

# Judging an excessive power as glitch or core-collapse supernova

With machine learning

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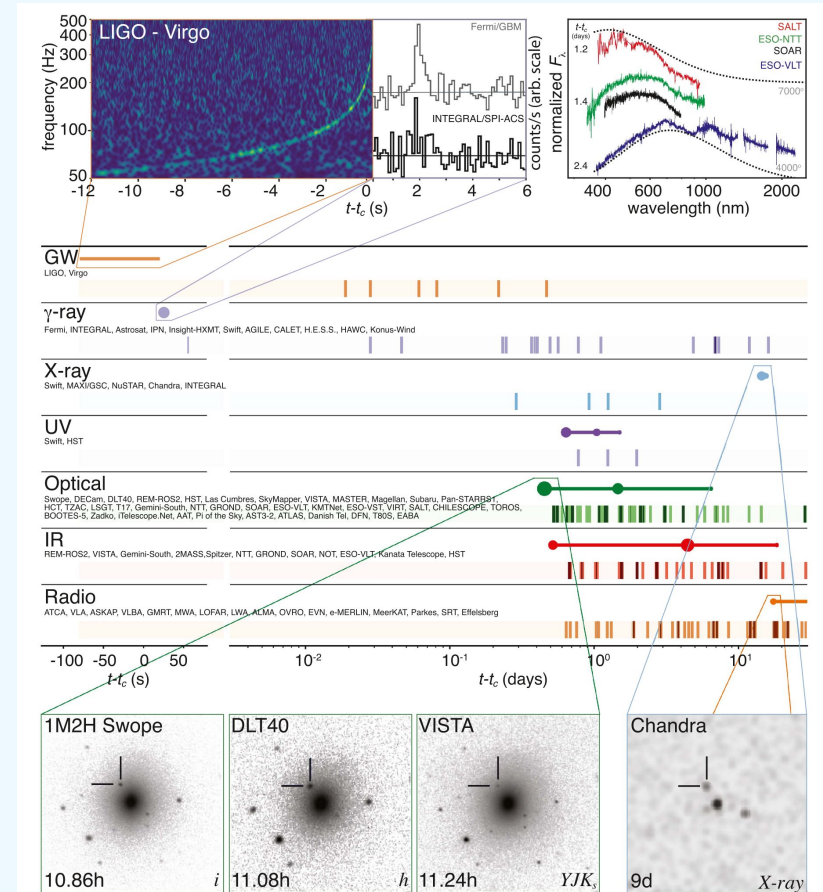


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  - Introduction to the ML4GW framework
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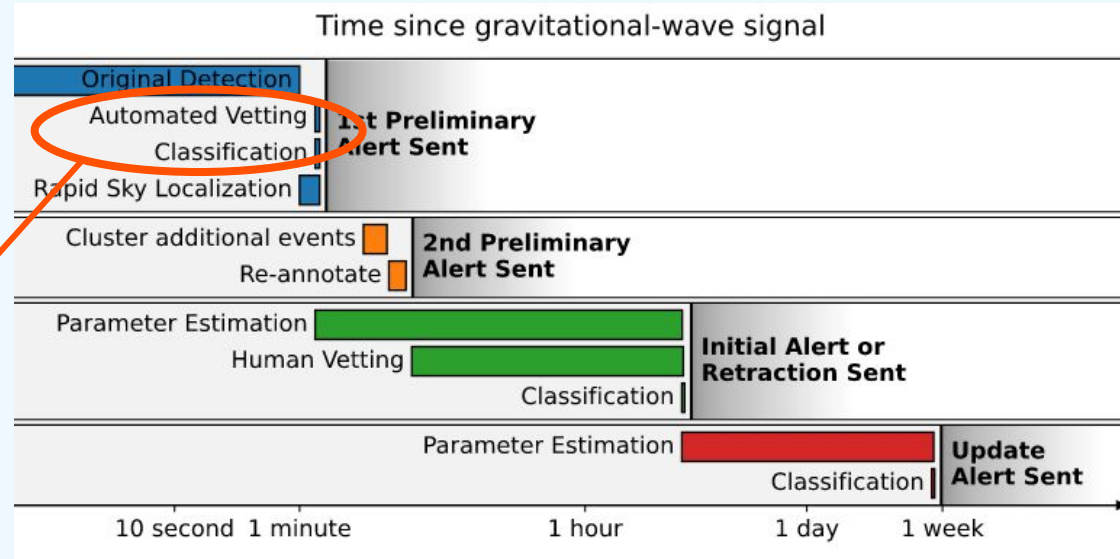
# MMA Introduction

- Early alert from the GW detector can provide prior information for EM counterparts.
- Possible candidate for MMA includes BNS merger, Core-Collapse Supernova (CCSN) etc.



