

Automated Variable Selection of Gamma-Ray Responses by Utilizing LASSO and Elastic Net

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Agenda

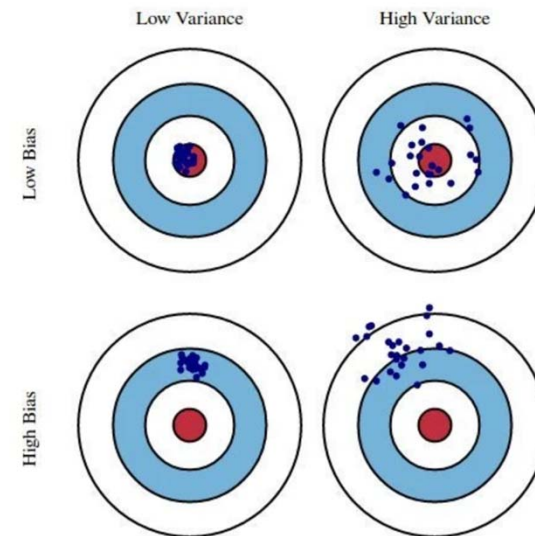
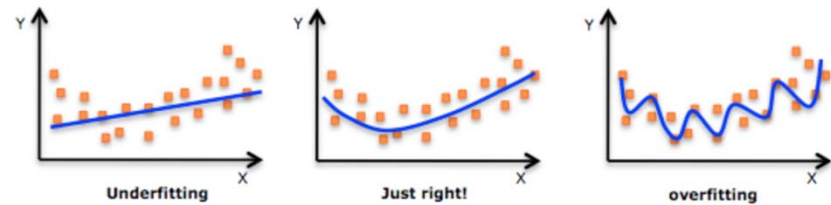
- Monte Carlo Library Least Squares (MCLLS) fitting
- Basics of supervised machine learning
- Regularization methods (LASSO and Elastic Net)
- Prompt Gamma Neutron Activation Analysis (PGNAA)
- Library generation and data processing
- Results

Monte Carlo Library Least Squares

- Same principle as ordinary least squares with library spectra used as the input
- $y_i = \sum_{j=1}^m x_j a_{ij} + E_i, i = 1, n$
 - y_i is the counts per channel i
 - x_j are linear coefficients for each element j
 - a_{ij} are the library spectra, or counts in channel i of element j
 - E_i is random error in counts in channel i
- How to deal with changing environments and unknown compositions?

Supervised Machine Learning

- Ordinary Least Squares traditionally suffers from overfitting
 - Too many model parameters
- Model selection should have the right parameters to accurately predict/fit correct solution
- Adding complexity to the model increases bias while reducing variance
- Supervised machine learning variable selection techniques aim to minimize unnecessary bias and variance simultaneously by selecting the correct parameters

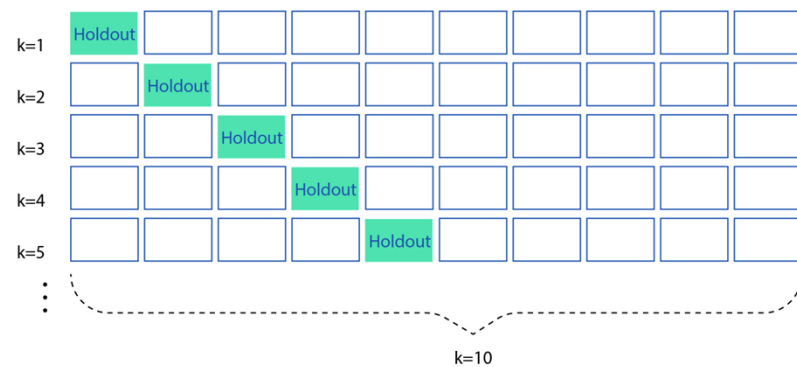
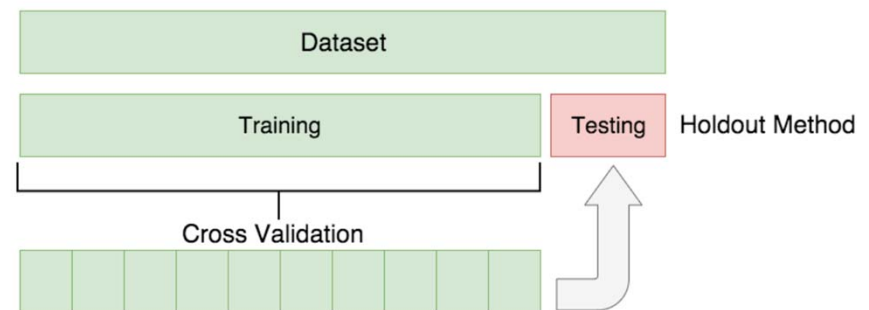


