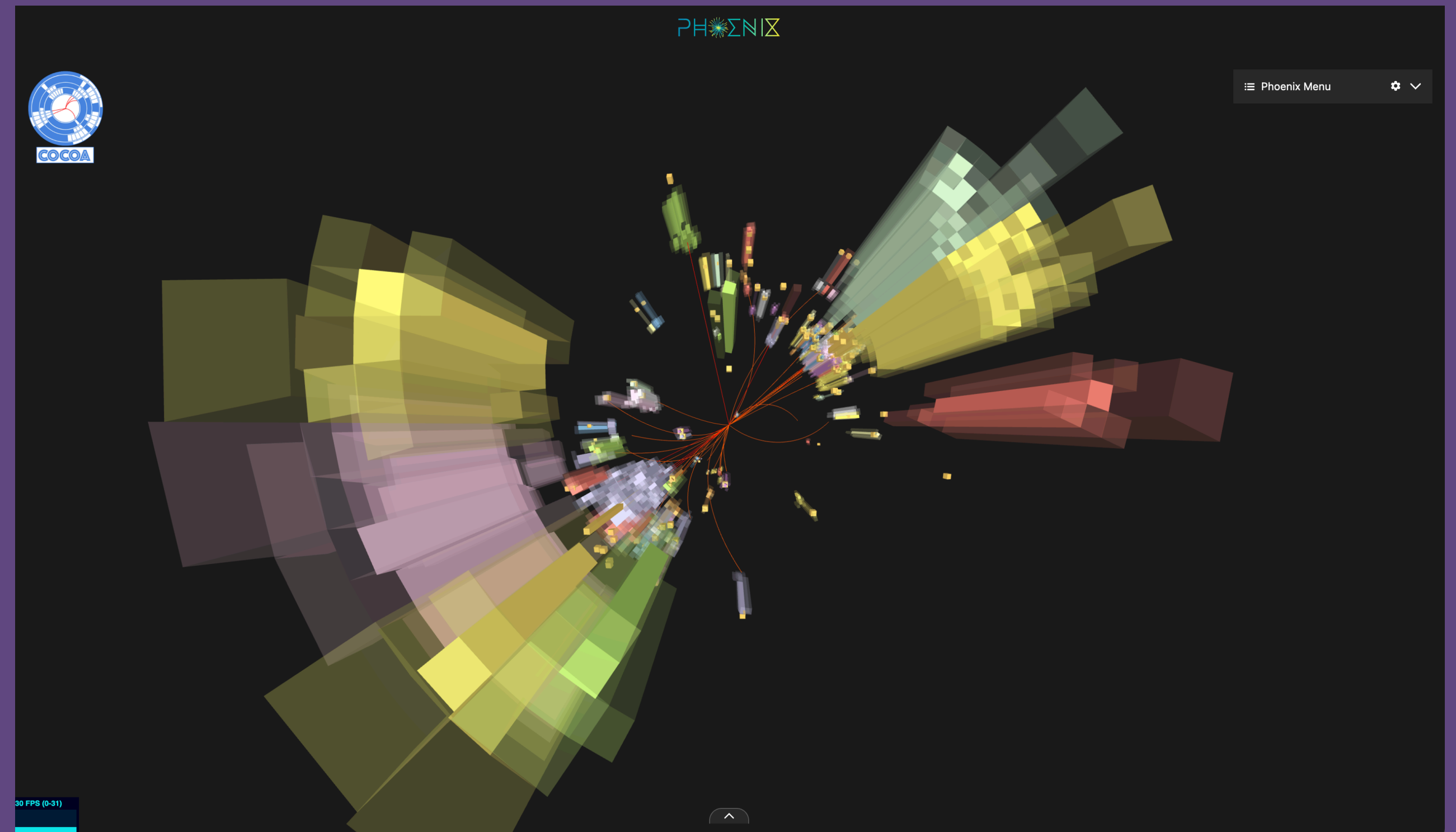


De-noising Graph Super Resolution with Diffusion Models and transformers



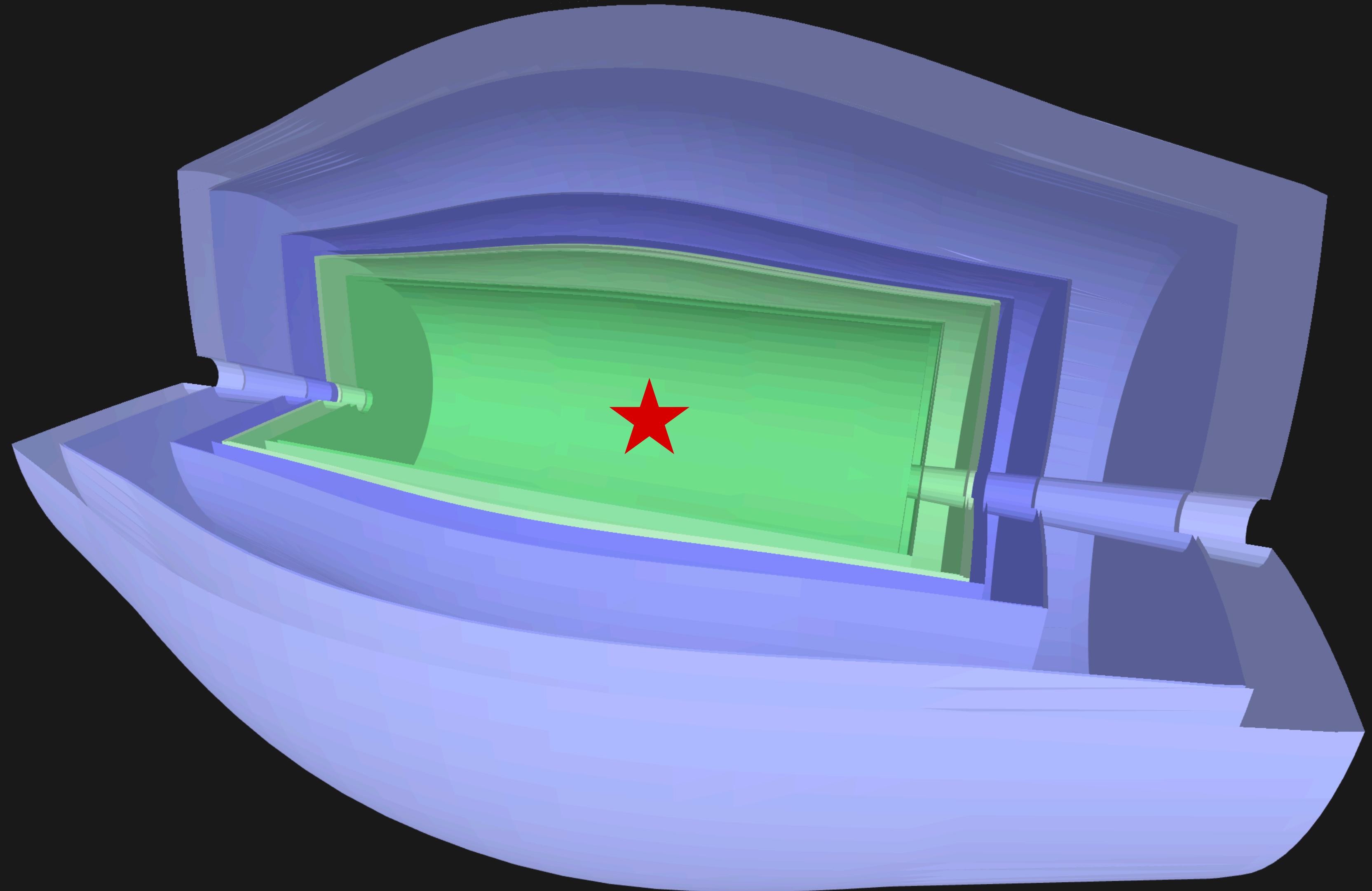
ML4Jets

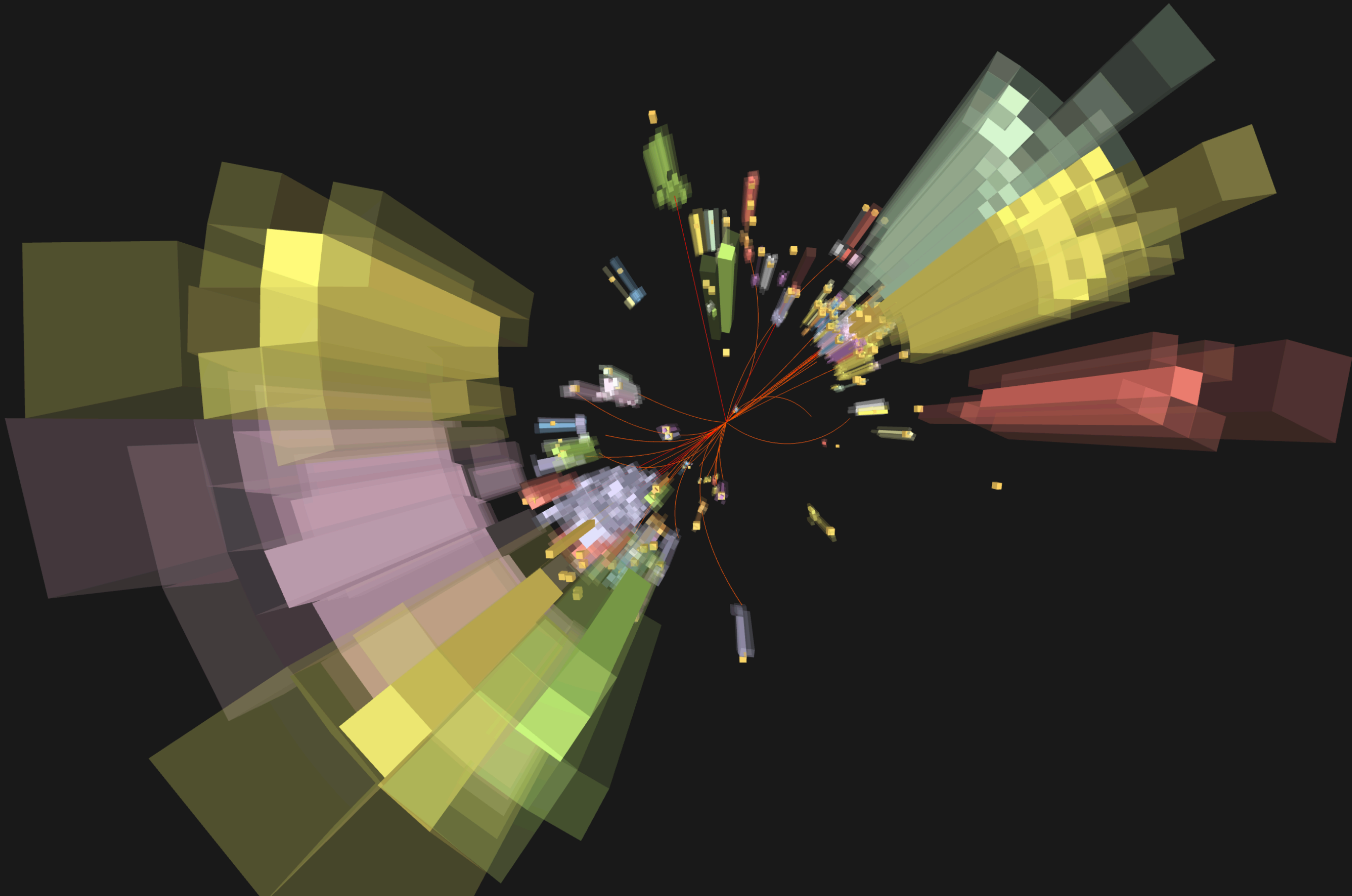
06 November, 2023

Super Resolution with Diffusion

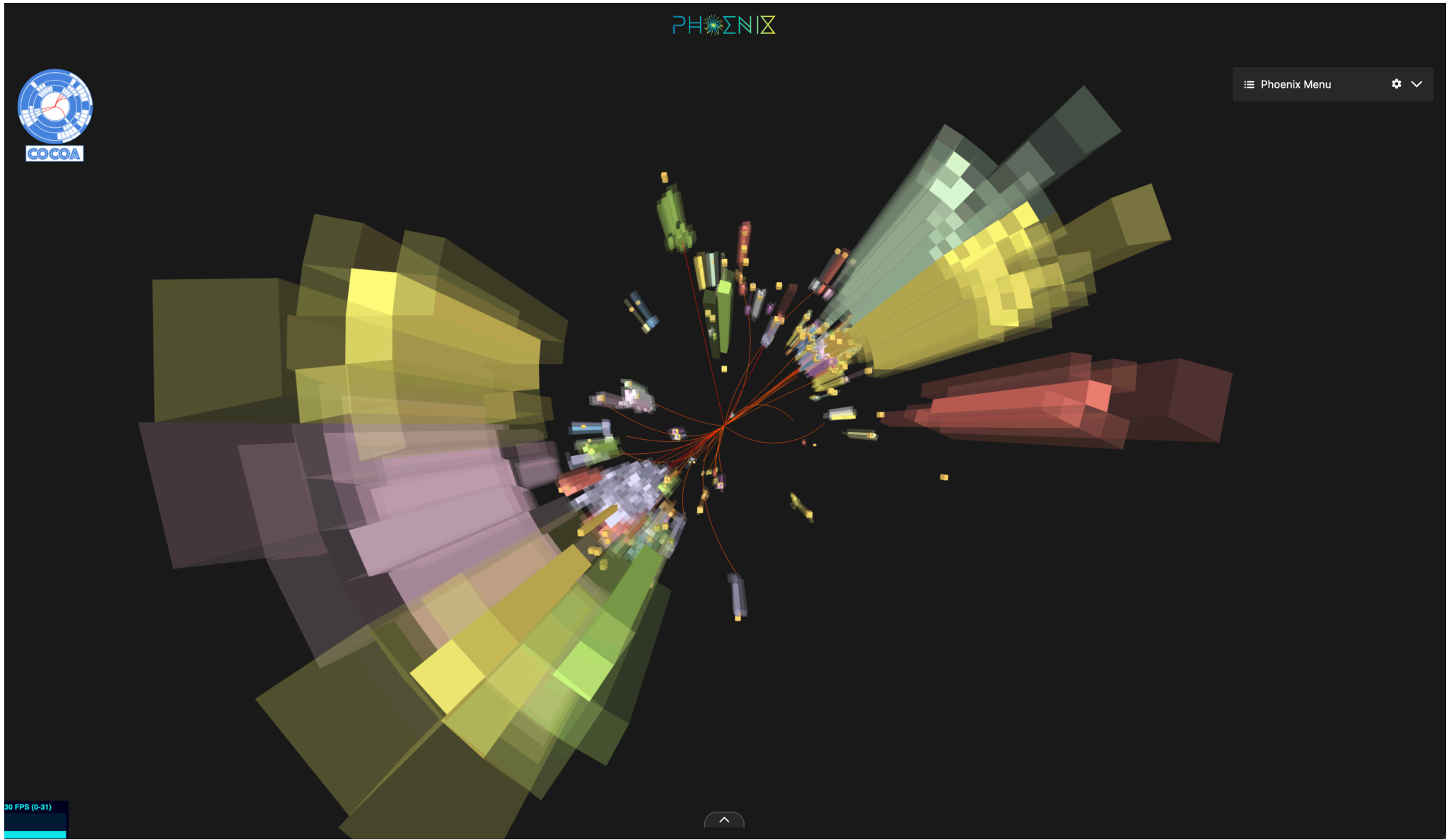
- ◆ Slightly less popular cousin of “text to image with Diffusion”
 - ➔ Still quite popular in CV
 - ➔ Not really studied in Particle Physics





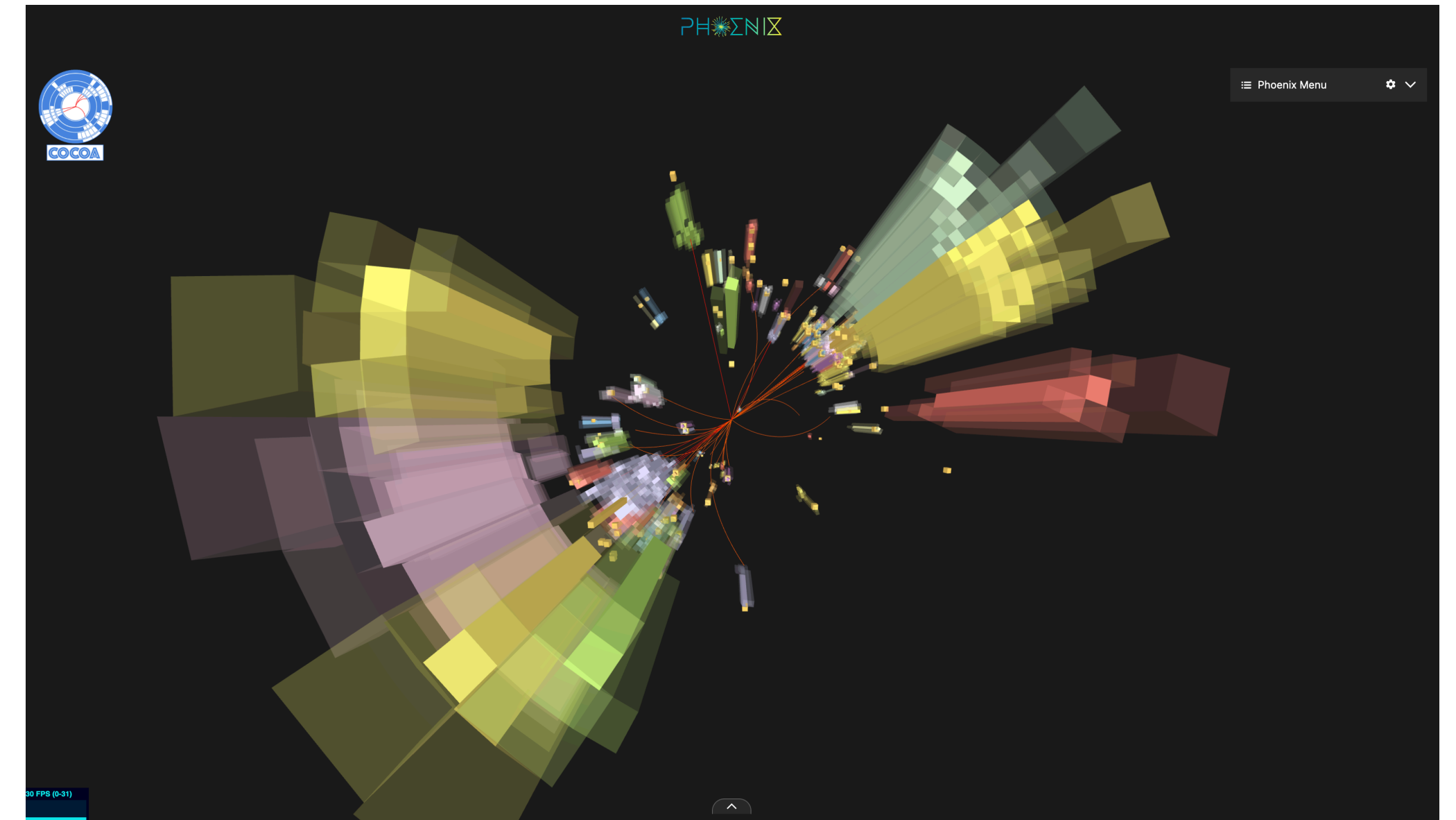


Why Super Resolution in Particle Physics?



