



Open Source Software for MRI Reconstruction

2026-05-09 – ISMRM26 | Cape Town

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Declaration of Financial Interests or Relationships

Speaker Name: Pierre-Antoine Comby

I have no financial interests or relationships to disclose with regard to the subject matter of this presentation.

Outline

■ MRI Reconstruction

1. Landscape of MRI Reconstruction

Overview of current usage in languages and toolboxes

2. MRI-NUFFT as a case study

3. Beyond toolboxes and software

Dataset, Benchmarking, Community Initiative

■ Good practice to take home

MRI Reconstruction



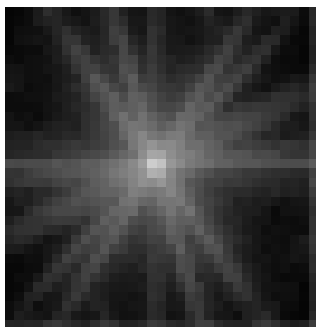
- Solving an (or many) *ill-posed* **inverse problem**

$$y = Ax + \varepsilon$$

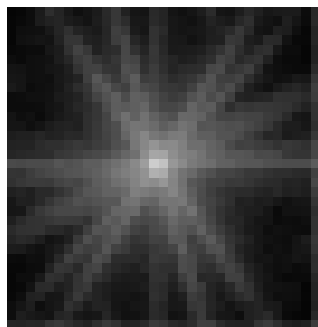
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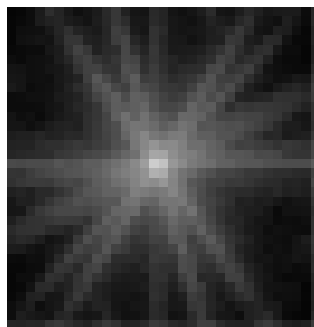
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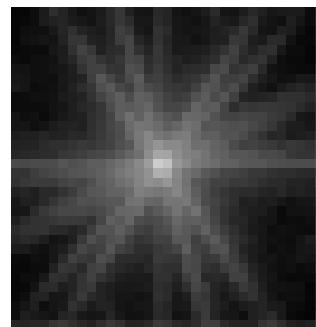
Anatomical MRI



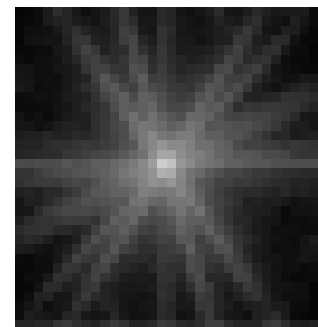
Cardiac /CINE



Quantitative MRI



Diffusion MRI

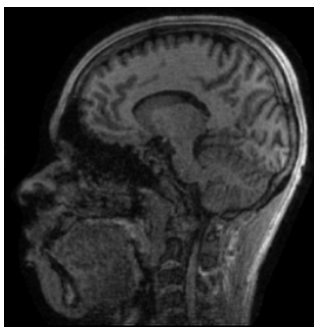


Functional MRI

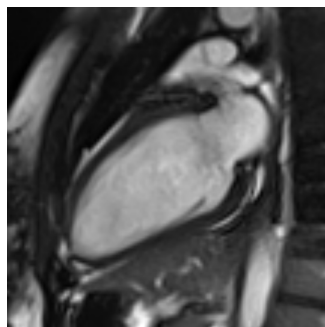
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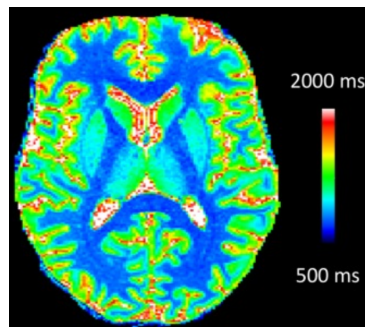
$$y = Ax + \varepsilon \implies x = \mathcal{M}(A, y, \text{priors})$$



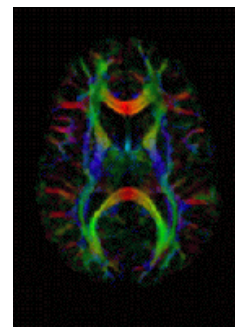
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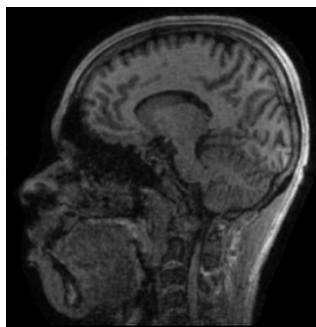


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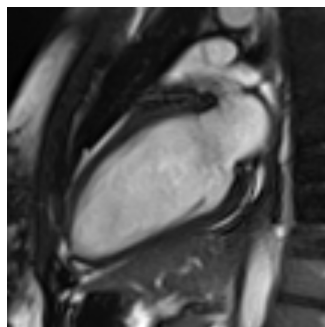
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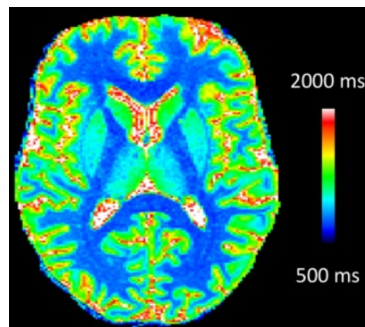
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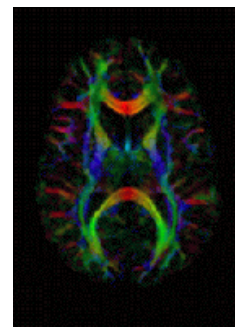
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Functional MRI

Common Goal: Solve the inverse problem

Design (Learn ?) and validate $\mathcal{M}$, A and *priors*

Open-source software

- Open source vs free software
 - “Free as in beer vs free as in freedom”

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- Open the black box
 - Access to source code, and (sometimes) freedom of modification
- Licences to (un)protect the code !

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- Licences to (un)protect the code !
- Good license choices
 - MIT, BSD-3-clause, (A)GPL¹
 - Or whatever your legal department requires

¹<https://opensource.stackexchange.com/questions/217/what-are-the-essential-differences-between-the-bsd-and-mit-licences>

Where do we stand with open-source in MRI ?

- Acquisition:
Pulseq, MTRK, ...

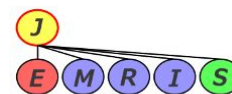


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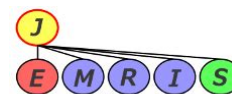


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■ Open-source Hardware
Low-Field MRIs

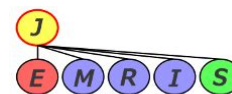


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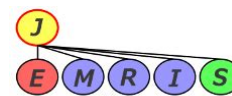


- Reconstruction Software

⇒ This talk !

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- Simulation
KomaMRI.jl, Jemris, **SNAKE**, ...
- Open-source Hardware
Low-Field MRIs
- Reconstruction Software
- Analysis
Nilearn, **f/s/d mriprep**, ...



⇒ This talk !





1. ■ MRI Reconstruction software

Software usage in the Literature

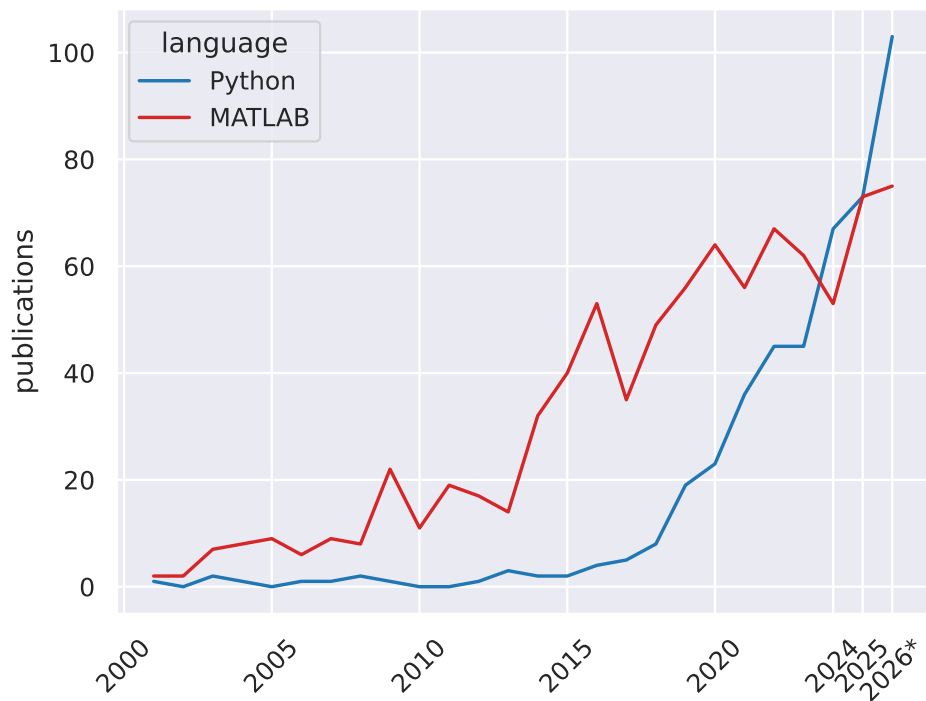


Figure 1: Published articles with “MRI”+*Language* in their abstract or title.

2026: Extrapolated results from Q1

Software usage in the Literature

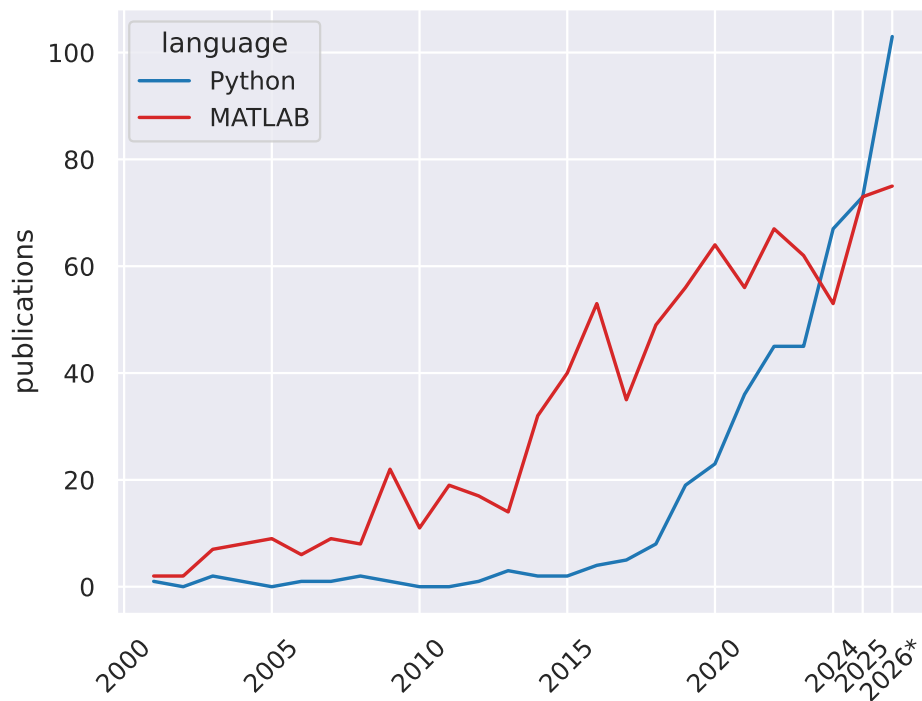


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- Python adoption accelerates, and has stronger ecosystem
 - LLMs loves it
 - PyTorch / JAX for deep learning and GPU based reconstruction
 - 2026 is (finally) the year of Python for MRI ?

