

# Aperture-based radiation imaging techniques

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Sandia National Laboratories

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MTV Nuclear Engineering Summer School

# Outline

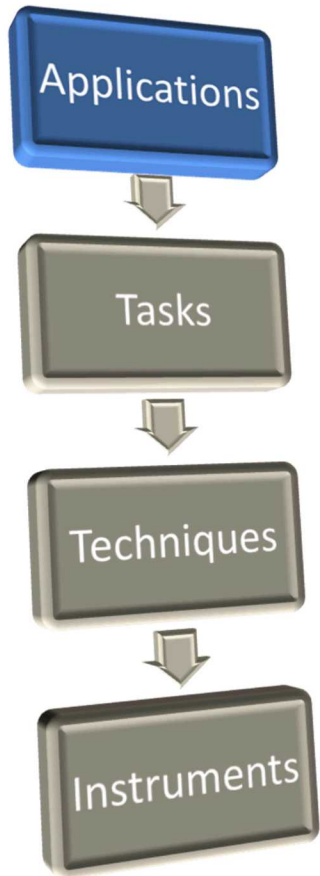
- Introduction
- Aperture imaging theory
- Image reconstruction
- Systems/results
  - Neutron coded aperture imager
  - Gamma coded aperture imager
  - Time-encoded imager (2D)
  - Time-encoded imager (1D)
  - Single-volume scatter camera with optical coded aperture



Focus on detector systems and results from my research (neutrons!)

# SNM detection/imaging

We develop systems for eventual application in a range of scenarios:



Standoff detection



Cargo screening



## SNM detection applications

- Low signal rate
  - Need large area detectors!
- Low signal to background
  - Need background discrimination!



Arms control treaty verification

Emergency  
response

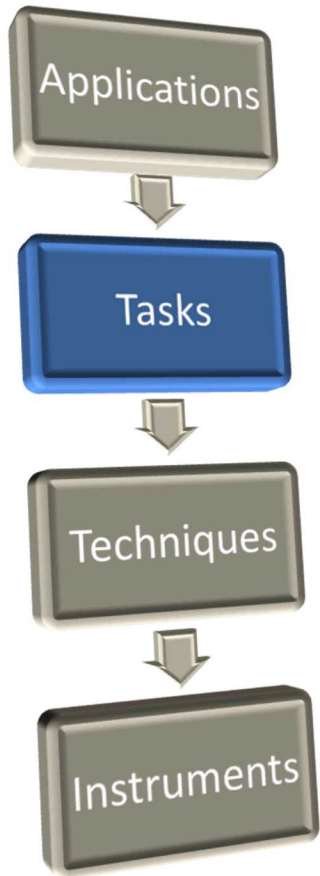


## SNM imaging applications

- High resolution required
  - Fine detector segmentation
- Multiple or extended sources



# Directional radiation detector “tasks”



## • Detection

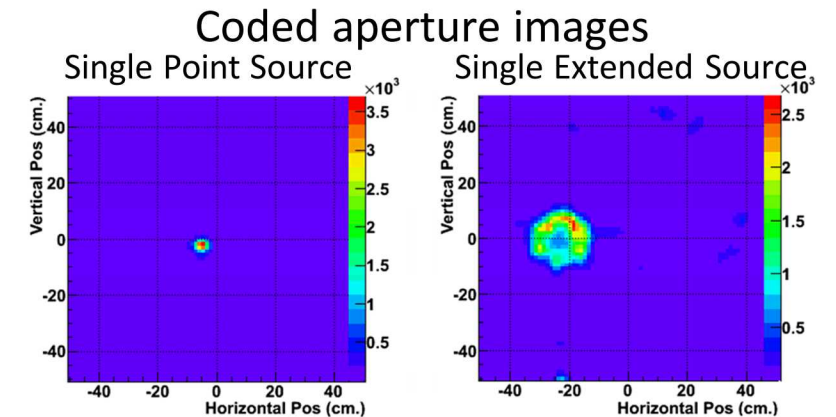
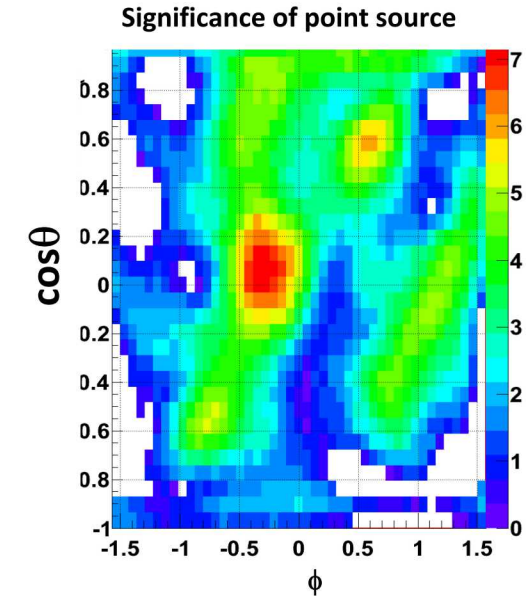
- Is there a localized source present w.r.t. diffuse background?
- Directional information can help, but in the limit no “image reconstruction” is needed

## • Localization

- What is the direction to a localized source?
- Can assume a single source on top of background
- Can perform image reconstruction (w/ constraints?) or query raw data directly

## • Imaging

- Most general case, includes possibility of multiple or extended sources
- Need full image reconstruction methods
- May want to include constraints (e.g. smoothness or sharp edges) in reconstruction
- May want to query image after the fact for e.g. dimensions or locations



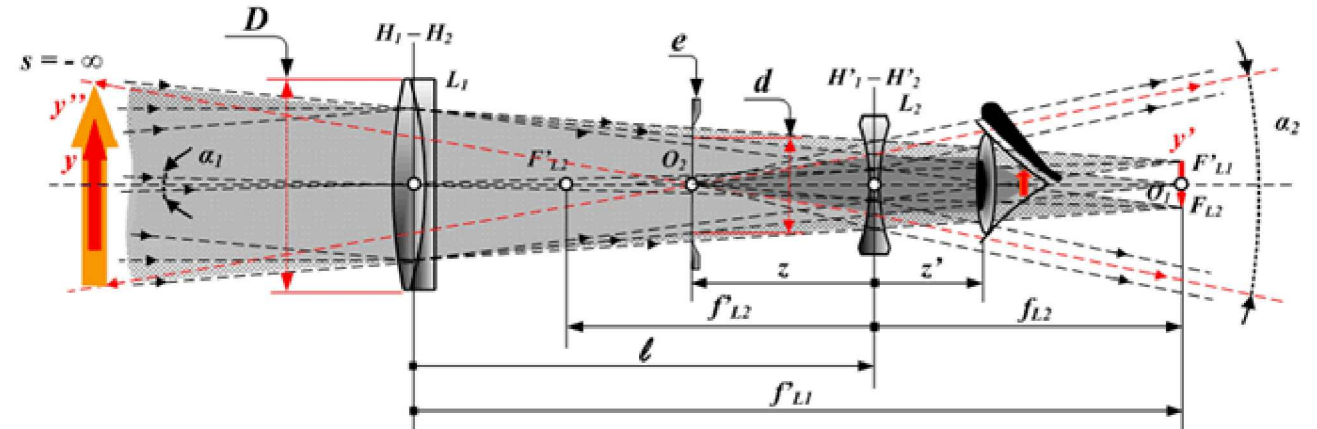




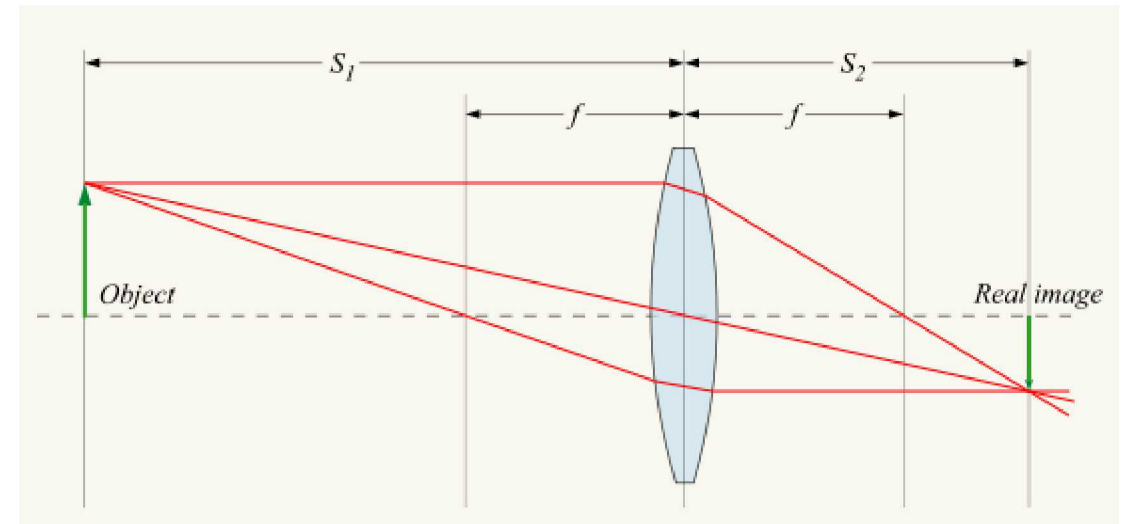
# Aperture imaging

# Optical cameras

- Lens converts position in the real world to position on a small position-sensitive detector
- Conveniently, 1-to-1 mapping of detected location to real-world location
- But we can't lens neutrons (or energetic photons, i.e. gammas)



<https://www.howtogeek.com/63409/htg-explains-cameras-lenses-and-how-photography-works/>



<https://aijaz.net/2010/01/23/how-camera-lenses-work/index.html>

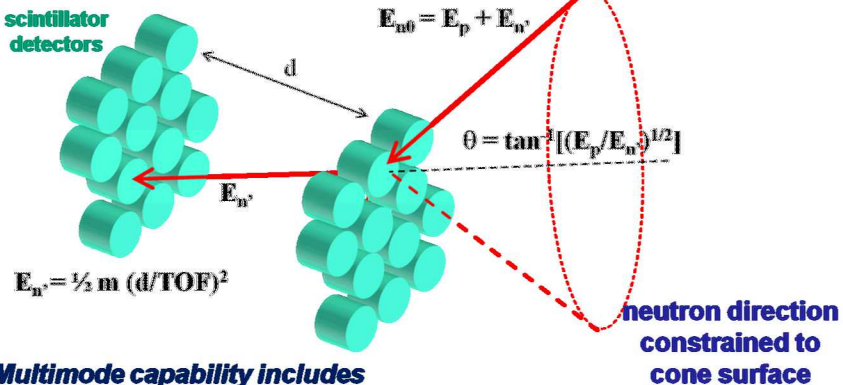
# Radiation emission imaging methods

## Kinematic imaging



**Fast neutron directions and energies constrained by double scatter geometry**

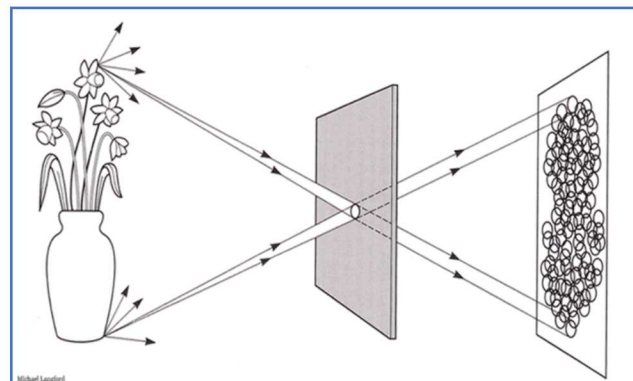
scintillator  
detectors



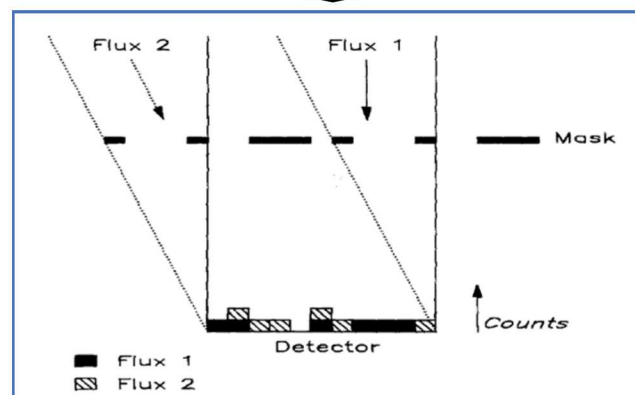
**Multimode capability includes**

- Neutron energy spectrum.
- Compton imaging.

23 July 2020

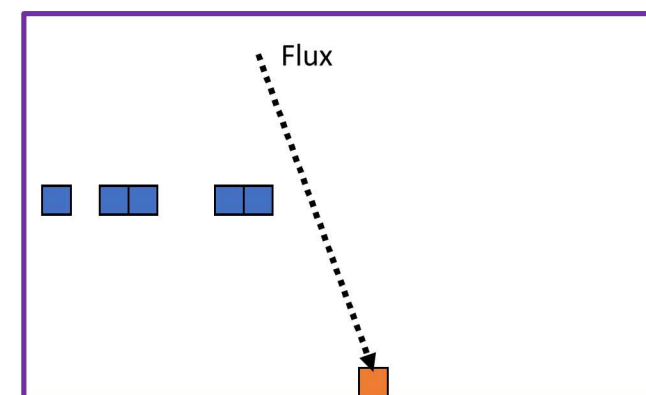


**Pinhole: High Resolution, Low Throughput**



**Coded aperture: High Resolution, High Throughput**

## Aperture imaging

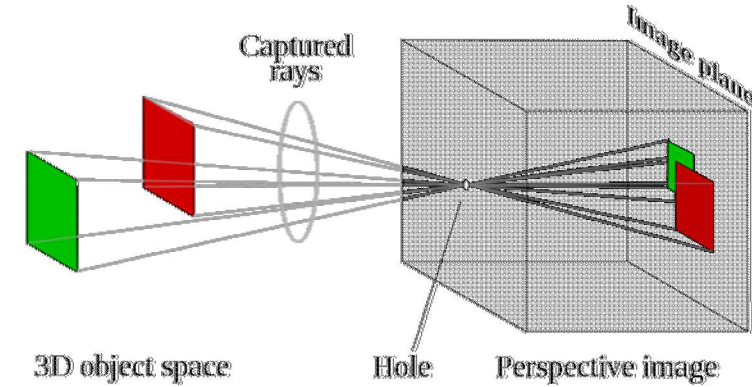


**Time-encoded imaging: Low Sensitivity, Low complexity**

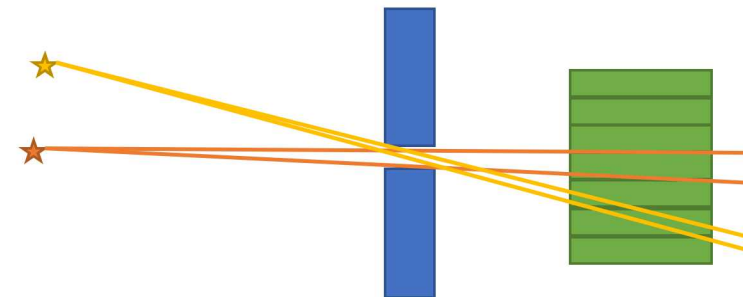
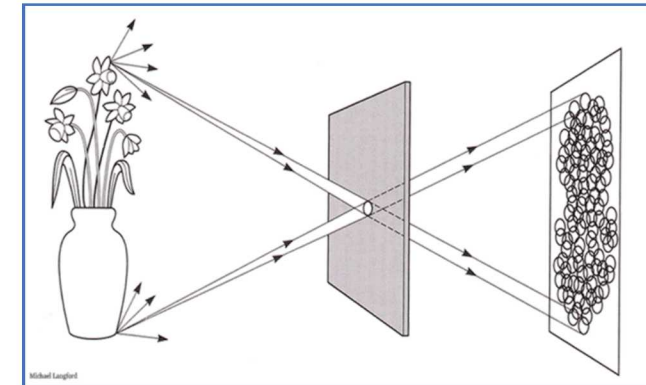


# Pinhole imaging

- Photons/particles stream through a small hole in an otherwise opaque mask
- Notice parameters of interest:
  - Pinhole size: resolution/sensitivity tradeoff
  - Mask-detector distance: magnification
  - Image plane size: field of view
- For energetic radiation, two more:
  - Image plane resolution (incl detector thickness): limits angular resolution
  - Mask thickness: vignetting (FOV)

