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# A Front-End Framework with Embedded ML Tools for Automating Neutron Scattering Experiments

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Boulder, Colorado USA | radiasoft.net



## Outline

- Motivation & background
  - Problem motivation
  - Partner beamlines
  - Image segmentation tasks
- Automation methods
  - Controls software
  - Image recognition
- Results & discussion



# **Background & motivation**

The value of automation on neutron scattering beamlines



## **Background & Motivation**

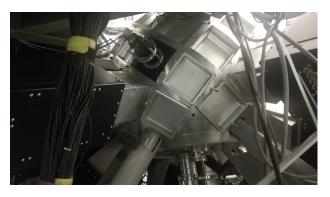
- Sample alignment is tedious, but critical
  - Currently requires human image processing
  - Limited neutron production time, schedule constraints
- Machine learning is a key automation tool
  - For computer vision, convolutional neural networks (CNNs)
  - For controls automated, reinforcement learning (RL)
- Alignment protocols are distinct to a beamline
  - Framework must be highly general & robust
  - Opportunity to deploy transfer learning



## Partner Beamlines: TOPAZ

- Fed by the Spallation Neutron Source (SNS)
- Sample in a chamber
  - Sample arm with 3 translational, 2 rotational axes
  - Neutron detectors, cameras, & environmental controls
- Highly developed controls system (EPICS)
  - Plenty of control channels (PVs already exist)
  - Pre-existing IOC software
    - Includes point-and-click sample alignment
    - Must consider interactions with our framework









## Partner Beamlines: HB-2A

- Fed by the High-Flux Isotope Reactor (HFIR)
- Samples in containers (cans) on a stage
  - 3 translational axes, stage & sample rotation axes
  - Partially encircled by a neutron detector
  - Diagnostics done with a **neutron** camera
- Controls executed with SPICE (no EPICS)
  - Required a "bridge" between software to use EPICS
  - Sample alignment completely manual



