

SCALABLE KERNEL METHODS

in machine learning

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Outline

- Discuss a very small set of machine learning methods
- Introduce kernel matrices and kernel methods
- Discuss kernel approximation (for scalability)
 - Nystrom/Nbody
- Apply to image analysis problems

Logistics

- Possible project at the end
- MATLAB, bring your laptops for practice
- <https://piazza.com/tum.de/summer2016/informaticsvsientificcomputing>
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Google's Artificial Neural Network of 16,000 CPUs Learns to Detect Cats

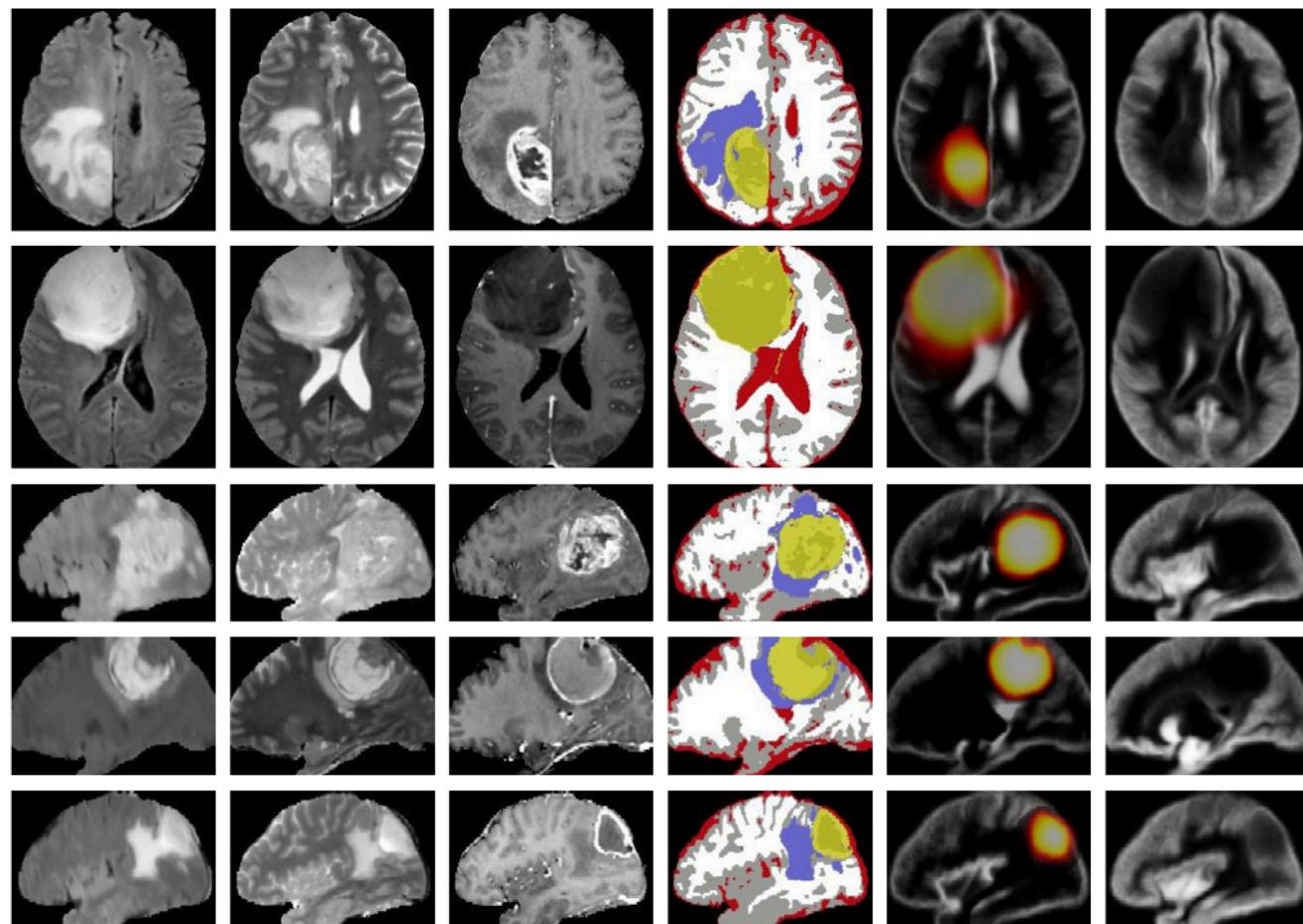
Posted on June 26, 2012



Digit recognition

0000000000000000
1111111111111111
2222222222222222
3333333333333333
4444444444444444
5555555555555555
6666666666666666
7777777777777777
8888888888888888
9999999999999999