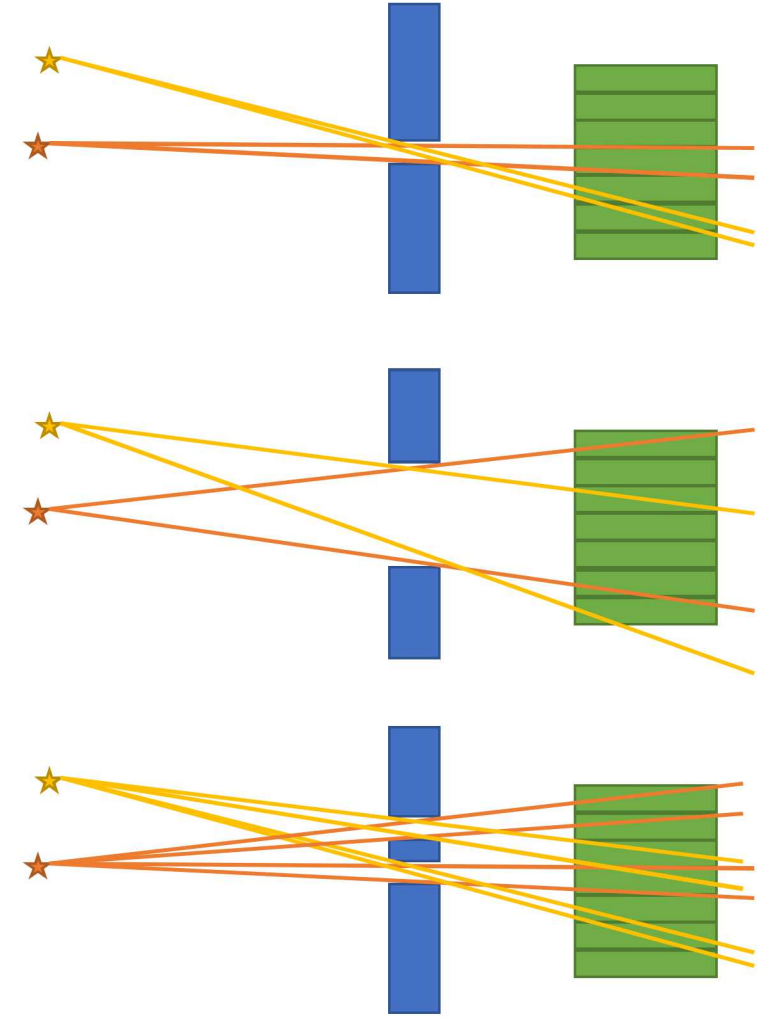


<https://www.howtogeek.com/63409/htg-explains-cameras-lenses-and-how-photography-works/>



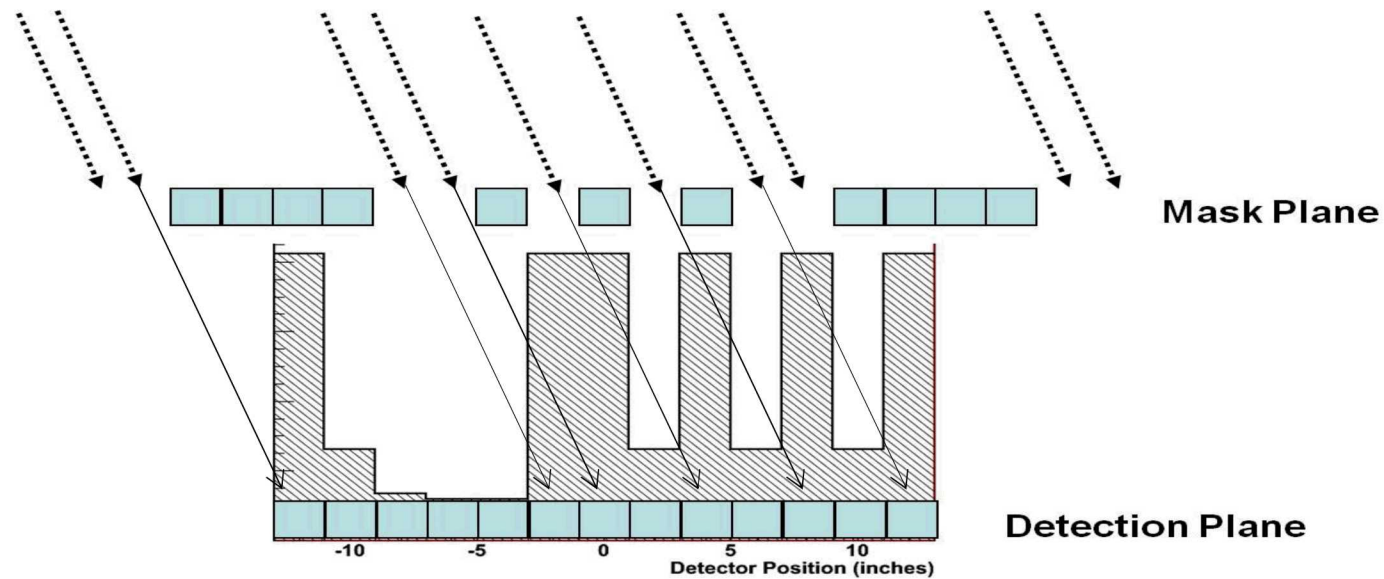
# Pinhole -> coded aperture

- What if we need more sensitivity but also good resolution?
- Diminishing returns on making pinhole larger
- Instead make many smaller holes
- Now have high sensitivity, high resolution, BUT also high complexity reconstruction



# Coded Aperture Imaging

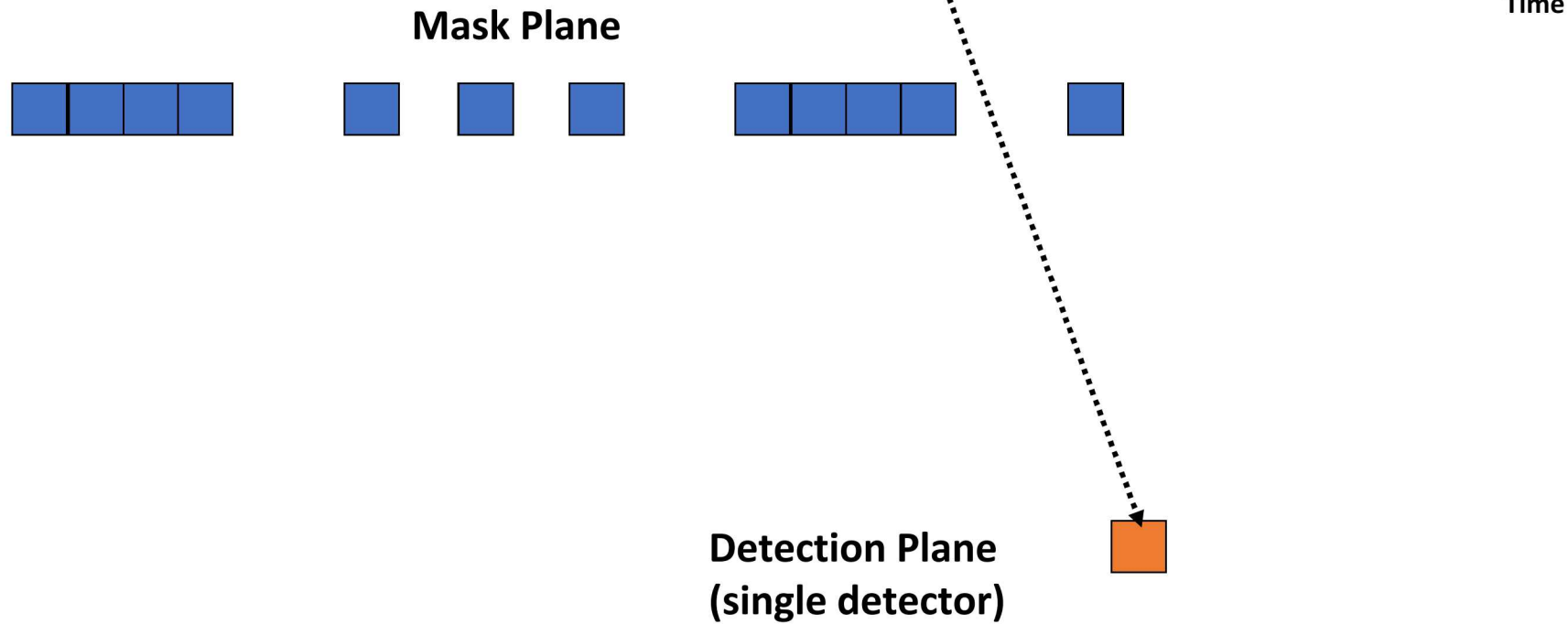
- Static aperture is used to modulate the flux emitted by an unknown source
  - Modulated flux intensity is measured at the detector plane by a position sensitive detector
- High-sensitivity extension of pinhole imaging
- Mask pattern is essential





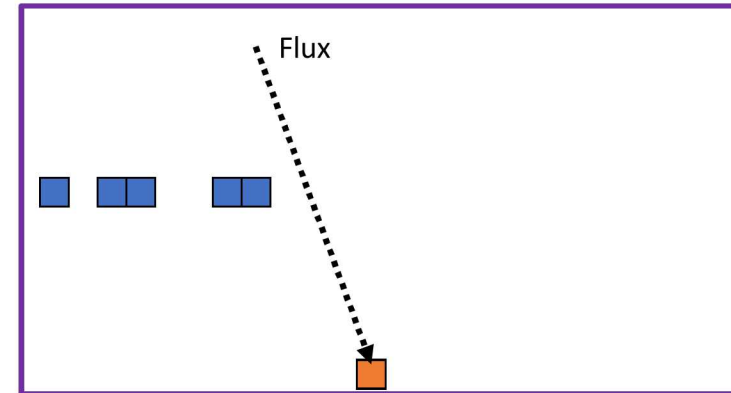
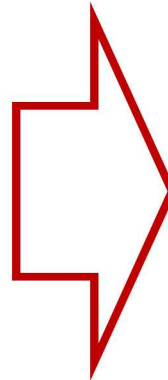
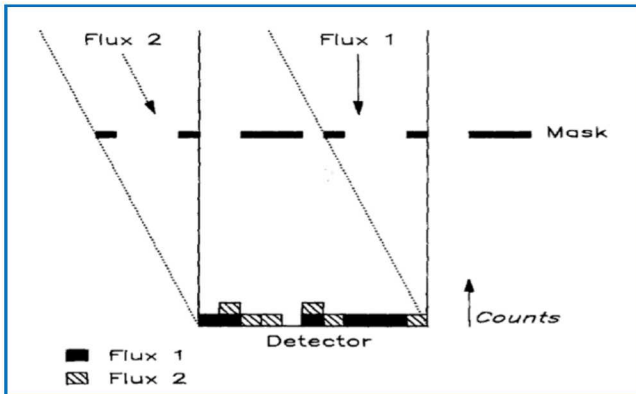
# Time-Encoded Imaging

- Moving aperture is used to modulate the flux emitted by an unknown source
  - Modulated flux intensity is measured by a time-sensitive detector



# Why TEI?

- Time-encoded imaging (TEI) is analogous to coded aperture imaging:
  - The *spatial* modulation of a particle flux induced by a *fixed mask* on a *position-sensitive image plane* is replaced by the *time* modulation of a particle flux induced by a *moving mask* on one or a few *time-sensitive detectors*.



- Spatial resolution in detectors is expensive. **Time resolution in detectors is ~free!**
- Accurately resolving spatial patterns requires precise inter-calibration of detector channels or regions. **In TEI, a single detector can be used;** or if multiple, they are independent.
- Simple** and **robust, low-channel-count** systems.
- Scalable to **large effective area**.
- Tradeoffs are **low sensitivity** and/or **low resolution**.



# Image reconstruction



# Mask patterns

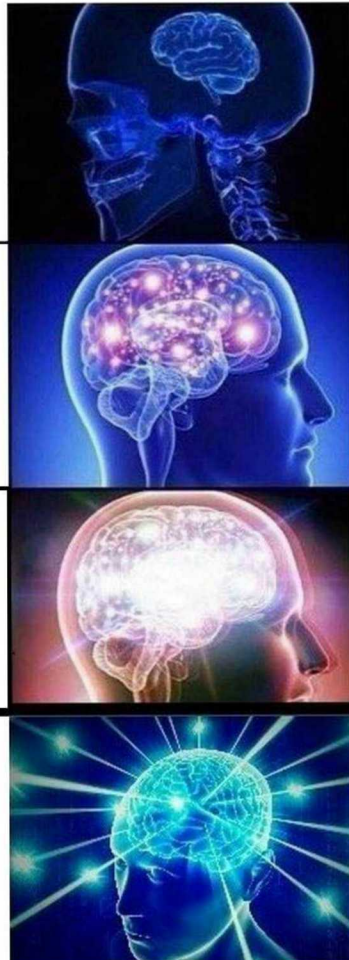
- Mask pattern is key to coded aperture image reconstruction

Lots 'o'  
apertures

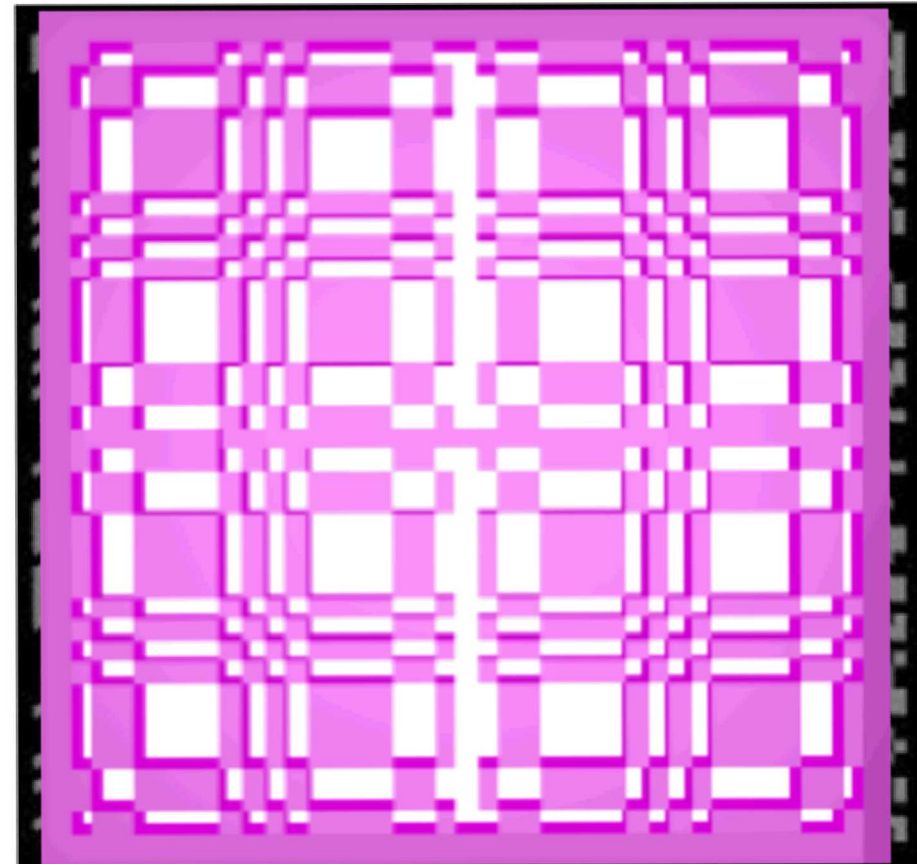
Code unique  
pattern for  
each source

Code orthogonal  
pattern for  
each source

Totally random  
mask pattern



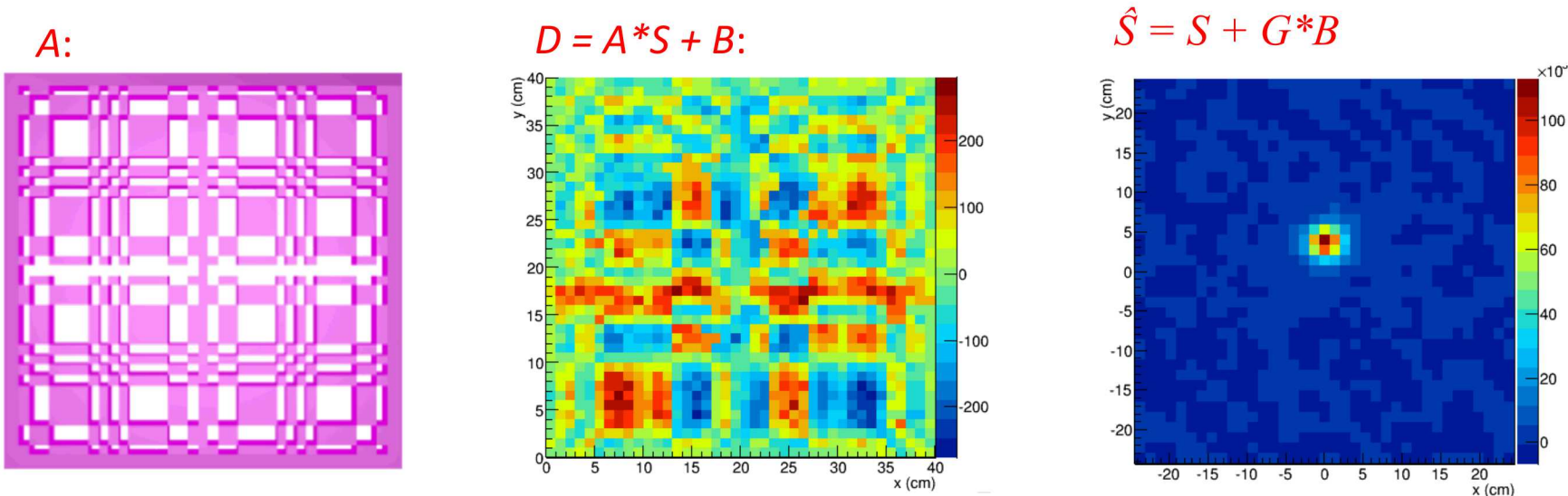
Rank 19 modified uniformly redundant array (MURA)

