

# Spectro-Perfectionism in SDSS-III

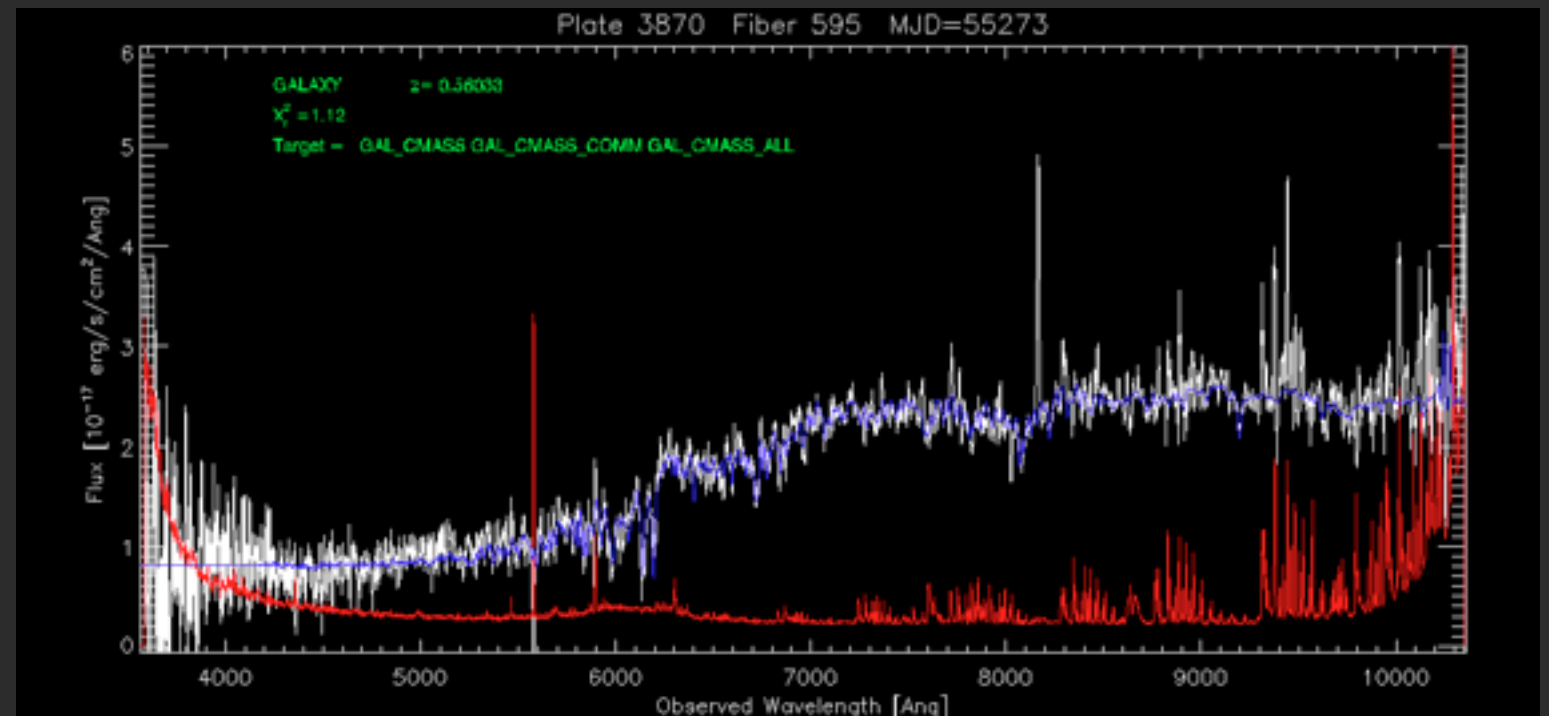
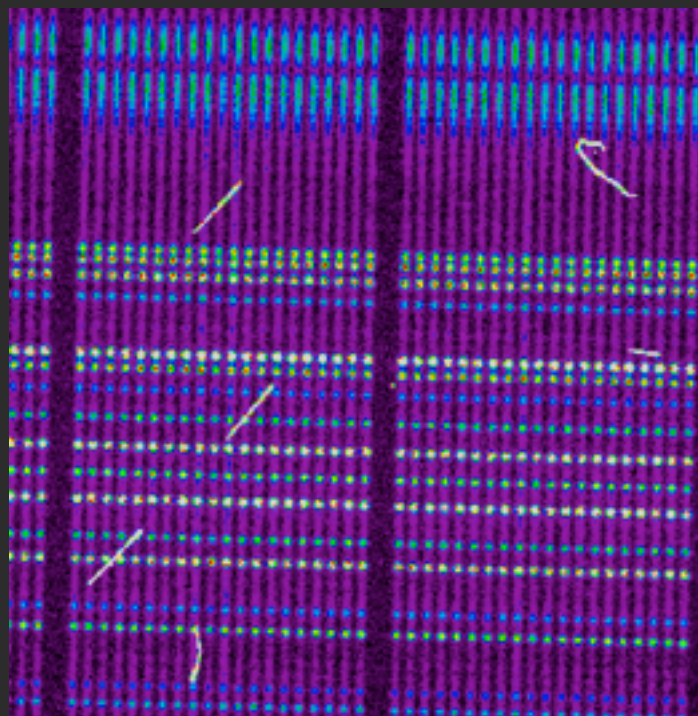
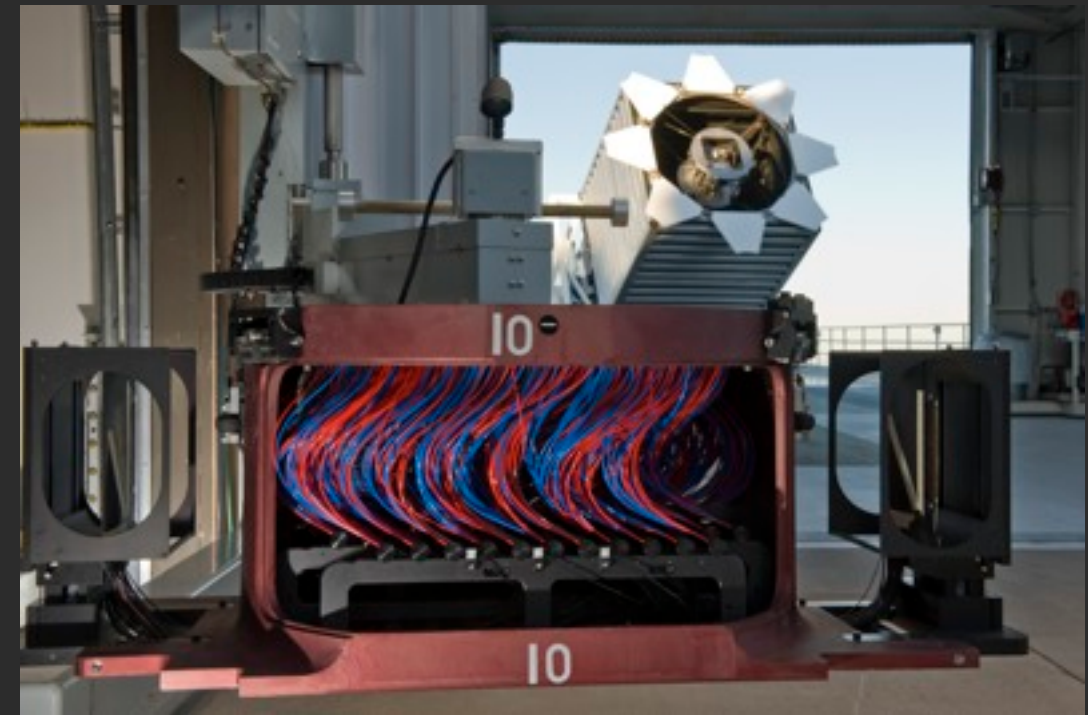
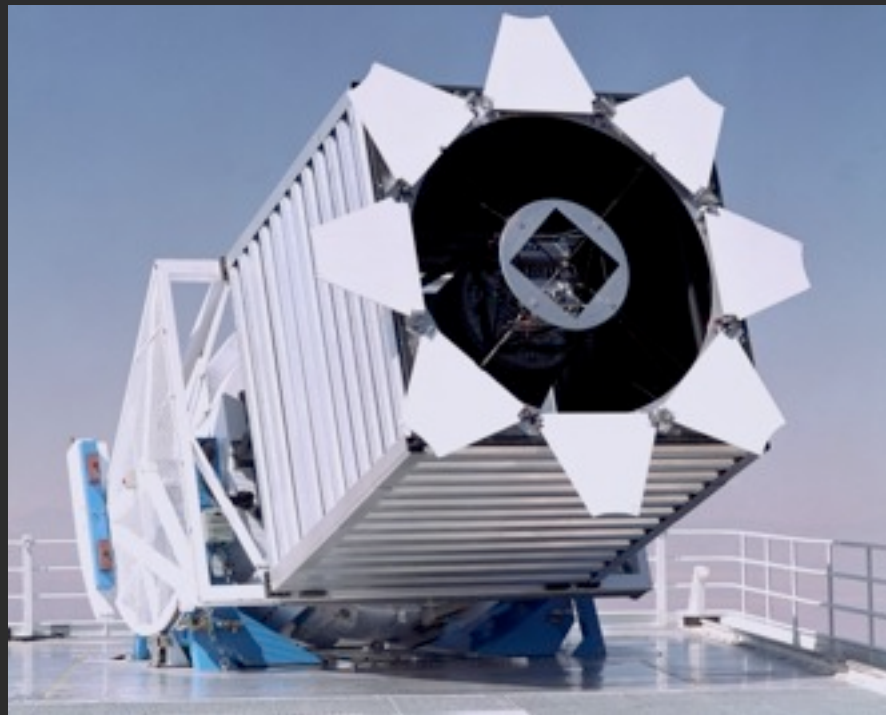
Adam S. Bolton

Department of Physics & Astronomy  
The University of Utah

Following the Photons - ROE - 2011-10-10

# What is SDSS-III?

Eisenstein et al. 2011



# What is SDSS-III?

Eisenstein et al. 2011

## BOSS: The Baryon Oscillation Spectroscopic Survey

- One of the four SDSS-III surveys
- 2009-2013 spectroscopic operations
- Redshifts of 1.5 million galaxies to  $z = 0.7$
- 160k quasars for Lyman- $\alpha$  forest
- Measurement of baryon acoustic feature vs.  $z$
- Constrain parameters of “dark energy”
- Largest spectro data set for massive galaxy evolution

# What is...

Spectro-Perfectionism

a.k.a.

2D PSF Extraction

a.k.a.

the Bolton & Schlegel algorithm

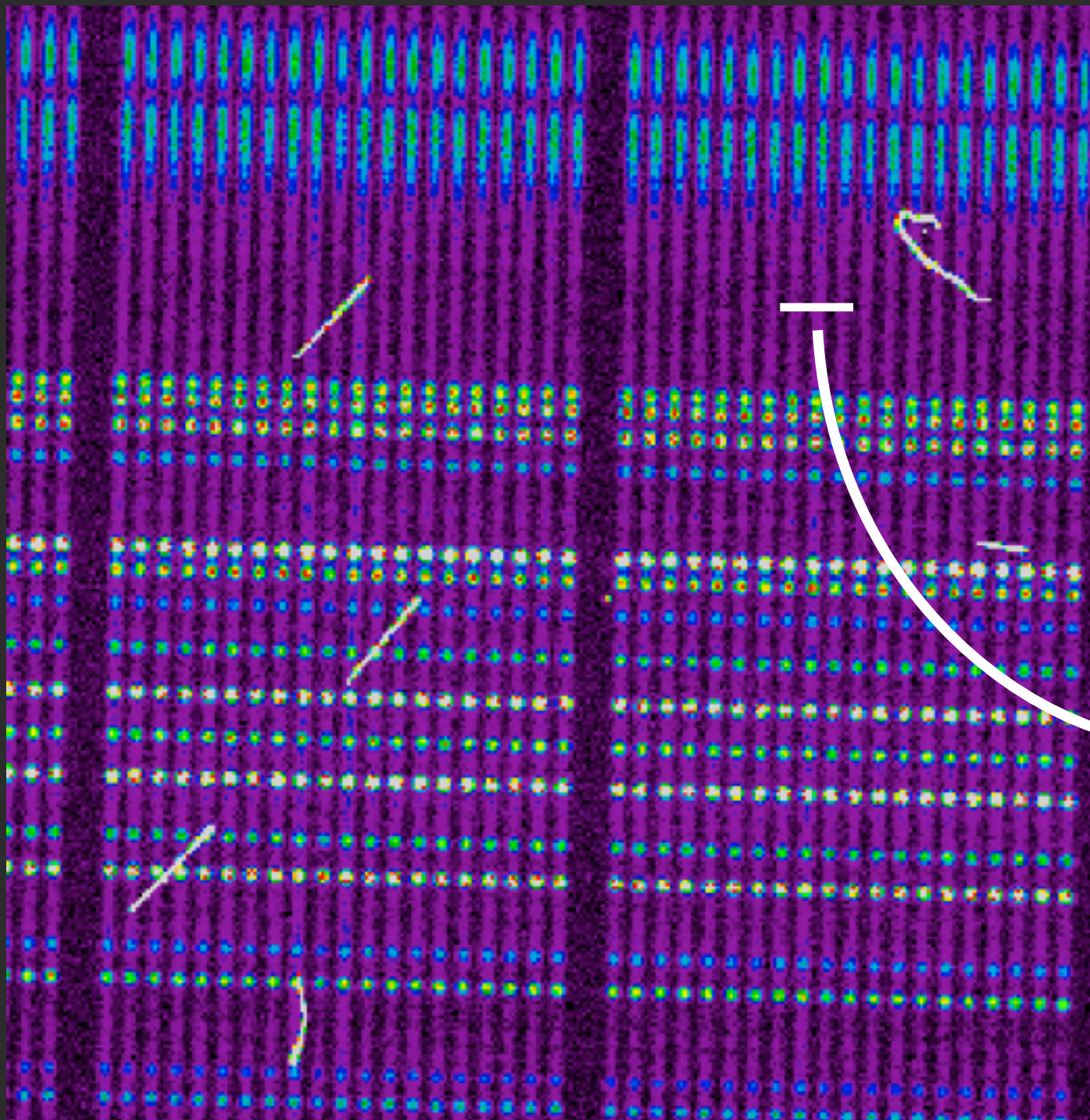
?

