# Learned Exascale Computational Imaging (LEXCI) overview

UCL ExCALIBUR Meetup

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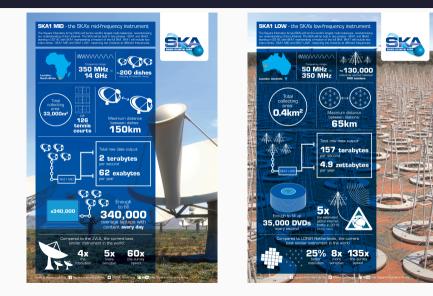
April 2022

## Canonical application: Square Kilometre Array (SKA)



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### SKA sites



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## Next-generation of radio interferometry rapidly approaching

**Next-generation of radio interferometric telescopes** will provide orders of magnitude improvement in sensitivity and resolution.

Unlock broad range of science goals.



Dark energy

General relativity





Epoch of reionization

Exoplanets

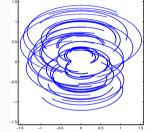
### SKA poses a considerable exascale computational imaging challenge



## Radio interferometric telescopes acquire "Fourier" measurements



"Fourier" Measurements ⇒



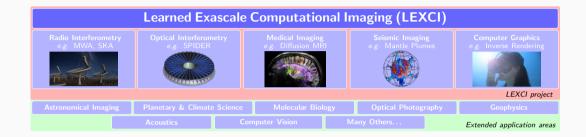
### Radio interferometric telescopes acquire "Fourier" measurements



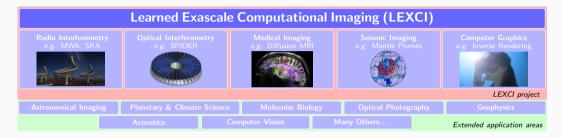
#### Interferometric imaging is an exascale computational inverse imaging problem:

Recover an image from noisy and incomplete "Fourier" measurements.

## LEXCI application domains more broadly



## LEXCI application domains more broadly



#### Partners

- Radio interferometry: Prof. Melanie Johnston-Hollitt (Curtin), Dr Luke Pratley (Toronto)
- SPIDER: Prof. Ben Yoo (UC Davis)
- Medical Imaging: Prof. Gary Zhang (CMIC, UCL)
- Seismic Imaging: Prof. Ana Ferreira (Earth Sciences, UCL)
- Computer Graphics & Virtual Reality: Kagenova
- (ExCALIBUR Benchmarking for AI for Science at Exascale; BASE)

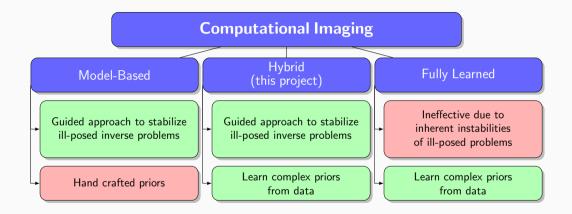
Classically, inverse imaging problems solved by **variational regularization**, where an optimization problem is posed that includes data fidelity and regularization terms:

$$\underset{\mathbf{x}}{\arg\min} \|\mathbf{y} - \mathbf{\Phi}\mathbf{x}\|_{2}^{2} + \lambda f(\mathbf{x}).$$

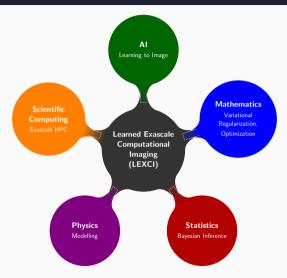
for observational model  $\Phi : \mathbb{R}^N \to \mathbb{R}^M$ , data y and underlying image x.

**Regularization functional**  $f : \mathbb{R}^N \to \mathbb{R}$  encodes prior knowledge.

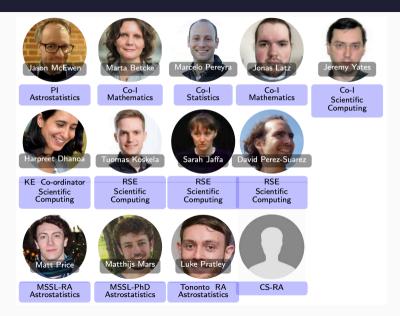
Typically **model-based regularizers** are used, *e.g.*  $f(\mathbf{x}) = \|\mathbf{\Psi}^{\dagger}\mathbf{x}\|_{1}$  to promote sparsity in some dictionary  $\mathbf{\Psi} : \mathbb{R}^{D} \to \mathbb{R}^{N}$ .



### Cross-cutting research areas



#### LEXCI team



Jason McEwen

### Methodological developments

- ▷ Hybrid deep learning & model-based approach
- ▷ Learned image model
- ▷ Learned instrument model
- ▷ Learned convex models to support uncertainty quantification
- ▷ Geometric imaging (*e.g.* spherical)

## Computing paradigms

- ▷ Data partitioning algorithms
- ▷ Distributed compute, storage & memory
- ▷ Stochastic distributed algorithms
- ▷ Parallelized & distributed uncertainty quantification
- ▷ Exploit mixed-precision arithmetic

### **ExCALIBUR** use cases

- ▷ Learned computational imaging
- ▷ Efficient data IO & workflows
- $\triangleright$  Visualization
- ▷ Mixed-precision arithmetic
- ▷ Fault tolerance
- ▷ Uncertainty quantification

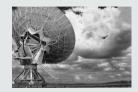
- ▷ Novel Hardware/Software Architecture Testbed (University of Birmingham, NextSilicon, Lenovo)
- ▷ Cerebras Testbed (EPCC, Cerebras, HPE)
- ▷ UCL Adaptable Cluster Testbed (UCL, Mellanox)
- ▷ FPGA Testbed (EPCC, UCL, University of Warwick, Xilinx)

▷ Traditional conference: *Computational Inverse Imaging* 

▷ Unconference: Applying LEXCI software to cross-cutting problems across domains

### Public open-source codes

#### PURIFY code



#### https://github.com/astro-informatics/purify

### Next-generation radio interferometric imaging

PURIFY is a highly distributed and parallelized open-source C++ code for radio interferometric imaging, leveraging recent developments in the field of variational regularization and convex optimisation.

#### SOPT code

https://github.com/astro-informatics/sopt



### **Sparse OPTimisation**

SOPT is a highly distributed and parallelized open-source C++ code for variational regularization and convex optimisation.