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A Framework For Rapid Parameter Inference of Kilonova Light Curves: Bayesian-Machine Learning Approach

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Parameter Inference Accelerated by 10³ times



INTRODUCTION: **Parameter inference** of the astrophysical events is an important aspect in concluding the proposed model while comparing it with the existing observation. In the domain of **multi-messenger astronomy** (MMA) associated with **gravitational waves** (GW), low-latency analysis forms an integral part. Low-latency parameter inference of the observed EM signal is thus beneficial for further studies. Here, in this work, we provide a framework for **rapid parameter inference of KNe** using a machine learning tool popularly known as **conditional variational autoencoder** (CVAE).



Schematics of the two combined morphologies of KNe model

Image Credits: Wollaeger et al, 2021



Saha et al. [in prep]

mw

Objective

Perform Rapid Parameter Inference Utilizing Bayesian-Machine Learning Framework



md: mass of low Y_e componentvd: velocity of low Y_e componentmw: mass of high Y_e componentvw: velocity of high Y_e component

and the decoder. The encoder compresses the multidimensional input and provides a data distribution in the latent space. The decoder extracts the distribution from the latent space to reconstruct the results.



Fig 1: The violin plot shows the comparison between the true and generated data for a set of g-band light curves.



Fig 2: The KDE plot shows the distribution of the true and generated data for a set of g-band light curves.



Fig 3: This polar plot shows the overlapping KDEs of all the physical parameters across all the filter bands. There is quite a good agreement between the true and the generated physical parameters.

Future Works: To include more KNe models to provide the most generalized PE.

References: 1. Wollaeger et al., The Astrophysical Journal, 918:10, 2021 September
2. Kawaguchi et al., The Astrophysical Journal, 825:52, 2016 July 1.
3. Saha et al, The Astrophysical Journal, Vol 961, Number 2