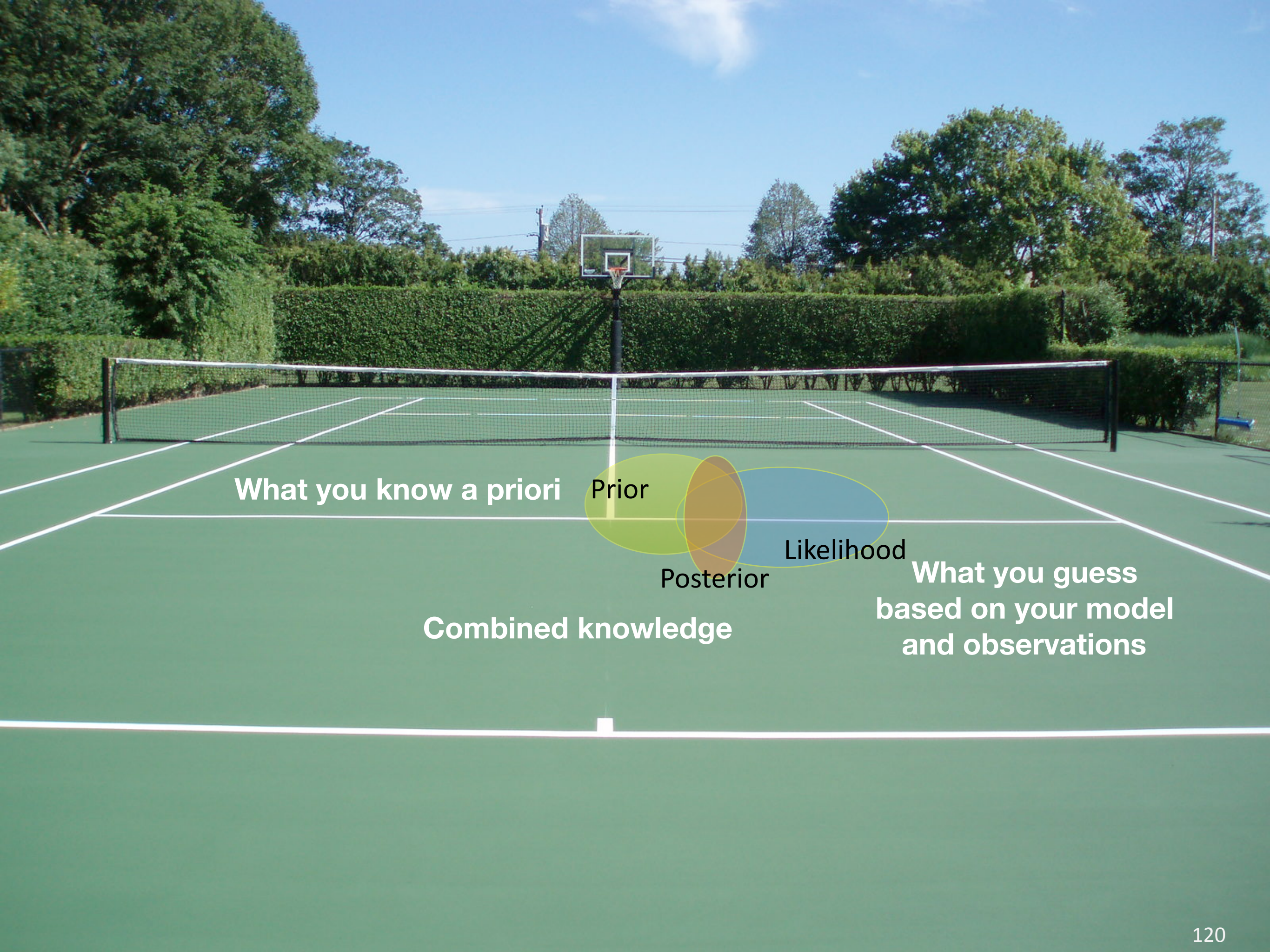


Where it looks “likely” that the ball will land given the position of the opponent and the orientation of the racket and our internal model of ball motion





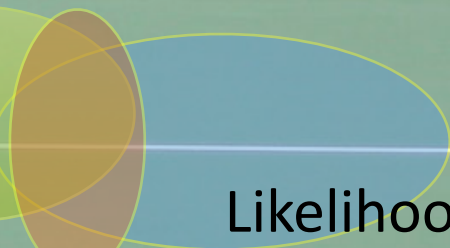
Plausible position of the ball given the combined prior knowledge, observations and our model of ball motion



What you know a priori



Prior



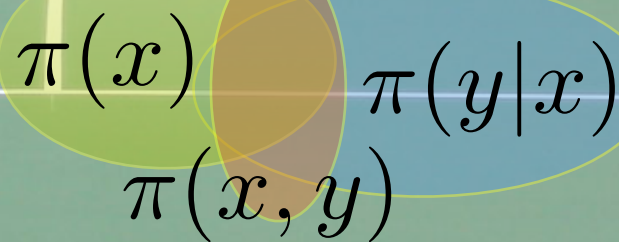
Likelihood

Posterior

Combined knowledge

What you guess based on your model and observations

What you know a priori



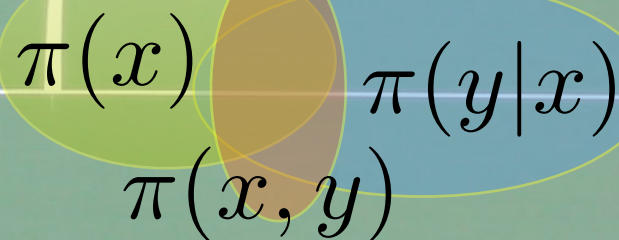
What you guess based on your model and observations

Combined knowledge

$$\pi(x, y) = \pi(y|x)\pi(x)$$

$$\text{posterior} = \frac{\text{prior} \times \text{likelihood}}{\text{evidence}}$$

What you know a priori



Combined knowledge

What you guess based on your model and observations

$$\pi(x, y) = \pi(y|x)\pi(x)$$

Bayes' theorem

$$\text{posterior} = \frac{\text{prior} \times \text{likelihood}}{\text{evidence}}$$

$$\pi(x|y) = \frac{\pi(x)\pi(y|x)}{\int \pi(x)\pi(y|x)dx}$$

A Venn diagram consisting of two overlapping circles. The left circle is light green and contains the text 'prior $\pi(x)$ '. The right circle is light blue and contains the text 'likelihood $\pi(y|x)$ '. The intersection of the two circles is shaded a darker green and contains the text 'posterior $\propto \pi(x|y)$ '.

prior $\pi(x)$

likelihood $\pi(y|x)$

posterior
 $\propto \pi(x|y)$

Parameter identification: Bayesian approach

Bayes' theorem

$$\pi(x|y) = \frac{\pi(x)\pi(y|x)}{\int \pi(x)\pi(y|x)dx}$$

$\pi(\cdot)$: probability distribution function

$\pi(\cdot|\cdot)$: conditional probability distribution function

x : material parameter

y : observations

Parameter identification: Bayesian approach

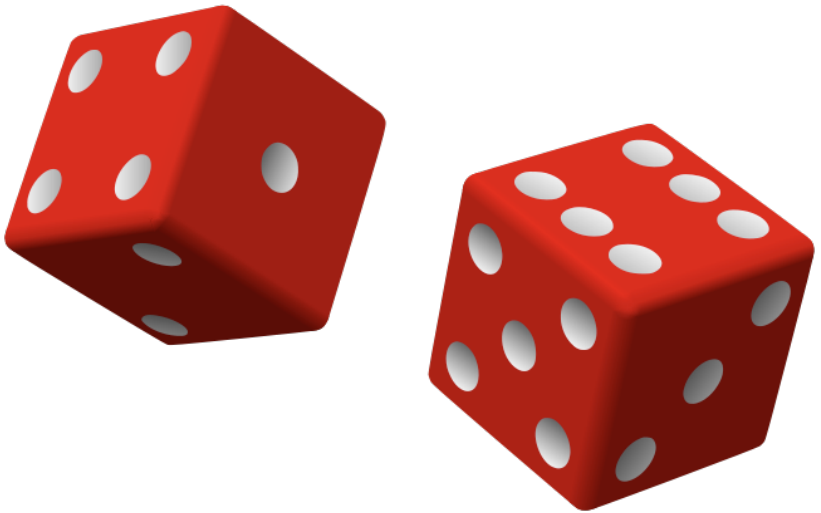
Bayes' theorem

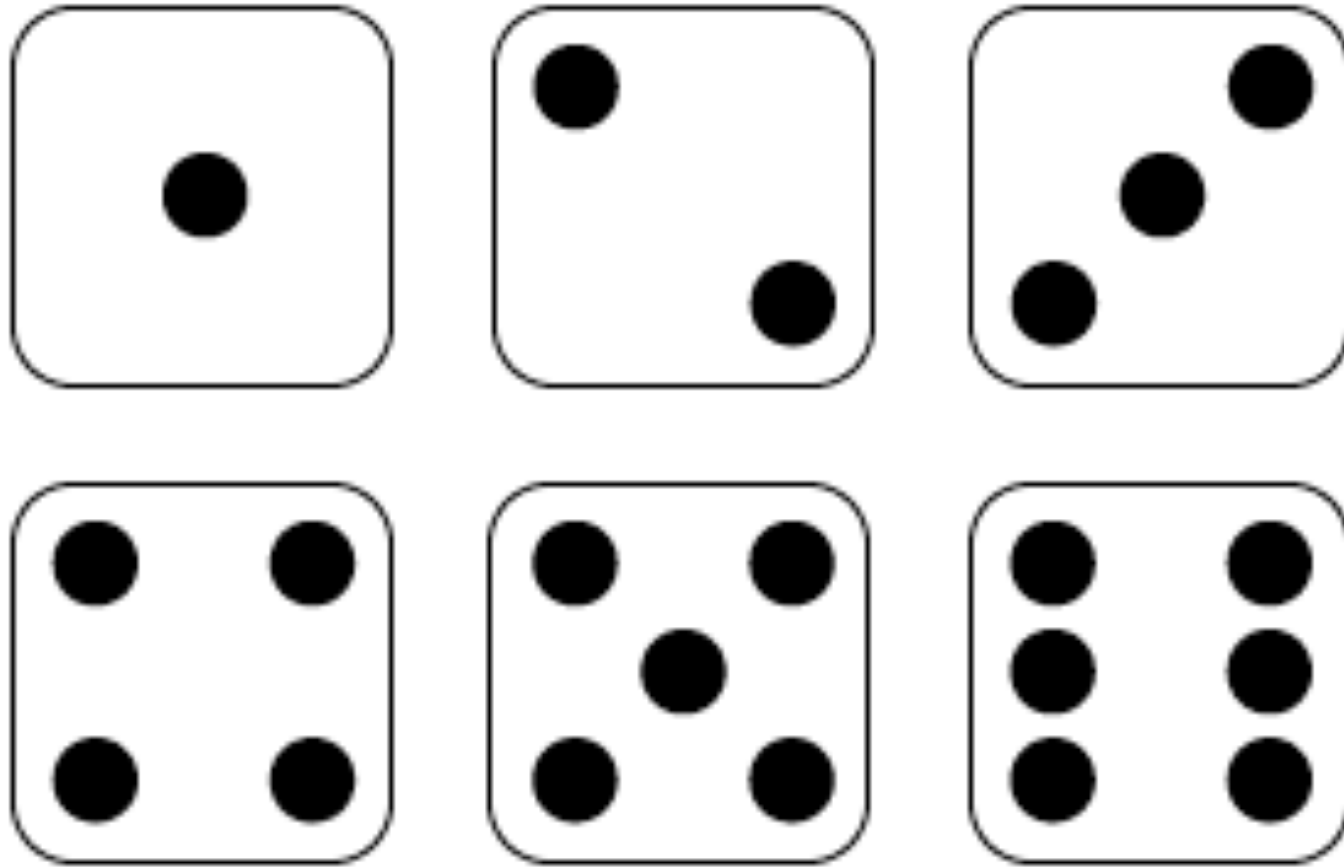
$$\pi(x|y) = \frac{\pi(x)\pi(y|x)}{\int \pi(x)\pi(y|x)dx}$$

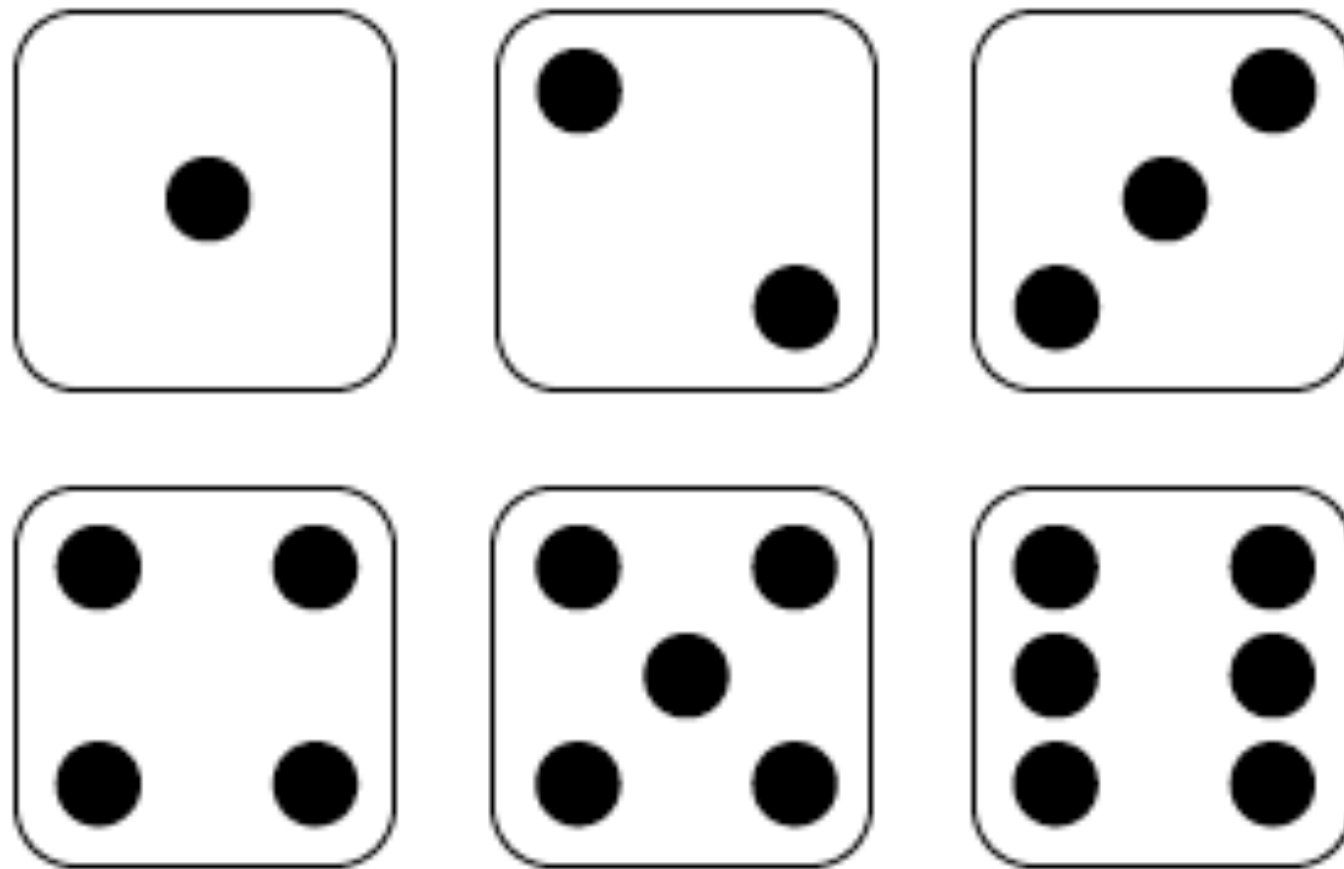
Descriptive formula

$$\text{Posterior} = \frac{\text{Prior} \times \text{Likelihood}}{\text{Evidence}}$$

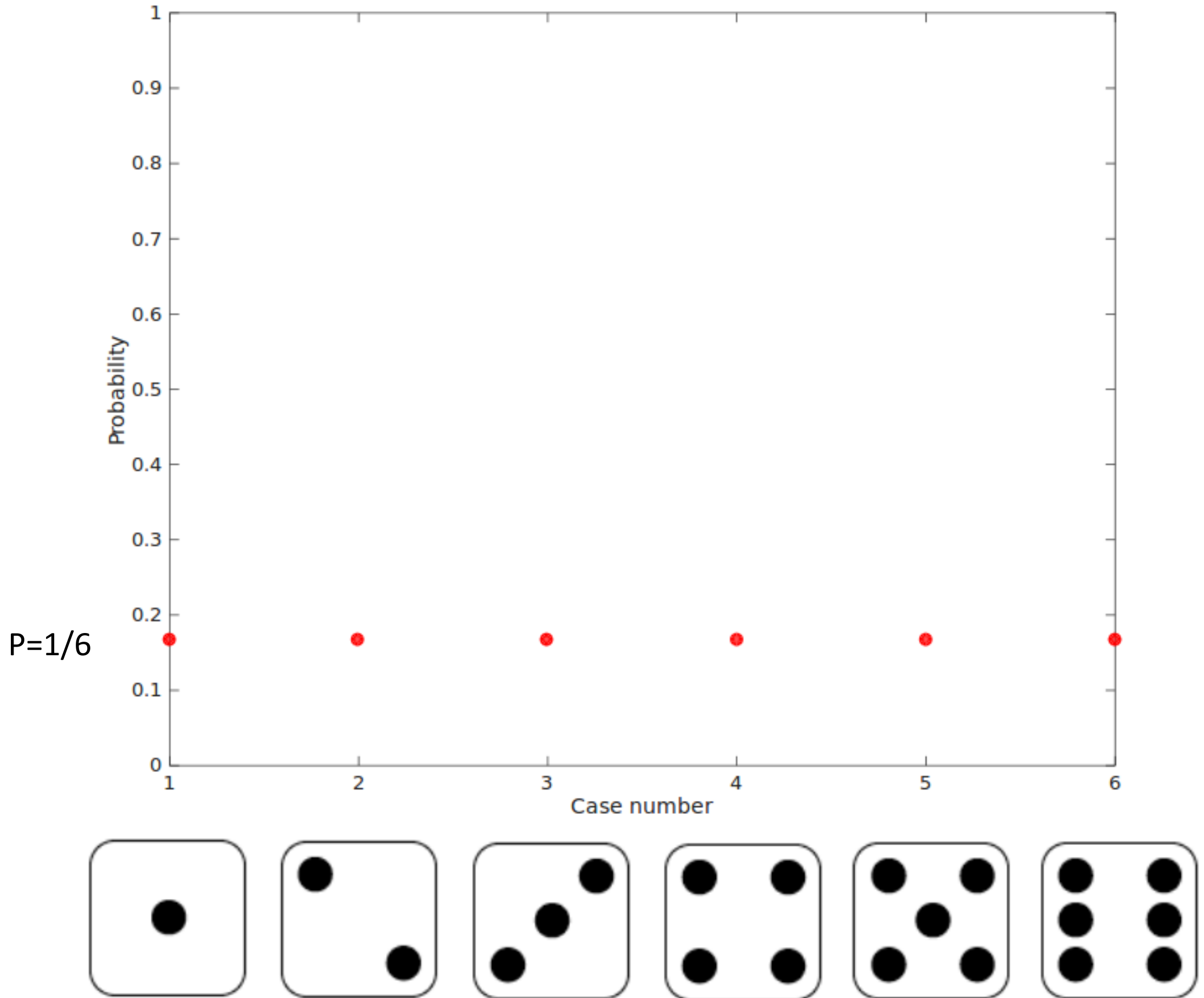
A discrete example of Bayes' theorem







This is our prior information for the probability of each face: $1/6$



$P=1/6$



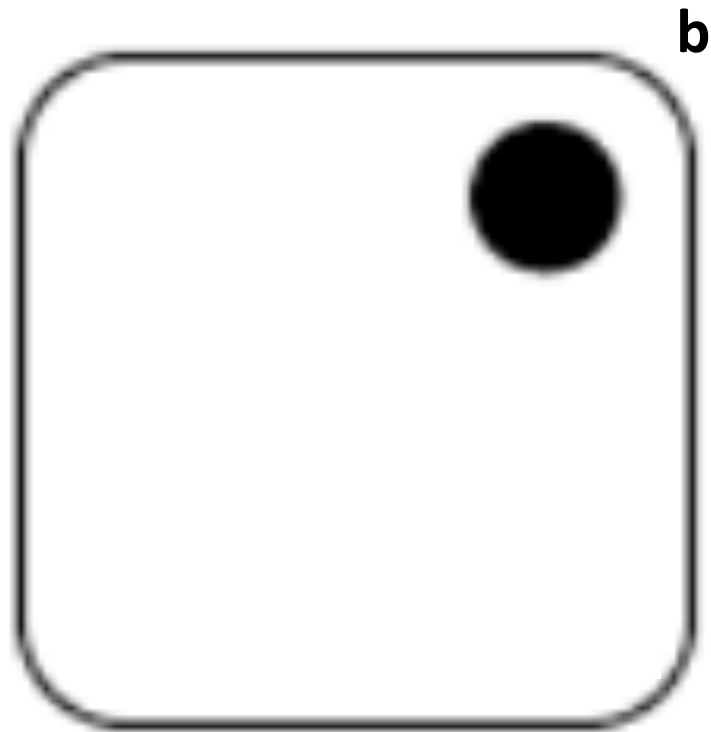
Assume that after throwing the dice, you see the above evidence



Goal: determine the probability of this evidence for each face of the dice

a



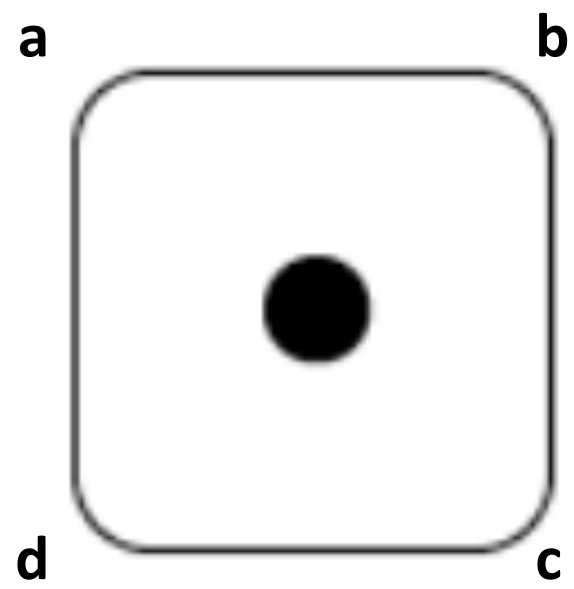


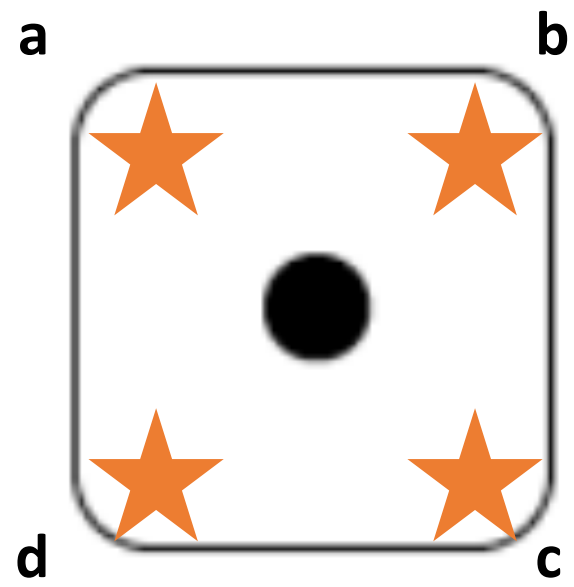


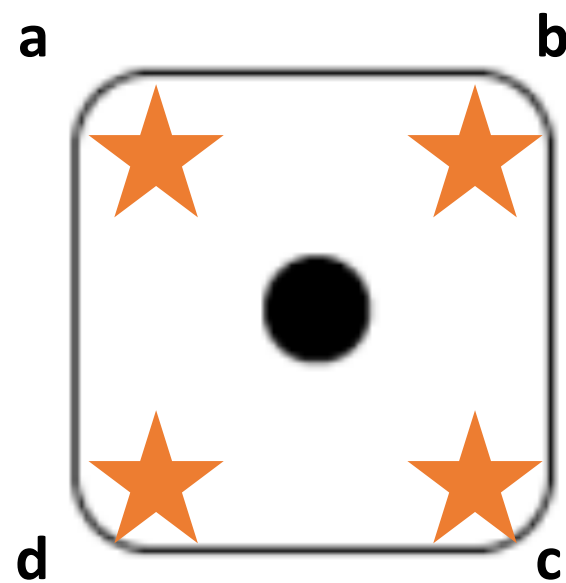
c



d

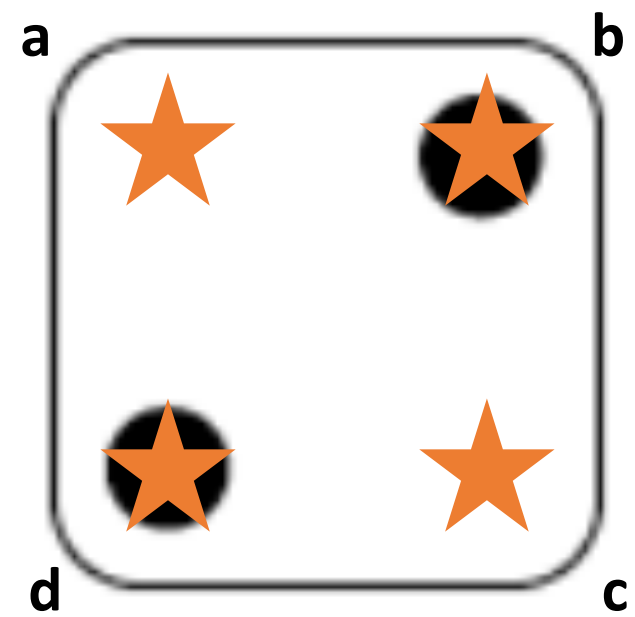
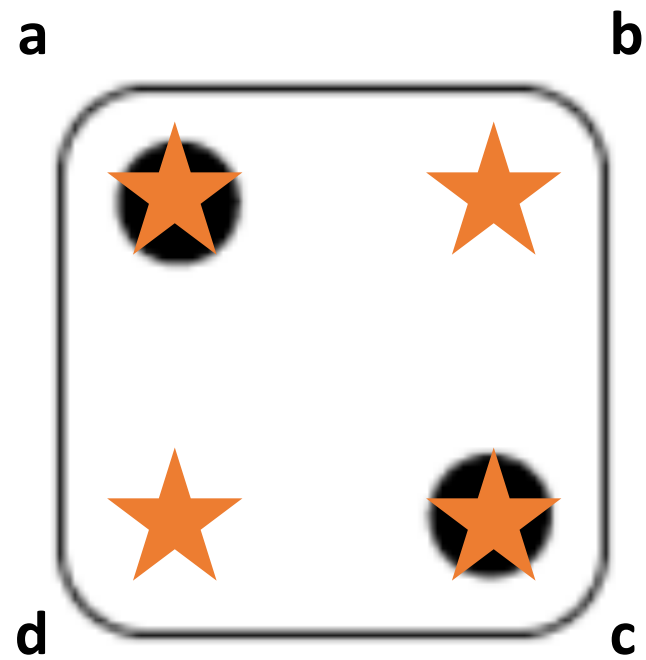






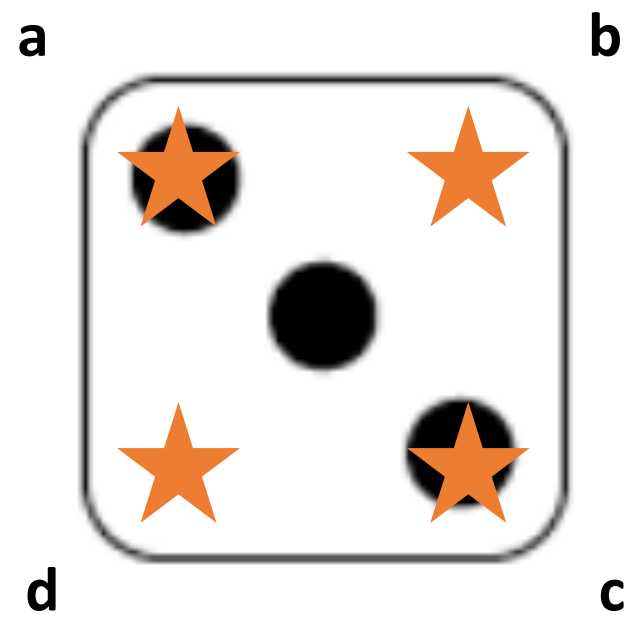
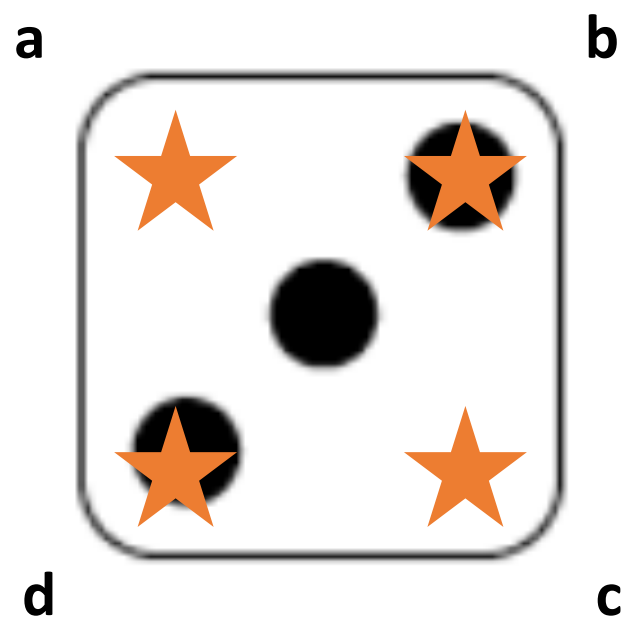
One would never see a dot at the star positions for this face
The probability of the evidence is *zero*



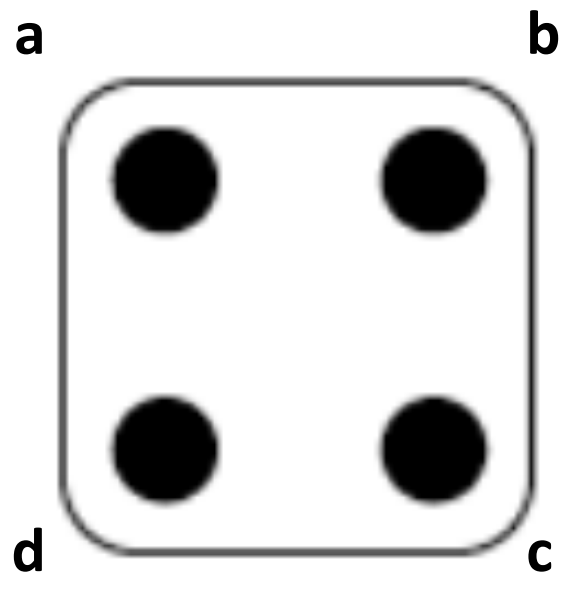


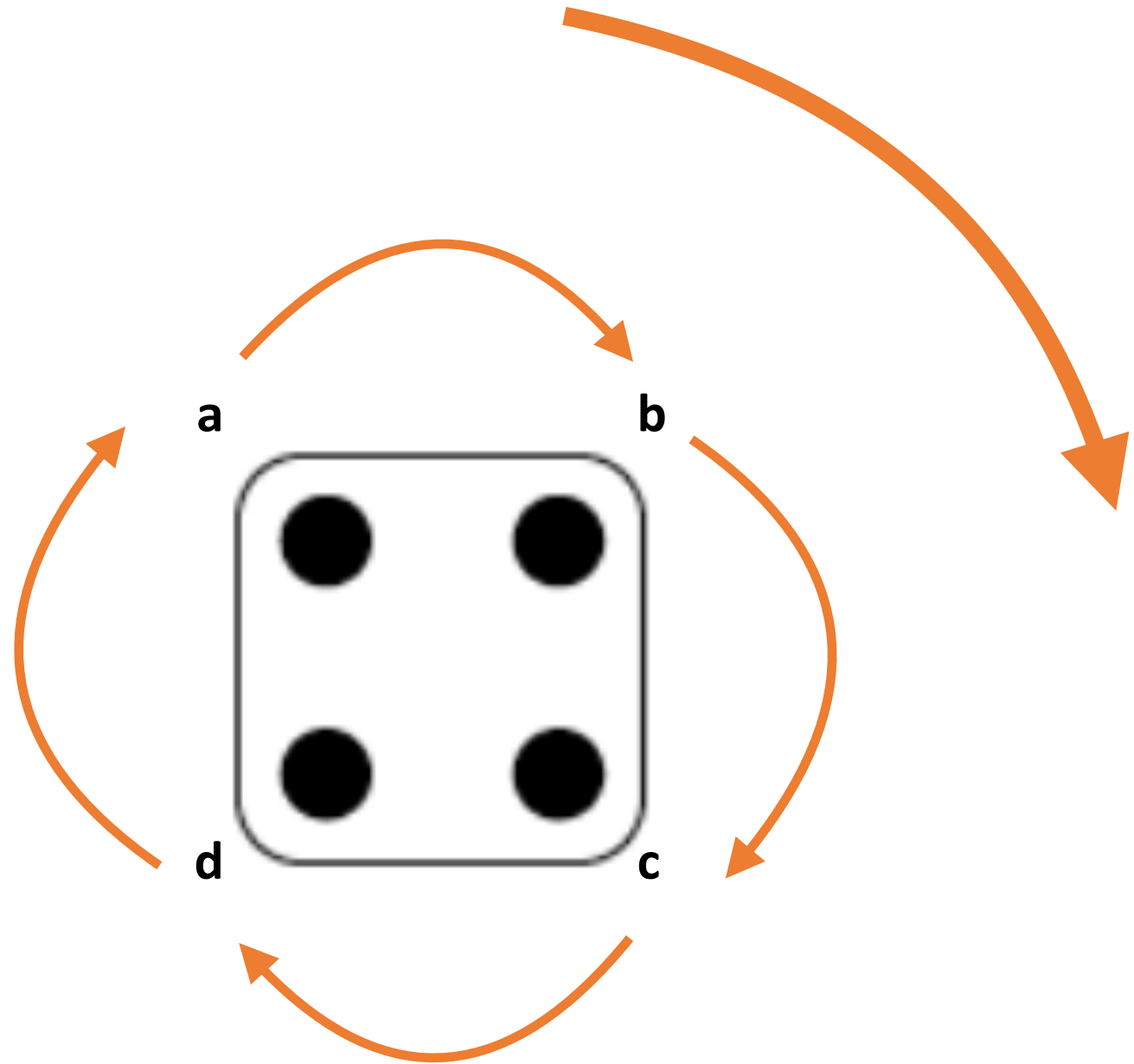
Two possibilities (a,c) and (b,d)

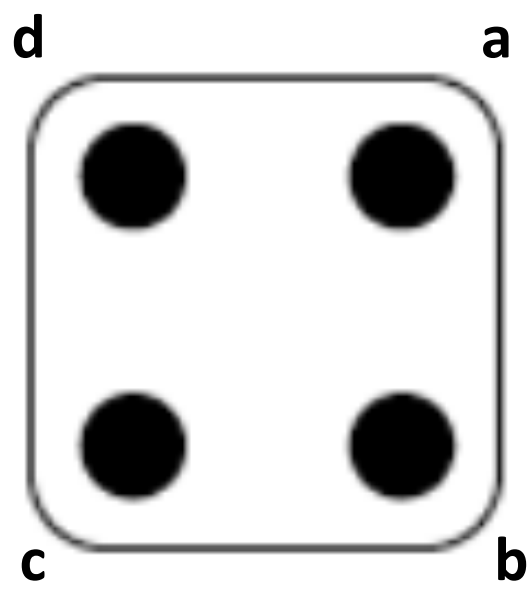


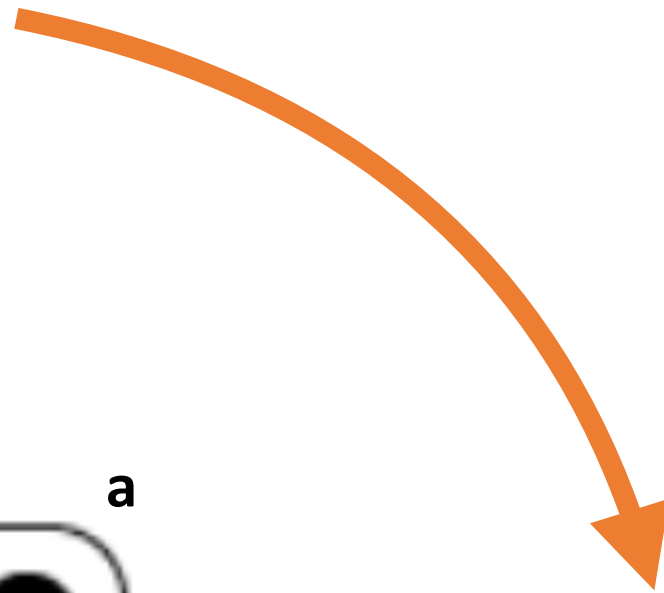
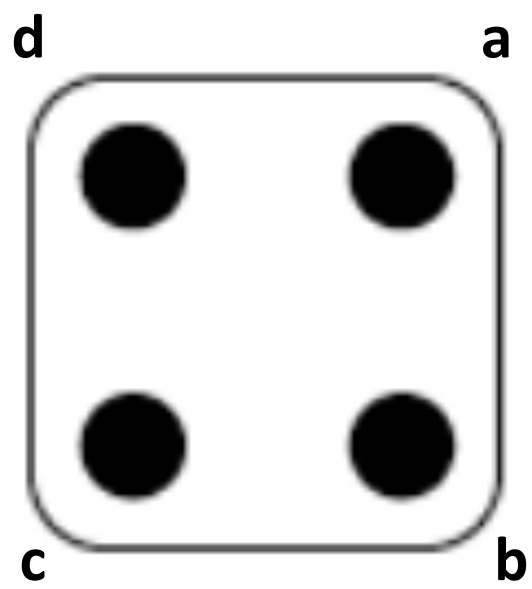


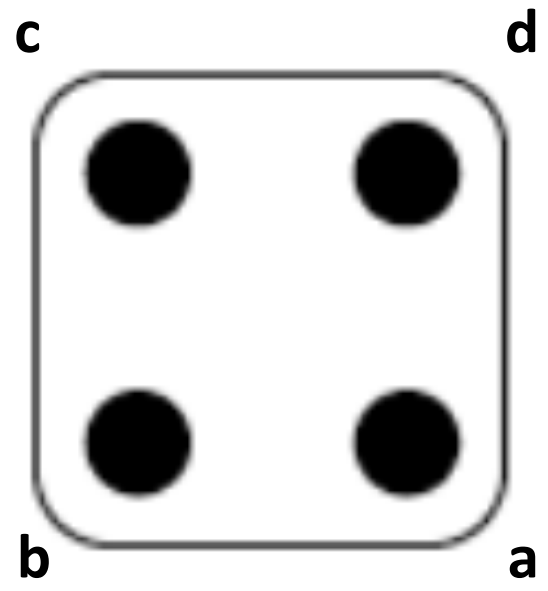
Also two possibilities (a,c) and (b,d)

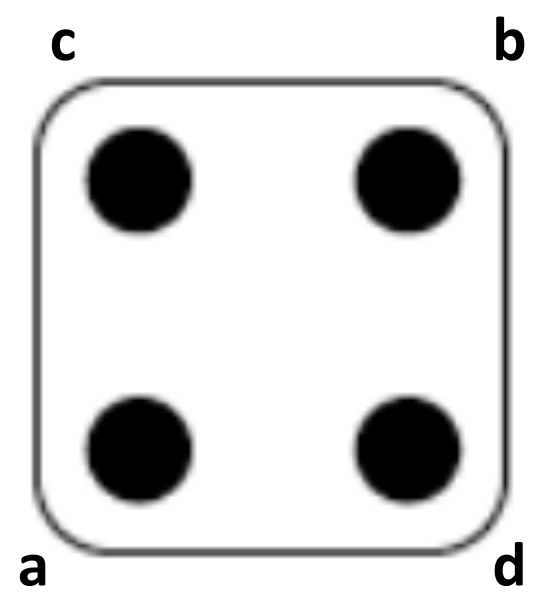
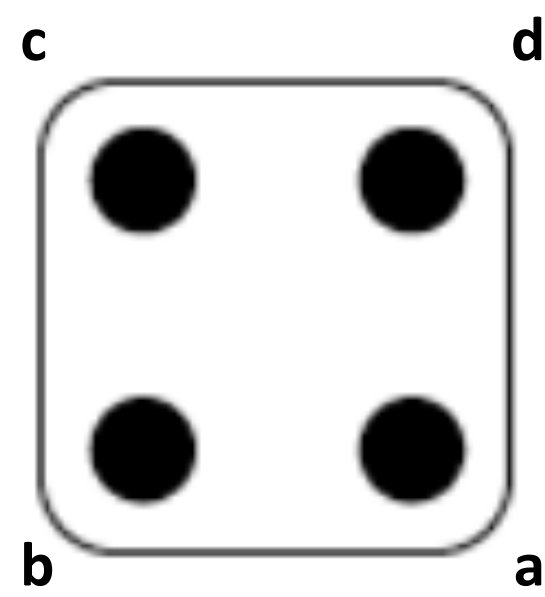
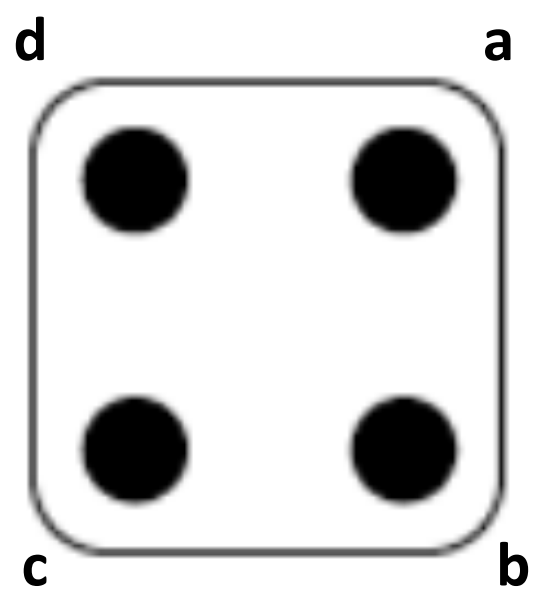
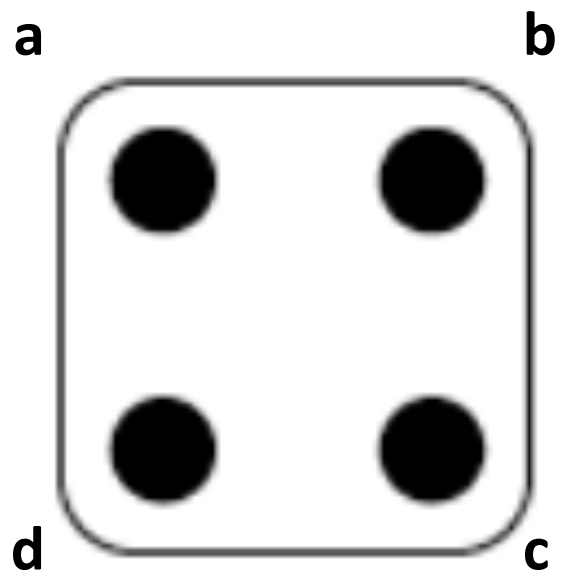


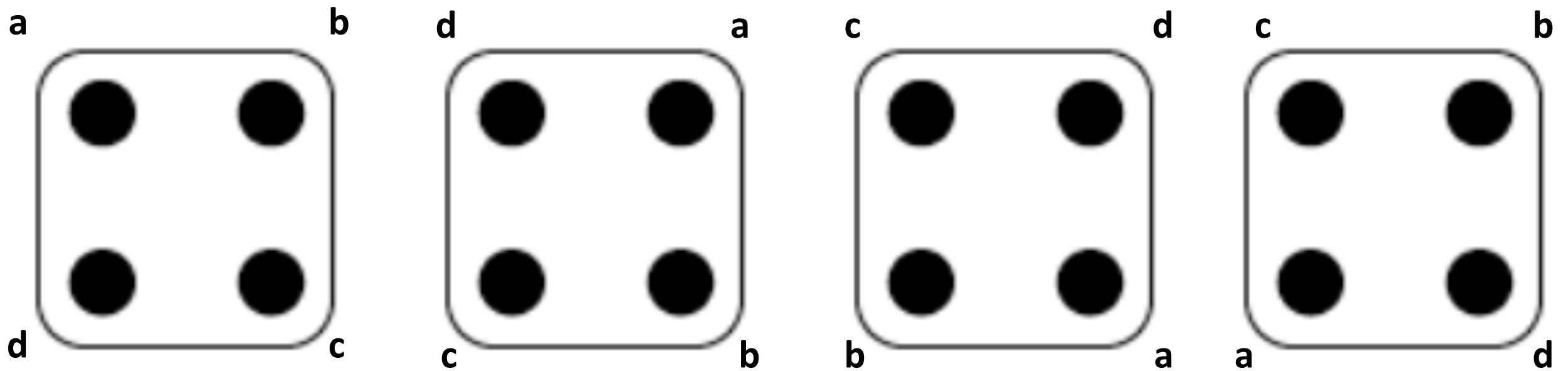




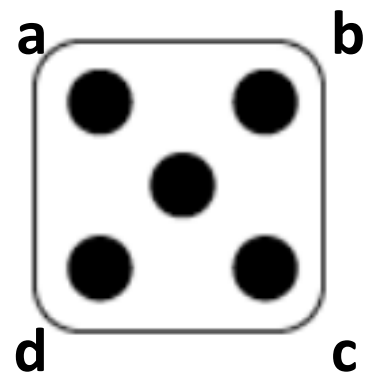




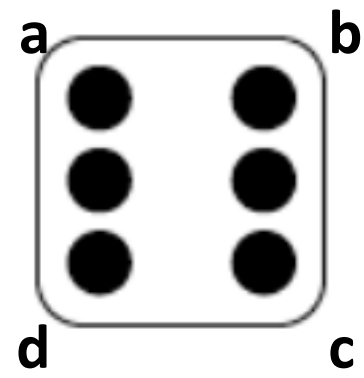




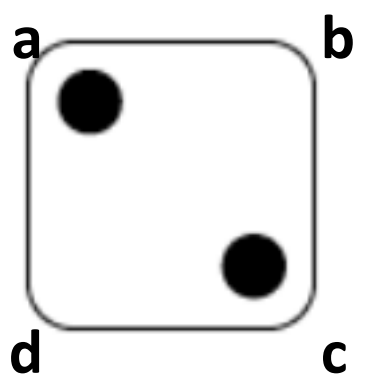
Four possibilities



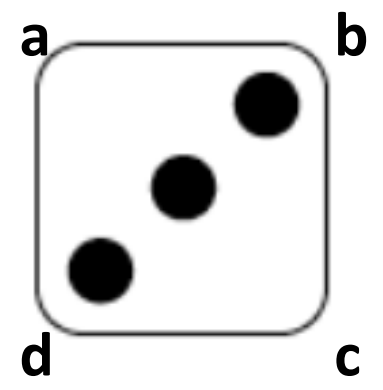
Four possibilities



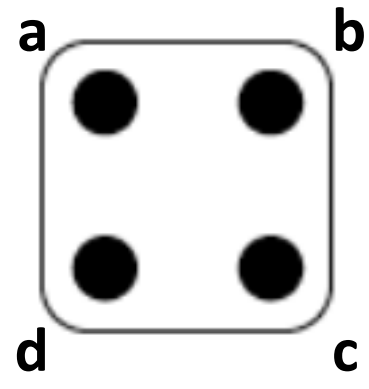
Four possibilities



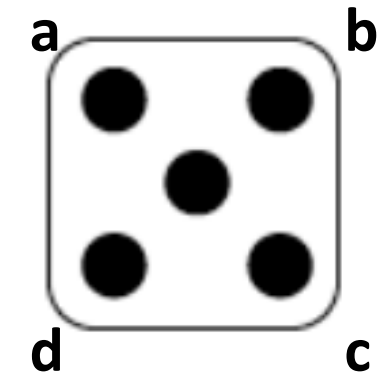
2



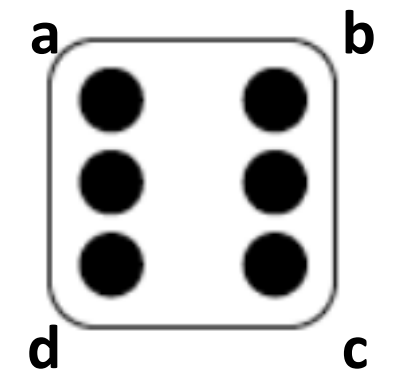
2



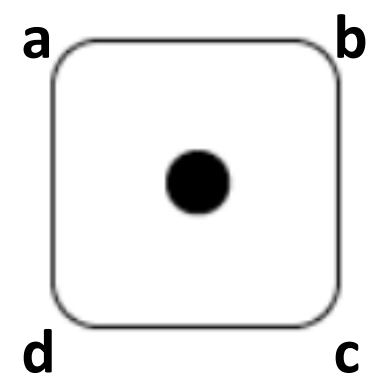
4



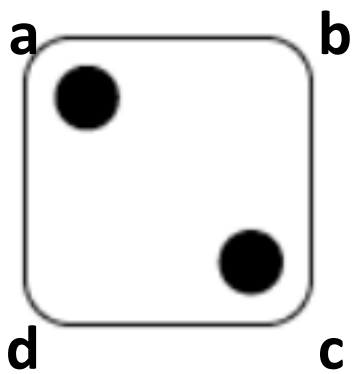
4



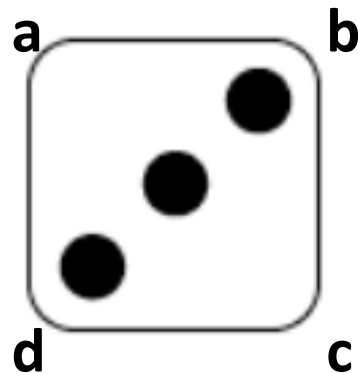
4



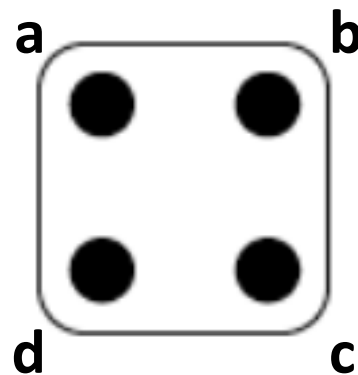
0



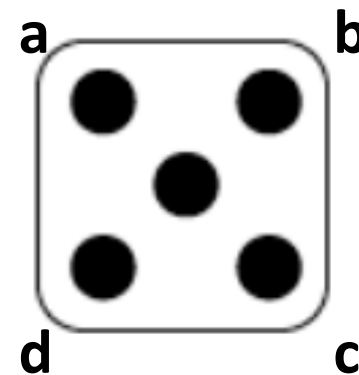
2



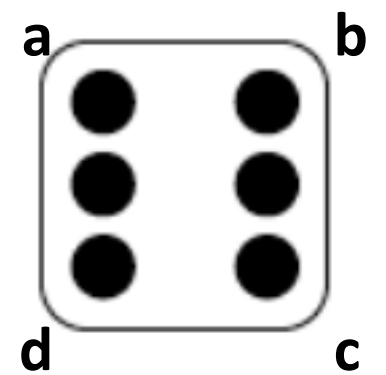
2



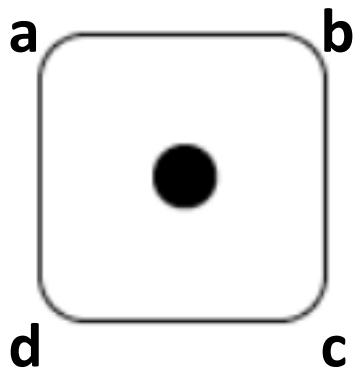
4



4



4



0

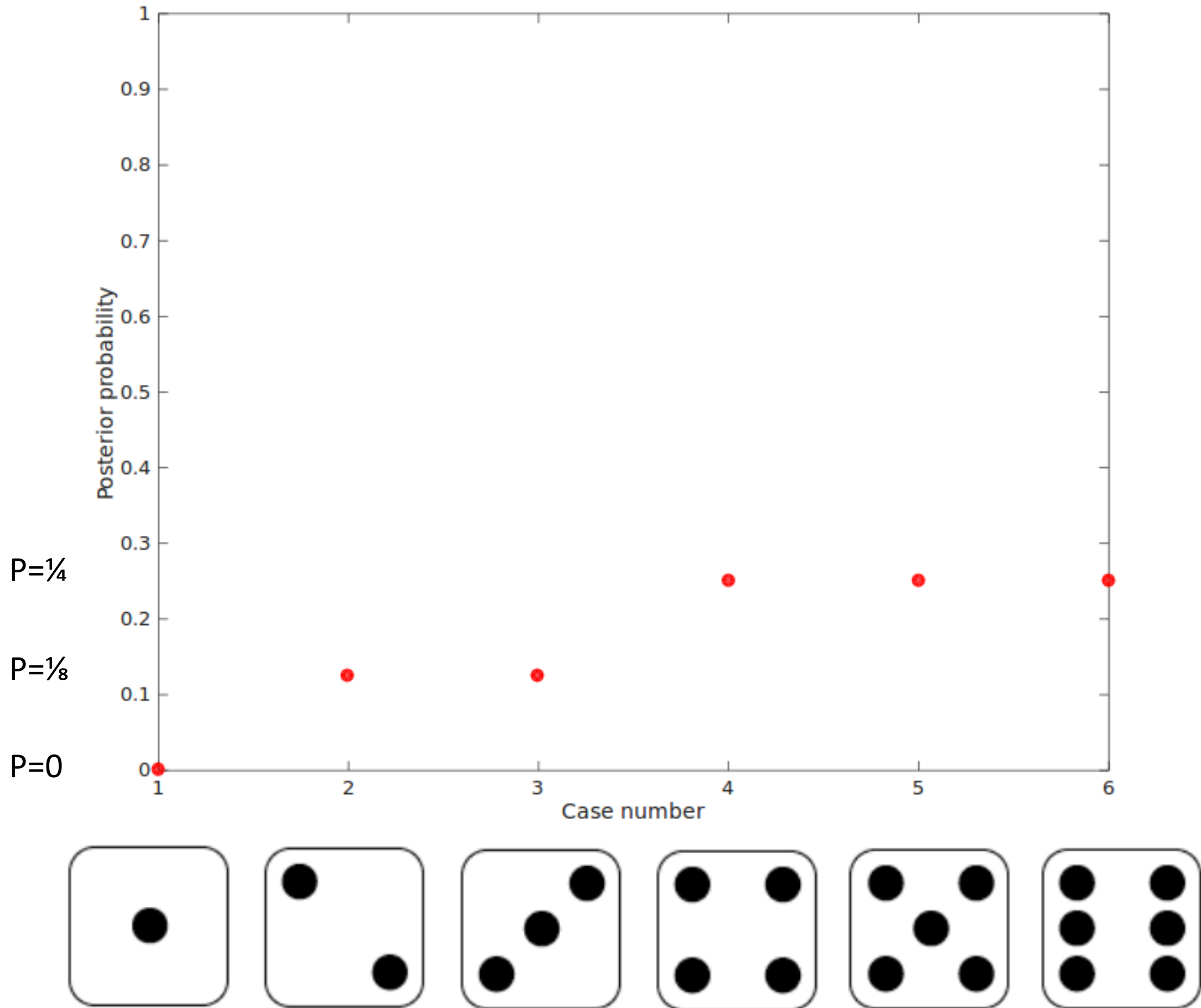
$$\pi(y) = \frac{0 + 2 + 2 + 4 + 4 + 4}{6 \times 4} = \frac{16}{24}$$

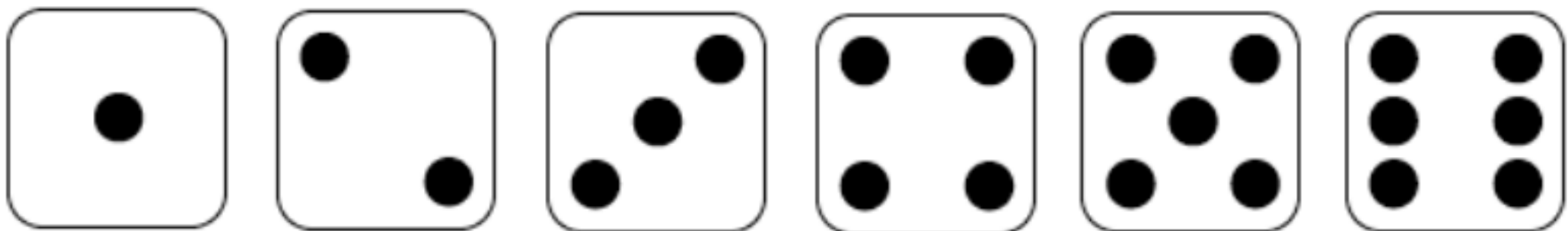
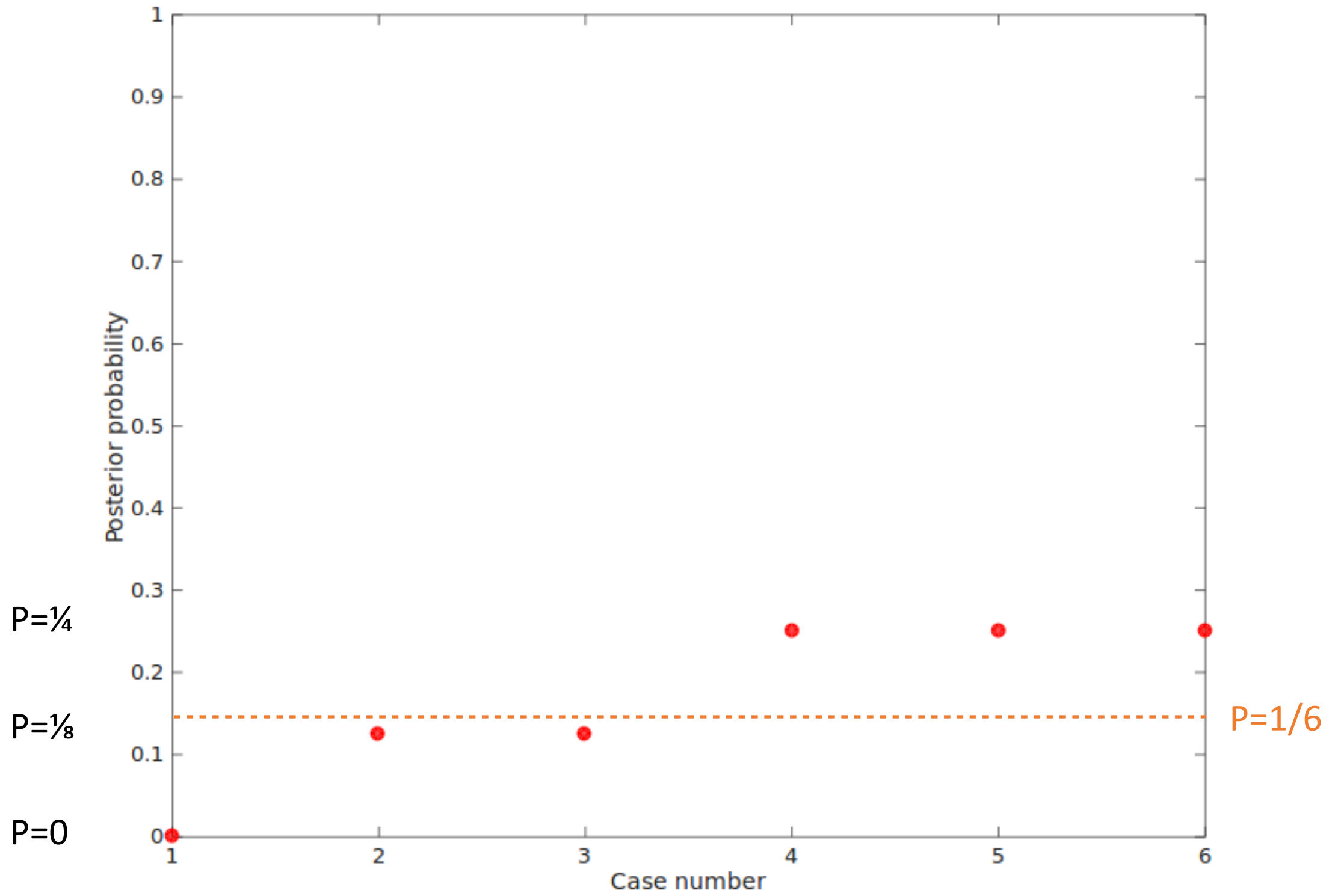


$$\pi(x|y) = \frac{\text{Prior} \times \text{Likelihood}}{\text{Evidence}} = \frac{\frac{1}{6} \times \frac{1}{2}}{\frac{16}{24}} = 0.125$$

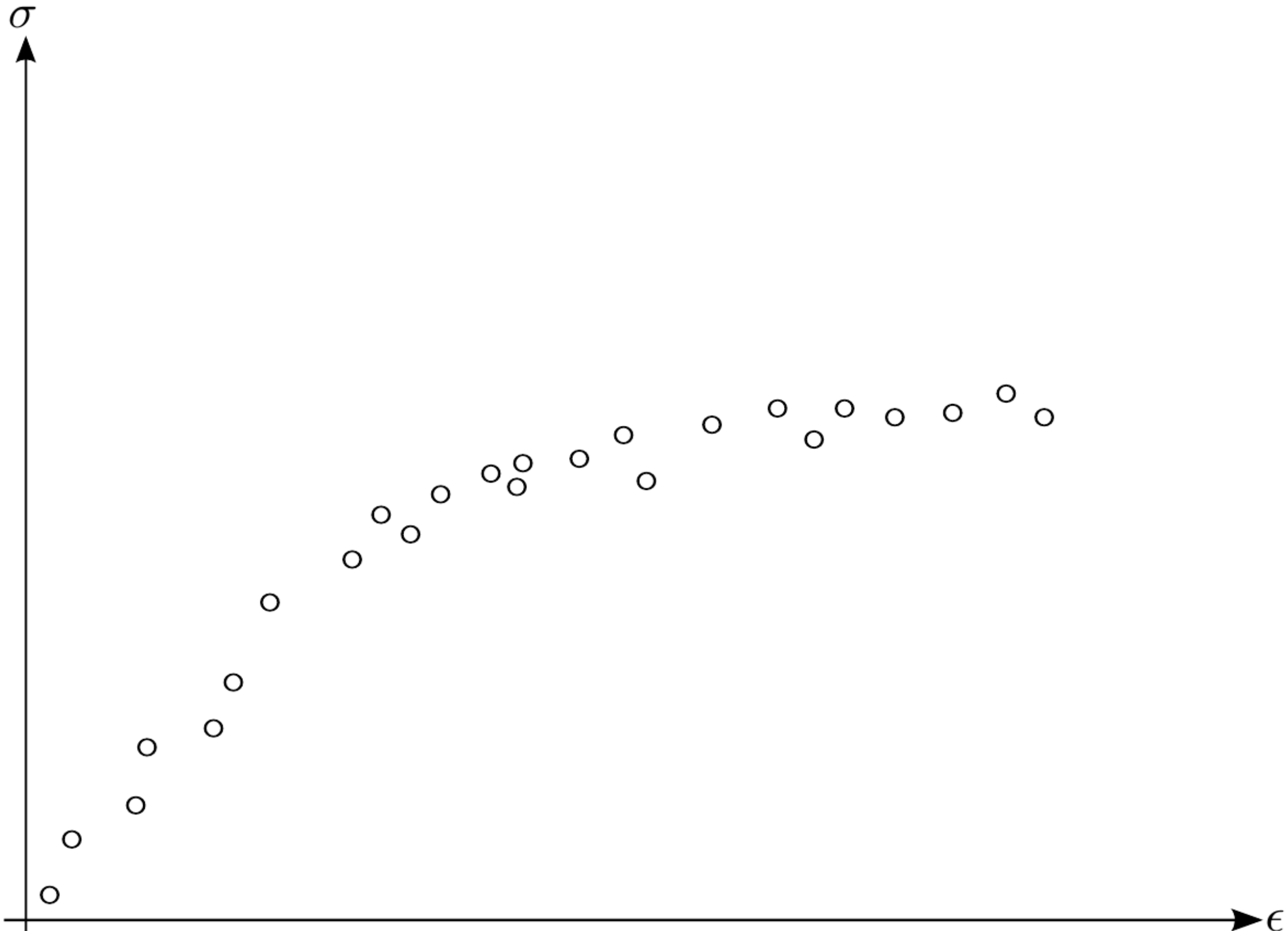
Probability that  was the face of the dice knowing 