LASSO and Elastic Net

- The LASSO (Least Absolute Selection and Shrinkage Operator) is defined as:
 - $\hat{\beta}^{lasso} = \underset{\beta \in \mathbb{R}^p}{\operatorname{argmin}} ||y - X\beta||_2^2 + \lambda \sum_{j=1}^p |\beta_j|$
or $= \underset{\beta \in \mathbb{R}^p}{\operatorname{argmin}} ||y - X\beta||_2^2 + \lambda ||\beta||_1$
 - Where the first term is the loss function and $\lambda ||\beta||_1$ serves as a penalty
 - Note: if $\lambda=0$, the equation is identical to that of Ordinary Least Squares (OLS)
 - Elastic Net is a modification of LASSO that adds a quadratic penalty as defined below:
 - $\hat{\beta}^{ElasticNet} = \underset{\beta \in \mathbb{R}^p}{\operatorname{argmin}} ||y - X\beta||_2^2 + \lambda_2 ||\beta||^2 + \lambda_1 ||\beta||_1$
 - As $\lambda_{1,2}$ increase, the penalty for each new coefficient grows, allowing variable selection to occur in linear models

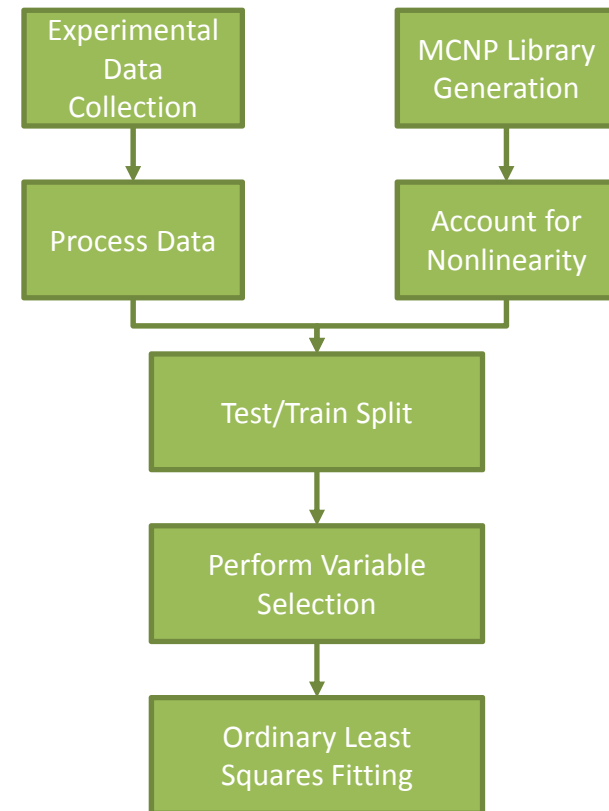
Cross Validation

- Each method relies on test/train split to tune model parameters
- This step introduces bias from selecting some data over others (random)
- Cross validation performs the test/train split multiple times, reducing random bias from the model training
 - Trade off - extra computing time



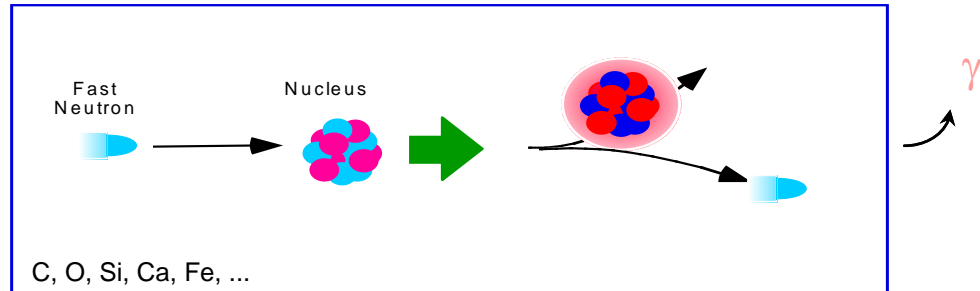
Full Run Flow Chart

- Simulated and experimental data is generated
- Each set is processed
 - Nonlinearities removed
 - Prompt/delayed responses are extracted
- Test/train split data for model selection
- Variable selection through LASSO or Elastic Net remove unnecessary model parameters and provides initial guesses
- Final ordinary least squares fitting to reduce bias from model selection procedure

