



# Why High Energy Data Are Different

## A View From *Chandra*

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2024 May 21

# Well, Why?



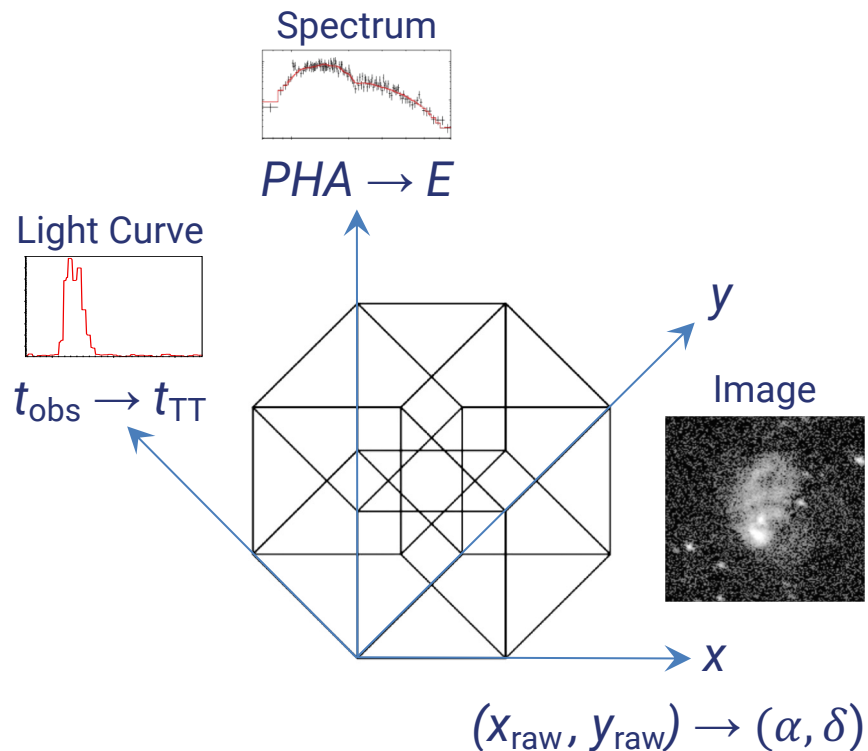
High energy photon



Optical/IR/radio photon

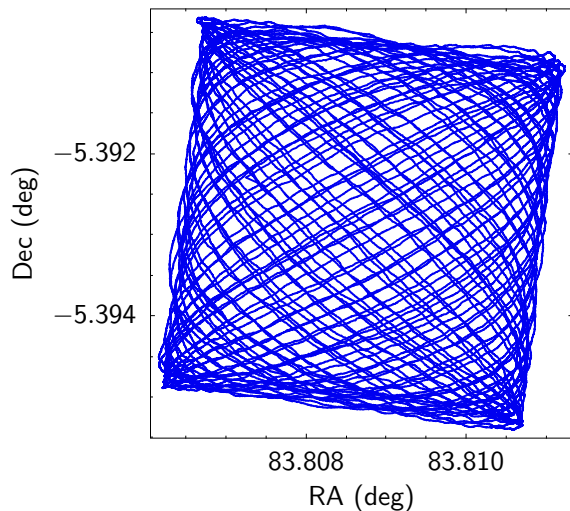
# The X-ray Data Hypercube

- *Chandra* detects *individual X-ray photons* with energies  $\sim 0.1\text{--}10.0$  keV  $\Rightarrow$  **photon counting**
- Each X-ray event records a (typically) 4-D set of observables that map to actual photon properties
  - Detector location  $(x_{\text{raw}}, y_{\text{raw}}) \rightarrow$  Sky position  $(\alpha, \delta)$
  - Spacecraft time  $t_{\text{obs}} \rightarrow$  Photon arrival time  $t_{\text{TT}}$
  - Pulse height  $PHA \rightarrow$  Photon energy  $E$
- A set of X-ray events (e.g., from a single observation) is termed an *event list*
- *Chandra* can reliably detect sources with as few as  $\sim 4\text{--}5$  counts  $\Rightarrow$  **Poisson regime**