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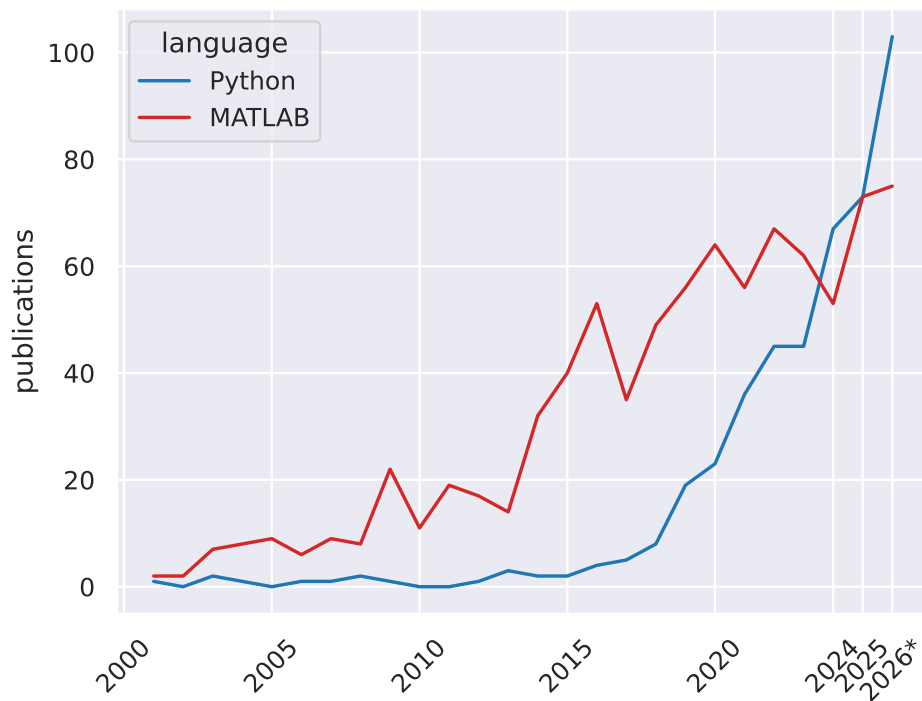


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 - 2026 is (finally) the year of Python for MRI ?
- MATLAB is still widely use (legacy code base)
 - But Julia grows !

Open Source Repositories

- ≈ 150 Repos filter with topic mri & mri-reconstruction

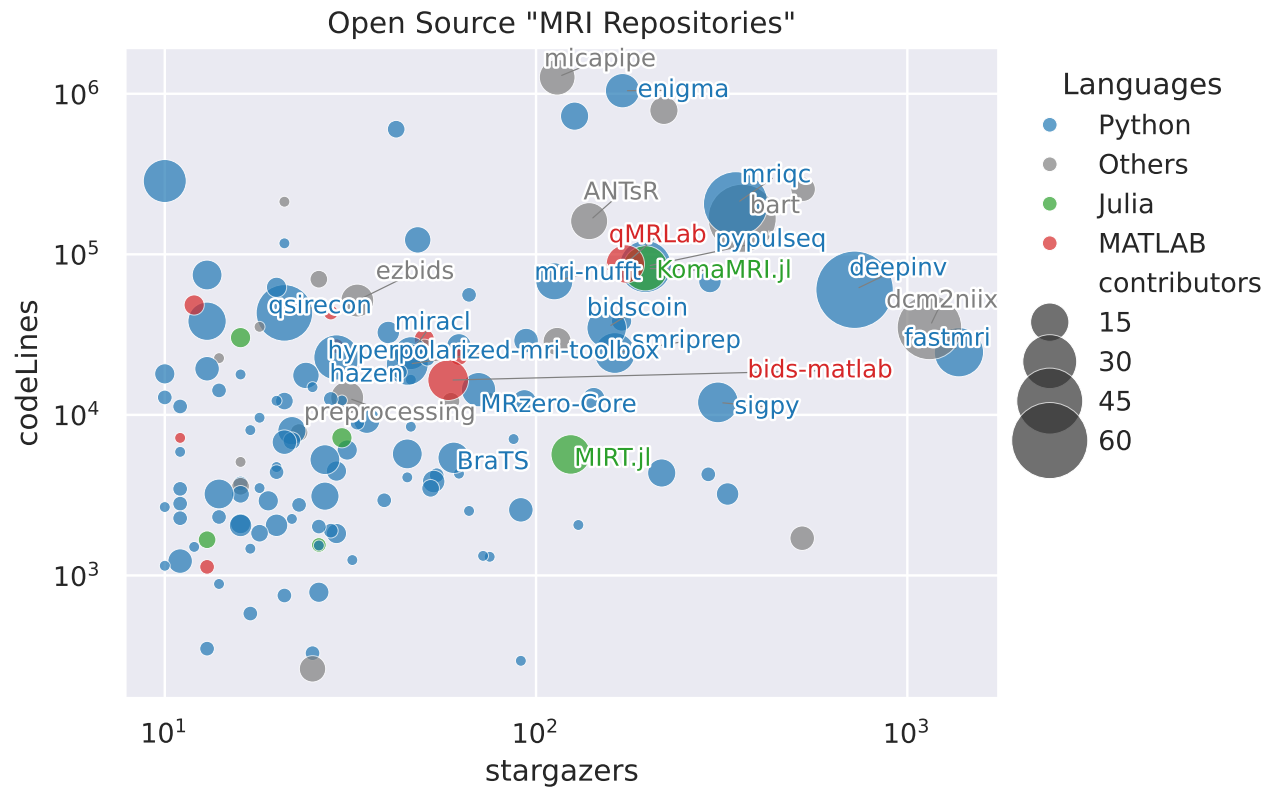


Figure 2: Main GitHub Repositories with 'mri' topic

Open Source Repositories

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- Python/Notebook is the most represented on GitHub
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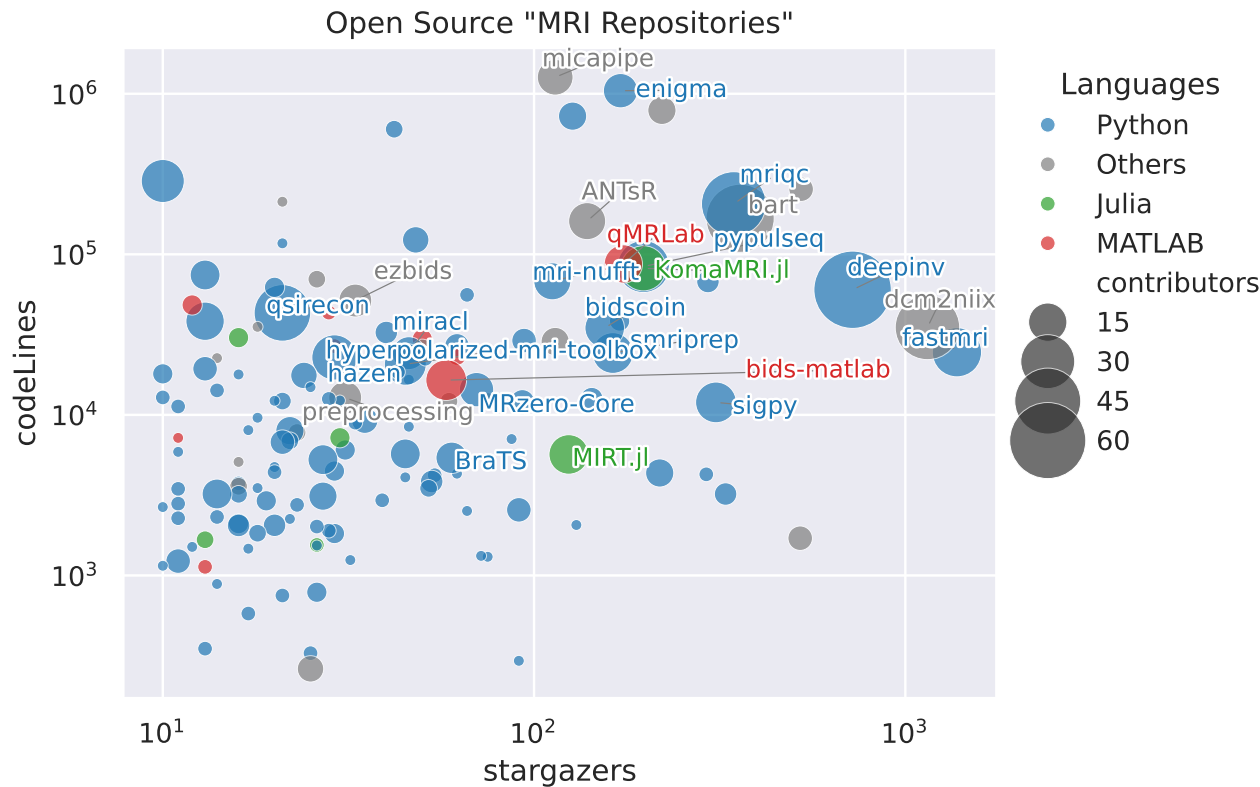


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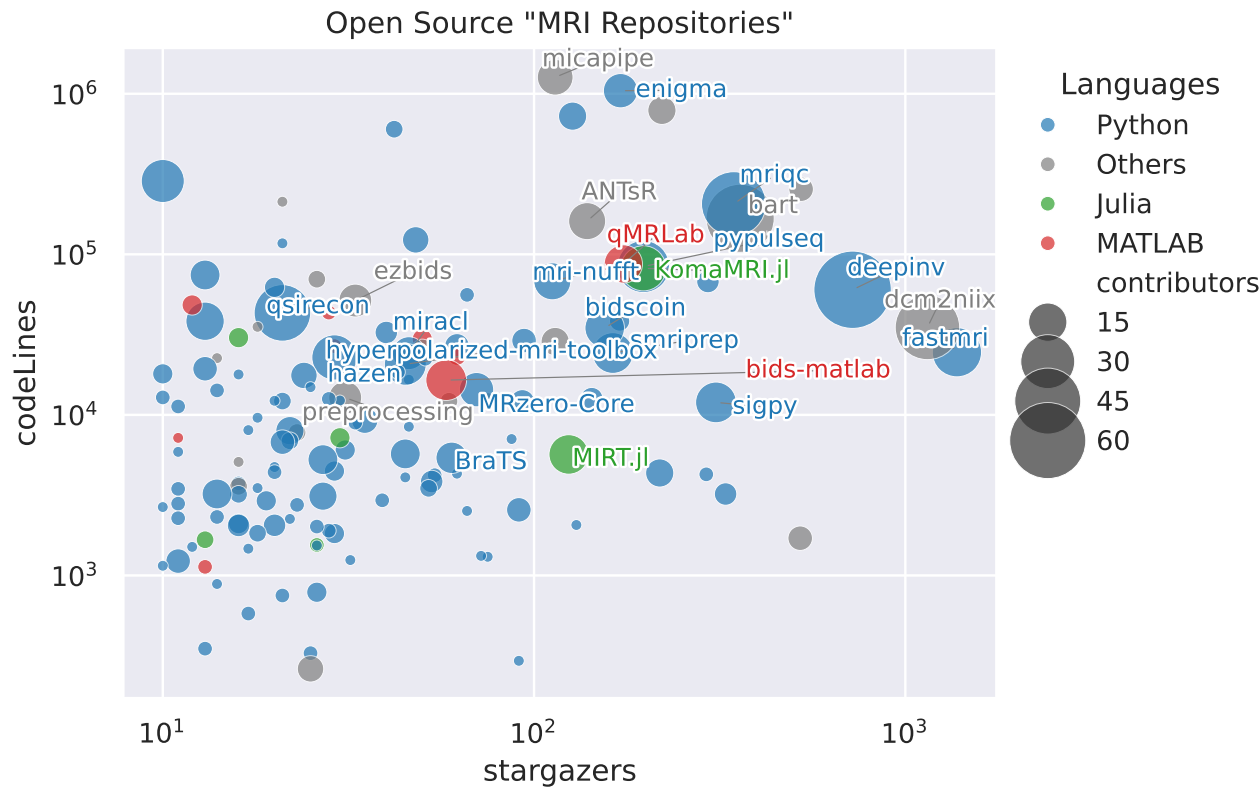


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- Only 56% report homepage/documentation!