Medical image analysis



Other examples

- Given DNA sequences
 - find genes
- Given web browsing history
 - recommend ads
- Given movie history
 - recommend new movies
- Given scientific articles
 - group them according to topics
- Given time series of pressure
 - recognize speech (words spoken)
- Given high-energy physics particle data
 - classify it as a Higgs boson
- Given MRI,CT, blood work, history profile, symptoms
 - diagnose

Notation

TRAINING SET

$$\underline{\mathcal{X}} = \{x_i\}_{i=1}^N, \quad x_i \in \mathbb{R}^d$$

$$\underline{\mathcal{Y}} = \{y_i\}_{i=1}^N; \quad y_i \in \mathbb{R}$$

N : NUMBER OF POINTS

d : DIMENSION

LEARNING

Given point x
 $y = f(x; w)$ parameters
(TRAINING).

↓ → LEARNING ALGORITHM
PREDICTION

How to learn? Optimization

Solve a minimization problem
propose a loss function

$$\min_w \mathcal{L}[\bar{\mathbf{y}}, \bar{\mathbf{x}}, w]$$

Function approximation

- Given \mathbf{X} and \mathbf{Y} construct $f()$ that approximates the function
- Assumes no noise
- Does not have a unique solution
- 200 years to figure everything out

Regression

- Term used in statistics
- Unclear what is the input and output
- Even if we know X and Y are noisy
- In these lectures approximation and regression are the same

Supervised classification

- Y is a categorical variable
- **Binary classification** $\{-1, 1\}$
- Multiclass $\{0, 1, 2, \dots\}$
- Trick for binary: $\text{sign}(f)$
- Trick for multiclass : one vs other

Unsupervised classification

- We are only given \mathbf{X}
- We want to group into k classes
 - sometimes k is not known
- The algorithms for this problem don't fit the function approximation.

Digit recognition

MNIST database; normalized digits from NIST institute in the US

<http://yann.lecun.com/exdb/mnist/>

<http://www.cs.nyu.edu/~roweis/data.html>

matlab format

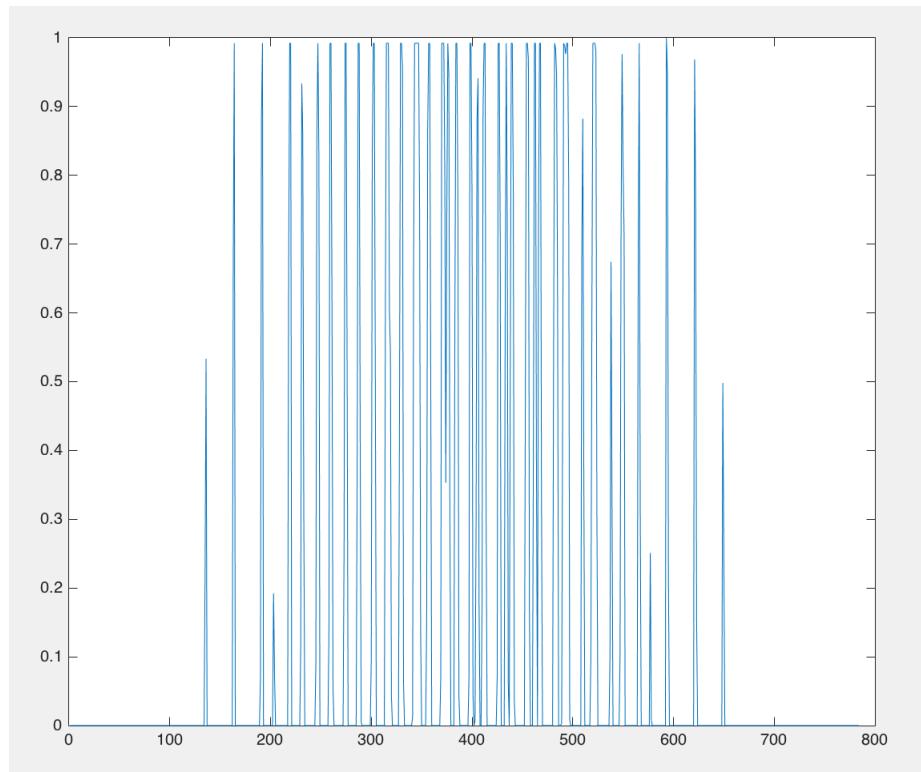
N= 60000

D = 784

Goal given training test, construct an algorithm that classifies correctly
unseen input