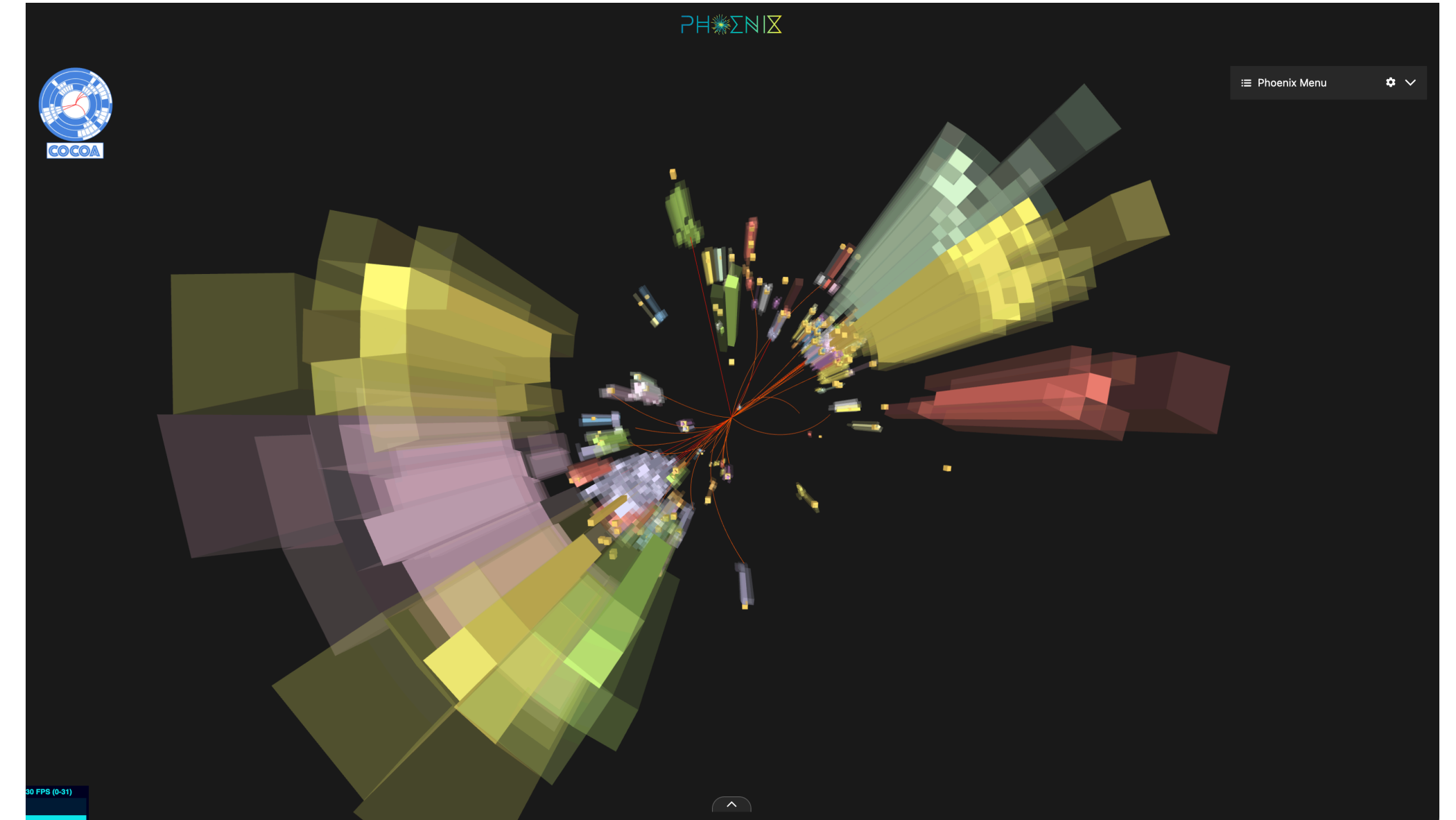
Why Super Resolution in Particle Physics?

- ◆ Reconstruction quality depends on the detector granularity
 - ➔ More granular -> better reconstruction
 - ➔ Granularity puts a cap on theoretical reconstruction capability



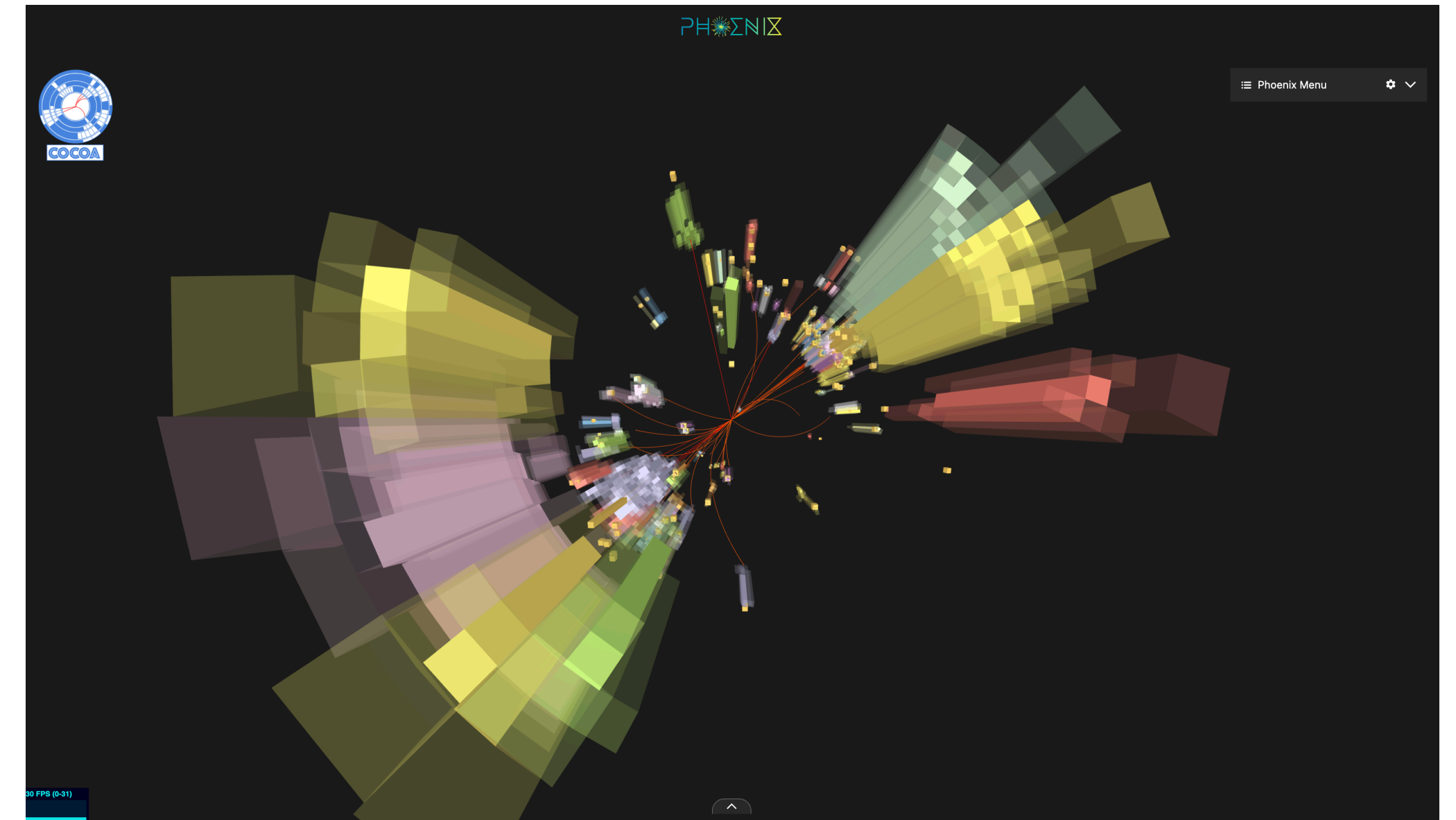
Why Super Resolution in Particle Physics?

- ◆ Reconstruction quality depends on the detector granularity
 - ➔ More granular -> better reconstruction
 - ➔ Granularity puts a cap on theoretical reconstruction capability
- ◆ High granularity detectors (simulations) are very expensive!
 - ➔ Increasing resolution in post can be a solution!



Why Super Resolution in Particle Physics?

- ◆ Reconstruction quality depends on the detector granularity
 - ➔ More granular -> better reconstruction
 - ➔ Granularity puts a cap on theoretical reconstruction capability
- ◆ High granularity detectors (simulations) are very expensive!
 - ➔ Increasing resolution in post can be a solution!
- ◆ Graph super resolution is not a common problem in general
 - ➔ Graphs are very natural in Particle Physics
 - ➔ Hence Graph Super resolution

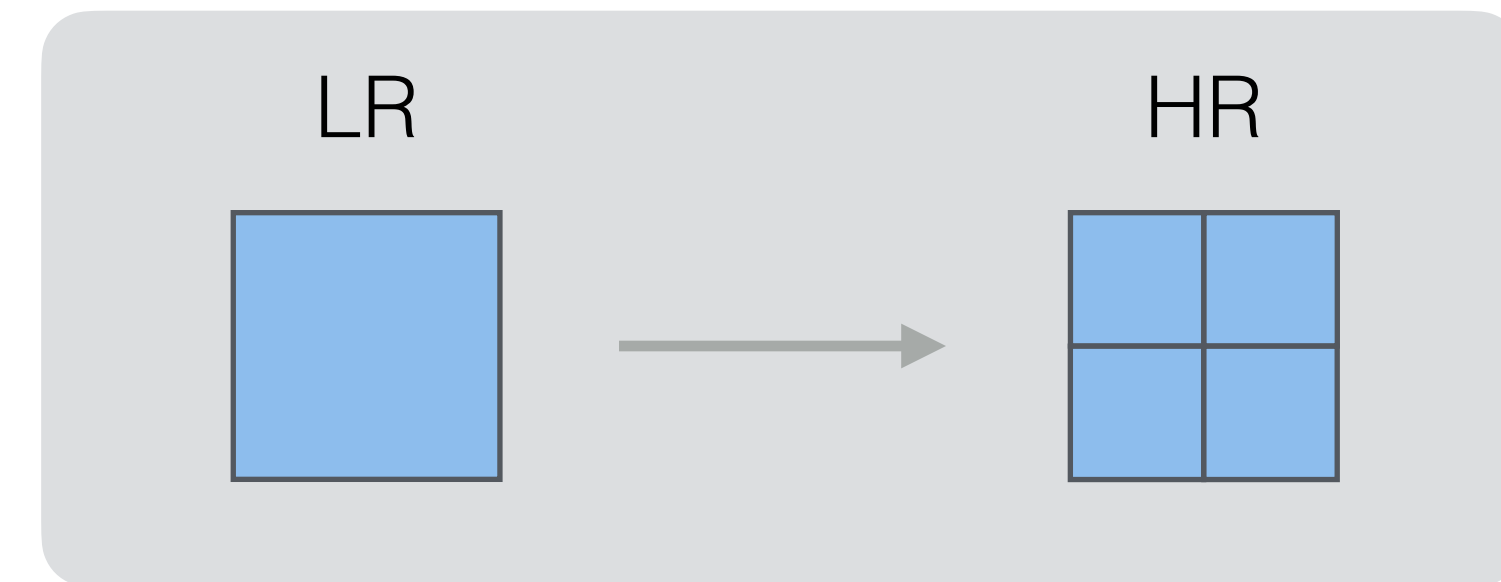


The SetUp

- ◆ COCOA mod (<https://iopscience.iop.org/article/10.1088/2632-2153/acf186/pdf>)
- ◆ Shooting single electron as a starting point

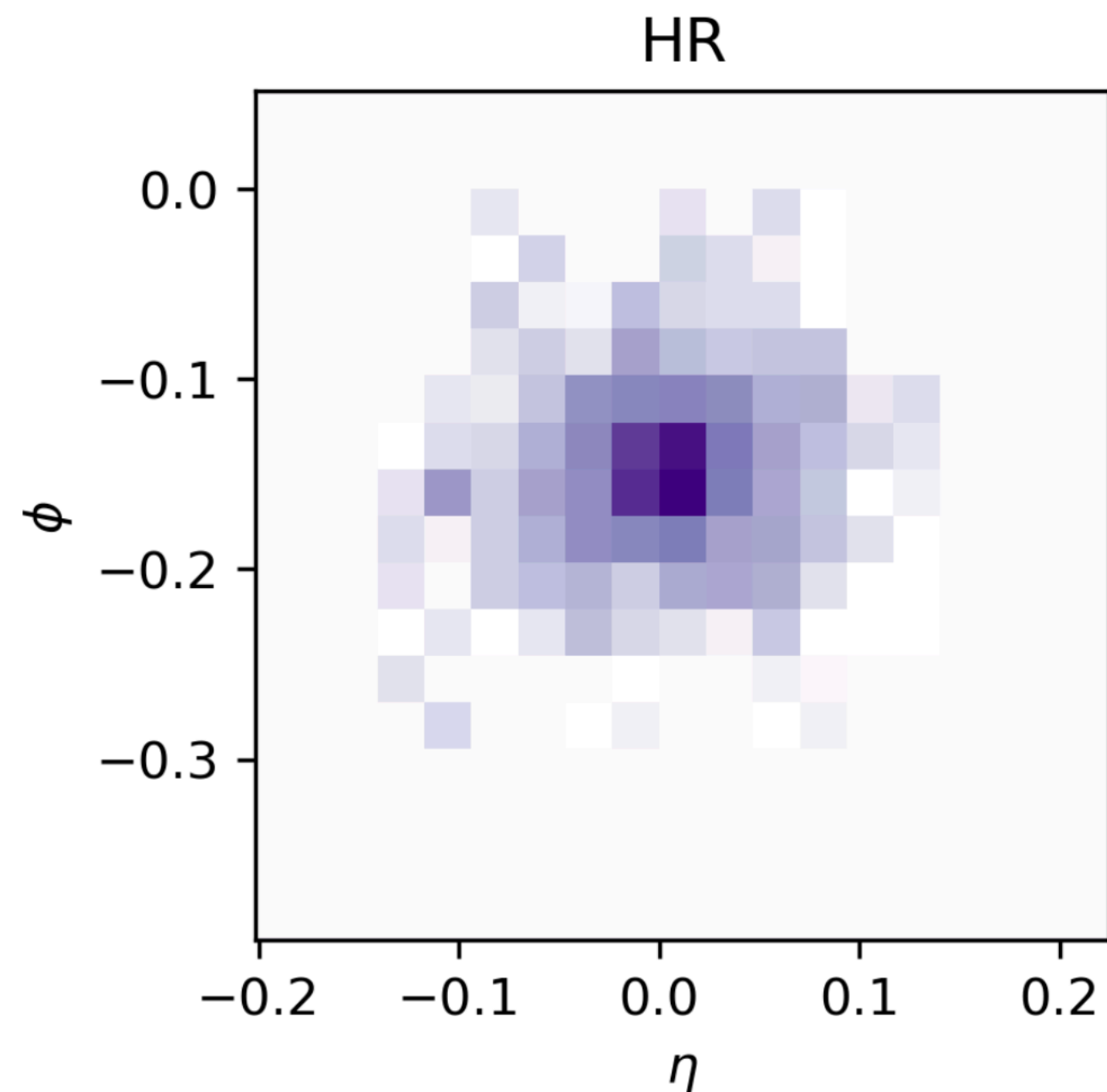
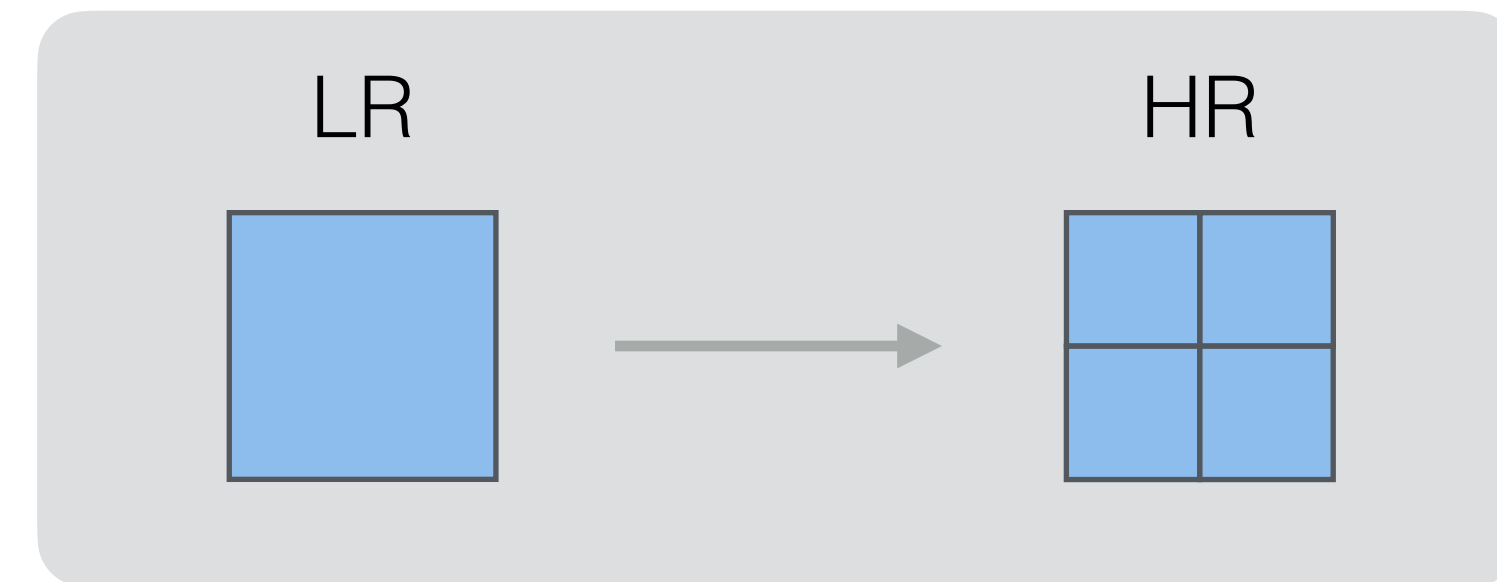
The SetUp

- ◆ COCOA mod (<https://iopscience.iop.org/article/10.1088/2632-2153/acf186/pdf>)
- ◆ Shooting single electron as a starting point



