COMPRESSIVE SPECTRAL IMAGING IN VIEW OF EARTH OBSERVATION APPLICATIONS

CLÉMENT THOMAS LAURENT JACQUES, MARC GEORGES

INTERNATIONAL SYMPOSIUM ON COMPUTATIONAL SENSING 2025 03/06/25









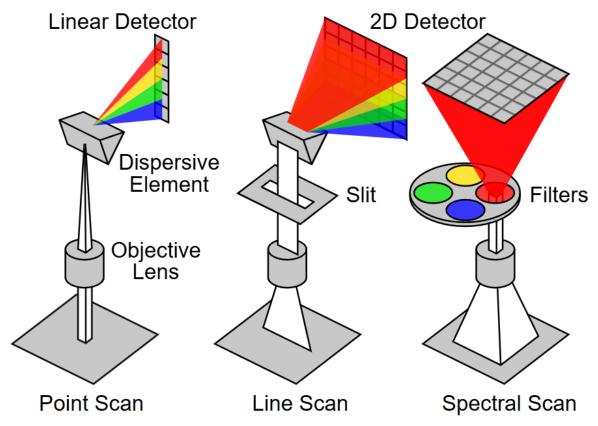


INTRODUCTION: EARTH OBSERVATION FROM SPACE



- Many applications: Meteorology, climatology, navigation, agriculture,...
- Large number of missions
- Need for high spatial and spectral resolutions (hyperspectral imagers)
- → Big Data: challenges for acquisition, storage, and transfer

TRADITIONAL SPECTRAL IMAGE ACQUISITION



Large data quantity:

$$N = n_x \times n_y \times n_\lambda$$

- Complex optics
- Spatially extended detectors
- Idea: use CS principles to build compact and fast spectral imagers

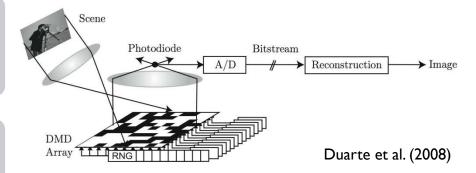
CHOICE OF ARCHITECTURE



Single-pixel camera: needs lots of time to acquire enough measurements



In Low Earth Orbit (LEO): time limited by the satellite's fast motion

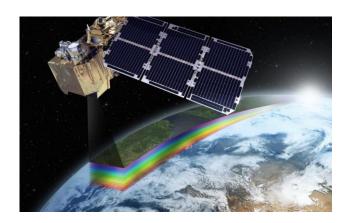




Need a compressive architecture with fast data acquisition

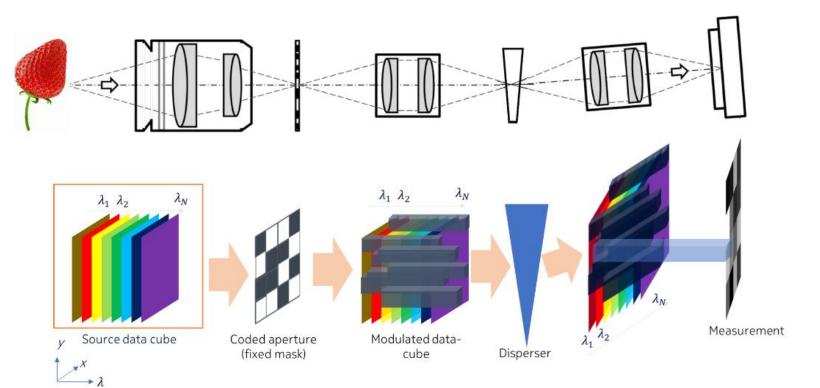


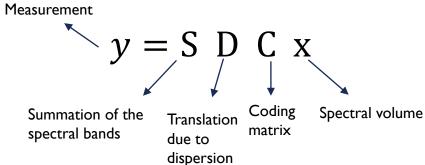
→ snapshot architectures



Source: ESA

CODED APERTURE SNAPSHOT SPECTRAL IMAGING (CASSI)



