

# Autoencoders for Compression in LUX-ZEPLIN

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January 16, 2025

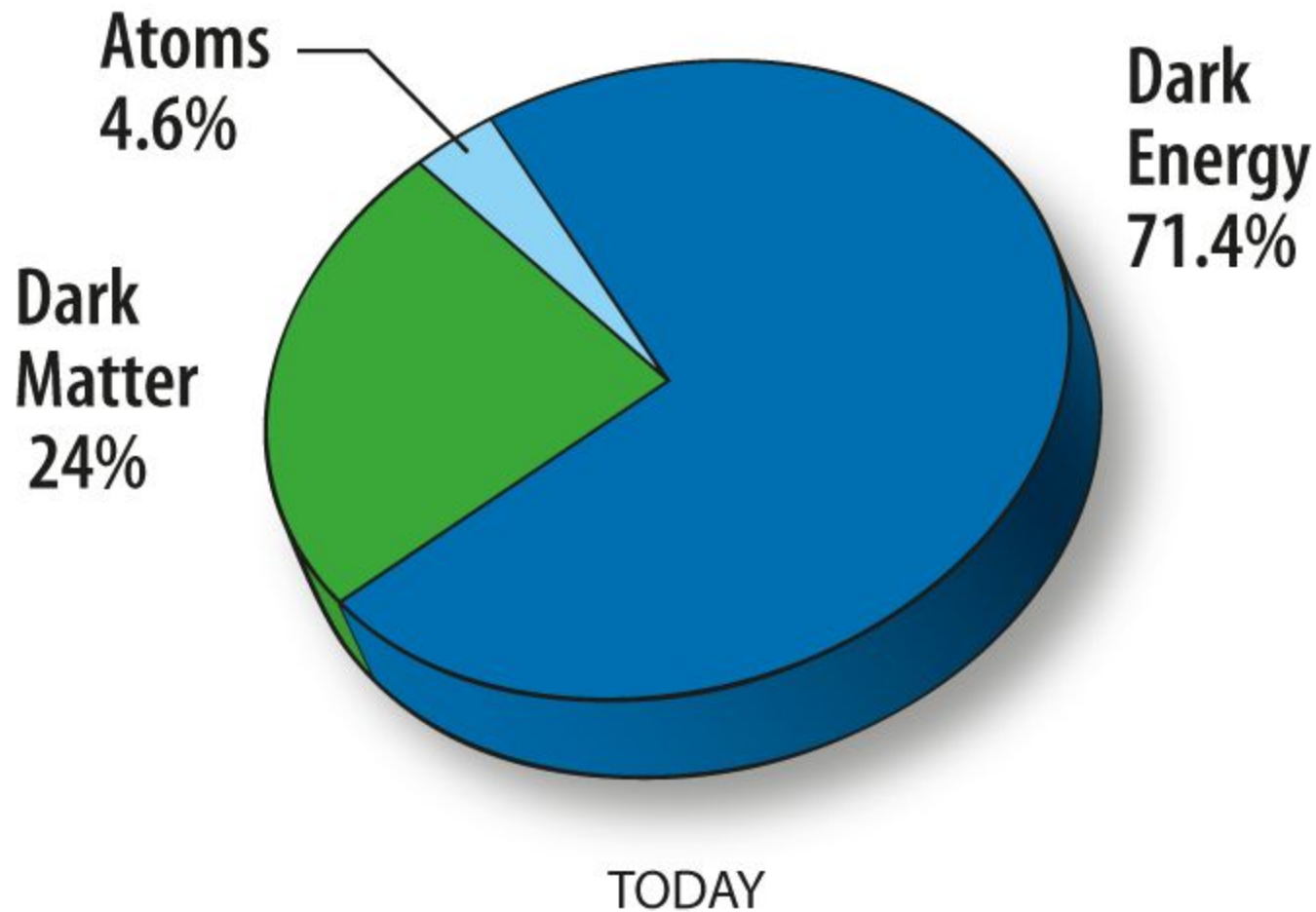


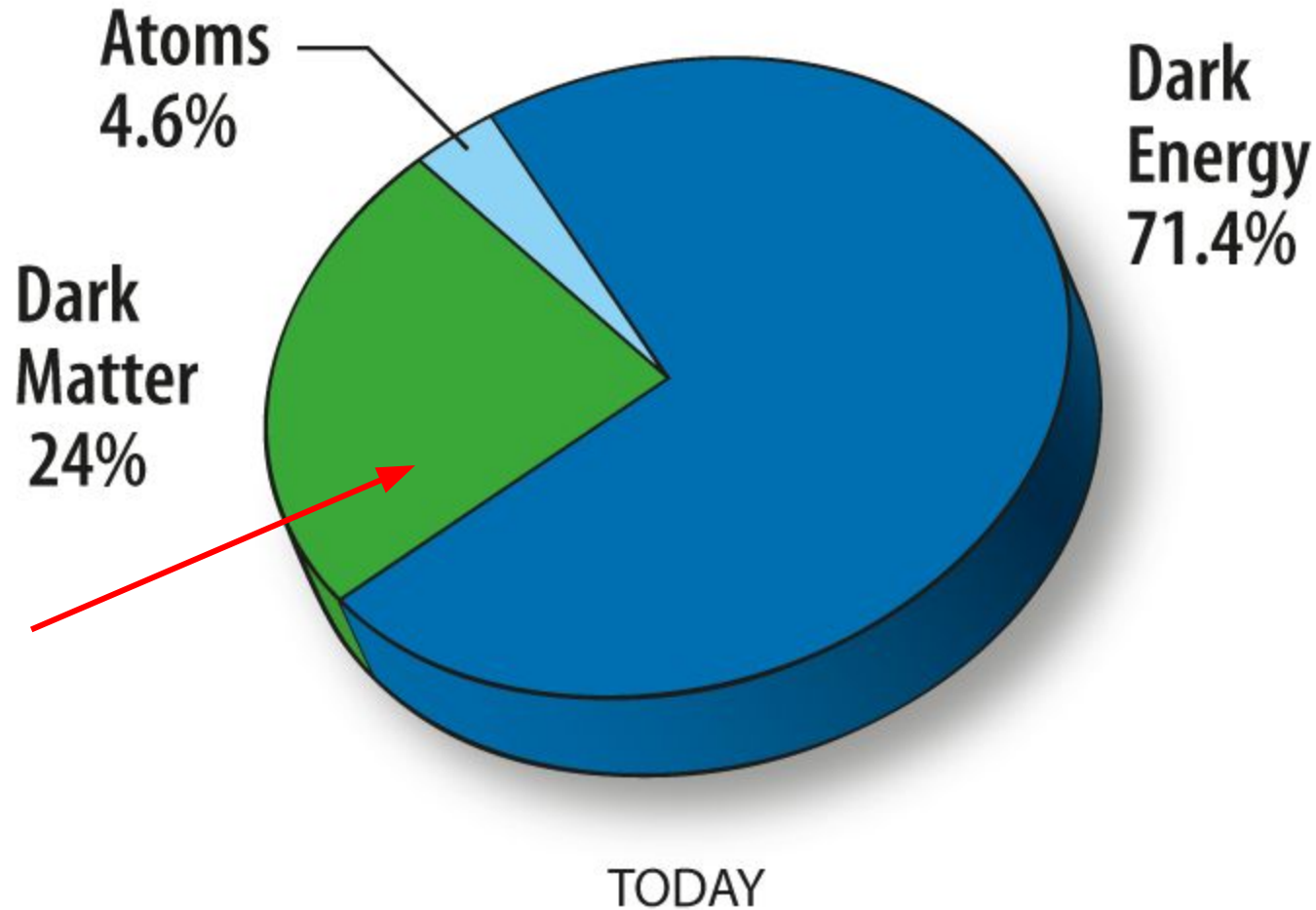
# Introduction

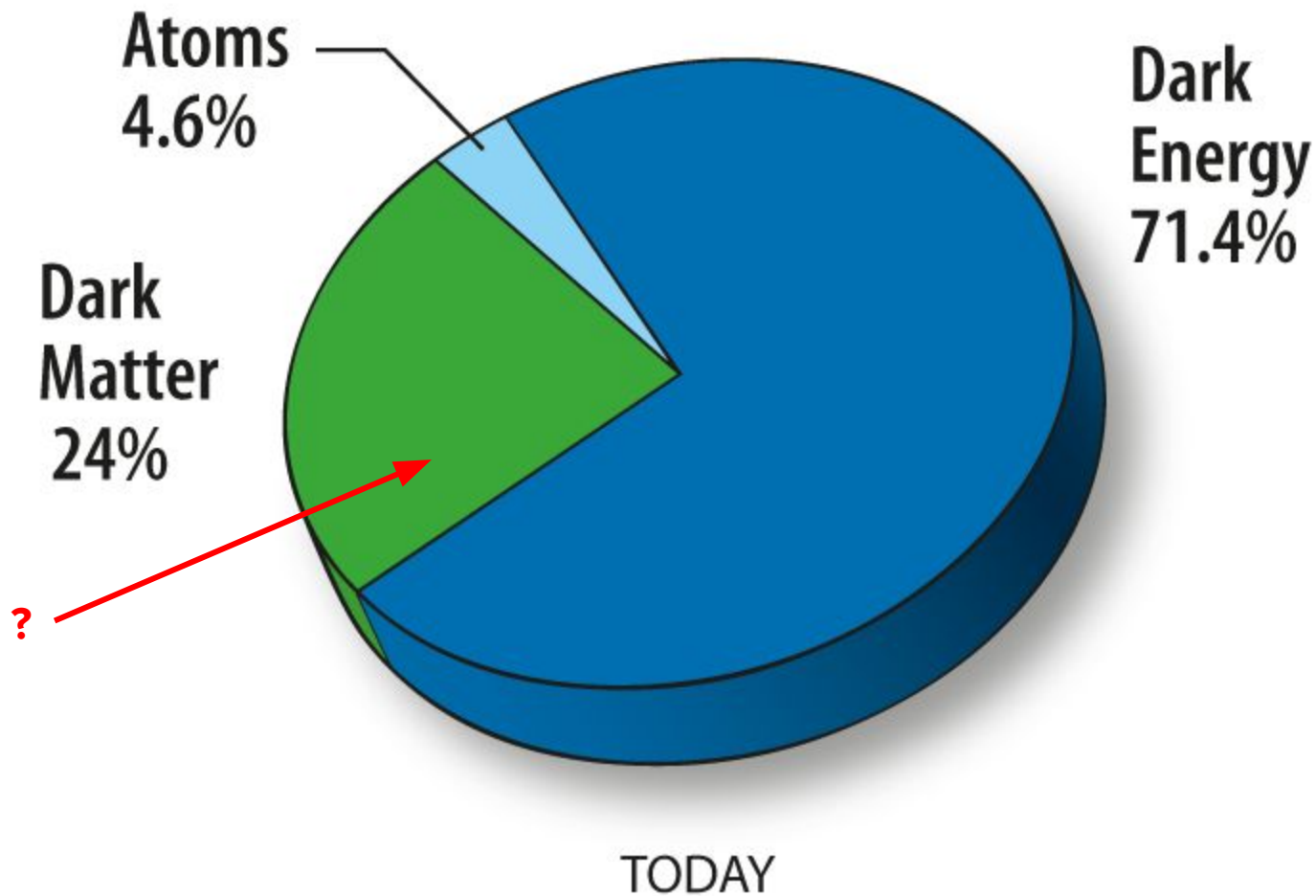
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# Introduction









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- How do we know this?

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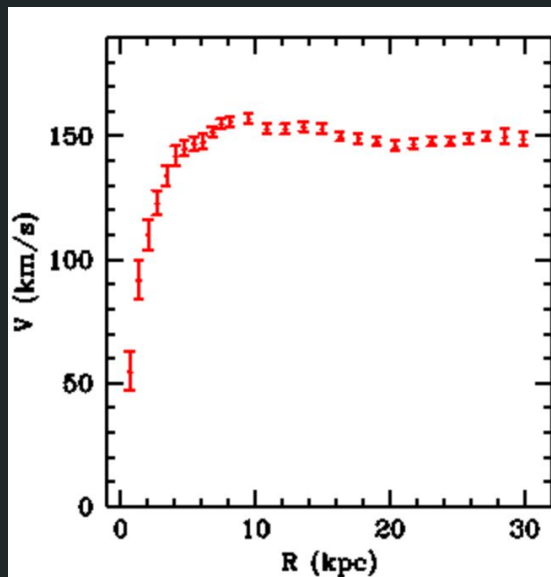
- How do we know this?
  - The most commonly discussed measurement is of *galaxy rotation curves*





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- How do we know this?
  - The most commonly discussed measurement is of *galaxy rotation curves*



*The rotation curve for the galaxy NGC3198 from Begeman 1989*

- By naive Newtonian gravity, velocity should fall off with distance