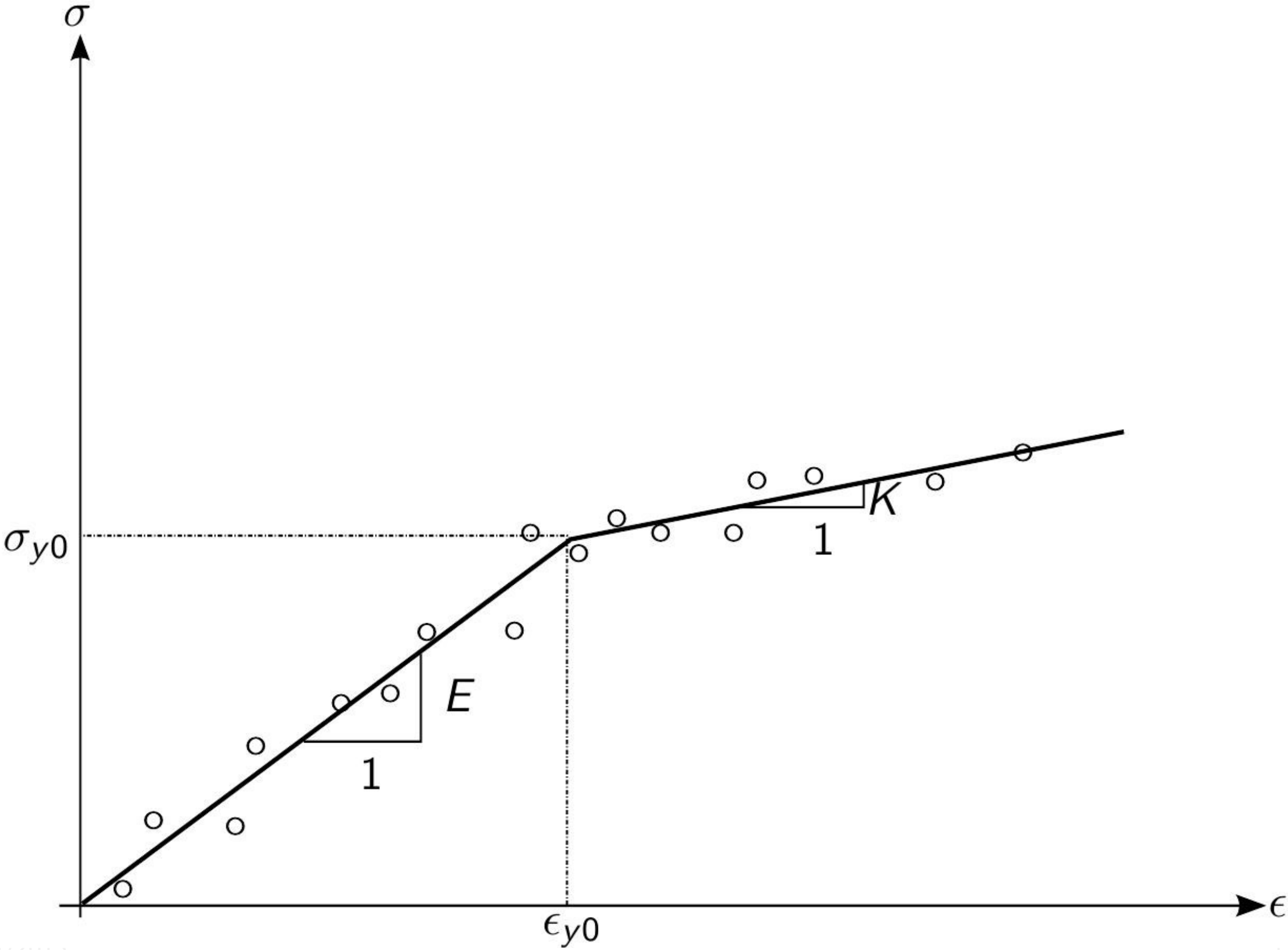




Stress-strain data



Identify the parameters



Construct the likelihood function

Model

$Y = f(X, \Omega)$ observations=f(parameters, error)

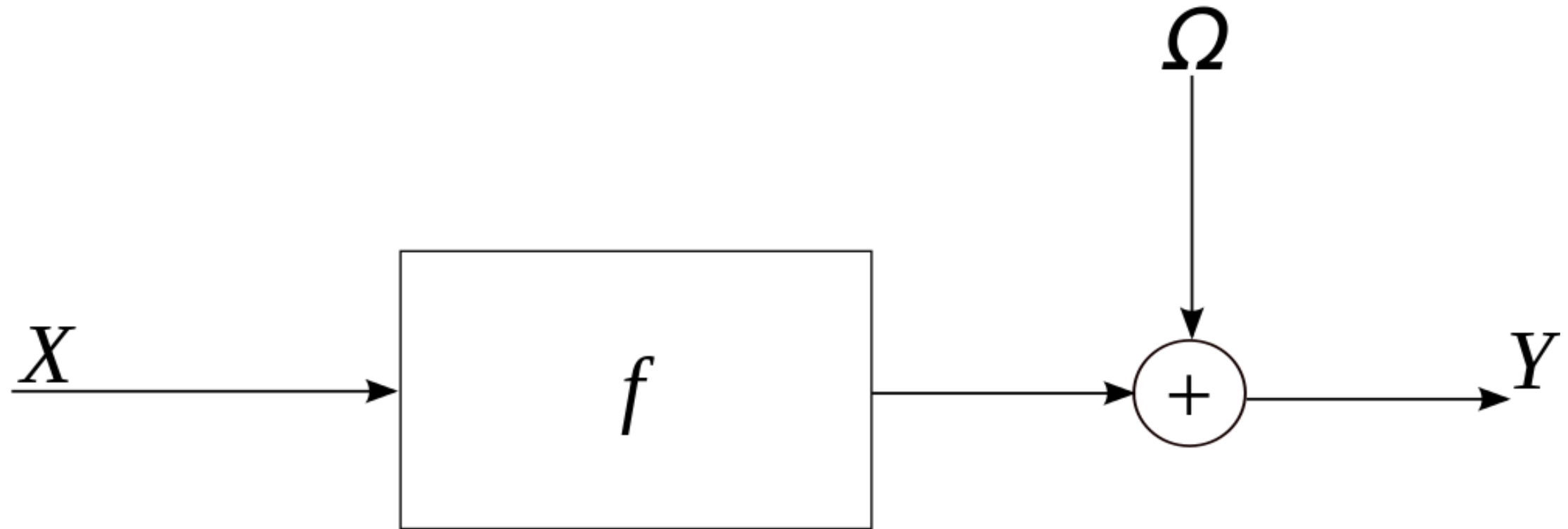
Ω : Error

X : Material parameter

Noise model

Additive noise model

$$Y = f(X) + \Omega$$



Likelihood function

Likelihood function for additive model

$$\pi(y|x) = \pi(\omega) = \pi(y - f(x))$$



$$Y = f(X) + \Omega$$

Constitutive law: linear elasticity

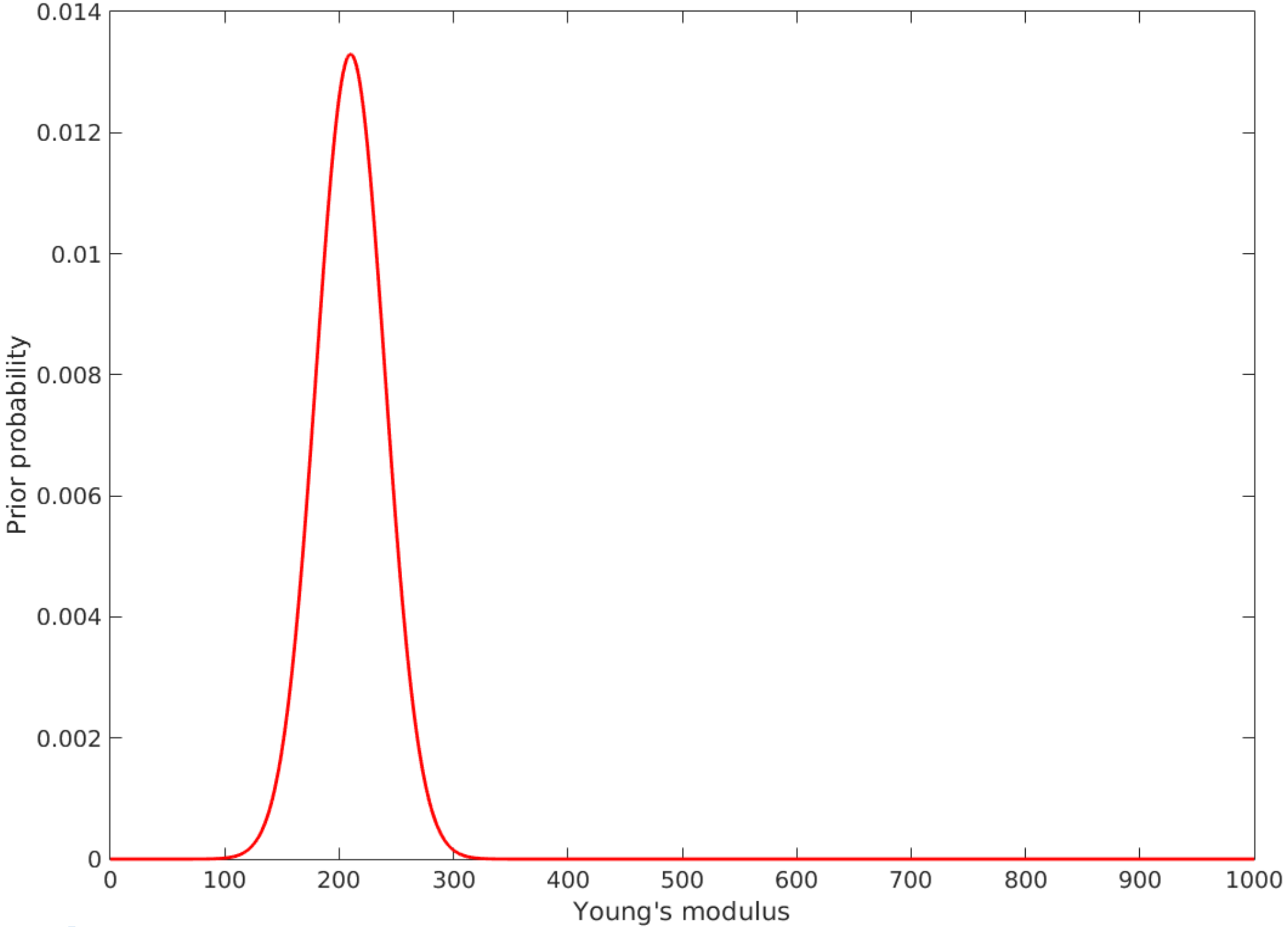
Constitutive model

$$\sigma = E\epsilon \text{ or } \sigma = x\epsilon$$

Observed data

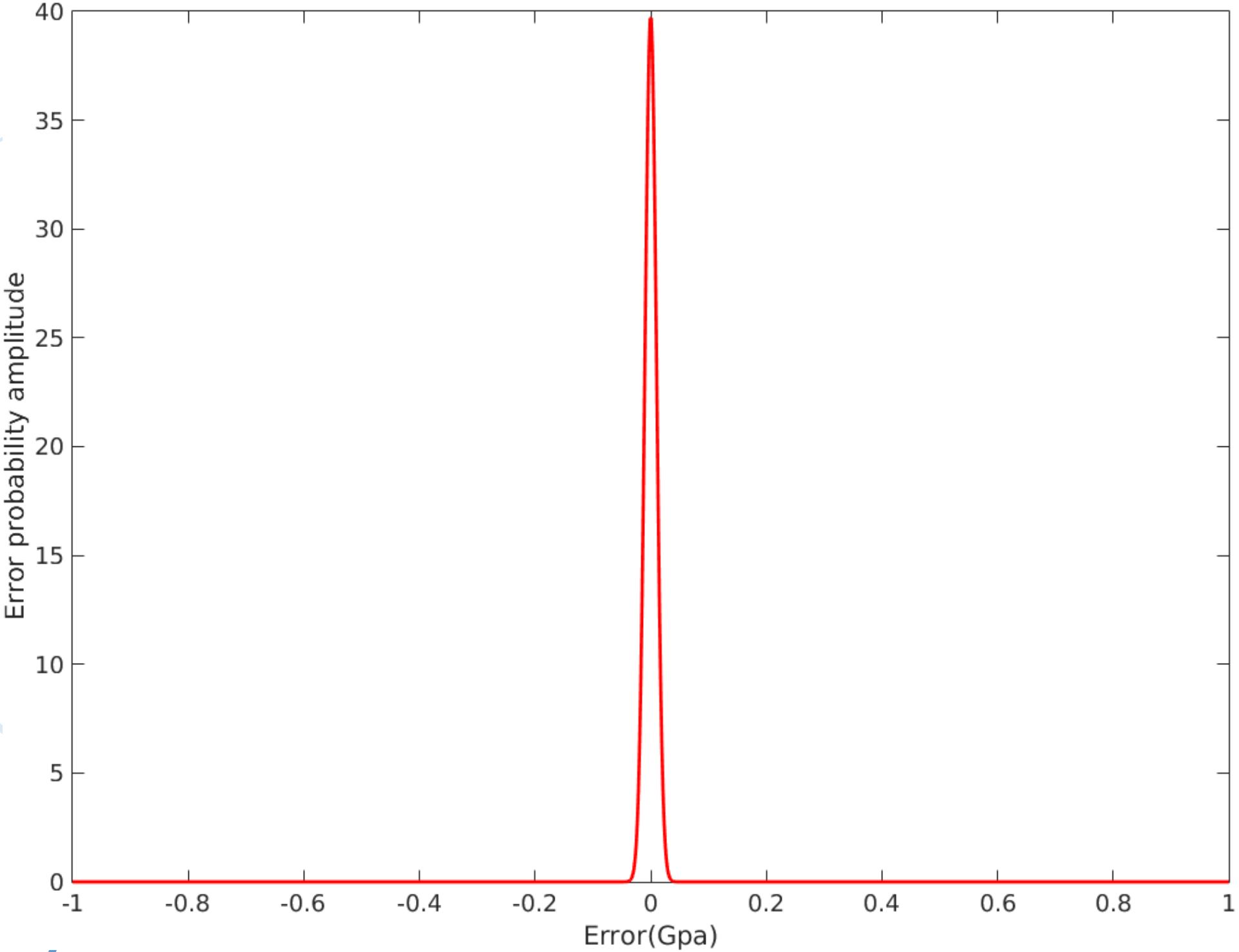
$$Y = X\epsilon + \Omega$$

Prior information on Young's modulus



$$\pi_{prior}(x) = N(210, 900)$$

Error model (noise)



$$\pi(e)_{error} = N(0, 0.0001)$$

Likelihood function



Likelihood function

$$\pi(y|x) = N(y - x\epsilon, 0.0001)$$

$$\pi(y|x) = \pi(\omega) = \pi(y - f(x))$$

Bayes' theorem: calculate the posterior

$$\text{posterior} = \frac{\text{prior} \times \text{likelihood}}{\text{evidence}}$$

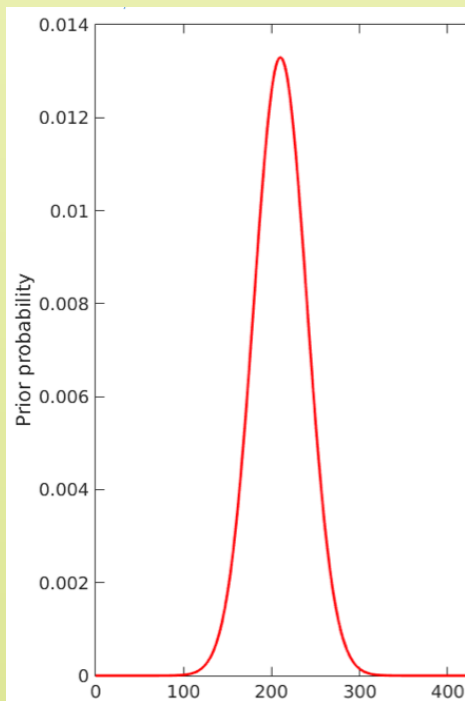
$$\pi(x|y) = \frac{\pi(x)\pi(y|x)}{\int \pi(x)\pi(y|x)dx}$$

Bayes' theorem: calculate the posterior

$$\text{posterior} = \frac{\text{prior} \times \text{likelihood}}{\text{evidence}}$$

$$\pi(x|y) = \frac{\pi(x)\pi(y|x)}{\int \pi(x)\pi(y|x)dx}$$

prior $\pi(x)$

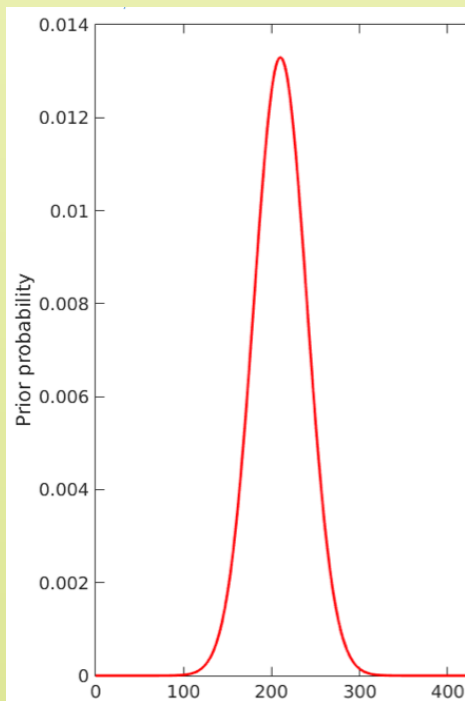


Bayes' theorem: calculate the posterior

$$\text{posterior} = \frac{\text{prior} \times \text{likelihood}}{\text{evidence}}$$

$$\pi(x|y) = \frac{\pi(x)\pi(y|x)}{\int \pi(x)\pi(y|x)dx}$$

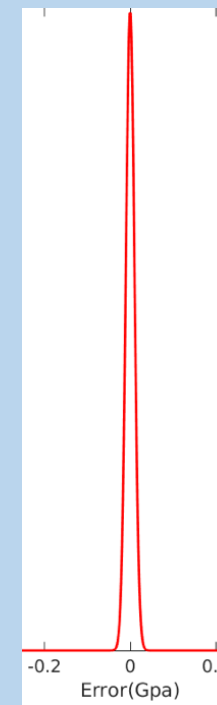
prior $\pi(x)$



likelihood $\pi(y|x)$

$$\pi(y|x) = N(y - x\epsilon, 0.0001)$$

$$\pi(y|x) = \pi(\omega) = \pi(y - f(x))$$

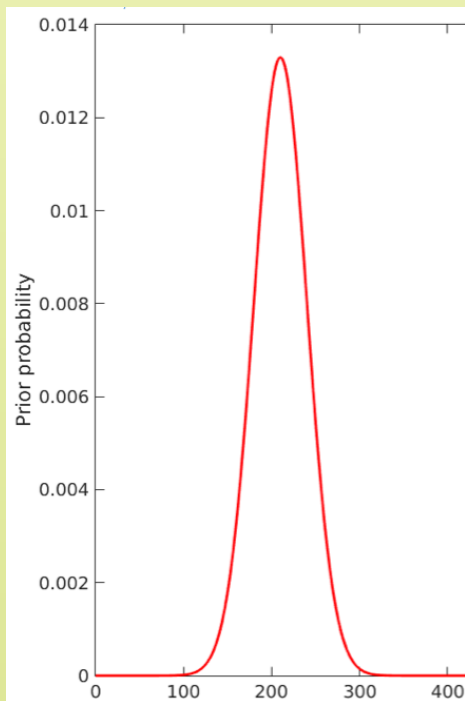


Bayes' theorem: calculate the posterior

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prior $\pi(x)$



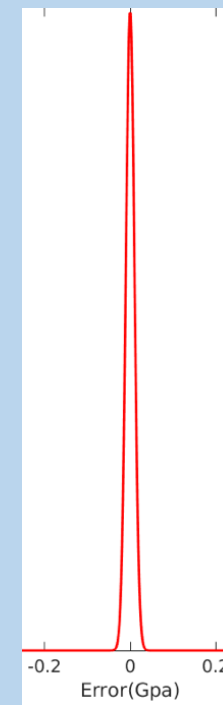
posterior

$$\propto \pi(x|y)$$

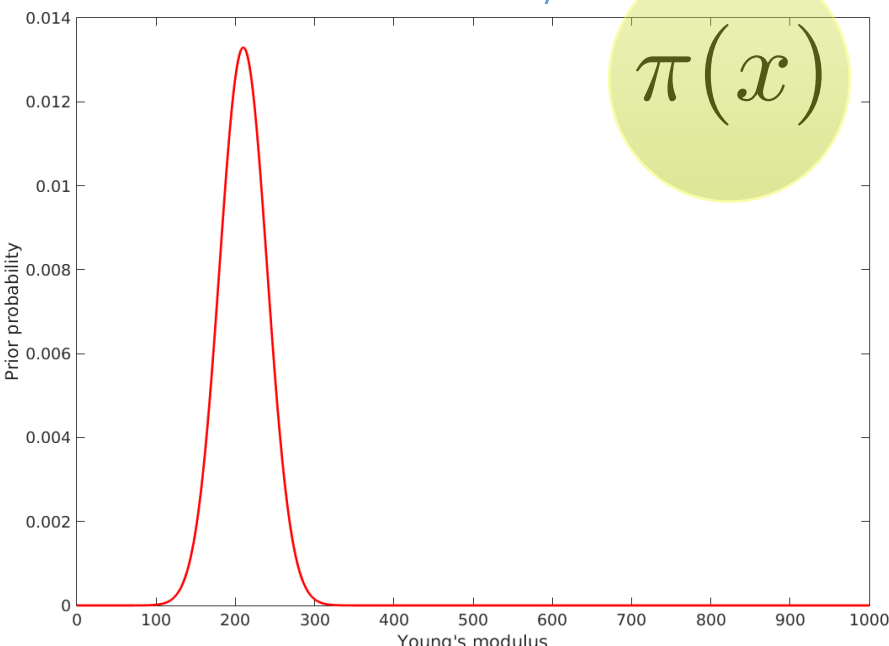
likelihood $\pi(y|x)$

$$\pi(y|x) = N(y - x\epsilon, 0.0001)$$

$$\pi(y|x) = \pi(\omega) = \pi(y - f(x))$$

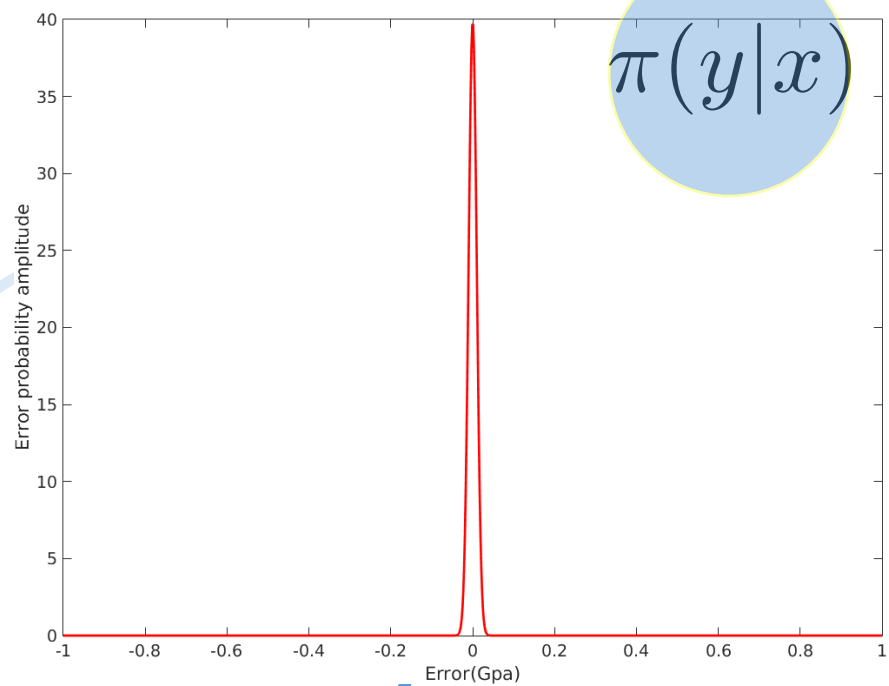


Posterior probability



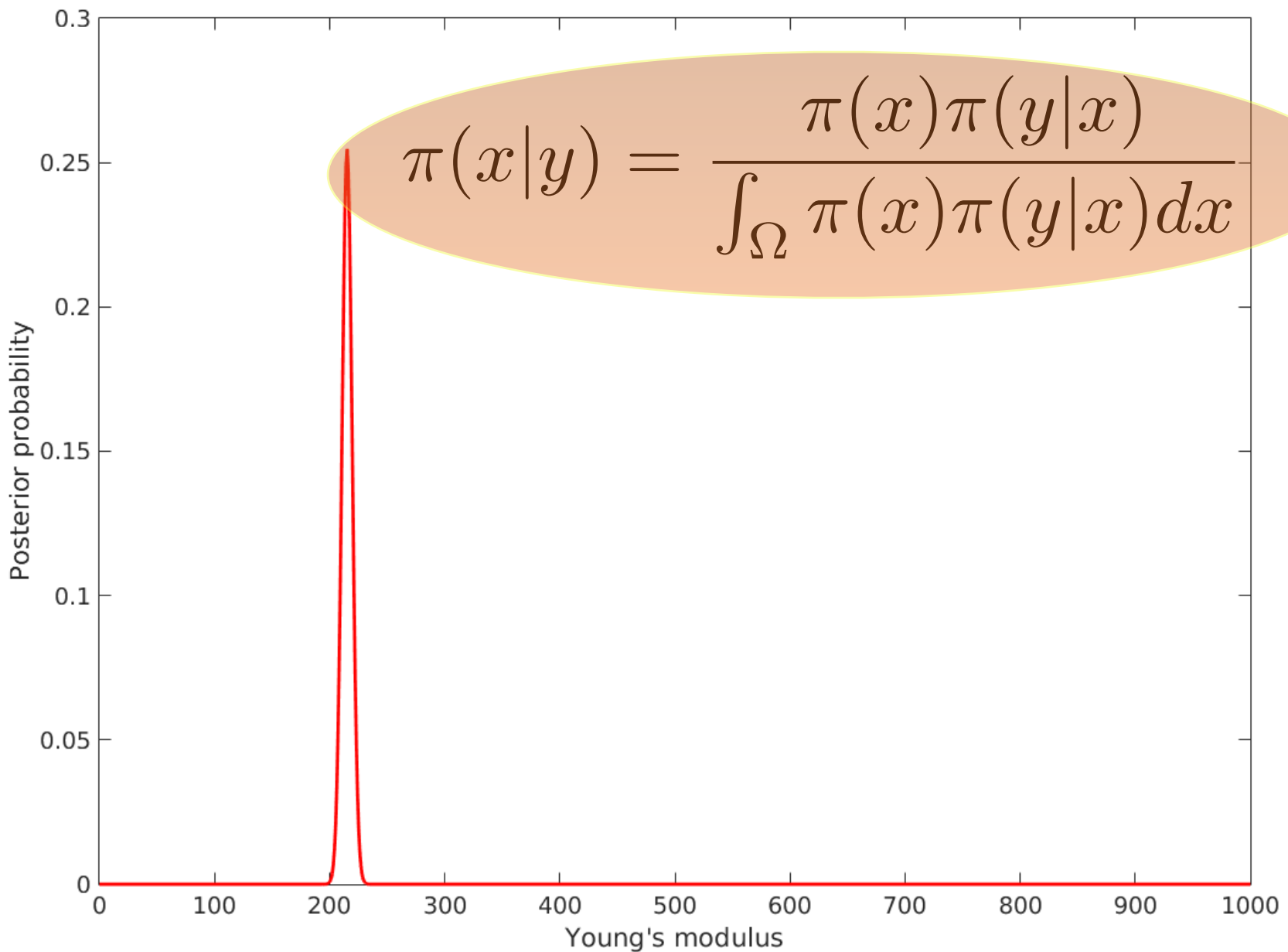
$\pi(x)$

$$\pi_{prior}(x) = N(210, 900)$$



$\pi(y|x)$

$$\pi(e)_{error} = N(0, 0.0001)$$



$$\pi(x|y) = \frac{\pi(x)\pi(y|x)}{\int_{\Omega} \pi(x)\pi(y|x)dx}$$

$$\pi_{posterior} = N(215.1533, 19.6168)$$

$$N_{sample} = 10$$

Outline

A focus on equation-based models

- How can we control the quality of simulations, verification and validation?
- Why are set-in-stone-models limited?
- How can we leverage statistical models to improve our models?

Outline

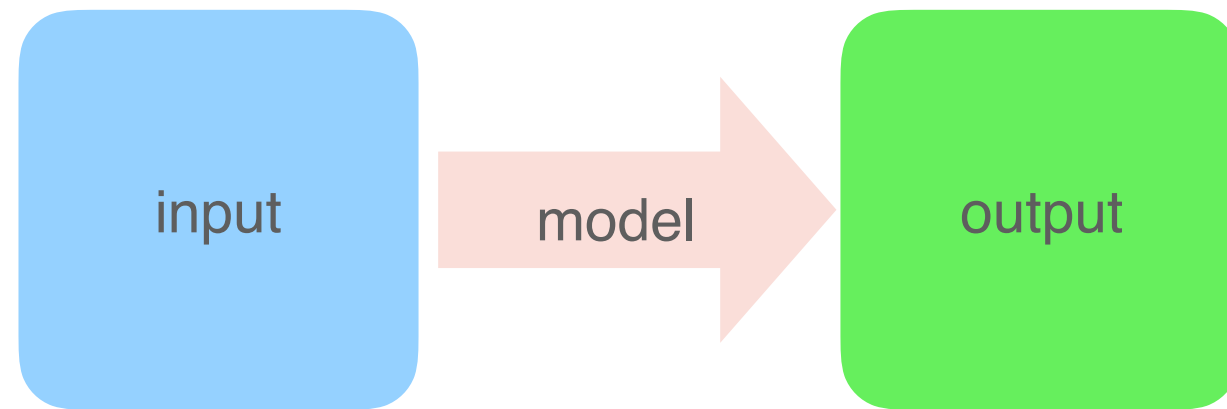
- **Data-driven modelling: Beyond setting models in stone**

- Data assimilation

- How can we learn from observations “on-the-fly”.
- The power of digital twins.

- **Future challenges**

Data-driven Modelling



$$f : \mathbf{x} \rightarrow \mathbf{y}$$

The structure of f is known but its parameters are not.

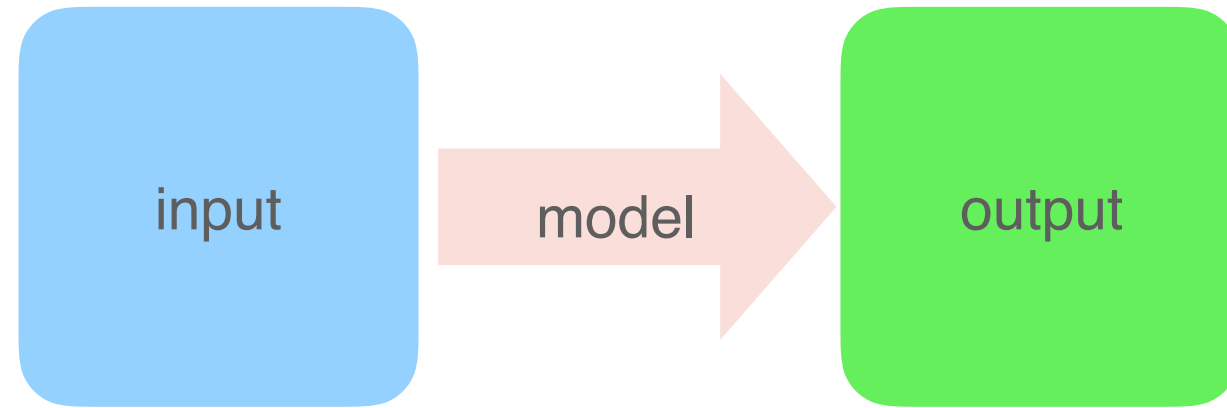
There is no a priori knowledge about the function f available.

model calibration

model identification

Embrace the conceptual shift from *"model through data abstraction"* to *"data is the model"*.

Data-driven Modelling



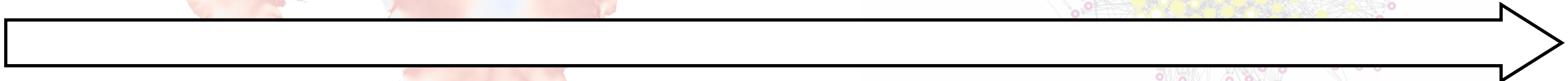
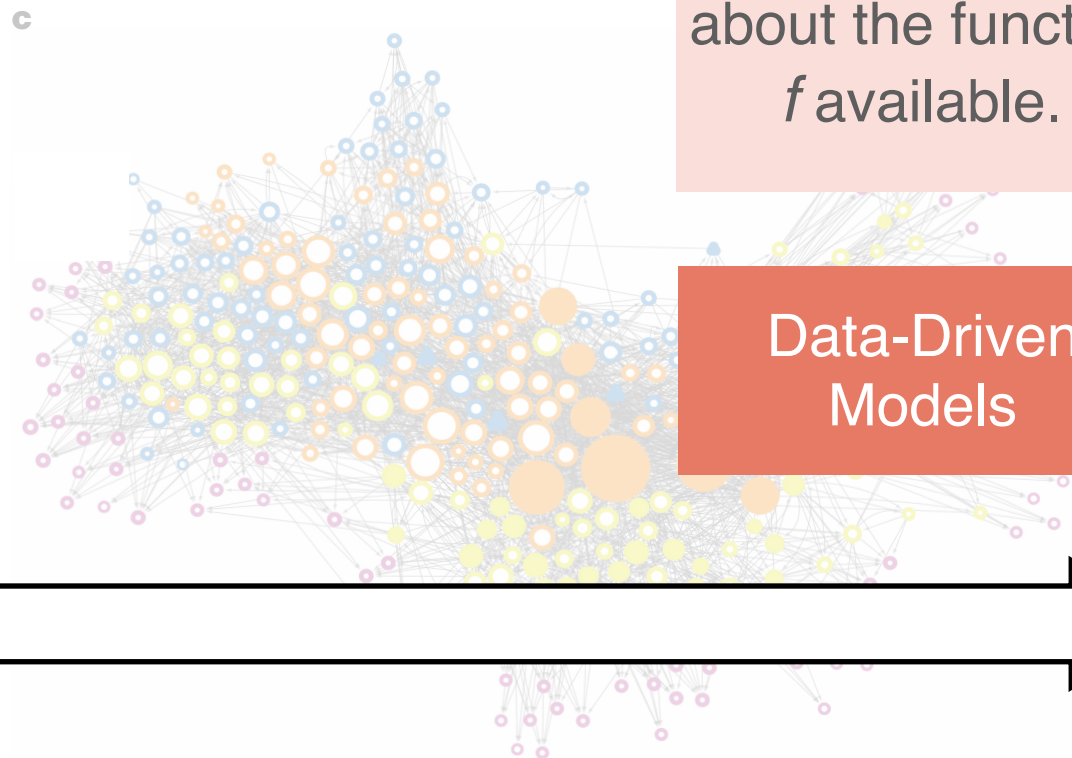
$$f : \mathbf{x} \rightarrow \mathbf{y}$$

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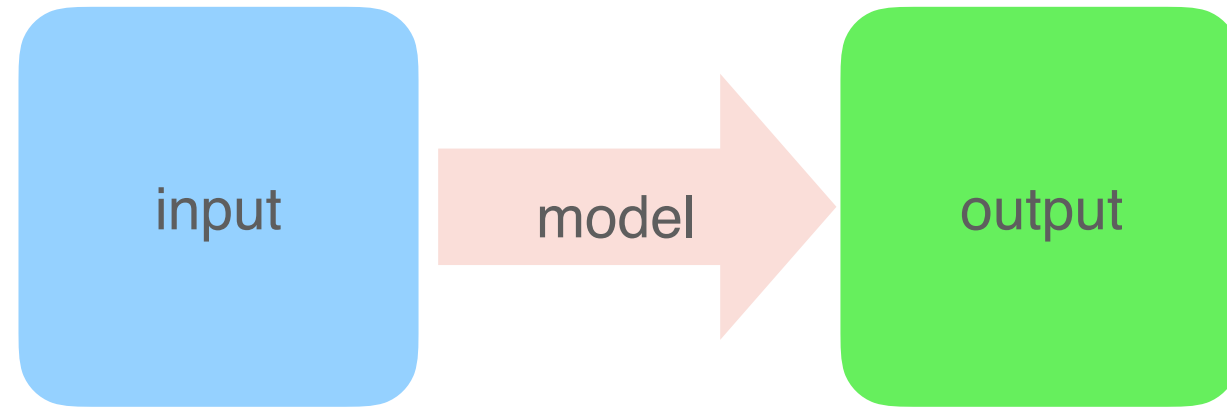
Hypothesis-based Models

There is no a priori knowledge about the function f available.

Data-Driven Models



Data-driven Modelling



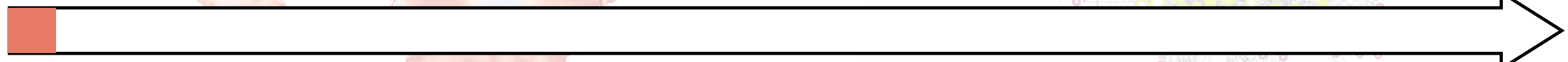
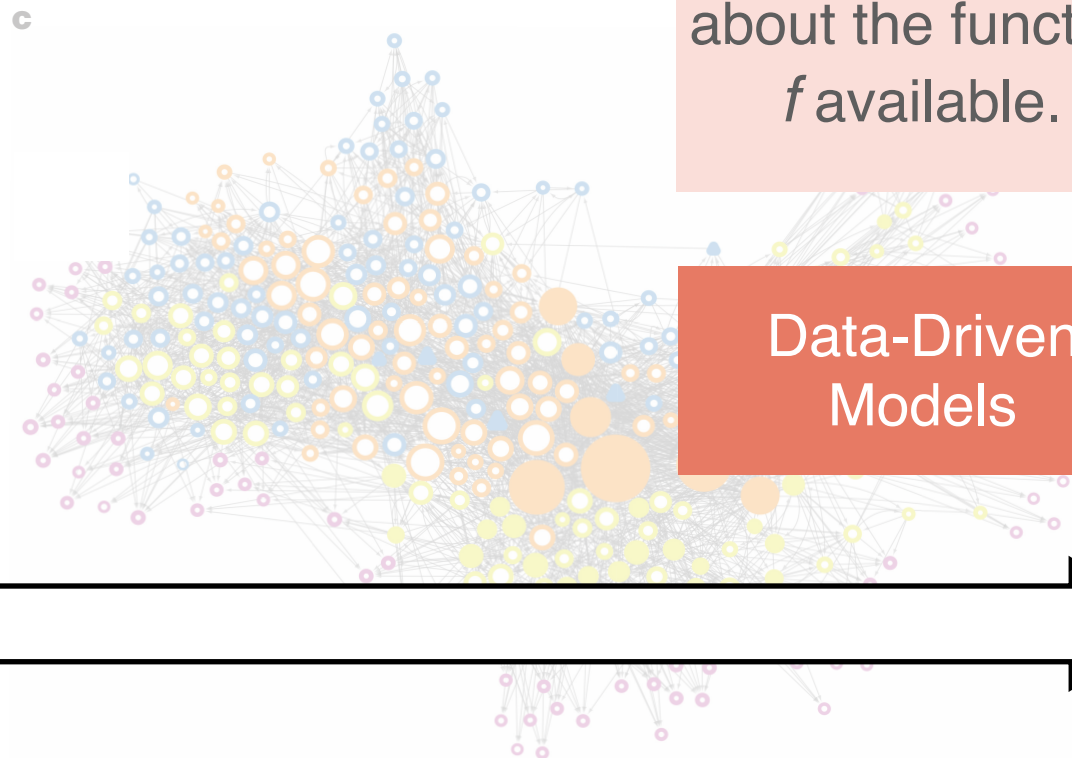
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Hypothesis-based Models

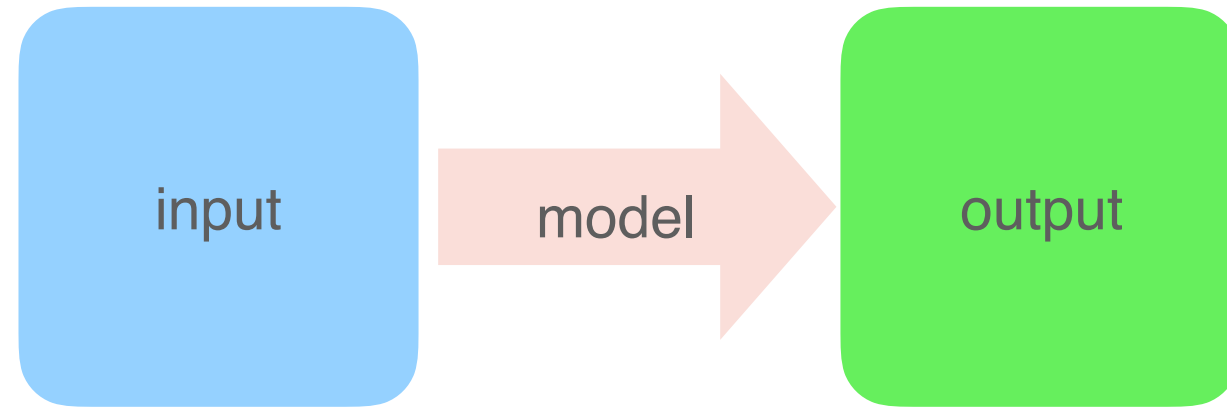
There is no a priori knowledge about the function f available.

Data-Driven Models



Small data

Data-driven Modelling



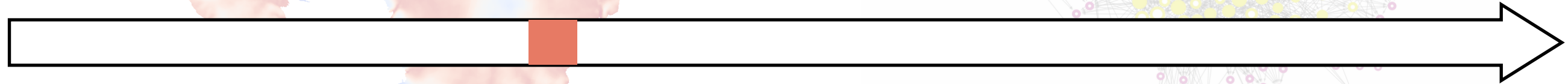
$$f : \mathbf{x} \rightarrow \mathbf{y}$$

The structure of f is known but its parameters are not.

Hypothesis-based Models

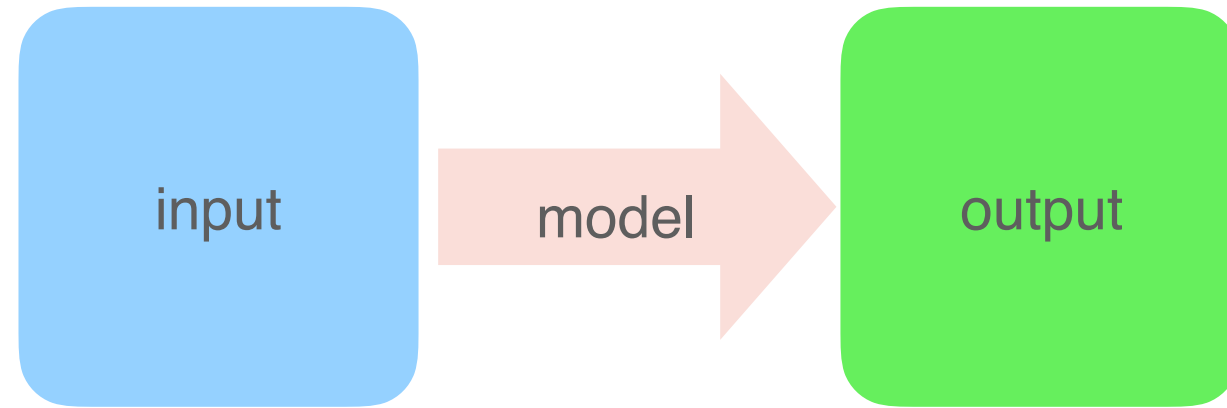
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Data-Driven Models



Small data

Data-driven Modelling



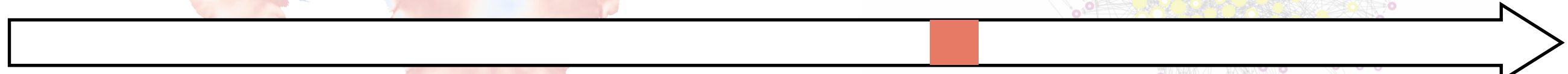
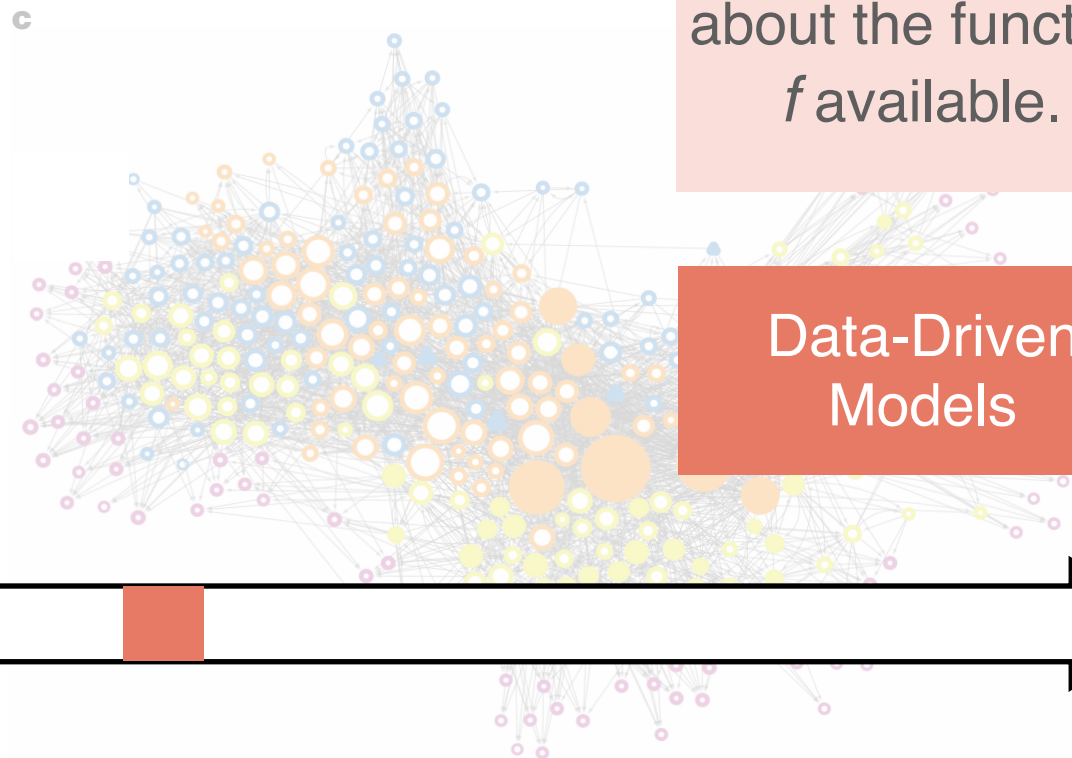
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Hypothesis-based Models

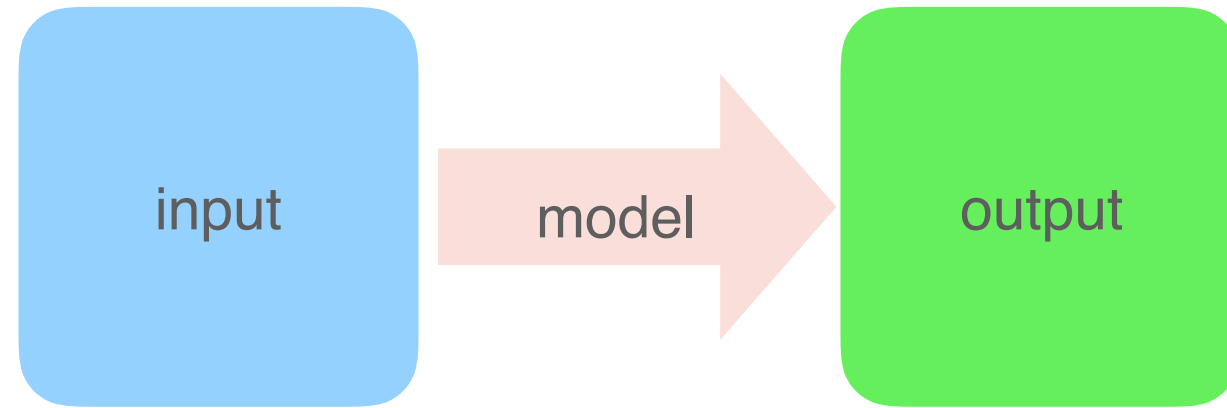
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Data-Driven Models



Small data

Data-driven Modelling



$$f : \mathbf{x} \rightarrow \mathbf{y}$$

The structure of f is known but its parameters are not.

Hypothesis-based Models

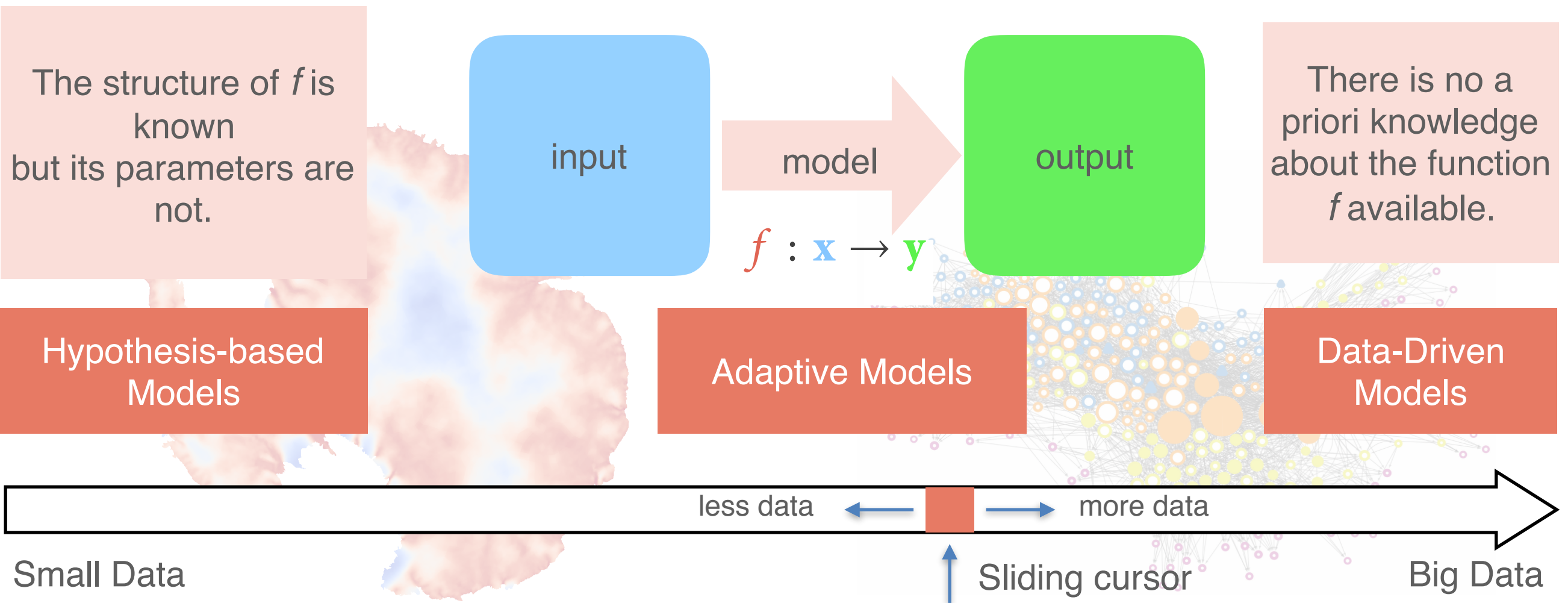
There is no a priori knowledge about the function f available.

Data-Driven Models

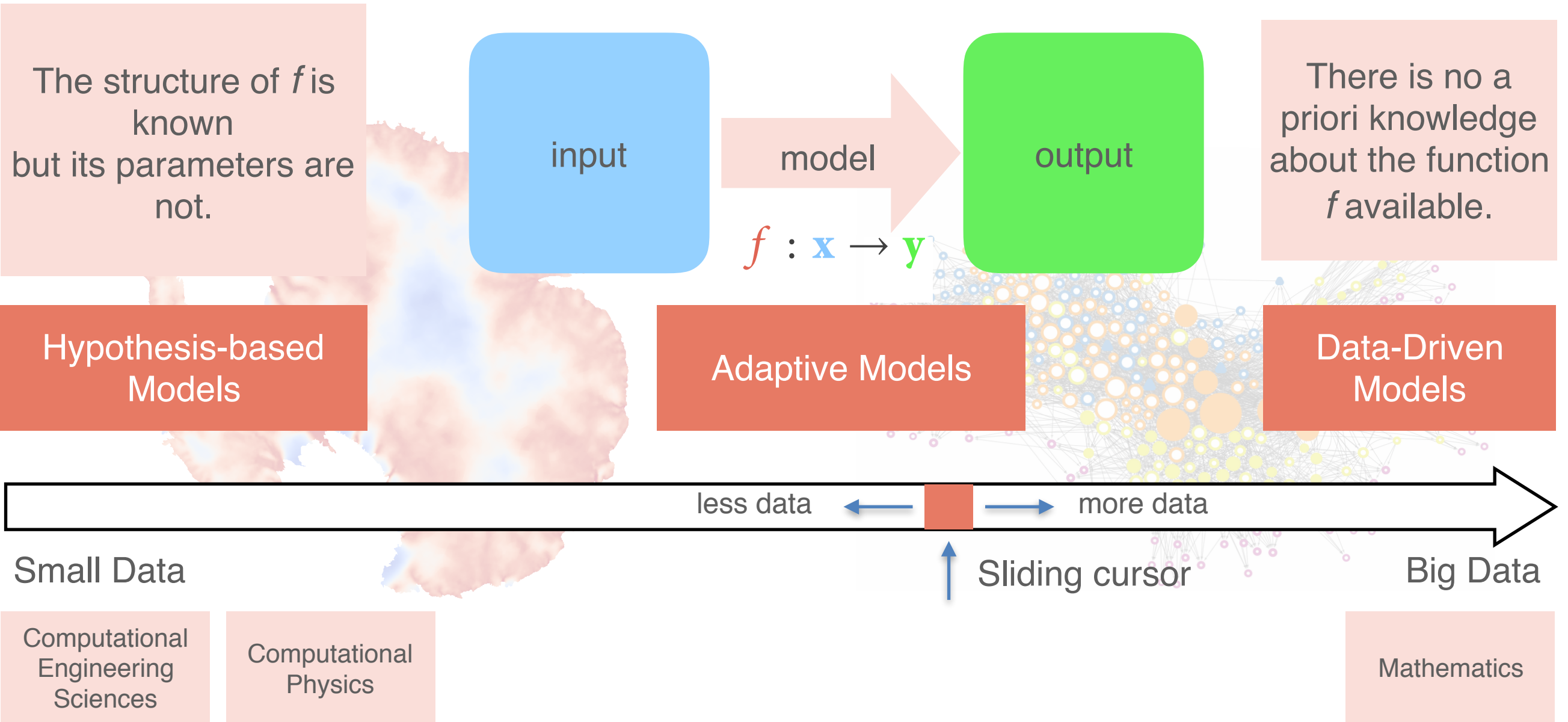
Small data

Big data

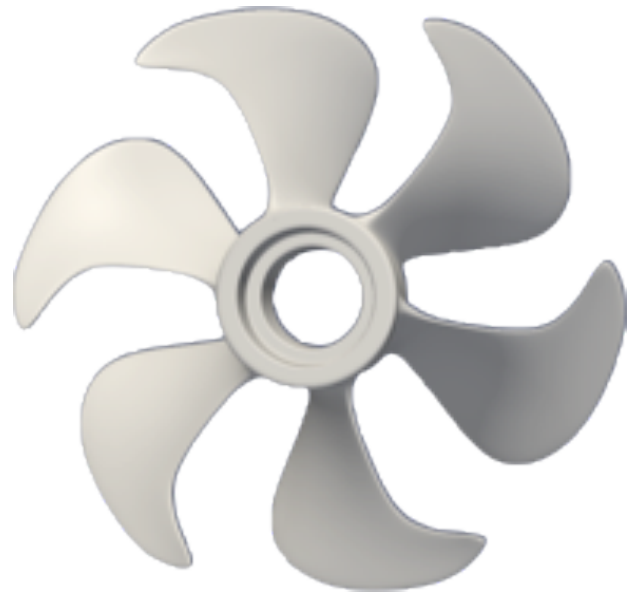
Model Discovery



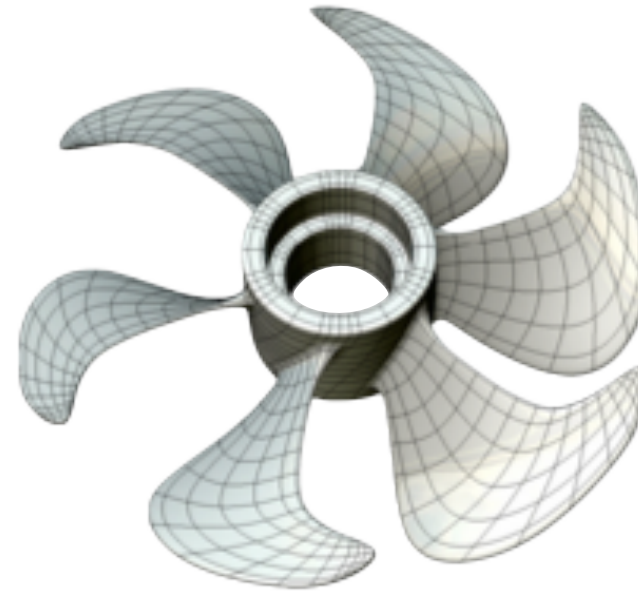
Model Discovery



GEOMETRY & BCs



DISCRETISATION

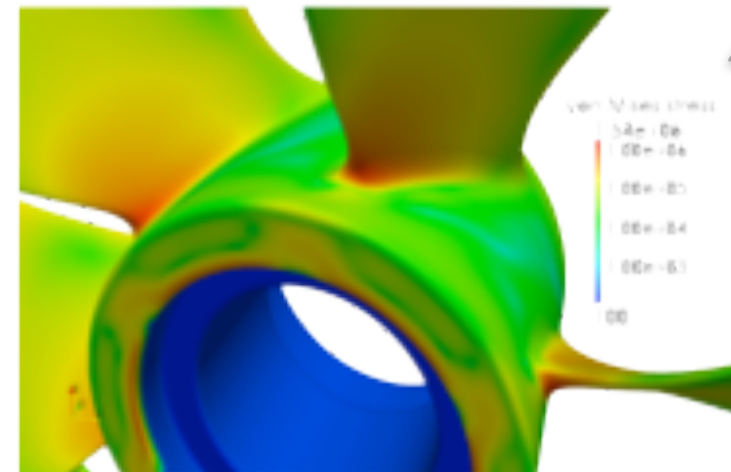


MODELS

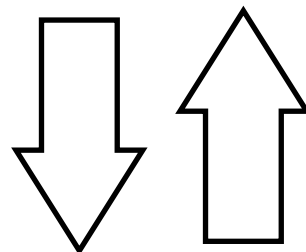
Elasticity/Plasticity, Crack growth law, Fracture energy, Maximum tensile strength, Multi-scale, Debonding, Fibre pull-out, Fibre breakage, interface fracture, grains, dislocations, MD, quantum...

VERIFICATION QUALITY CONTROL

NUMERICAL SOLUTION



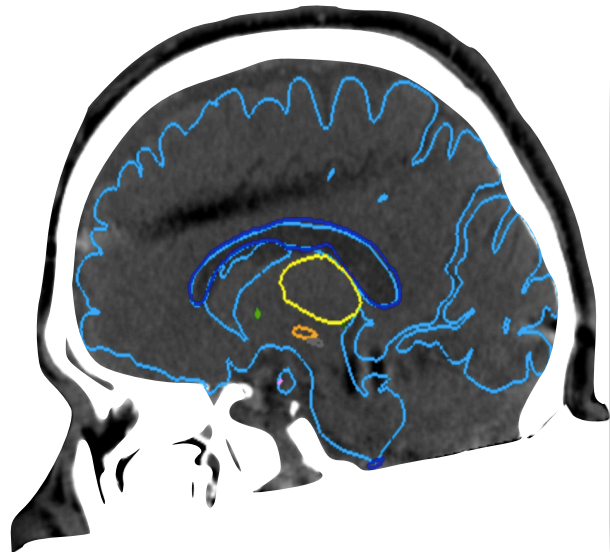
Validation



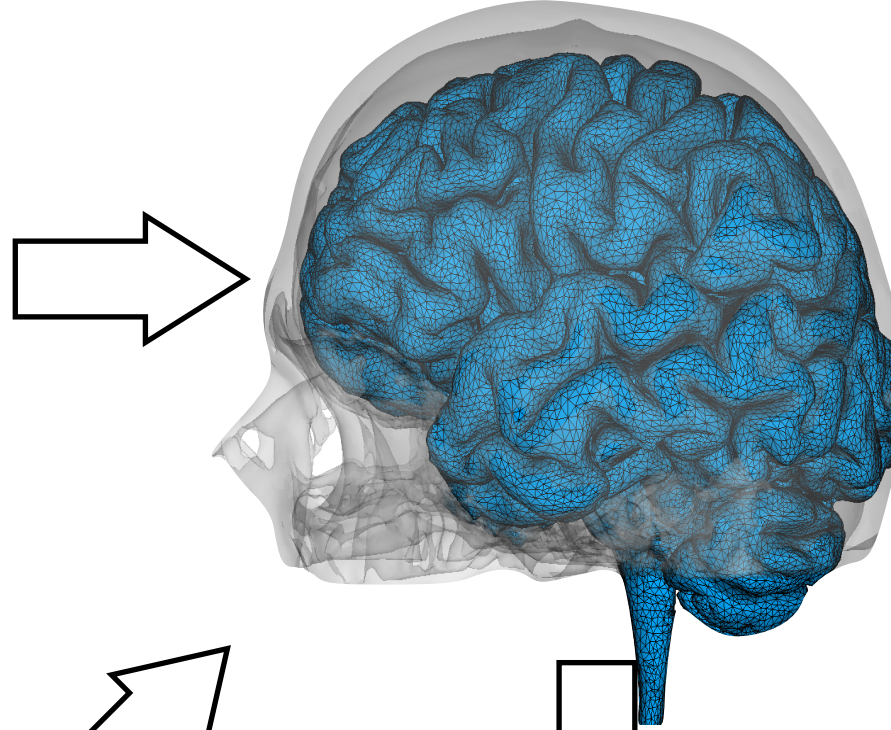
Parameter
identification

EXPERIMENTS

IMAGE/MODEL/BCs



DISCRETISATION



MODELS

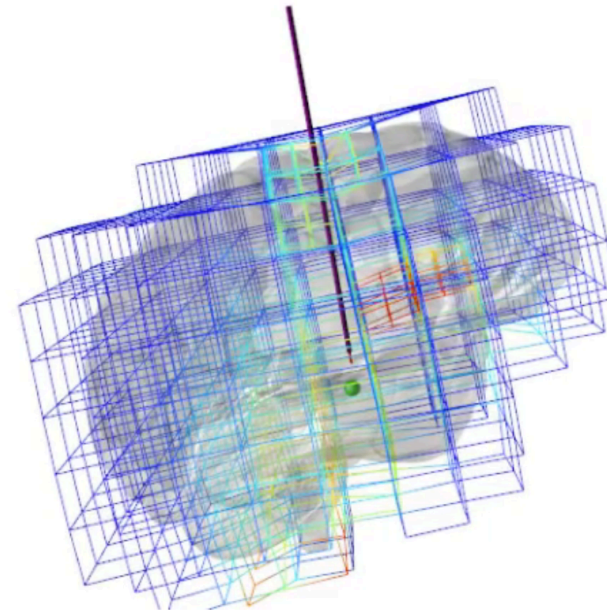
Phenomenological
Neo-Hookean, Ogden, ...

