



XENON



清華大學

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# Accidental Coincidence Background in XENONnT

## for Low Energy Nuclear Recoil Searches

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Tsinghua University

On behalf of XENON collaboration



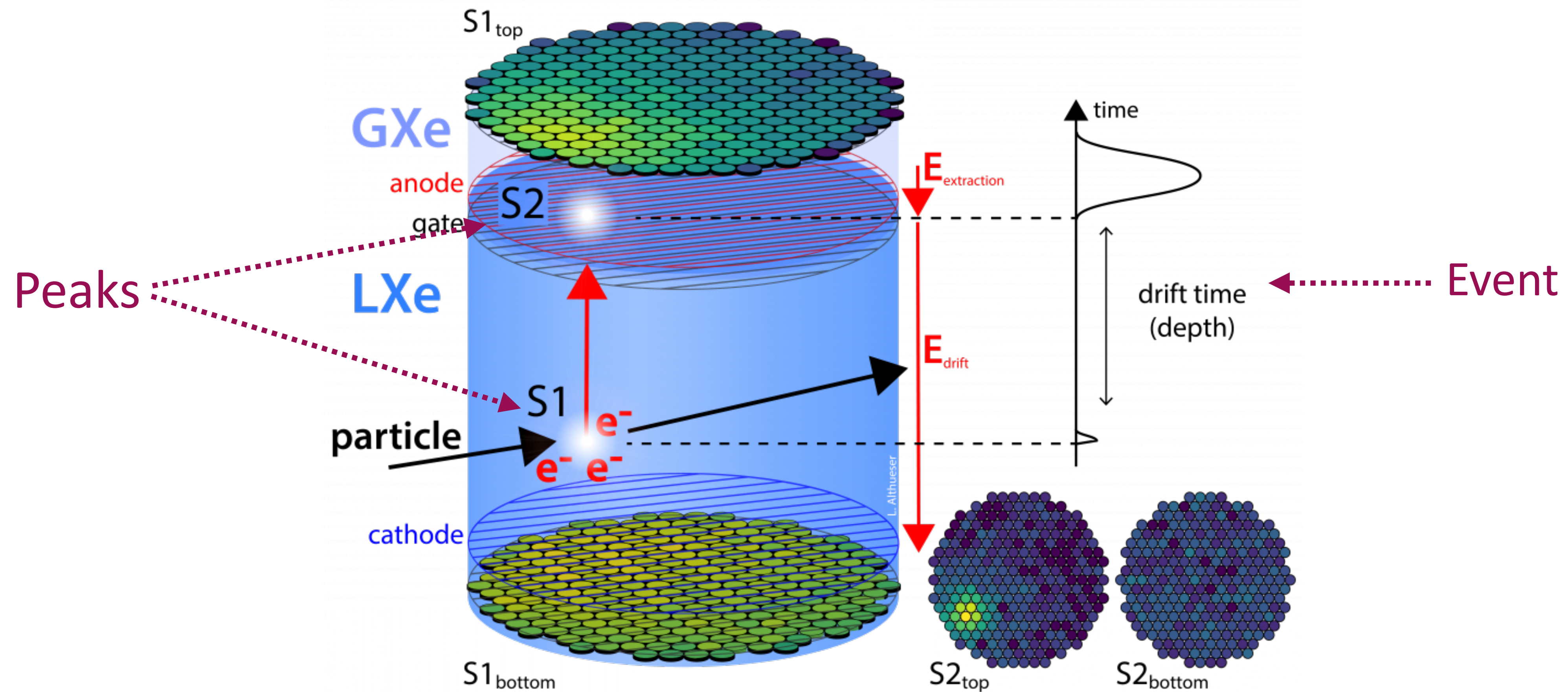
15th International Workshop on the Identification of Dark Matter  
July 8-12, 2024, L'Aquila



# XENONnT Detector and Event Building



## Two-Phase LXe TPC



# Accidental Coincidence



- In **XENON1T** B8/light WIMP Search  
Phys. Rev. Lett. 126, 091301 (2021)

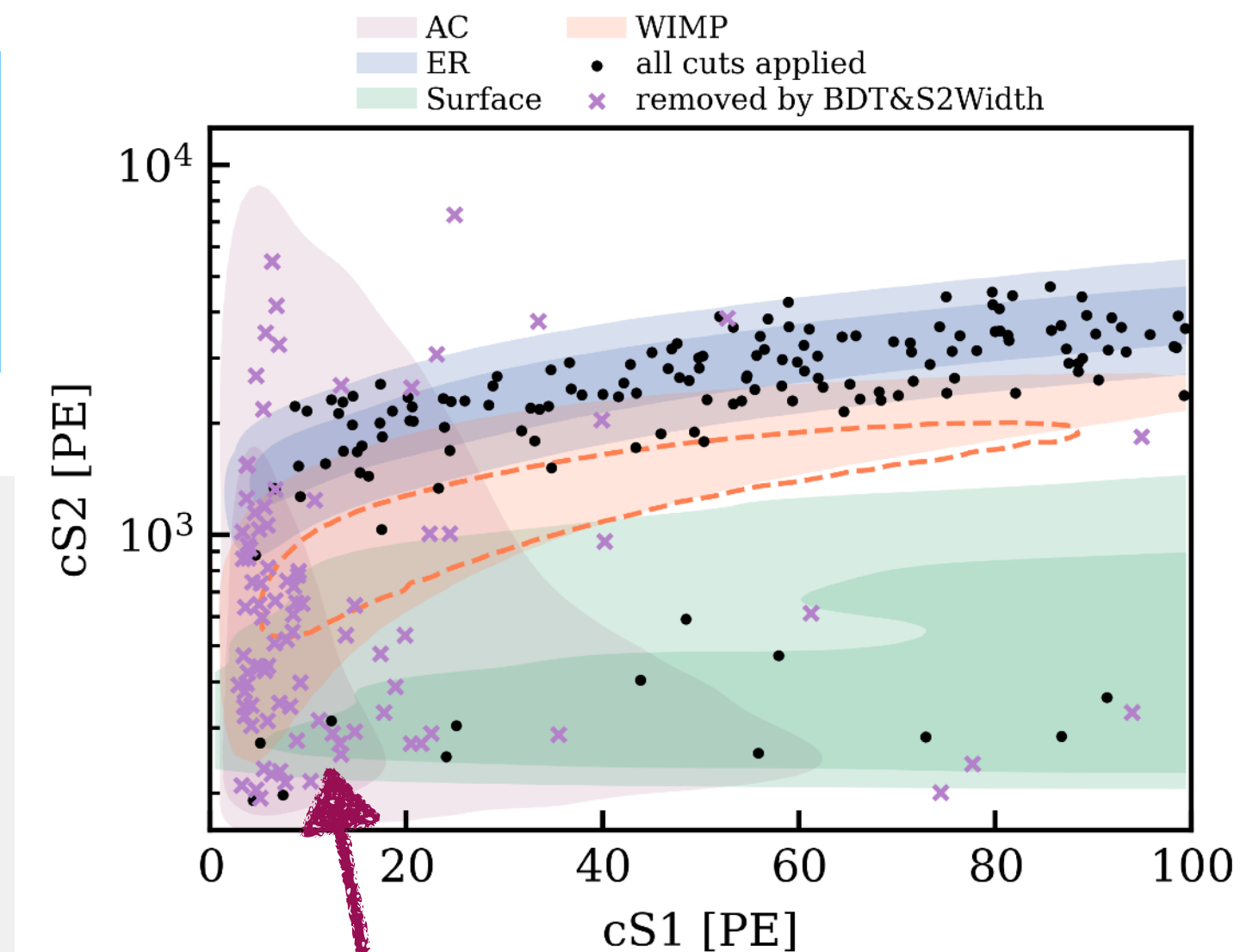
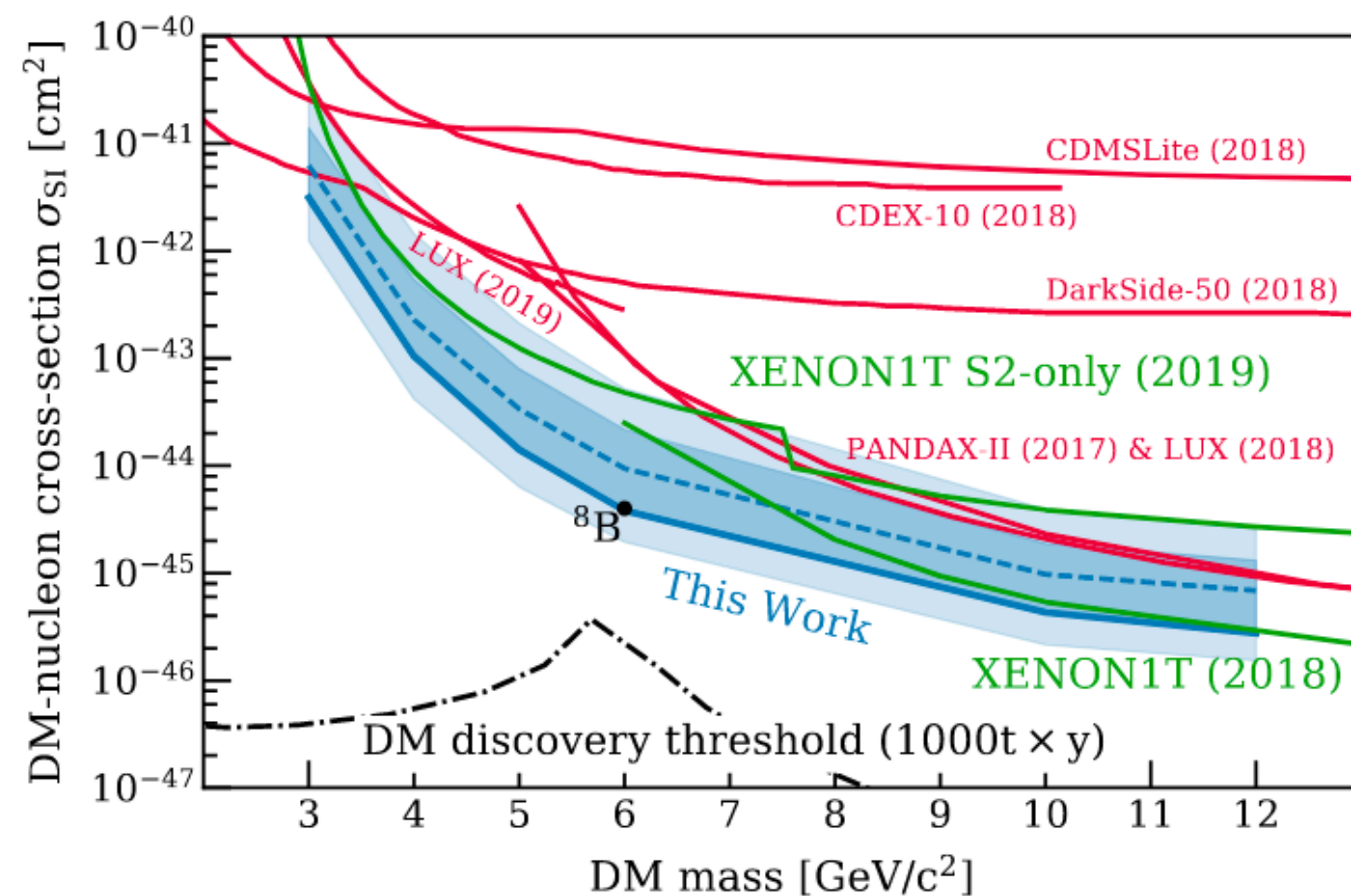
Source	Expectation
<b>CEvNS</b>	<b>2.11</b>
<b>Accidental</b>	<b>5.14</b>
ER	0.21
Radiogenic	0.03
Total	7.65
Observed	6

- In **XENONnT** WIMP Search  
Phys. Rev. Lett. 131, 041003 (2023)

WIMP Dark Matter searches with the XENONnT experiment  
Henning Schulze Eißing  
Today-14:20

**WIMP search results**

	Nominal	Best Fit (200 GeV/c <sup>2</sup> )
ER	134	135(+12)(-11)
Neutrons	1.1(+0.6)(-0.5)	1.1 ± 0.4
CEvNS	0.23 ± 0.06	0.23 ± 0.06
AC	4.3 ± 0.2	4.32 ± 0.16
Surface	14 ± 3	12(+0)(-4)
Total Background	154	152 ± 12
WIMP	-	2.6
Observed	-	152



AC events removed by dedicated cuts.

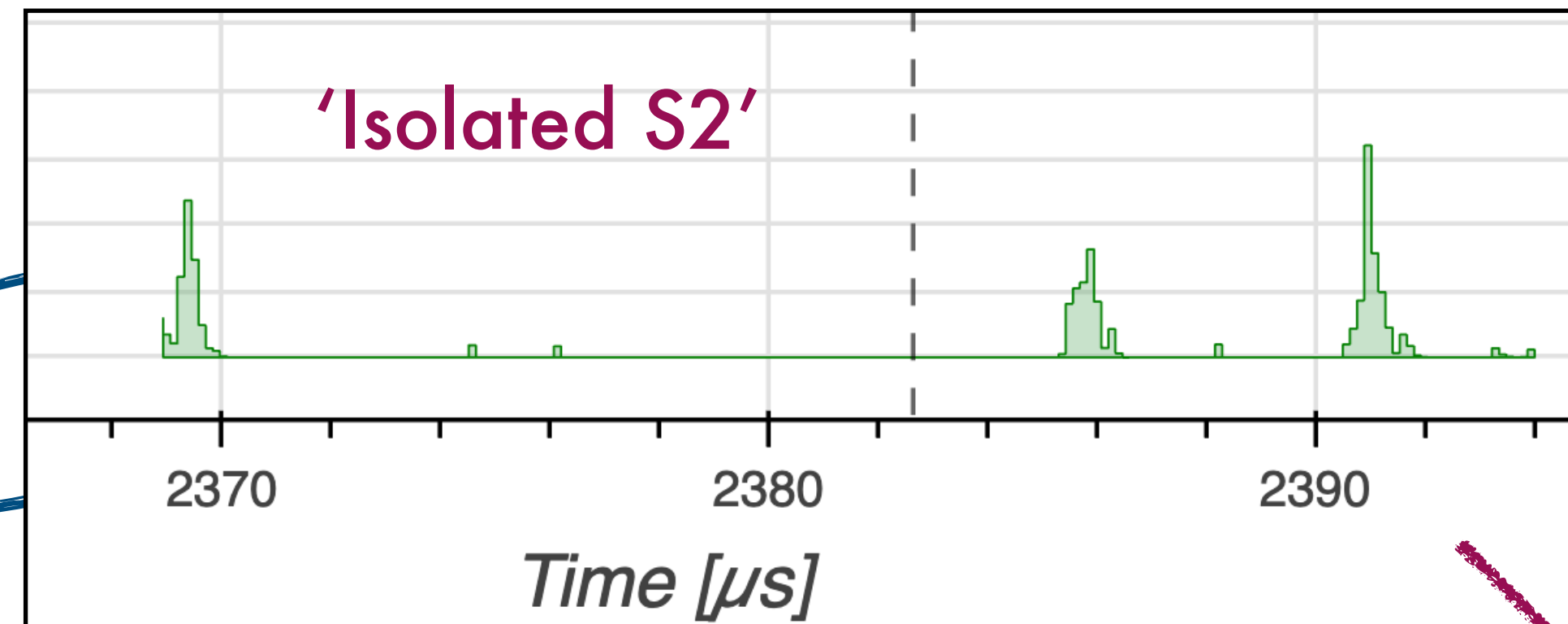
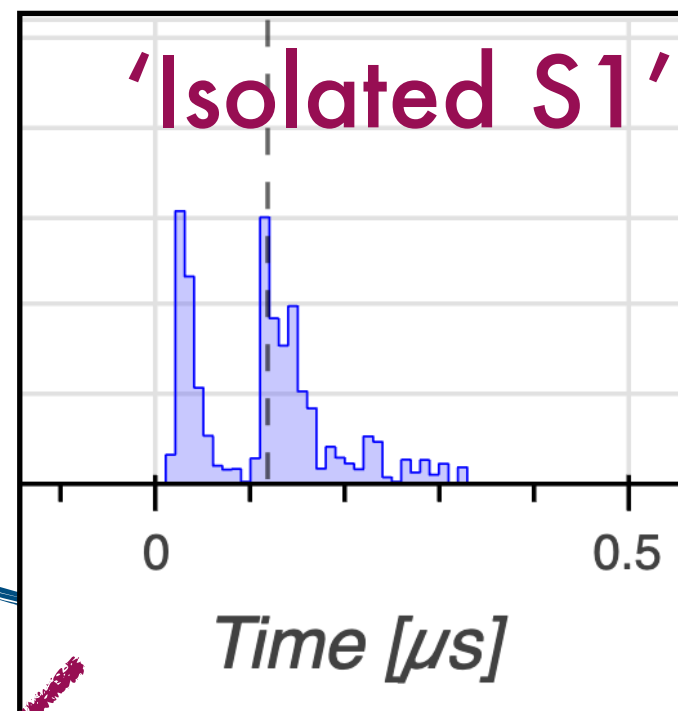


