

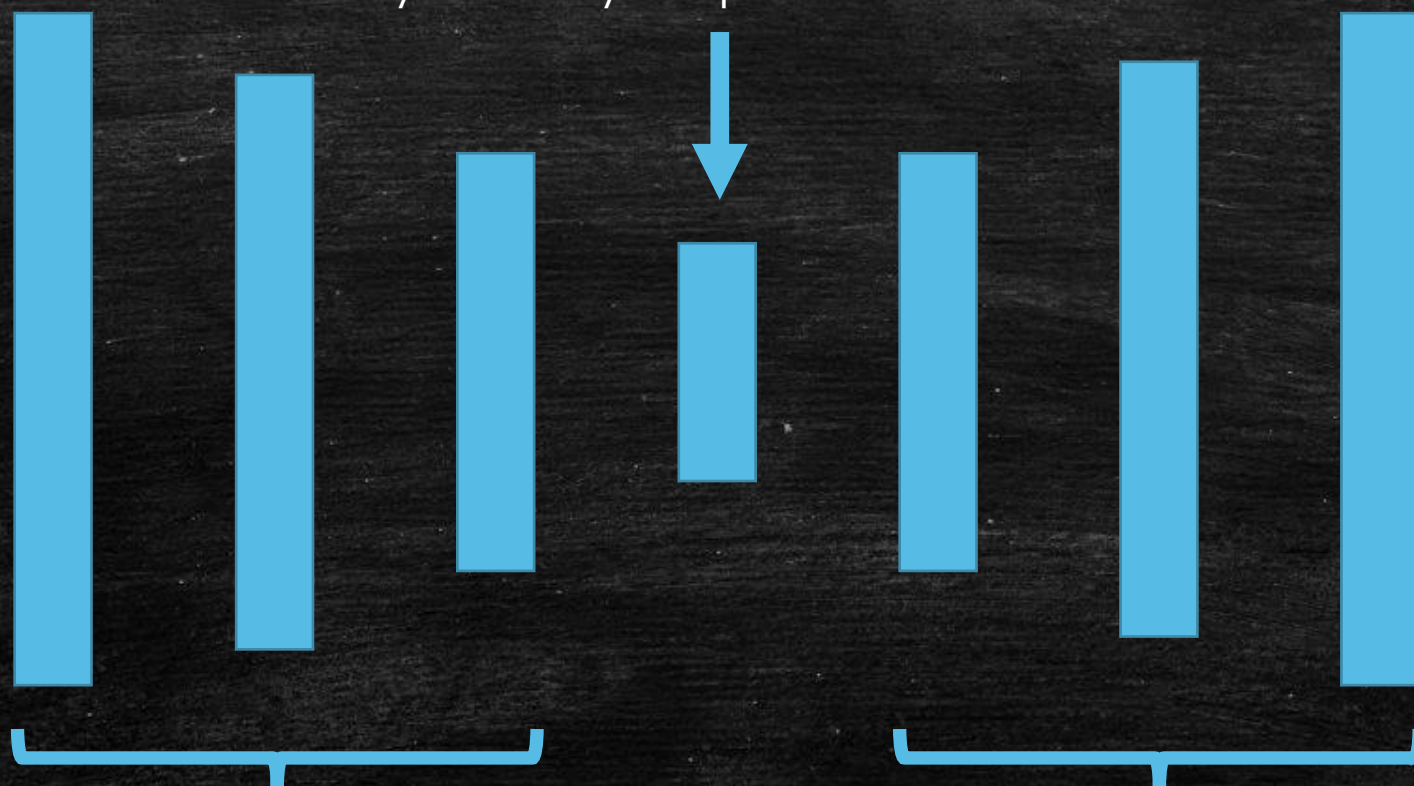
# Anomalous Resonance Identification with Autoencoders

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Andru Quiroga

# What is an Autoencoder?

Latent Space – Reduced Dimensionality  
Stores only necessary components for reconstruction



Encoder

Decoder

# What is an Autoencoder?

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Encoder

The Encoder's job is to reduce the dimensionality of the feature space.

# What is an Autoencoder?

Latent Space – Reduced Dimensionality  
Stores only necessary components for reconstruction

The Decoder's job is to take this encoded latent representation and reconstruct the original feature space.