Multi-scale
cutting, fracture,

???

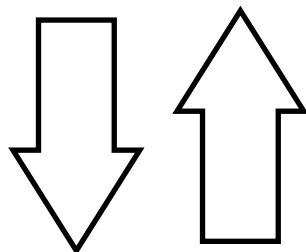
Patient specific ???

NUMERICAL SOLUTION



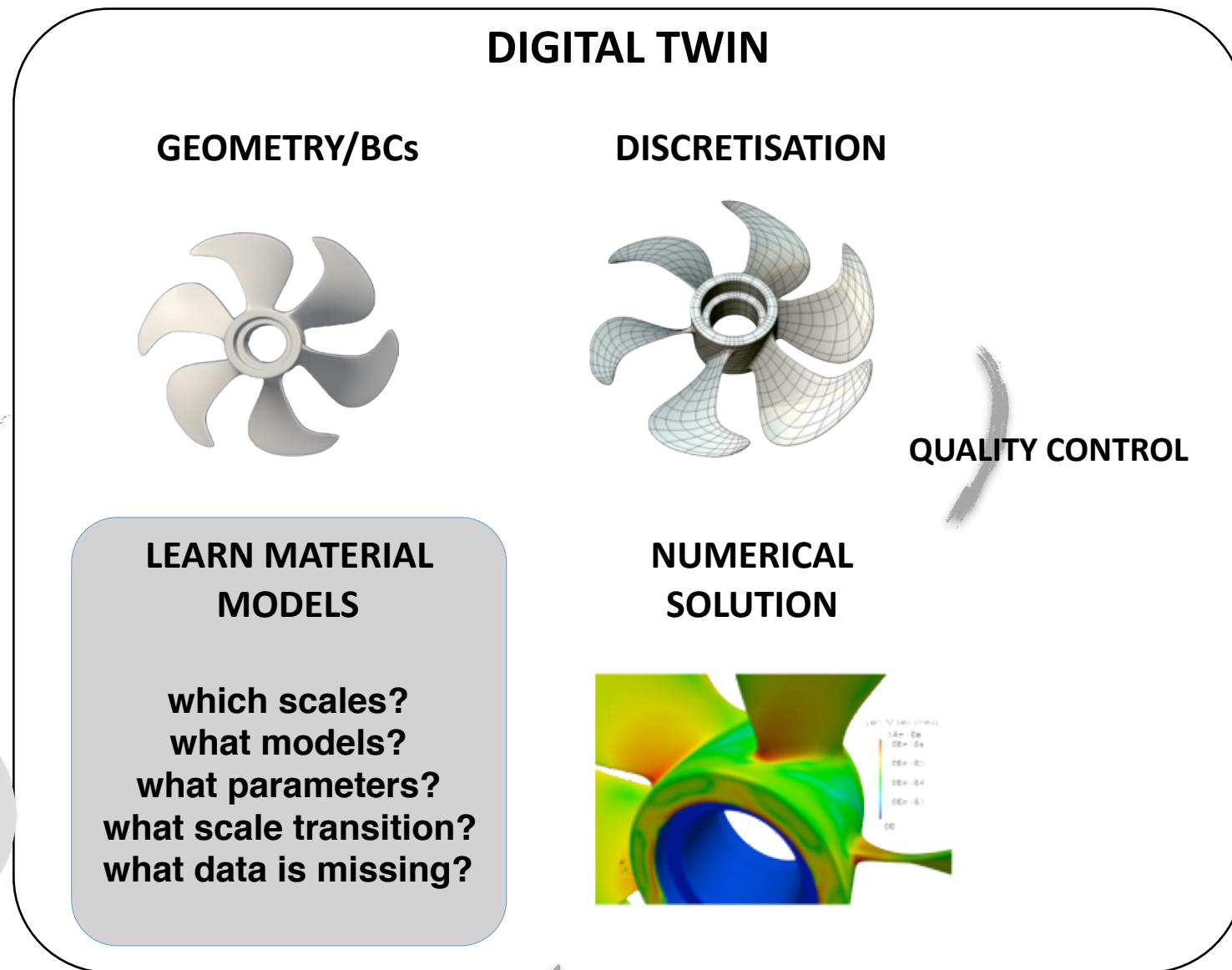
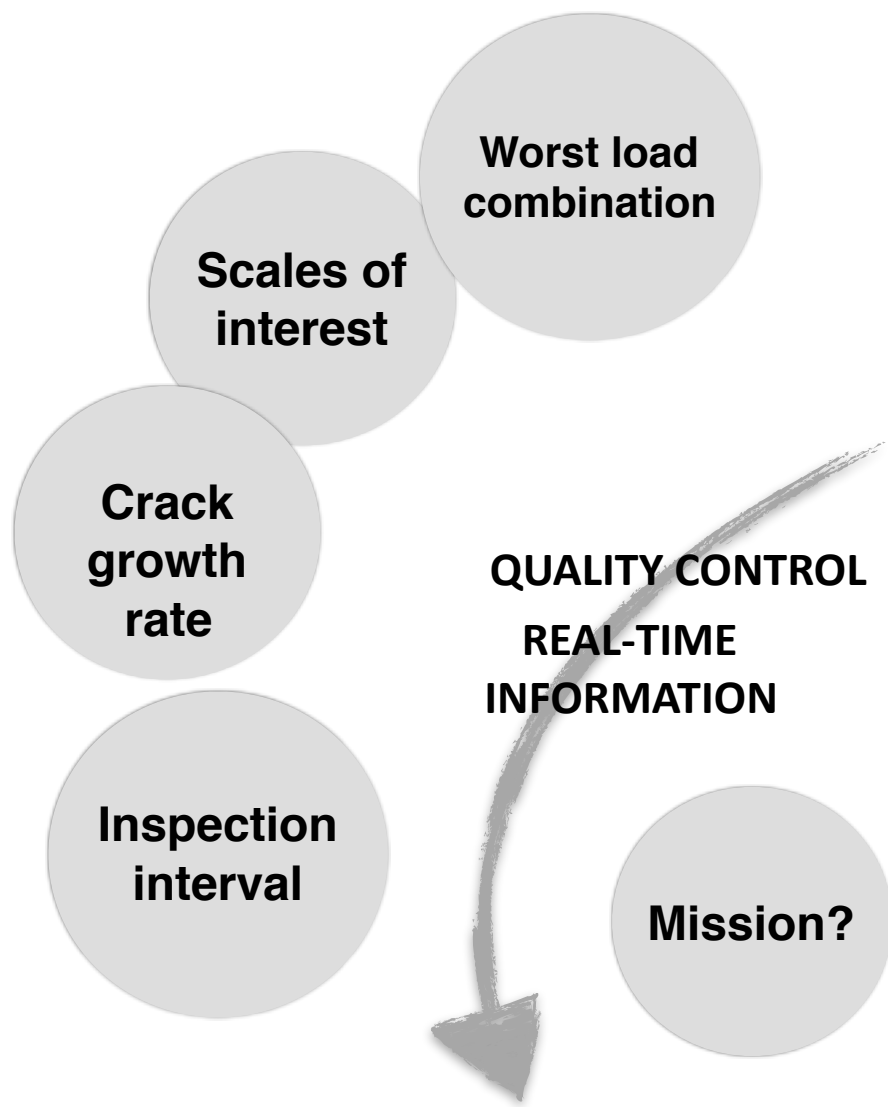
VERIFICATION
QUALITY CONTROL

Validation

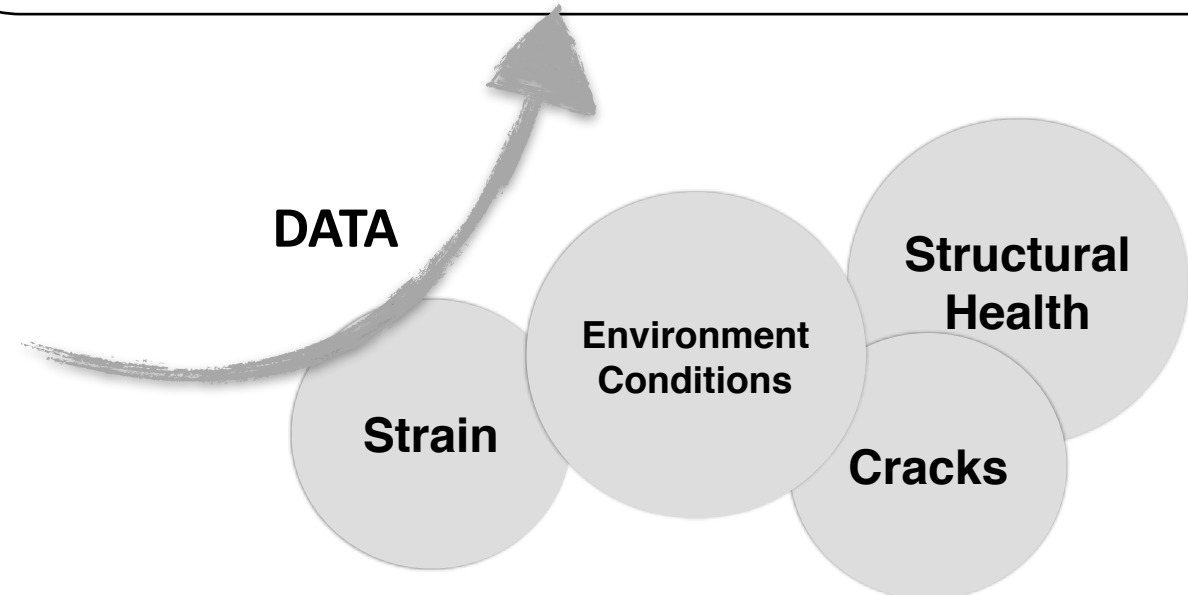
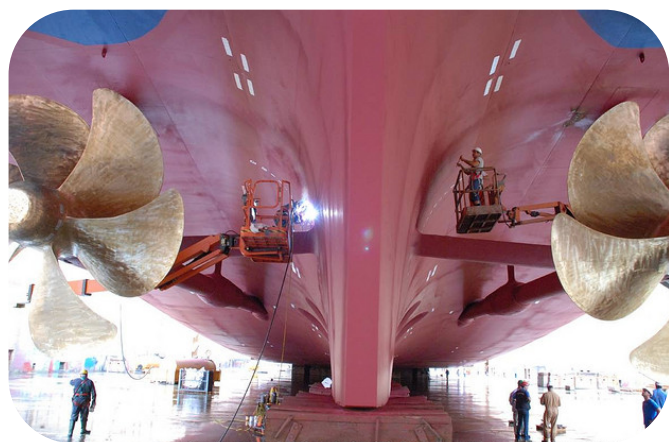


Parameter
identification

EXPERIMENTS



REAL SYSTEM



DIGITAL TWIN OF THE PATIENT



Alex Garland, *Ex Machina*, 2015

Treatment simulation

Scales of interest

Disease evolution

“Inspection” interval

QUALITY CONTROL
REAL-TIME
INFORMATION

Fitness

REAL PATIENT



DATA

Environment
Conditions

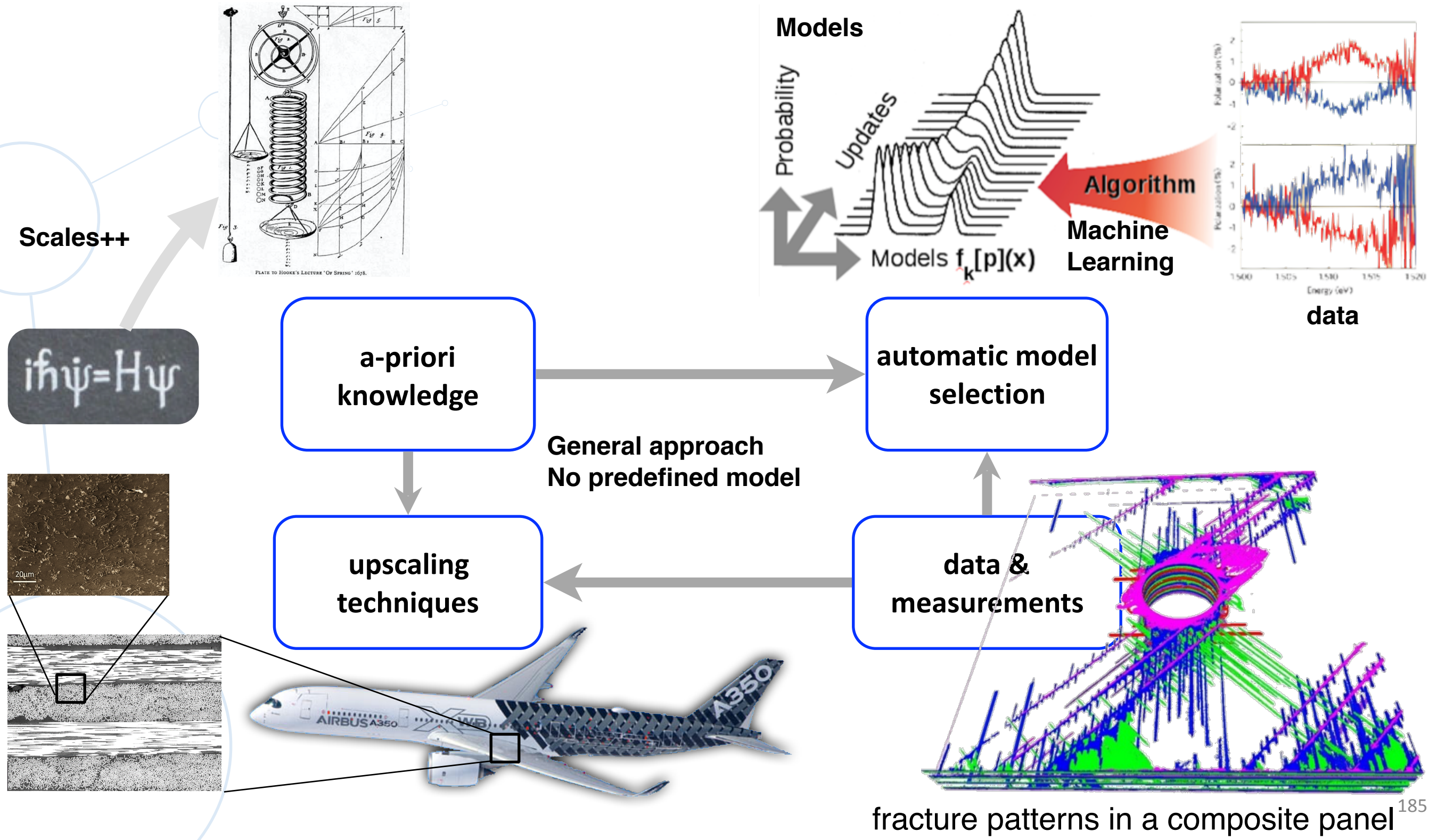
Organ
state

Disease

Health



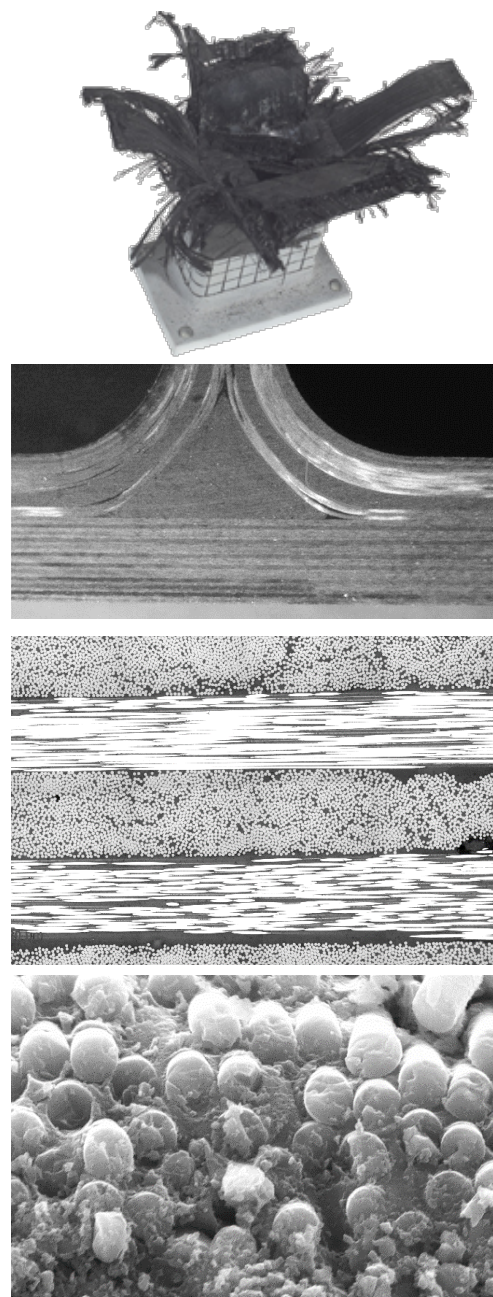
VISION



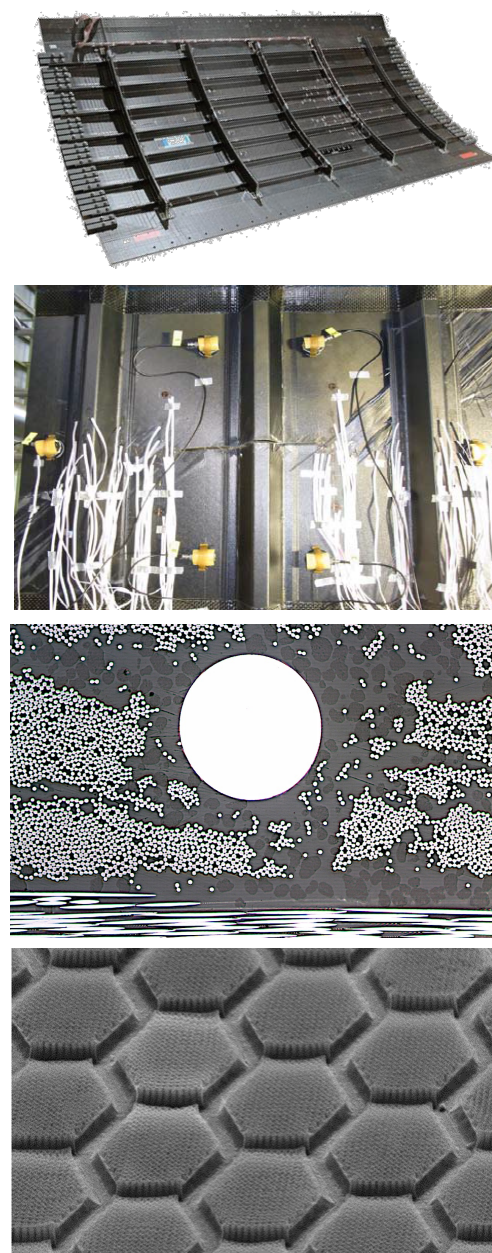


Digital Twins...

Characterisation



Monitoring



Multiscale models are unreliable

Quantitative predictions ?

Learn better models

Fracture/lack of scale separation

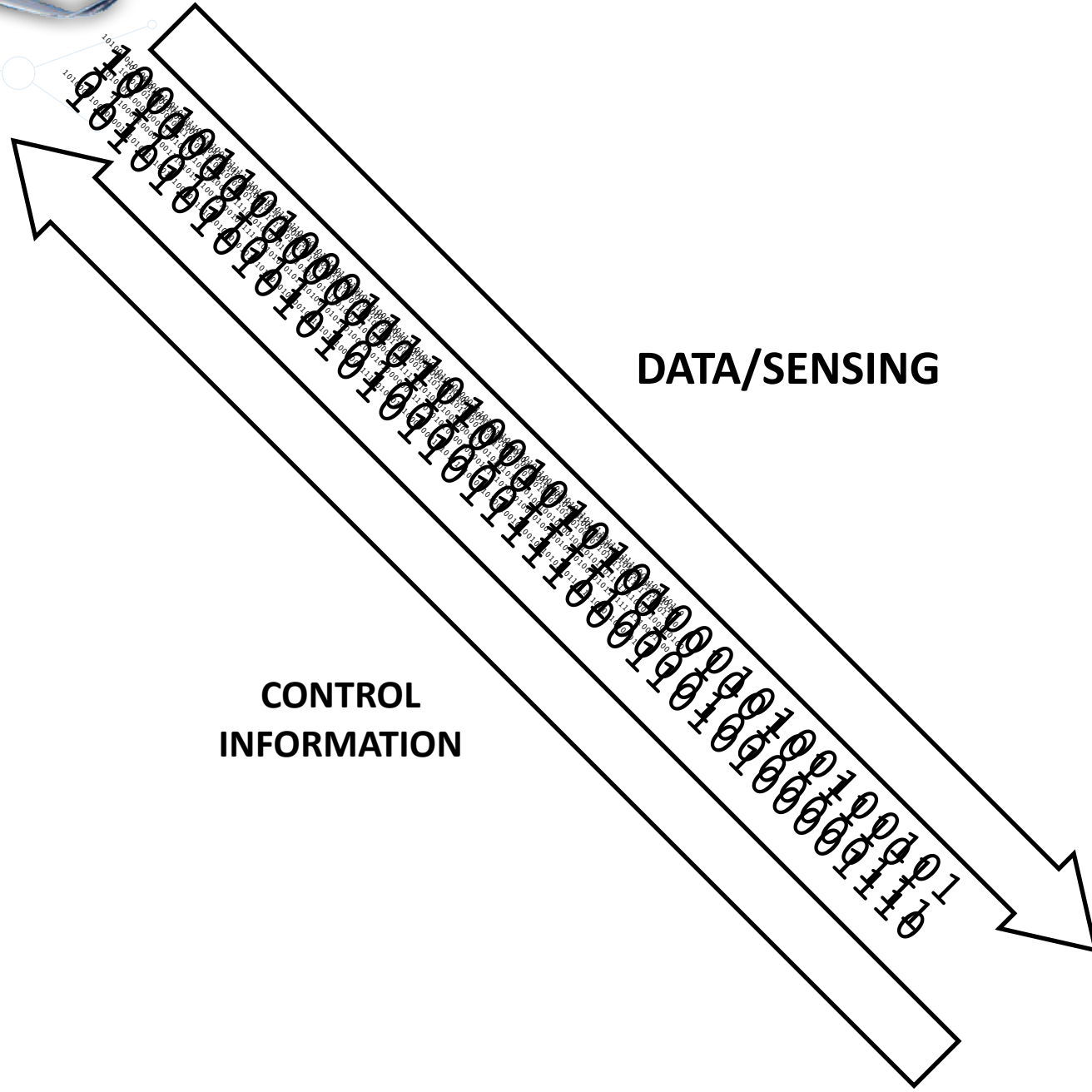
Measure

Analyse & Learn from **data**

Improved model

Identify missing data

Validate







Prior Knowledge



Prior Knowledge



Hypothesis

Domain
expert



Prior Knowledge



(Big) Data



Hypothesis

Domain
expert

Prior Knowledge

**Computational
Science**

HPC

(Big) Data

Hypothesis

Domain
expert

Prior Knowledge

**Computational
Science**

HPC

Conclusions

(Big) Data

Hypothesis

Domain
expert

Prior Knowledge



Data-driven modelling

(Big) Data

Hypothesis

Domain expert

Conclusions

Patient-specific material models

1. Prior

Prior knowledge

Material parameters inc. distribution from general population

Noise

- ▶ Model (e.g. additive)
- ▶ Distribution (Gauss)
- ▶ Characterisation

Constitutive model

$$\psi^{\text{eq}}(\mathbf{F}) = \frac{\mu}{2} \left(\frac{\text{tr} \mathbf{B}}{J^{2/3}} - 3 \right) + \frac{1}{2} K (J - 1)^2$$

Patient-specific material models

1. Prior

Prior knowledge

Material parameters inc. distribution from general population

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Constitutive model

$$\psi^{\text{eq}}(\mathbf{F}) = \frac{\mu}{2} \left(\frac{(\text{tr} \mathbf{B})}{J^{2/3}} - 3 \right) + \frac{1}{2} K (J - 1)^2$$

Data assimilator Bayesian inference



$$\Pr(M|D) = \frac{\Pr(D|M) \Pr(M)}{\Pr(D)}$$

Patient-specific material models

1. Prior

Prior knowledge
Material parameters inc. distribution from general population

- Noise**
- ▶ Model (e.g. additive)
 - ▶ Distribution (Gauss)
 - ▶ Characterisation

Constitutive model

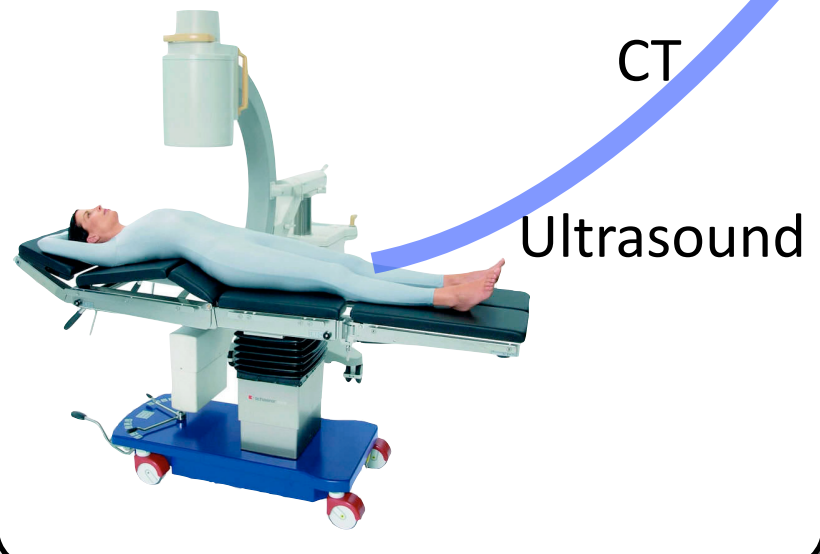
$$\psi^{eq}(\mathbf{F}) = \frac{\mu}{2} \left(\frac{(\text{tr} \mathbf{B})}{J^{2/3}} - 3 \right) + \frac{1}{2} K (J - 1)^2$$

Data assimilator Bayesian inference



$$\Pr(M|D) = \frac{\Pr(D|M) \Pr(M)}{\Pr(D)}$$

2. Posterior data



MRI
Stereo-cameras

Patient-specific material models

1. Prior

Prior knowledge
Material parameters inc. distribution from general population


Noise

- ▶ Model (e.g. additive)
- ▶ Distribution (Gauss)
- ▶ Characterisation

Constitutive model

$$\psi^{eq}(\mathbf{F}) = \frac{\mu}{2} \left(\frac{\text{tr} \mathbf{B}}{J^{2/3}} - 3 \right) + \frac{1}{2} K (J - 1)^2$$


2. Posterior data



CT

Ultrasound

Data assimilator
Bayesian inference

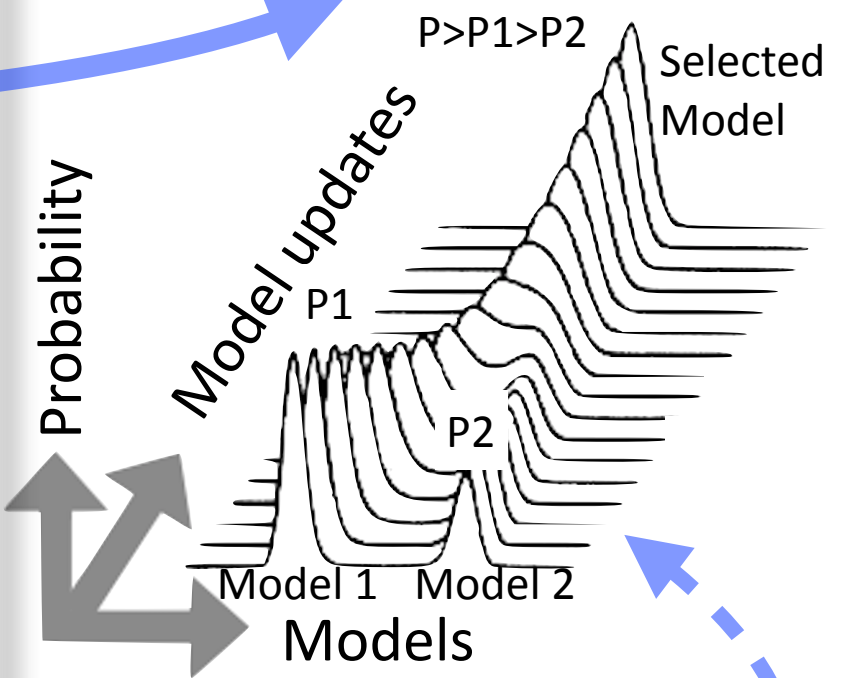


$$\Pr(M|D) = \frac{\Pr(D|M) \Pr(M)}{\Pr(D)}$$

MRI
Stereo-cameras

3. Simulator

Model selector



Probability

Model updates

P1

P2

Model 1

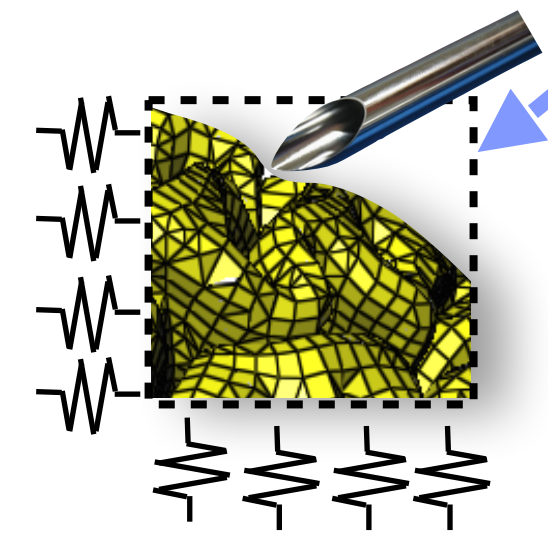
Model 2

Models

P > P1 > P2

Selected Model

Mechanical solver



Patient-specific material models

1. Prior

Prior knowledge
Material parameters inc. distribution from general population


Noise

- ▶ Model (e.g. additive)
- ▶ Distribution (Gauss)
- ▶ Characterisation

Constitutive model

$$\psi^{eq}(\mathbf{F}) = \frac{\mu}{2} \left(\frac{\text{tr} \mathbf{B}}{J^{2/3}} - 3 \right) + \frac{1}{2} K (J - 1)^2$$

Data assimilator
Bayesian inference



$$\Pr(M|D) = \frac{\Pr(D|M) \Pr(M)}{\Pr(D)}$$

3. Simulator

Model selector

Probability

Model updates

P1 P2


Model 1 Model 2 Models

P > P1 > P2 Selected Model

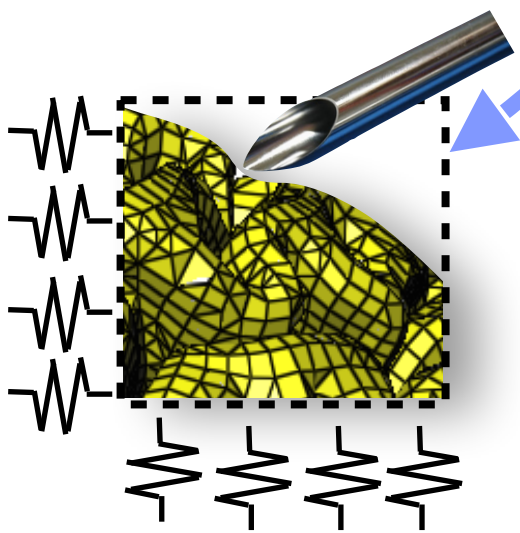
2. Posterior data

CT

Ultrasound




Mechanical solver



4. Feedback

Patient-specific material models

1. Prior

Prior knowledge
Material parameters inc. distribution from general population


Noise

- ▶ Model (e.g. additive)
- ▶ Distribution (Gauss)
- ▶ Characterisation

Constitutive model

$$\psi^{eq}(\mathbf{F}) = \frac{\mu}{2} \left(\frac{\text{tr} \mathbf{B}}{J^{2/3}} - 3 \right) + \frac{1}{2} K (J - 1)^2$$

Data assimilator
Bayesian inference



$$\Pr(M|D) = \frac{\Pr(D|M) \Pr(M)}{\Pr(D)}$$

3. Simulator

Model selector

Probability

Model updates

P1 P2


Model 1 Model 2 Models

P > P1 > P2 Selected Model

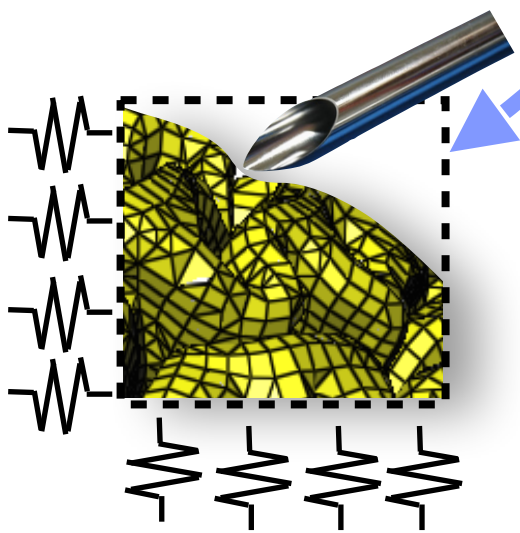
2. Posterior data

CT

Ultrasound




Mechanical solver

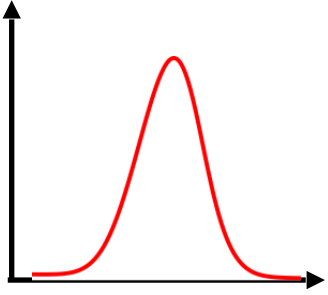


5. Action

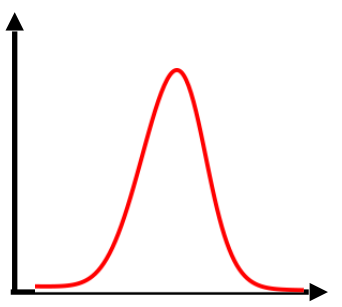
4. Feedback

Prior

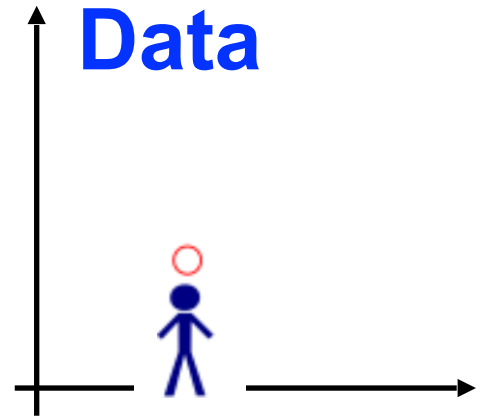
Knowledge



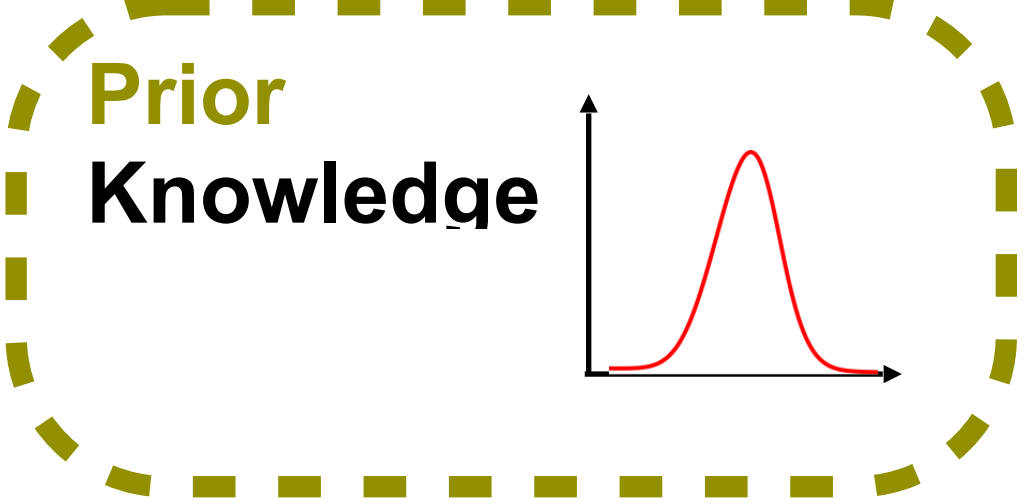
Prior Knowledge



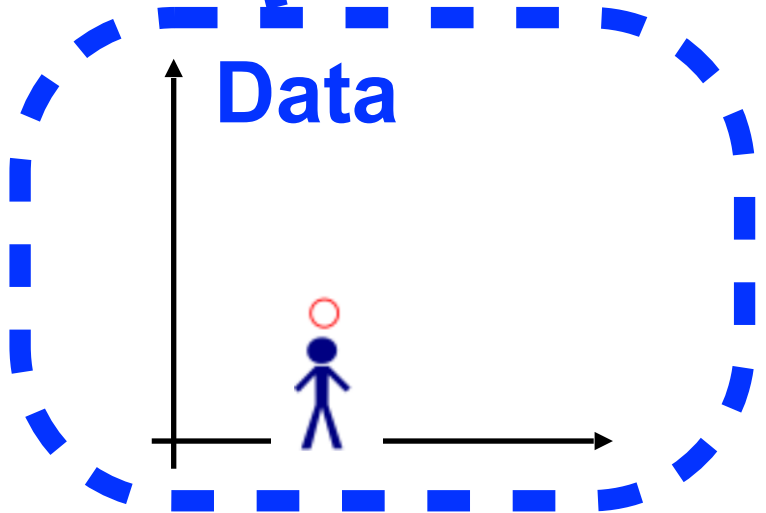
Data



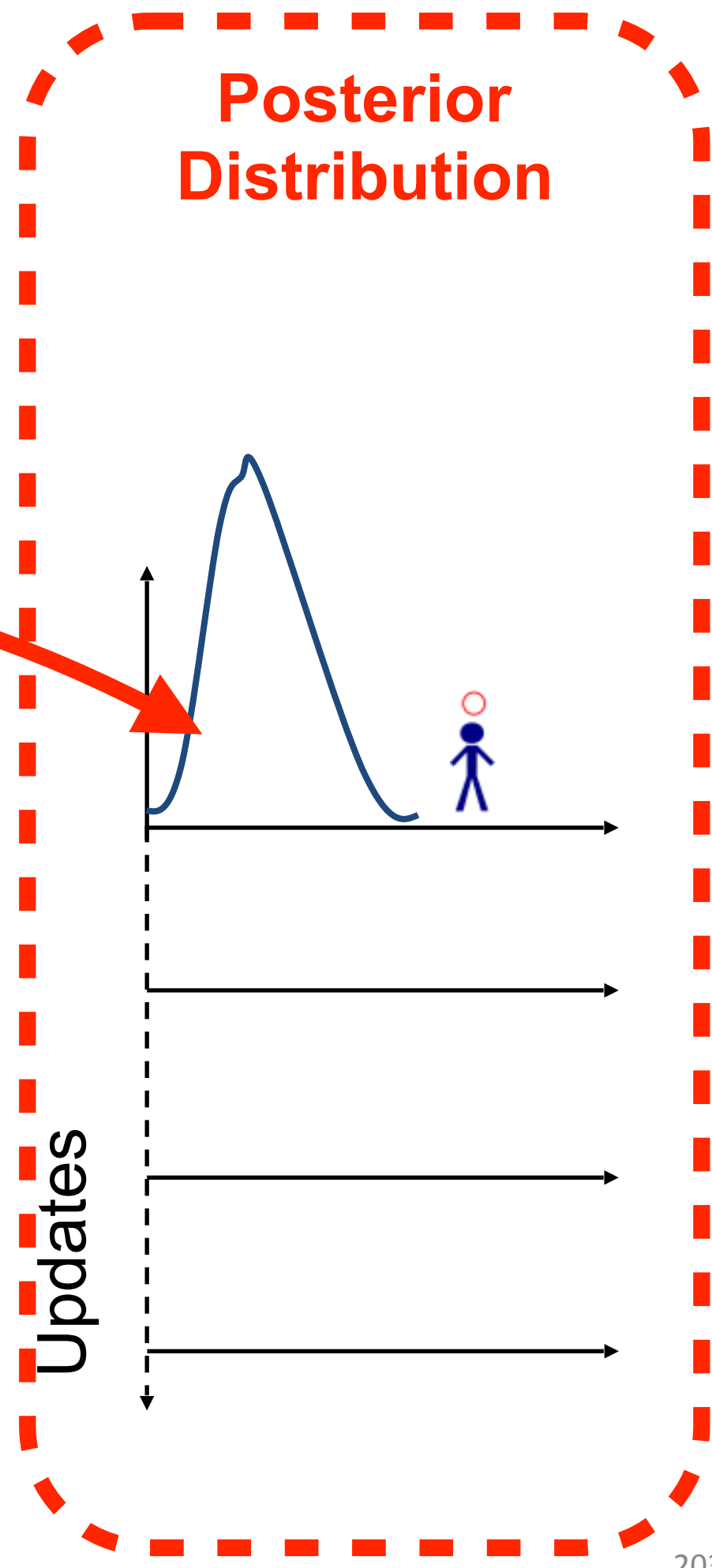
Hypothesis

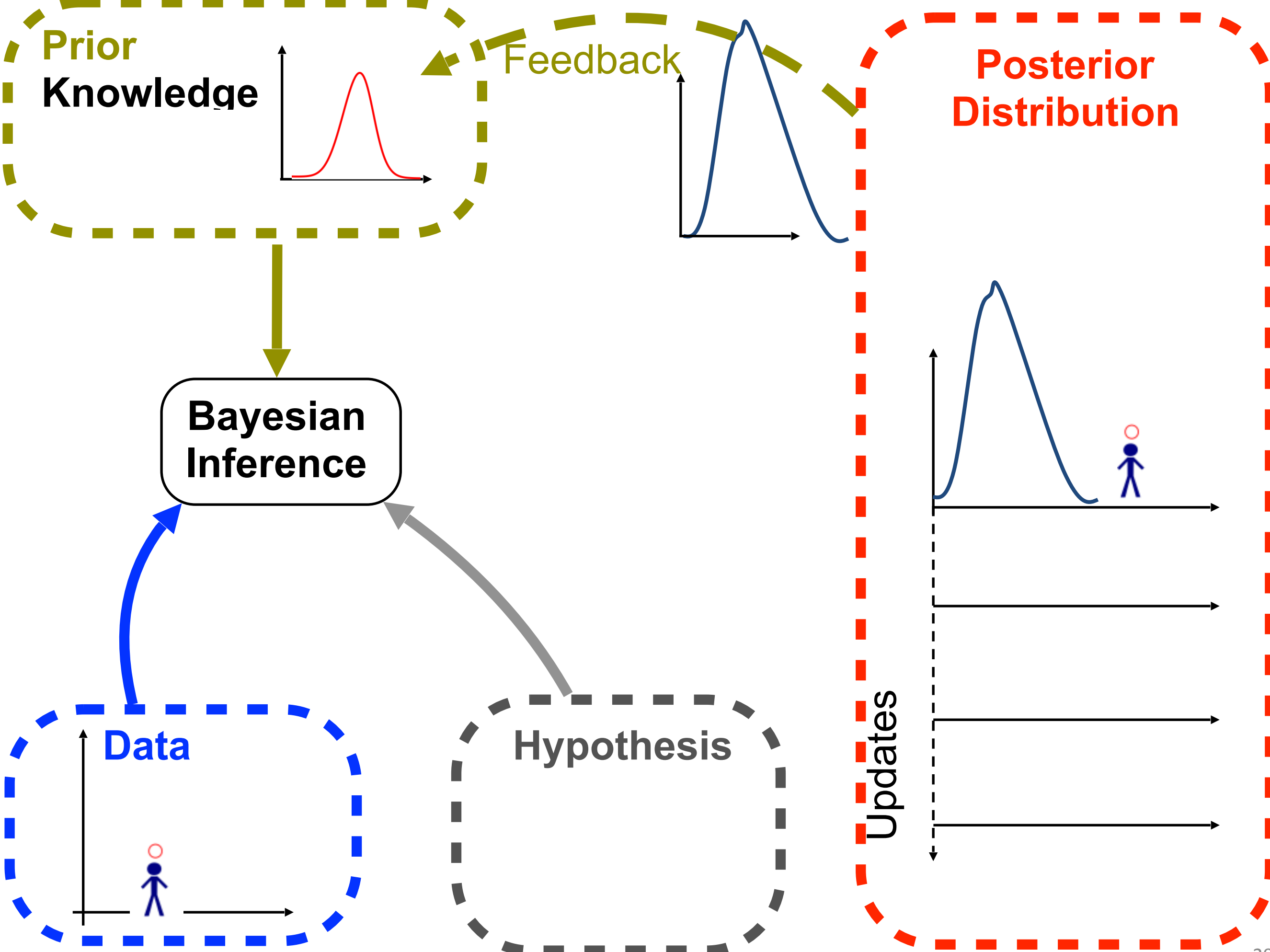


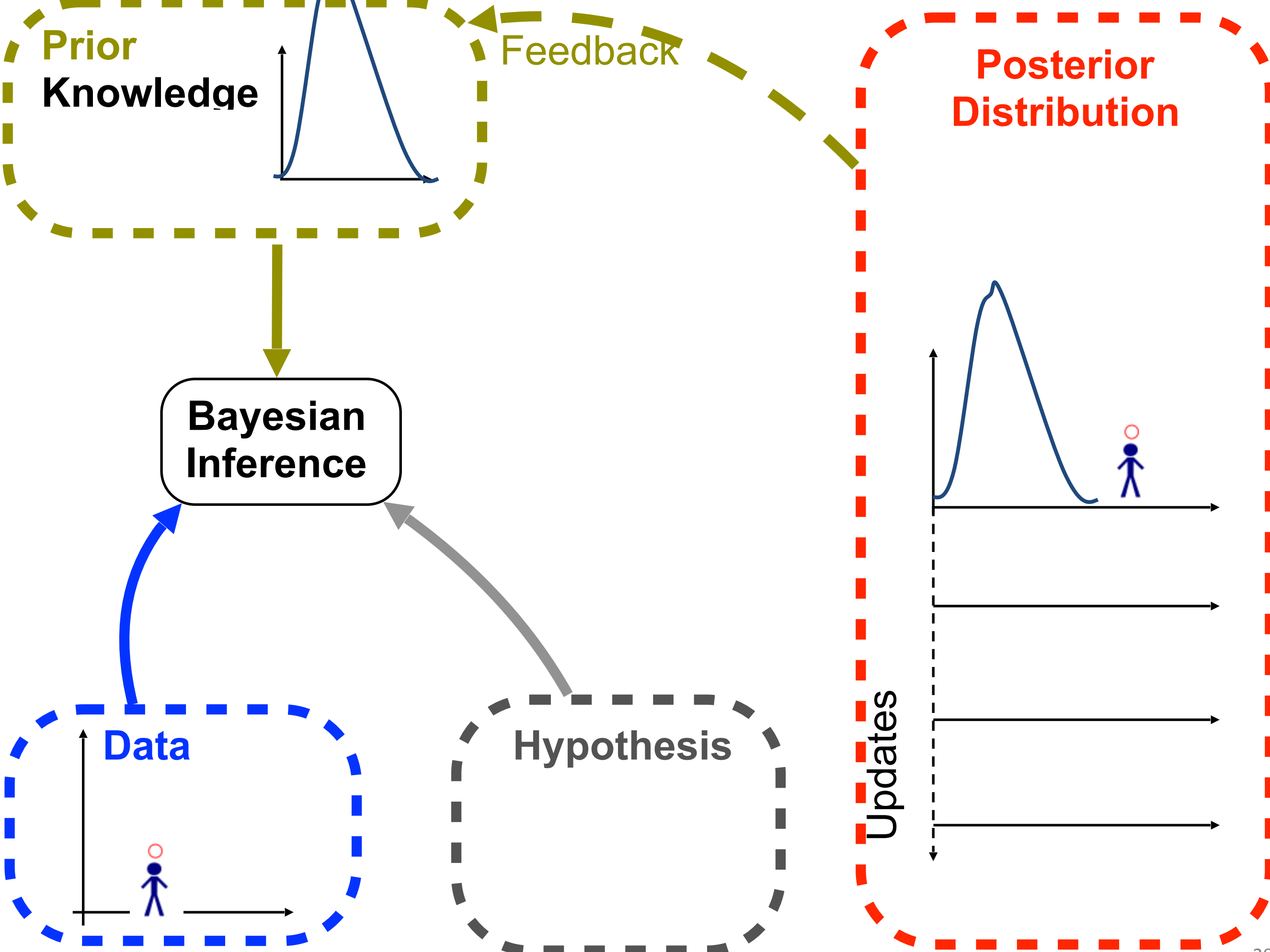
Bayesian Inference

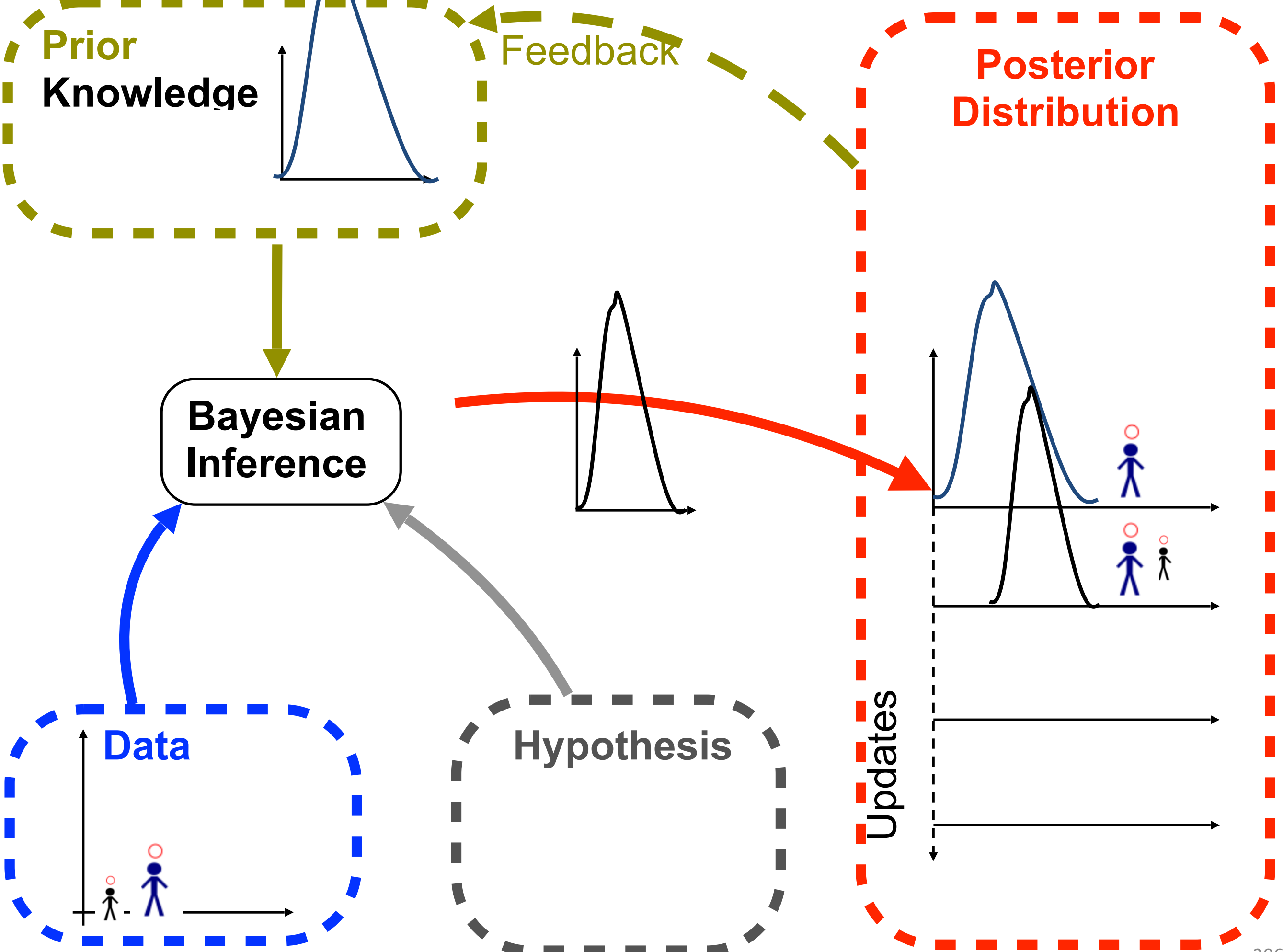


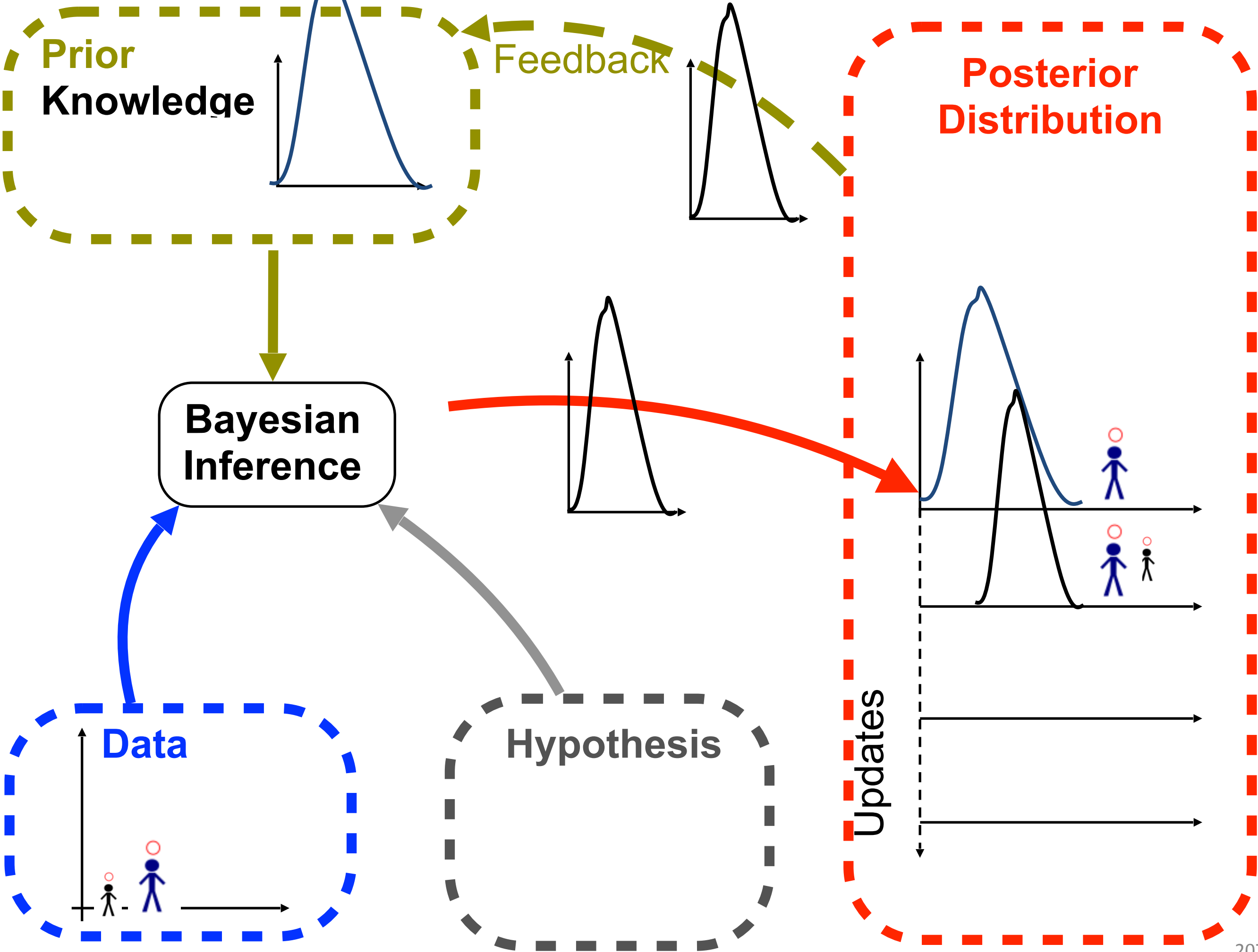
Hypothesis

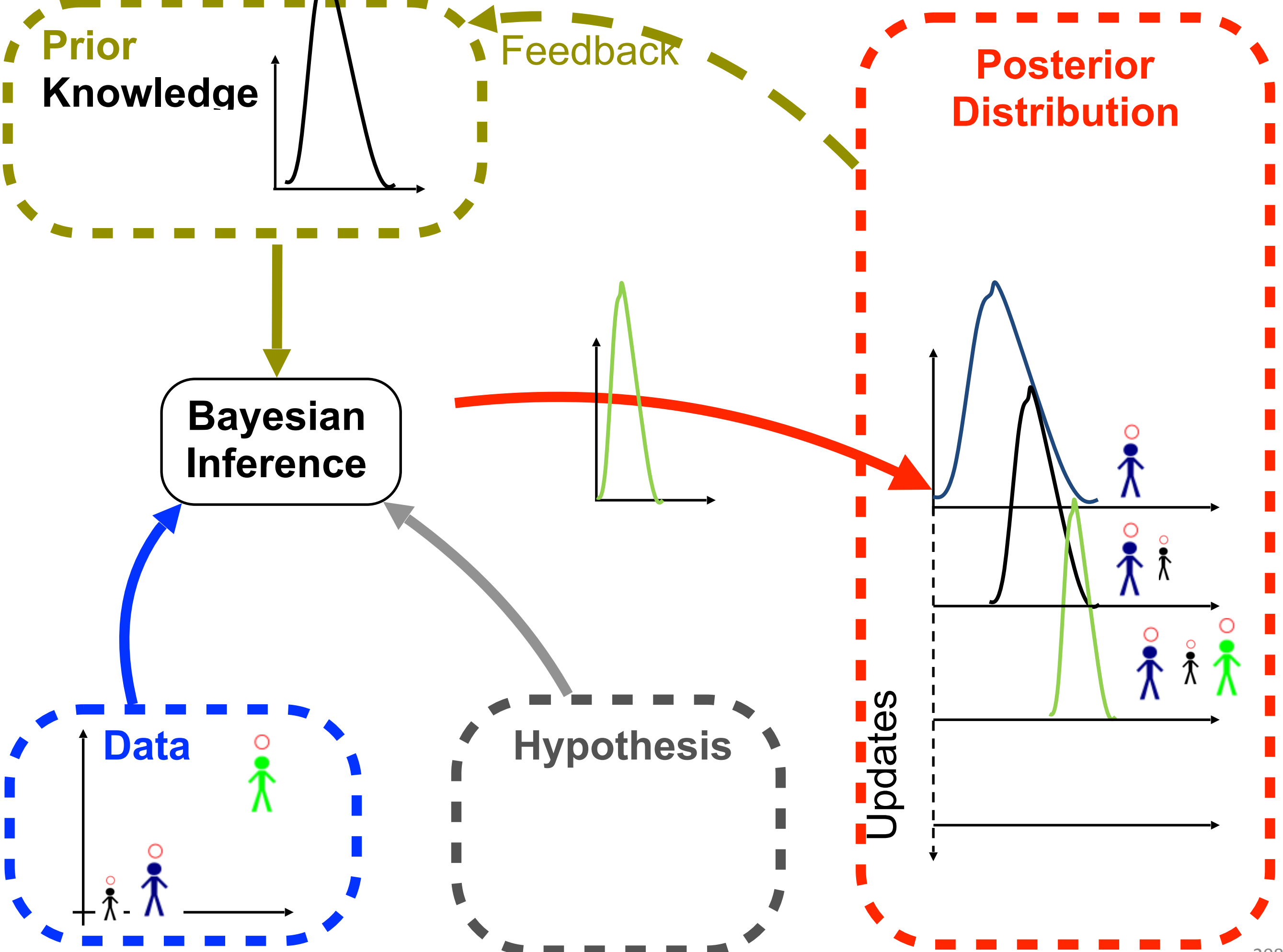


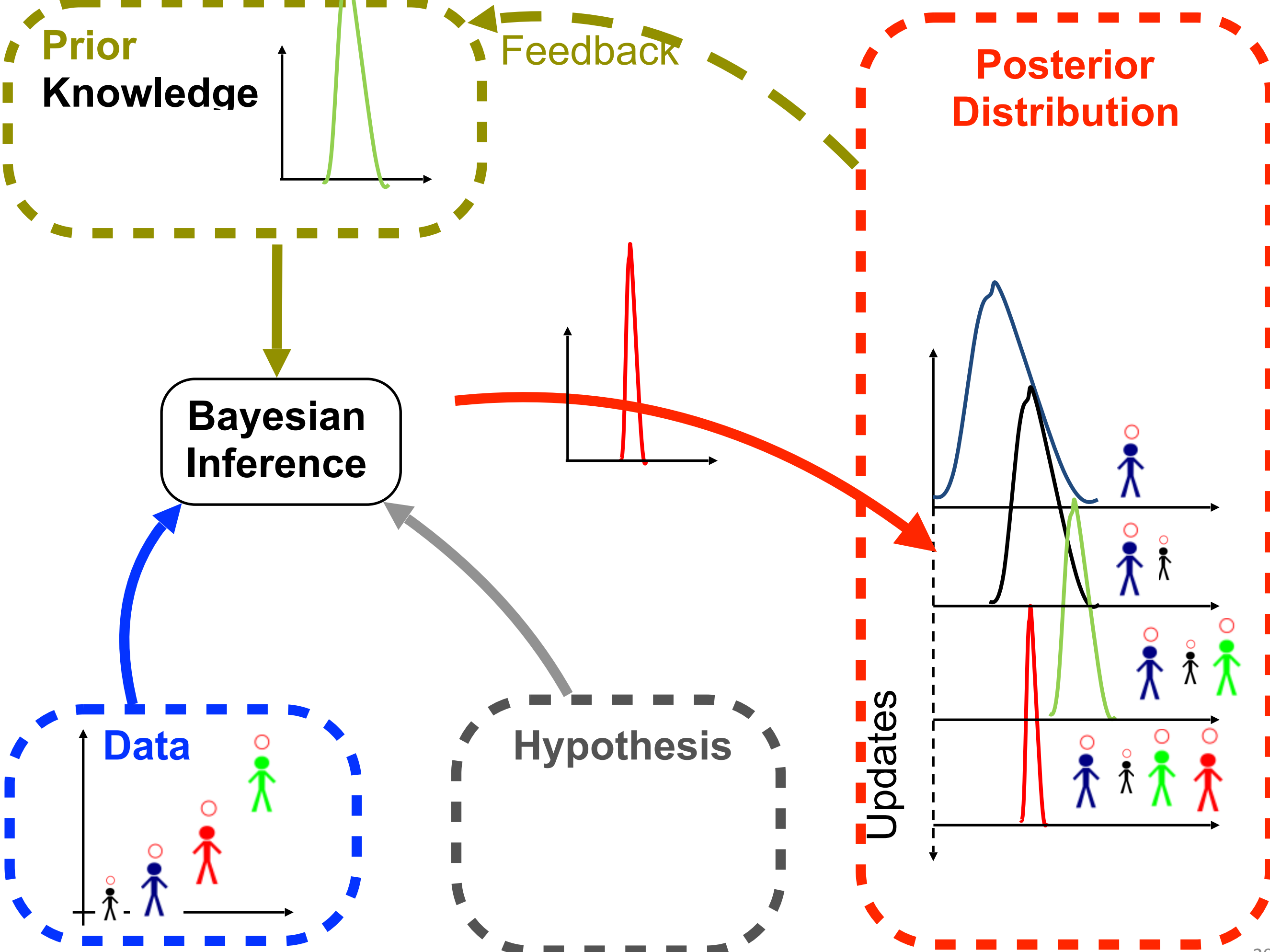


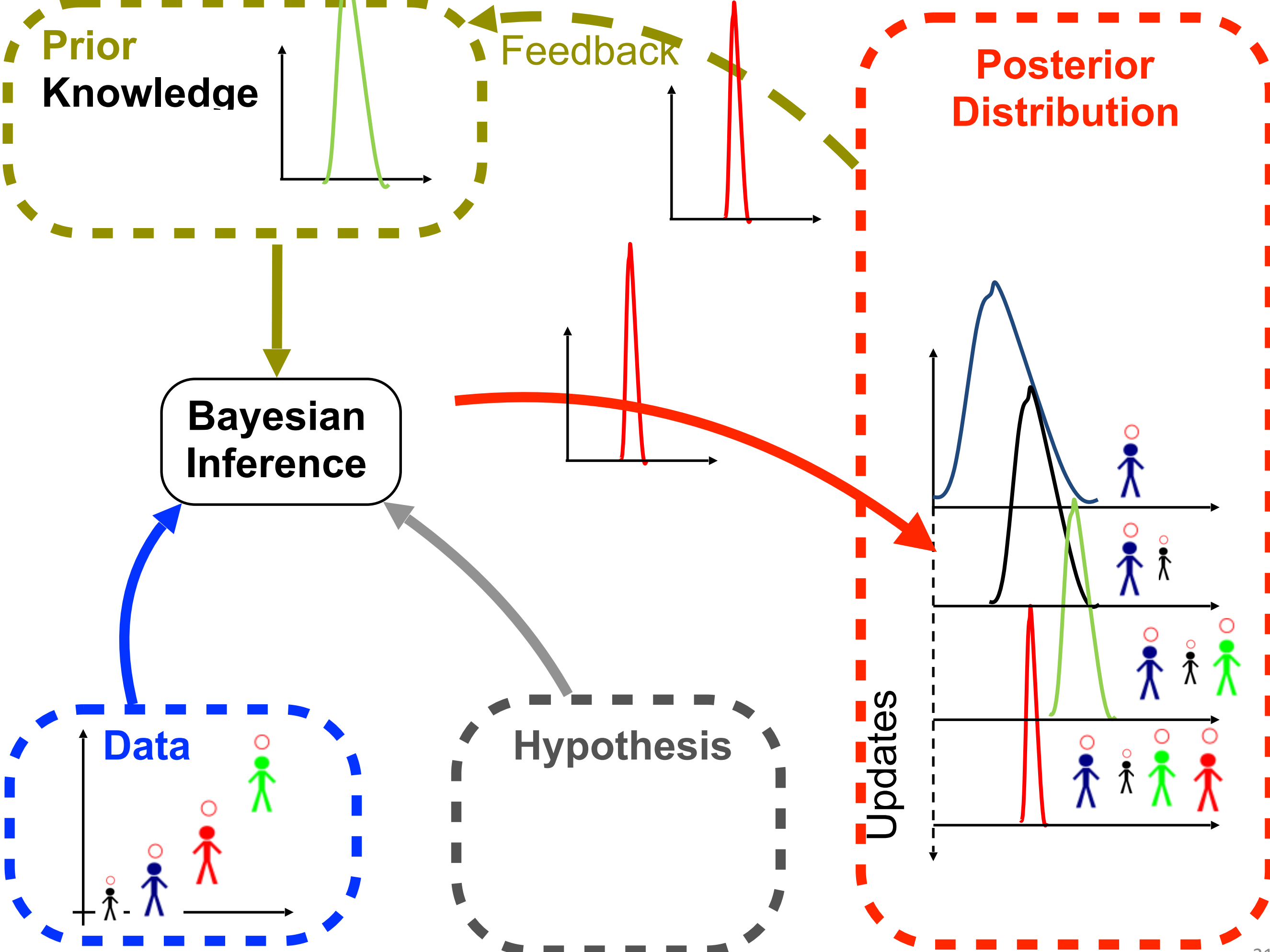


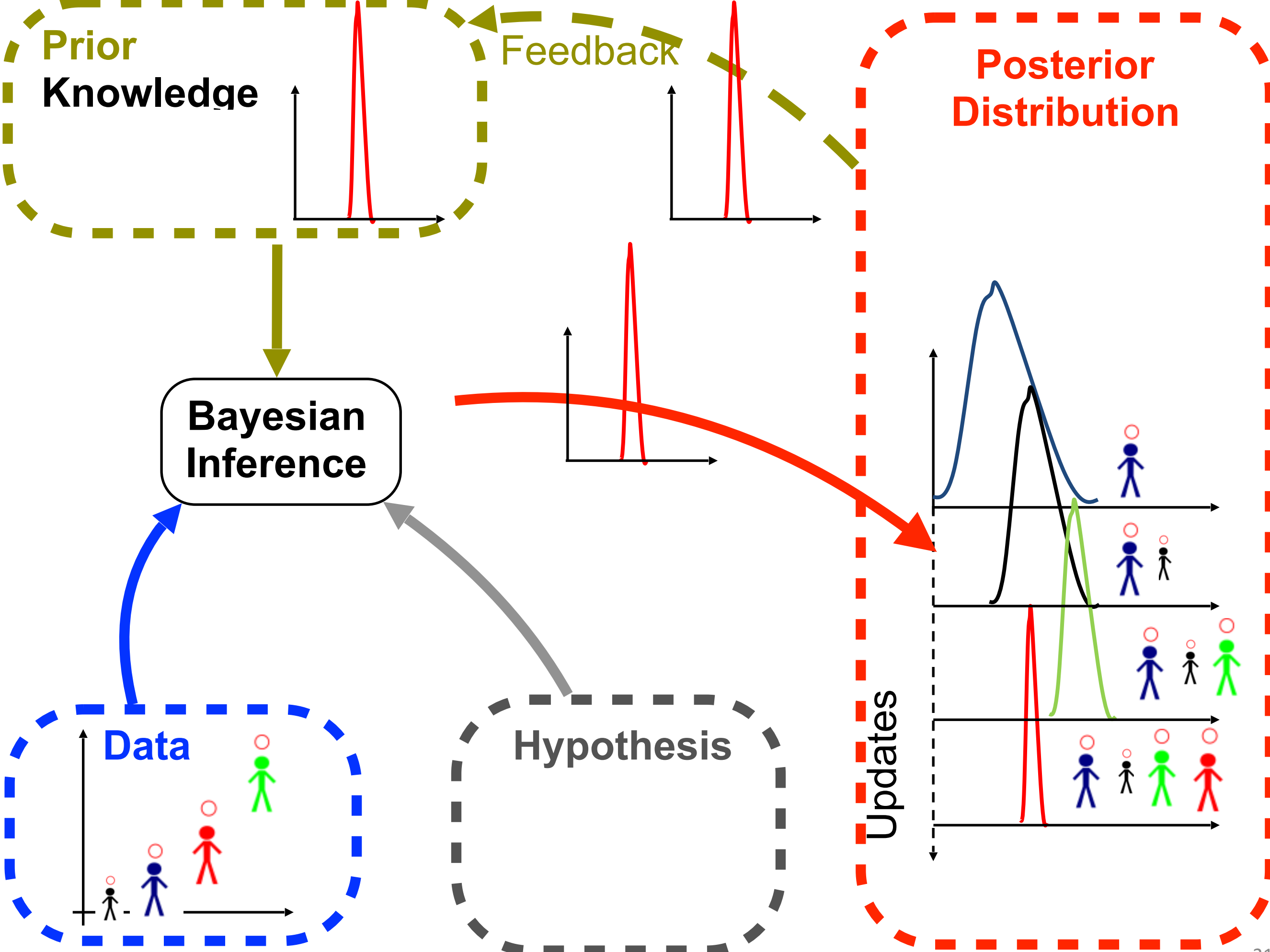












Some applications

- Focus on the finite element method
- Applications in materials science
- Applications in fluid dynamics
- Applications in manufacturing
- Applications in biomechanics
- Applications in real-time simulation for surgical training and surgical guidance



Cut Finite Element Methods for Contact Problems

Susanne Claus

Department of Computer Science, University of Copenhagen, Denmark.

