

A New Paradigm in X-ray Spectral Analysis

Deconvolving Spectra using Machine Learning

Carter Rhea Ph.D Candidate,¹ Jutie Hlavacek-Larrondo¹,
Akos Bogdan², & Ralph Kraft²

May 17, 2022

¹L'Université de Montréal

²Harvard Smithsonian Center for Astrophysics

Fonds de recherche
Nature et
technologies

Québec 



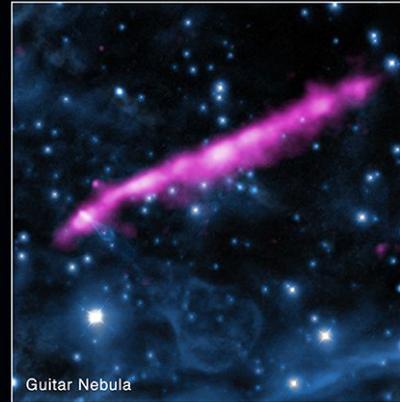
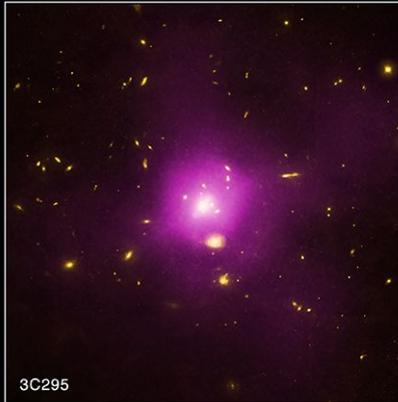
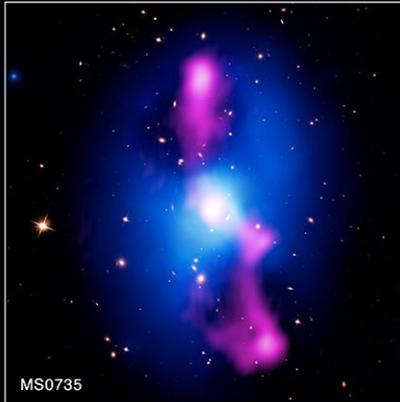
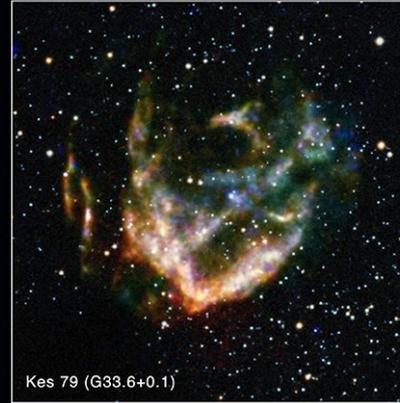
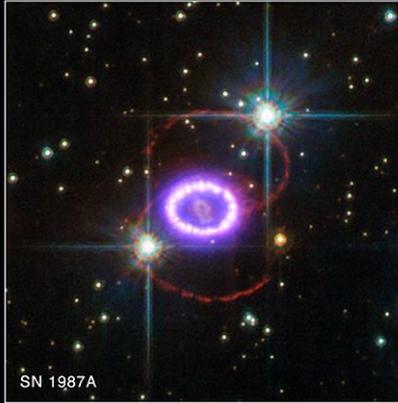
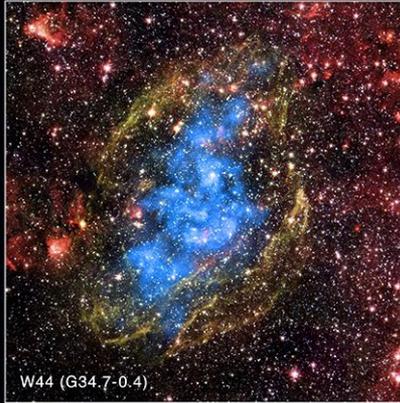
IVADO

