

Surojit Saha Graduate Student Institute of Astronomy National Tsing Hua University, Taiwan

7th KAGRA International Workshop, 18-20 December 2020 @ National Central University, Taiwan



國立情華大學

19th December, 2020

NATIONAL TSING HUA UNIVERSITY

Collaborators







Surojit Saha Prof. Albert Kong (Supervisor)

Laurence Datrier Michael Williams Prof. Ik Siong Heng Prof. Martin Hendry Dr. Daniel Williams Fergus Hayes Nicola De Lillo **Dr. Matt Nicholl**



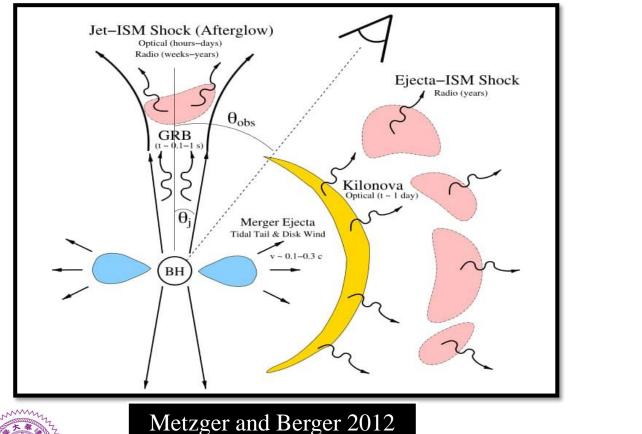
Outline

- Background and Motivation
 Planning
 Data Used
 Roadmap
 Results
- Discussion

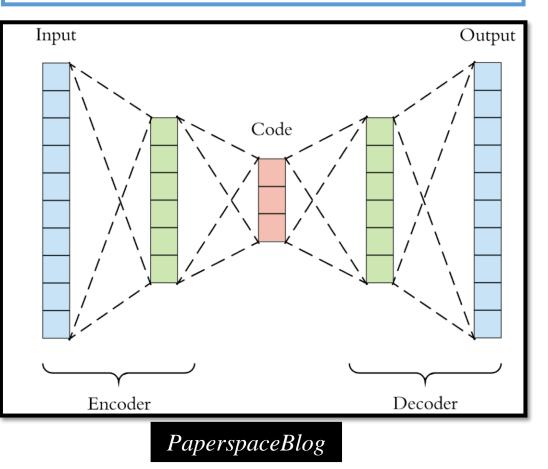


Background and Motivation

Kilonova: Transient electromagnetic counterpart resulting from merger of NS-NS or BH-NS, powered by the radioactive decay of *r-process* nuclei.



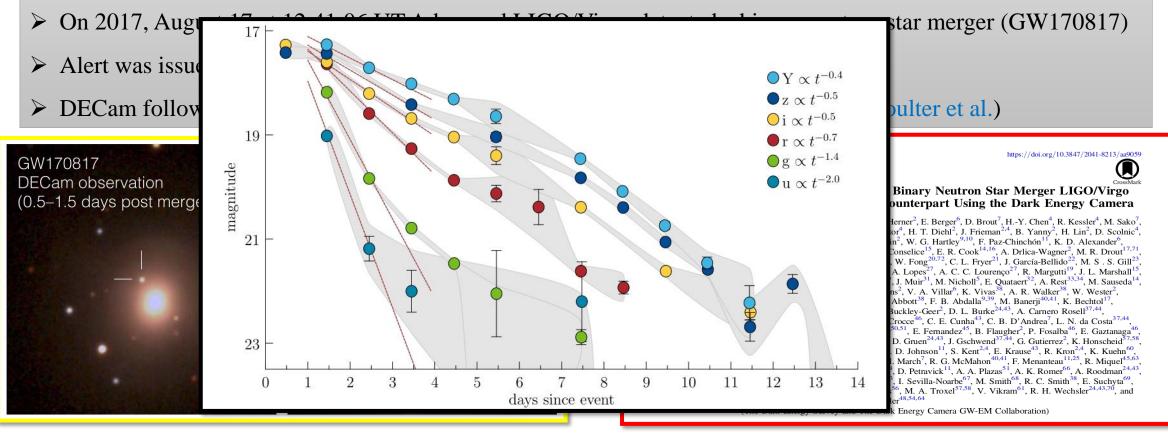
Autoencoder: Unsupervised learning technique where neural networks are used for representation learning.





Background and Motivation

Birth of Multi-Messenger Astronomy



NGC4993 grz color composites

Kilonova from follow up observation of GW170817

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Follow up observations: HST, GS/F2, Swift/UVOT

Motivation and Planning

Looking For Some Answers

How to relate the ejecta parameters with the kilonova (KN) light curves (LC) from the follow-up observations using autoencoder?