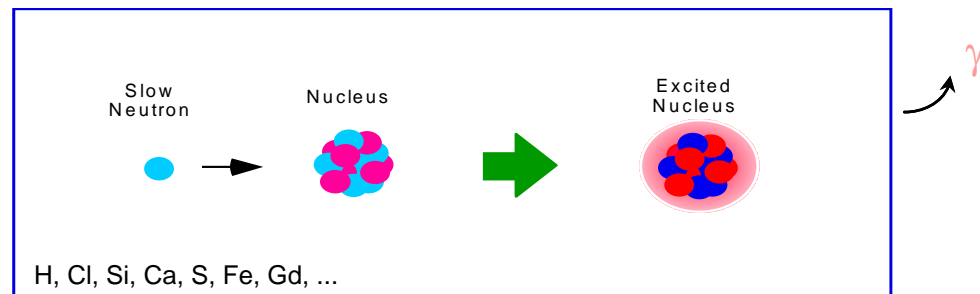


Neutron Interactions

INELASTIC SCATTERING



THERMAL CAPTURE



NEUTRON ACTIVATION

Inelastic or capture reaction that leads to a radioactive element and decay.
Examples:

O-activation	$T_{1/2} = 7.1 \text{ s}$	$(n + {}^8_8\text{O}^{16} \rightarrow {}^7_8\text{N}^{16} + \dots \rightarrow {}^8_8\text{O}^{16} + \gamma + \dots)$
Al-activation	$T_{1/2} = 2.3 \text{ m}$	$(n + {}^{13}_{13}\text{Al}^{27} \rightarrow {}^{13}_{13}\text{Al}^{28} \rightarrow {}^{14}_{14}\text{Si}^{28} + \gamma + \dots)$
Si-activation	$T_{1/2} = 2.3 \text{ m}$	$(n + {}^{14}_{14}\text{Si}^{28} \rightarrow {}^{13}_{13}\text{Al}^{28} + \dots \rightarrow {}^{14}_{14}\text{Si}^{28} + \gamma + \dots)$

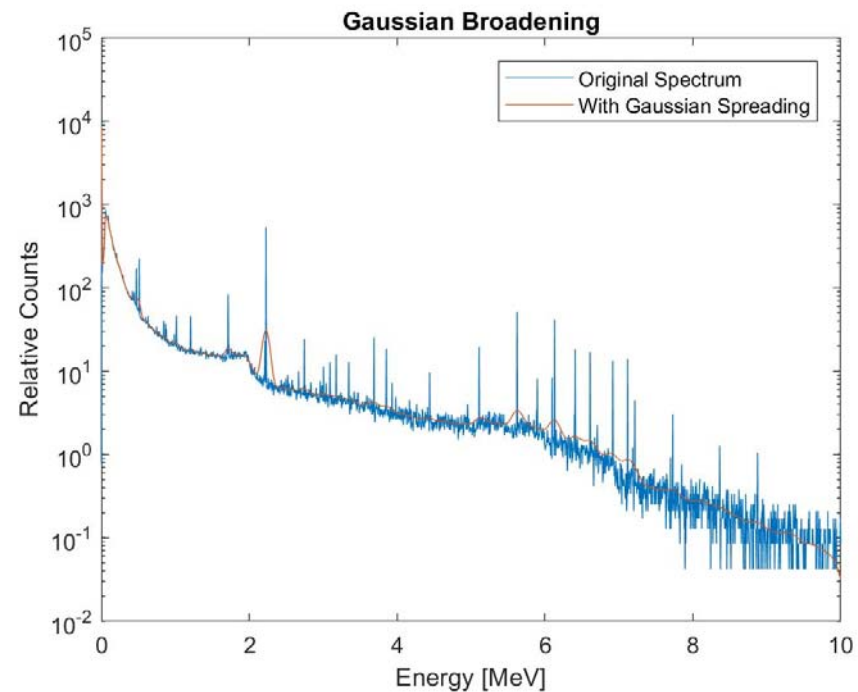
(Na, Cu, Fe..... and many more).

