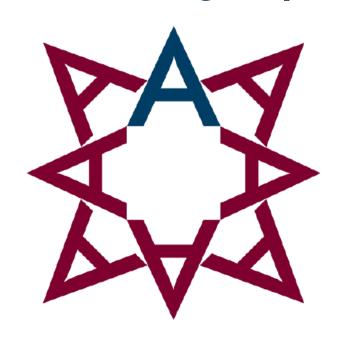




The ASTRA tomography toolbox 3D tomography



Willem Jan Palenstijn, Tim Elberfeld





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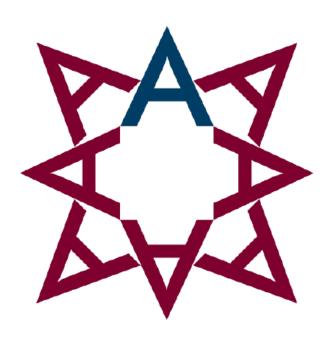
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About us



Tim Elberfeld

Doctoral Researcher imec-Vision Lab
Dept. Physics, University of Antwerp

Developing reconstruction algorithms with model priors



Scientific Software Developer Centrum Wiskunde & Informatica (CWI), Amsterdam



Lead developer on ASTRA Toolbox

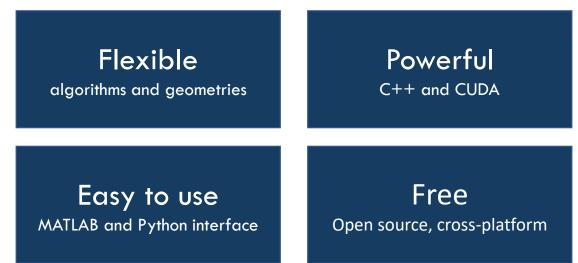




Introduction to ASTRA

What is ASTRA?

ASTRA provides fast and flexible building blocks for 2D/3D tomographic reconstruction, aimed at algorithm developers and researchers.







History

The ASTRA toolbox was started at the Vision Lab of the University of Antwerp in Belgium by PhD students and post-docs



Initial goal: reduced implementation work for internal PhD projects

Later on: interest from external labs and companies

▶ E.g. ESRF (France) and FEI (The Netherlands, now ThermoFisher)

First open source release in August 2012

Active development continues

Now jointly by Vision Lab and CWI

2.0 Release in 2019





Features

Geometries:

- ▶ 2D parallel and fan beam
- ▶ 3D parallel and cone beam
- All with fully flexible source/detector positioning

Basic reconstruction algorithms:

- ▶ FBP, FDK reconstruction
- Iterative SIRT, CGLS reconstruction

Primitives for building your own algorithms:

▶ FP, BP





Non-Features

Data I/O

Data alignment, preprocessing

Matched GPU FP, BP operators

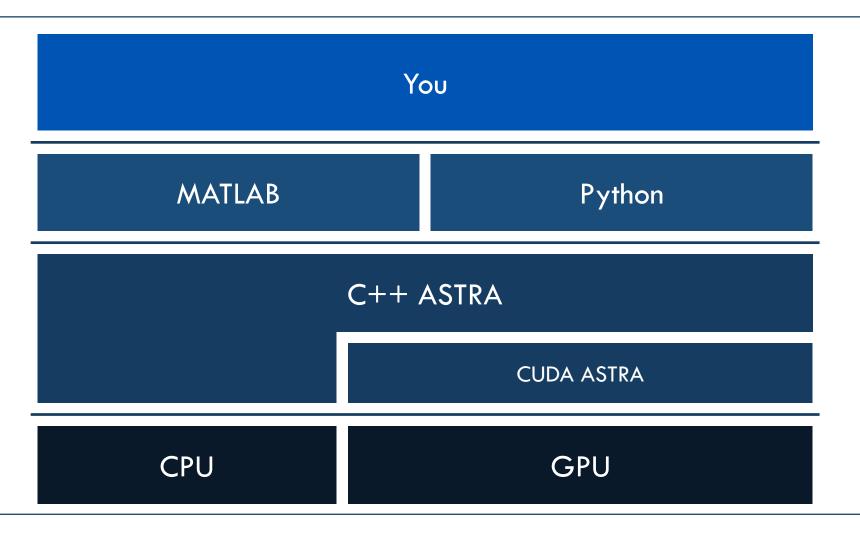
▶ CPU FP, BP are matched; GPU are not

Modern advanced reconstruction algorithms





ASTRA Architecture

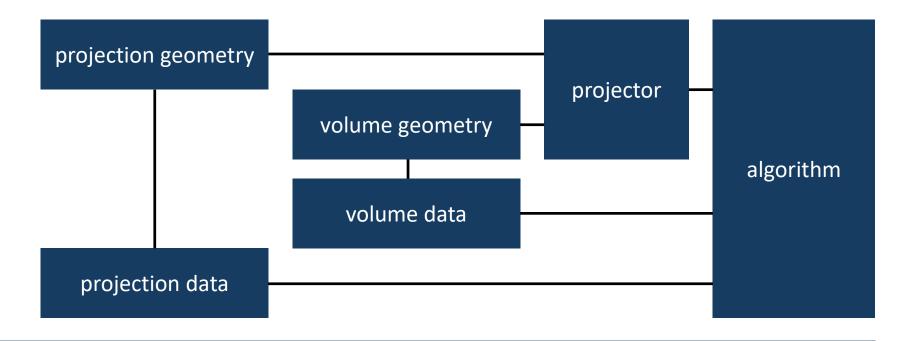






Modeling of the x-ray scanning setup

- Phantom / volume
- ▶ Source
- ▶ Detector

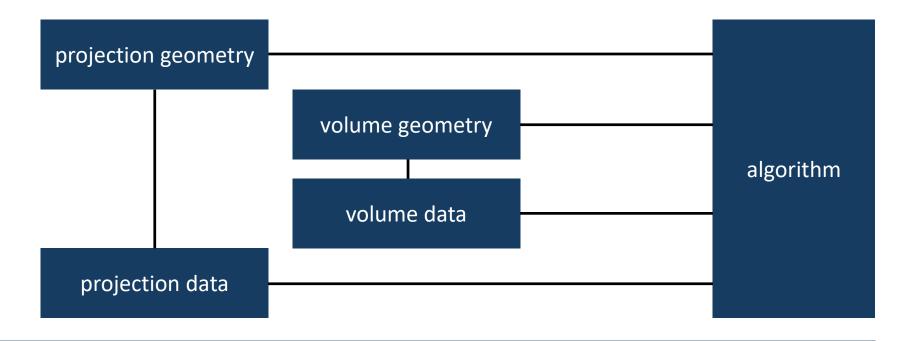






Modeling of the x-ray scanning setup

- Phantom / volume
- ▶ Source
- ▶ Detector





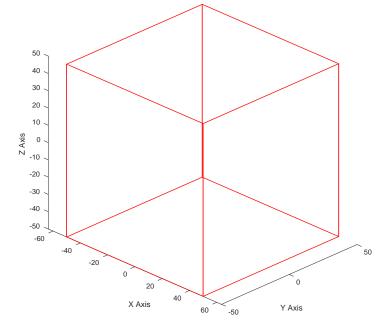


Volume geometry and volume data

vol_geom = astra.create_vol_geom(ny, nx, nz)

limitation:

- all lengths and sizes defined relative to voxel size
- voxel size defaults to 1 unit (cube voxels)







Volume geometry and volume data

rec_id = astra.data3d.create(`-vol`, vol_geom)

Allocates float32 array in C++ interface

ID for reference in Python

Storage for volume data (e.g. the reconstruction)

Needs volume geometry as initializer, is tied to it





Volume data

```
rec_id = astra.data3d.create('-vol', vol_geom)

rec_id = astra.data3d.create('-vol', vol_geom, 0)
astra.data3d.store(rec_id, 0)

rec_id = astra.data3d.create('-vol', vol_geom, V)
astra.data3d.store(rec_id, V)

V = astra.data3d.get(rec_id)

astra.data3d.delete(rec_id)
```





