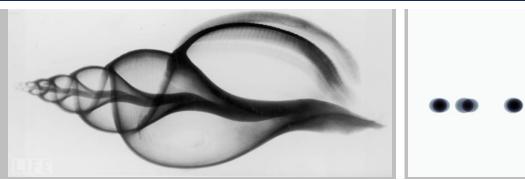
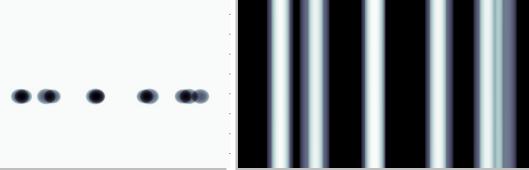
Exceptional service in the national interest







Object Composition Identification via Mediated-Reality
Supplemented Radiographs

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Elevator Speech



- Summary:
 - This work presents an exploratory method to post-process radiographs to allow for maximum information extraction. Specifically, can one identify material contain within the field-of-view.
- What benefit could TSA obtain from this technology?
 - Increased detection capacity
 - Reduced false-positive rates
 - Precise material classification
- So what?
 - This technology potentially leads to increased separation between materials with similar properties.
- Who cares?
 - Medical, security, manufacturing, industrial NDT/NDE.

Detection



- Object is scanned
 - X-ray Image(s) produced
 - Displayed on a monitor (LCD-type)
- Image is processed by designated algorithm
 - Pixel Thresholding
 - Usually utilize Dual-Energy Radiography
 - Color coding

Exploratory work

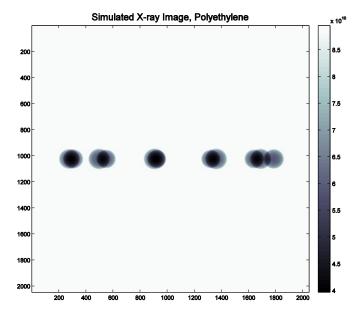


- Avoid the approximation $\bar{I} = \bar{I_0}e^{-\bar{\mu}x}$, and try to estimate $\mu(\varepsilon, x)$ directly.
- Very challenging, is it possible?
- Radiation detection hardware has evolved significantly in recent years.
- Decreasing cost has allowed for new types of data acquisition.

Past Work



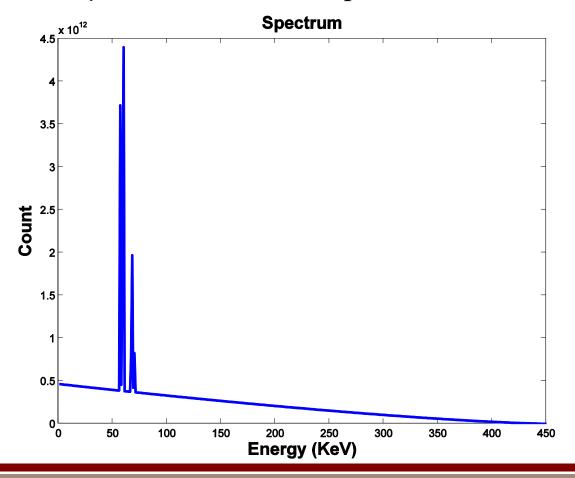
- 1st: acquire images at multiple energies, solve effective attenuation (does not work!)
- Next: Approach: Use basis function to create candidate materials, simulate the image and compare it to the acquired image.
 - Limited success