# What is an Autoencoder?

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Encoder

Decoder

# What is an Autoencoder?

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Encoder

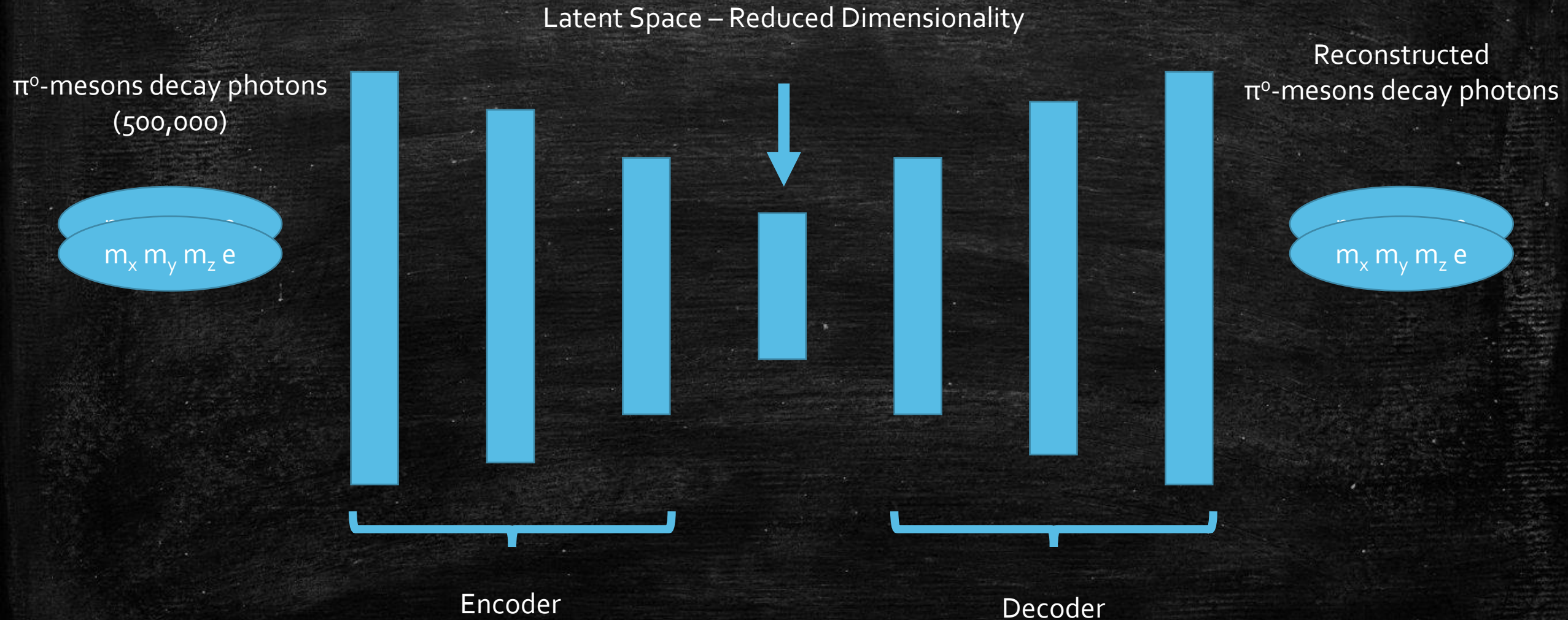
Decoder

# Anomalous Resonance Identification?

- The goal is to identify several resonances with only one known resonance.
- I will be using exclusively the decaying photons to allow the autoencoder more freedom and flexibility to encode the resonance.
- In this method we intentionally overtrain a specific known resonance, such as  $\pi^0$ , during training.
- By doing this, the latent representation of resonances will only make sense for our overtrained resonance.



