



Visualization of LEGEND-200 data for high-dimensional classification algorithms

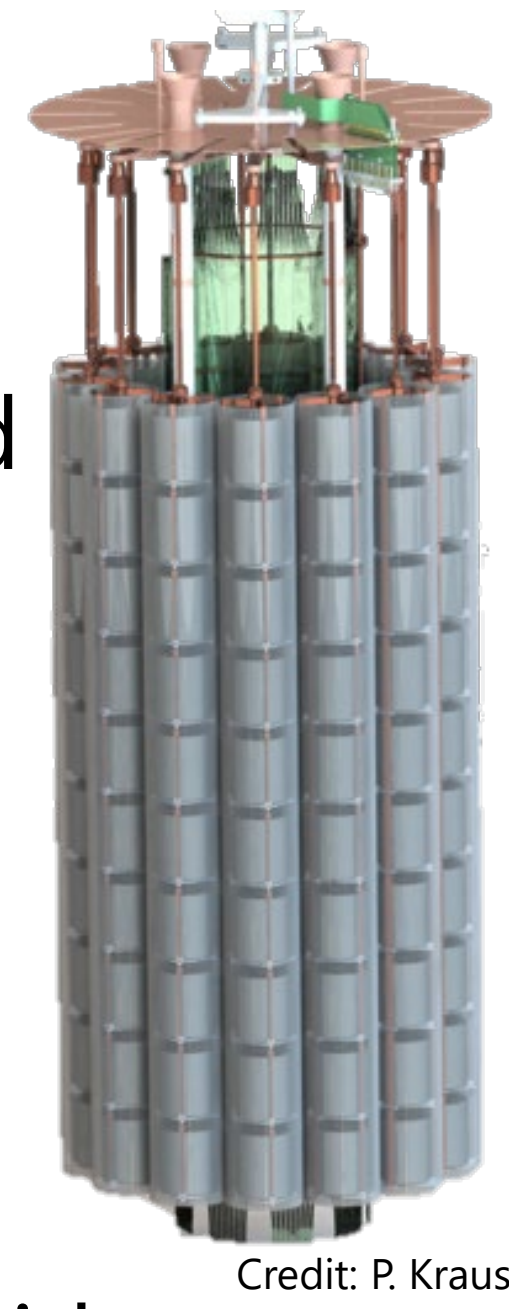
Miguel A. B. Schott, Esteban León, Julieta Gruszko, Aobo Li

Department of Physics and Astronomy



LEGEND - 200

- The Large Enriched Germanium Experiment for Neutrinoless $\beta\beta$ (LEGEND) follows from previous neutrino experiments GERDA and MAJORANA.
- The goal is to detect the hypothetical nuclear decay $(A, Z) \rightarrow (A, Z+2) + 2e^-$
- LEGEND-200 overcomes backgrounds through a configuration of ^{76}Ge detectors with a total mass of 200kg.
- Observation of $0\nu\beta\beta$ would demonstrate Lepton Number Violation[1].