Projection geometry and data

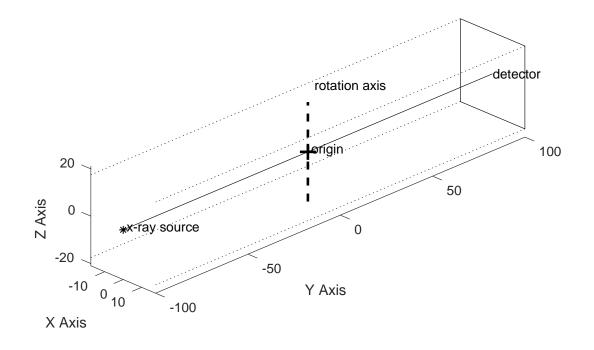
- trajectory of source and detector plane
- number of detector elements
- ▶ detector size

- 1. parallel beam
- 2. cone beam





parallel projection geometry

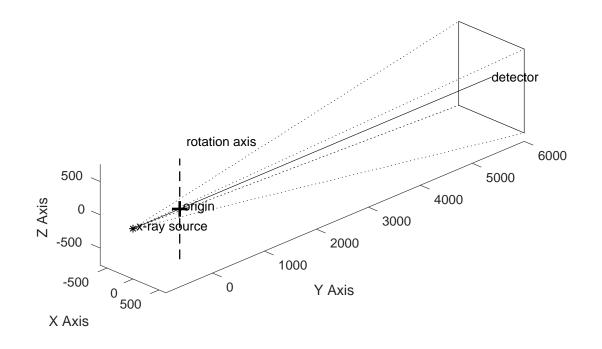


```
angles = np.linspace(0, np.pi, 180, False)
proj_geom = astra.create_proj_geom('parallel3d', det_width, det_height, det_rows, det_cols, angles)
```





cone beam projection geometry

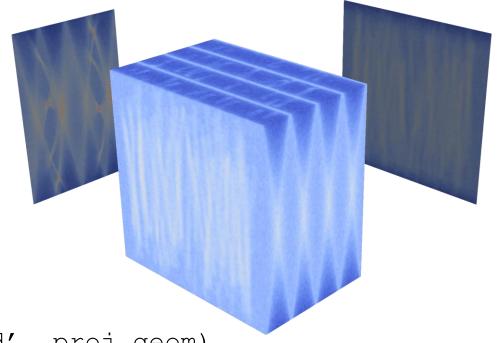






projection data

- place to store projections
- ▶ similar to volume data
- b tied to projection geometry



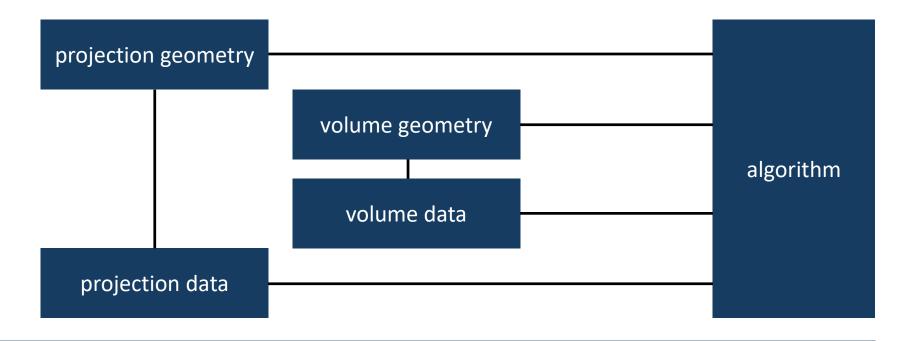
```
proj_id = astra.data3d.create('-proj3d', proj_geom)
proj_id = astra.data3d.create('-proj3d', proj_geom, 0)
proj id = astra.data3d.create('-proj3d', proj geom, V)
```