- Can infer that the M/L ratio is larger than can be expected from just the luminous matter

<https://w.astro.berkeley.edu/~mwhite/darkmatter/rotcurve.html>



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  - “Normal” matter, (mostly) what you see around you
  - Dark Energy (don’t know what it is)
  - **Dark Matter** (don’t know what it is)



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      - Some type of new particle(s)
      - Modified/new theories of gravity



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        - Axions
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          - How do we directly detect particles like these?



# A BIG VAT OF LIQUID



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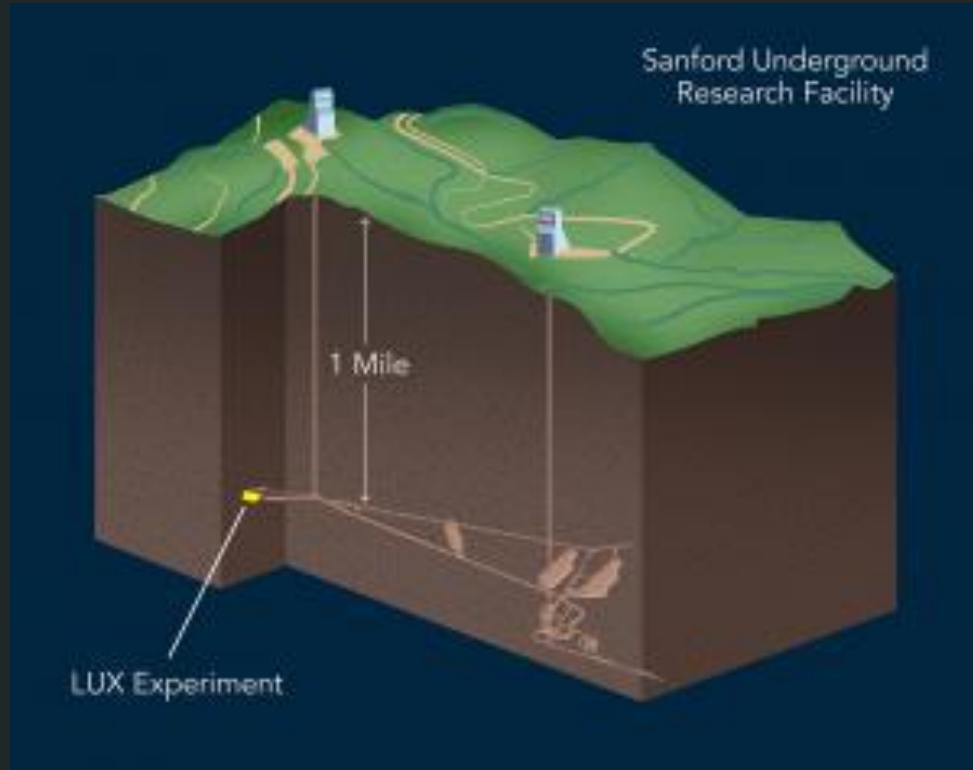
# A BIG VAT OF LIQUID

Specifically, liquid xenon



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# Sanford Underground Research Facility (SURF) Lead, South Dakota, USA