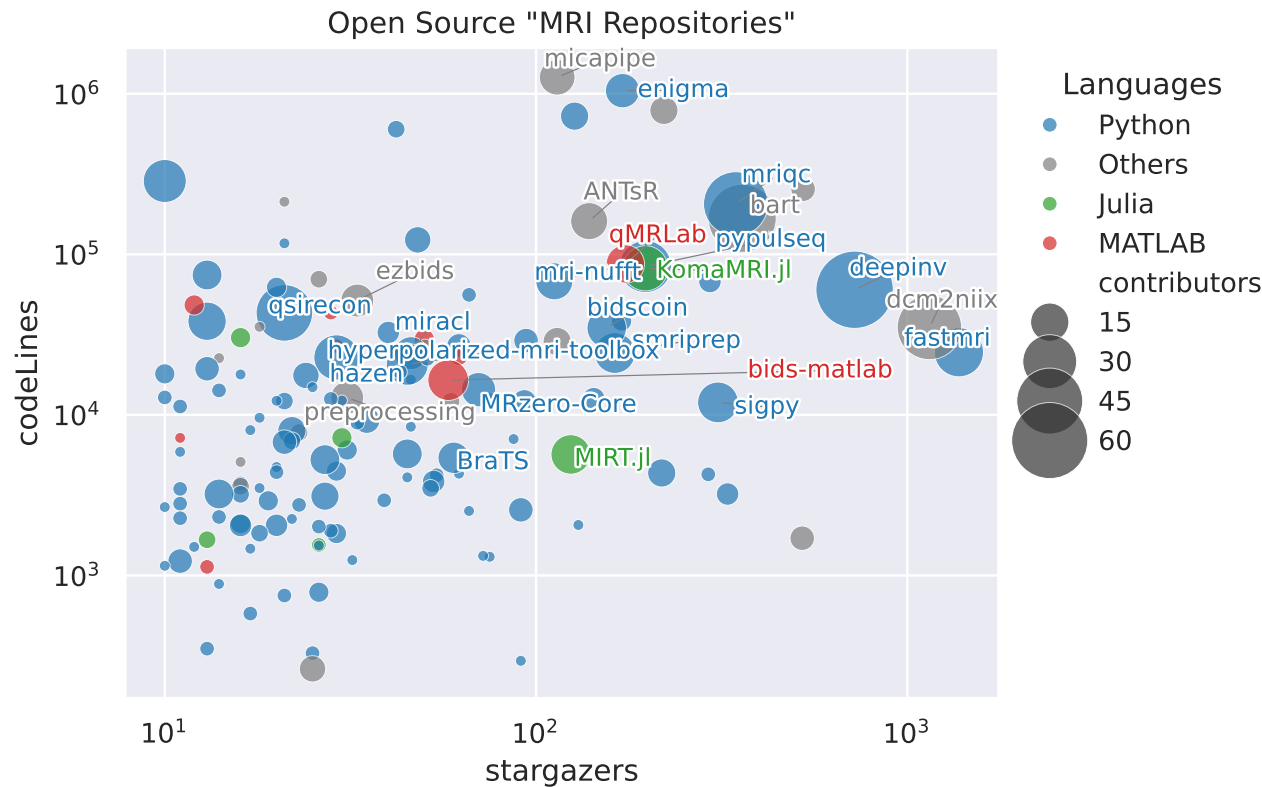


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Catch them all !



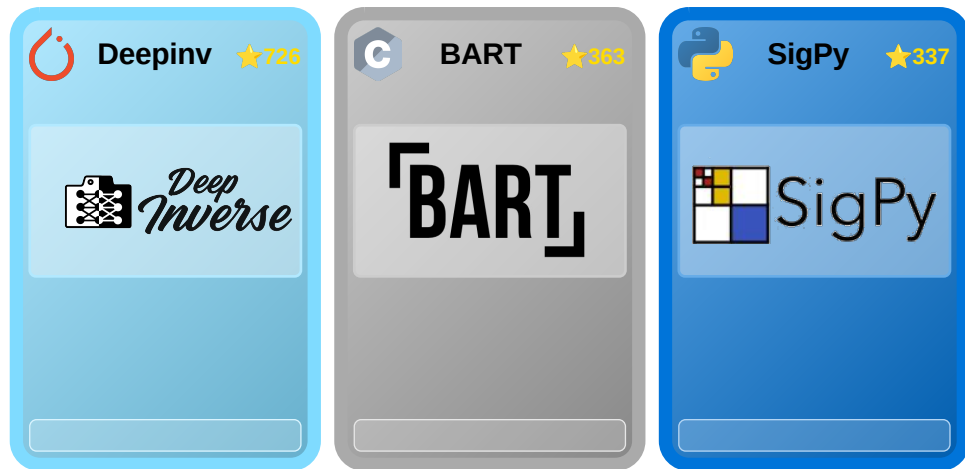
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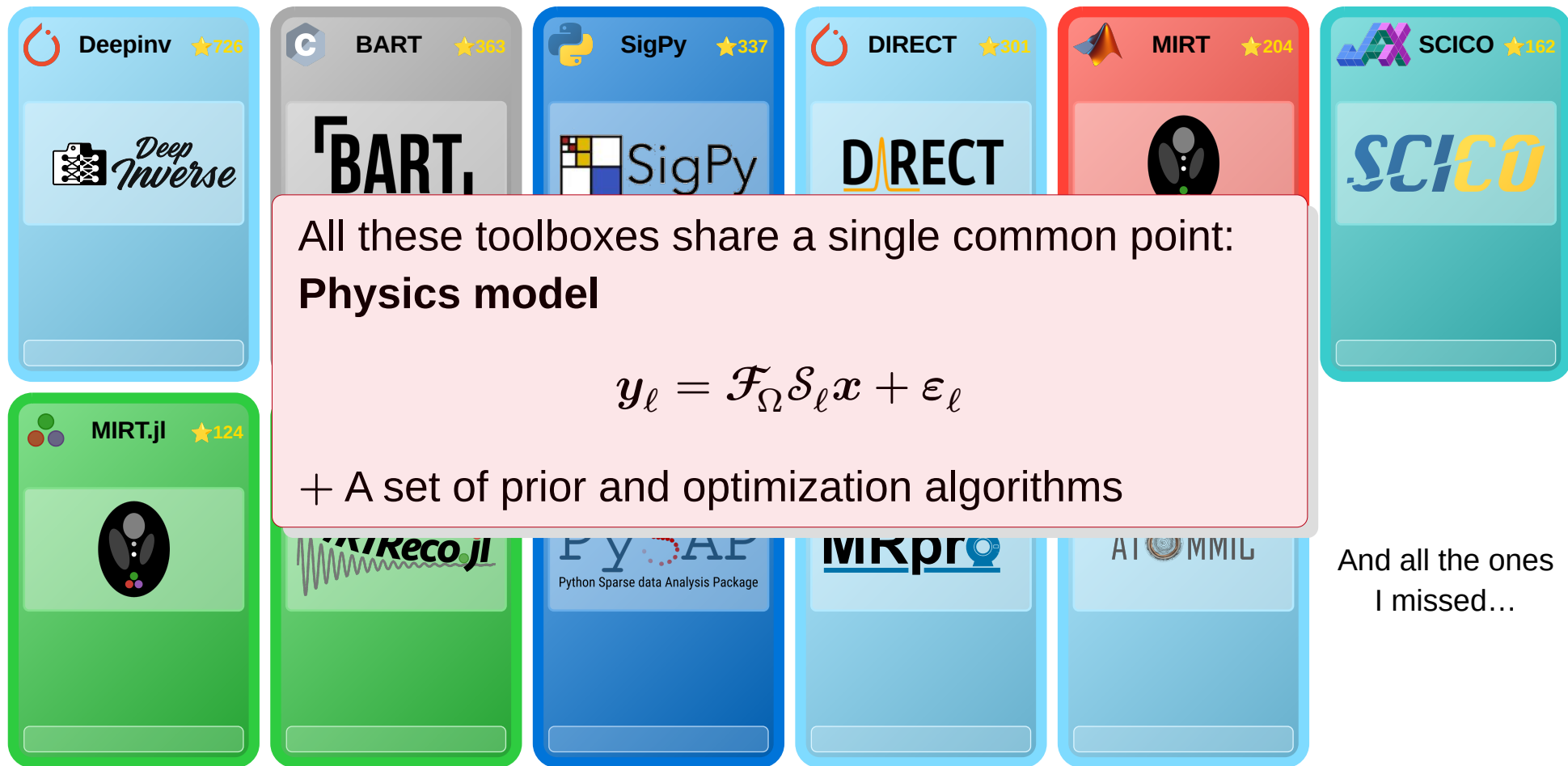


Catch them all !



And all the ones
I missed...

Catch them all !





2. ■ MRI-NUFFT: Case study

Introducing MRI-NUFFT



- Most MR Toolboxes focuses on Cartesian sampling

Introducing MRI-NUFFT



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- Non-Cartesian Reconstruction has great potential
 - ✓ Freedom of sampling locations
 - ✓ Increased sampling efficacy
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 - ✗ Off-Resonance effects
 - ✗ (Re)-design non-Cartesian trajectories for sequences
 - ✗ MR System constraints

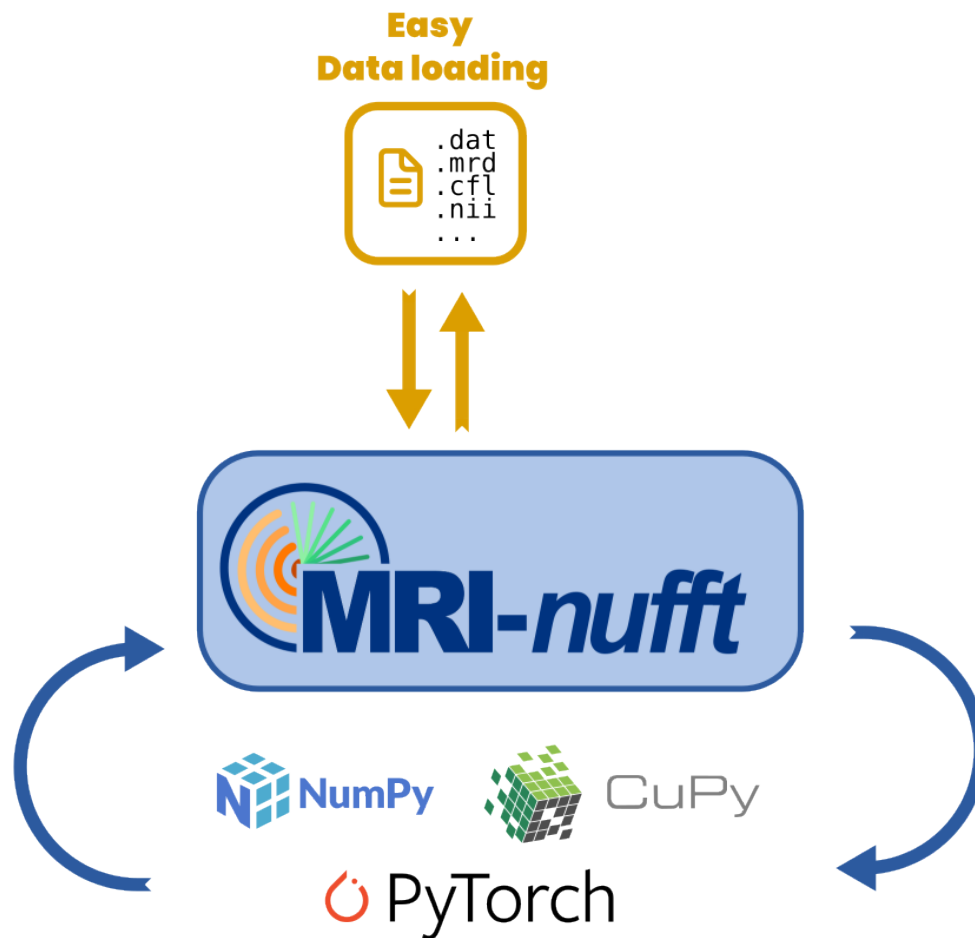
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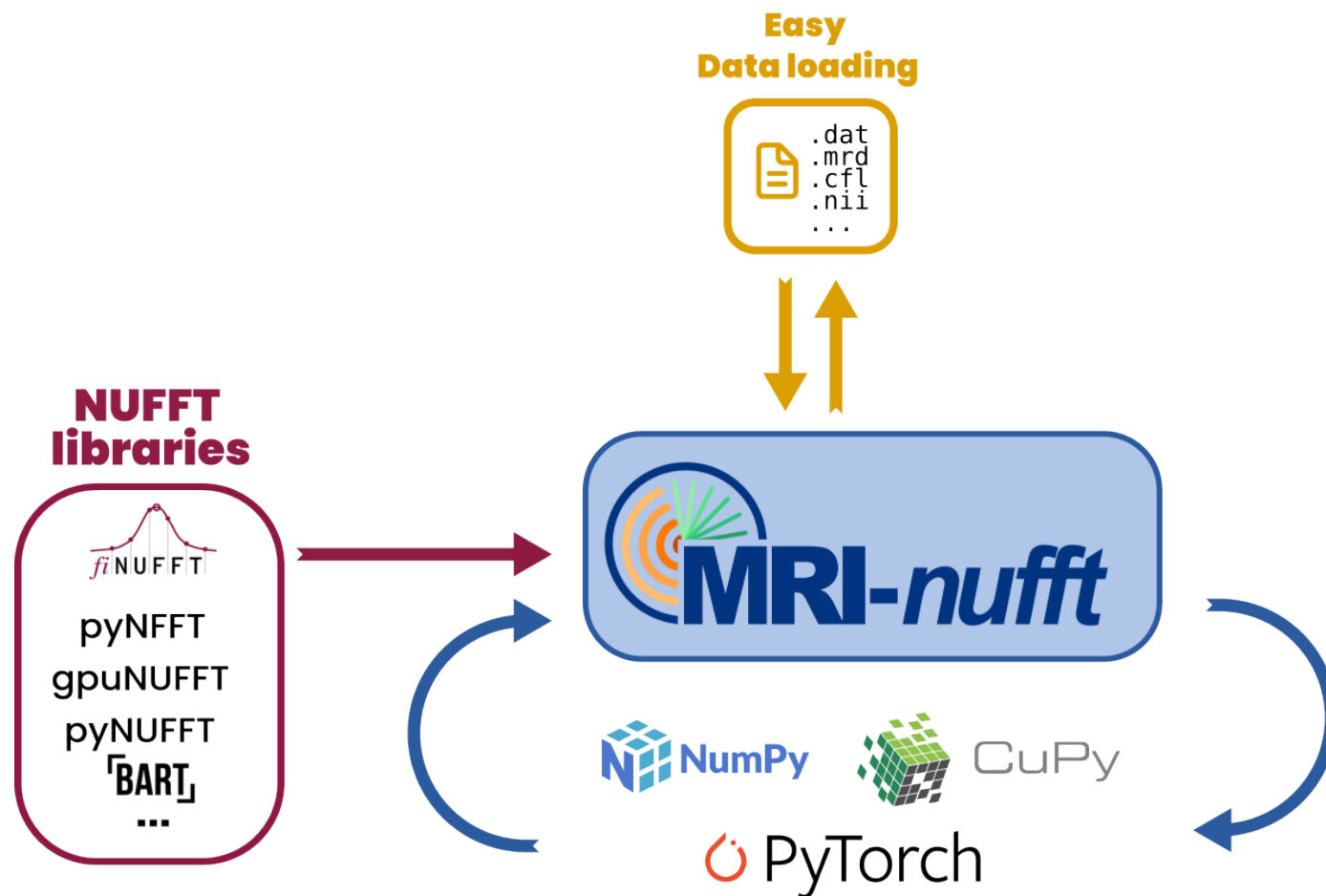
Let's fix this with:



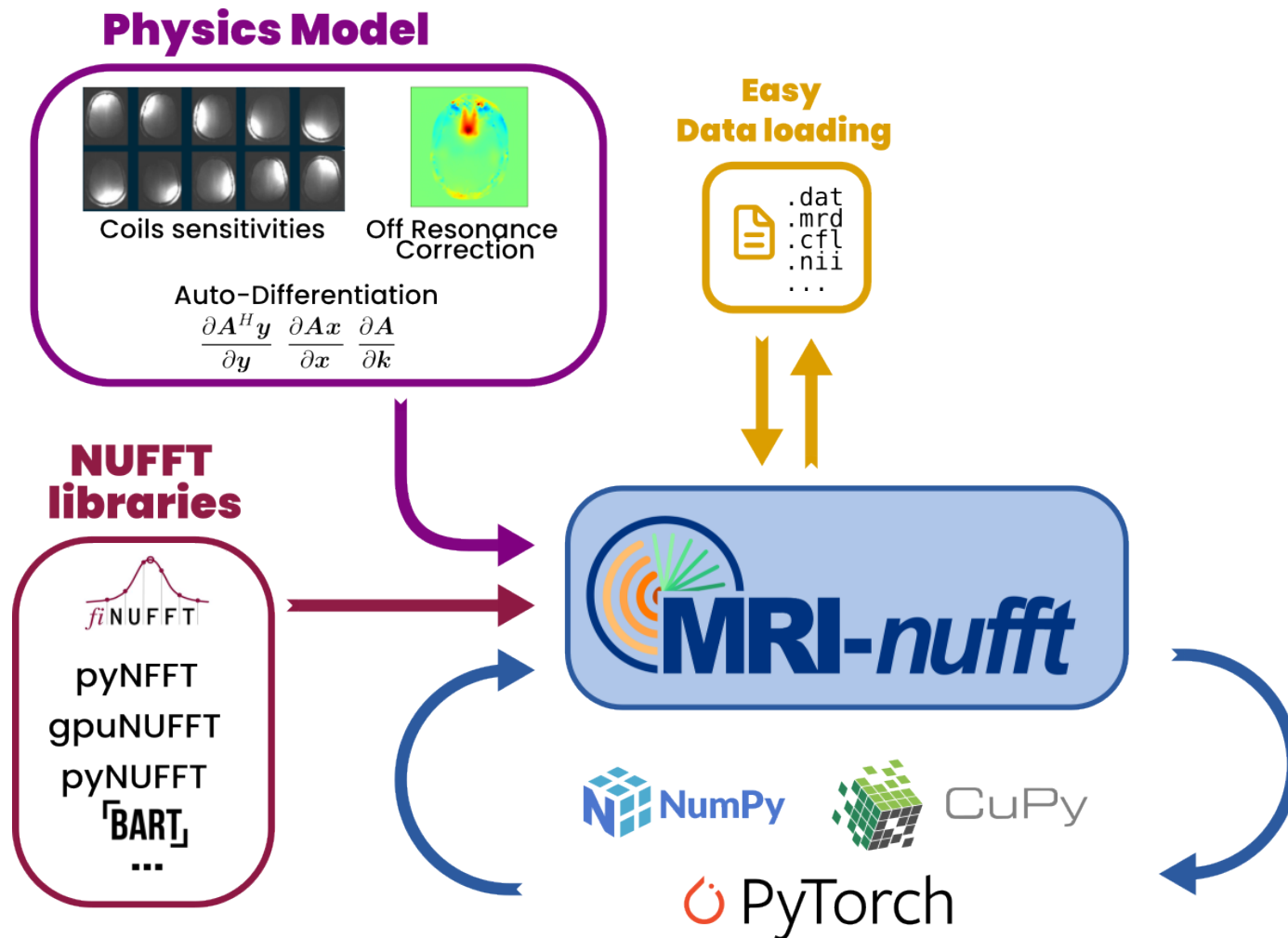
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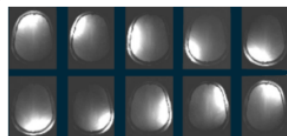
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Learn more

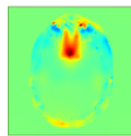
AI Driven k-space
Sampling

Mon. 11/05 7:30-8 AM
Chaithya GR.

Physics Model



Coils sensitivities



Off Resonance
Correction

Auto-Differentiation

$$\frac{\partial A^H y}{\partial y} \quad \frac{\partial A x}{\partial x} \quad \frac{\partial A}{\partial k}$$

NUFFT libraries

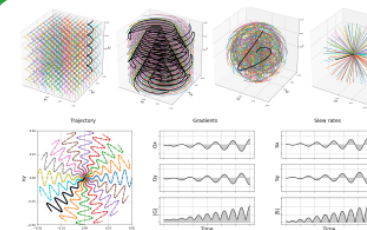


pyNUFFT
gpuNUFFT
pyNUFFT
「BART」
...

Easy
Data loading



≥ 50 k-space
trajectories



NumPy



CuPy



PyTorch

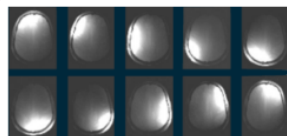
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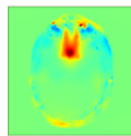
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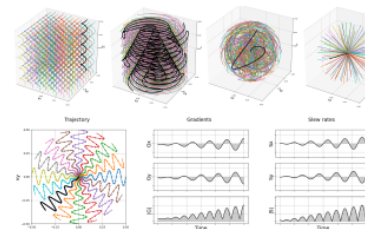


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PyTorch

NUFFT for MRI Reconstruction



- Inverse problem formulation for MRI

$$\hat{\boldsymbol{x}} = \arg \min_{\boldsymbol{x} \in \mathbb{C}^N} \frac{1}{2} \sum_{\ell=1}^L \|\mathcal{F}_{\Omega} \mathcal{S}_{\ell} \boldsymbol{x} - \boldsymbol{y}_{\ell}\|_2^2 + \lambda g(\boldsymbol{x})$$

NUFFT for MRI Reconstruction

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*“Priors and Networks are temporary, ...
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MRI-NUFFT provides *efficient* implementation for the forward model operator:

- based on the NUFFT backend of your choice.
- Handles CPU/GPU transfers and array conversion (torch/numpy/cupy)
- autograd support $\frac{\partial \mathbf{A}x}{\partial x}$, $\frac{\partial \mathbf{A}^H y}{\partial y}$, $\frac{\partial \mathbf{A}_{\theta} x}{\partial \theta}$, ...