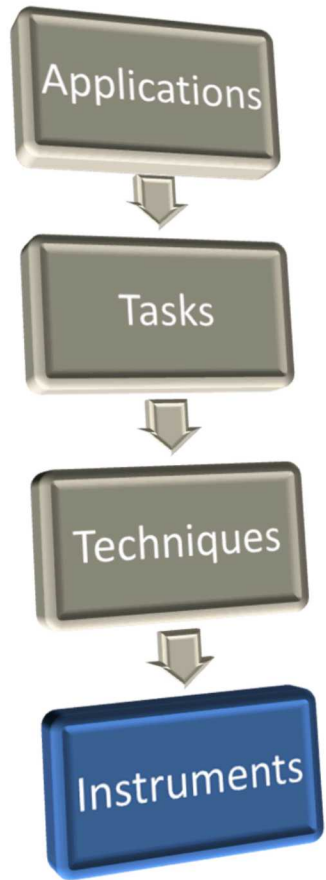


DOI: 10.1007/s11214-017-0454-5

# Coded aperture image reconstruction

- Reconstruction procedure:
  - with known mask pattern ( $A$ )
  - measure the projected pattern ( $D$ ) with a pixelated detector: includes an un-modulated background term ( $B$ ) in addition to modulated term ( $A*S$ )
  - reconstruct the source distribution ( $\hat{S}$ ) by decoding with suitable matrix ( $G*A = \delta$ ):
 
$$\hat{S} = G*D = S + G*B$$
- Modified Uniformly Redundant Array (MURA):
  - optimizes signal to noise ratio for point source
  - approximate delta-function point spread response, so there will be some artifacts
  - Tile MURA to get anti-mask ( $A'$ ) via rotation—adds second pattern ( $D' = A'*S + B$ ), and subtracted pattern ( $D'' = D - D' = A*S - A'*S = 2A*S$ )





# Neutron coded aperture

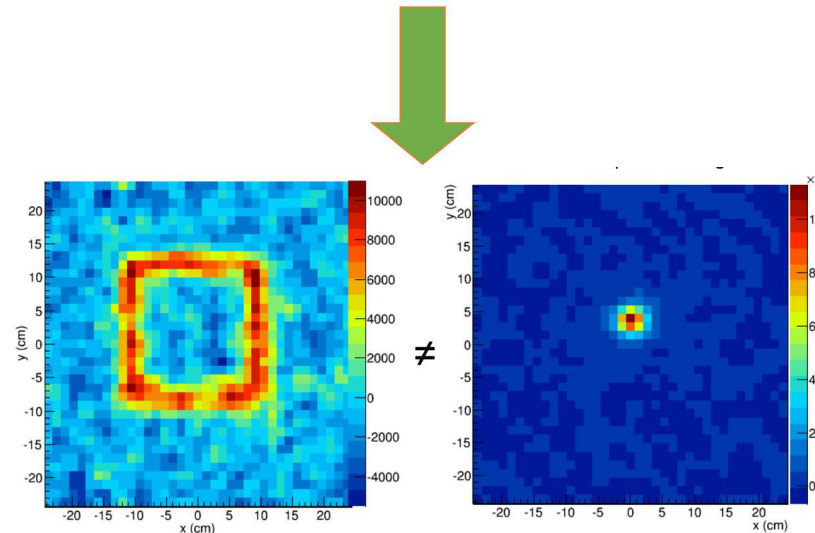


- Fast neutrons, and to a lesser degree gammas, modulated by rank-19 MURA made of HDPE
- Mask rotates to acquire mask/anti-mask data
- Mask thickness, detector-mask separation are adjustable to trade off field of view, contrast, and imaging resolution
- PSD capable plastic scintillator (EJ-299-34) based fast neutron detector
  - 16 pixelated block detectors with 100 pixels each: 1600 one cm pixels total
  - Each block detector read out by four PMTs, with pixel determined by Anger logic
- Enables characterization of SNM based on geometrical layout

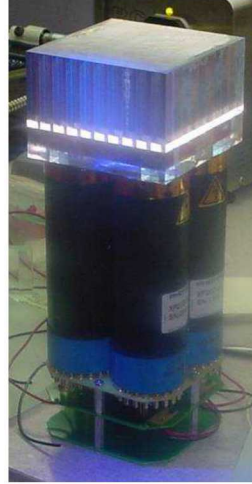
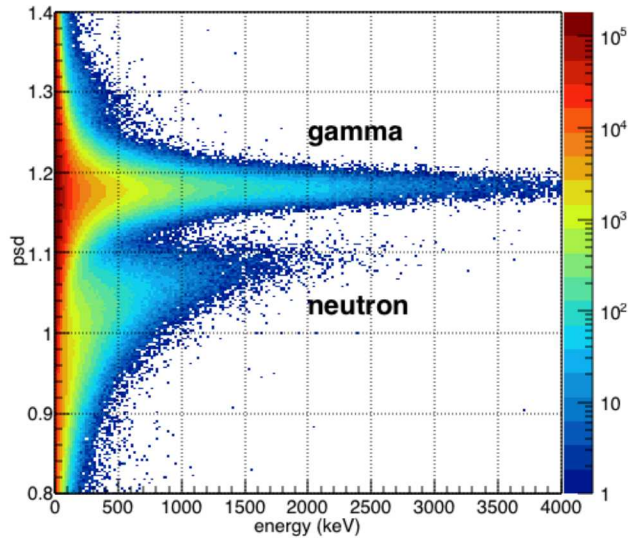


Slides based on presentation by Melinda Sweany (SNL)

ORNL collaborators on NCAI:  
Paul Hausladen  
Jason Newby  
Matthew Blackston



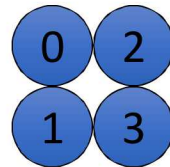
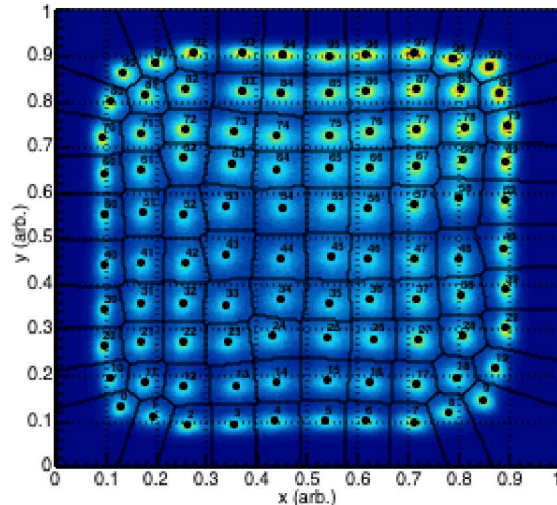
# Pixelated fast neutron/gamma detection



- Gamma/neutron discrimination with EJ-299-34
- Scintillation from optically separated 1x1 cm pixels propagate through light guide to PMTs
- Normalized sum of two orthogonal PMTs yield pixel responses in x and y:

$$x = (2+3)/(0+1+2+3)$$

$$y = (0+2)/(0+1+2+3)$$

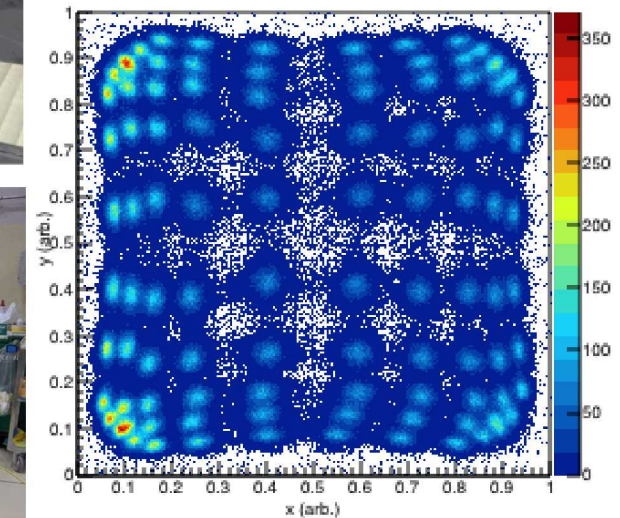
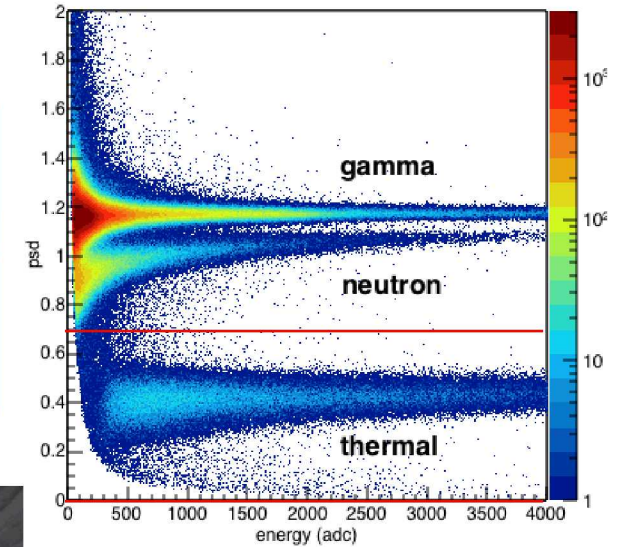
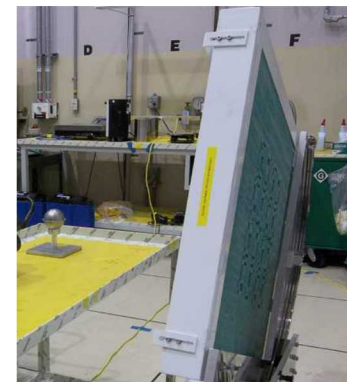
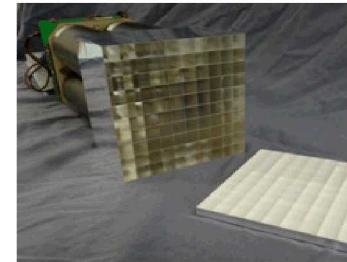
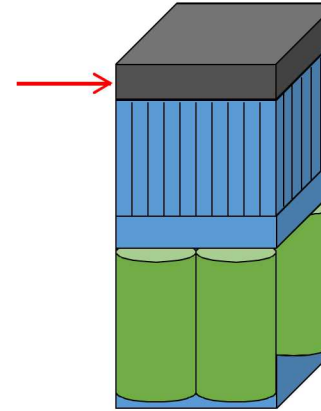


- Algorithm creates a Look Up Table (LUT) that maps x and y centroids to pixel numbers



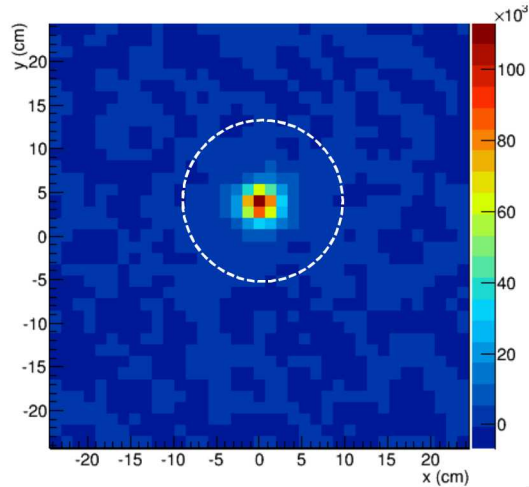
# Thermal Neutron detection/modulation

- Thermal neutrons add information on nearby moderating material—image the surface of last scatter
  - High explosives
  - Shielding materials
- Thermal detection from added layer of EJ-426HD ( $^6\text{Li}/\text{ZnS}:\text{Ag}$ ):
  - $n+^6\text{Li} \rightarrow \alpha + t$
  - Pulse shape easily separable from gamma & fast neutron
- Thermal neutrons are modulated with borated polyethylene mask, situated between HDPE (fast neutron mask) and detector array

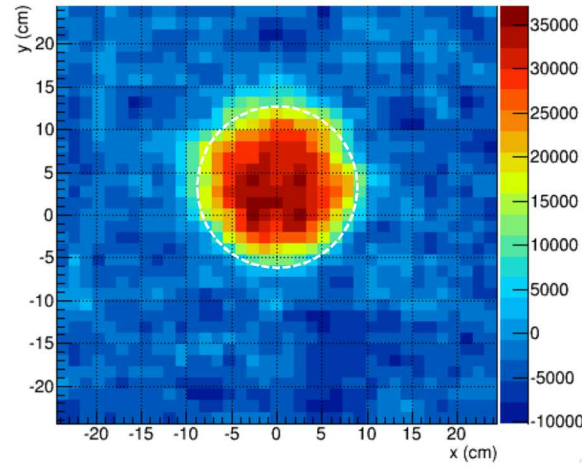


# $^{252}\text{Cf}$ inside 7" polyethylene vs. 7" borated polyethylene sphere

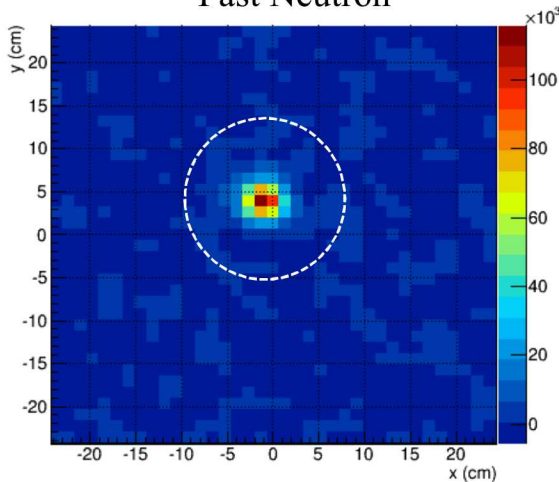
Fast Neutron



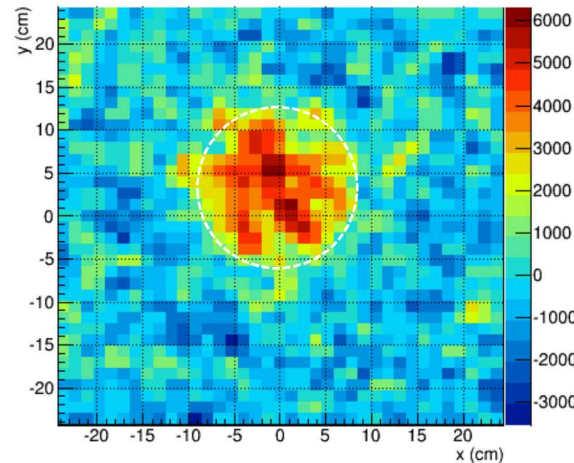
Thermal Neutron



Fast Neutron



Thermal Neutron

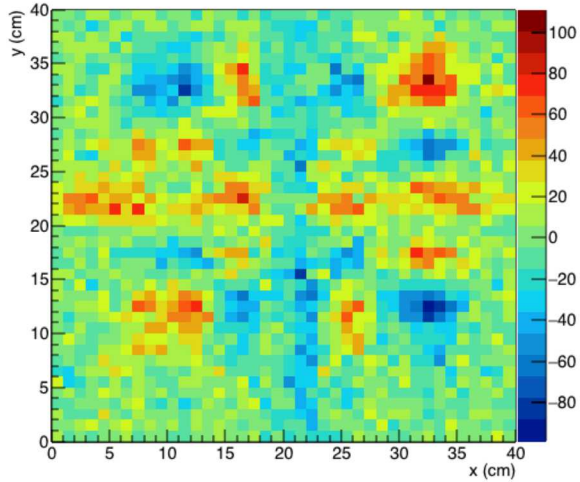


- Fast neutron image shows point source from  $^{252}\text{Cf}$
- Thermal image shows polyethylene moderator edge
- Borated polyethylene sphere shows some leakage, but thermal source strength is cut by  $\sim 6\times$