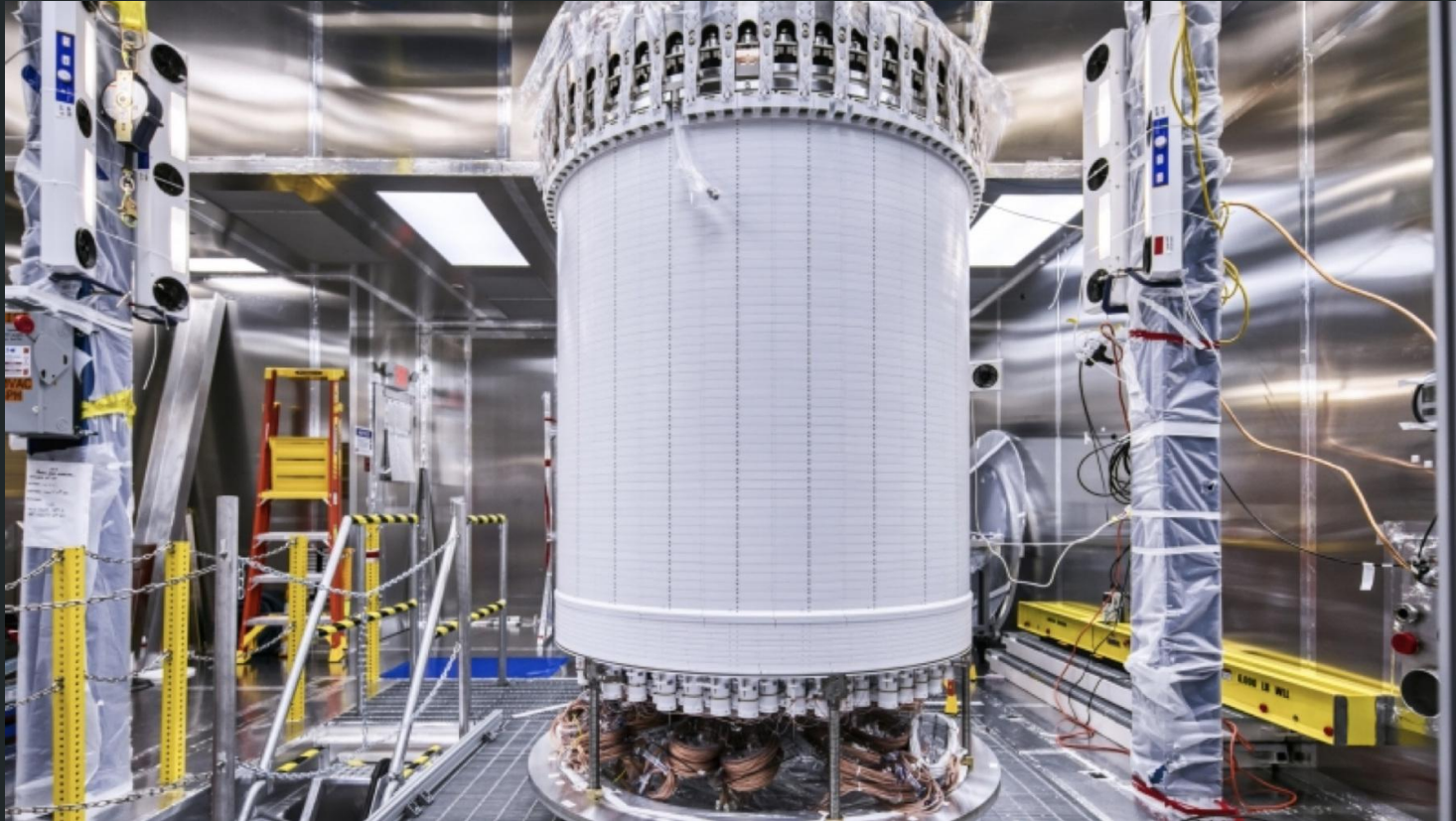


- 4850 ft (1478.28 m)
- 4300 meter water equivalent rock



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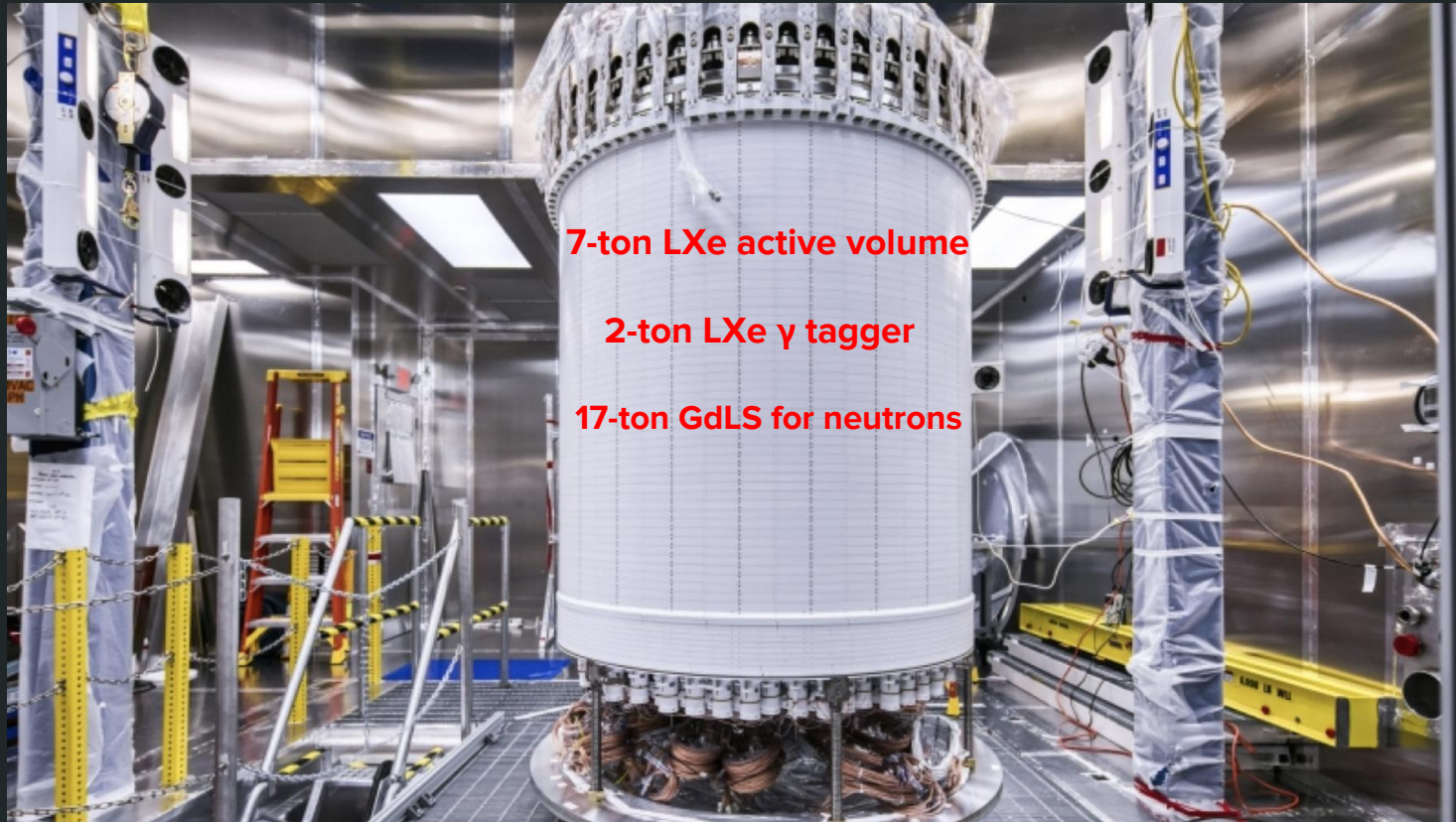
# Large Underground Xenon-ZonEd Proportional scintillation in LIquid Noble gases LUX-ZEPLIN (LZ) <https://lz.lbl.gov/>