Université 
de Montréal

CRAQ
Comprendre l'Univers
Understanding the Universe

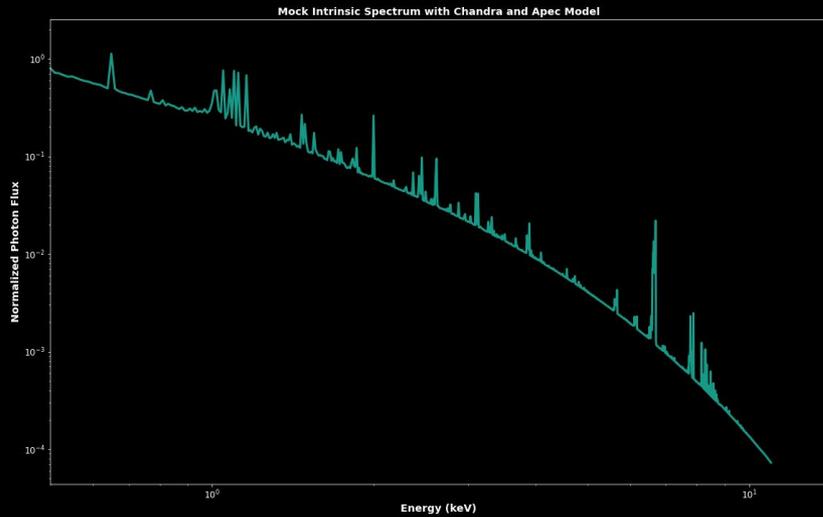


X-ray Astronomy

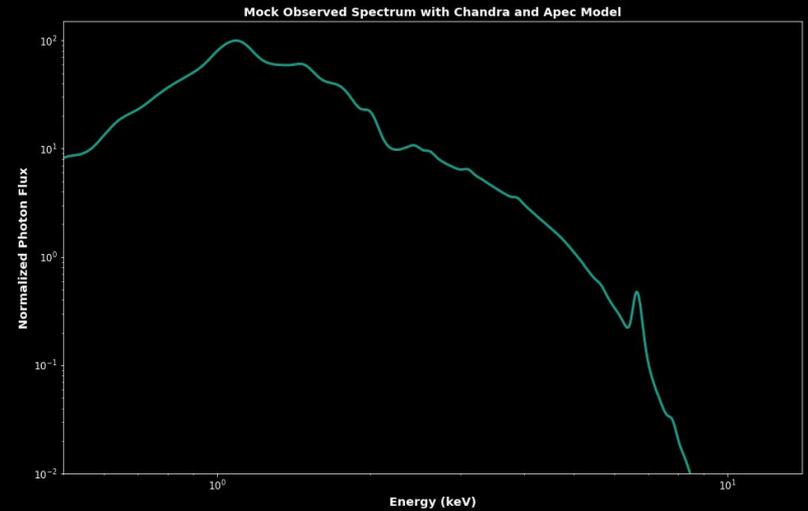


Observations vs Reality

The Source's Intrinsic Spectrum



What we observe

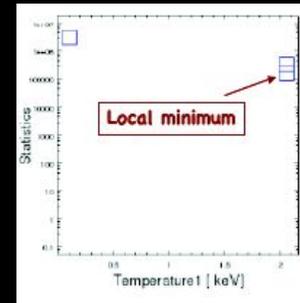


Why do we need this?

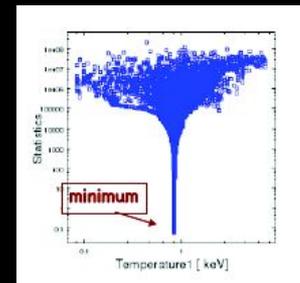
1. Initial Parameter Estimation for Traditional Fitting Methods

LevMar

Fit Space



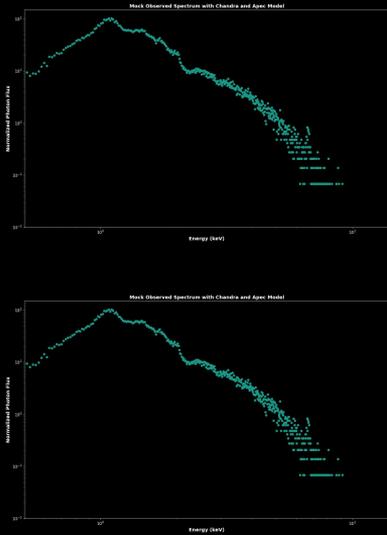
Monte Carlo



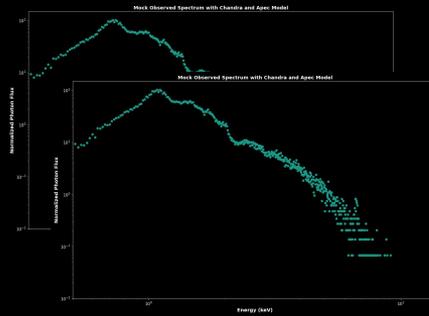
Why do we need this?

1. Initial Parameter Estimation for Traditional Fitting Methods
2. Stacking of X-ray Data

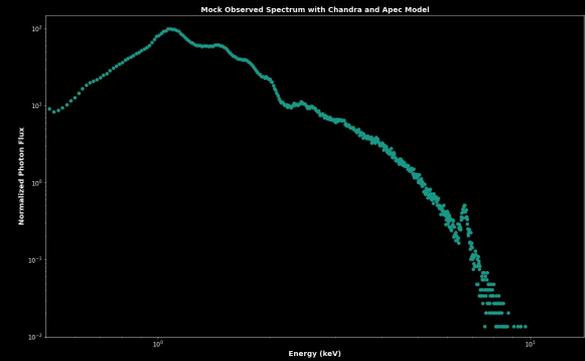
Individual Spectra



Combine Spectra



Stacked Spectra



Outline of Presentation

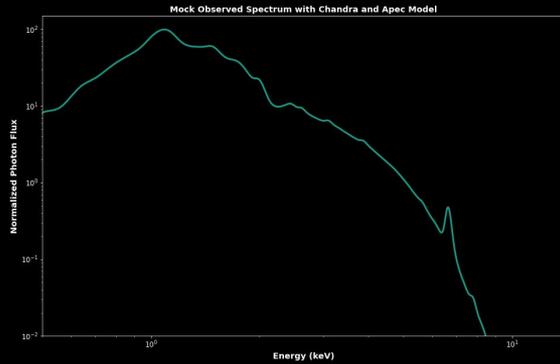
- Introduction
 - Formulation of observed spectrum
 - Effect of Response Matrix in Observations
 - Previously Attempted Solutions
- Recurrent Inference Machine
 - Introduction
 - Basic Example
 - In the Context of X-ray Spectra
- Next Steps

