

# Machine learning applications in NEWS-G

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#### **NEWS-G – Spherical Proportional Counter** een's E Field [V/m] 10<sup>6</sup> **Signal Generation:** 1. Primary ionization 10<sup>5</sup> 2. Electron drift 10<sup>4</sup> 3. Townsend avalanche $10^{3}$ lons 4. Positive ion drift $10^{2}$ Analysis data taken from 10<sup>1</sup> SEDINE at LSM in France

### **Problem Definition**



## Goals:

- Utilize machine learning methods to remove noise from recorded detector signals
- Model implementation should aid in measuring important signal characteristics, such as amplitude and risefeatures





#### **Methods – Model Training**



#### Trained on a simulation-based dataset modeled after our real detector







Energy: ~1370eV

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#### **Example Pulses**





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### **Simulated Energy Resolution**

















- Other single-output predictions
  - Direct energy prediction, pulse shape classification
- Double-deconvolution layer implementation
  - Explicitly add preprocessing steps to network layers
  - Learn to return primary electron arrival times
- Different model architectures for improved performance
  - Adversarial networks (see Tianai's poster)



# Thank you!



## Additional Slides

#### **Simulated Energy Resolution Results** /ew Leen's Ideal Reconstructed Energy based on Simulated Number of Secondary Electrons (eV) Noisy Double Deconvolution Denoised Double Deconvolution Clean Double Deconvolution Standard Deviation over Mean $10^{-1}$

Number of Secondary Electrons





- Triggering efficiency test on simulated data
- 10000 events with a simulated pulse, 10000 noise traces
- Modelled electronic triggering
- Preliminary results









#### **Model Architecture**



Layer	Stride	Window	Output
Input			4096, 1
Convolution	1	1	4096, 8
Convolution	1	9	4088, 16
Average Pooling	2	2	2044, 16
Convolution	1	17	2028, 32
Average Pooling	2	2	1014, 32
Convolution	1	33	982, 64
Average Pooling	2	2	491, 64
Convolution	1	33	459, 32
Transpose Convolution	1	33	491, 32
Upsampling	2	2	982, 64
Transpose Convolution	1	33	1014, 64
Upsampling	2	2	2028, 64
Transpose Convolution	1	17	2044, 32
Upsampling	2	2	4088, 32
Transpose Convolution	1	9	4096, 16
Convolution (output)	1	1	4096, 1





- Developed an effective deep learning noise removal model
  - Resulting in more accurate energy measurements, primary electron counting
- Developed a single output model based on denoising architecture
  - Offers improvements in energy measurements
- Outlined further avenues for machine learning applications in NEWS-G