



Survey of Hyperspectral Imaging Techniques

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October 30, 2009

Compared Systems

- **Baseline – Scanning Filter**
- **Baseline – Simple Pushbroom**
- **Gehm (Brady) – Multiplexed Pushbroom**
 - “High-throughput, multiplexed pushbroom hyperspectral microscopy”
- **Wagadarikar (Brady) – Single Disperser**
 - “Single disperser design for coded aperture snapshot spectral imaging”
- **Gehm (Brady) – Dual Disperser**
 - “Single-shot compressive spectral imaging with a dual-disperser architecture”
- **Descour – CTIS**
 - “Computed-tomography imaging spectrometer: experimental calibration and reconstruction results”
- **Mooney – Prism Tomographic**
 - “High-throughput hyperspectral infrared camera”
- **Gentry – ISIS**
 - “Information-Efficient Spectral Imaging Sensor”
- **Mohan (Raskar) – Agile Spectrum Imaging**
 - “Agile Spectrum Imaging: Programmable Wavelength Modulation for Cameras and Projectors”

Points of Comparison

- Data volume
- Physical volume
- Architectural impact on acquisition time
- Computational reconstruction and scaling
- Photon efficiency (noise, sensitivity, etc.)
- Compression (Information efficiency)

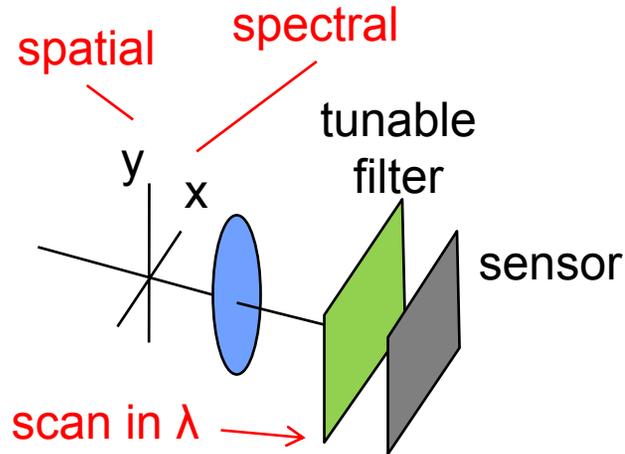
Caveats

- Many quantities (like physical volume and reconstruction scaling) depend heavily on the specific implementation. Interpret these results as expected limits.
- Data quality metric – there is none. Different techniques can be expected to produce different amounts and types of artifacts. These are discussed qualitatively herein.

Baseline – Scanning Filter

Summary:

- Data Cube: $N_x \times N_y \times L$
- Volume: $1f * D^2$
- Acquisition time: scanning.
- Reconstruction: None
- Photon Efficiency: $1/L$
- Compression: 1

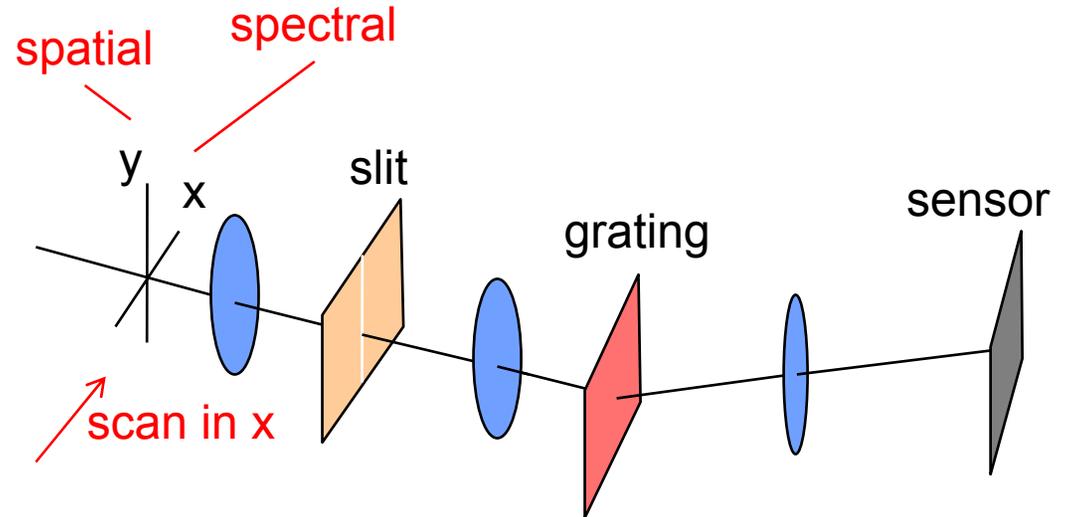


Scan in λ using an electronically-tunable filter. Typically, the filter is based on either liquid crystals or acousto-optic principles.