Library Generation

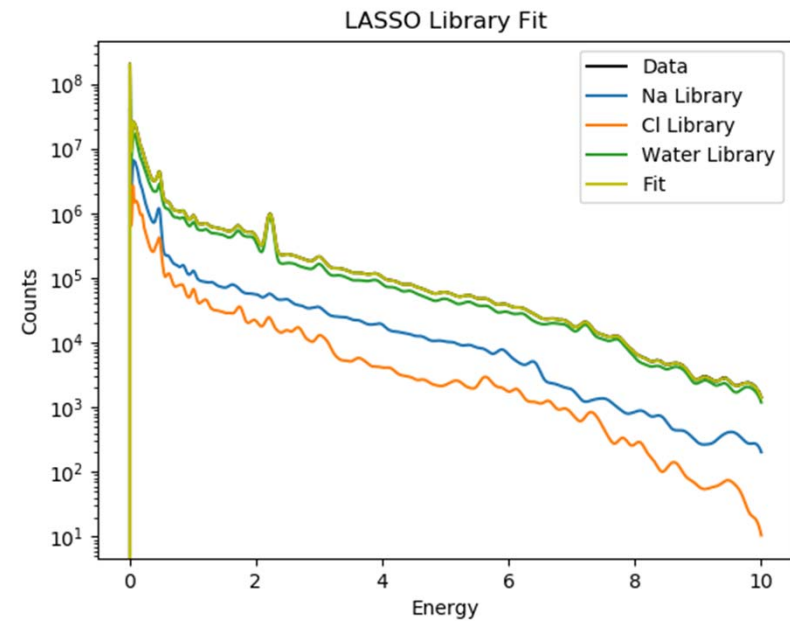
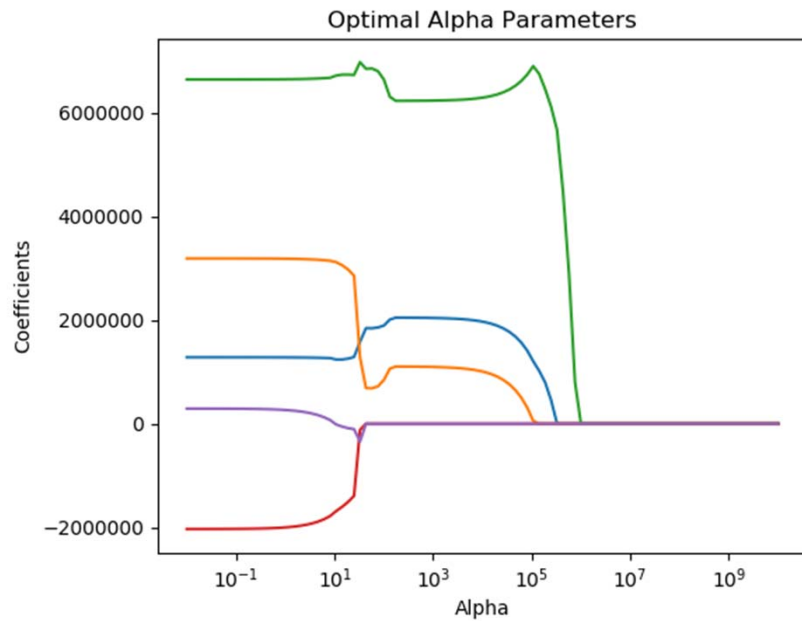
- Extensive MCNP simulations
 - Center for Engineering Applications of Radioisotopes (CEAR) cluster – reduced run times from 1.5-3 days to ~1 hour
 - F8 tally for gamma detection
 - No GEB
- Detector Response Function (DRF)
 - Response generated in a detector from incident radiation
 - Non-linearities resolved to treat problem as a linear combination of library inputs
 - Gaussian broadening fit of full width/half max using calibration sources
 - Energy to channel conversion (2nd order polynomial in the following examples)
 - Both of these will be upgraded by the work performed by Aaron Feinberg

Processing of Data

- MCNP simulations are broadened post processing (faster than GEB tally)
- Nonlinear NaI response requires adjustment during channel to energy conversion
- $Energy = a + b * channel + c * channel^2$