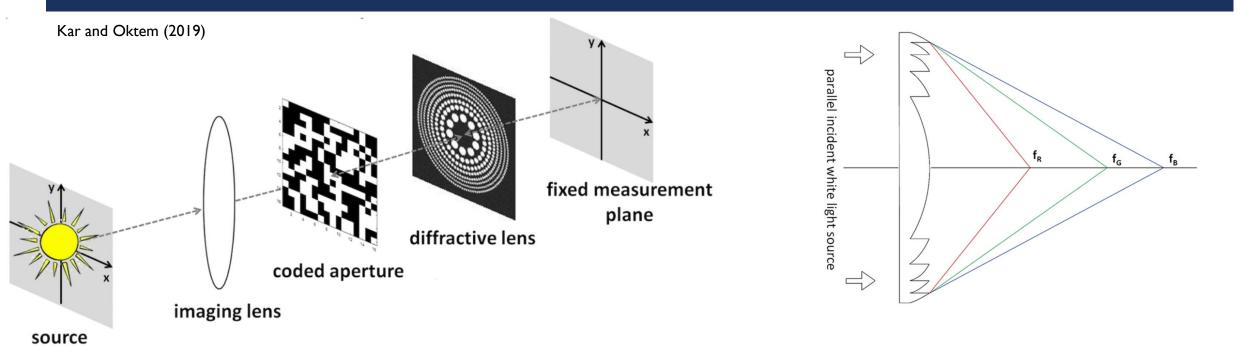


Compression ratio $\sim \frac{\#snapshots}{\#spectral\ bands}$

5

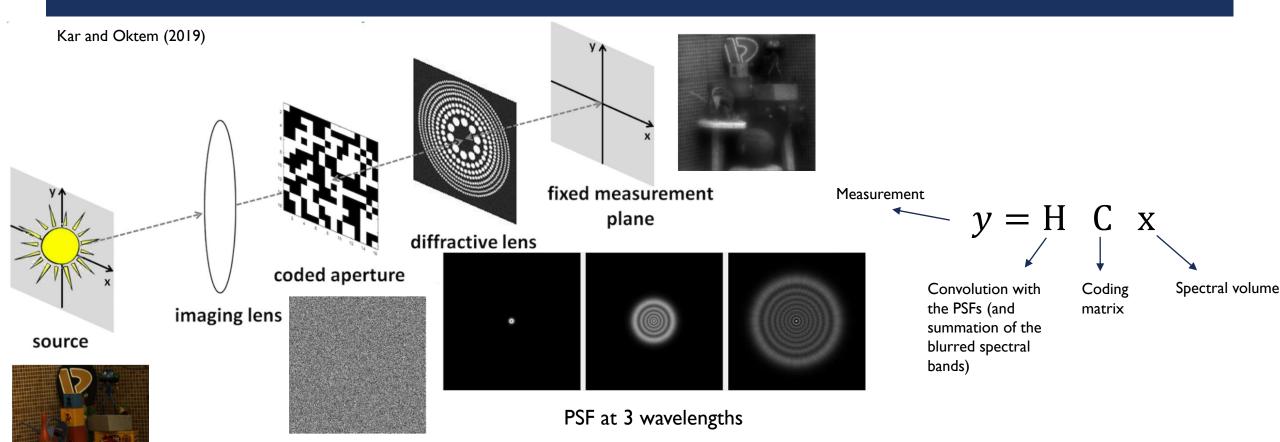
Yuan et al. (2021)

COMPRESSIVE SPECTRAL IMAGING USING A DIFFRACTIVE LENS (CSID)



- Idea: using a diffractive lens to simultaneously disperse and focus light
- Knowledge of the Point Spread Functions (PSFs) at the different wavelengths enables reconstruction

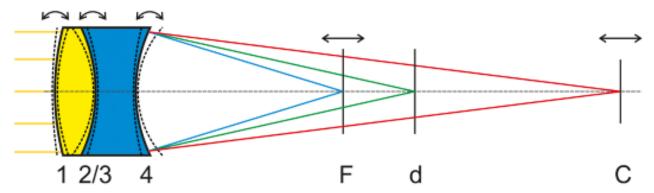
COMPRESSIVE SPECTRAL IMAGING USING A DIFFRACTIVE LENS (CSID)