# Accidental Coincidence in XENONnT



The dominant background in low energy nuclear recoil search

Accidentally pair S1 and S2 peaks

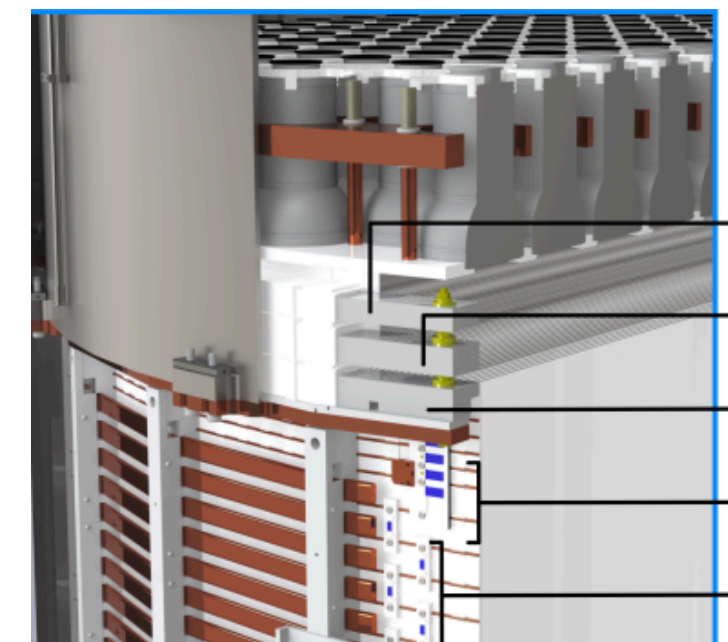
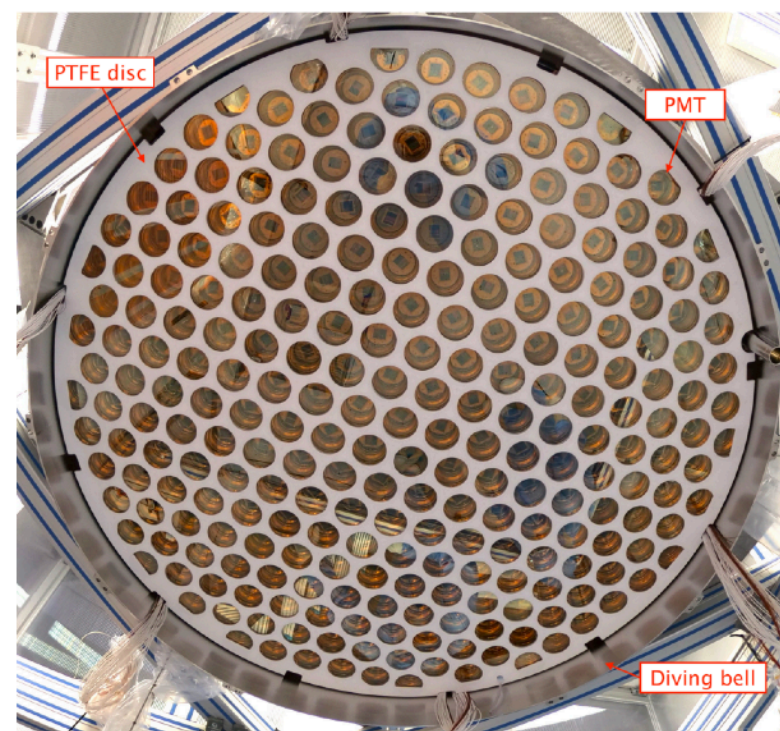


$$Rate_{AC} = \int_{t_0}^{t_1} R_{S1}(t) \times R_{S2}(t) \times T_{max} dt$$

In low energy NR ROI: (S1 2/3 hits, S2 from few to dozens electrons)

Iso-S1 Rate	Iso-S2 Rate	T max	Raw AC Rate
~ 15 Hz	~ 0.15 Hz	2.25 ms	5 mHz (~400/day)

20 V/cm drift field

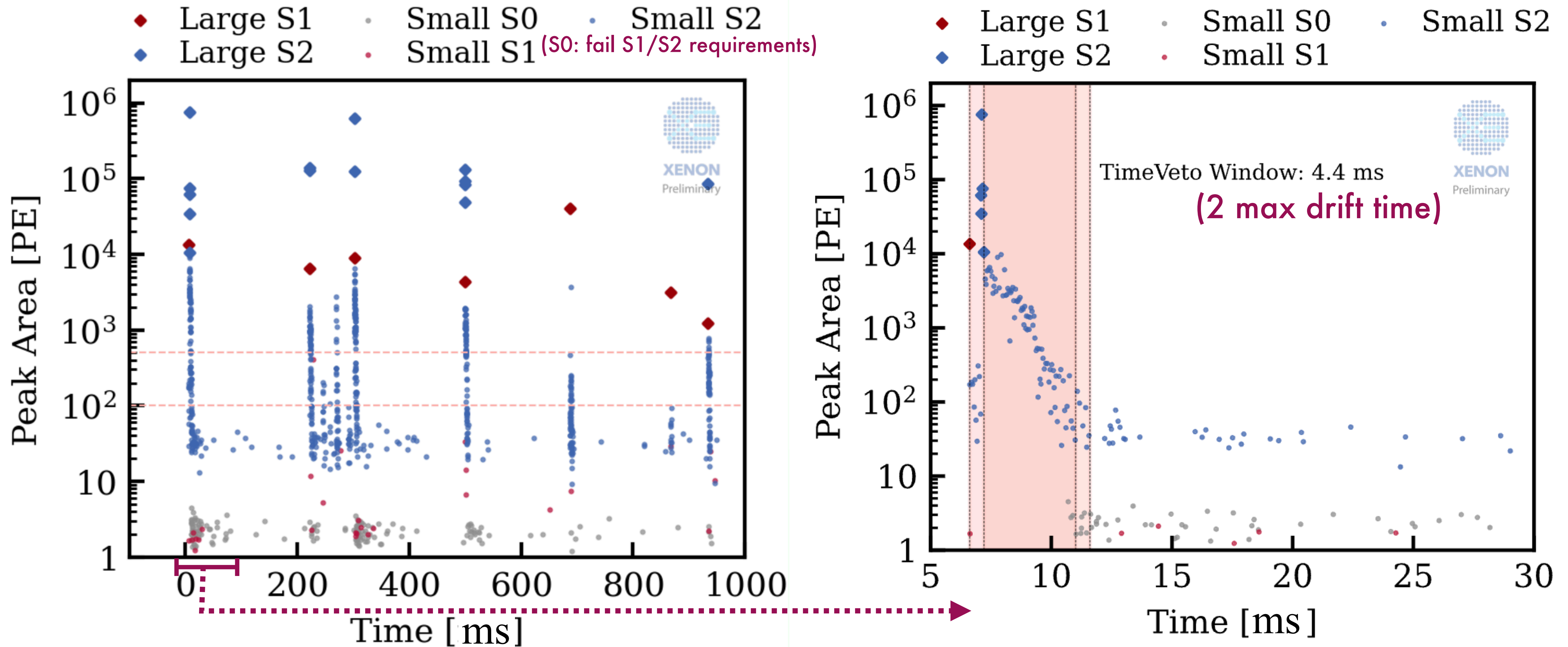




# 1 Second in the data ...



## How to find clean enough exposure?

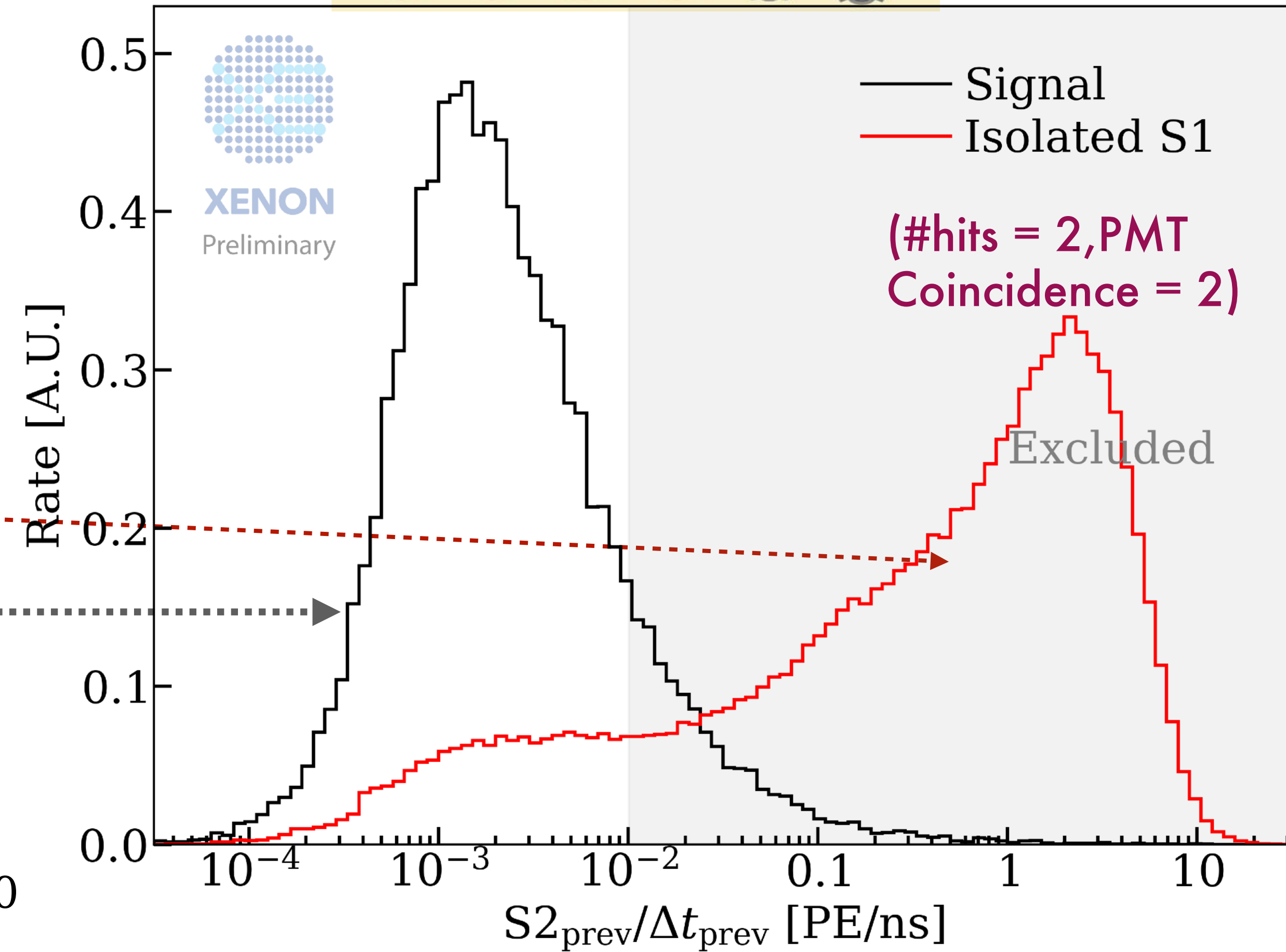
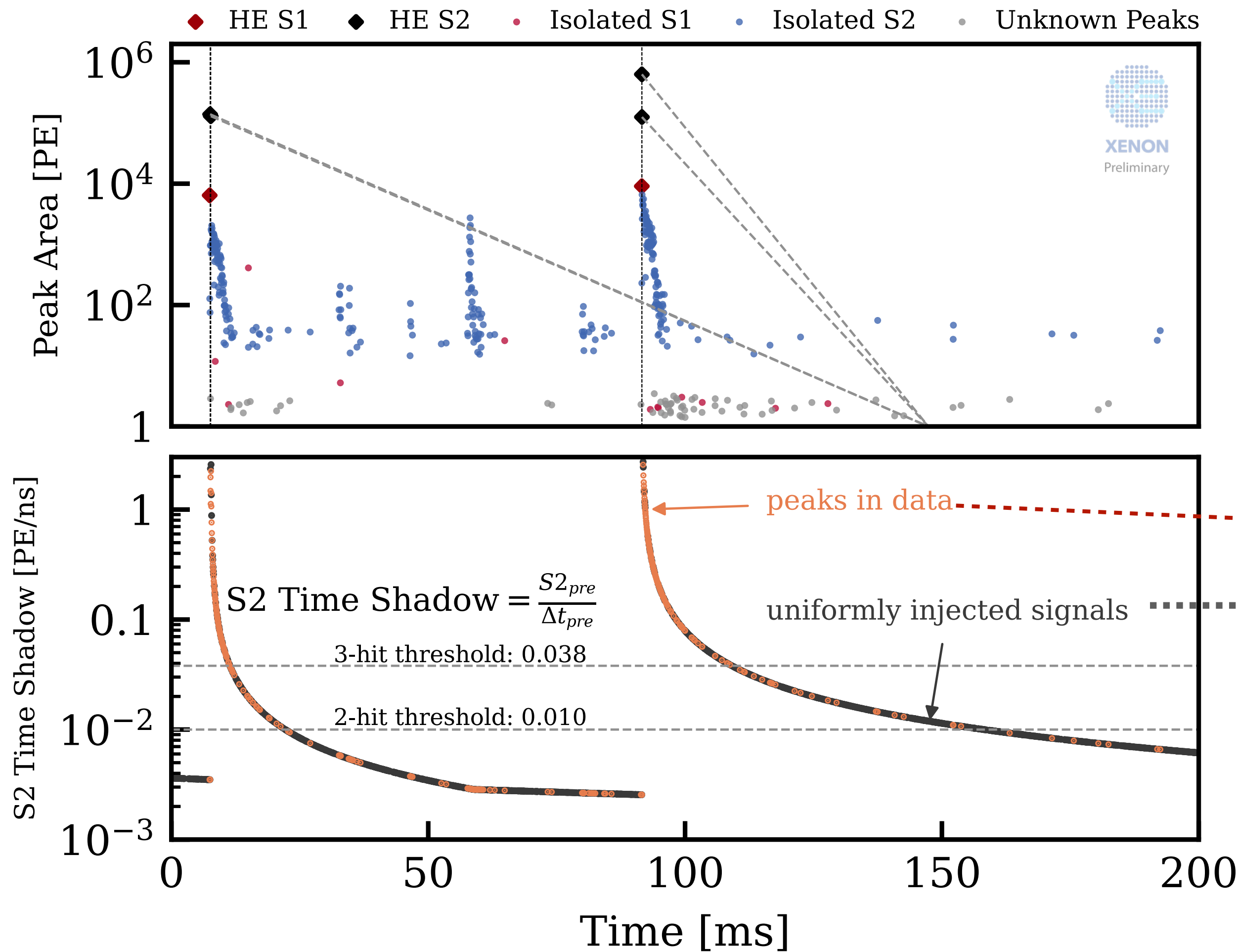