Introduction

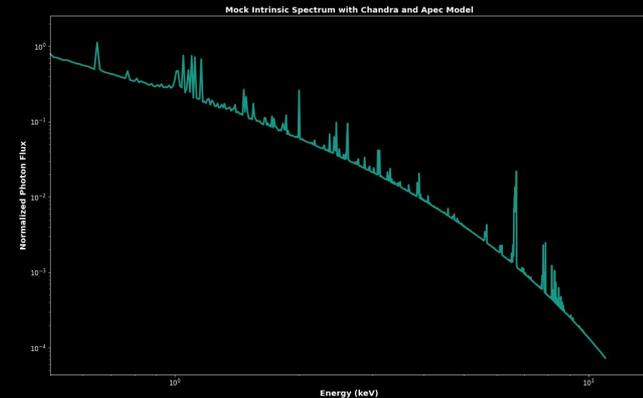
Potential Solutions for the Response Matrix

E = Photon Energy Space
 E' = Detector Energy Space

$$S_{obs}(E) = \int_0^{\infty} R(E', E) S_{true}(E) dE$$



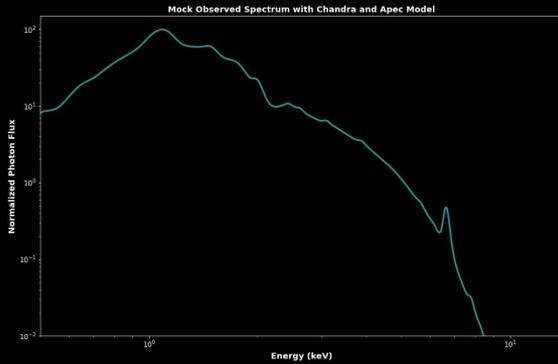
=



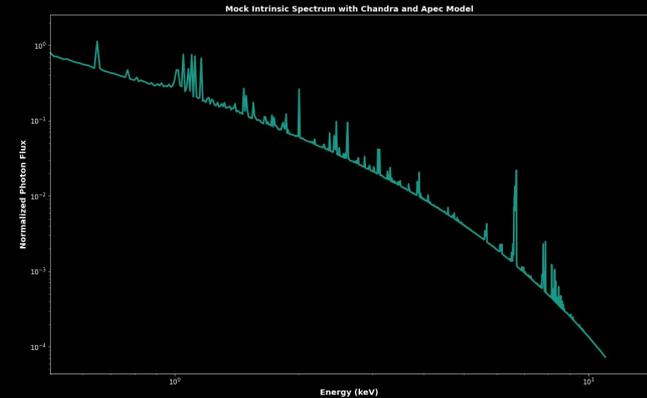
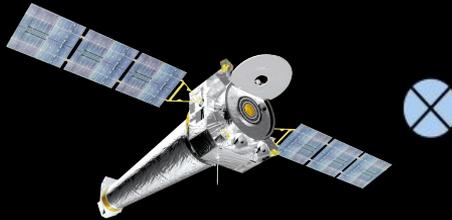
Potential Solutions for the Response Matrix