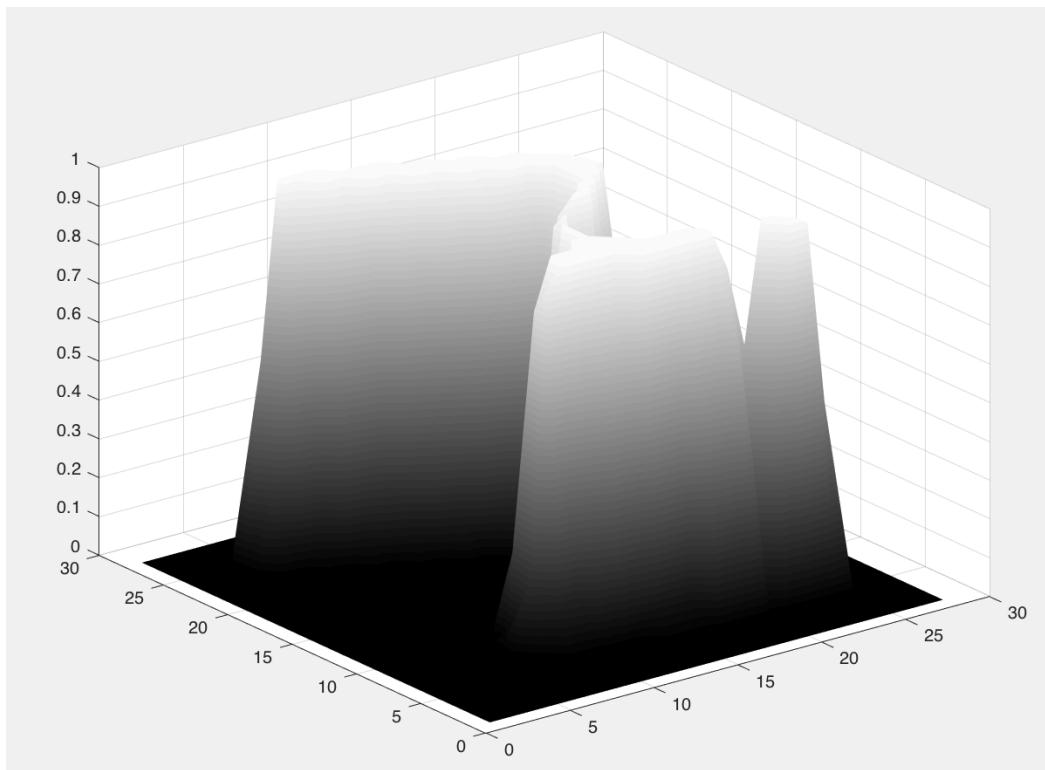
Digit recognition (mnist)

plot(x)



Digit recognition (mnist 60K)