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```
from mrinufft import get_operator
nufft = get_operator("cufinufft", samples, shape, smaps=smaps,
density="pipe")
ksp = nufft.op(image)          #  $y = Ax$ 
adjoint = nufft.adj_op(ksp)    #  $x = A^H y$ 
pinv = nufft.pinv_solver(ksp)  #  $x = A^{\dagger} y$ , using CG
```

Extended Fourier Model

- Multi-coil, subspace representations, ...
-



Extended Fourier Model

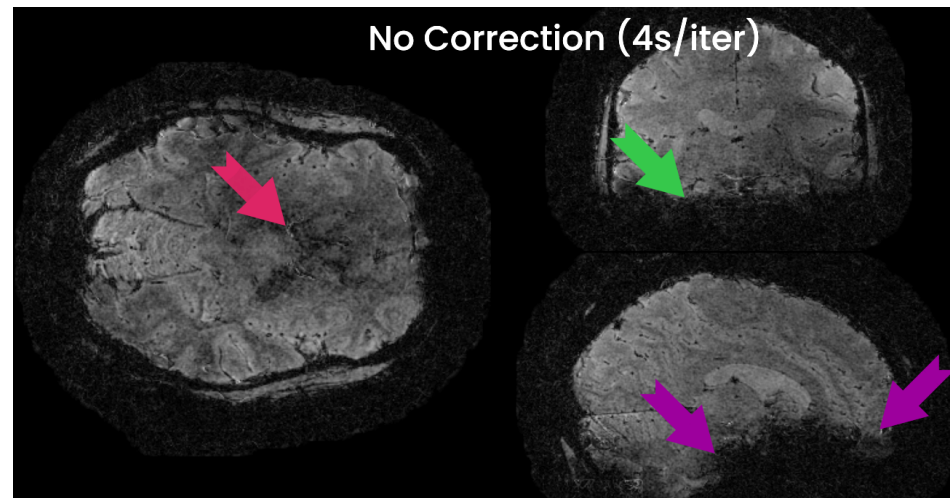
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- Density Compensation (voronoi, pipe, ...)
-



Extended Fourier Model

- Multi-coil, subspace representations, ...
- Density Compensation (voronoi, pipe, ...)
- Off-Resonance

$$\mathbf{y}_\ell(t) = \int_{FOV} \mathbf{x}(\mathbf{r}) \mathcal{S}_\ell(\mathbf{r}) e^{-\Delta\omega(\mathbf{r})t} e^{-2i\pi\mathbf{k}(t)\cdot\mathbf{r}} d\mathbf{r}$$

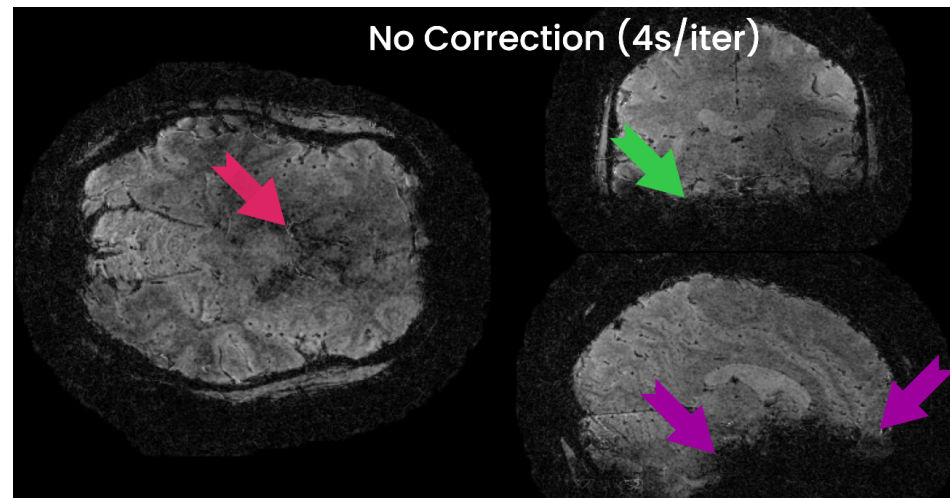


Extended Fourier Model

- Multi-coil, subspace representations, ...
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$$\mathbf{y}_\ell \approx \sum_{p=1}^P \mathbf{b}_p \sum_{\ell=1}^L \mathcal{F}_\Omega(\mathbf{c}_p \mathcal{S}_\ell \mathbf{x})$$

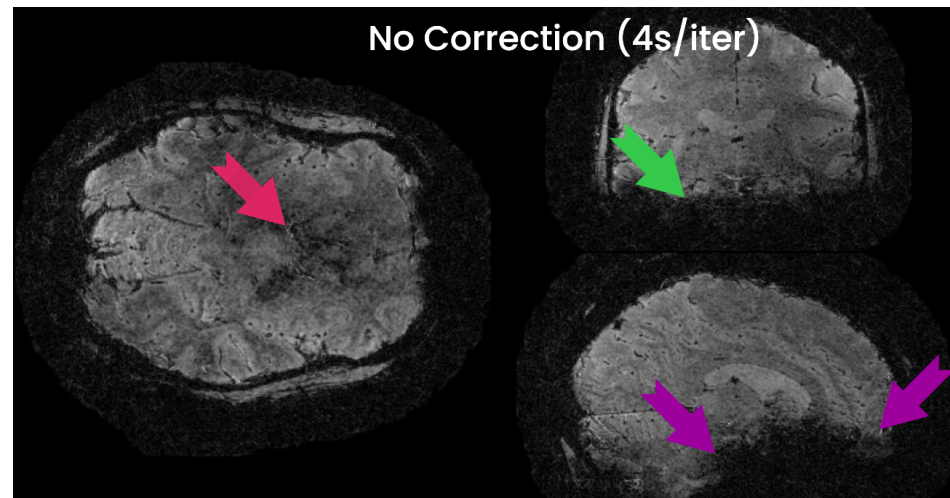


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$$\min_{\substack{\mathbf{B} \in \mathbb{C}^{M \times P} \\ \mathbf{C} \in \mathbb{C}^{P \times N}}} \|e^{-\Delta\omega(\mathbf{r})t} - \mathbf{BC}\|_2^2 \Rightarrow \mathbf{y}_\ell \approx \sum_{p=1}^P \mathbf{b}_p \sum_{\ell=1}^L \mathcal{F}_\Omega(\mathbf{c}_p \mathcal{S}_\ell \mathbf{x})$$



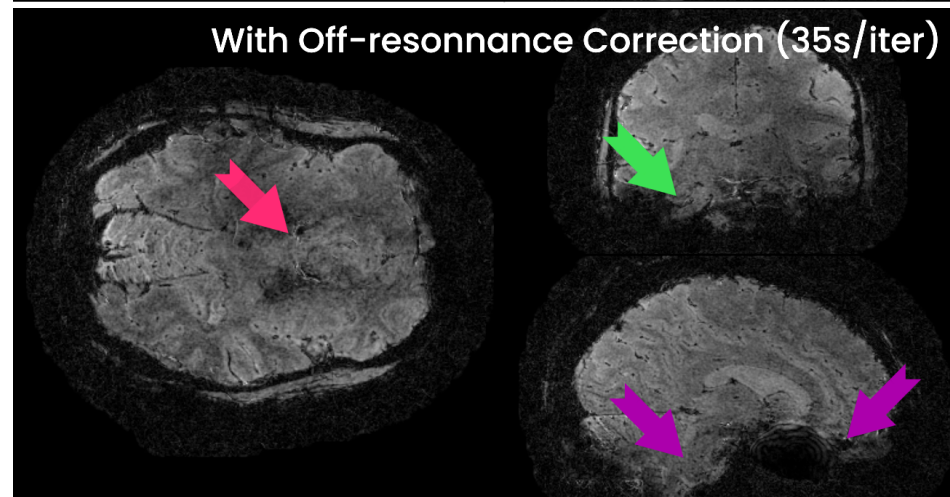
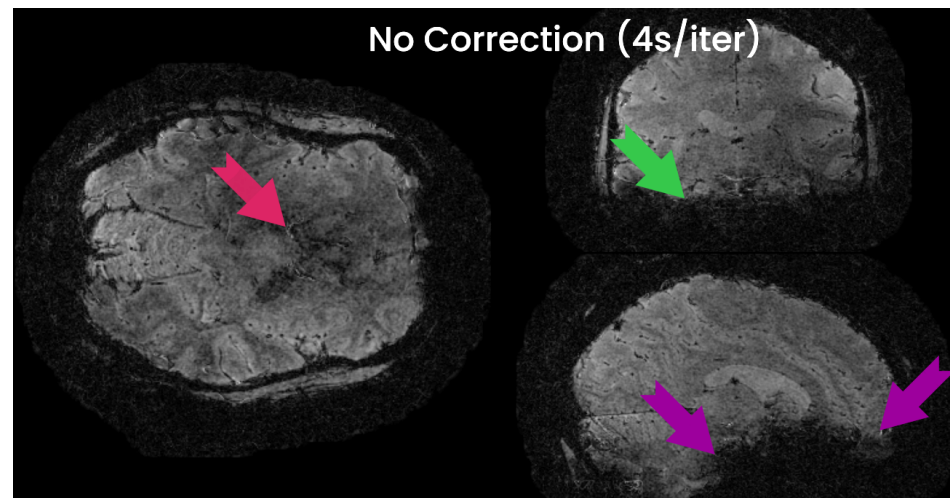
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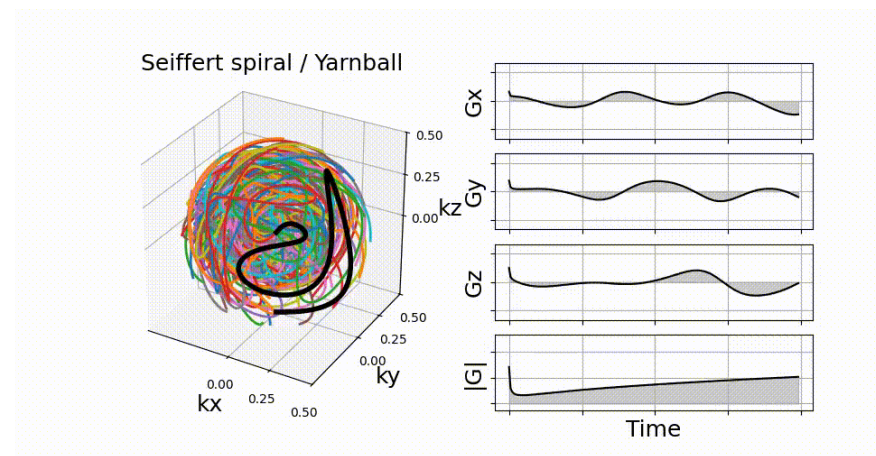
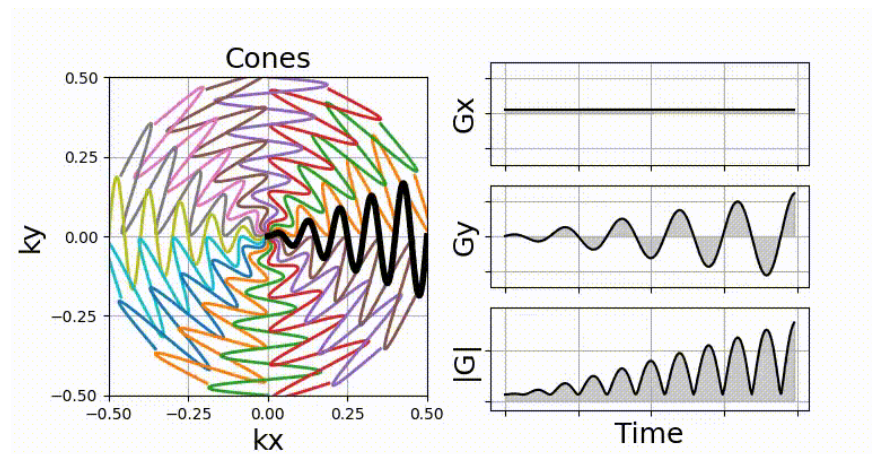
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```
nufft_orc = nufft.with_off_resonance_correction(  
    field_map=field_map,  
    readout_time=readout_time,  
    interpolators=20)  
adj_orc = nufft_orc.adj_op(ksp) #  $x = A_{\Delta\omega}^H y$   
pinv_orc = nufft_orc.pinv_solver(ksp) #  $x = A_{\Delta\omega}^\dagger y$ 
```



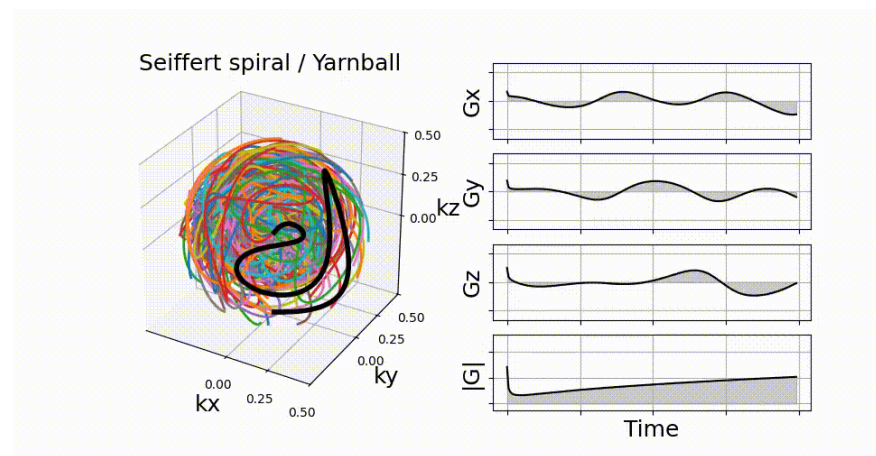
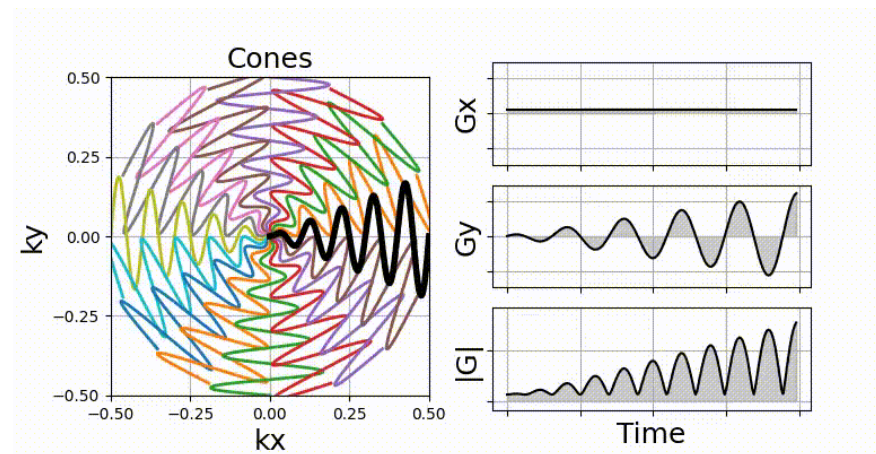
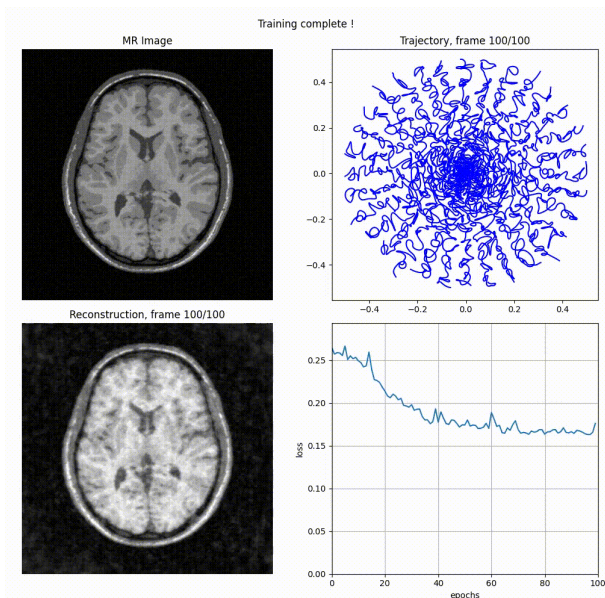
Trajectories & Gradients