# Chandra-Specific Considerations

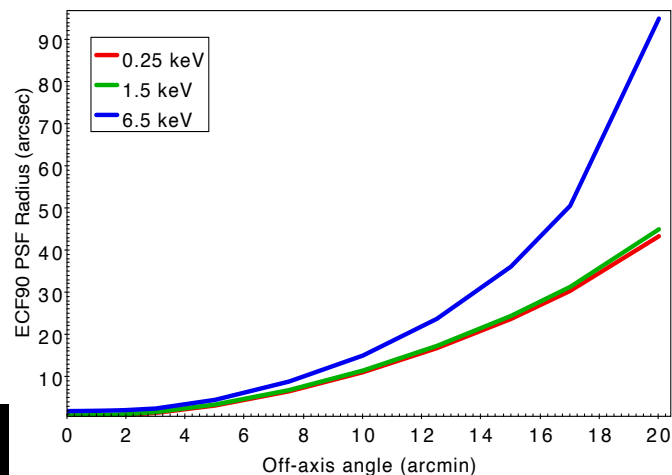
- Telescope pointing *dithers* on the sky  
⇒ Source samples different regions of the detector with different calibrations during the exposure



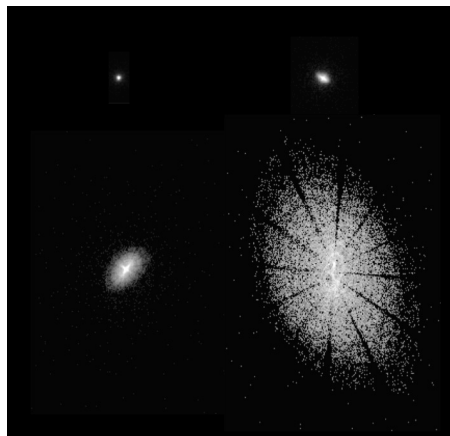
Above: Aspect solution (RA, Dec) vs. time corrects for dither motion of source on detector

- PSF varies with energy and position within FoV  
⇒ PSF depends on (unknown) source spectrum

Right: 90% Encircled Counts Fraction PSF radius vs. off-axis angle for 3 energies



Left: Chandra PSF for a specific energy on-axis, and 4, 8, 18 arcmin off-axis (log scale, various azimuthal angles)



# X-ray Pulse Height Spectroscopy

- The expected observed channel distribution of detected source counts  $M(E', \hat{p}', t)$  is:

$$\overset{\text{What we observe}}{M(E', \hat{p}', t)} = \int dE d\hat{p} R(E'; E, \hat{p}, t) P(\hat{p}'; E, \hat{p}, t) A(E, \hat{p}, t) \overset{\text{What we want}}{S(E, \hat{p}, t)}$$

$S(E, \hat{p}, t)$  is the physical model that describes the physical energy spectrum, spatial morphology, and temporal variability of the source