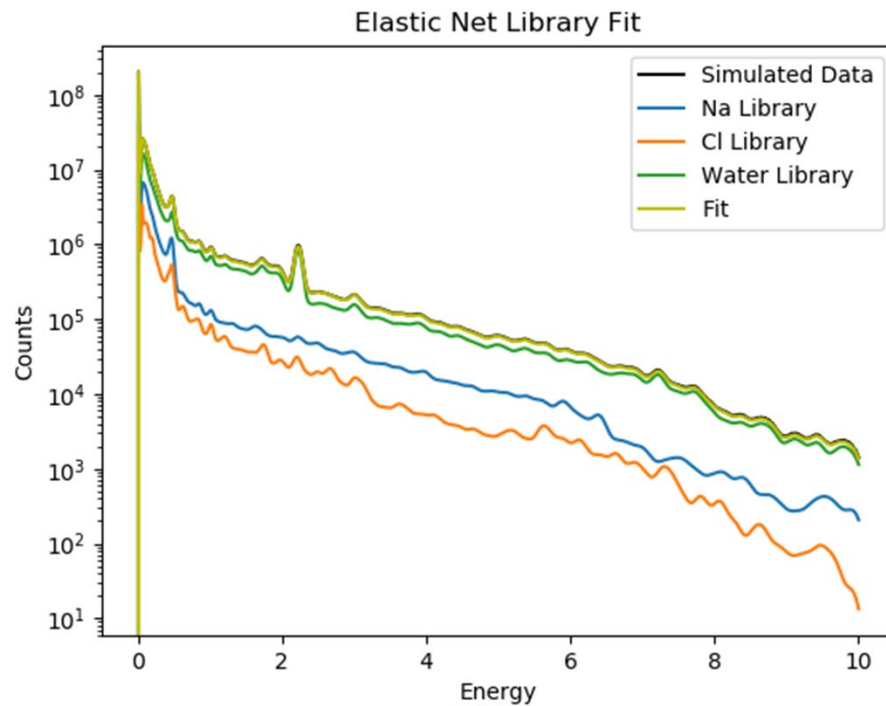


LASSO Simulated Example



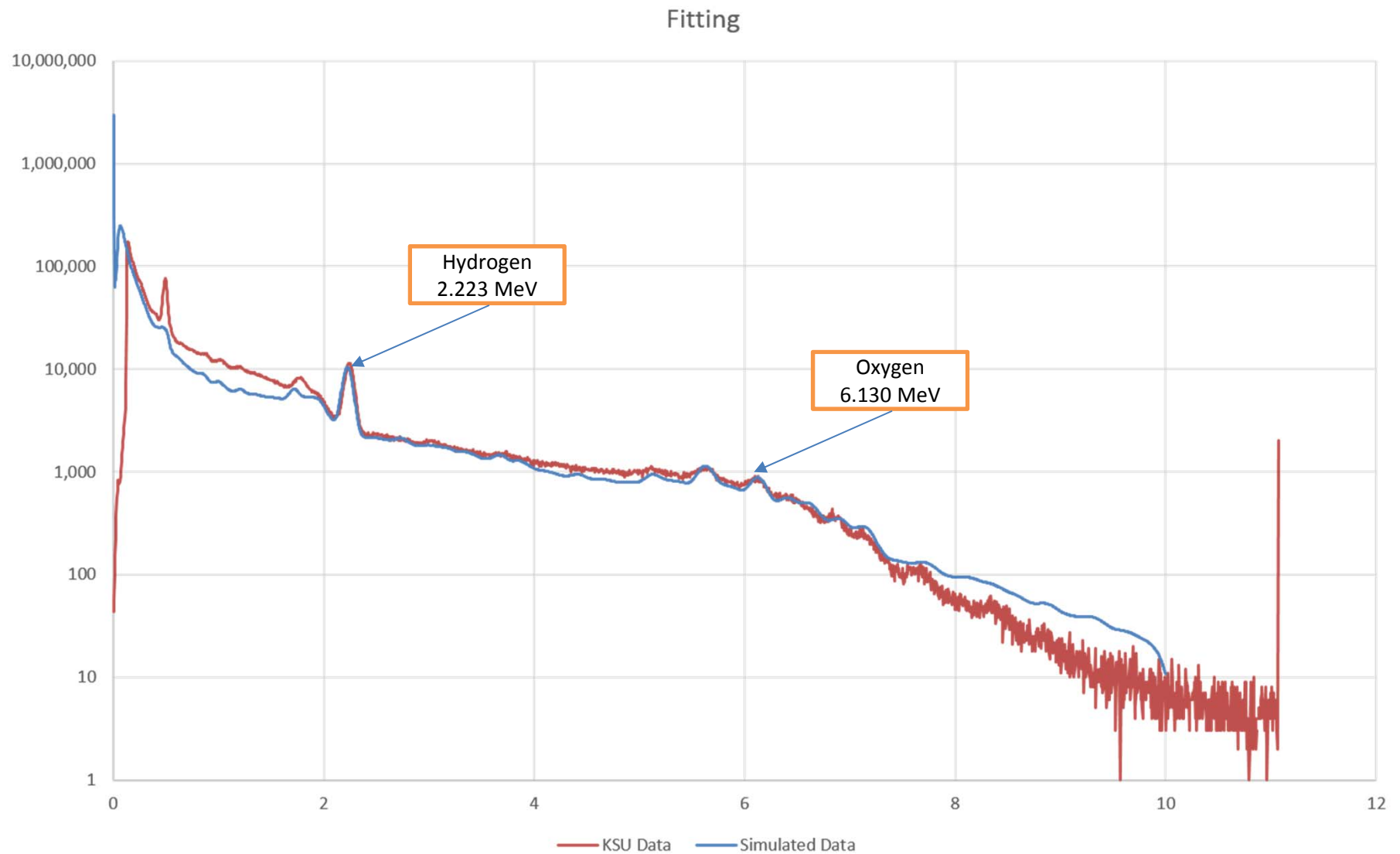
Note: 5 libraries are used. Fe and Cu do not contribute to the total spectrum, are given a 0 contribution and selected out of the final model.