Baseline – Pushbroom

Summary:

- Data Cube: $N_x \times N_y \times L$
- Volume: $5f * D^2$
- Acquisition time:
Mechanical motion is required between lines (resulting in photon dead-time) but object motion is treated stably.
- Reconstruction: None
- Photon Efficiency: $1/N_x$
- Compression: 1

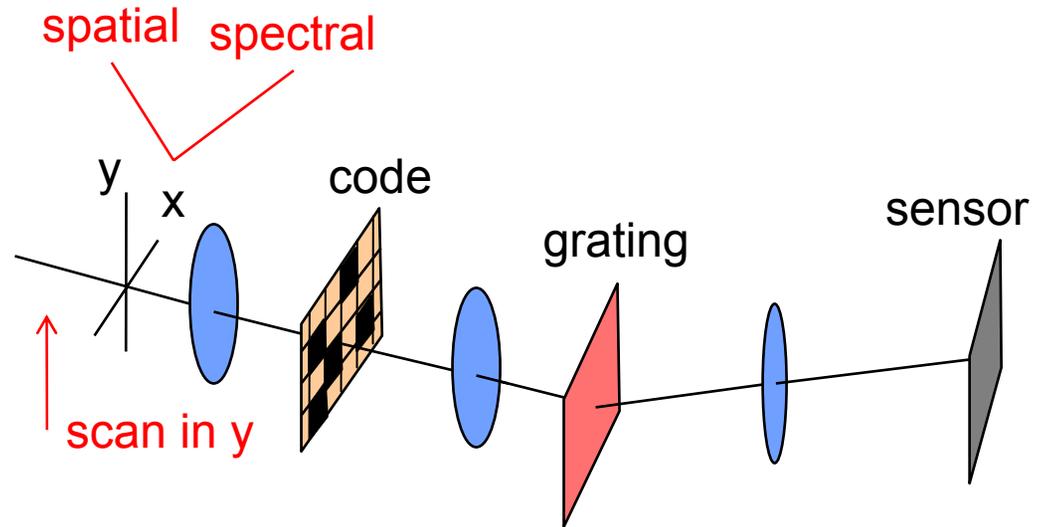


Each row on the sensor provides a spectrum at that y value. Scanning in x provides the other spatial dimension.

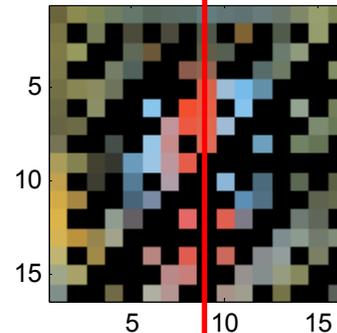
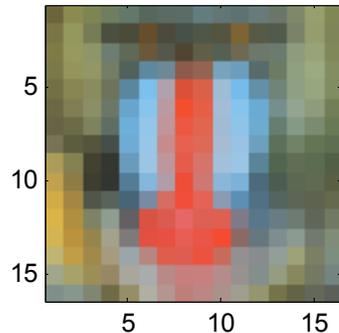
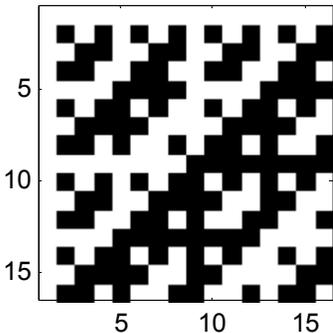
Gehm (Brady) – Multiplexed Pushbroom

Summary:

- Data Cube: $N_x \times N_y \times L$
- Volume: $5f * D^2$
- Acquisition time:
Mechanical motion is required between lines.
- Reconstruction: $O(N_x N_y^2 L)$
- Photon Efficiency: $\sim 1/2$
- Compression: ~ 1



code/decode orthogonality requires scene uniformity in y .



by sliding code over scene vertically (or vice versa) one can mix rows to synthesize columns of uniform scene value.

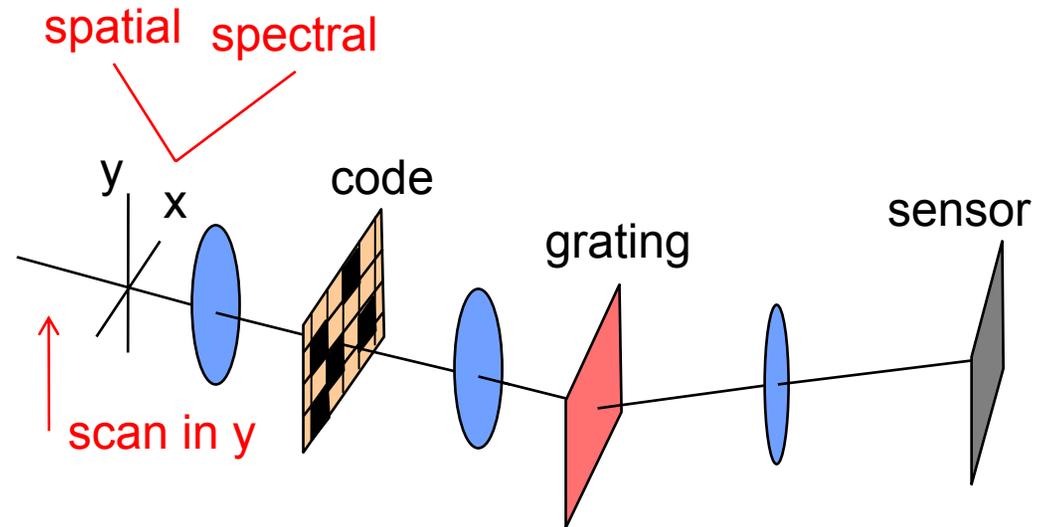
Gehm (Brady) – Multiplexed Pushbroom (2)

- Reconstruction: $O(N_x N_y^2 L) = O(N_x N_y L \times N_y)$
Every point in the data cube is a dot-product of length- N_y vectors.
- Scanning options:
 - Scan scene over code for “continuous” pushbroom mode, requiring slightly more complex data re-mapping, or
 - Circularly scan code through the field stop for fixed-field capture
- In prototype systems, resolution was set by code size to order 6x6 CCD pixels for processing/sampling convenience. The re-binning and digital aberration (smile) correction was not included in the reconstruction scaling.