AI Seminar Series, Copenhagen, Jan 2019

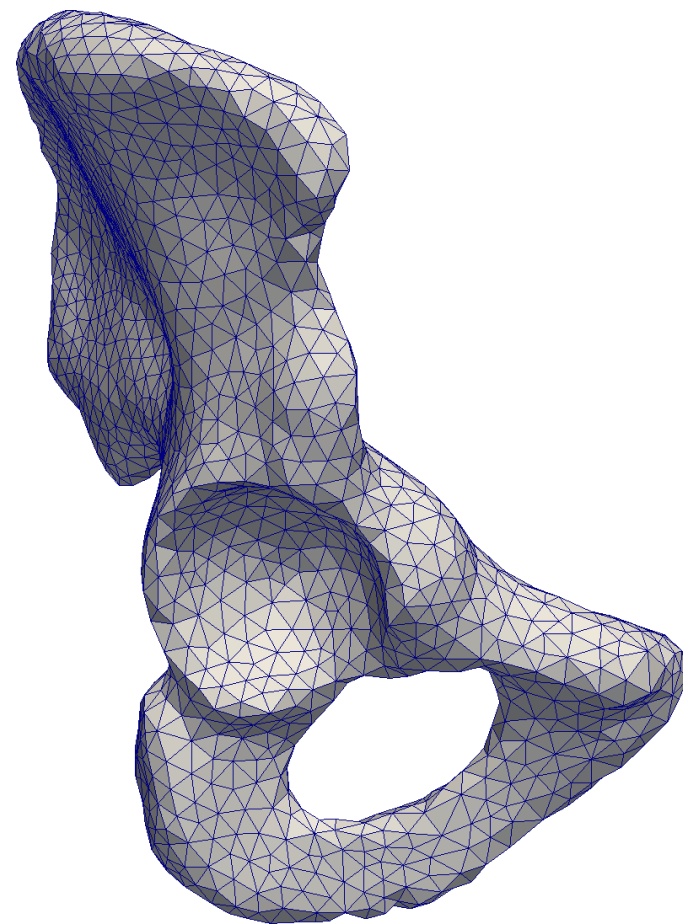


THE ALFRED BENZON FOUNDATION

Geometry Discretisation in Finite Element Methods

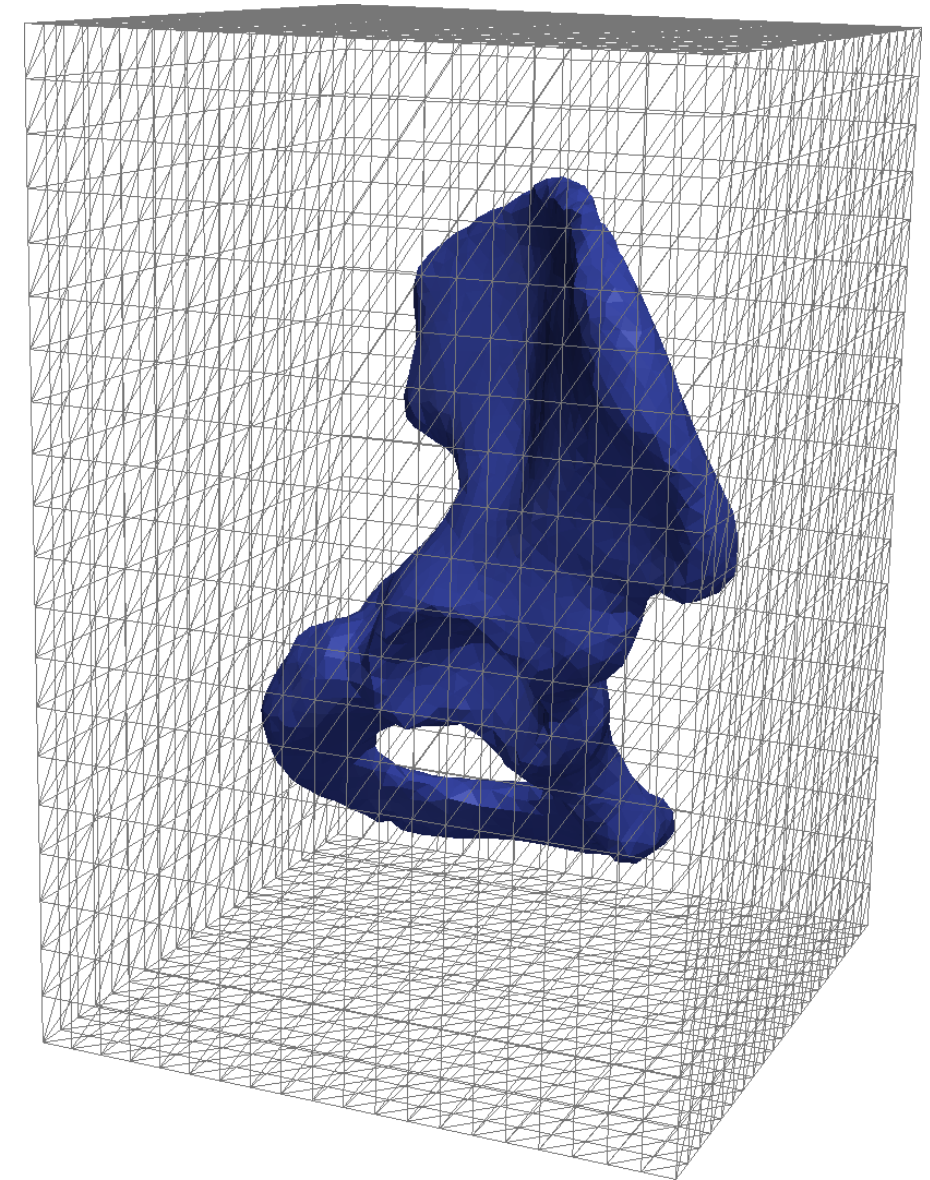


Classical FEM



Geometry is meshed

CutFEM



Geometry is embedded in fixed background grid and described by a function (e.g. level set function)

Finite Element Methods

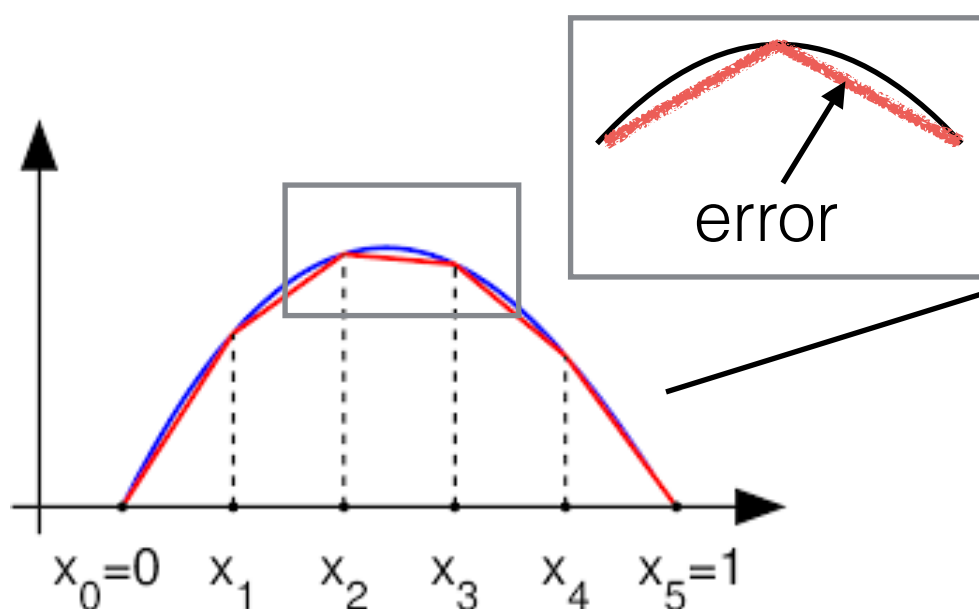
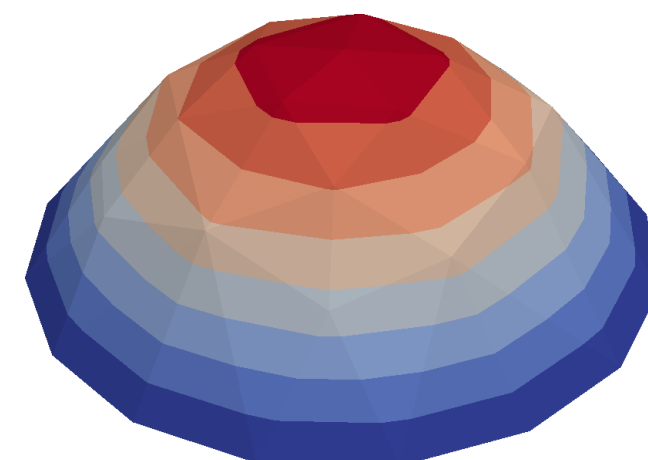
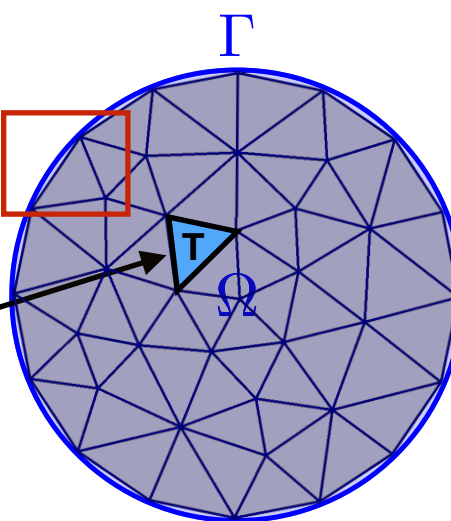
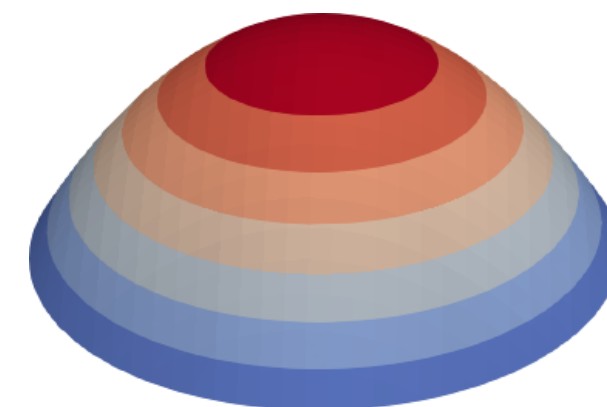
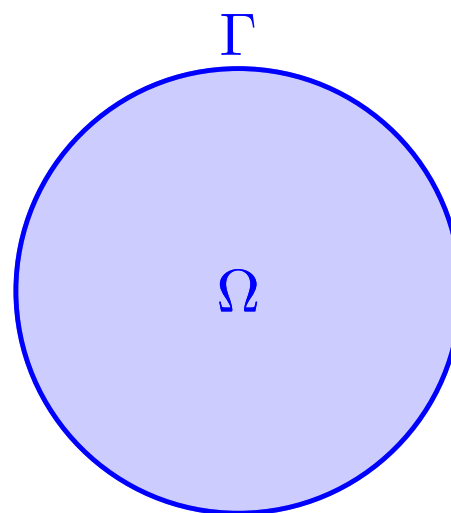
Consider the following diffusion partial differential equation (PDE)

$$-\Delta u = f \text{ in } \Omega$$

$$u = 0 \text{ on } \Gamma$$

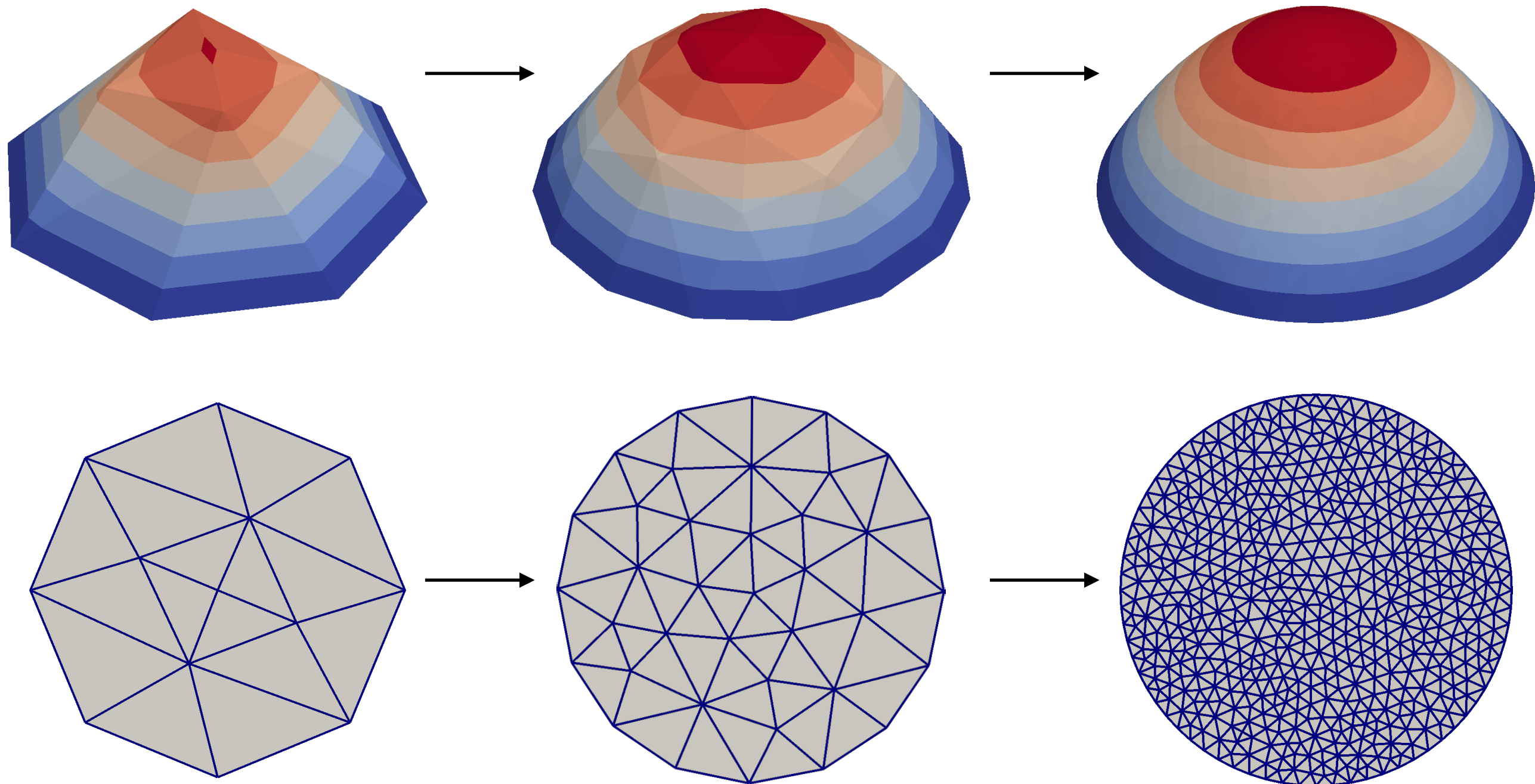
Find $u \in V$ such that

$$\int_{\Omega} \nabla u \cdot \nabla v dx = \int_{\Omega} f v dx \quad \forall v \in V$$



Convergence with mesh refinement

The error decreases with mesh refinement. However, how fast the error decreases with mesh refinement (convergence order) depends on multiple factors.



Convergence strongly depends on

Accuracy

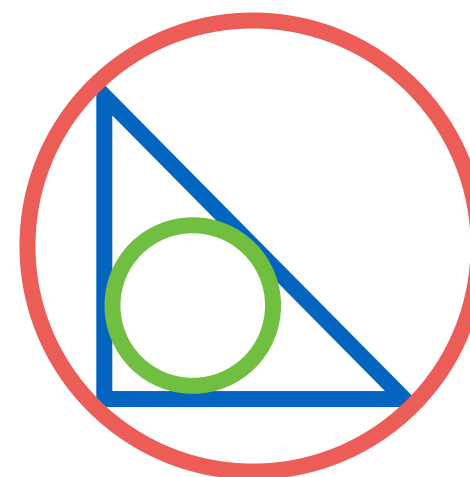
- **Numerical error:** error from piecewise polynomial approximation of the solution and of the geometry.
- **Mesh Quality:** The quality and size of the mesh has a significant impact on the accuracy of the solution

Stability

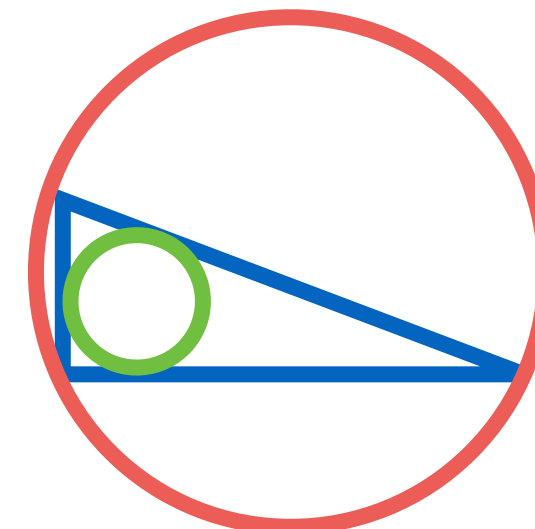
- **Instabilities** frequently occur in **simulations as numerical errors can grow in the solution process. Numerical error growth needs to be controlled and stabilised** carefully. Too much stabilisation leads to inaccuracies.

Convergence = Accuracy + Stability

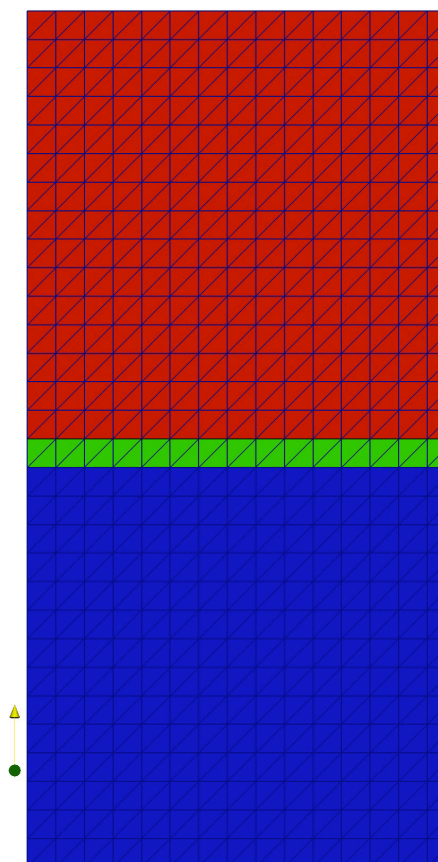
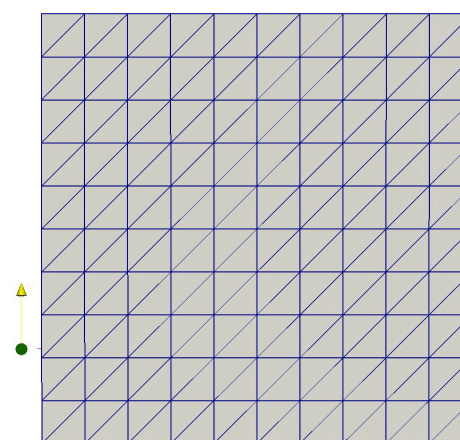
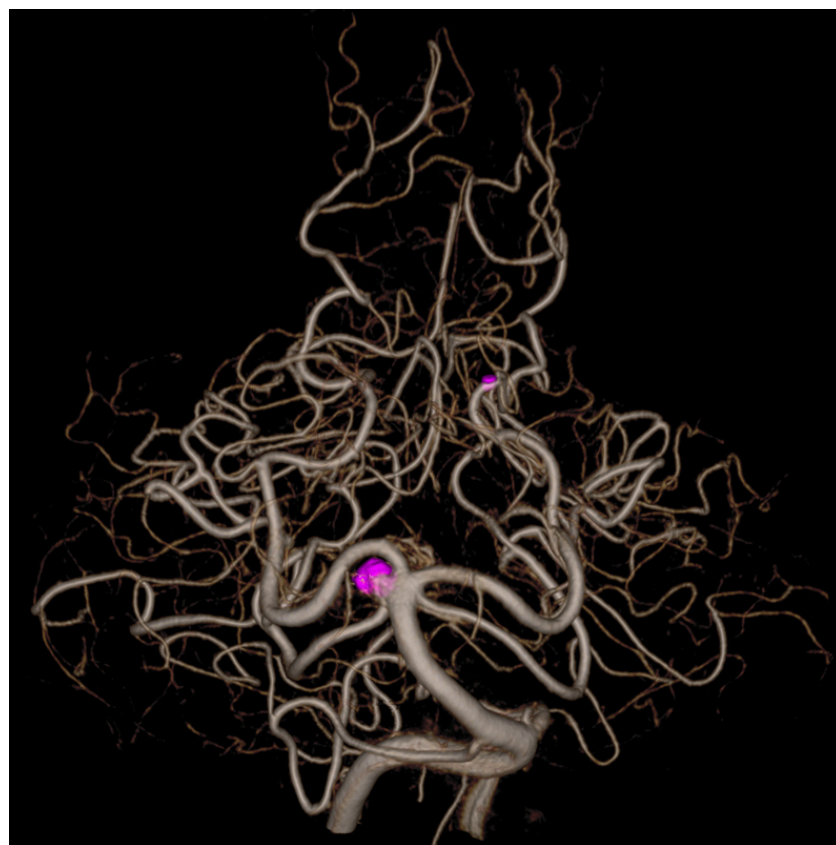
Good Element



Bad Element



Difficulty of maintaining a high quality mesh



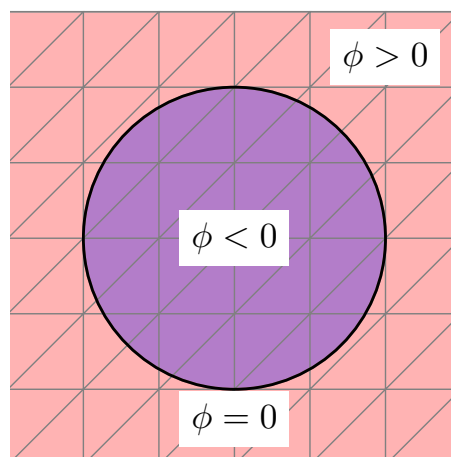
Advantages of Mesh Independent Geometry Descriptions

1. reduces the computational cost for preprocessing or transformation of acquired geometry descriptions
2. efficient and robust for problems involving evolving geometries undergoing large deformations

Important aspects of implicit geometry/cut finite element methods

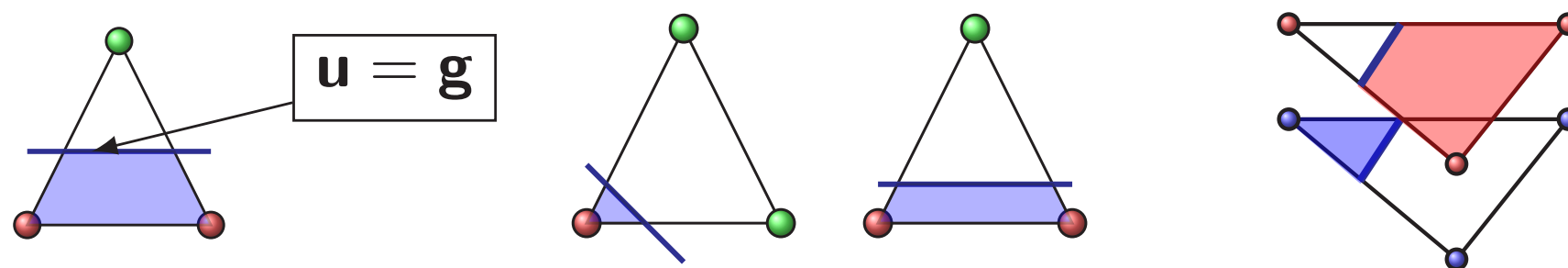
Geometry Algorithms

Discretisation of the geometry based on implicit interface description



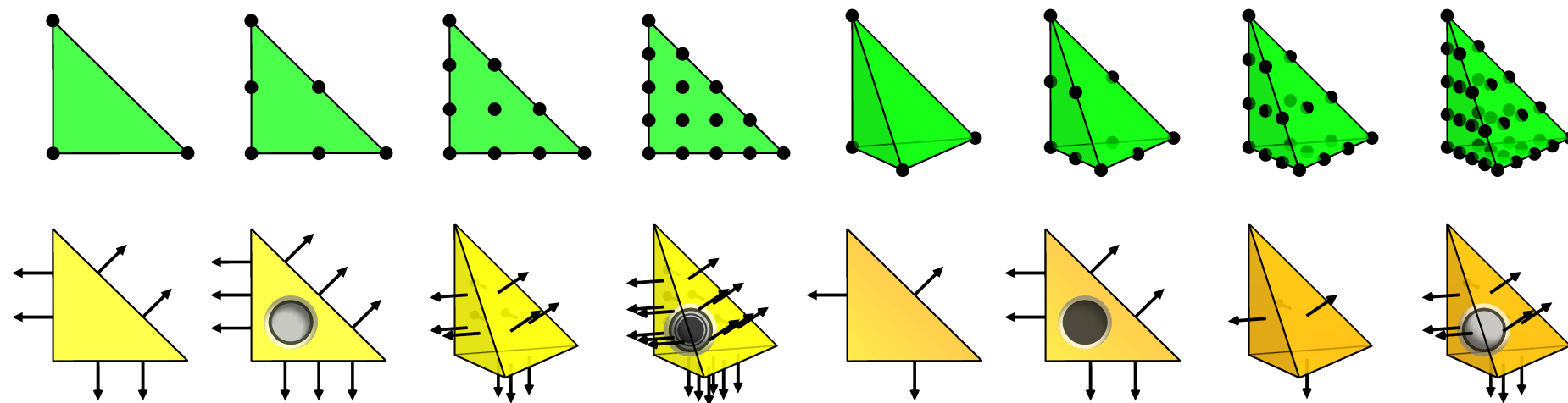
Accuracy and Stabilisation

Construction of stable and accurate finite element methods independent of how the interface intersects the mesh.



Open Source Finite Element Library for the Automated Solution of PDEs

- high level mathematical input language
- generates efficient C++ code from these mathematical inputs
- supports a wide range of different finite element types
- supports simulations in 2D and 3D
- fully parallelised
- active world wide developer community, e.g. Simula Research Laboratory, University of Cambridge, University of Chicago, University of Texas at Austin, KTH Royal Institute of Technology, Chalmers University of Technology.



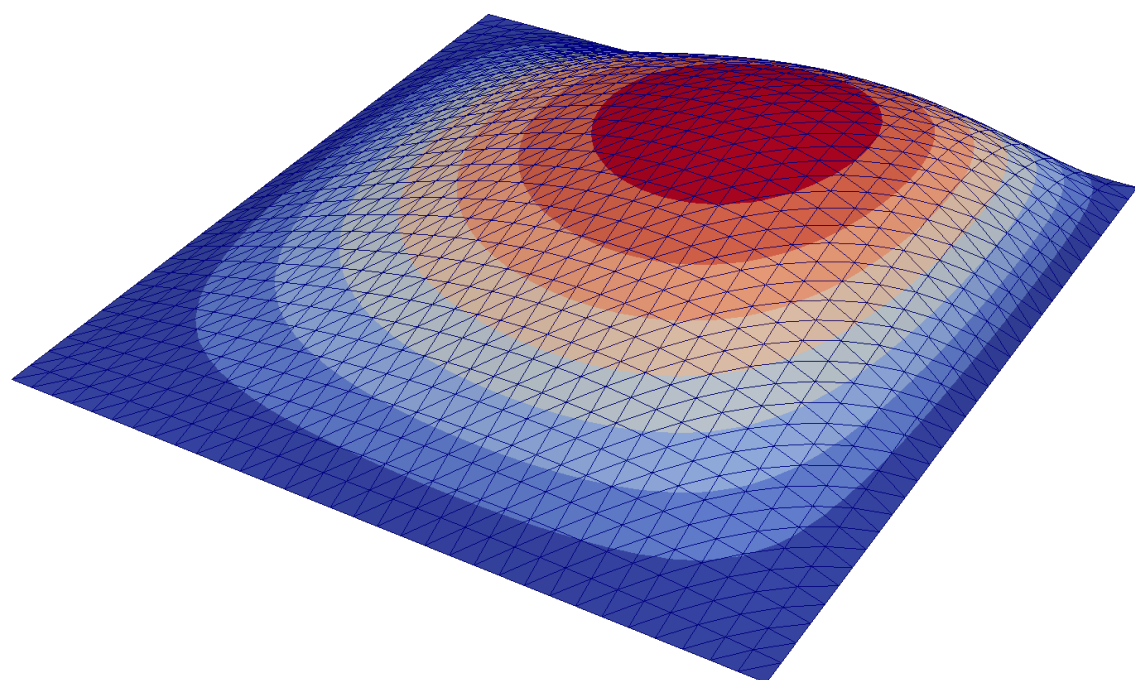
FEniCS Example: Poisson Equation

Consider the elliptic problem

$$\begin{aligned} -\Delta u &= f \text{ in } \Omega, \\ u &= 0 \text{ on } \Gamma. \end{aligned}$$

Find $u \in V$ such that

$$\int_{\Omega} \nabla u \cdot \nabla v dx = \int_{\Omega} f v dx \quad \forall v \in V$$



```
from dolfin import *

# Create mesh and define function space
mesh = UnitSquareMesh(32, 32)
V = FunctionSpace(mesh, "CG", 1)

# Define boundary condition
bc = DirichletBC(V, 0.0, DomainBoundary())

# Define variational problem
u = TrialFunction(V)
v = TestFunction(V)
f = Expression("x[0]*x[1]")

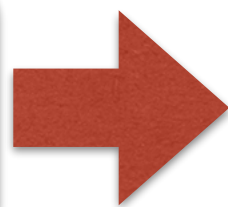
a = inner(grad(u), grad(v))*dx
L = f*v*dx

# Compute solution
u = Function(V)
solve(a == L, u, bc)
```




FEniCS: Under The Hood

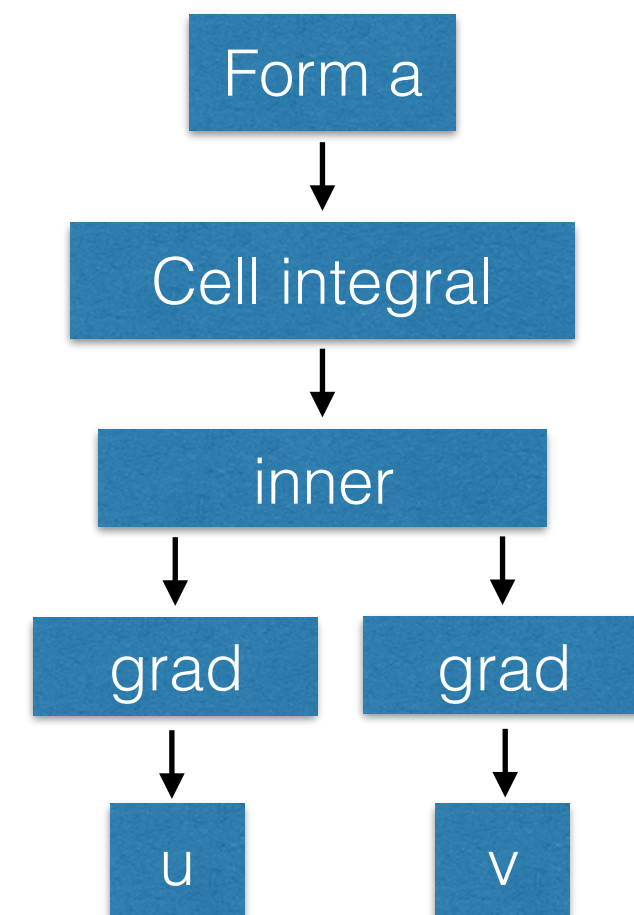
Input: File in Python
and Unified Form
Language



Unified Form Language (UFL)
Interprets expressions close to
mathematical notation

```
a=inner(grad(u), grad(v))*dx
```

$$a(u, v) = \int \nabla u \nabla v \, dx$$





FEniCS: Under The Hood

Input: File in Python and Unified Form Language

```
a=inner(grad(u), grad(v))*dx
```

$$a(u, v) = \int \nabla u \nabla v \, dx$$

Unified Form Language (UFL)
Interprets expressions close to mathematical notation

FEniCS Form Compiler (FFC)
Generates Header file with information about
Elemental matrices (form)
Degrees of Freedom Map (element)

Poisson.h

DOLFIN

(Mesh, Communicator and Assembler)

LibCutFEM

Form a

Cell integral

inner

grad

u

grad

v

```
/// Evaluate basis function i at given  
point x in cell  
_evaluate_basis(std::size_t i, double*  
values, const double* x, const double*  
coordinate_dofs)
```

```
/// Tabulate the tensor for the  
contribution from a local cell  
virtual void tabulate_tensor(double* A,  
const double * const * w, const double*  
coordinate_dofs)
```

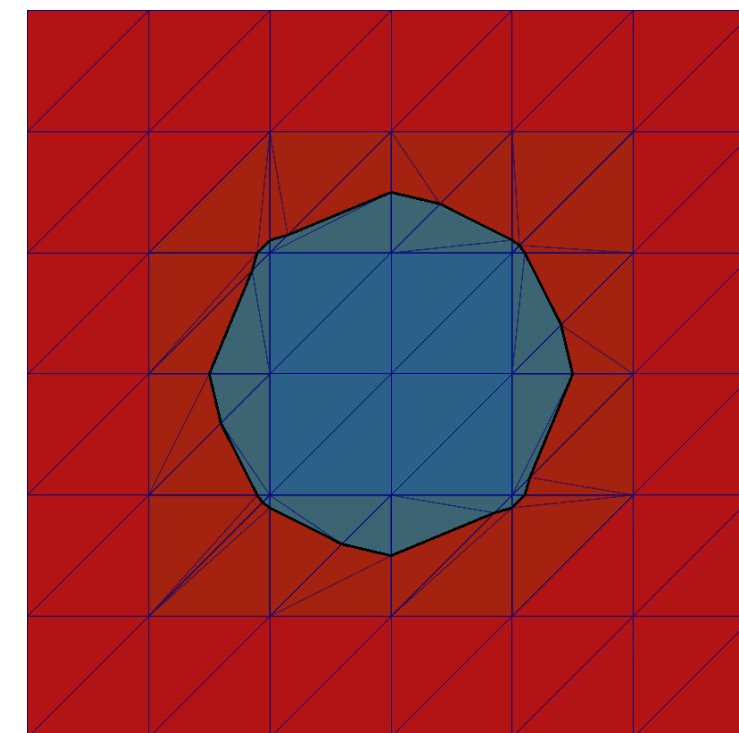
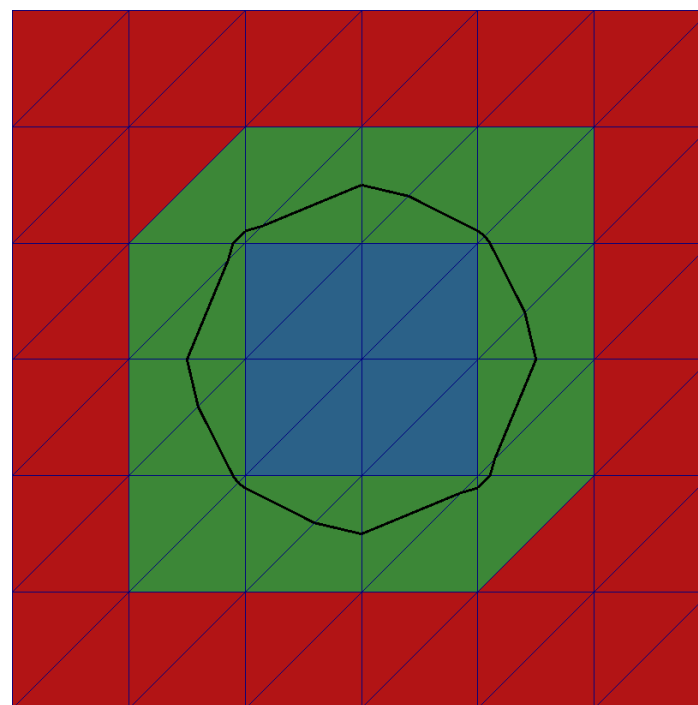
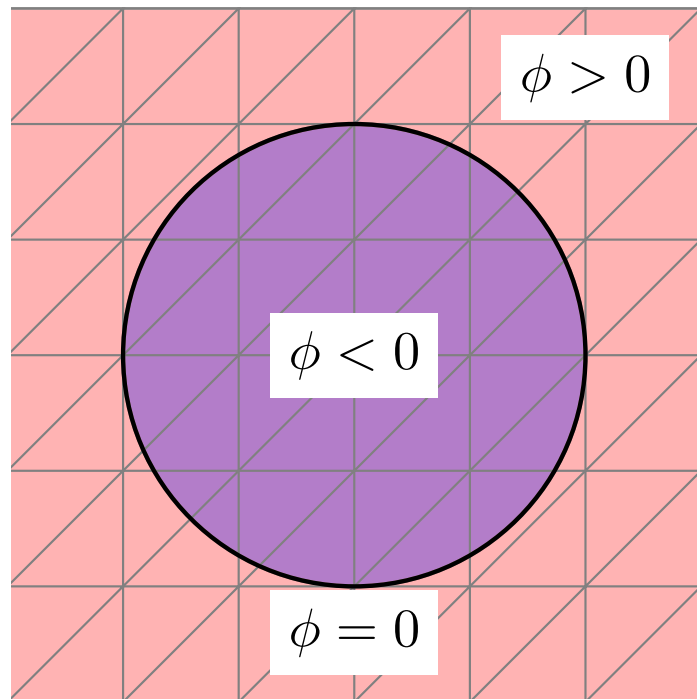
Geometry Algorithm



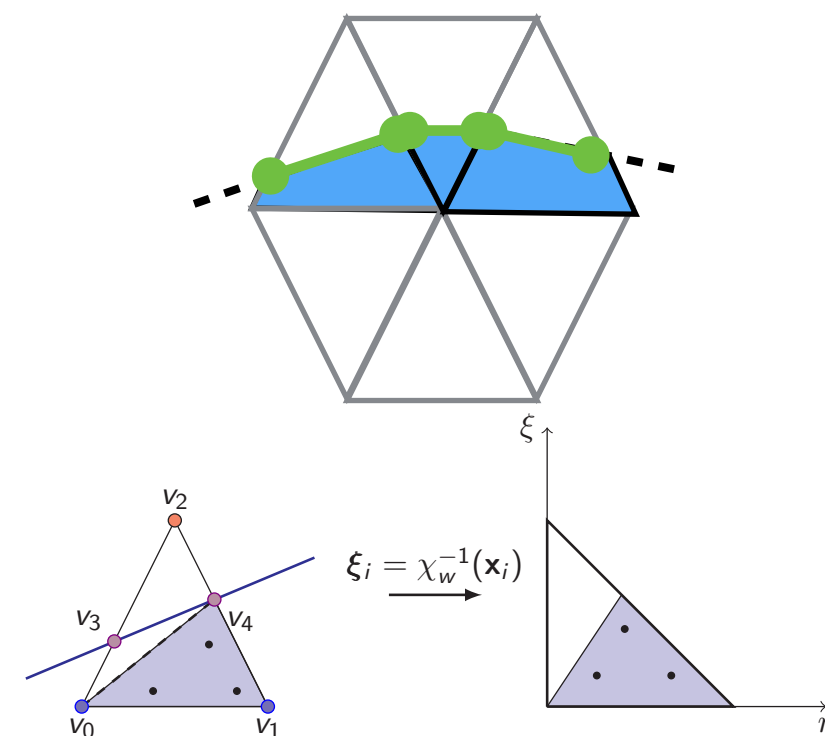
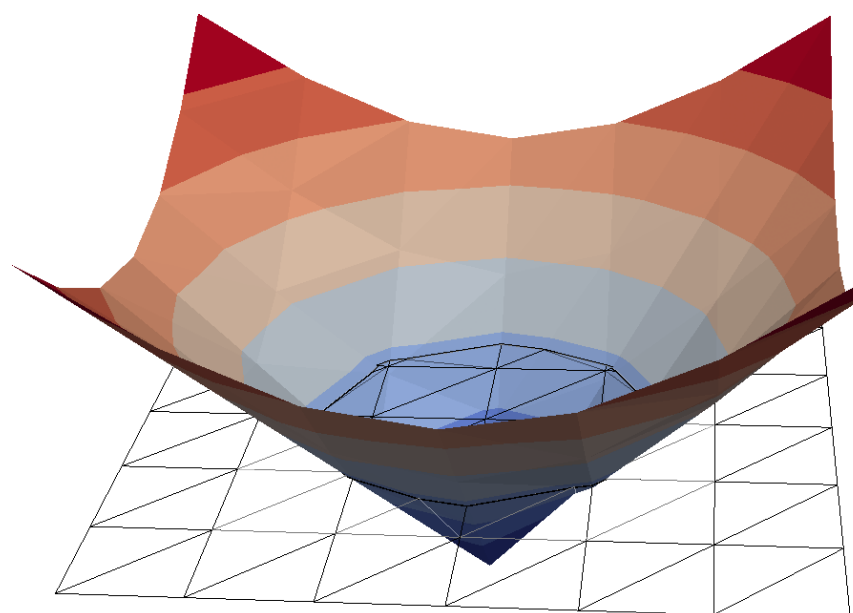
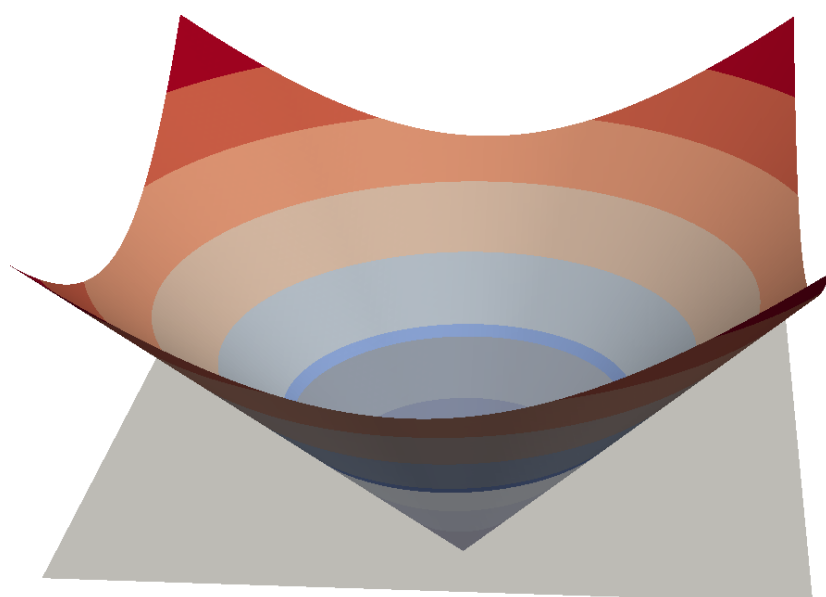
Level Set Function

Finite Element Approximation

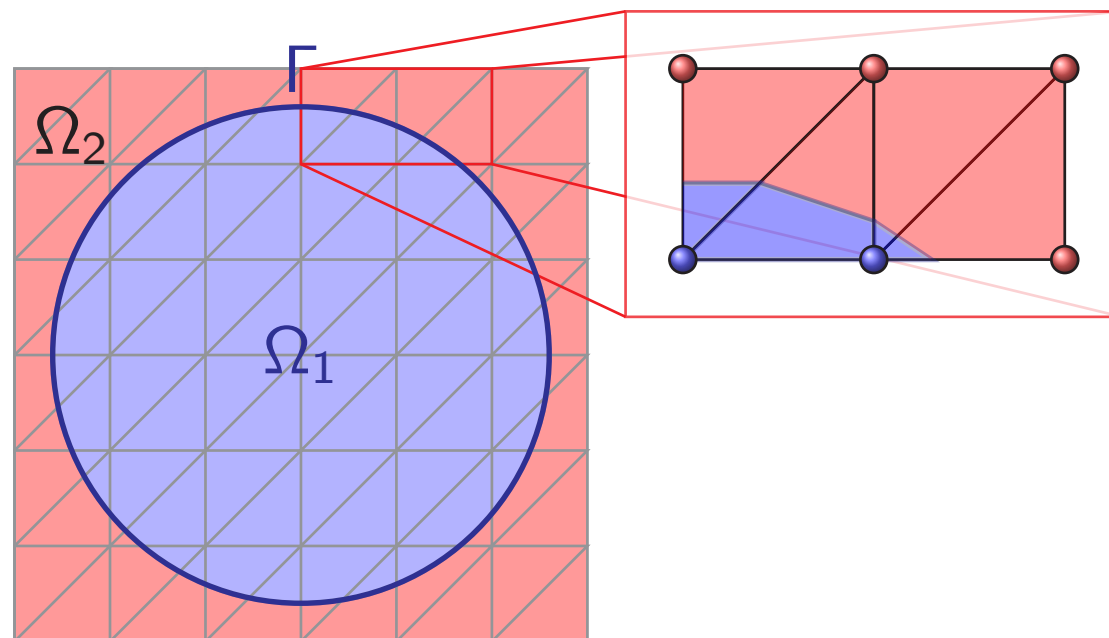
Mesh/Levelset intersection for integration



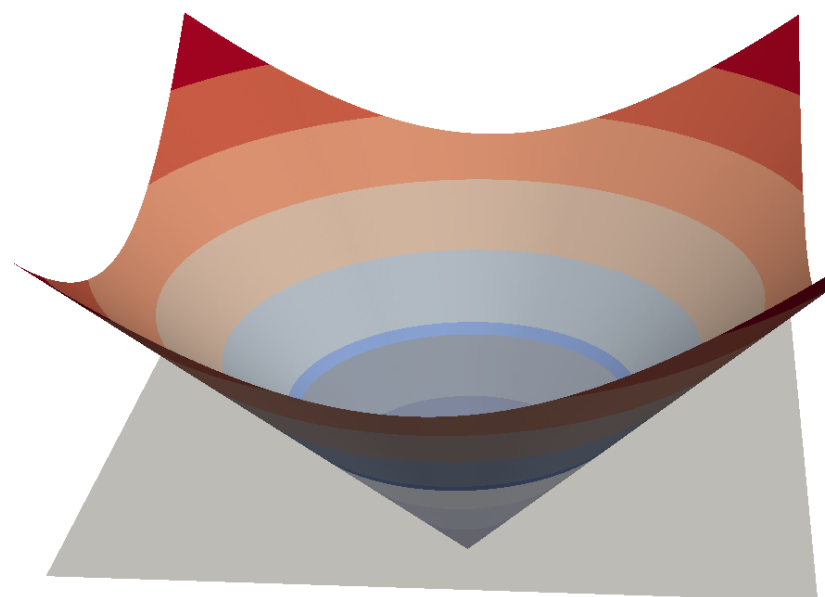
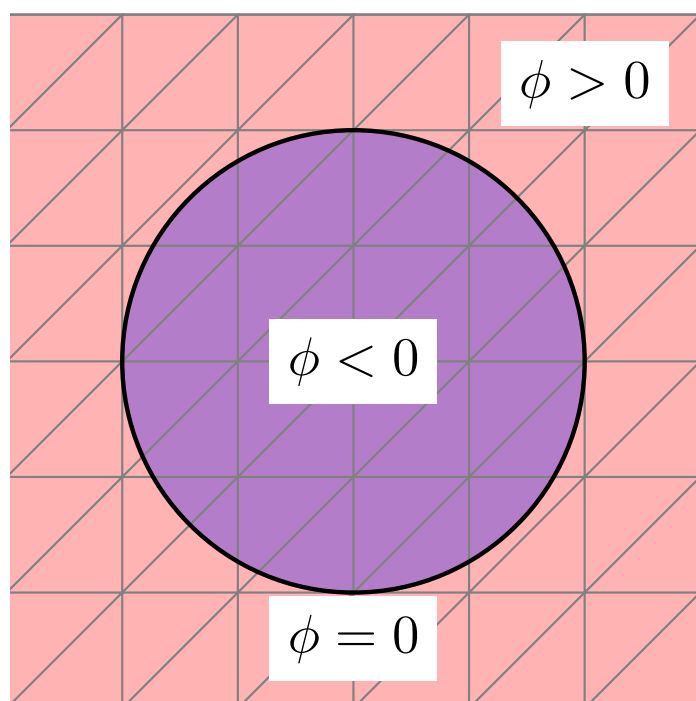
Signed Distance Function



Level Set Geometry Description



Describe geometry using Level-Set Function



Normal

$$\mathbf{n}_\Gamma = \frac{\nabla \phi}{\|\nabla \phi\|_0}$$

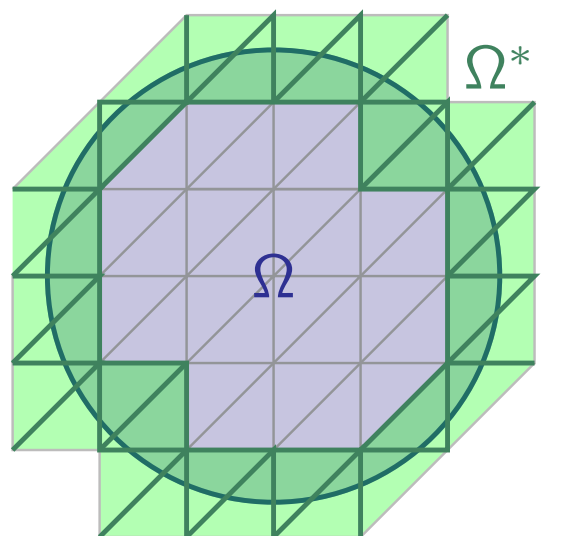
Curvature

$$\kappa = \nabla \cdot \mathbf{n}$$

Sense of Vicinity

$$|\phi(x)| < \delta$$

Fictitious Domain Poisson Problem



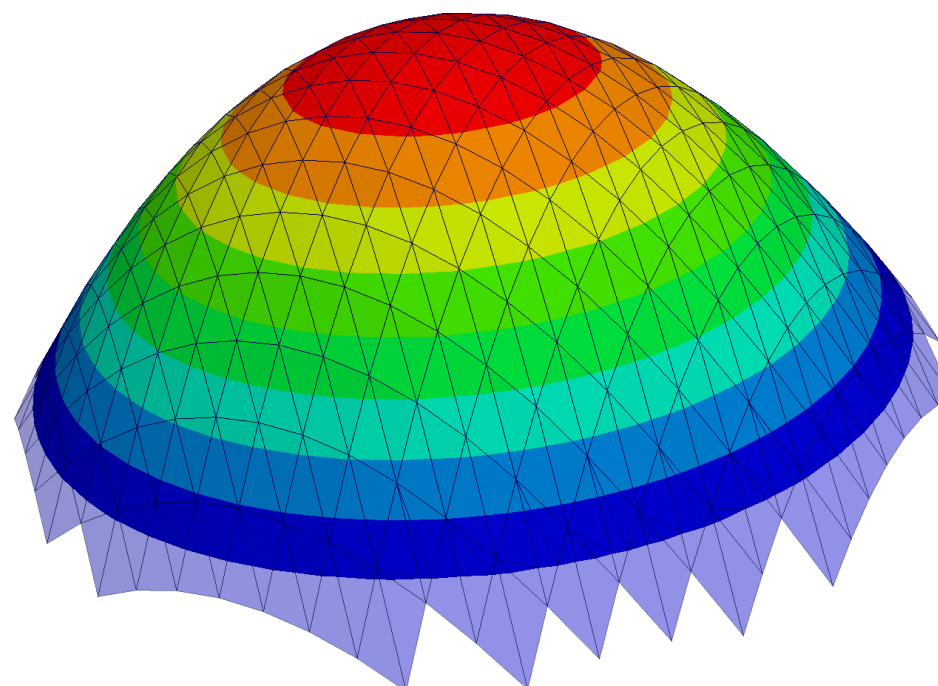
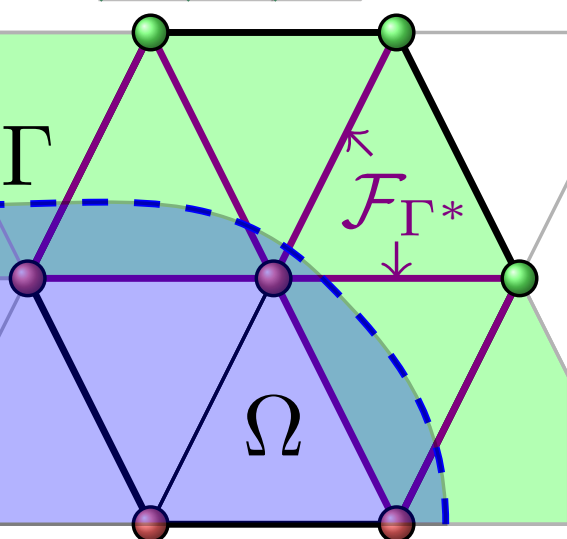
Find $u_h \in V_h$ such that for all $v_h \in V_h$

$$A(u_h, v_h) = a(u_h, v_h) + j(u_h, v_h) = L(v_h)$$

$$a(u_h, v_h) = \int_{\Omega} \nabla u_h \cdot \nabla v_h \, dx + \int_{\Gamma} \underbrace{(-\nabla u_h \cdot n v_h)}_{\text{consistency}} - \underbrace{\nabla v_h \cdot n u_h}_{\text{symmetry}} + \underbrace{\frac{\gamma}{h} u_h v_h}_{\text{coercivity}} \, ds$$

$$L(v_h) = \int_{\Omega} f v_h \, dx + \int_{\Gamma} (-g \nabla v_h \cdot n + \frac{\gamma}{h} g v_h) \, ds$$

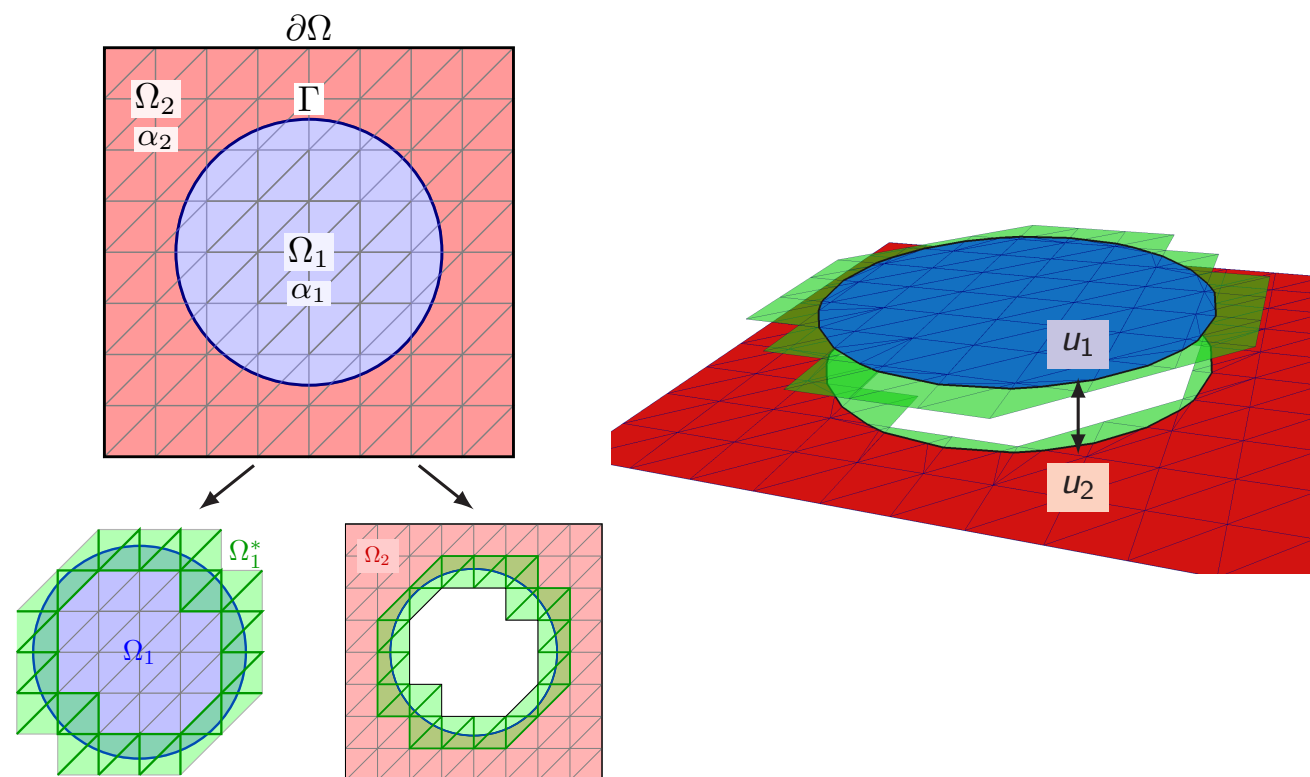
$$j(u_h, v_h) = \gamma_1 \sum_{F \in \mathcal{F}_{\Gamma^*}} h_F ([\partial_n u_h], [\partial_n v_h])_F$$



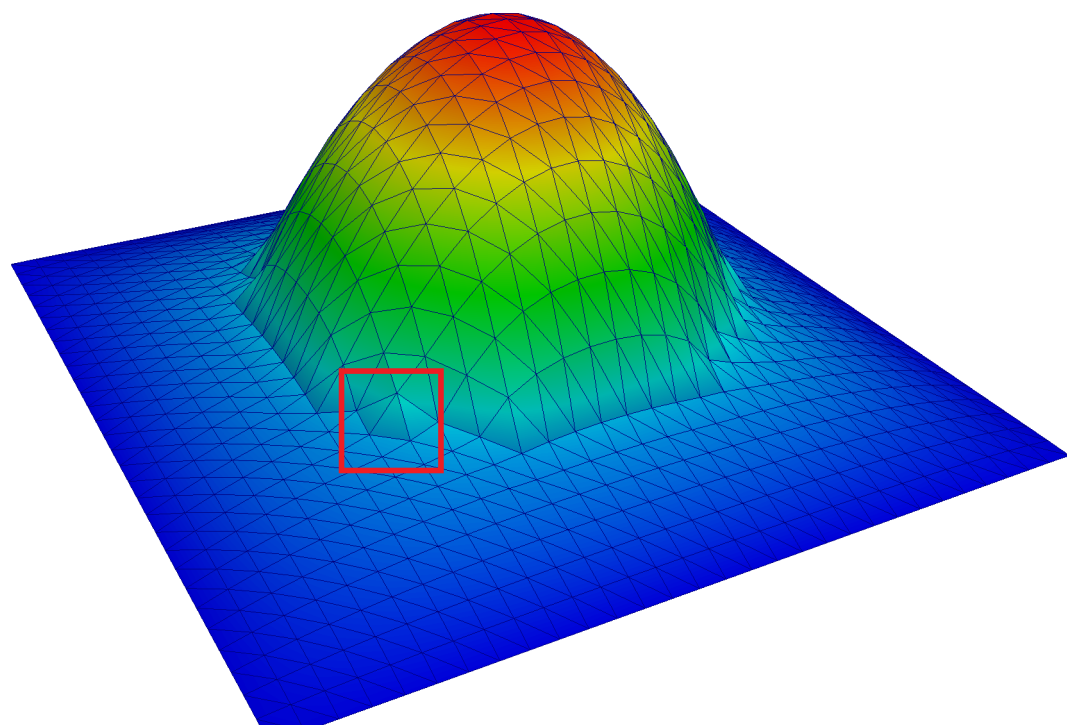
Poisson with contrast in diffusivities

$$\begin{aligned}
 -\nabla \cdot (\alpha \nabla u) &= f & \text{in } \Omega_1 \cup \Omega_2, \\
 u &= 0 & \text{on } \partial\Omega, \\
 [[u]] &= 0 & \text{on } \Gamma, \\
 [[-\alpha_i \nabla u_i \cdot \mathbf{n}]] &= 0 & \text{on } \Gamma.
 \end{aligned}$$

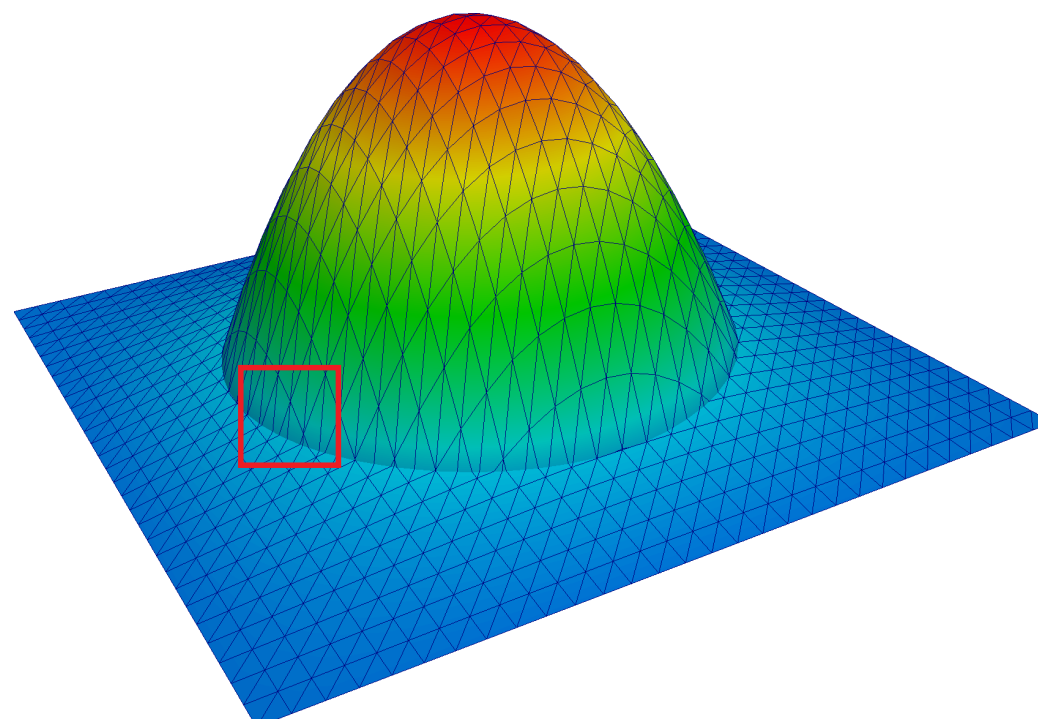
Choose $\alpha_1 = 1, \alpha_2 = 10, f_1 = f_2 = 1$.