# Time Shadow



Quantify the cleanliness of the exposure

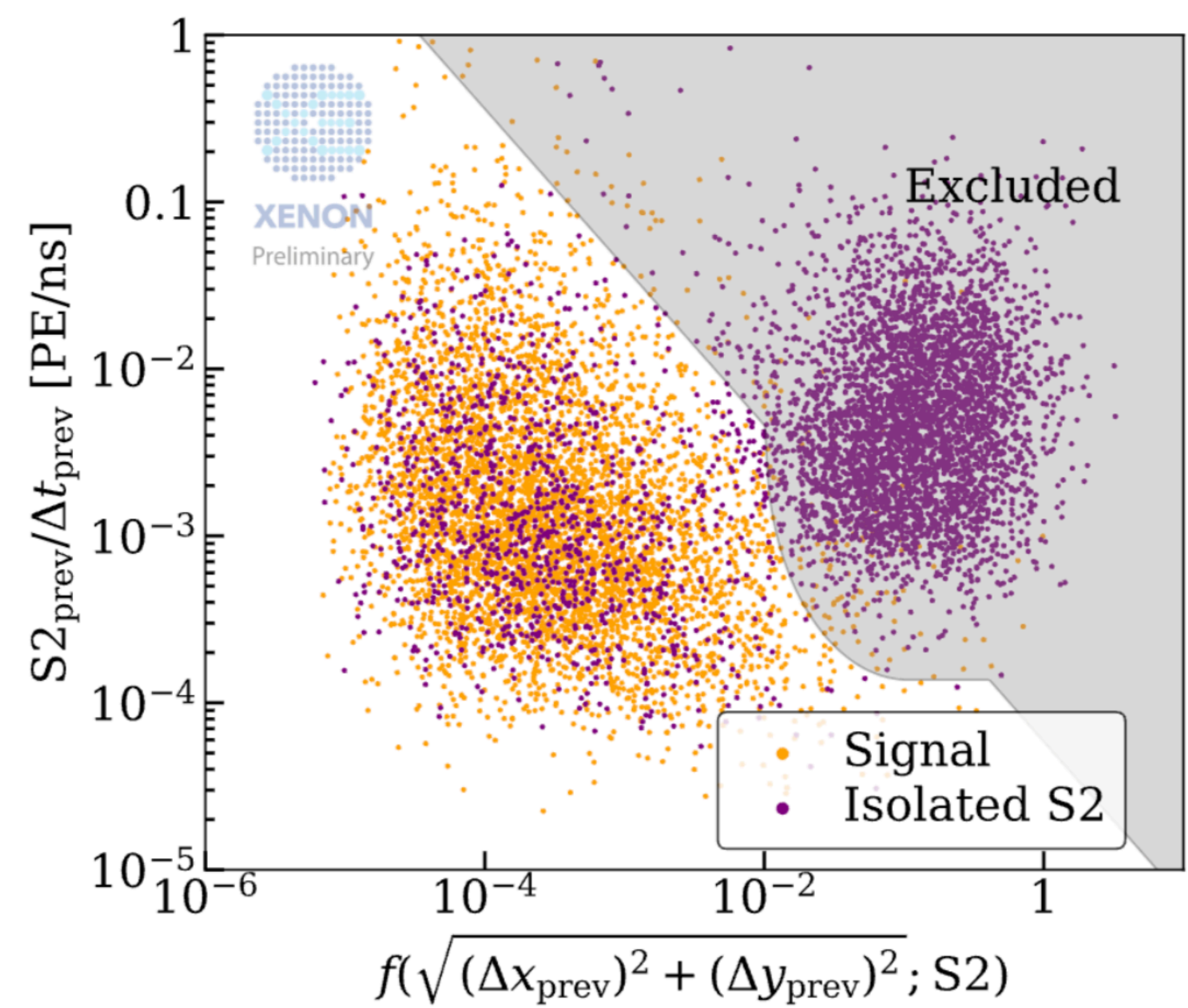
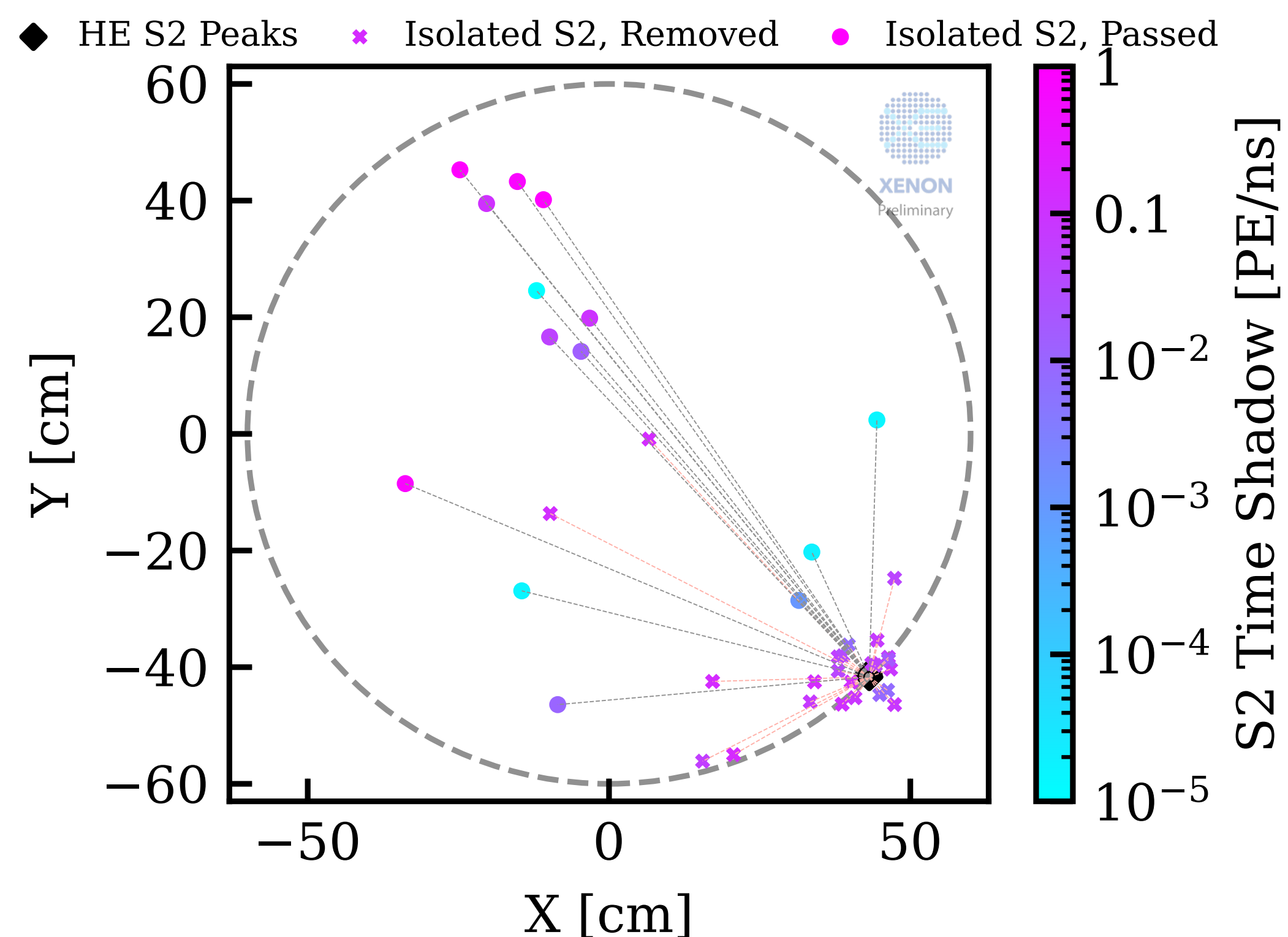




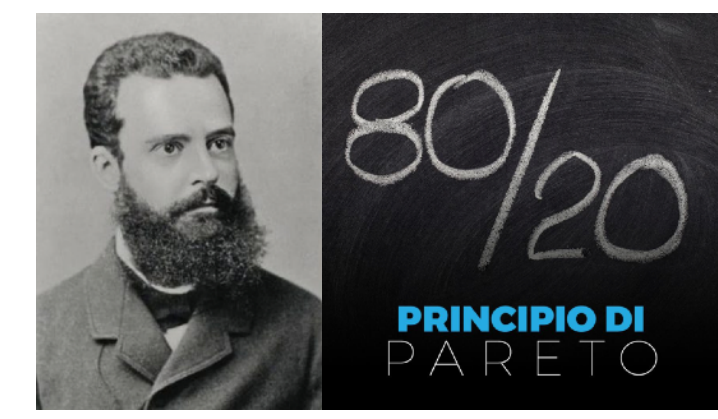
# Shadow Cut (Time + Position)



Remove the time&space contains the most IsoS1/S2



- Cut threshold set to remove the worst 20% of time & space.

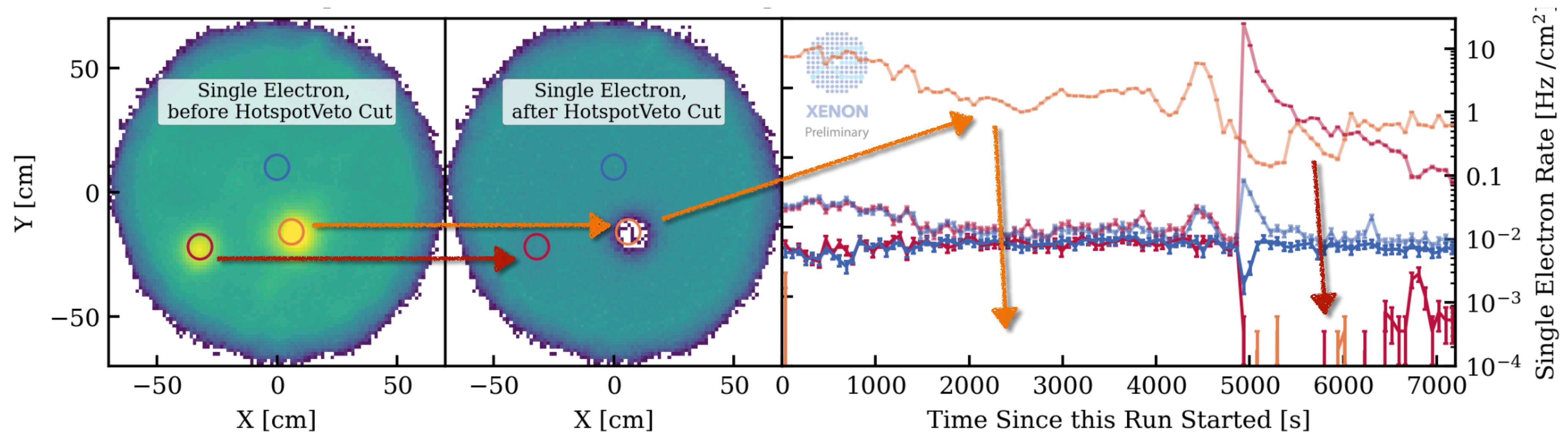




# Following this idea



## To deal with detector's imperfections— hotspot veto

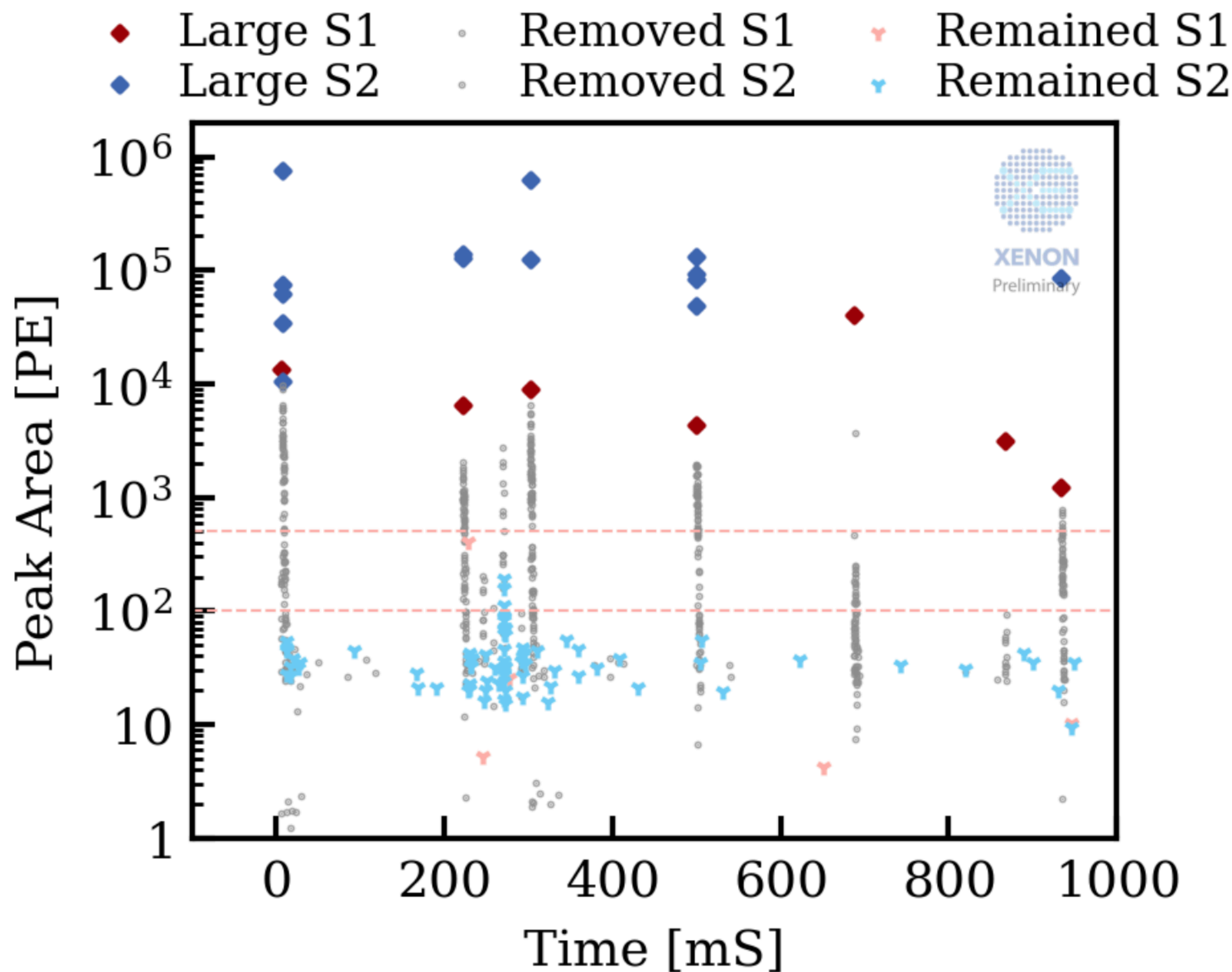


- Sit in **hotspot** (localized single electron burst)? — “Hotspot Veto” selection: Remove those with more adjacent **single electrons** than normal; **only** remove the contaminated **time&space**.
- Sit in ‘**fuzzy**’ environment? — “Ambience” selection: remove those peaks/events with too much **S1**, **S2**, **S0**, **lone hits** in the neighborhood

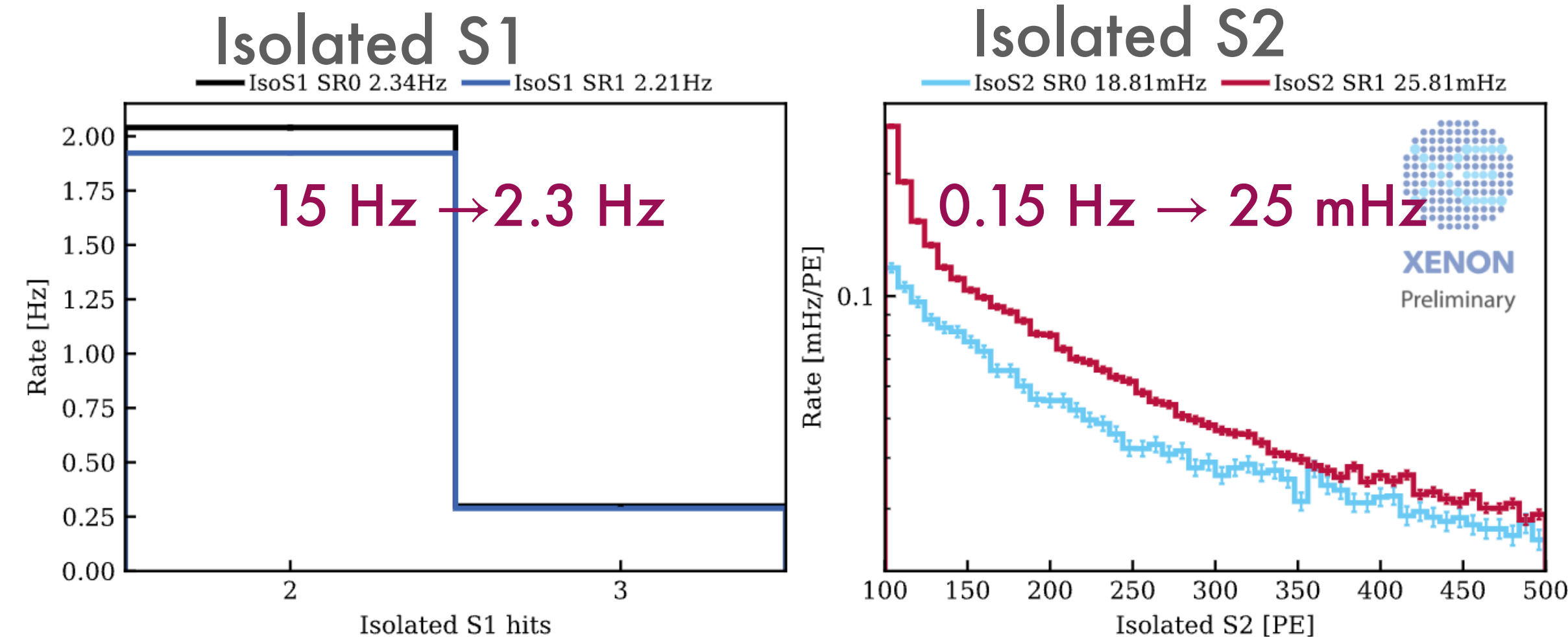
# Combined efforts of suppress isolated peaks



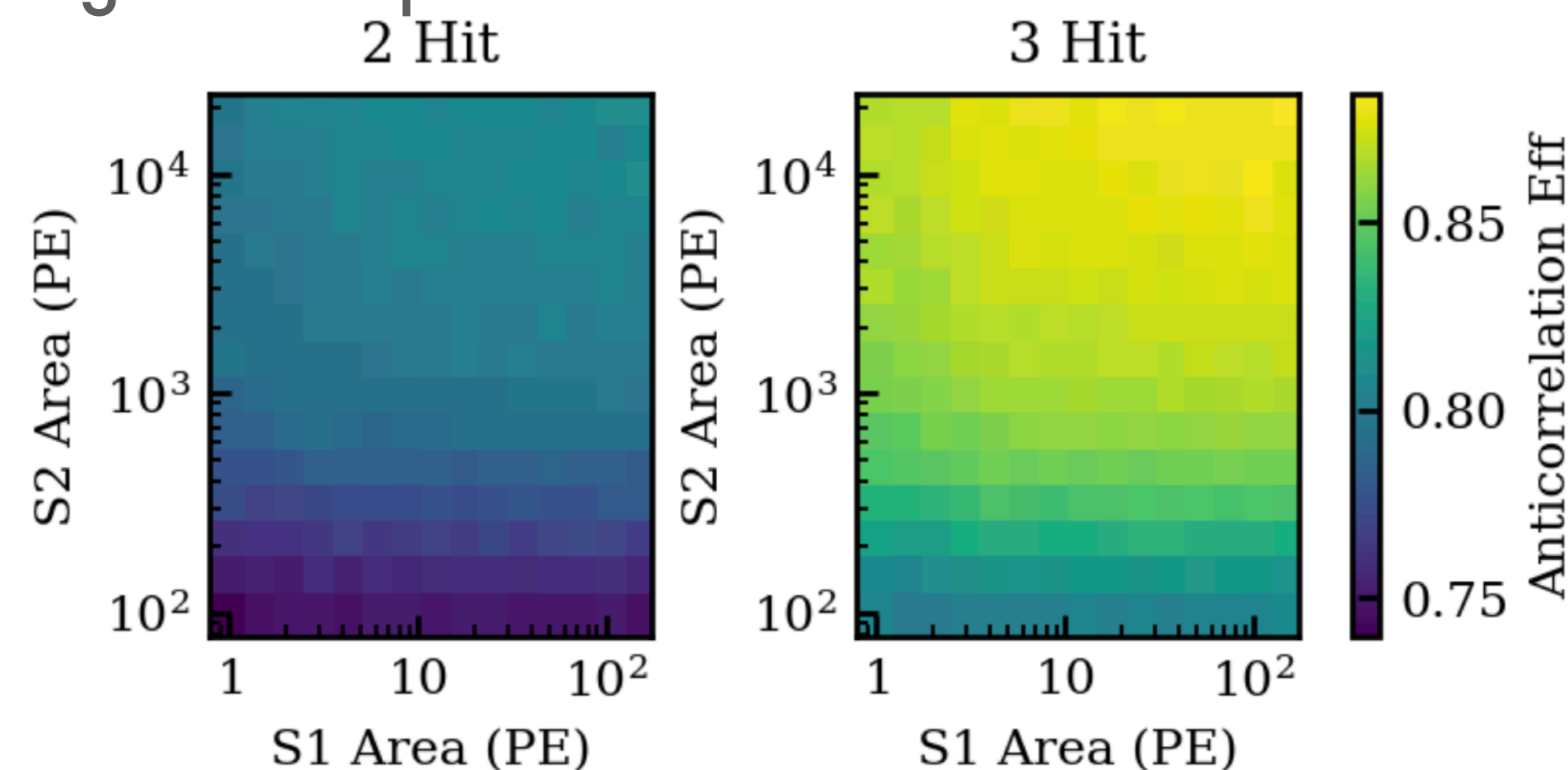
## Isolated peaks suppressed vs. signal acceptance



After the time&space correlation cuts, the majority of isolated peaks is removed.



