

# Optimal Photometry of Faint Galaxies

Kenneth M. Lanzetta  
Stony Brook University

# Collaborator:

- Stefan Gromoll (Stony Brook University)

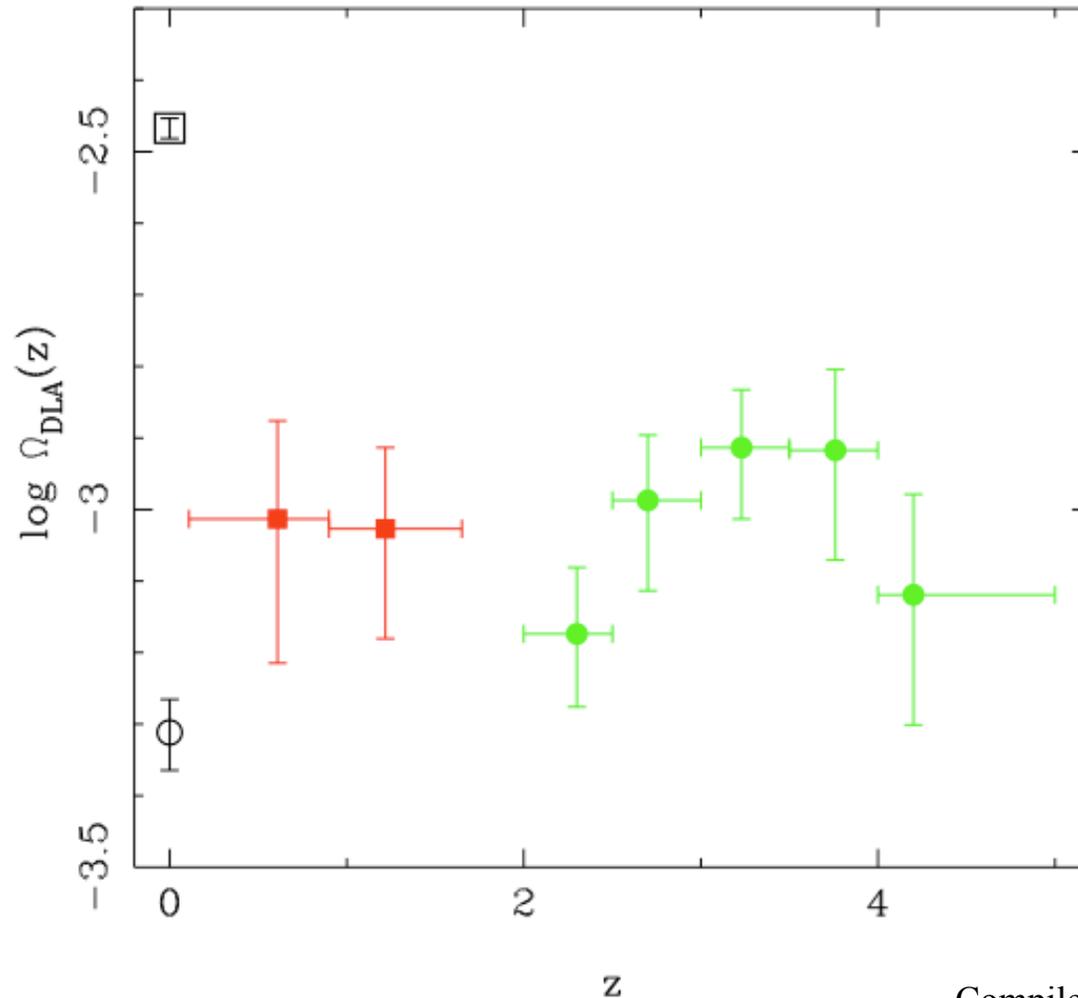
# Outline

- scientific motivation
- data
- photometric redshift technique
- optimal photometry and photometric redshifts of faint galaxies

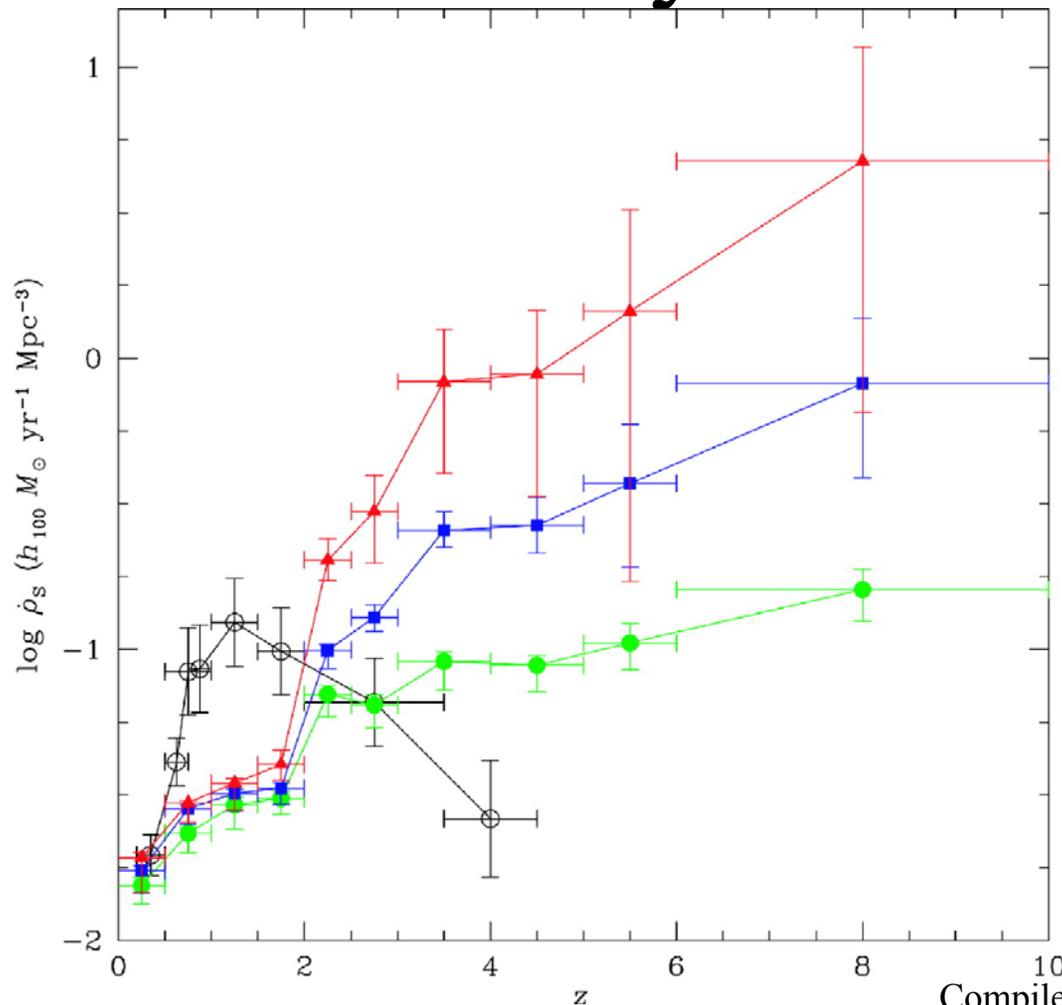
# Cosmic chemical evolution

- We are interested in the quantities of cosmic chemical evolution...
  - $\Omega_g$  (damped Ly $\alpha$  absorbers)
  - $\psi$  (rest-frame ultraviolet, H $\alpha$  emission)
  - $Z$  (damped Ly $\alpha$  absorbers)
  - $\Omega_s$  (rest-frame near-infrared emission)
- ...which are the quantities of galactic chemical evolution averaged over cosmic volumes

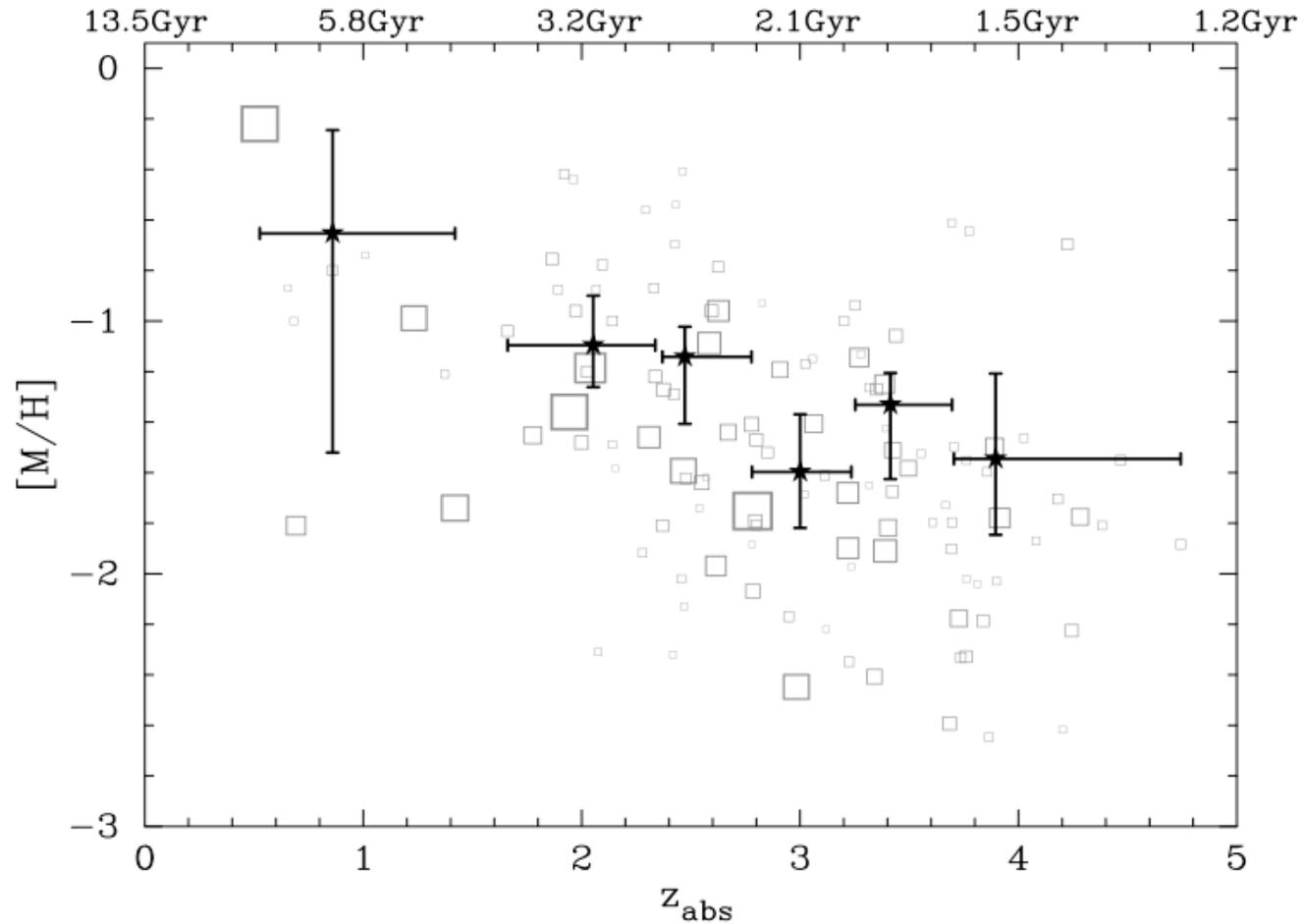
# Comoving mass density of gas



# Comoving star formation rate density



# Cosmic metallicity



# Outstanding issues

- very limited statistics
- cosmic variance
- selection biases
  - damped Ly $\alpha$  absorbers: obscuration by dust of QSOs behind high-column-density absorbers
  - ultraviolet emission: dust extinction, cosmological surface brightness dimming