Modeling of the x-ray scanning setup

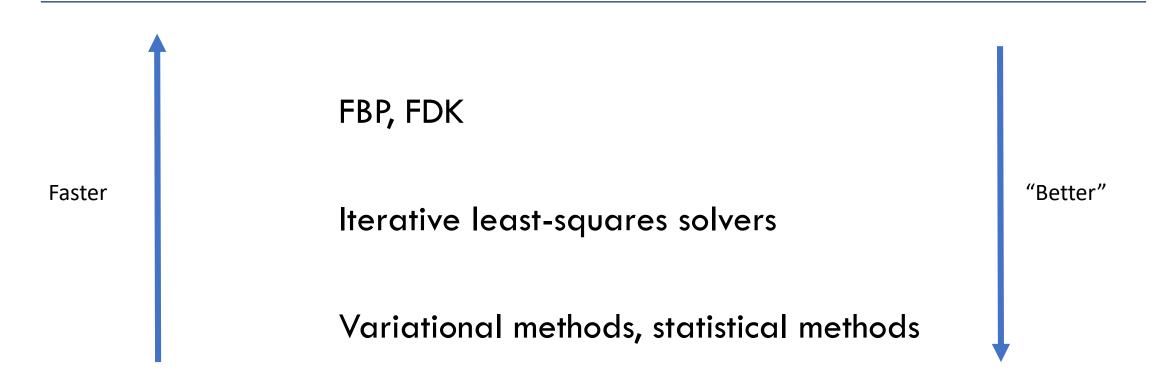
- Phantom / volume
- ▶ Source
- ▶ Detector







Reconstruction Algorithms







Reconstruction Algorithms

ASTRA provides

3D GPU

- FP3D_CUDA, BP3D_CUDA
- FDK_CUDA
- SIRT3D_CUDA, CGLS3D_CUDA





Reconstruction - SIRT

```
# Configure geometry
angles = np.linspace(0, 2*np.pi, nAngles, False)
proj geom = astra.create proj geom('cone', 1.0, 1.0,
                detectorRowCount, detectorColCount, angles,
                originToSource, originToDetector)
vol geom = astra.create vol geom(vy, vx, vz)
# Data objects for input, output
proj id = astra.data3d.create('-proj3d', proj geom, P)
rec Id = astra.data3d.create('-vol', vol geom)
# Configure algorithm
cfg = astra.astra dict('SIRT3D CUDA')
cfg['ReconstructionDataId'] = rec id
cfg['ProjectionDataId'] = proj id
alg id = astra.algorithm.create(cfg)
# Run
astra.algorithm.run(alg id, 100)
rec = astra.data3d.get(rec id)
```





SciPy linear operator

```
vol geom = astra.create vol geom(256, 256, 256)
proj geom = astra.create proj geom('parallel3d', 1.0, 1.0, 256, 256,
                                   np.linspace2(0, np.pi, 180, False))
proj id = astra.create projector('cuda3d', proj geom, vol geom)
# Create OpTomo operator
W = astra.OpTomo(proj id)
# Forward projection
s = W * P
s = s.reshape(astra.geom size(proj geom))
# Reconstruction using scipy's lsqr
output = scipy.sparse.linal.lsqr(W, s.ravel(), iter lim=100)
rec = output[0].reshape(astra.geom size(vol geom))
```



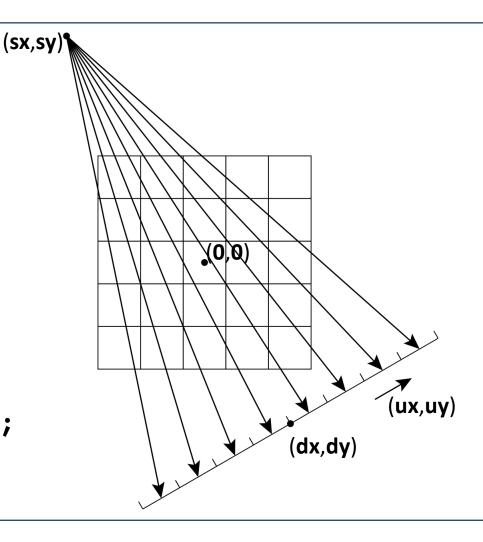


Fan beam – vector form

Three 2D parameters per projection: s, d, u

These form a 6 element row vector.

Recall: sample is stationary; source and detector move







Cone beam - vector form

```
# Cone - vector form
proj_geom = astra.create_proj_geom('cone_vec', 256, 256, vectors)
```

vectors consists of a 12 element row vector per projection.