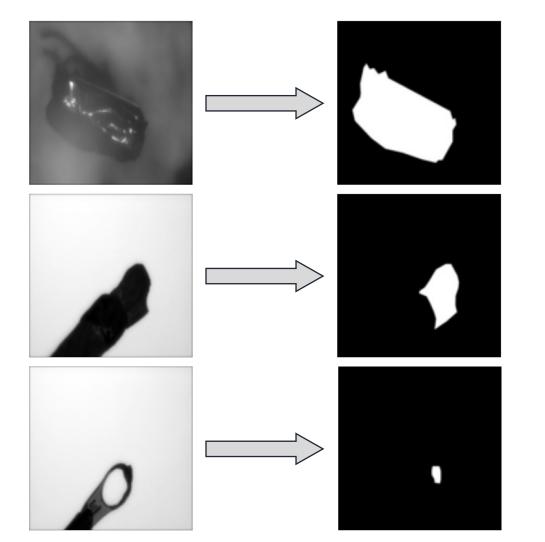




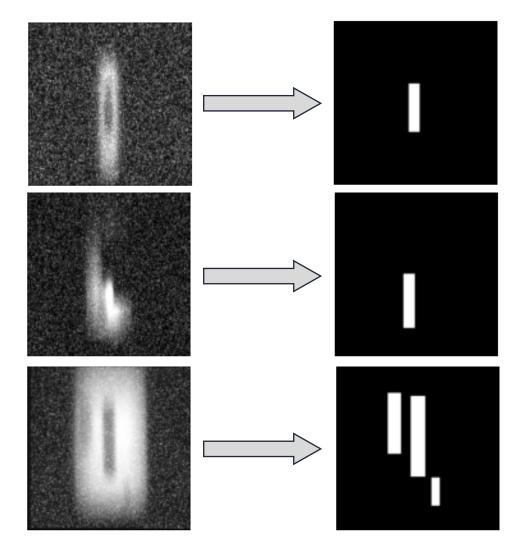


## Image Segmentation Tasks

TOPAZ



HB2A





# **Automation methods**

Approaches for removing humans from the loop

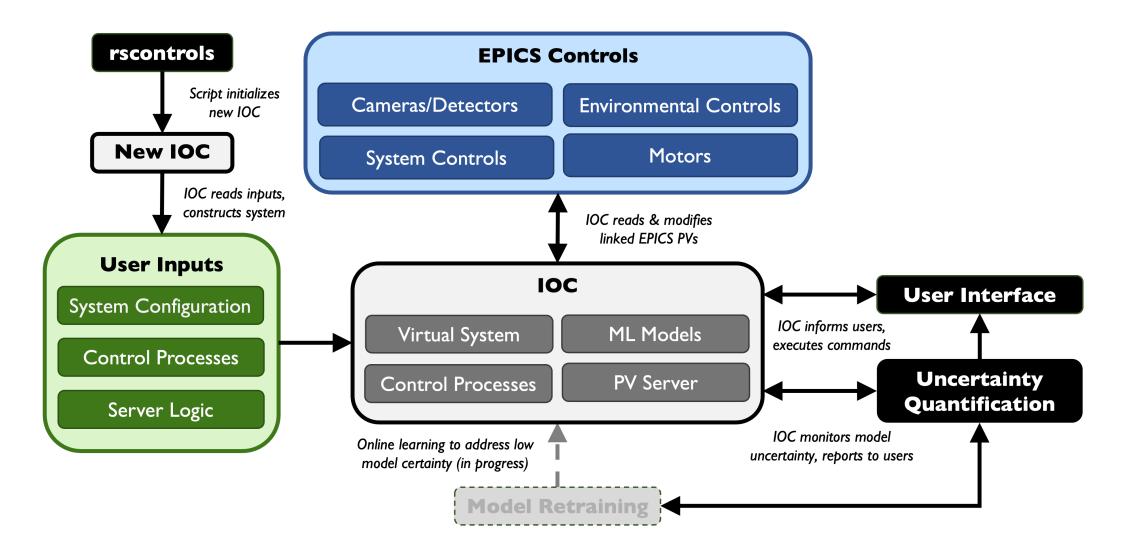


## **Controls Software**

- Require a means to deploy developed automation tools
  - Machine learning models, control loops (e.g., PID), etc.
  - Must be flexible to apply broadly to controls environments
- Development & testing resulted in rscontrols
  - Written in Python using existing EPICS tools (PyEpics, PCASPy, & P4P)
  - Hardware (detectors, motors, etc.) represented in thin virtualization layer
    - Connected via EPICS process variables (PVs)
  - Handles for user-defined controls processes & served PVs
    - Written & imported as Python modules, on the fly
  - Dedicated embeddings for ML models



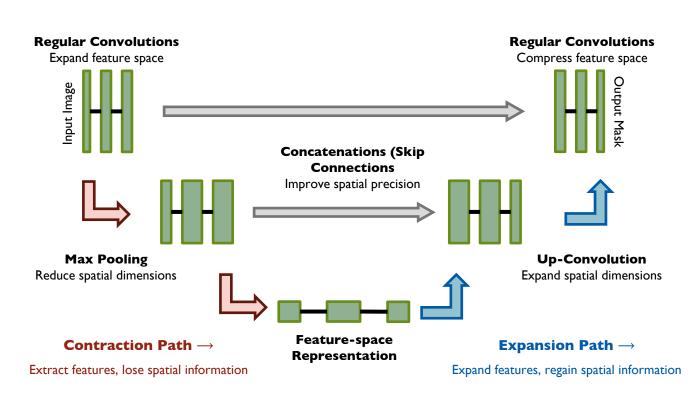
## **rscontrols** Workflow





# **Image Recognition**

- Classic use-case for convolutional neural networks (CNNs)
  - UNet specifically designed for image & semantic segmentation
  - Labor-intensive supervised learning regimen
- Must be robust to use cases
  - No sample (TOPAZ & HB-2A)
  - Multiple samples (HB-2A)





# **Results**

## rscontrols UI, model uncertainties, and sample alignment



## rscontrols, Simulated Beamline Example

## IOC : protocol : CA modes : [primary, secondary]

System : name : Example Beamline prefix : EXBL

# PrimaryCam : type : Detector prefix : PCam modes : [primary] dimensions : [100, 100] data\_pv : Image

#-----(break)-----

Server :
 path : server\_functions.py
 prefix: EXBL
 pvs:

PCam:CleanIm :
 count: 10000
 get:
 function: denoise\_cam
 args:
 model: DenoiseUNet
 cam: PrimaryCam
 put: None

\_\_\_\_\_\_

Models :
 DenoiseUNet:
 type : UNet
 weights: model.h5
 architecture: parameters.pkl
#-----(break)------

#-----(break)------

Processes :
 path : exbl\_processes.py

# align\_sample : primary: function : auto\_align args : model: MaskUNet cam : PrimaryCam controls: PrimaryControls secondary: function : auto\_align args : model: MaskUNet cam : SecondaryCam controls: SecondaryControls

#### RSControls Command Line Interface

#### EPICS Server: OFF

1)	Run a control process
2)	List control processes

3) List control elements

0) Exit

Enter a # for one of the commands listed above: -> 3

#### Control elements for 'Example Beamline IOC'

Name	Туре	PVs
PrimaryCam	CADetector	Mask: Array, length: 10000, min: 0.000000, max: 1.000000 ImData: Array, length: 10000, min: 0.038875, max: 282.920188
PrimaryControls	CAElement	<pre>x: 0.0 y: 0.0 z: 0.0 theta: 0.0 Rotate90: 0 Rotate180: 0</pre>
LED	CAElement	power: 1

------

#### EPICS Server: OFF

Run a control process
 List control processes
 List control elements

#### 0) Exit

Enter a # for one of the commands listed above:

->



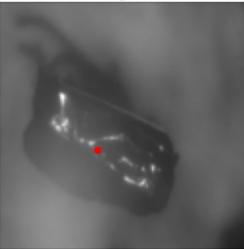
## rscontrols, CS Studio UI

		CS-Studio		
Example Beamline UI ×				
<b>A</b> radiasof	t			
Primary Mode Controls	Primary Motors		Primary Sample Camera	Ô
Secondary Mode Controls Beamline IOC	X Motor         RBV:       5.000         Input:       0.000         Tweak:       1.000         Input:       0.000         Set       Tweak:         Tweak:       1.000         Input:       Set         Tweak:       1.000         Input:       Input:         Set       Tweak:         Tweak:       Tweak:         Set       Tweak:         Set       Tweak:         Set       Tweak:         Set       Tweak:         Set       Tweak:         Set       Tweak:         Set	Y Motor         RBV:       -5.000         Input:       0.000       Set         Tweak:       1.000       -       +         Rotation Angle		
mhenderson				Update