# Equations of cosmic chemical evolution

$$\frac{d\Omega_s}{dt} = (1 - R)\psi$$

$$\frac{d\Omega_g}{dt} = -(1 - R)\psi + \phi$$

$$\Omega_g \frac{dZ}{dt} = y(1 - R)\psi - \phi Z$$

# Comoving mass density of stars

- existing surveys target very large numbers of galaxies (statistics) across many fields (cosmic variance)
- measurement is based upon rest-frame near-infrared emission (dust extinction)
- *objective: determine the comoving mass density of stars versus cosmic epoch with the accuracy needed to obtain a statistically meaningful time derivative*

# Our program

- measure optimal photometry (at observed-frame near-ultraviolet through mid-infrared wavelengths) and photometric redshifts of faint galaxies in GOODS and SWIRE surveys
- use rest-frame near-infrared luminosities and rest-frame optical and near-infrared colors to estimate stellar mass densities
- construct comoving mass density of stars versus cosmic epoch

# GOODS survey

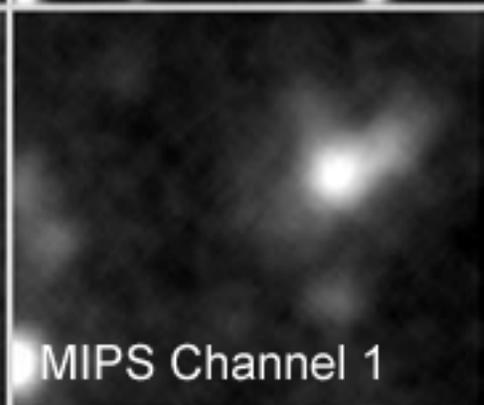
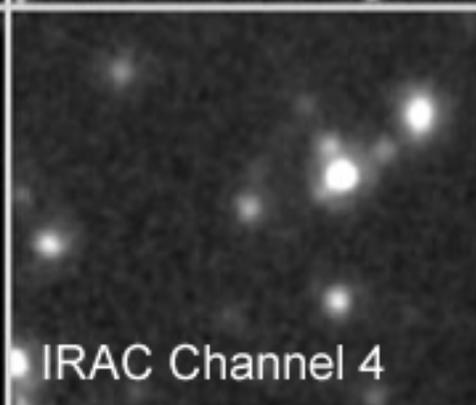
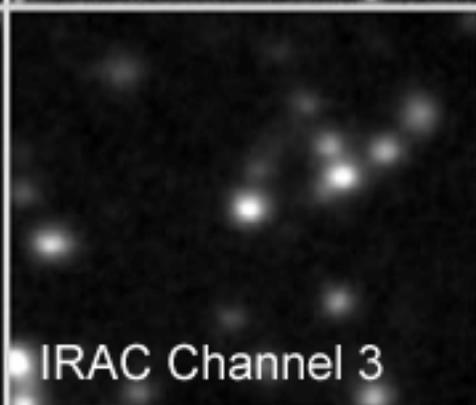
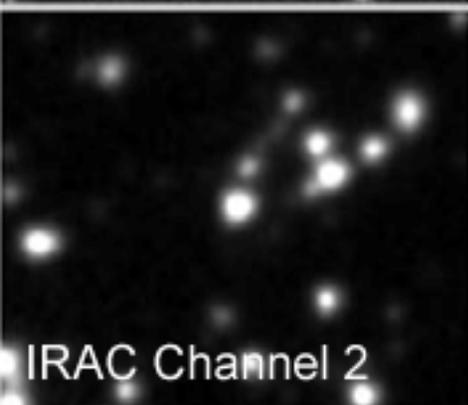
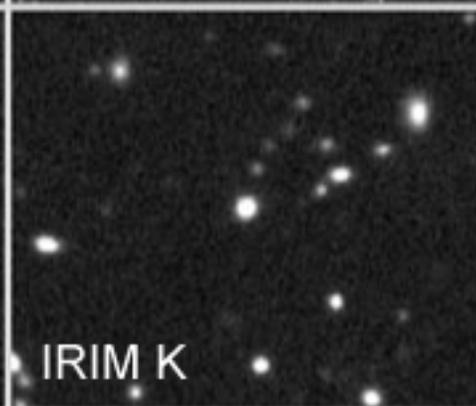
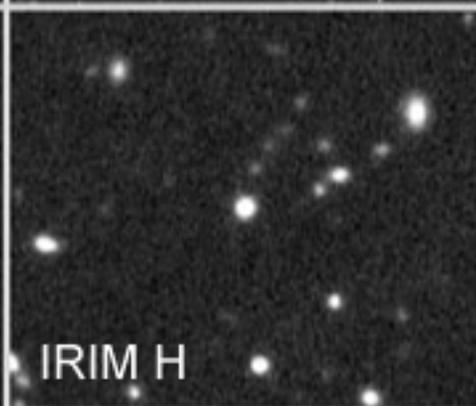
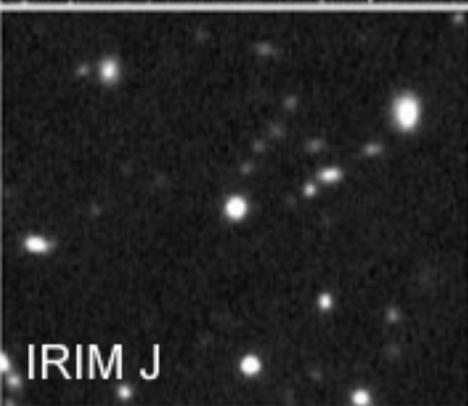
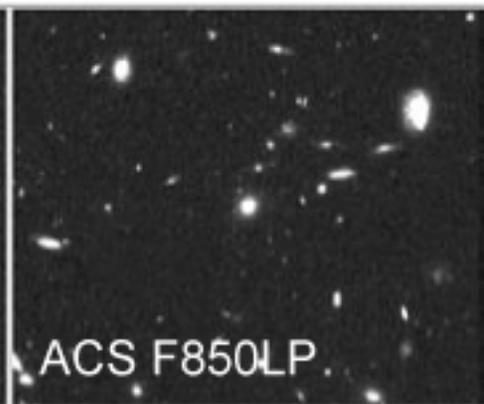
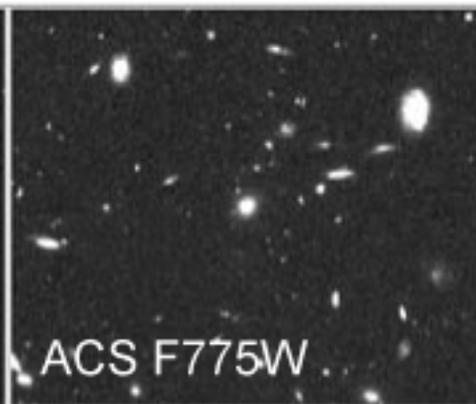
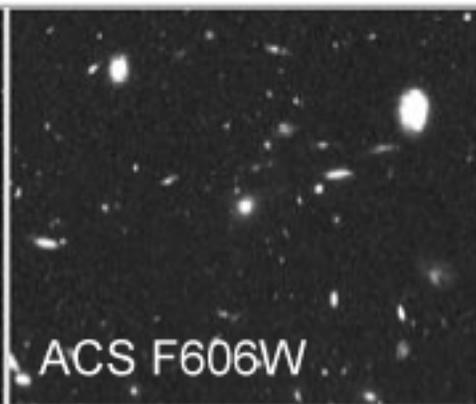
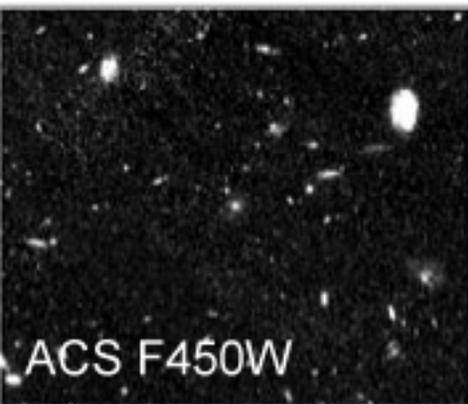
- two fields spanning 320 arcmin<sup>2</sup>
- Spitzer IRAC images at 3.6, 4.5, 5.8, and 8.0  $\mu\text{m}$  and MIPS images at 24  $\mu\text{m}$
- HST and ground-based images at observed-frame optical and near-infrared wavelengths
- roughly 10,000 IRAC images and 10,000 MIPS images
- roughly 200,000 galaxies at  $z \approx 0 - 6$

# SWIRE survey

- six fields spanning 49 deg<sup>2</sup>
- Spitzer IRAC images at 3.6, 4.5, 5.8, and 8.0  $\mu\text{m}$  and MIPS images at 24, 70, and 160  $\mu\text{m}$
- ground-based images at observed-frame optical wavelengths
- roughly 100,000 IRAC images and 500,000 MIPS images
- roughly 8,000,000 galaxies at  $z \approx 0 - 2$