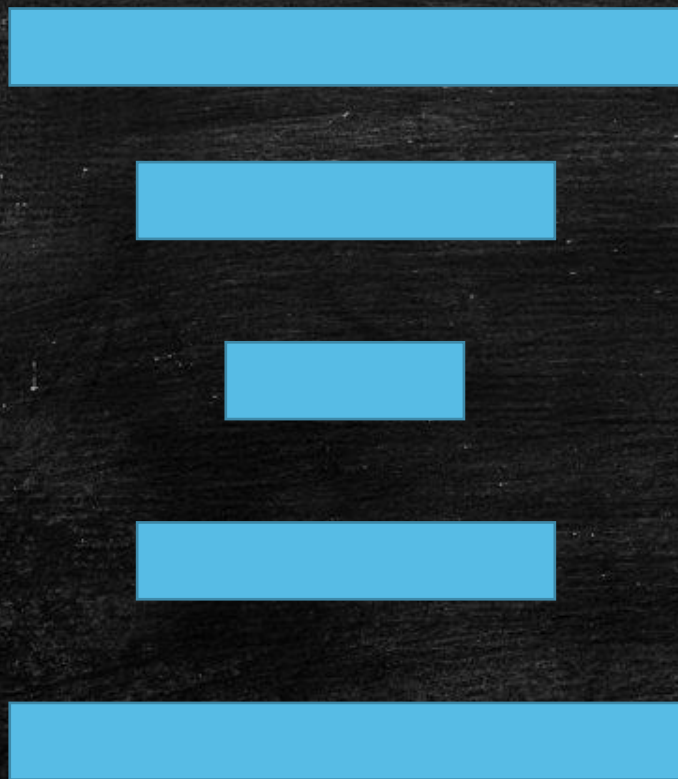
# Training Data





# Theory

$\pi^0$ -meson decay photons



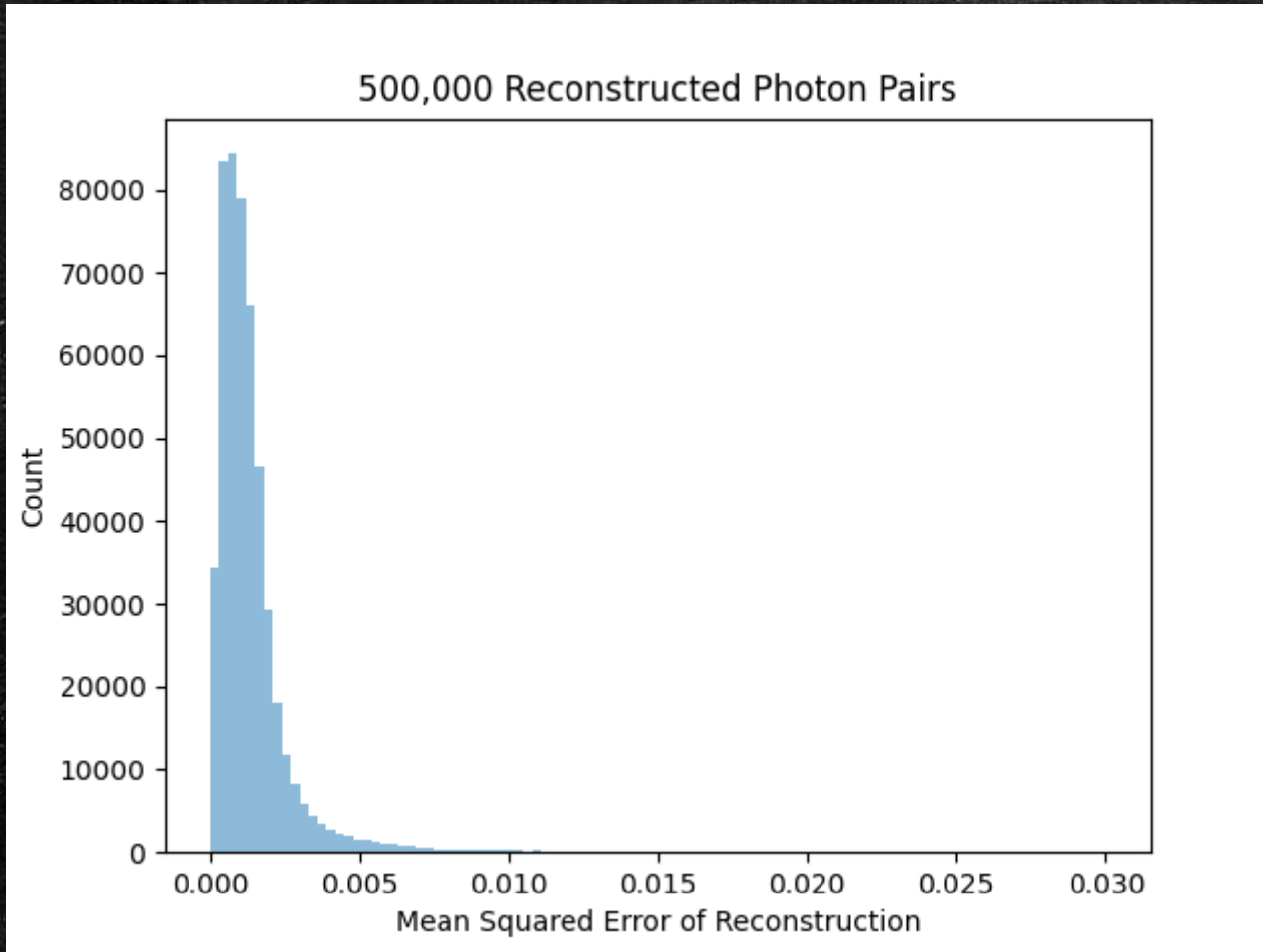
A well reconstructed  $\pi^0$ -meson decay photons

