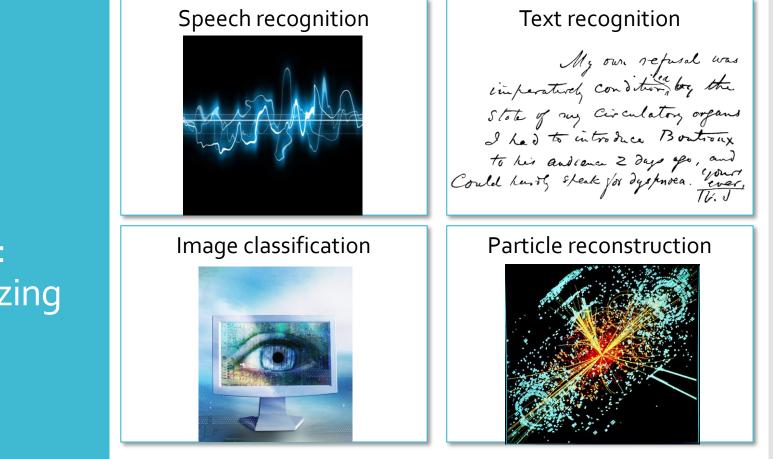
Deep learning based pulse shape discrimination for GERDA

Philipp Holl, Max-Planck-Institut für Physik DPG Frühjahrstagung Münster, 28.03.2017

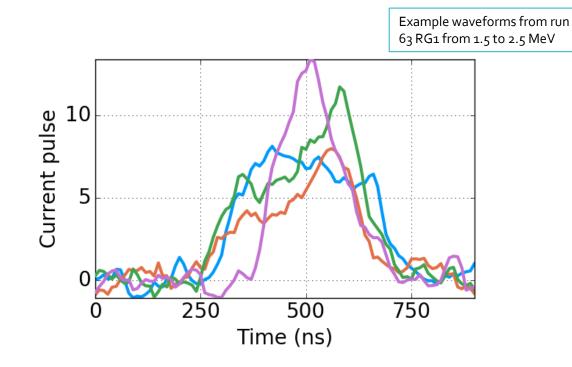
https://commons.wikimedia.org/wiki/File:Sound\_wave.jpg https://www.flickr.com/photos/downloadsourcefr/16361602656 https://commons.wikimedia.org/wiki/File:CMS\_Higgs-event.jpg



Deep learning is about recognizing patterns

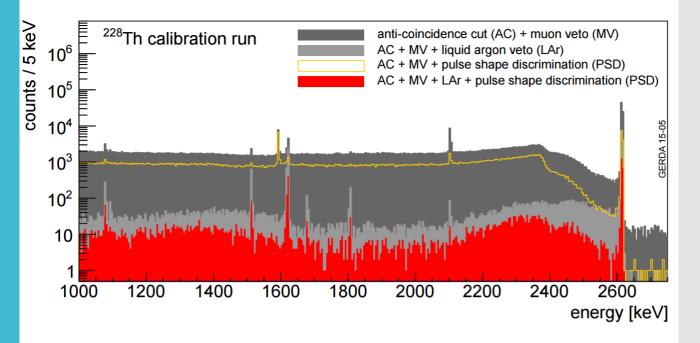
Deep learning: Recognizing patterns

### Pulse shapes and deep learning



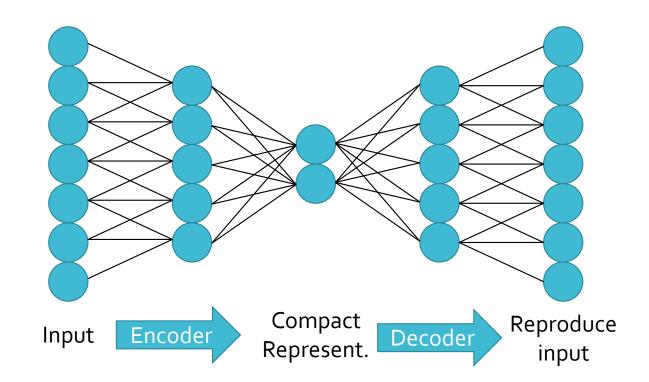
- Signal encodes type of pulse in ~1000 samples
  - Background rejection
- Patterns highly complex  $\rightarrow$  deep learning for recognition