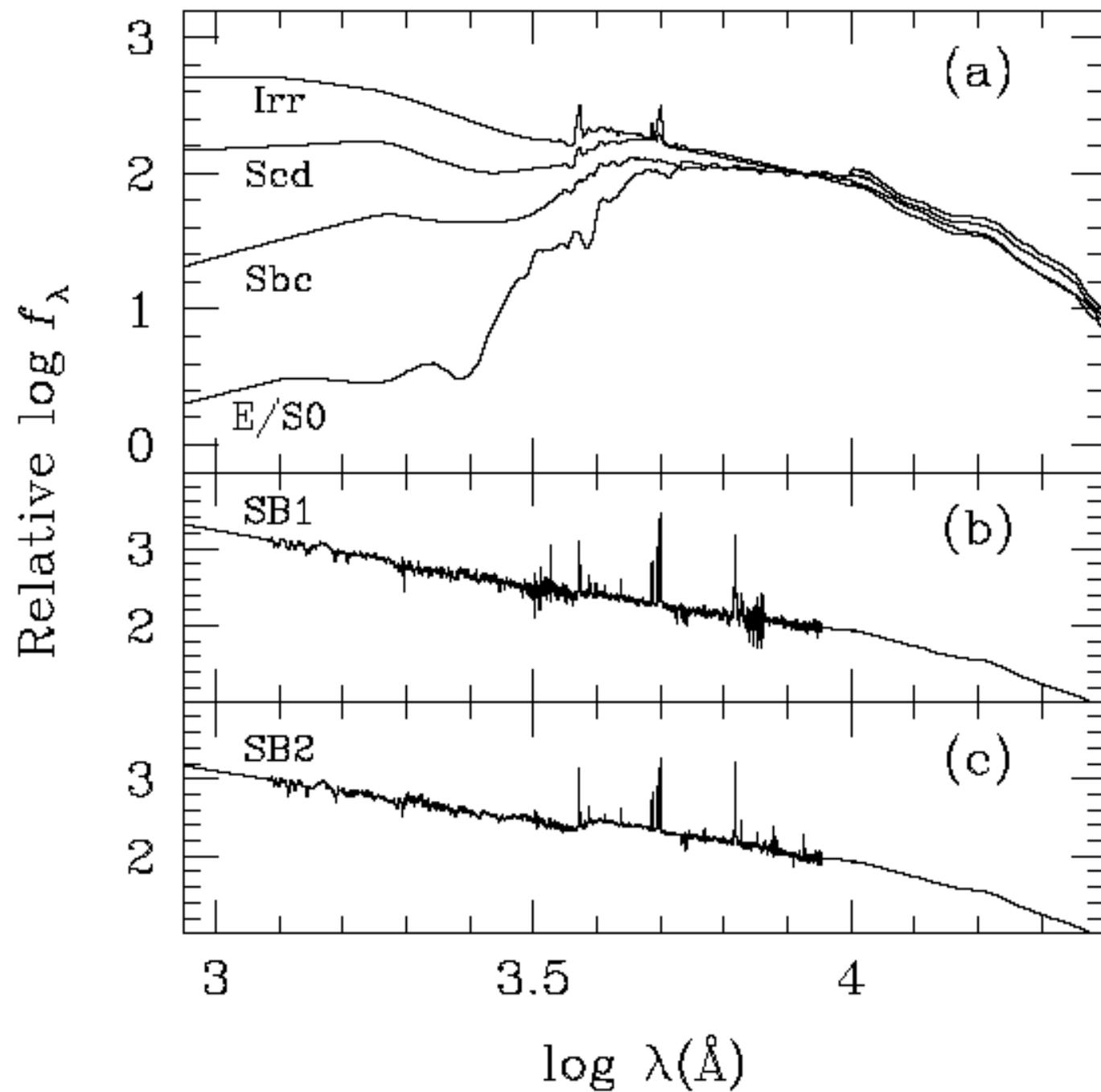
# Why the measurement is difficult

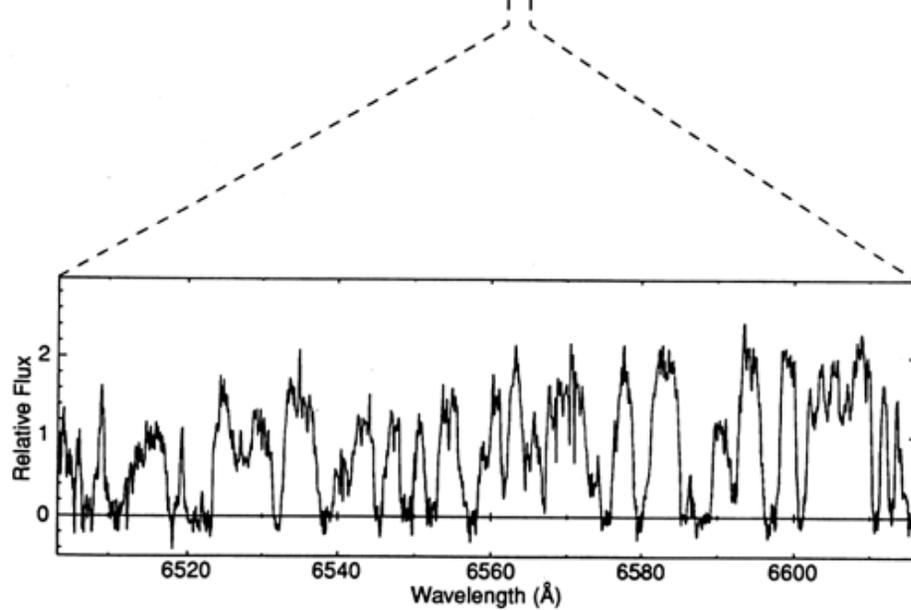
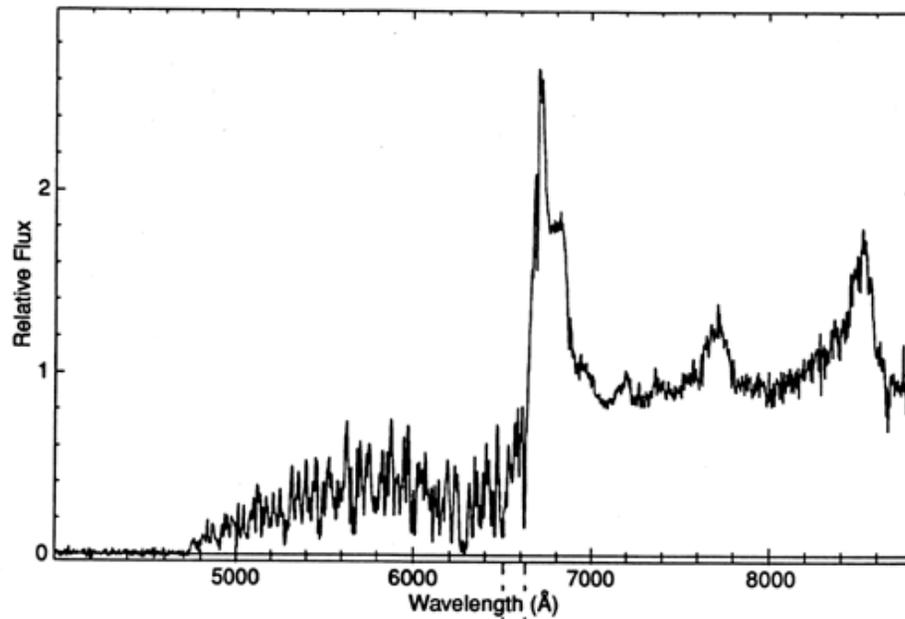
- characteristic scale of high-redshift galaxies: 0.1 arcsec
- characteristic scale of Spitzer PSF: 2.5 arcsec (or larger at longer wavelengths)
- Spitzer images are undersampled
- almost all galaxies overlap other galaxies
- how to measure faint galaxies that overlap bright galaxies?

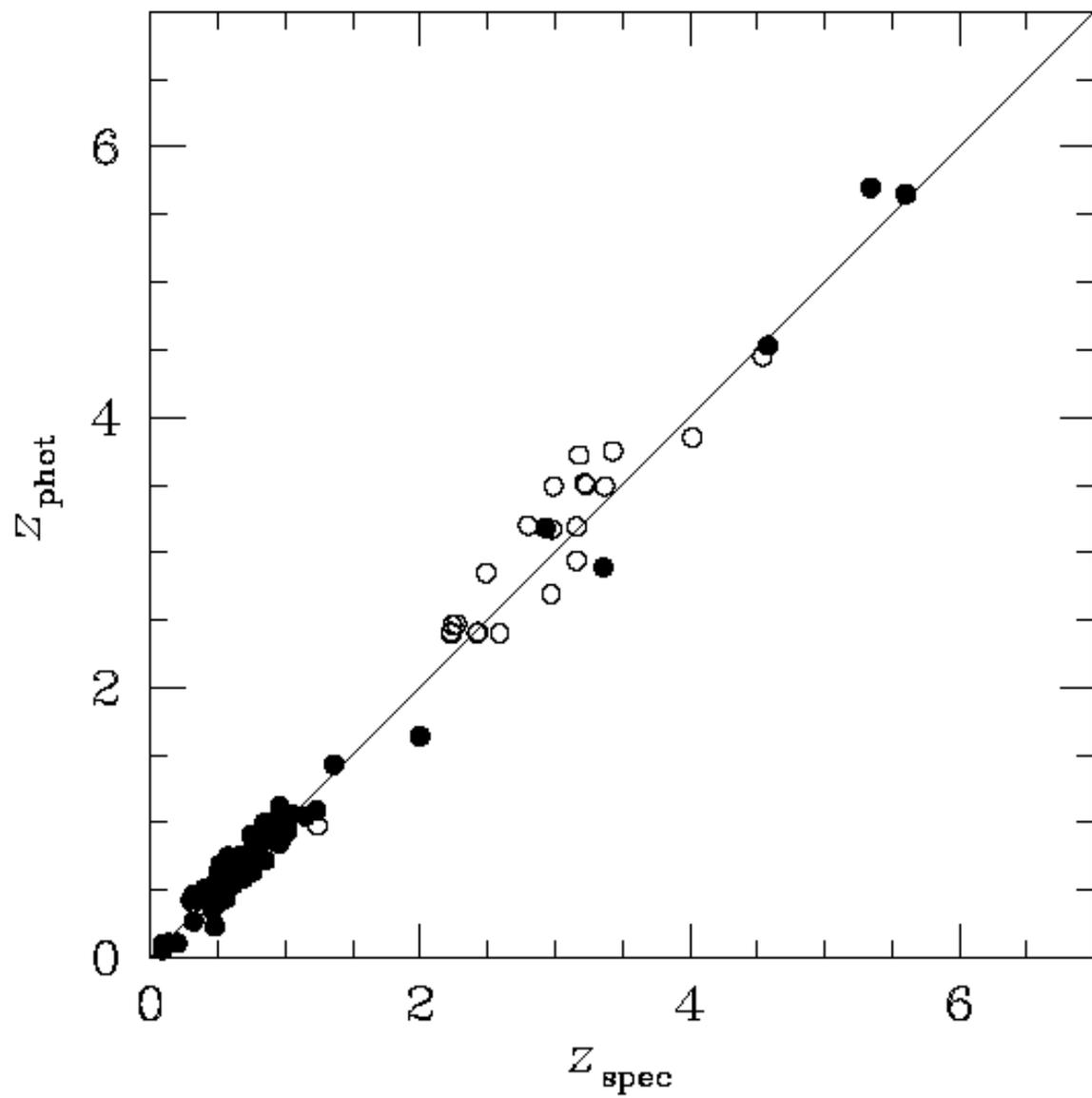


# Photometric redshift technique

- Determine redshifts by comparing measured and modeled broad-band photometry
- Six galaxy spectrophotometric templates
- Effects of intrinsic (Lyman limit) and intervening (Lyman-alpha forest and Lyman limit) absorption
- Redshift likelihood functions
- Demonstrated accurate ( $\Delta z / (1 + z) < 6\%$ ) and reliable (no outliers) at redshifts  $z = 0$  through 6