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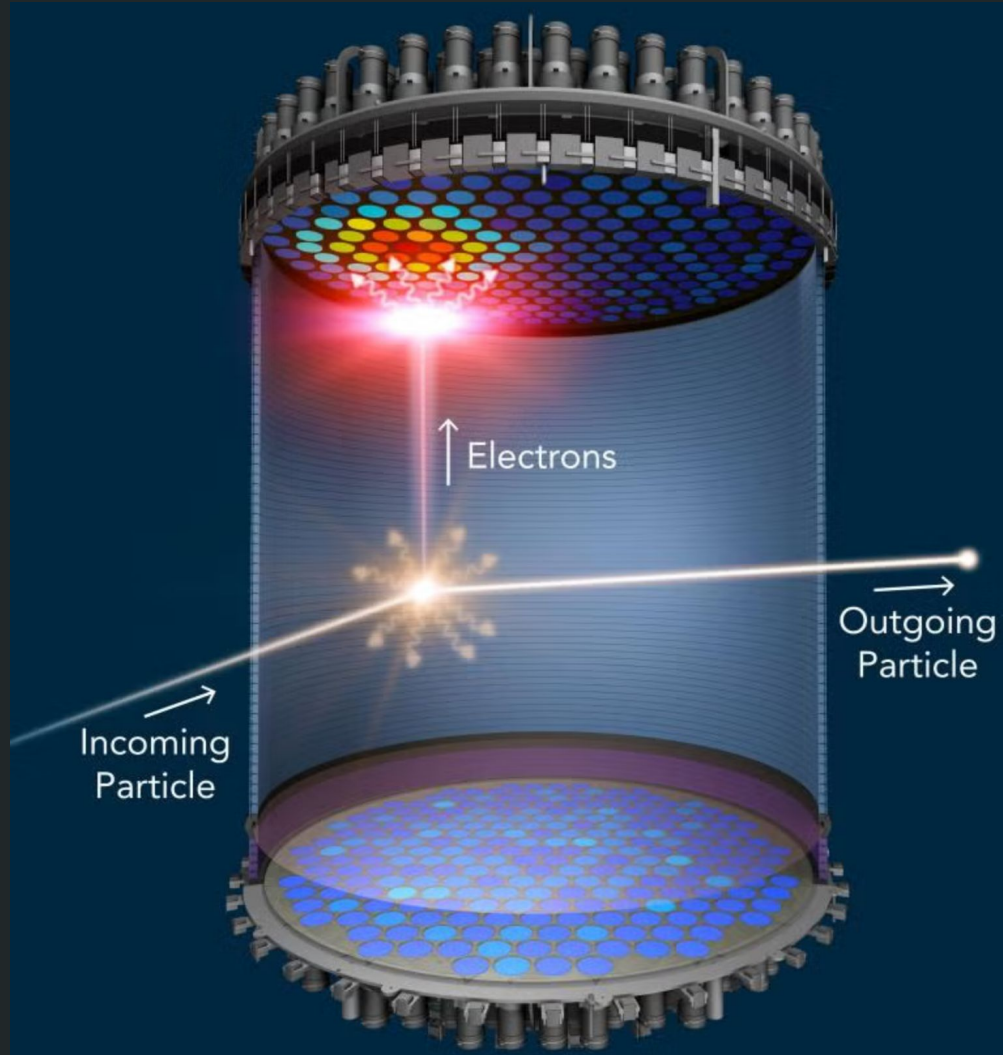
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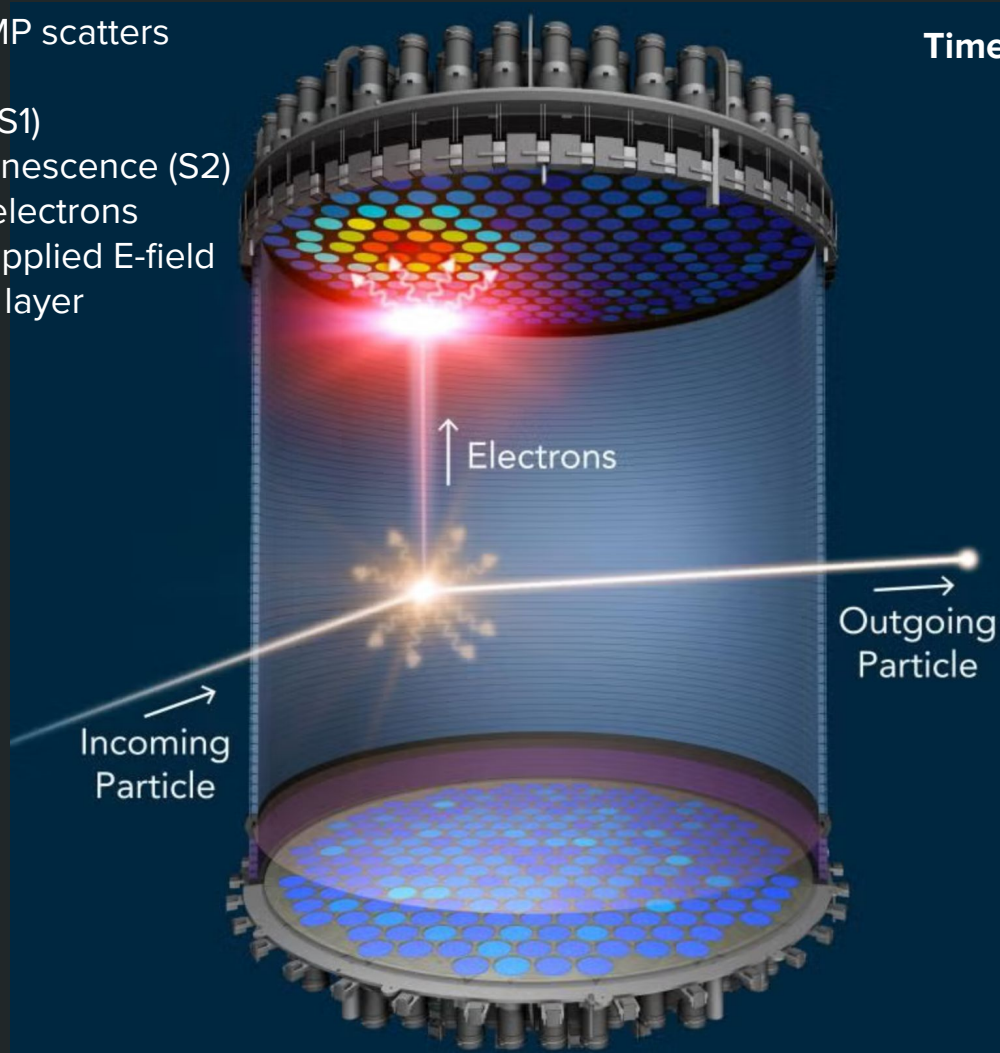


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When a particle like a WIMP scatters

- Prompt UV photon (S1)
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  - Produced by electrons drifting in an applied E-field through a gas layer

Time Projection Chamber (TPC)