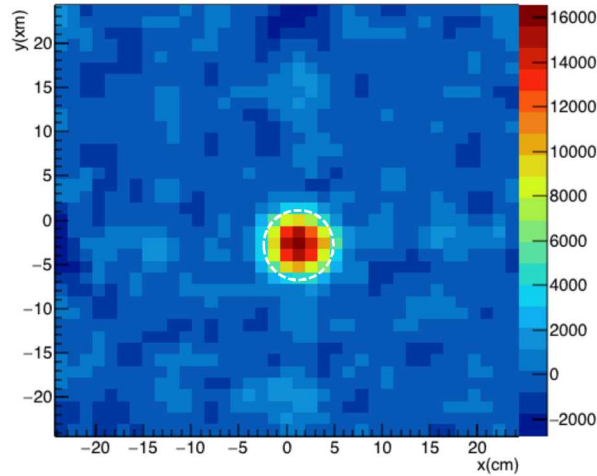


# Bare BeRP Ball – 12 mins total

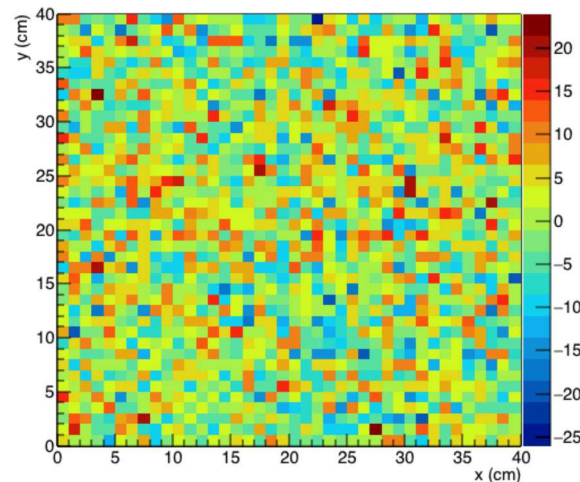
Raw Fast Neutron Data



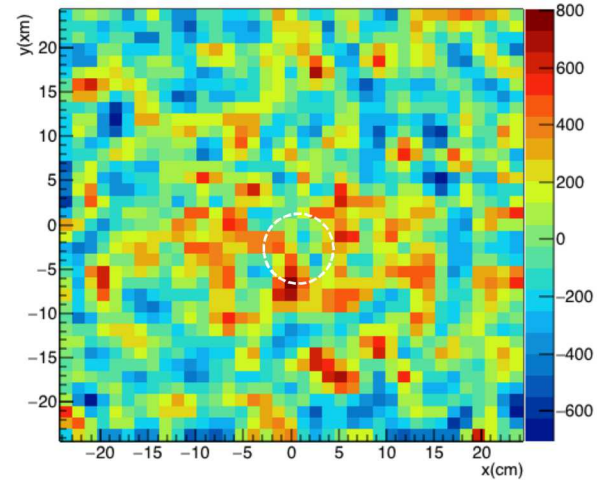
Fast Neutron



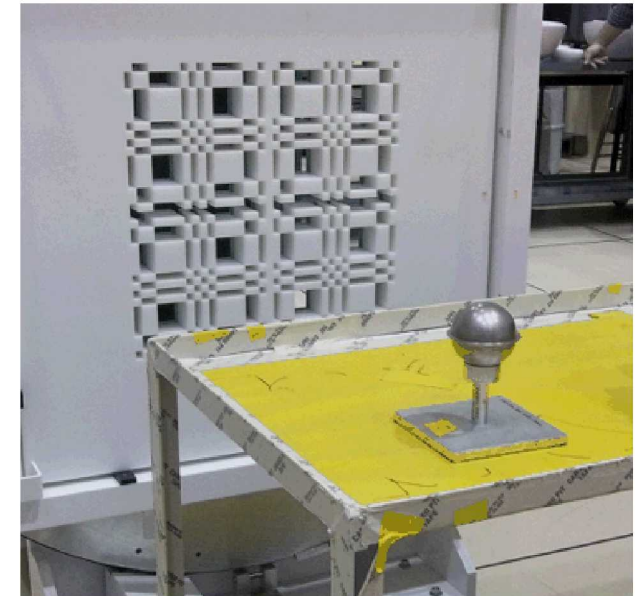
Raw Thermal Data



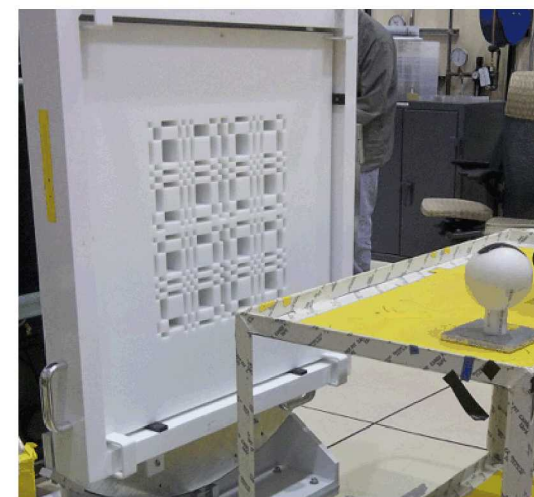
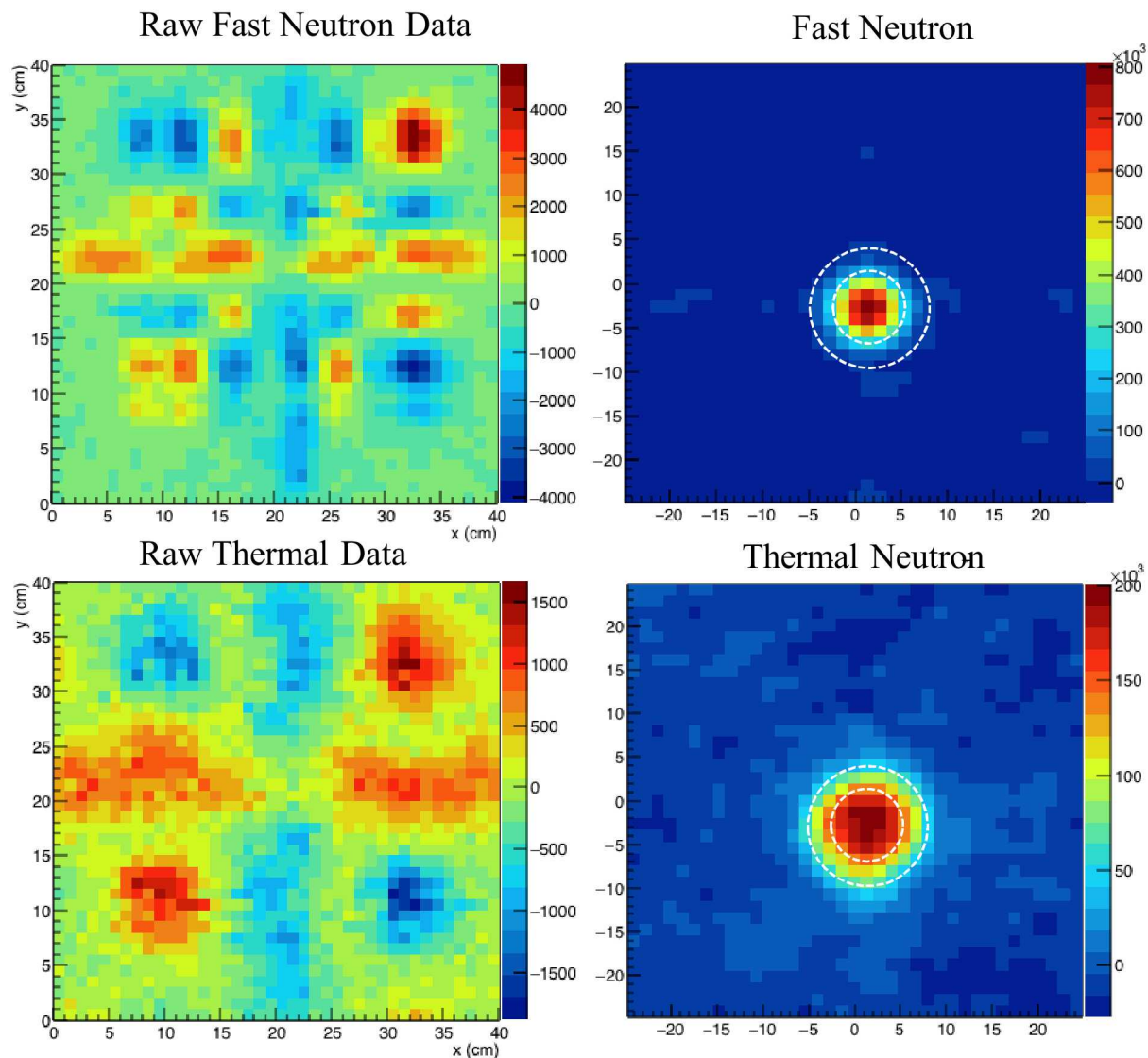
Thermal Neutron



- BeRP Ball:
  - 3.8 cm radius sphere
  - 4.5 kg WGPu



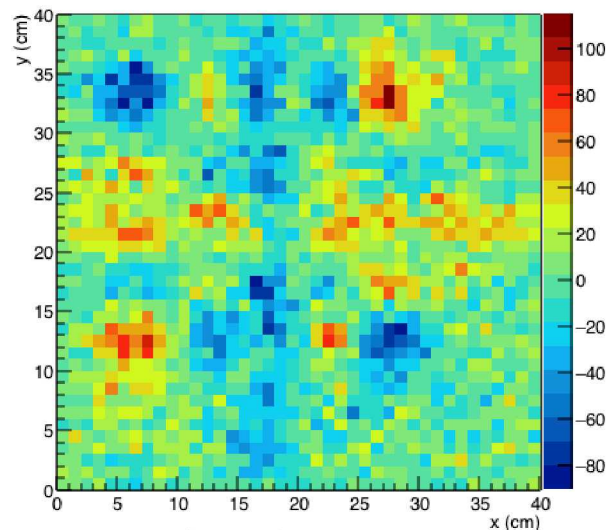
# BeRP Ball with 1" Poly – 15 hours total



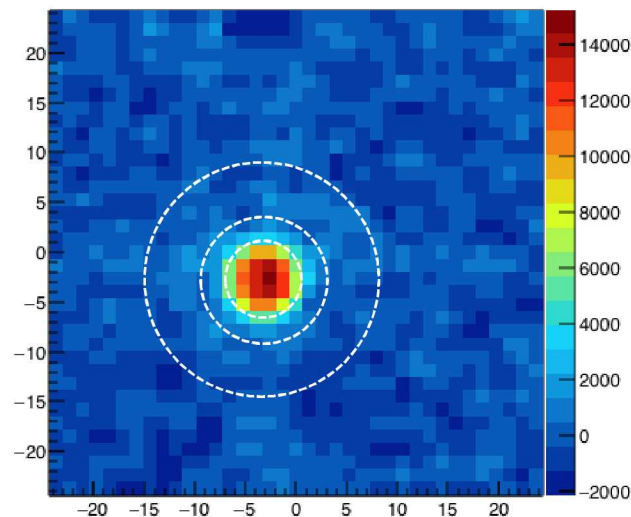


# BeRP Ball with 1" void, 2" Poly – 28 mins total

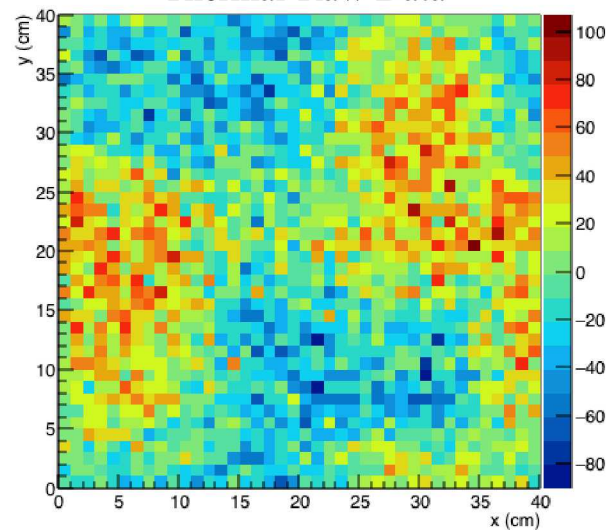
Fast Neutron Raw Data



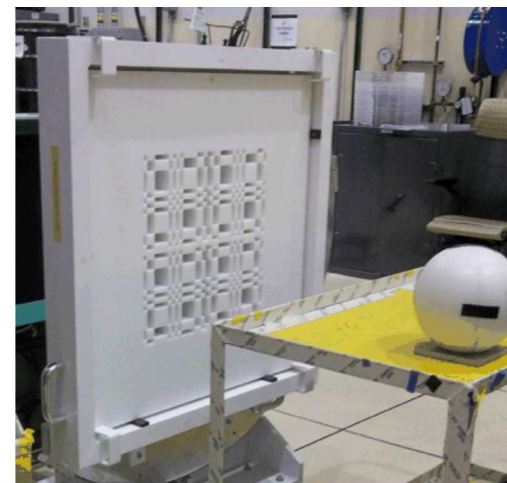
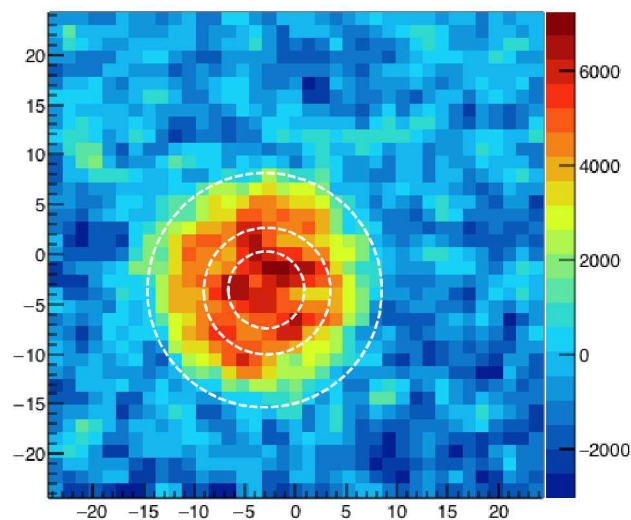
Fast Neutron



Thermal Raw Data

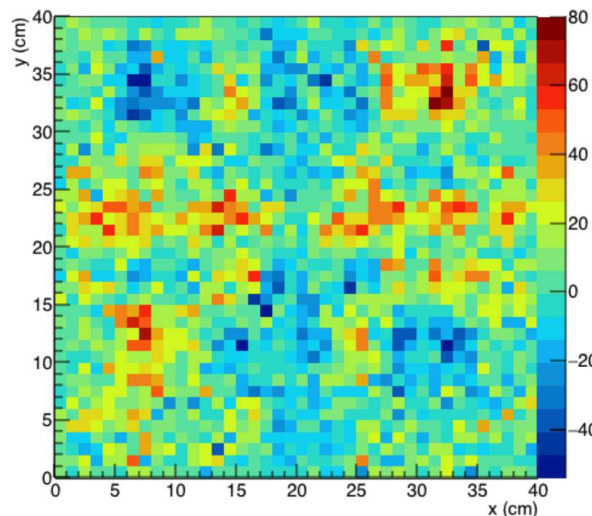


Thermal Neutron

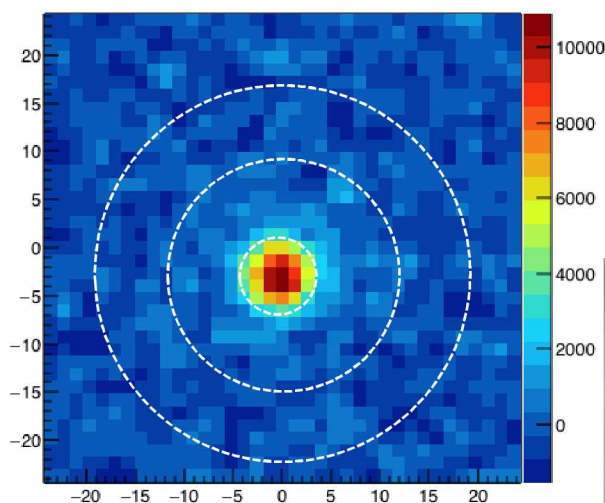


# BeRP Ball with 3 inch void, 3 inch Poly – 48 mins total

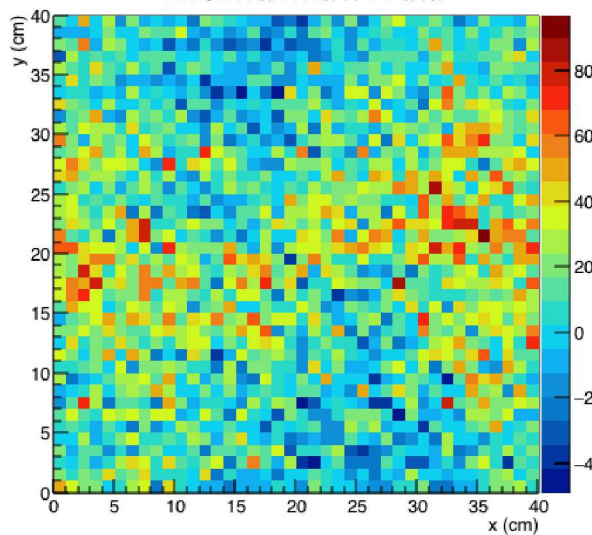
Fast Neutron Raw Data



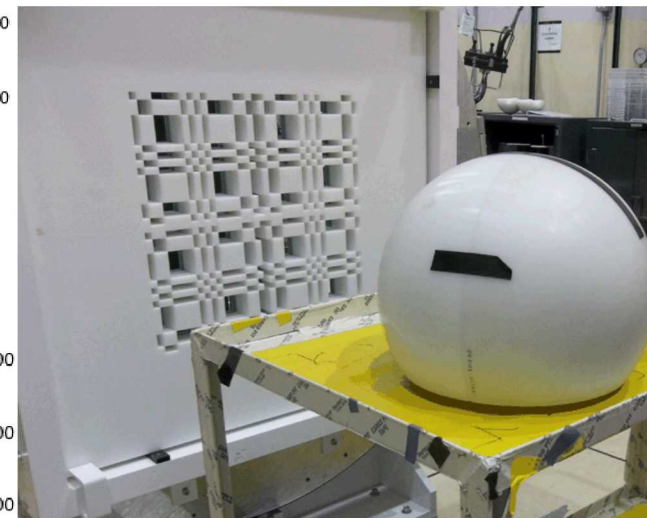
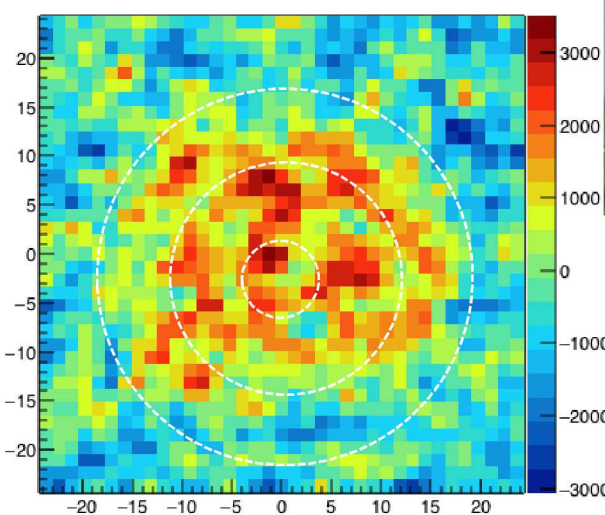
Fast Neutron