`surf(reshape(x,[28,28]))`



`imshow(reshape(x,[28,28]))`



- BREAK

Austin, Texas



Location in the United States of America

Coordinates:  30°15'0"N 97°45'0"W



Keep Austin Weird

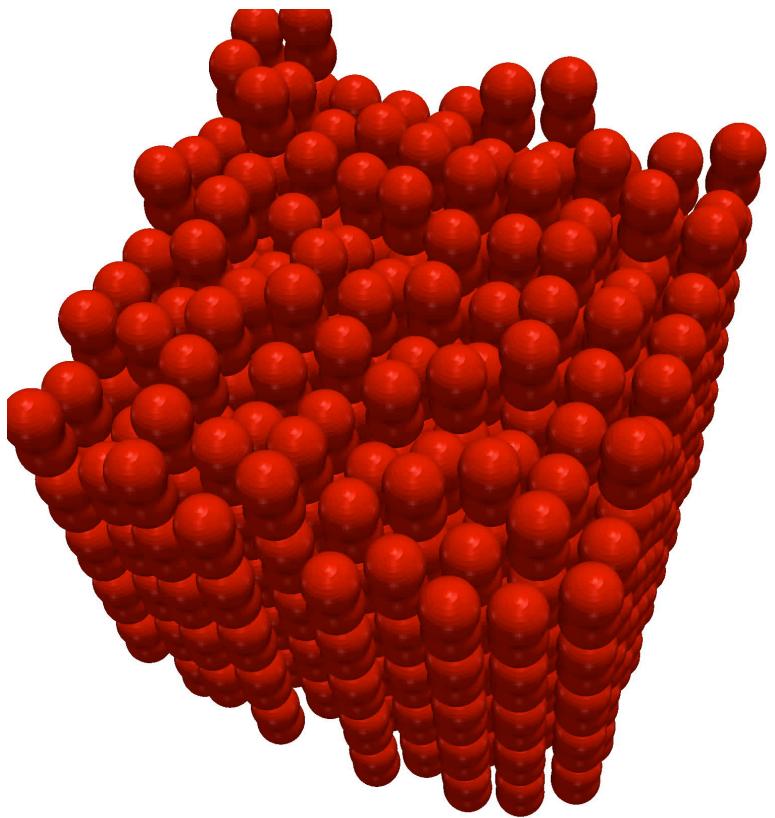
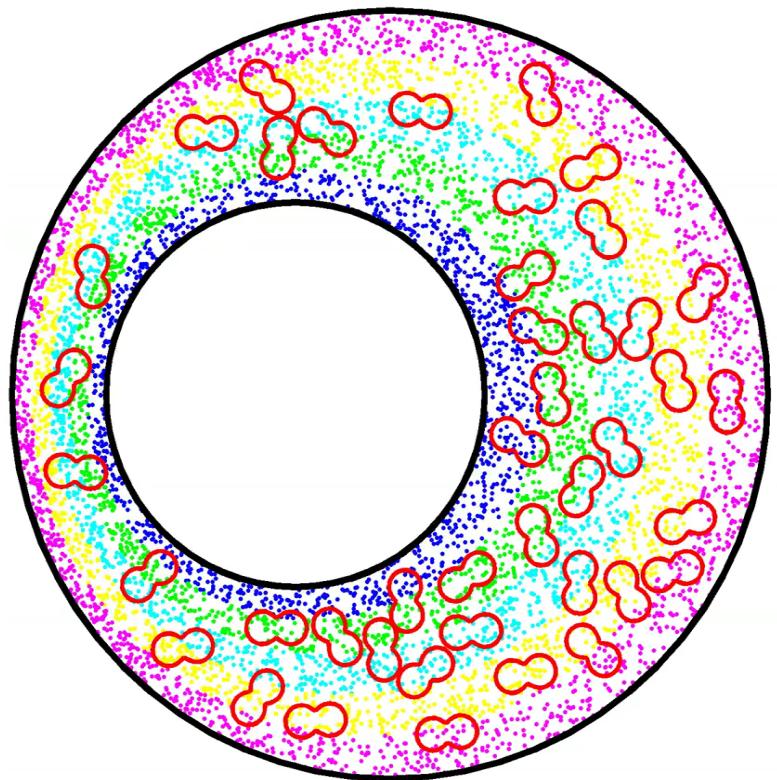