# MMA Introduction

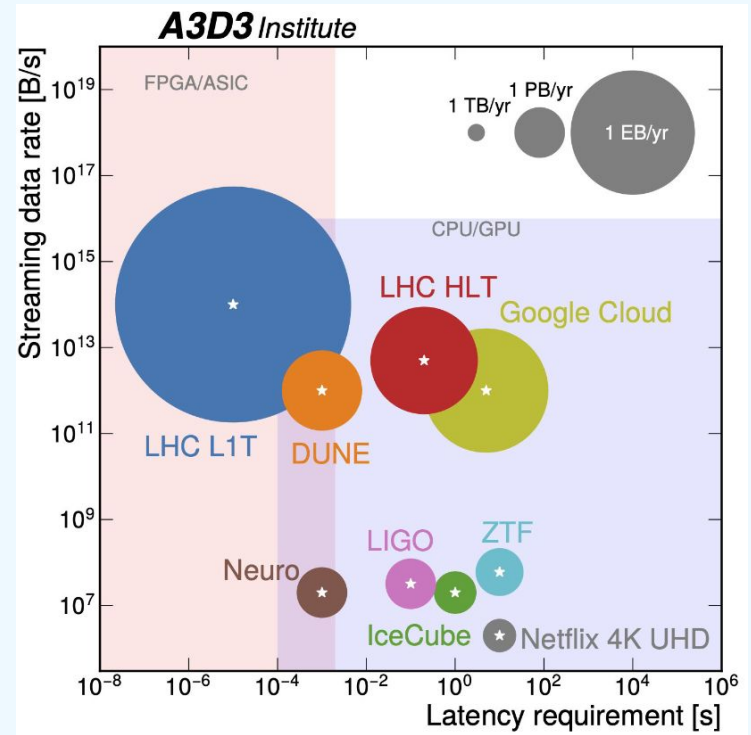
- We focus on recognizing the difference of glitch and GW signals after the detection



<https://rtd.igwn.org/projects/userguide/en/latest/analysis/index.html>

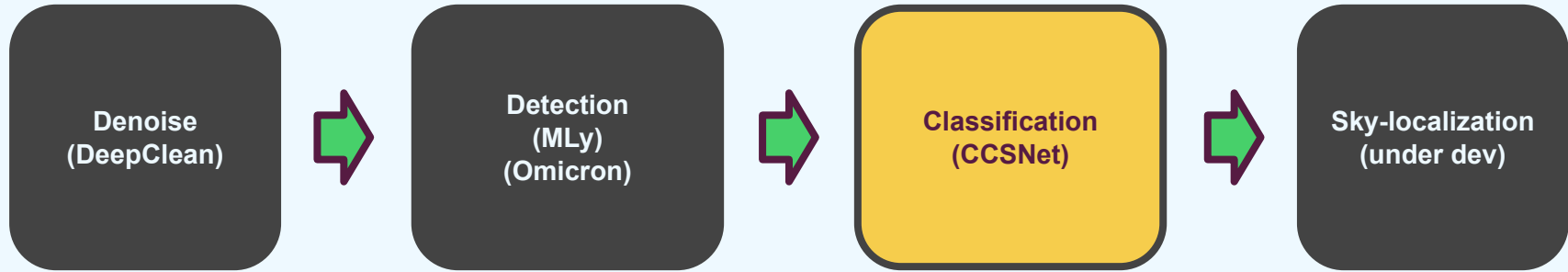
# MMA Introduction

- GW Inference-as-a-Service(GW-iaaS) provide efficient machine learning usage for speedy and accurate analysis.
- GW-iaaS:
  - Hardware: Nvidia Triton inference server
  - Software: ML4GW packages