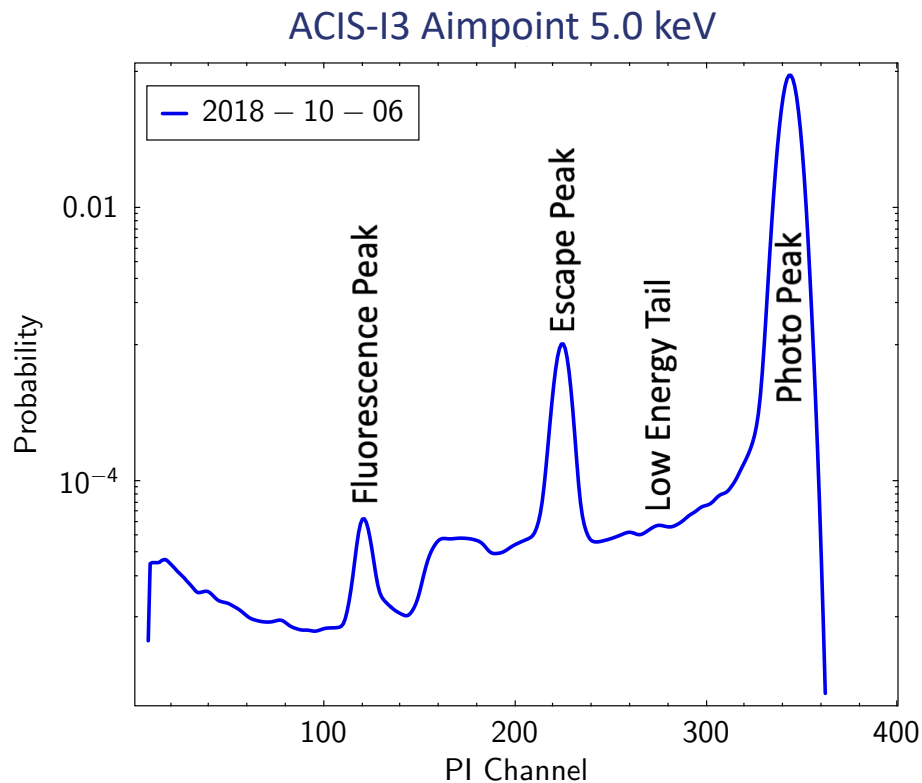
$R(E'; E, \hat{p}, t)$  is the *redistribution matrix* that defines the probability that a photon with actual energy  $E$ , location  $\hat{p}$ , and arrival time  $t$  will be observed with apparent energy  $E'$  and location  $\hat{p}'$

$A(E, \hat{p}, t)$  is the instrumental *effective area* (sensitivity)

$P(\hat{p}'; E, \hat{p}, t)$  is the photon spatial dispersion transfer function (the instrumental PSF)

# X-ray Pulse Height Spectroscopy — “Line Spread Function”

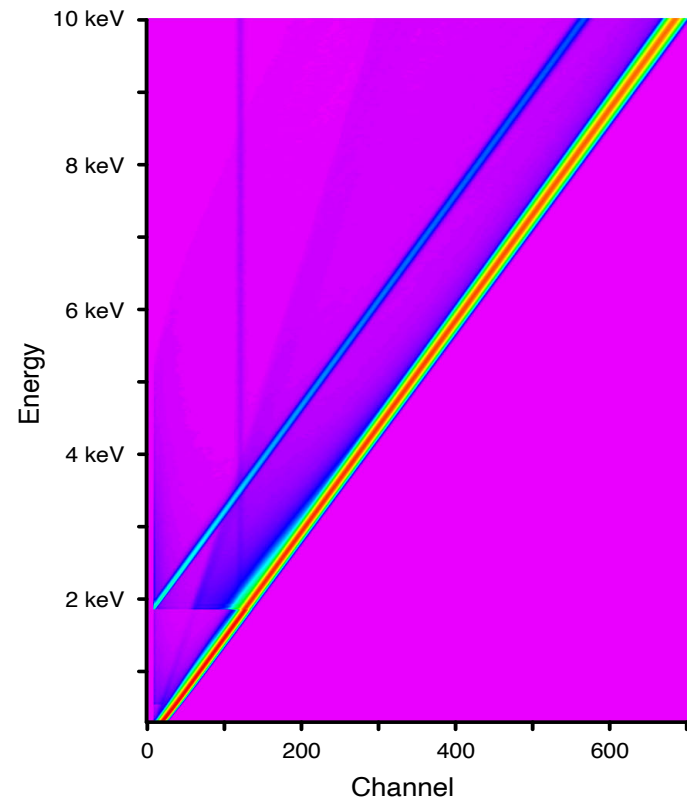
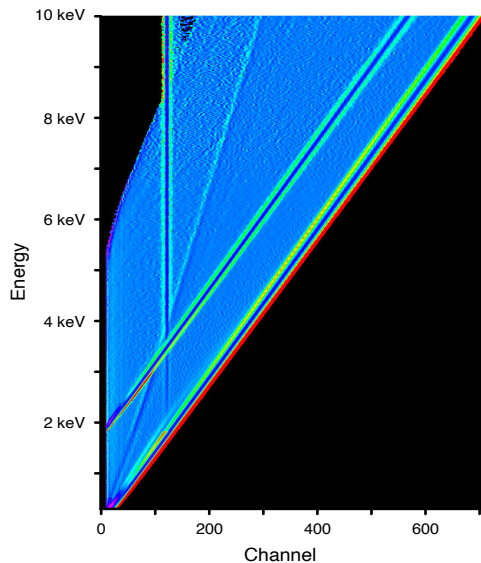
- X-ray photons of a given energy may be recorded in different detector channels with varying probabilities due to interactions with the telescope and detector optics
- Analogous to the line spread function in optical astronomy
- For *Chandra*, this response varies with *location on the detector* and *observation epoch*