Multi-Energy Attempt

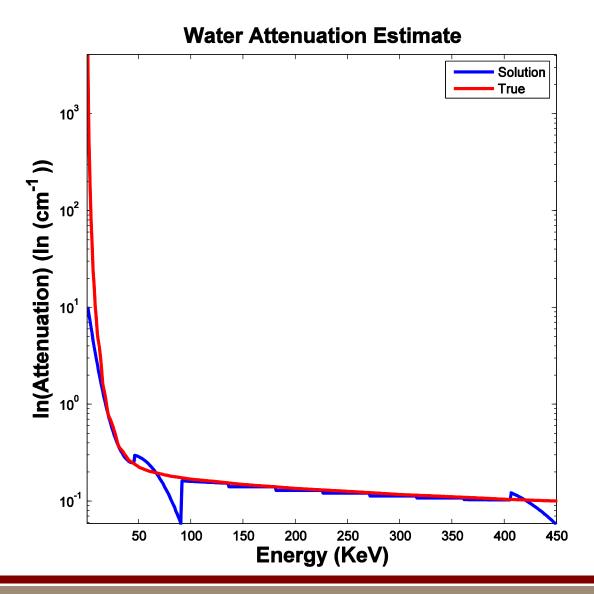


 Acquire images with an energy discriminating detector and apply mediated-reality simulation-based optimization



Results

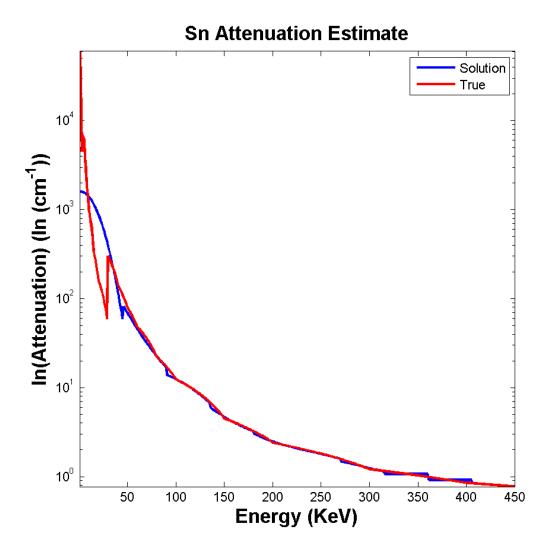
Water:



Results:



Tin:



Results cont.



Lead:

Pb Attenuation Estimate Solution • True 10⁴ In(Attenuation) (In (cm⁻¹)) 10¹ 100 150 200 250 50 300 350 400 450 **Energy (KeV)**

Conclusion



- Even for narrow energy ranges, direct evaluation of attenuation values introduce non-trivial error.
- Leveraging simulation-based optimizations could lead to increased numerical stability.
- Future work: Laboratory-based validation experiments underway.
- Thanks! Questions?

Dual Radiography



- X-rays follow Beer's law, for images:
 - $I = \iint_{\varepsilon \in E, x \in X} I_0(\varepsilon) e^{-\mu(\varepsilon, x)x} d\varepsilon dx$
- But, is approximated as:

$$\bar{I} = \bar{I_0}e^{-\bar{\mu}x}$$

- Using this approximation, 2-4 radiographs measured
 - Low energy (object present, absent)
 - High energy (object present, absent)

Continued



A ratio image of high-to-low energies is created

$$R = \frac{\frac{I_H}{I_{0,H}}}{\frac{I_L}{I_{0,L}}} \approx \frac{\overline{\mu_H}}{\overline{\mu_L}}$$

- This ratio approximates "effective atomic number"
- This ratio can then be subject to a threshold to separate material types
- Drawback: due to noise and other factors, only general classifications can be made.

SNL Applications

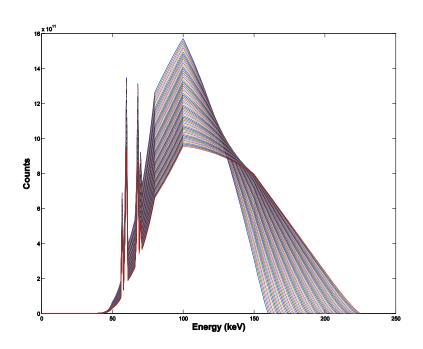


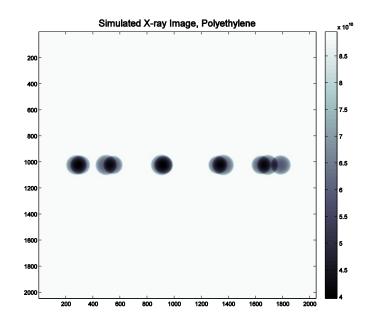
- Current Radiography technology leveraged at SNL
 - 3D Computed Tomography
 - Digital Radiography
 - Computed Radiography
 - Flash Radiography
- Applications
 - Defect detection
 - Anomaly detection
 - Materials characterizations
 - Feature extraction
- Sandia National Laboratories has many sources, up to 6 MeV.