PADAS Lab: padas.ices.utexas.edu

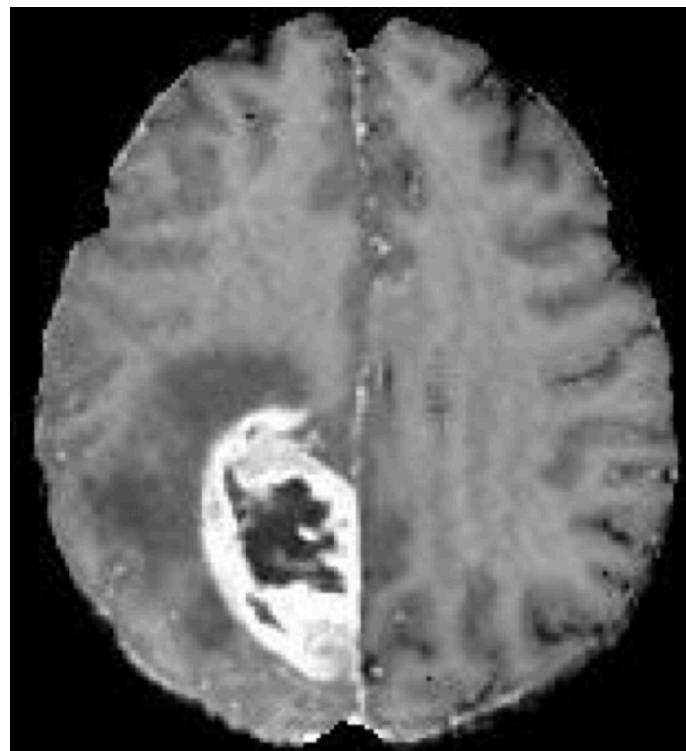
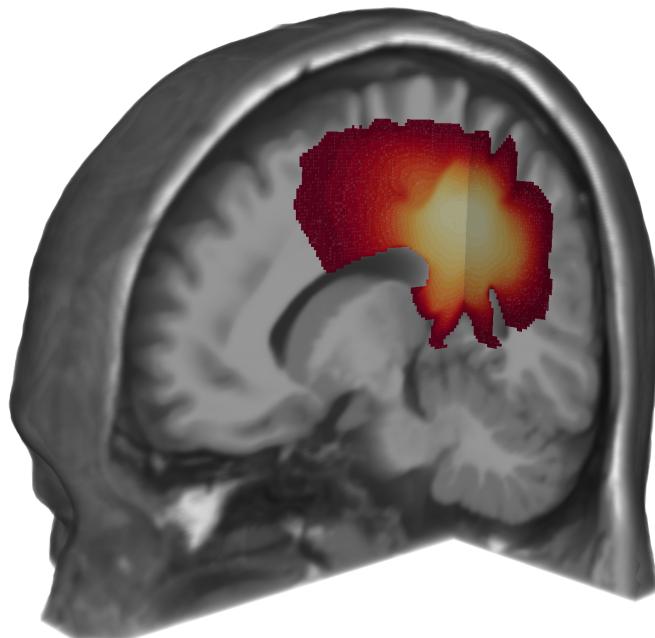
Parallel Algorithms for Data Analysis and Simulation

PI: George Biros (ICES, CS, Mechanical Engineering)