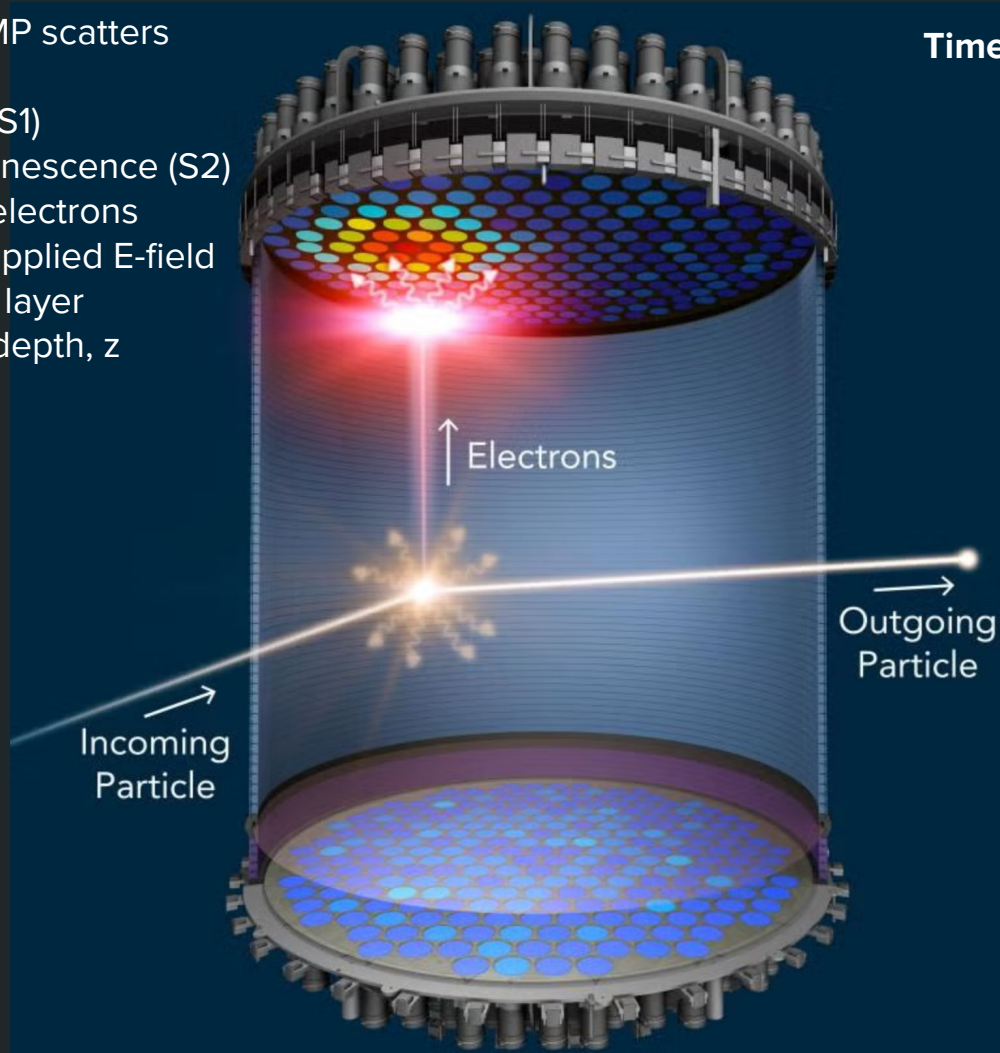


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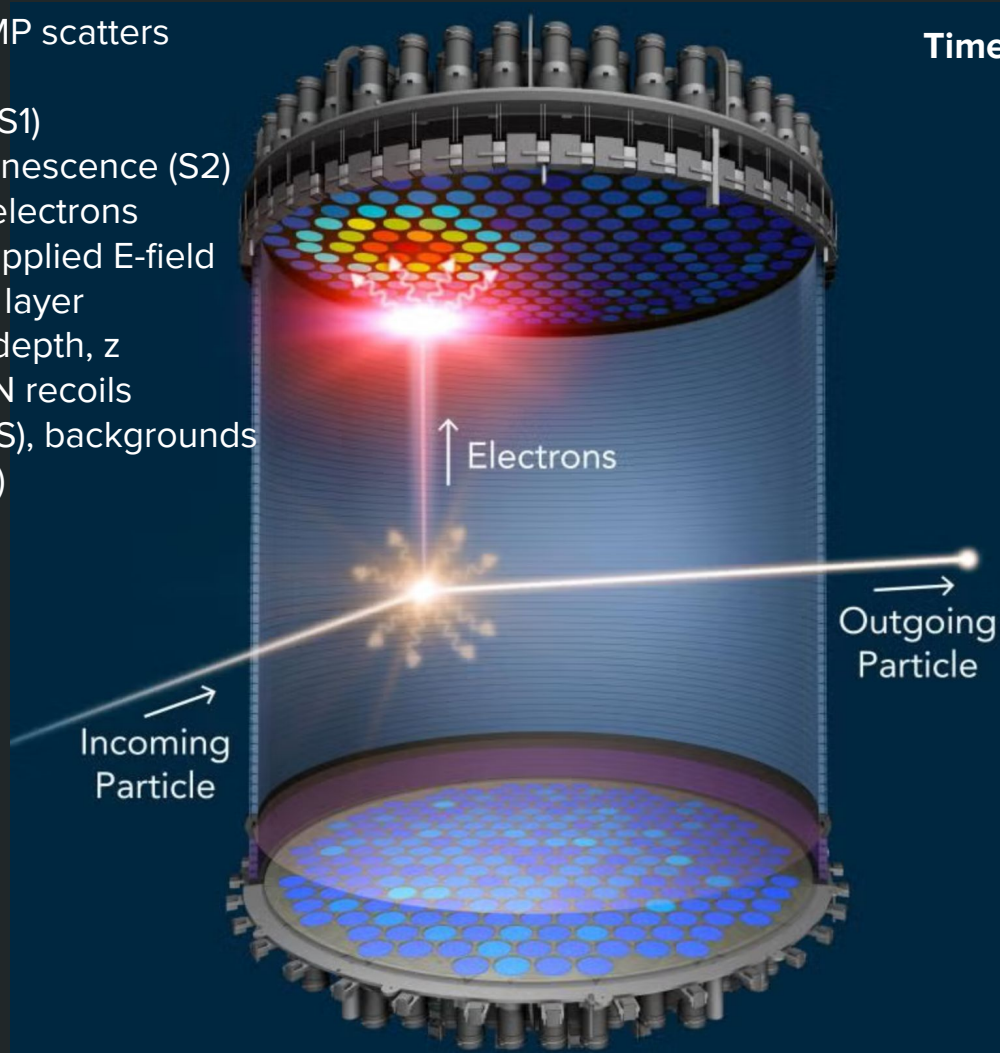


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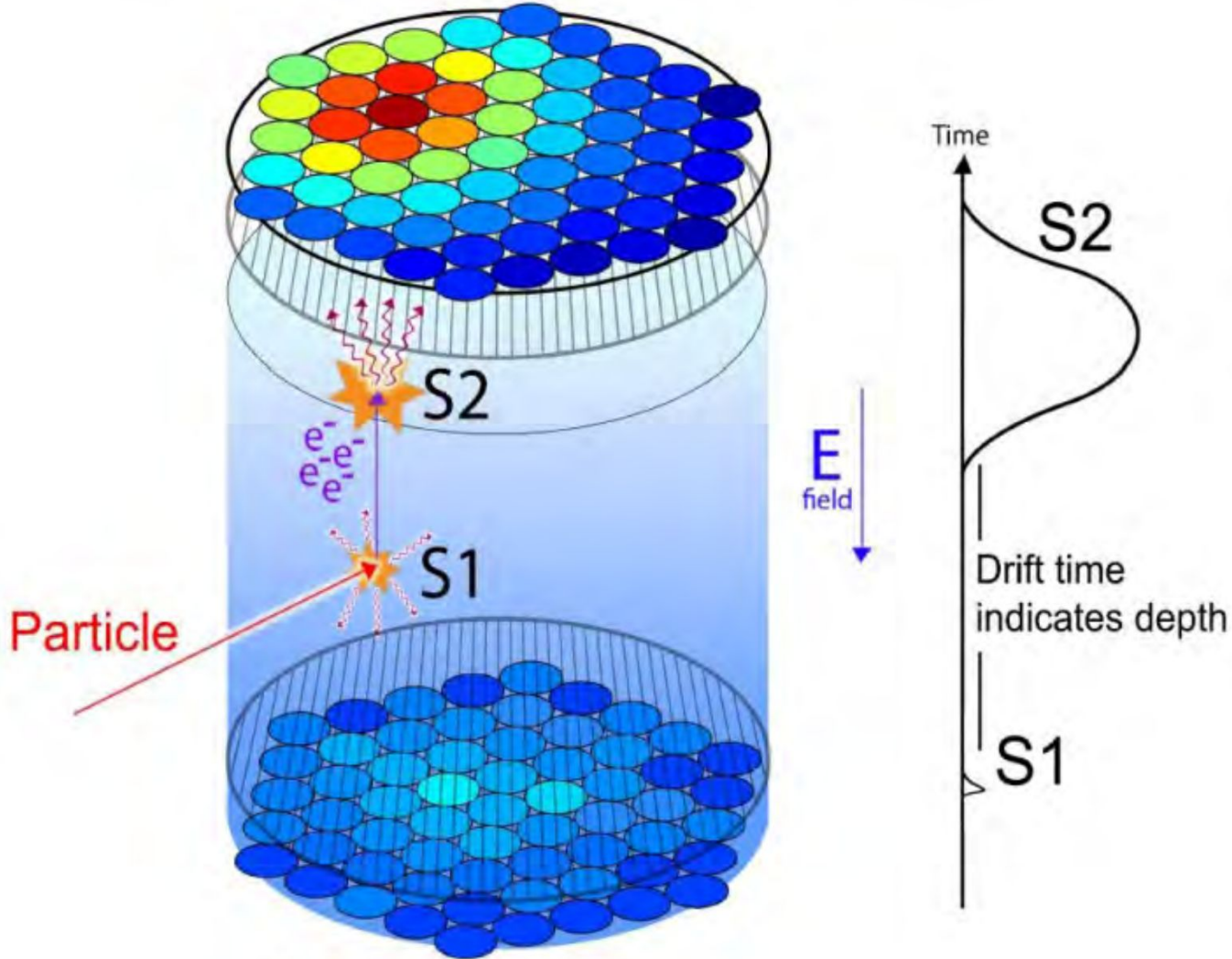
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- Signals from WIMP-N recoils are single scatter (SS), backgrounds are multiscatter (MS)