Wagadarikar (Brady) – Single Disperser

Summary:

- Data Cube: $N_x \times N_y \times L$
- Volume: $5f * D^2$
- Acquisition time: Mechanical motion is required between lines (if any).
- Reconstruction: $O((N_x N_y L)^3)$, L_1 minimization
- Photon Efficiency: $\sim 1/2$
- Compression: $1/L$ to 1
- Identical hardware to Multiplexed Pushbroom
- Skip scan steps or don't scan at all
- Reconstruct via L_1 minimization
- Reduced spatial information in single-shot mode – object pixels imaged to closed code addresses are completely lost



Gehm (Brady) – Dual Disperser

Summary:

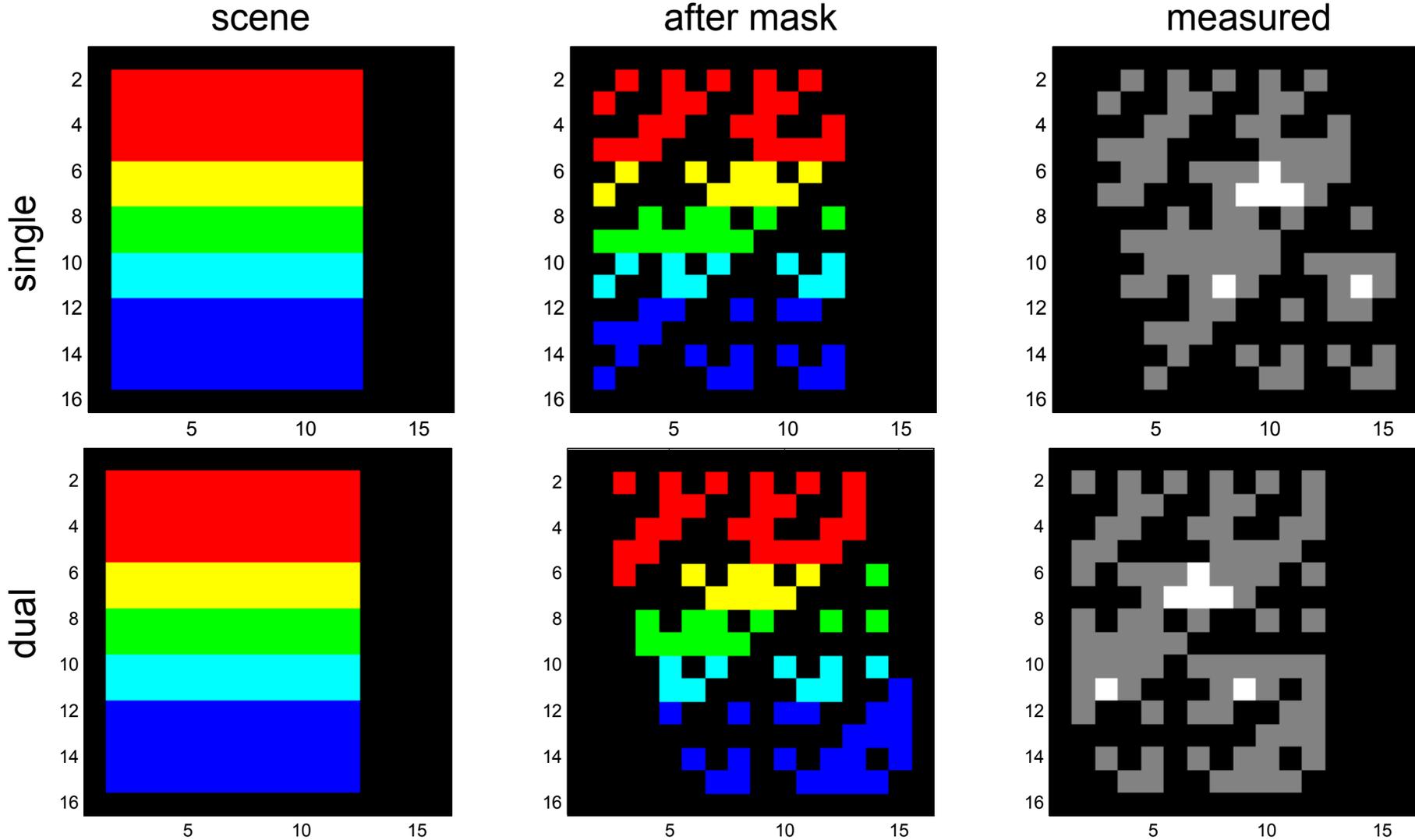
- Data Cube: $N_x \times N_y \times L$
- Volume: $9f * D^2$
- Acquisition time: Snapshot
- Reconstruction: $O((N_x N_y L)^3)$,
 L_1 minimization
- Photon Efficiency: $\sim 1/2$
- Compression: $1/L$

- Raw measured frames are spatially isomorphic with scene – each pixel is a spectral projection.