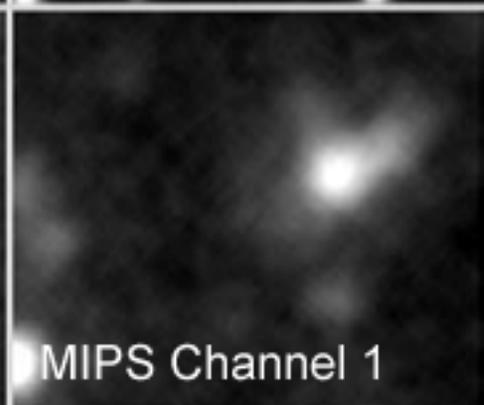
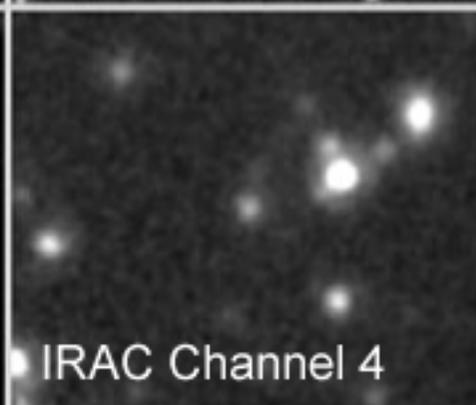
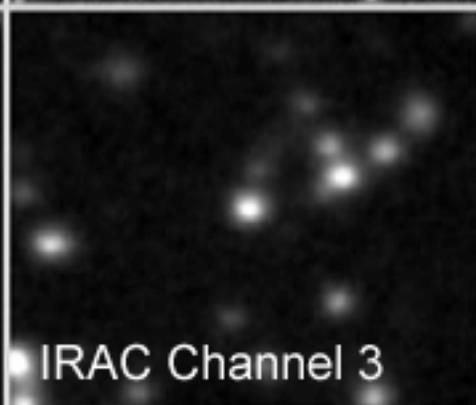
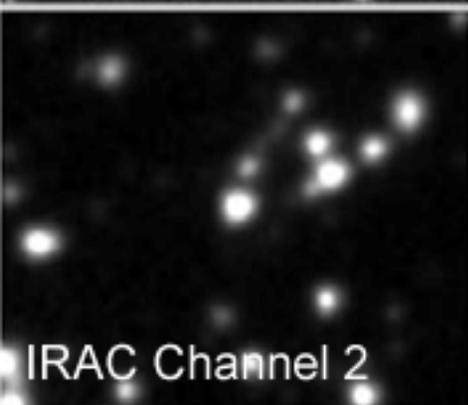
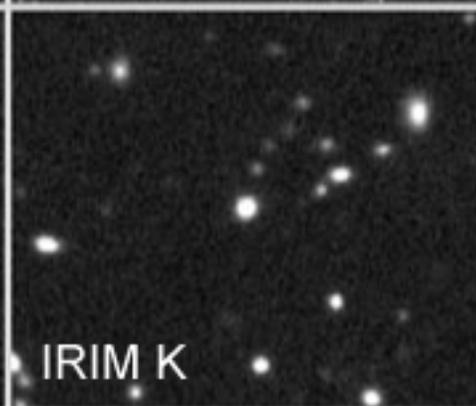
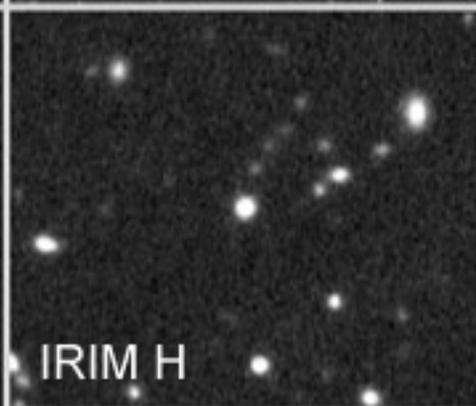
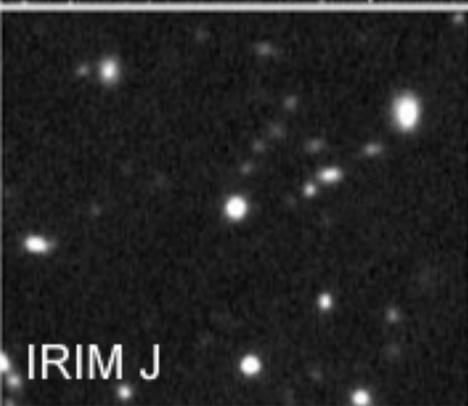
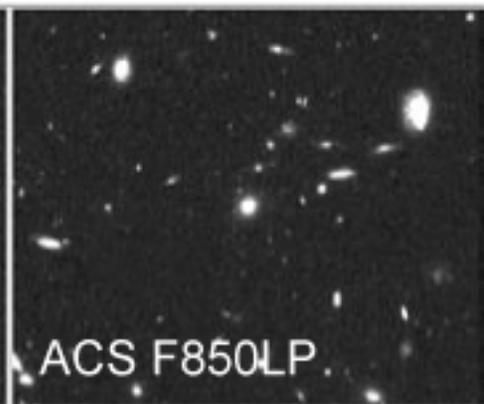
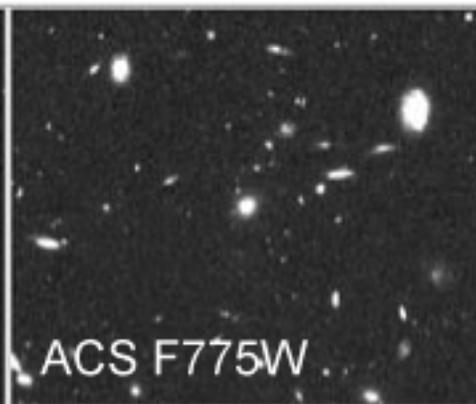
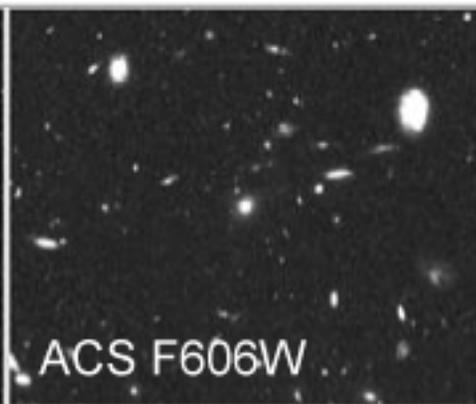
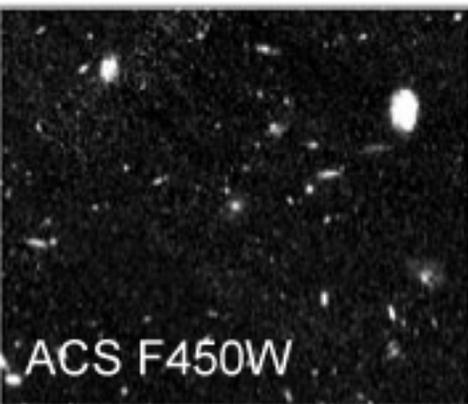






# “Redshift spatial profile fitting” technique

- “deconvolve” a sequence of “source” images (typically higher-resolution images at optical wavelengths) to obtain photometric redshifts and spatial models of galaxies
- use spatial models to fit for energy fluxes in a sequence of “target” images (typically lower-resolution images at near- or mid-infrared wavelengths)

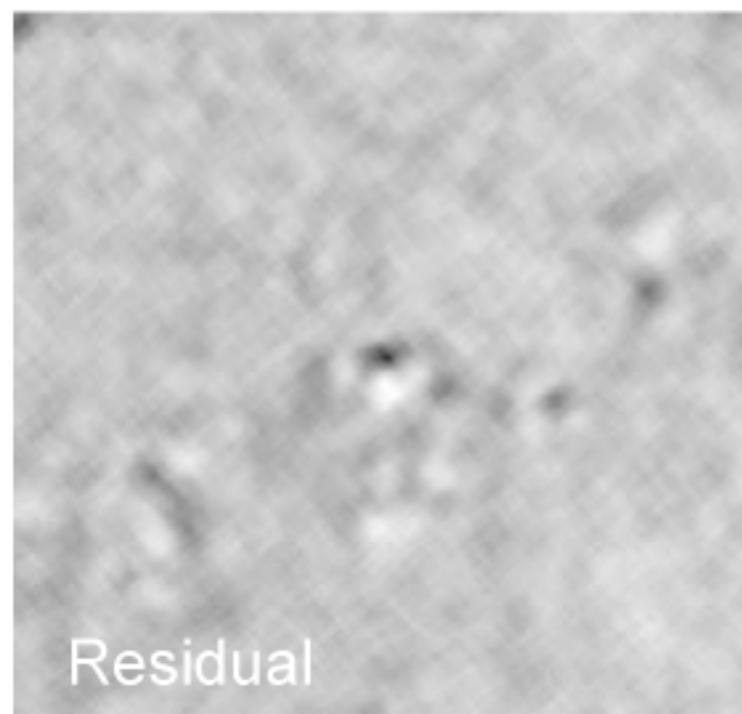
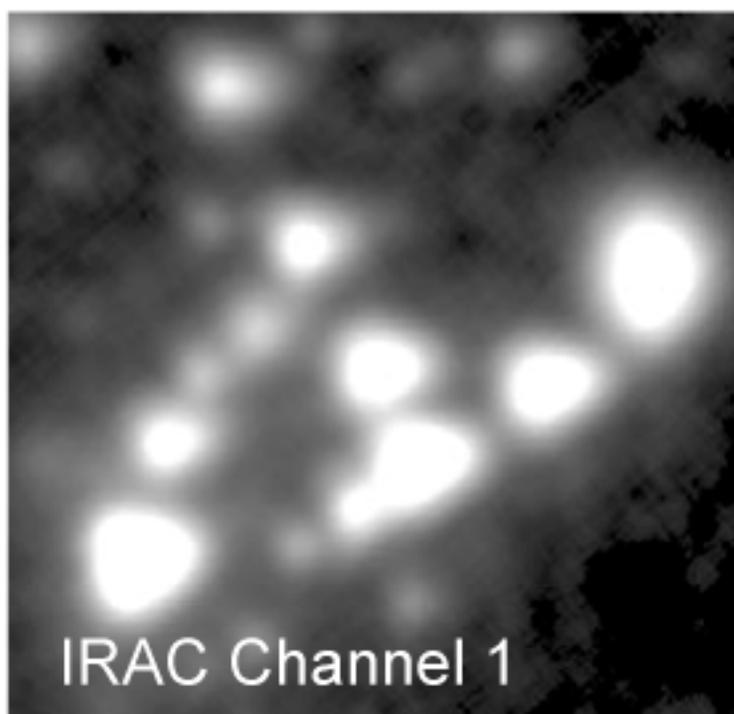
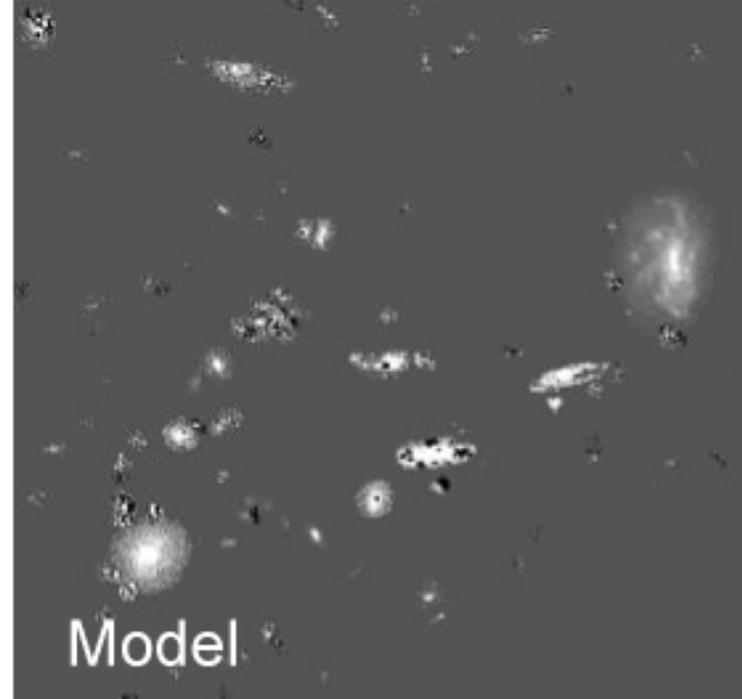
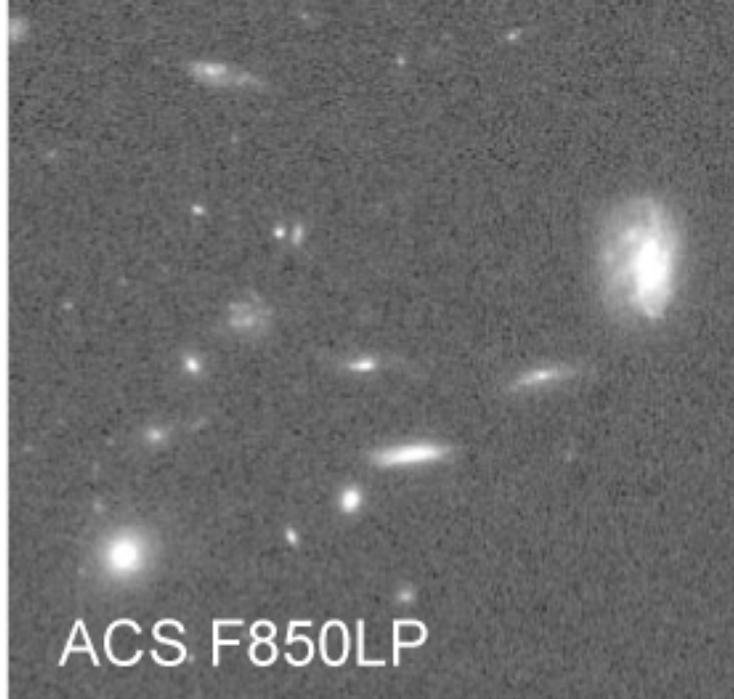