$\eta$ -mesons decay photons



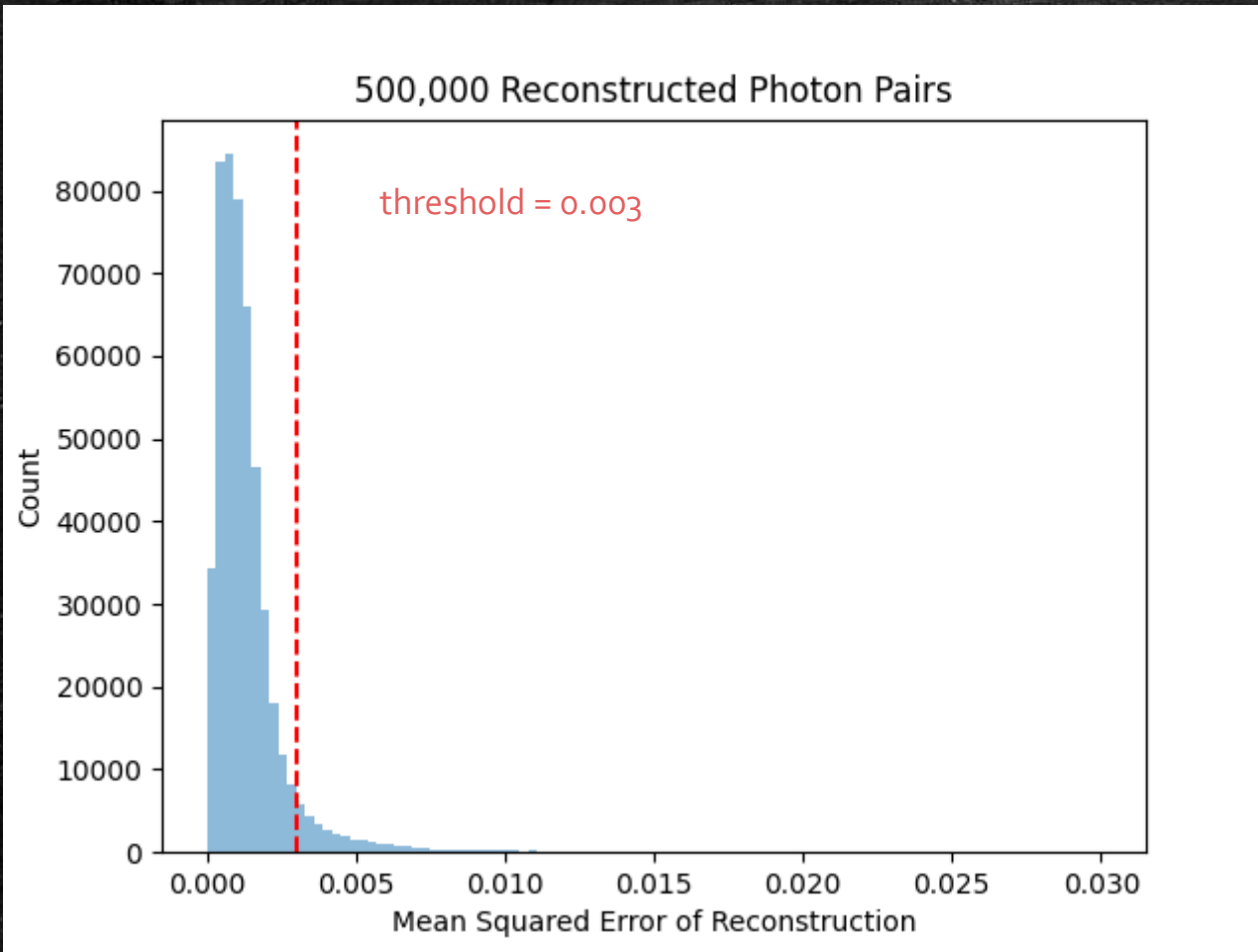
? Jumbled mess  
(Anomaly!)