Elastic Net Simulation and Comparison

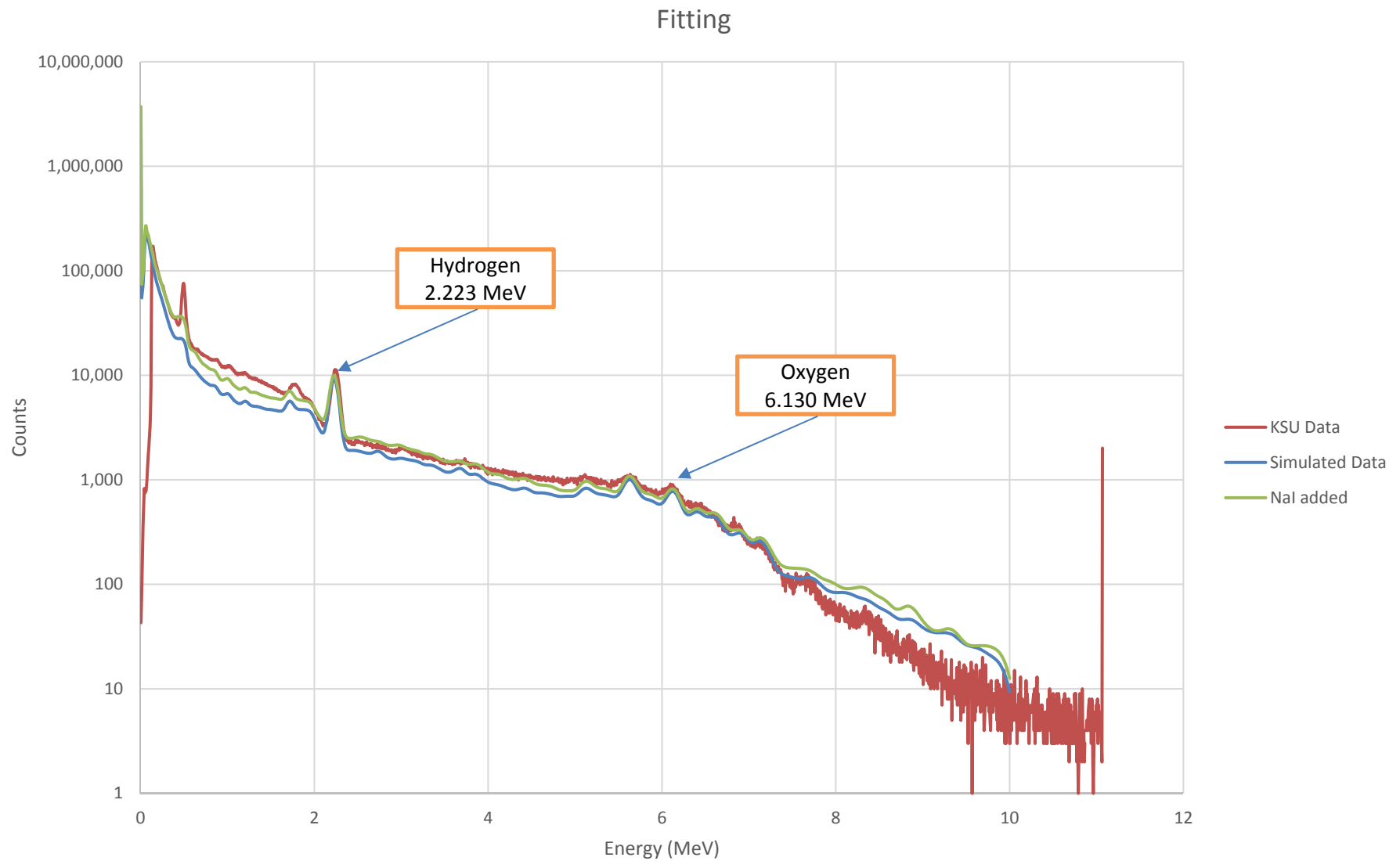


Relative Error		
	Elastic Net	LASSO
Water	5.01%	0.07%
Na	2.95%	1.32%
Cl	29.95%	3.65%