i = Photon Energy Space
 j = Detector Energy Space

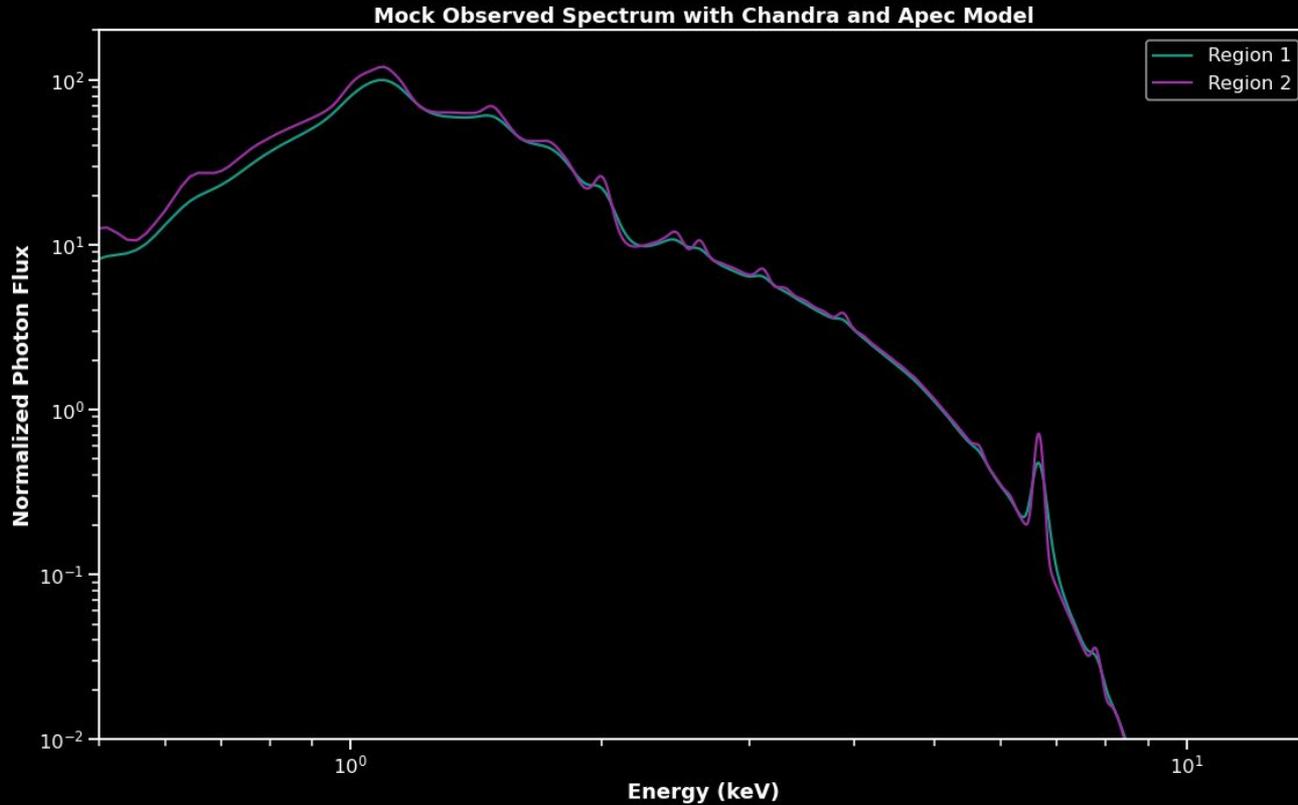
$$S_{obs_i} = \sum_{ij} R_{ij} S_{true_j}$$



=



Effects of Response Files



Solving the Matrix Equation

$$R S_{true} = S_{obs}$$

Pseudo-Inverse

Tikhonov Regularization

DOES NOT WORK

$$S_{true} = R^\dagger S_{obs}$$

$$\|R S_{true} - S_{obs}\|_{L_2} - \mu \|D\|_{L_2}$$

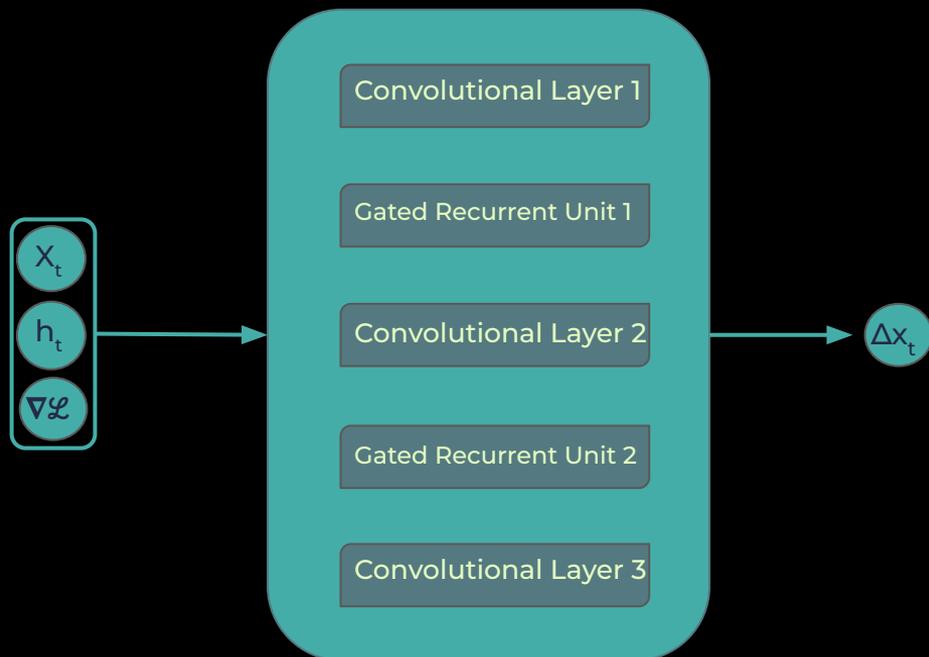
$$S_{true} = [R R^T + \mu I]^\dagger [R^T S_{obs}]$$

Recurrent Inference Machine

Recurrent Inference Machines

How does a Recurrent Inference Machine work:

Solve the linear equation $Ax=b$ iteratively by using an RNN to update solution



Putzky & Welling 2017
arxiv.org/abs/1706.04008

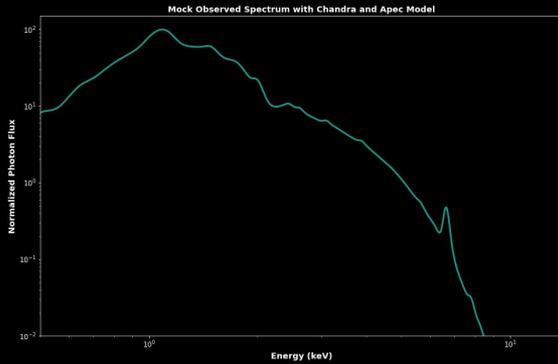
Morningstar et al. 2018
arxiv.org/abs/1808.00011

Morningstar et al. 2019
arxiv.org/pdf/1901.01359.pdf

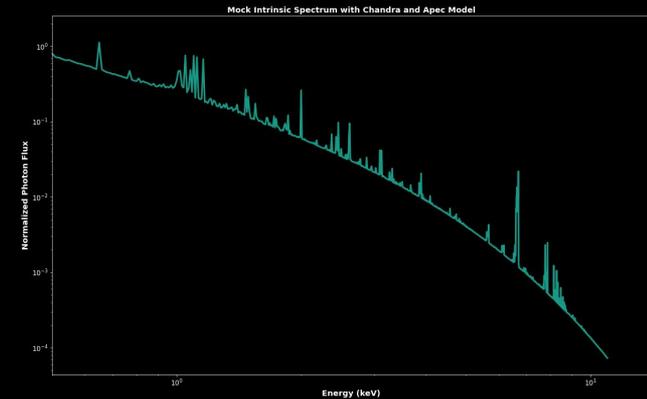
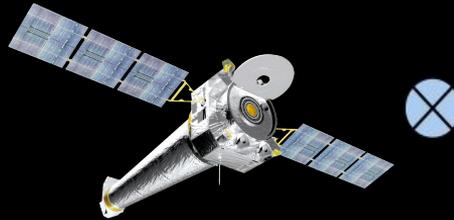
A quick reminder about our problem

i = Photon Energy Space
 j = Detector Energy Space

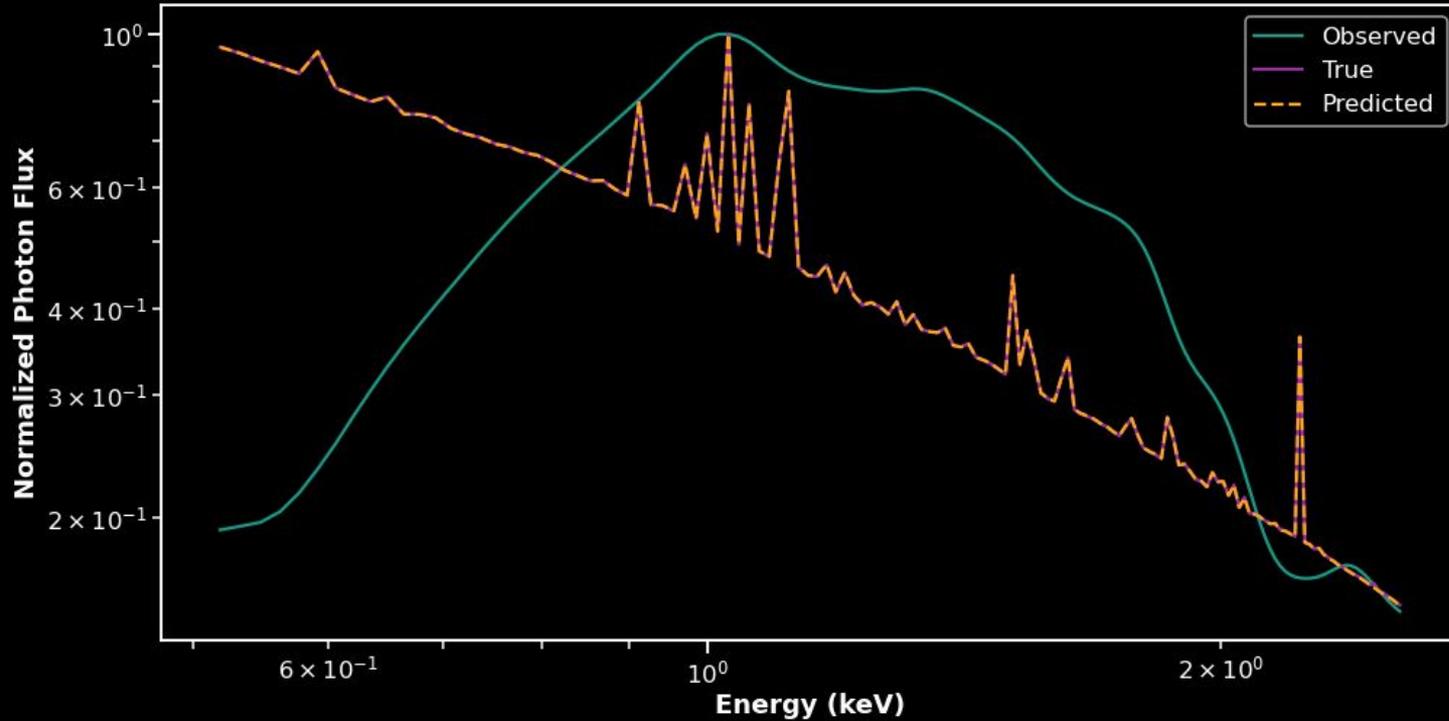
$$S_{obs_i} = \sum_{ij} R_{ij} S_{true_j}$$



=



Application to X-ray Spectra



What can we do with this?