# Results



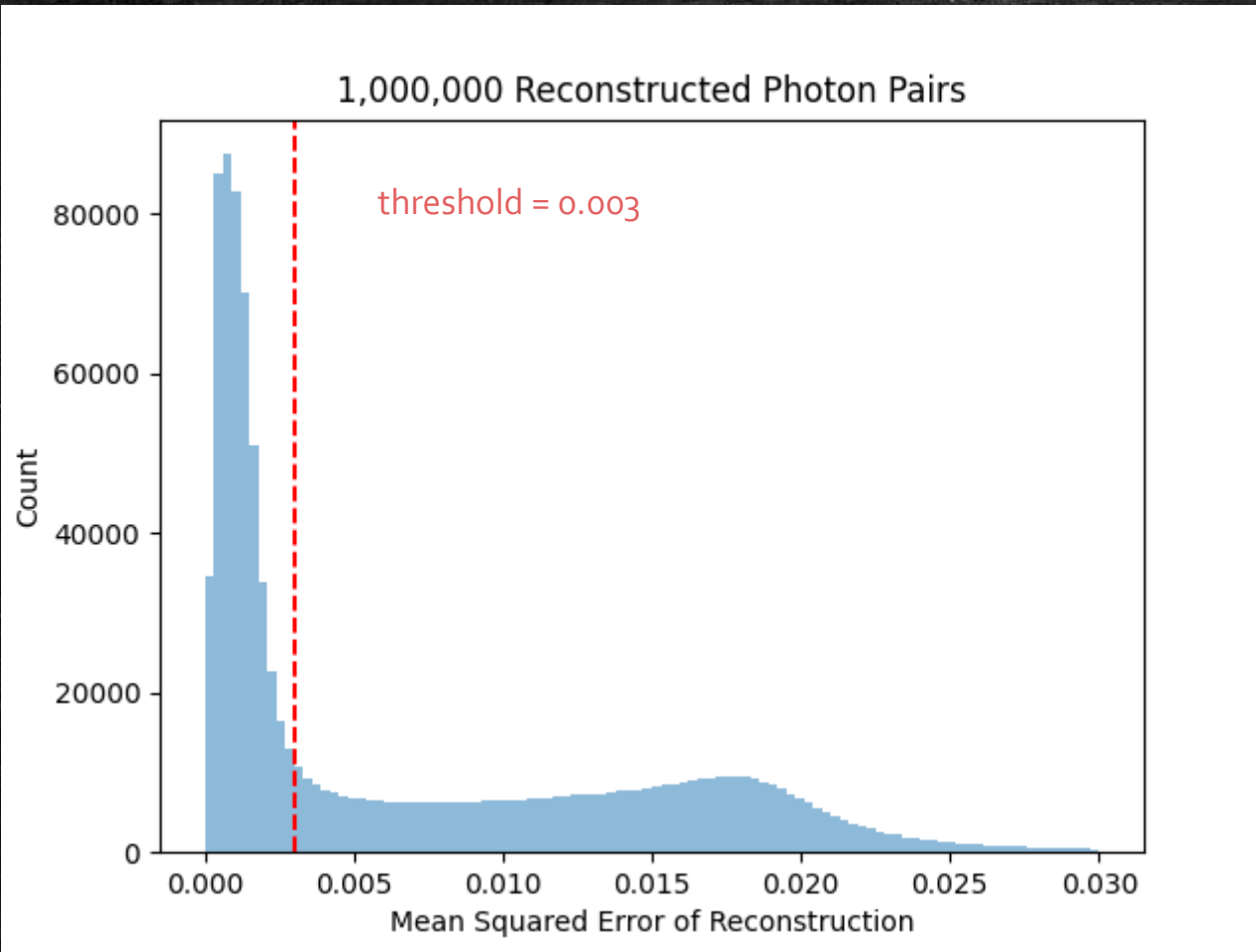
- After Training, we can plot a histogram of the photon pair's reconstruction loss.
- Now let's eyeball a threshold to justify an "Anomaly".

# Results



- After Training, we can plot a histogram of the photon pair's reconstruction loss.
- Now let's eyeball a threshold to justify an "Anomaly".