(Bolton & Schlegel 2010,  
PASP, 122, 248)

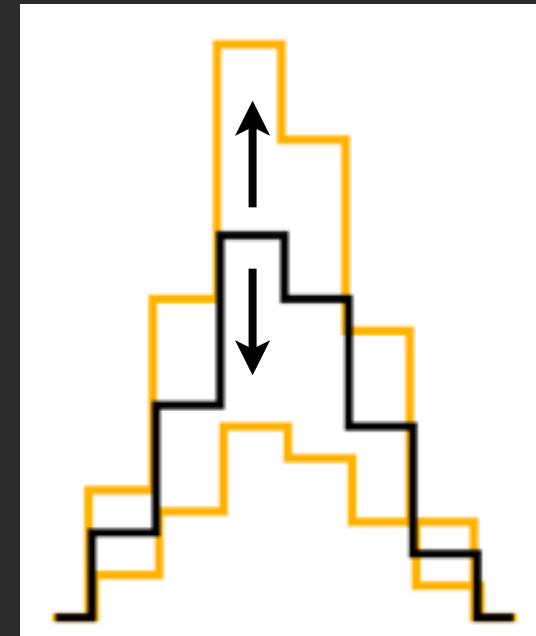
Spectroscopic extraction via mathematically correct forward modeling of the raw data via the 2D spectrograph point-spread function (PSF).



# Doesn't "optimal extraction" do this?

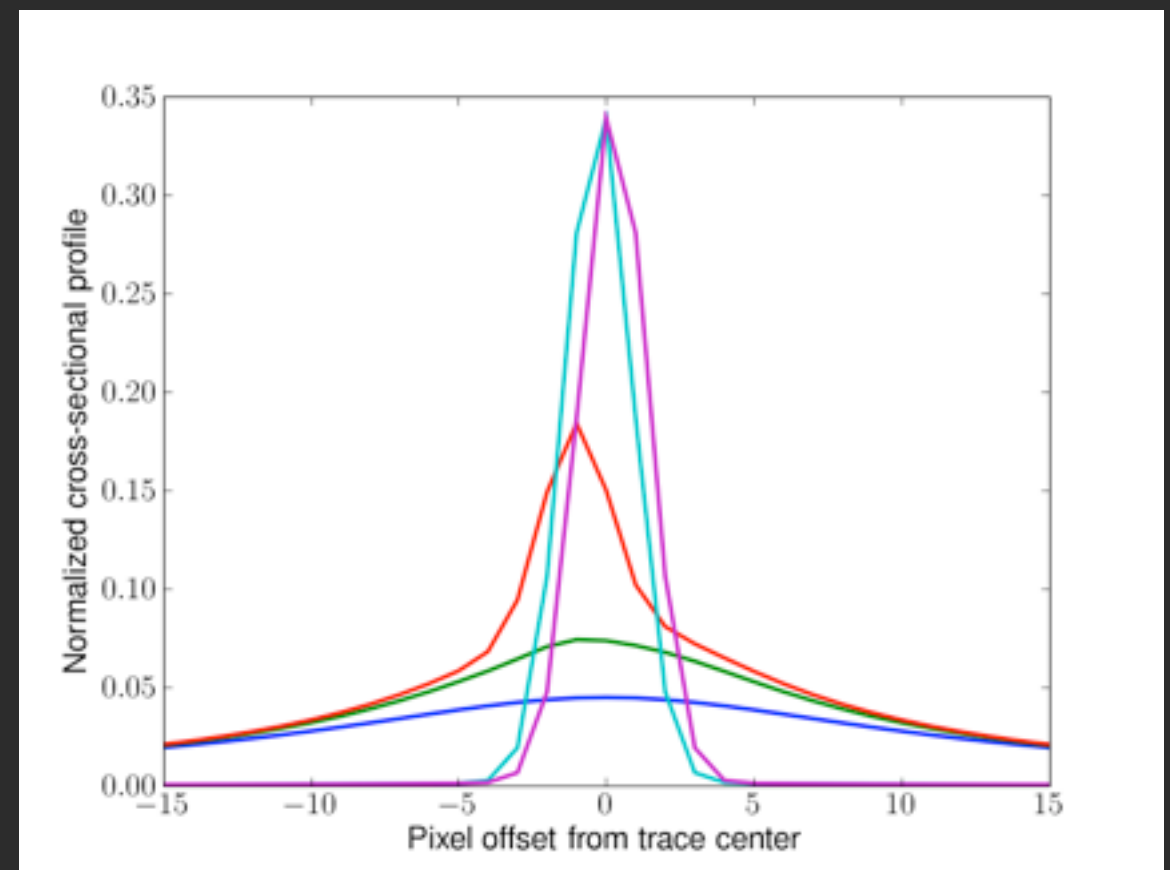
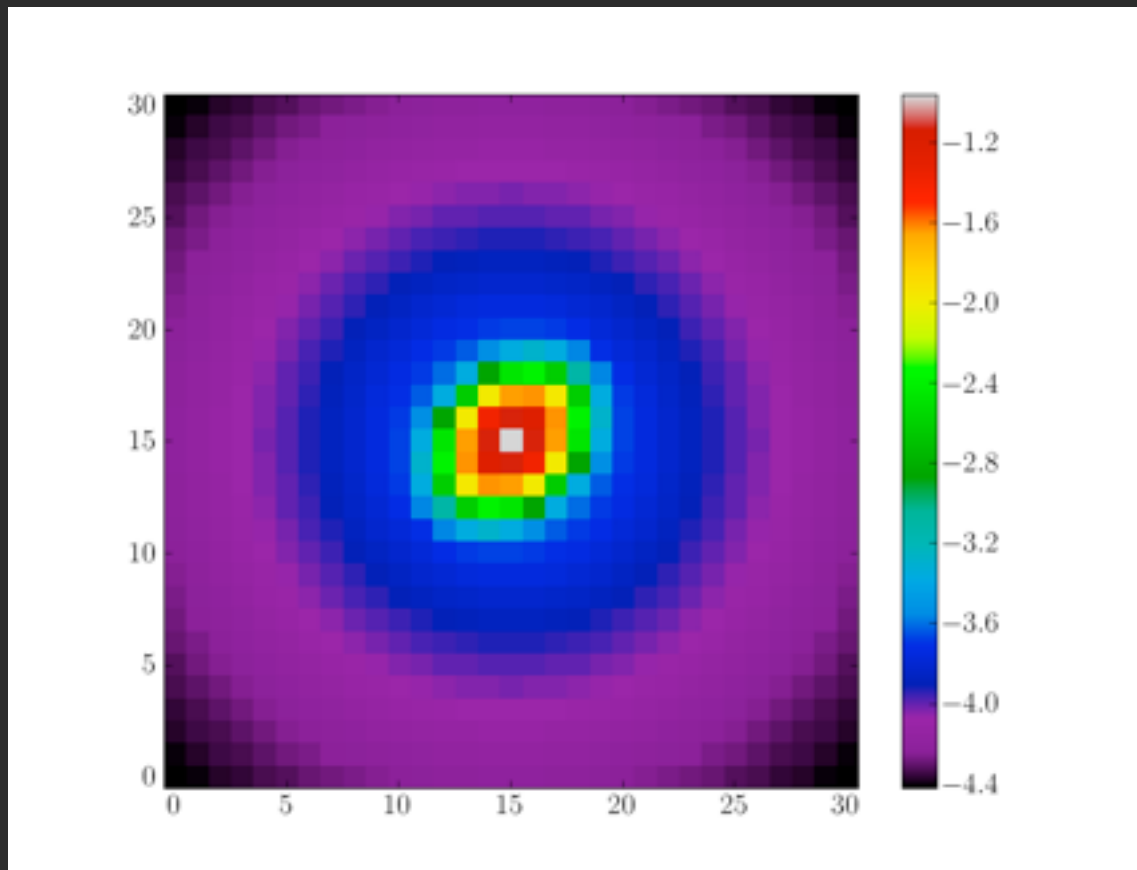


Hewett et al. 1985;  
Horne 1986



- Determine cross-sec'n
- Weighted amplitude fit
- Call that your spectrum

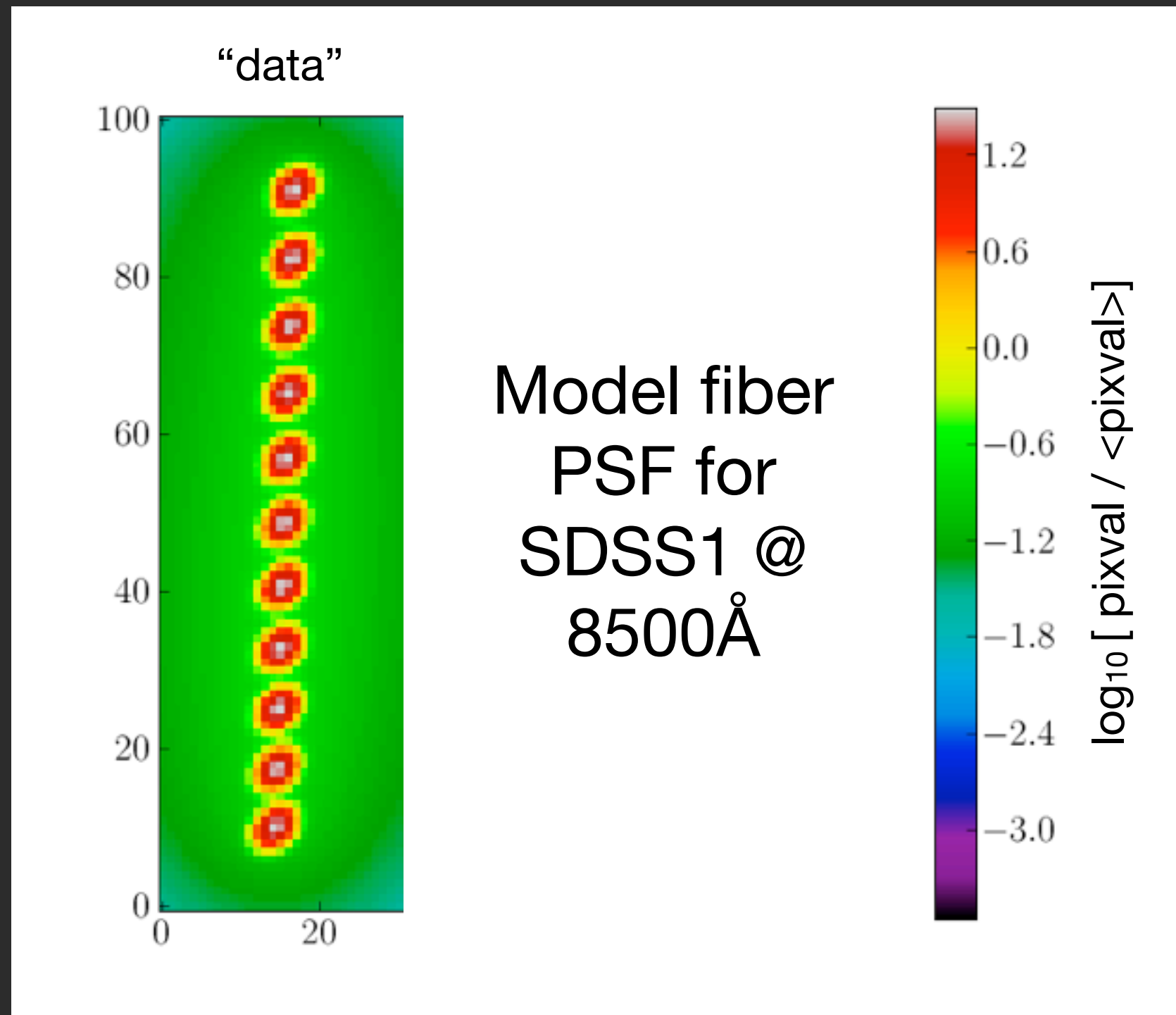
# How do you extract an emission line?



Row-by-row optimal extraction can only be correct when the spectrograph PSF is a *separable* function of  $x$  and  $y$ .