Water – First Results

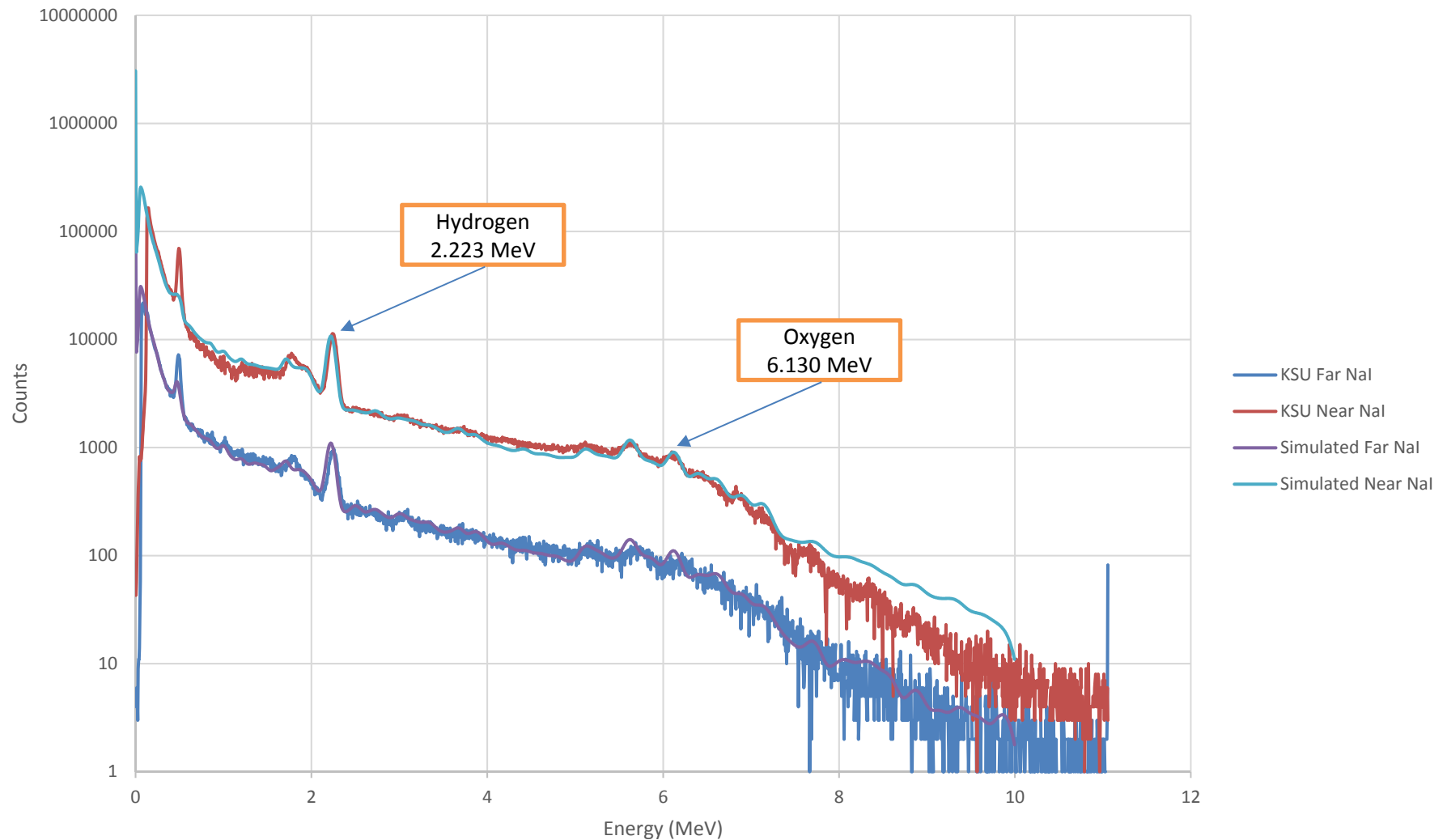


Results With NaI Activation - Water

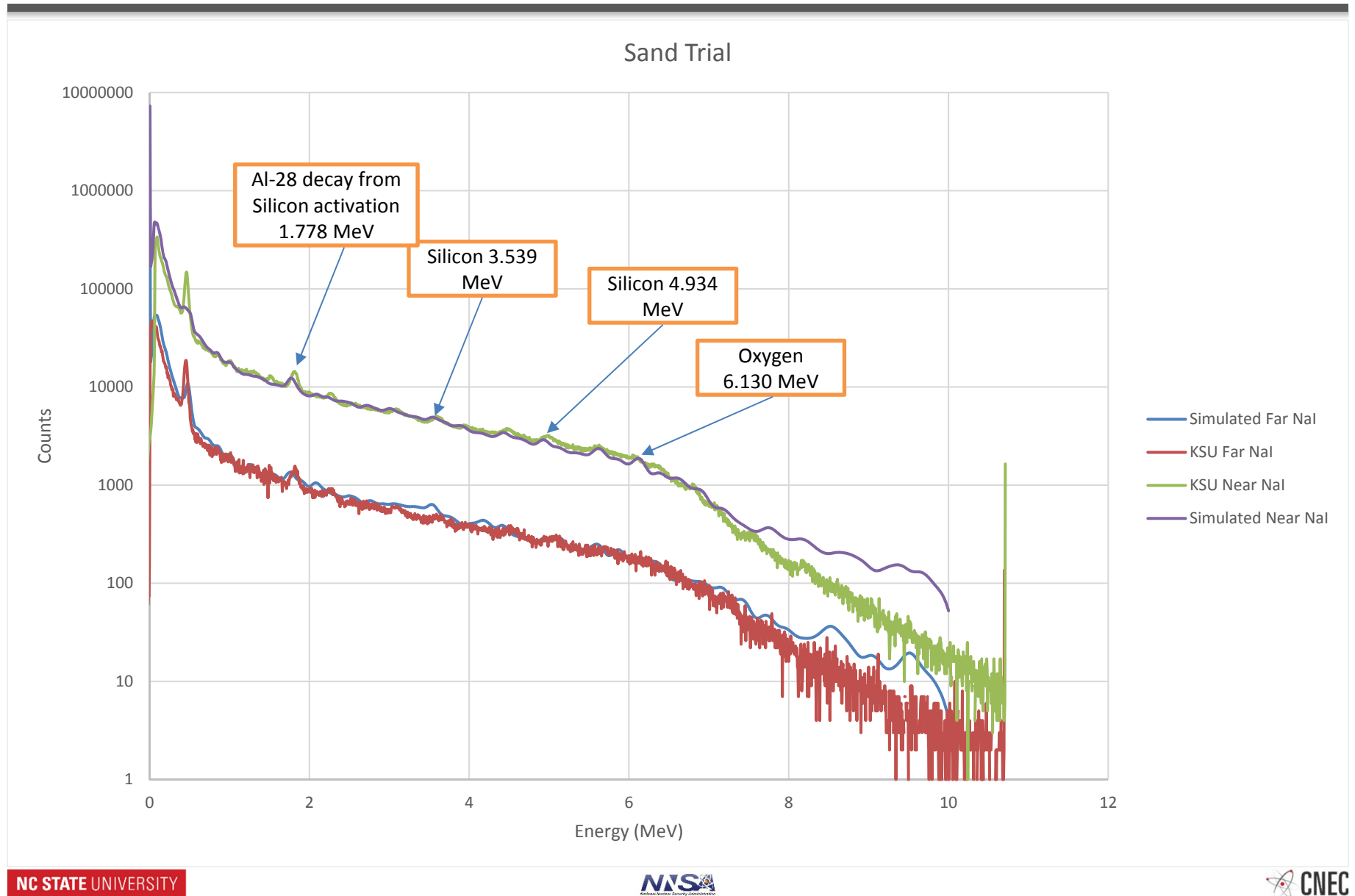


Water Trial – Background Removed

Pure Water Trial



Sand Trial – Background Removed



Continuing Work/Next Steps

- Incorporate work from Aaron Feinberg and Long Vo
 - Bayesian approach to fit all non-linear components
 - Time dependent digitizer data
- Analyze additional trials run at KSU
 - Limestone
 - Multiple porosities, water content
- Extensive testing of the limitations of LASSO and Elastic Net using simulated radioisotope data
 - High, medium, low counting situations
 - Vary number of radioisotopes
 - Vary number of channels
 - Shielding situations

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Discussion

- Questions?



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