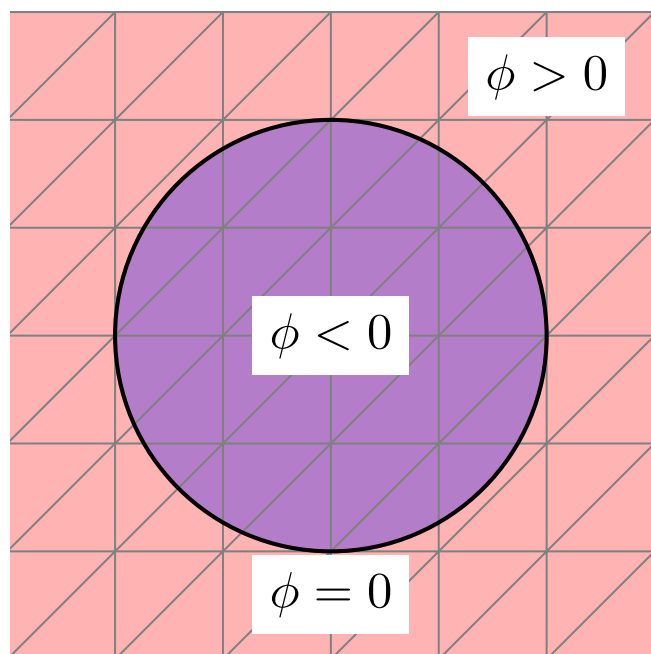


Without Enrichment



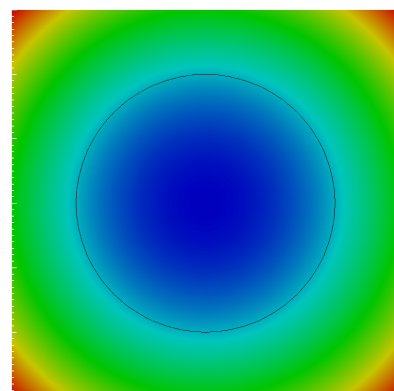
With Enrichment





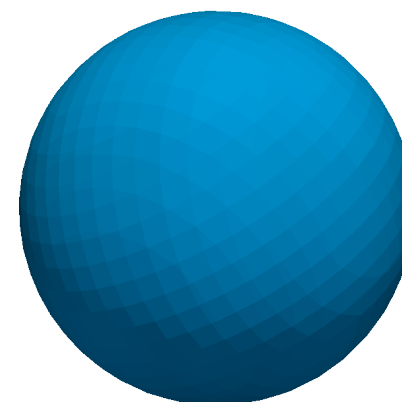
Circle

$$\phi(\mathbf{x}) = x^2 + y^2 - 1$$



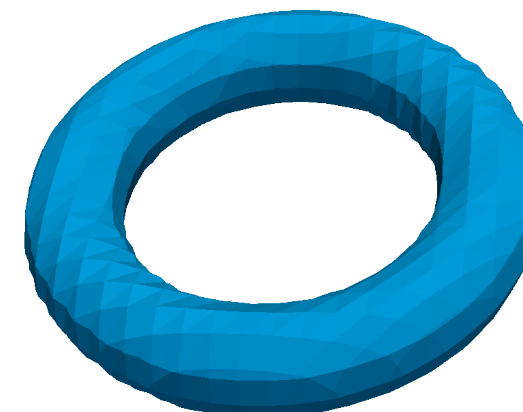
Sphere

$$\phi(\mathbf{x}) = x^2 + y^2 + z^2 - 1$$



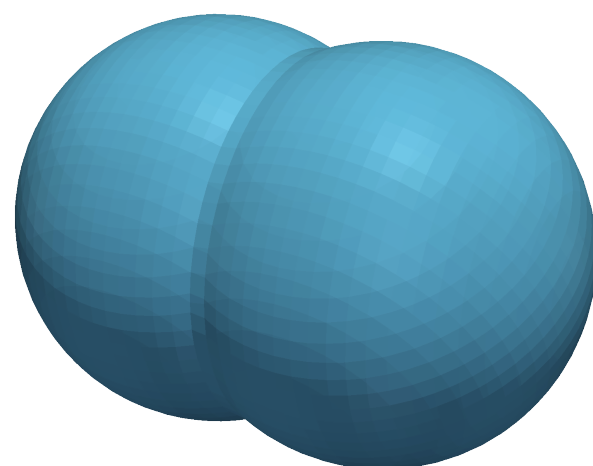
Torus

$$\phi(\mathbf{x}) = (R - \sqrt{x^2 + y^2})^2 + z^2 - r^2$$



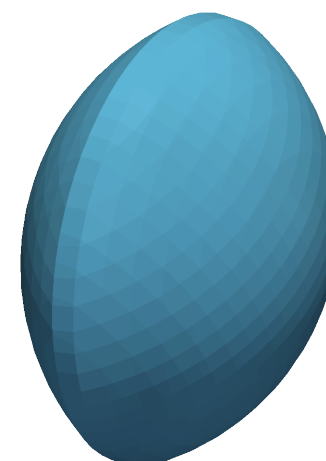
Union:

$$\phi(\mathbf{x}) = \min(\phi_1(\mathbf{x}), \phi_2(\mathbf{x}))$$



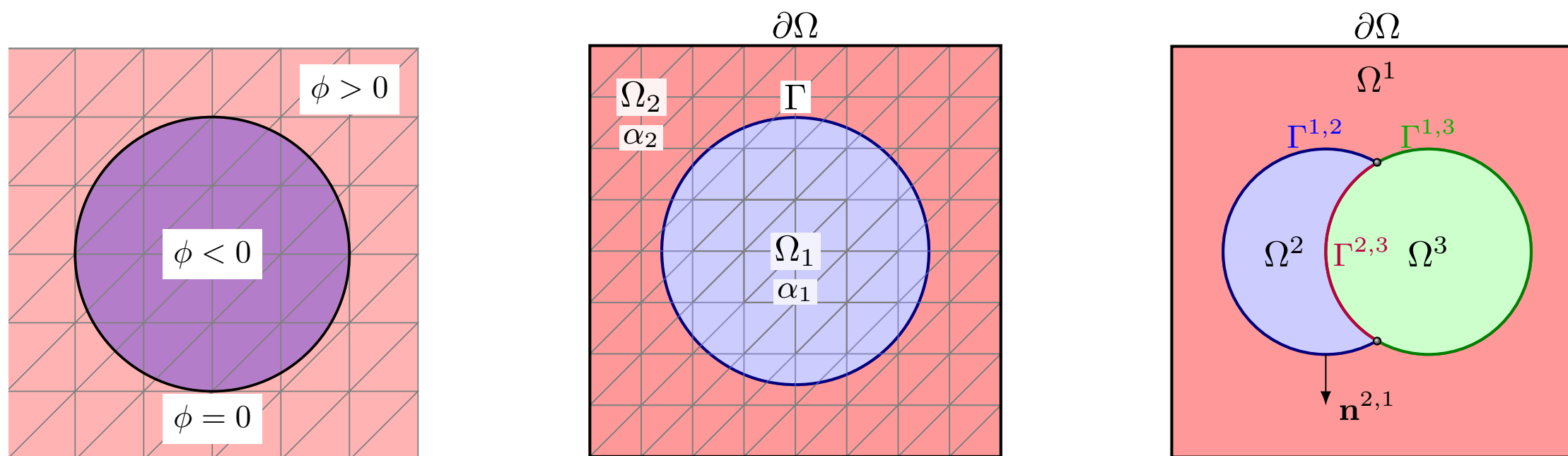
Intersection:

$$\phi(\mathbf{x}) = \max(\phi_1(\mathbf{x}), \phi_2(\mathbf{x}))$$

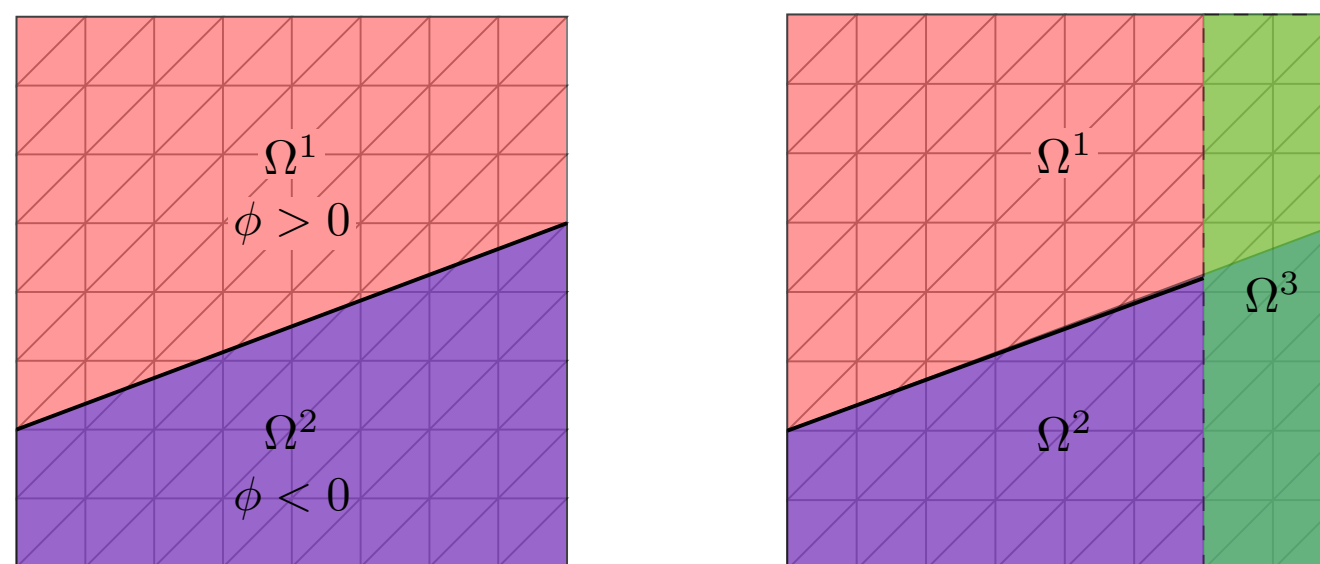


Limitations with single level set function

Partitions the domain into one inside and one outside domain (max. 2 different materials)

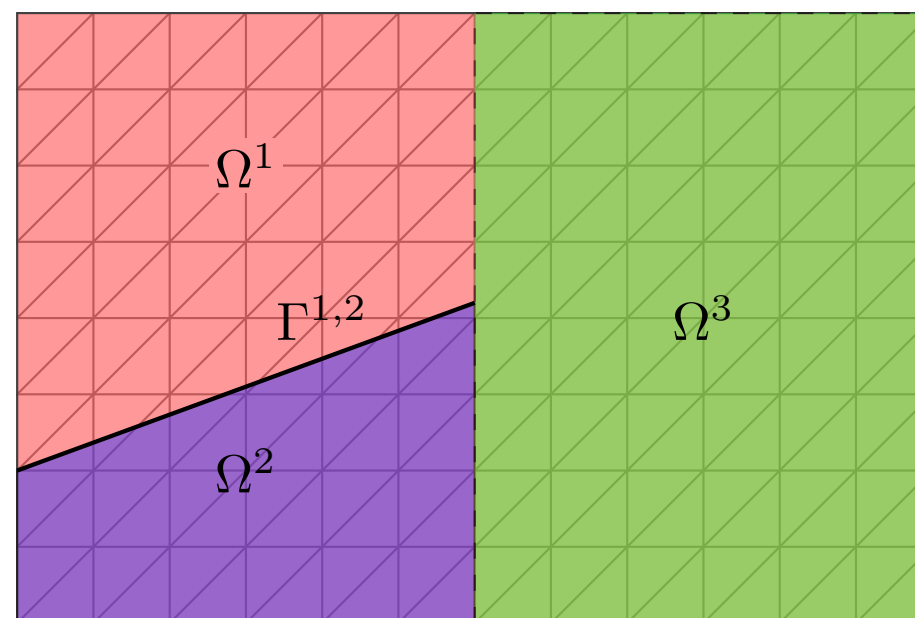
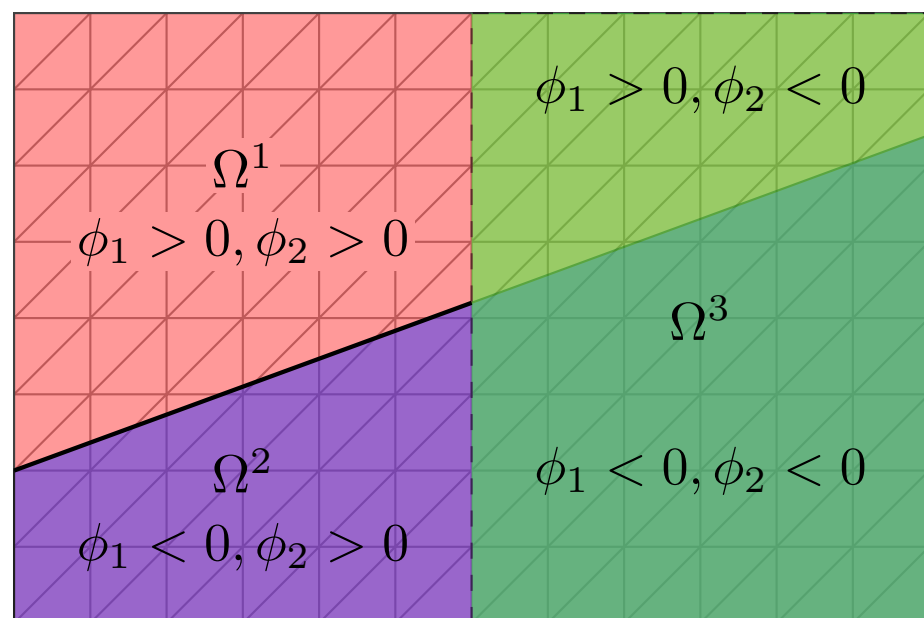
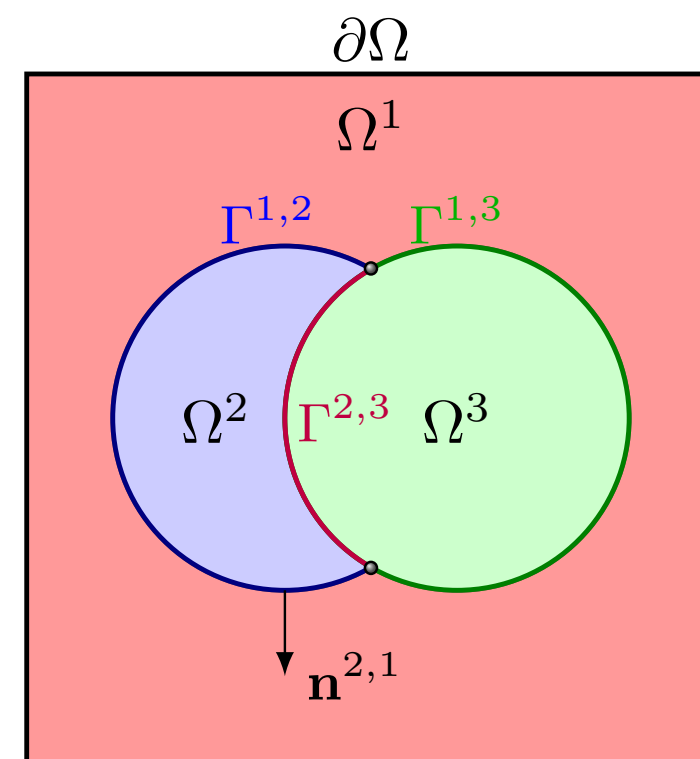
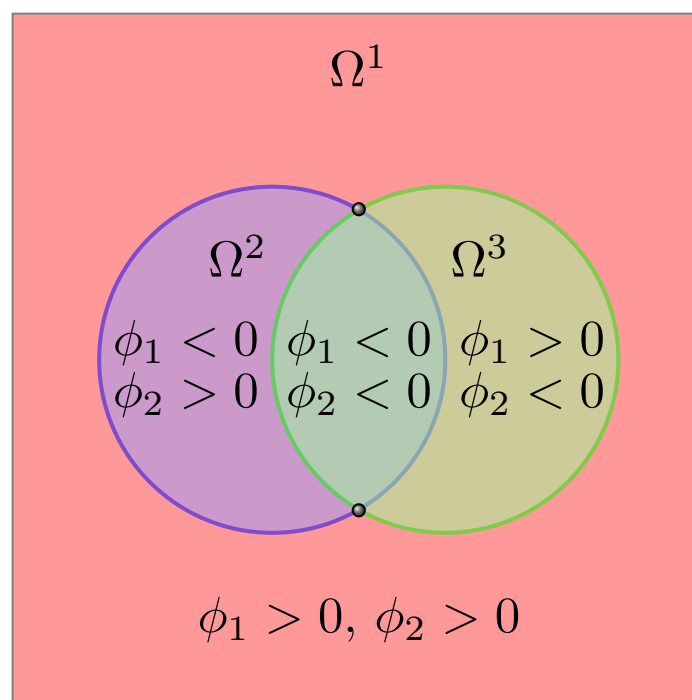
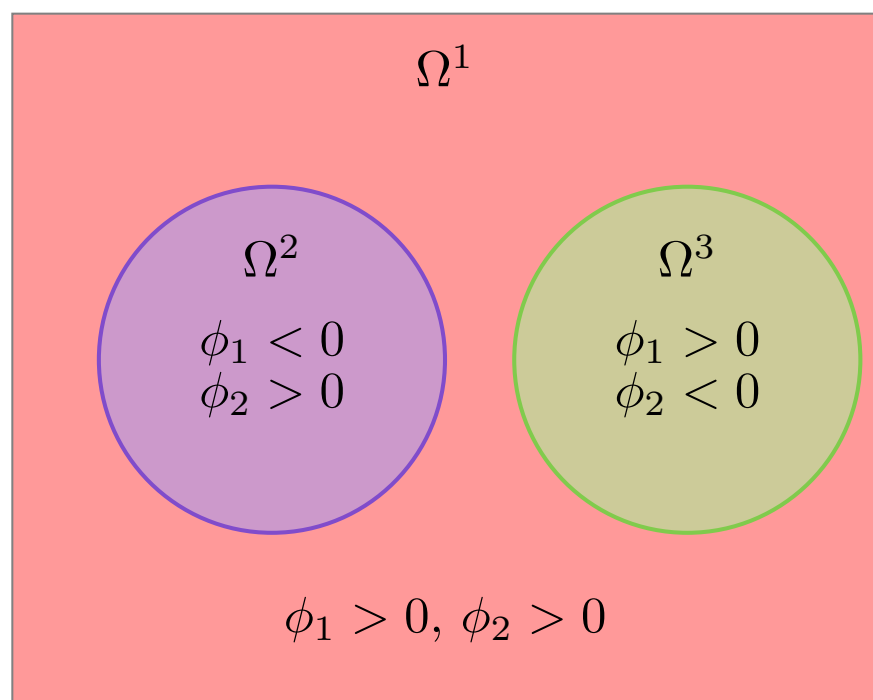


Zero level set surface contour is a closed surface, i.e it is not possible to describe geometries with open boundaries such as cracks

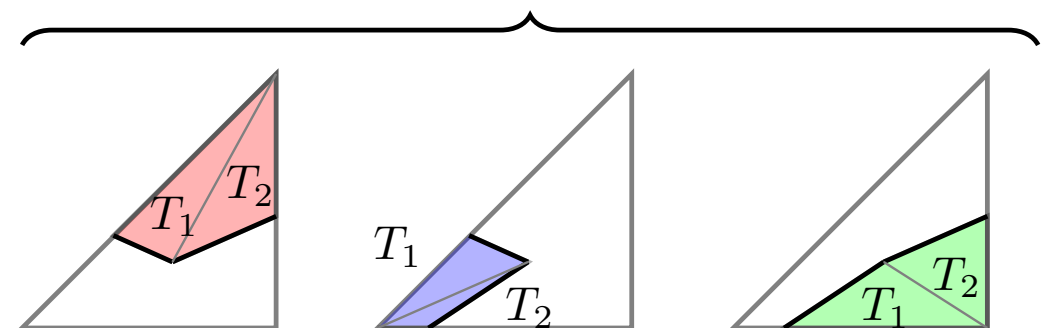
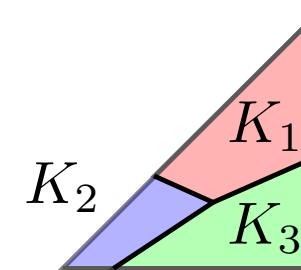
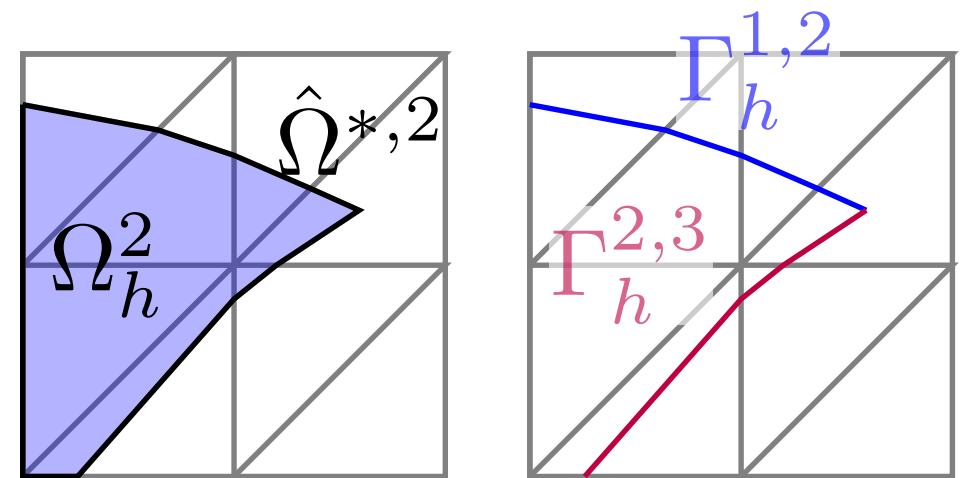
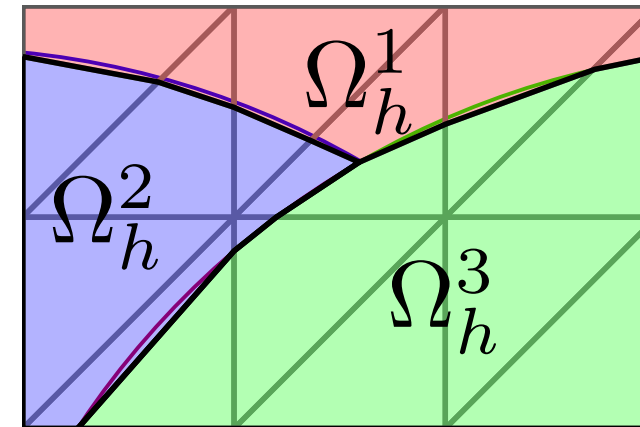
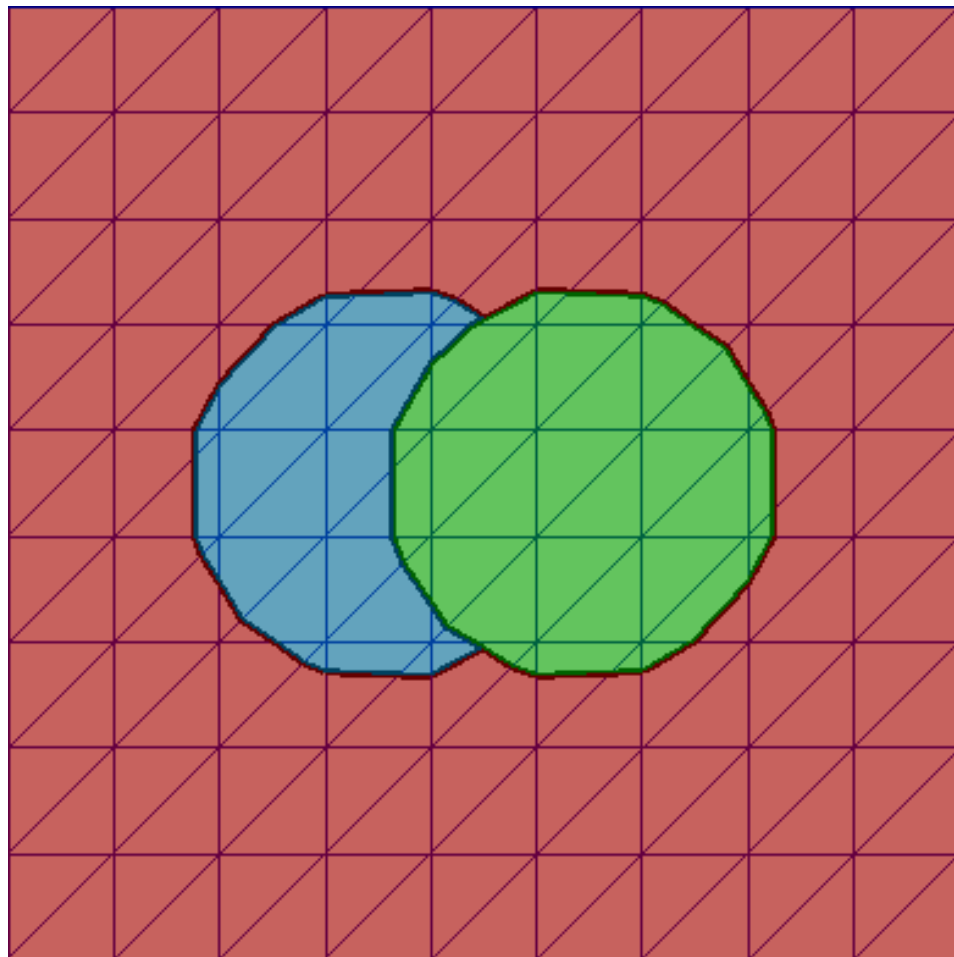


Use multiple level set functions for complex geometries

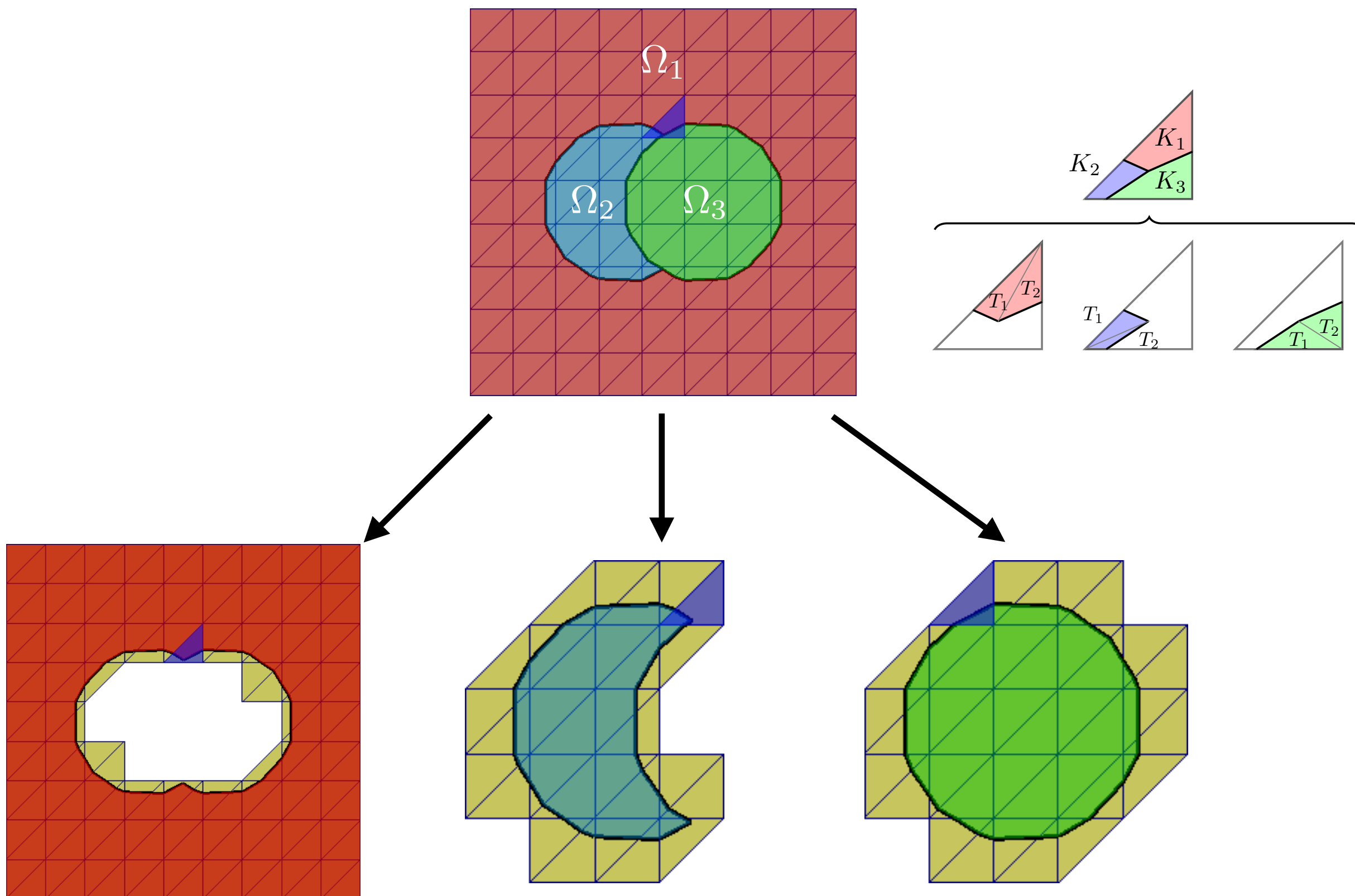
N-1 level set functions can describe N different subdomains



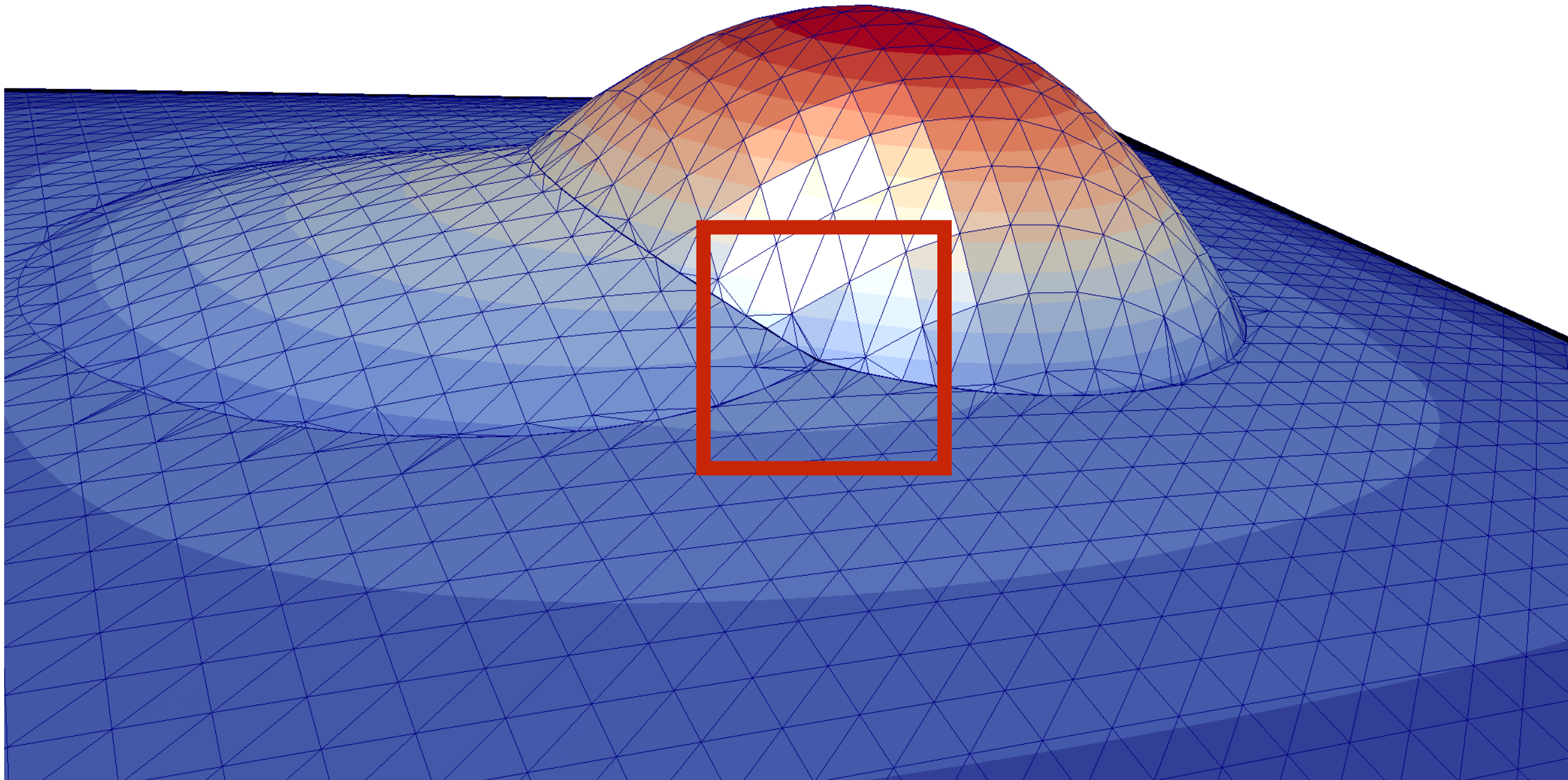
Level Set Mesh Intersection



Enrichment for jump and kink representation



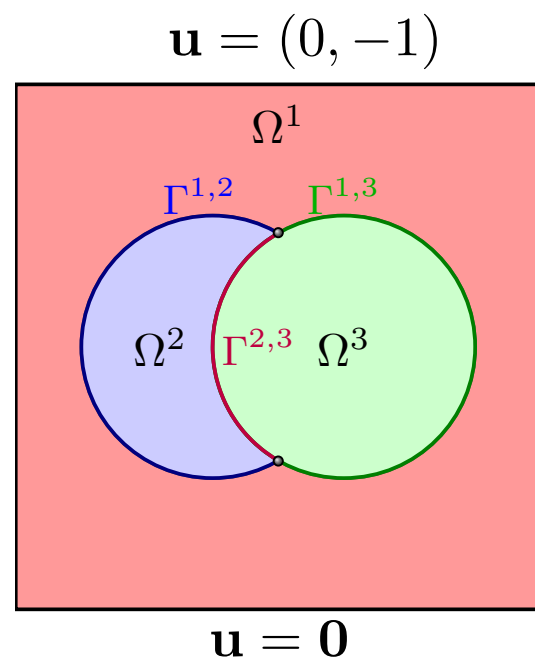
Triple Poisson Problem





Contact Problems

Contact Problem in linear Elasticity



Bulk Problem

For all Ω_i , find the displacement fields $\mathbf{u}^i : \Omega_i \rightarrow \mathbb{R}$, such that

$$-\nabla \cdot \boldsymbol{\sigma}(\mathbf{u}^i) = \mathbf{f} \quad \text{in } \Omega_i$$

$$\boldsymbol{\sigma}(\mathbf{u}^i) = \lambda^i \text{tr}(\boldsymbol{\epsilon}(\mathbf{u}^i)) \mathbf{I} + 2\mu_i \boldsymbol{\epsilon}(\mathbf{u}^i) \quad \text{in } \Omega_i$$

$$\mathbf{u}^i = \mathbf{g} \text{ on } \partial\Omega_D, \quad \boldsymbol{\sigma}(\mathbf{u}_i) \cdot \mathbf{n} = \mathbf{F}_N \text{ on } \partial\Omega_N$$

Here, $\boldsymbol{\epsilon}(\mathbf{u}) = \frac{1}{2} (\nabla \mathbf{u} + \nabla \mathbf{u}^T)$ is the strain tensor, \mathbf{f} is the body force, \mathbf{F}_N is the surface load, \mathbf{g} the Dirichlet boundary condition, λ^i and μ^i are the two Lamé coefficients (E^i is the Young's modulus, $\nu = 0.3$ is the Poisson's ratio)

$$\mu^i = \frac{E^i}{2(1+\nu)}, \quad \lambda^i = \frac{E^i \nu}{(1+\nu)(1-2\nu)}.$$

Unilateral contact for isotropic linear elasticity

Contact Conditions

For any displacement field \mathbf{u}_i , we decompose the surface traction $\mathbf{F}^i = \boldsymbol{\sigma}(\mathbf{u}_i) \cdot \mathbf{n}^{i,j}$ on the interface $\Gamma^{i,j}$ into its normal and tangential components

$$\mathbf{F}^i = \mathbf{F}_n^i + \mathbf{F}_t^i.$$

Then, the conditions of contact with Coulomb friction reads

$$(\mathbf{u}^j - \mathbf{u}^i) \cdot \mathbf{n}^{i,j} \geq 0,$$

$$\mathbf{F}^i \cdot \mathbf{n}^{i,j} \leq 0,$$

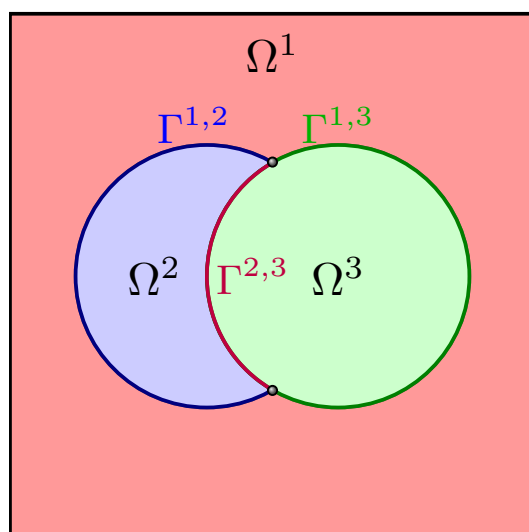
$$((\mathbf{u}^j - \mathbf{u}^i) \cdot \mathbf{n}^{i,j}) \cdot (\mathbf{F}^i \cdot \mathbf{n}^{i,j}) = 0,$$

$$\|\mathbf{F}_t^i\| \leq c \mathbf{F}^i \cdot \mathbf{n}^{i,j} \quad \text{if } \|\hat{\mathbf{g}}_t^i\|_2 = 0$$

$$\mathbf{F}_t^i = -c \mathbf{F}^i \cdot \mathbf{n}^{i,j} \frac{\hat{\mathbf{g}}_t^i}{\|\hat{\mathbf{g}}_t^i\|_2} \quad \text{if } \|\hat{\mathbf{g}}_t^i\|_2 > 0$$

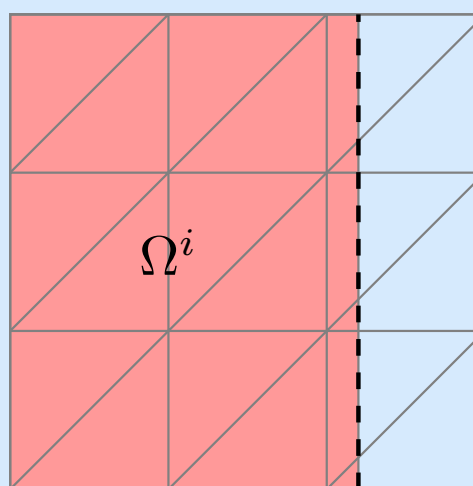
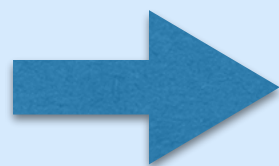
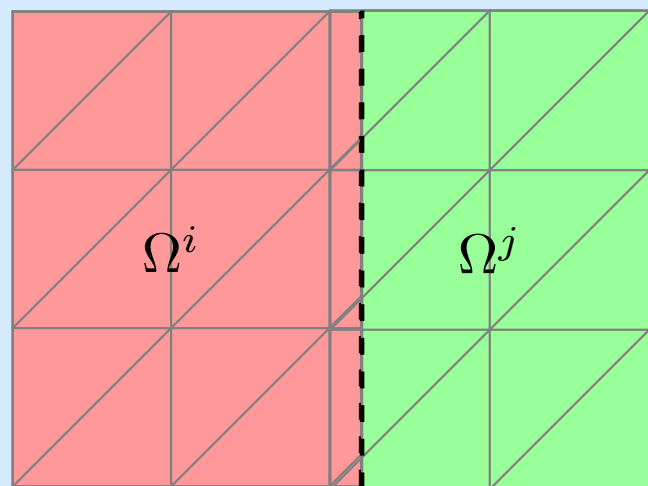
Here, $\mathbf{n}^{i,j}$ is the normal pointing from Ω_i to Ω_j , c is the Coulomb friction coefficient, and $\hat{\mathbf{g}}_t^i := (\mathbf{I} - \mathbf{n}^{i,j} \otimes \mathbf{n}^{i,j}) \cdot (\dot{\mathbf{u}}^j - \dot{\mathbf{u}}^i)$ is the relative tangential velocity

$$\mathbf{u} = (0, -1)$$

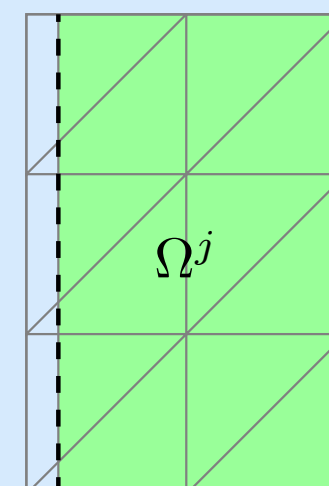
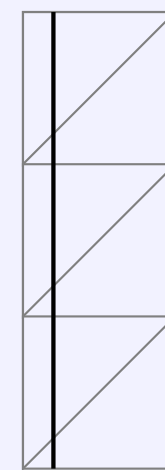


$$\mathbf{u} = \mathbf{0}$$

LaTin Algorithm



$\Gamma^{i,j}$

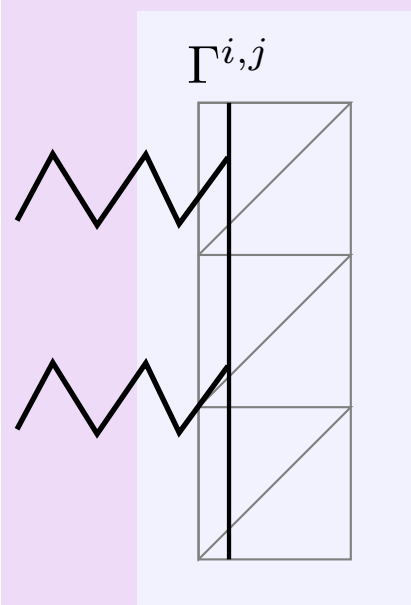


Linear Elasticity

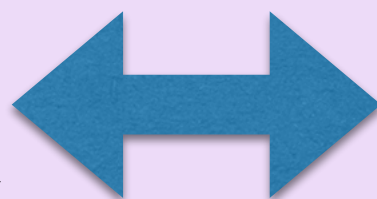
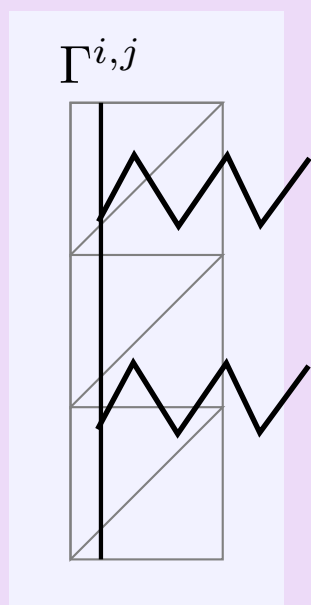
Contact

Linear Elasticity

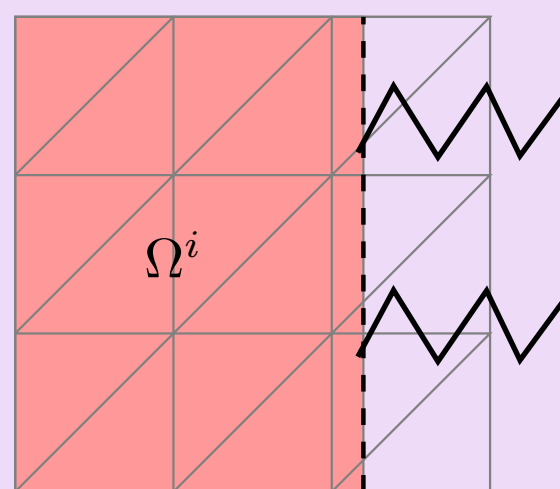
Local Stage



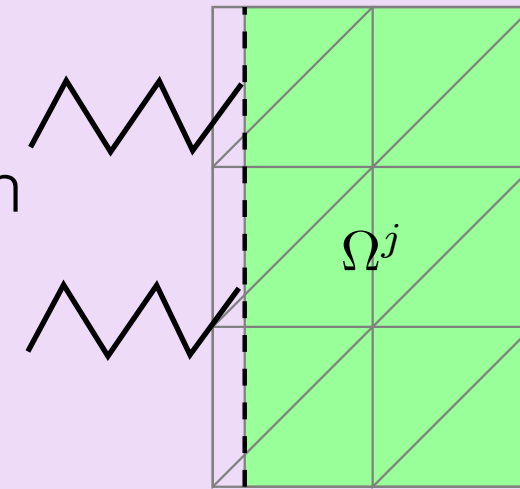
Contact Law



Linear Stage



Robin BC



$$\sigma(\mathbf{u}^i) \cdot \mathbf{n} = \mathbf{F}^i \text{ on } \Gamma^{i,j}$$

$$\mathbf{u}^i = \mathbf{w}^i \text{ on } \Gamma^{i,j}$$

$$\mathbf{F}^i + k^- \mathbf{w}^i = \hat{\mathbf{F}}^i + k^- \hat{\mathbf{w}}^i$$

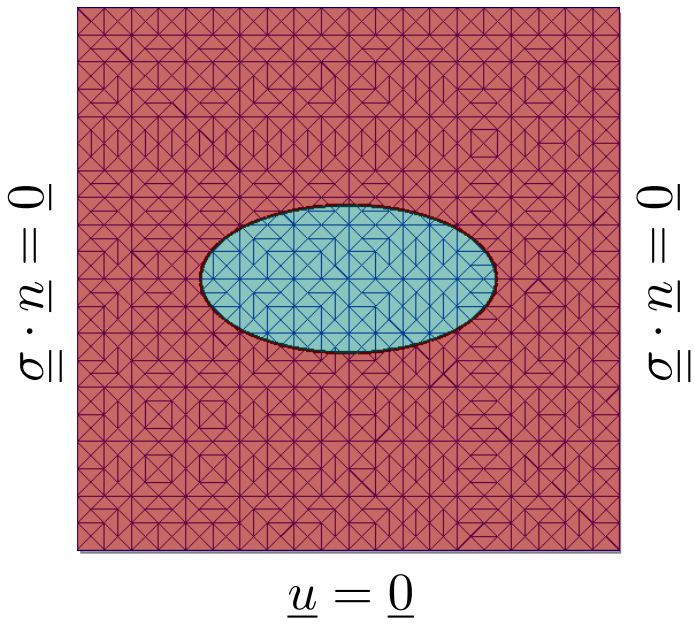
$$\hat{\mathbf{F}}^i - k^+ \hat{\mathbf{w}}^i = \mathbf{F}^i - k^+ \mathbf{w}^i$$

where $\hat{\mathbf{F}}^i$ and $\hat{\mathbf{w}}^i$ satisfy contact

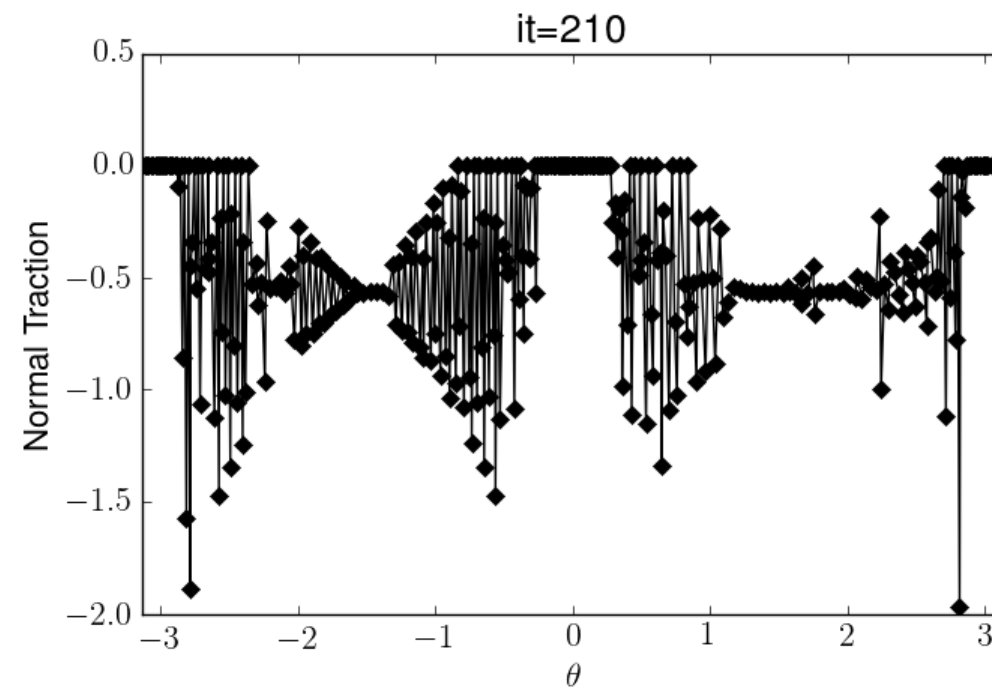
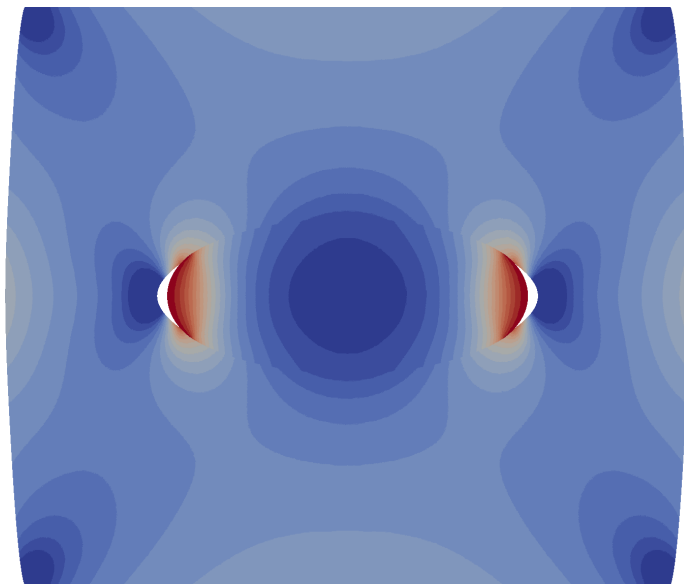
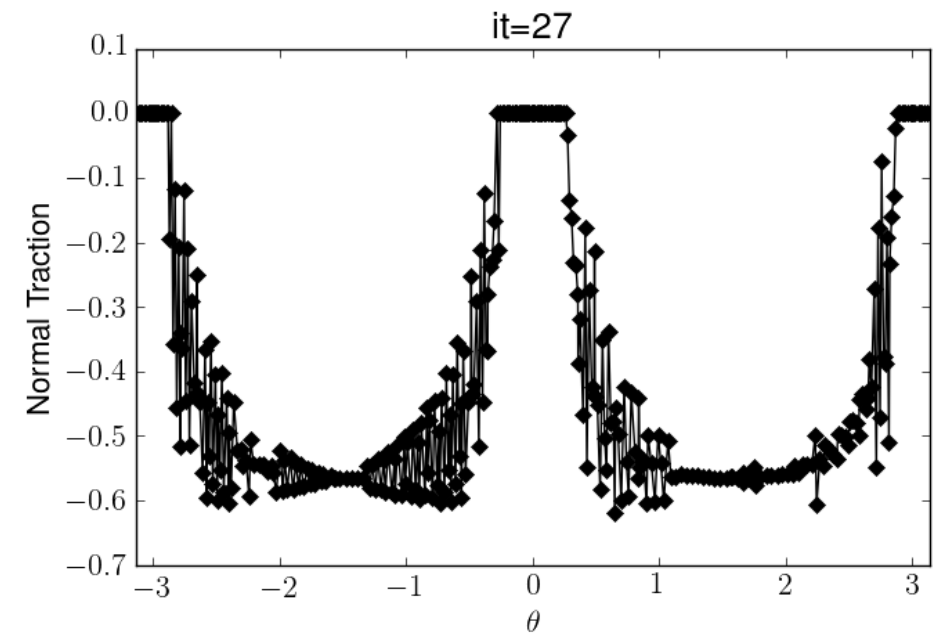
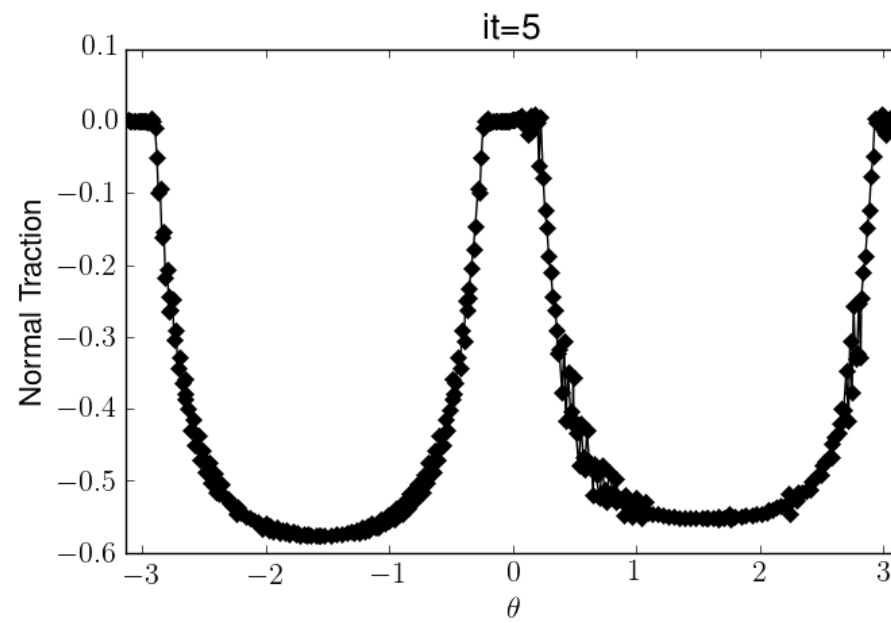
LaTin Algorithm: Stability



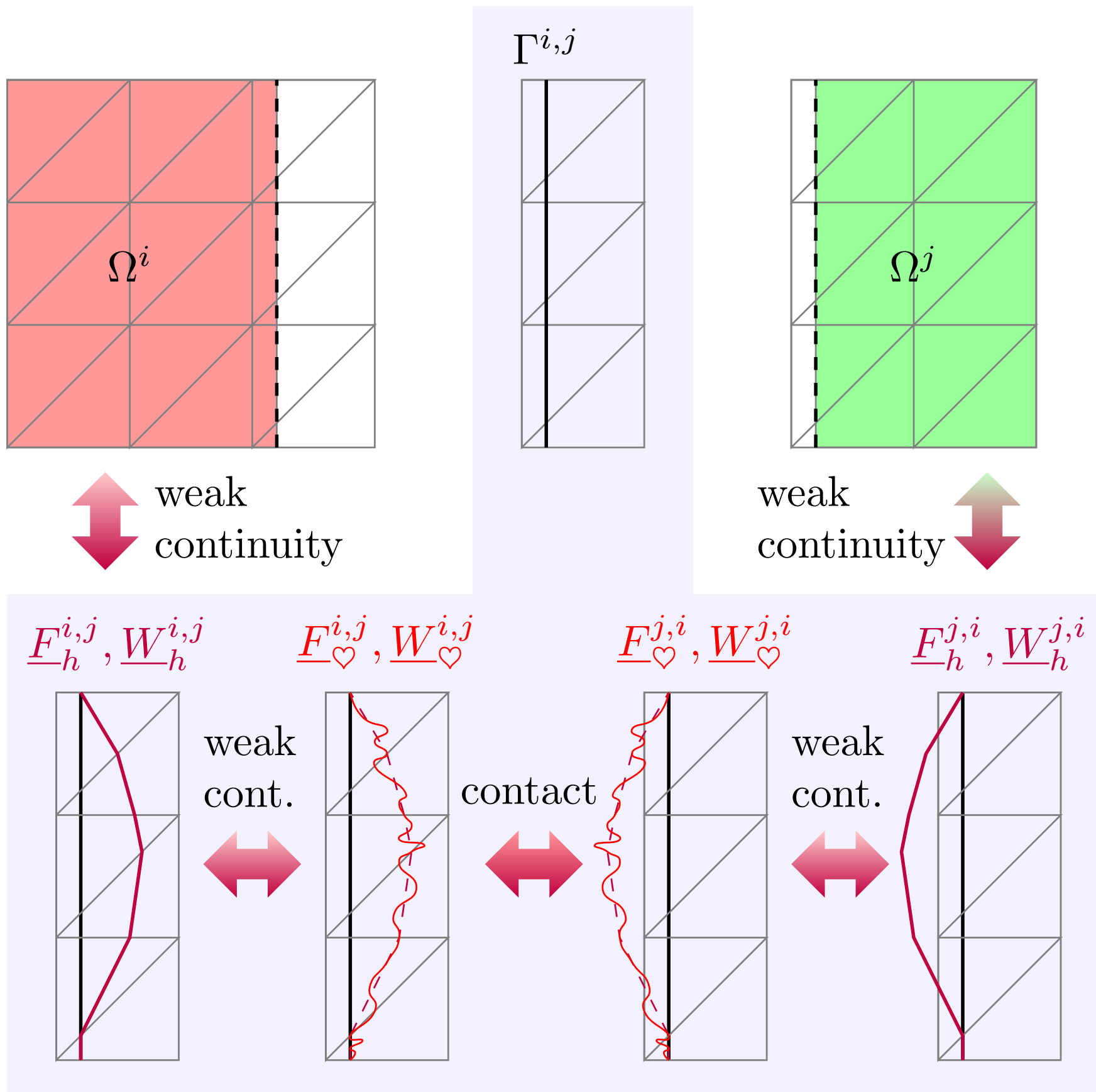
$$\underline{u} = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$