# Redistribution Matrix

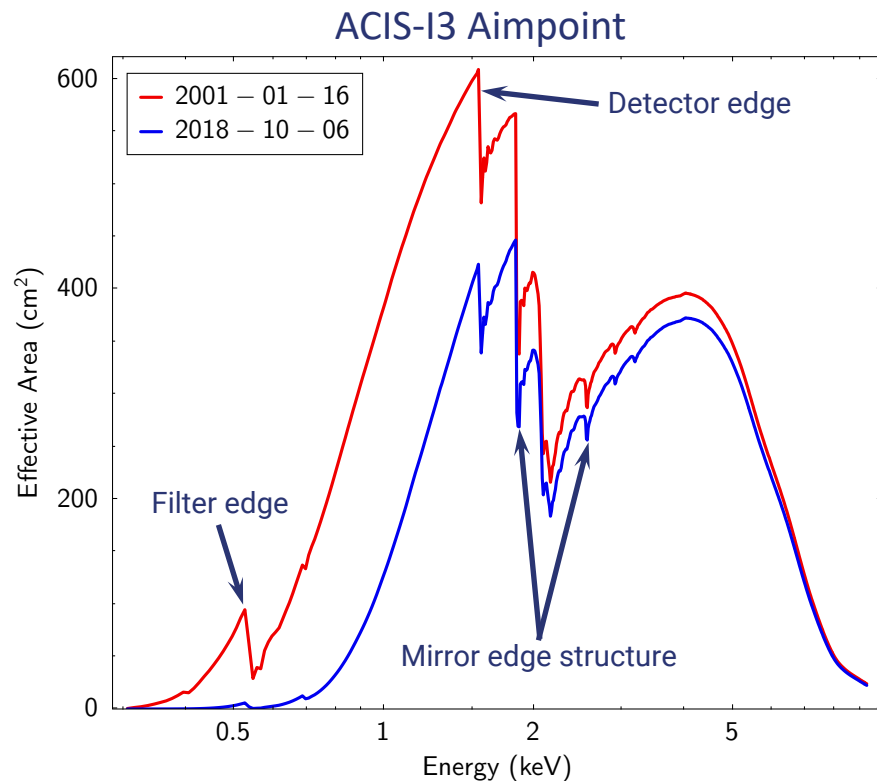
- The redistribution matrix  $R(E'; E, \hat{p}, t)$  maps the relationship between the incident photon energy and the detected signal distribution over detector channels (i.e., the event pulse height)
- *Chandra* uses the NASA HEASARC OGIP-standard RMF (Redistribution Matrix File) FITS file format

*Right: Example variation of the RMF across the Chandra dither pattern*



# X-ray Pulse Height Spectroscopy — “Sensitivity”

- The *ancillary response*  $A(E, \hat{p}', t)$  records the effective area of the telescope/detector combination
  - *Chandra* uses the NASA HEASARC OGIP-standard ARF (Ancillary Response File) FITS file format
- Analogous to the sensitivity curve in optical astronomy
- Includes geometric collecting area  $\times$  optics, gratings, detector efficiencies

