SPIE 2008 :: Observatory Operations :: 7016-23 :: 16:20 Wednesday 25 June 2008

How to handle calibration uncertainties in high-energy astrophysics

Vinay Kashyap

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How to handle calibration uncertainties in high-energy astrophysics

Vinay Kashyap

CHASC Astrostatistical Collaboration Chandra X-ray Center Smithsonian Astrophysical Observatory

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bottom line

there is now a way to include calibration uncertainty in astrophysical data analysis in a flexible way for any instrument, mission, or detector. Part I calibration uncertainty in data analysis

Part II practical issues of storage, retrieval, flexibility The three most important effects that affect data analysis astrophysical model uncertainty statistical uncertainty (measurement error) calibration uncertainty (systematic error)

examples of calibration uncertainty

power-law residuals with current calibration





power-law residuals with 20Å contamination overlayer

general model of HEA data

 $M(E', \mathbf{p}', t) = \int dE \, d\mathbf{p} \, S(E, \mathbf{p}, t; \theta) \, A(E, \mathbf{p}'; \mathbf{p}, t)$ $R(E, E', \mathbf{p}'; t) \, P(\mathbf{p}, \mathbf{p}', E; t)$

(E,p,t): photon energy, location, arrival time
(E',p'): detector channel, chip location
5: astrophysical source model
R: energy redistribution function (RMF)
P: position redistribution function (PSF)
A: effective area (ARF)
M: predicted model counts

effect on model parameter uncertainty



but how exactly?

MCMC









but where do the effective areas come from?



Principal Components



$$\begin{array}{c|c} \mathsf{new} & \mathsf{nominal} & \mathsf{bias} \\ \mathcal{A}'(E_j) &= \mathcal{A}_0(E_j) + \overline{\delta \mathcal{A}(E_j)} + \sum\limits_{k=1}^{N_{max}} r_k \ e_k \ \nu_k(E_j) \\ &+ r_{(N_{max}+1)} \ \xi(E_j) & \mathsf{PCs} \end{array}$$

 $A = A_0 + bias + components + residual$

Part II practical issues of storage, retrieval, flexibility

 $A = A_0 + bias + components + residual$



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3	0.24	0.25	134.755	0.591328	49.5941	51.6605					
4	0.25	0.26	156.063	0.689149	47.6866	49.5941					
5	0.26	0.27	175.412	0.726106	45.9204	47.6866					
6	0.27	0.28	196.854	0.706191	44.2804	45.9204					
7	0.28	0.29	57.5639	0.0294997	42.7535	44.2804	∇				
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what have we got so far?

- realistic error bars
- implemented in BLoCXS
- 500x speed up in analysis
- Sherpa on the way
- unified file format for use in XSPEC/Sherpa
- 100x drop in storage
- generalize to any instrument
- extendable to model lacunae
- roll your own

summary

there are a number of steps between a calibration scientist saying "the error on the effective area is X% at energy Y" to then have it fold into a spectral model fit and inflate the error bars on the parameters, and we believe that we have connected the dots.

uncertainty in energy response



one more thing..

what's next?

unified file format implemented in XSPEC/Sherpa other schemes of dimensionality reduction RMFs: 2D PCA, within and between PCA PSFs: multiscale residuals

The Three Poisson Model

Paul Edlefsen

A Dempster-Shafer Bayesian Solution to the Banff A1 Challenge