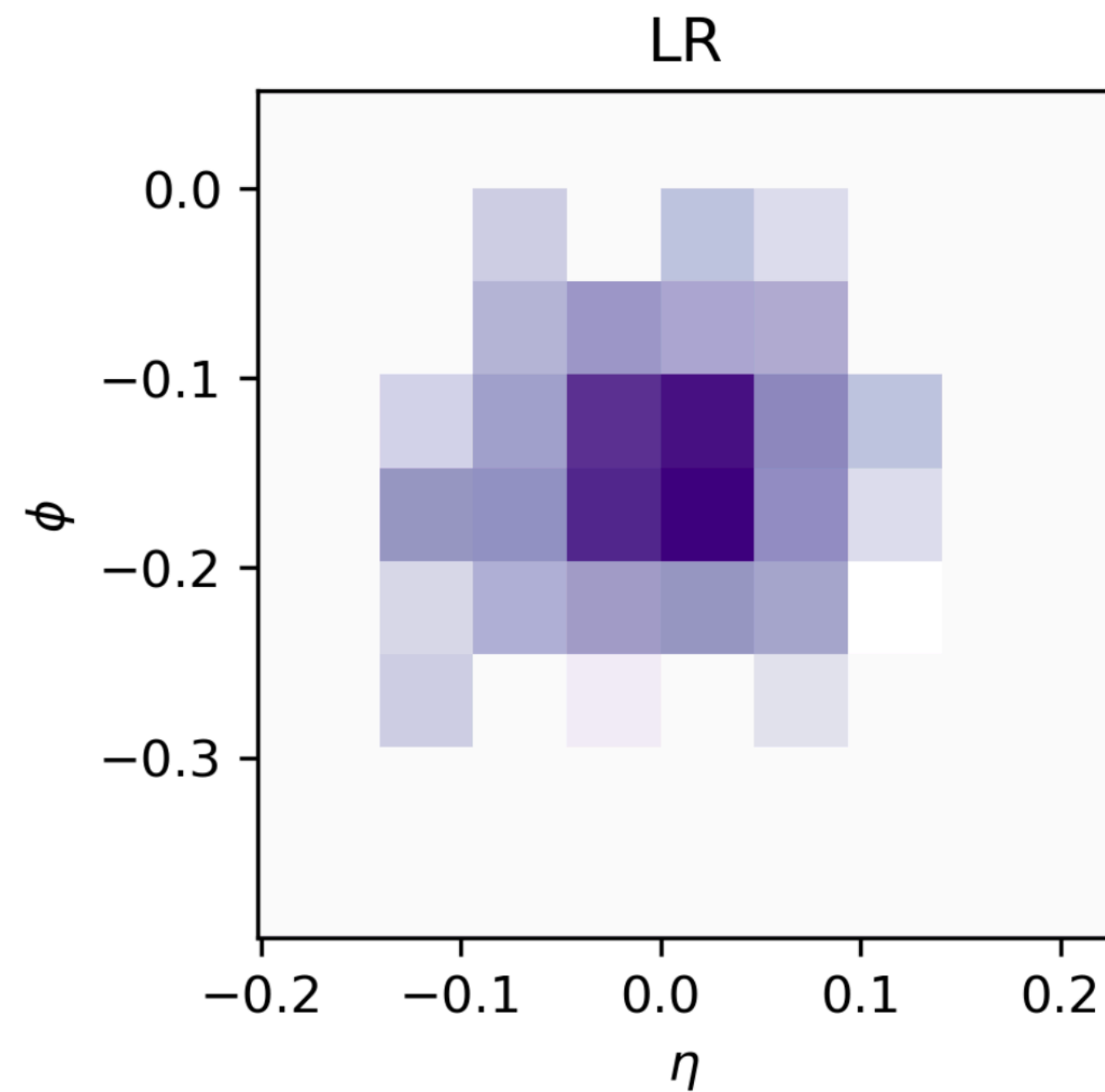
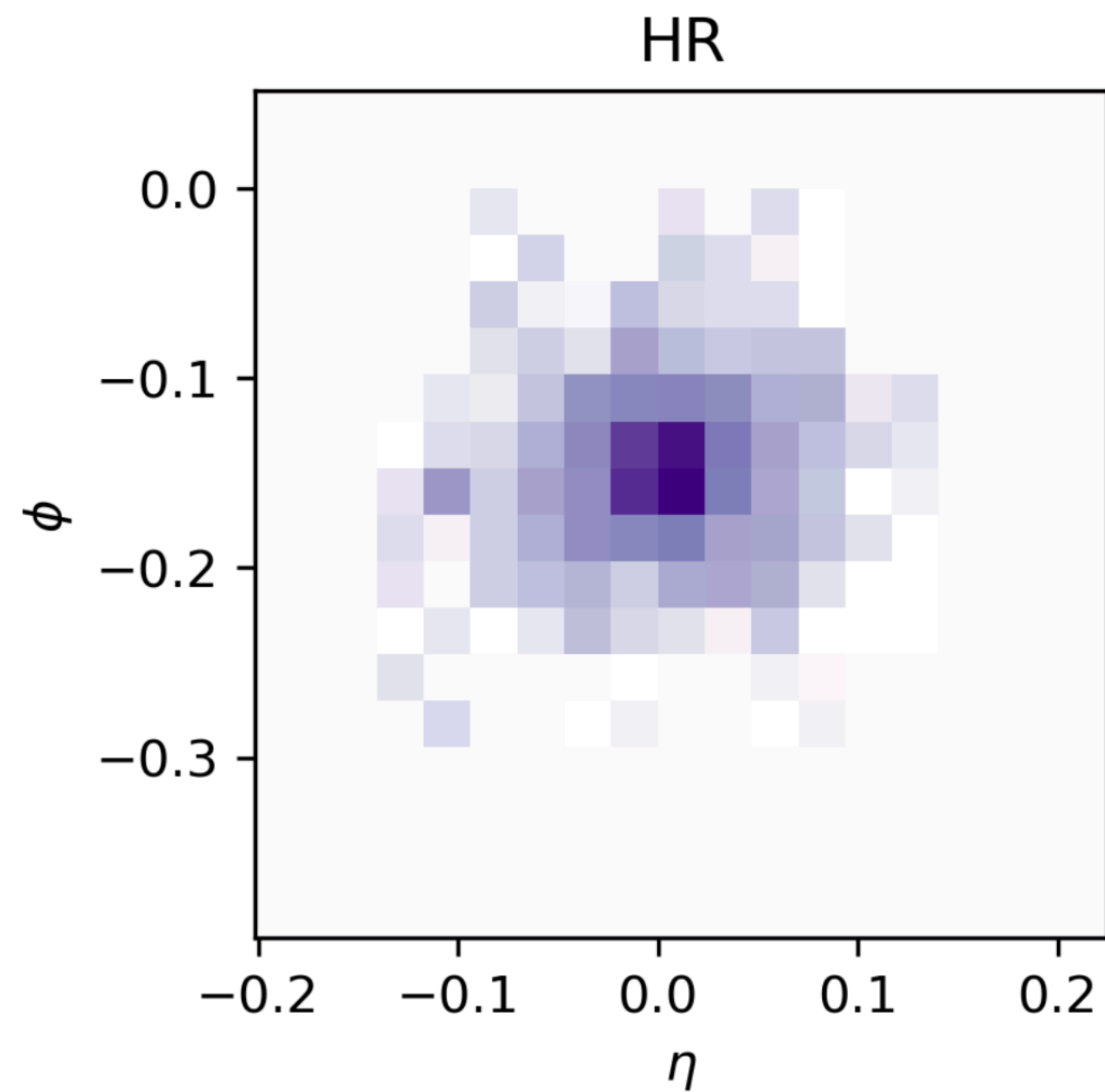
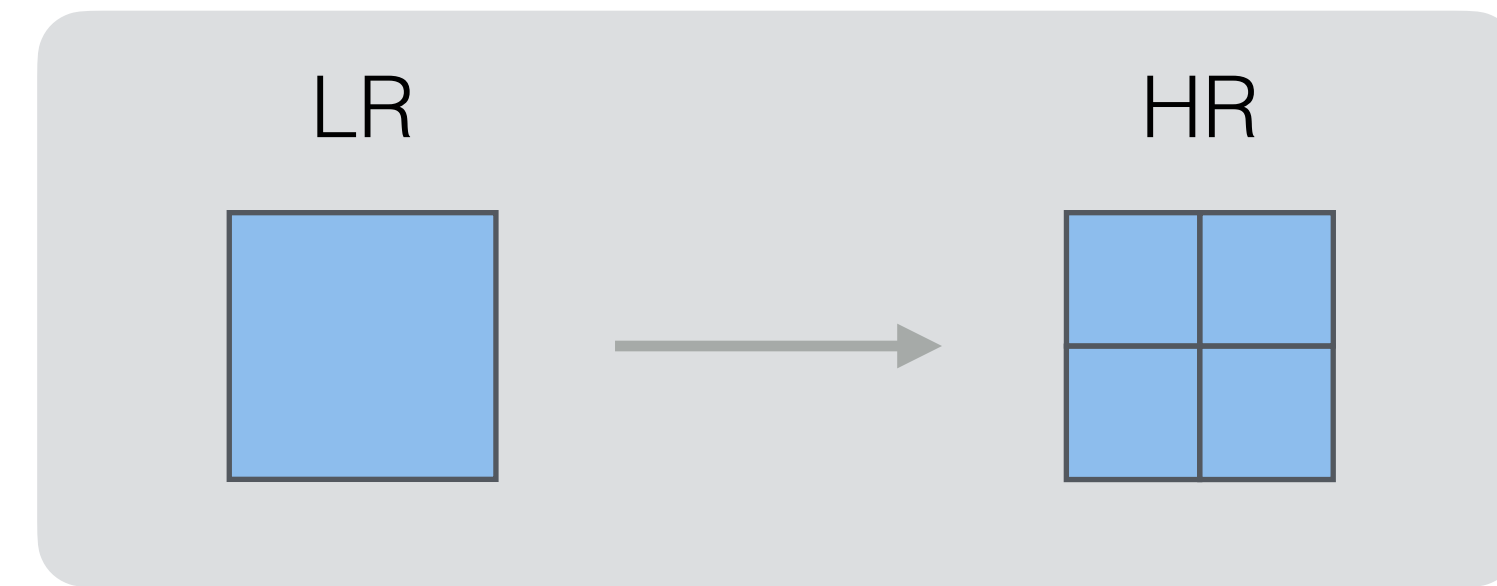
The SetUp

- ◆ COCOA mod (<https://iopscience.iop.org/article/10.1088/2632-2153/acf186/pdf>)
- ◆ Shooting single electron as a starting point



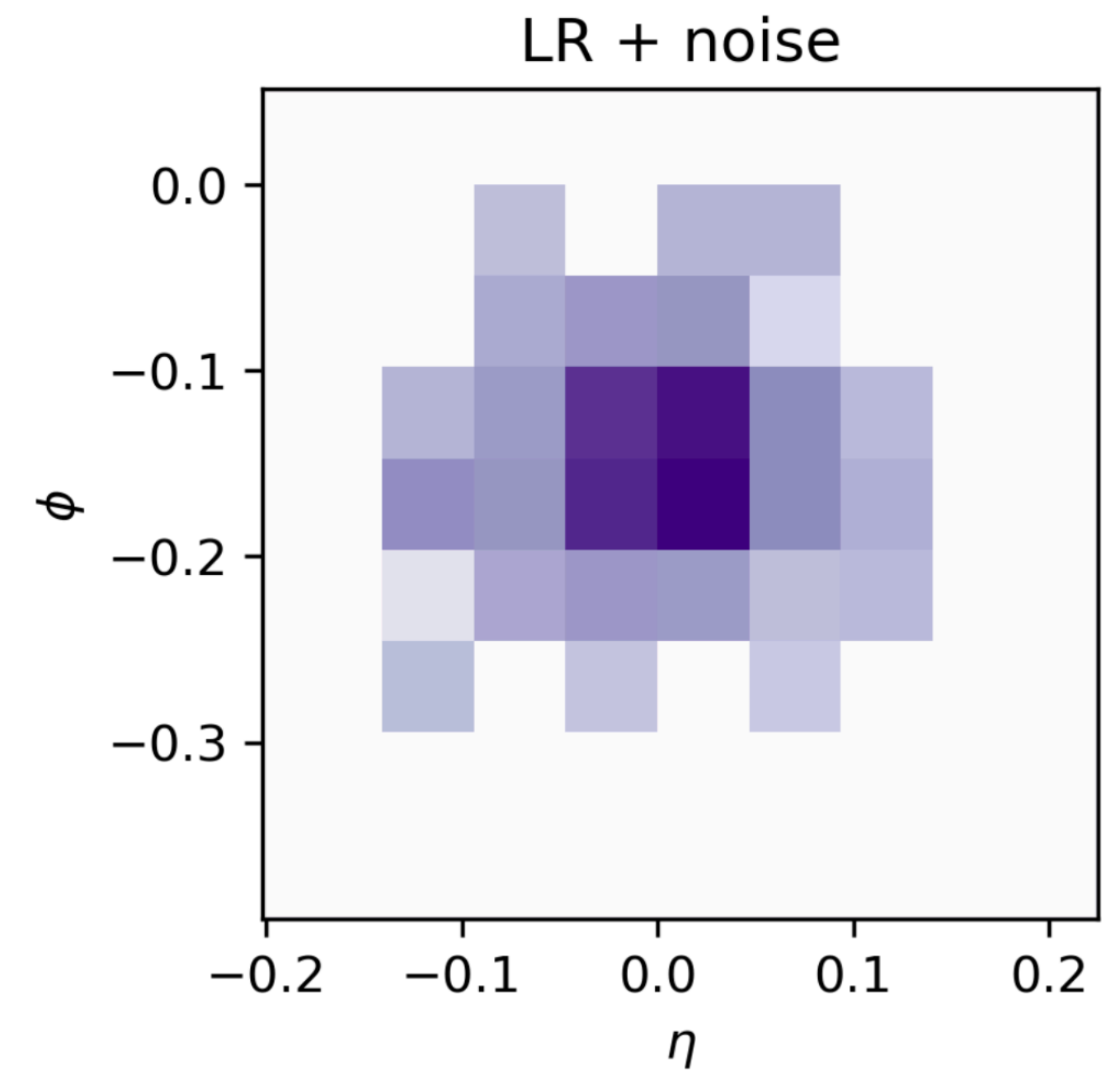
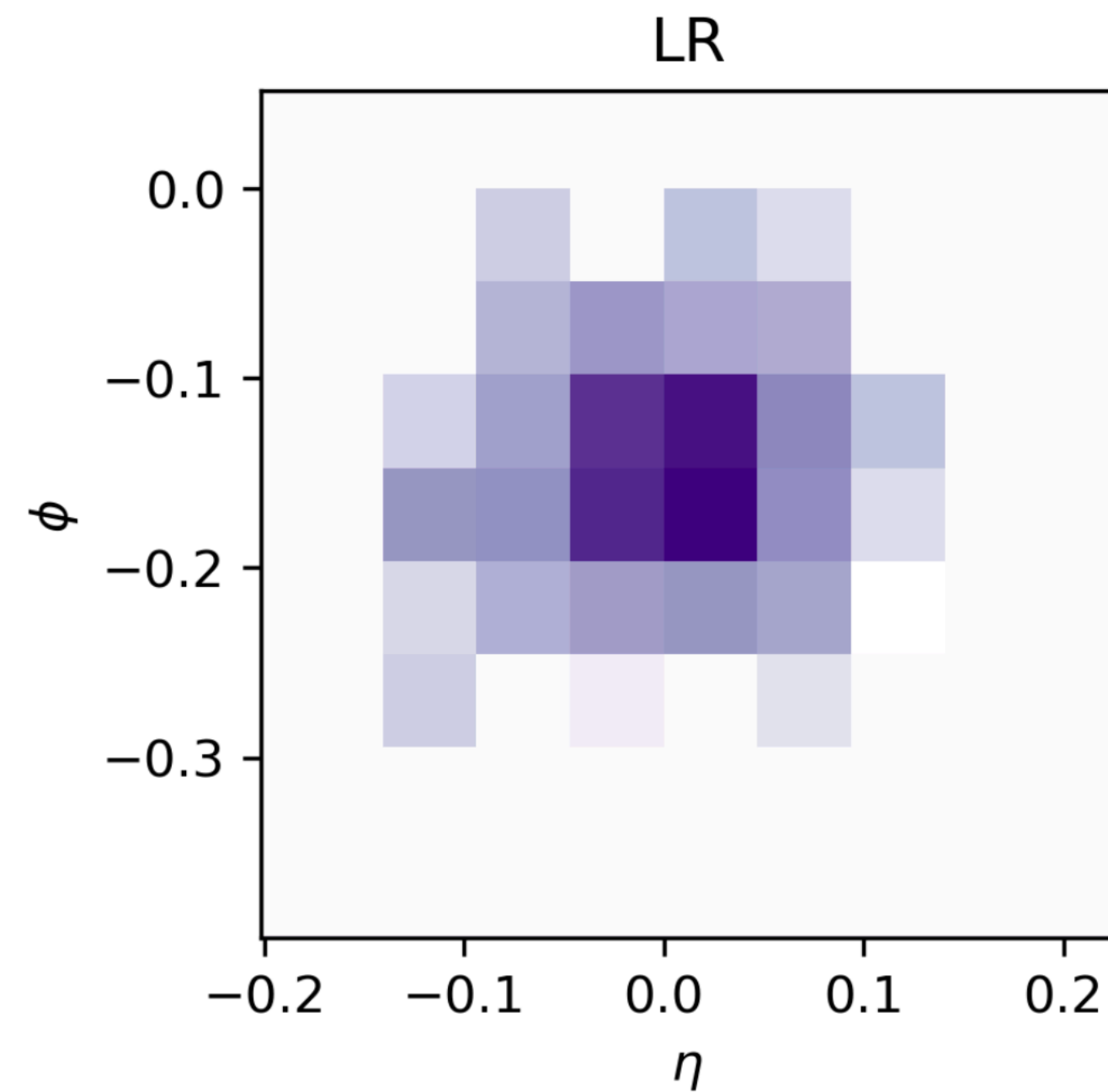
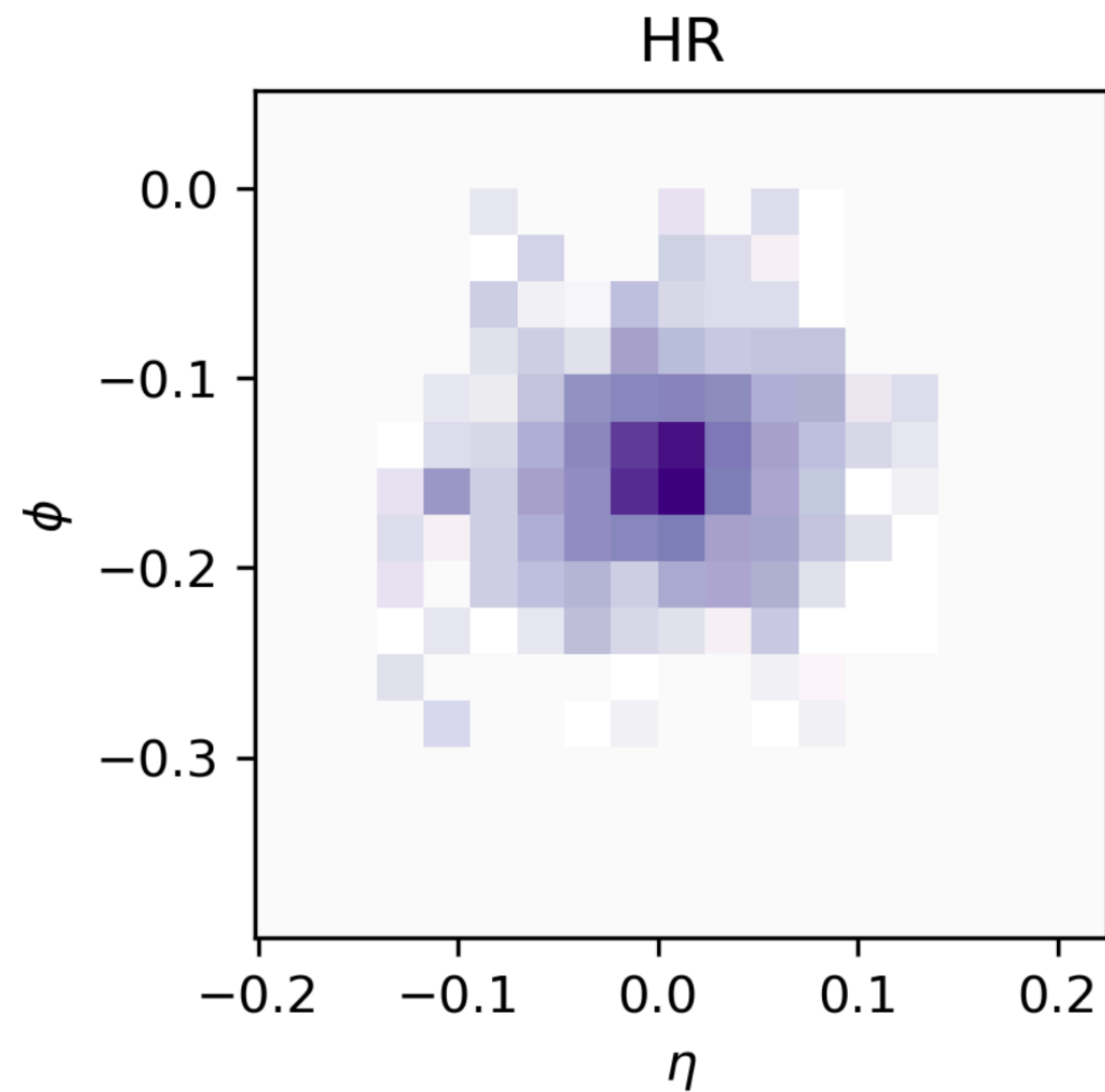
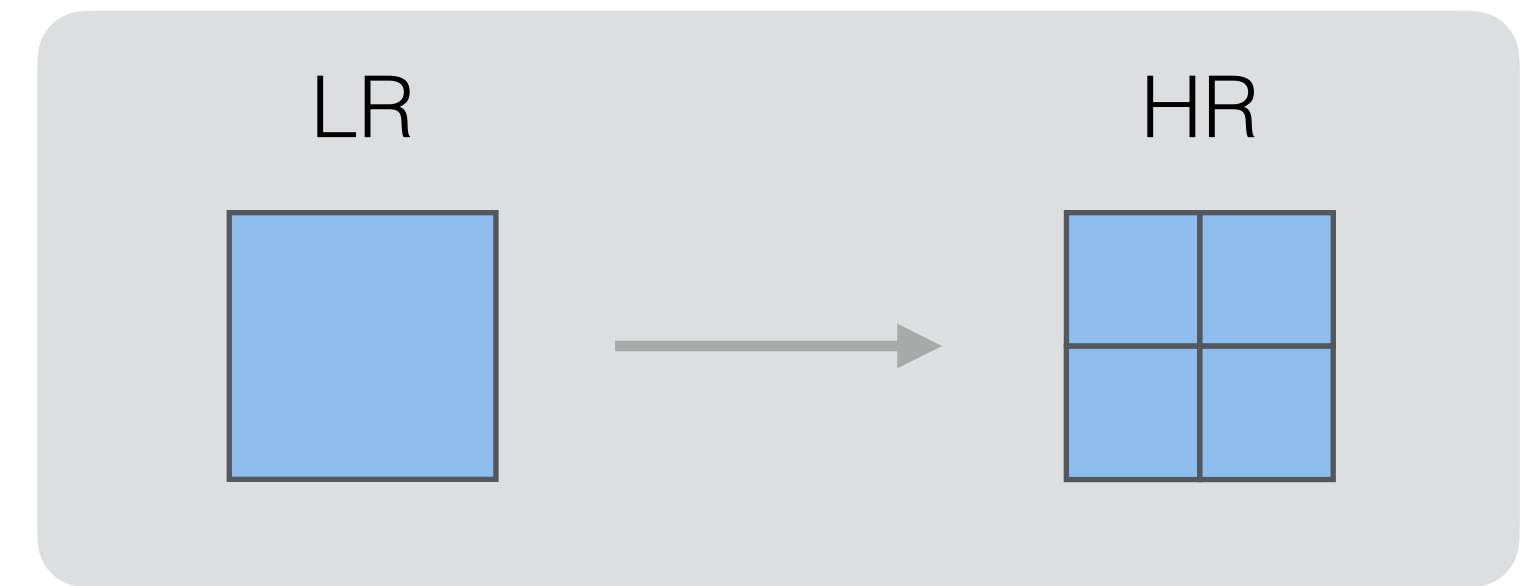
The SetUp

- ◆ COCOA mod (<https://iopscience.iop.org/article/10.1088/2632-2153/acf186/pdf>)
- ◆ Shooting single electron as a starting point



The SetUp

- ◆ COCOA mod (<https://iopscience.iop.org/article/10.1088/2632-2153/acf186/pdf>)
- ◆ Shooting single electron as a starting point