Signal acceptance evaluated:





# Data-driven simulation for AC Model



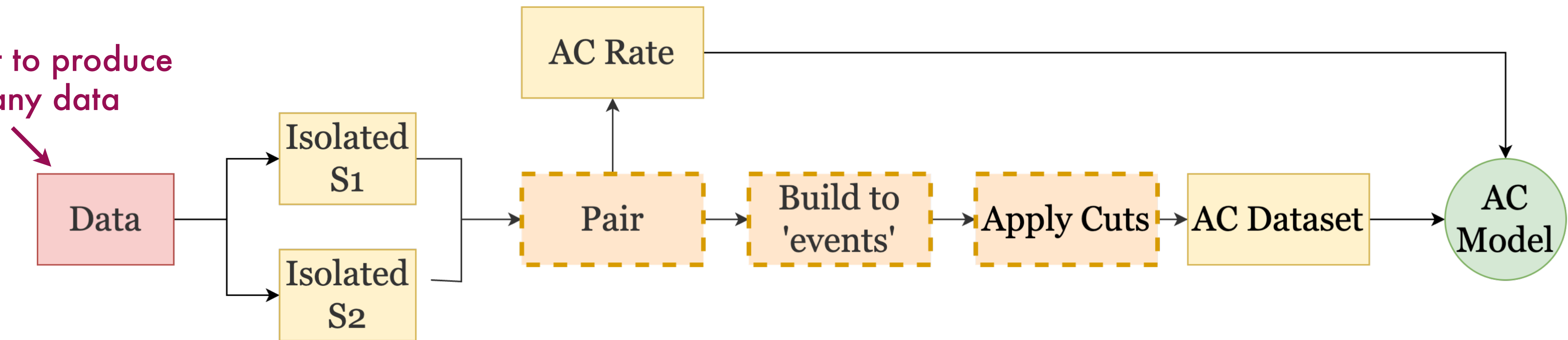
**axidence 0.3.1**

`pip install axidence`

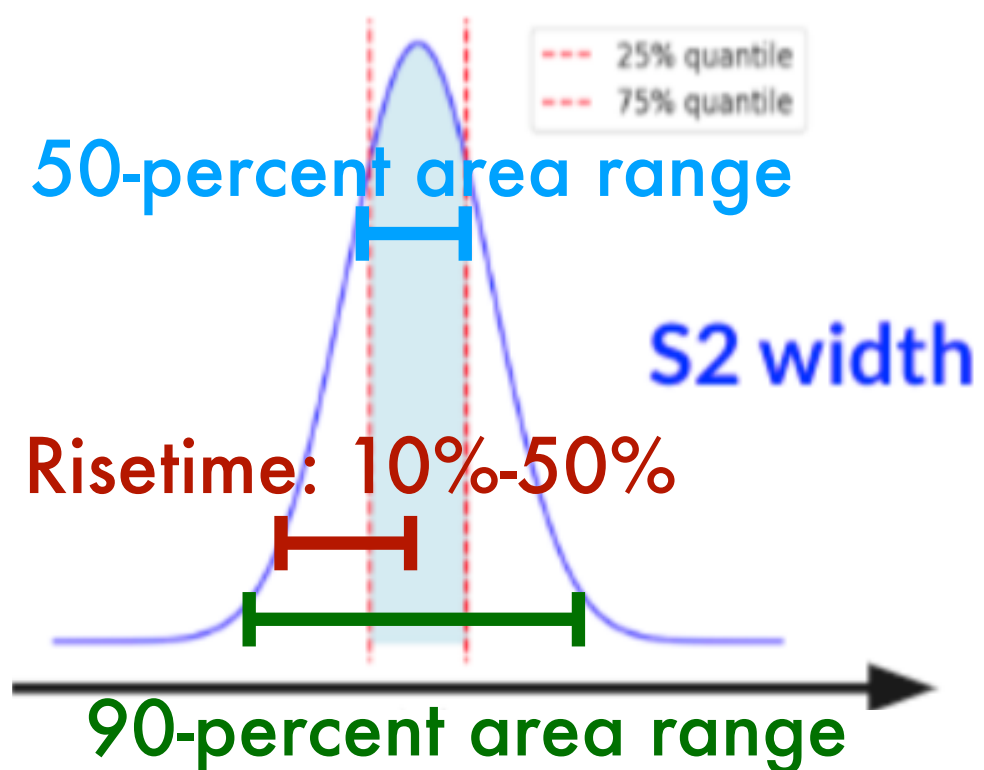
**dachengx**

- Stream-like structure: Bootstrap, pair(assign drift time + match time shadow), then go through event-building.

Change the input to produce AC datasets for any data



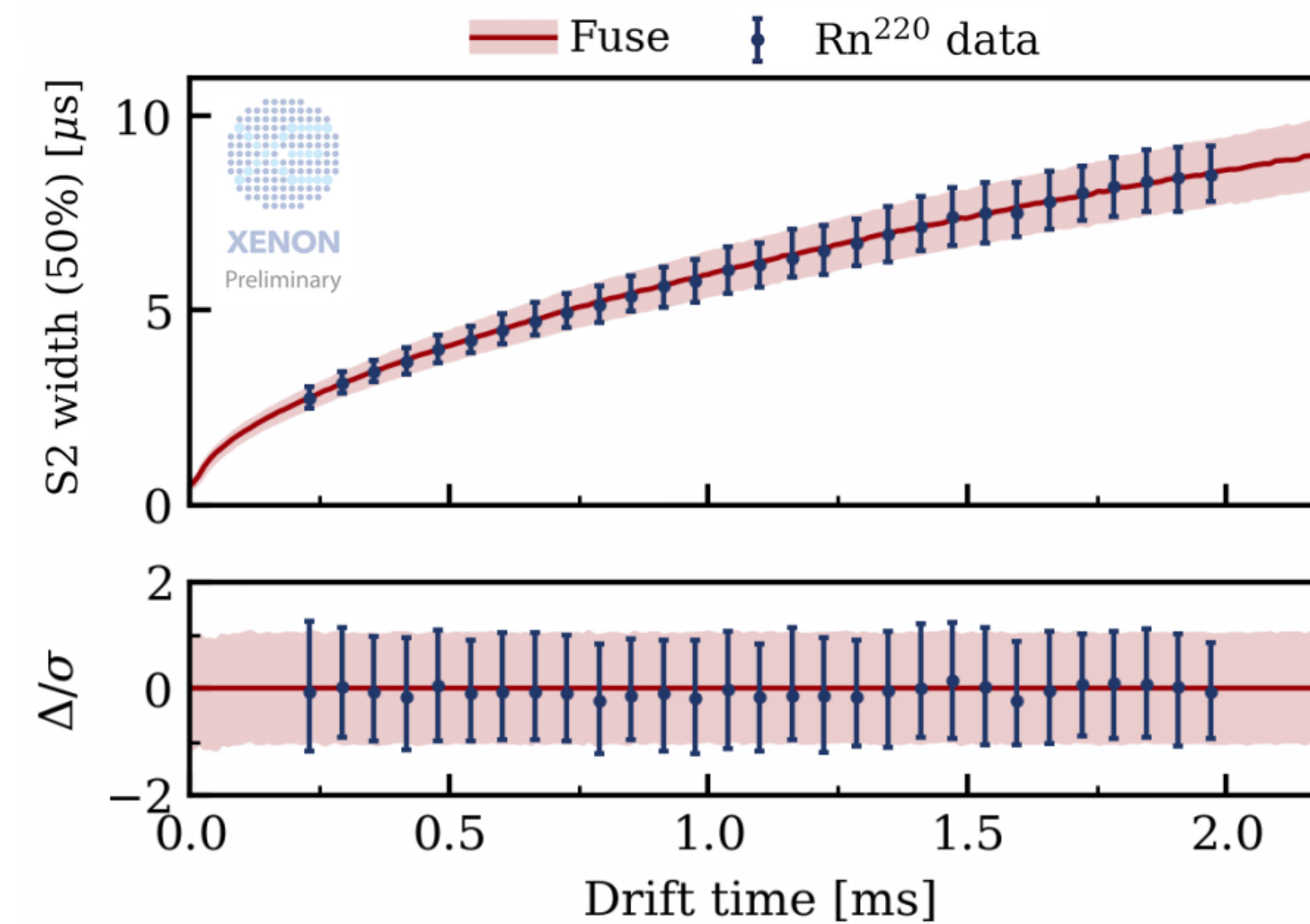
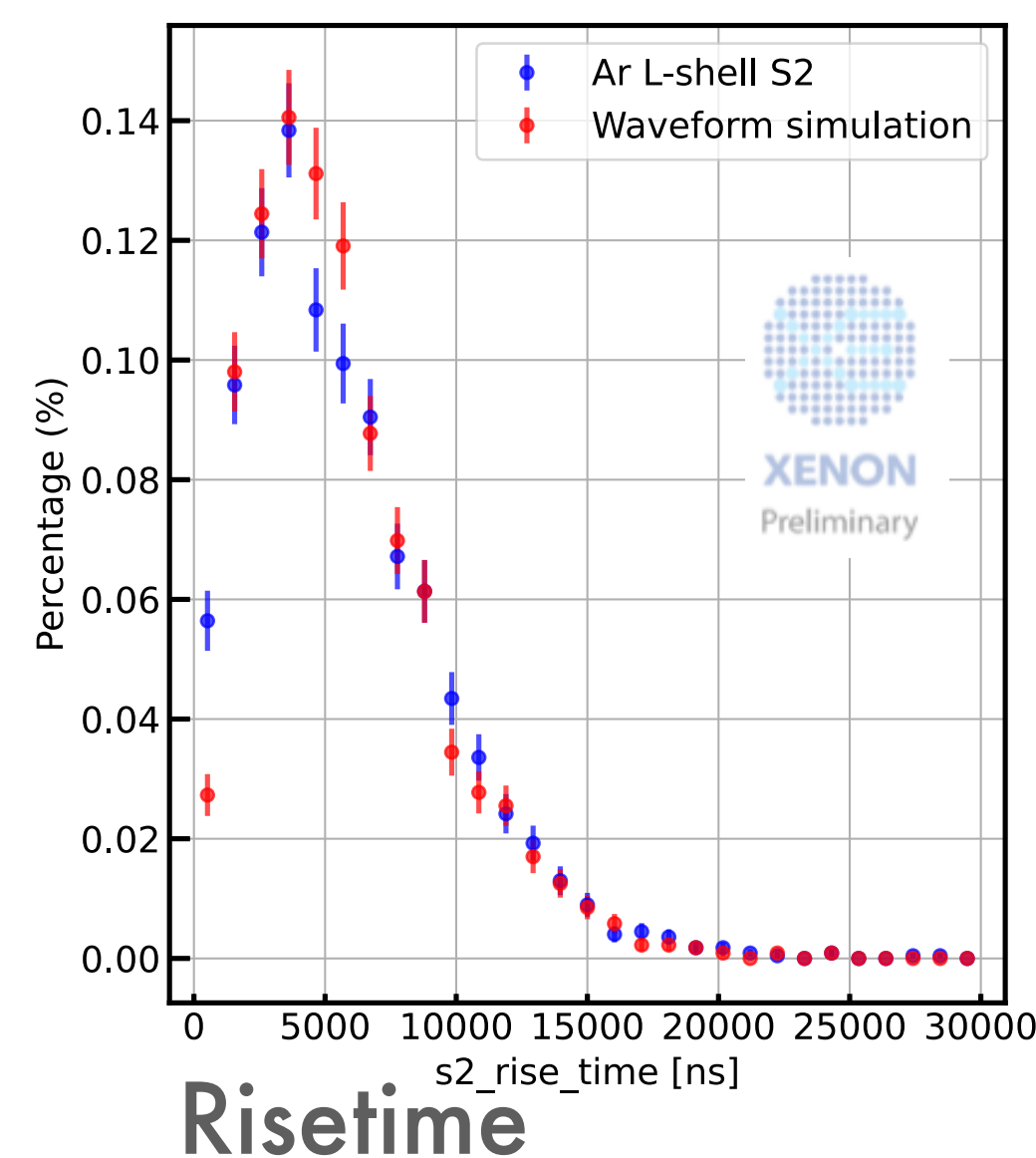
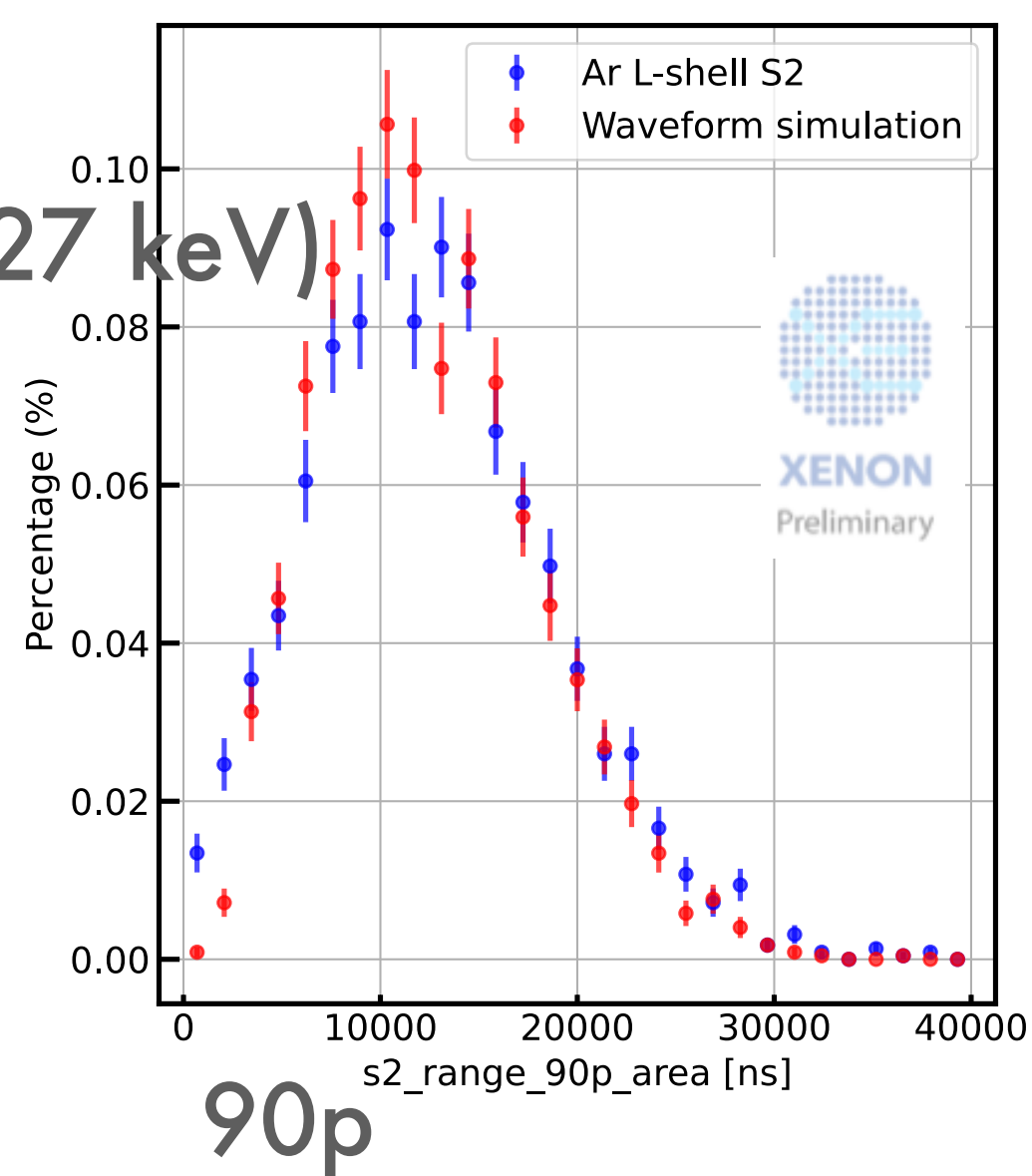
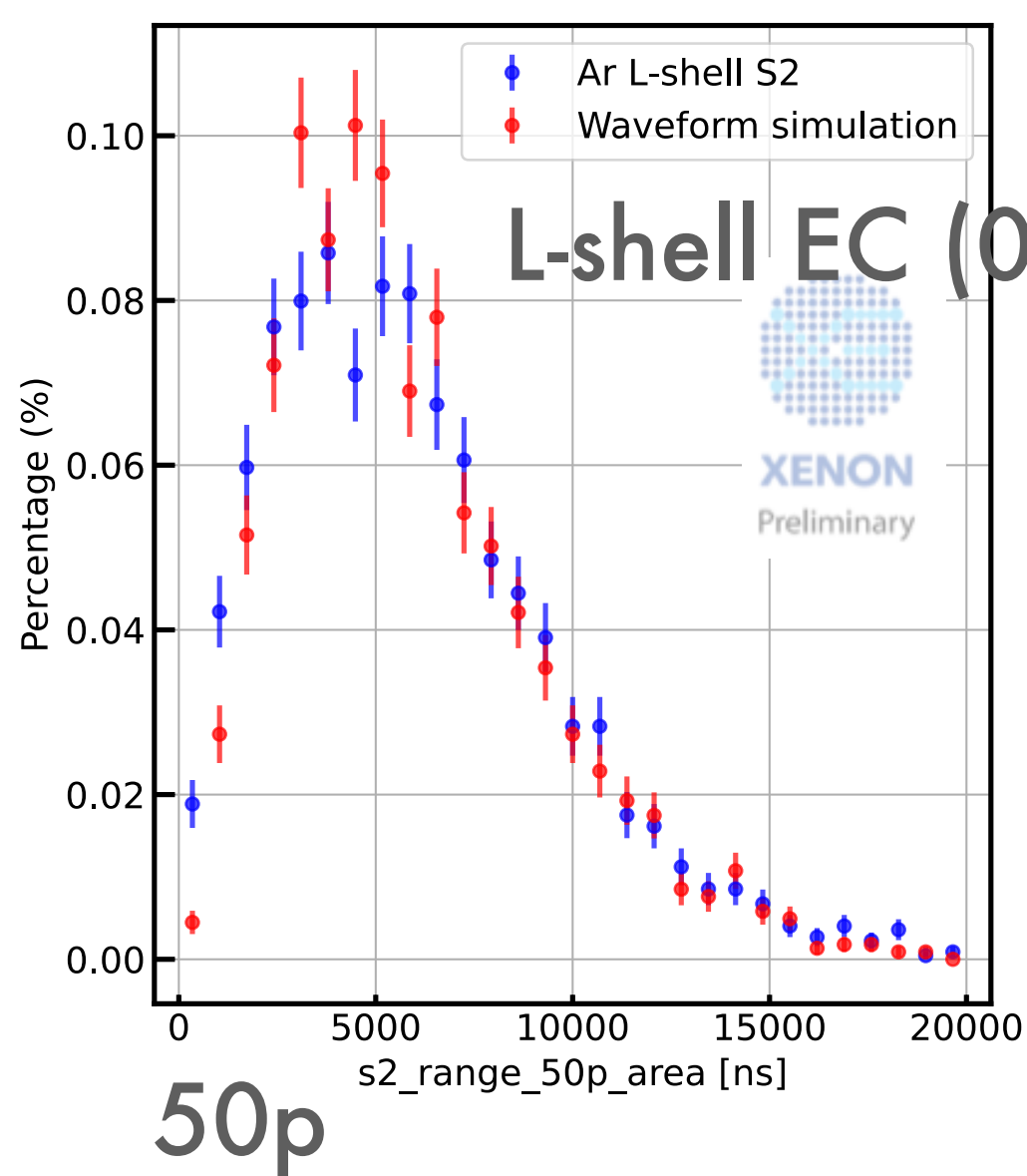
# XENON Fuse: Framework for Unified Simulation of Events