P1/P0 scheme polluted \mathbf{F}^i



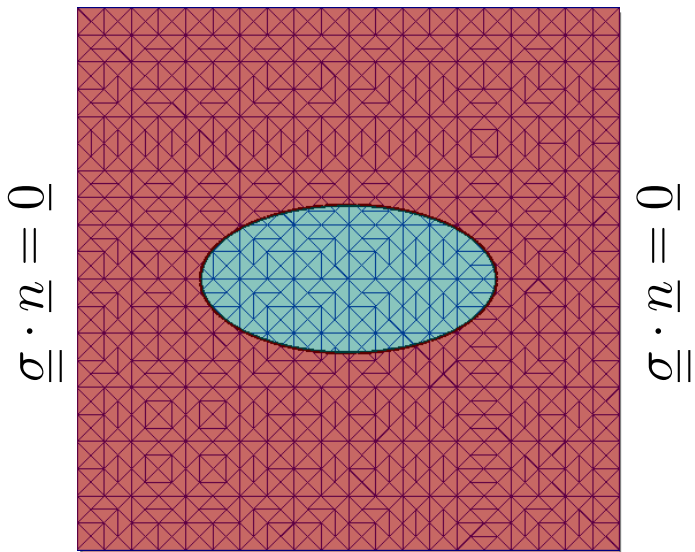
P1/P1 Stabilised Projection



LaTin Algorithm: Stability

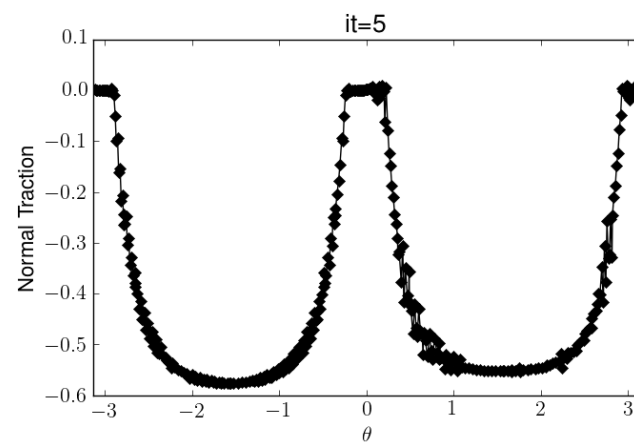


$$\underline{u} = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$

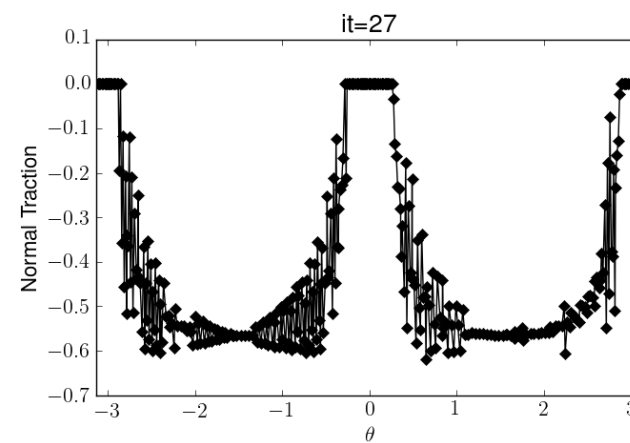


$$\underline{u} = \underline{0}$$

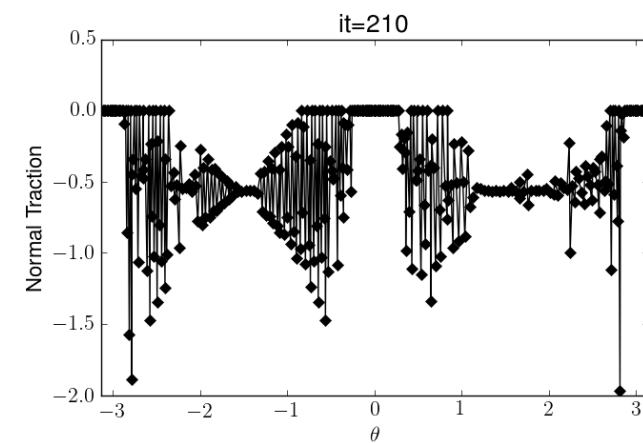
P1/P0 scheme polluted



(a) P1/P0

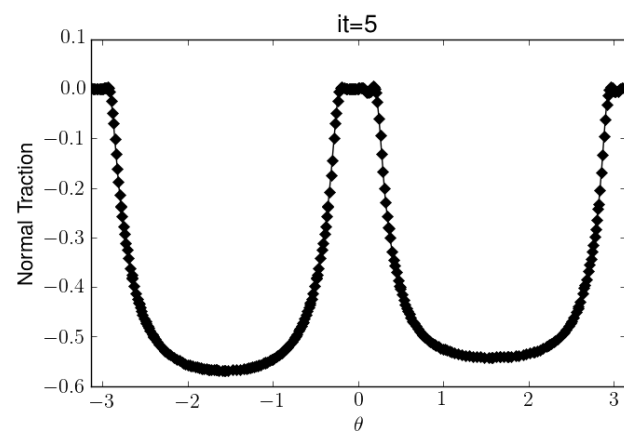
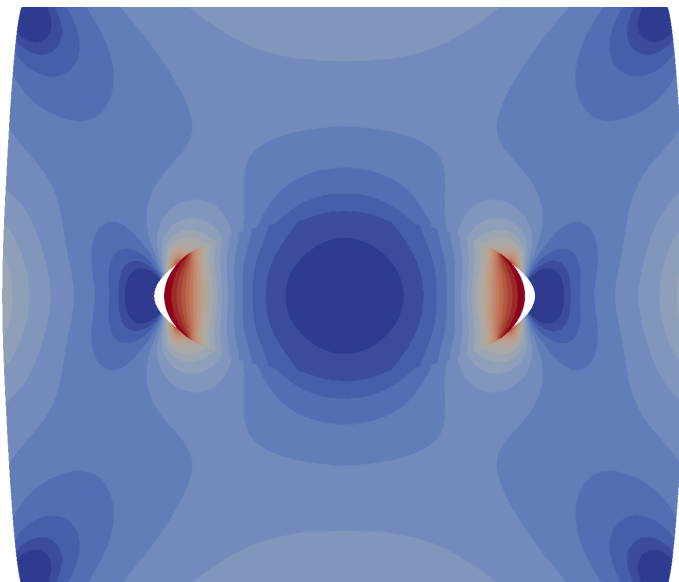


(b) P1/P0

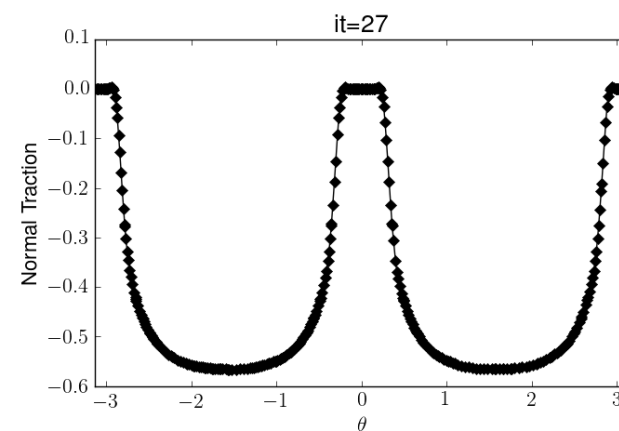


(c) P1/P0

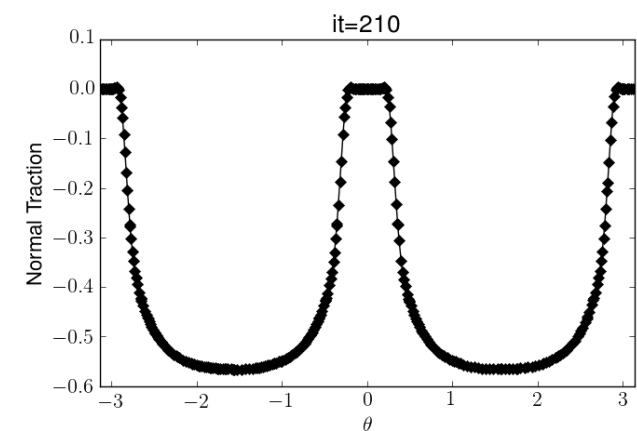
Proposed Solution: P1/P1 scheme with stabilisation



(d) P1/P1



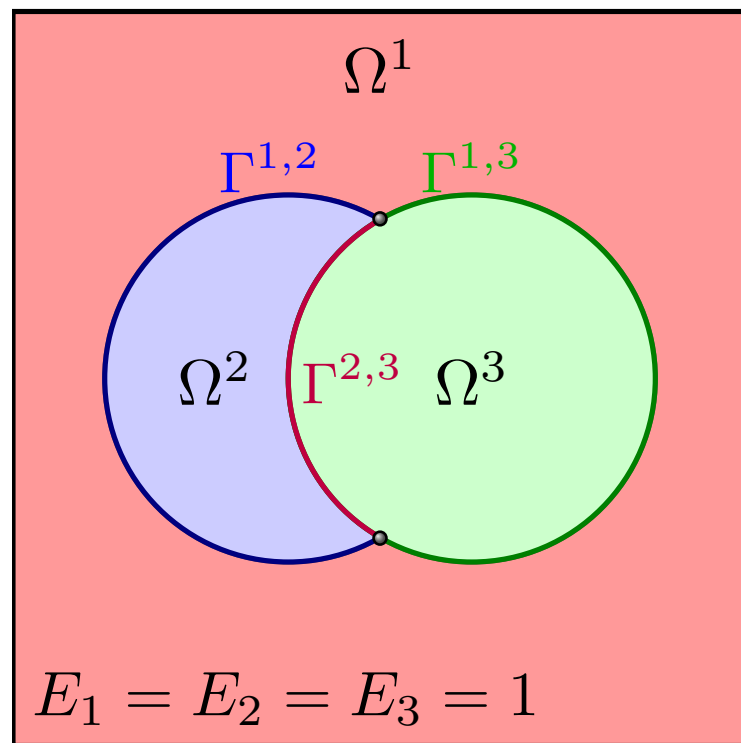
(e) P1/P1



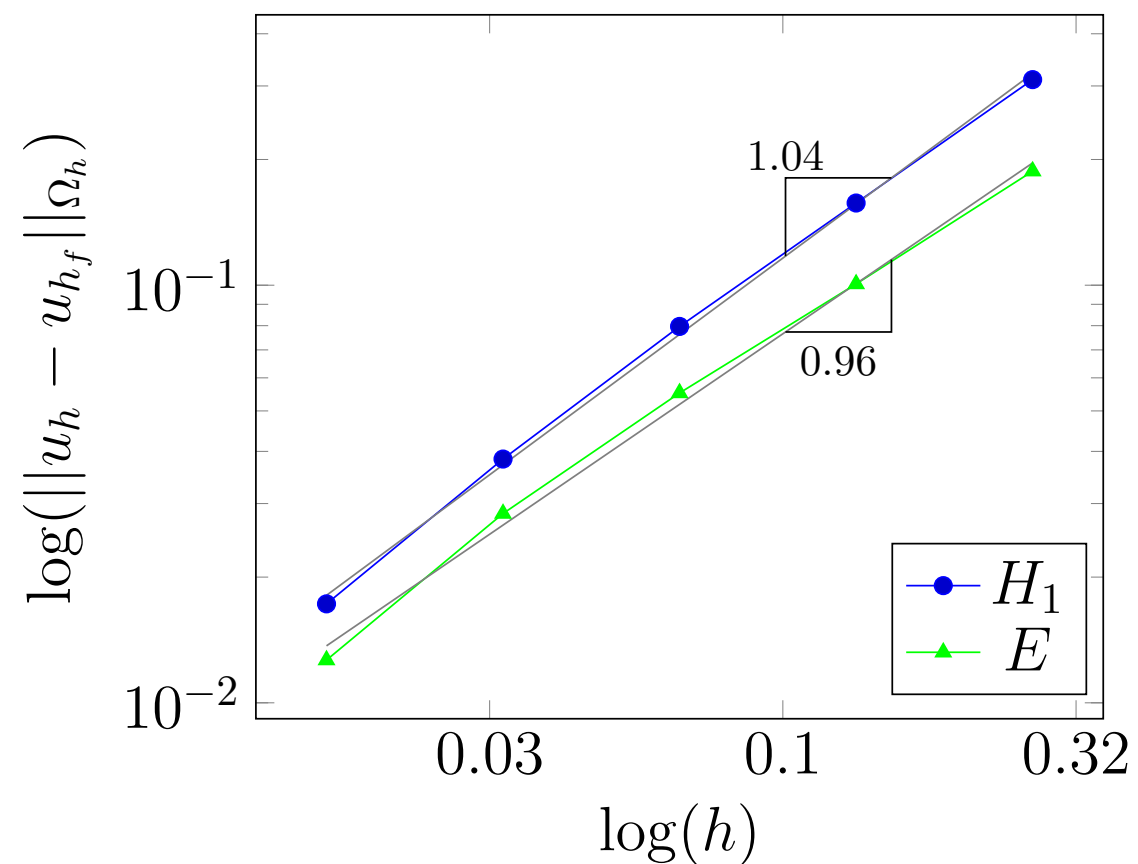
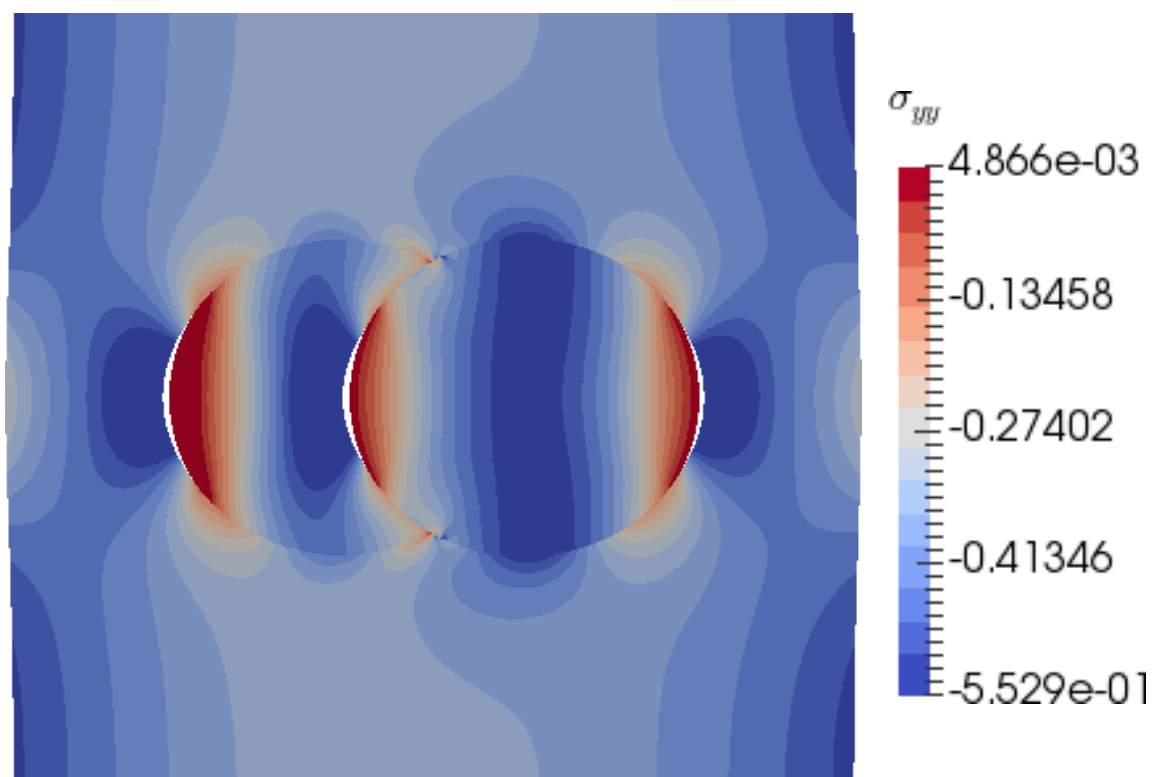
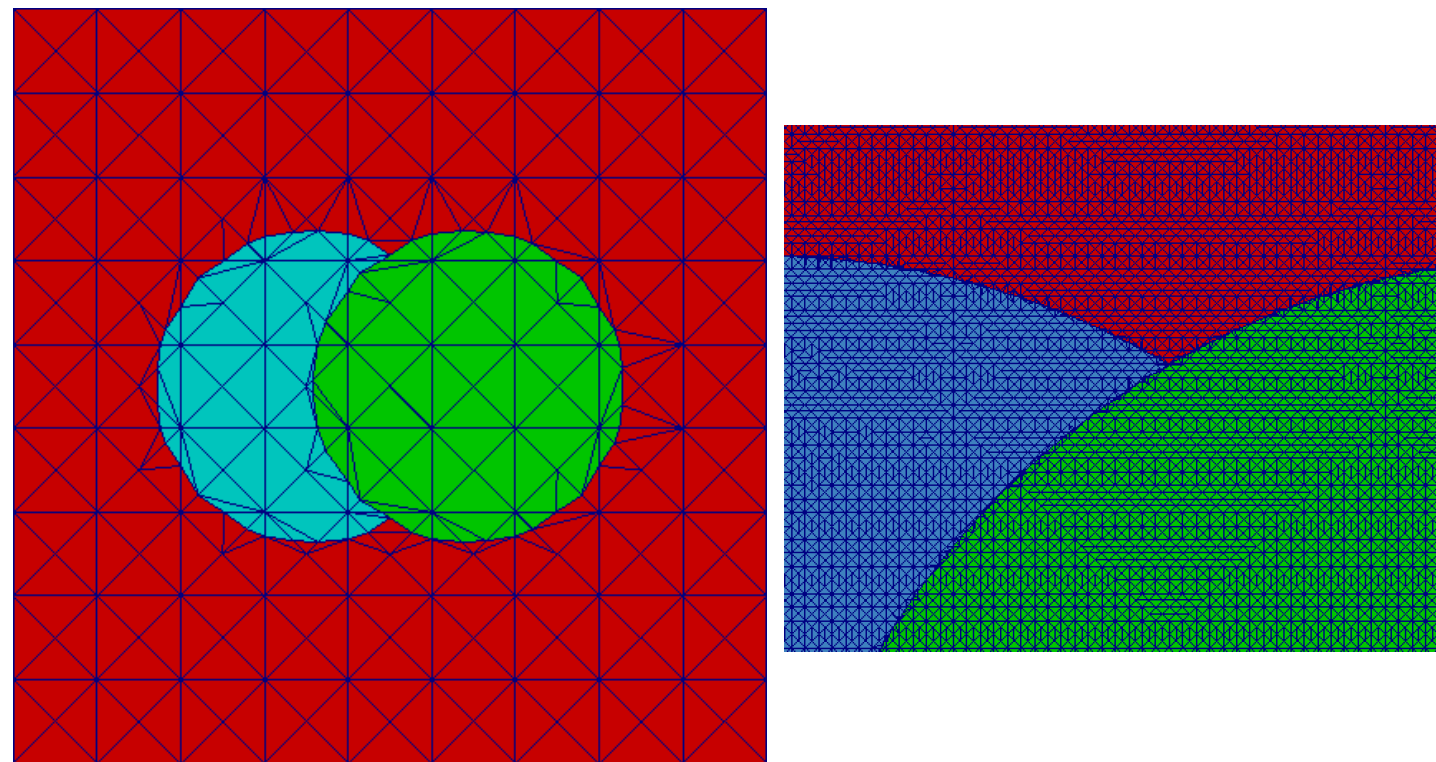
(f) P1/P1

Two Inclusions Frictionless Contact

$$\mathbf{u} = (0, -1)$$



$$\mathbf{u} = \mathbf{0}$$





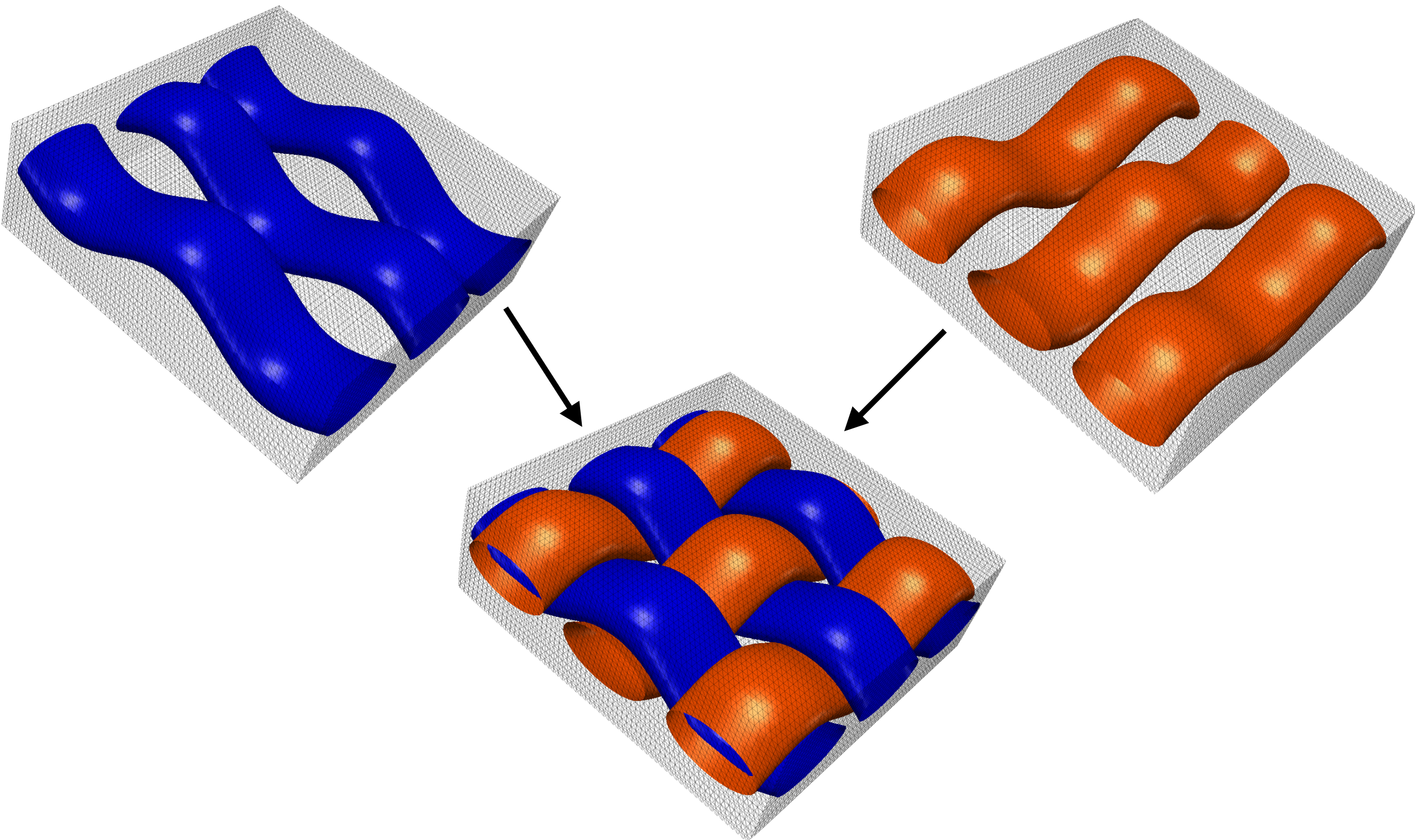
Applications in Engineering

Braided Composite

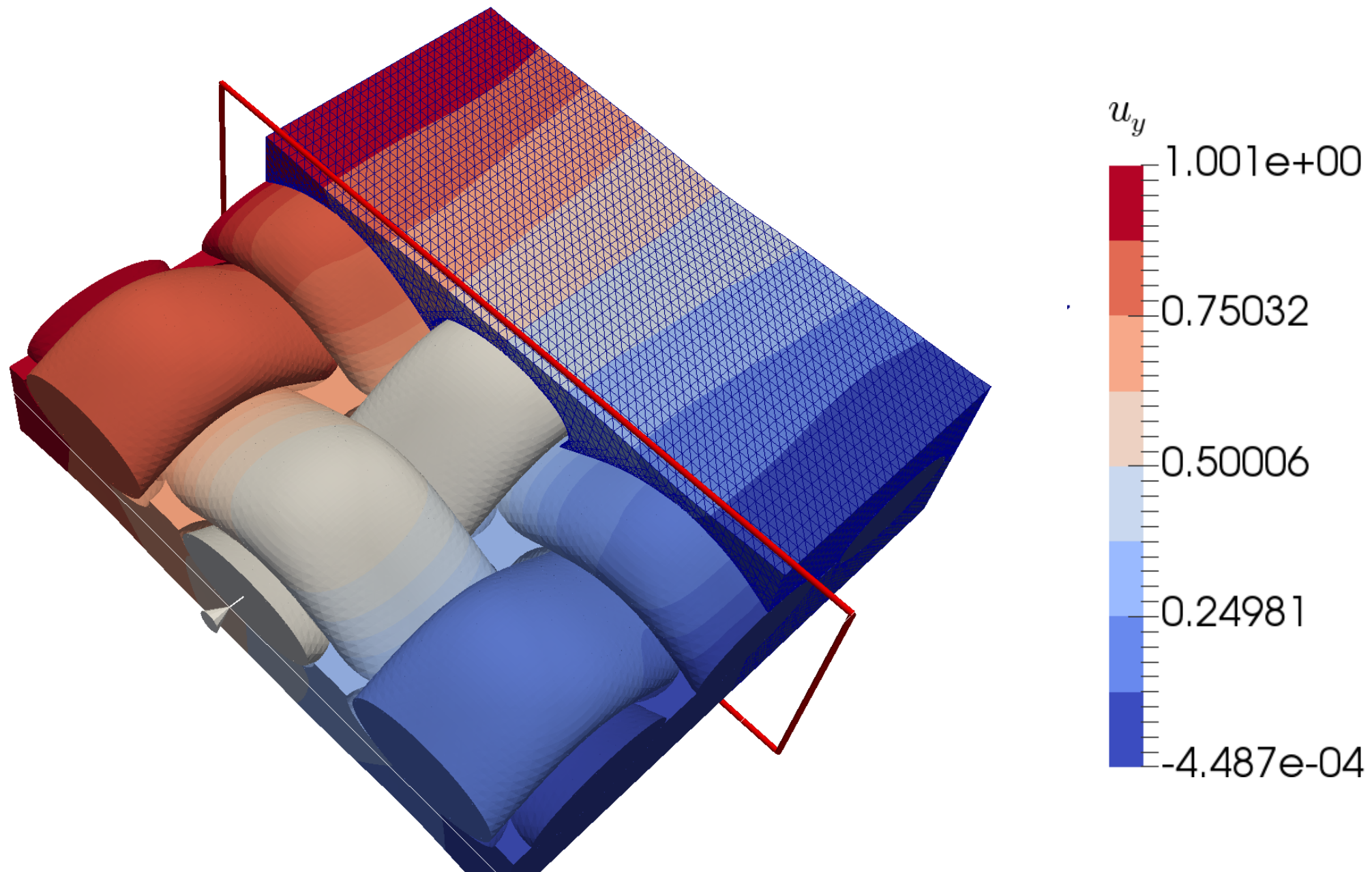


$\phi_1(\mathbf{x})$

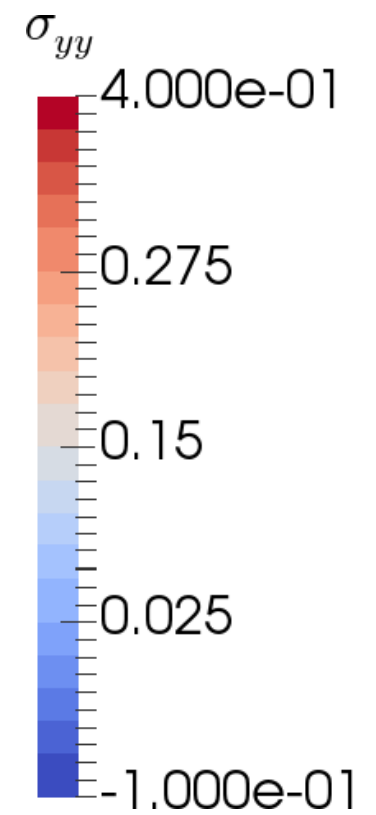
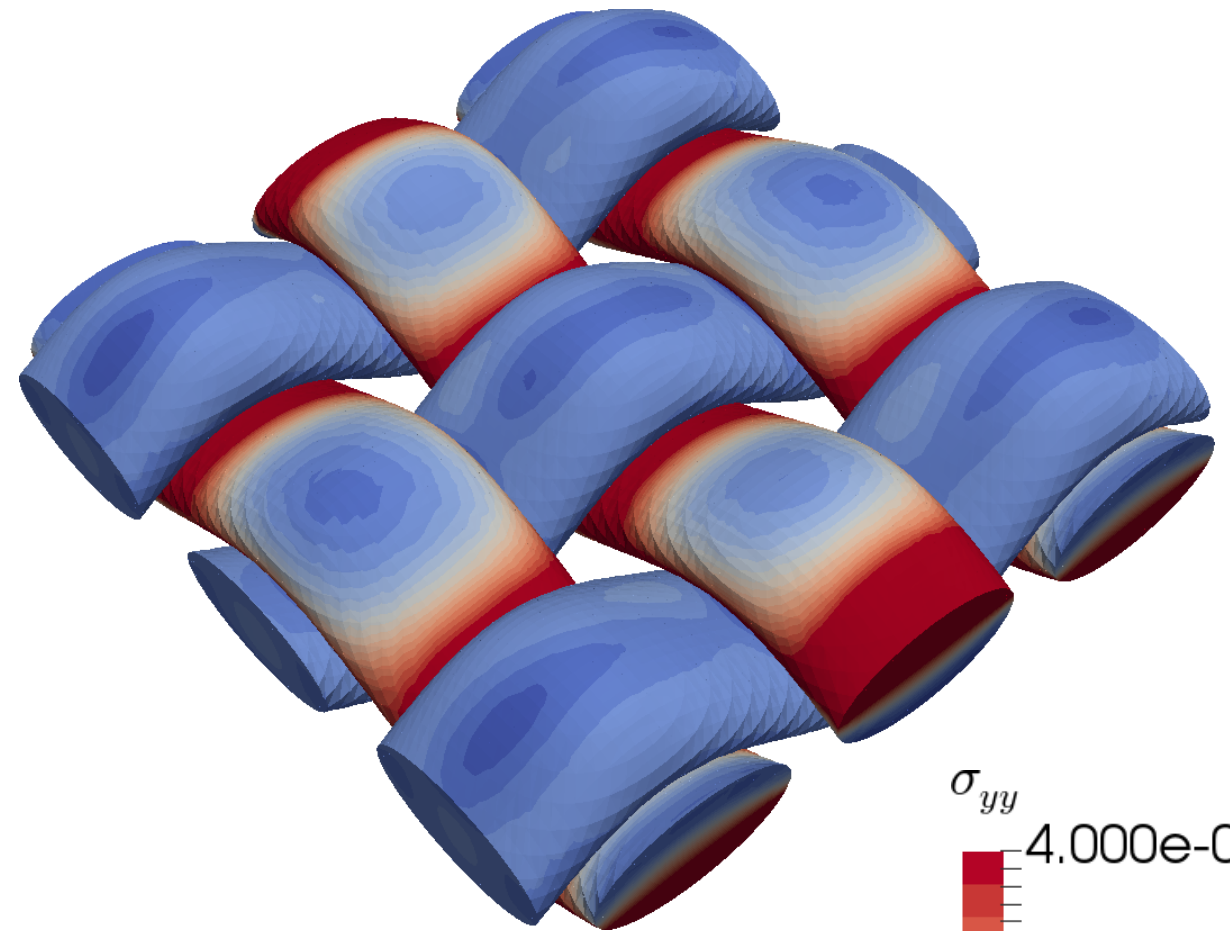
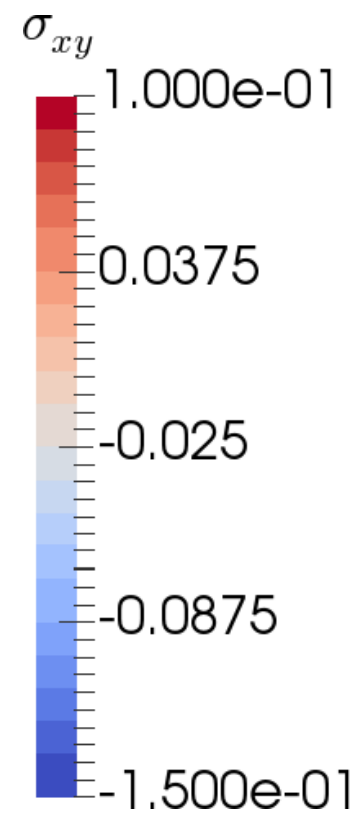
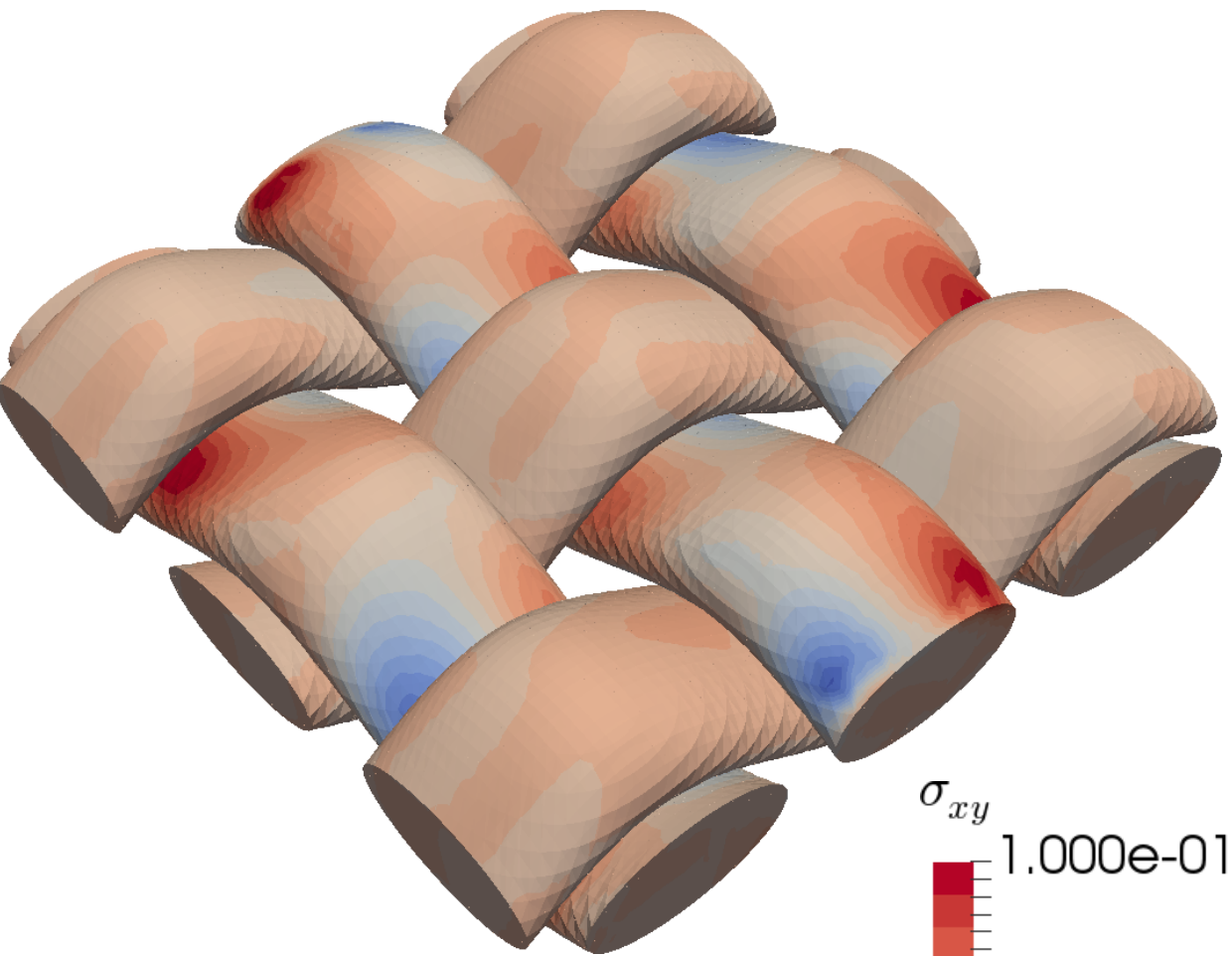
$\phi_2(\mathbf{x})$



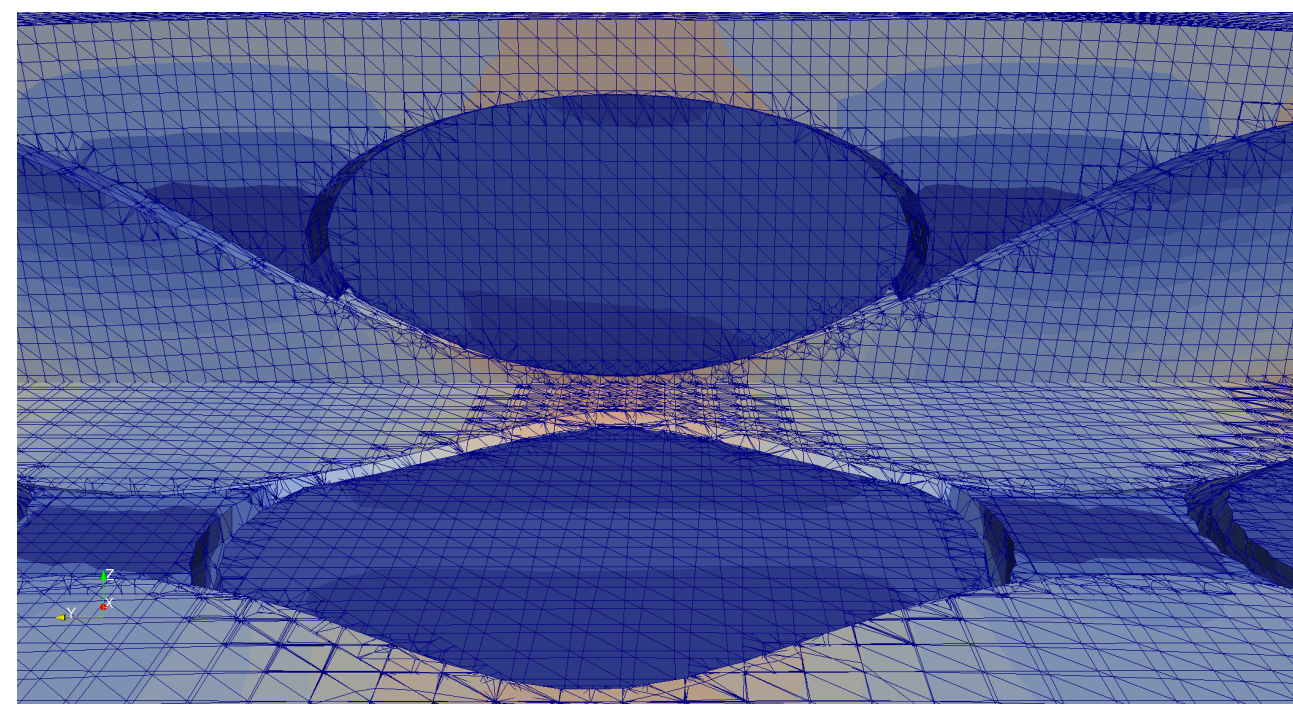
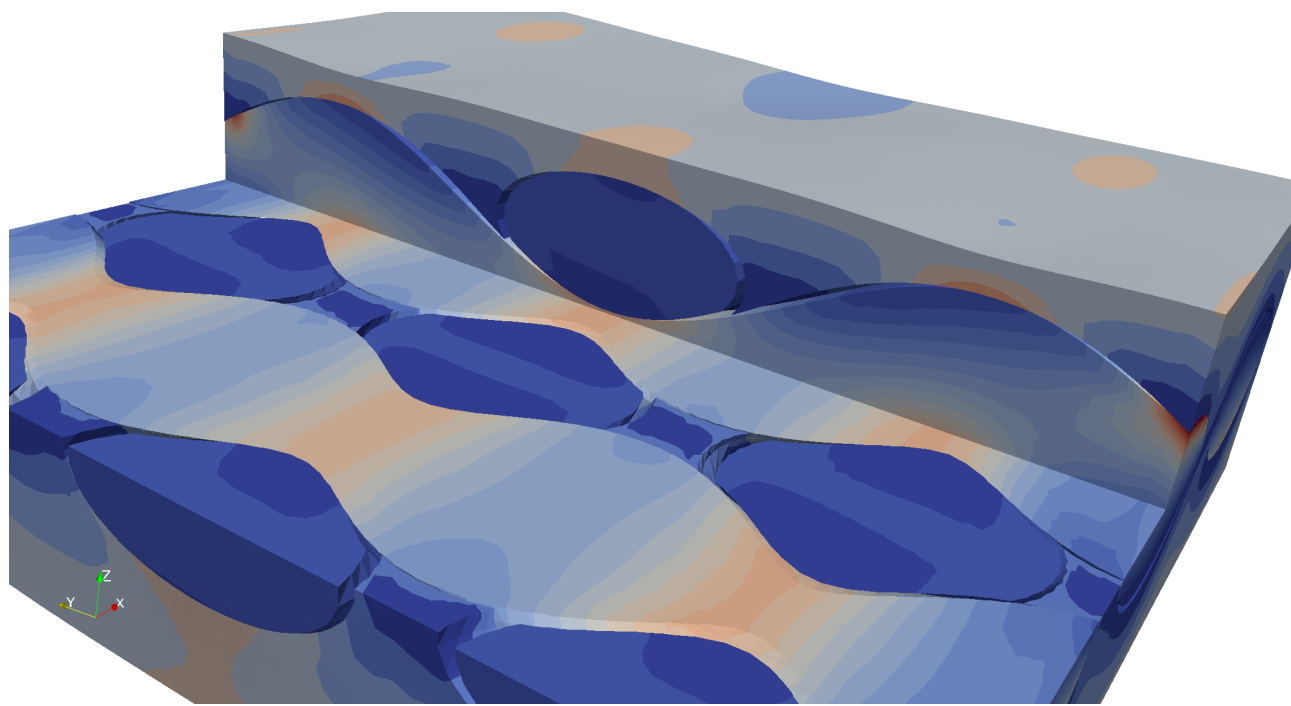
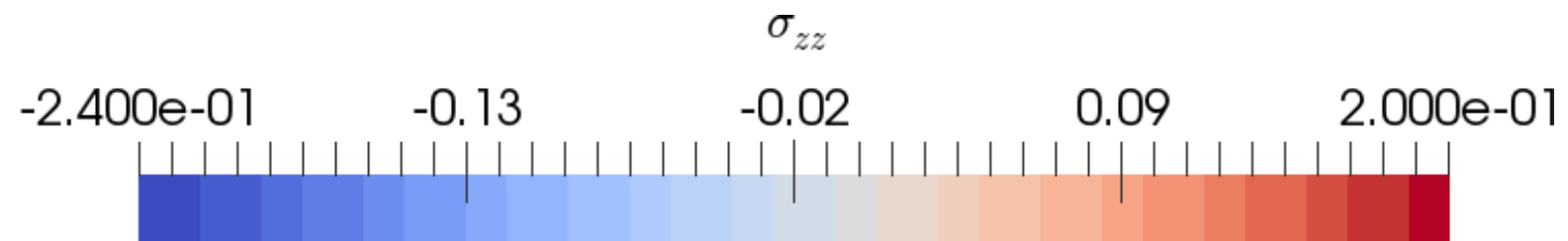
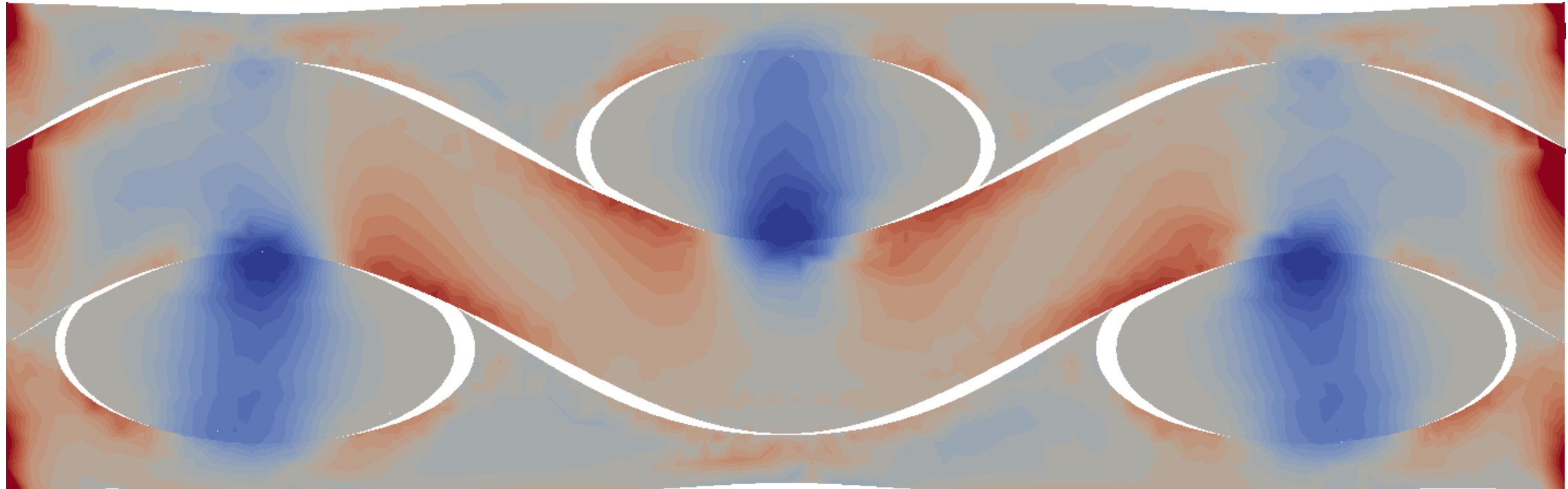
Braided Composite



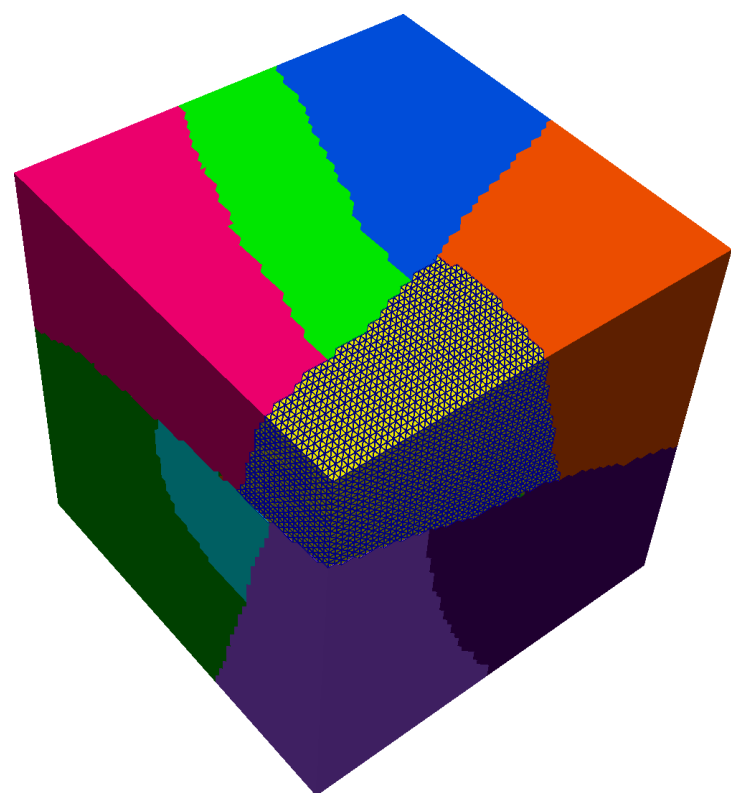
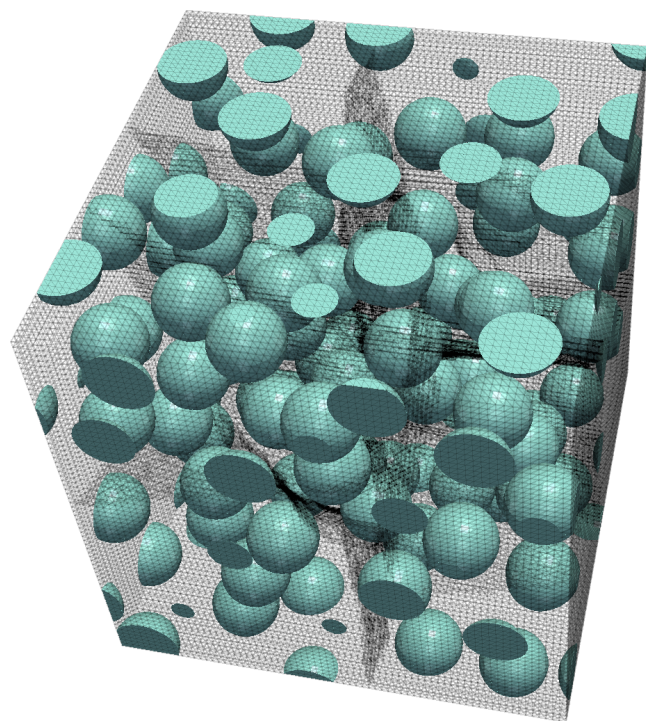
Braided Composite



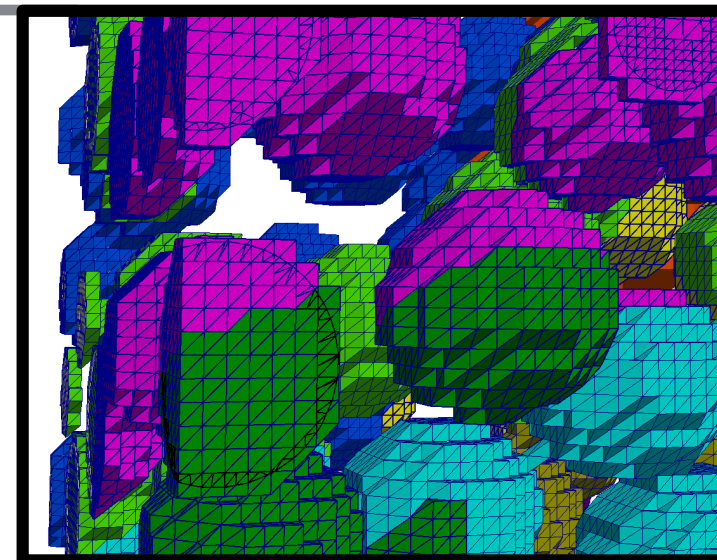
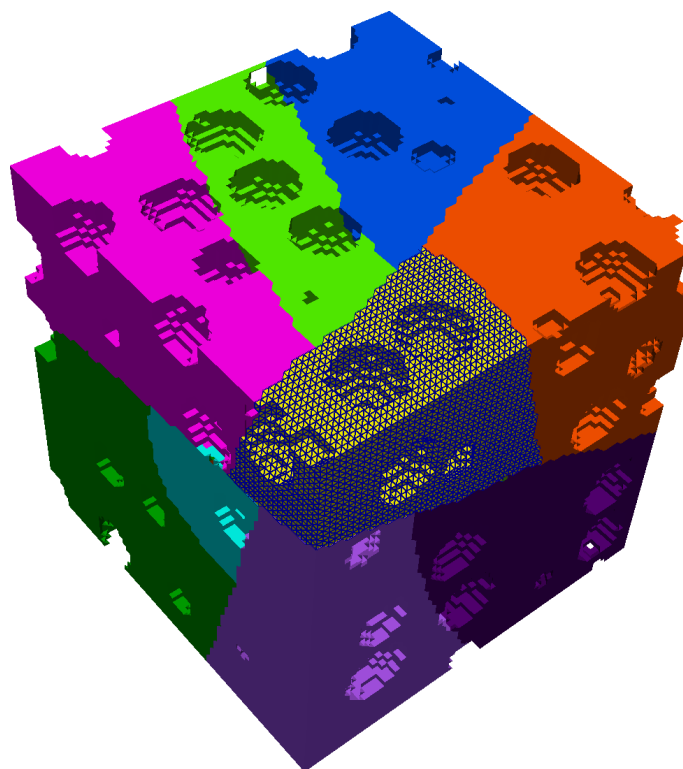
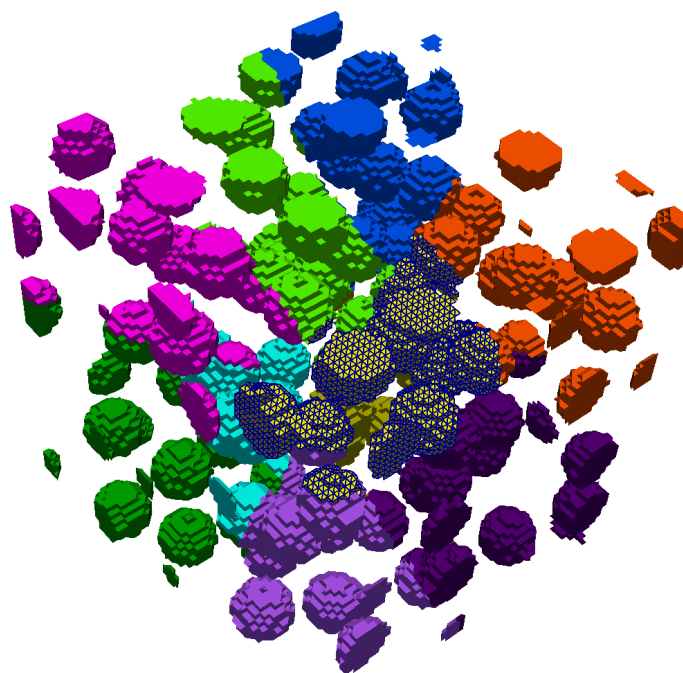
Braided Composite



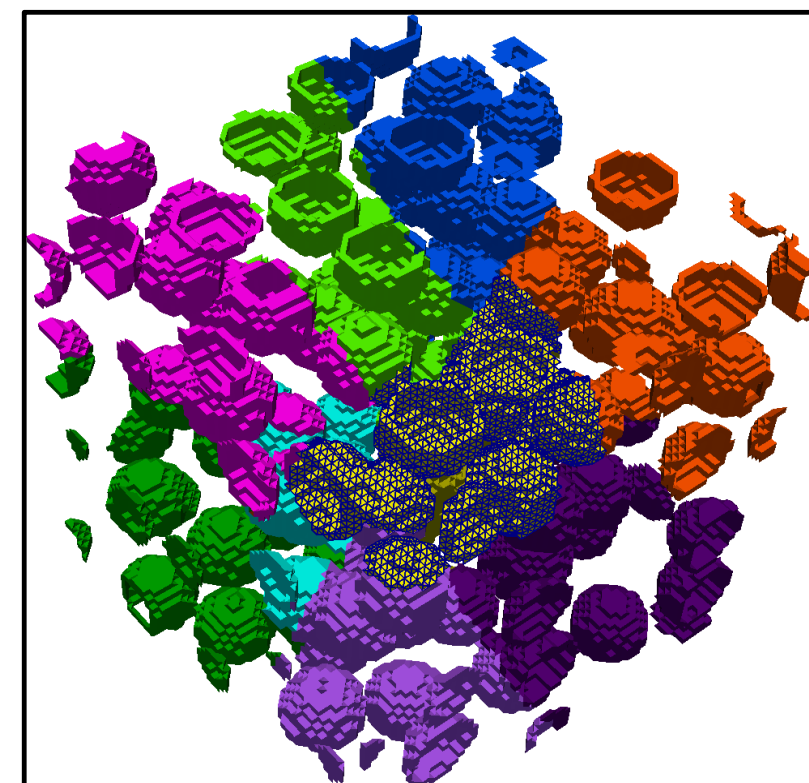
Damage in Concrete: Parallelisation



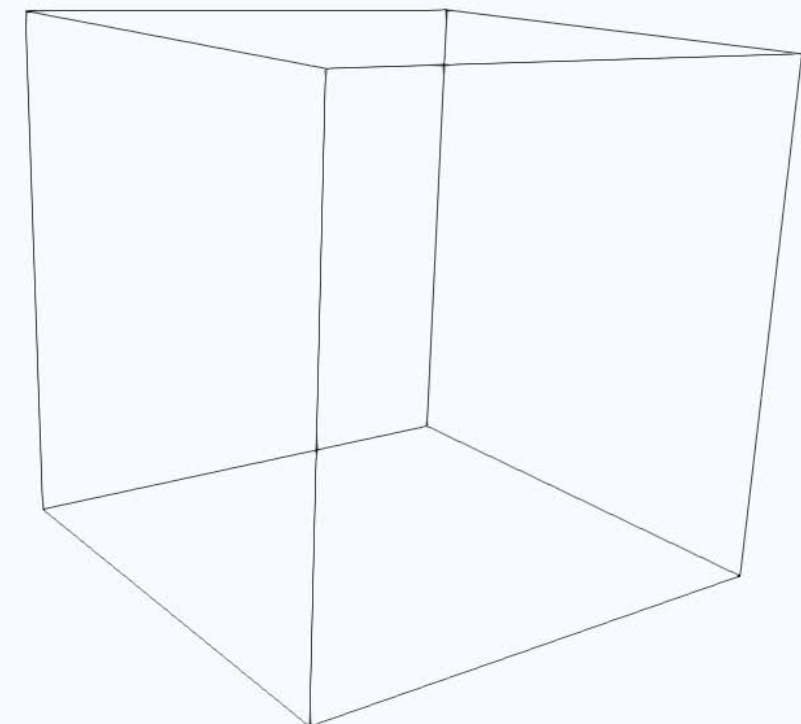
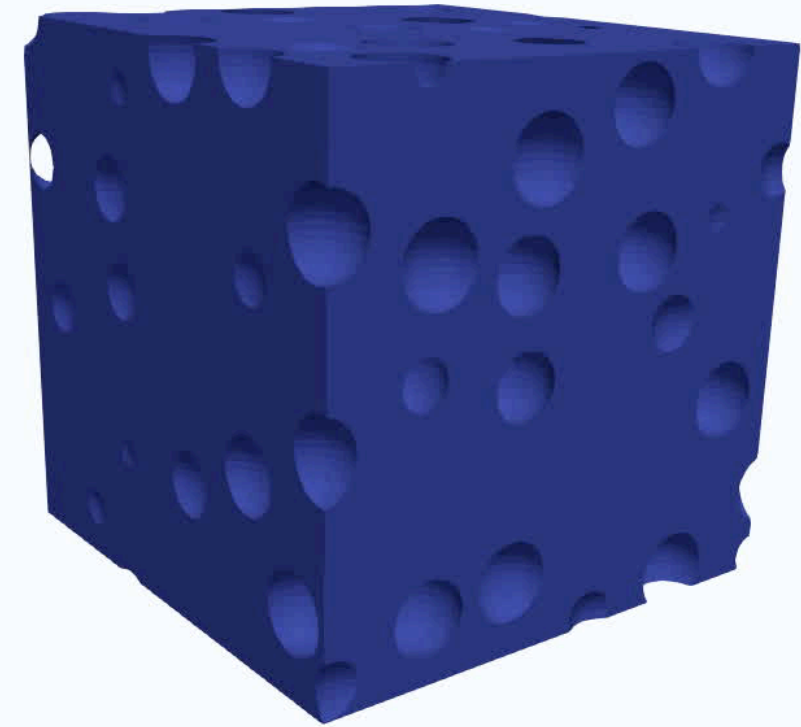
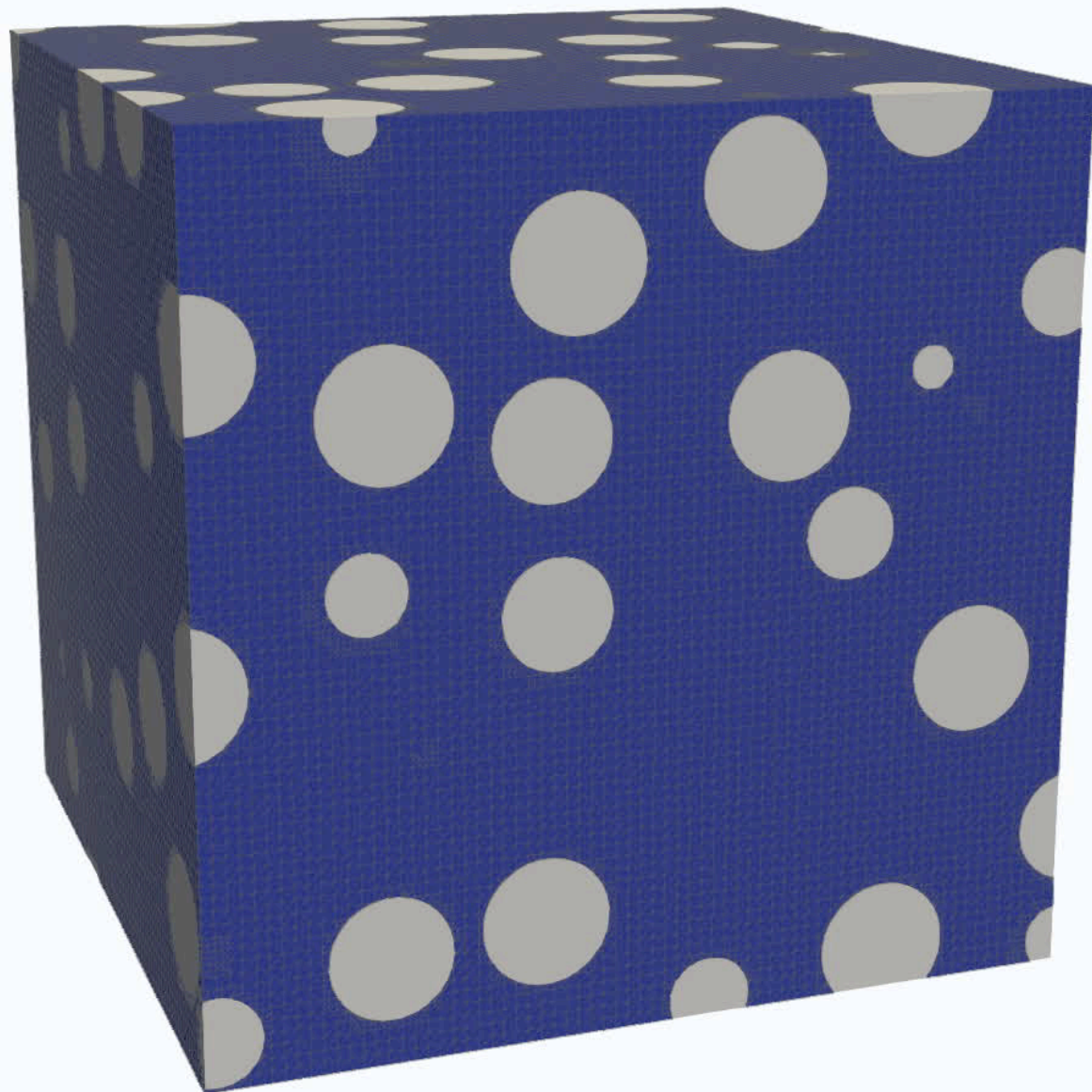
Linear Stage