Thermal Raw Data

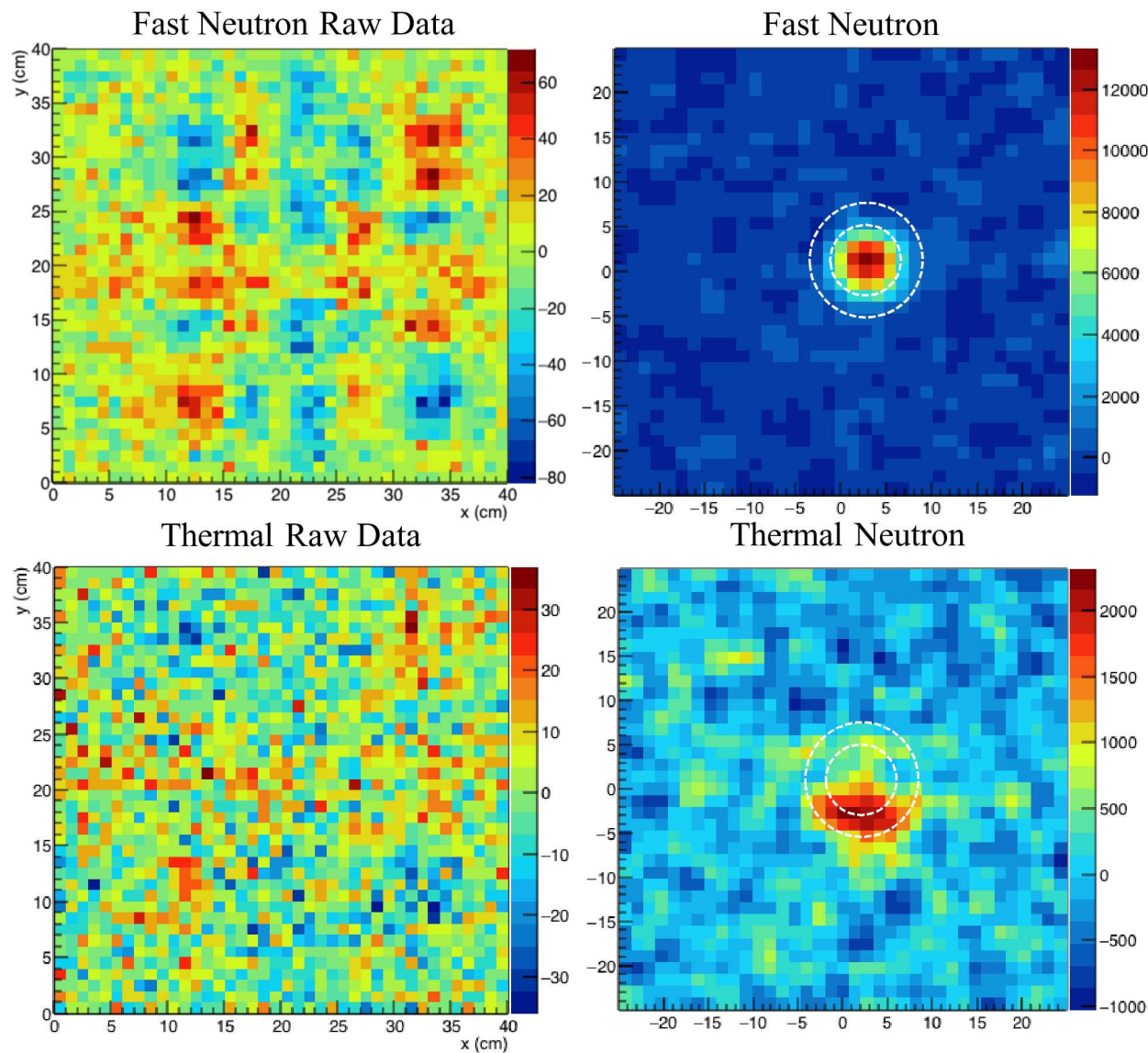


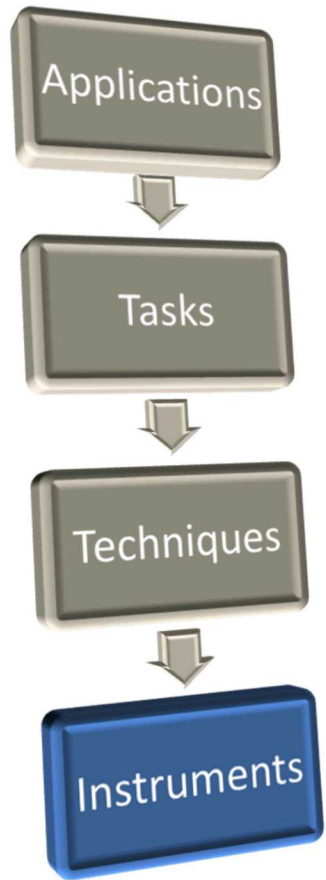
Thermal Neutron





# BeRP Ball with 1.25" Poly Hemi – 16 mins total





# Gamma coded aperture

# What is different for gammas?

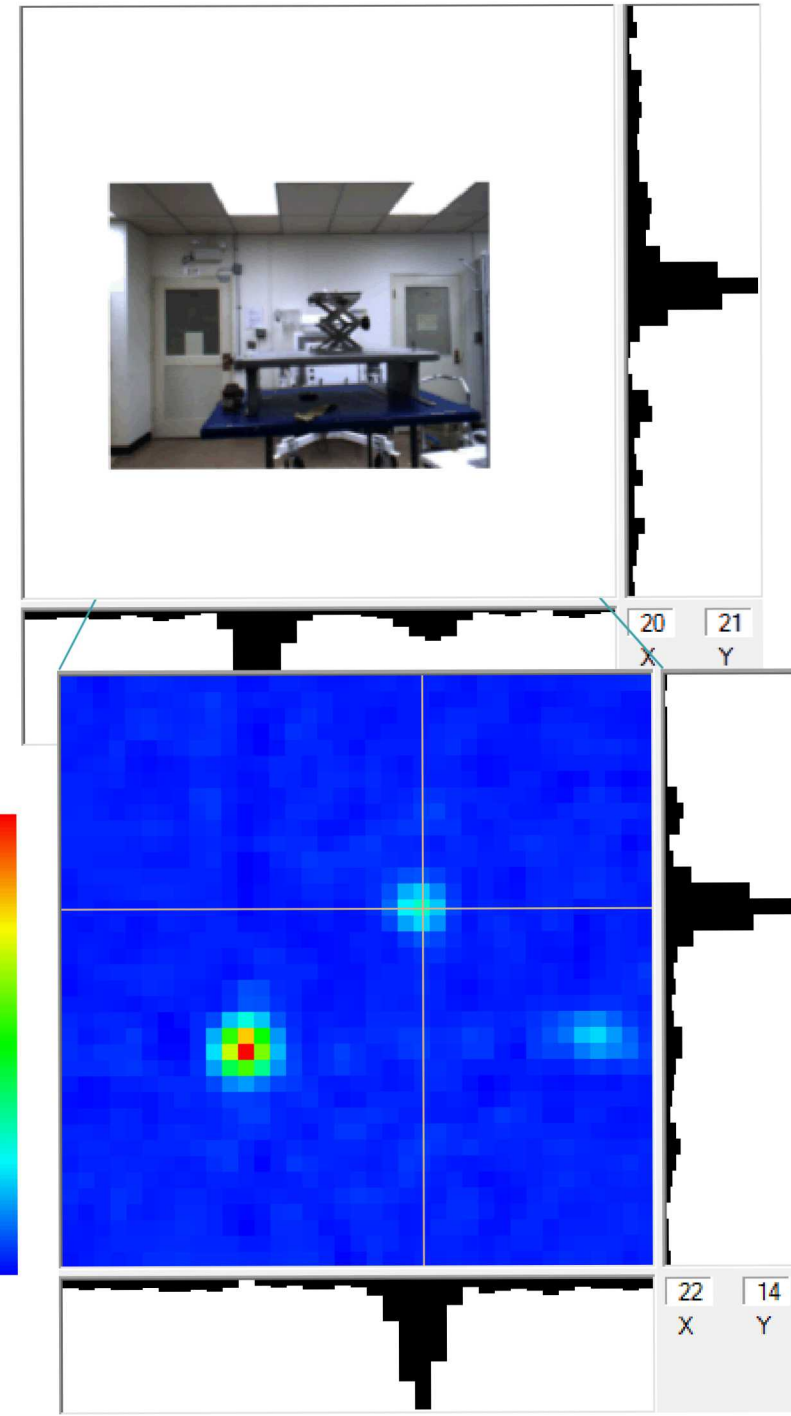
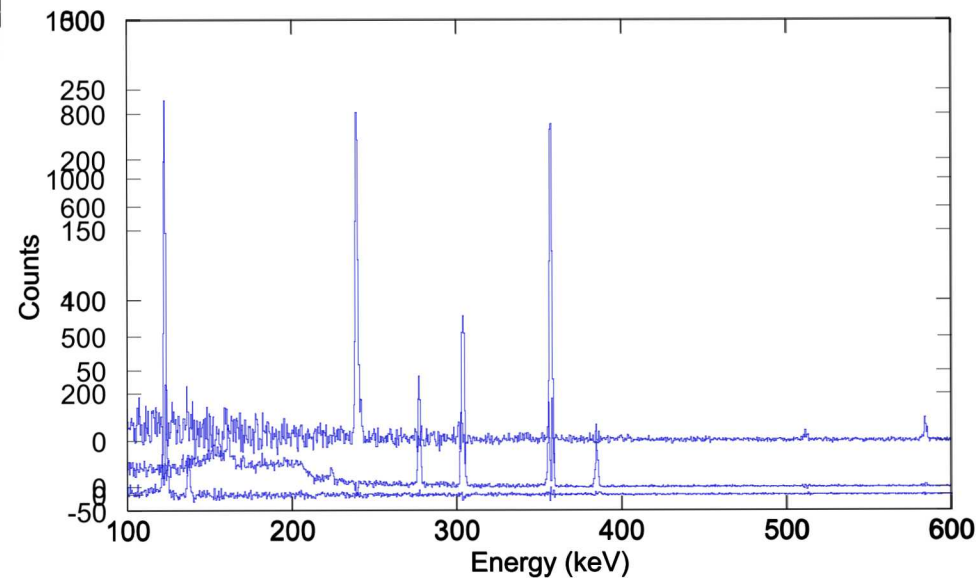
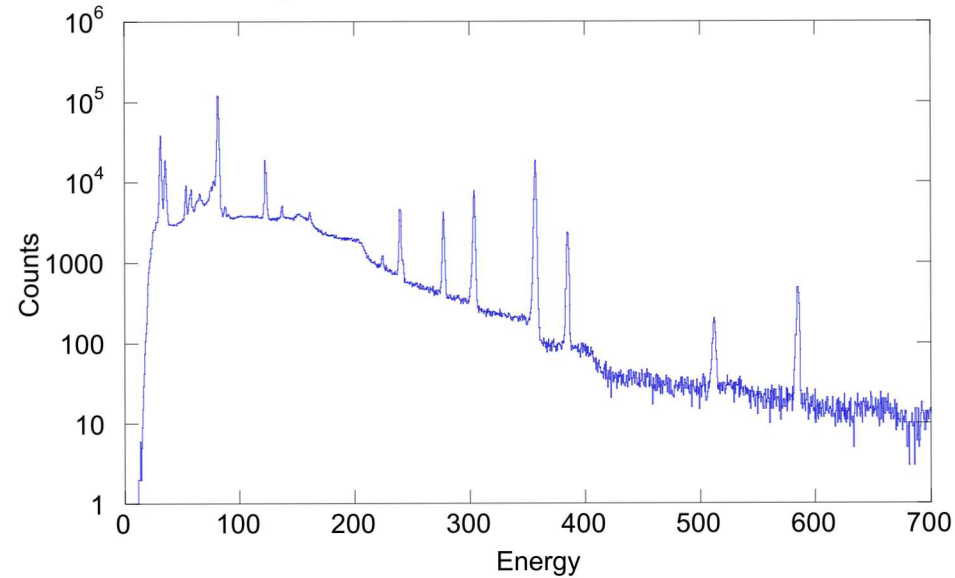
- Physics of gamma production (well-defined decay lines)
- Physics of gamma interactions
- Typically higher rates



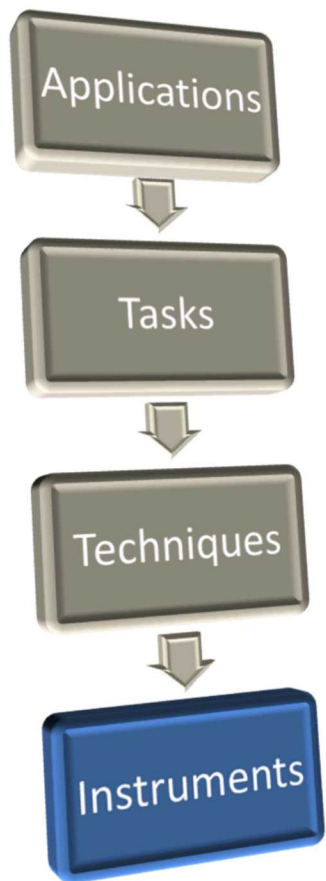
- Detector material: NaI, HPGe, CZT
- Mask material: high-Z (lead, tungsten, etc.)
- Detector/mask thickness
- Spectroscopy

# ORNL HPGe Coded-Aperture Imager

Thanks to  
Klaus Ziock (ORNL)