- Simulation include the microphysics, detector physics, PMT&DAQ response to get events.
- simulation & data match well

**xenon-fuse 1.3.0**

```
pip install xenon-fuse
```



- Enormous & reliable datasets for machine learning training!



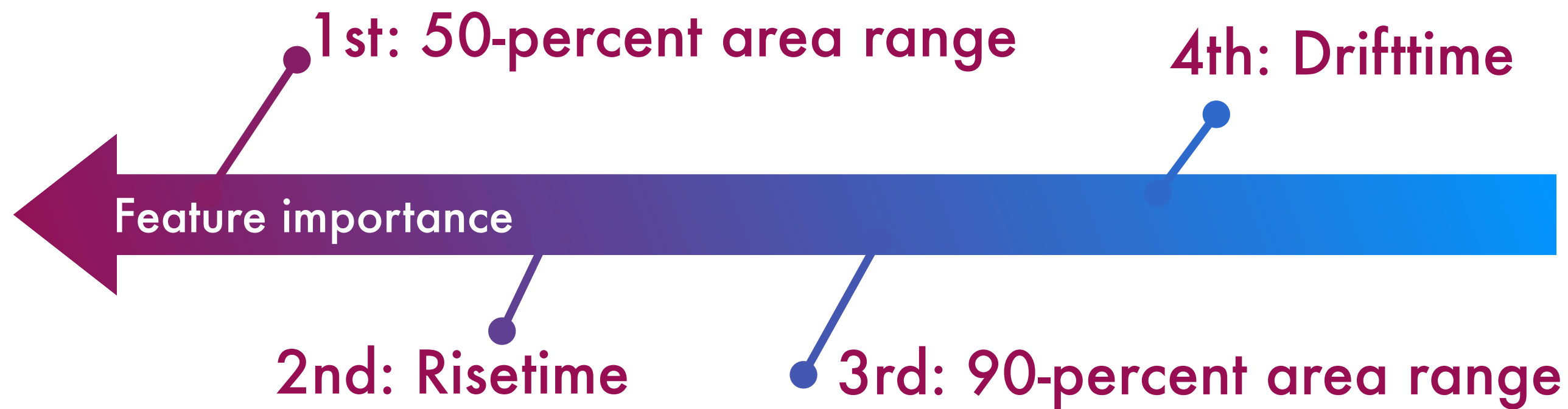
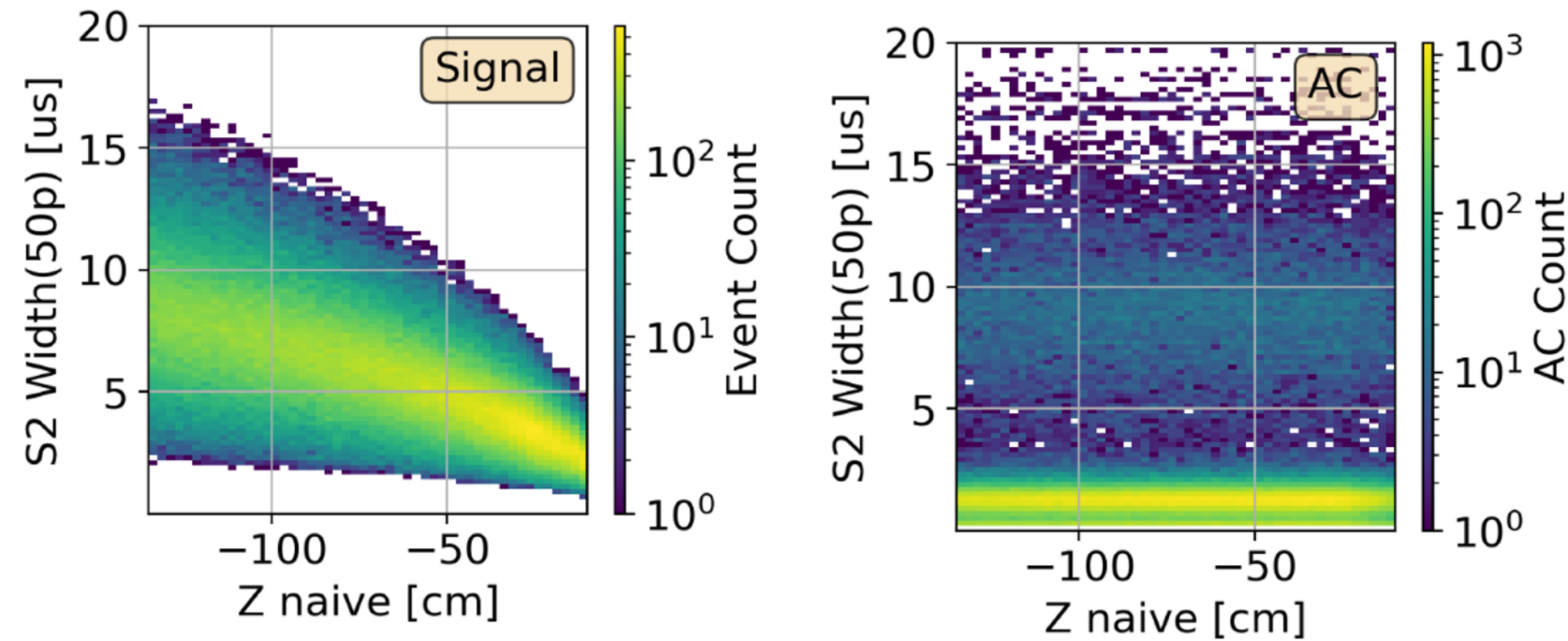
# S1/S2 Pulse shape into Gradient Boosting Decision Tree



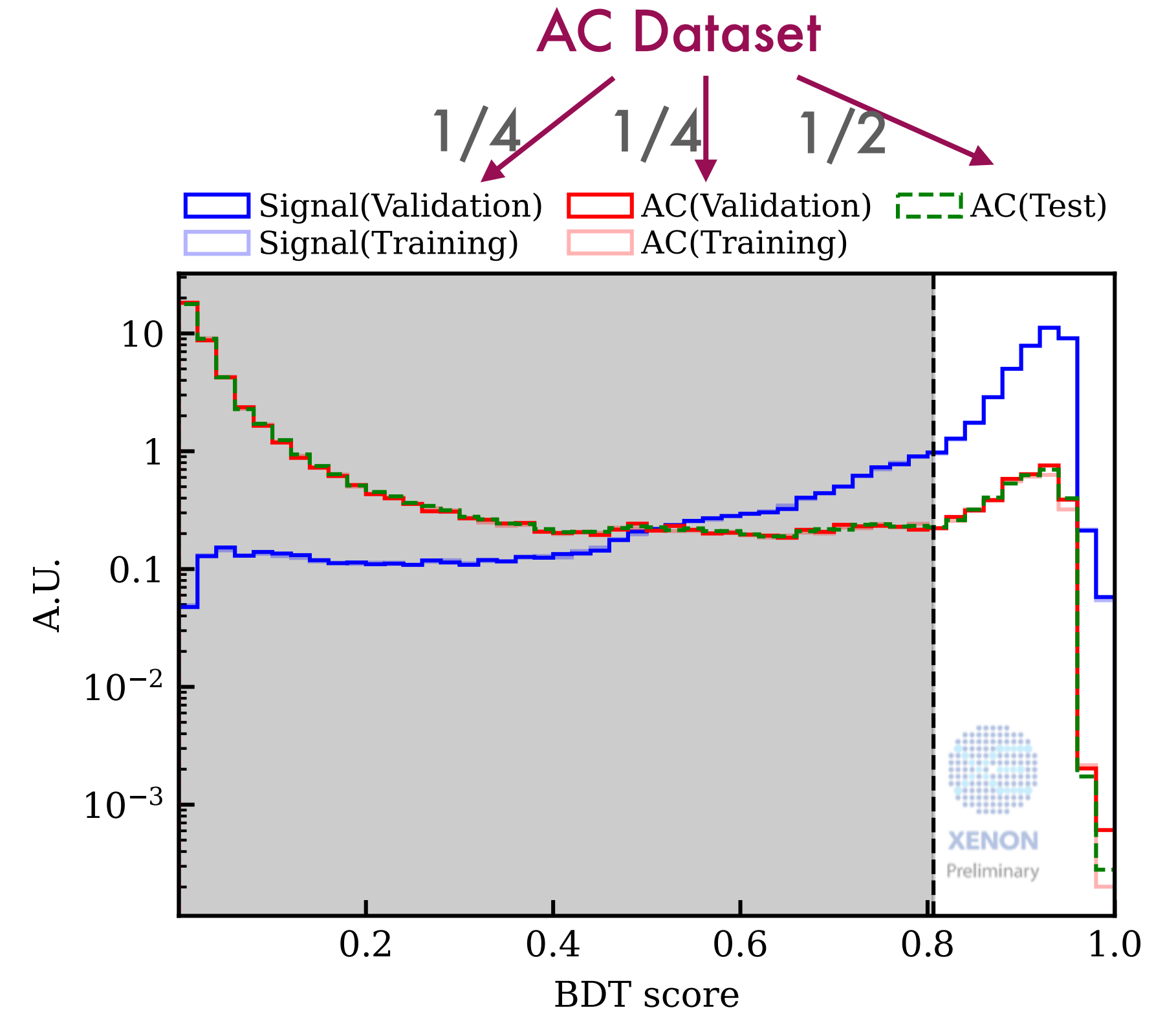
## S2BDT: Diffusion model & Abnormal Width

Diffusion model:

$$t_{50p} \propto \sqrt{T_{drift}}$$



- Split datasets to prevent overfitting

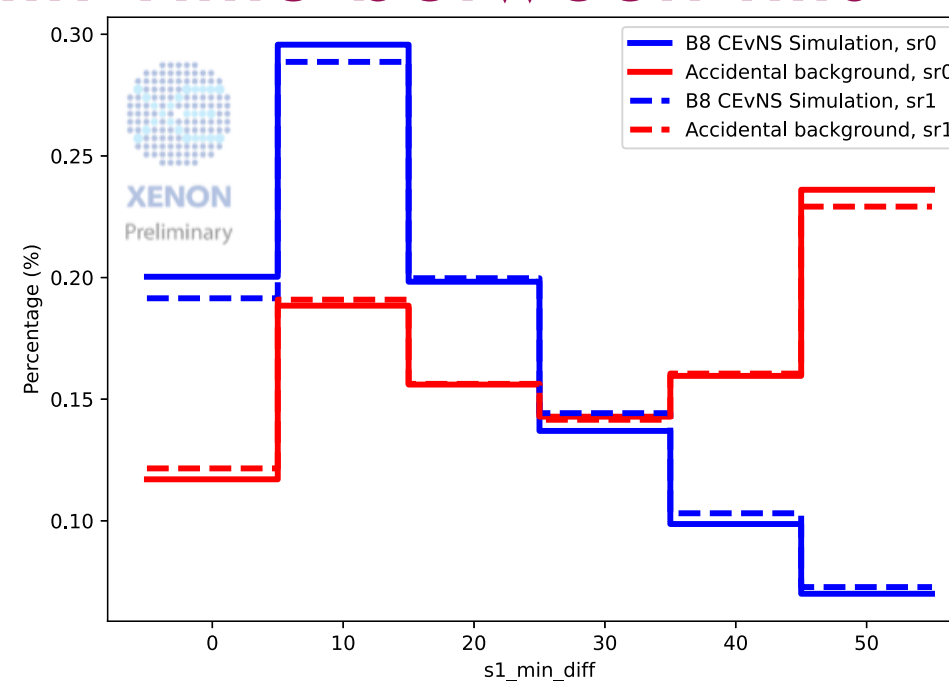
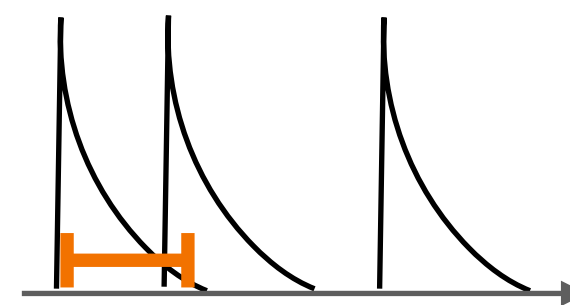
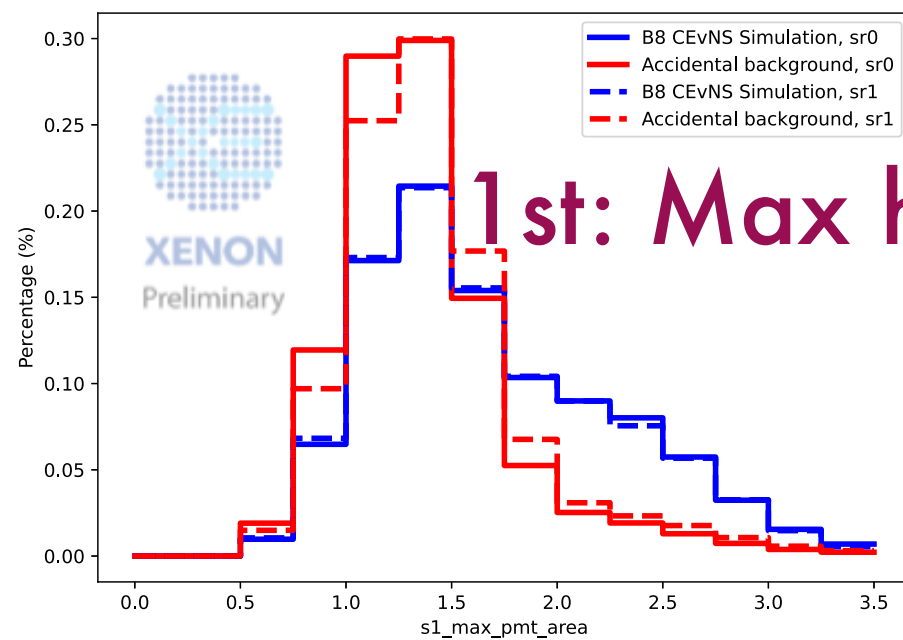


- Reject about **90%** AC with **80%** signal acceptance
- The remaining part still have discrimination power.