Diffusion set up

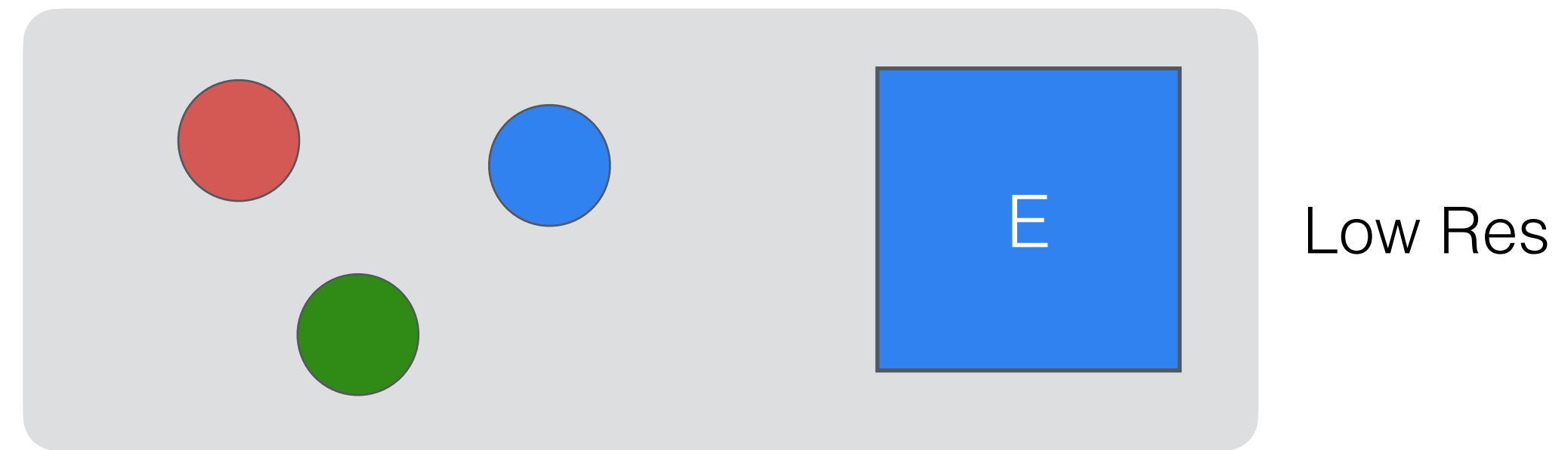
- ◆ Inspired by the SR3 paper

Image Super-Resolution via Iterative Refinement (<https://arxiv.org/pdf/2104.07636.pdf>)

Diffusion set up

- ◆ Inspired by the SR3 paper

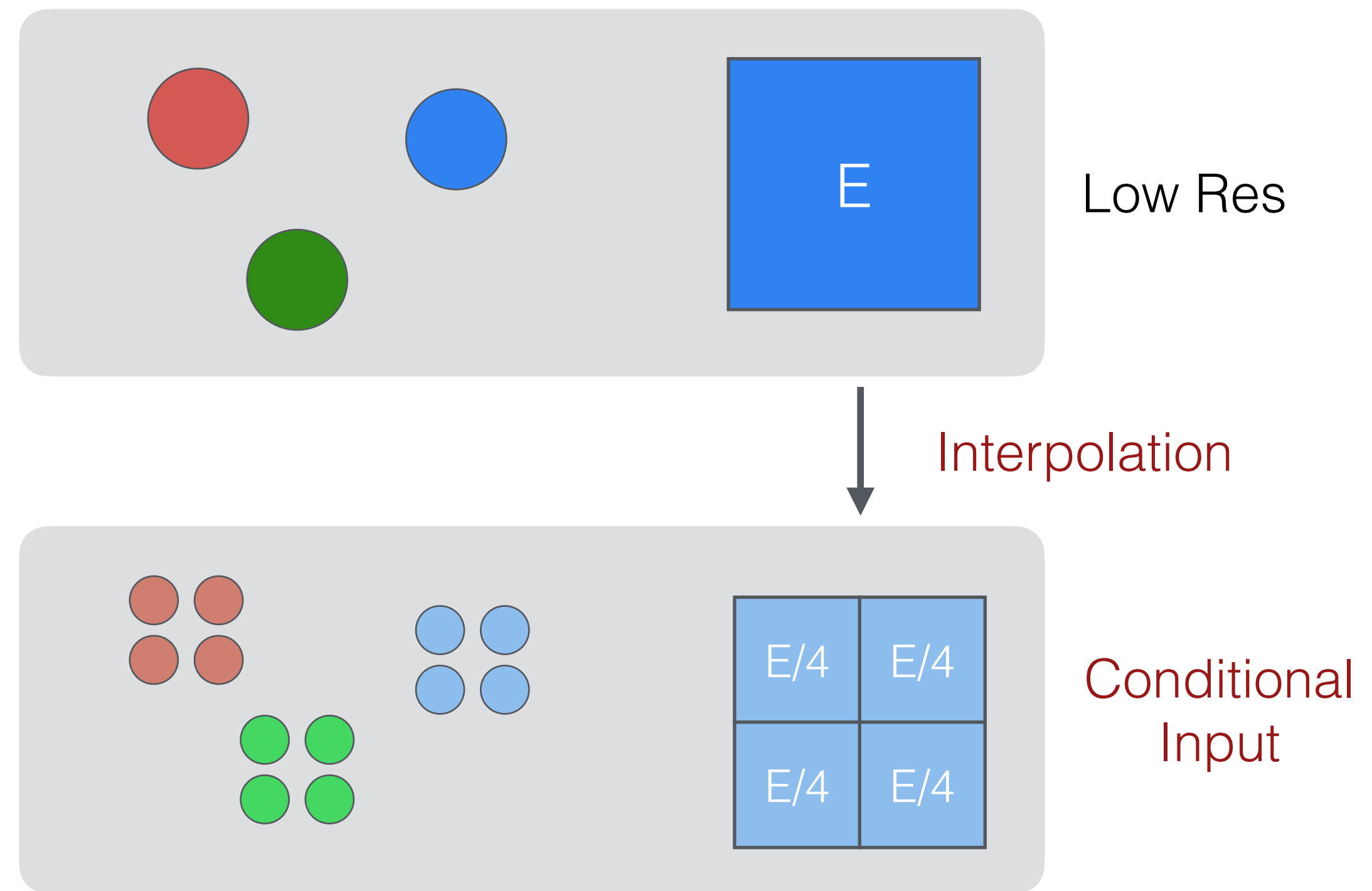
Image Super-Resolution via Iterative Refinement (<https://arxiv.org/pdf/2104.07636.pdf>)



Diffusion set up

- ◆ Inspired by the SR3 paper

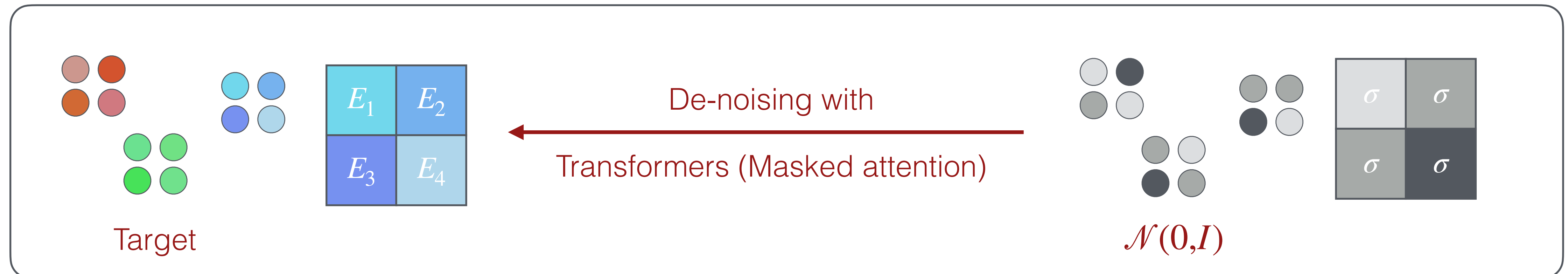
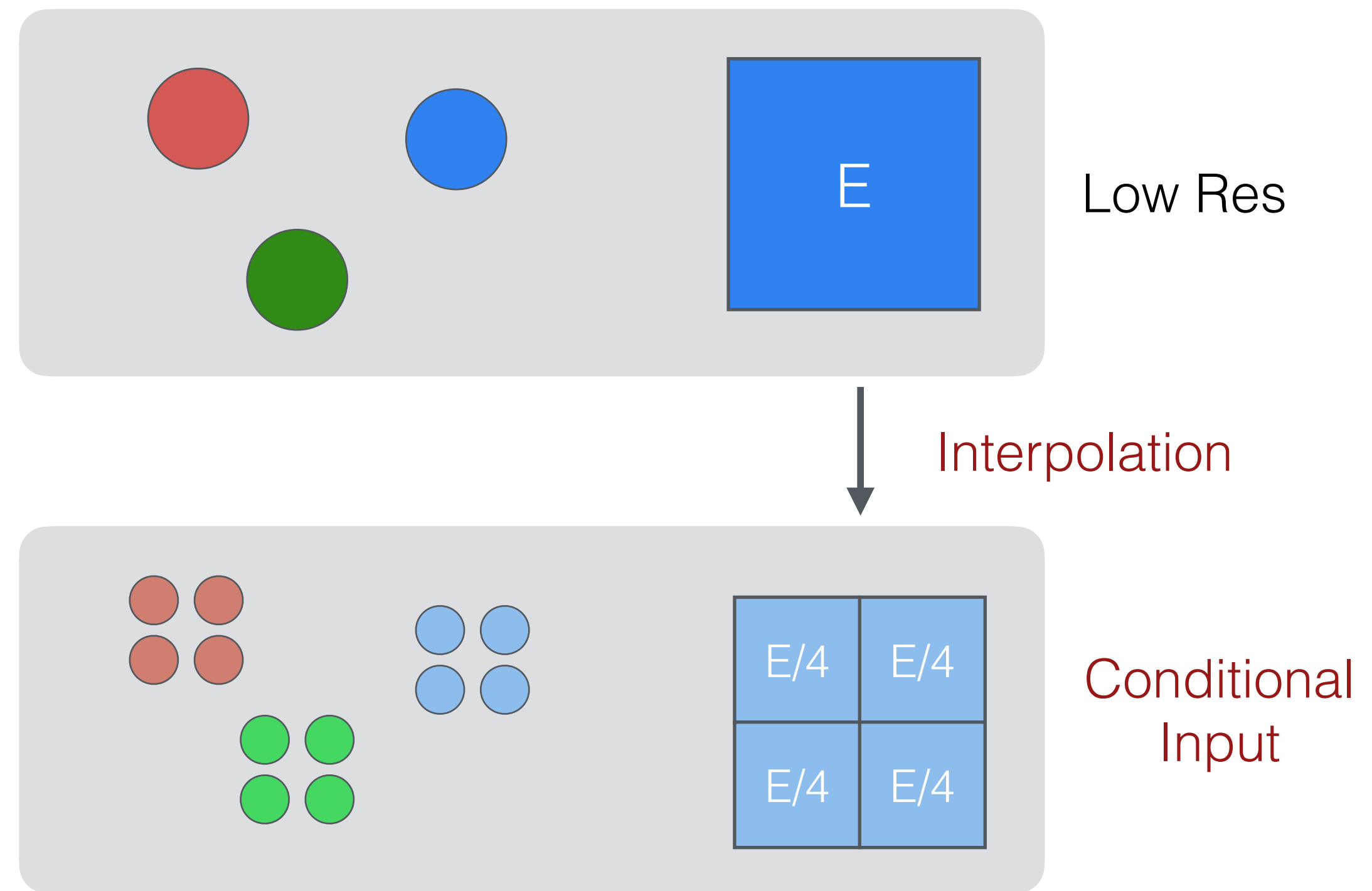
Image Super-Resolution via Iterative Refinement (<https://arxiv.org/pdf/2104.07636.pdf>)



Diffusion set up

- ◆ Inspired by the SR3 paper

Image Super-Resolution via Iterative Refinement (<https://arxiv.org/pdf/2104.07636.pdf>)



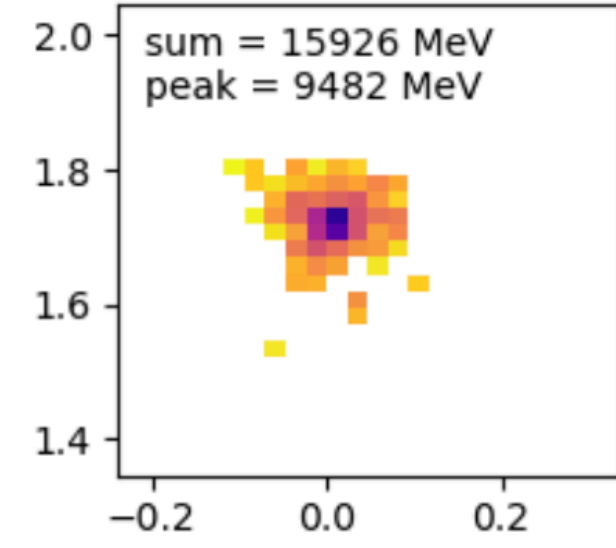
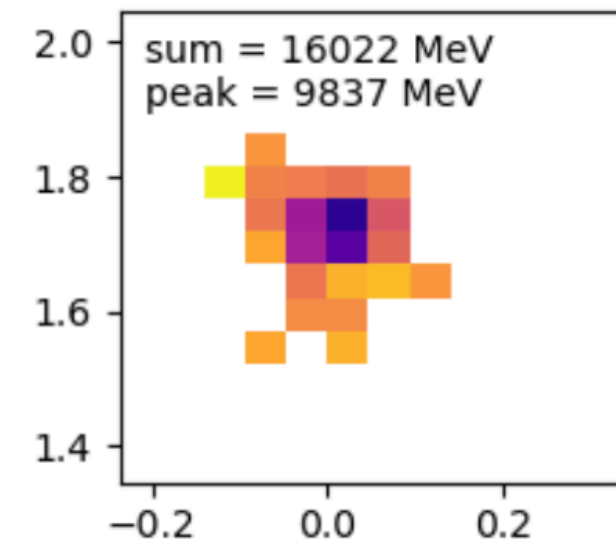
Sneak peak into the results

LR

HR (target)

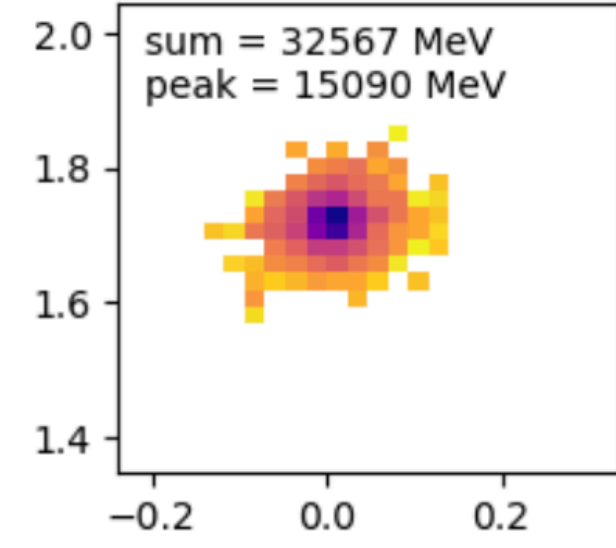
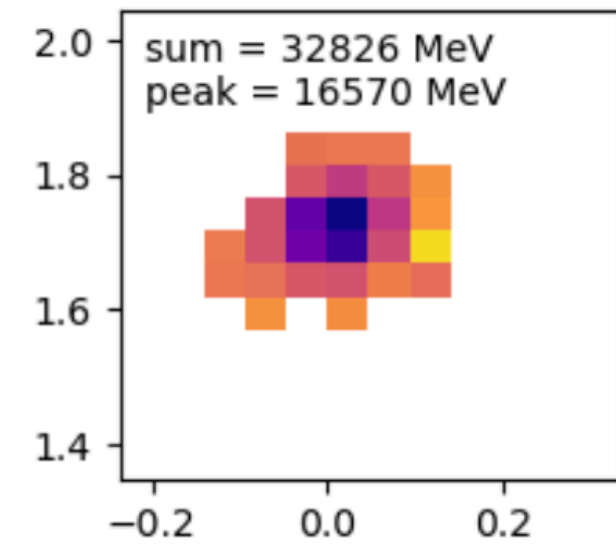
LR

HR (target)



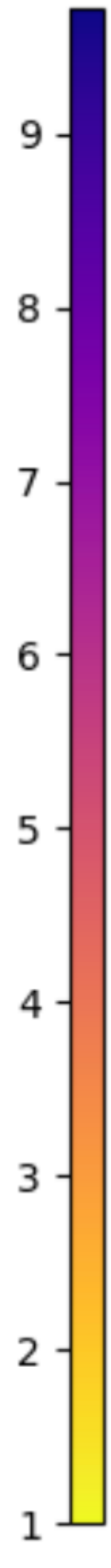
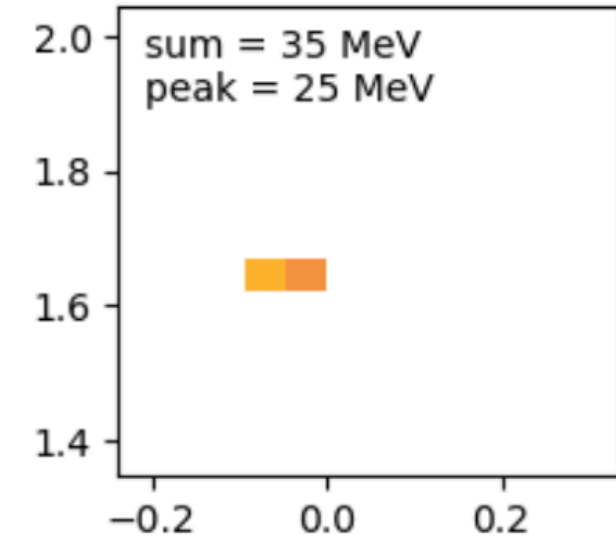
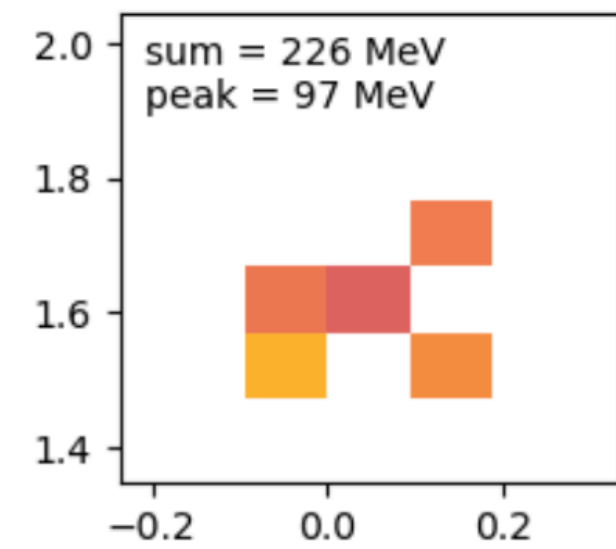
LR

HR (target)