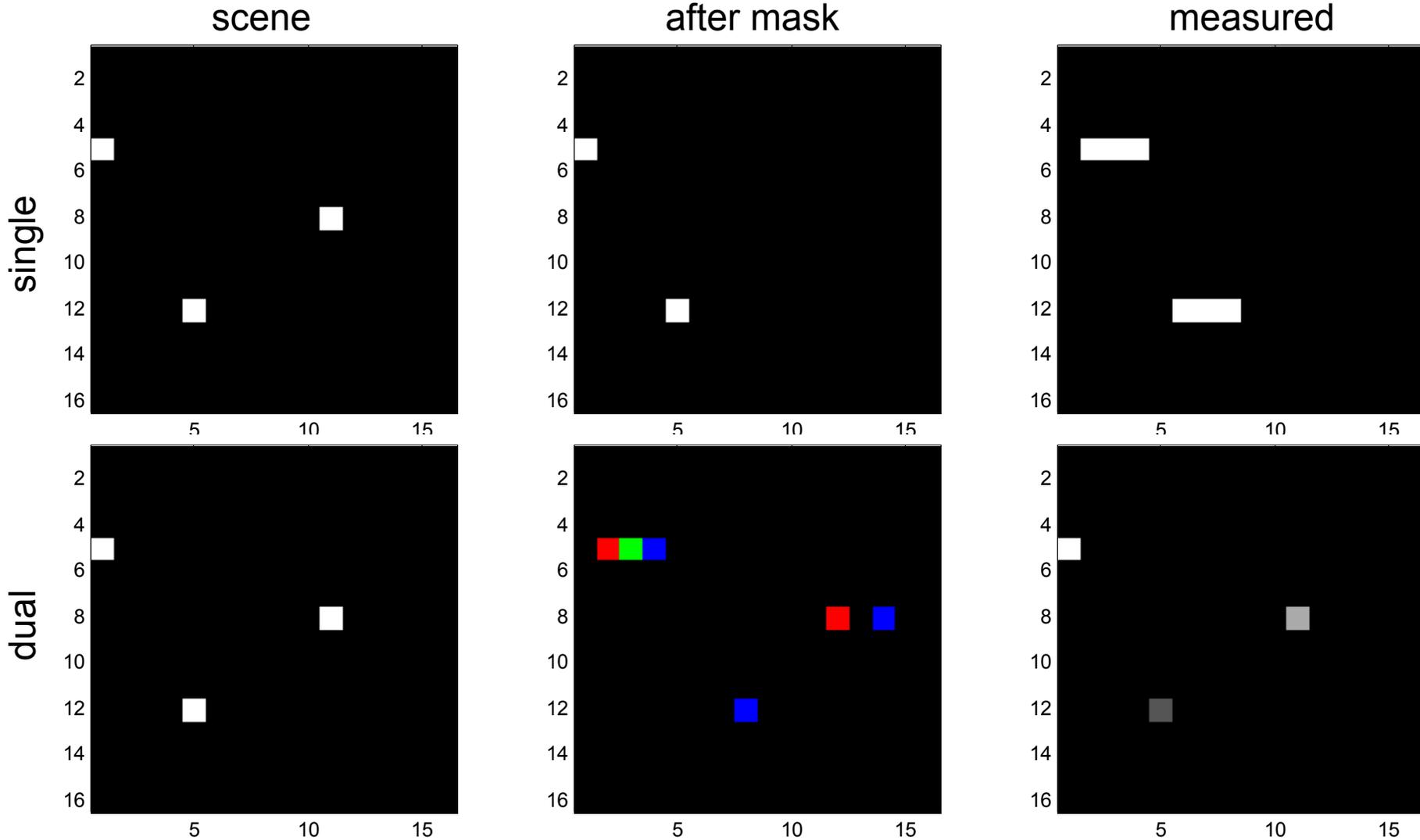
Images removed due to copyright restrictions.

Source: Gehm, M. E. et al. "Single-shot Compressive Spectral Imaging with a Dual-disperser Architecture." *Optics Express* 15, no. 21 (2007): 14013-14027.