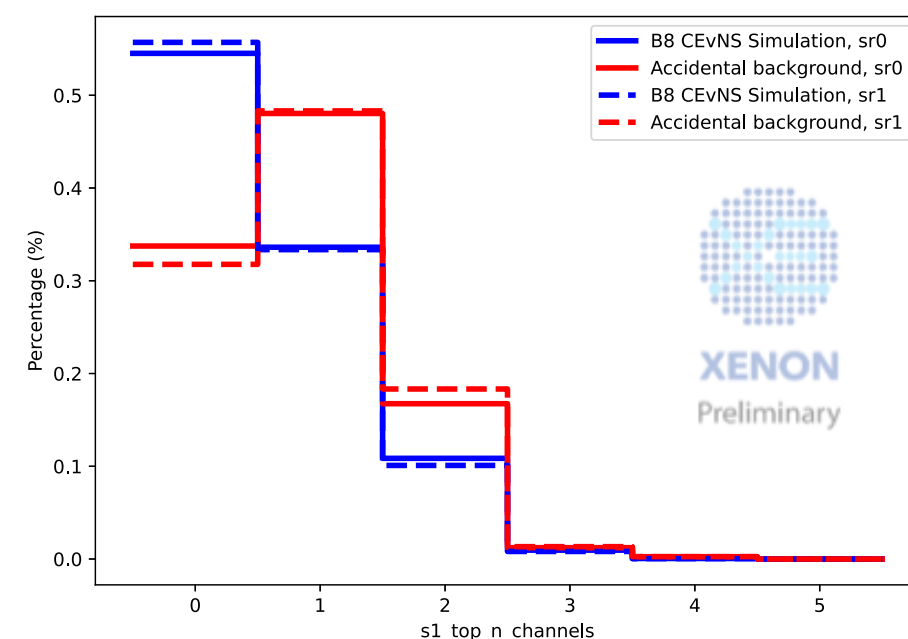
# S1/S2 Pulse shape into Gradient Boosting Decision Tree



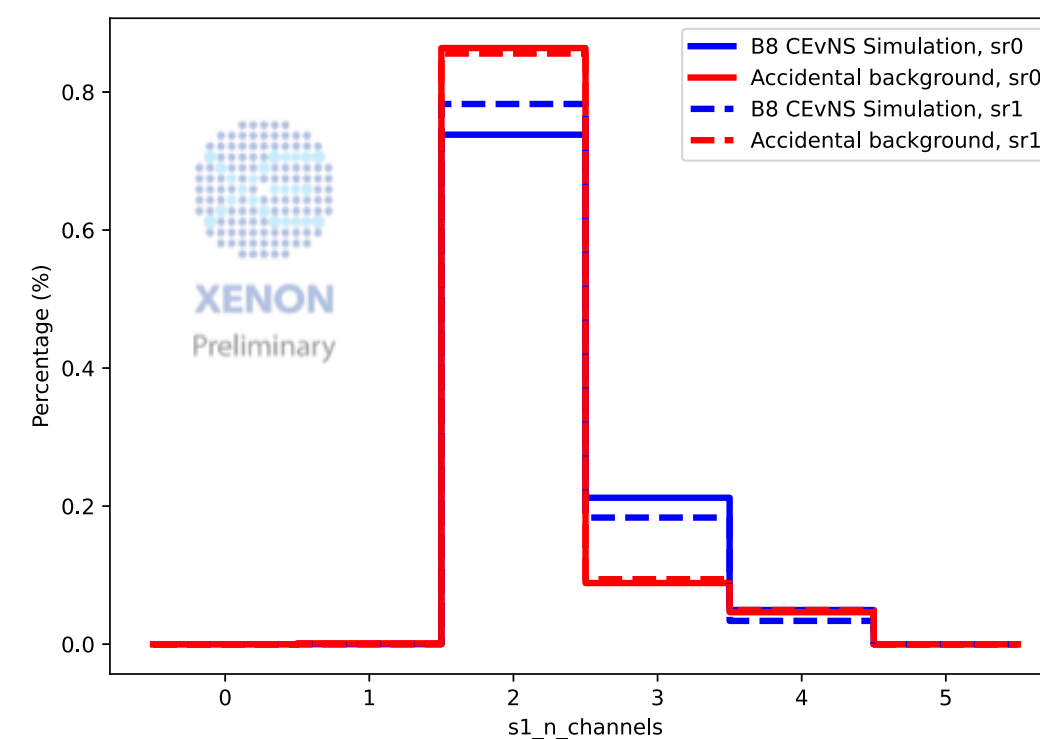
## S1BDT: LXe light + signal spectrum vs. lone hits pile-up



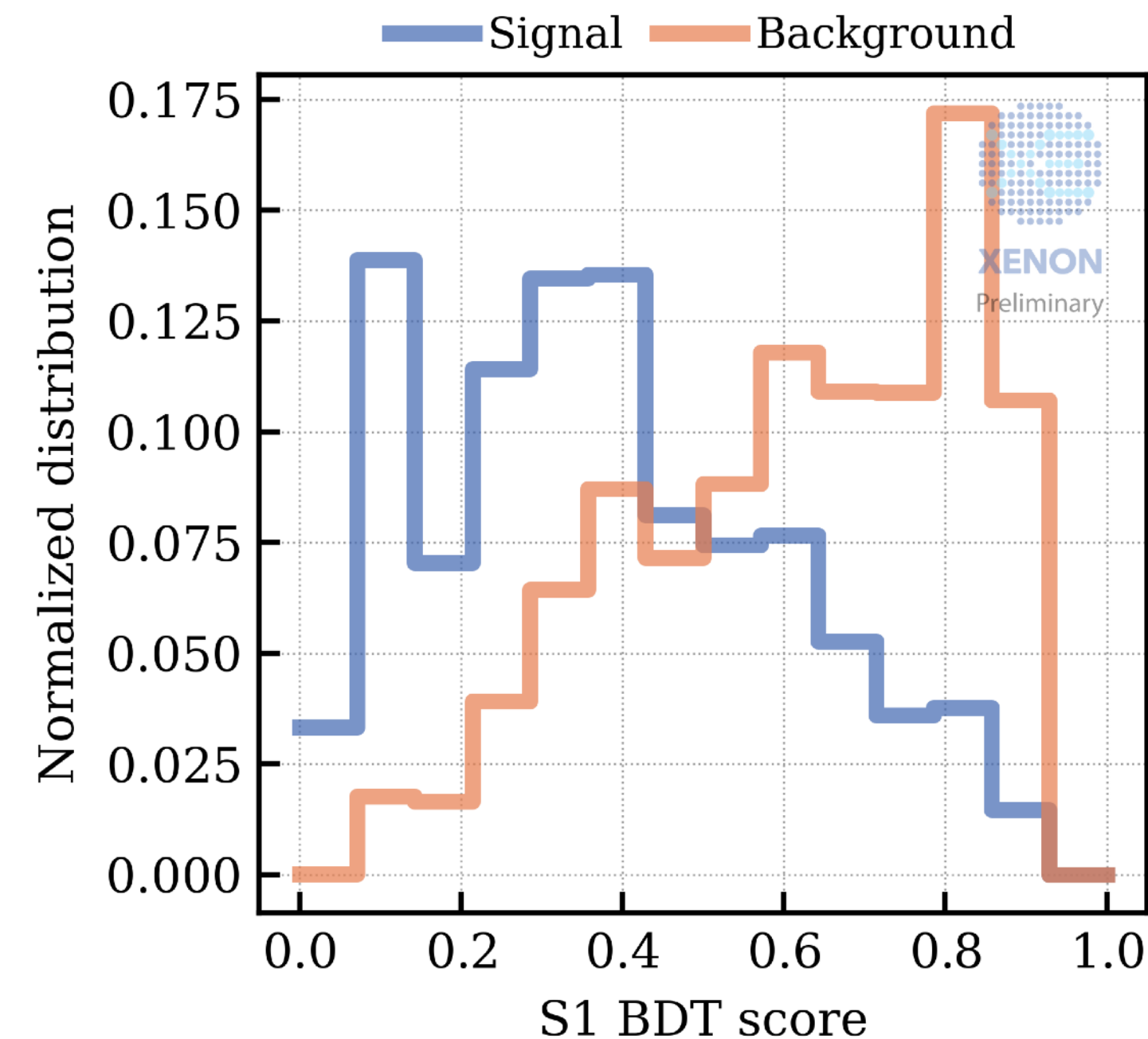
3rd: # of hits from Top Array



4th: total # of hits



Feature importance



- Trained with IsoS1 vs. Simulated B8 S1
- Utilize this discrimination power in the inference. So do the remaining parameter space of the TimeShadow and S2BDT cut.



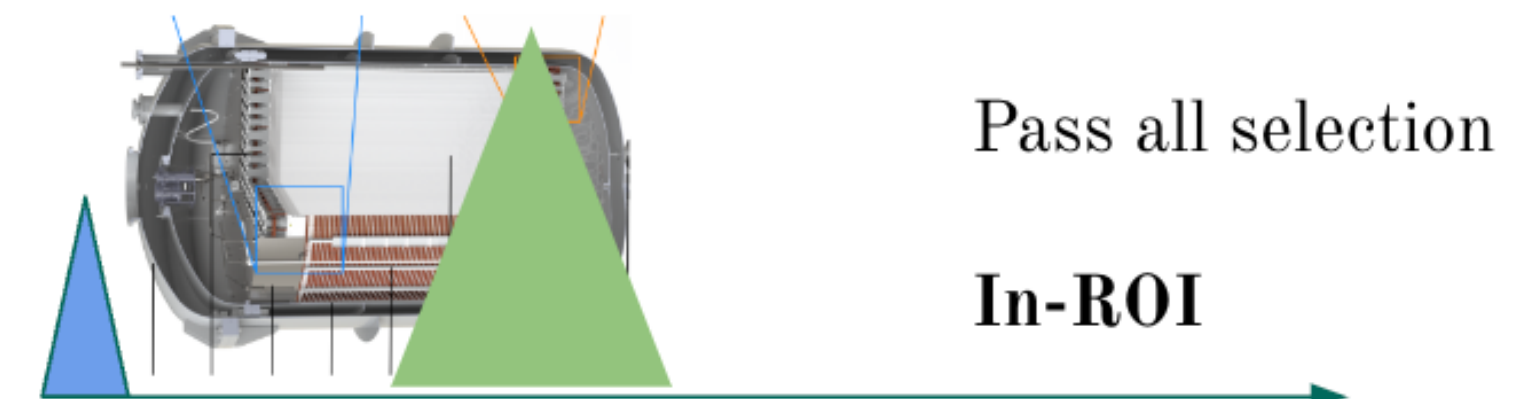
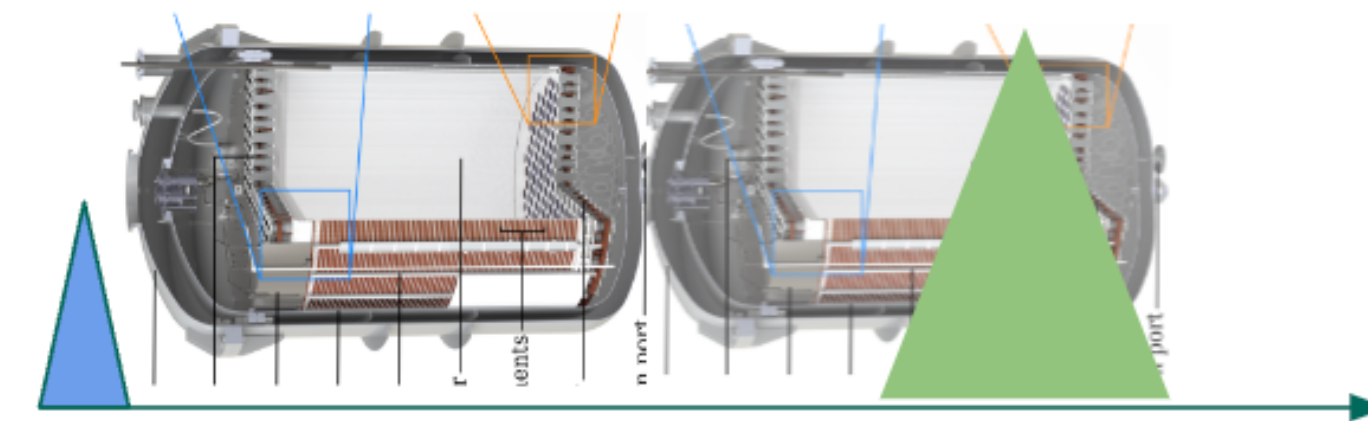
# Model Validation & Systematic Error



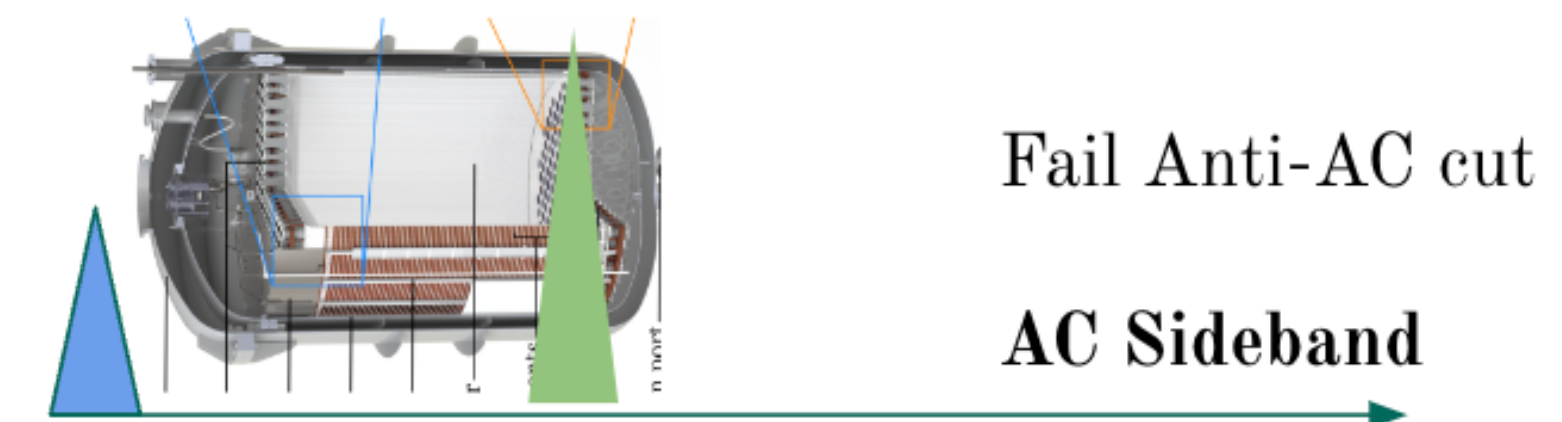
## Test the mode with AC-rich datasets

- statistical error is small (less than 1%). Systematic error need to be tested by data
- How to find suitable dataset contain large and pure AC events?