A3D3, A3D3

# Example of Machine learning Alert system

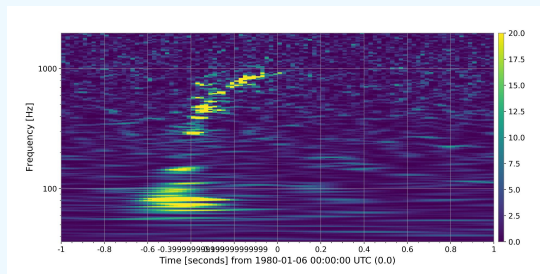
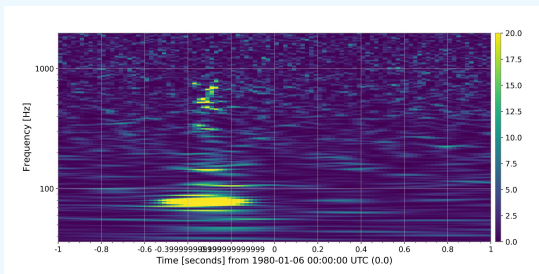


- DeepClean: Non-gaussian noise reduction according to environmental noise [2005.06534](#), [2108.12430](#)
- MLy: Machine Learning based coherent detection method [2009.14611](#)
- CCSNet: Detection trigger classifier for glitch vs CCSN
- Sky-location: Still Under development

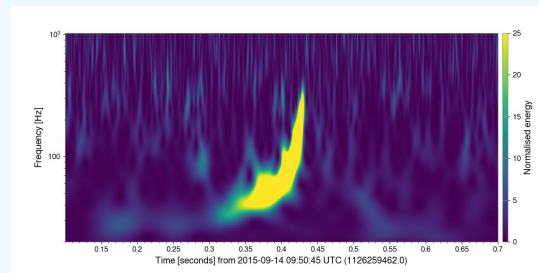
# Data preparation

# Signals in detector

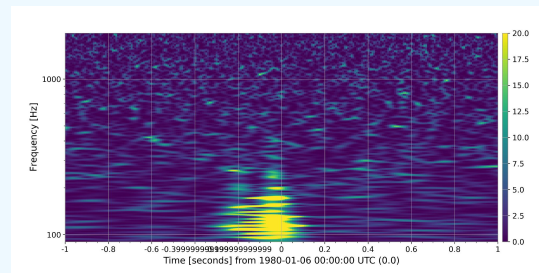
## CCSNs



## BBH



## Glitch



Unlike CBC's chirp like structure in time frequency space  
CCSN waveform have various time frequency structure



# Distribution of CCSN

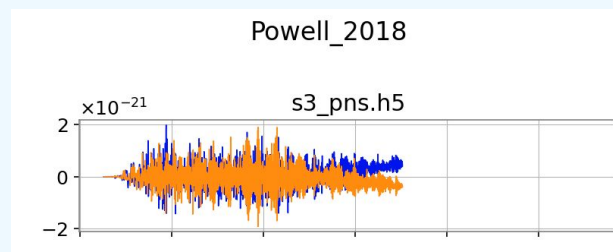
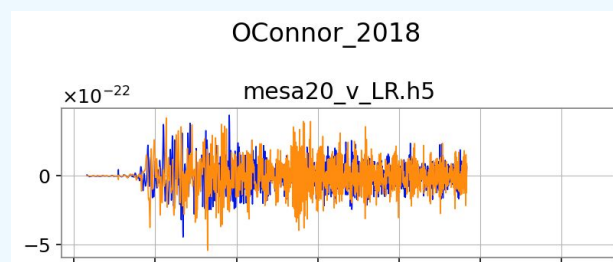
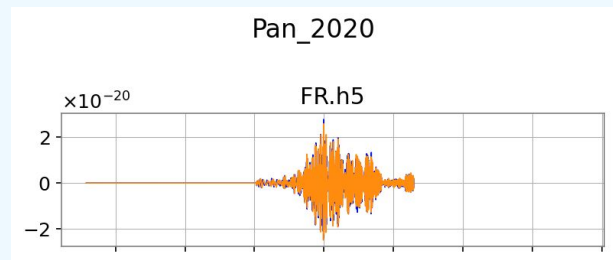
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## Generating the two polarizations

- 34 different progenitor star (3D simulation)
- Uniform distribution for orientation, polarization, and sky-location in its parameter space