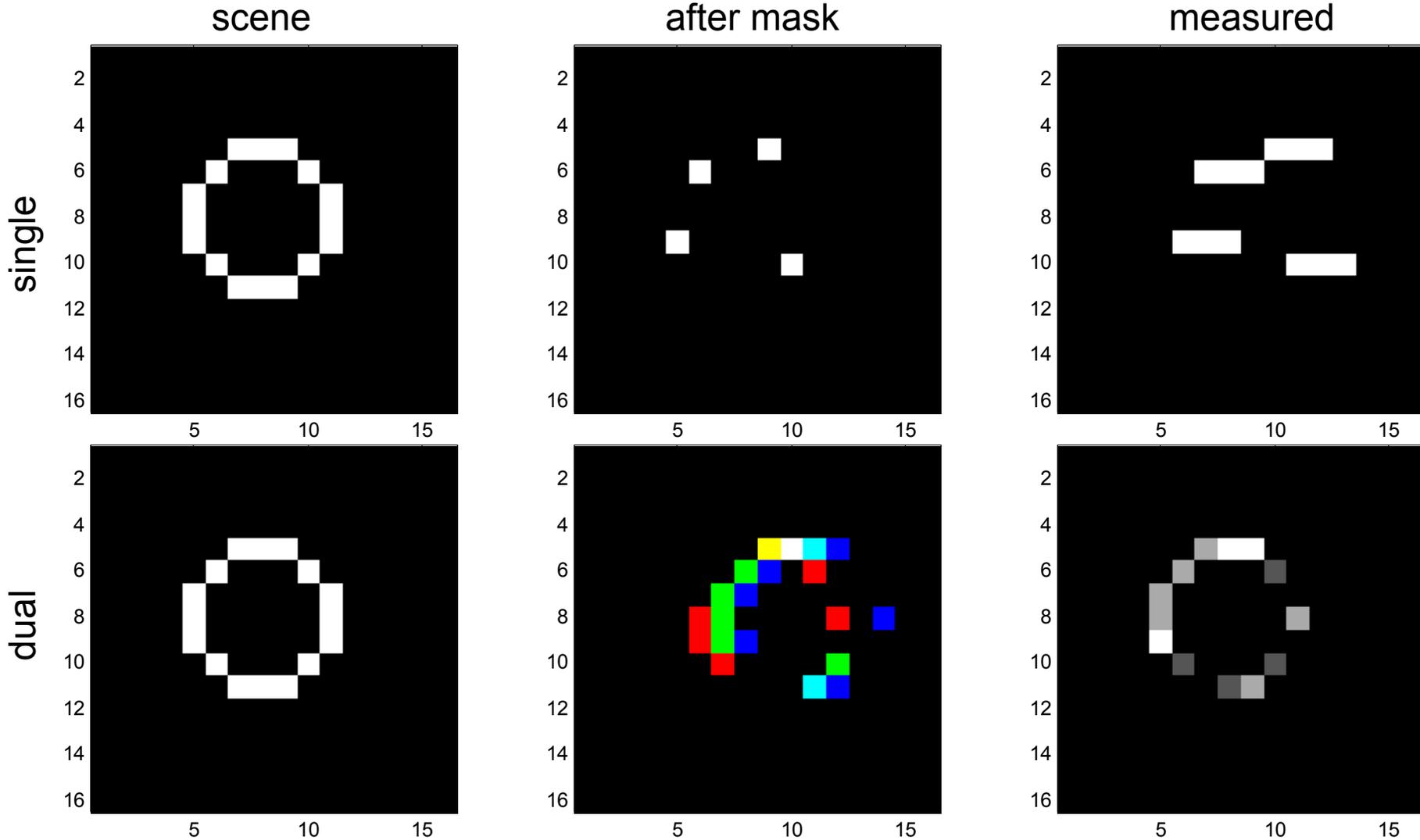
Single/Dual Disperser Comparison



Single/Dual Disperser Comparison



Single/Dual Disperser Comparison



Descour – CTIS

Summary:

- Data Cube: $N_x \times N_y \times L$
- Volume: $4f * D^2$
- Acquisition time: Snapshot
- Reconstruction:
 - $O(n^3)$, FBP
 - $O(n^2 \log n)$, Fourier
- Photon Efficiency: 1
- Compression: ~ 1

Images removed due to copyright restrictions.

Source: Descour, M., and E. Dereniak. "Computed-tomography Imaging Spectrometer: Experimental Calibration and Reconstruction Results."

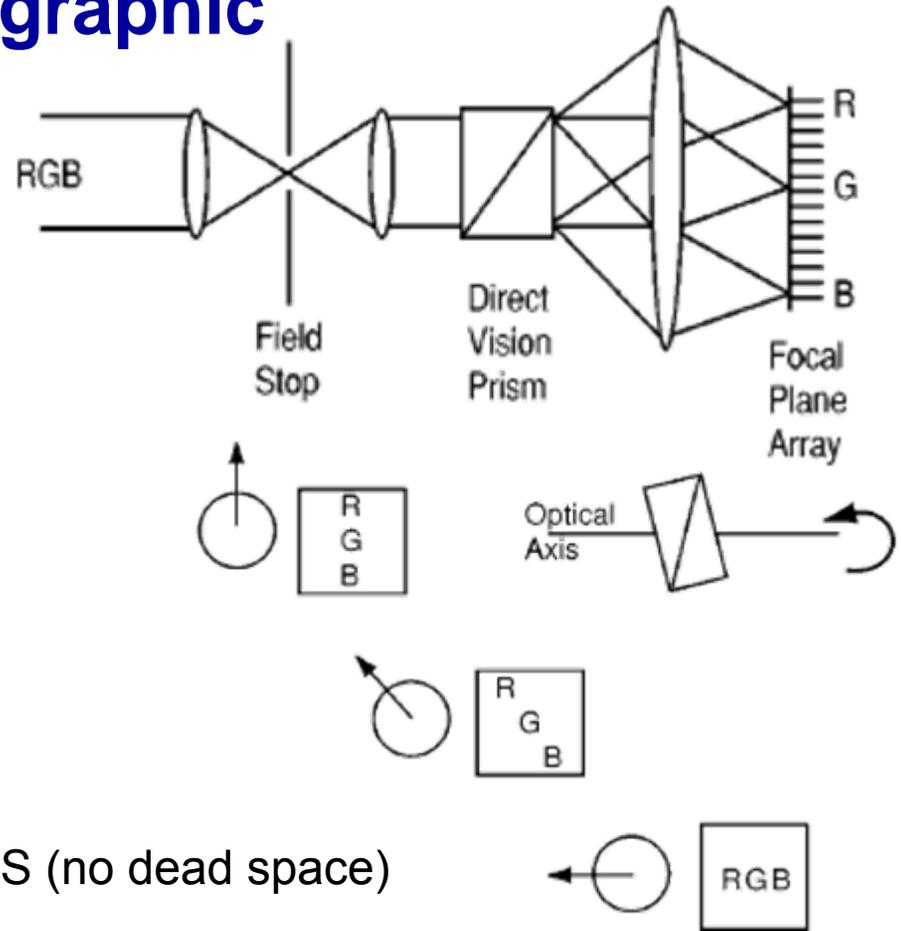
Applied Optics 34, no. 22 (August 1, 1995): 4817-4826.