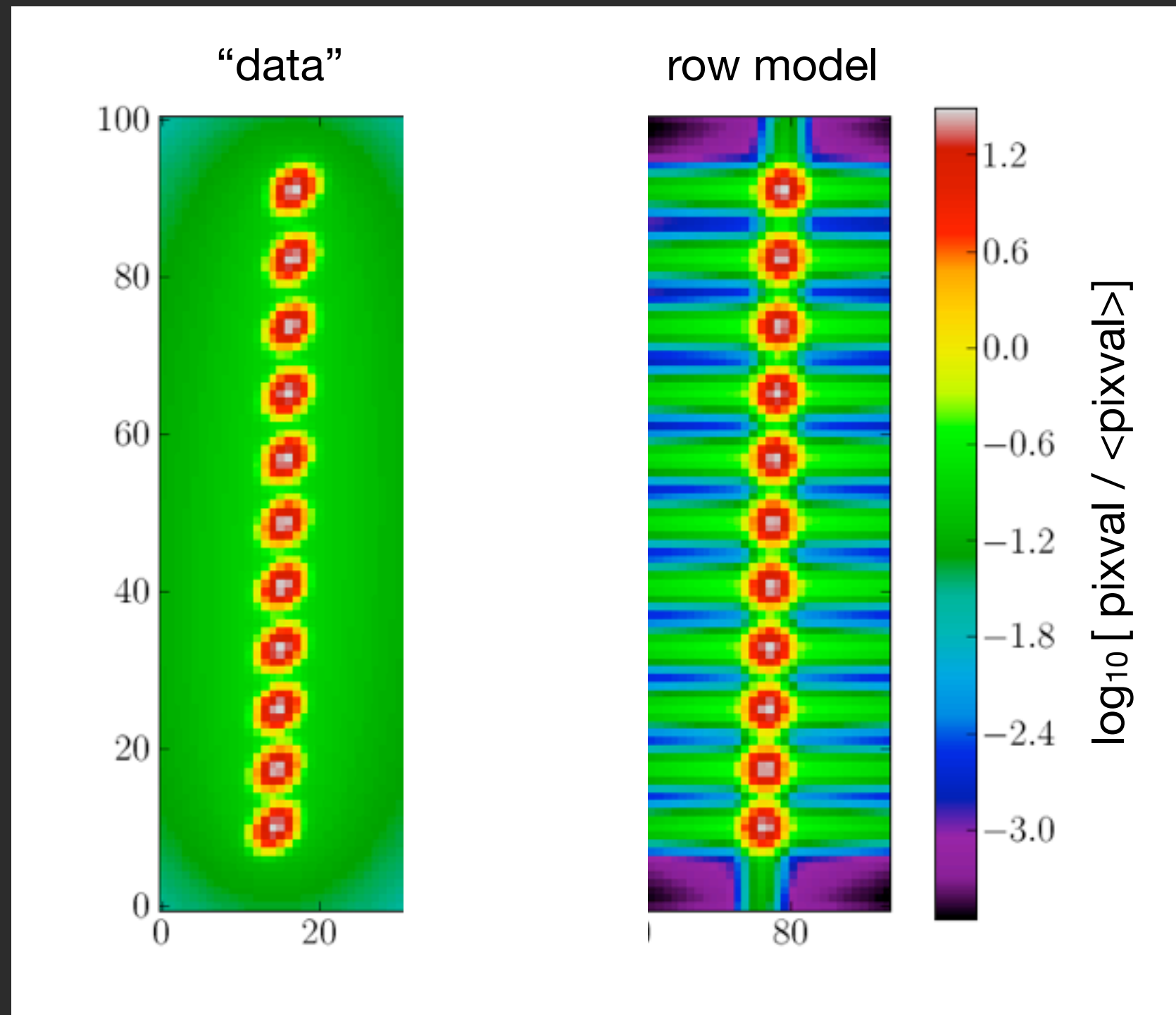
***2D PSF extraction correct for arbitrary PSF shape.***

# Extraction as image modeling



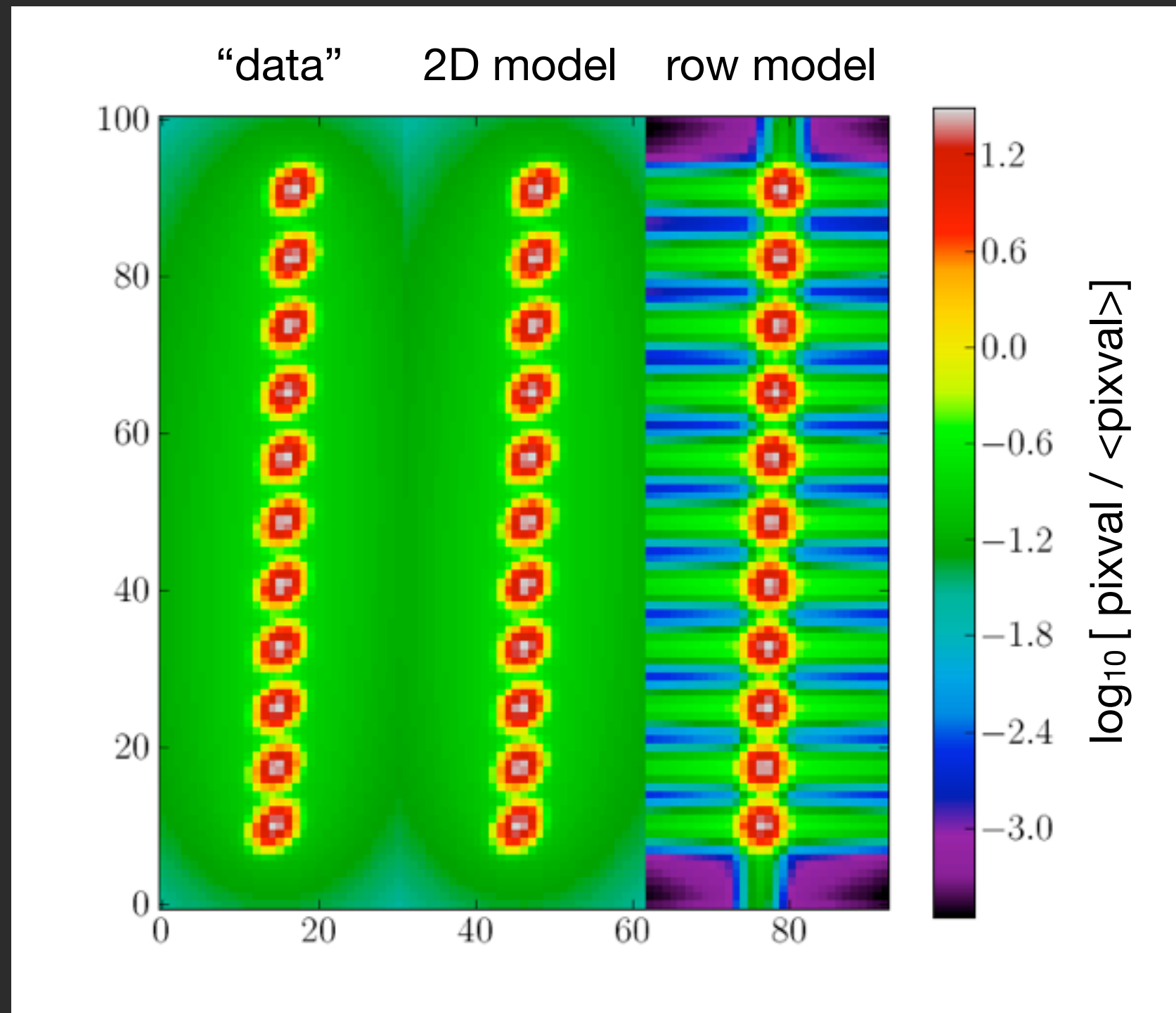


# Extraction as image modeling

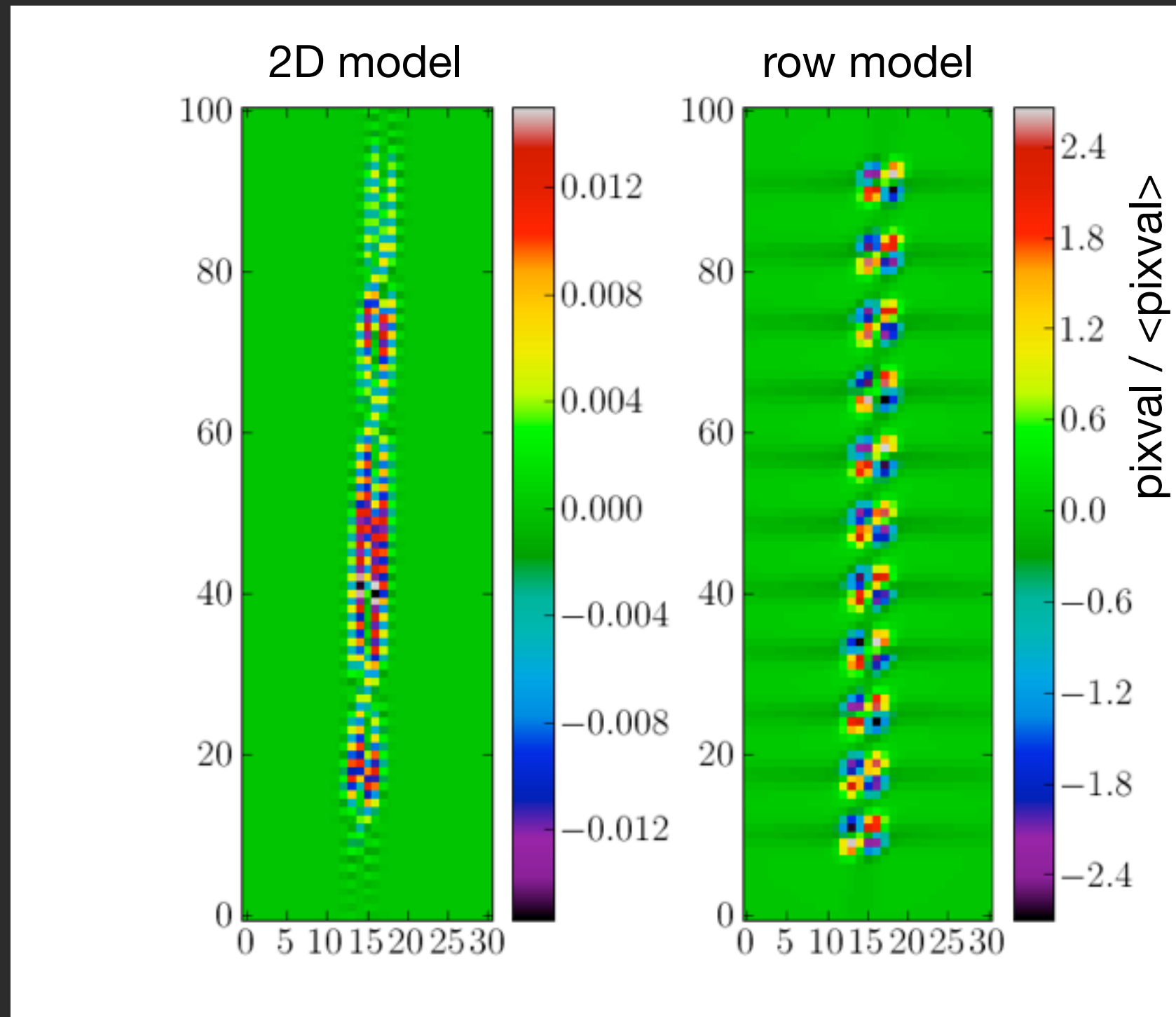




# Extraction as image modeling



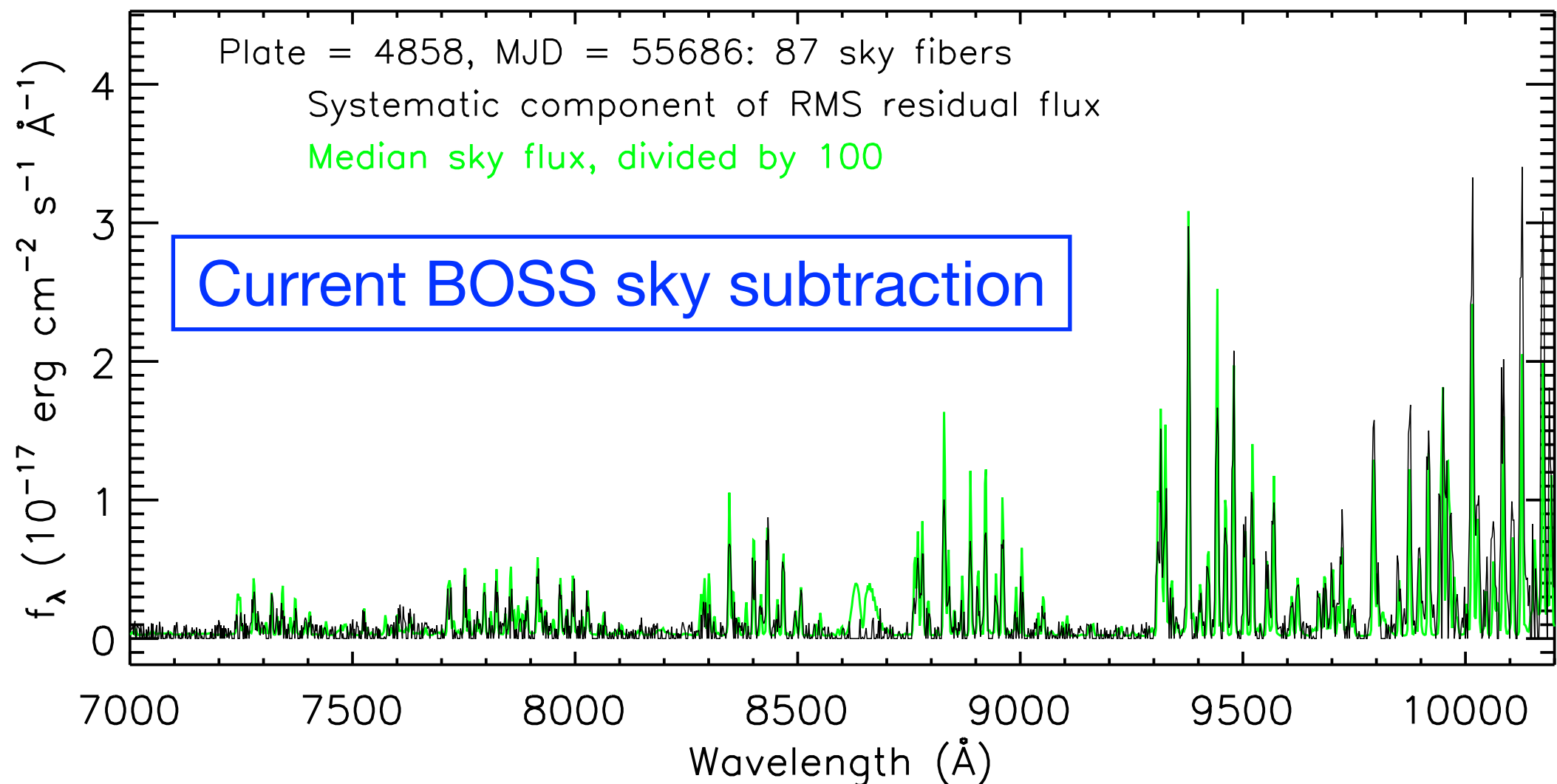
# 2D extraction model residuals



# Why does this matter?

## 1) Poisson-limited sky subtraction

⇒ Current and future faint-galaxy redshift surveys  
(E.g., BOSS, BigBOSS -- esp. [OII] ELG sample, ...)