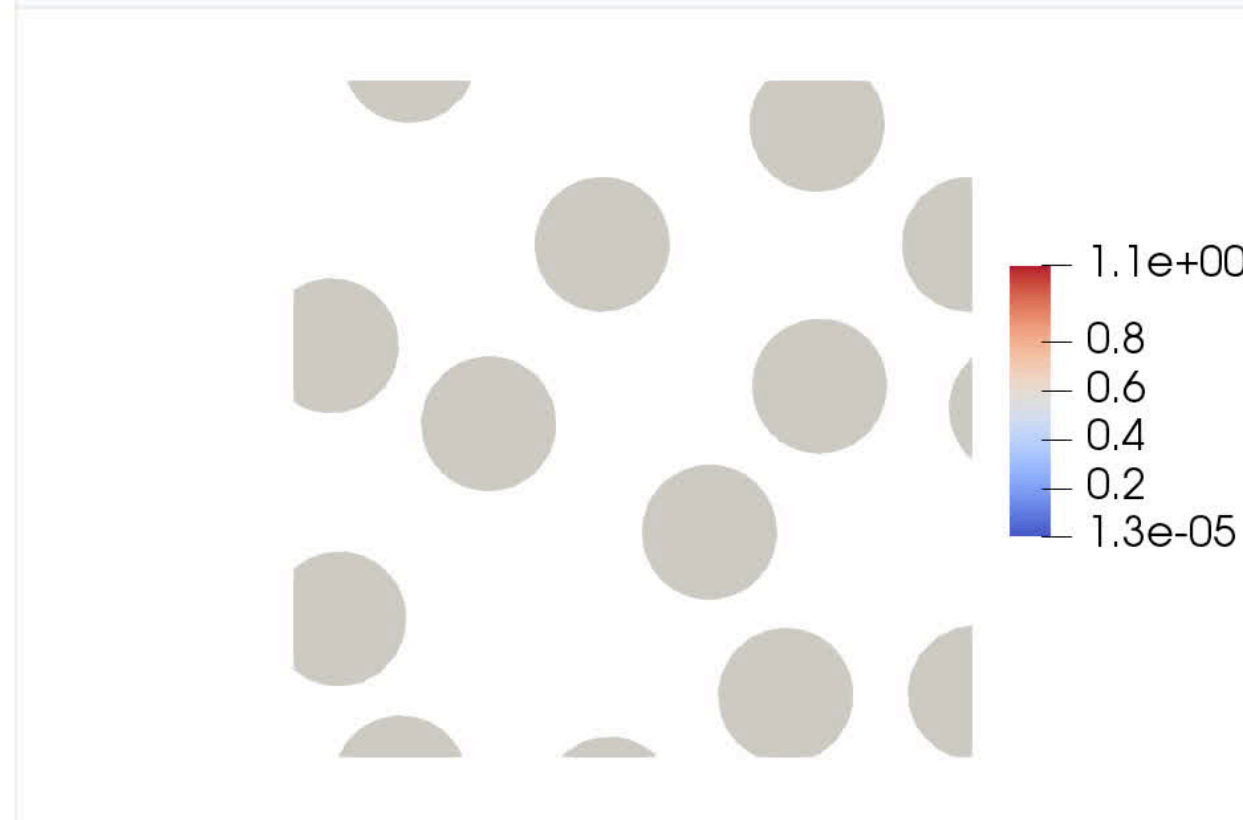
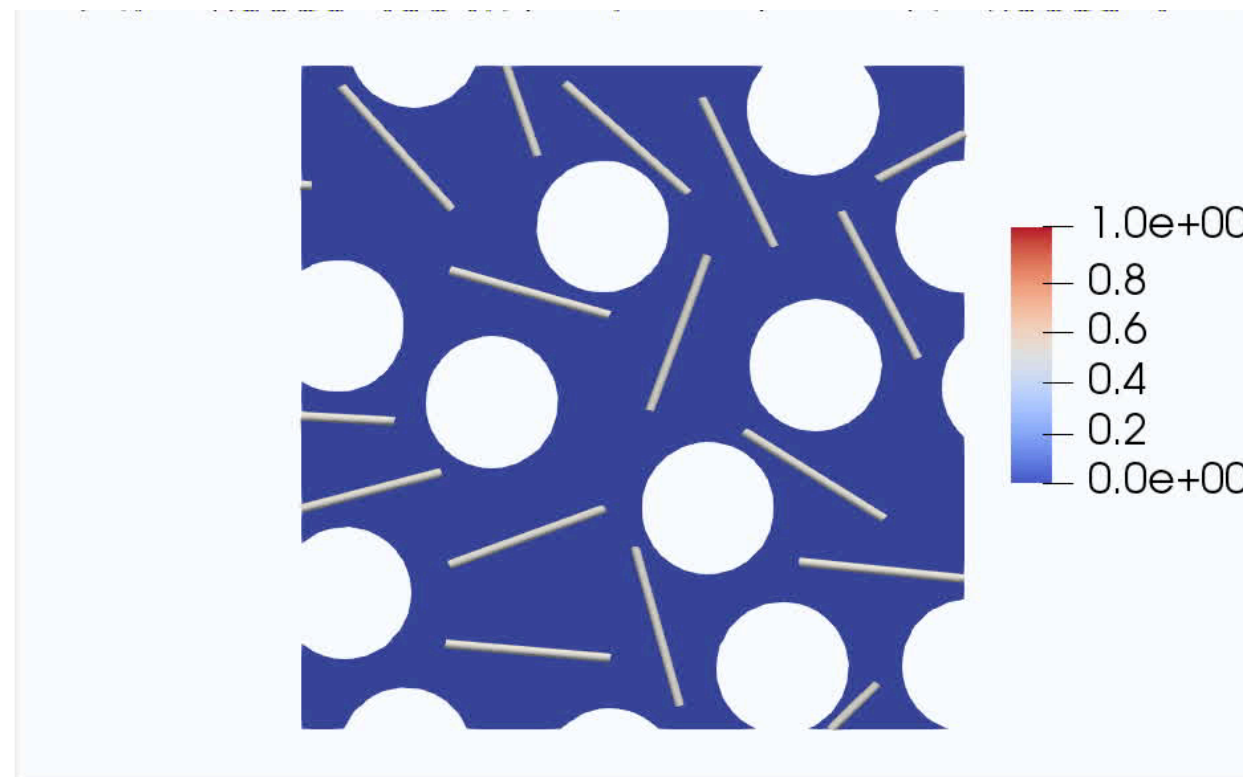
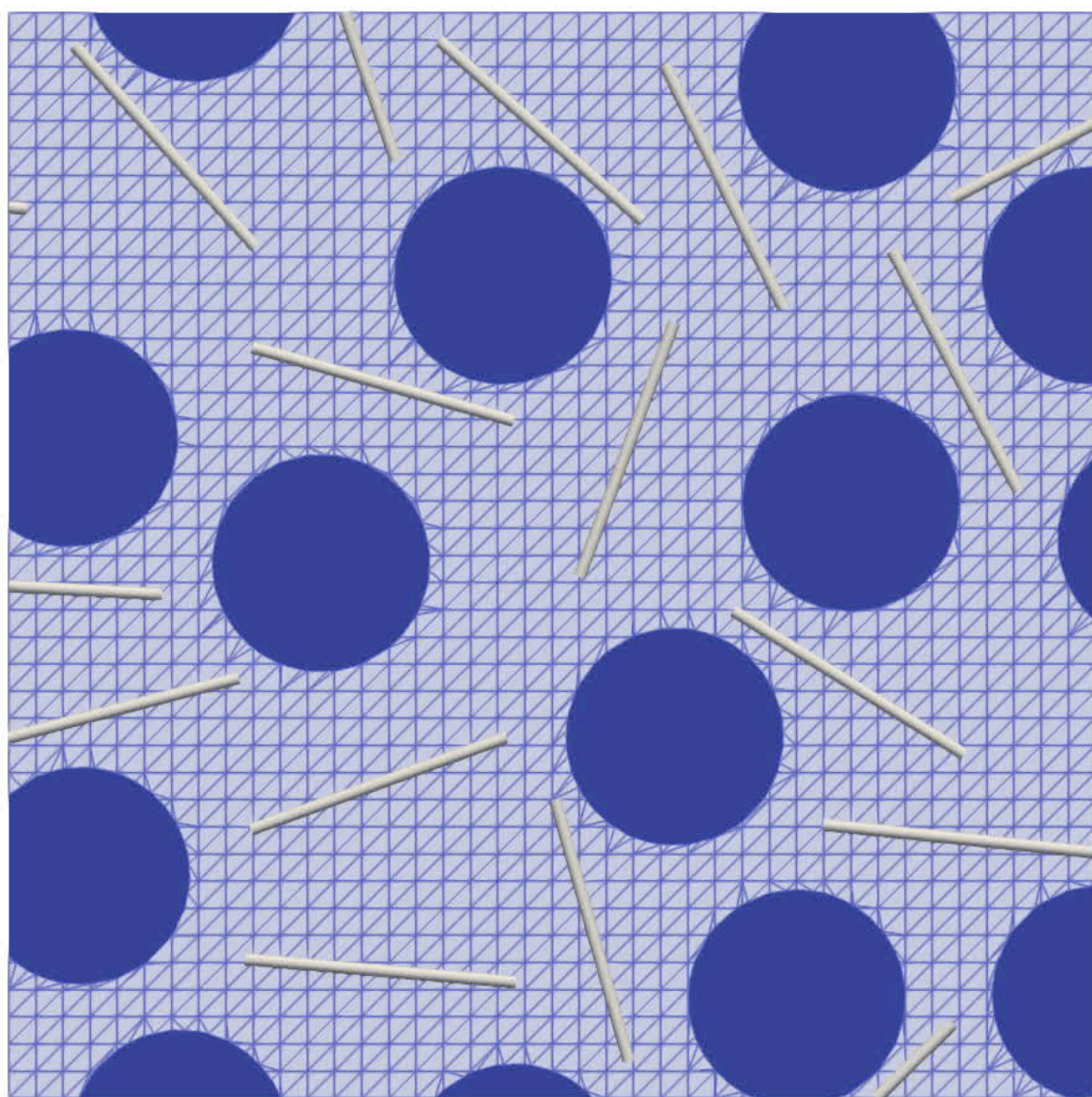


Local Stage

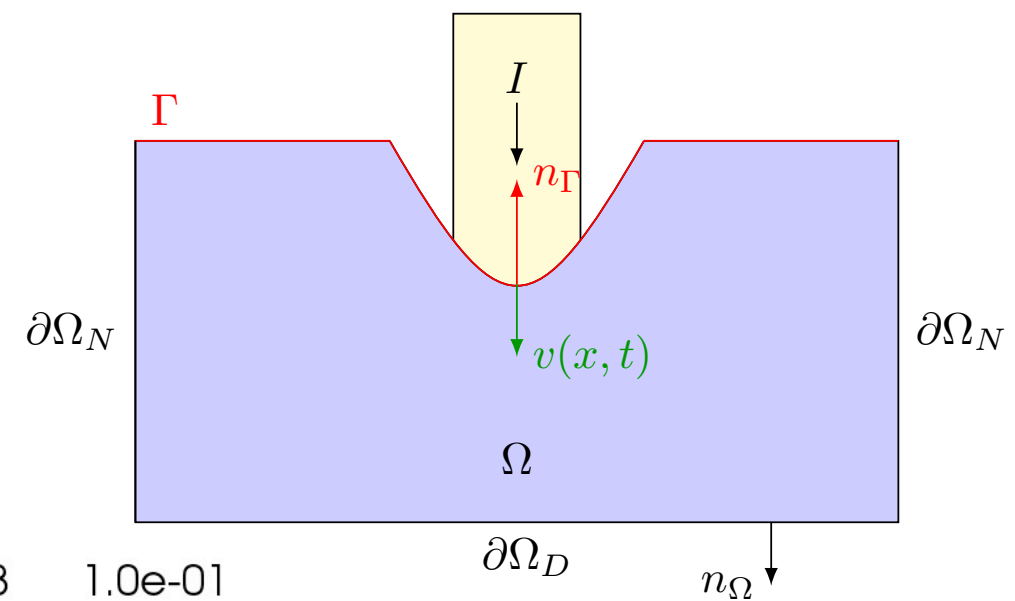
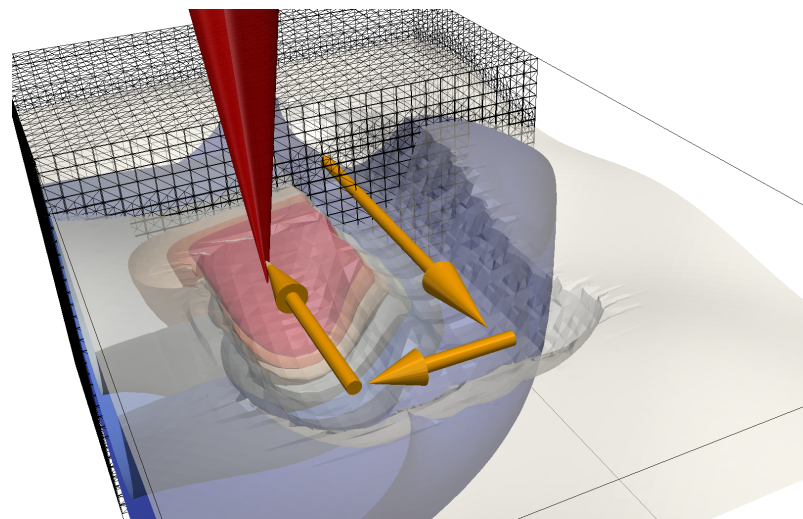


Damage in Concrete

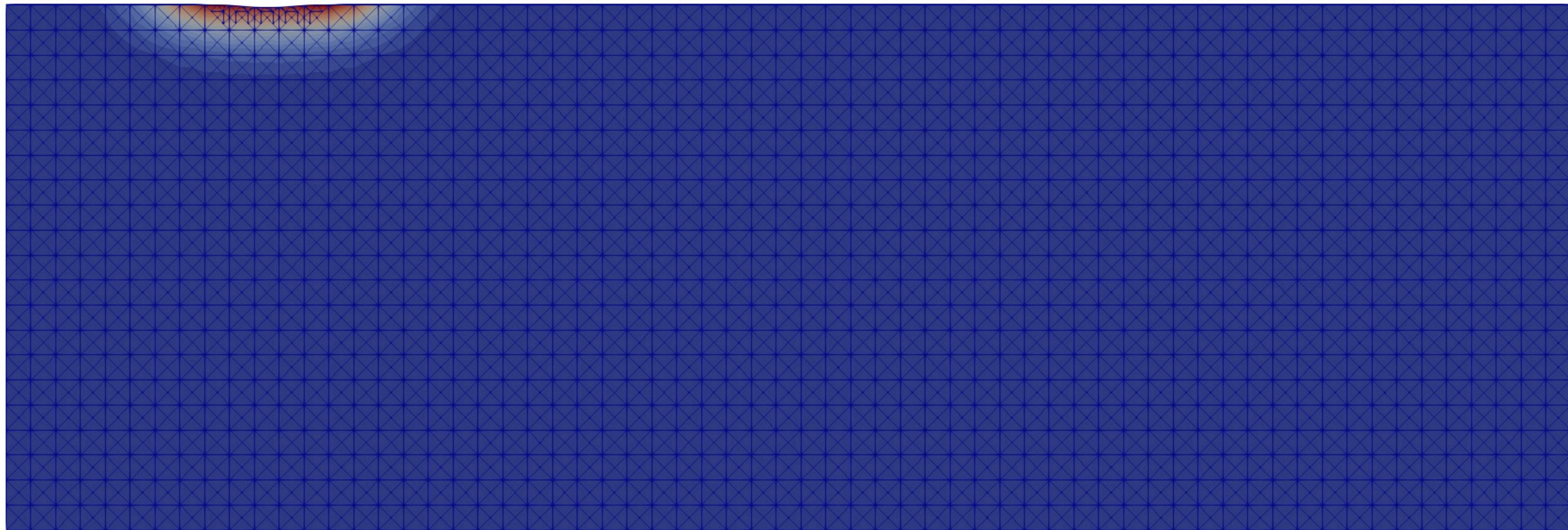




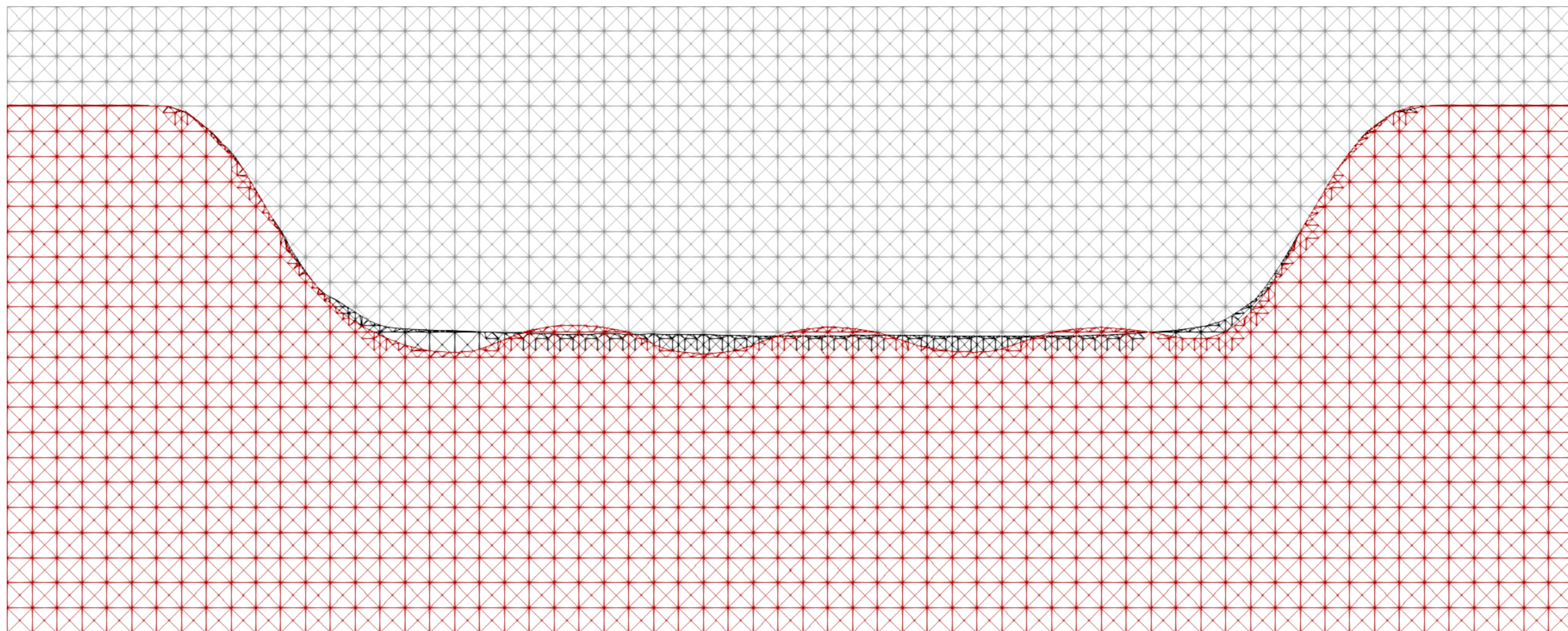
Pulsed Thermal Ablation



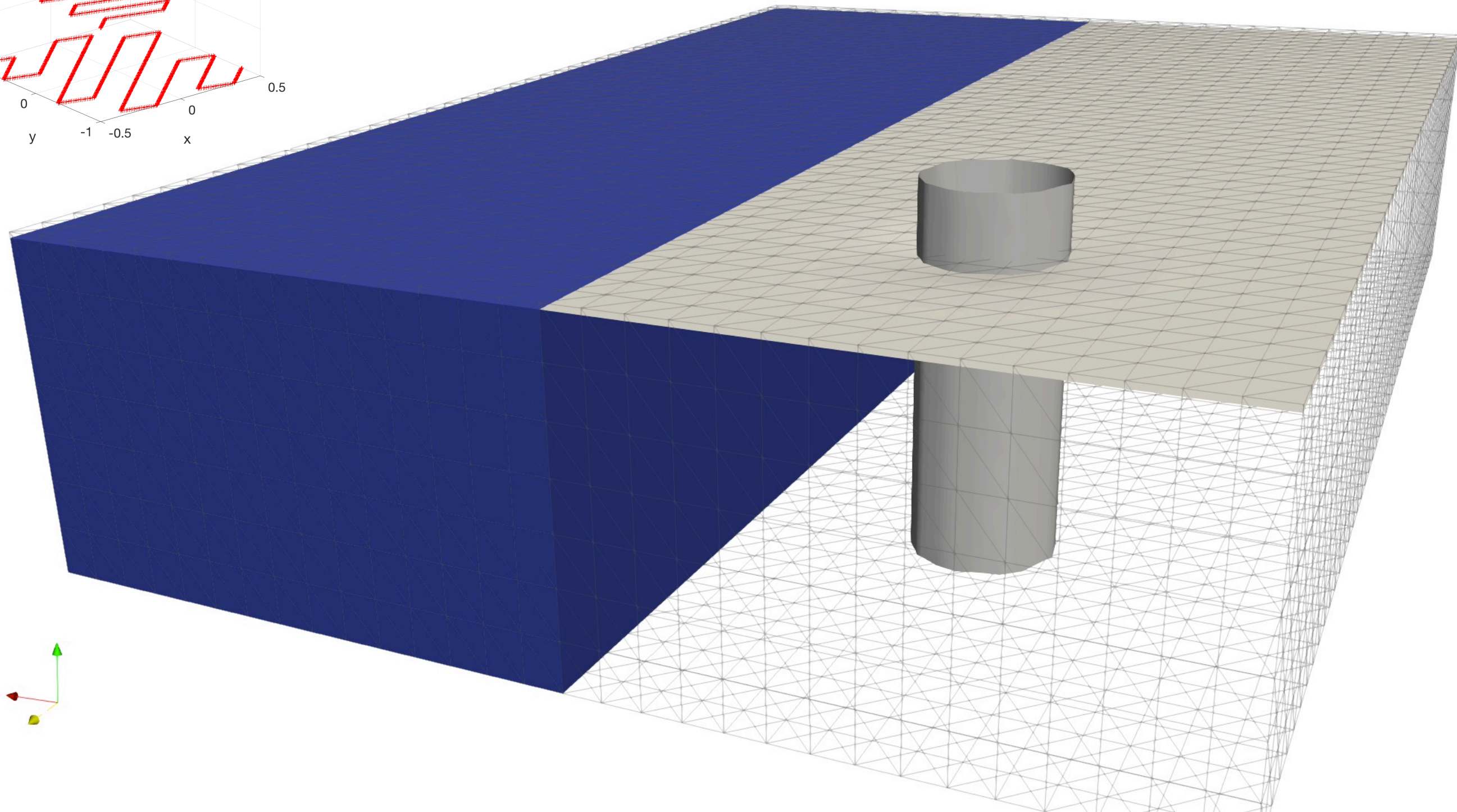
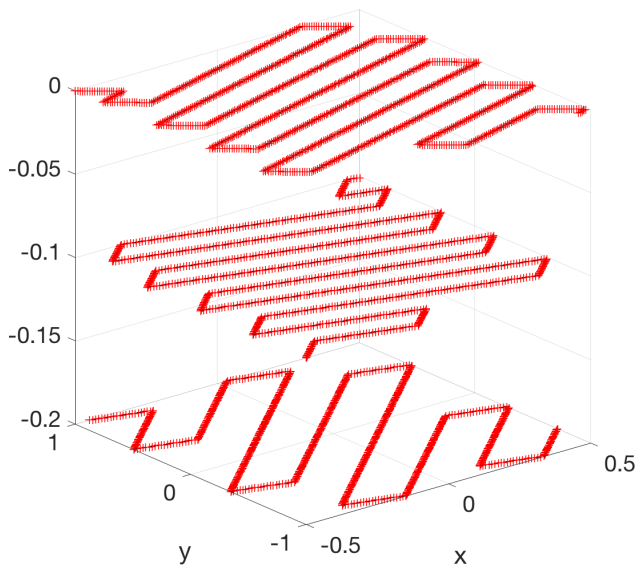
Pulsed Thermal Ablation



Pulsed Thermal Ablation



3D Machining Path





Applications in Biomechanics

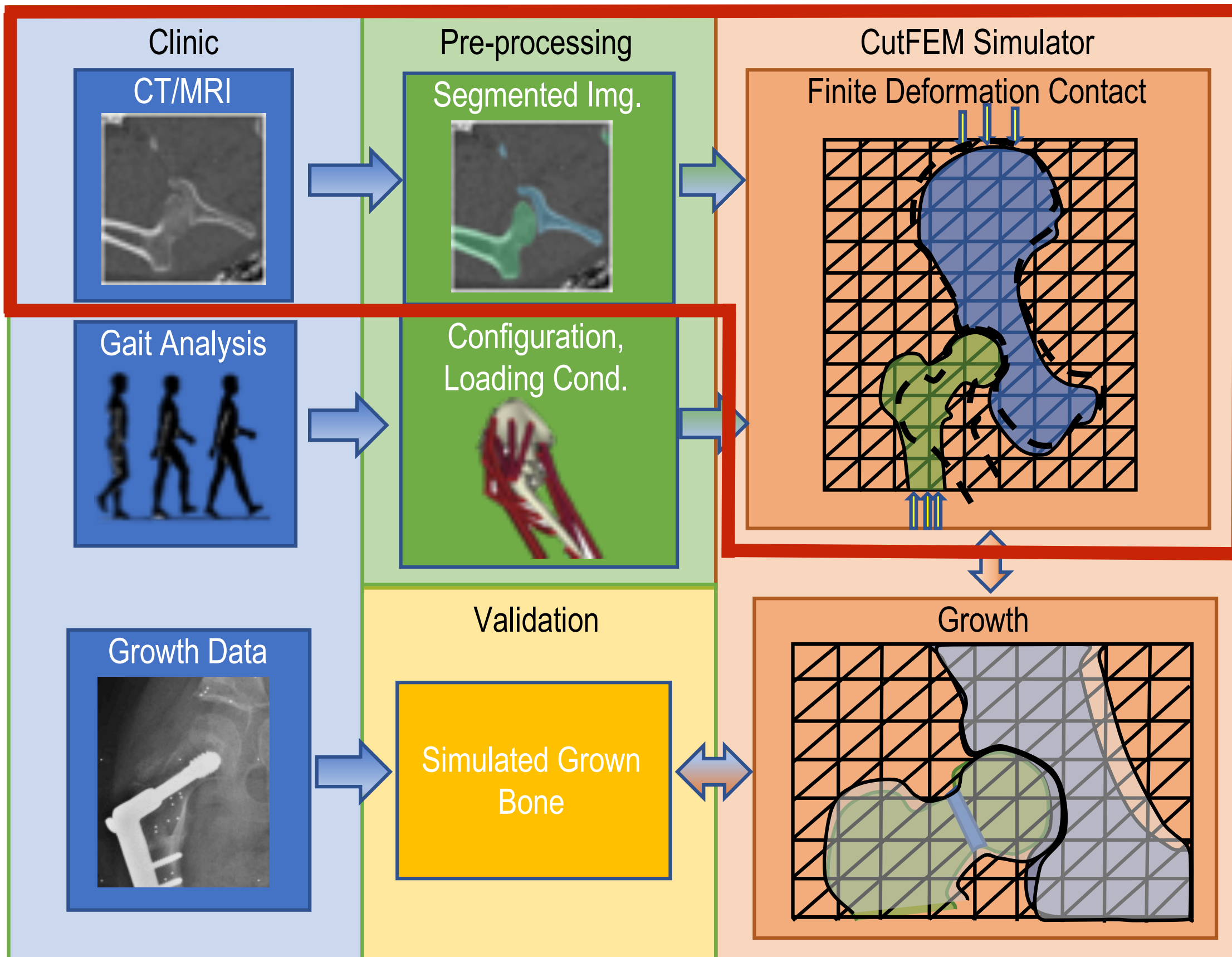
Cut Finite Element Hip Modelling Motivation



Treatment options for hip malformations: (left) untreated hip deformity of a 4-year-old child, (middle) well-formed hip 8 years post guided growth surgery¹, i.e. insertion of one screw through the growth plate of the femur bones, (right) well-formed hip after an invasive femur and hip osteotomy, i.e. cutting through the bones and insertion of screws and plates.

Study stress in hip joint using FE Modelling to enhance understanding of bone growth and bone shape changes

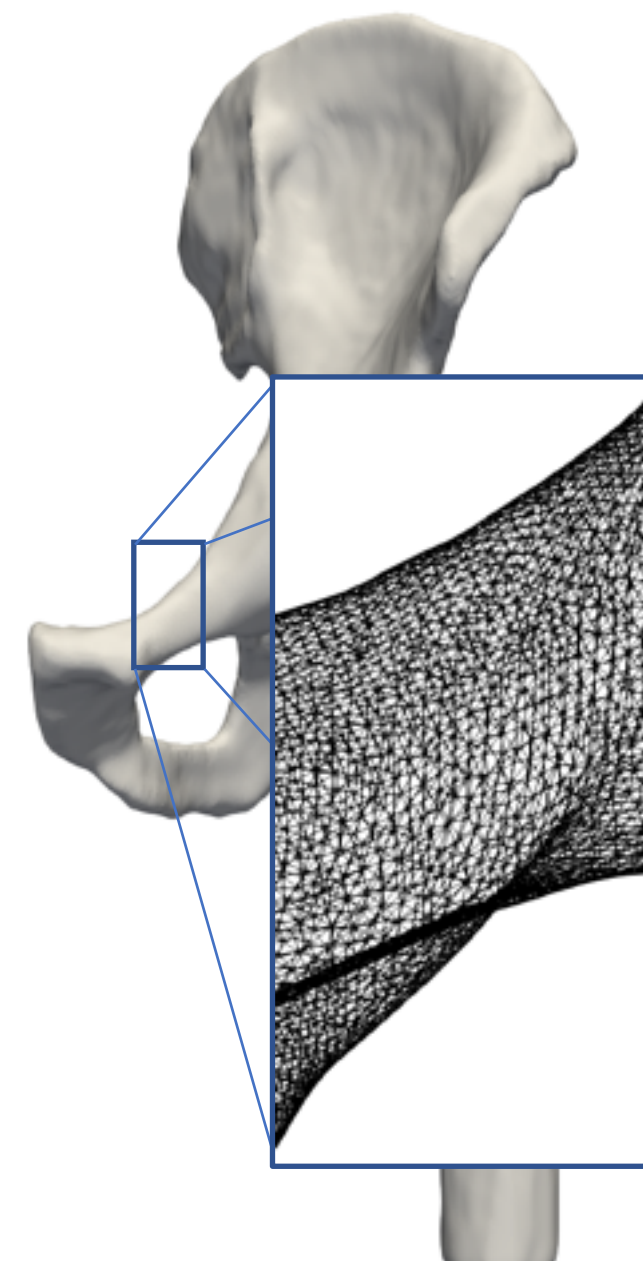
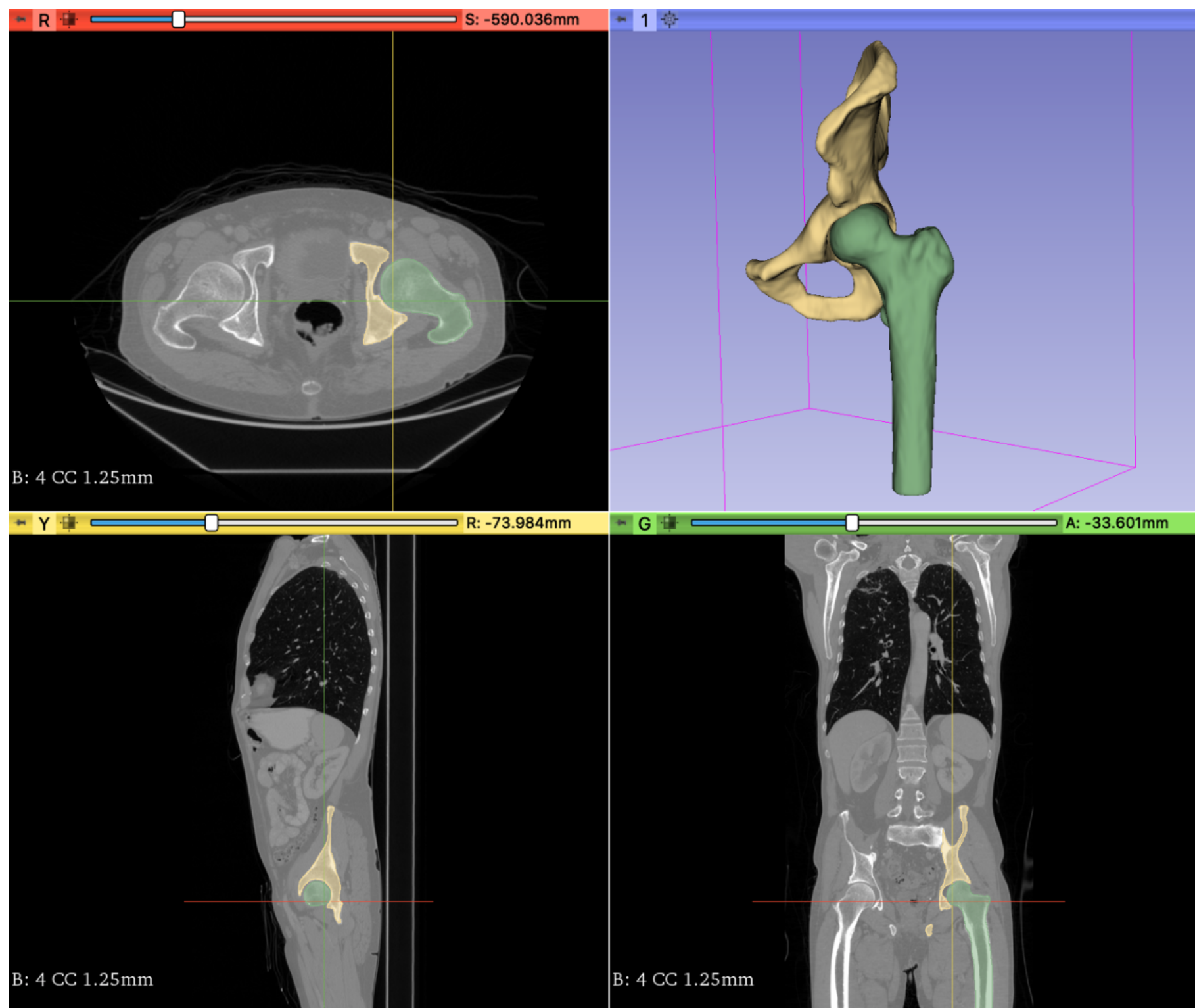
[1] Lee, W-C et al. "Guided growth of the proximal femur for hip displacement in children with cerebral palsy." *Journal of Pediatric Orthopaedics* 36.5, 511-515, (2016).



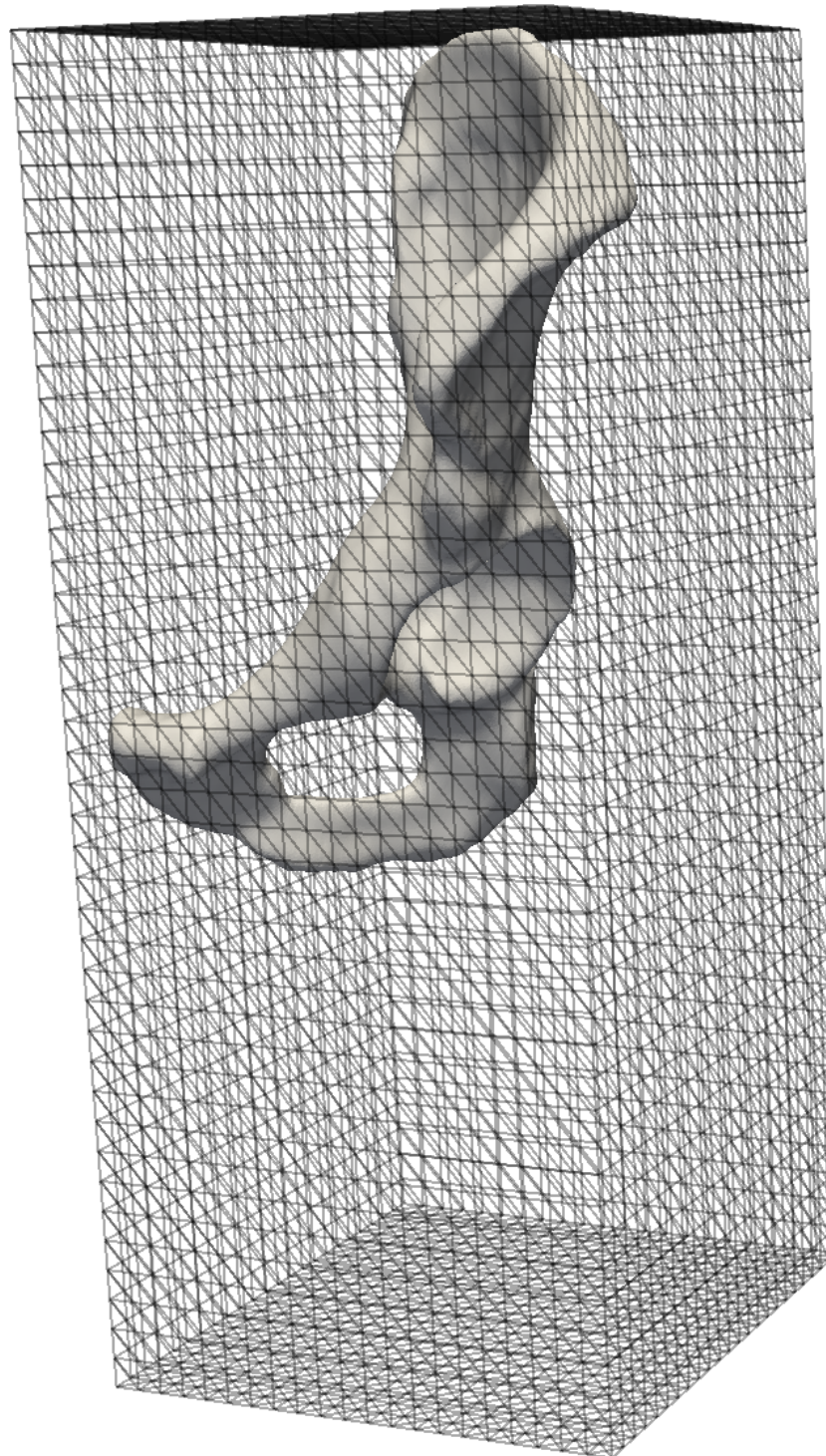
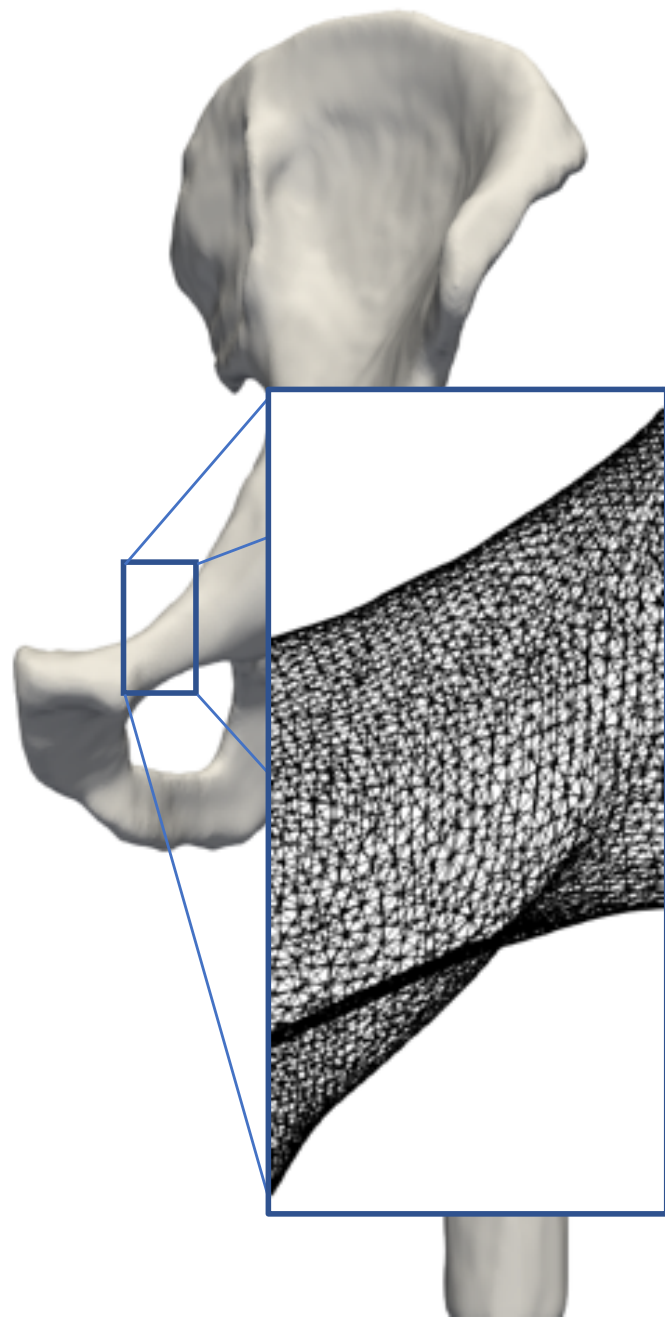
Surface Triangulation to CutFEM pipeline

Segmentation with 3D Slicer (Faezeh Moshfeghifar) of CT-image from the cancer imaging archive (TCIA)

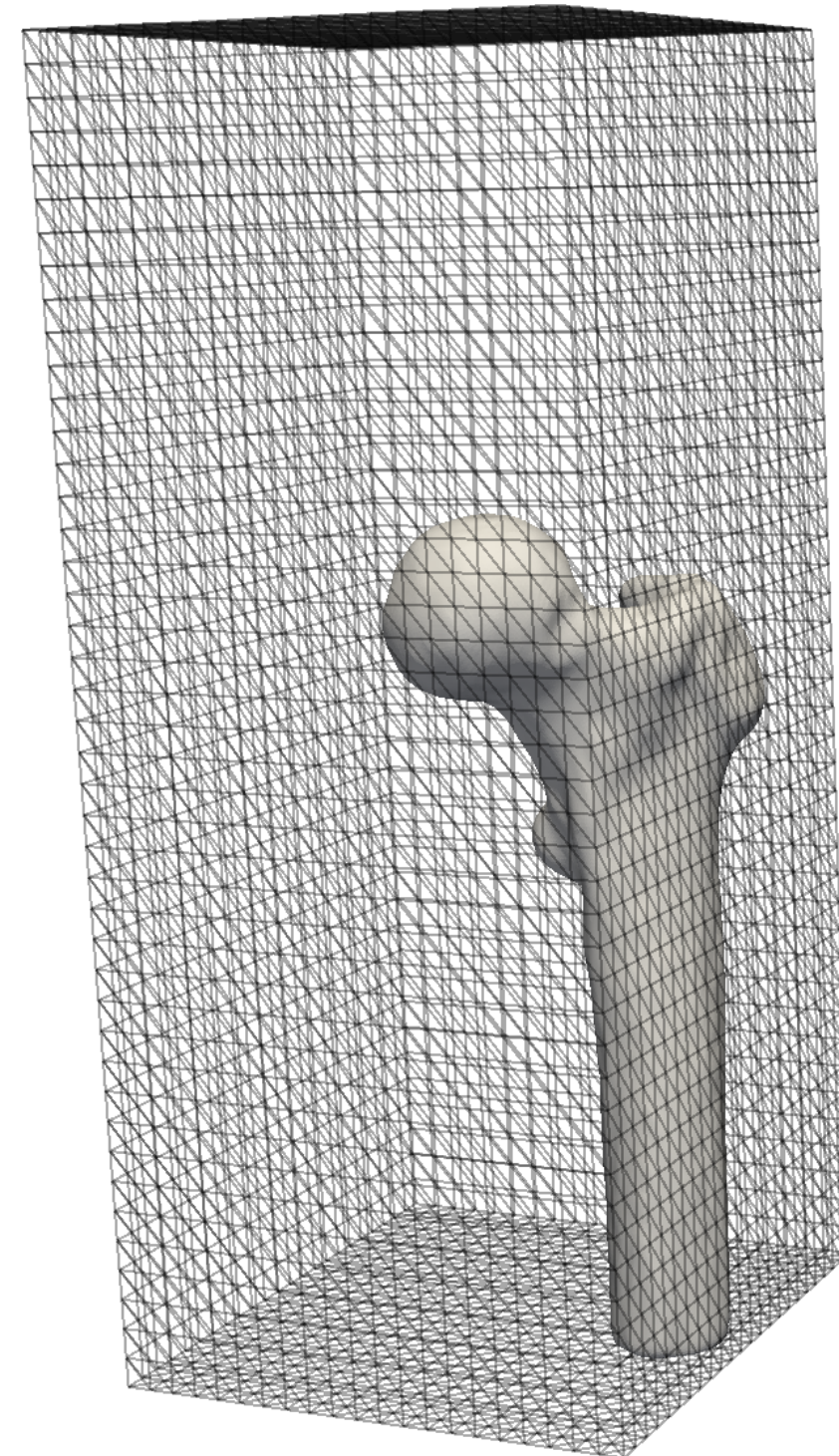
Surface triangulation



Create Regular Background mesh

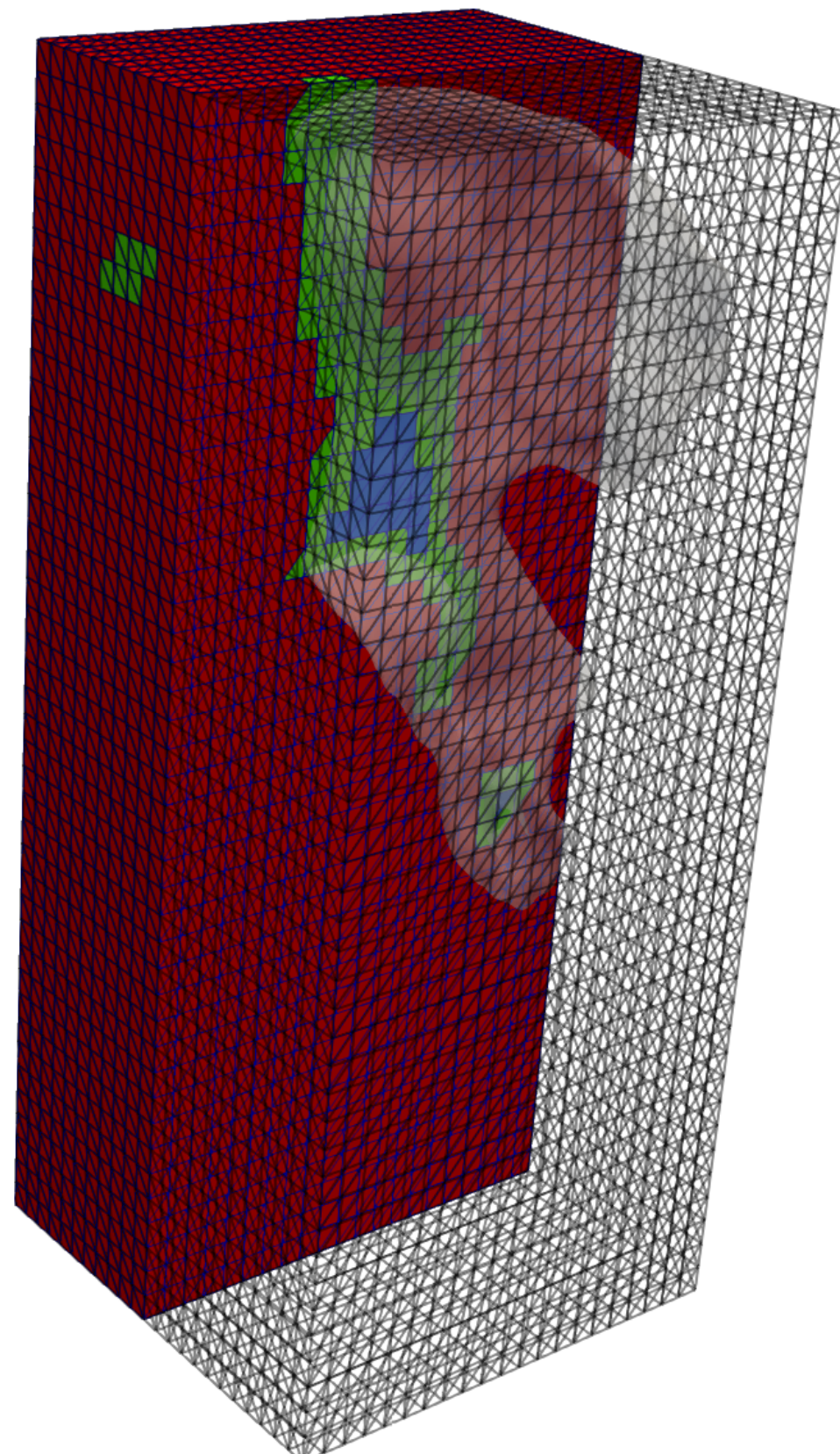
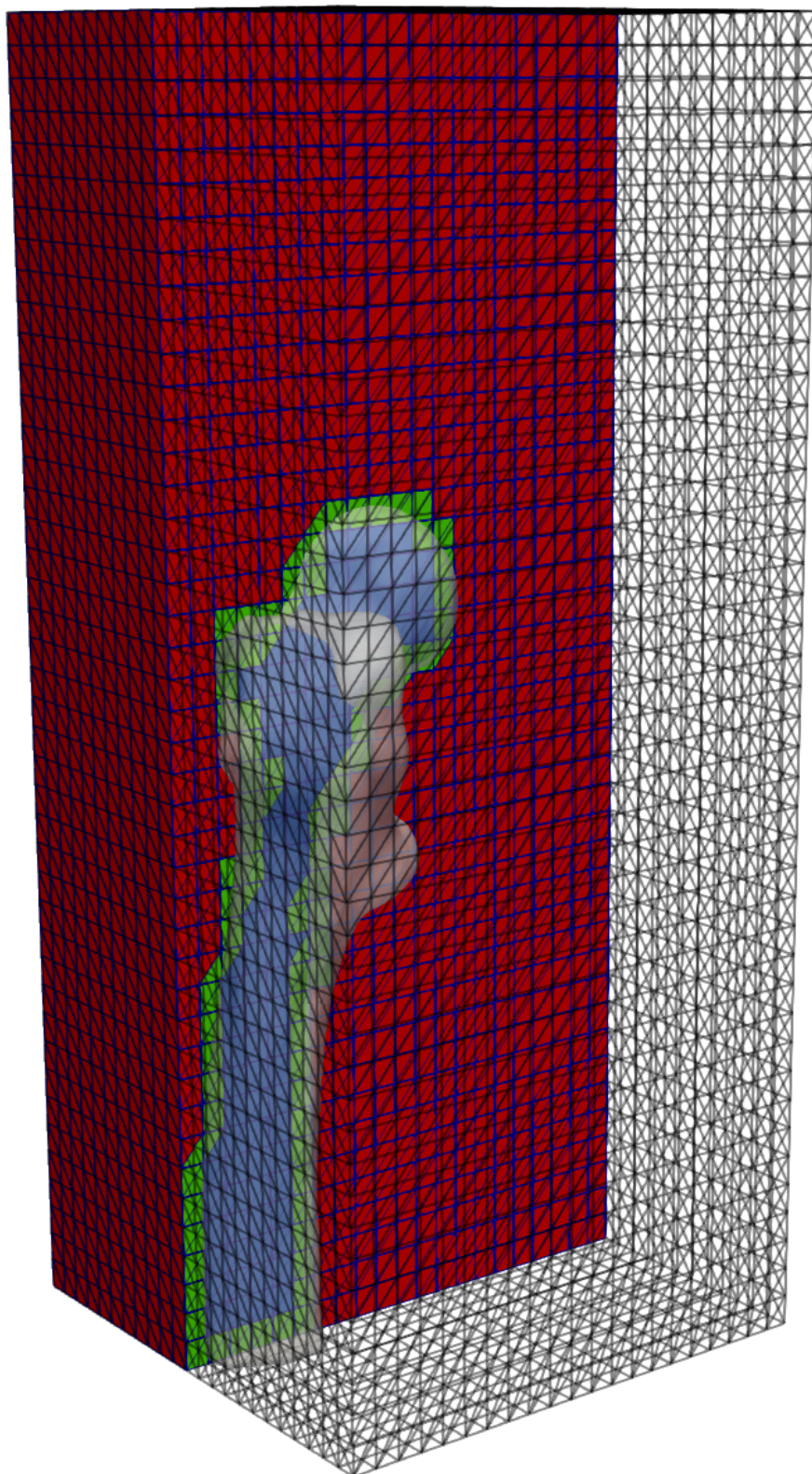


Hip bone surface
triangulation embedded in
regular background mesh



Femur bone surface
triangulation embedded in
regular background mesh

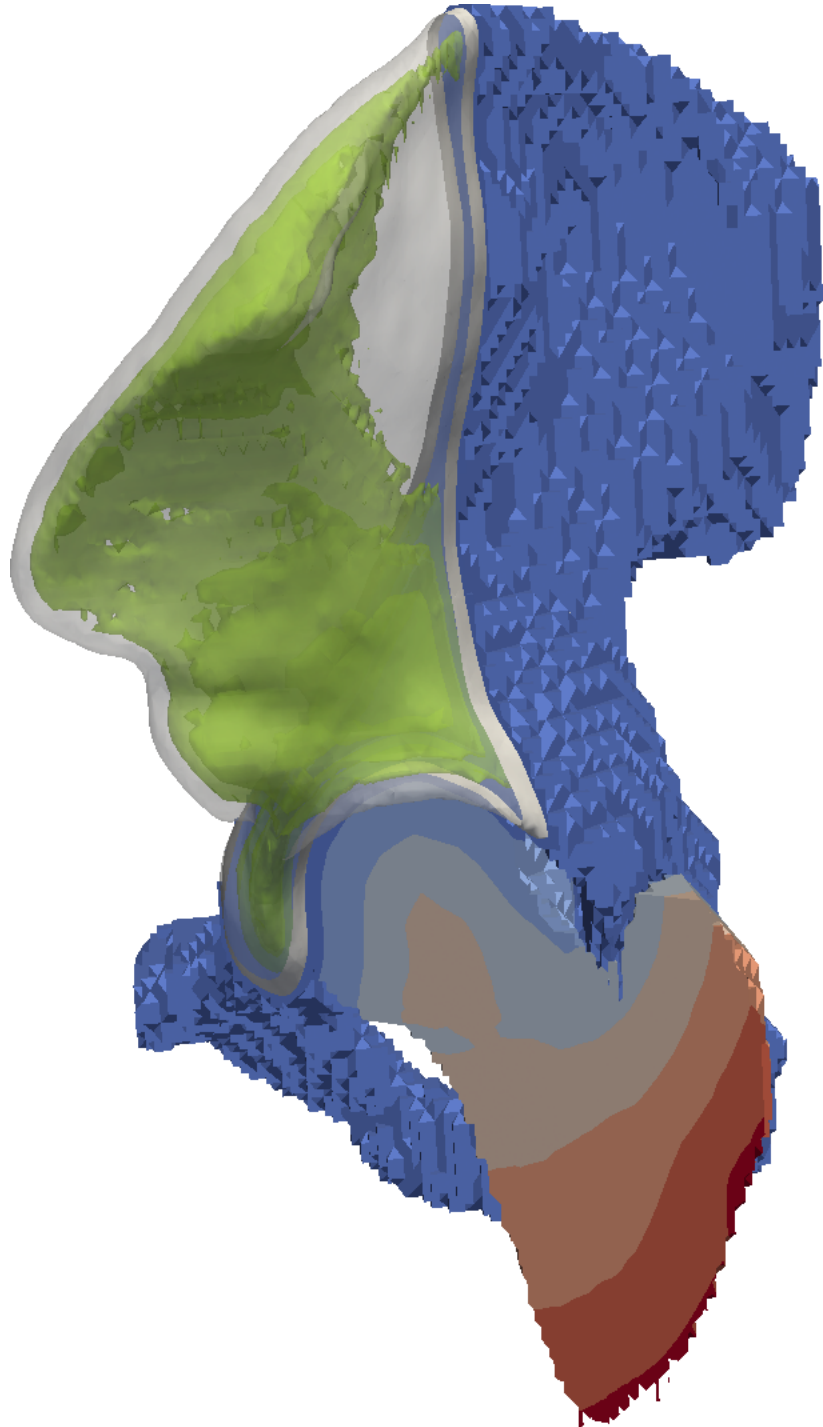
Determine inside, outside and intersected cells



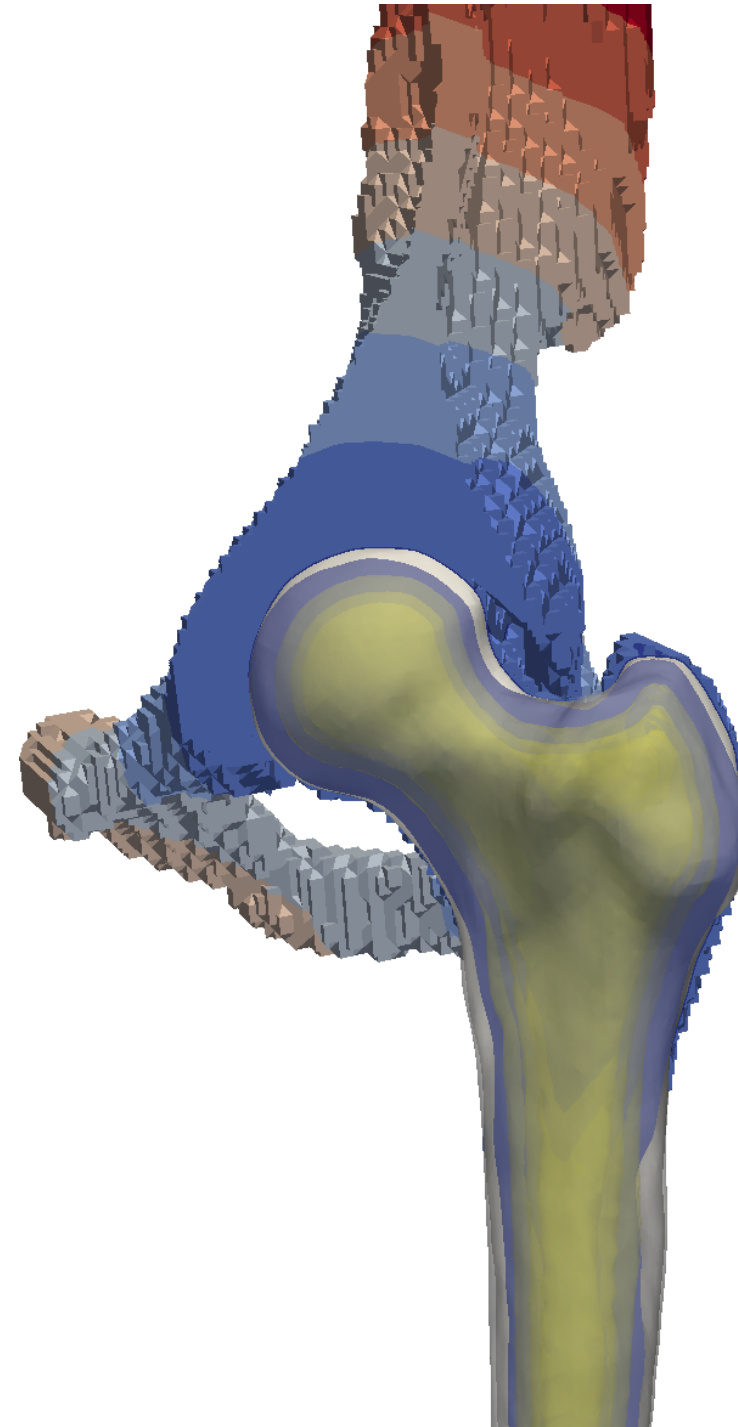
Compute signed distance function for each bone



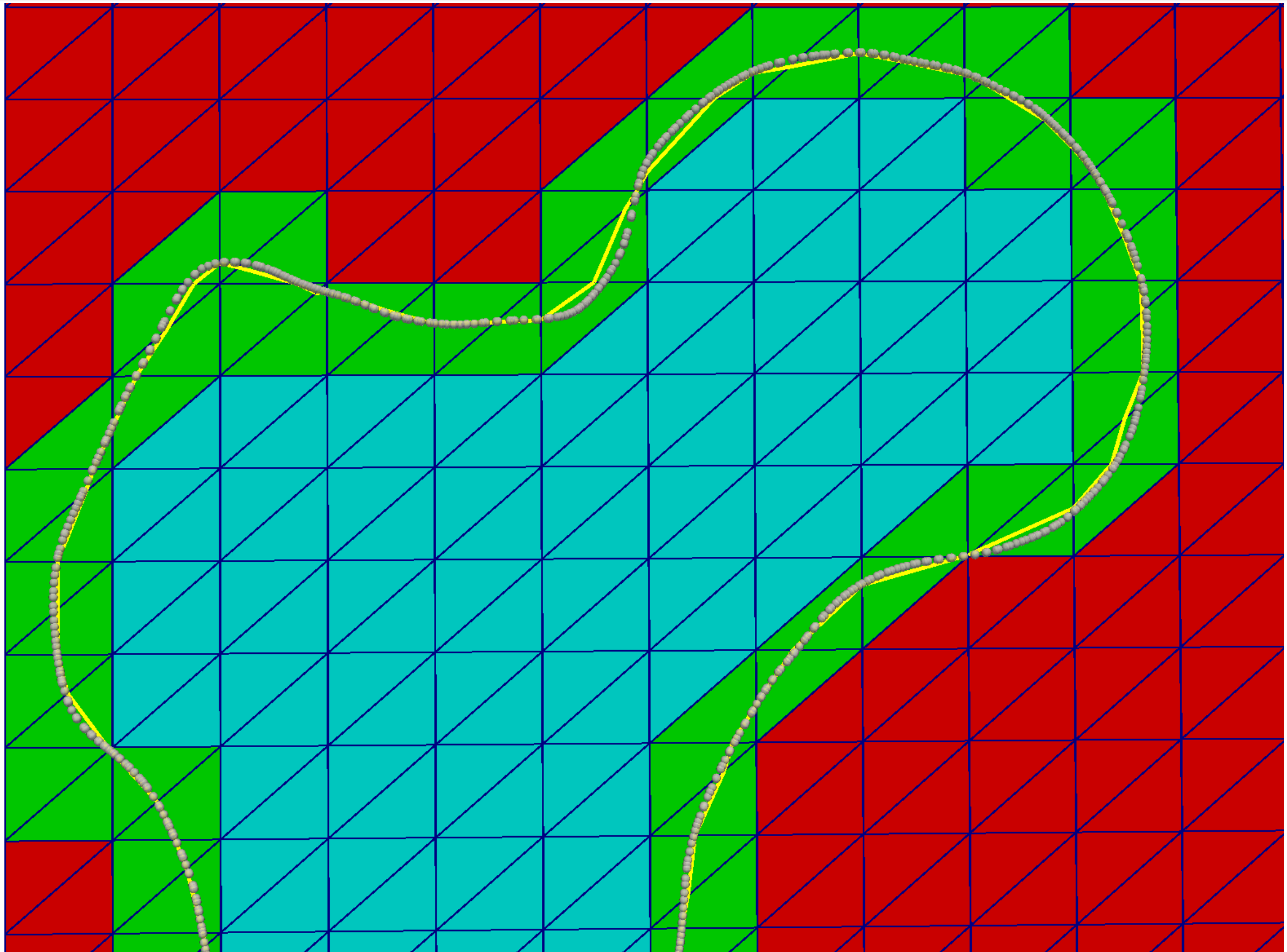
ϕ_1



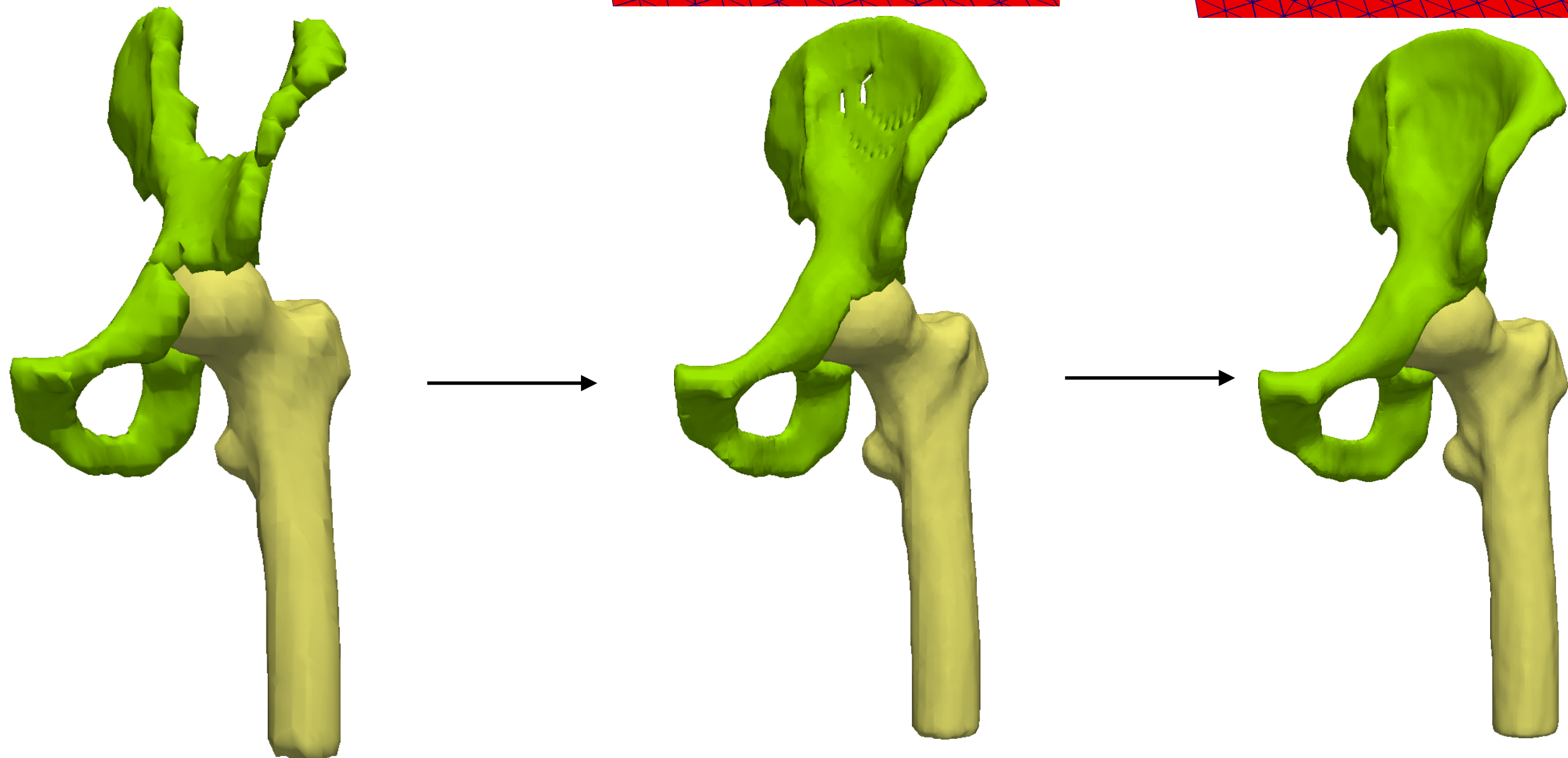
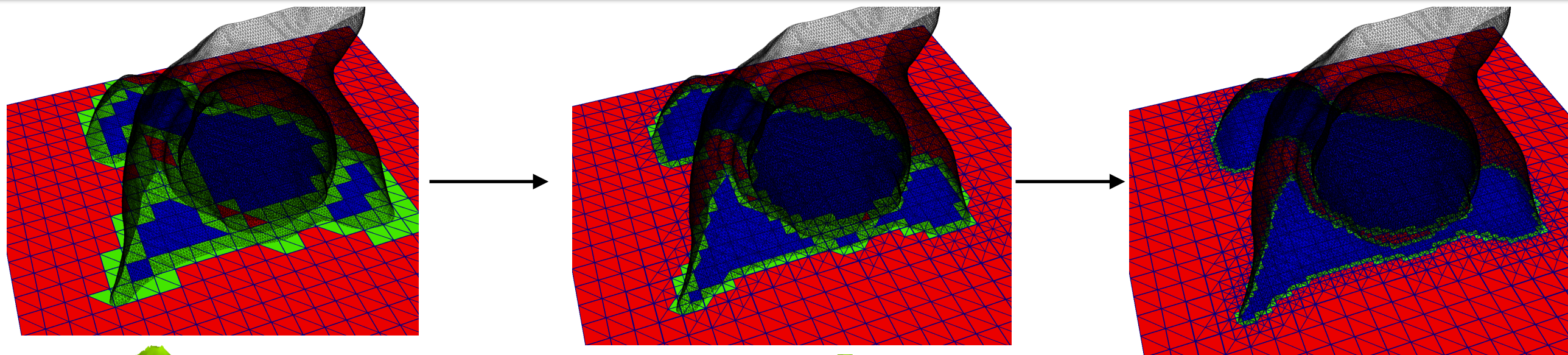
ϕ_2