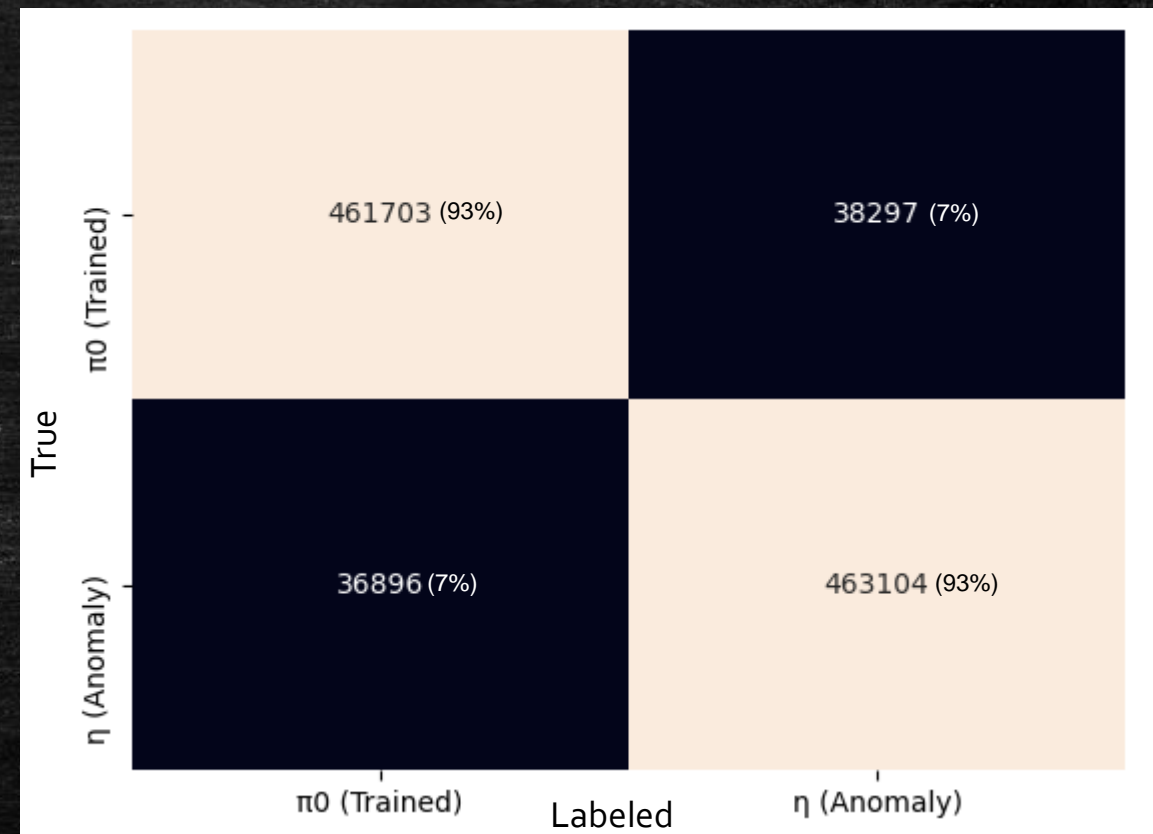
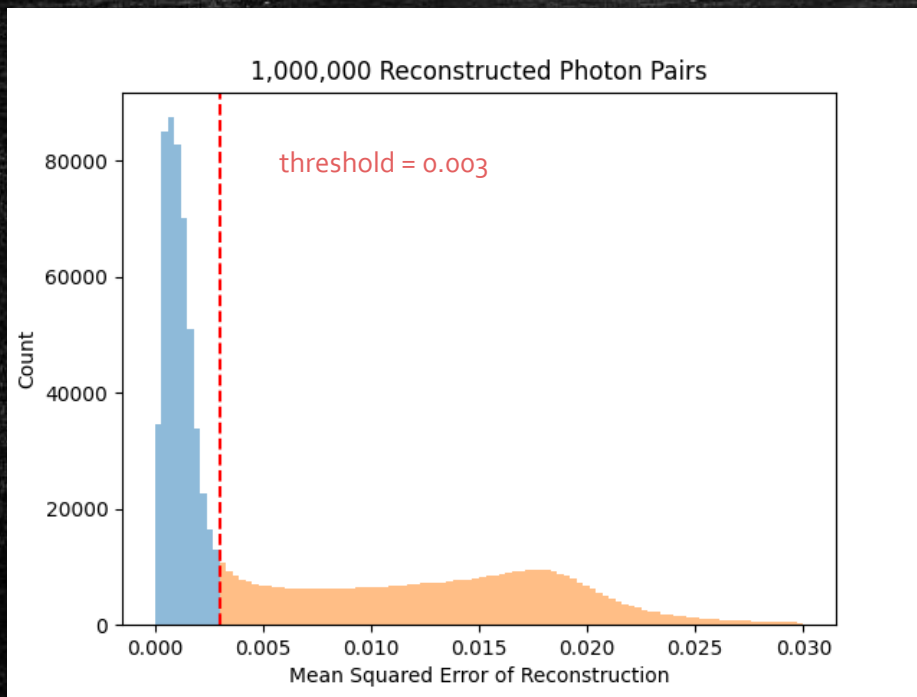
# Results



- After Training, we can plot a histogram of the events and their reconstruction loss.
- Now let's eyeball a threshold to justify an "Anomaly".
- Adding in some new data:  $\eta$ -mesons