## Injection to Handford, Livingston

- 4096 sampling rate
- Core bounce at different time
- **Dynamic SNR**



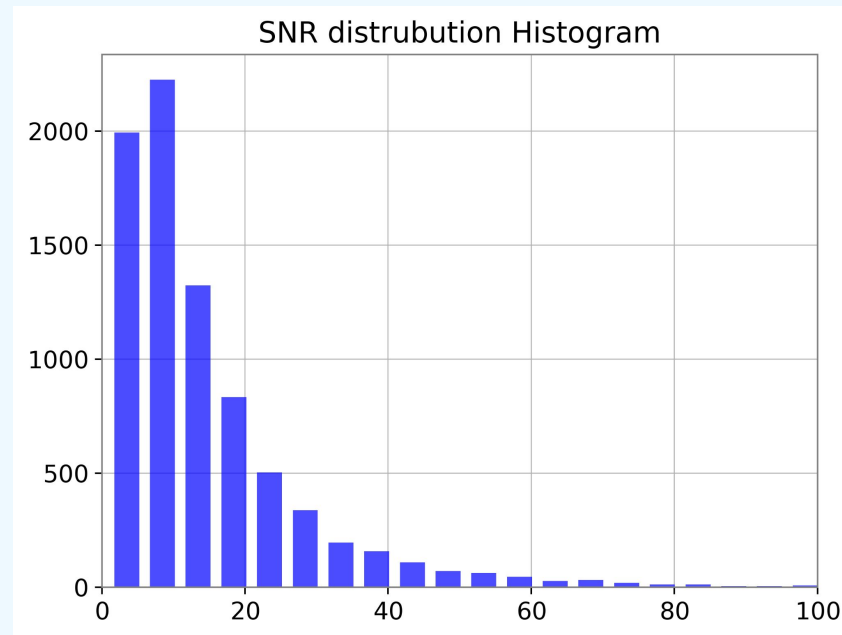
# Distribution of CCSN

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ML4GW provide fast SNR rescaling during training

Log normal SNR distribution:

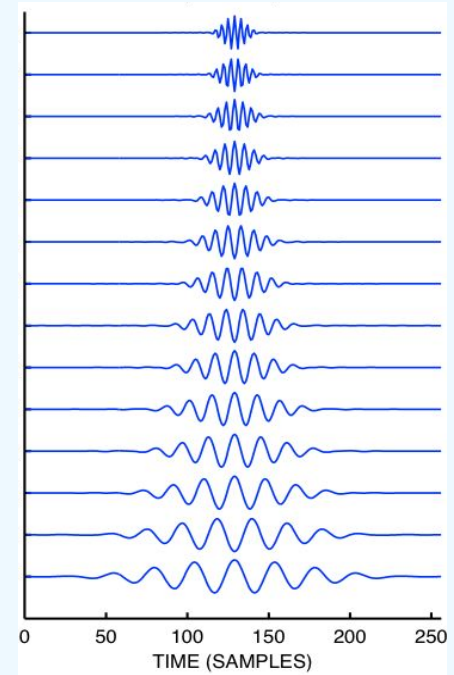
- Mean = 15
- Variance = 15
- Low bound cut off = 1



# Collecting Glitch using Omicron

- Omicron pipeline pick up “excessive power” using Q-Transform
- Q-Transform are similar to applying wavelet transformation to whitened signals
- Wavelet transformation is like a modified Fourier transformation, it replace the sine wave with a sine Gaussian wavelet for integration.

Wavelets with same q value

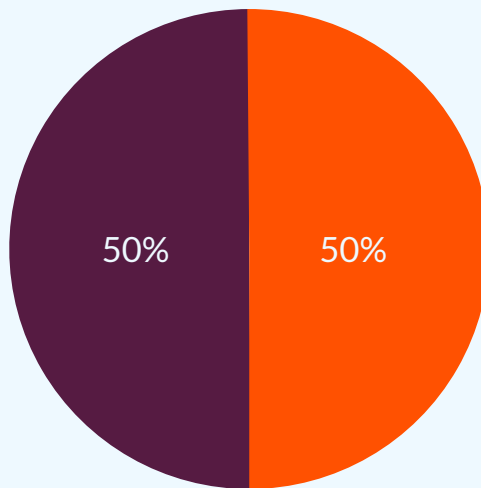


# Training & Validation Data

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- Glitch from 53K seconds of strain data from O3
  - Hanford 8K
  - Livingston 12K
- 4096Hz, 1 Second, 2 IFOs, whiten strain data
- 16K of CCSN injection in total for training and validation
- 75% for training 25 % for validation