# Why does this matter?

## 1) Poisson-limited sky subtraction

=> Current and future faint-galaxy redshift surveys  
(E.g., BOSS, BigBOSS -- esp. [OII] ELG sample, ...)

## 2) Extraction as lossless compression

=> All high-precision spectroscopic science  
(Up to and including, e.g., RV planet surveys?)



# What is a spectrum, anyway?

Not just

$\mathbf{f}$  = extracted spectrum vector

but also

$\mathbf{R}$  = band-diagonal line-spread function matrix  
and

$\mathbf{C}$  = spectrum covariance matrix

Together, these encode the likelihood of a given input spectrum model  $\mathbf{m}$  via:

$$\chi^2 (\mathbf{m} \mid \text{data}) = (\mathbf{f} - \mathbf{R} \mathbf{m})^T \mathbf{C}^{-1} (\mathbf{f} - \mathbf{R} \mathbf{m})$$

# How do we do this?

*Projection of input spectrum to CCD pixel frame of raw data via “calibration matrix” **A***

$$\text{(CCD pixel counts)} = \mathbf{A} \text{ (input spectrum counts)} + \text{(noise)}$$

(That is,  $A_{jk}$  = predicted counts in pixel “j” from monochromatic input at wavelength “k”).

*Generalizes and incorporates:*

- Trace solution
- Wavelength solution
- 2D spectrograph PSF and its variation (i.e., aberrations)
- Relative and absolute throughput variation
- CCD pixel sensitivity variations
- Etc.

# How do we do this?

*Projection of input spectrum to CCD pixel frame of raw data via “calibration matrix”  $A$*

$$\text{(CCD pixel counts)} = A \text{ (input spectrum counts)} + \text{(noise)}$$

(That is,  $A_{jk}$  = predicted counts in pixel “j” from monochromatic input at wavelength “k”.)

Likelihood of any model spectrum  $m$  then encoded by:

$$\chi^2 (m | p) = (p - A m)^T N^{-1} (p - A m)$$

*This is forward-modeling to the raw pixels.*

# How do we do this?

“De-convolved” minimum- $\chi^2$  spectrum solution would be

$$m = (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} (\mathbf{A}^T \mathbf{N}^{-1}) \mathbf{p}$$



# How do we do this?

“De-convolved” minimum- $\chi^2$  spectrum solution would be

$$m = (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} (\mathbf{A}^T \mathbf{N}^{-1}) p$$

Now define resolution  $R$  and covariance  $C$  via:

$$(\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A}) = \boxed{\mathbf{Q} \mathbf{Q}} = (\mathbf{R}^T \boxed{\mathbf{C}^{-1}} \mathbf{R})$$

diagonal

Symmetric matrix root

# How do we do this?

“De-convolved” minimum- $\chi^2$  spectrum solution would be

$$m = (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} (\mathbf{A}^T \mathbf{N}^{-1}) \mathbf{p}$$

Now define resolution  $\mathbf{R}$  and covariance  $\mathbf{C}$  via:

$$(\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A}) = \boxed{\mathbf{Q} \mathbf{Q}} = (\mathbf{R}^T \boxed{\mathbf{C}^{-1}} \mathbf{R})$$

diagonal

Symmetric matrix root

And define extracted spectrum as:

$$\mathbf{f} = \mathbf{R} (\mathbf{A}^T \mathbf{N}^{-1} \mathbf{A})^{-1} (\mathbf{A}^T \mathbf{N}^{-1}) \mathbf{p}$$

(Like a “re-convolution” of the de-convolved solution)

# How do we do this?

Likelihood of any model spectrum  $m$  encoded by

$$\chi^2 (m | f) = (f - R m)^T C^{-1} (f - R m)$$

is then mathematically equivalent to

$$\chi^2 (m | p) = (p - A m)^T N^{-1} (p - A m)$$

(up to a constant offset)

*Forward-modeling to a spectrum extracted in this manner is information-equivalent to forward-modeling to the raw CCD pixels.*

# What is extraction?

Calibration:

Likelihood functional determination

Extraction:

Likelihood functional compression

Measurement:

Likelihood functional projection



# Summary of 2D PSF extraction

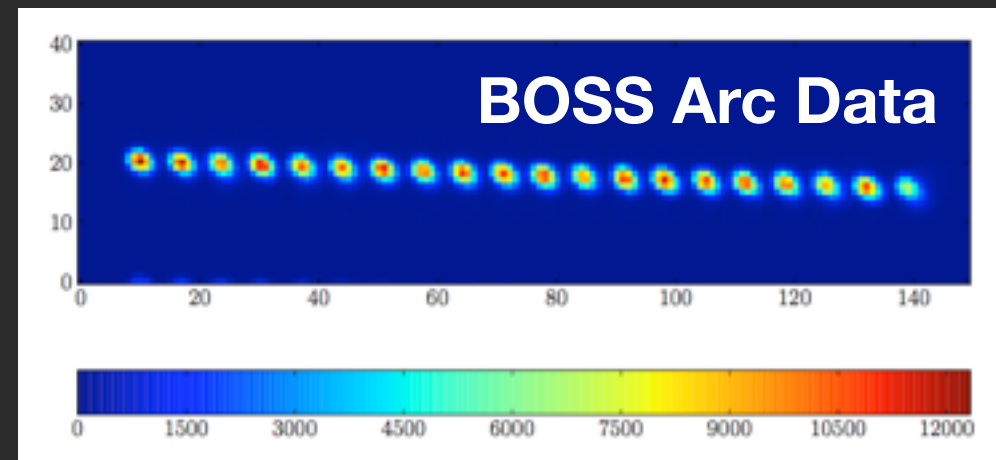
## Major Advantages:

- Extraction as lossless compression
- Mathematically correct even for non-separable PSF
- Incorporates explicit model of 2D data
- Poisson-limited sky subtraction
- Data products “look & feel like spectra”

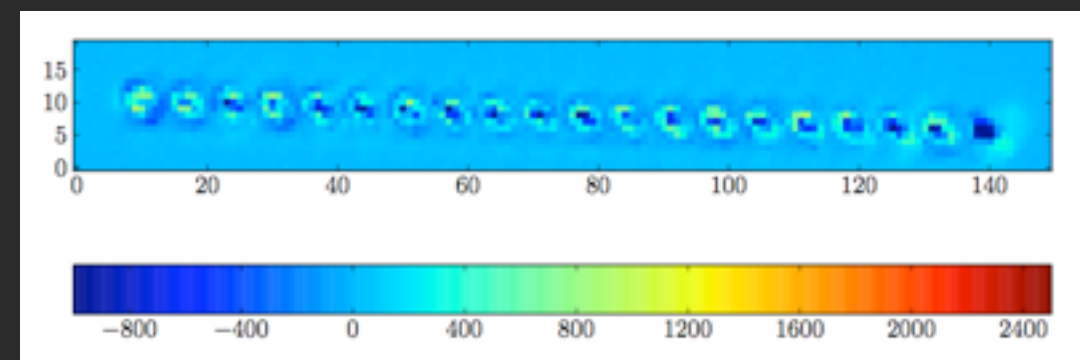
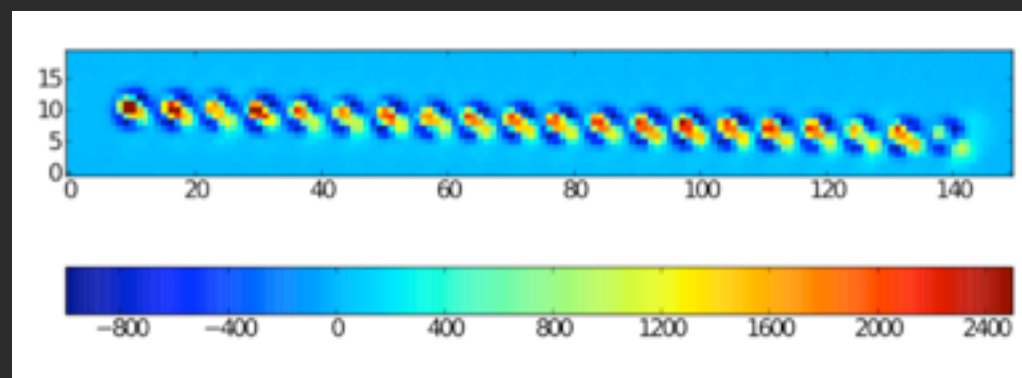
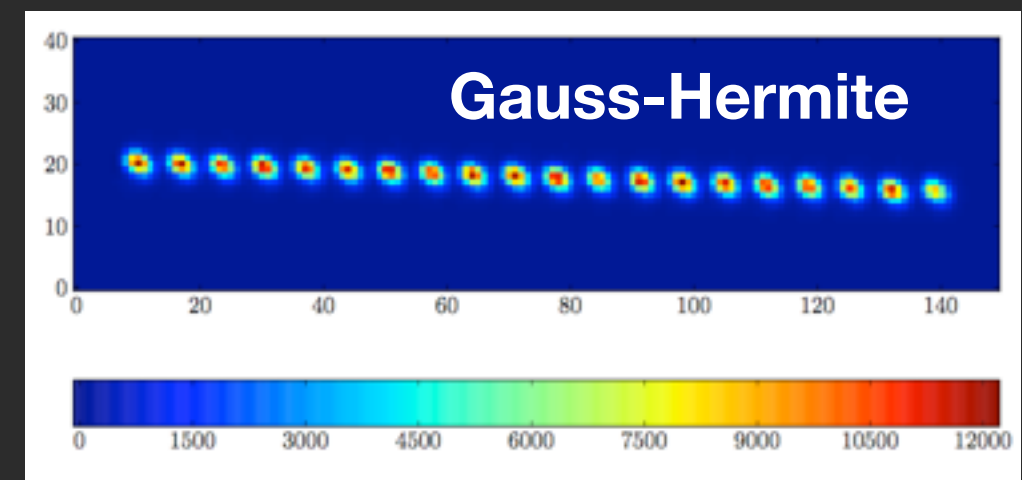
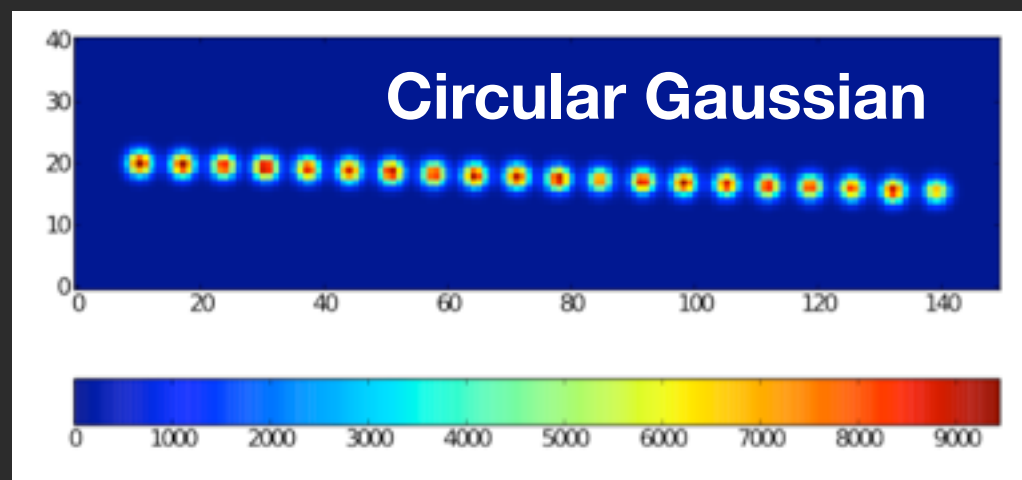
## Major Challenges:

- Extraction coupled across wavelengths
- Requires exquisite calibration
- Some subtlety related to flux normalization