First Attempt



- How far off would we be using approximations?
- Approach: acquire images at multiple energies, solve effective attenuation

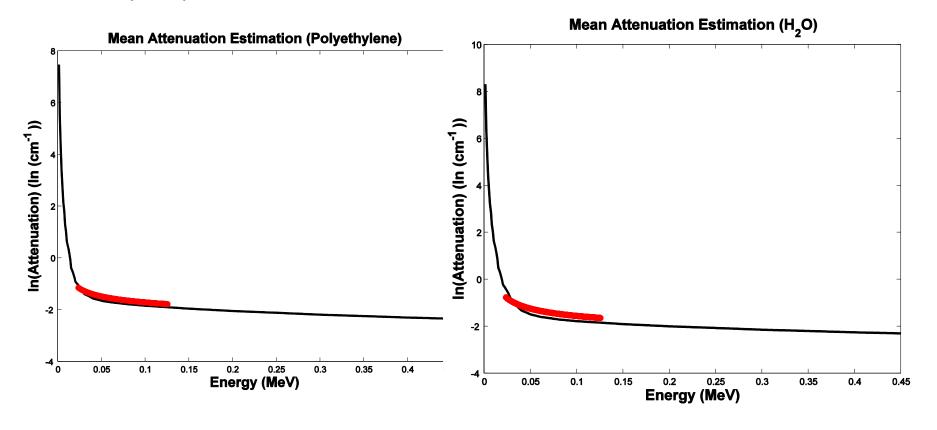




Results



Polyethylene and Water

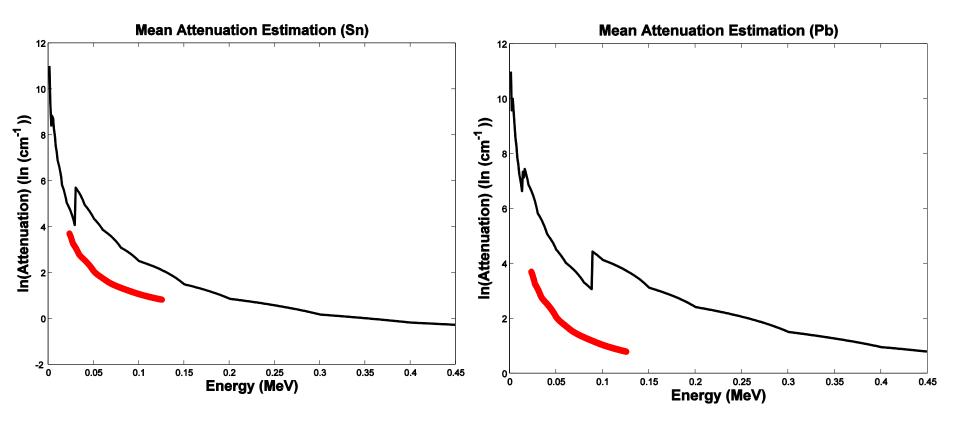


...not too bad!

Results Cont.