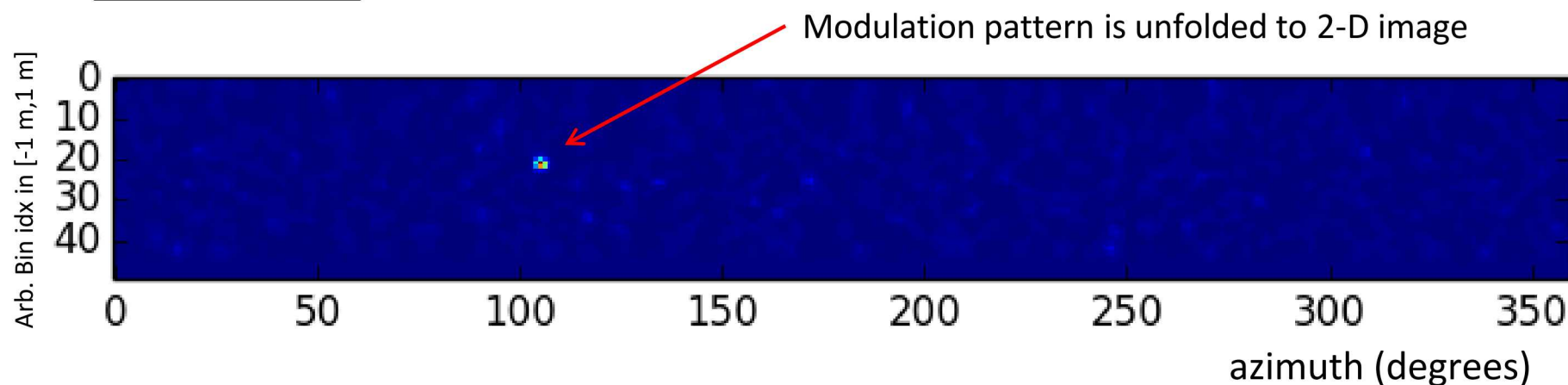
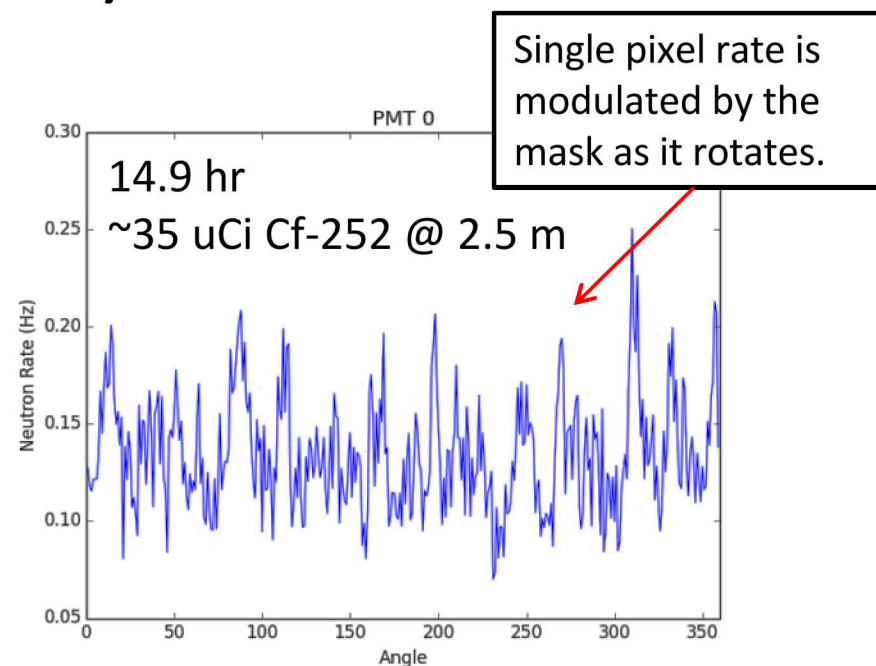
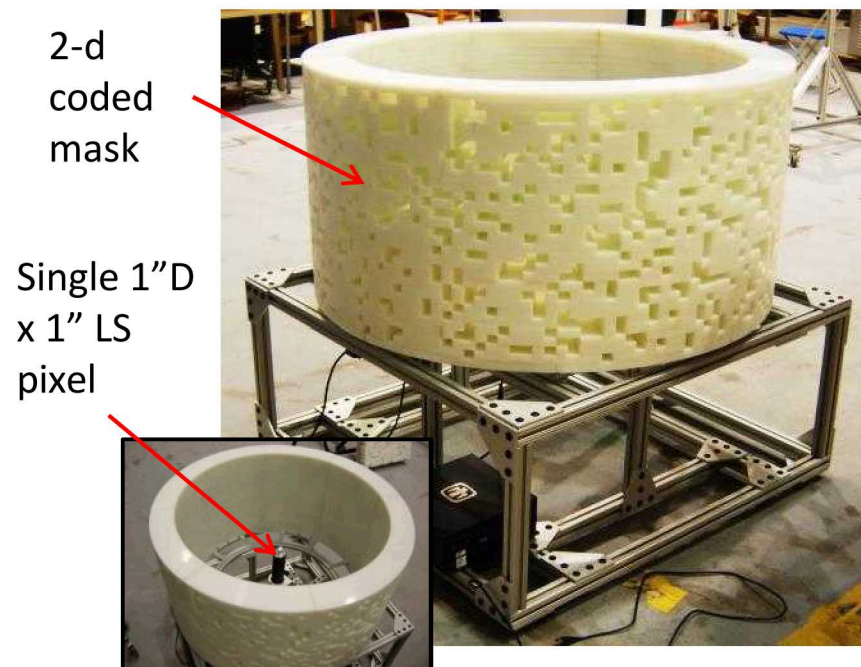


# TEI-2D

Many SNL contributors:

Peter Marleau  
Melinda Sweany  
Nathalie Le Galloudec  
Mark Gerling  
Aaron Nowack  
Jim Brennan  
Etc.

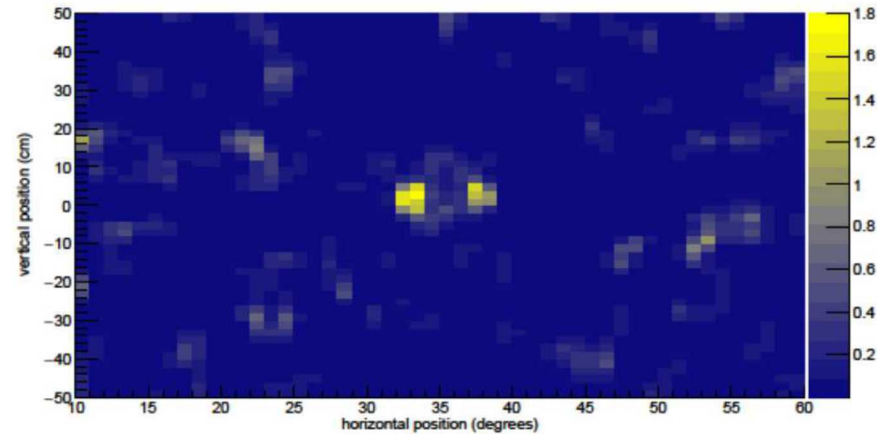
# TEI-2D demonstration system



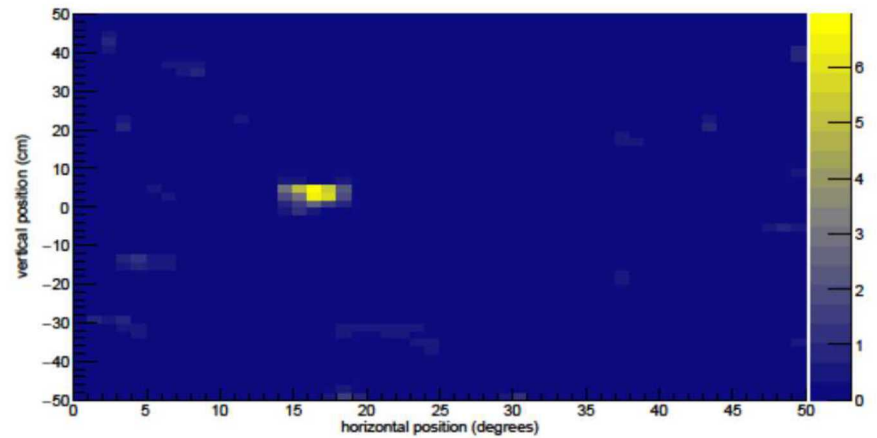
# TEI-2D results (1)

Two  $1.4\text{e}5$  n/s  $^{252}\text{Cf}$  point sources at 2 m distance from center of detector.

5 degree separation in 1 hour  
(50 mlem iterations)



2 degree separation in 24 hours  
(250 mlem iterations)

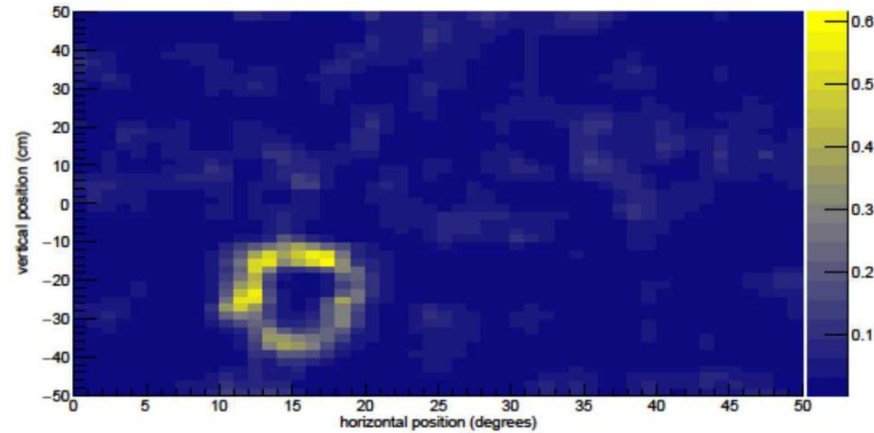


NIM A, **802** (2015) p. 76

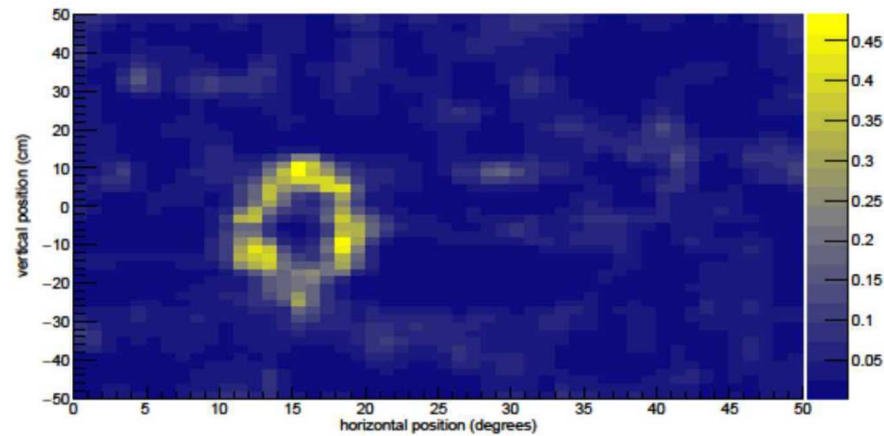
# TEI-2D results (2)

Single  $1.4\text{e}5$  n/s  $^{252}\text{Cf}$  point source moved through a ring pattern at 2 m distance.

Location 1: 72 hr  
(100 MLEM iterations)



Location 2: 94 hr  
(100 MLEM iterations)

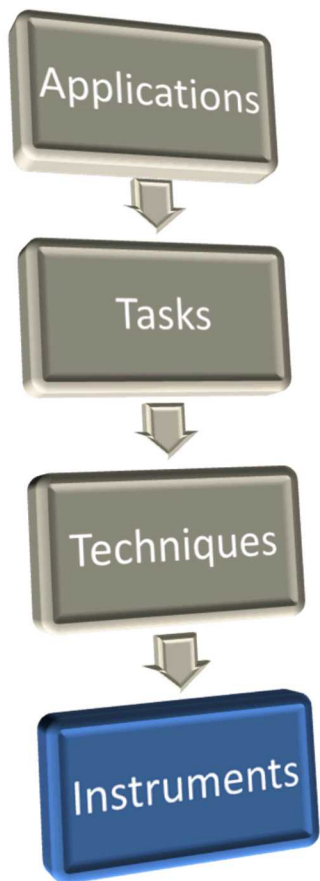


NIM A, **802** (2015) p. 76



# What is it good for?

- TEI-2D provides cm-scale neutron emission image resolution with a single detector channel
- Requires long measurement times
  - Reduce somewhat by adding cells (more detectors)
  - Or by worsening resolution (larger detector)
- Potentially applicable when simple, robust detector is attractive, long dwell time acceptable
  - Arms control treaty verification (CONFIDANTE)
  - Material holdup measurements
  - Monitored material storage

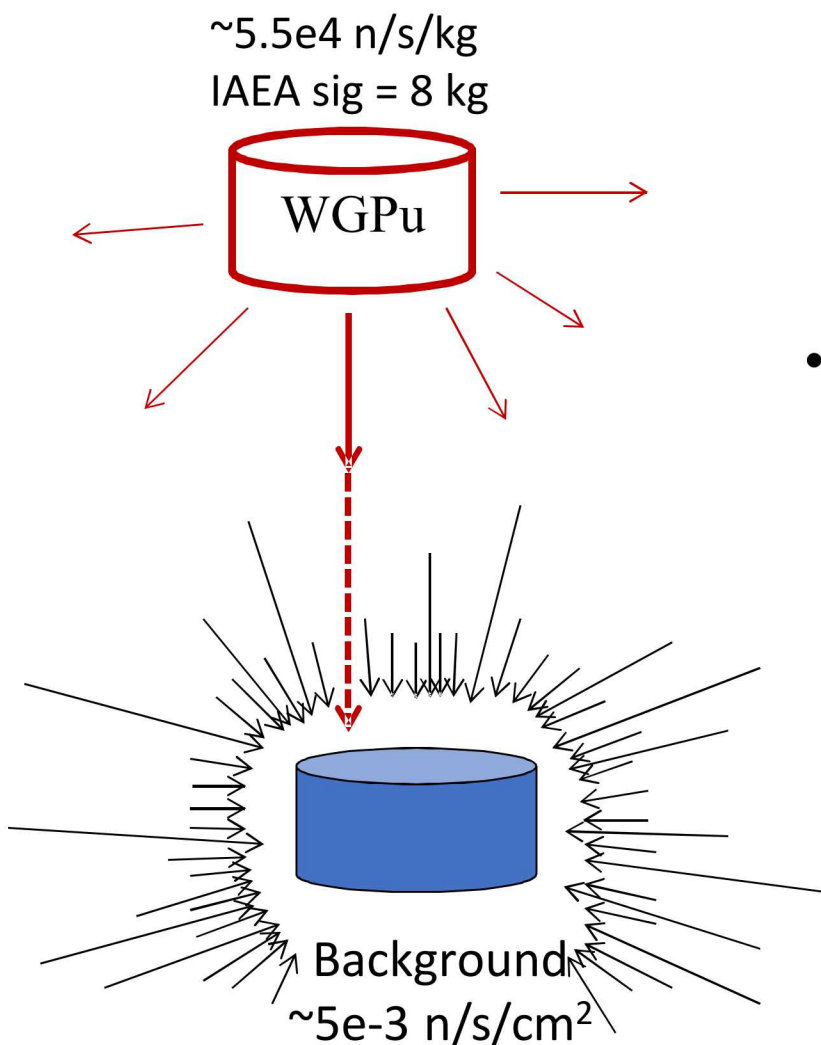


# TEI-1D

Many SNL contributors:

Peter Marleau  
Melinda Sweany  
Patricia Schuster  
Aaron Nowack  
Mateusz Monterial  
Jim Brennan  
Dan Throckmorton  
Etc.

# Standoff detection



- Example: Large stand-off application (100 m)

- 8 kg WGPu = ~4.4e5 n/s →  
 $4.4e5 \cdot \exp(-R/100)/4\pi R^2 \approx 1.3 \text{ n/s/m}^2$
- Background = ~50 n/s/m<sup>2</sup> (at sea level)
- 100% efficient, 1 m<sup>2</sup> detector →  
 5σ detection in **~13 minutes**
- 10% efficient, 1 m<sup>2</sup> detector →  
 5σ detection in **~2 hours**
- 10% efficient, 1 m<sup>2</sup> detector,  
*3% bg rate systematic* →  
 5σ detection in **never**



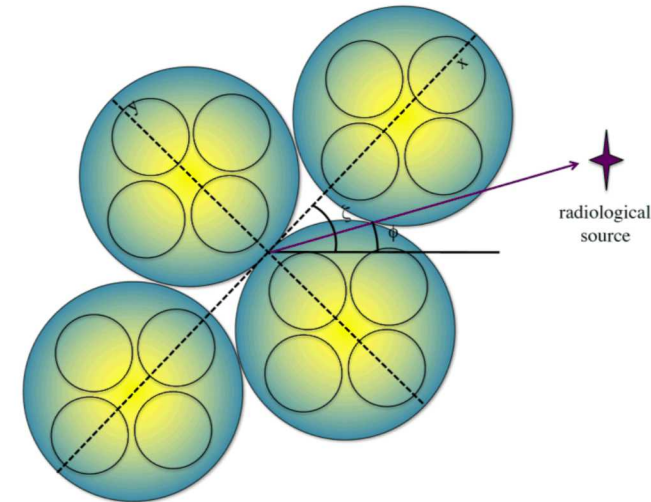
Directional information, however, allows to simultaneously measure signal and background, change **never** to **< never**.