# “Deconvolving” source images

- build one spatial model image on a fine pixel scale
- relate spatial model image to each data image via geometric transformation, convolution, and scaling by spectral templates on a galaxy-by-galaxy basis
- simultaneously determine spatial models and photometric redshifts

# Fitting target images

- do not “add” target images (undersampling, correlated noise)
- instead, relate spatial model image to each data image via geometric transformation, convolution, and scaling by unknown energy flux on a galaxy-by-galaxy basis
- determine energy fluxes



# Computational requirement

- each step of “deconvolving” or fitting requires transformation and convolution of the spatial model image to each data image
- registration of each data image must be fitted for as part of the process
- since there are a lot of data images, this is computationally very expensive

# Computer setup

- 50 Xeon 3.06 GHz processors (donated by Intel Corporation)
- 20 cluster nodes, four workstations, one file server
- two Itanium 1.4 GHz processors (donated by Ion Computers)
- one database server
- 2 TB disk storage, 10 TB local disk caches
- custom job control and database software



# What is needed to measure faint galaxies in deep Spitzer images

- accurate image alignment
  - geometric distortion, registration
  - better than 0.01 pixel
- accurate spatial models
  - deconvolution of source images
  - convolution of target images
- “color segmentation”
  - segment galaxy profiles by color

# Part I:

---

To measure photometry of faint galaxies:

1. Construct profiles (**models**) of all galaxies using **source images** (e.g. ACS)
2. Use **models** to simultaneously solve for best-fit photometry of all galaxies in **target images** (e.g. SST IRAC, MIPS)

# z Spatial Profile Fitting Technique

---

For **source images** in bandpasses  $\beta$ :

$$D^\beta = T(F^\beta \cdot M) \otimes P^\beta + \text{noise}$$

( $D$  data,  $F$  template,  $T(F \cdot M)$  transformed scaled model,  $P$  psf)

and:

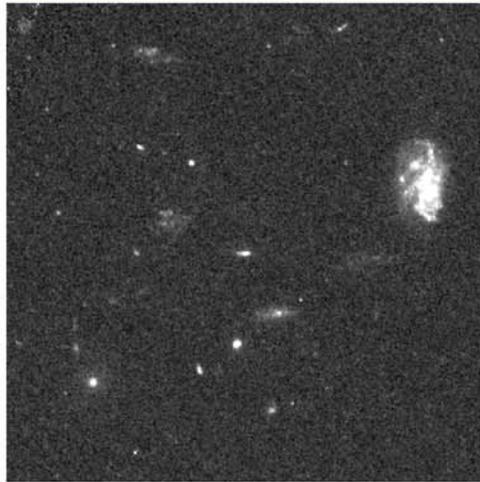
$$\chi^2 = \sum \left( \frac{T(F^\beta \cdot M) \otimes P^\beta - D^\beta}{\sigma^\beta} \right)^2$$

Solve  $\nabla \chi^2 = 0$  across all bandpasses for:

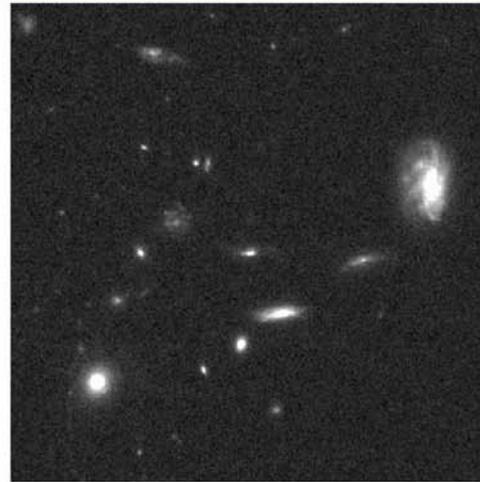
- **pixels of  $M$**  on model grid
- Per-galaxy **spectral templates  $F$**  of model
- **transformation  $T$**  of model to each image

# Example: Source Images

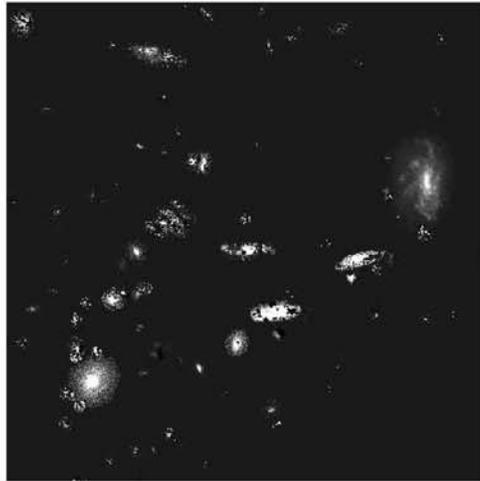
---



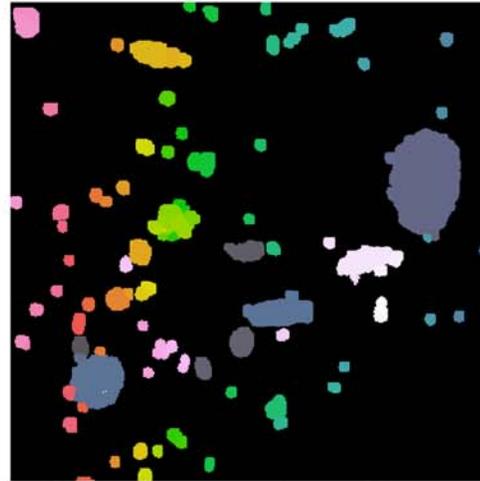
F450W Data



F850LP Data



Model



Segmentation Map

# z Spatial Profile Fitting Technique

---

For **target images** in a single bandpass  $\gamma$ :

$$D^\gamma = T(f^\gamma \cdot \bar{M}) \otimes P^\gamma + \text{noise}$$

( $f$  per-galaxy photometry,  $T(\bar{M})$  transformed normalized model)

and:

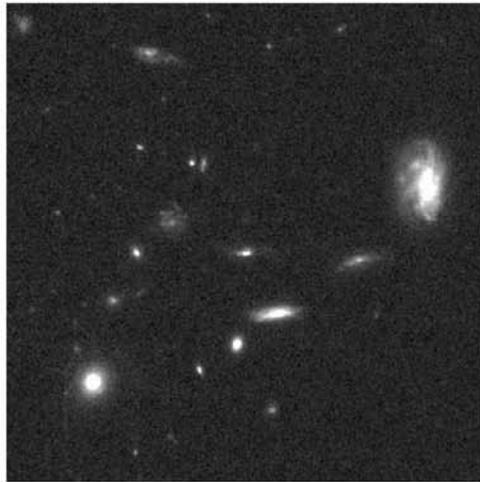
$$\chi^2 = \sum \left( \frac{f^\gamma \cdot T(\bar{M}) \otimes P^\gamma - D^\gamma}{\sigma^\gamma} \right)^2$$

Given model  $M$ , solve  $\nabla \chi^2 = 0$  for:

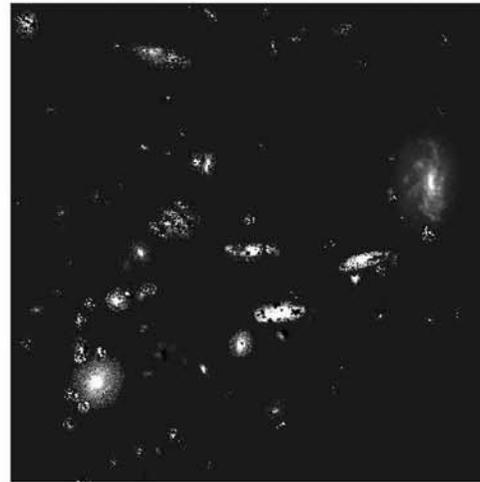
- Per-galaxy **fluxes**  $f$
- **transformation**  $T$  of model to each image