- A large collection of non-Cartesian trajectories
 - 2D and 3D, fully parameterized



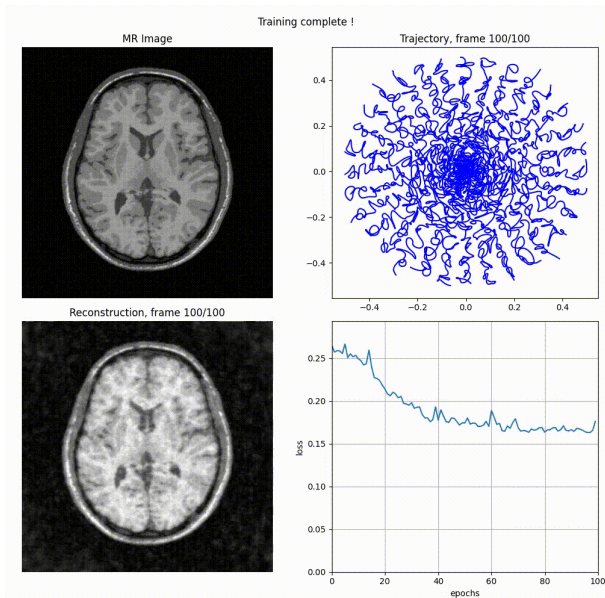
Trajectories & Gradients

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 - Hardware compliance enforcement.
 - You can even learn trajectories in a deep Learning setting with autodifferentiation !



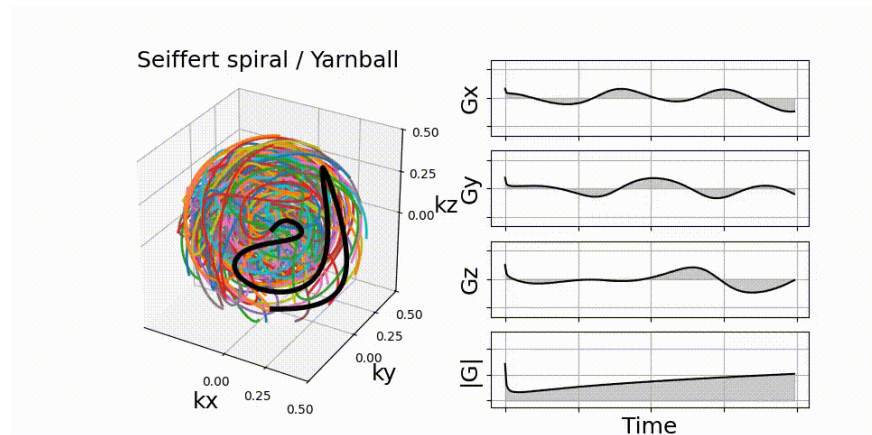
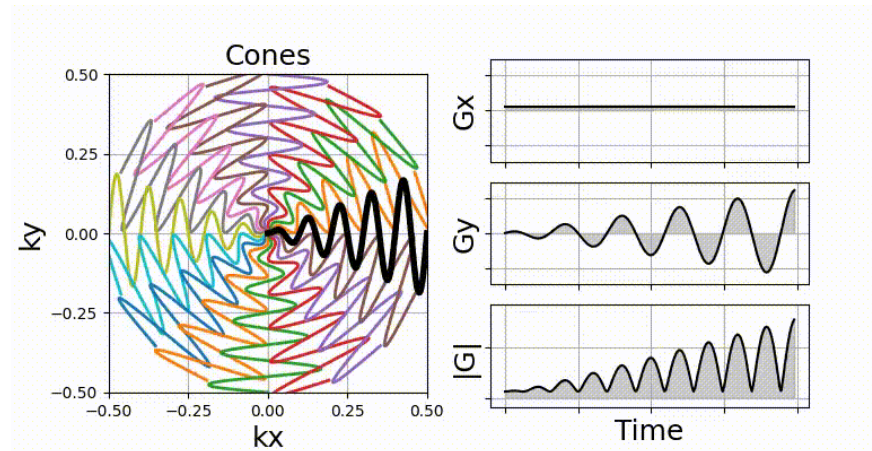
Trajectories & Gradients

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Live Demo !

[Open in Colab](#)



Integration with other libraries



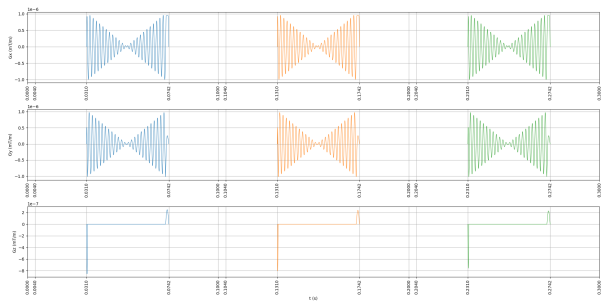
- ≥ 20 Projects depends on MRI-NUFFT¹
- Integration with external packages

¹<https://github.com/mind-inria/mri-nufft/network/dependents>

Integration with other libraries

- ≥ 20 Projects depends on MRI-NUFFT¹
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 - (Py)Pulseq

```
from mrinufft.trajectories import (pulseseq_gre,  
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traj = ...  
seq = pulseseq_gre(traj, acq=acq)  
seq.write("spiral_gre.seq")  
traj_read = read_pulseseq_traj(seq)
```

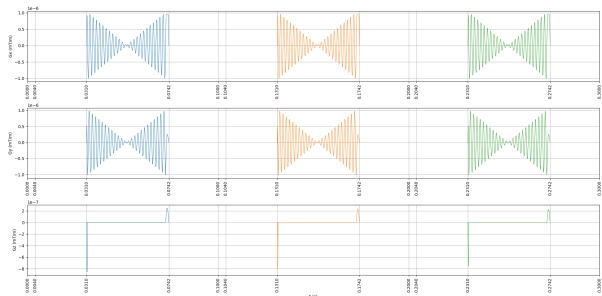


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Integration with other libraries

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Example for integration with deepinv

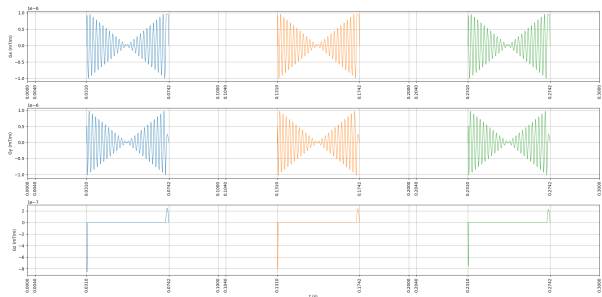
```
from mri_nufft import get_operator
from deepinv.optim import WaveletPrior, L2, optim_builder
# MRI-NUFFT setup
fourier_op = get_operator("cufinufft", samples_loc, shape=shape)
y = fourier_op.op(ground_truth) + noise #  $y = Ax + \epsilon$ 
```

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y = fourier_op.op(ground_truth) + noise #  $y = Ax + \epsilon$ 

# DeepInverse setup
physics = fourier_op.make_deepinv_phy()
wavelet = WaveletPrior(wv="sym8", is_complex=True, p=1) #  $\|\Psi \cdot\|_1$ 
wavelet_fista = optim_builder(
    iteration="FISTA",
    prior=wavelet, # swap with ANY prior !
    data_fidelity=L2(),
    params_algo={"lambda": 0.1}
)