- Build events longer than the TPC, thus build **pure AC** events
- In high rate **calibration** data
- In science search data, select events which only failed anti-AC cuts: **ACSideband**

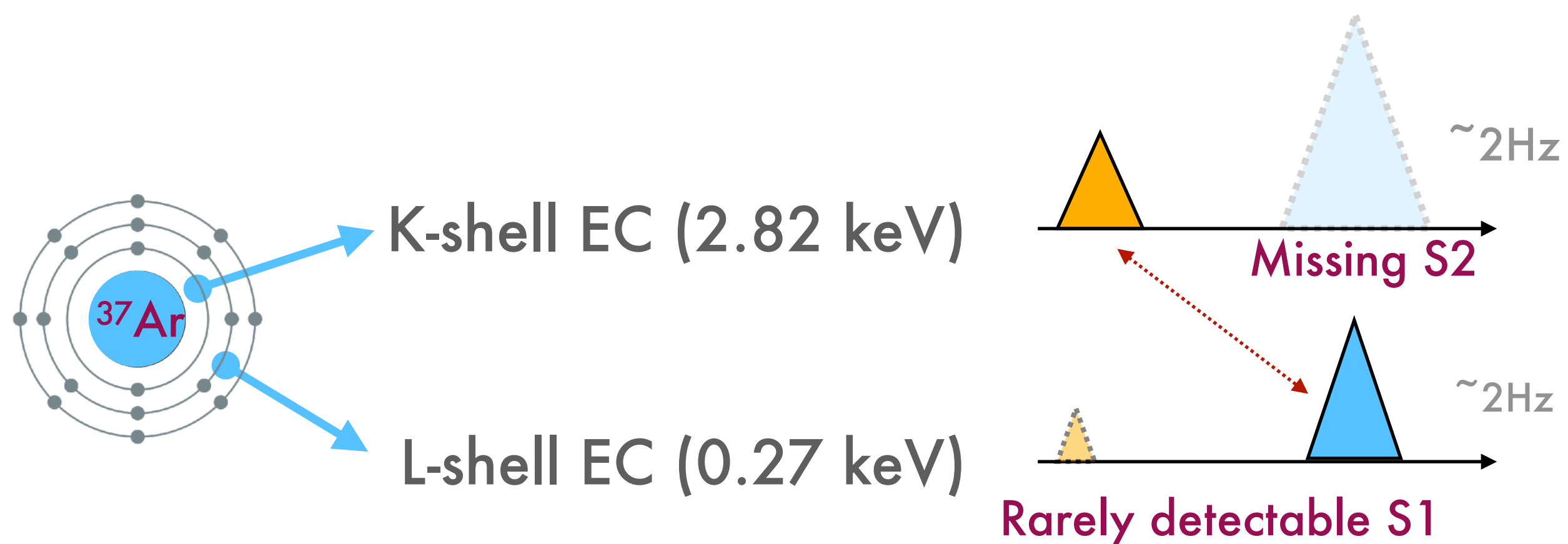


Drift time



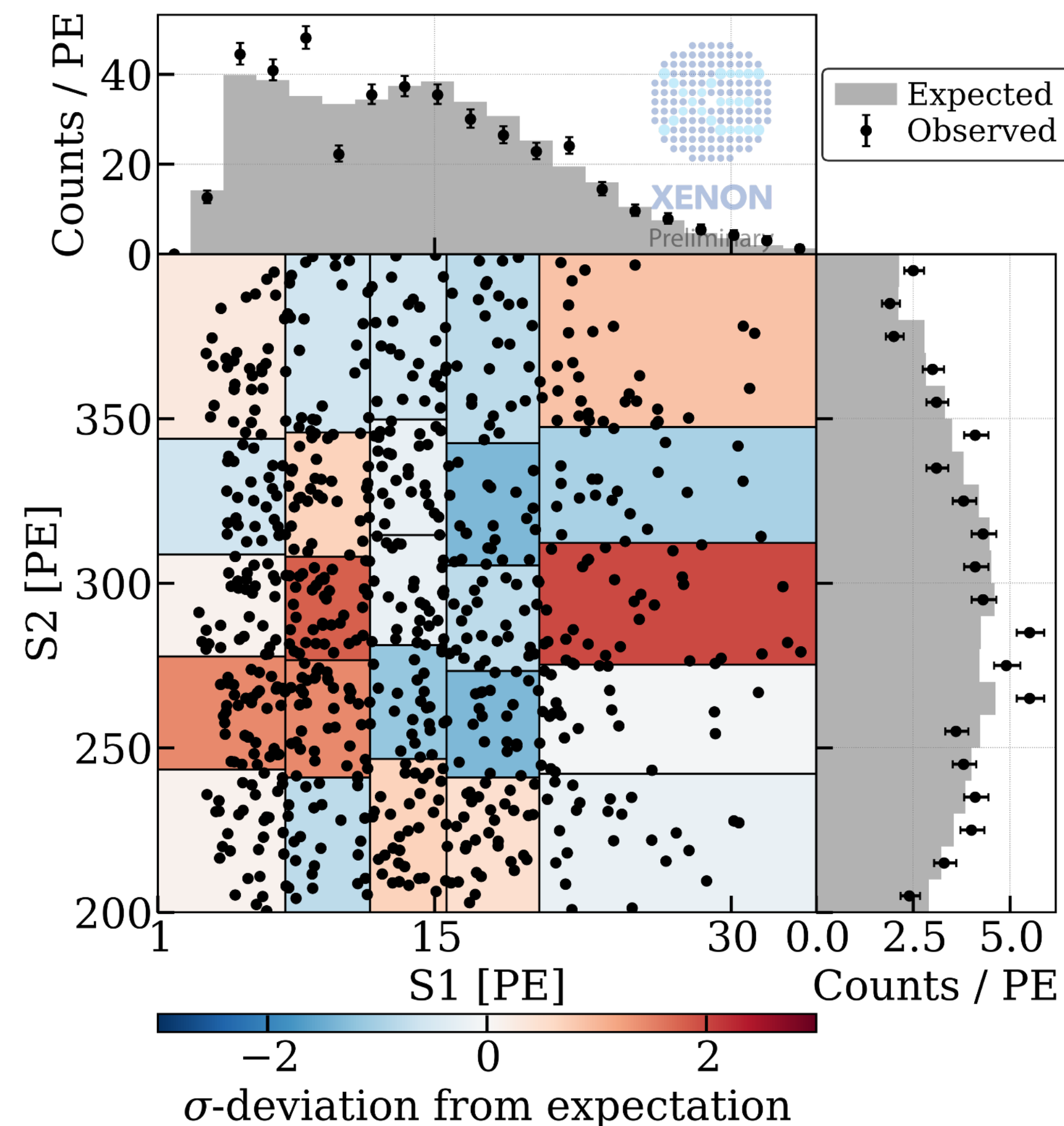
# Find AC in $^{37}\text{Ar}$ datasets

Provide High AC Counts to validate the framework



- Prediction vs. Observation, within **5%** deviation

Dataset	Predicted	Observed
PureAC	1522.7	1459
ACSideband	349.7	366
In-ROI	731.6	733

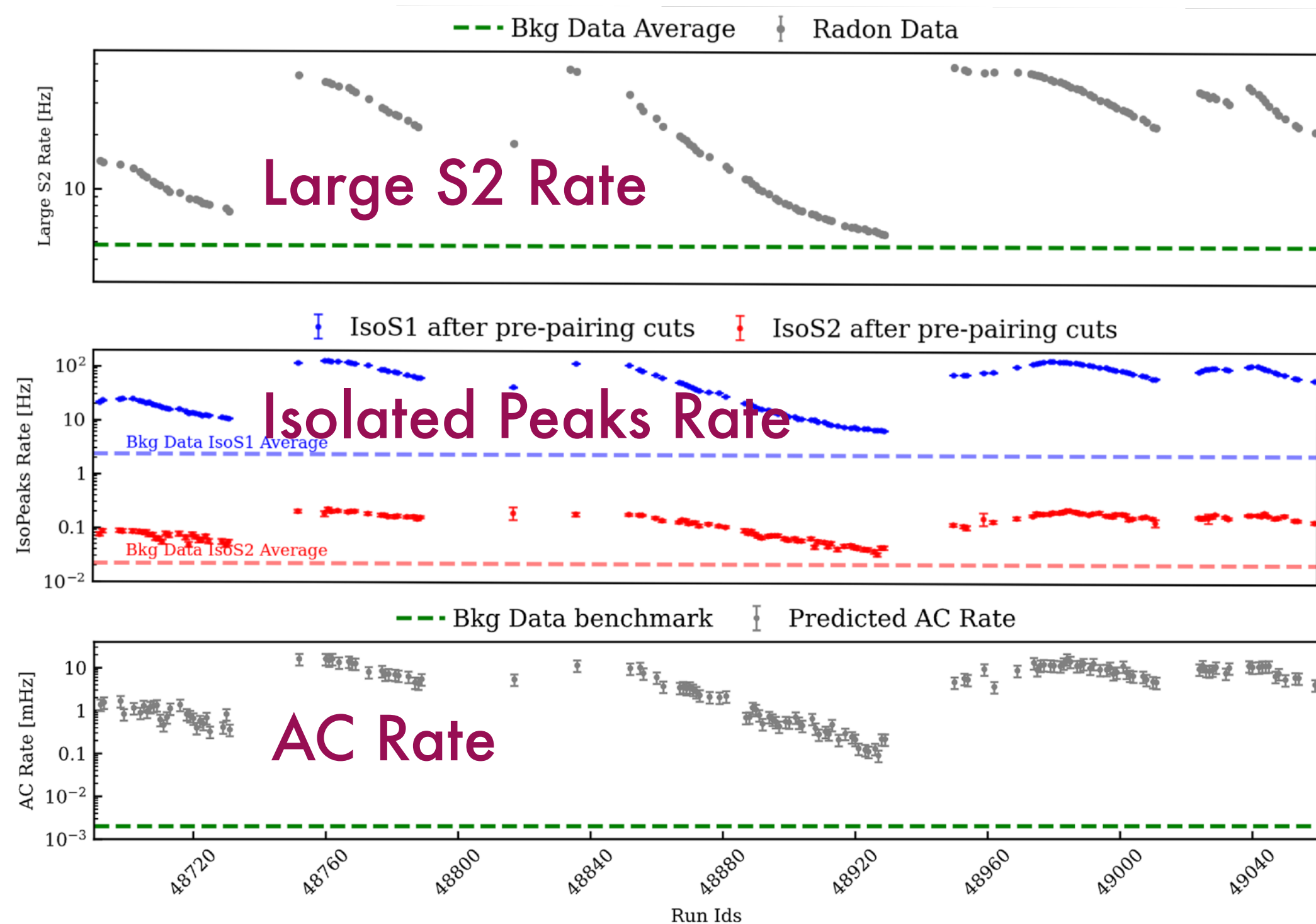






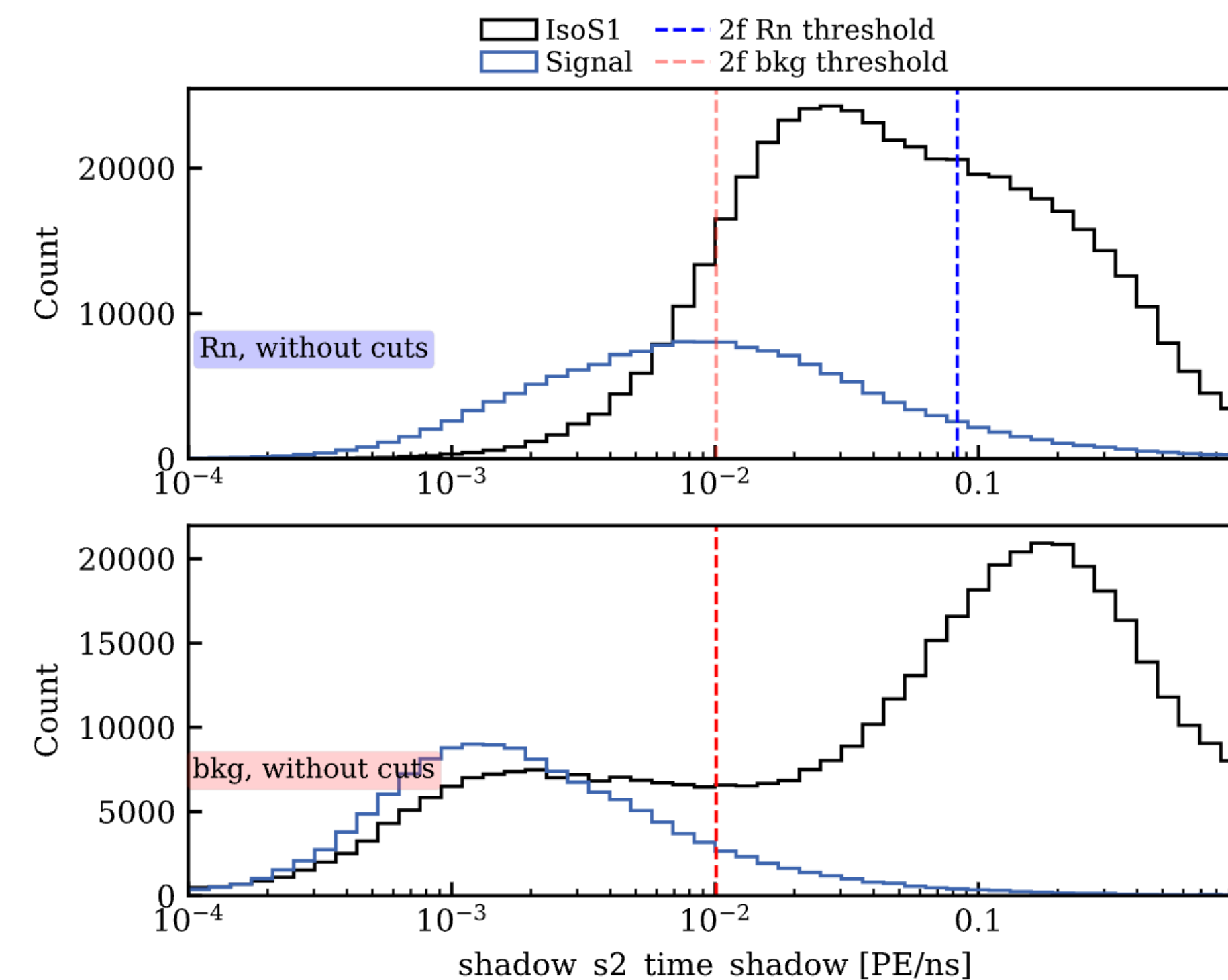
# Validation in $^{220}\text{Rn}$ dataset

## Heavy 'shadow' challenges the AC Model



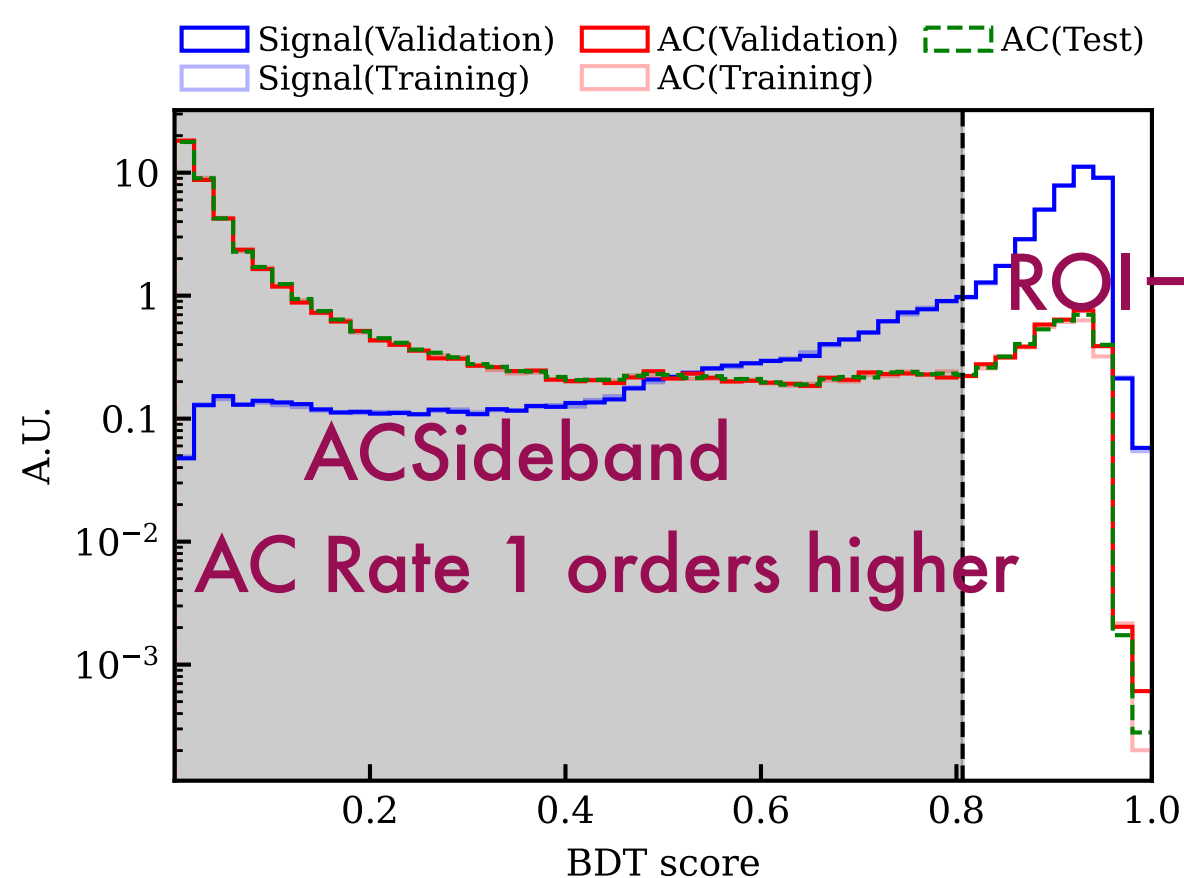
- Featured with high **large peak** rate, thus larger shadow remains after applying cut
- Mis-modeling in S2 in few-electron region(100-120 PE)

Dataset	S2 [PE]	Predicted	Observed
In-ROI	[120, 500]	54.6	57
	[100, 120]	201.0	<b>271</b>