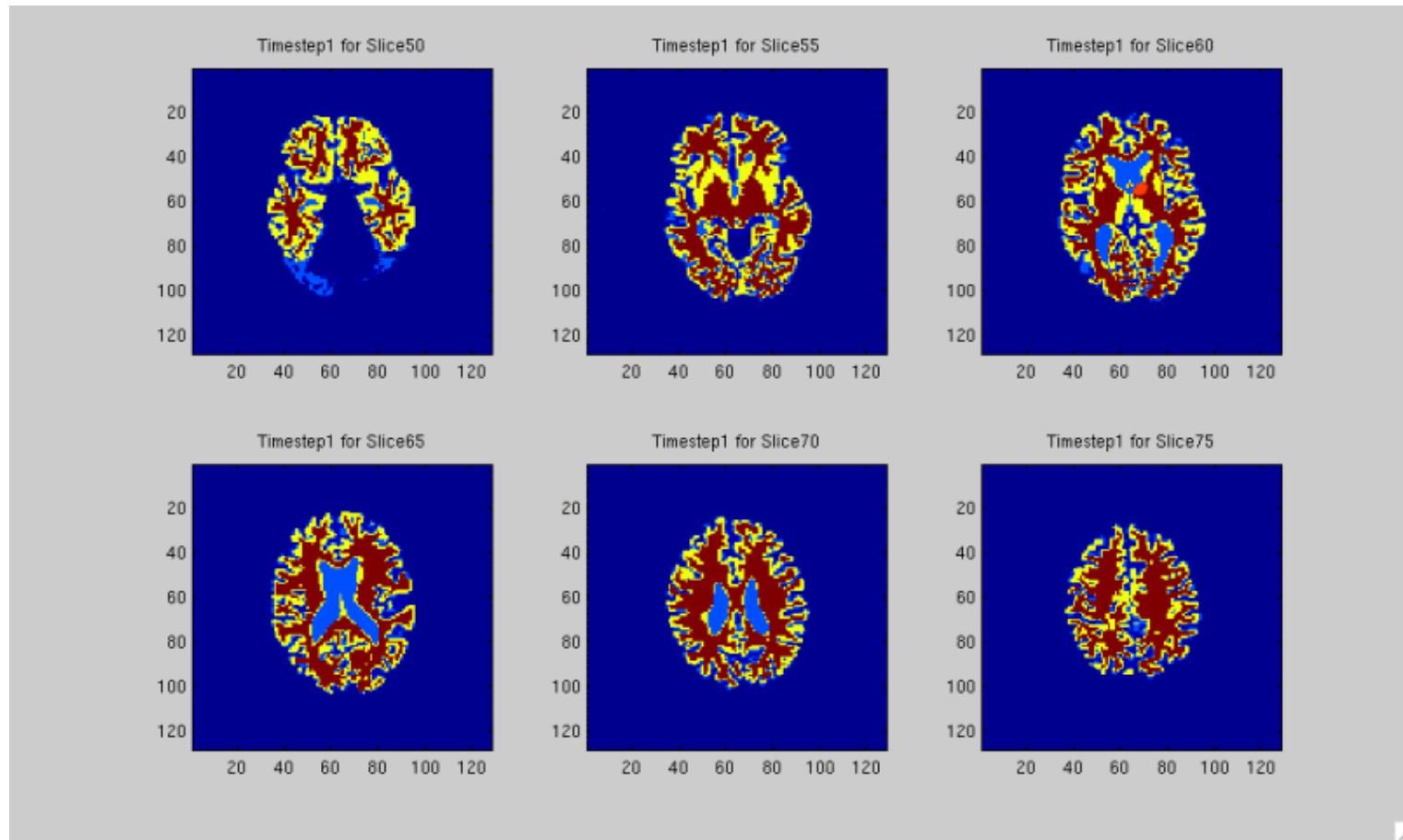
- **MISSION:** High-performance algorithms and software for core problems in numerical analysis, machine learning, and simulation; scalability on millions of heterogeneous cores
- **ALGORITHMS:** N-body algorithms, high-order methods, fast PDE solvers, fast approximation, inverse problems, regression, classification, model reduction, uncertainty quantification
- **APPLICATIONS:** biomechanics, fluid mechanics (microfluidics, nanofluidics, complex fluids), image analysis (registration, segmentation, classification), inverse scattering, convection-reaction-diffusion systems, rare event detection, data assimilation
- **SOFTWARE:** padas.ices.utexas.edu/software