- Inefficiently uses sensor; dead spaces required to avoid overlap.
- Requires $P > N_x \times N_y \times L$ pixels
- Limited information efficiency; missing cone problem
- Reconstruction approaches have been proposed to improve missing cone (extrapolation and model-based approaches)

Mooney – Prism tomographic

Summary:

- Data Cube: $N_x \times N_y \times L$
- Volume: $4f * D^2$
- Acquisition time: Scanning
- Reconstruction:
 - $O(n^3)$, FBP
 - $O(n^2 \log n)$, Fourier
- Photon Efficiency: 1
- Compression: ~ 1



- More efficiently uses pixels than CTIS (no dead space)
- Requires $P = N_x \times N_y$ pixels.
- Limited information efficiency; missing cone problem
- Reconstruction approaches have been proposed to improve missing cone (extrapolation and model-based approaches)

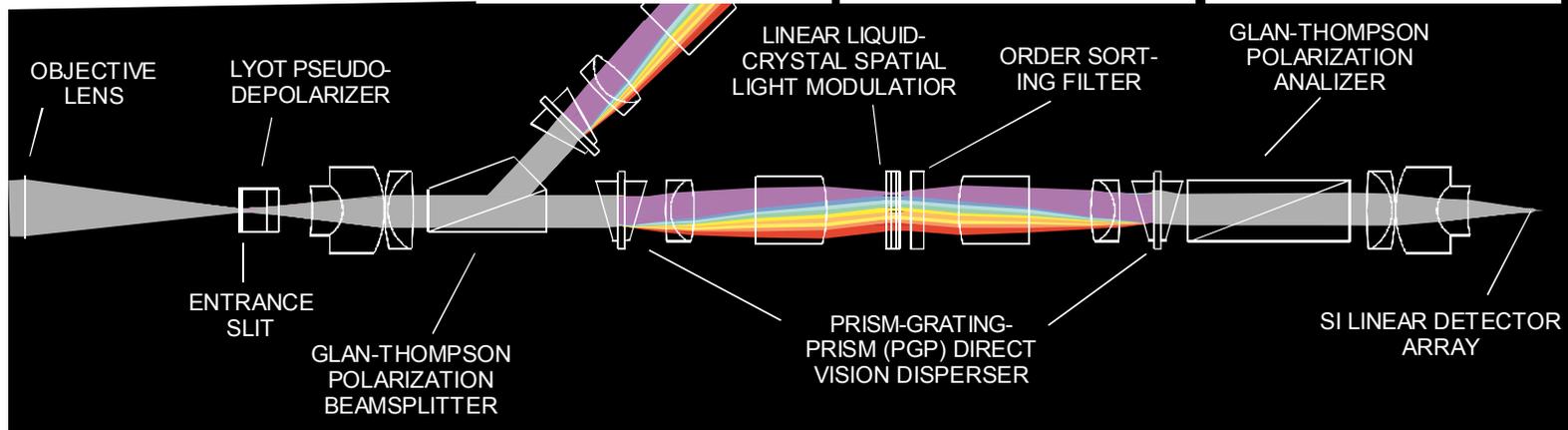
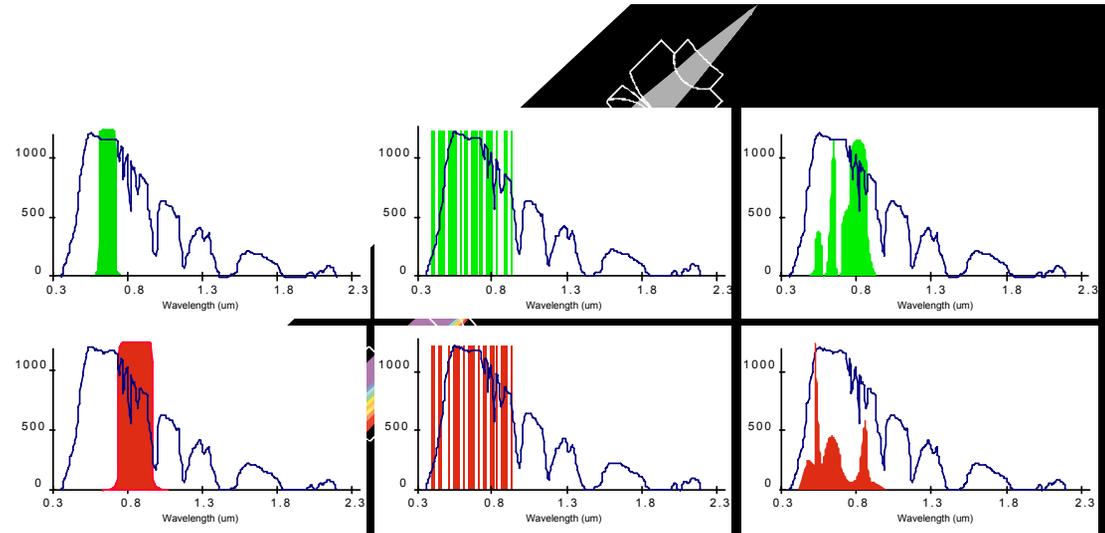
Image from Mooney, JM et al. "High-throughput hyperspectral infrared camera." *JOSA A* 14, no. 11 (1997): 2951-2961. (All authors with US Air Force.)