Time Projection Chamber (TPC)

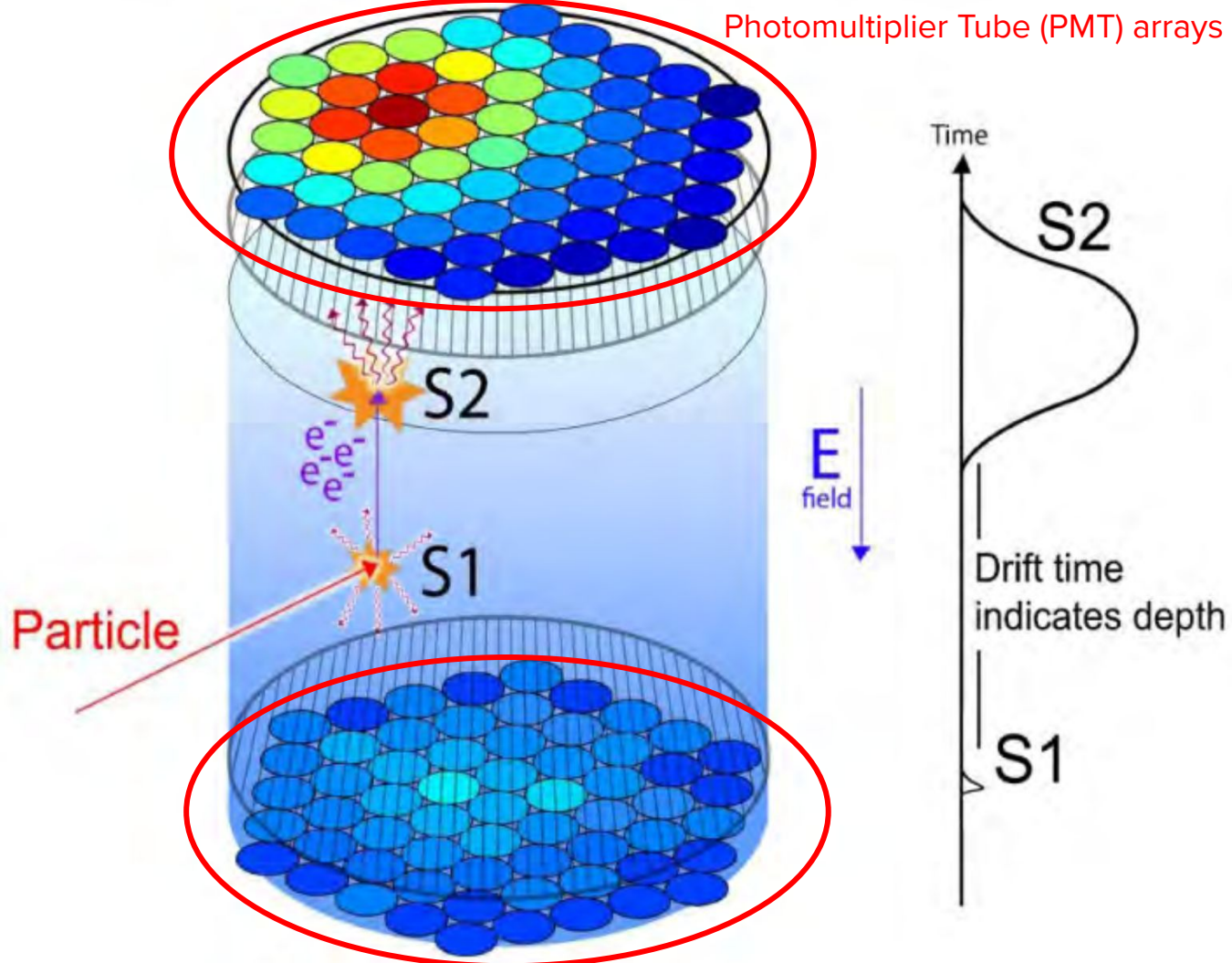


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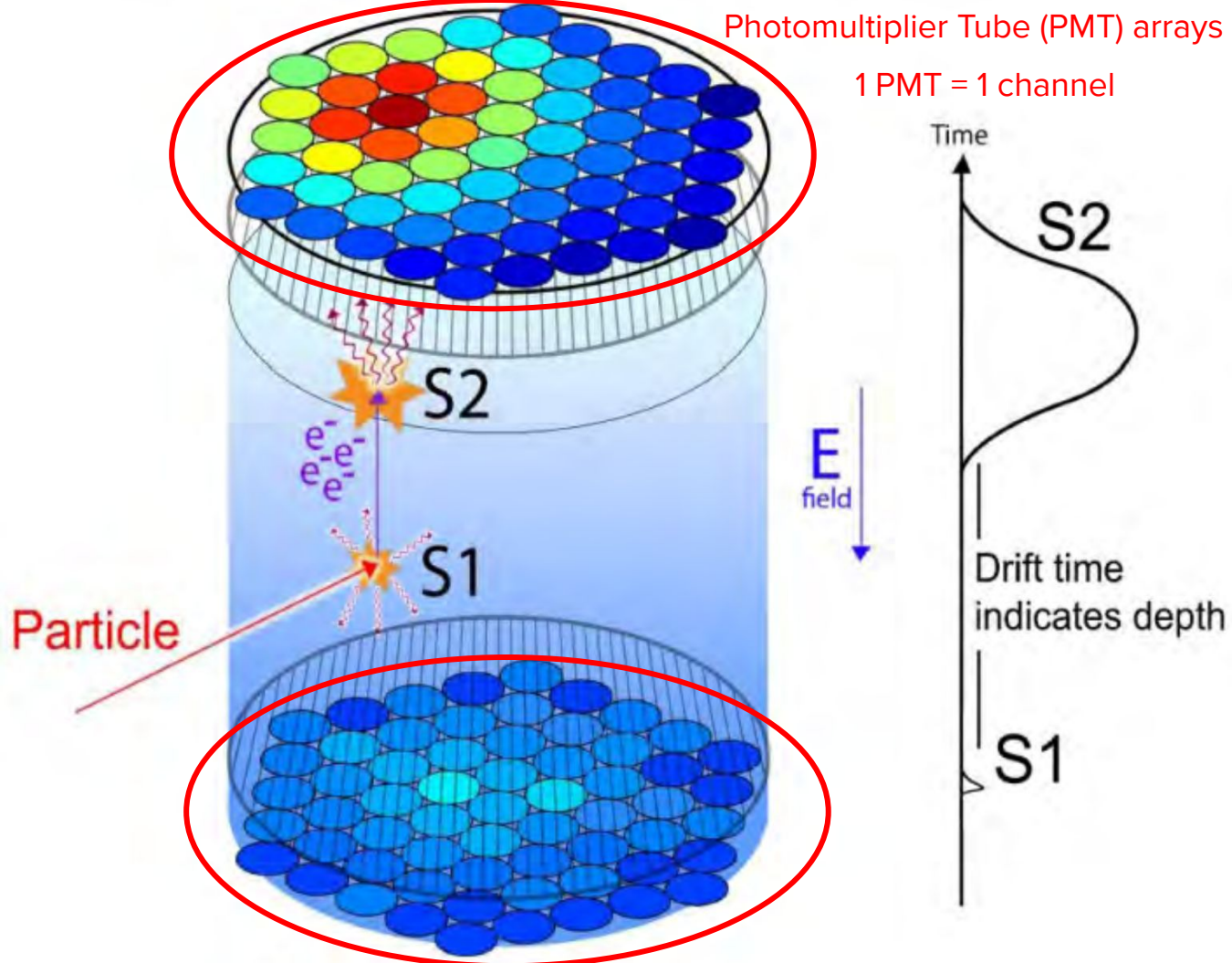