# Reconstruction
x_pinv = physics.A_dagger(y) #  $\arg \min_x \|Ax - y\|_2^2$ 
x_wavelet = wavelet_fista(y, physics) #  $\arg \min_x \|Ax - y\|_2^2 + \lambda \|\Psi x\|_1$ 
```

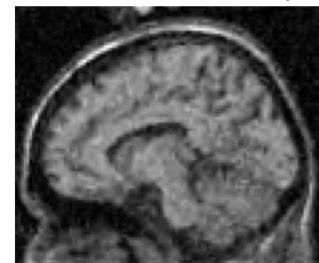
Ground truth



Adjoint reconstruction



Reconstruction with wavelet prior



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Packaging an Open Source Software



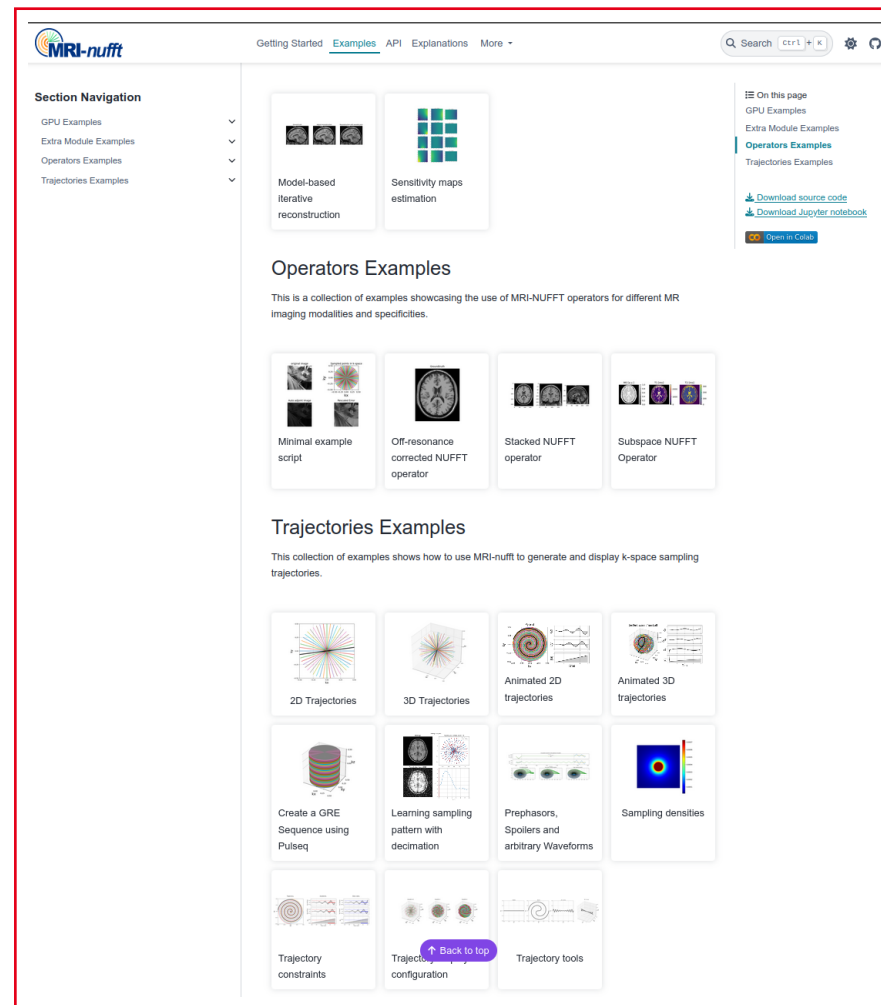
Documentation

- No documentation, no utilization



Examples, tutorials

- Onboarding new-comers  [Open in Colab](#)



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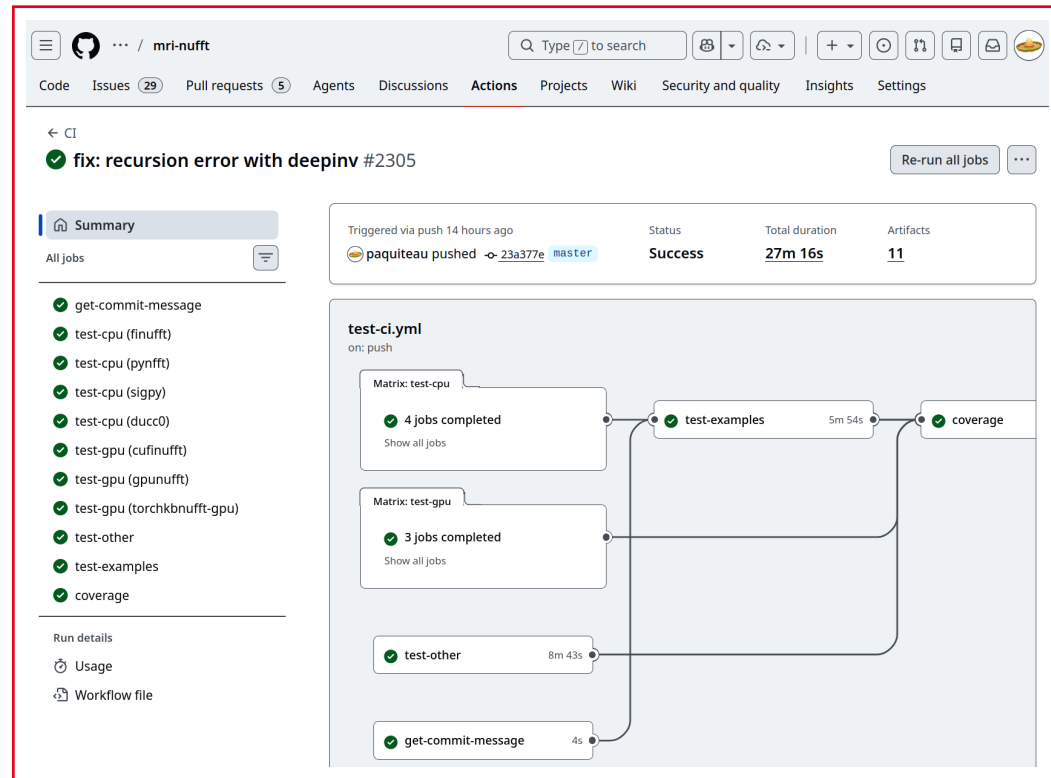


Unit Tests

- Catch bug if you can



Continuous integration and quality linters



The screenshot shows the GitHub Actions interface for the repository `mri-nufft`. The workflow `fix: recursion error with deepinv #2305` is displayed, triggered by a push to the `master` branch. The workflow status is **Success**, with a total duration of **27m 16s** and **11** artifacts. The workflow file `test-ci.yml` is shown, defining a matrix of tests for CPU and GPU, and a `test-other` job. The matrix jobs are `test-cpu (finufft)`, `test-cpu (pynfft)`, `test-cpu (sigpy)`, `test-cpu (ducc0)`, `test-gpu (cufinufft)`, `test-gpu (gpnufft)`, `test-gpu (torchknufft-gpu)`, `test-other`, `test-examples`, and `coverage`. The `test-examples` job is shown with a duration of 5m 54s. The `test-other` job is shown with a duration of 8m 43s. The `get-commit-message` job is shown with a duration of 4s.

Workflow file: `test-ci.yml`

```
on: push

matrix:
  test-cpu:
    - finufft
    - pynfft
    - sigpy
    - ducc0
  test-gpu:
    - cufinufft
    - gpnufft
    - torchknufft-gpu
  test-other:
    - test-other

jobs:
  get-commit-message:
    runs-on: ubuntu-latest
    steps:
      - run: get-commit-message

  test-cpu:
    runs-on: ubuntu-latest
    steps:
      - run: test-cpu (${{matrix.test-cpu}})

  test-gpu:
    runs-on: ubuntu-latest
    steps:
      - run: test-gpu (${{matrix.test-gpu}})