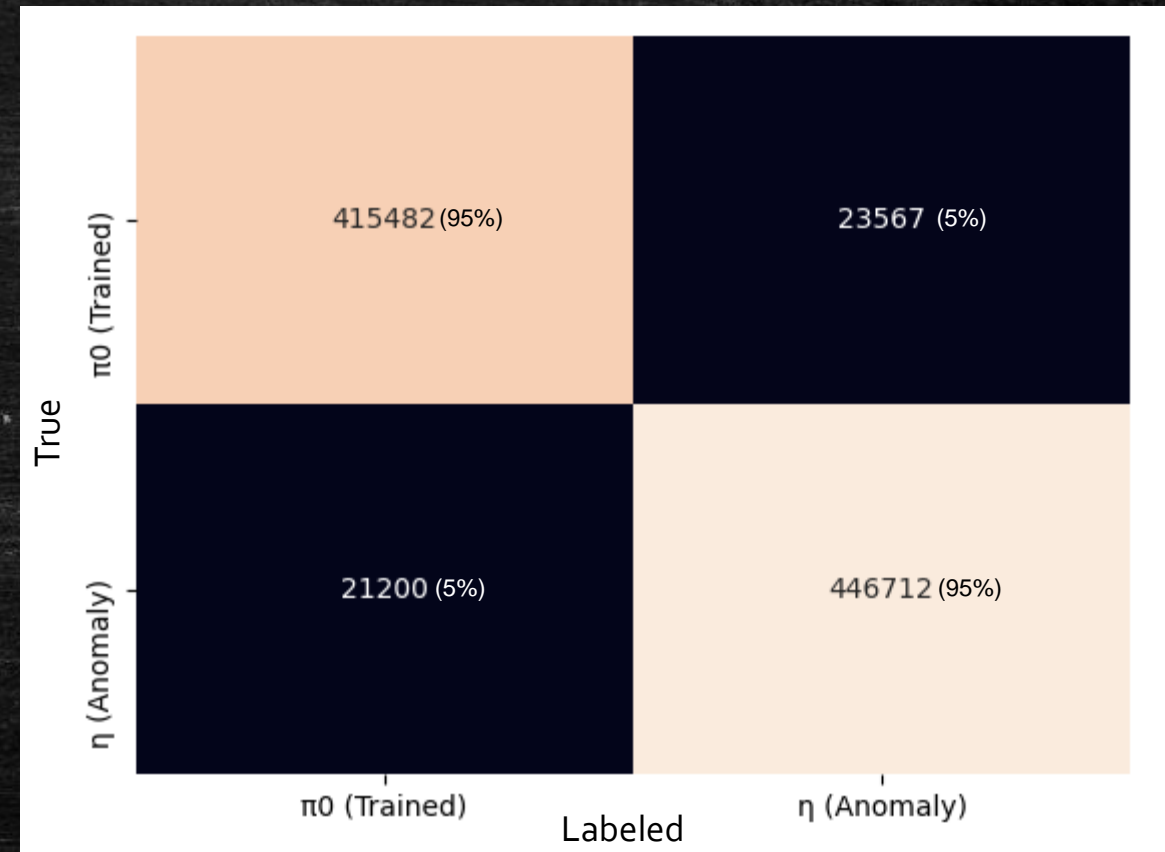
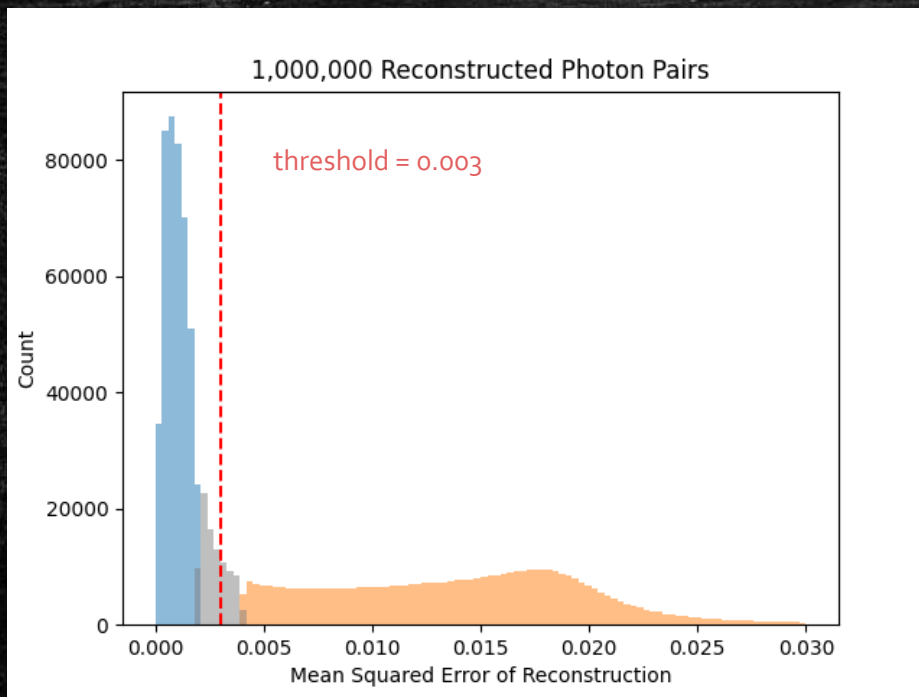
# Results

We can very crudely classify using this threshold.  
By doing so we get the following results:



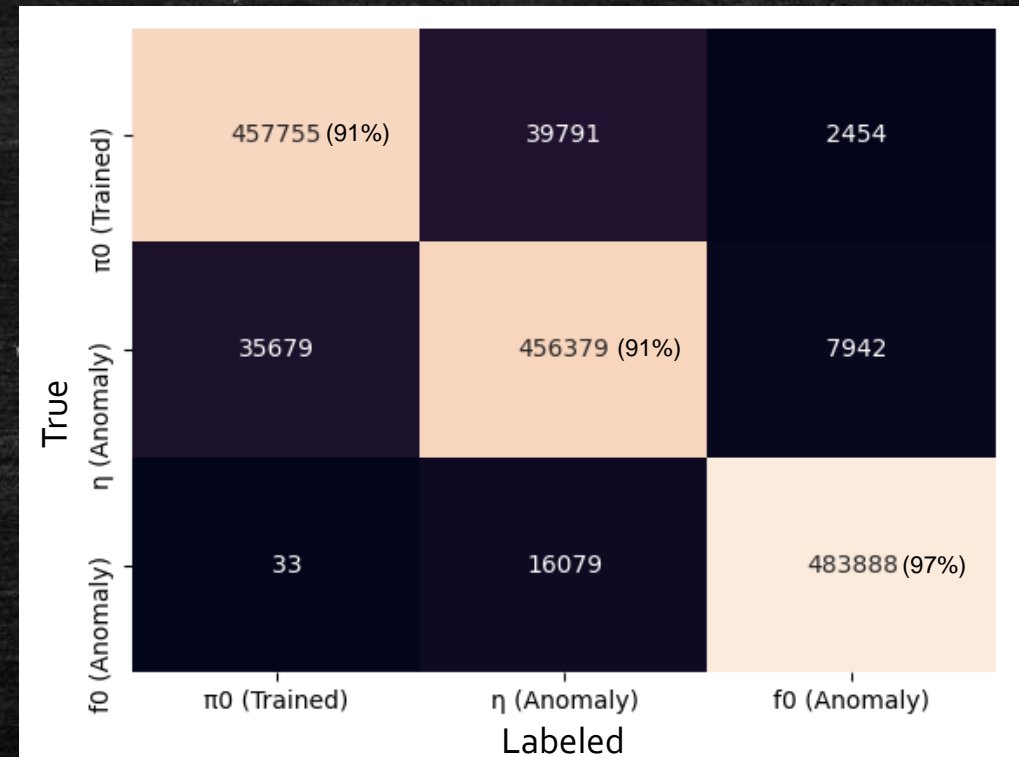
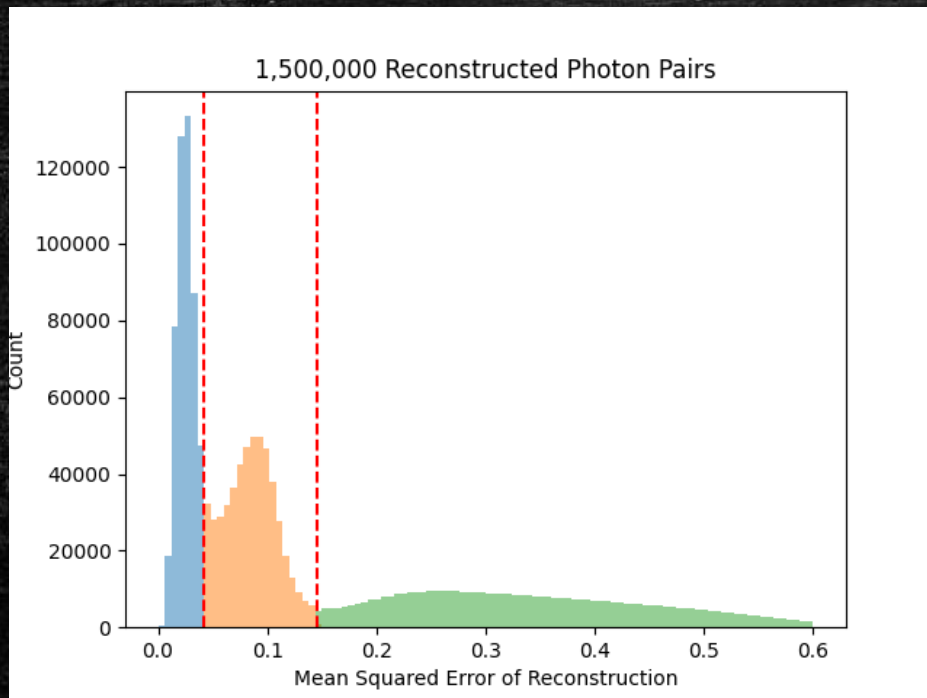
# Results

We can also take out a portion of the loss space for 'uncertainty':



# Does this expand beyond 2 Resonances?

Let's see what happens when we add another Resonance:



# Summary

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- Unsupervised classification of decays with only being trained on one resonance
  - With multiple categories!
- What can we do with this?
  - Would it be possible to use as a way of selecting signal?
  - Is it possible to find evidence of new resonances using this technique?
- Preliminary work with promising results!