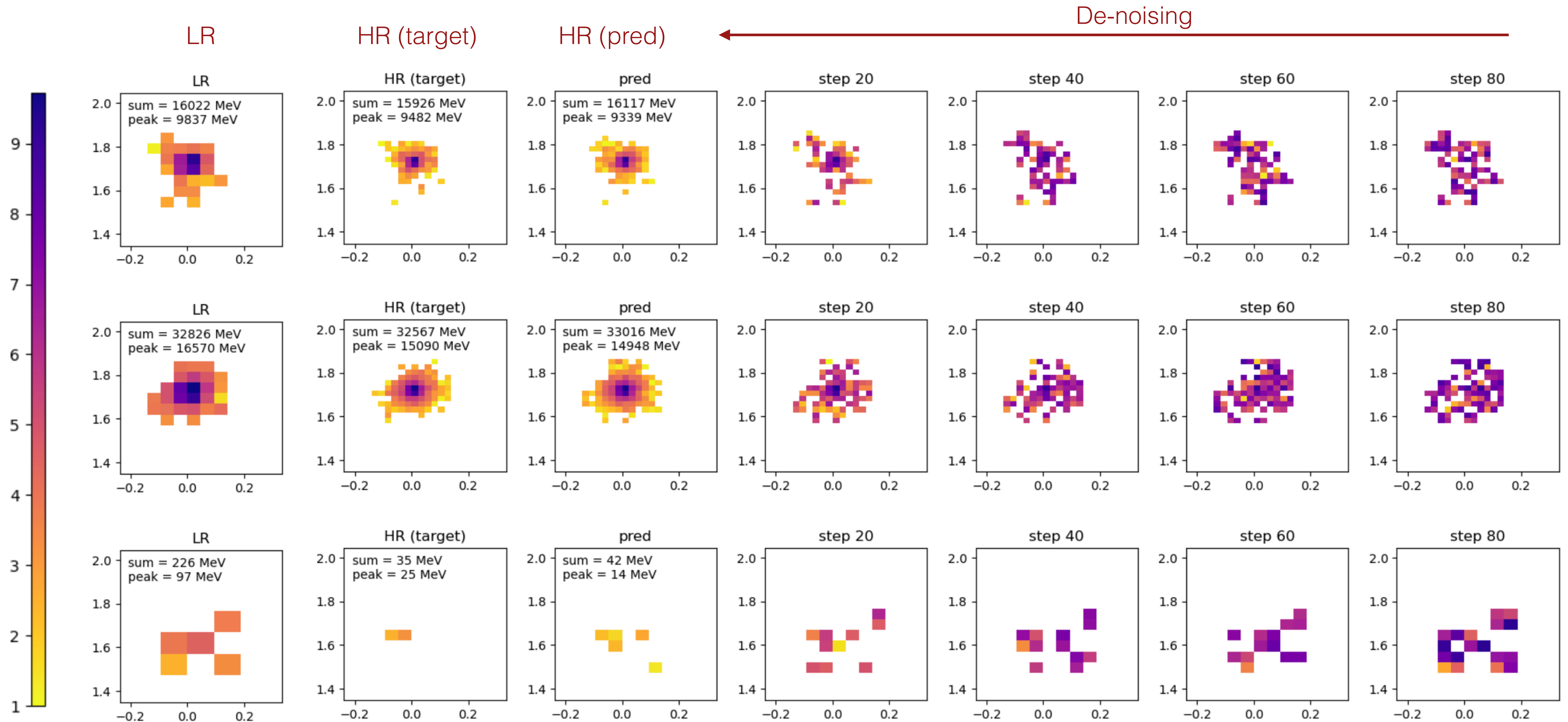


LR

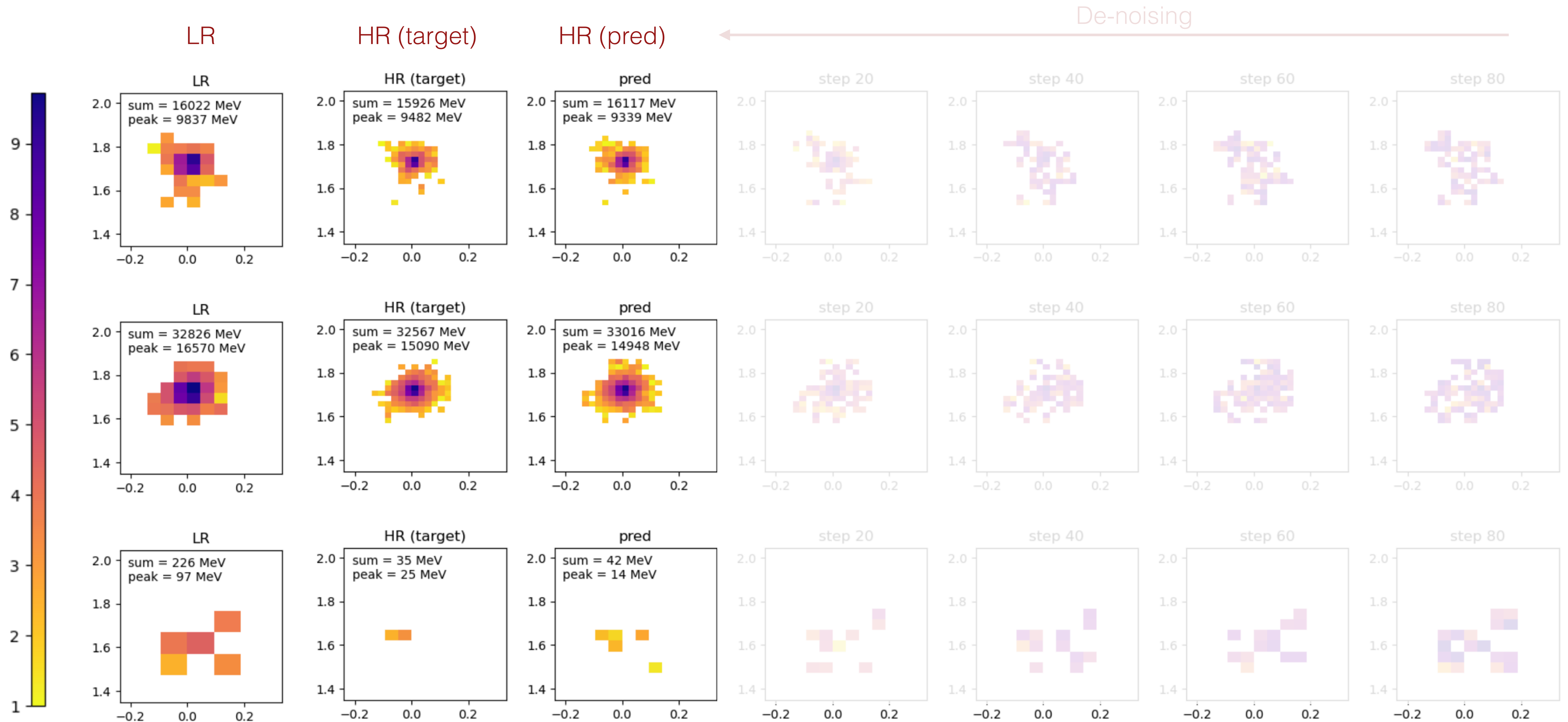
HR (target)



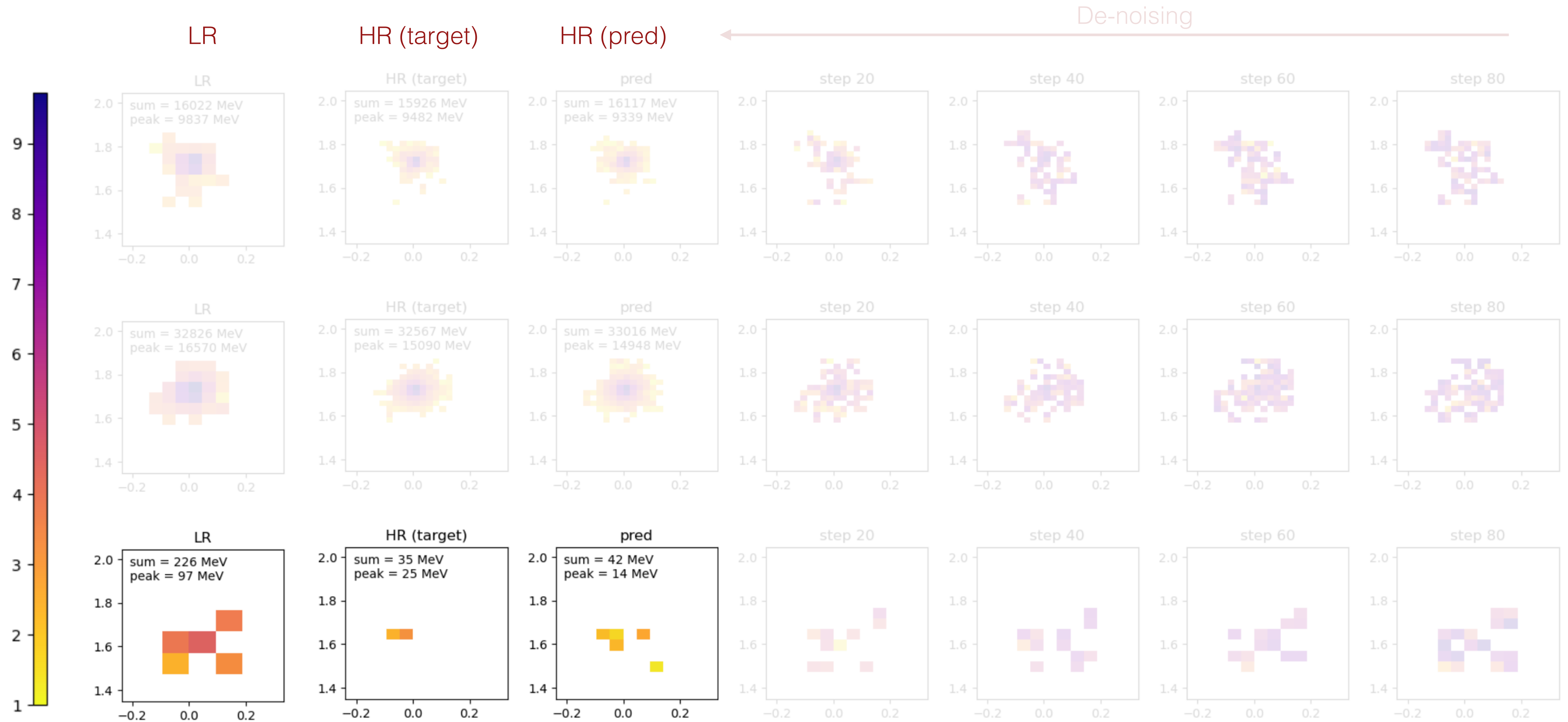
Sneak peak into the results



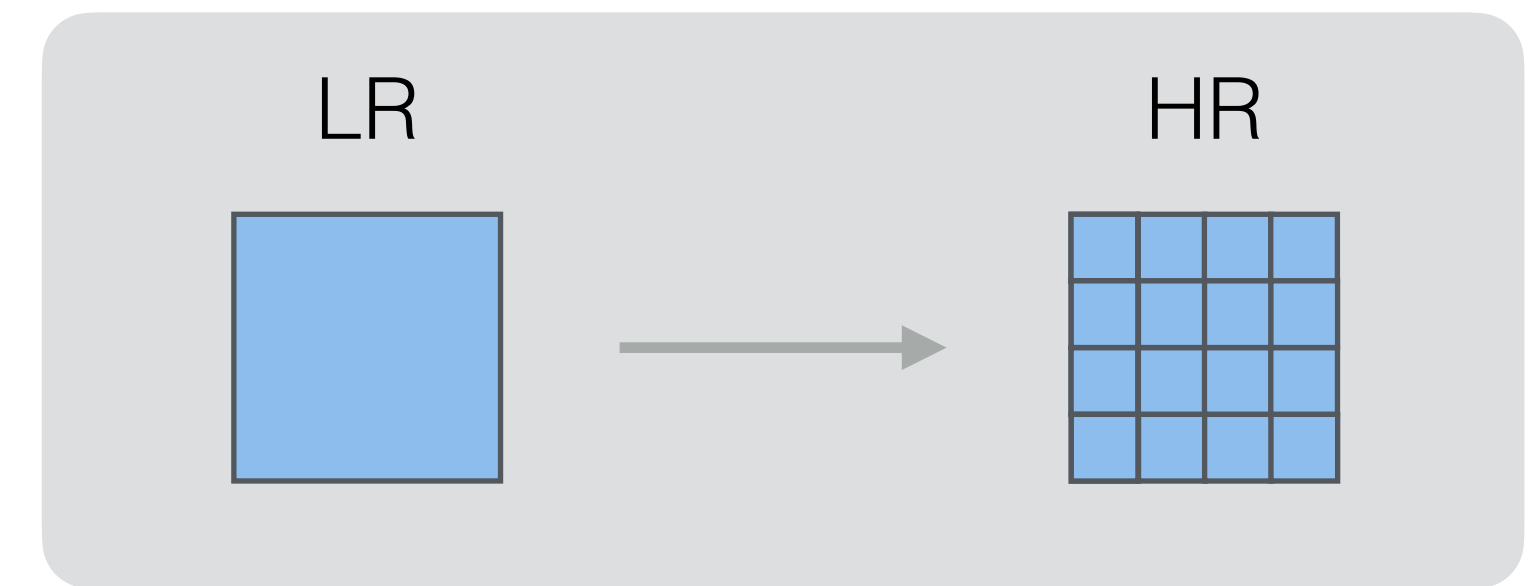
Sneak peak into the results



De-noising



A more Interesting case!

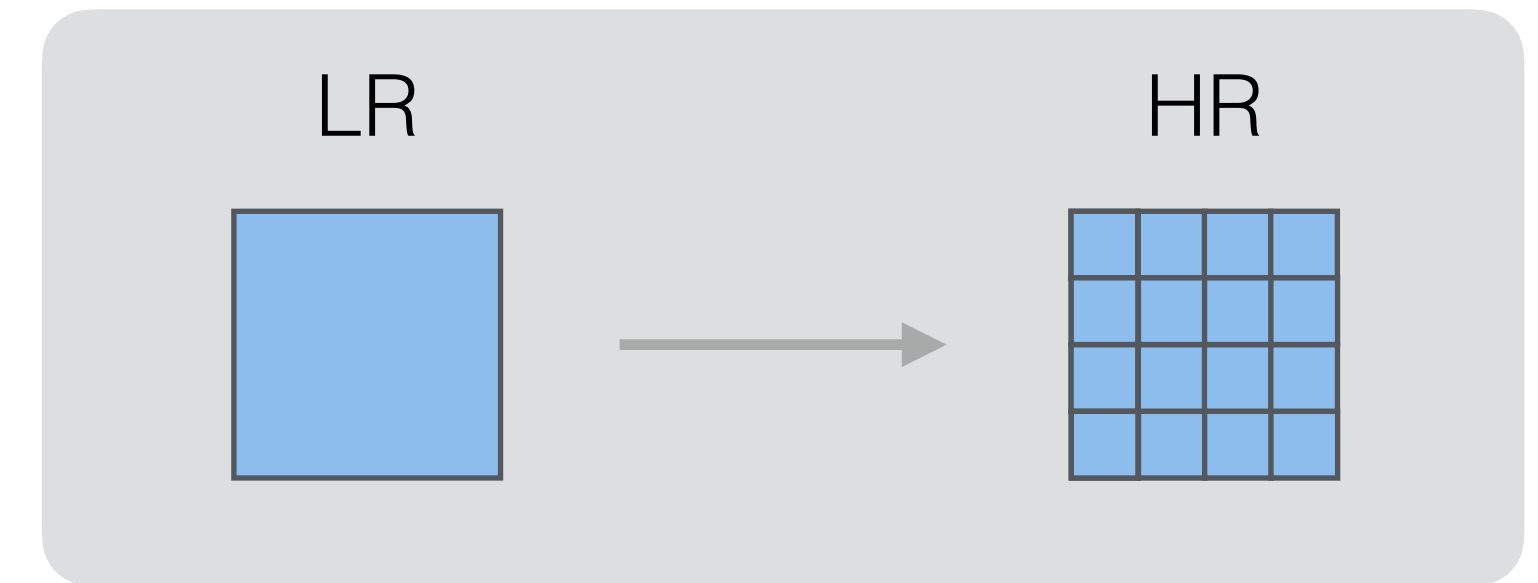
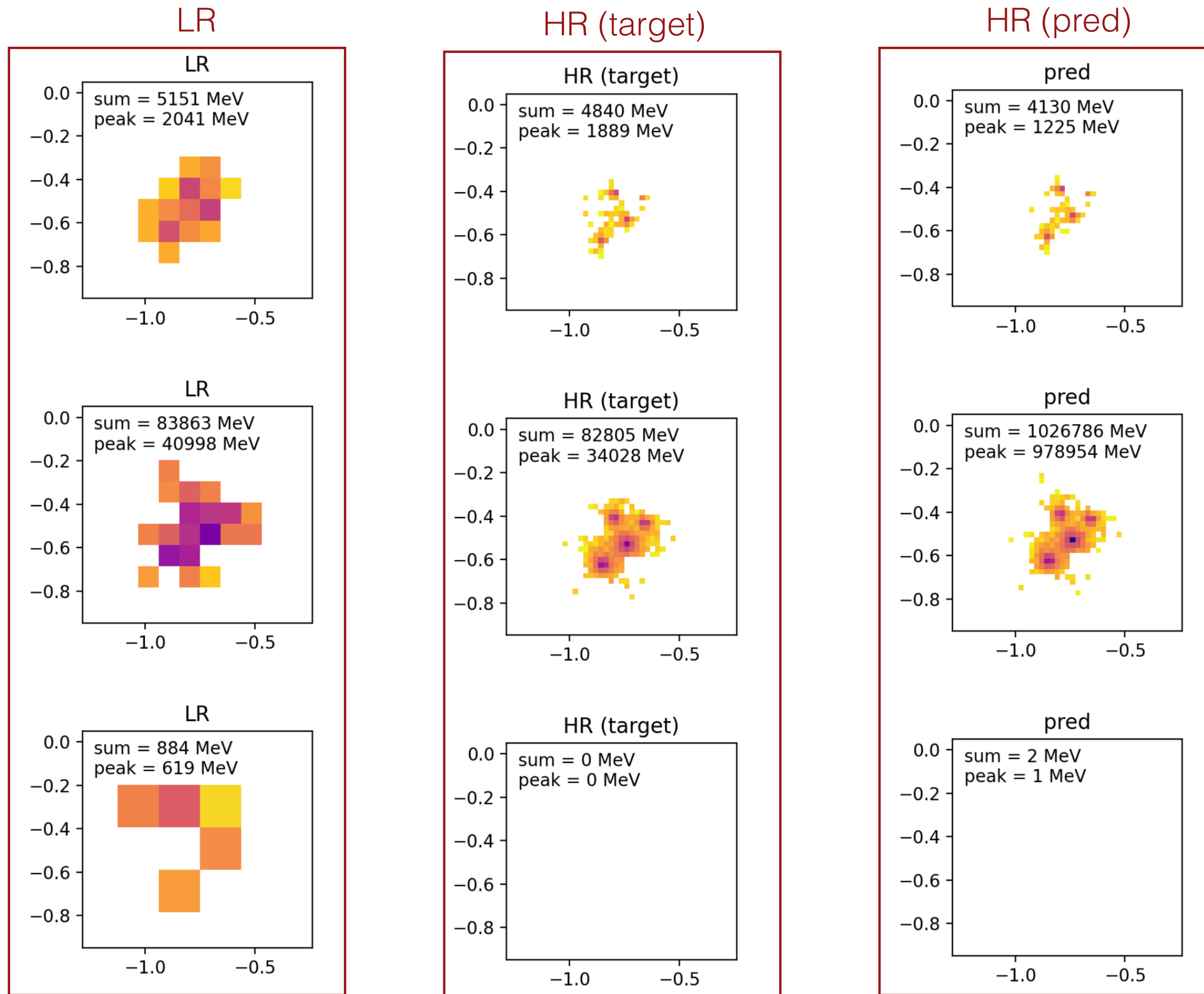


- Multiple particles
- 1-5 particles
- Electrons and photons

Overtraining!

(Couldn't get the full training ready in time)

A more Interesting case!

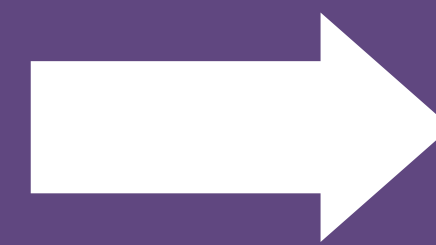


- Multiple particles
- 1-5 particles
- Electrons and photons

Overtraining!

(Couldn't get the full training ready in time)

**Is it
hallucination?**

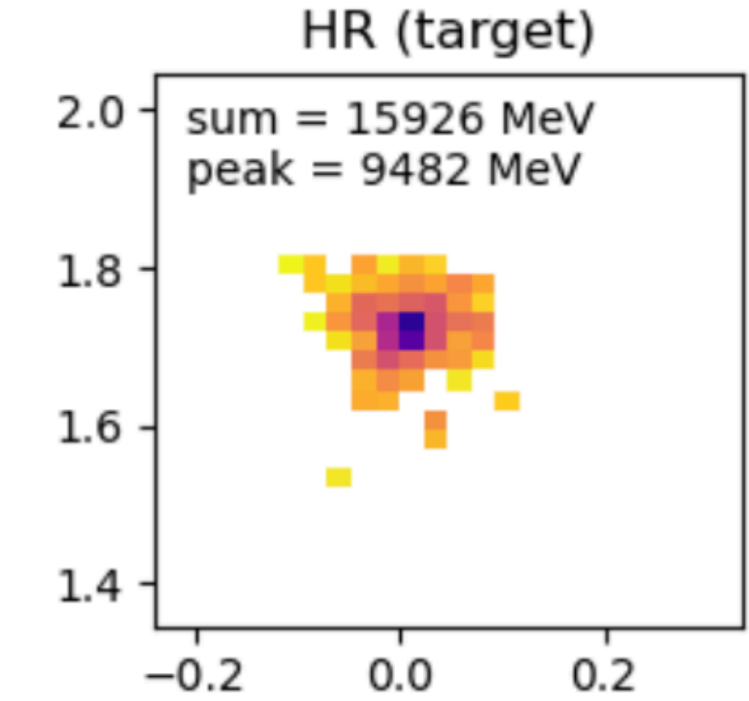
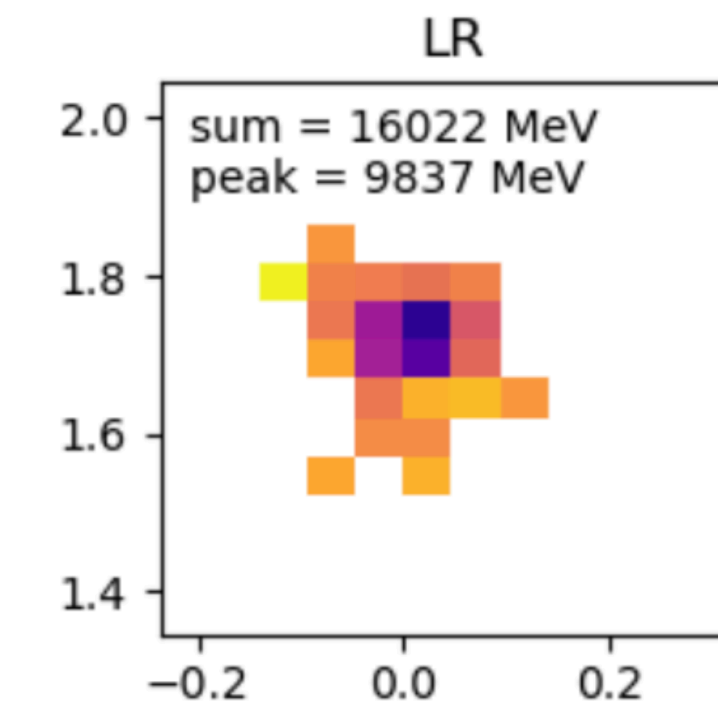


Creating information out of nowhere?