# Development & Implementation Status



Images from  
Parul Pandey  
M.S. Thesis  
U. of Utah

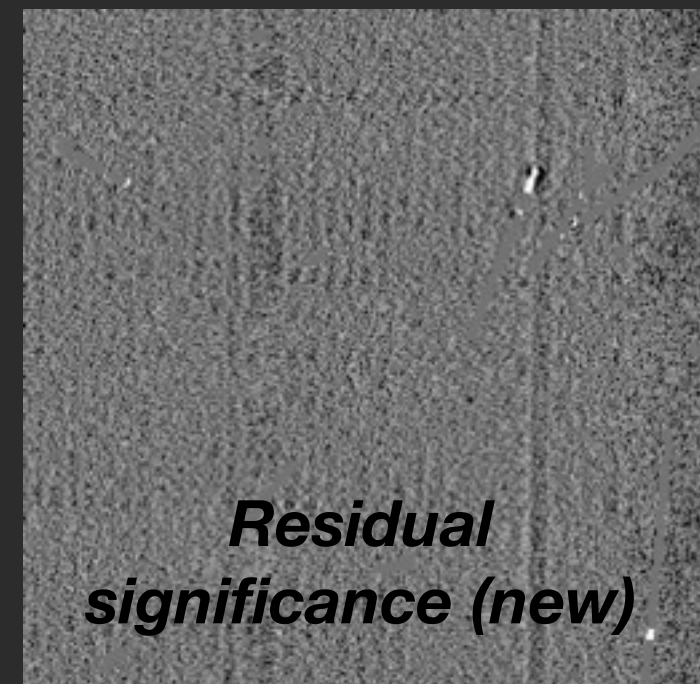
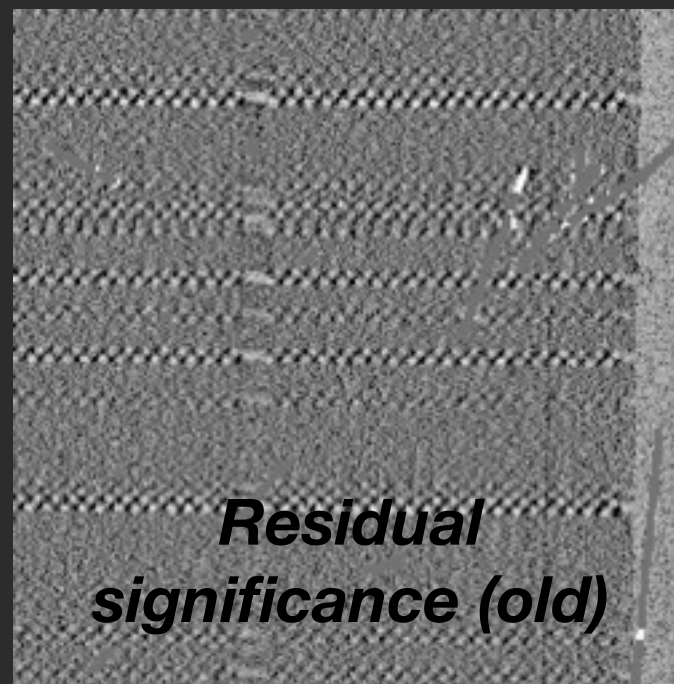
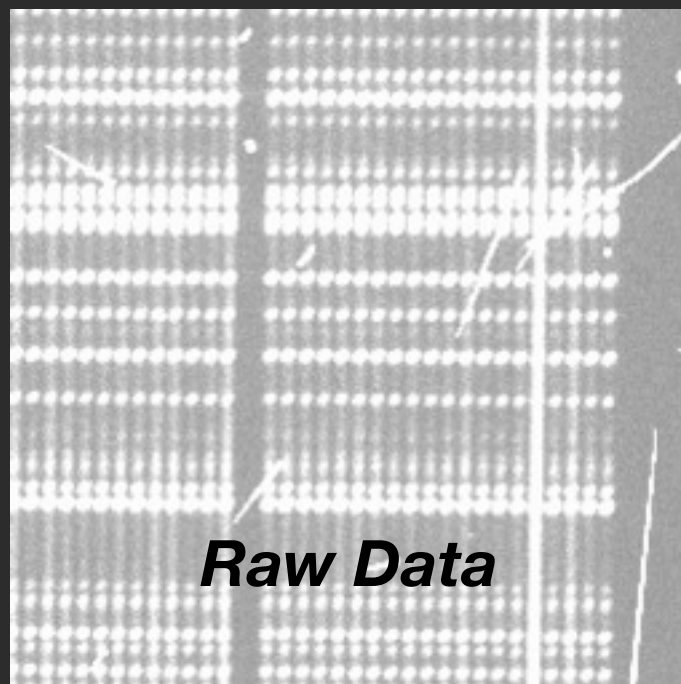


*Also: wing component, higher order GH, pixelized PSF*

# Development & Implementation Status

Demonstrated path for computational tractability:

- Decompose among bundles, exposures, spectrographs, and wavelength ranges



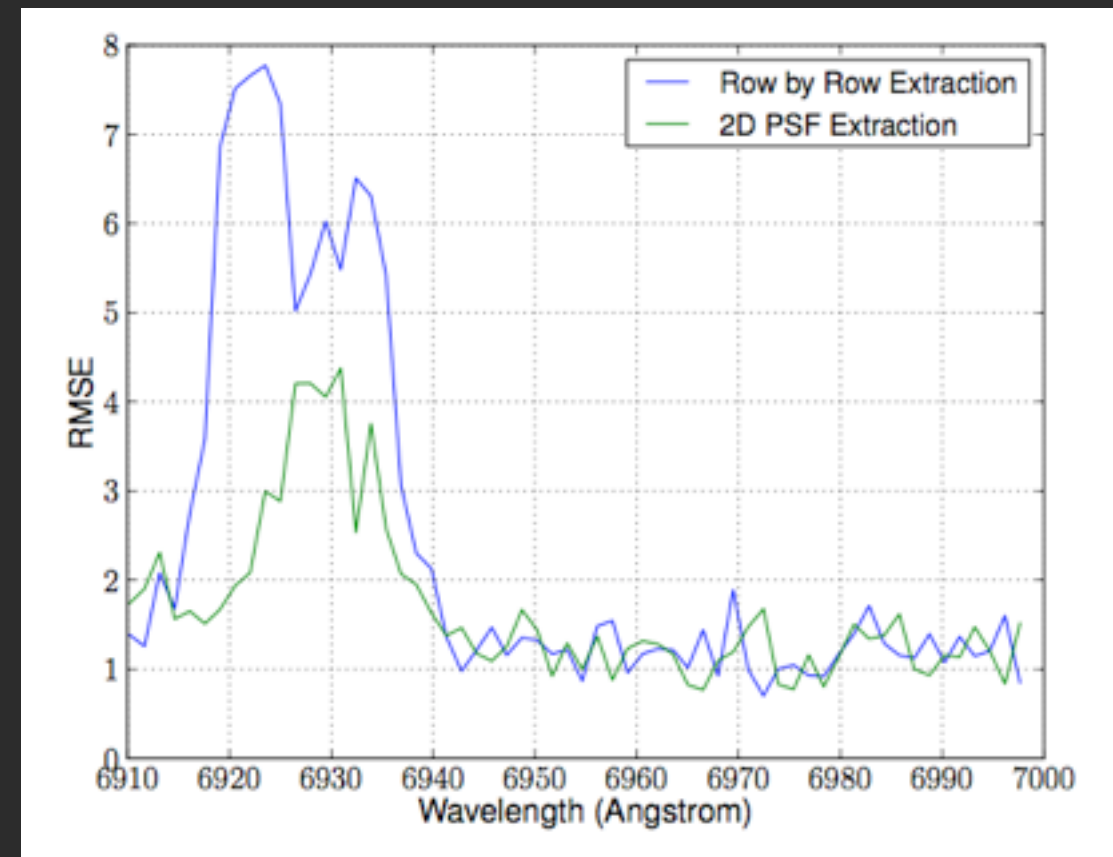
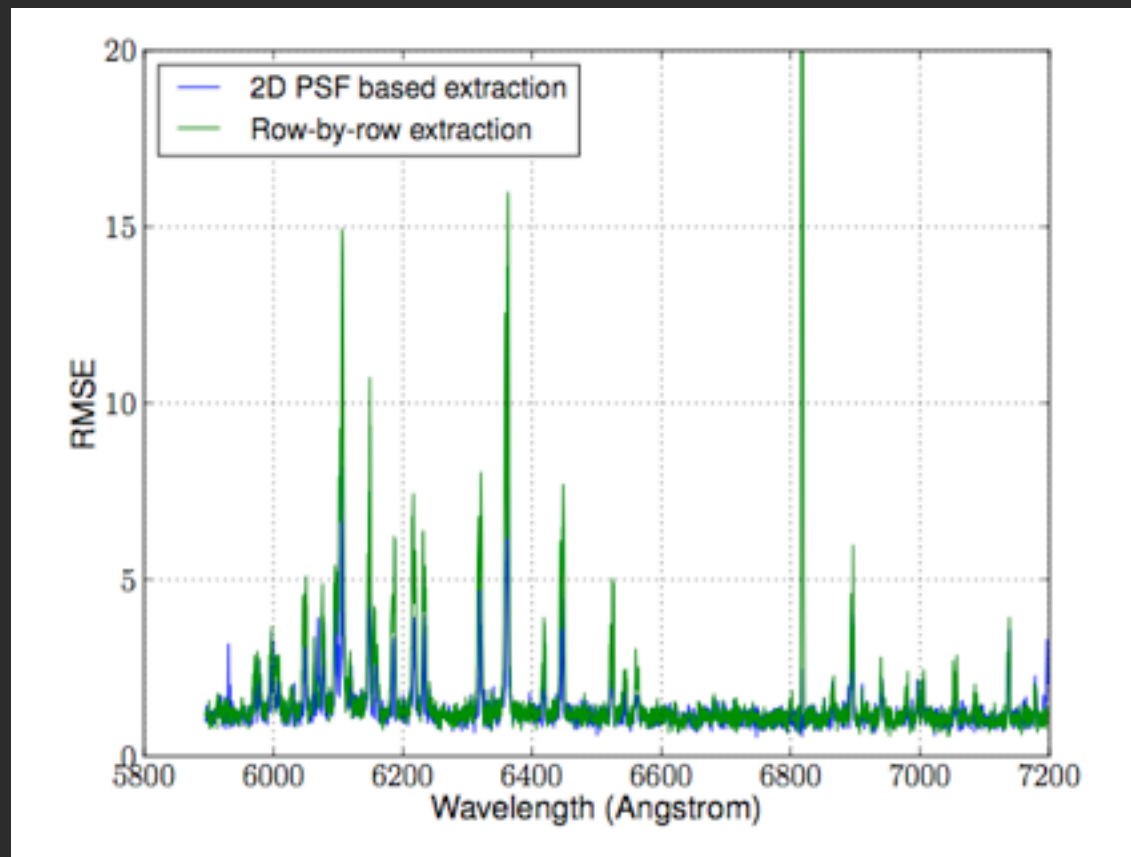
Effort in summer 2011 and ongoing by:

ASB, Joel Brownstein, Parul Pandey (U. of Utah)

Stephen Bailey, Ted Kisner, David Schlegel (LBNL)

# Benefits in extracted-spectrum frame

*Sky subtraction, as simulated by arc-lamp data*



Images from Parul Pandey, M.S. Thesis 2011, U. of Utah



# Software Requirements on Hardware

**Separability:** we absolutely need gaps between bundles of fibers where cross-talk goes to zero

**True resolution:** metric is not camera spot EE or flux-weighted  $r^2$ , but *wavelength autocorrelation of PSF:*

$$\left[ \int p(x,y;\lambda) (x,y;\lambda+\Delta\lambda) dx dy \right] / \left[ \int p^2(x,y;\lambda) dx dy \right]$$

(N.B.: Rayleigh criterion is autocorrelation of 1/4)

**Calibration:** tunable monochromatic system for mapping out system calibration matrix?

**Stability:** fractional spectrum bias for assuming wrong PSF  $q(x,y)$  instead of right PSF  $p(x,y)$  is:

$$b = 1 - \left[ \int p(x,y) q(x,y) dx dy \right] / \left[ \int p^2(x,y) dx dy \right]$$

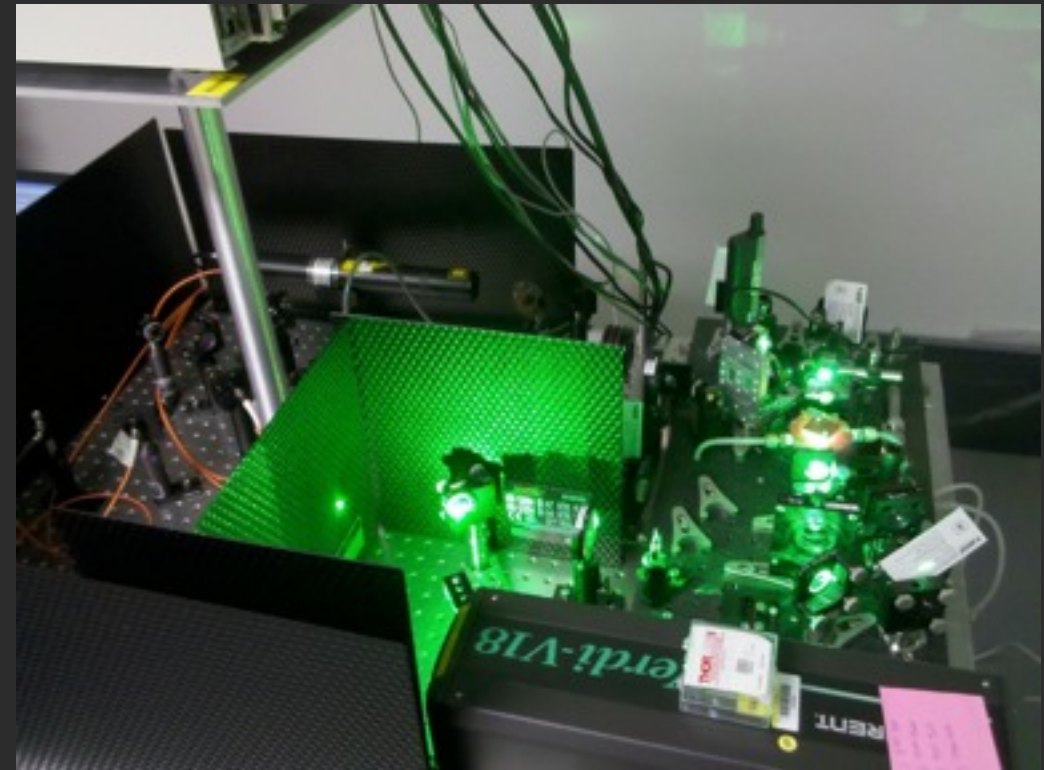
# Software Requirements on Hardware

Ultimately calls for a full integration of data analysis software with instrumental design software