# Validation on Science data ACSideband

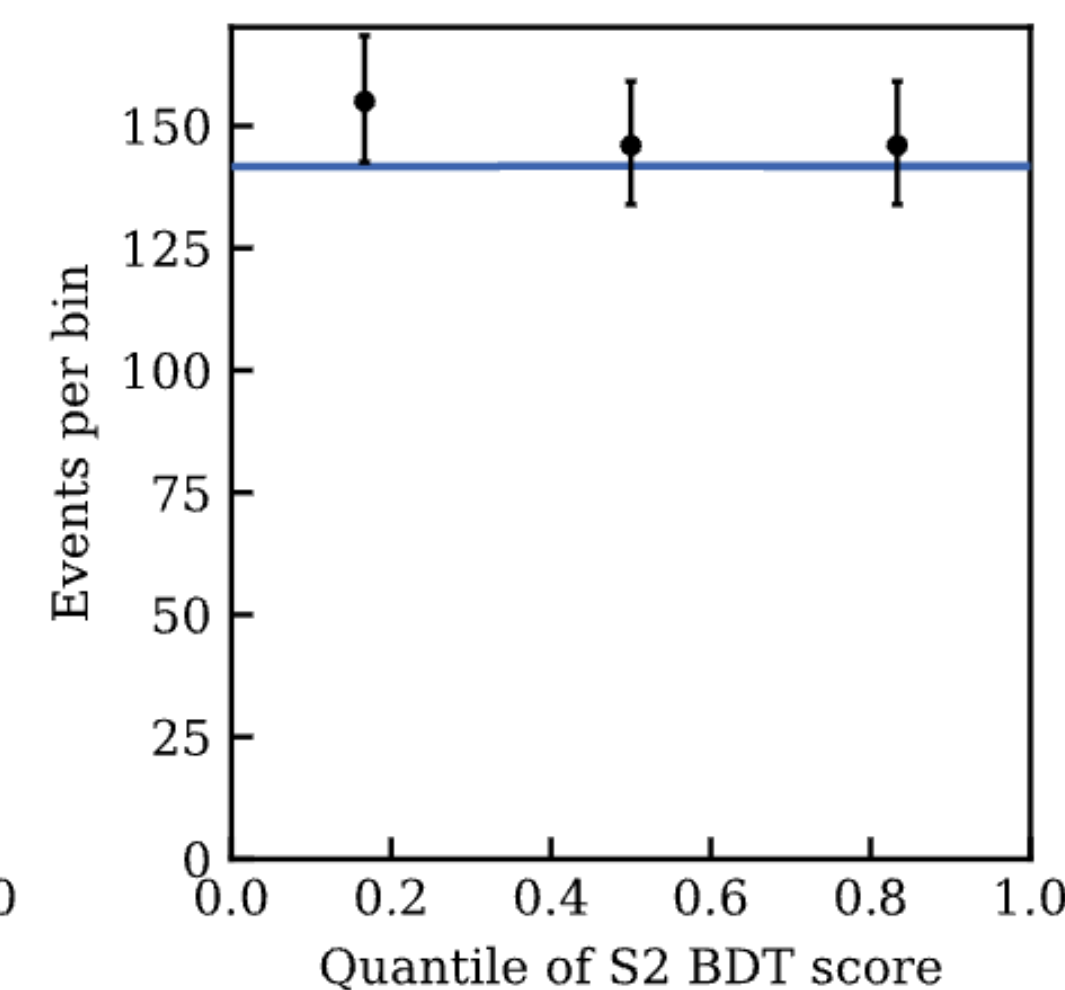
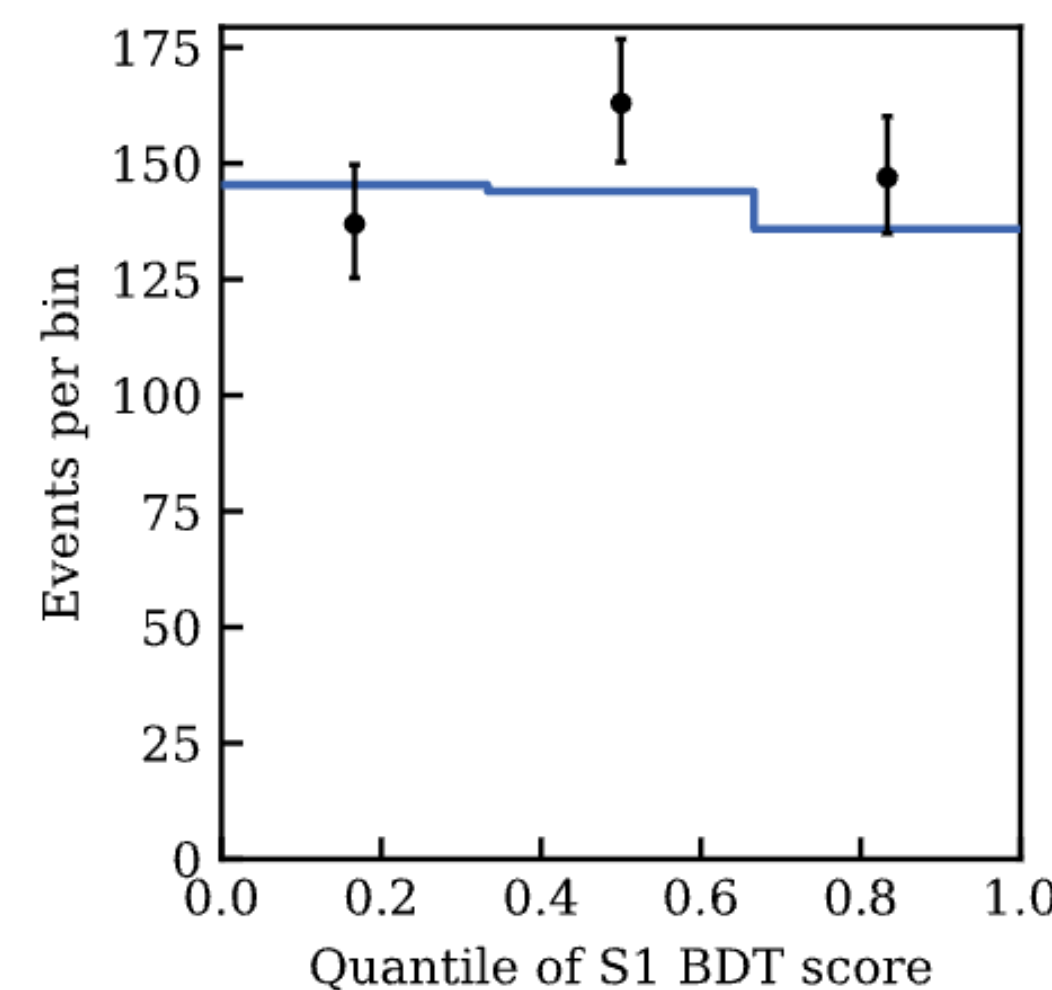
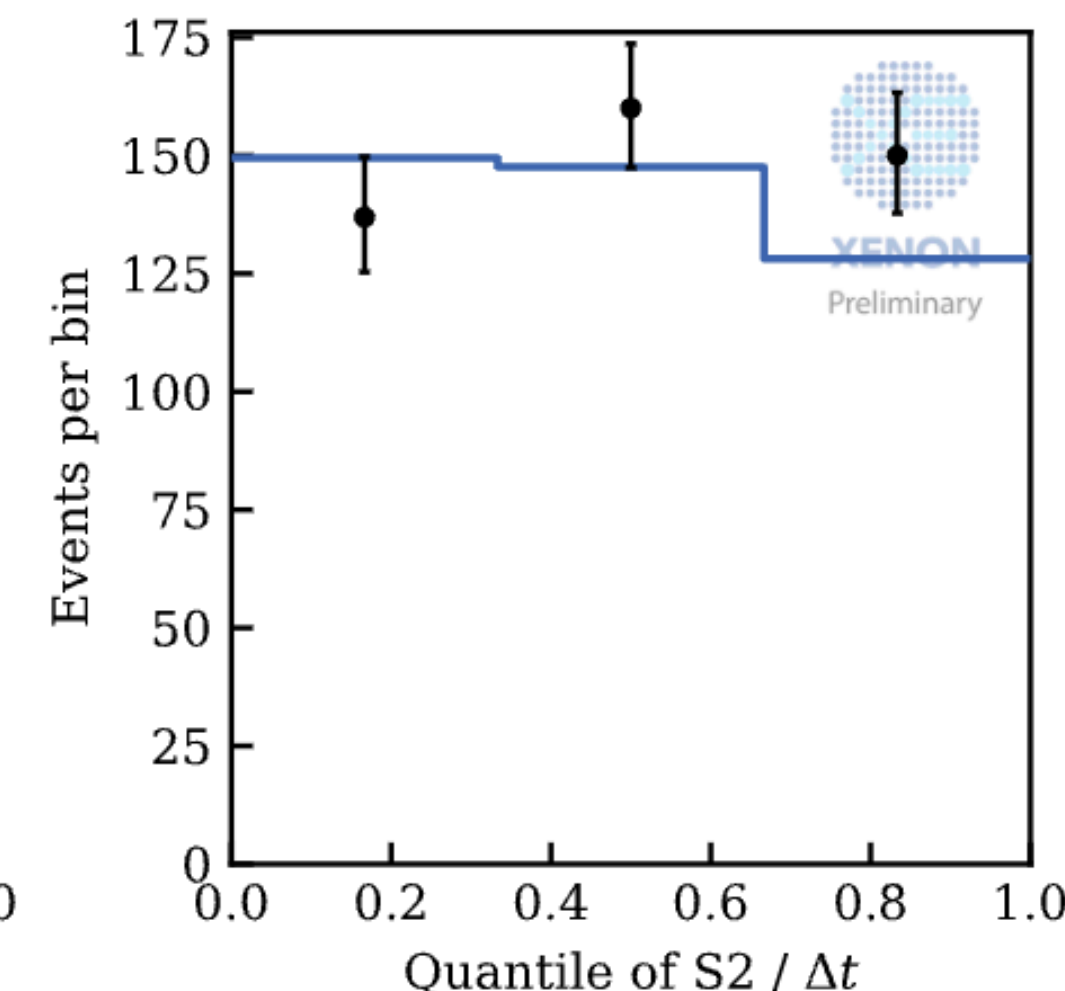
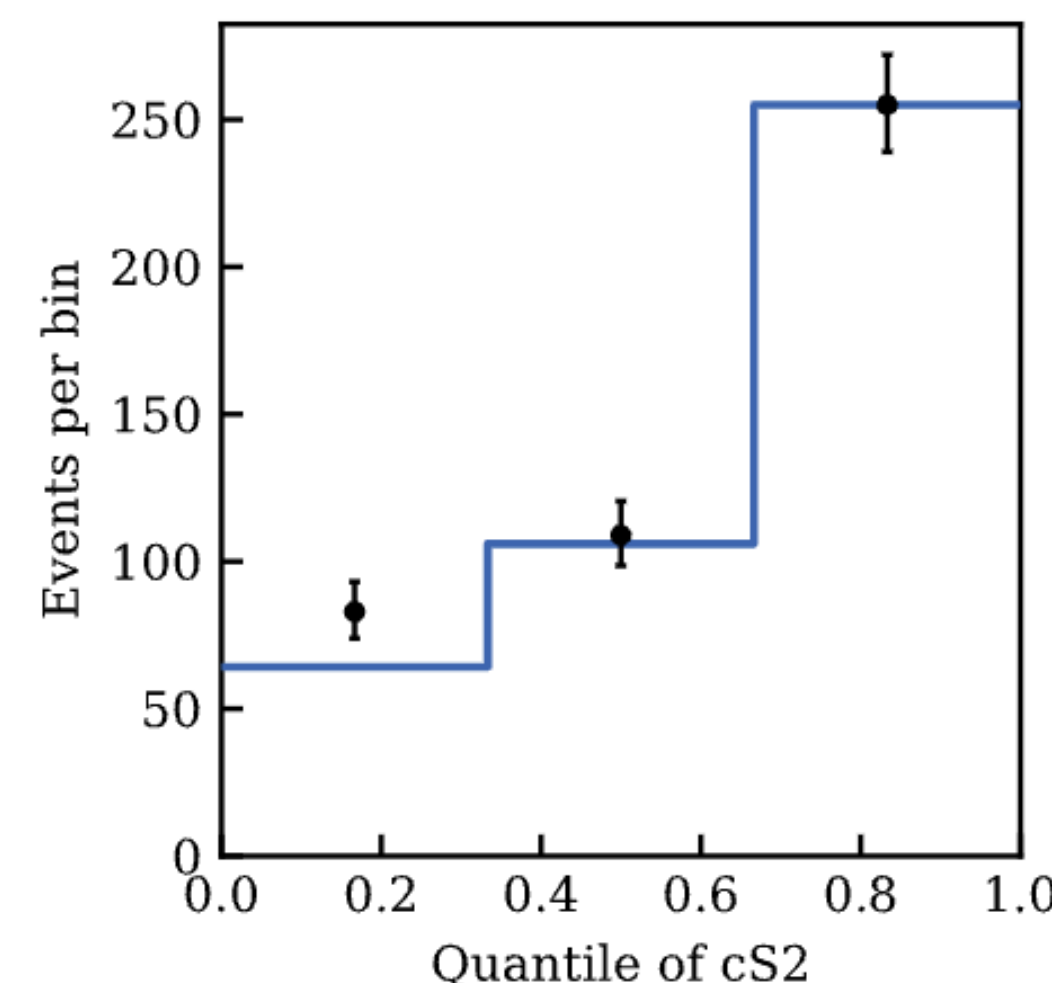
## Determine Systematic Uncertainty



	AC Rate[/t/y]
SR0	6.37
SR1	7.58

- ‘Unblind’ shows within 2-sigma, use the statistic uncertainty of ACSideband to be the systematics
- 4D GoF test: p-value=0.17

Dataset	Predicted	Observed	Relative Uncertainty
SR0	122.7	121	9.04%
SR1	302.5	326	5.76%
Total	425.2	447	/

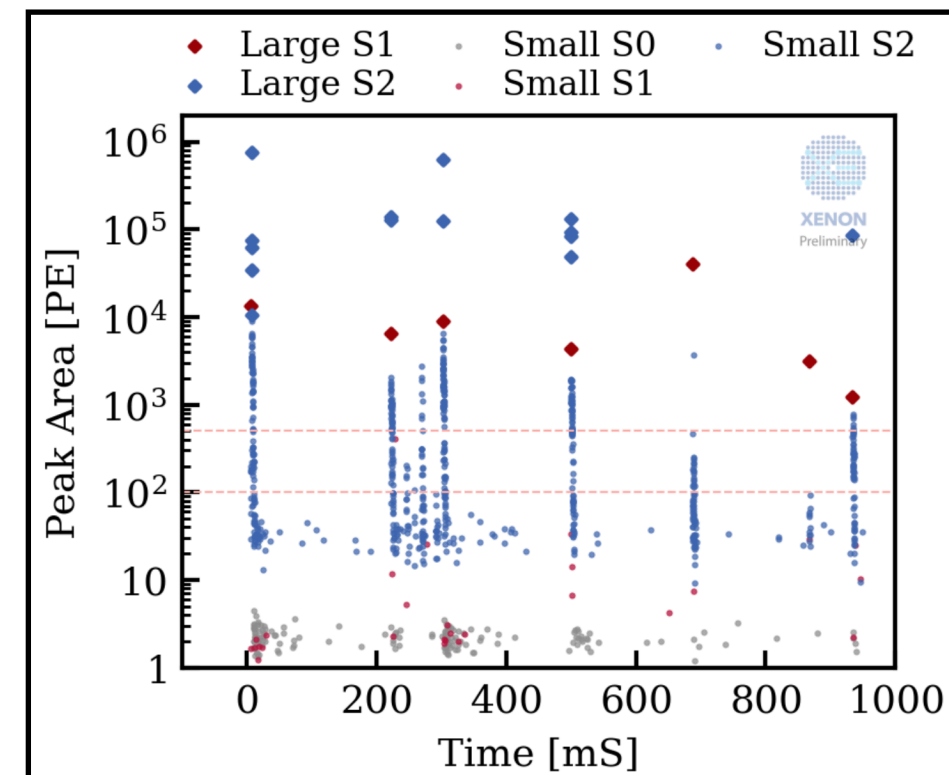




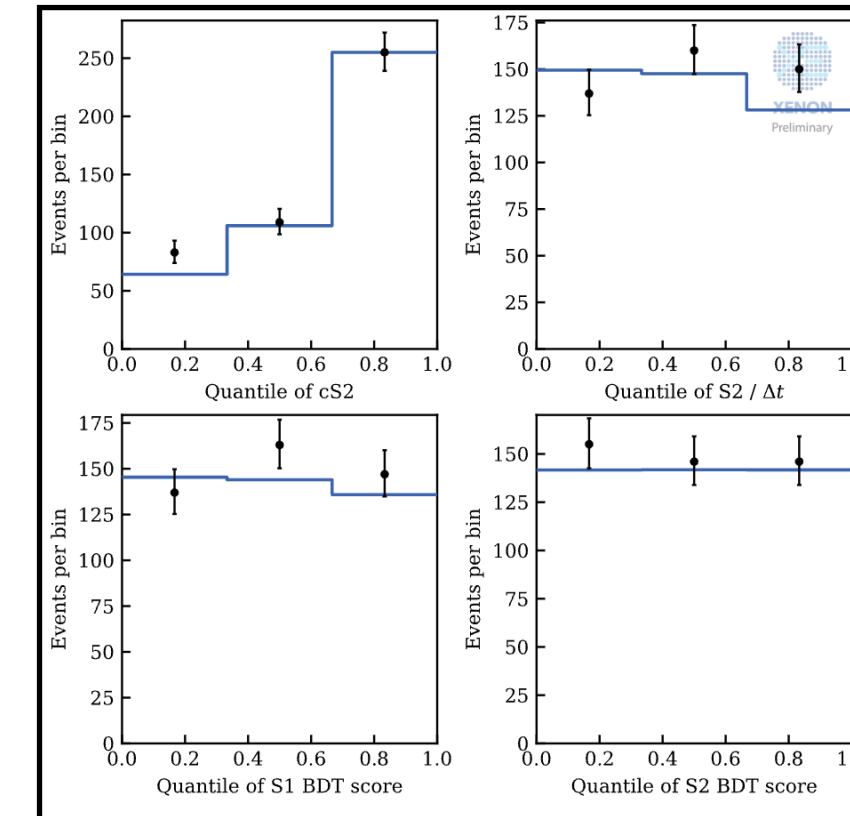
# Summary



From Chaotic Isolated Peaks



To reduced & validated AC model



- **A**ccidental **C**oincidence background is the dominant background in low energy NR search with a two-phase LXe TPC.
- What we can do with the AC background:
  - **S**uppress isolated peaks by time+space correlation, and suppress AC events by comprehensively use features which have discrimination power.
  - **P**redict the rate and distribution in high dimensions with data-driven simulation.
  - **V**alidate the model with real data to set the systematics uncertainty

Thanks!