Where does the extra info come from?

Where does the extra info come from?

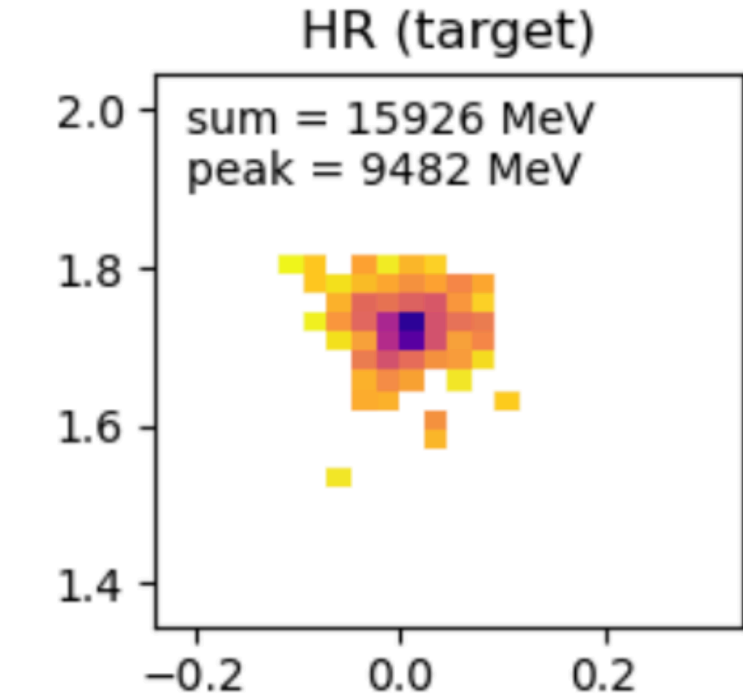
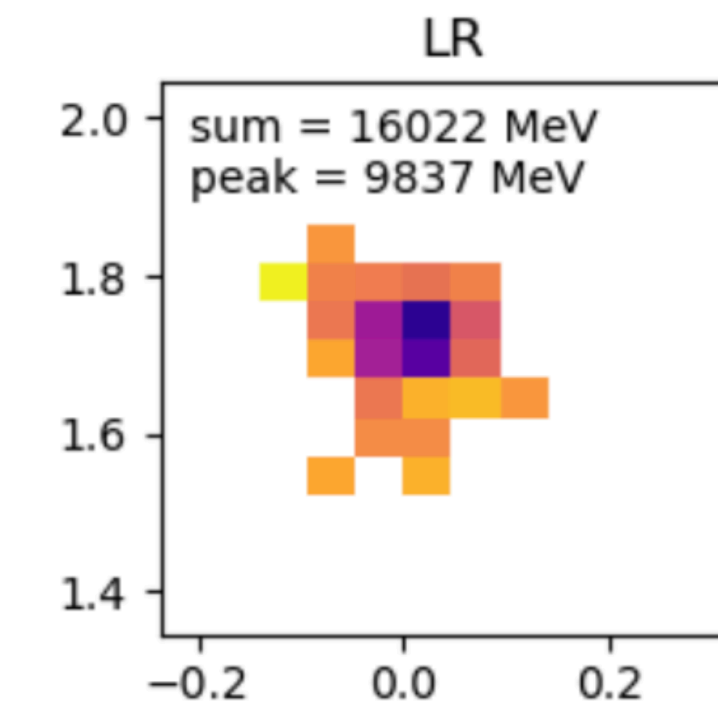
- ◆ From training data!
 - ➔ Energy deposition, by let's say a photon, is not random
 - ➔ Model can learn the HR distribution conditioned on the LR distribution



Where does the extra info come from?

- ◆ From training data!
 - ➔ Energy deposition, by let's say a photon, is not random
 - ➔ Model can learn the HR distribution conditioned on the LR distribution

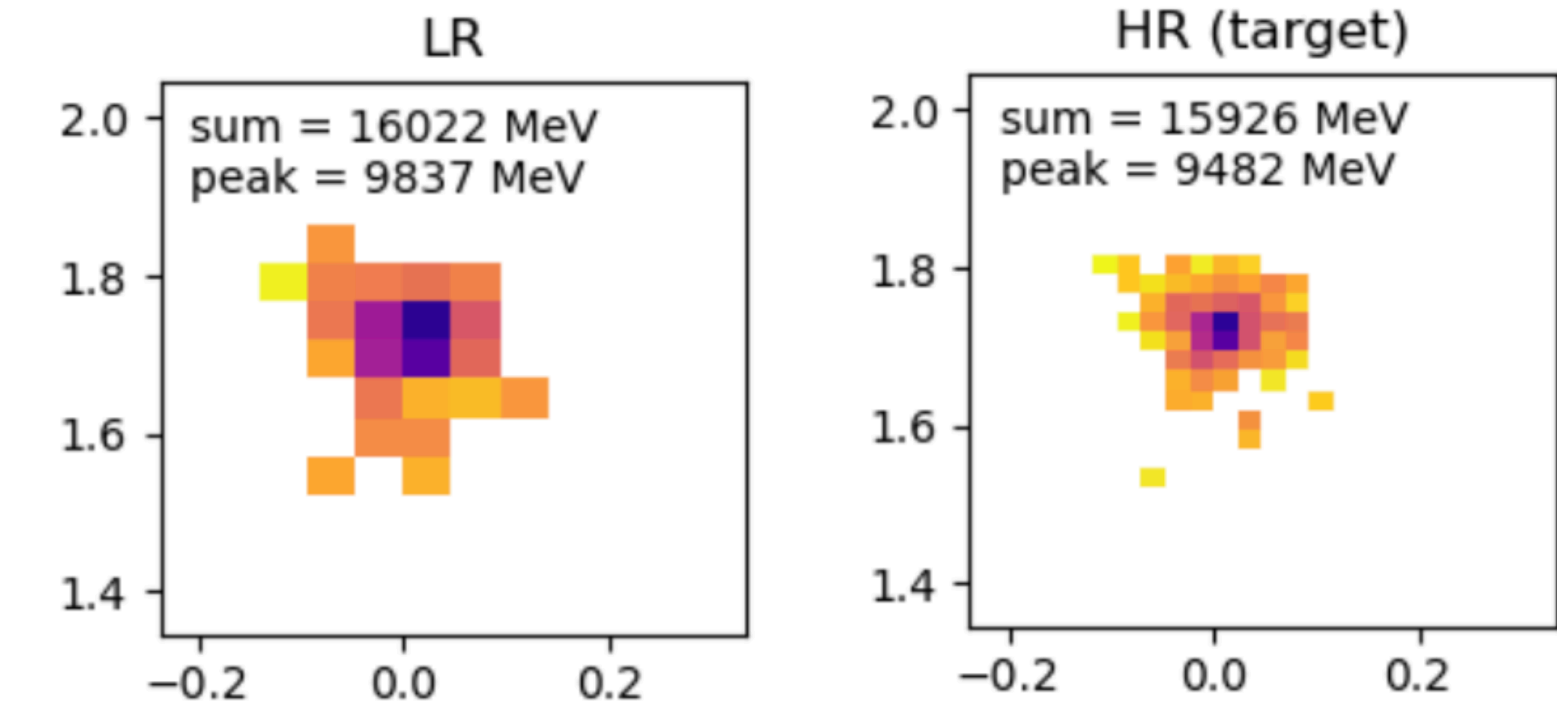
- ◆ HR output = **Educated estimation** of the model based on the **patterns learned from the training data**



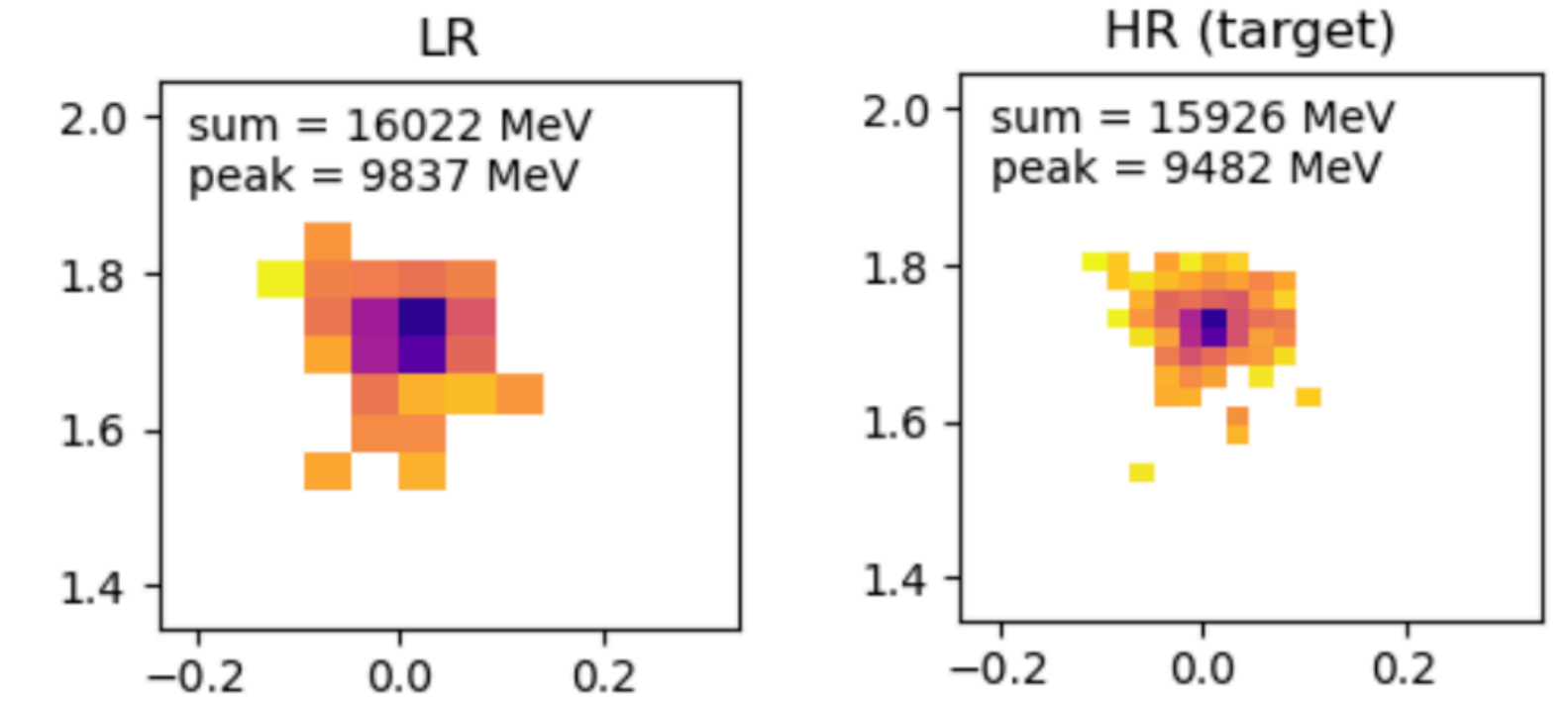
Where does the extra info come from?

- ◆ From training data!
 - ➔ Energy deposition, by let's say a photon, is not random
 - ➔ Model can learn the HR distribution conditioned on the LR distribution

- ◆ HR output = **Educated estimation** of the model based on the **patterns learned from the training data**
 - ➔ Similar to how SR work in Computer vision

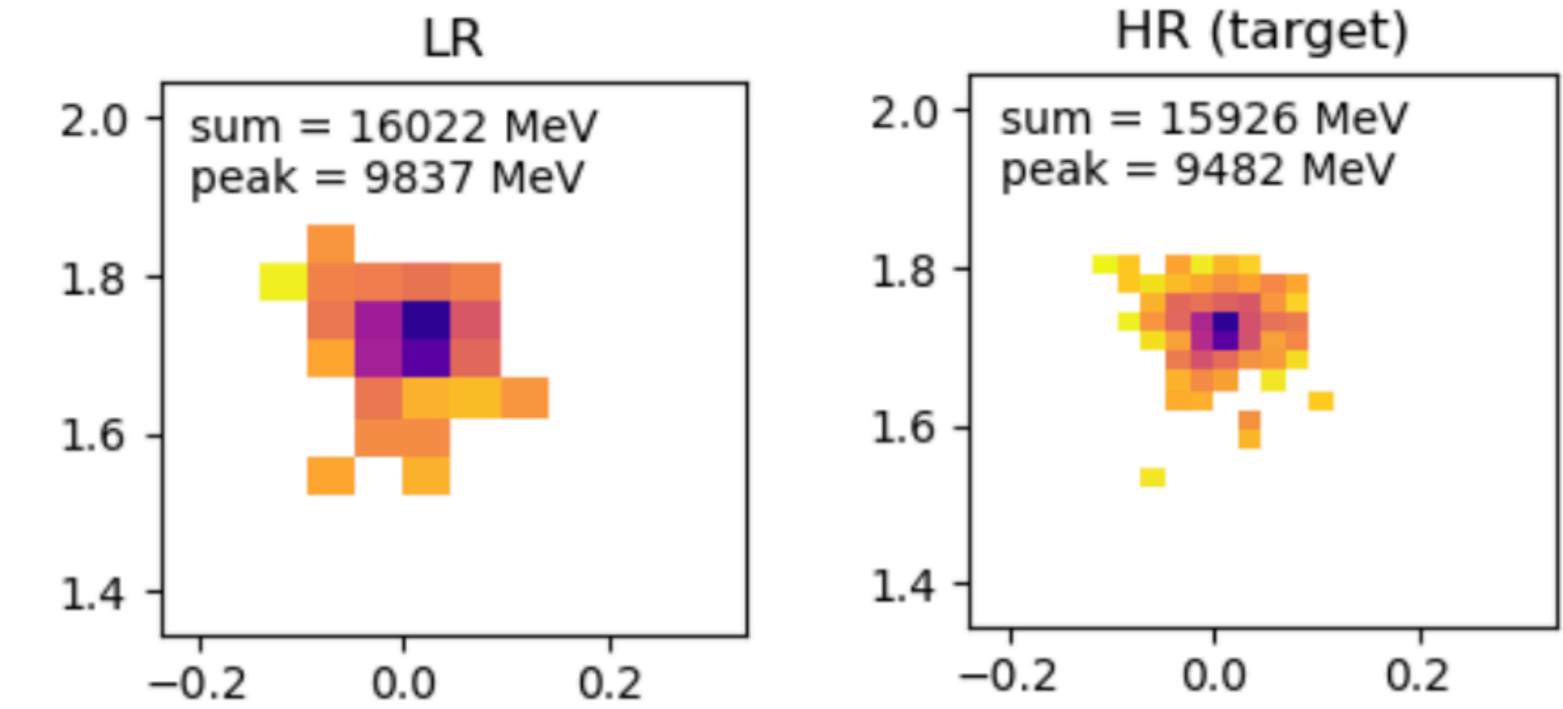


But, how do we still verify we are not hallucinating?



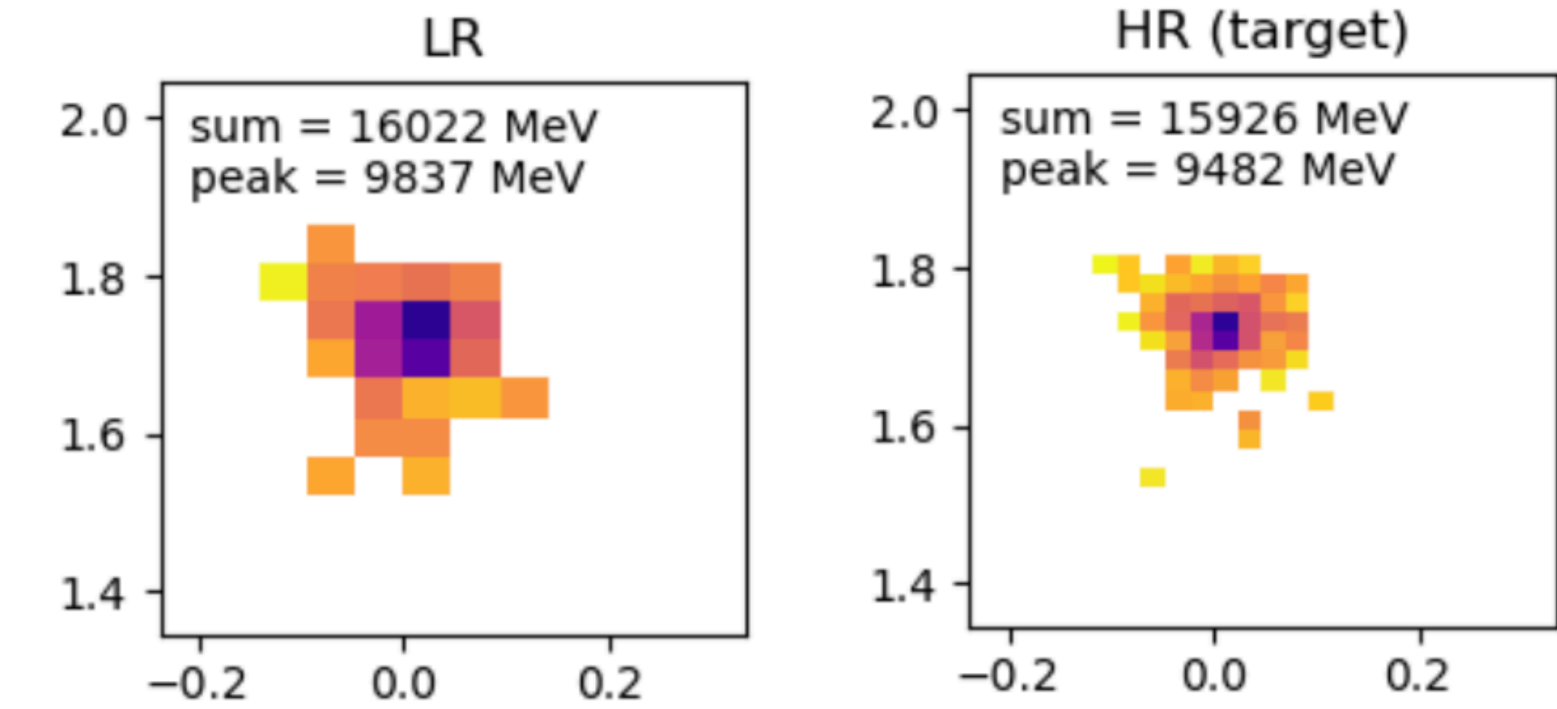
But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?