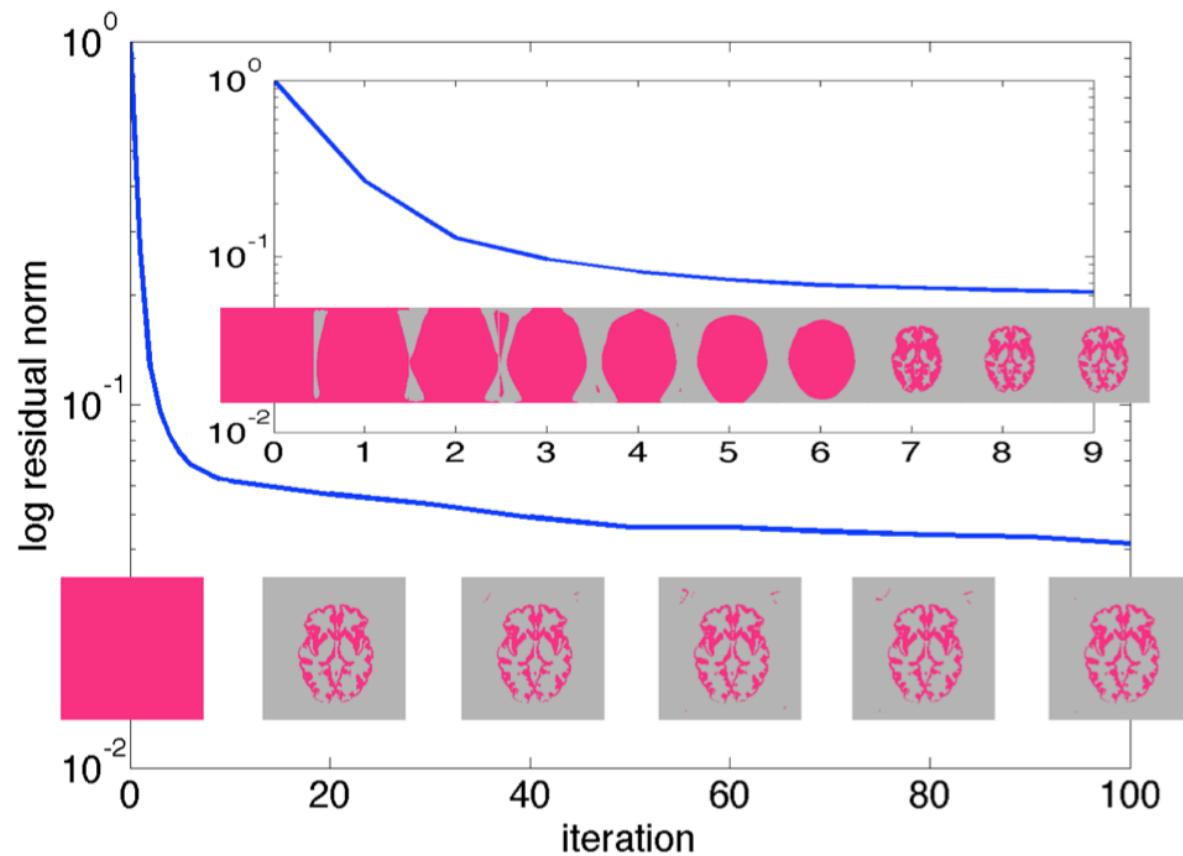
Biophysical models and image analysis



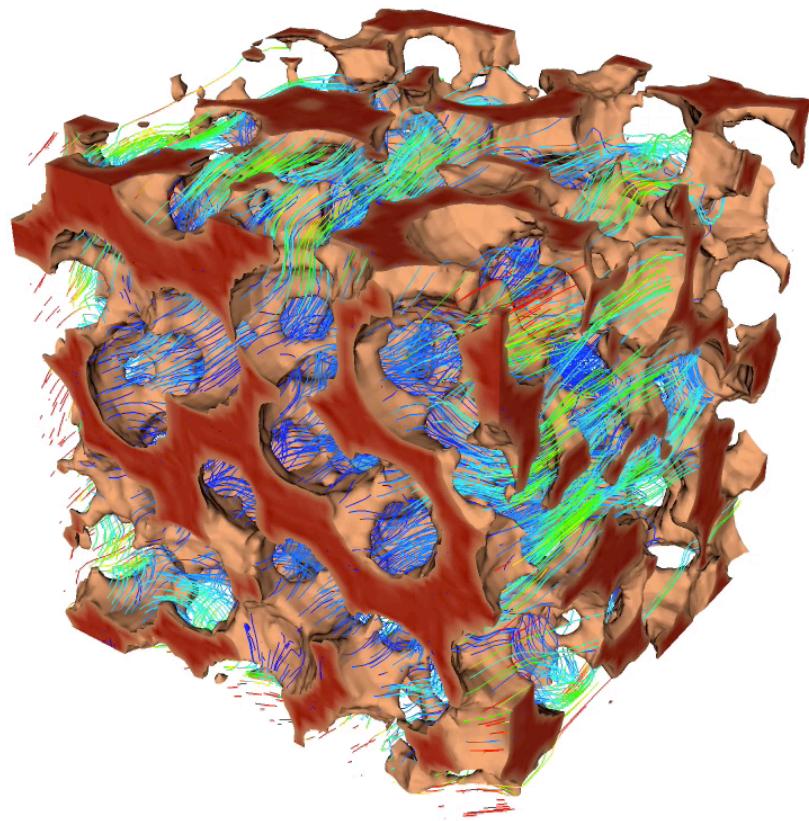
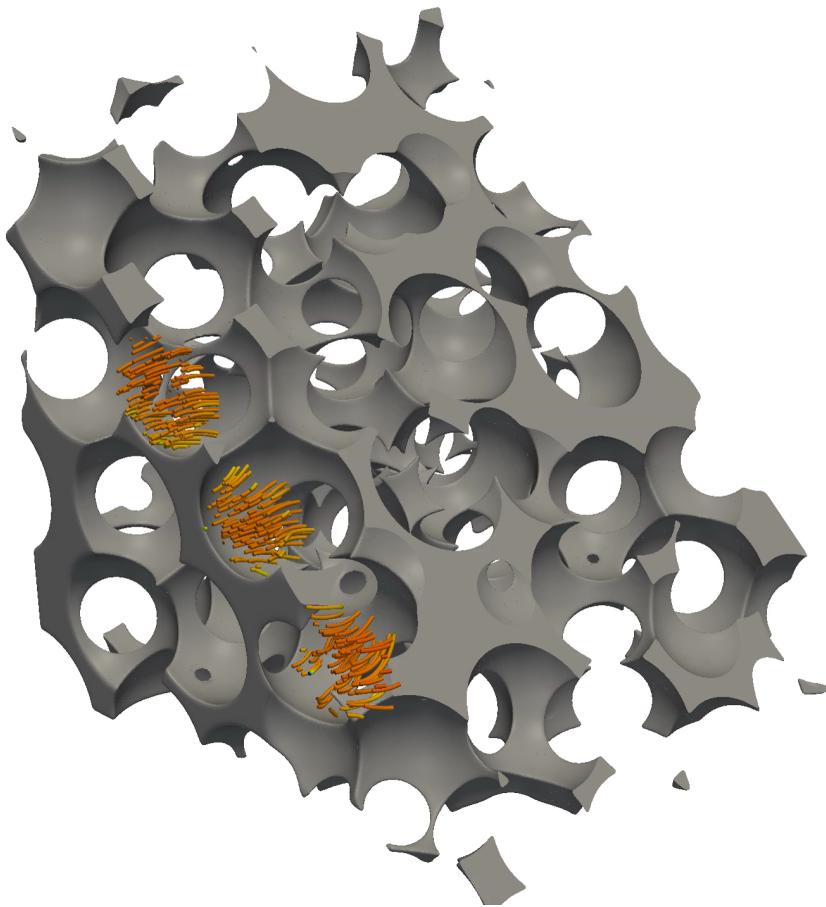
Tumor simulations