# Photomultiplier Tube (PMT) arrays



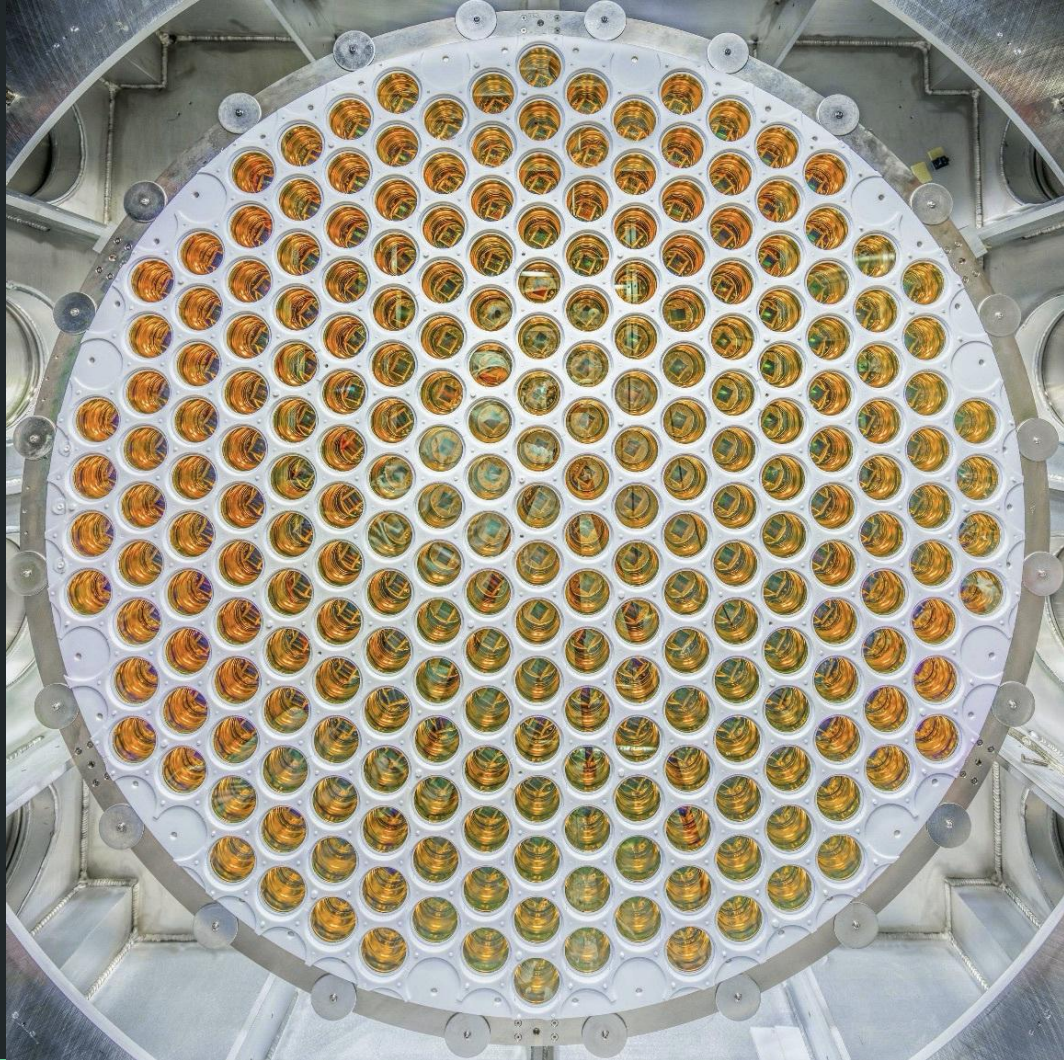
Photomultiplier Tube (PMT) arrays

1 PMT = 1 channel

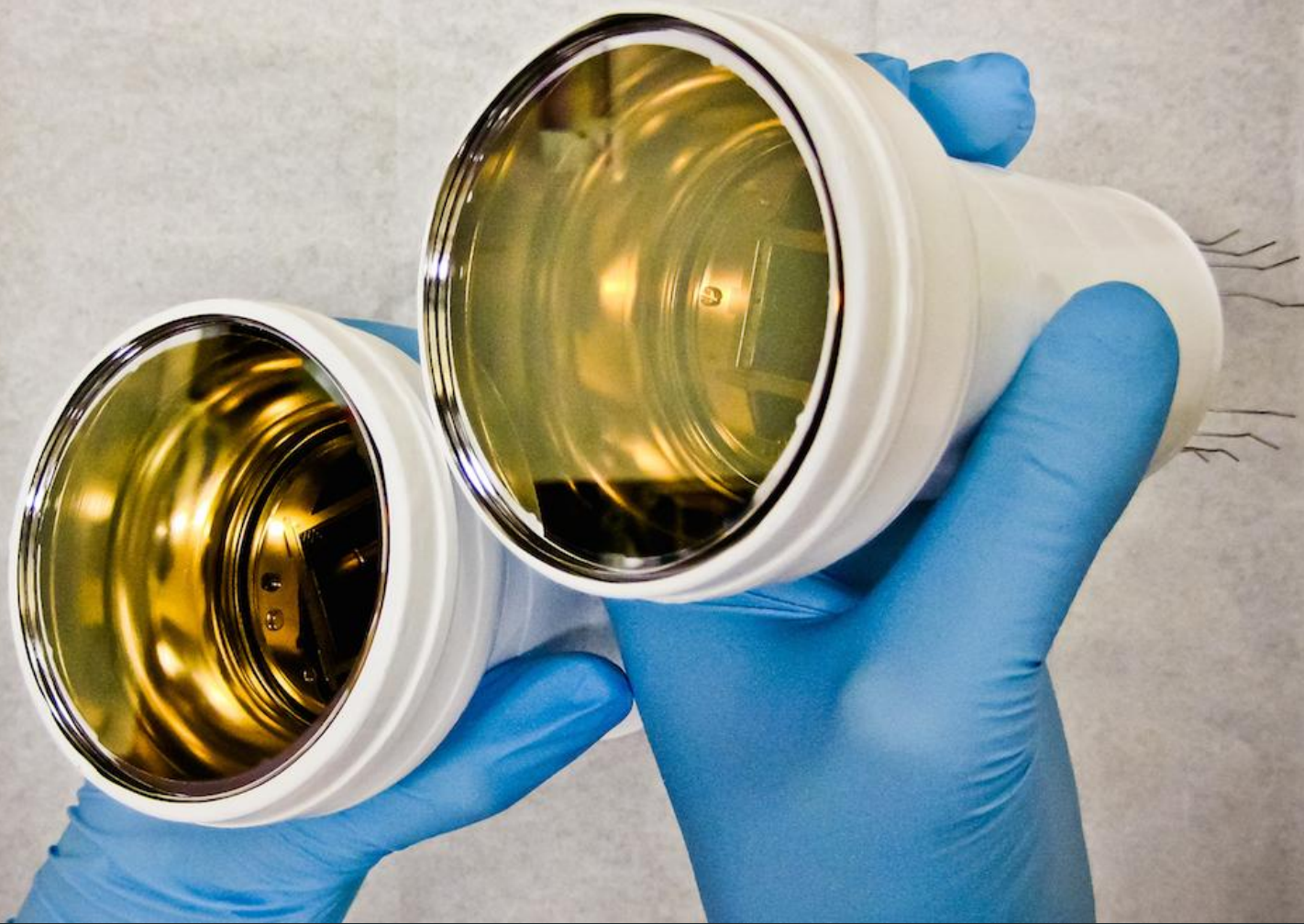


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# Introduction

- Data from these PMTs are recorded by the readout electronics
- Data is stored on disk for further analysis
- Large quantities of data (1 PB/year)
  - Not all of the data, or features of the data, are necessarily useful to perform physics analyses that search for dark matter candidates, e.g. WIMPs
  - Can the amount of data that is saved be reduced while still having high-fidelity physics searches?