ALTERNATIVE TO THE DIFFRACTIVE LENS

- Diffractive lenses can be complex to manufacture and sometimes expensive: is there an alternative?
- Idea: use chromatic aberration of refractive lenses

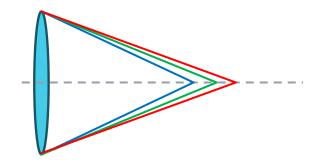
Hyperchromatic doublet



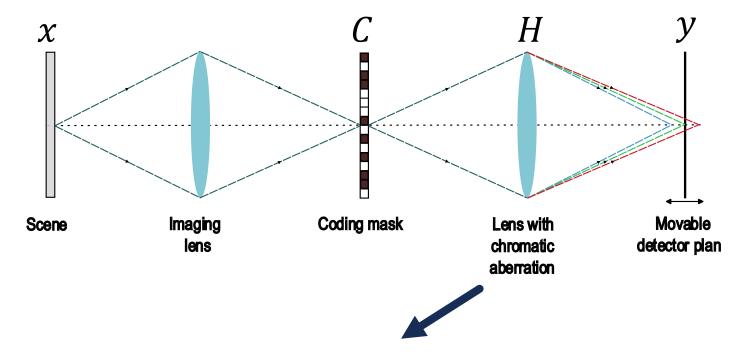
Refractive index:

$$n(\lambda) = A + \frac{B}{\lambda^2}$$

Single lens

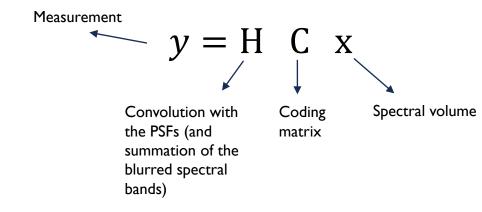


PROPOSED ARCHITECTURE



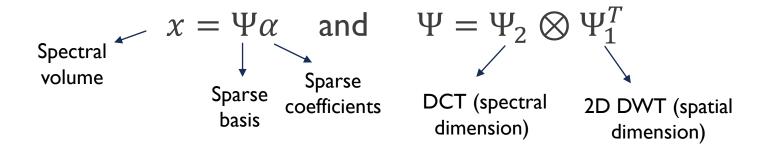
Simple NBK7 catalog lens:

- Cheap
- Easily procurable



RECONSTRUCTION

Saprsifying basis:



Forward model:

$$y = H C \Psi \alpha \rightarrow y = A \alpha$$
Convolution with Coding the PSFs mask Sensing operator

RECONSTRUCTION

Basis Pursuit:

Primal-dual formulation:

Where p is the dual variable

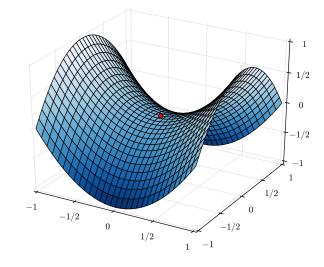
$$g(\alpha) = ||\alpha||_{1}$$

$$\downarrow$$

$$prox_{\lambda g}(s) = soft_{\lambda}(s)$$

$$\min_{\alpha} ||\alpha||_1 s.t. ||y - A\alpha||_2^2 \le \epsilon$$

$$\max_{p} \min_{\alpha} \langle A\alpha, p \rangle + g(\alpha) - f^{*}(p)$$



$$f(p) = \begin{cases} 0 & \text{if } ||p||_2 \le \epsilon \\ +\infty & \text{otherwise} \end{cases}$$

 $prox_f = \text{projection on the Euclidean ball } B_{\epsilon}(y)$

RECONSTRUCTION

Accelerated Chambolle-Pock algorithm

Iterations:

$$p^{k+1} = p^k + \sigma^k A \overline{\alpha}^k - prox_{\sigma^k f} \left(p^k + \sigma^k A \overline{\alpha}^k \right)$$