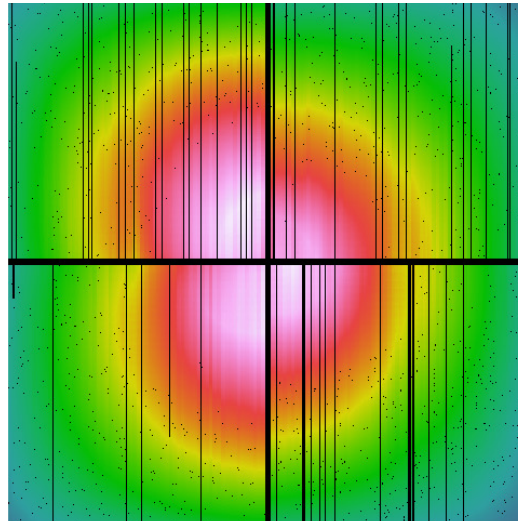




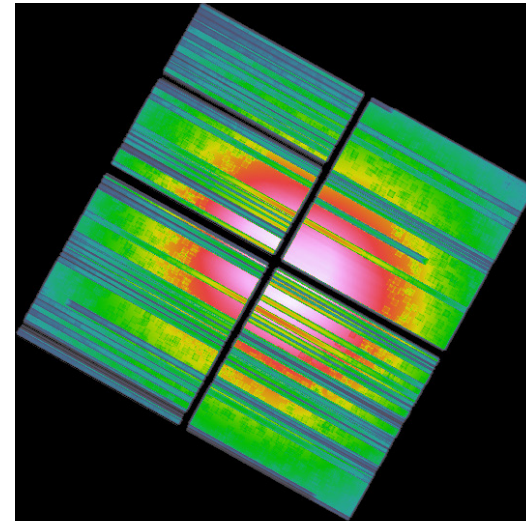
# How Does The Response Vary Over The Field?

- *Chandra* instrument & exposure maps
  - Vary with location because of vignetting and detector non-uniformities
  - Vary with observation epoch
  - Depend on energy

⇒ Maps integrated over energy depend on source spectrum



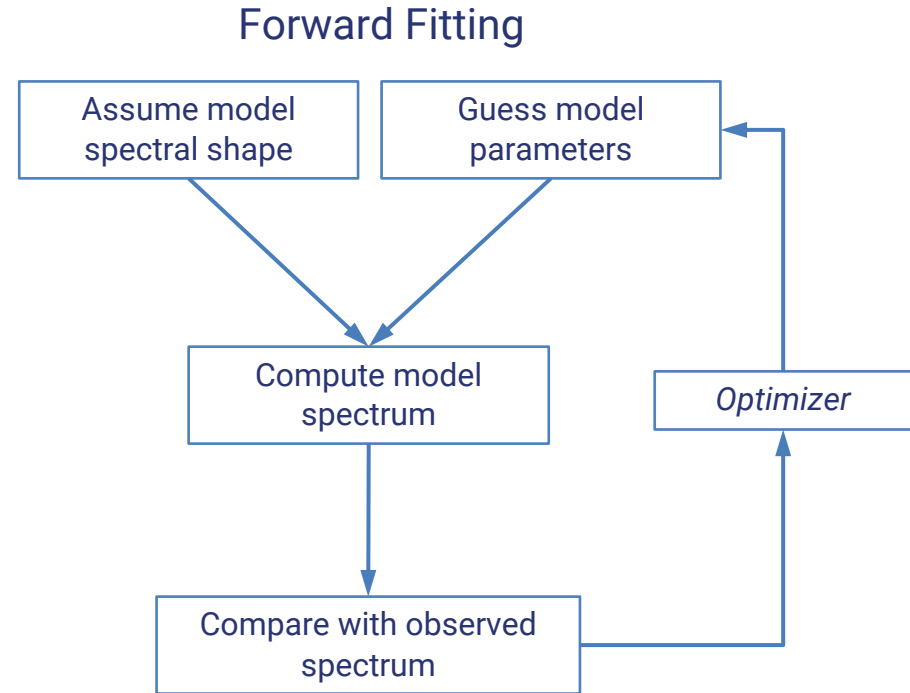
Instrument map records the instrument sensitivity in detector coordinates