➢ Is it possible to generate KN LC based on certain ejecta parameters, hence helping in better parameter estimation?

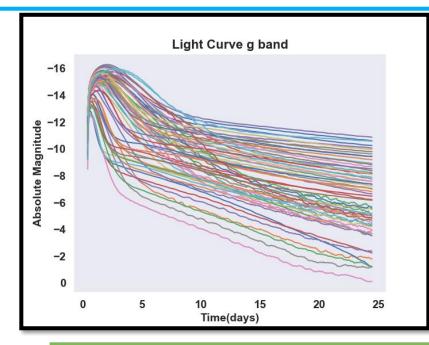
Build an autoencoder (AE) model of KN using some simulated data and reconstruct the LC, learn the physical parameters, use the model to extract KN LC from other data (if present).



Data Used

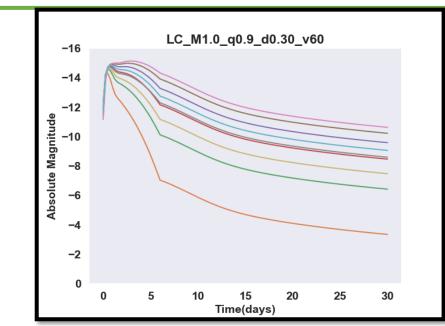
Data 1

- Simulated LC using Kasen Model
- Each LC has its own dynamical ejecta properties
- Time, Absolute magnitude, ejecta mass, ejecta velocity, lanthanide fraction.
- LC curves are available in *g*,*i*,*r*,*y* and *z* bands of DECam.



Data 2

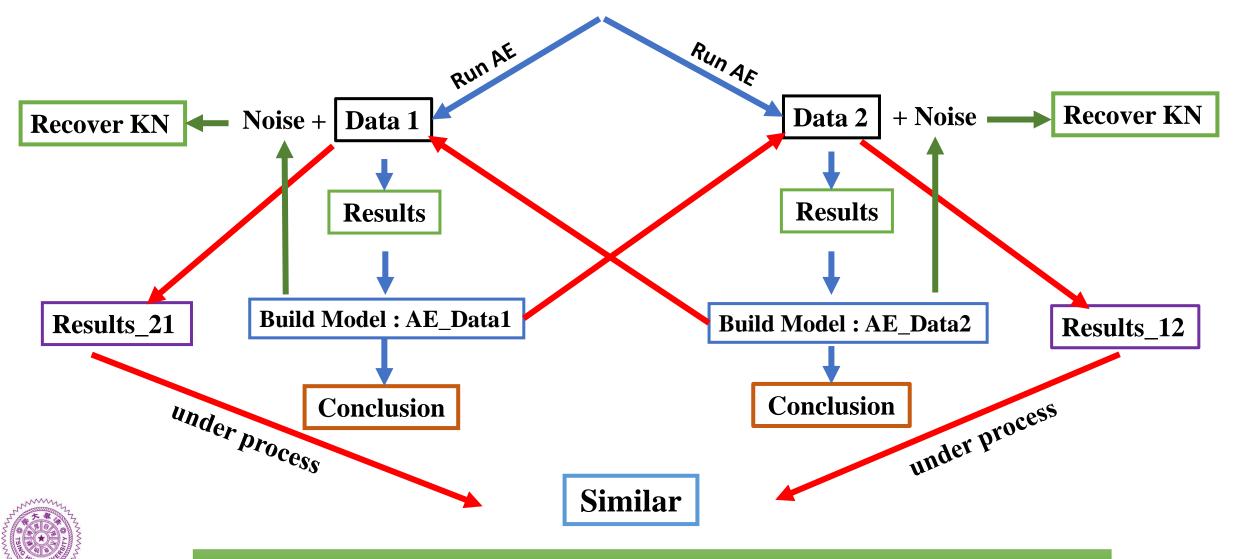
- Simulated from his own KN model.
- LC are generated based on chirp mass, mass ratio, viewing angle fraction of the ejected remnant disk.
- Each data set has a particular value for the above parameters and has values for *LSST* (*u*,*g*,*r*,*i*,*z*,*y*), *PanSTARRS* (*w*) and *ATLAS* (*c*,*o*) bands.