Four 3D parameters per projection:

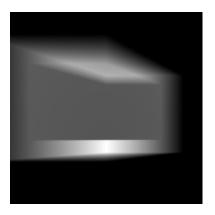
- ▶ s: source location
- ▶ **d**: detector center
- bu: horizontal detector pixel basis vector
- v: vertical detector pixel basis vector

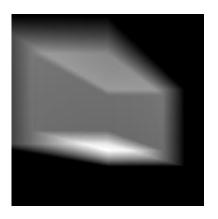




Cone beam - vector form

```
vectors = [[ 192, 30, 192, 0, 0, 0, 1, 0, 0, 0, 1, 0 ], [ 192, 90, 192, 0, 0, 0, 1, 0, 0, 1, 0 ]]
```









Parallel beam - vector form

```
# Parallel beam - vector form
proj_geom = astra.create_proj_geom('parallel3d_vec', 256, 256, vectors)
vectors consists of a 12 element row vector per
projection.
```

Four 3D parameters per projection:

- r: Ray direction
- ▶ **d**: detector center
- bu: horizontal detector pixel basis vector
- v: vertical detector pixel basis vector





Conversion to vector form

Using utility function:

```
# Standard cone beam
proj_geom = astra.create_proj_geom('cone', 1.0, 1.0, 256, 256, angles,
2000, 0)

# Convert to vector form
proj_geom_vec = astra.geom_2vec(proj_geom)

# Center-of-rotation correction (by -3.5 pixels horizontally)
proj geom cor = astra.geom postalignment(proj geom, [-3.5, 0])
```





Using ASTRA via ODL

"ODL is a python library for fast prototyping focusing on (but not restricted to) inverse problems.

The main intent of ODL is to enable mathematicians and applied scientists to use different numerical methods on real-world problems without having to implement all necessary parts from the bottom up. ODL provides some of the most heavily used building blocks for numerical algorithms out of the box, which enables users to focus on real scientific issues."

Developed by KTH Stockholm





ASTRA for large data sets

Historically, a major limitation of ASTRA has been the requirement that data fits in GPU memory.

With ASTRA 1.8, we've started removing this limit.

Current status:

▶ 3D FP, BP and FDK transparently handle this; limited by host memory.

Other reconstruction algorithms don't directly





ASTRA for large data sets

For FP, BP, FDK this works transparently:

```
proj_geom = astra.create_proj_geom('parallel3d', 1.0, 1.0, 1024, 1024, angles)
vol_geom = astra.create_vol_geom(1024, 1024, 1024)
id, projdata = astra.create_sino3d_cuda(voldata, proj_geom, vol_geom)
id, voldata = astra.create_backprojection3d_cuda(projdata, proj_geom, vol_geom)
```

Also works transparently with OpTomo operator

To help with performance, you can also use multiple GPUs:

astra.set_gpu_index([0, 1])





ASTRA for large data sets – Titan RTX

2 D	FP per slice (ms)	Full volume FP (s)	BP per slice (ms)	Full volume BP (s)
512	2.43	1.24	2.75	1.41
1024	6.80	6.96	7.59	7.72
2048	28.8	58.9	42.0	86.0
4096	194	795	300	1229

3D	Full volume FP (s)	Full volume BP (s)
512	1.00	0.69
1024	13.3	7.75
2048	179	100
4096	1736	1460

NxNxN volume, N projections of NxN, parallel beam





ASTRA for large data sets

Reducing memory usage:

- Use dtype=np.float32 to use single-precision floats instead of double
- Share memory between numpy and ASTRA:
 - ▶astra.data3d.get shared(id)
 - astra.data3d.link (instead of astra.data3d.create)





ASTRA for large data sets

SIRT3D_CUDA and CGLS3D_CUDA require all data (plus temporaries) to fit in GPU memory.

As a workaround use (provided) SIRT, CGLS algorithms that operate in host memory:

```
astra.plugin.register(astra.plugins.SIRTPlugin) astra.plugin.register(astra.plugins.CGLSPlugin)
```

Then use algorithms SIRT-PLUGIN, CGLS-PLUGIN Also see sample script s018_plugin.py





Major new feature in ASTRA 1.7 (Dec 2015), in separate mpi branch:

Distributed ASTRA: run ASTRA on a (GPU) cluster

- Process larger data sets faster
- A toolbox of building blocks for your (and our) own algorithms
- ▶ (Most of) the flexibility of ASTRA
- Python interface (only)

Developed by Jeroen Bédorf at CWI.