- => Optimize *scientific* metrics in hardware design
- => Tune instrument directly from science CCD data
- => “Use what you know” during analysis

# Monochromatic calibration

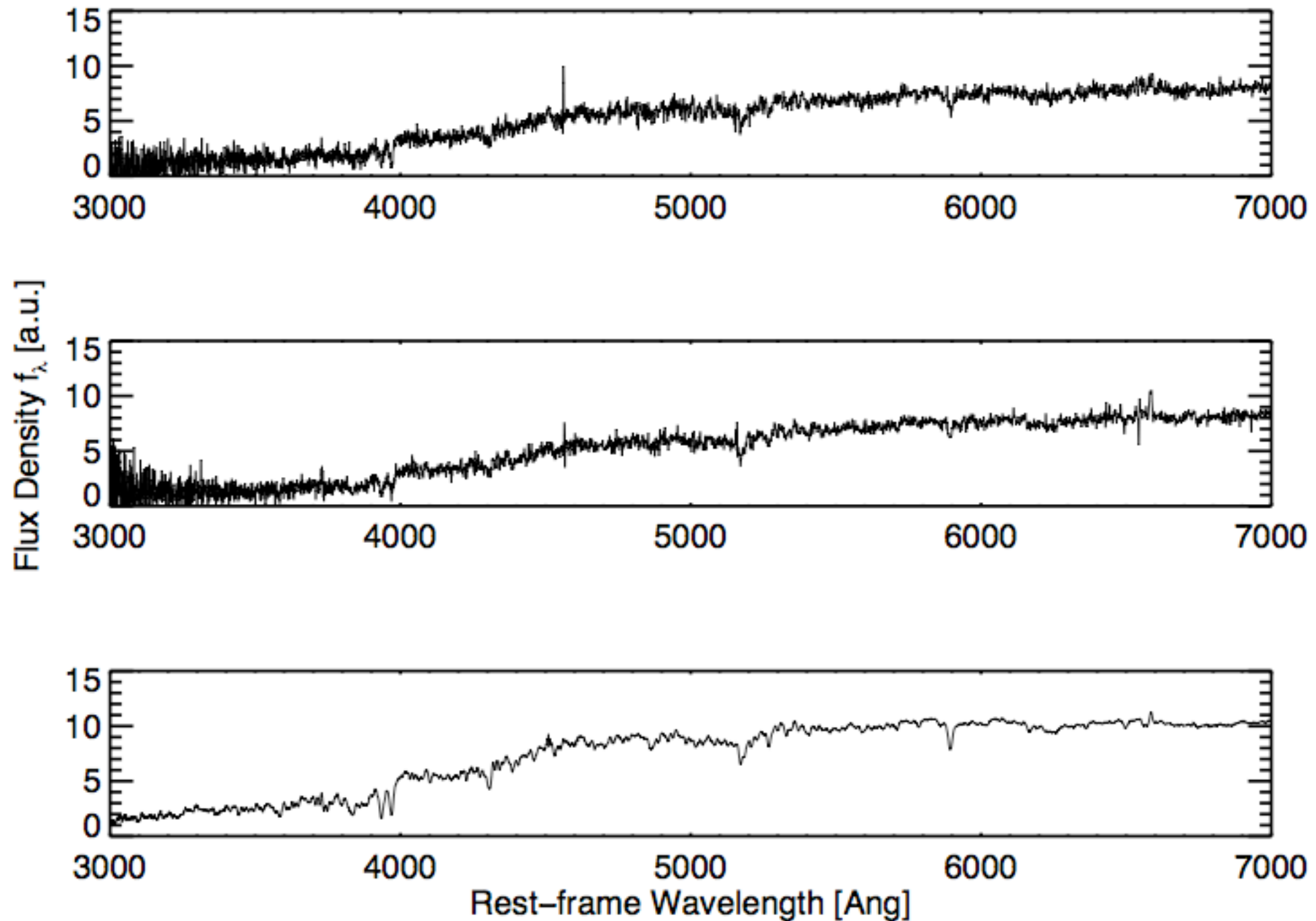
NIST-BOSS tunable laser experiment (w/ C. Cramer, K. Lykke)  
(Also see G. Tarle “Line-O-Matic”)



**VS.**



# Application: Bayesian stacking



Shu, ASB, et al., submitted (arXiv 1109.6678)

# Application: Bayesian stacking

Model vdisp distribution at fixed  $z$  and  $M$  as a log-normal distribution (c.f. Bernardi et al. 2003):

$$p(\log_{10}\sigma|m, s) = \frac{1}{\sqrt{2\pi}s} e^{-\frac{(\log_{10}\sigma - m)^2}{2s^2}} \quad (1)$$

Constrain parameters in  $(z, M)$  bins by integrating over all spectra and all vdisp values:

$$\begin{aligned} \mathcal{L}(m, s|\{\vec{d}\}) &= p(\{\vec{d}\}|m, s) \\ &= \prod_i p(d_i|m, s) \\ &= \prod_i \int d\log_{10}\sigma p(d_i|\log_{10}\sigma) p(\log_{10}\sigma|m, s) \end{aligned} \quad (2)$$

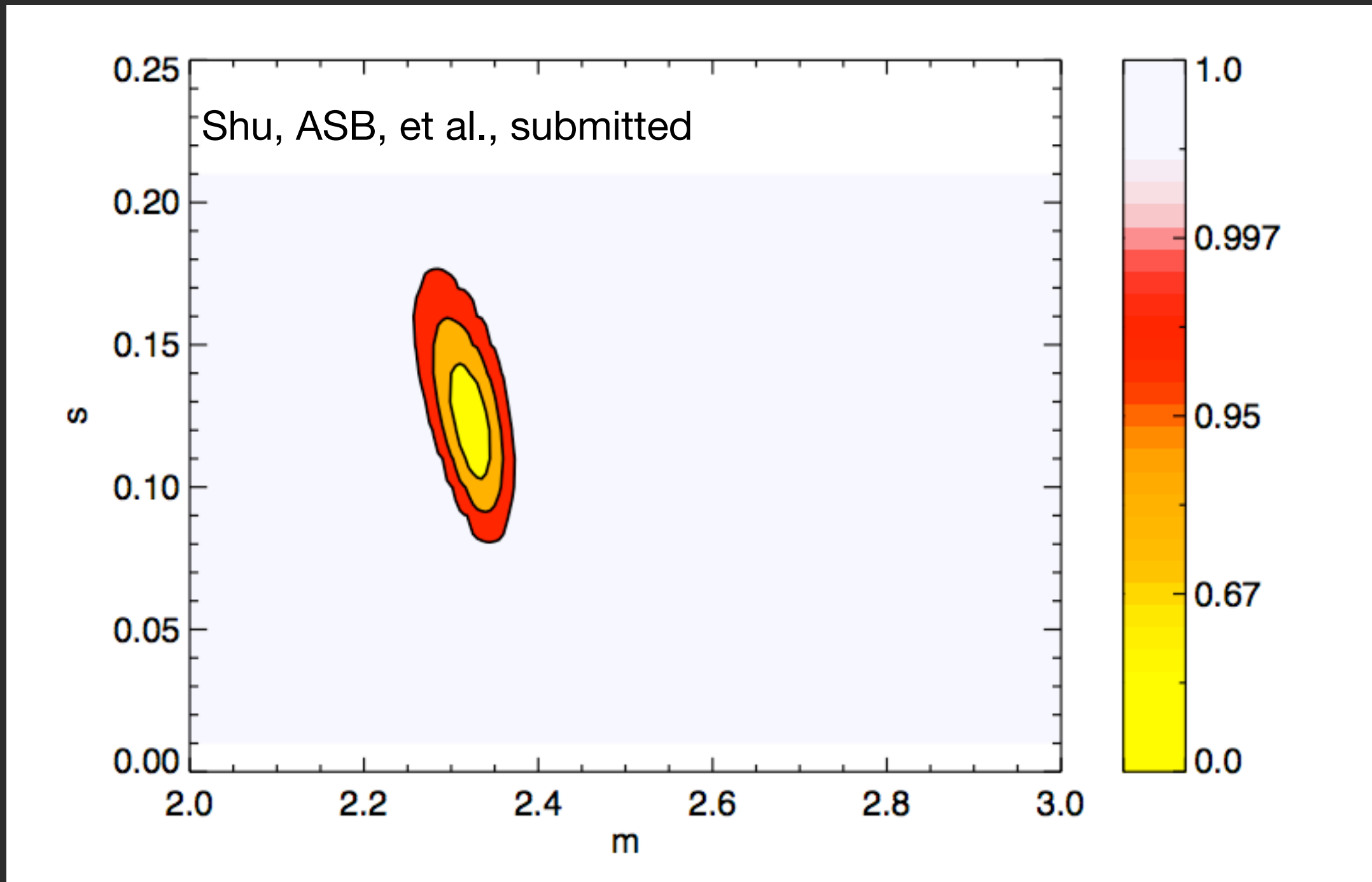
$$p(m, s|\{\vec{d}\}) = \frac{p(\{\vec{d}\}|m, s)p(m, s)}{p(\{\vec{d}\})} \quad (3)$$

Shu, ASB, et al., submitted  
(arXiv 1109.6678)

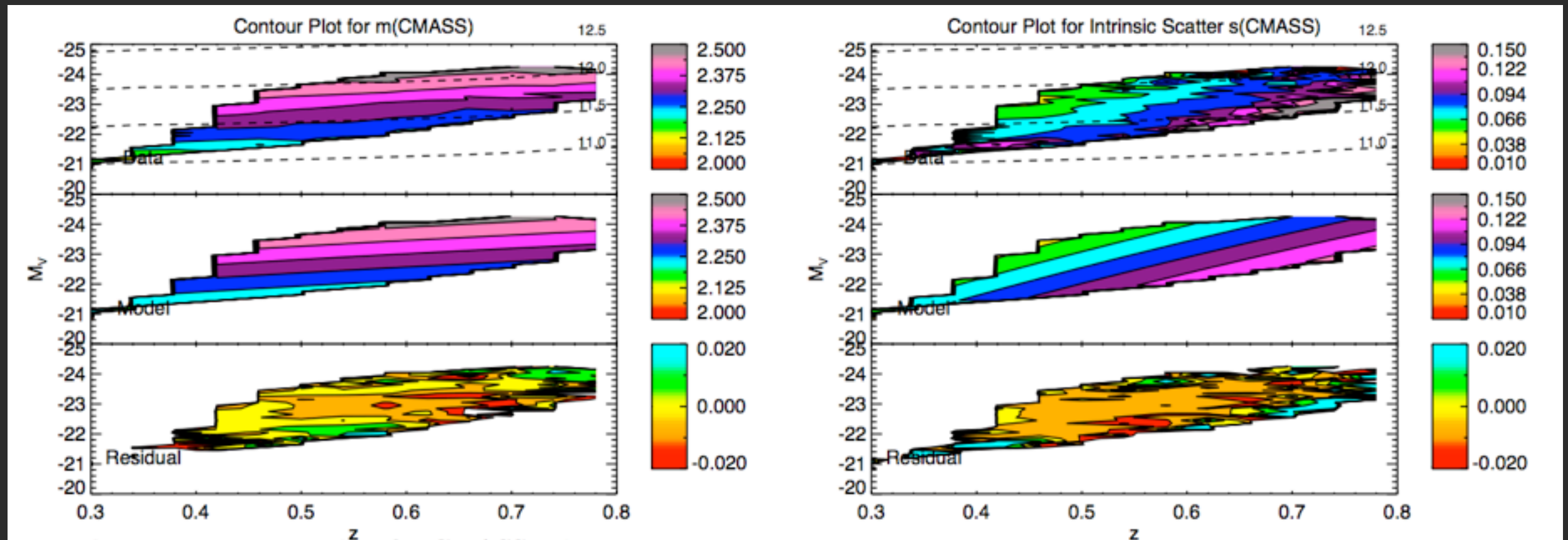
N.B.: if you stack directly, you will measure  $\sigma = 10^{[m + s^2 \ln(10)]}$



# Posterior probability for a single bin



# Distribution results: population evolution



Shu, ASB, et al., submitted

# Summary and conclusions

- Full 2D forward modeling of raw data is the way of the future for spectroscopic extraction
- Poisson-limited sky subtraction for ground-based faint-galaxy redshift surveys (BOSS, BigBOSS)
- Lossless compression of spectrum likelihood functional
- We have the algorithmic framework, and are currently putting it into practice
- Major challenges are in calibration, computation, and integration of data analysis with hardware design

