# Emulating Photon Pile-up Effects on X-ray Spectra with a Neural Network

Justina R. Yang

with Daniela Huppenkothen, Rafael Martínez-Galarza, Vinay Kashyap, Douglas Finkbeiner, Aneta Siemiginowska

> Wednesday, 6 November 2024 justina.yang@cfa.harvard.edu

#### X-ray telescopes enable us to examine distant objects



X-RAY

Cygnus X-1: > 6000 light-years away (original image: NASA/CXC/M. Weiss) 2

#### X-ray telescopes enable us to examine distant objects



Cygnus X-1 is a black hole X-ray binary (original image: NASA/CXC/M. Weiss)

#### X-ray data arrives in event lists from which we can extract spectra



Images: NASA/CXC/M. Weiss, NASA/CXC/MPI/Brusa et al. 2005

X-ray spectra contain information about astrophysical object properties





### Photon pile-up

is an instrumental effect related to detector frame rate that is common in X-ray detectors. It affects observed events and distorts spectra.

Accurately understanding pile-up is important for measuring physical properties such as black hole spin.





Image: "The *Chandra* ABC Guide to Pileup"



Image: NASA/CXC/M. Weiss, NASA/CXC/SAO/Miller et al. 2002

### Chandra X-Ray Observatory, ACIS-S instrument



and produces an electric signal

linearly related to photon energy

The ACIS focal plane layout in 'sky' coordinates with each CCD labelled with its identification number.

## Photon pile-up occurs when a detector registers multiple photons as a single event



### Pile-up affects both events and spectra

#### Pile-up causes

- energy migration (events are associated with higher energies)
- decrease in event count rates
- grade migration (events may be assigned "grades" of poorer quality and may be rejected)



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#### Pile-up is only one part of the instrument response



## Parts of the instrument response are described by the ARF and RMF, which are well-understood and calibrated



**Detector sensitivity** is parametrized by the **ARF** (ancillary response file): includes effective area, quantum efficiency



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## Pile-up is an important effect to model, but it is fundamentally photon-level and is difficult to model analytically



#### The current standard is the Davis 2001 pile-up model

Incident spectrum Detector response: ARF Pile-up Detector response: RMF Output spectrum

Davis 2001:

Probability (piled-up combination of n photons has a good grade, given that the combination of photons 1, ..., n - 1 had a good grade) =  $\alpha$ 

- Often, analysis assumes  $\alpha = 0.5$
- Statistical model does not hold for severe pile-up
- Standard method, e.g. implemented in MARX ("Model of AXAF Response to X-rays"), a ray-trace Monte Carlo simulator of Chandra observations.

## Can a neural network emulate a simulated version of pile-up, and eventually the empirical *Chandra/*ACIS pile-up?

 Neural networks are good function approximators and may be able to directly learn the function that distorts the idealized spectrum into an observed spectrum



 Neural networks can be evaluated much faster than MARX simulations

### Neural Network Model



A trained neural network can be incorporated into different parameter inference methods



## Details of the neural network

Training data: >17,000 pairs of (incident spectrum with instrumental response, MARX-simulated spectrum with pileup and observational noise)

### Incident spectrum with instrumental response



#### The neural network can closely reproduce a MARXsimulated piled-up spectrum



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## Preliminary work using the neural network in an MCMC is promising

- Parameter inference with **emcee** MCMC code (Foreman-Mackey et al. 2012)
- Optimization quantity = log(P) = log[product over channels k of Probability (MARX counts in channel k, given Poisson mean = NN-predicted counts in channel k)]



## Preliminary work using the neural network in an MCMC is promising



#### Preliminary work using the neural network in an MCMC is promising



## Future work: can the neural network emulate the empirical *Chandra/*ACIS pile-up?

Transfer learning to real Chandra observations with HETG data:



Images: Canizares et al. 2005, TGCat (Huenemoerder et al. 2011)

- Observations taken with *Chandra*/HETG (High-Energy Transmission Grating) distribute photons across more pixels and reduce pile-up in dispersed spectra. We will use 1st-order dispersed spectra as a proxy for unpiled "ground truth" spectra.
- New training data set: ~ 1000 pairs of (unpiled 1<sup>st</sup>-order HETG spectra, piled-up ACIS-S spectra)

### Possible extensions of work

- Relevance for other non-CCD detectors
  - E.g. transition edge sensor (TES) microcalorimeters have a different pile-up effect, related to finite time resolution



### Conclusions

- A simple, fully-connected neural network can emulate a simulated version of pile-up.
- We are investigating how to use it for robust and accurate parameter inference from piled-up data.
- Future work will include transfer learning with ~1000 Chandra observations.
- Does your data experience a distortion like the Chandra/ACIS version of pile-up?

#### **Neural Network Prediction**