# TEI-1D demonstration system



- Four large cells: 12"D x 15" LS
- Read out with 4x 5" PMTs each
- Configured in diamond pattern
- Placed on rotating turntable
- Each cell is analyzed independently

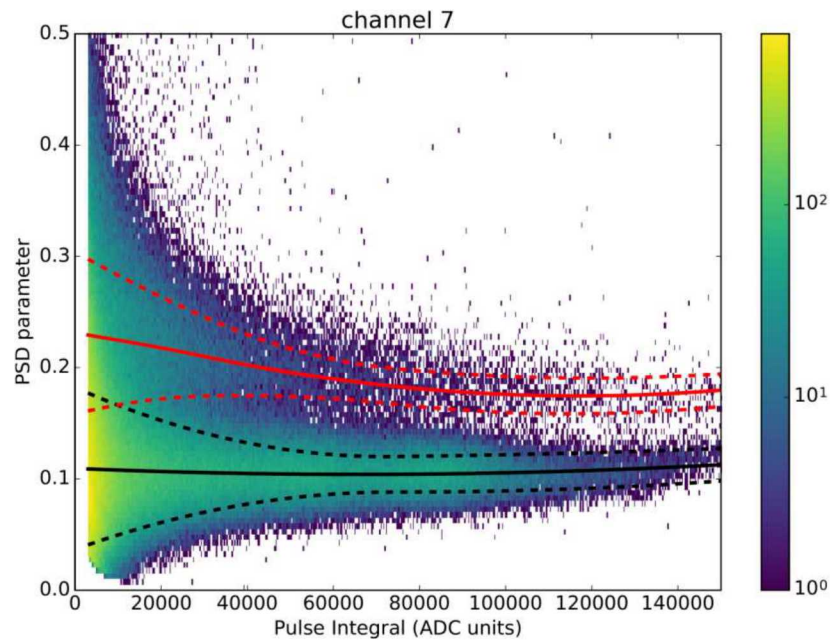
- Detector effective area is key.
  - 27 L of scintillator x 4 cells
- Use self-modulation to add some directional information—discriminate localized threat source from diffuse background



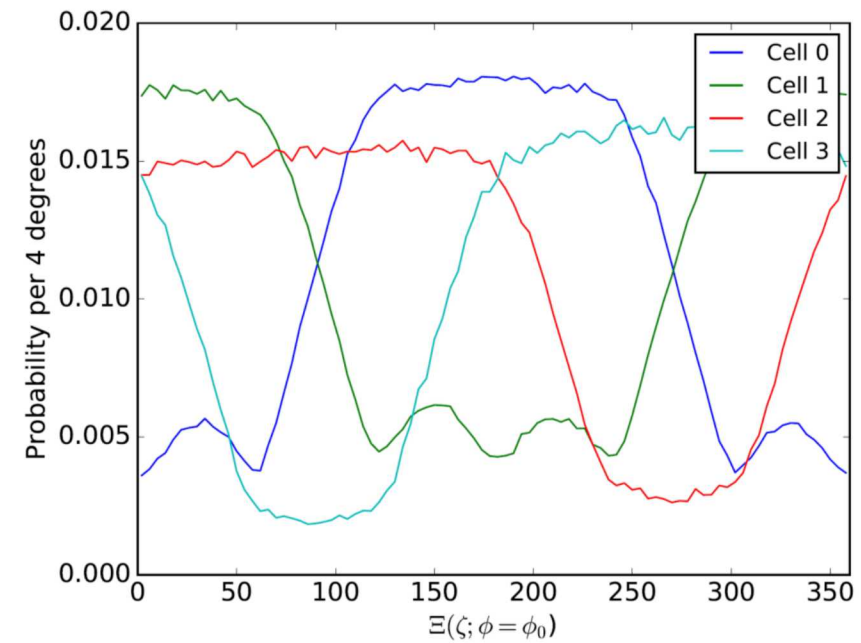


# Detector calibration

Pulse shape discrimination:  
biggest challenge in large-  
volume cells



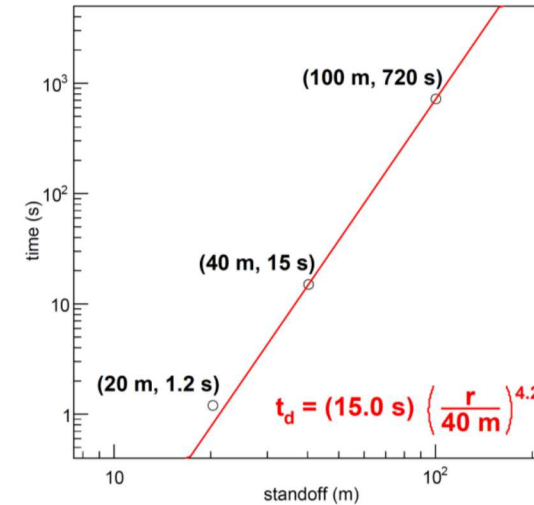
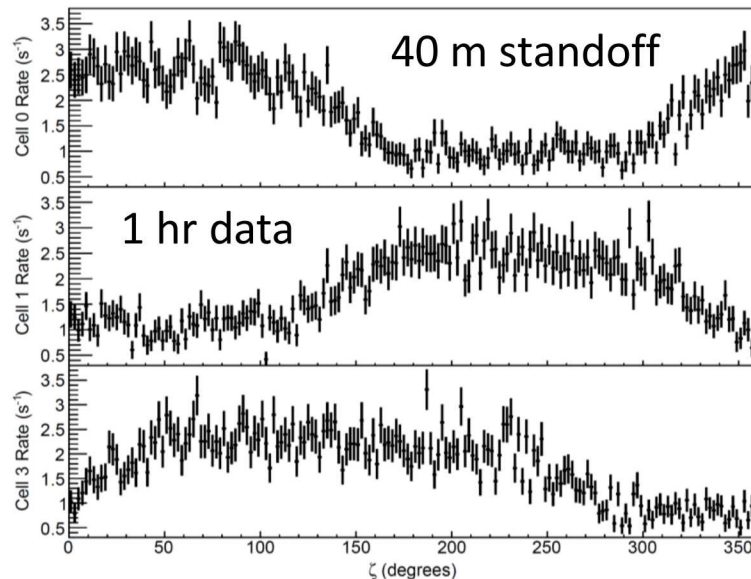
Modulation function:  
broad features, goal is  
detection, not resolution



# Standoff detection results



- ~1 mCi Cf-252 source at 20 m, 40 m, 100 m standoff
- Only 3 working detectors
- Detection (90% TP, 10% FP) at 100 m in 720 s



NIM A, doi:10.1016/j.nima.2017.09.052

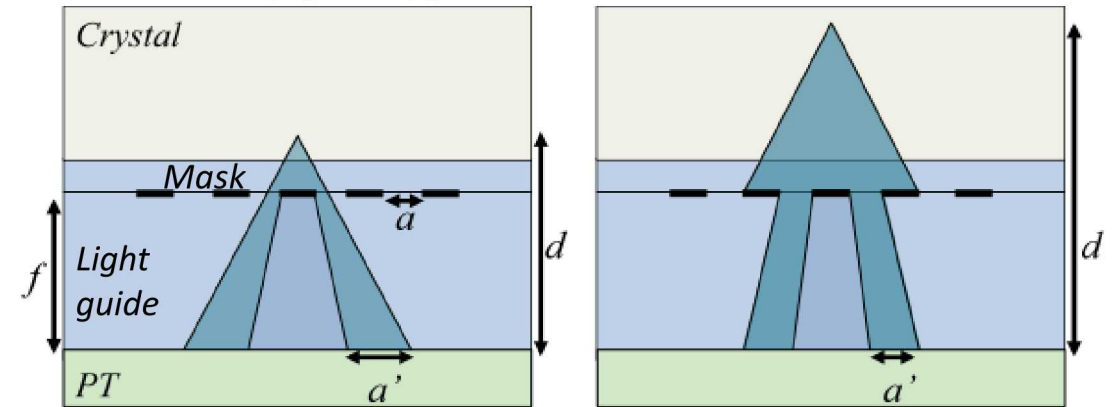


SVSC/OCA



# Interaction resolution via optical coded aperture

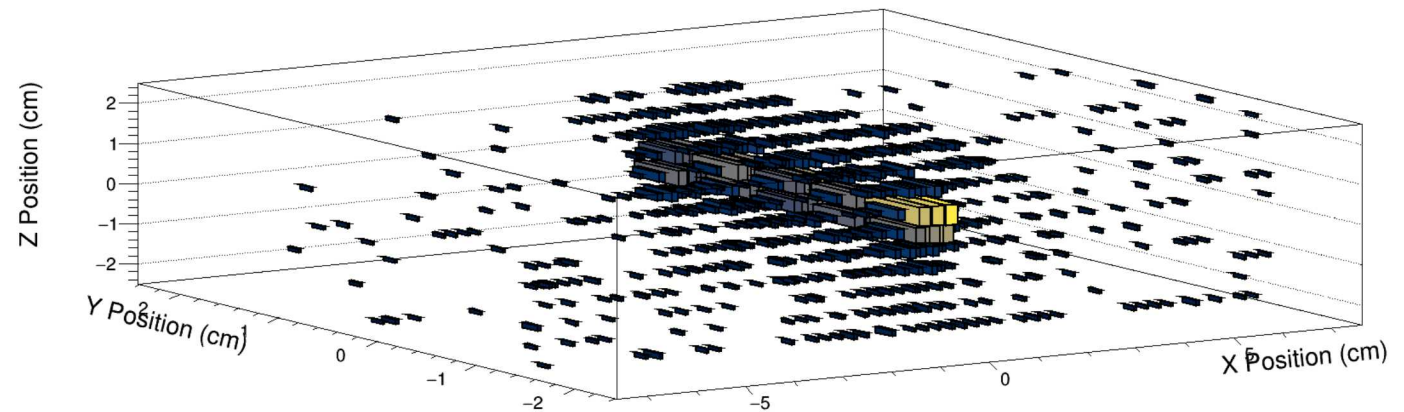
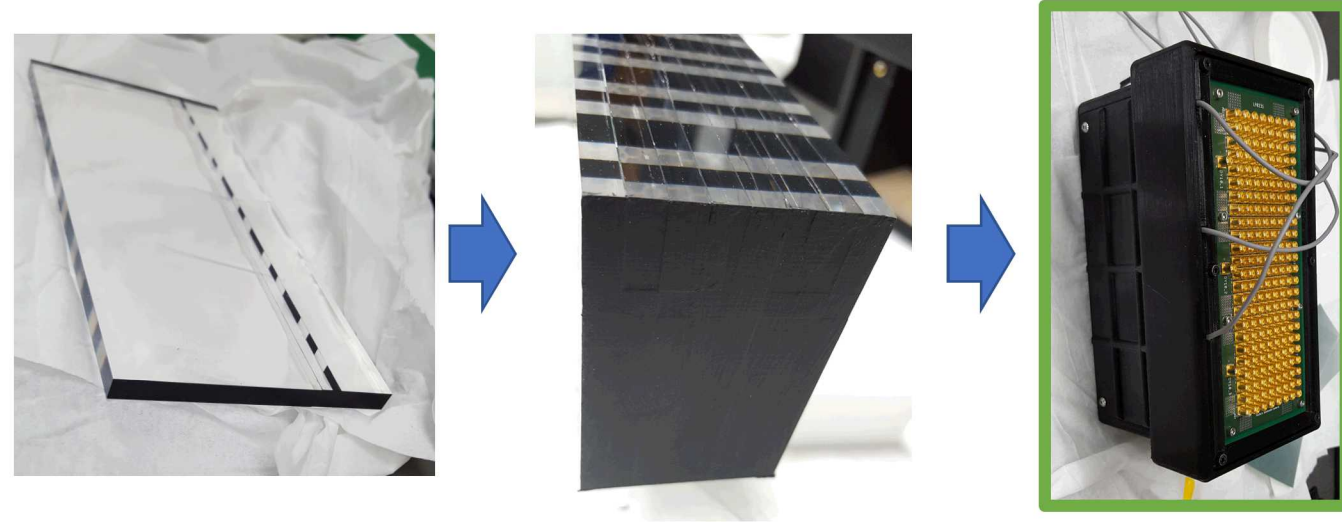
- So far, coded aperture was for the energetic radiation ( $n, \gamma$ )
- But can use optical coded aperture (OCA) for scintillation light itself -> resolve interaction position in a bulk scintillator
- Supports various imaging modalities



Klaus Ziock / Micah Folsom (ORNL/UTK)

# Neutron scatter camera using OCA

- Micah Folsom (UTK/ORNL) built a detector comprising a stack of organic scintillator slabs, each with a 1-D OCA
- Position resolution in x via coded pattern shifts, y via slab ID, z via coded pattern magnification
- Calibrated position results using gamma beam from collimated Ba-133 source



# Conclusions

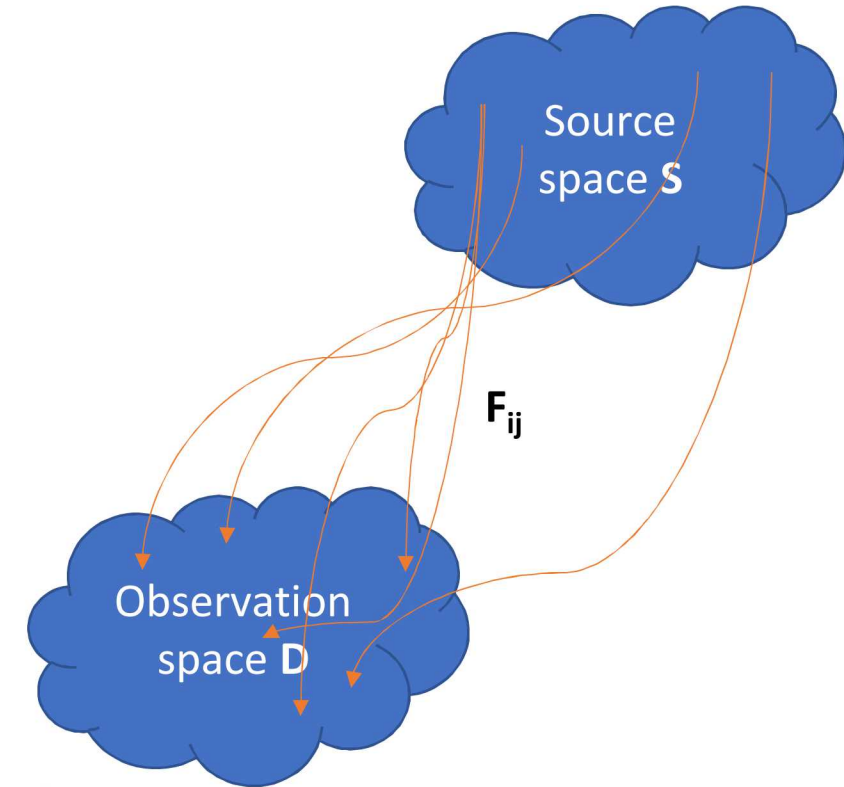
- System parameters (esp. physical dimensions) drive tradeoffs:
  - Sensitivity vs resolution
  - Field of view vs zoom
  - Capability vs complexity
  - Etc.
- Wide range of aperture-based radiation imaging systems:
  - High-resolution coded aperture imaging
  - Single-detector time-encoded imaging
  - Large, high-sensitivity time-encoded imaging
- Together with double-scatter imaging (Will's presentation) and transmission imaging (x-ray, neutron radiography), elements of toolkit for imaging with energetic radiation



# Other topics

# General image reco problem

- Consider a scenario defined by:
  - Consider binned spaces for simplicity
  - A source space  $\mathbf{S}$  ( $s_j$ )
    - e.g. 1 m x 1 m plane at 3 m from the detector
    - e.g. spherical coordinates (theta,phi) in the far field
  - An observation space  $\mathbf{D}$  ( $d_i$ )
    - e.g. detector pixel ID in a coded aperture system
    - e.g.  $(d\mathbf{x}, dt, E_1)$  for a SVSC ( 5 dimensional space!)
  - A set  $\mathbf{F}$  ( $F_{ij}$ ) of probability density functions connecting them.
    - $\text{Sum}_i(F_{ij}) = 1$  (or  $= \varepsilon_j$ ): For each source location  $j$ , the probability to observe the event in the various detector bins is a p.d.f. (normalized or unnormalized)
- Then for a given source distribution  $\mathbf{s}$ , we expect to observe  $\mathbf{d} = \mathbf{F}\mathbf{s}$
- How to solve the inverse? Given  $\mathbf{d}$ , what is  $\mathbf{s}$ ?



# Iterative methods

- Then for a given source distribution  $\mathbf{s}$ , we expect to observe  $\mathbf{d}=\mathbf{F}\mathbf{s}$
- How to solve the inverse? Given  $\mathbf{d}$ , what is  $\mathbf{s}$ ?
- Simple!  $\mathbf{s}=\mathbf{F}^{-1}\mathbf{d}$ , right?
  - Oops,  $\mathbf{F}^{-1}$  may be difficult to compute
  - Even more oops, in many cases  $\mathbf{F}$  is only approximate, and inverse is ill-conditioned, so  $\mathbf{F}^{-1}$  is not reliable
- Need a solution that takes into account response  $\mathbf{F}$ , and also measurement noise—in our case mostly Poisson stats
  - $\mathbf{d}=\mathbf{F}\mathbf{s}$  is just the expected value; really  $d_i \sim \text{Pois}((\mathbf{F}\mathbf{s})_i)$
  - Likelihood  $L = \prod_i (\text{Poisson}(d_i, (\mathbf{F}\mathbf{s})_i))$
  - Just need to maximize  $L$  over  $\mathbf{s}$ :  $\hat{\mathbf{s}} = (\arg \max)_{\mathbf{s}} L$
  - Pushes us toward iterative methods



# Maximum likelihood expectation maximization (MLEM)

- ML vs MLEM
- Shepp and Vardi (1982):

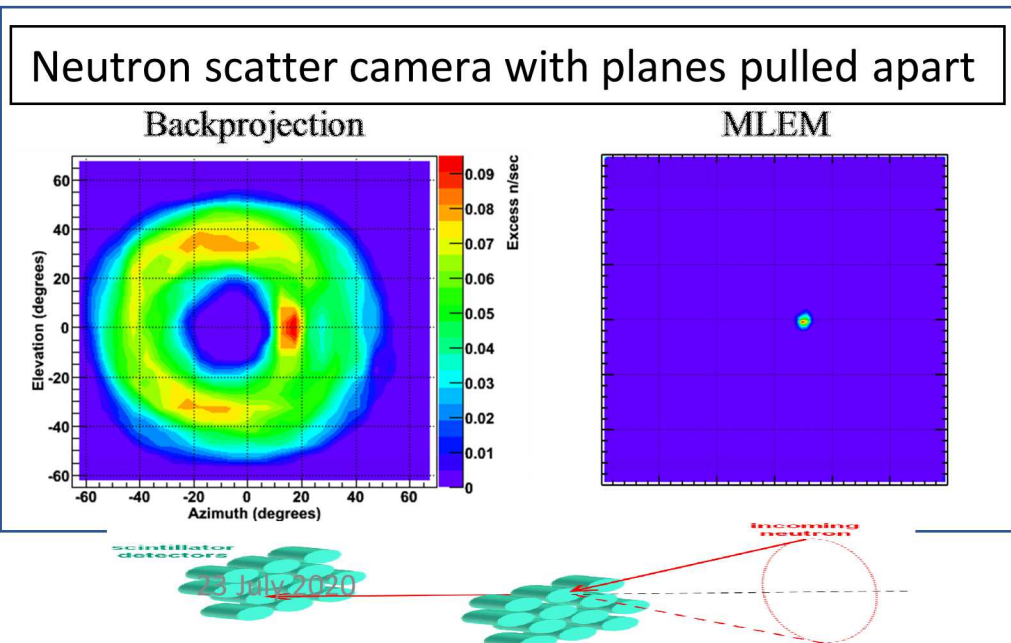
the likelihood function is given by

$$L(\lambda) = P(\mathbf{n}^* | \lambda) = \sum_A \prod_{\substack{b=1, \dots, B \\ d=1, \dots, D}} e^{-\lambda(b,d)} \frac{\lambda(b,d)^{n(b,d)}}{n(b,d)!}$$

Tedious, extremely obvious math that I won't bore you with

Under fairly general conditions, this iterative update equation converges to mathematically (!) optimal solution:

$$\hat{\lambda}^{\text{new}}(b) = \hat{\lambda}^{\text{old}}(b) \sum_{d=1}^D \frac{n^*(d)p(b,d)}{\sum_{b'=1}^B \hat{\lambda}^{\text{old}}(b')p(b',d)}, \quad b = 1, \dots, B.$$



# System response function

- All more complex image reconstruction methods (iterative techniques) require a system response function to be defined.
- First question, what are the source and observation spaces  $\mathbf{s}, \mathbf{d}$ ?
- How to estimate the response  $\mathbf{F}$ ?
  - Data (best fidelity, tedious, depends on environment)
  - Monte Carlo (lots of CPU, large uncertainties in low-probability regions)
  - Pseudo-MC (equal uncertainties across regions, not necessarily faster)
- This is the hardest part! Once you have  $\mathbf{F}$ , there are lots of options for the reconstruction method itself!