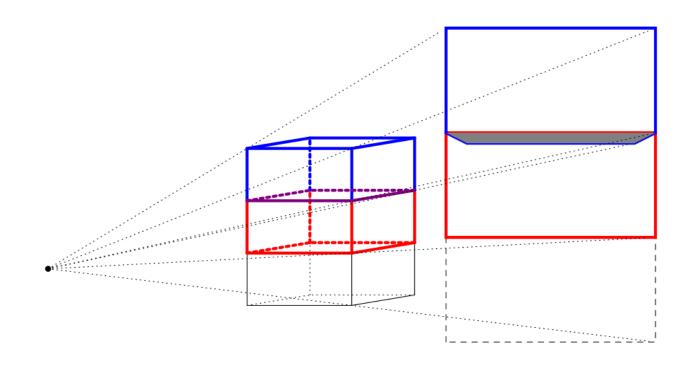
Our approach:

Data objects are distributed, and (some) 3D algorithms will automatically run distributed:

- ▶ SIRT
- ▶ CGLS
- Forward projection
- Backprojection











(Un)distributed ASTRA

```
# Configure geometry
angles = np.linspace(0, 2*np.pi, nAngles, False)
proj geom = astra.create proj geom ('cone', 1.0, 1.0, detectorRowCount, detectorColCount,
                angles, originToSource, originToDetector)
vol geom = astra.create vol geom(vx, vy, vz)
# Data objects for input, output
proj id = astra.data3d.create('-proj3d', proj geom, P)
rec id = astra.data3d.create('-vol', vol geom)
# Configure algorithm
cfg = astra.astra dict('SIRT3D CUDA')
cfg['ReconstructionDataId'] = rec id
cfg['ProjectionDataId'] = proj id
alg id = astra.algorithm.create(cfg)
# Run
astra.algorithm.run(alg id, 100)
rec = astra.data3d.get(rec id)
```





```
# Configure geometry
angles = np. [inspace (0, 2*np.pi, nAngles, False)
proj_geom = astra.create proj_geom('cone', 1.0, 1.0, detectorRowCount, detectorColCount, angles, originToSource, originToDetector)
vol geom = astra.create vol geom(vx, vy, vz)
# Set up the distribution of data objects
proj geom, vol geom = mpi.create(proj_geom, vol_geom)
# Data objects for input, output
proj id = astra.data3d.create('-proj3d', proj geom, P)
rec id = astra.data3d.create('-vol', vol geom)
# Configure algorithm
cfg = astra.astra dict('SIRT3D CUDA')
cfg['ReconstructionDataId'] = \overline{rec} id
cfg['ProjectionDataId'] = proj id
alg id = astra.algorithm.create(cfg)
# Run
astra.algorithm.run(alg id, 100)
rec = astra.data3d.get(rec id)
```





The mpi branch of 1.7 is very experimental and a bit out of date.

Currently developing new distributed version, with more efficient partitioning/distribution of data, with Jan-Willem Buurlage (CWI)





Useful links

Webpage, for downloads, docs:

https://www.astra-toolbox.com/

▶ Github, for source code, issue tracker:

https://github.com/astra-toolbox/

▶ Email:

astra@astra-toolbox.com

willem.jan.palenstijn@cwi.nl tim.elberfeld@uantwerpen.be





Publications

Palenstijn et al, "Performance improvements for iterative electron tomography reconstruction using graphics processing units (GPUs)", J. of Structural Biology, 2011

van Aarle, Palenstijn et al, "The ASTRA Toolbox: A platform for advanced algorithm development in electron tomography", **Ultramicroscopy**, 2015

van Aarle, Palenstijn et al, "Fast and flexible X-ray tomography using the ASTRA Toolbox", Optics Express, 2016

Palenstijn, Bédorf et al, "A distributed ASTRA Toolbox", Adv. Struct. & Chem. Imag. 2016





Code example





Now: Q&A and exercises

Installation instructions and exercises:

https://www.astra-toolbox.com/ictms.html

Also, feel free to ask us anything