Exposure map is the instrument map convolved with the aspect solution

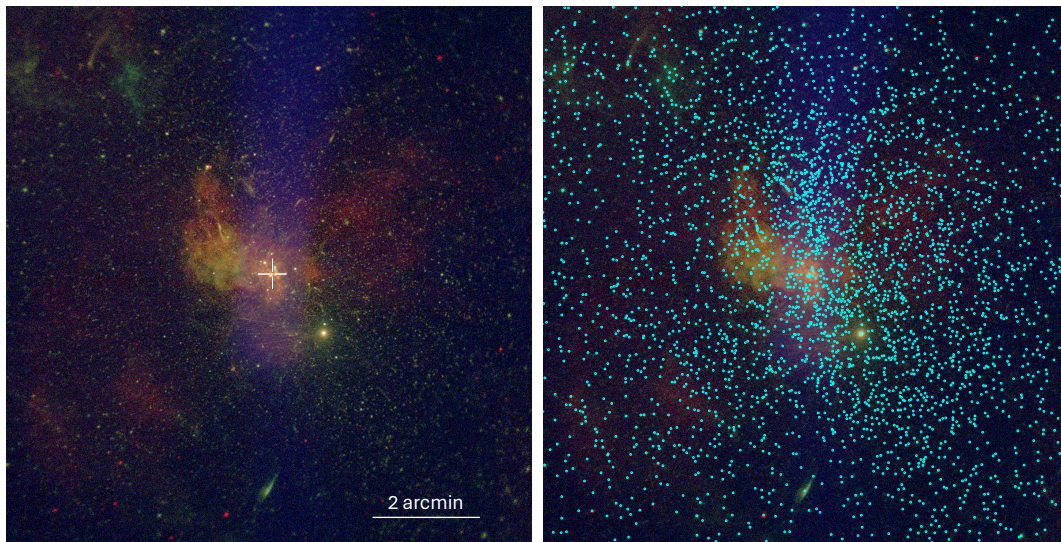
# Solving For The Source Model Parameters

- The transformation between  $S(E, \hat{p}, t)$  and  $M(E', \hat{p}', t)$  is not easily invertible
- Forward fitting
  - Propose a model for  $S(E, \hat{p}, t)$
  - Fold the model through the responses
  - Optimize the parameters of  $S(E, \hat{p}, t)$  by comparing with the observed channel counts distribution



# Chandra Source Catalog Release 2.1

- Source positions, extents, multi-band aperture photometry, temporal variability, hardness ratios, spectral fits
- All measurements have associated confidence intervals
- Imaging data released publicly through 2021
- Stacked observations for fainter detections
- Tied to Gaia-CRF3 astrometric frame
- IVOA compliant interfaces
- ~408K unique X-ray sources on the sky
- ~2.1M detections and photometric upper limits
- ~730 square degrees on the sky
- ~45TB science ready FITS format data products



A cutout of a roughly 3 Ms observation stack (a co-add of 86 observations) from CSC 2.1, centered on Sgr A\* (identified by the cross). The positions of roughly 3,300 X-ray point sources in this region from CSC 2.1 are identified.