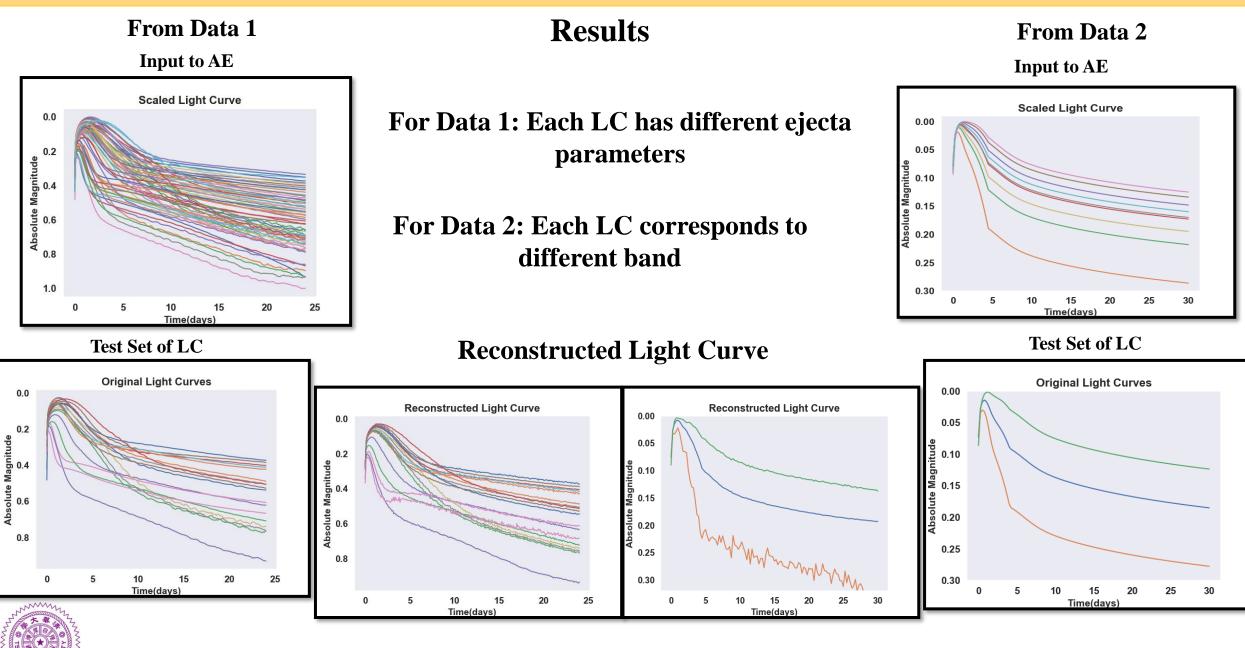




Roadmap

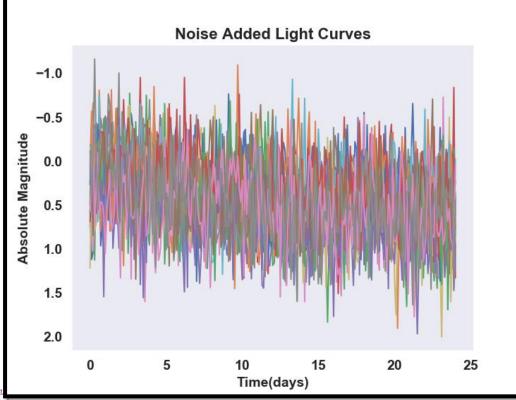
Build The Autoencoder

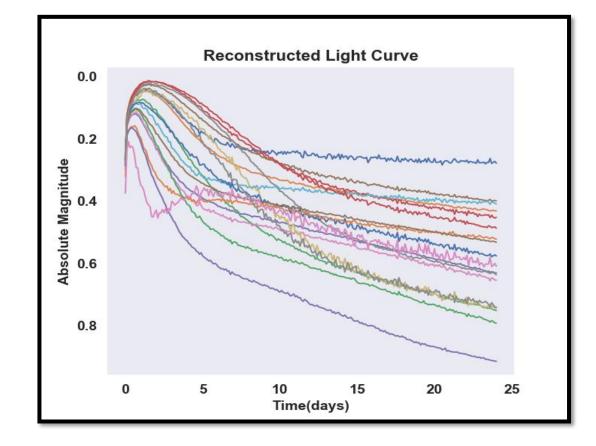




Model Test With Denoising Autoencoder

- To test the validity and accuracy of the model, denoising autoencoder has been used
- Random noise has been added to the data
- Model is run on it to extract the LC.







Discussion

- > Currently the autoencoder provides quite satisfactory results in reconstructing the LC
- \succ In some cases there are still some issues in reconstructing LC
- Even though the noise added to the is not physical, but the recovery shows good results with the model-test run.
- > The model built AE_data1 and AE_data2 depends on the shape of the input.
- > More realistic noise need to be added for achieving better physical significance.
- \succ It is possible to recover hidden KN from a noisy data set.



Future Works

 \succ Make the autoencoder model more general.

Add GRB afterglow data (generated with suitable time steps) to the autoencoder to perform the same analysis.

Generate KN LC based on physical parameters (long term!!!)

That's Great

I am Useless

Thank you for your Attention