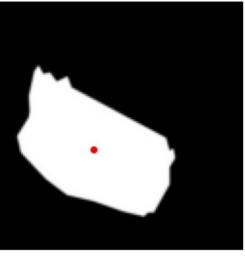


## **Unoptimized Ensemble Predictions**

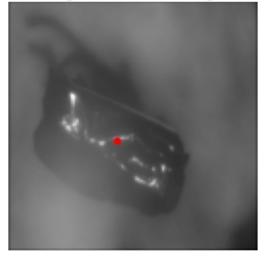
Test Image & CoM



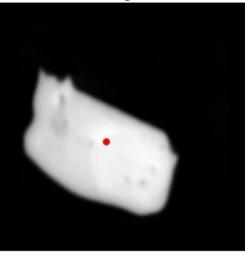
Test Mask & CoM



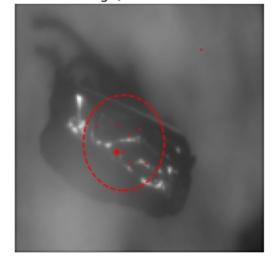
#### Test Image, Ensemble Average CoM



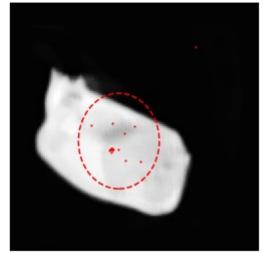
#### Ensemble Average Mask & CoM



Test Image, Ensemble CoMs



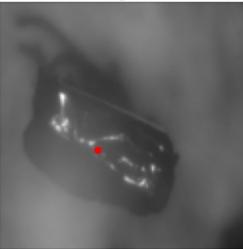
Ensemble Mask Variance & CoMs



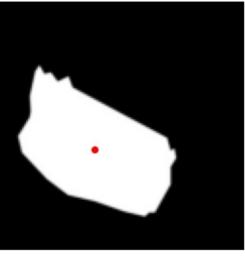


## **Optimized Ensemble Predictions**

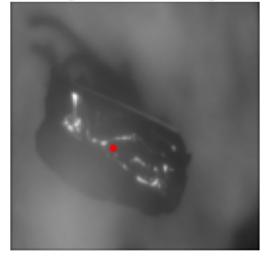
Test Image & CoM



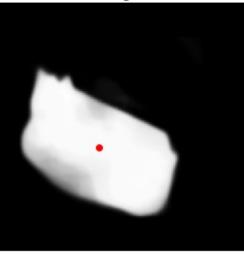
Test Mask & CoM



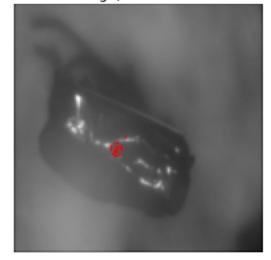
#### Test Image, Ensemble Average CoM



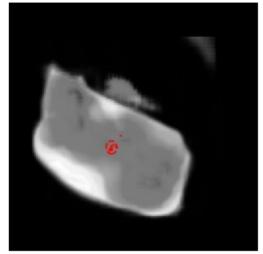
#### Ensemble Average Mask & CoM



Test Image, Ensemble CoMs



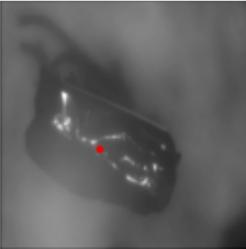
Ensemble Mask Variance & CoMs





## **Optimized & Transferred Ensemble Predictions**

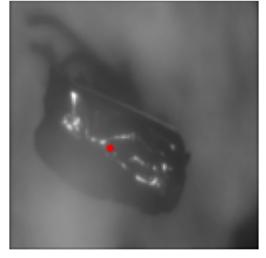
Test Image & CoM



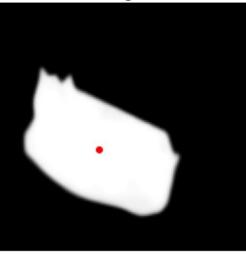