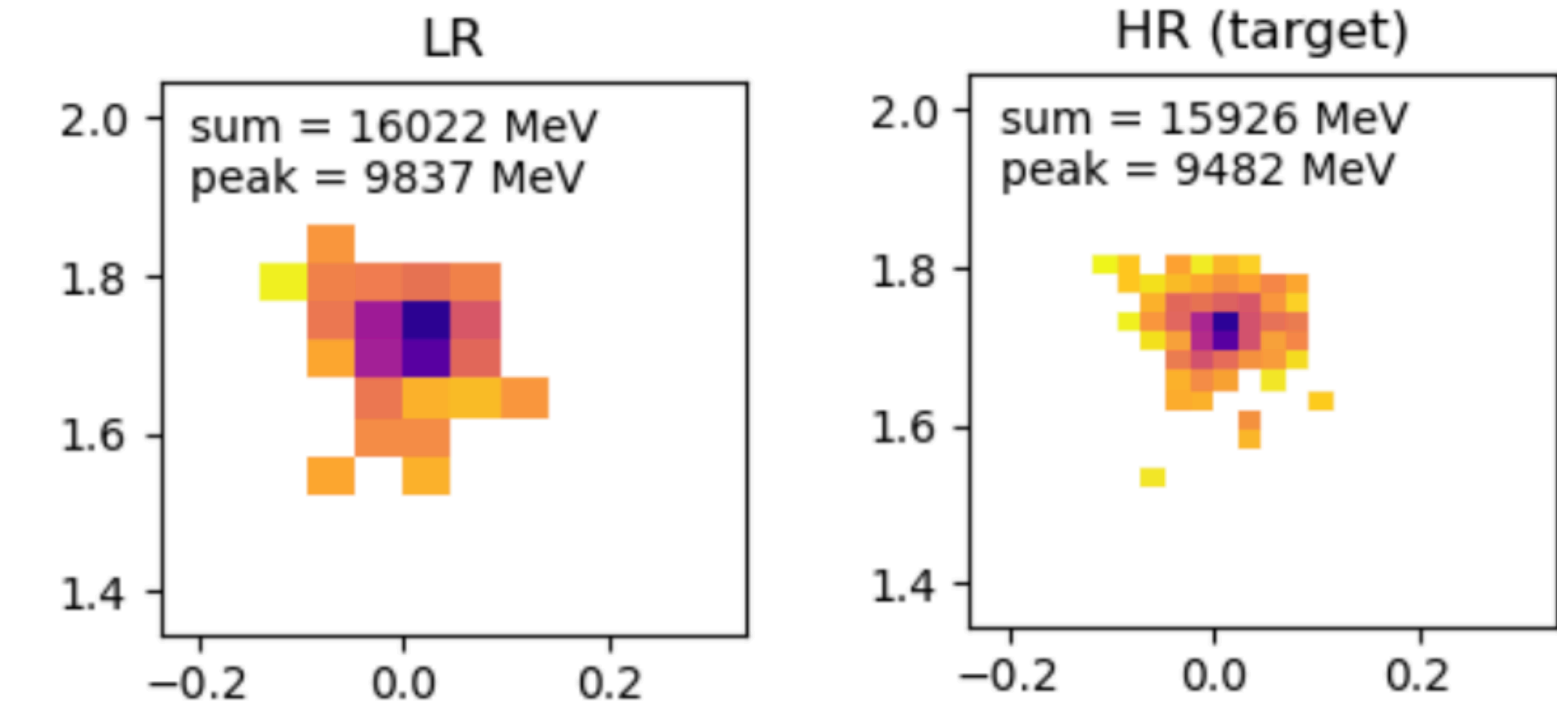
But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?
 - ➔ Ans: Of course, they are not perfect, **just like any other ML model**



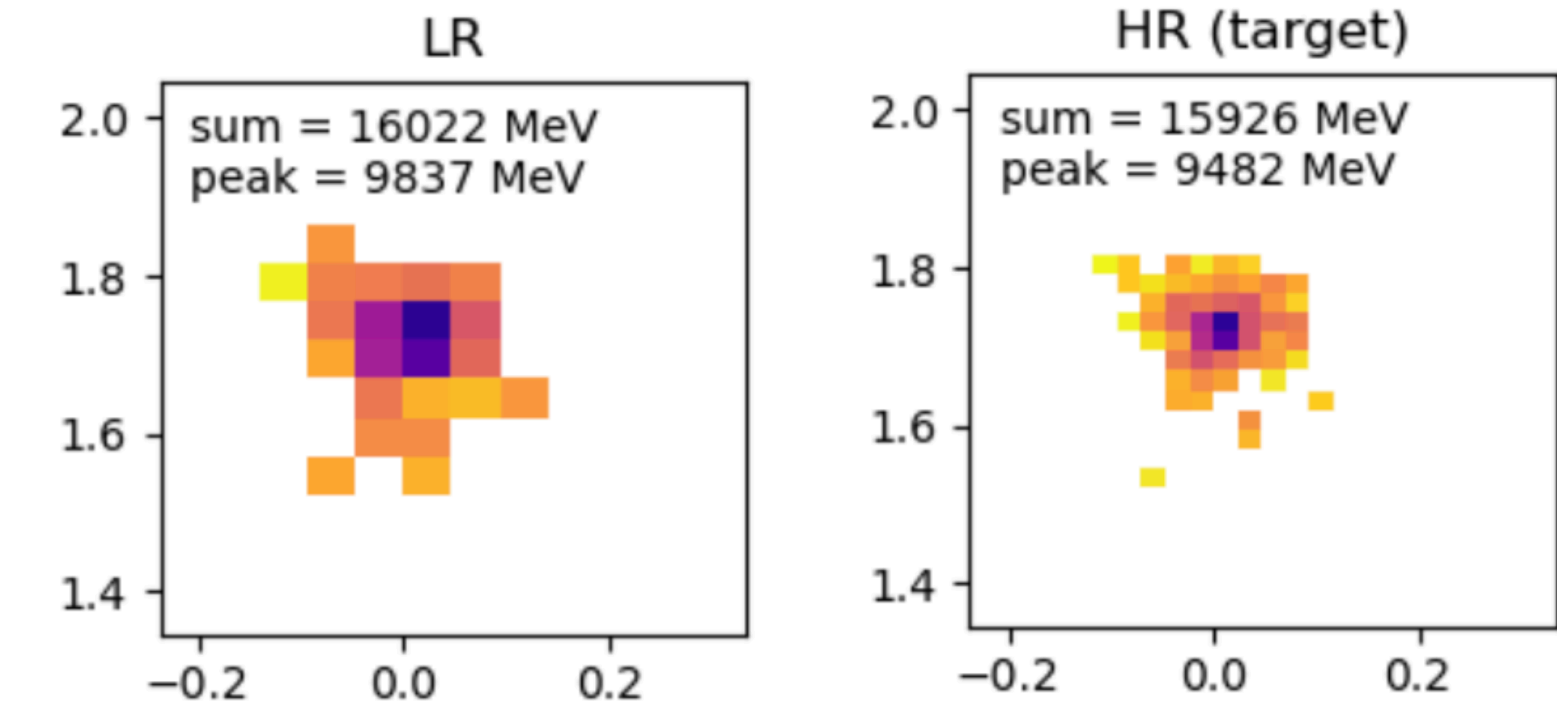
But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?
 - ➔ Ans: Of course, they are not perfect, **just like any other ML model**
- ◆ How do we estimate how good/bad are the HR estimations?



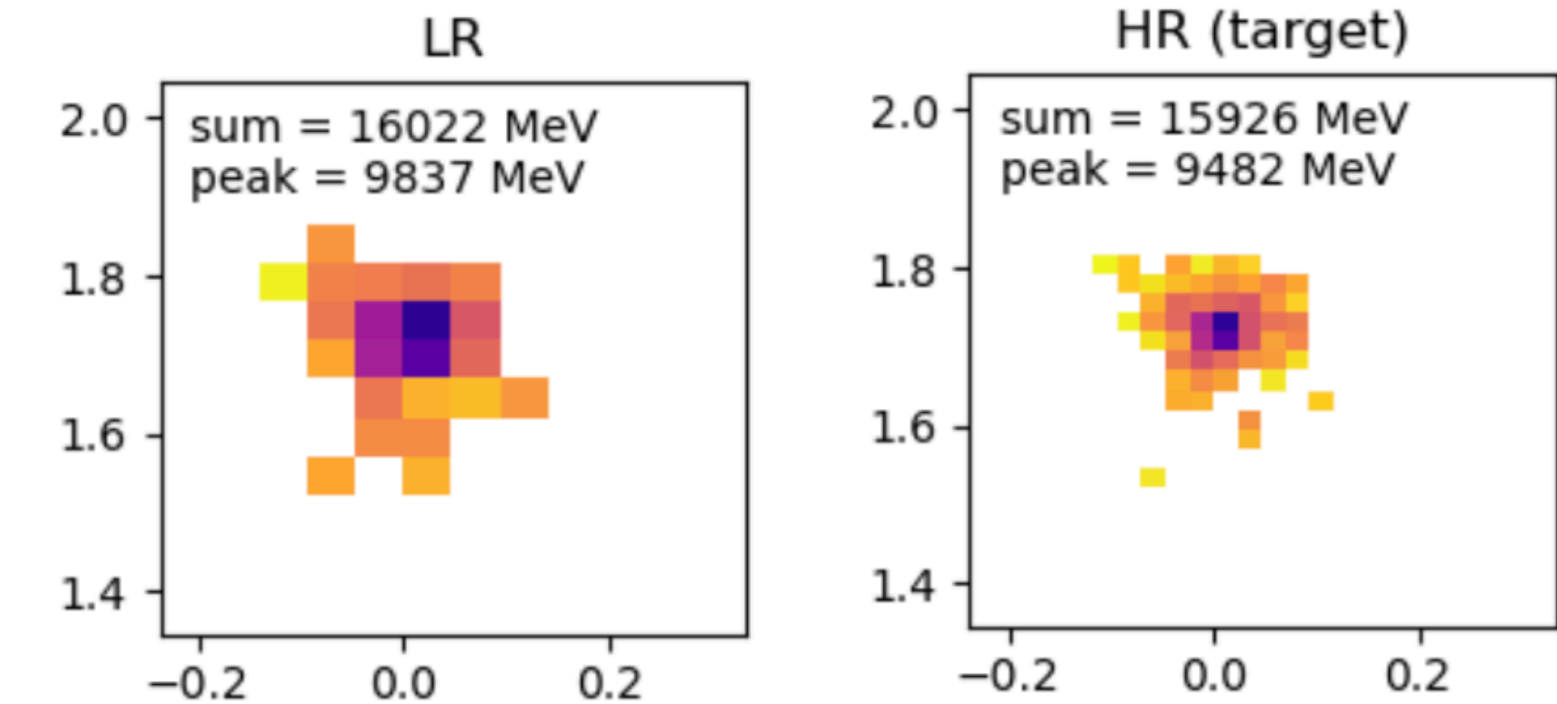
But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?
 - ➔ Ans: Of course, they are not perfect, **just like any other ML model**
- ◆ How do we estimate how good/bad are the HR estimations?
- ◆ With simulation,



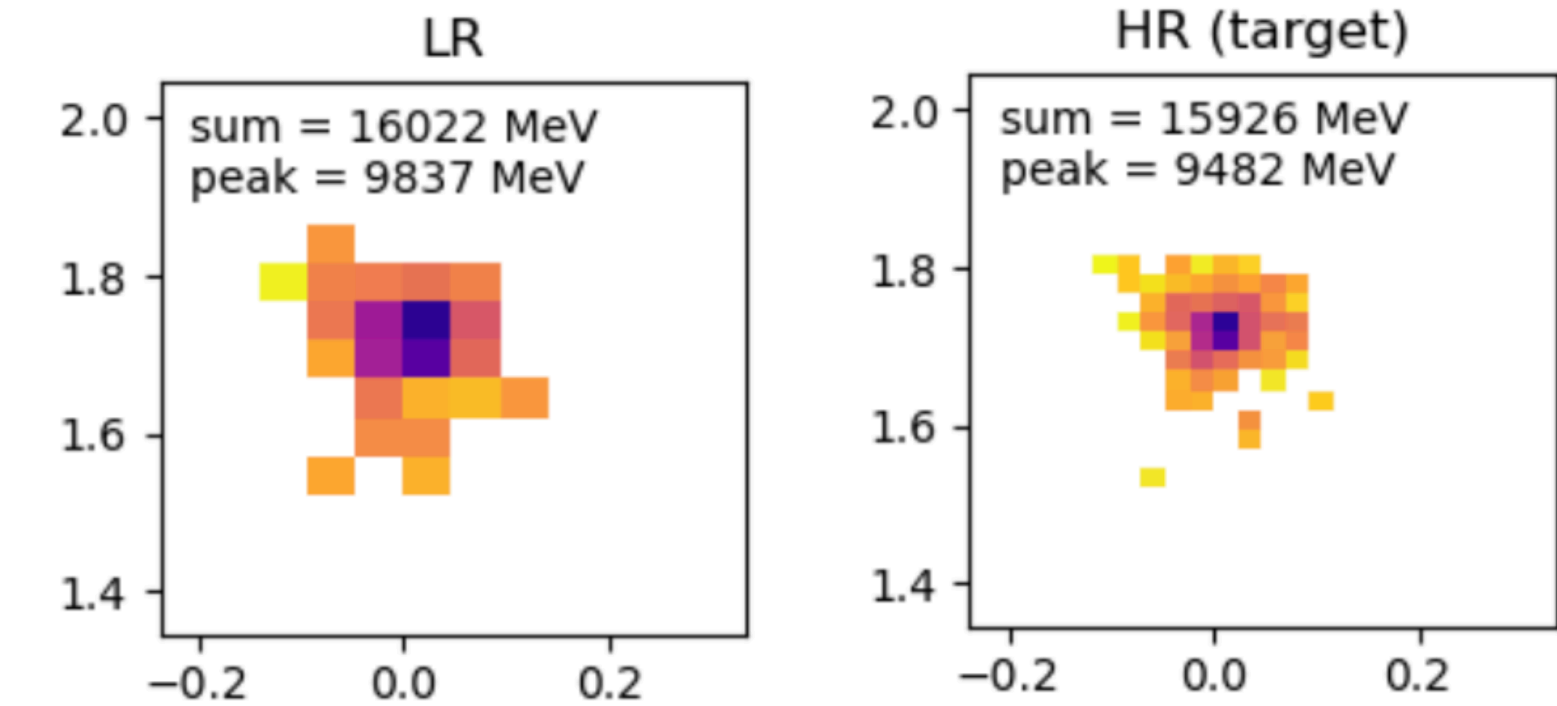
But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?
 - ➔ Ans: Of course, they are not perfect, **just like any other ML model**
- ◆ How do we estimate how good/bad are the HR estimations?
- ◆ With simulation,
 - ➔ it’s easier, we can have the truth targets



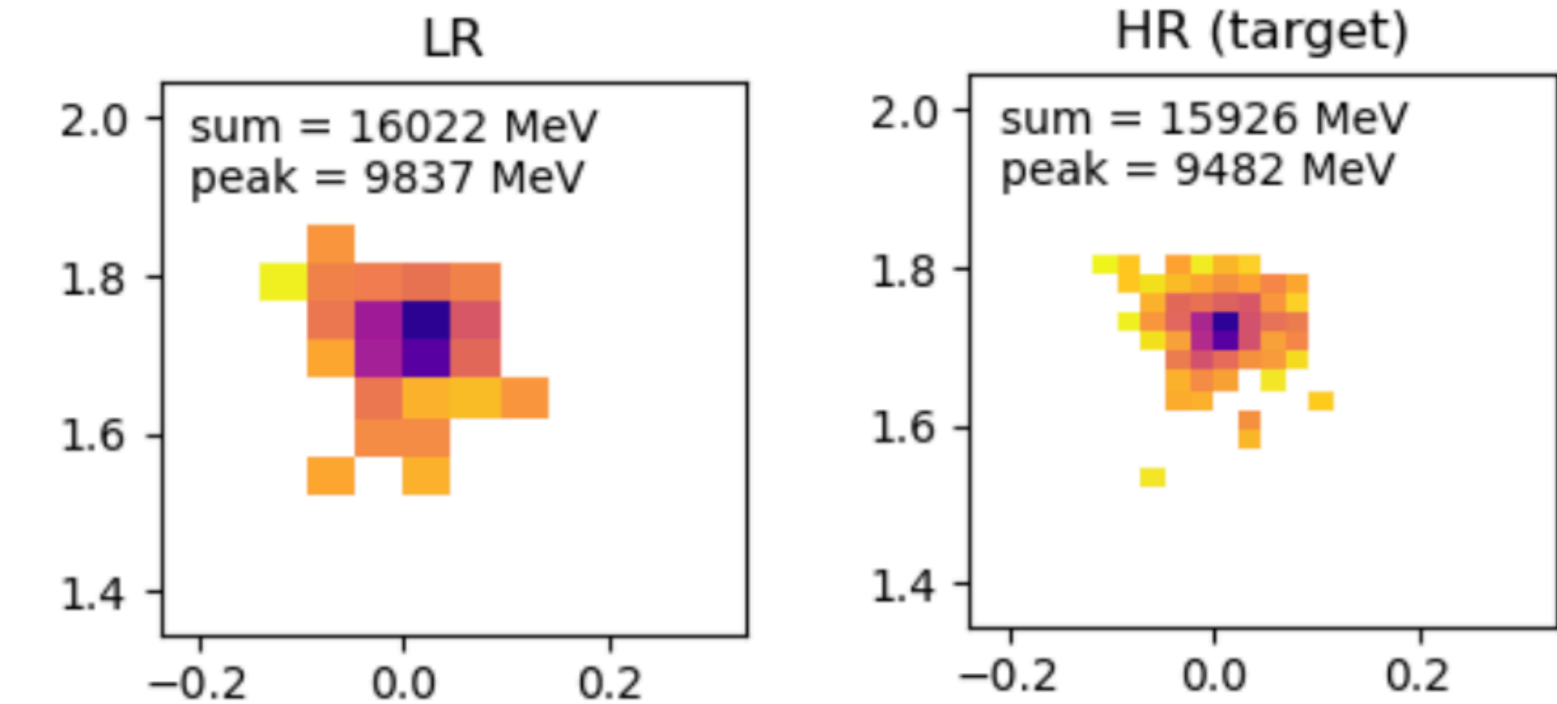
But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?
 - ➔ Ans: Of course, they are not perfect, **just like any other ML model**
- ◆ How do we estimate how good/bad are the HR estimations?
- ◆ With simulation,
 - ➔ it’s easier, we can have the truth targets
- ◆ With actual data,



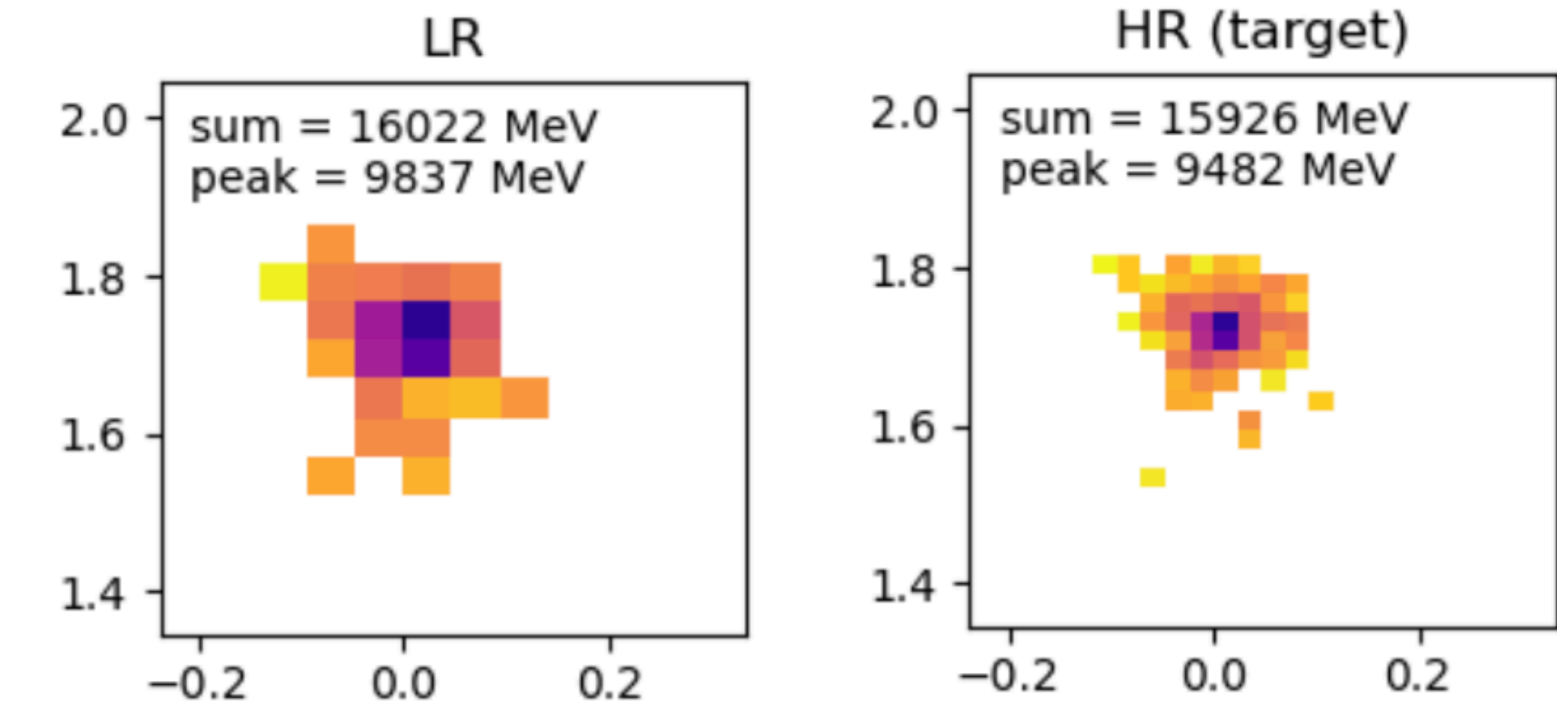
But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?
 - ➔ Ans: Of course, they are not perfect, **just like any other ML model**
- ◆ How do we estimate how good/bad are the HR estimations?
- ◆ With simulation,
 - ➔ it’s easier, we can have the truth targets
- ◆ With actual data,
 - ➔ Not so easy. (How much we trust our simulations?)