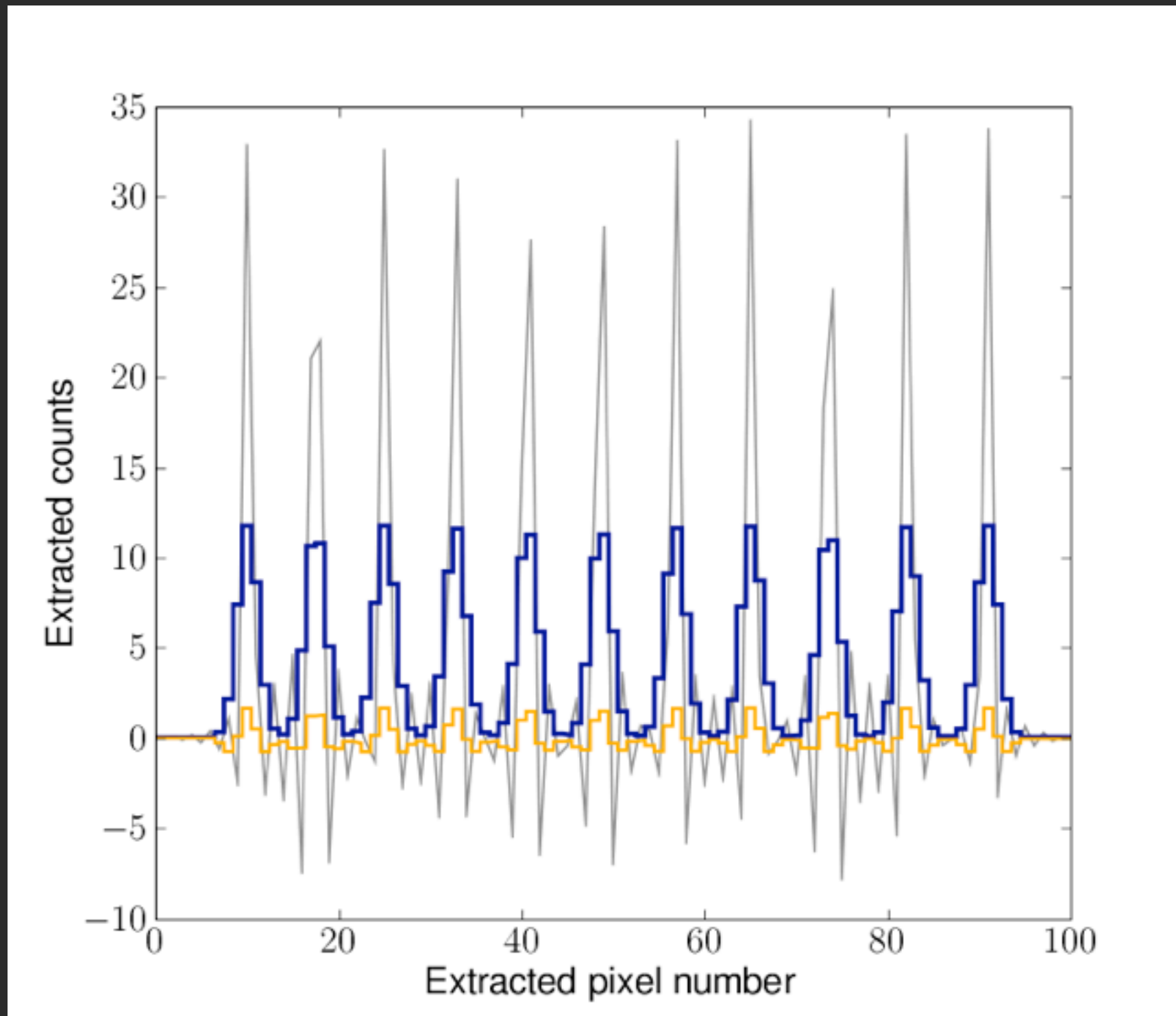




***Thank You!***

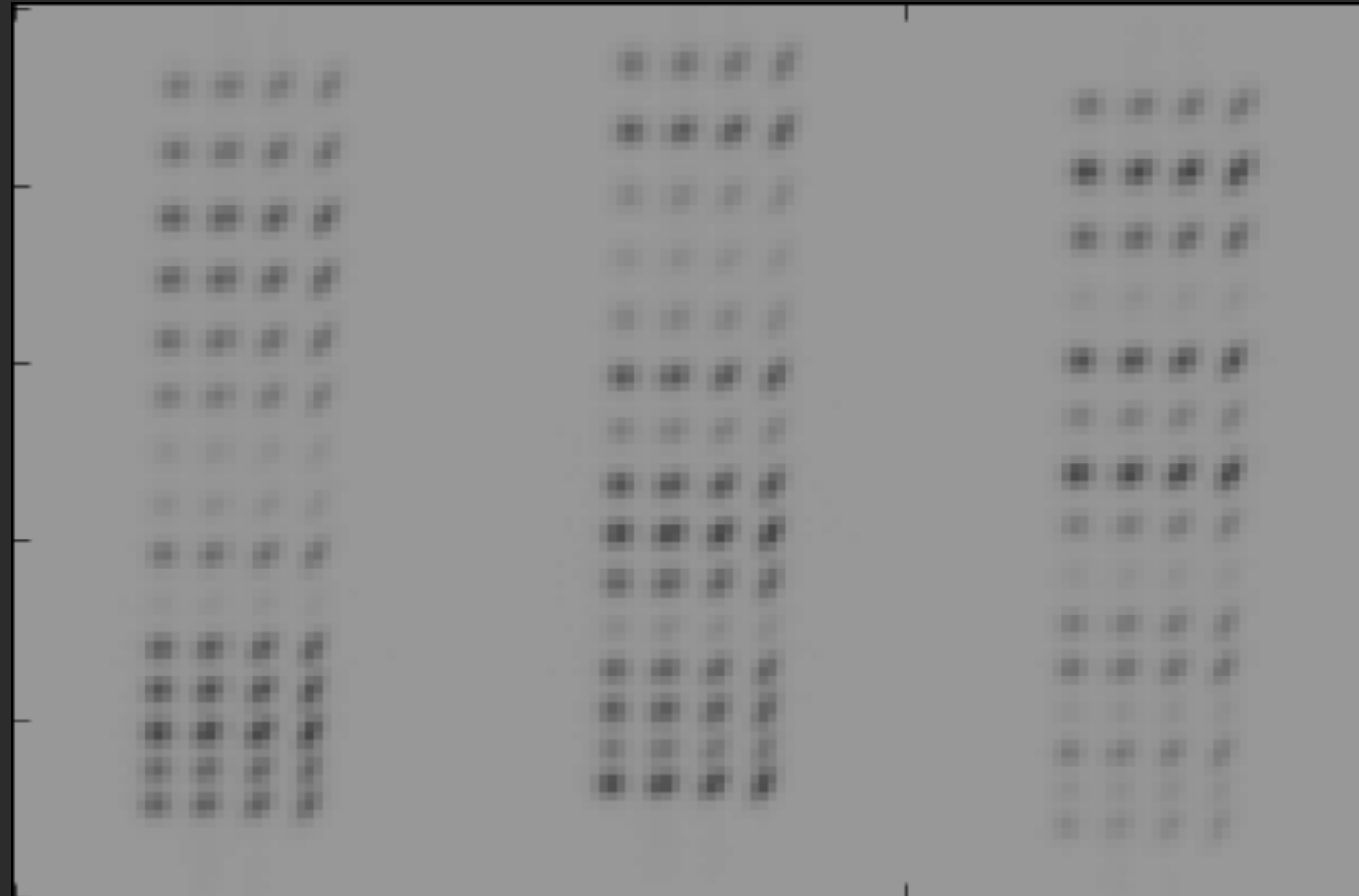


# Deconvolution and reconvolution

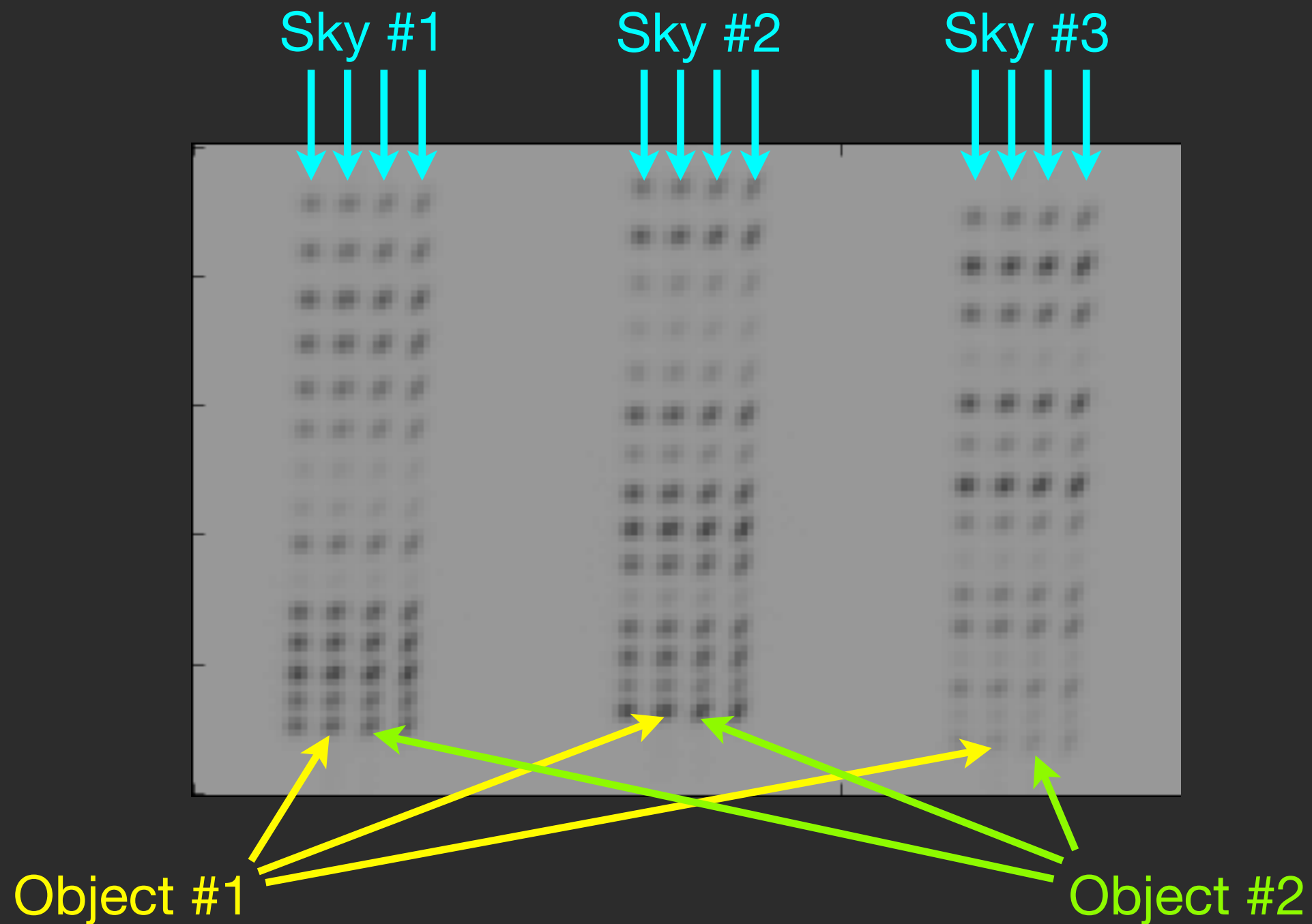




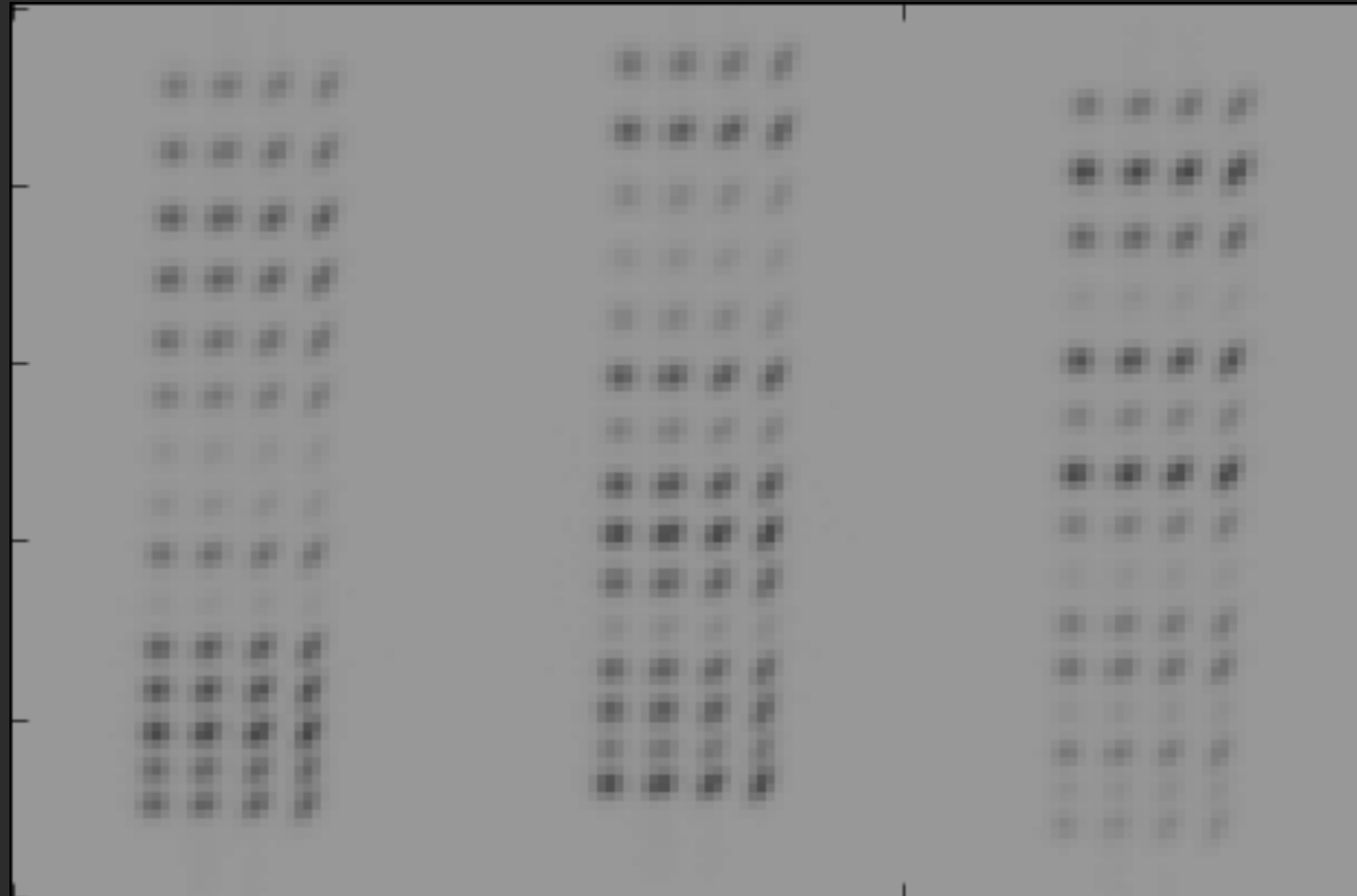
# Multi-frame, multi-fiber simulated data