Gentry – ISIS

Sandia National Laboratories, US Department of Energy

Summary:

- Data Cube: $N_x \times N_y \times 1$
- Volume: $9f * D^2$
- Requires SPM/SLM
- Acquisition time: Scanning
- Reconstruction: $N_x N_y$
- Photon Efficiency: $\sim 1/(4N_y)$
- Compression: 2

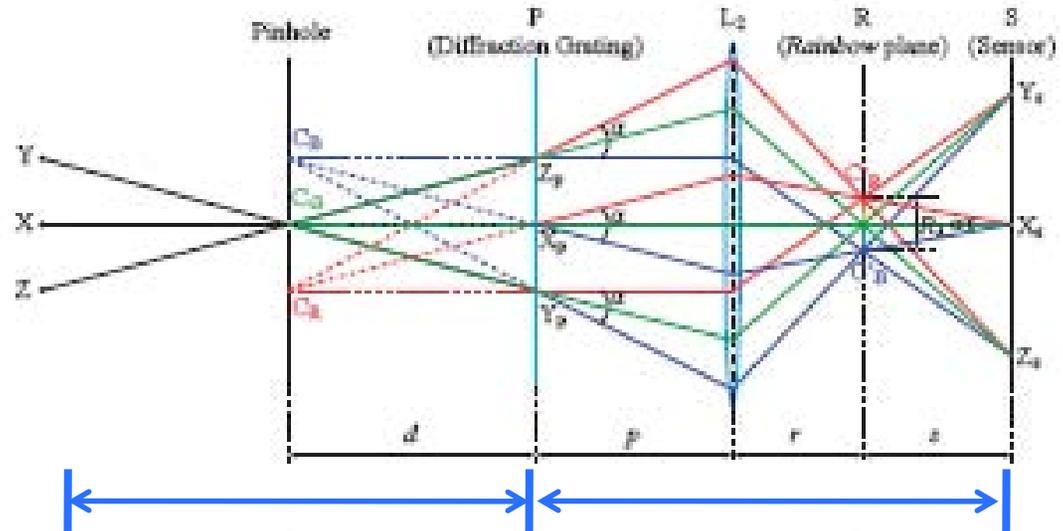


- Reconstruction: subtraction required for every $N_x N_y$ point
- Photon efficiency: for any given pixel-channel band, one arm is always zero (losing half the light) and the other will in general be between 0 and 1.

Mohan (Raskar) – Agile Spectrum Imaging

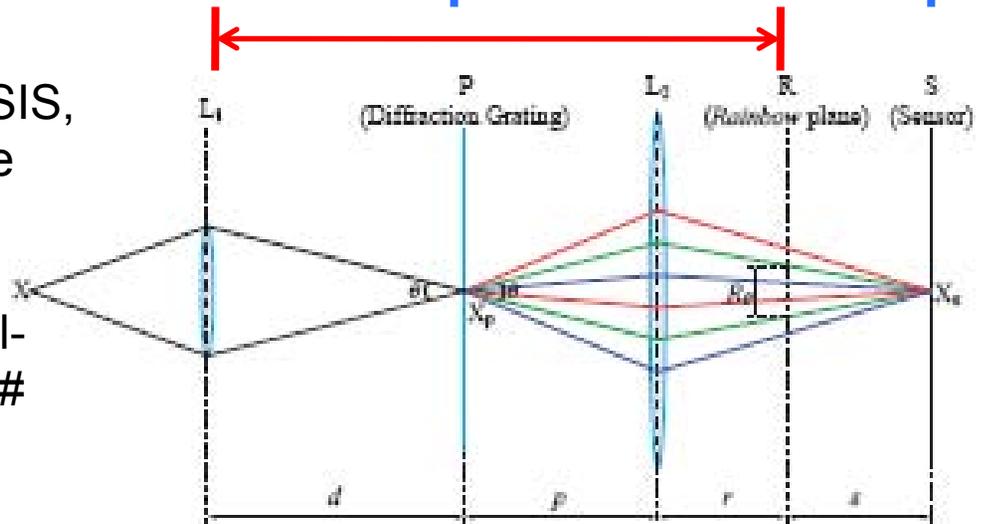
Summary:

- Data Cube: $N_x \times N_y \times 1$
- Volume: $5f * D^2$
- Requires SLM
- Acquisition time: Snapshot
- Reconstruction: None
- Photon Efficiency: $\sim 1/2$
- Compression: 1