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- It is possible to use *machine learning* to compress the data.
  - Specifically, we want to use **autoencoders**



# Autoencoders

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- Types of ANNs that perform an unsupervised learning task to learn representations of the data called **latent representations**



# Autoencoders

- Types of ANNs that perform an *unsupervised* learning task to learn representations of the data called **latent representations**
  - This *latent space* typically has a lower dimensionality than the original data
- Learns by trying to copy the input to the output
  - I.e., tries to learn the identity function
  - But this done under some constraints, e.g. forcing the latent space to be of smaller dimension or adding some noise
  - Forces the model to learn efficient representations of the input data



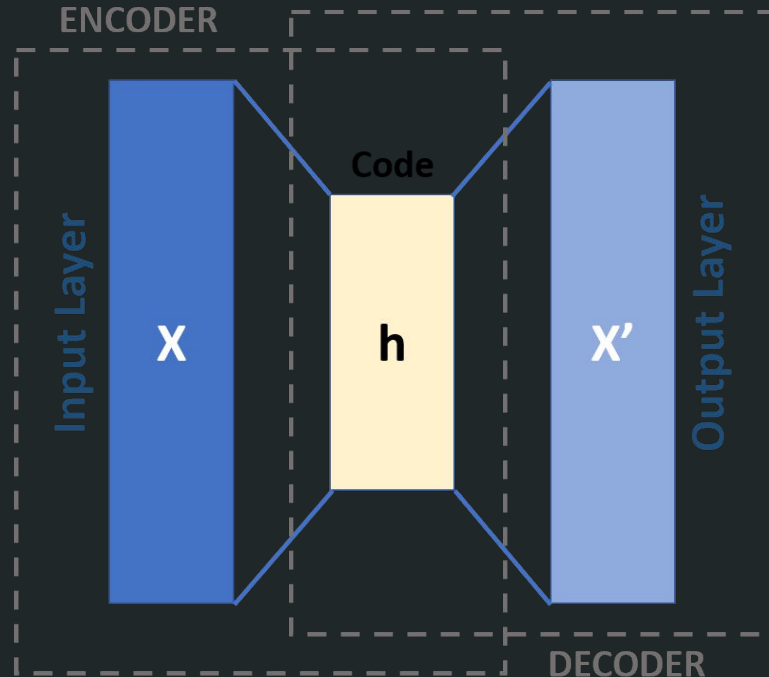
# Autoencoders

- Autoencoders are composed of two parts:
  - Encoder
  - Decoder



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# Application of Autoencoders to LZ Channel Data

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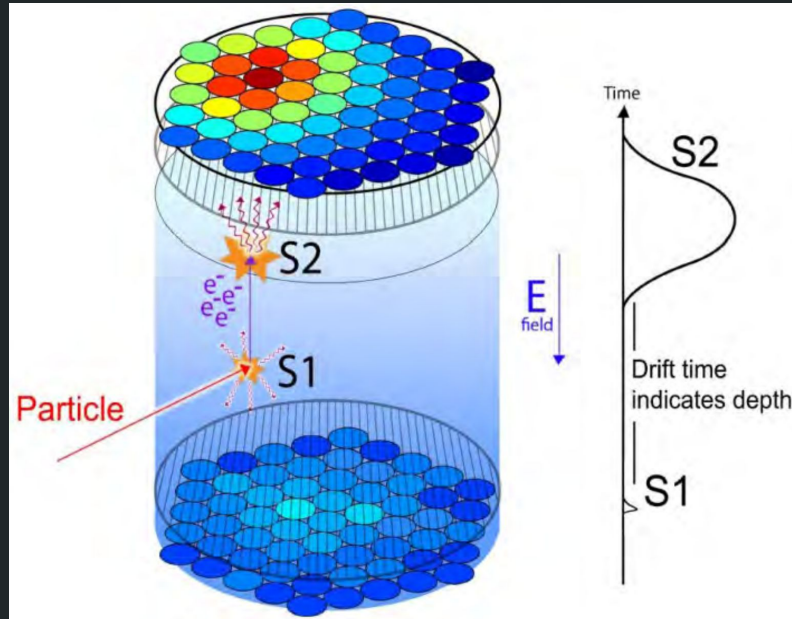
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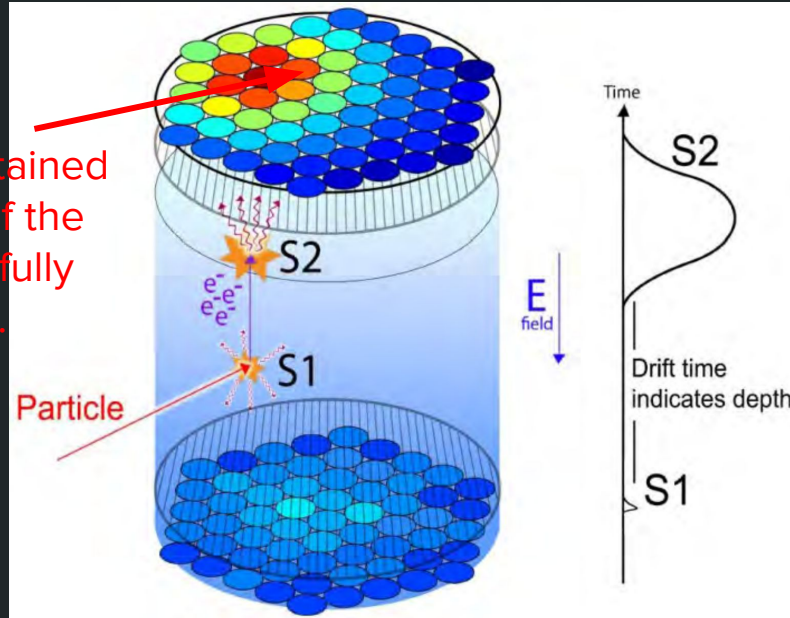
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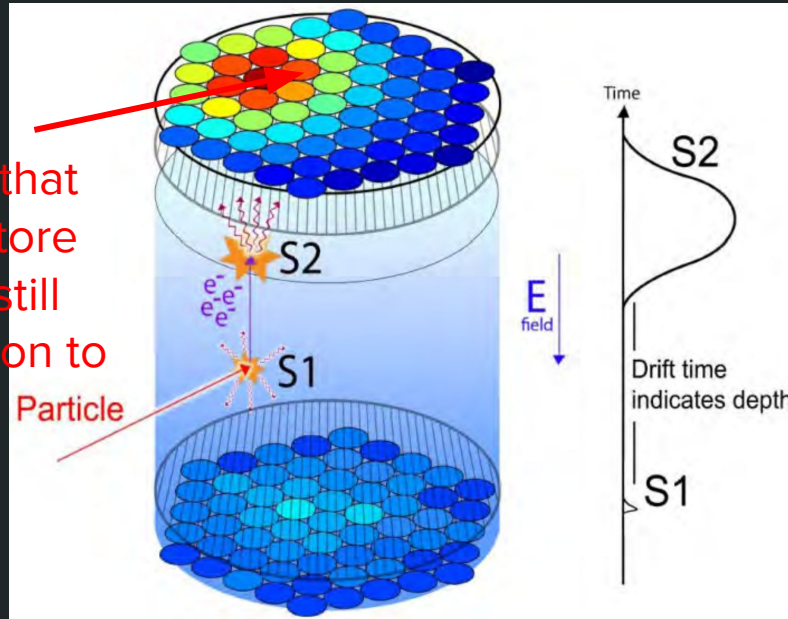
Want to compress the data obtained from simulated PMTs and see if the compressed data can still faithfully represent the the original data.



# Application of Autoencoders to LZ Channel Data

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If we can, then we have demonstrated a technique that can be used to efficiently store data, reducing costs while still providing enough information to search for dark matter.



For this hands-on we will use *\*simulated\**, LZ-like events, not real LZ data.

# Summary and Conclusion

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- Have observed astrophysical phenomena, such as seen in galaxy rotation curves, of unknown origin, called “Dark Matter”
- Hypothesize that a new class of particle exists to try and explain the phenomena
  - WIMPs
- Built a large LXe detector to try and detect these particles coming from outer space
- Want to use machine learning, specifically autoencoders, to compress the detector data to perform efficient searches for DM
  - Generally applicable technique



# References

1. [https://wmap.gsfc.nasa.gov/universe/uni\\_matter.html](https://wmap.gsfc.nasa.gov/universe/uni_matter.html) (energy content of the universe)
2. <https://w.astro.berkeley.edu/~mwhite/darkmatter/rotcurve.html> (galaxy rotation curves)
3. LZ Sites
  - a. <https://lz.lbl.gov/>
  - b. <http://lz.ac.uk/>
  - c. <https://lz.slac.stanford.edu/>
4. <https://arxiv.org/abs/2410.17036> (latest results)
5. Non-LZ direct detection experiment websites
  - a. <https://xenonexperiment.org>, [http://xenon.astro.columbia.edu/XENON10\\_Experiment/](http://xenon.astro.columbia.edu/XENON10_Experiment/)
  - b. <https://depts.washington.edu/admx/>