Glitch appears in at least one detector: ■  
Signal on normal noise: ■

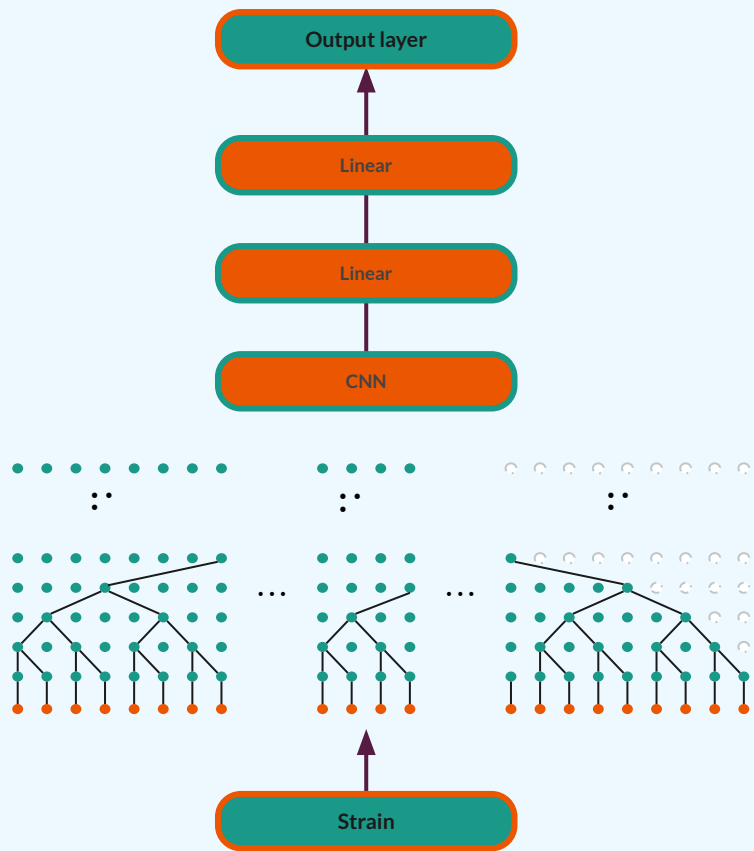


# The ML model and training method

# WaveNet

We use dilation feature from WaveNet

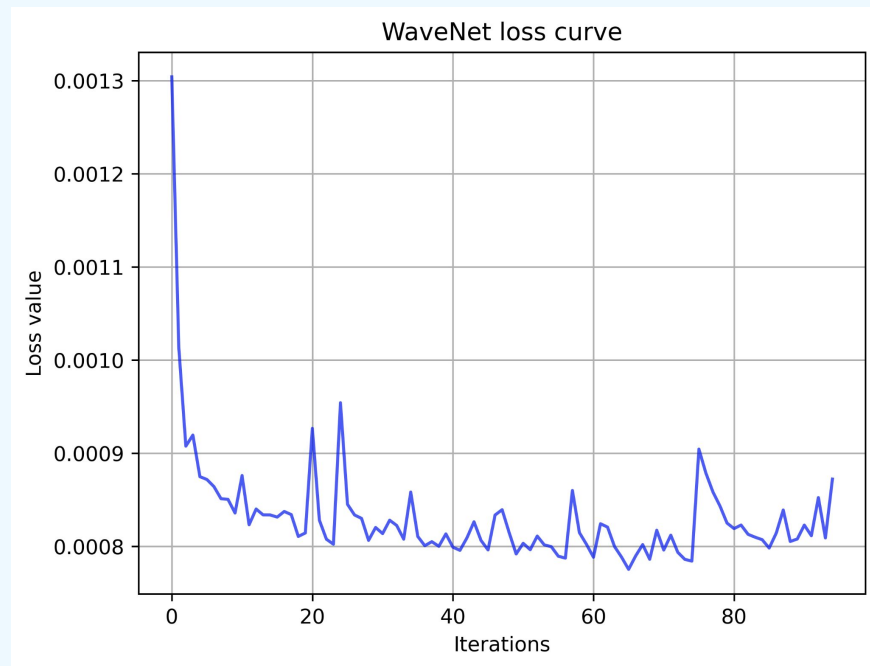
- The dilation feature provide a larger receptive field for deep CNN layers
- For deep layer of CNN less neuron is required to capture same kernel width of the input data



# Training performance

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- Loss function: Cross Entropy
- Optimizer: Adam
- Scheduler: Learning rate schedule every  
batch