Test Mask & CoM



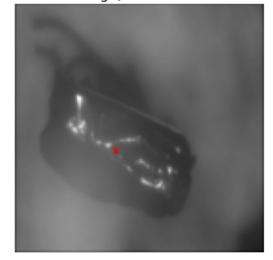
Test Image, Ensemble Average CoM



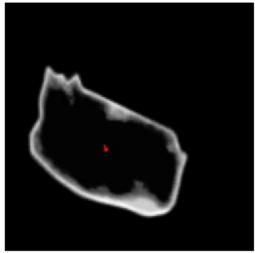
Ensemble Average Mask & CoM



Test Image, Ensemble CoMs

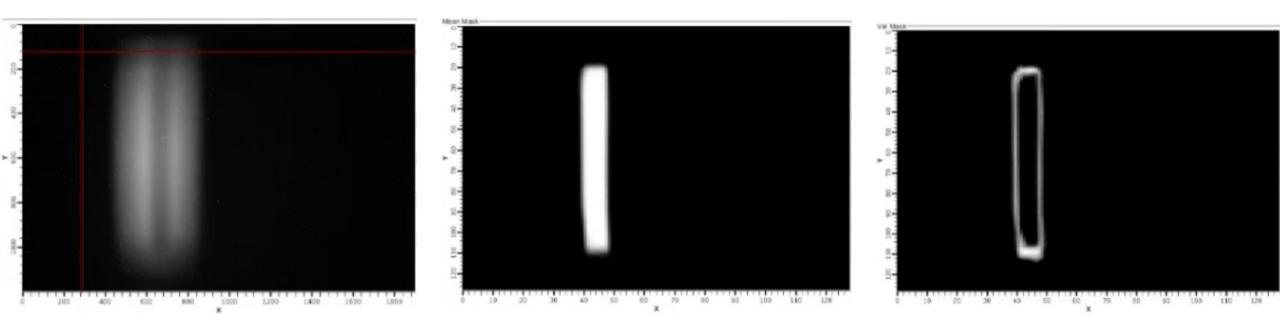


Ensemble Mask Variance & CoMs





## Live Uncertainty Reporting, HB-2A



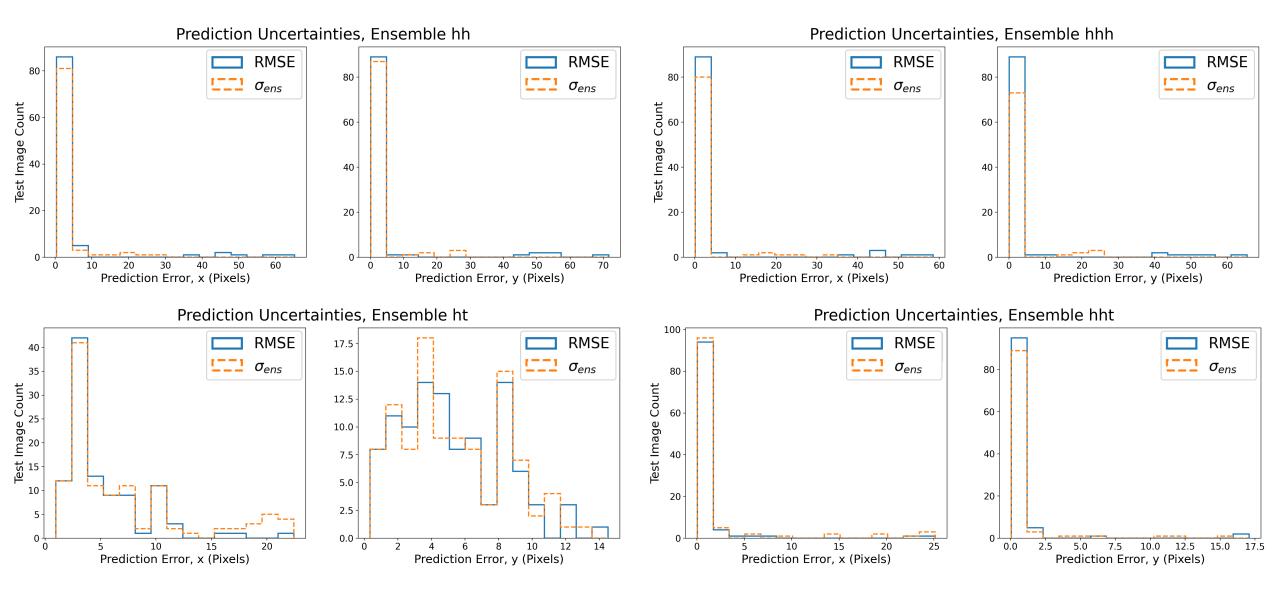
Live Sample Image

## Ensemble Mask Average

## Ensemble Mask Variance



## **UNet Prediction Uncertainties**





# Automated Alignment, TOPAZ

rocess (press any key)	+
Beamline Processes: lightswitch_cryo align_cryo temp_ramp	
Interface Actions: 0) Exit 1) Print beamline state 2) Print beamline element state 3) Time a beamline process	
Please choose a beamline processes or interface action:	





## Automated Alignment, HB-2A

-> 2

#### Control processes for 'MB2A IGC'

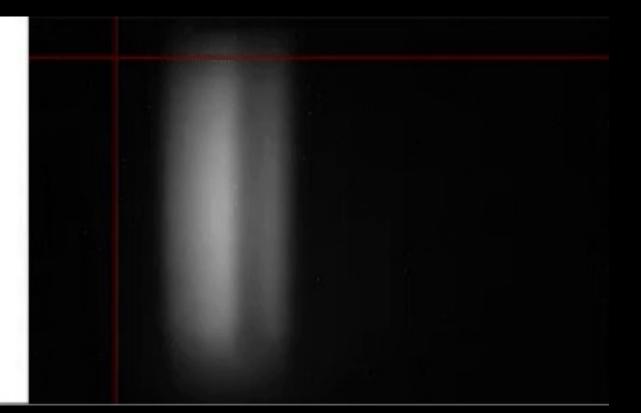
control biore		
Name		
**********	*********	**********
test_ncam test_stage align_sample slit: Slit stage: Stage	motor_test autoalign	stage: Stage
mask_unet: op		
***********		

