

The blessing and the curse of the big image

Carola Schönlieb, Jan Lellmann
Cambridge Image Analysis

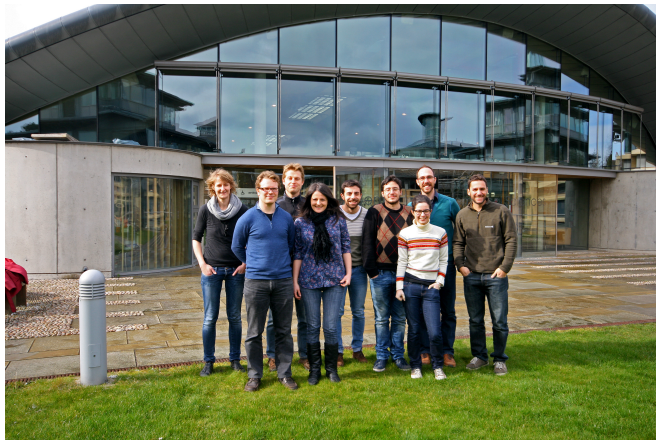
Department for Applied Mathematics and Theoretical Physics
University of Cambridge, UK



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1 PI, 3 PostDocs, 6 PhD students, 3 Masters students

www.damtp.cam.ac.uk/research/cia/

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The blessing

Images are *big*

HD video	6 MB/image uncompressed 180MB/second
Digital cameras	80-200 million pixels 480-1,200MB/image
MRI	16 MB/volume at 1mm ³ 590MB/volume at 0.3mm ³

- High number of dimensions:
 - 2D–3D astronomical imaging, global terrestrial seismic tomography, arts applications ([Fitzwilliam Museum](#))
 - 3D dynamic microscopy ([CAIC](#))
 - 3D–4D (dynamic) MRI ([MRRC](#)), PET/SPECT ([WBIC](#))
 - 5D Diffusion MRI



Image credit: Wikimedia/Selbmay

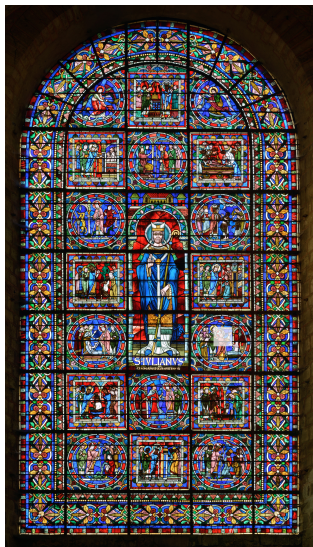
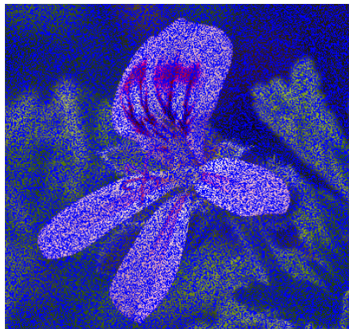


Image credit: Wikimedia/Selbymay



The curse

- Images are *redundant*



Challenges

- 1 How to **extract** the interesting information?
 - classification, image labelling, segmentation, sparse representation, image parametrisation
 - methods from **machine learning, statistics**

Dynamic image analysis for light microscopy

Optical flow estimation:

Cell tracking:

Computations by Hendrik Dirks/CIA,
image data from Goldstein Lab

References: Möller et al. '12; ...; cell
tracking through mitosis with Burger,
Grah, Reichelt (CRUK CI)

Challenges

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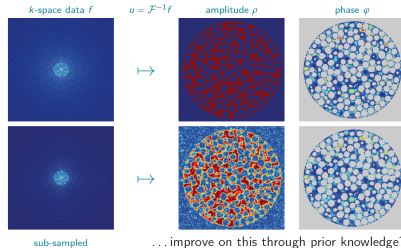
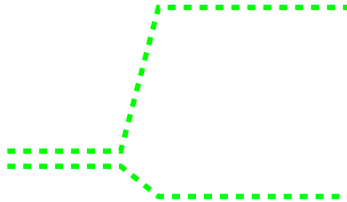
Challenges

- 1 How to **extract** the interesting information?
 - classification, image labelling, segmentation, sparse representation, image parametrisation
 - methods from **machine learning, statistics**

- 2 How to deal with **incomplete data**?
 - only **limited measurements** available, or
 - deliberate reduction for **speeding up** the imaging process
 - methods from **statistics, inverse problems, partial differential equations, variational methods, regularisation, large-scale optimisation**

Image reconstruction from phase-encoded MR data

Phase difference $\phi_1 - \phi_2$ of two images u_1 and u_2 proportional to velocity of a fluid along a gradient.