- Train a convolutional neural network to estimate the underlying parameters (Rhea et al. in prep.)
- Explore the transiency of X-ray sources (Rhea et al. in prep.)
- Study the calibration of the Chandra X-ray Observatory (Prunier & Rhea et al. in prep.)
- Investigate Metallicity in the outskirts of galaxies

A New Paradigm in X-ray Spectral Analysis

Deconvolving Spectra using Machine Learning

Carter Rhea ^{Ph.D Candidate, 1}, Julie Hlavacek-Larrondo¹, Akos Bogdan², & Ralph Kraft²
May 17, 2022

¹ L'Université de Montréal

² Harvard Smithsonian Center for Astrophysics

Fonds de recherche
Nature et
technologies

Québec 



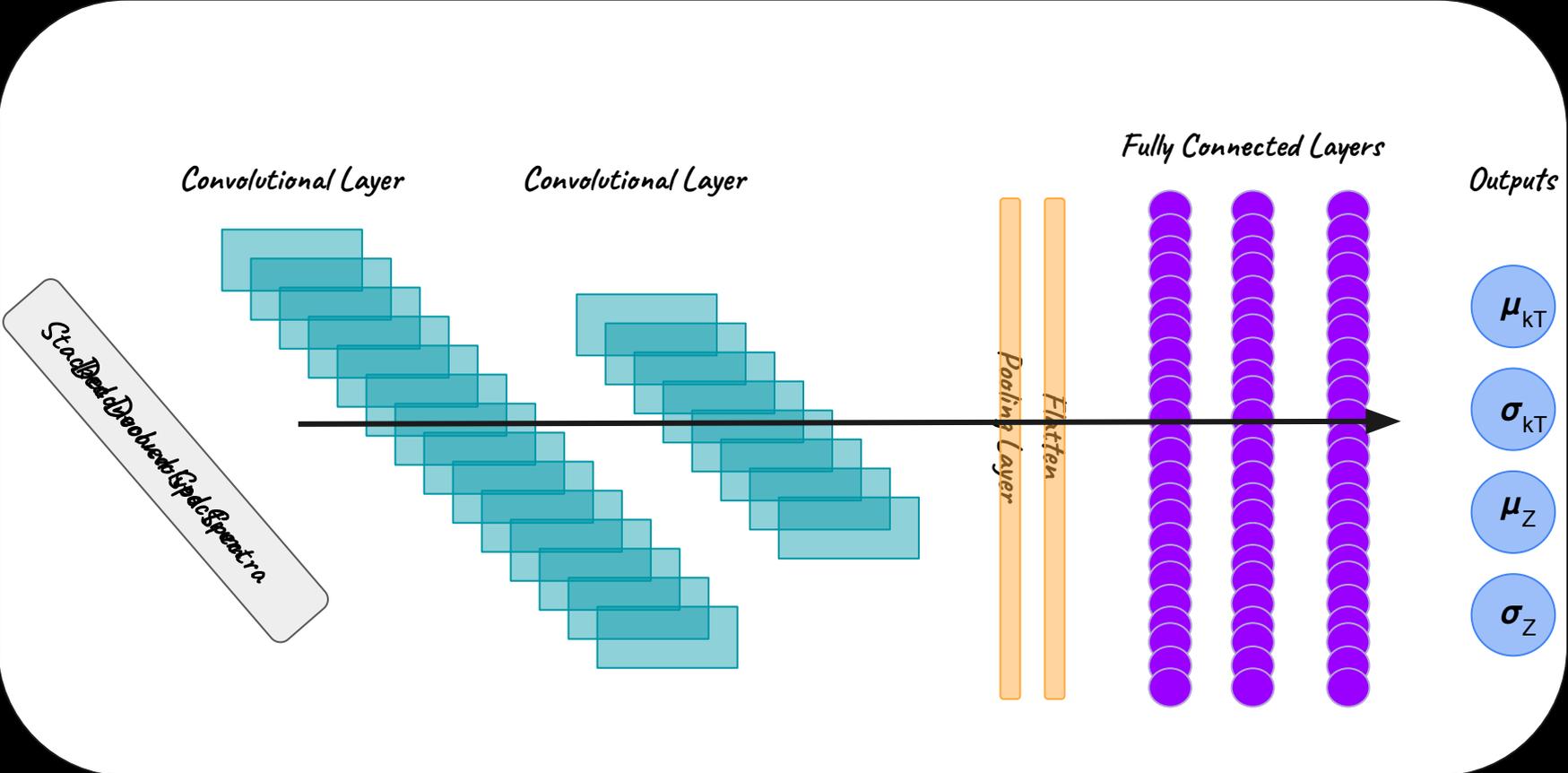
IVADO

Université 
de Montréal

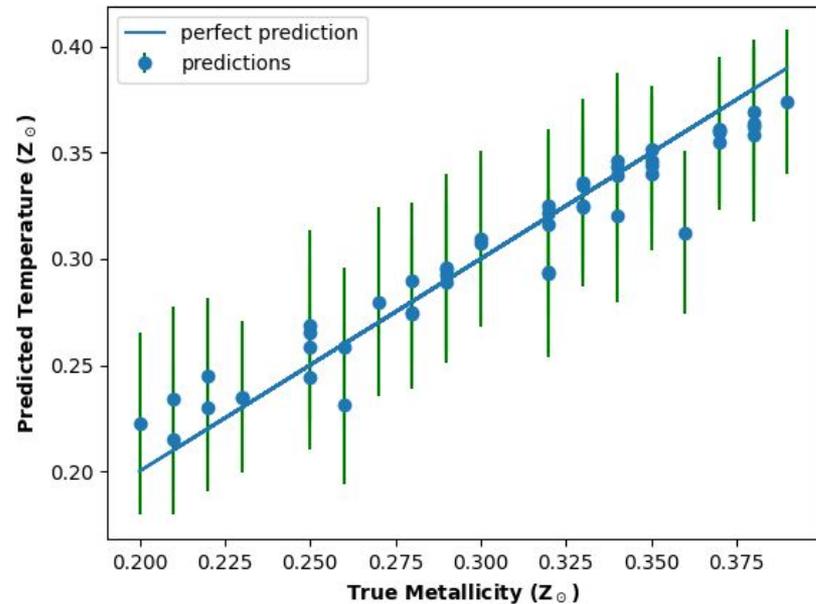
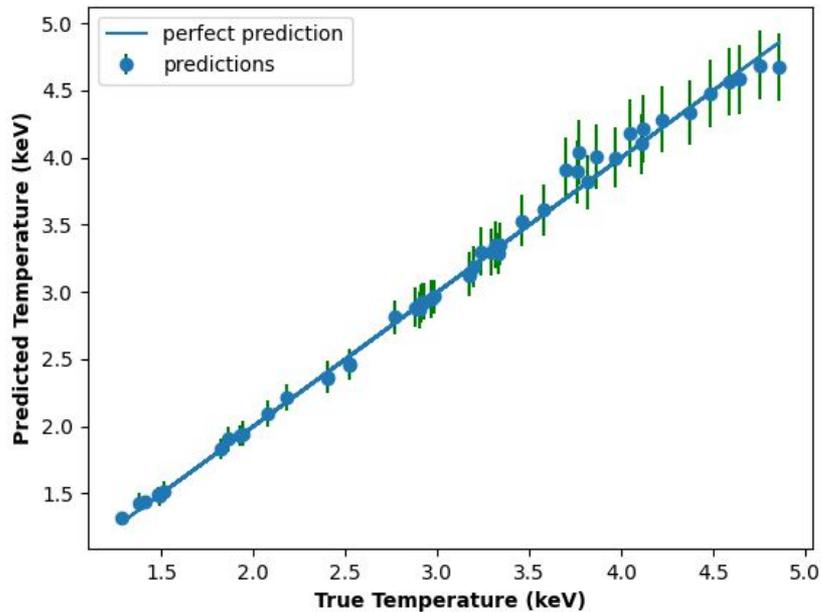
CRAQ
Comprendre l'Univers
Understanding the Universe