EPICS Server: 0 Type: CA PV Prefix: <mark>HB2A</mark>

Run a control process
 List control processes
 List control elements
 Print EPICS server configuration

0) Exit

nter a # for one of the commands listed above:







# **Ongoing & Future Efforts**

- Further improve rscontrols UI
- Automate complete/more complex procedures
  - Temperature ramping with constant realignment at TOPAZ
  - Beam refinement at HB-2A
- Develop reinforcement learning solutions
  - Fuller automation at TOPAZ (less reliance on existing software)
  - Detector-based alignment at TOPAZ & HB-2A
- Synergize with RS 3D visualizer, also deployed at ORNL
- Deploy at more partner beamlines



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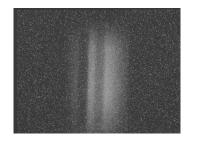


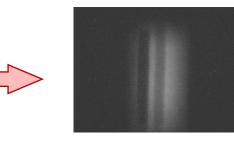
# **Unused Slides**



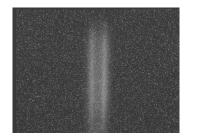
# **Denoising Results**

## Image denoising, regular filter

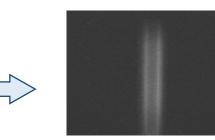




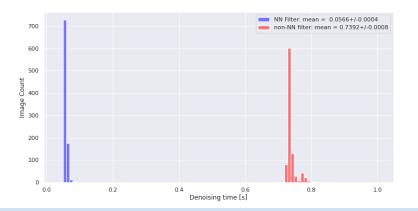
### Image denoising, CNN filter

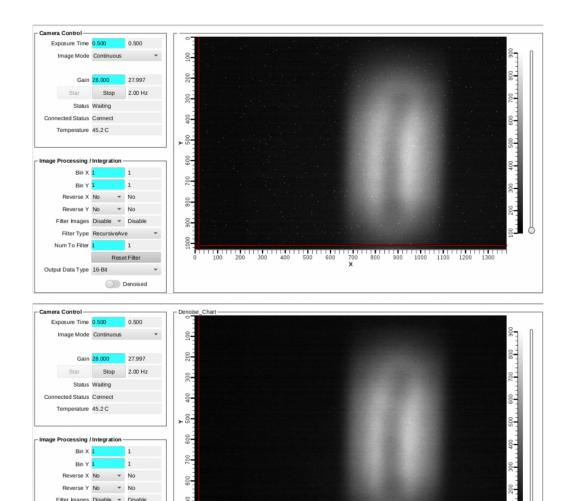


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## Execution time, regular vs. CNN filter





100 200 300 400 500

600 700

800 900 1000 1100 1200 1300

## ICFA ML 2024 | Gyeongju, South Korea

Filter Type

Output Data Type 16-Bit

Num To Filter

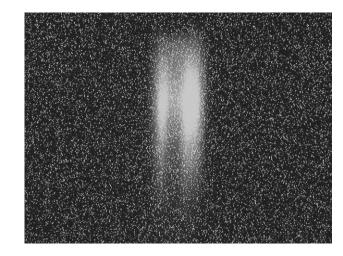
RecursiveAve

Reset Filter

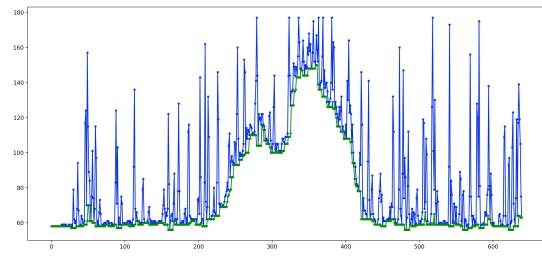
loc://boolTest(0

## **Neutron Camera Denoising**

- Neutron cameras are highly sensitive to noise
  - Characteristic "salt-and-pepper" speckle pattern
  - Signal condition degrades over time
- Image recognition models are sensitive to noise patterns
  - Trained on original (noisy)
- Multiple possible solutions
  - Simple analytic denoising algorithm
    - e.g. "inscribed envelope"
  - ML-based denoising solution
    - Potentially faster execution



Typical raw image from HB-2A



Horizontal pixel trace & inscribed envelope