For more information see  
<https://cxc.cfa.harvard.edu/csc/>

# Chandra Source Catalog X-ray Property Measurements

**Measurements are only meaningful if we understand the measurement confidence**

- Optimization methods and Bayesian inference are used extensively for X-ray data analysis
  - ⇒ Measurements may be represented in ways that are less commonly used in other wavebands
- Example: Bayesian X-ray aperture photometry (Primini & Kashyap 2014 ApJ 796, 24)

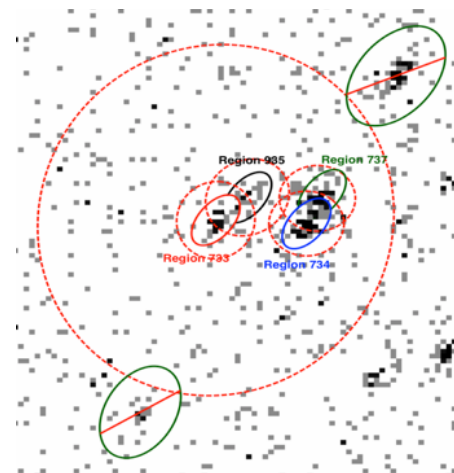
- Solve for multiple detections/overlapping apertures and background simultaneously
- Compute joint posterior probability density function (JPDF) for source and background fluxes

$$P(\mu_{s_1} \dots \mu_{s_n}, \mu_b | C_1 \dots C_n, B) = K P(\mu_b) \text{Pois}(B | \mu_b) \prod_{i=1}^n P(\mu_{s_i}) \text{Pois}(C_i | \mu_{s_i})$$

- Calculating *marginalized posterior probability density function (MPDF)* for a single source by integrating the JPDF is computationally complex

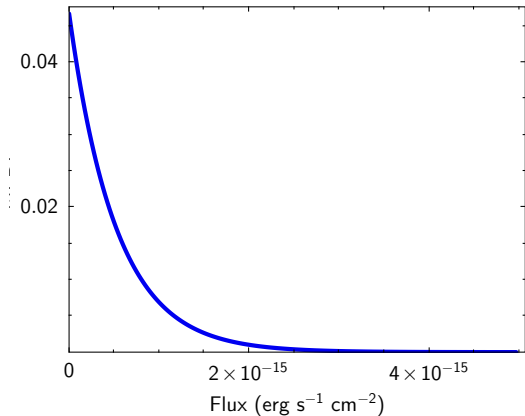
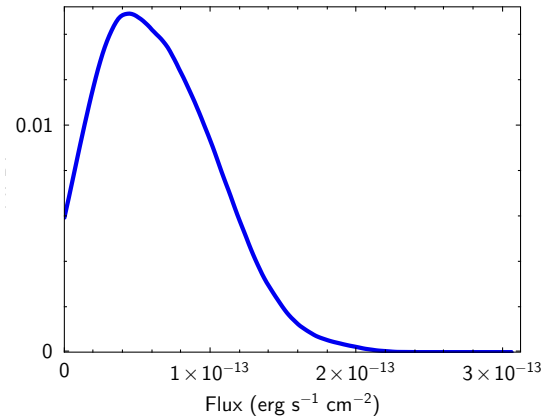
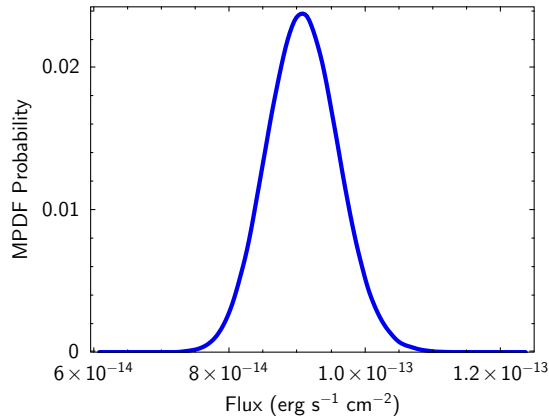
$$P(s_i | C_1 \dots C_n, B) = \int \dots \int_{b, s_j \neq s_i} db \left( \prod_{j \neq i} ds_j \right) P(\mu_{s_1} \dots \mu_{s_n}, b | C_1 \dots C_n, B)$$