Convergence



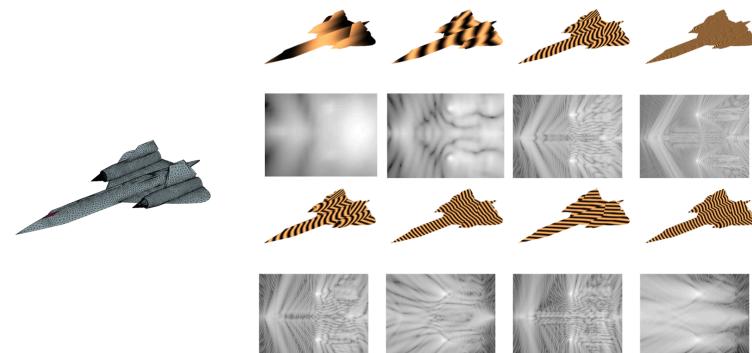
Multiphysics



FaIMS: $\mathcal{O}(N_\omega N)$

SVD: $\mathcal{O}(N^2 N_s N_d N_\omega)$

Lanczos: $\mathcal{O}(N_i N N_s N_\omega)$



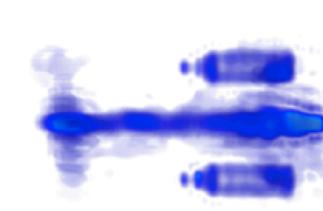
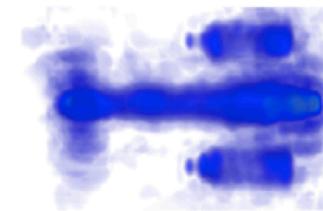
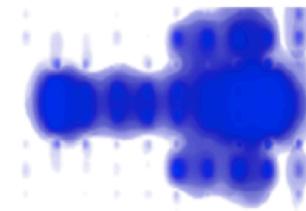
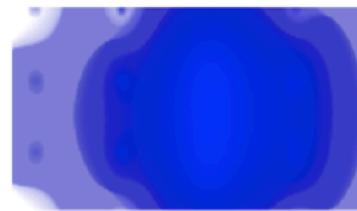
Low frequency, small number of events

0.5L/1 event

1L/8 events

4L/8 events

4L/32 events



8L/512 events