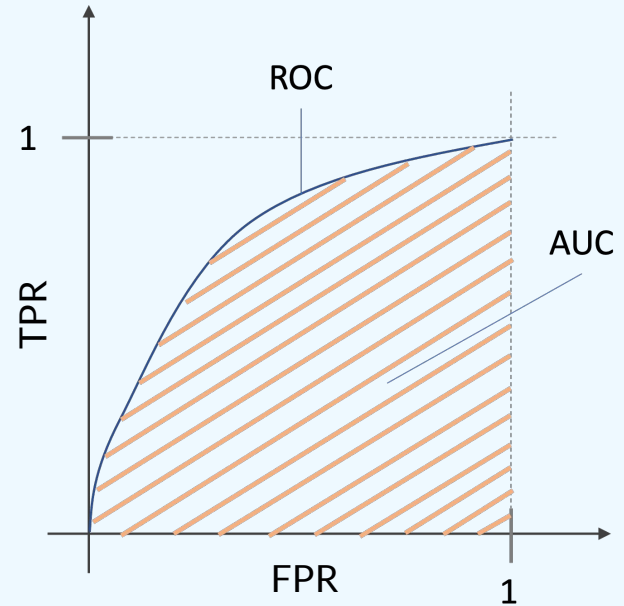


# Validation metric

		True Class	
Predict Class	True	True Positive (TP)	False Positive (FP)
	False	False Negative (FN)	True Negative (TN)

- Recall, True Positivity Rate (TPR) =  $TP / (TP + FN)$
- False Positivity Rate (FPR) =  $FP / (TN + FP)$

- Receiver operating characteristic (ROC)
- **AUC**: Area Under ROC

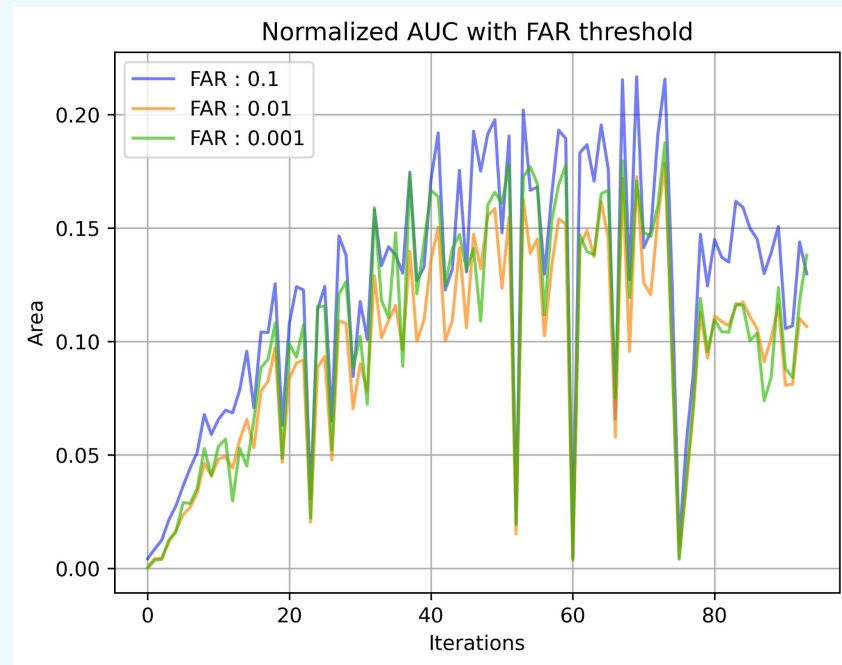




# Model Performance

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- We normalize the AUC with its max possible area, so that the max area is still equal to one
- The performance may look worse but can be improved by hyperparameter tuning



# Summary

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- Machine learning method is at booming stage for Multi-messenger Astrophysics.
- The GW Inference-as-a-service(GW IaaS) can provide an end-to-end alert pipeline for MMA.
- We trained a classifier for distinguish lots of different CCSN waveforms from unmodeled glitches.
- The performance of the current NN model is poor but it can be improved by hyperparameter tuning.
- We would like to test the model with Omicron triggered CCSN injection versus glitch in near future.

**Thank You!!**

# Appendix

# ML4GW framework

- ML4GW provides the software utils that accelerate the training and inference process.
- The Triton server helps to save the latest pre-trained model and schedule the computational resource for inference.
- ML4GW provide the software tools for data preprocess on GPU, and high throughput inference.