Not designed to be a HSI, but like ISIS, allows for spectrally-weighted image acquisition. Differences from ISIS:

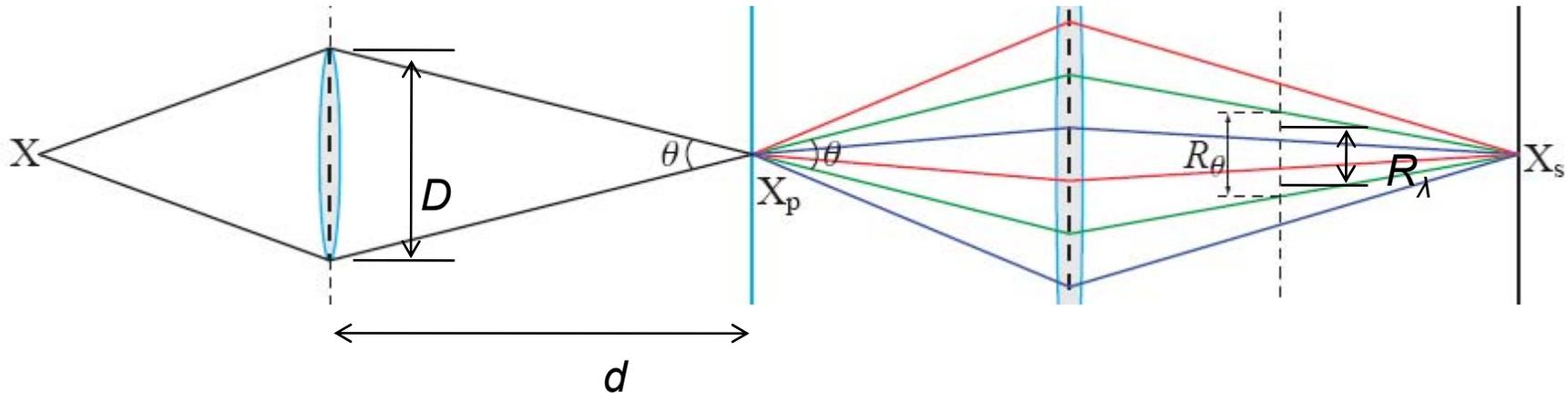
- Limited spectral filtering and spatial-spectral coupling as a function of $F/\#$
- Positive-only filter functions



Images courtesy of Ramesh Raskar. Used with permission.

Source: Mohan, A., R. Raskar, and J. Tumblin. "Agile Spectrum Imaging: Programmable Wavelength Modulation for Cameras and Projectors" *Eurographics* 2008, Vol 27 no. 2 (2008).

Mohan (Raskar) – Agile Spectrum Imaging spectral selectivity



R_θ = width of one wavelength in rainbow plane

R_λ = distance between centers of extreme wavelengths

$$\text{Maximum number of distinct wavelengths} = \frac{R_\lambda}{R_\theta} + 1 = \frac{d}{D} + 1 = F + 1$$

Where F is the F-number of the objective lens. Therefore, high spectral selectivity requires a very slow system.

Image courtesy of Ramesh Raskar. Used with permission.

Source: Mohan, A., R. Raskar, and J. Tumblin. "Agile Spectrum Imaging: Programmable Wavelength Modulation for Cameras and Projectors" *Eurographics* 2008, Vol 27 no. 2 (2008).

Summary

| | Data Cube | Physical Volume | Acquisition | Reconstruction | Photon Efficiency | Compression |
|-----------------------|---------------------------|-----------------|-----------------------|---|-------------------|-------------|
| Scan. Filter | $N_x \times N_y \times L$ | $1f * D^2$ | Scanning | None | $1/L$ | 1 |
| Pushbroom | $N_x \times N_y \times L$ | $5f * D^2$ | Scanning | None | $1/N_x$ | 1 |
| Multiplexed Pushbroom | $N_x \times N_y \times L$ | $5f * D^2$ | Scanning | $O(N_x N_y^2 L)$ | $\sim 1/2$ | 1 |
| Single Disperser | $N_x \times N_y \times L$ | $5f * D^2$ | Scanning/ Snapshot | $O((N_x N_y L)^3)$, L_1 minimization | $\sim 1/2$ | $1/L$ to 1 |
| Dual Disperser | $N_x \times N_y \times L$ | $9f * D^2$ | Snapshot | $O((N_x N_y L)^3)$, L_1 minimization | $\sim 1/2$ | $1/L$ |
| CTIS | $N_x \times N_y \times L$ | $4f * D^2$ | Snapshot | $O(n^3)$, FBP $O(n^2 \log n)$, Fourier | 1 | ~ 1 |
| Prism Tomographic | $N_x \times N_y \times L$ | $4f * D^2$ | Scanning | $O(n^3)$, FBP $O(n^2 \log n)$, Fourier | 1 | ~ 1 |
| ISIS | $N_x \times N_y \times 1$ | $9f * D^2$ | Scanning | $N_x N_y$ | $\sim 1/(4N_y)$ | 2 |
| Agile Spectrum | $N_x \times N_y \times 1$ | $5f * D^2$ | Snapshot | None | $\sim 1/2$ | 1 |

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MAS.531 Computational Camera and Photography
Fall 2009

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