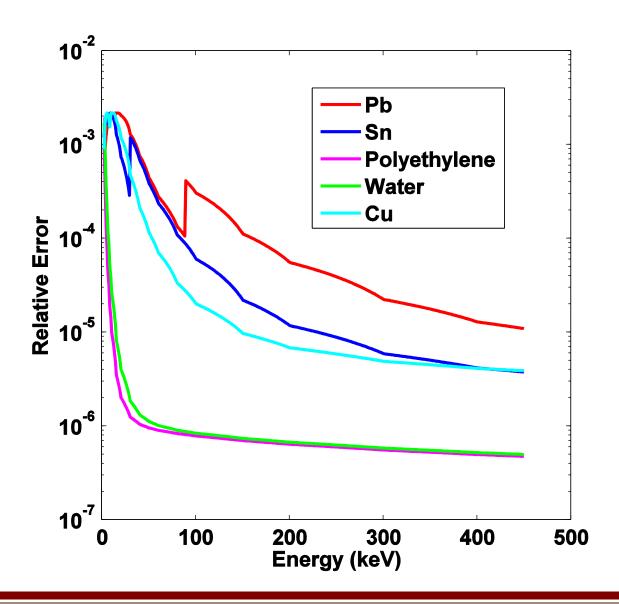
Tin and Lead



•This is problematic

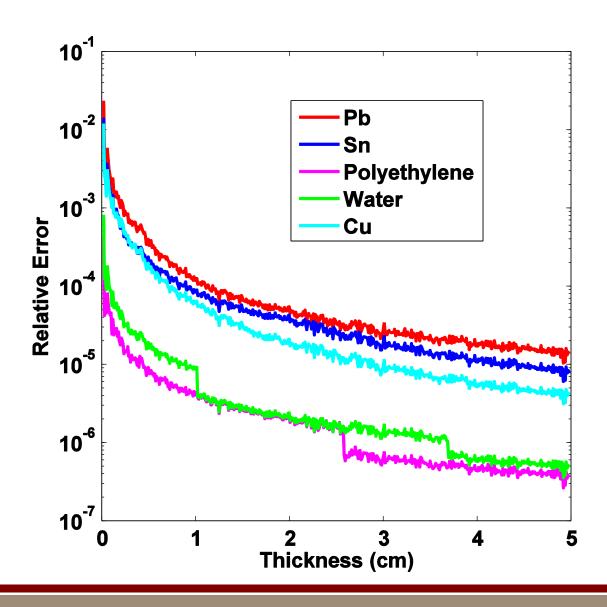
Relative error WRT Energy – Direct Evaluation





Relative error WRT Thickness – Direct Evaluation





Second Attempt



- Mediated Reality and Simulation-based Optimization
- Approach: Use basis function to create candidate materials, simulate the image and compare it to the acquired image.
- Goal: Try different basis functions, try to resolve discontinuities in the attenuation profile.
- Optimize:

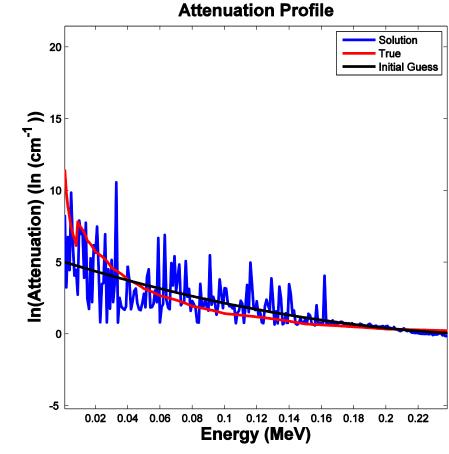
$$\underset{\hat{\mu}(x,\varepsilon)}{\operatorname{argmin}} \| \log(\overrightarrow{g}_{\mu(x,\varepsilon)}) - \log(\hat{g}_{\hat{\mu}(x,\varepsilon)}) \|_2$$

Results:



Using a very large search space (~450 dimensions)

Copper:



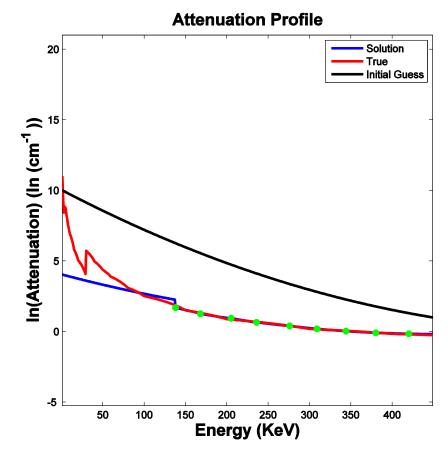
...Problematic

Results cont.



Smaller space, interval-based basis functions

Tin:



No k-edge resolved...but promising!