# Example: Target Images

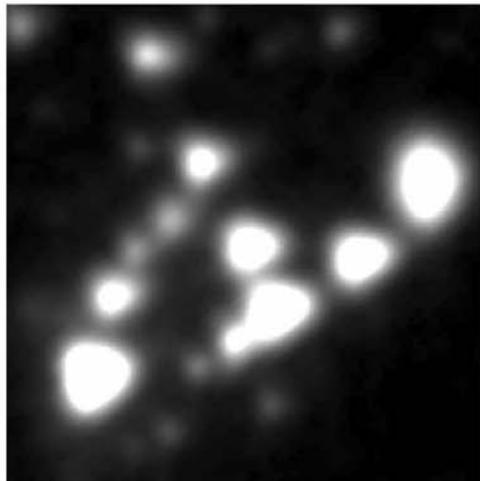
---



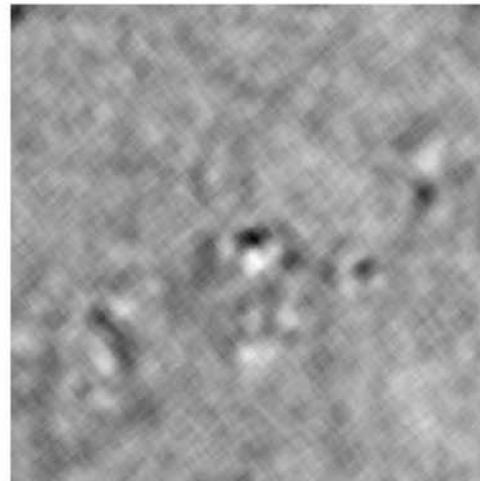
F850LP Data



Model



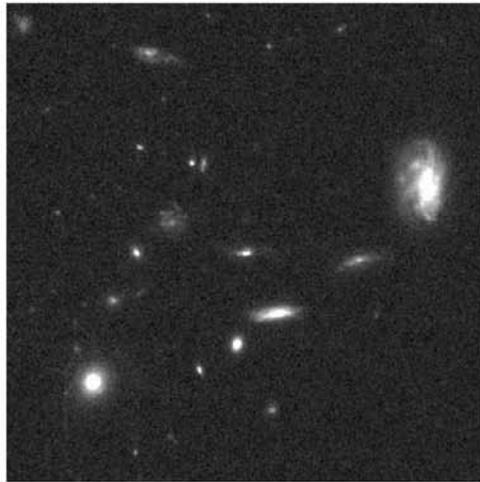
IRAC 3.6  $\mu\text{m}$  Data



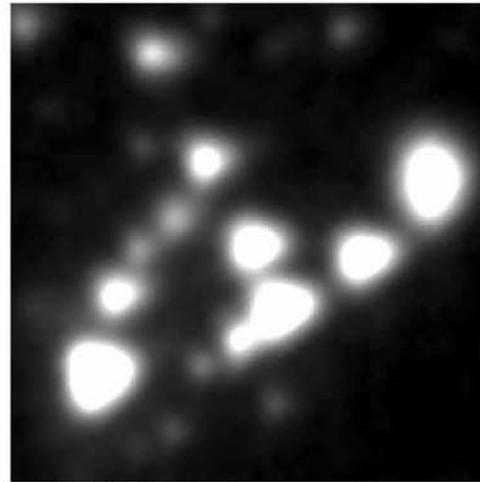
Residual

# Problem: Target Image Residuals

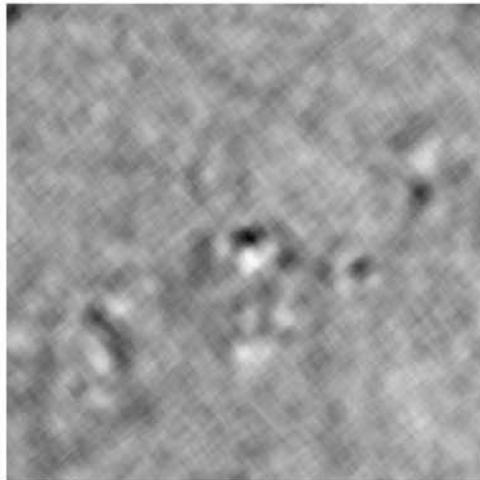
---



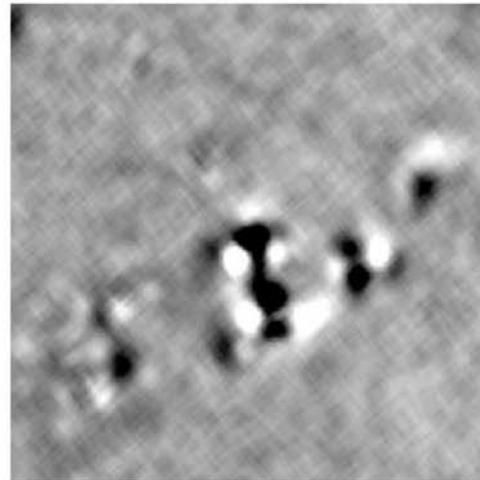
F850LP Data



IRAC 3.6  $\mu\text{m}$  Data



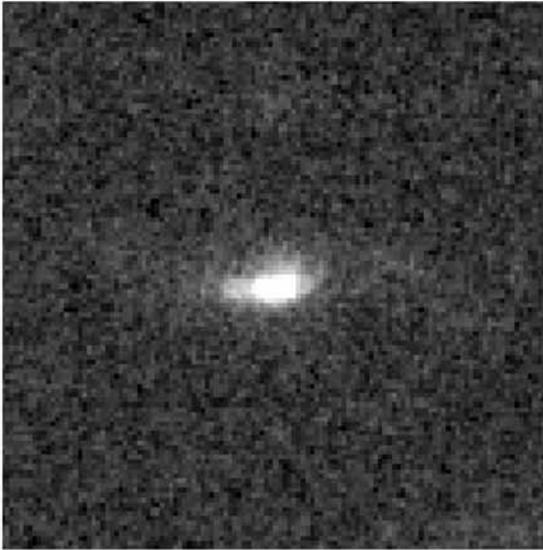
Residual



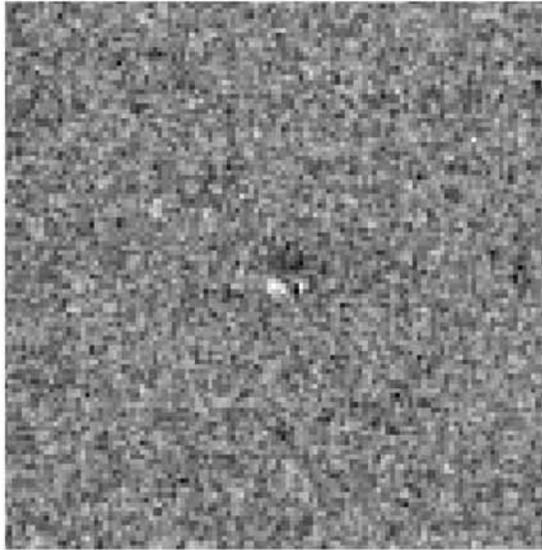
Residual per Eqns

# Problem: Color Variations

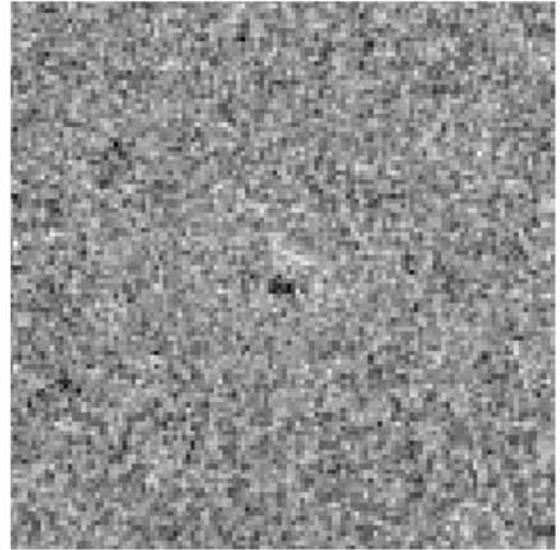
---



F850LP data



F850LP residual

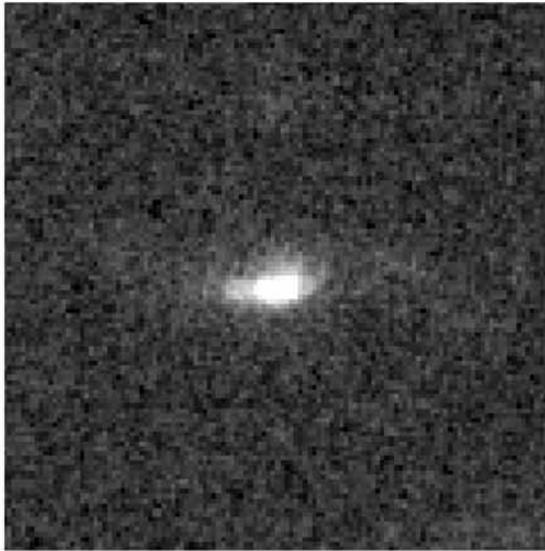


F450W residual

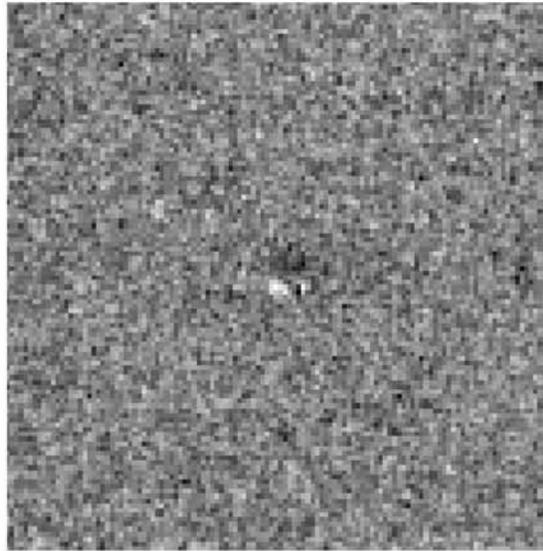
**Problem:** Galaxy profile changes with wavelength

# Solution: Color Segmentation

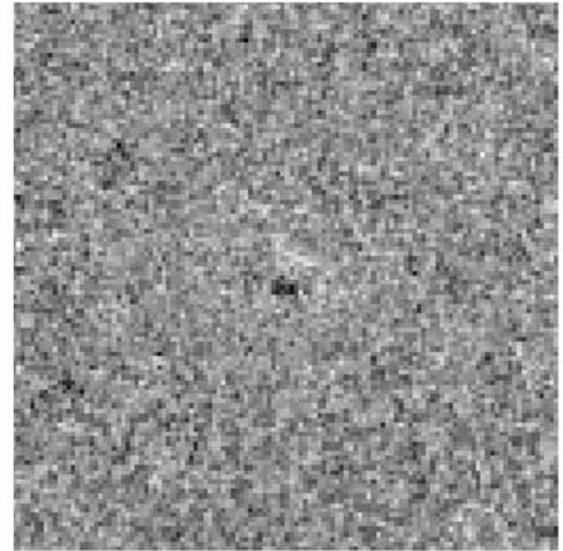
---



F850LP data



F850LP residual



F450W residual

**Problem:** Galaxy profile changes with wavelength

**Solution:** Segment models into **color components**

# Color Segmentation

Recall, for **source images**:

$$D^\beta = T(F^\beta \cdot M) \otimes P^\beta \quad (\beta \equiv \text{bandpass})$$

and:

$$\chi^2 = \sum \left( \frac{T(F^\beta \cdot M) \otimes P^\beta - D^\beta}{\sigma^\beta} \right)^2$$

Solve  $\nabla \chi^2 = 0$  for pixels of  $M$  on model grid

$\Rightarrow$  at minimum,  $\nabla \chi^2 = 0$ , but  $\nabla \chi_\beta^2 \neq 0$

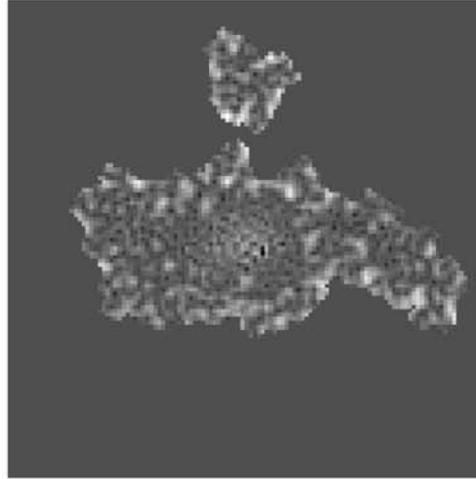
$$F^\beta \cdot M \Rightarrow F_1^\beta \cdot M_1 + F_2^\beta \cdot M_2 + \dots$$