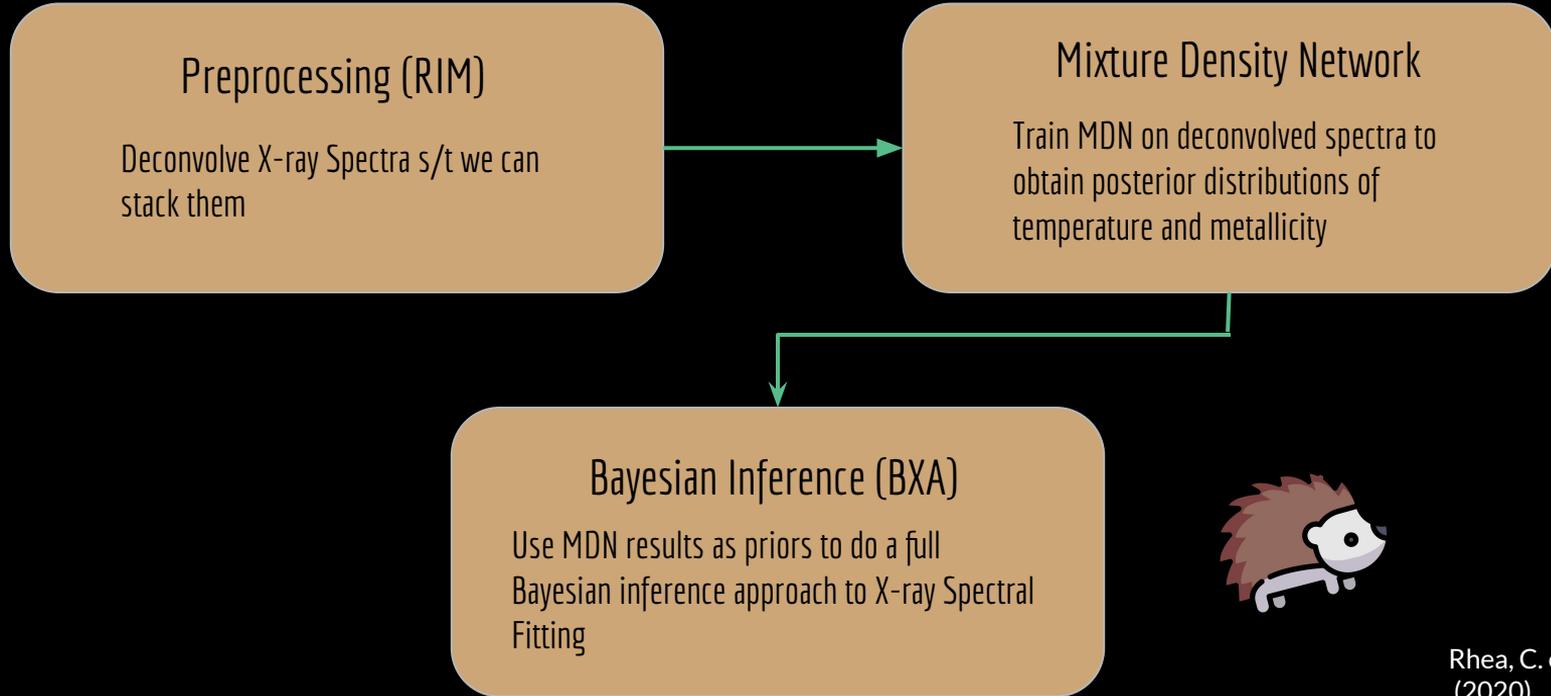
Next Step: Estimate Temperature and Metallicity



Estimate Temperature and Metallicity



Putting it all together!



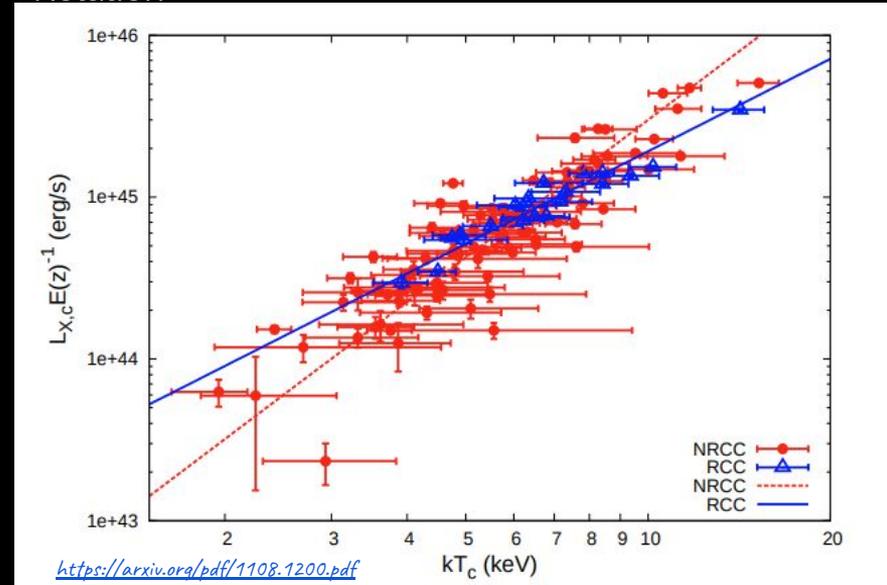
Rhea, C. et al, AJ 160, 5 (2020)

Buchner, J. et al, A&A 564, A125 (2014)

Next Steps

1. Converge RIM on larger dataset
 - a. Redo CNN analysis on larger dataset
2. Apply to real data (Self-Similarity)
 - a. Reduce Chandra Data
 - b. Use RIM to deconvolve
 - c. Apply CNN to estimate parameters
3. Potential other ideas?

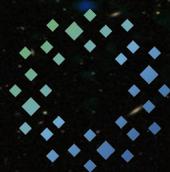
Galaxy Cluster X-ray Luminosity-Temperature Relation



A New Paradigm in X-ray Spectral Analysis

Deconvolving Spectra using Machine Learning

Carter Rhea, Julie Hlavacek-Larrondo, Akos Bogdan, & Ralph Kraft
April XX, 2022



IVADO

Université 
de Montréal

CRAQ
Comprendre l'Univers
Understanding the Universe