  test-other:
    runs-on: ubuntu-latest
    steps:
      - run: test-other

  test-examples:
    runs-on: ubuntu-latest
    steps:
      - run: test-examples

  coverage:
    runs-on: ubuntu-latest
    steps:
      - run: coverage
```

Packaging an Open Source Software



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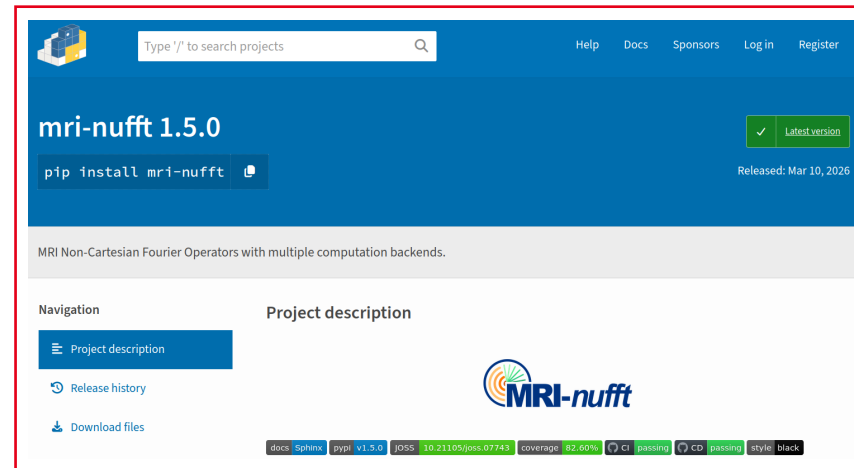
Continuous integration and quality linters

- ruff, black, ty, pyright, pyrefly, ...



Python Package:

- `pip install mri-nufft`



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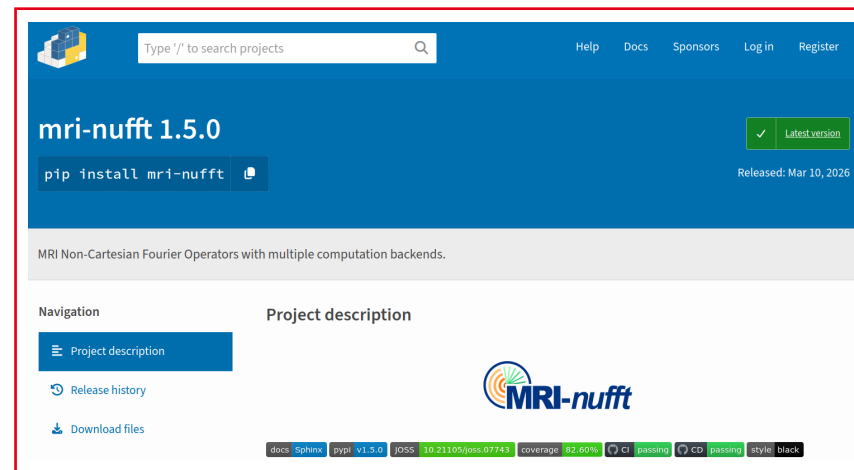


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Collaborators (and stickers !)





3. Beyond Toolboxes and Packages

Open data for Reproducibility



Dataset for training and validation of reconstruction methods

FastMRI

- Brain, Knee, Prostate, Breast
- stack of 2D images
- <https://fastmri.med.nyu.edu/>

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- Cardiac Imaging 2D+T, 3D+T
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Open Format and API

- BIDS: Dataset file structure for MRI (and Neuro Imaging)
 - No k-space standard (yet ?)
- ISMRMRD: Standard raw k-space data format (Instead of .dat, .raw, "P-files", ...)

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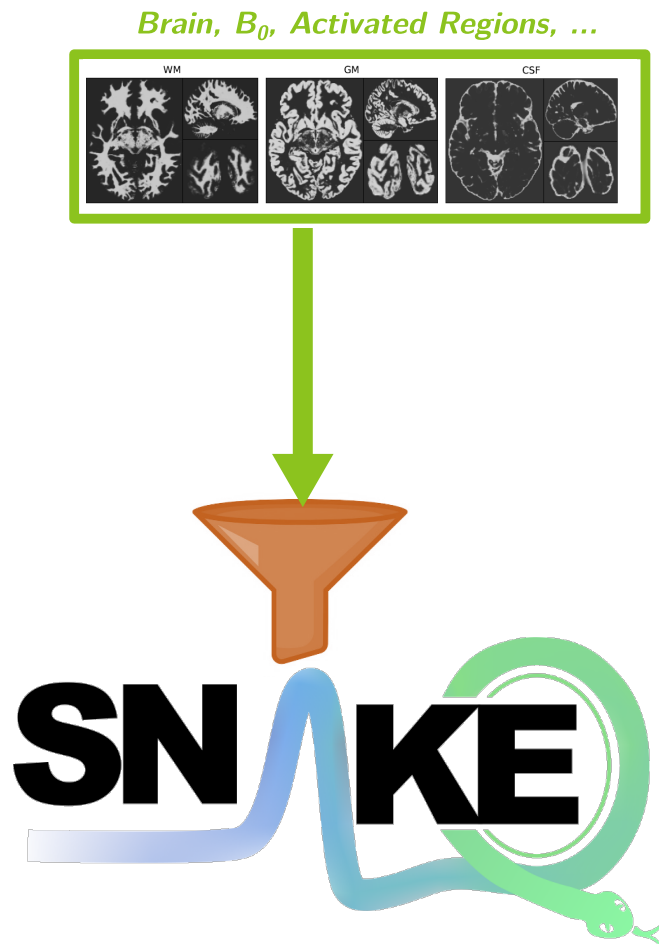
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No dataset or Ground-Truth available ?
fMRI, space/time resolution tradeoff

SNAKE fMRI k-space Simulator



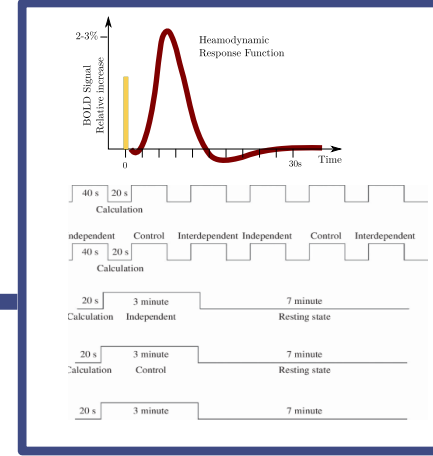
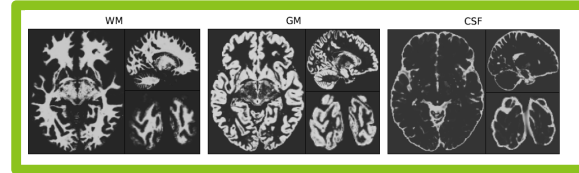
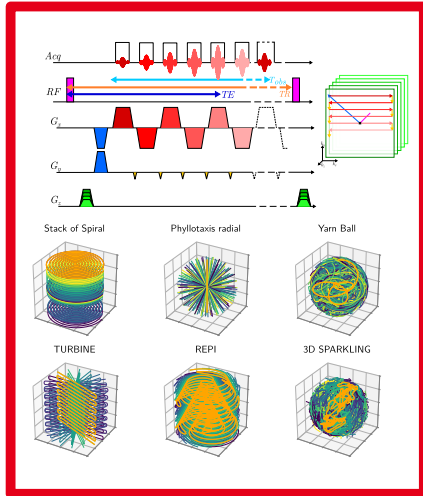


SNAKE fMRI k-space Simulator

Brain, B_0 , Activated Regions, ...

Paradigm Design & HRF model

Acquisition Setting

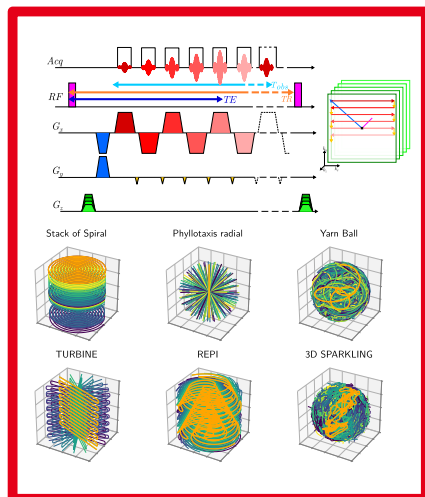
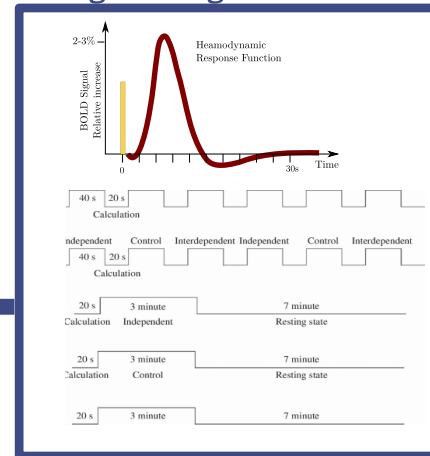
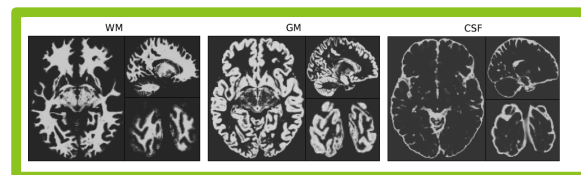


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SN  **KE**

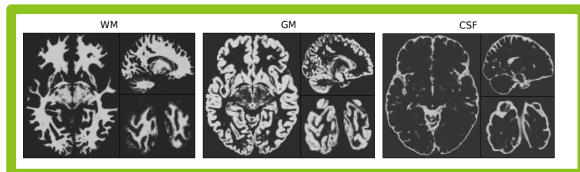
RECONSTRUCTION

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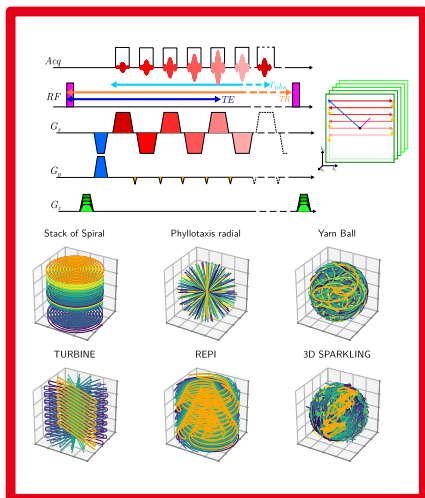
Learn more

Comby et al., Imaging Neuroscience, 2025.

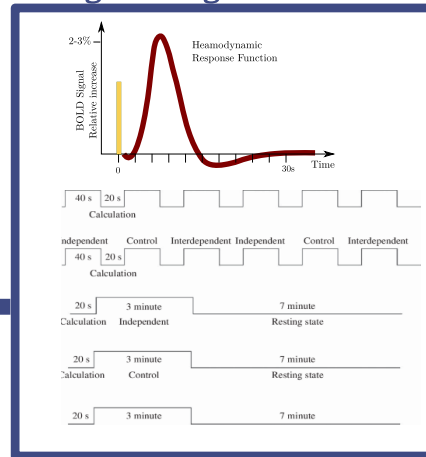
Brain, B_0 , Activated Regions, ...



Acquisition Setting



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SNKE

RECONSTRUCTION

STATISTICAL
ANALYSIS
VALIDATION

A word about benchmarks



MRM Editorial



EDITORIAL |  Free Access

The “State of the Art” in MR Image Reconstruction? Knowledge, Culture, and What We Leave Behind in An Era of Big Data and Machine Learning

[Justin P. Haldar](#) 

First published: 10 April 2026 | <https://doi.org/10.1002/mrm.70377> | [VIEW METRICS](#)

- Is PSNR and SSIM really what we should optimize for ?
- Leaderboards vs clinical impact

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State of the art $\neq$ State of the practice

- Make the toolbox usable by non-expert of the k-space.

Community initiatives



Workshop on MRI Acquisition and Reconstruction in Sedona, AZ.

Community initiatives



Workshop on MRI Acquisition and Reconstruction in Sedona, AZ.



Reproducible research group

- MRM Code Review
- Join the mailing list !



Group Meeting

Sunday 9:45-11:45
Ballroom East

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Commits to h-index convertors:

- Journal of Open-Source Software (JOSS)
- Zenodo
- MRM Toolbox Articles (be the first ?)



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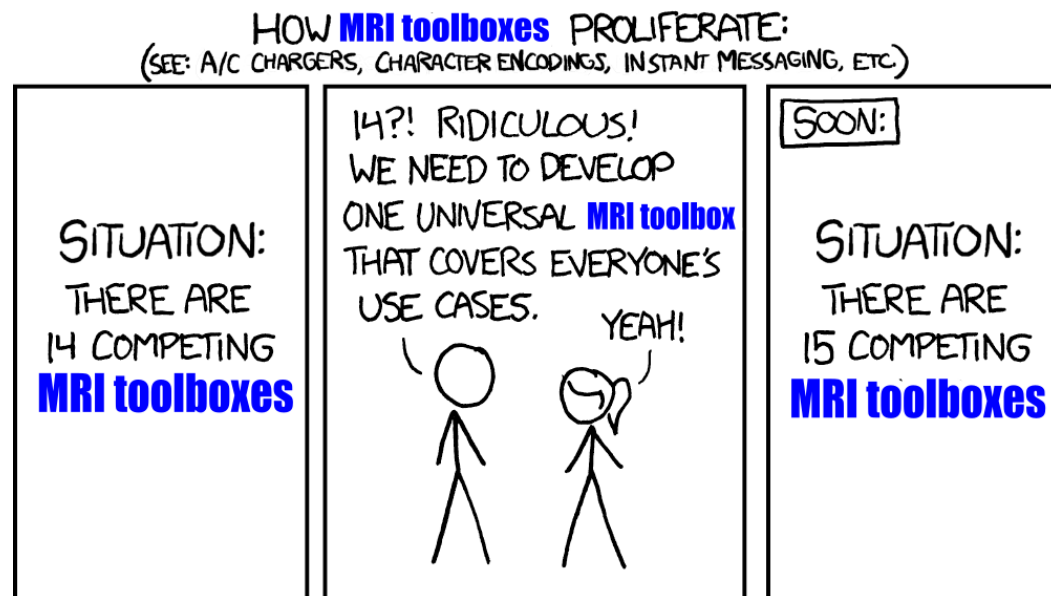
Take home messages

- MRI Reconstruction is a challenging but *fun* problem
 - Interface between Math, Physics and Computer Science

¹MRI-NUFFT + DeepInv goes a long way: https://mind-inria.github.io/mri-nufft/generated/autoexamples/extras/example_recon.html

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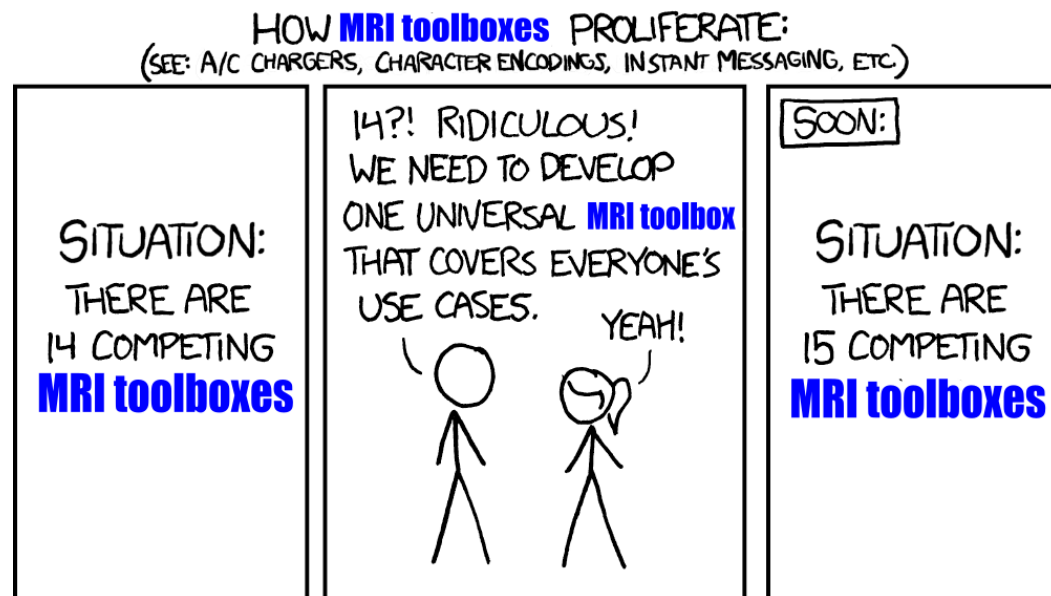
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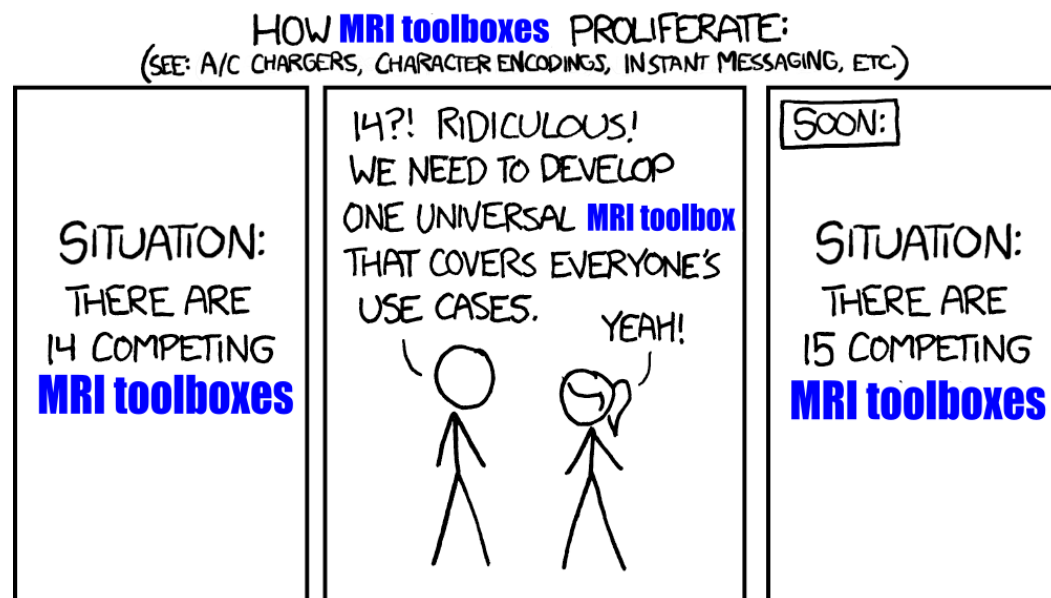
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 - don't forget the instructions to run it.
 - NO HARDCODED ABSOLUTE FILEPATH!



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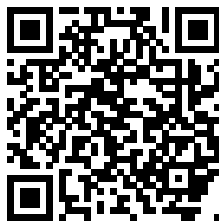
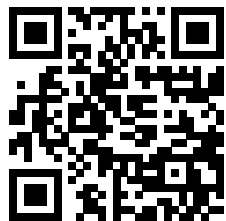
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Check out Our awesome **open-source** software:



Thank you for your attention

pierre-antoine.comby@cea.fr

Neurospin, CEA, Paris-Saclay University