# An Example

---



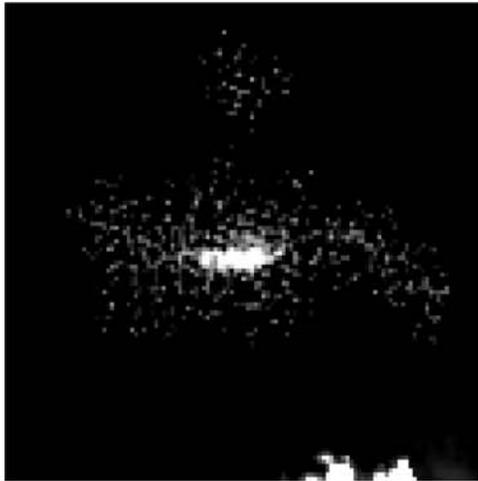
Model



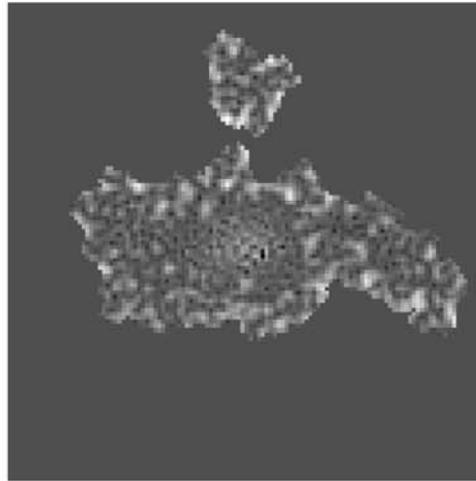
Model gradient

# An Example

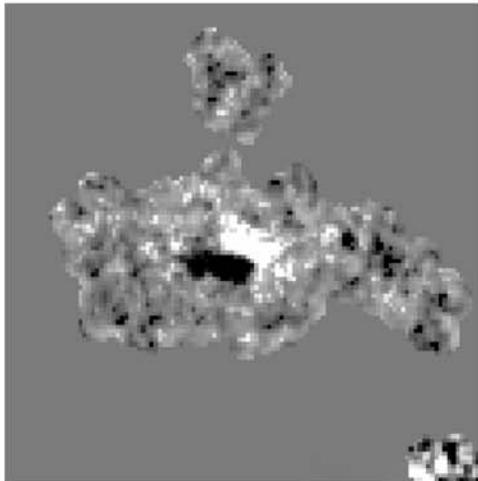
---



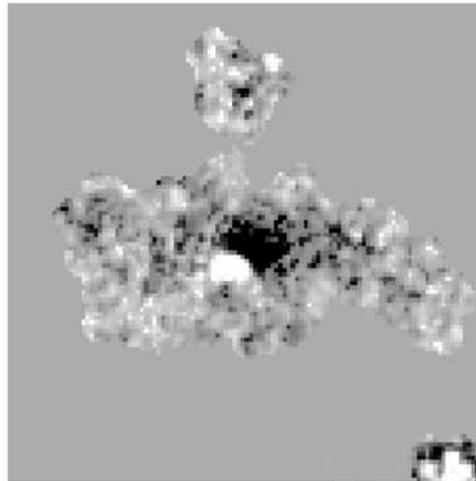
Model



Model gradient

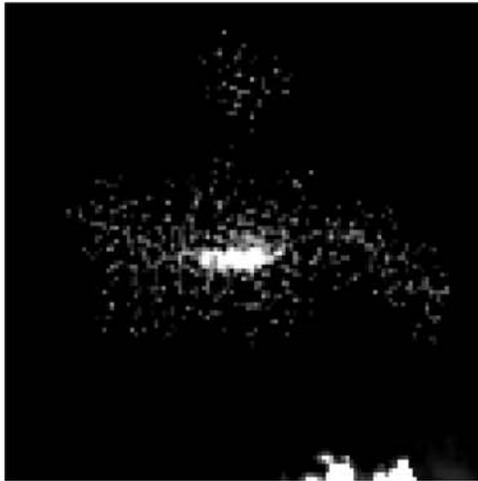


Gradient in F450W

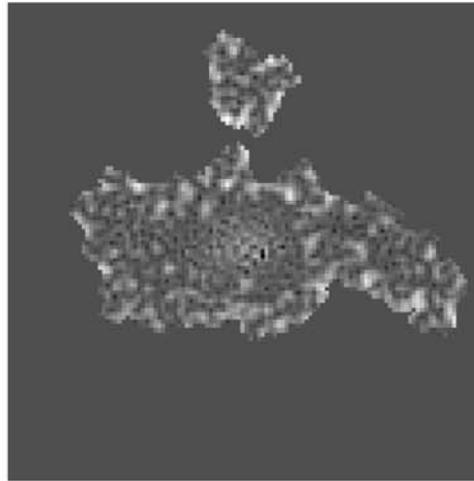


Gradient in F850LP

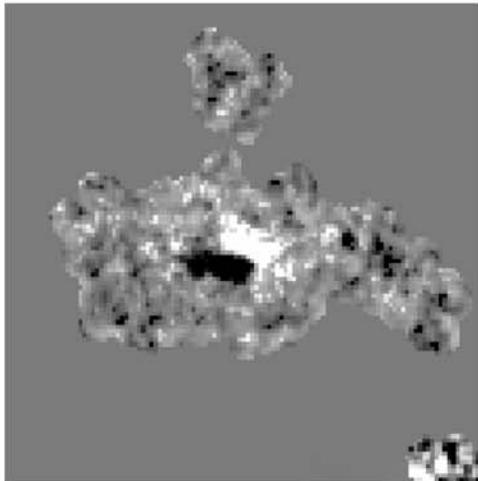
# An Example



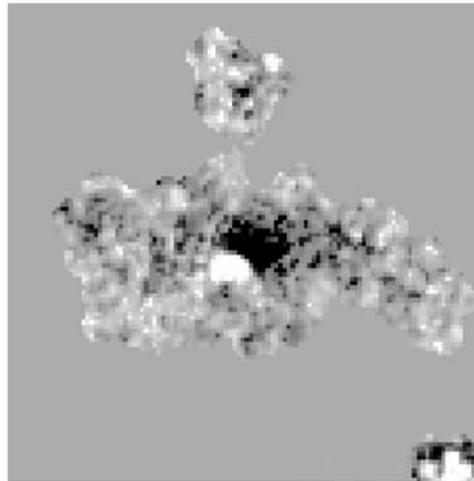
Model



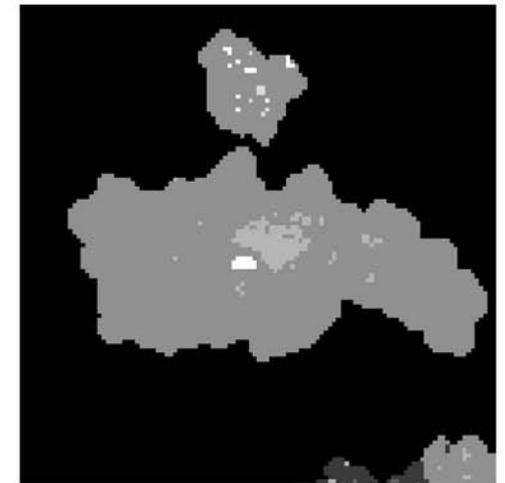
Model gradient



Gradient in F450W



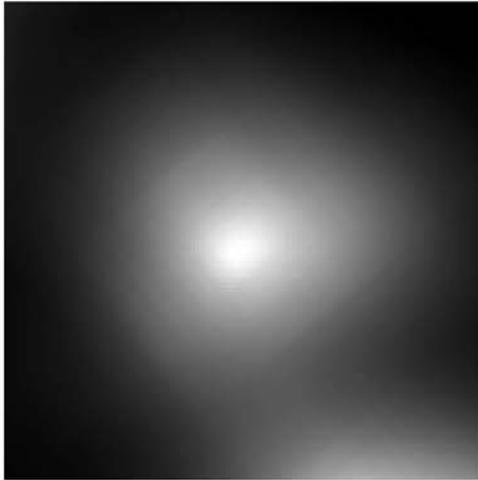
Gradient in F850LP



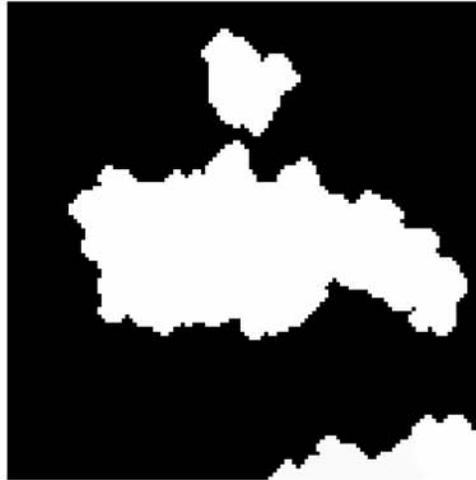
Segmentation

# An example: Spitzer

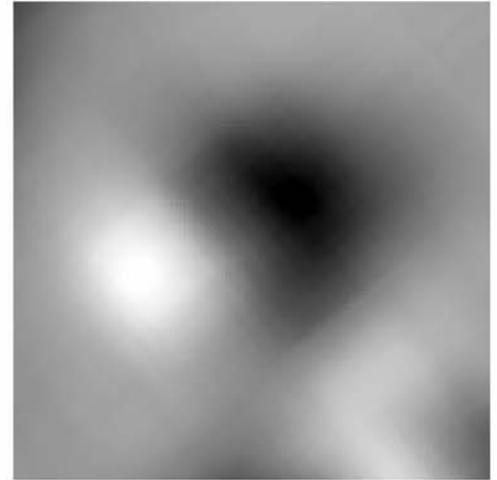
---



SST 3.6  $\mu\text{m}$  Data



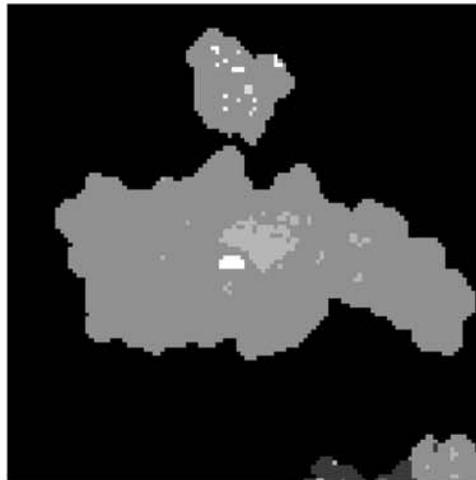
Segmentation Map



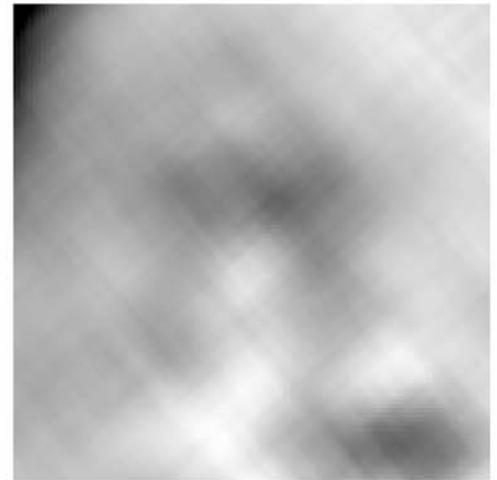
SST 3.6  $\mu\text{m}$  Residual



SST 3.6  $\mu\text{m}$  Data



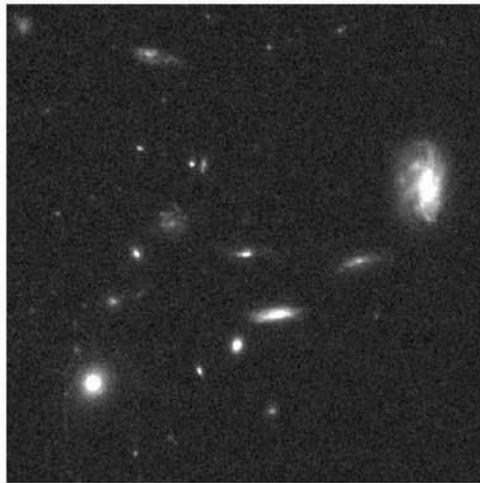
Type Map



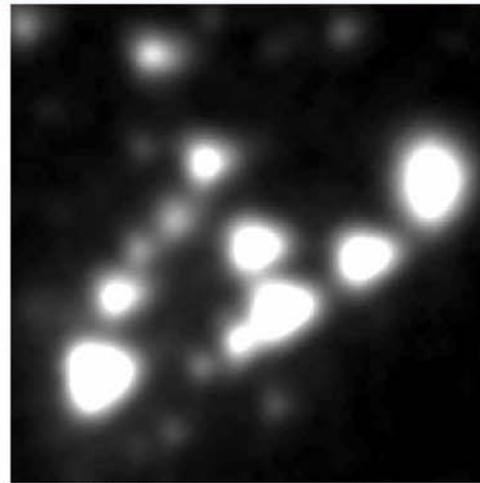
SST 3.6  $\mu\text{m}$  Residual

# Spitzer PSF Determination

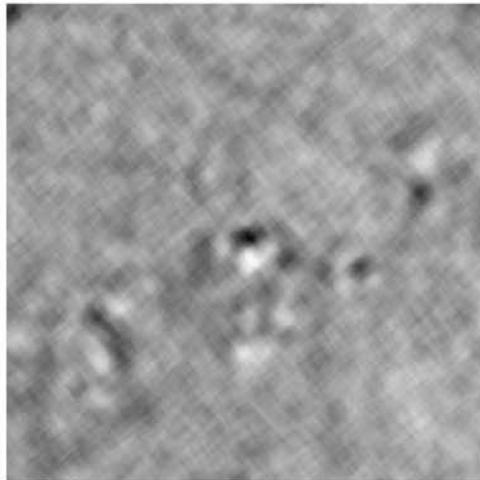
---



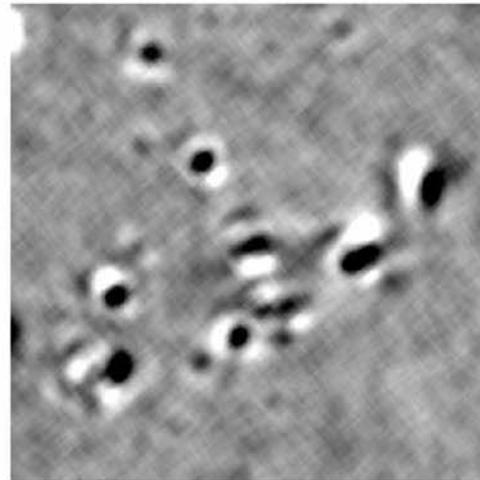
F850LP Data



IRAC 3.6  $\mu\text{m}$  Data



Residual



Residual w/ SST PSF

# PSF Fitting

Recall, once again, for **source images**:

$$D^\beta = T(F^\beta \cdot M) \otimes P^\beta \quad (\beta \equiv \text{bandpass})$$

and:

$$\chi^2 = \sum \left( \frac{T(F^\beta \cdot M) \otimes P^\beta - D^\beta}{\sigma^\beta} \right)^2$$

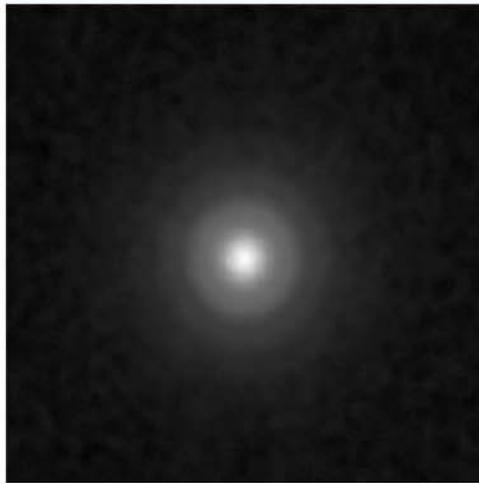
Given model  $M$ , solve  $\nabla \chi^2 = 0$  for:

- **pixels of  $P$**  on PSF grid
- **transformation  $T$**  of model to each image

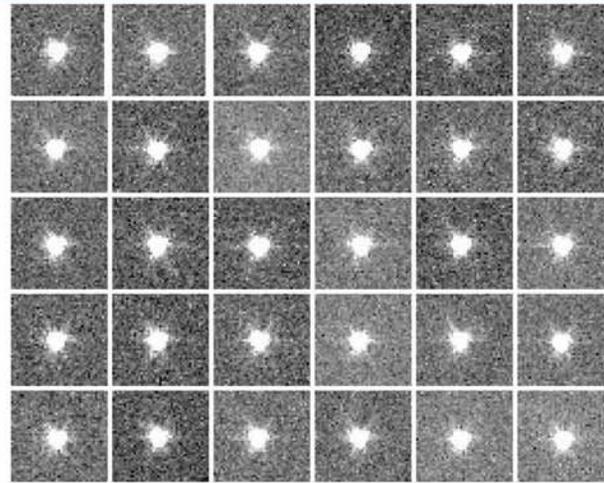
⇒ Use IRAC PSF calibration data (BD+67 1044)

# Spitzer PSF Determination

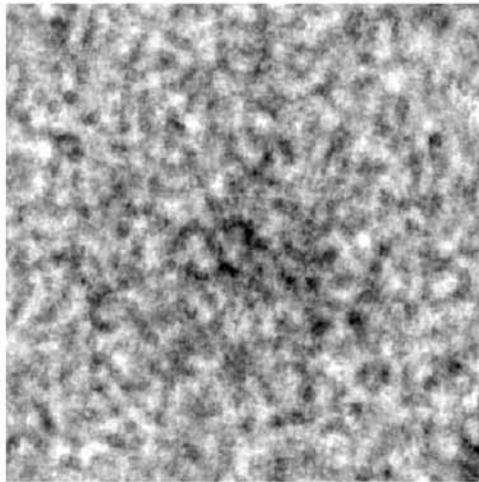
---



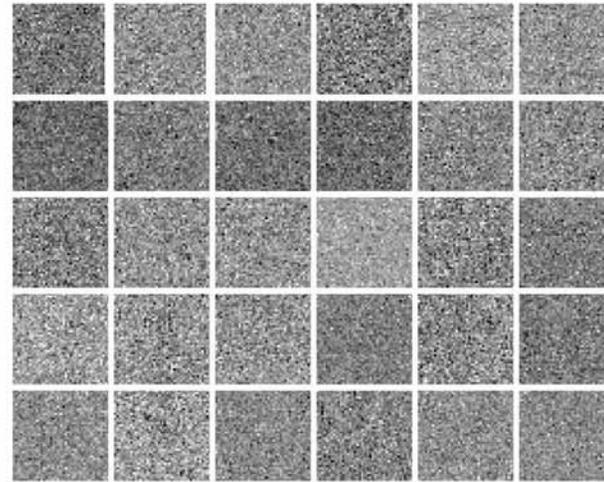
IRAC 3.6  $\mu\text{m}$  Data



Data Mosaic



Residual



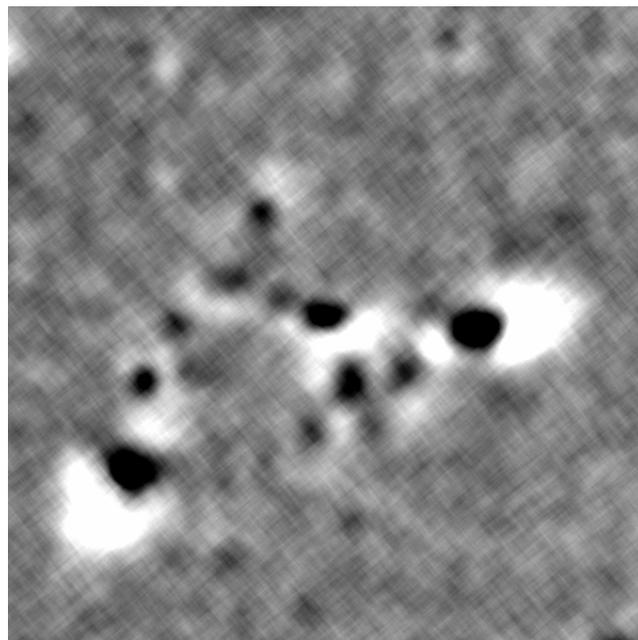
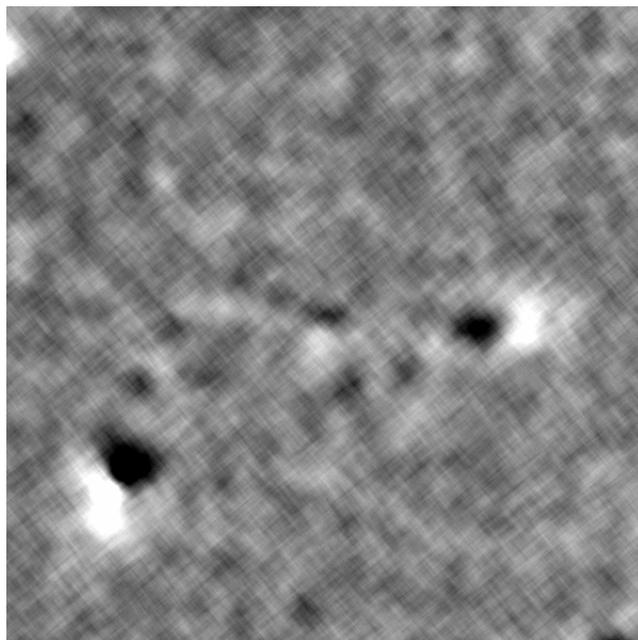
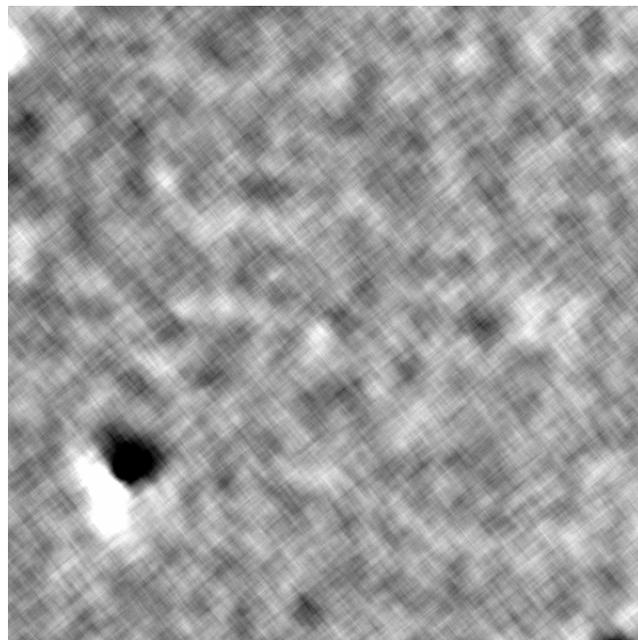
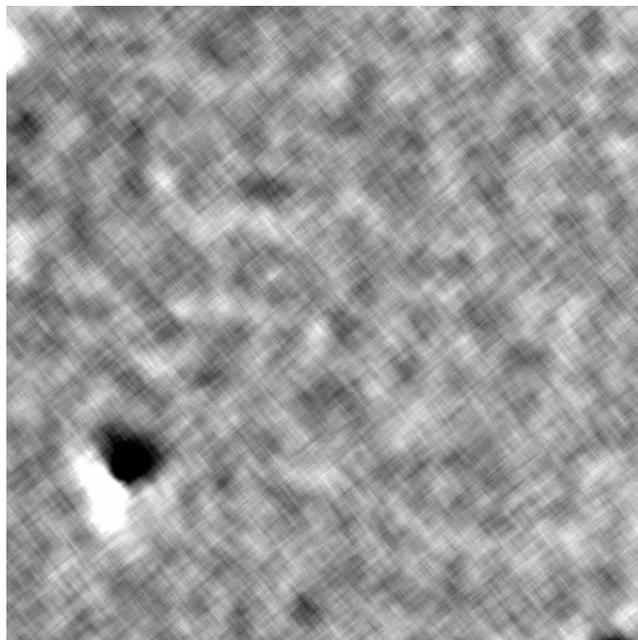
Residual Mosaic

# Image alignment

- Geometric distortion and registration—how to calibrate?
- $S/N = 500$  for a typical SST “pixel”
- Required image alignment accuracy better than 0.01 pixel
- *More or less solved*

# Noise in source images

- $S/N = 500$  for a typical SST pixel
- $S/N = 200$  over a comparable region of sky for ACS
- *noise in source images is the limiting systematic effect in measuring SST images*
- *SST images cannot be measured to within noise given current ACS images*



# Summary

- We believe that faint galaxies can be measured in deep Spitzer images only with...
- ...accurate spatial models (alignment, deconvolution and convolution, color segmentation)...
- ...and computational expense