References: Gladden et al. '90-; ...; Benning, Gladden, Holland, CBS, Valkonen, Journal Magnetic Resonance '13. Animations courtesy of Alex Tayler.

Mathematical framework

The problem

Given data f , find the image information u that solves

$$f = T(u) + n$$

where T models the relation between u and f , and n is noise.

Issues

In most real-world applications, reconstructing f from u is

- not **unique** (T “forgets” data), or
- not **stable** (small variations in $f \rightarrow$ large variations in u),
- not **clear**, because the noise n is unknown

We need additional **prior knowledge** on u ...

Mathematical framework

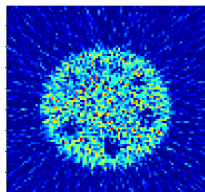
Energy minimisation framework

Reconstruct image information u from data f by **minimising**

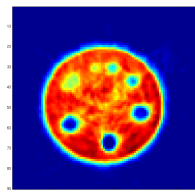
$$\min_u \underbrace{R(u)}_{\substack{\text{prior knowledge} \\ \text{class-specific}}} + \underbrace{D(T(u); f)}_{\text{fit to measurements } f}$$

→ state of the art **optimisation for large-scale computing**

- convex optimisation, operator splitting
- subspace decomposition, parallelisation, GPU computing



naive PET reconstruction



result of energy minimisation

Closing gaps: restoration of paintings



Figure: Adoration of the Shepherds, Sebastiano Del Piombo (1519)

References: Bachelor thesis Marie Autume (intern from ENS Cachan) in collaboration with Spike Bucklow (Hamilton Kerr Institute)

Closing gaps: restoration of paintings

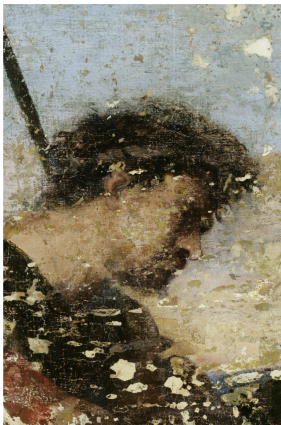


Figure: Detail

References: Bachelor thesis Marie Autume (intern from ENS Cachan) in collaboration with Spike Bucklow (Hamilton Kerr Institute)

Closing gaps: restoration of paintings

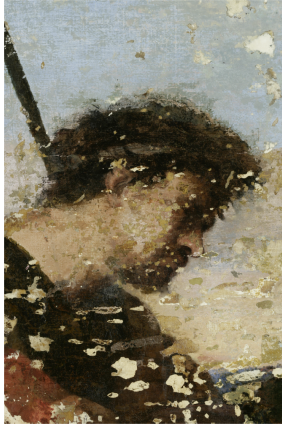


Figure: Small damages restored with structure inpainting

References: Bachelor thesis Marie Autume (intern from ENS Cachan) in collaboration with Spike Bucklow (Hamilton Kerr Institute)

Closing gaps: restoration of paintings



Figure: Large damages restored with structure & texture inpainting

References: Bachelor thesis Marie Autume (intern from ENS Cachan) in collaboration with Spike Bucklow (Hamilton Kerr Institute)

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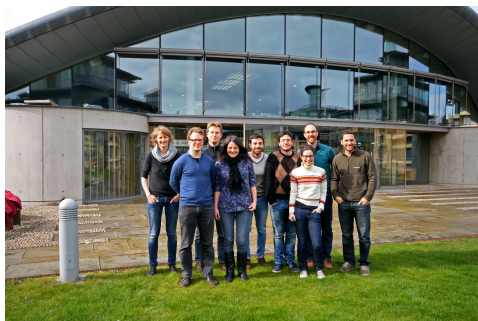
- Dr Carola Schönlieb
- Dr Jan Lellmann
- Dr Martin Benning (MRRRC)
- Dr Xiaohao Cai (Forest Ecology)
- Dr Kostas Papafitsoros (CCA)
- Dr Tuomo Valkonen
- Luca Calatroni (CCA)
- Joana Grah (Cancer Research)
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