$$\alpha^{k+1} = prox_{\tau^k g} (\alpha^k - \tau^k A^T p^{k+1})$$

$$\theta^k = \frac{1}{\sqrt{1 + 2\gamma \tau^k}}$$

$$\tau^{k+1} = \theta^k \tau^k$$

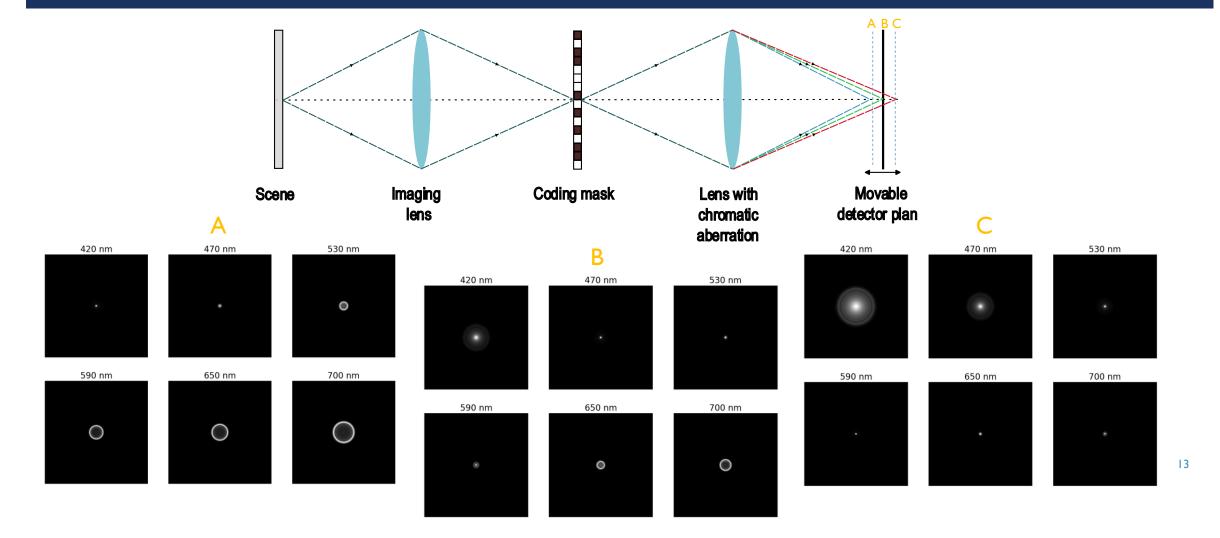
$$\sigma^{k+1} = \frac{\sigma^k}{\theta^k}$$

$$\overline{\alpha}^{k+1} = \alpha^{k+1} + \theta^k (\alpha^{k+1} - \alpha^k)$$

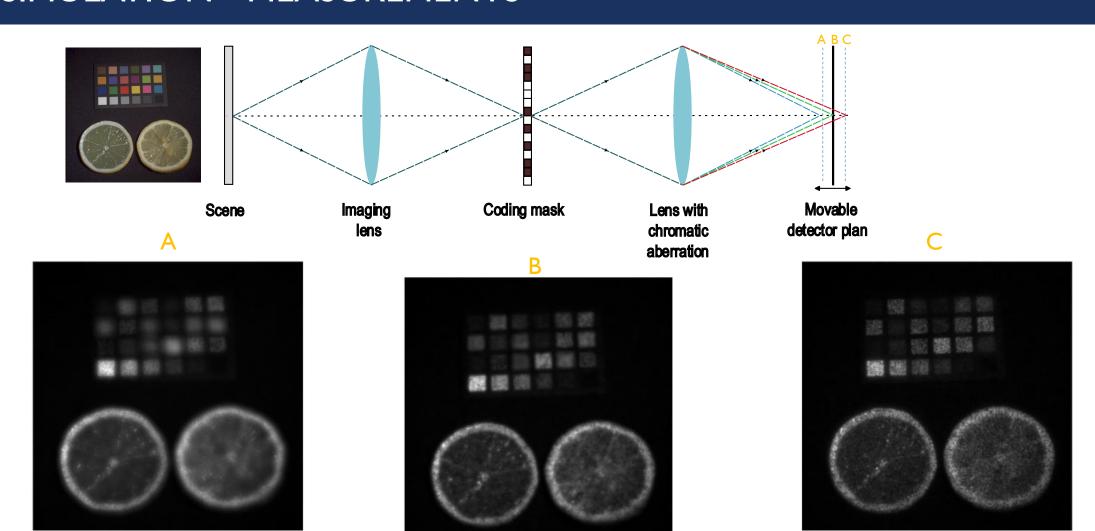
Final output: $\overline{\alpha}$

Final reconstructed datacube:

SIMULATION - PSF



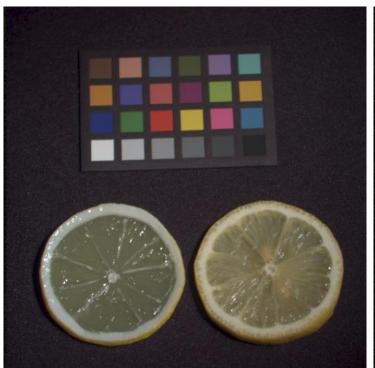
SIMULATION - MEASUREMENTS

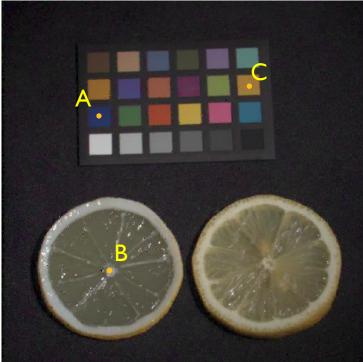


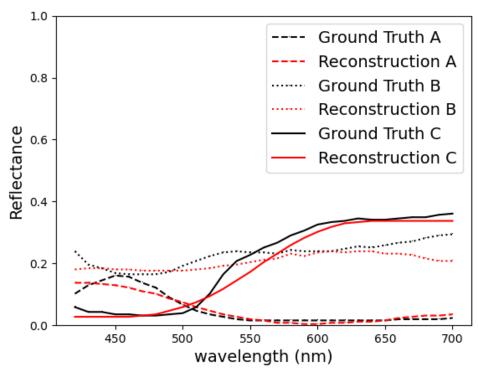
RECONSTRUCTION FROM 10% SUBSAMPLING

Original

Reconstruction (noiseless measurement) PSNR = 35.46 dB





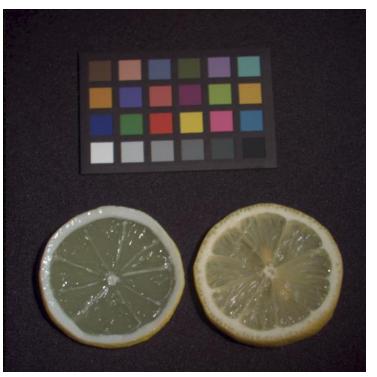


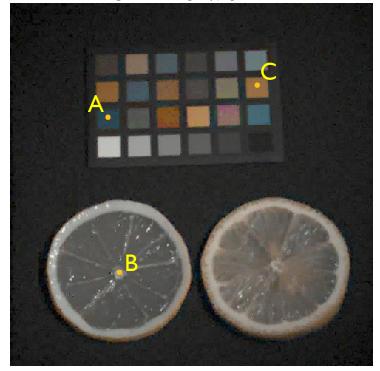
RECONSTRUCTION FROM 10% SUBSAMPLING

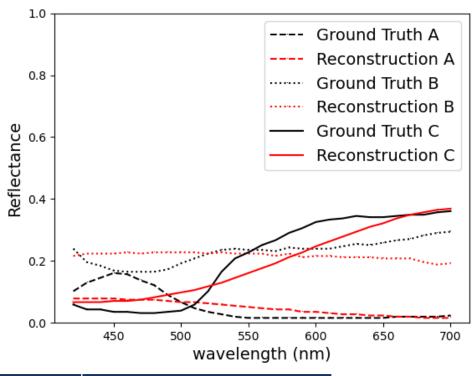
Original

Reconstruction (measurement SNR = 28 dB)

PSNR = 31.76 dB







16

Measurement SNR	22 d B	28 d B	34 dB
Reconstruction PSNR	30.58 dB	31.76 dB	32.82 dB

CONCLUSION AND FUTURE WORK



Simulations show the feasibility of the method

Compact compressive spectral imager using:

- Two lenses
- One SLM
- One panchromatic detector



Laboratory setup is under development



Areas of improvement:

Reconstruction method

Coding patterns