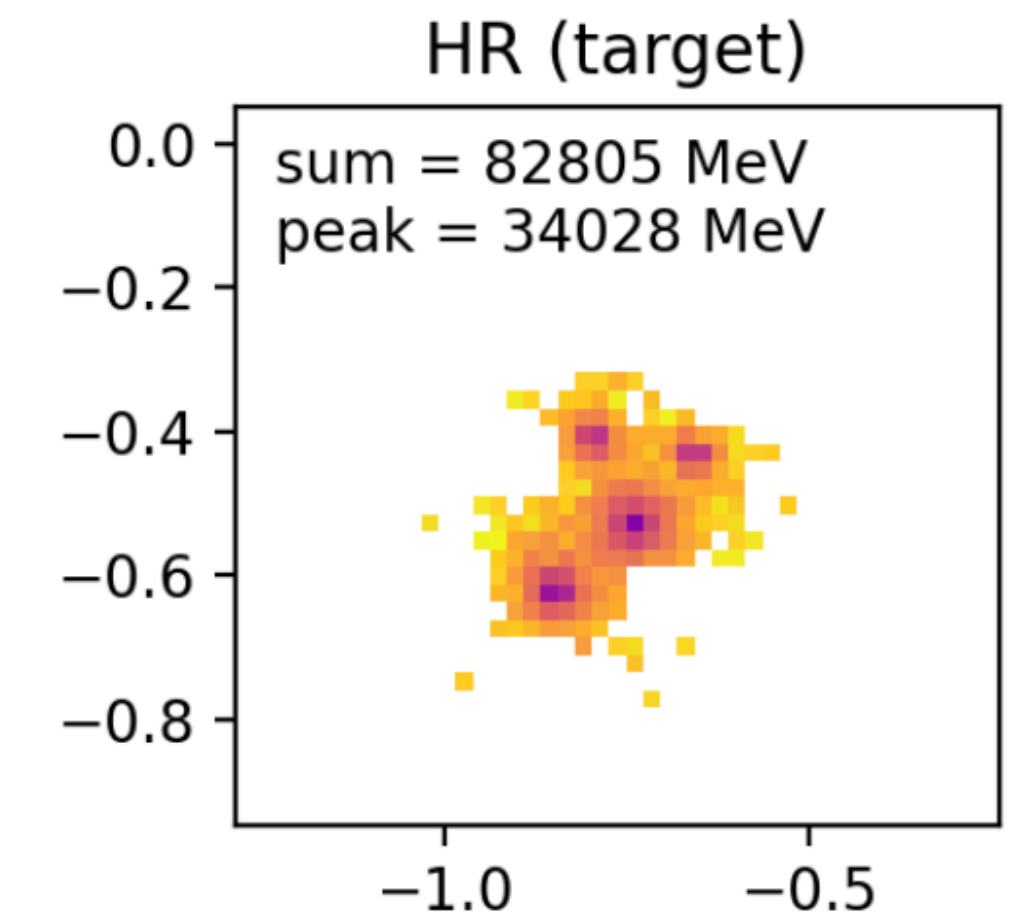
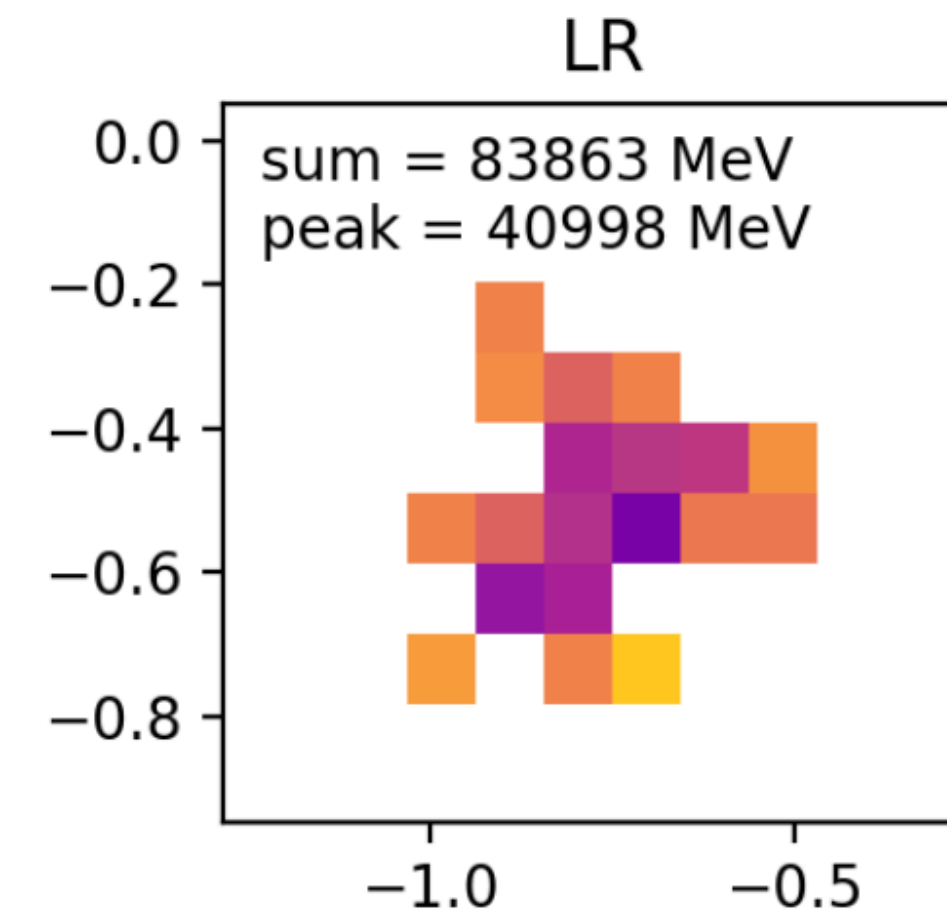


But, how do we still verify we are not hallucinating?

- ◆ Qn: The model can predict “realistic looking” outputs, but are they correct?
 - ➔ Ans: Of course, they are not perfect, **just like any other ML model**
- ◆ How do we estimate how good/bad are the HR estimations?
- ◆ With simulation,
 - ➔ it’s easier, we can have the truth targets
- ◆ With actual data,
 - ➔ Not so easy. (How much we trust our simulations?)
 - ➔ **Calibration problem** (tricky, but I believe doable)

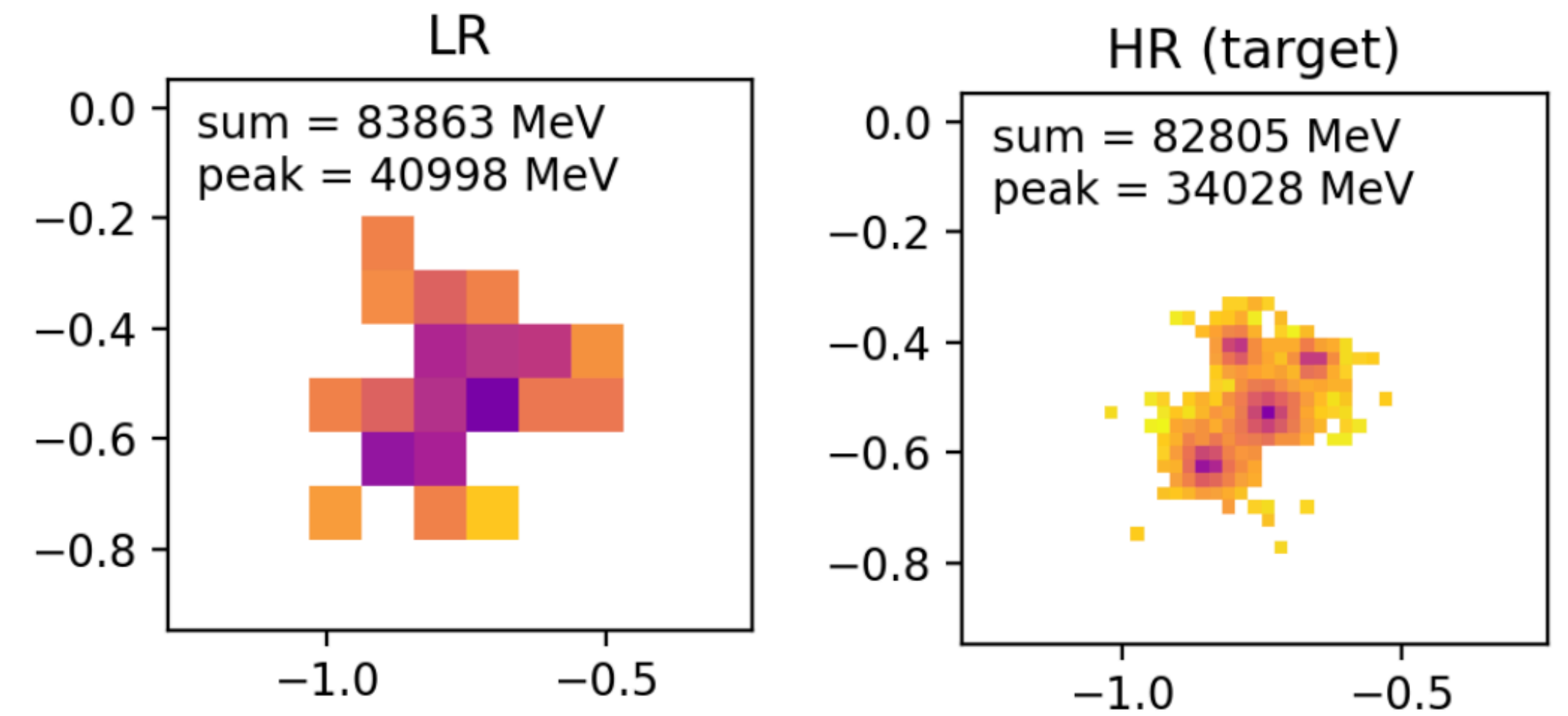


But, most importantly,



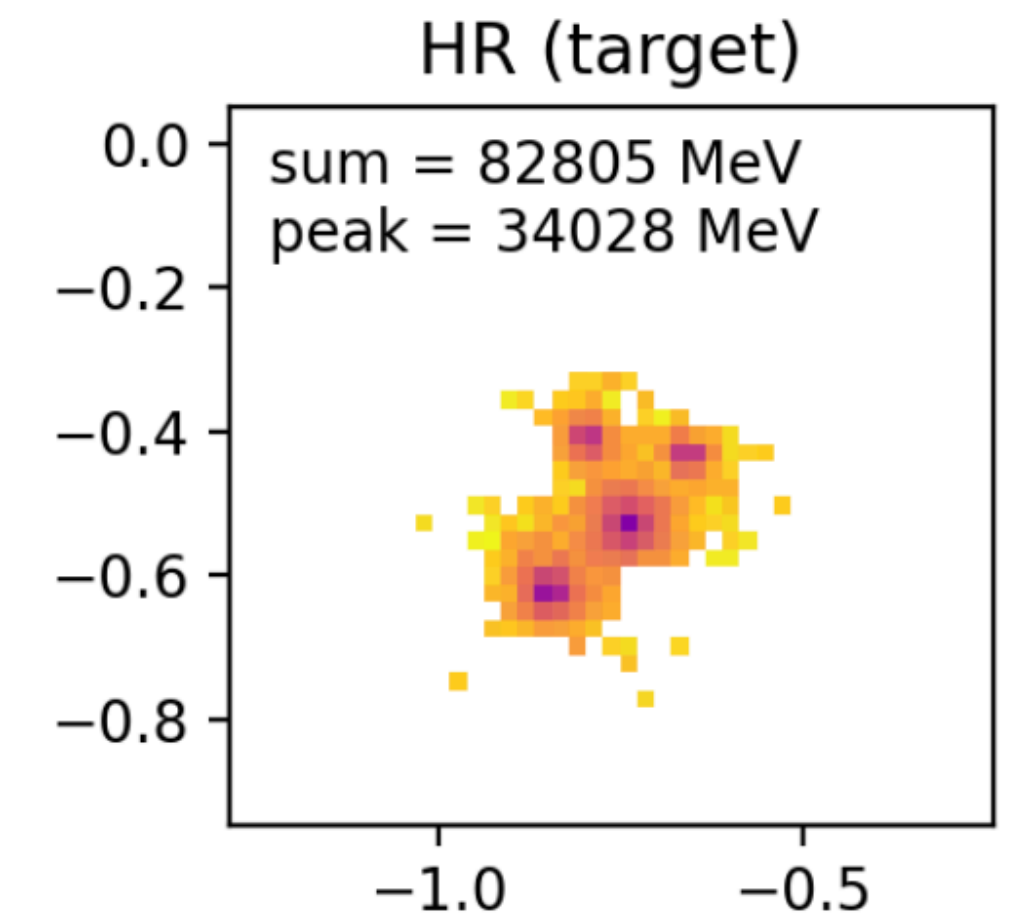
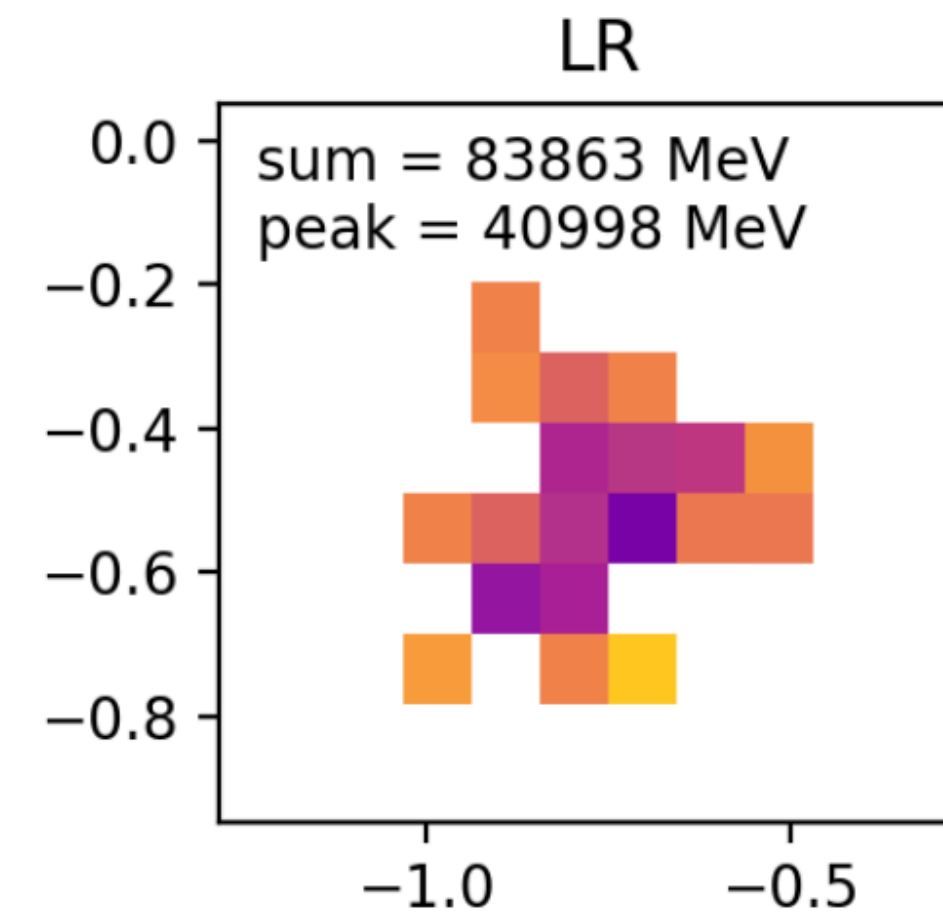
But, most importantly,

- ◆ We shouldn't look at it in isolation



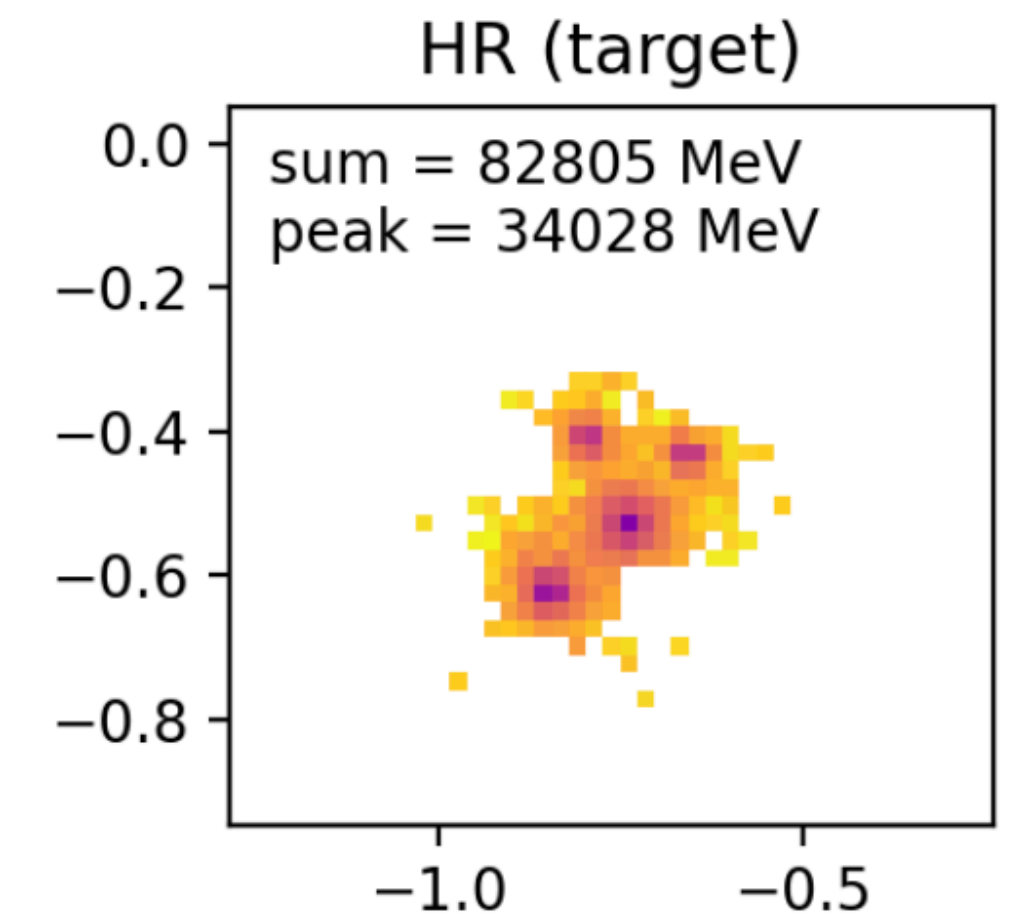
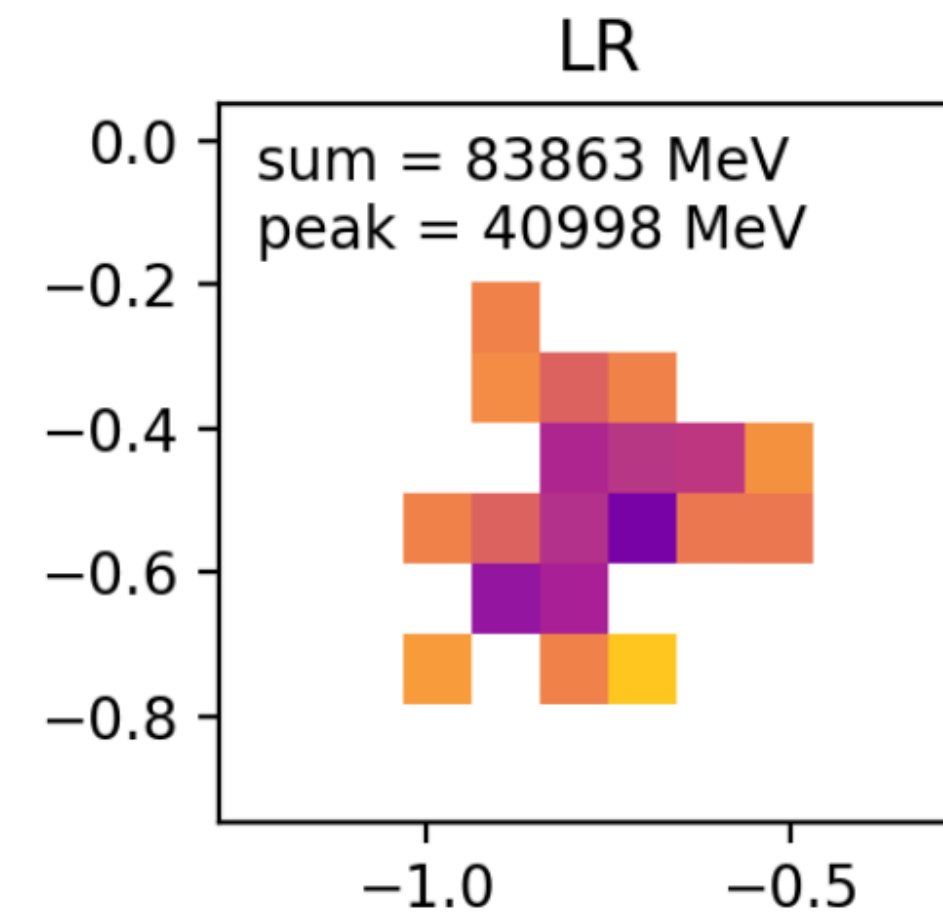
But, most importantly,

- ◆ **We shouldn't look at it in isolation**
- ◆ Primary goal -
 - ➔ **Assist downstream reconstruction task**
 - ➔ Multi-particle example demonstrates **cardinality**
 - ➔ Plenty more to look at (ongoing study)



But, most importantly,

- ◆ **We shouldn't look at it in isolation**
- ◆ Primary goal -
 - ➔ **Assist downstream reconstruction task**
 - ➔ Multi-particle example demonstrates **cardinality**
 - ➔ Plenty more to look at (ongoing study)



Thanks!