⇒ Optimize using Markov chain Monte Carlo sampler



# Example: Bayesian X-ray Aperture Photometry

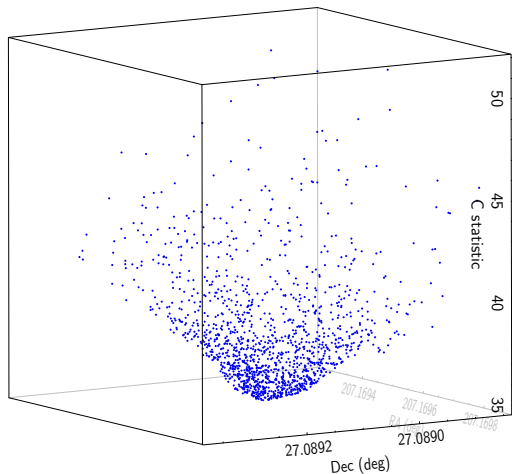
- Posteriors provide an estimate of the measurement value (*e.g.*, the mode of the distribution) and arbitrary lower and upper confidence intervals can be evaluated
  - MPDFs are unlikely to be Gaussian in the Poisson regime
  - Posteriors must be accompanied by information about models and prior probability distributions
  - Note there are many types of measurements for which multi-dimensional posteriors are appropriate



Three example X-ray photometry MPDFs for different source fluxes

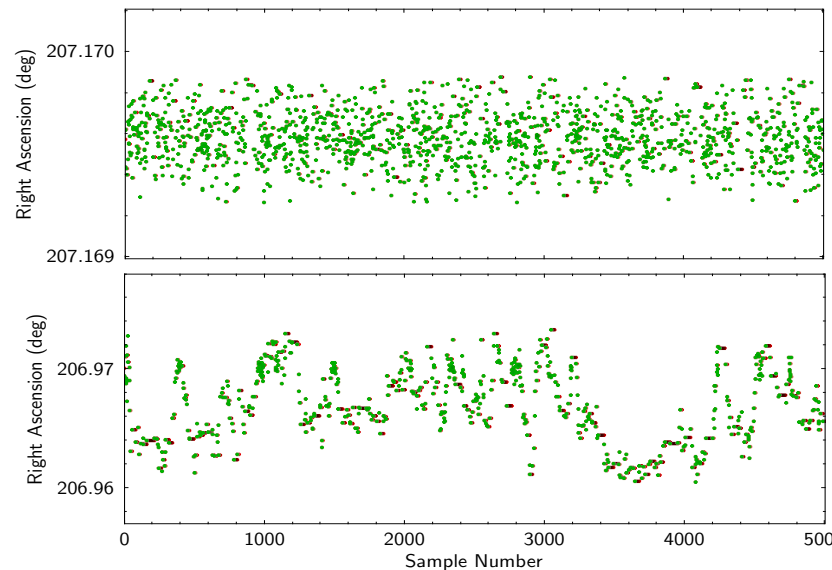
# Example: Markov Chain Monte Carlo Position Confidence

- Markov chain Monte Carlo draws provide more information but are less directly interpretable
  - Mode of sample values after burn-in is the measurement value
  - Distribution of the sample values is the MPDF from which confidence intervals can be calculated
  - Convergence can be quickly evaluated visually from the distribution of samples
  - Sample subsets can be used to compute numerical convergence criterion,  $\hat{R}$



*Above:* Position confidence MCMC draws

*Right:* Example converged and un-converged MCMC draws





# Conclusions

- Responses (ARF, RMF, and possibly PSF, time filter) are ancillary data products that are *necessary* to understand/interpret X-ray pulse height spectroscopy measurements
  - These products depend on *user's choice* of spectral model
- ⇒ **Data models must be able to capture and encode these ancillary products *and* metadata**
- Many current analysis techniques used in X-ray astronomy use Bayesian analysis methods that work robustly in the Poisson regime
  - These analyses may produce MPDFs for the parameters of interest or in some cases MCMC draws
- ⇒ **Data models must be able to capture and encode these representations of measurements *and* associated metadata necessary for their meaningful interpretation**
- Even quoting measurement and confidence limits don't forget that the *shape of the distribution* (e.g., Gaussian) and the *confidence level* (e.g., 68%, 95%) must be captured for the measurement to be useful