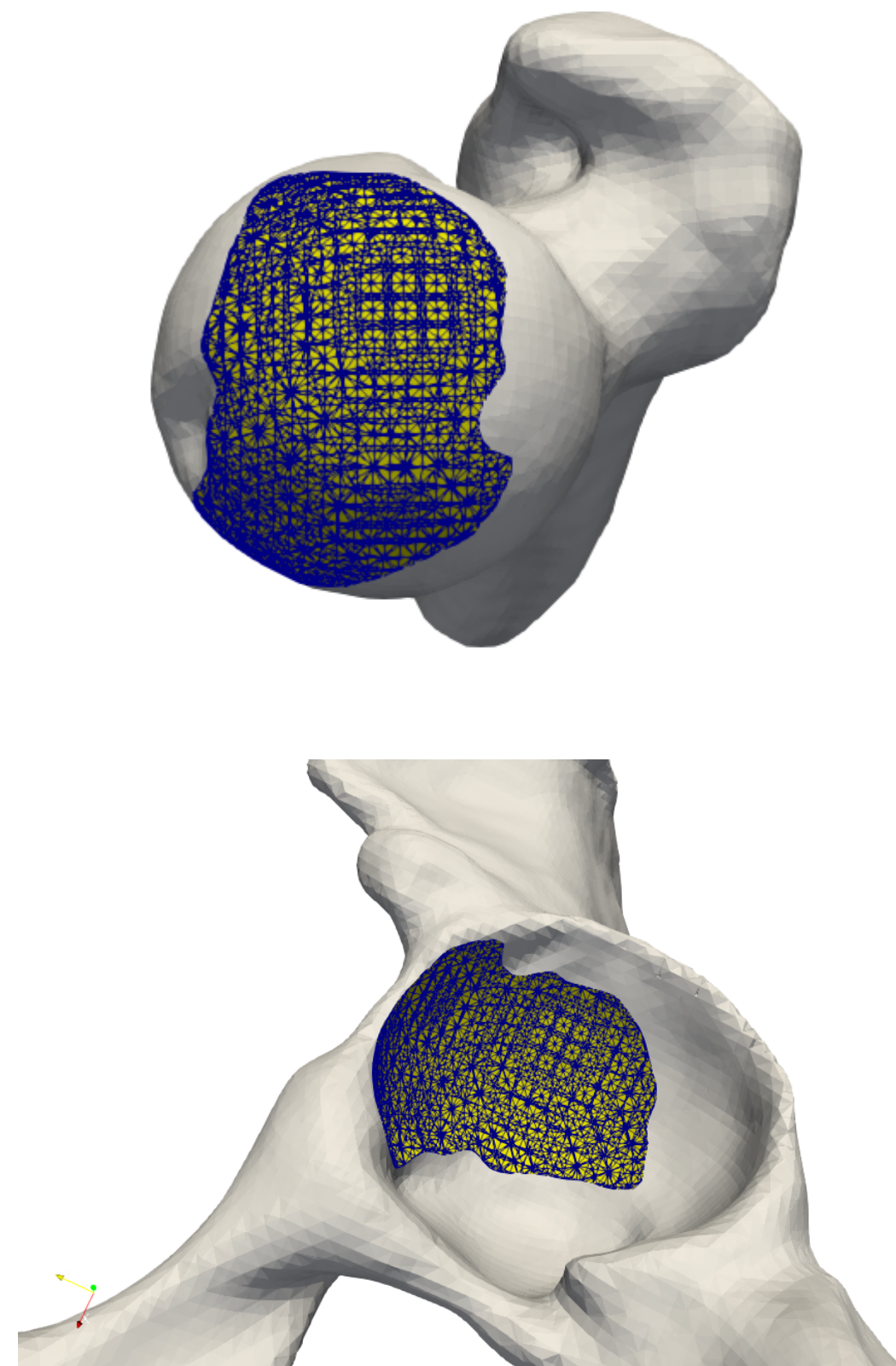
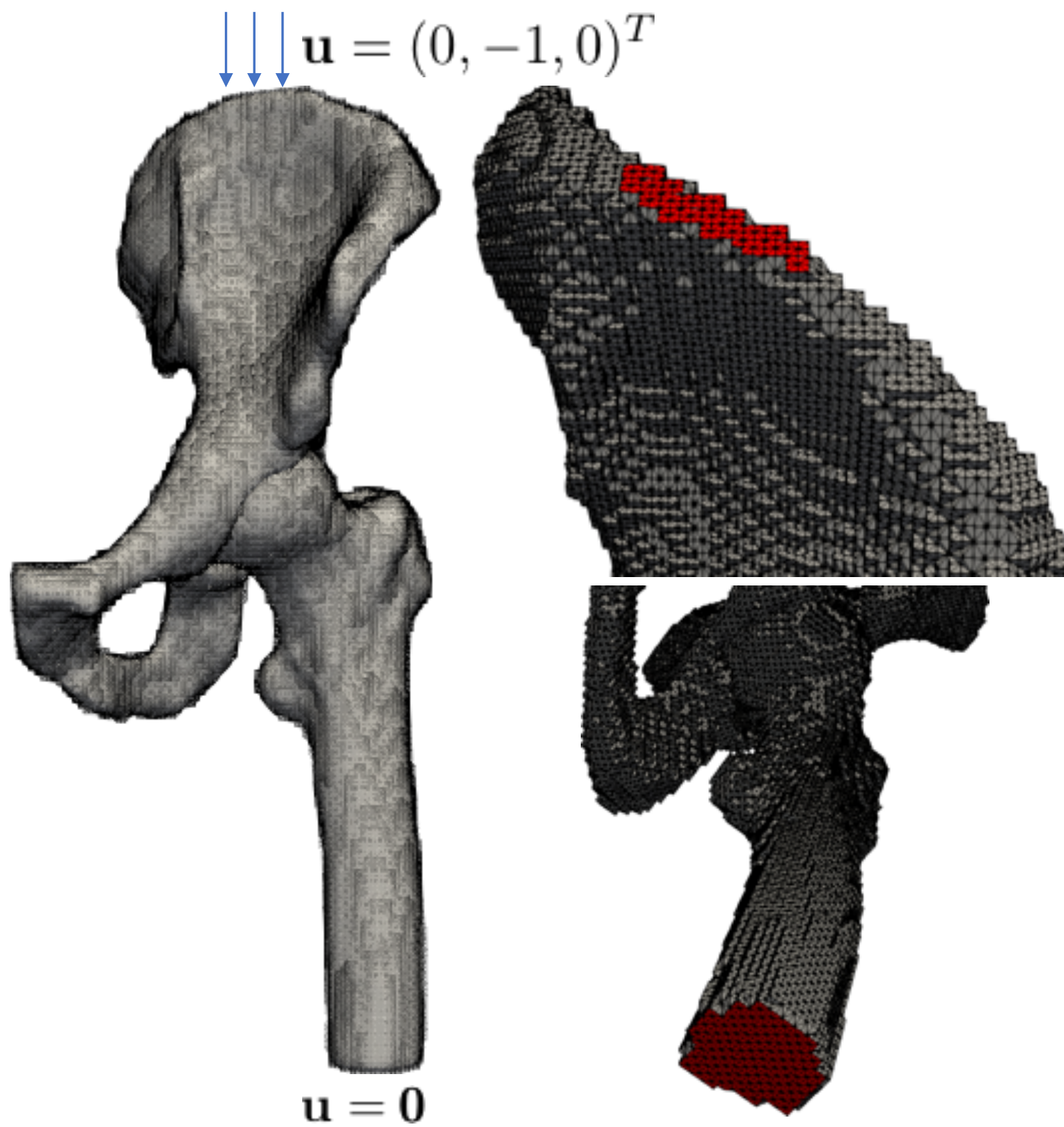
Geometrical Error (Linear Approximation)



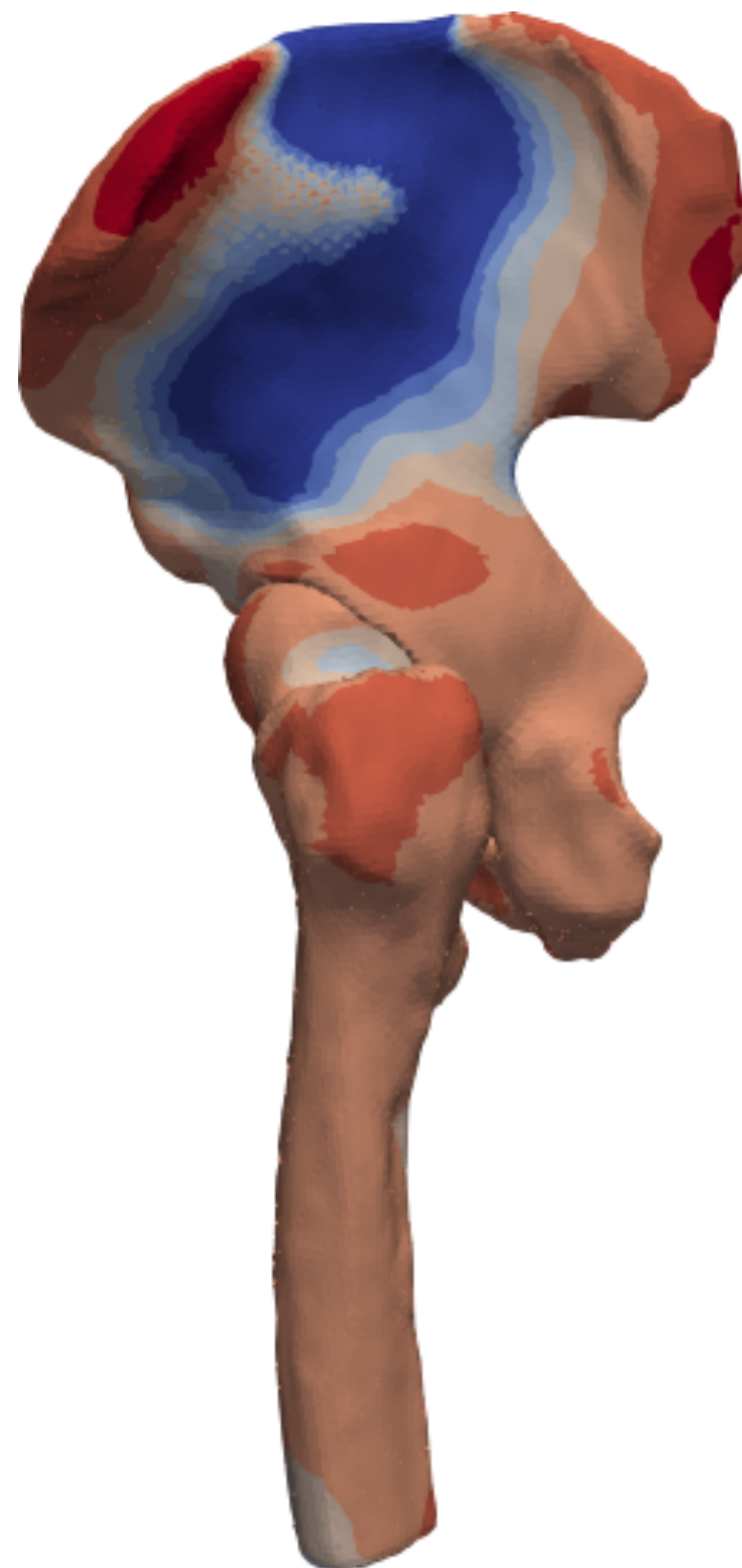
Refine elements that are intersected by surface triangulation



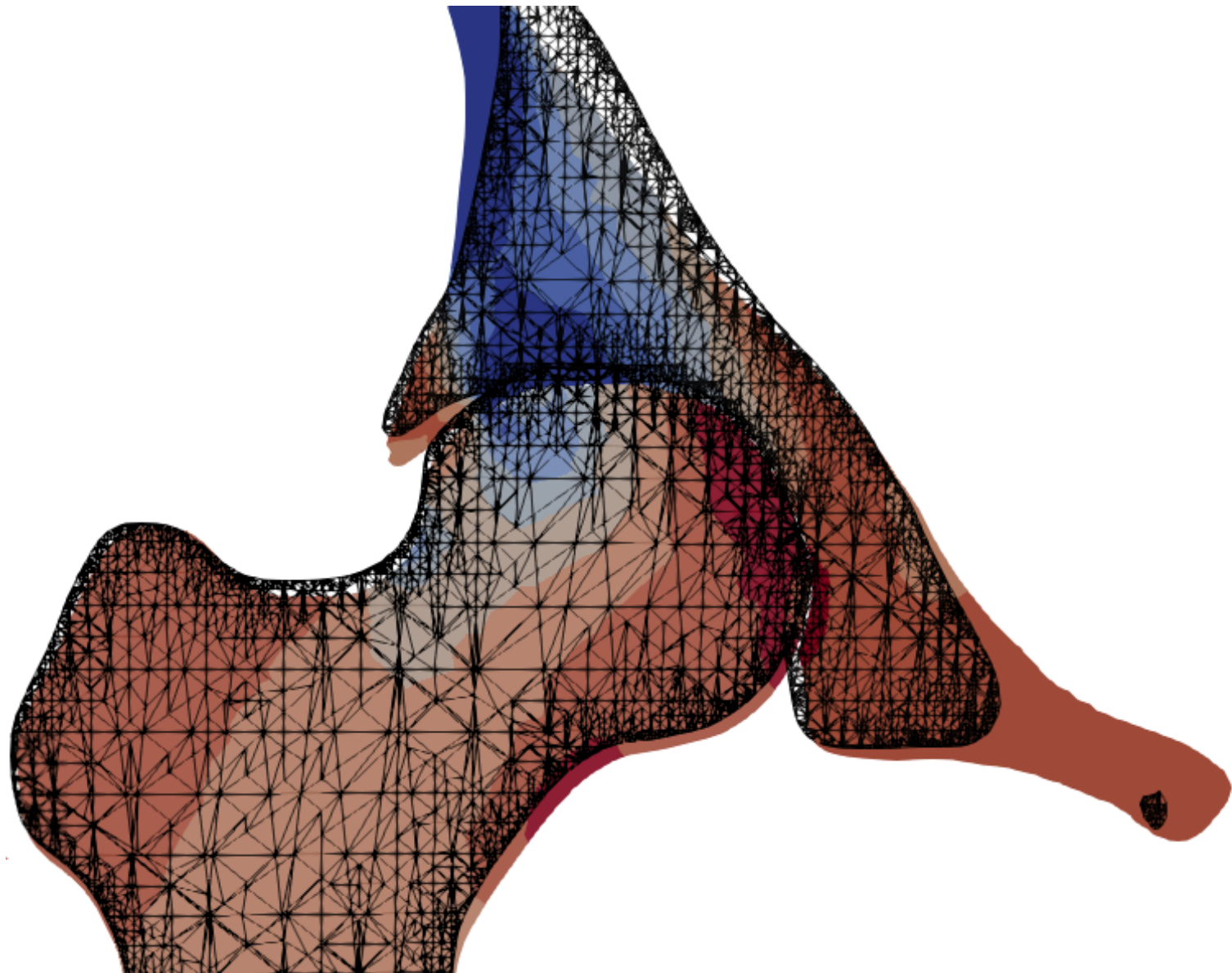
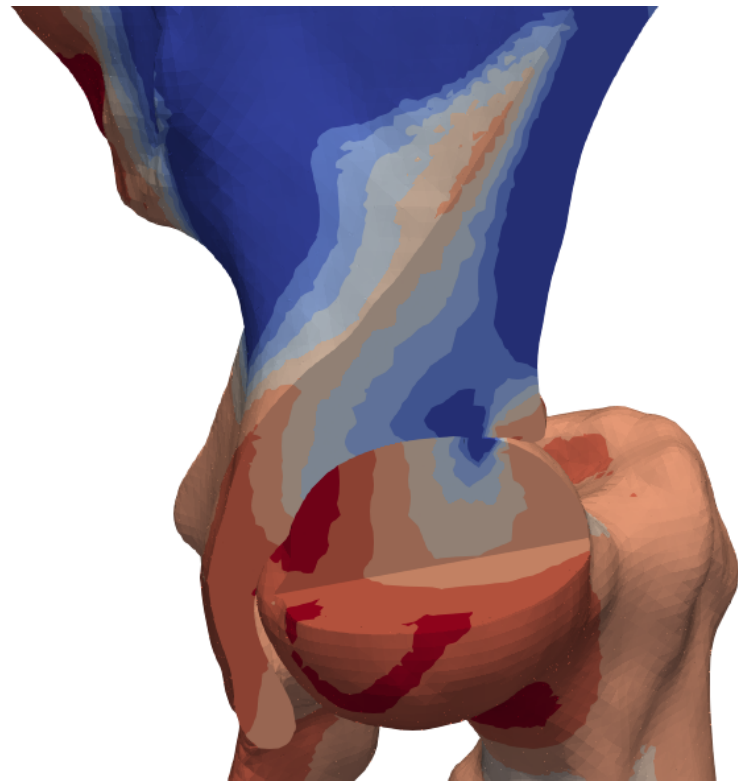
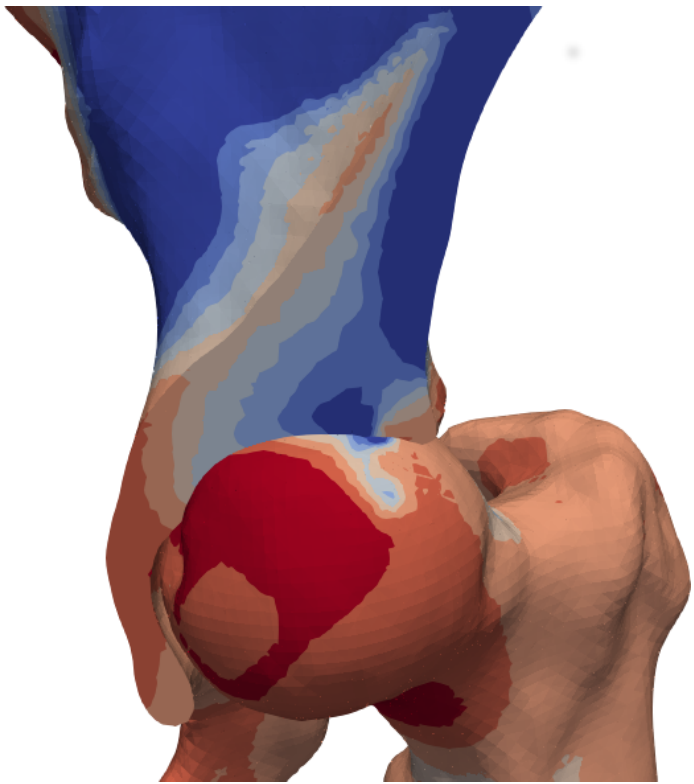
Extract elements and set boundary conditions



Stress Profile σ_{yy}



Stress Profile σ_{yy}





Patient-Specific Data

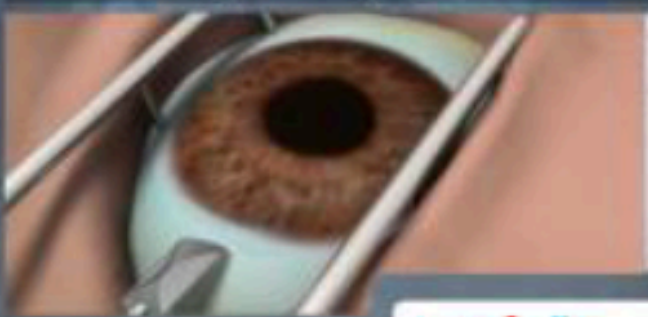


Expert Knowledge

<https://rainbow.ku.dk>



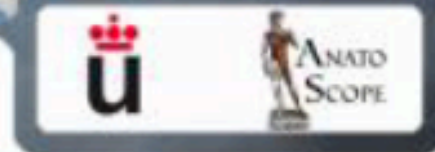
Eye surgery



Neurology



Spine Braces



Intraoperative radiotherapy



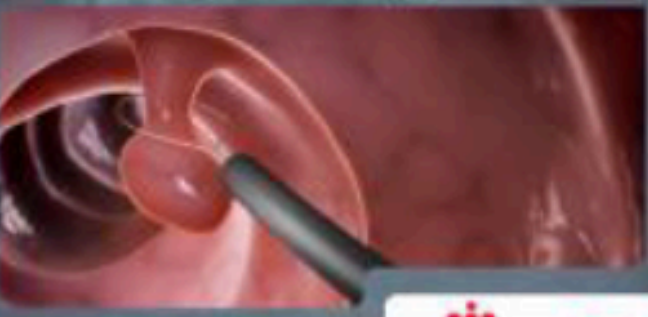
Surgical guidance



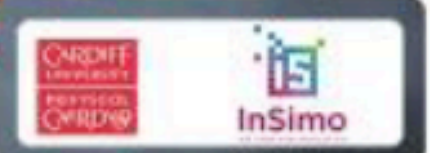
Soft organ diagnosis



Surgical navigation



Surgical planning



Apps

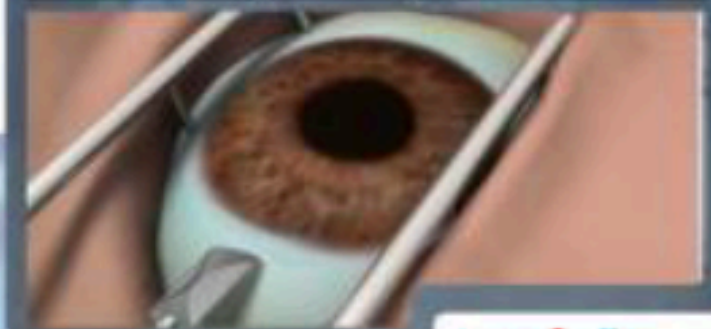
Cardiovascular Devices



Scoliosis



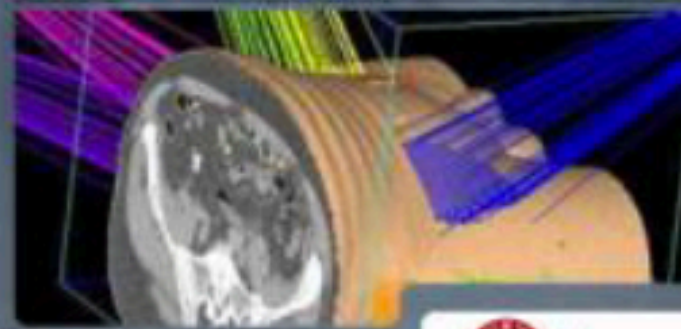
Eye surgery



Hip growth



Prostate Cancer



Intraoperative radiotherapy



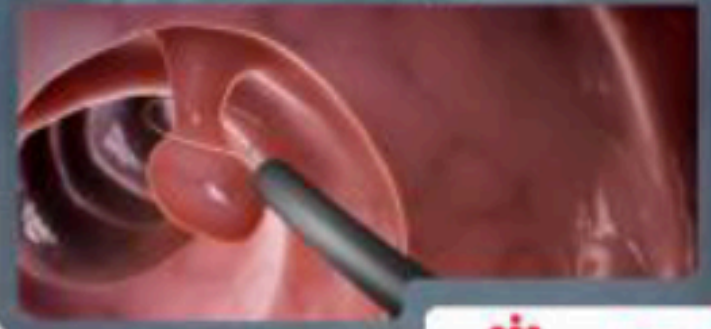
Dental prostheses



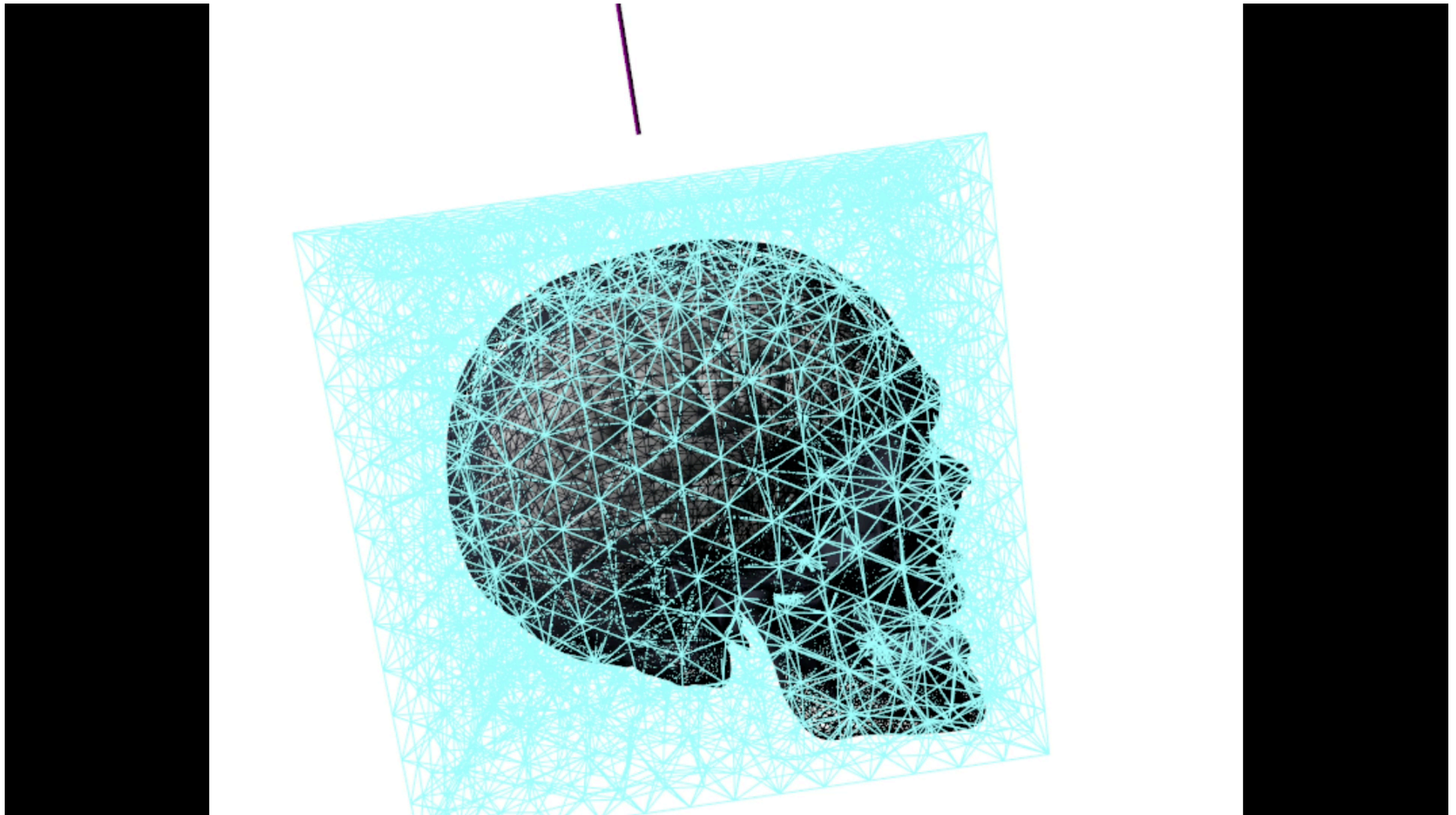
Breast Cancer



Surgical navigation



Brain shift and electrode implantation



Controlling the Error on Target Motion through Real-time Mesh Adaptation: Applications to Deep Brain Stimulation, HP Bui et al, Int J Numer Meth Bio, 2017.

Error estimation and adaptivity

Controlling the Error on Target Motion through Real-time Mesh Adaptation: Applications to Deep Brain Stimulation, HP Bui et al, Int J Numer Meth Bio, 2017.

Stéphane Pierre Alain BORDAS, Department of Computational Engineering & Sciences University of Luxembourg - stephane.bordas@uni.lu

Superconvergence recovery

Controlling the Error on Target Motion through Real-time Mesh Adaptation: Application to Deep Brain Stimulation

H. P. Bui, S. Tomar, H. Courtecuisse, M. Audette, S. Cotin and S. P. A. Bordas

Controlling the Error on Target Motion through Real-time Mesh Adaptation: Applications to Deep Brain Stimulation, HP Bui et al, Int J Numer Meth Bio, 2017.

Stéphane Pierre Alain BORDAS, Department of Computational Engineering & Sciences University of Luxembourg - stephane.bordas@uni.lu

Goal-oriented error estimate

$$Q(\mathbf{u}) = \frac{1}{|\omega|} \int_{\omega} \nabla \mathbf{u} d\omega$$

$$Q(\mathbf{u}) - Q(\mathbf{u}_h)$$

$$|Q(\mathbf{u}) - Q(\mathbf{u}_h)| \leq \epsilon$$

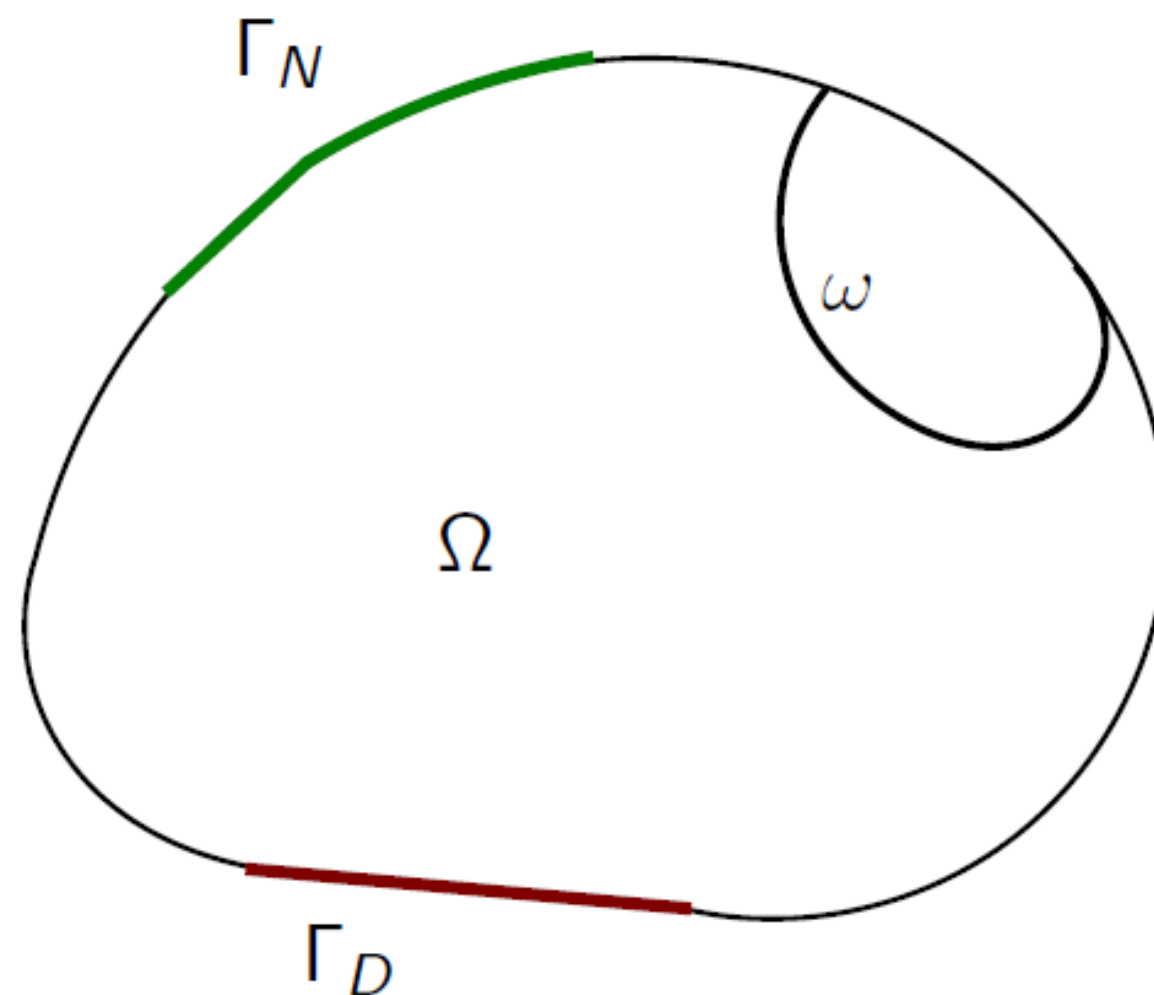


Figure 1: If we are interested in some quantity of interest defined on a subdomain ω , what is the optimal mesh?

Goal-oriented error estimate

Primal problem: $a(\mathbf{u}, \mathbf{v}) = l(\mathbf{v}) \quad \forall \mathbf{v} \in \mathbf{V}$

Solve by FEM: $a(\mathbf{u}_h, \mathbf{v}_h) = l(\mathbf{v}_h) \quad \forall \mathbf{v}_h \in \mathbf{V}_h$

$\longrightarrow \mathbf{u}_h$

Weak residual: $r(\mathbf{v}) = l(\mathbf{v}) - a(\mathbf{u}_h, \mathbf{v}) \quad \forall \mathbf{v} \in \mathbf{V}$

If we define a dual problem:

Find $\mathbf{z} \in \mathbf{V}$ such that $a(\mathbf{v}, \mathbf{z}) = Q(\mathbf{v}) \quad \forall \mathbf{v} \in \mathbf{V}$

We observe:

$$Q(\mathbf{u}) - Q(\mathbf{u}_h) = a(\mathbf{u}, \mathbf{z}) - a(\mathbf{u}_h, \mathbf{z}) = l(\mathbf{z}) - a(\mathbf{u}_h, \mathbf{z}) = r(\mathbf{z})$$

Corotational Cut Finite Element Method for real-time surgical simulation: application to needle insertion simulation, HP Bui et al, arXiv:1712.03052[cs.CE]

Hyperelasticity

Equilibrium equations in initial configuration:

$$\begin{aligned}
 -\operatorname{div} \boldsymbol{\Pi} &= \mathbf{B} & \text{in } \Omega^0 \\
 \mathbf{u} &= \mathbf{0} & \text{on } \Gamma_D^0 \\
 \boldsymbol{\Pi} \cdot \mathbf{N} &= \mathbf{T} & \text{on } \Gamma_N^0
 \end{aligned}$$

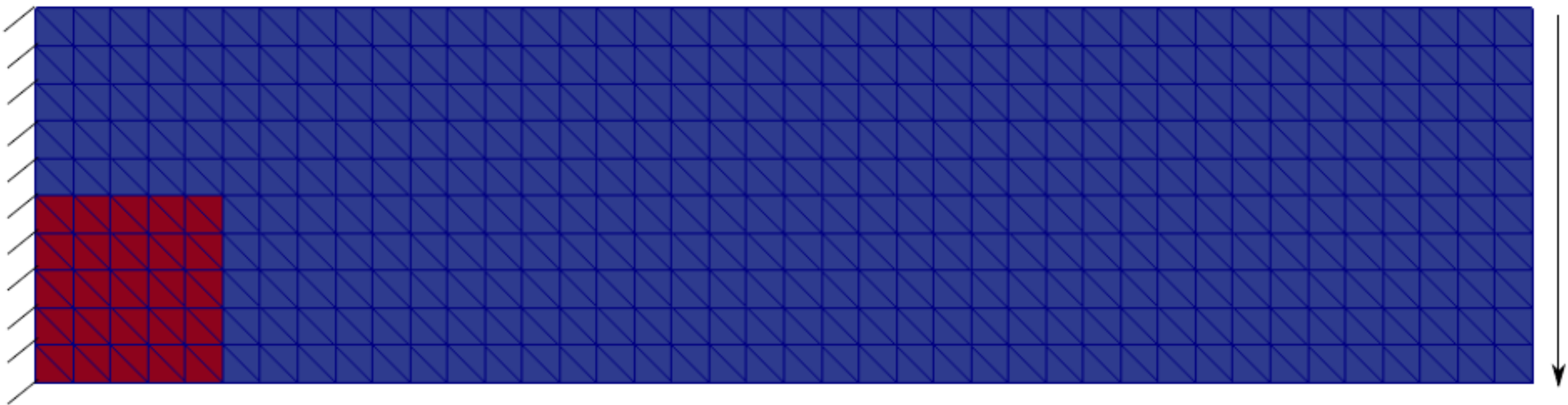
- $\boldsymbol{\Pi} = \boldsymbol{\Pi}(\mathbf{u})$ is the first Piola-Kirchhoff stress tensor
- \mathbf{B} is a given body force per unit volume
- \mathbf{u} is the displacement
- \mathbf{T} is a given boundary traction

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Cantilever beam

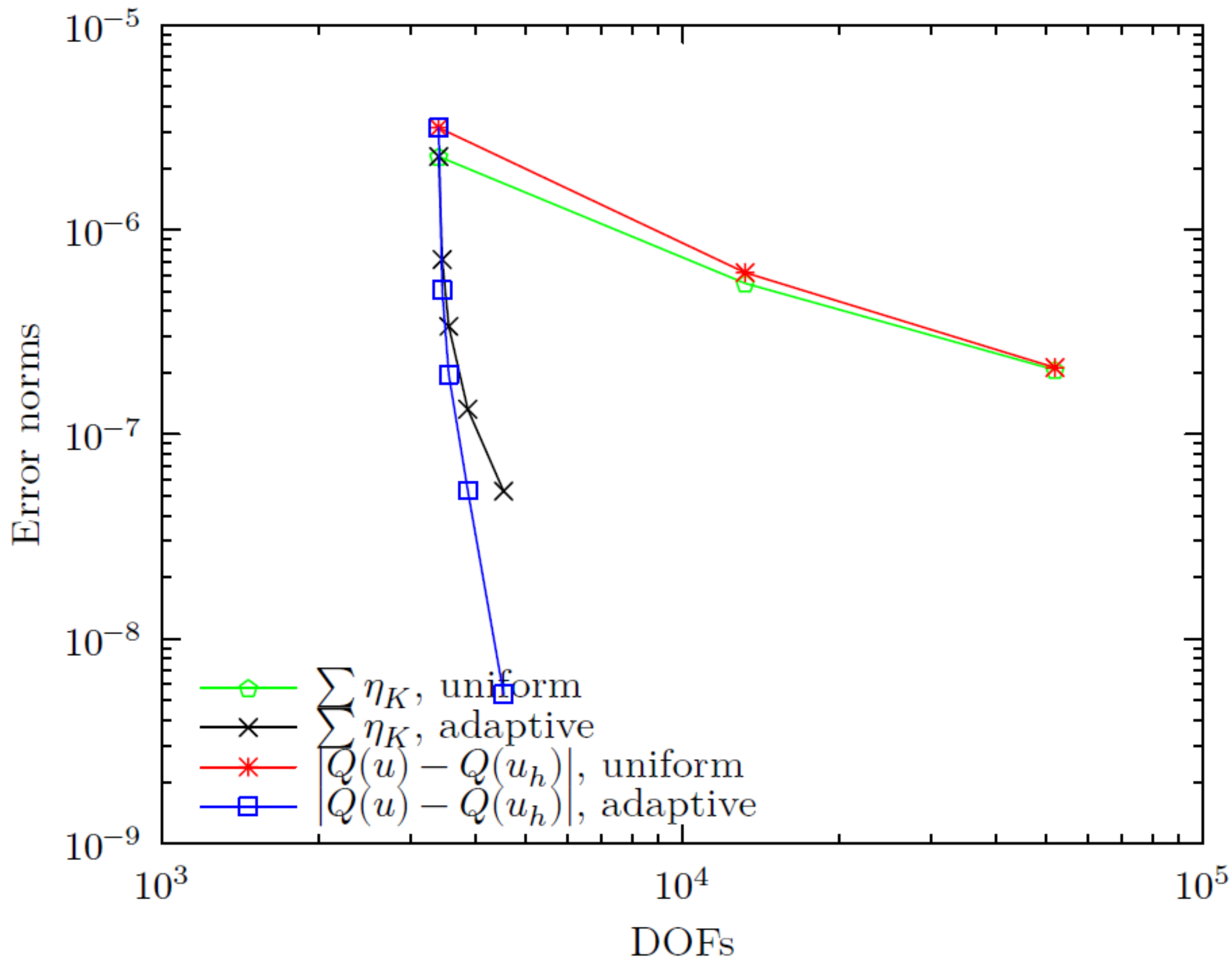
Parameters

Saint Venant-Kirchhoff material: $E = 1000$, $\nu = 0.4$

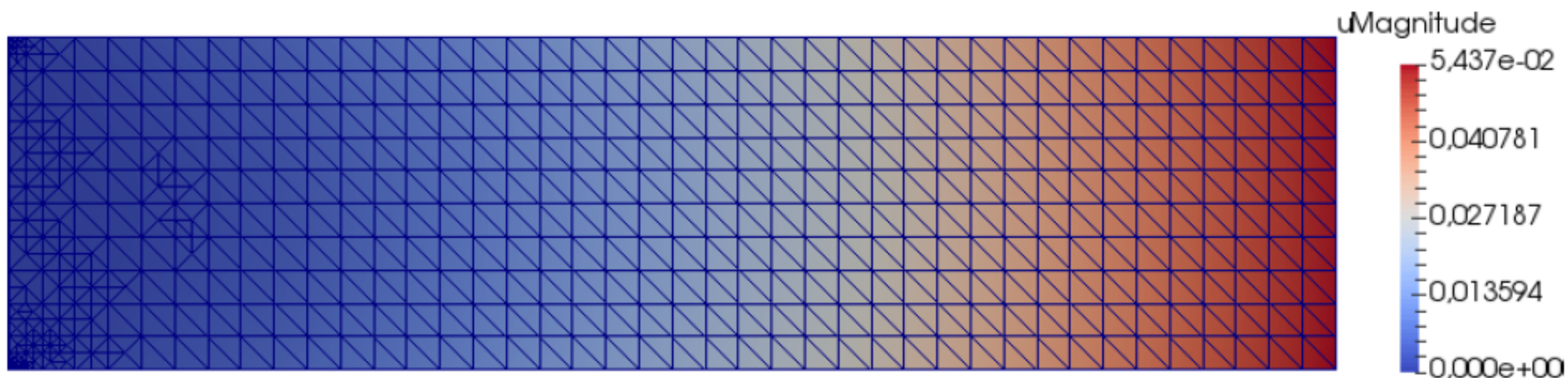


Corotational Cut Finite Element Method for real-time surgical simulation: application to needle insertion simulation, HP Bui et al, arXiv:1712.03052[cs.CE]

Cantilever beam (2)

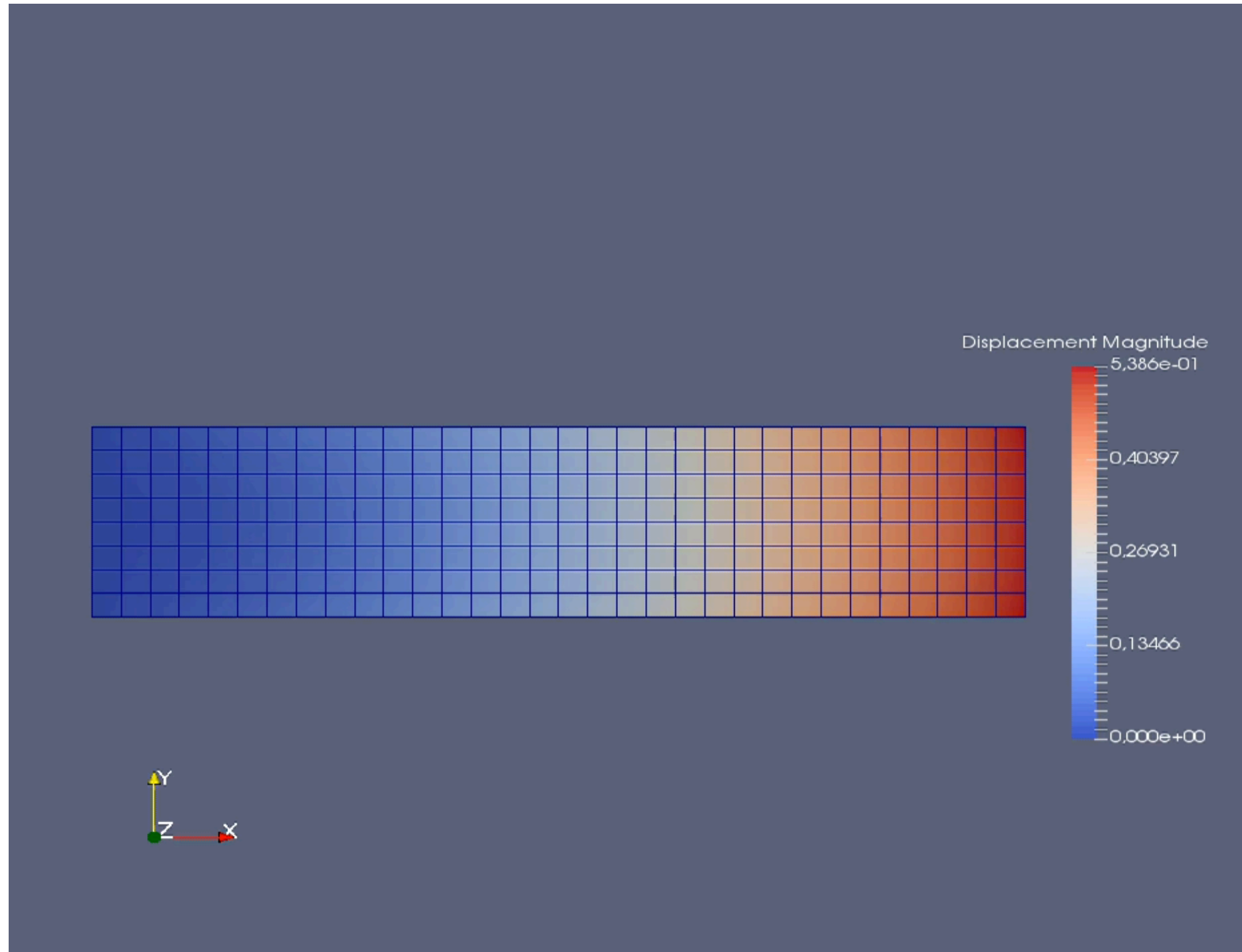


Cantilever beam (3)



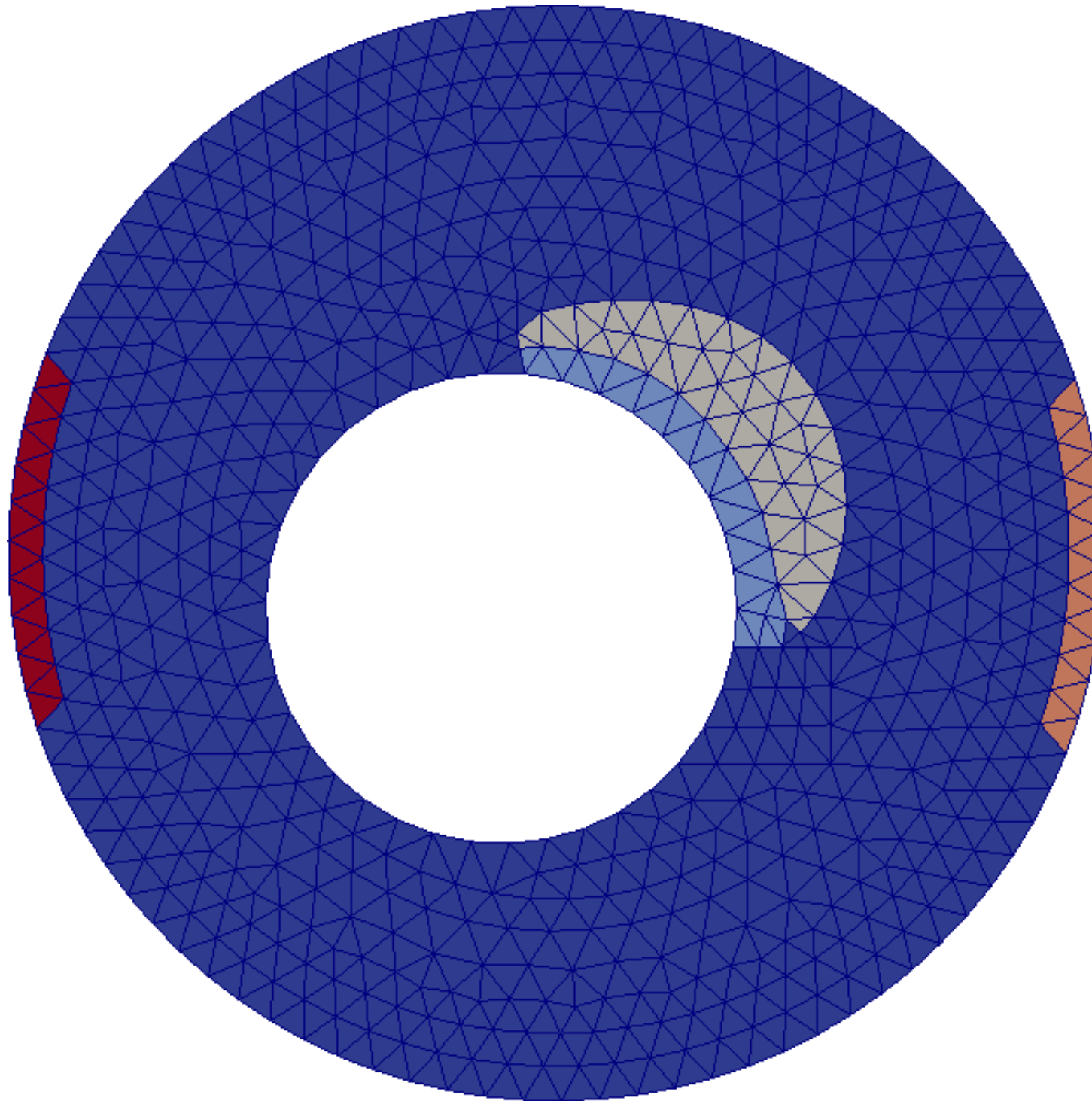
Corotational Cut Finite Element Method for real-time surgical simulation: application to needle insertion simulation, HP Bui et al, arXiv:1712.03052[cs.CE]

Adaptivity using quadrilaterals



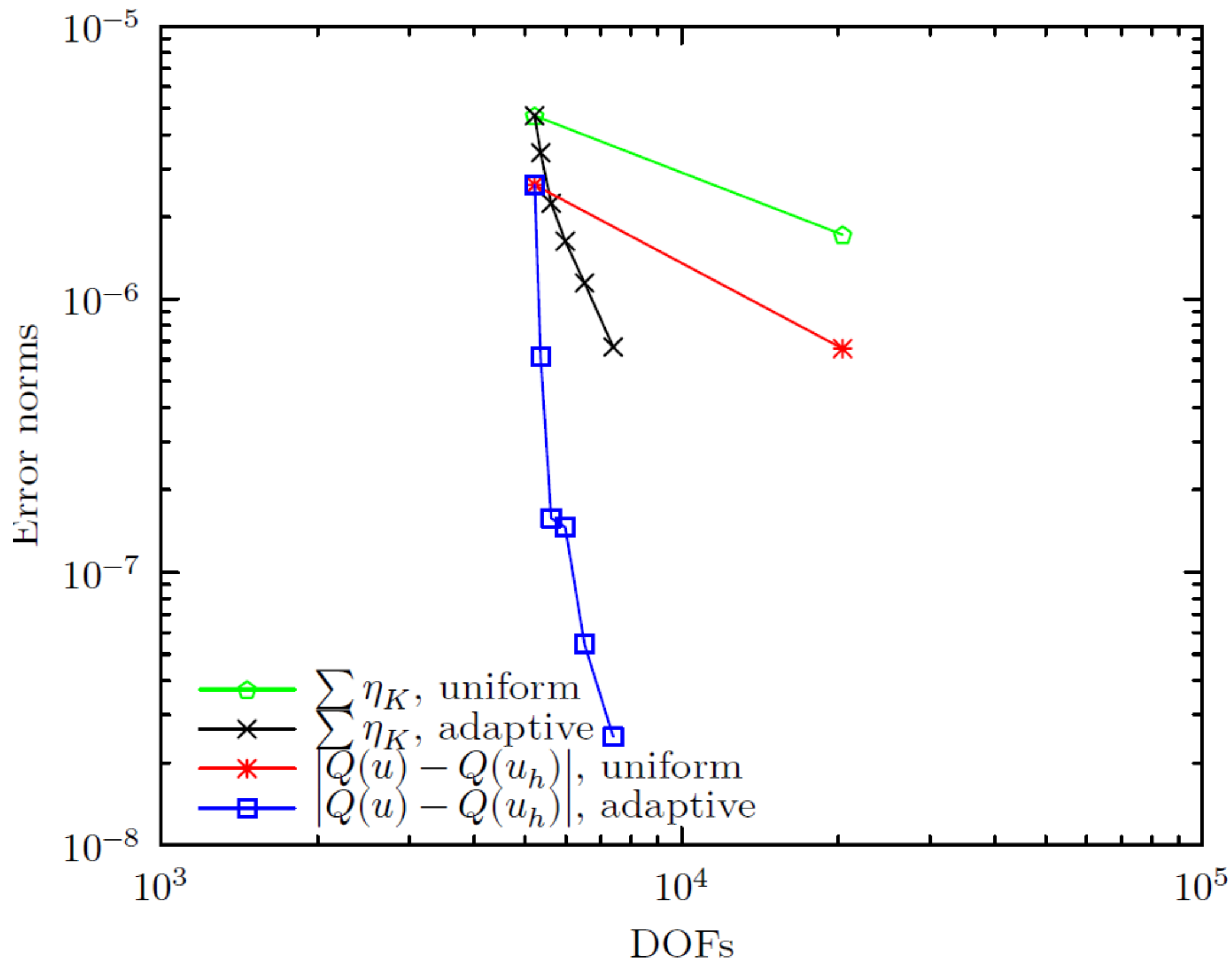
Corotational Cut Finite Element Method for real-time surgical simulation: application to needle insertion simulation, HP Bui et al, arXiv:1712.03052[cs.CE]

Human artery

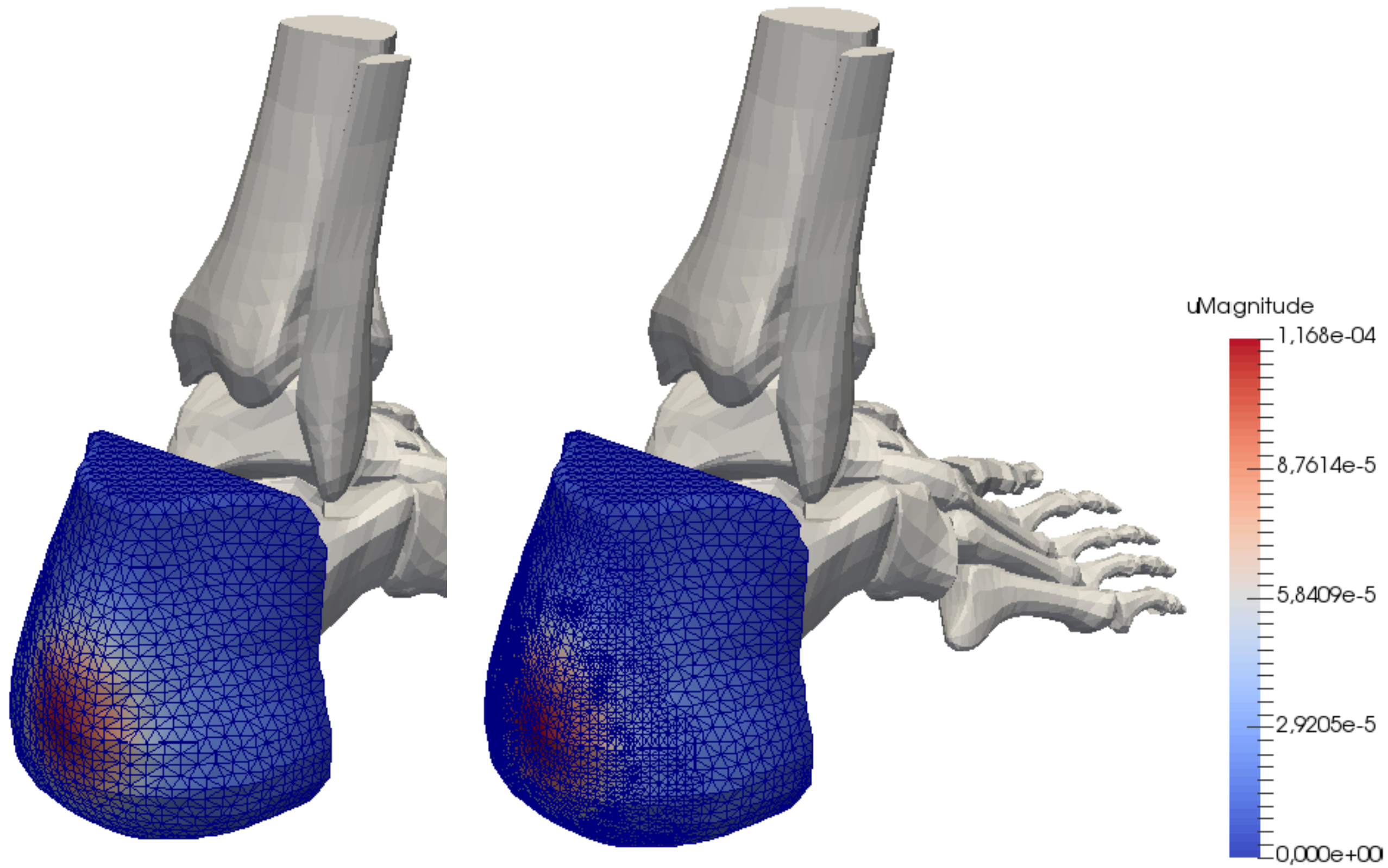


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Human artery



Human heel



Corotational Cut Finite Element Method for real-time surgical simulation: application to needle insertion simulation, HP Bui et al, arXiv:1712.03052[cs.CE]

Conclusions

Cut FEM/XFEM for surgical simulations with complex geometries

Making the discretization as independent as possible from geometric description

Verification of convergences with optimal rates

Cut FEM is suitable for real-time and patient specific simulations

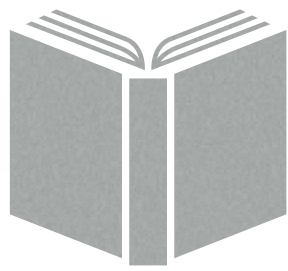
Corotational Cut Finite Element Method for real-time surgical simulation: application to needle insertion simulation, HP Bui et al, arXiv:1712.03052[cs.CE]

Perspectives

Higher order cut elements

Alexei Lozinski and Franz Chouly: avoid
integration on cut elements

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References

Real-time Error Control for Surgical Simulation, HP Bui *et al*, *IEEE Trans. Biomed. Eng.*, 2016.

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Acknowledgements

University of Strasbourg USIAS (BPC 14/Arc 10138)

ERC-StG RealTCut (grant N° 279578)

European project RASimAs (FP7 ICT-2013.5.2 No610425)

Luxembourg National Research Fund (INTER/MOBILITY/14/8813215/
CBM/Bordas and INTER/FWO/15/10318764)

Legato team (Luxembourg) and MIMESIS team (Strasbourg)

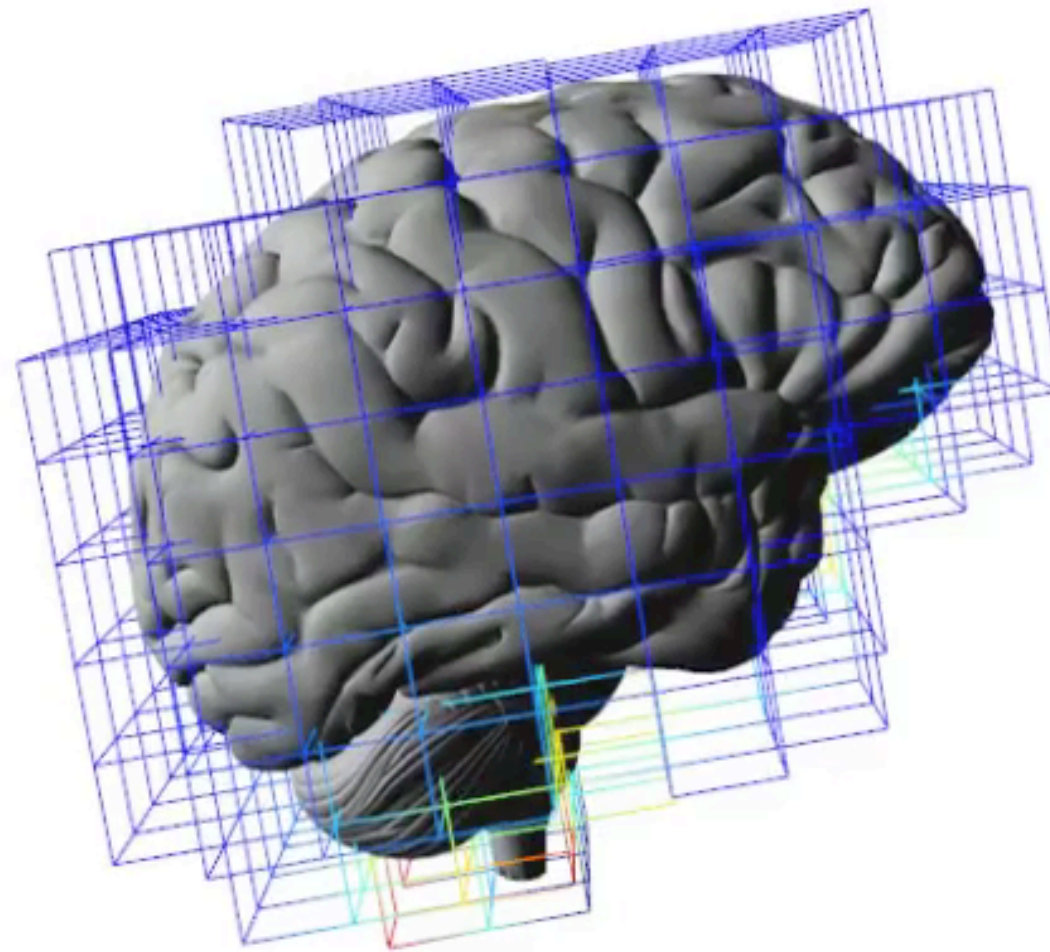
SOFA community

AMIES

LmB, University of Franche-Comté, CNRS

Real-time needle steering

Brain shift occurs
prior to cannula insertion



Skip

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