#### GERDA data: Mostly unlabelled



- Terabytes of high-dimensional data, only tiny fraction labelled
- Not enough labelled training data → Hand-written parameter extraction (preprocessing)

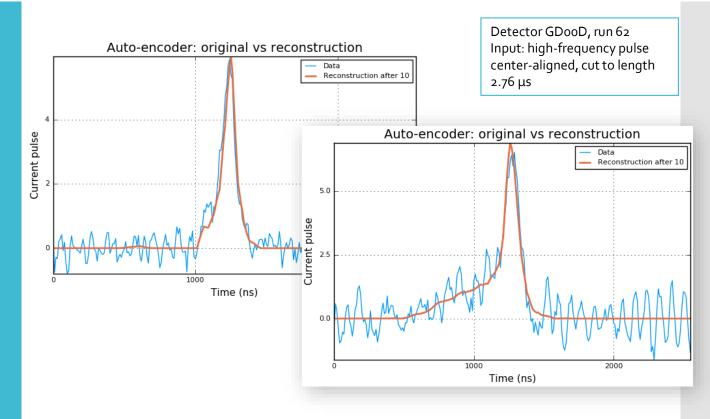
### Autoencoder: Unsupervised learning



- Autoencoders learn the best compression of unlabelled data
- Encoder, decoder & compact representation are learned from data
- Compact representation ~2 orders of magnitude smaller than original pulse

## Autoencoder in action

10 compact parameters

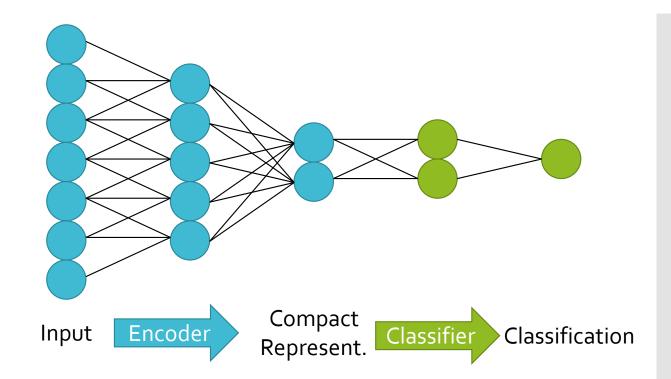


- Reconstruction from compact representation
  - Less information than original pulse
- Noise is not learned as it follows no pattern

Advantages of the autoencoder

- Unsupervised
  - Can be trained on unfiltered data sets
- Compact representation
  - Low-dimensional
  - No noise information

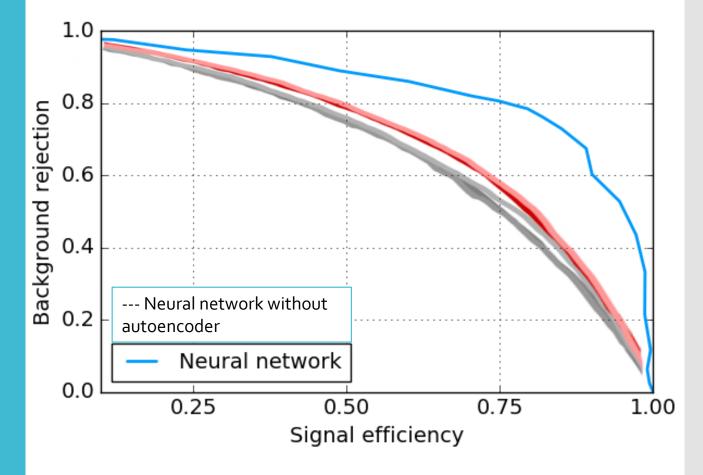
Classification with autoencoder and neural network



- Train autoencoder on ALL data
- Train neural network on compact representation of labelled data

Detector RG1, run 62, Autoencoder and PSD network with 2 hidden layers trained on run 63

Autoencoder + neural network for Coax



• Performance with autoencoder much better

# Summary & Outlook

- New method for pulse analysis
  - Key technique: Unsupervised information reduction
  - Competitive compared to current methods
- Outlook
  - Use simulated pulses for training