# More MLEM

- Standard MLEM issue: what is the stopping condition?
  - More iterations converge to mathematically optimal answer, but that may not be real-world optimal!
  - Can stop after a certain number of iterations, but no generally accepted way to choose that
- Can add regularization terms to the likelihood.
  - E.g. a term that penalizes high-frequency changes
  - Or, edge-preserving prior
  - But, for generic terms added to the likelihood, Shepp & Vardi iterative update equation no longer applies; need more complicated minimization methods e.g. surrogate functions
  - Alenius, Sakari, and Ulla Ruotsalainen. "Generalization of median root prior reconstruction." *IEEE Transactions on Medical Imaging* 21.11 (2002): 1413-1420.
- What is list-mode likelihood?

# Stochastic Origin Ensembles (SOE)

- Based on Markov Chain MC methods (Metropolis-Hastings)
- Source density  $s$  as the thing to be estimated
- Each event came from one particular location on its cone
- Use the density itself to “walk through” the posterior distribution of the density
  - First set of iterations converges on the right neighborhood
  - Further iterations map out posterior distribution

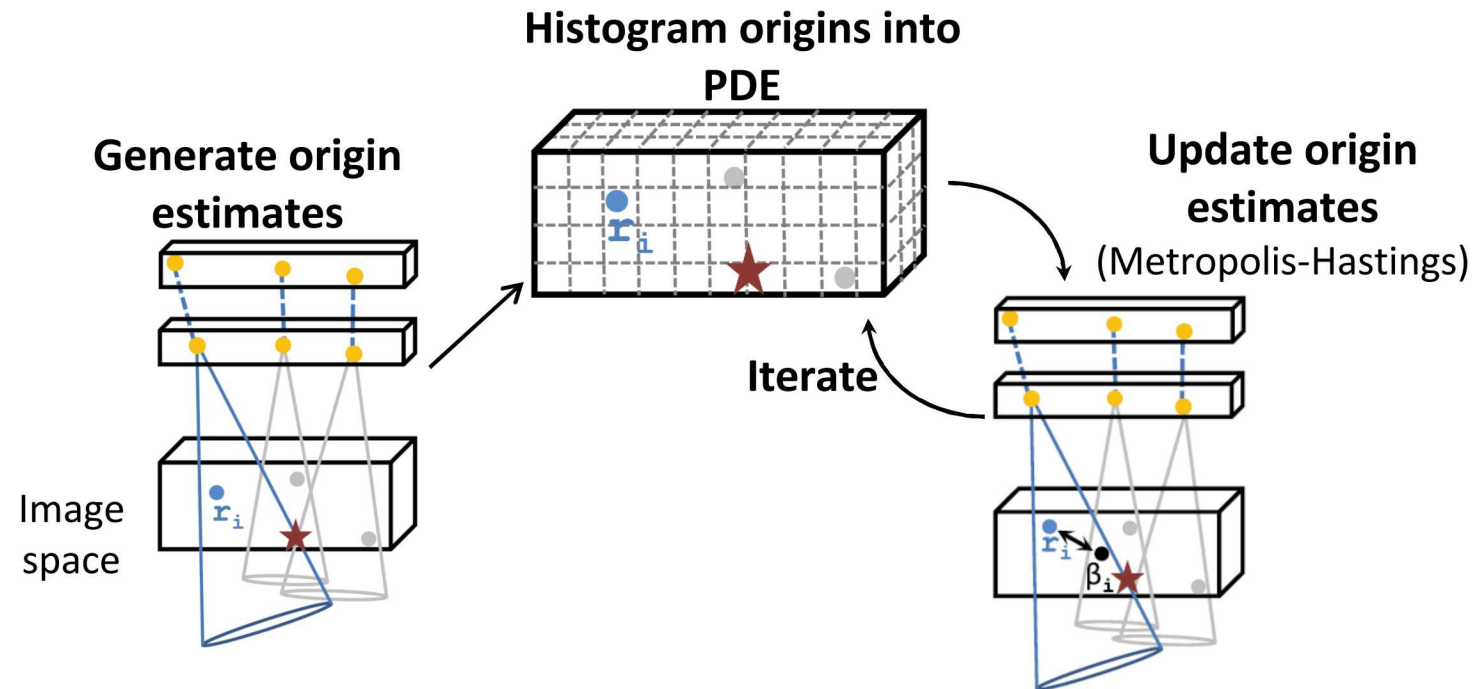


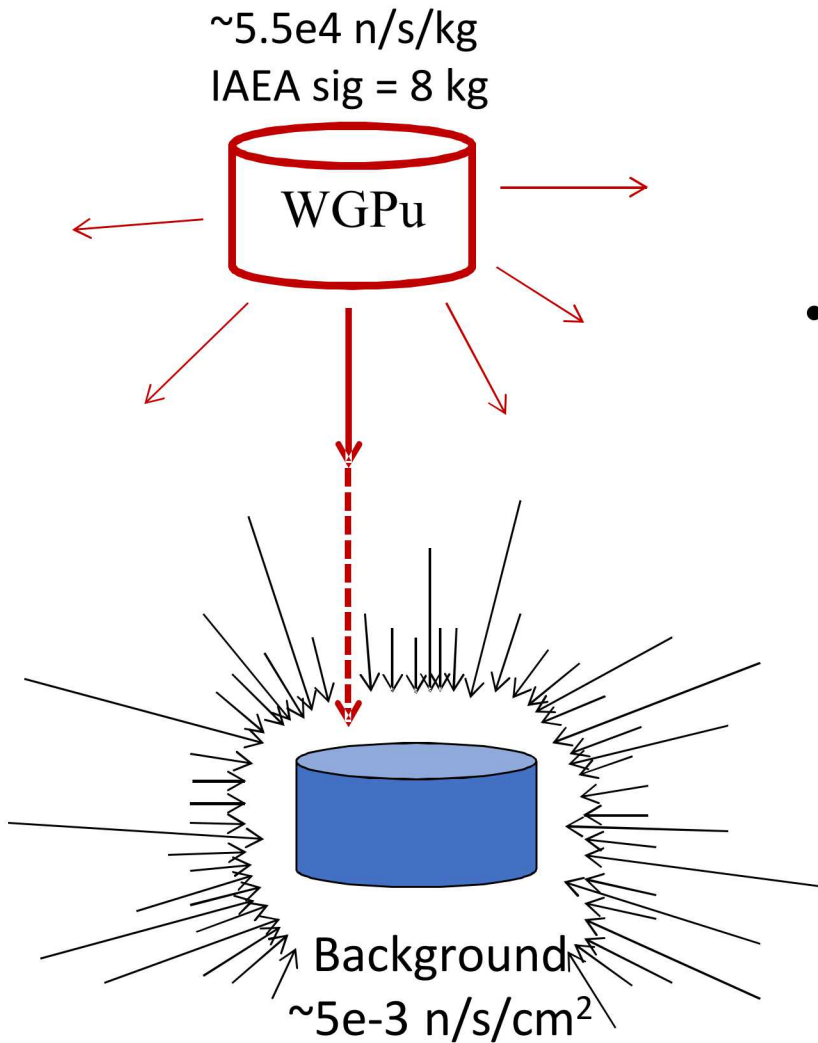
Image: D. Goodman  
[1] D. Mackin et al. 2012



# Additional image reconstruction methods

- SOE: stochastic origin ensemble
- MLE: maximum likelihood estimation (for point source reconstruction)
- Sparsity condition (compressive sensing)
- Machine learning—e.g. (de)convolutional neural network
  - Only one that may not require high-quality system response function

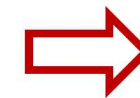
# Standoff detection



Case: background unknown

- Example: Large stand-off application (100 m)

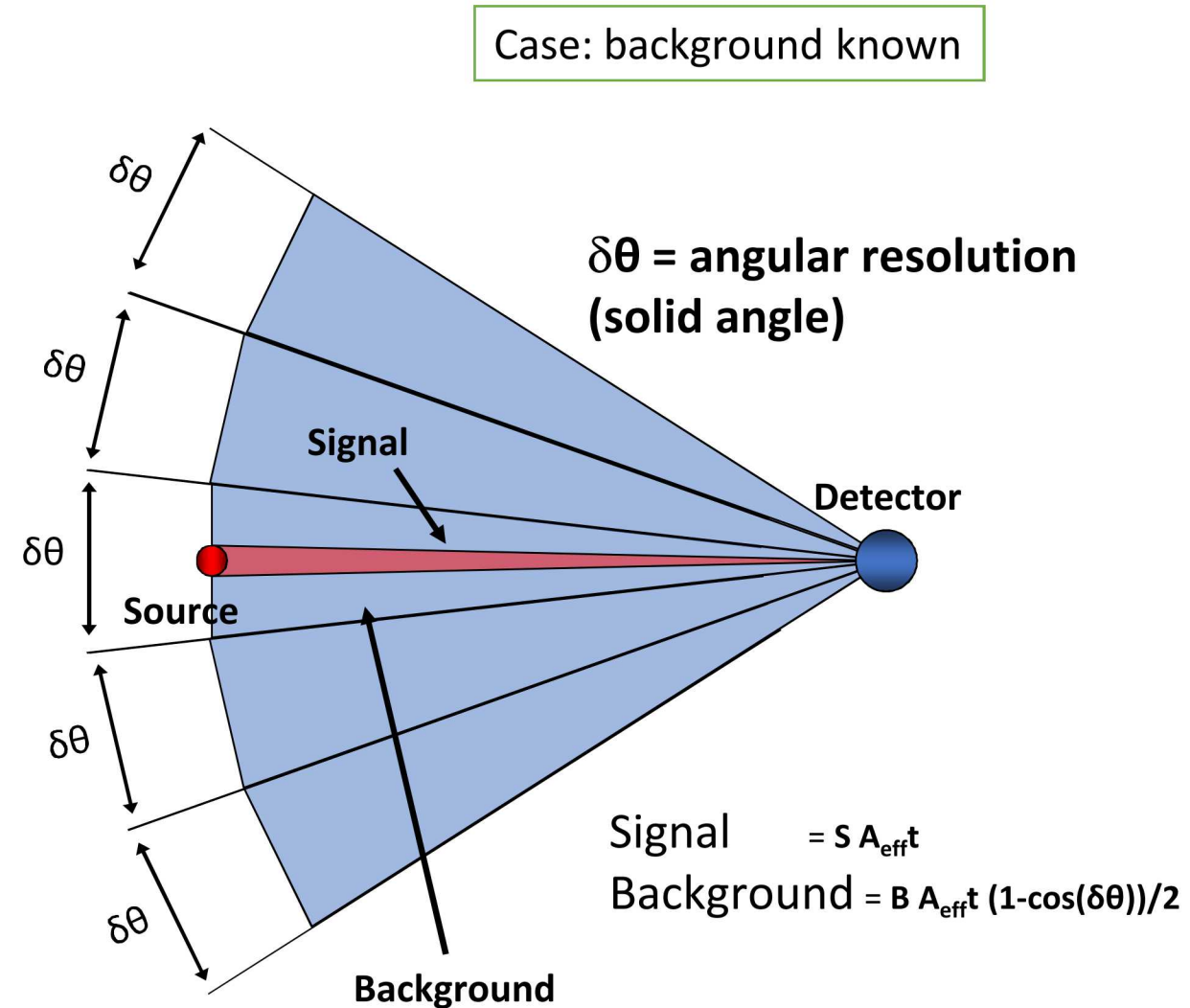
- 8 kg WGPu =  $\sim 4.4e5 \text{ n/s}$  →  
 $4.4e5 * \exp(-R/100) / 4\pi R^2 \approx \mathbf{1.3 \text{ n/s/m}^2}$
- Background =  $\sim 50 \text{ n/s/m}^2$  (at sea level)
- 100% efficient, 1 m<sup>2</sup> detector →  
5 $\sigma$  detection in  **$\sim 13$  minutes**
- 10% efficient, 1 m<sup>2</sup> detector →  
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- 10% efficient, 1 m<sup>2</sup> detector,  
*3% bg rate systematic* →  
5 $\sigma$  detection in **never**



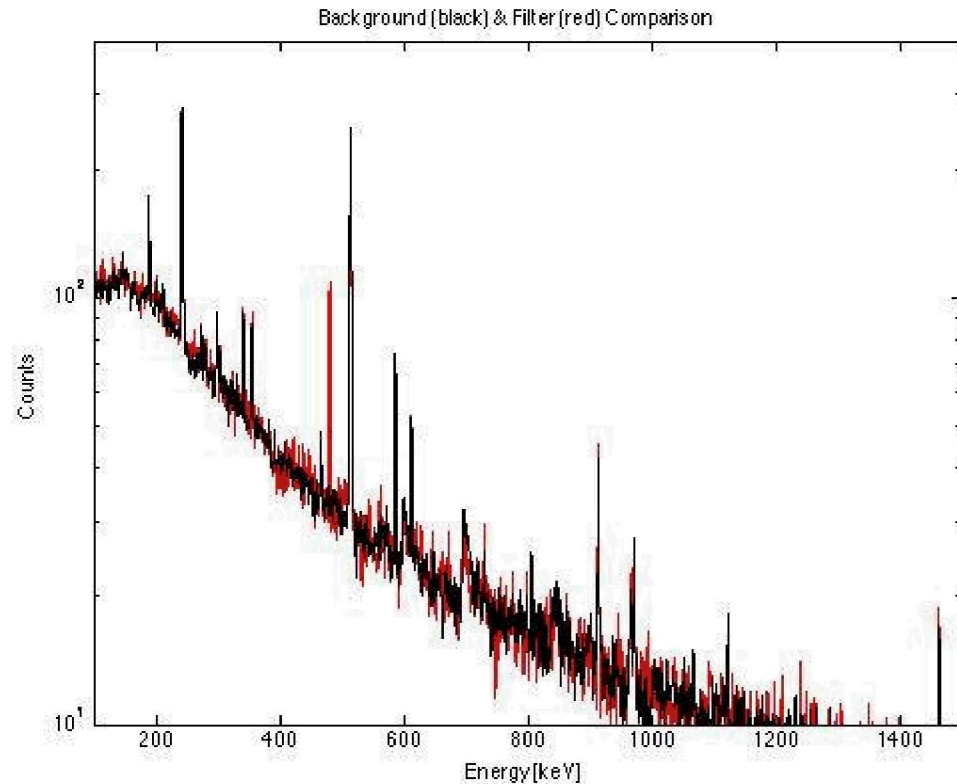
Directional information, however, allows to simultaneously measure signal and background, change **never** to **< never**.

# Detection again

- What about when the background is independently known?
  - Example: portal monitor. Effectively have repeated background measurements in between occupancies.
  - Example: building monitoring. Looking for changes in the rad field due to an approaching source.
- Now is there an advantage from imaging?
  - Cartoon at right due to Peter Marleau.
  - But real imagers do not have simple angular resolution like cartoon.
  - Also generally take a hit on efficiency.



# Gamma spectral analogy



- Analogy to information from gamma spectrum.
  - Estimate background (systematic).
  - Ignore most background (statistical).



# The neutron imager zoo