# Multi-frame, multi-fiber simulated data



# Multi-frame, multi-fiber simulated data

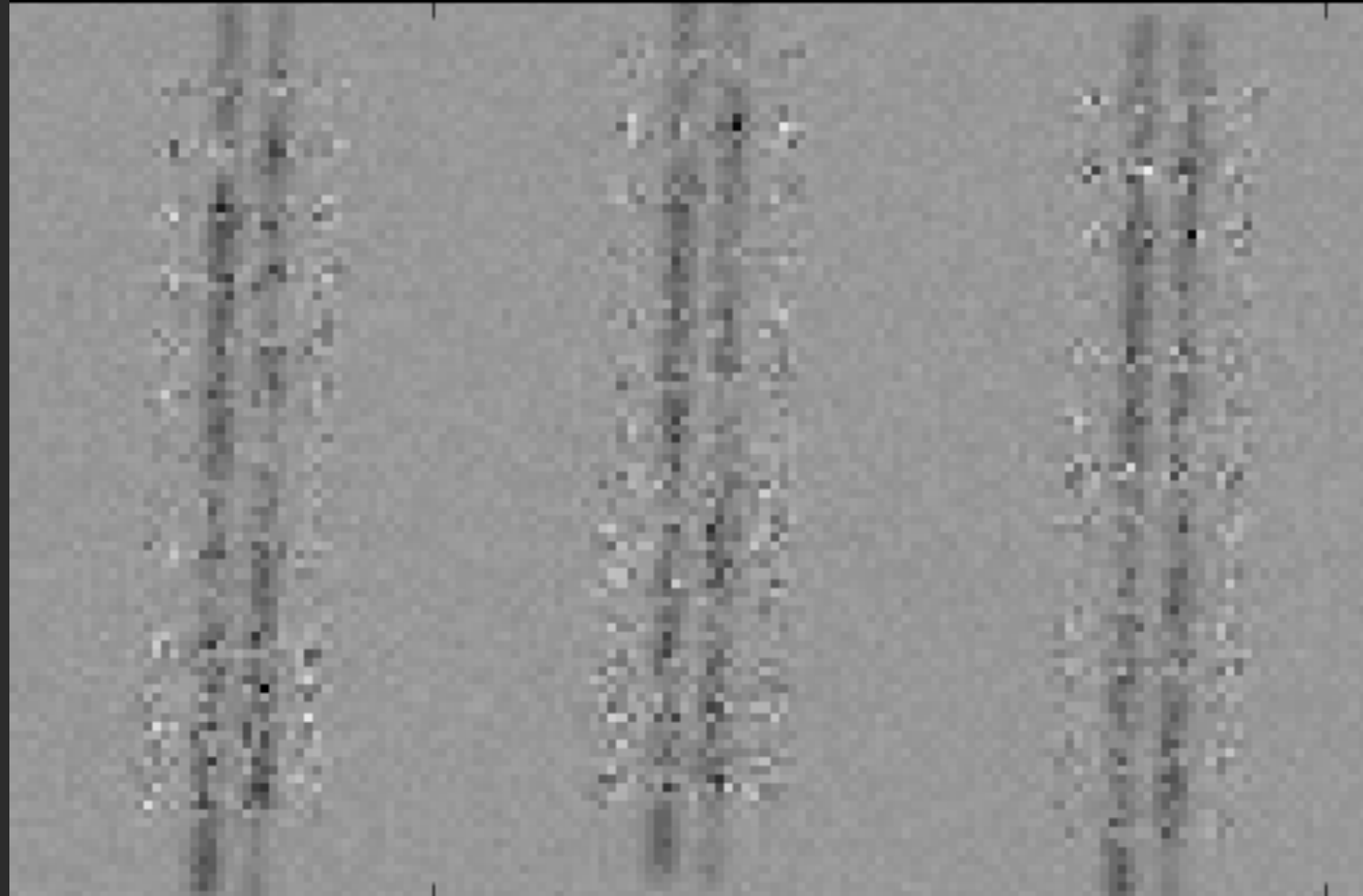


$$\text{Objflux} = \text{Skyflux} / 20$$

ObjSNR  $\approx$  5 (per extracted sample, sky-noise limited)

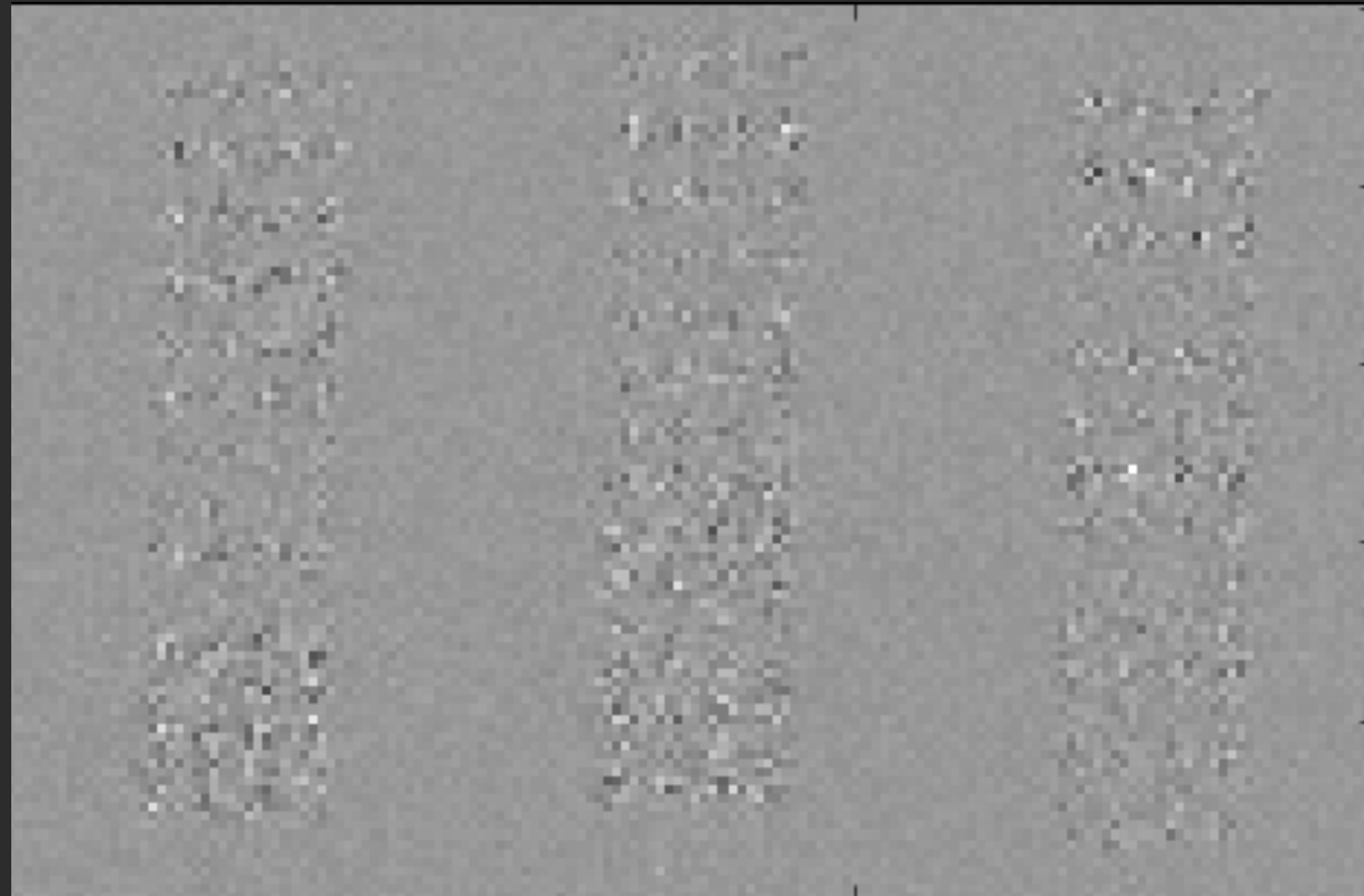
# Sky model decomposed & removed

Sky spectrum is modeled “upstream” from optical heterogeneities between fibers



(Grayscale stretch X 40 relative to previous)

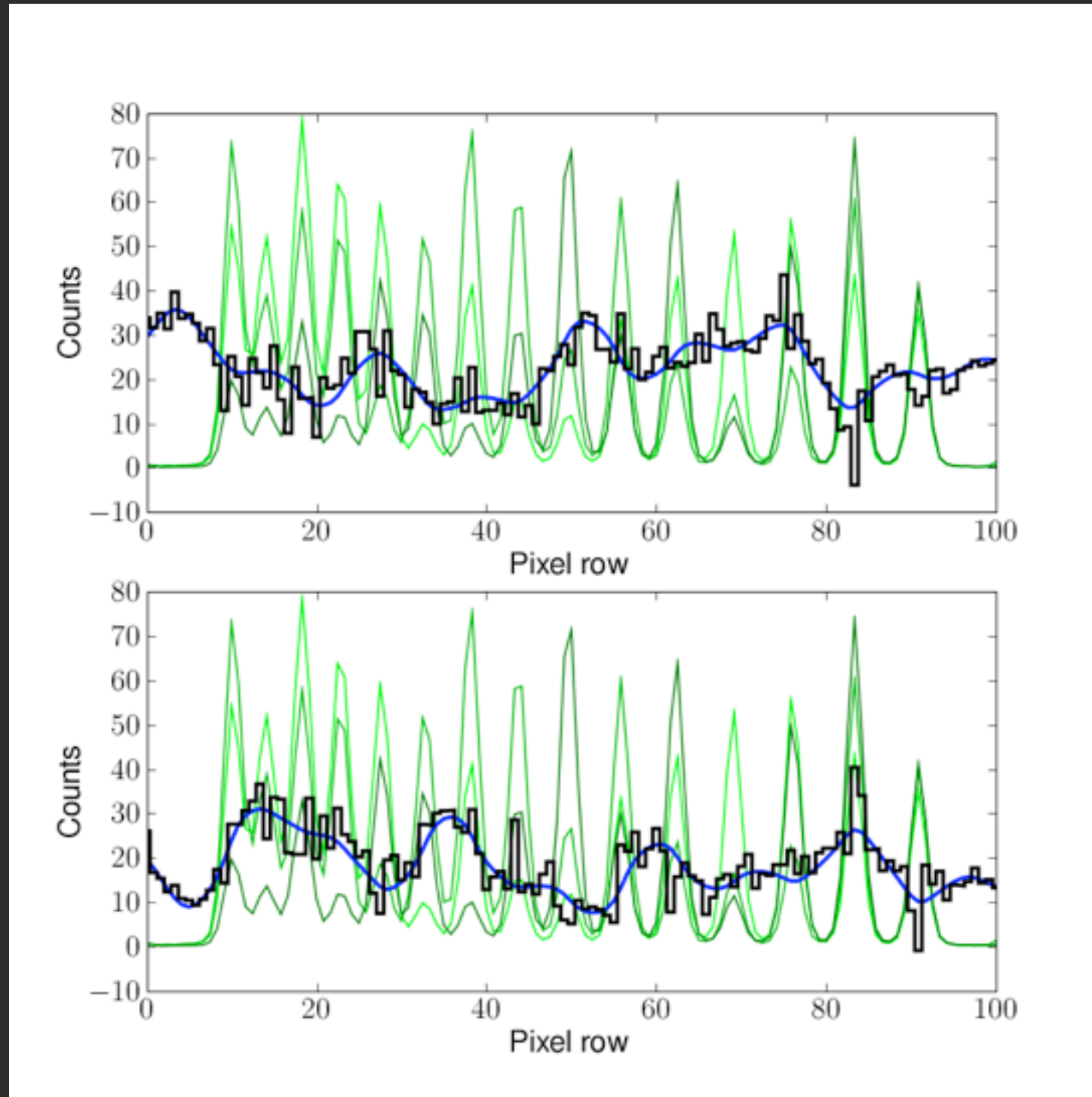
# All models removed



Consistent with pure noise

# Extracted objects + skies

Sky scaled  
down by a  
factor of 20  
in plot



RMS error-  
scaled  
residuals of  
unity