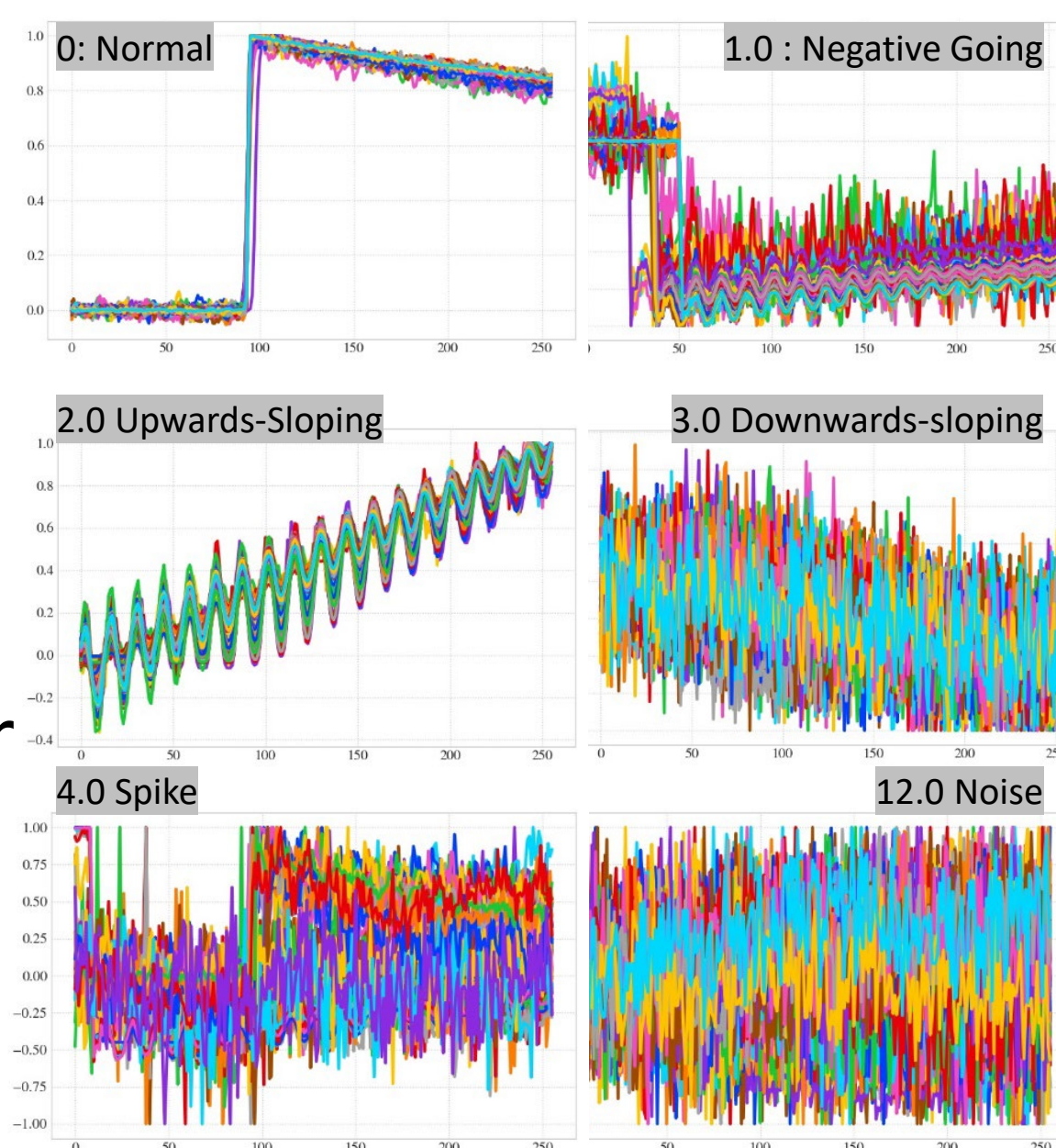


Credit: P. Krause

Data Cleaning

3-step traditional data cleaning process:

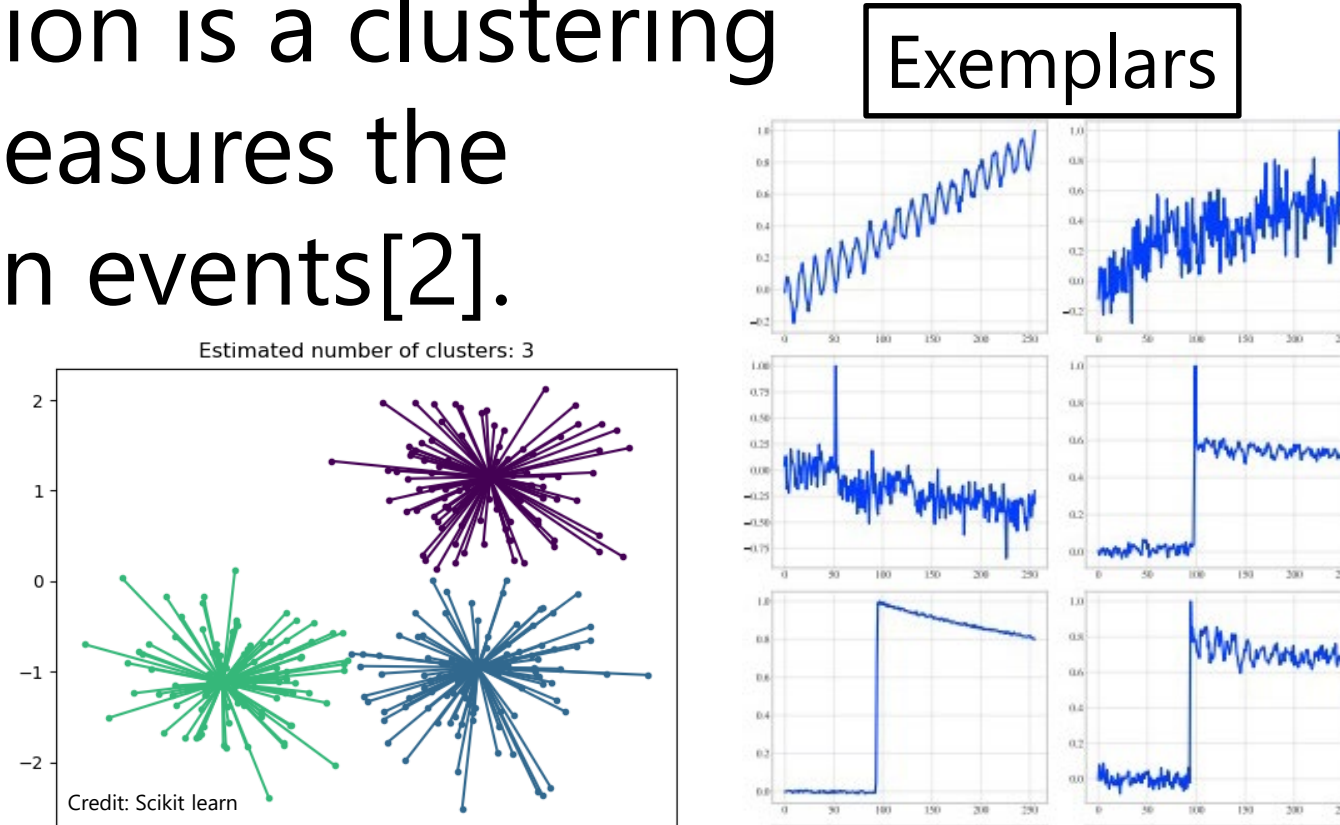
- Identify undesirable data.
- Create a parameter to isolate desirable results.
- Cut data that falls outside the created parameters.



- Traditional data cleaning may unintentionally remove desirable waveforms
- The goal of the machine learning implementation is to reduce sacrifice while automating the classification of all physical events
- The machine learning approach uses Affinity Propagation and Support Vector Machine (SVM) algorithms.

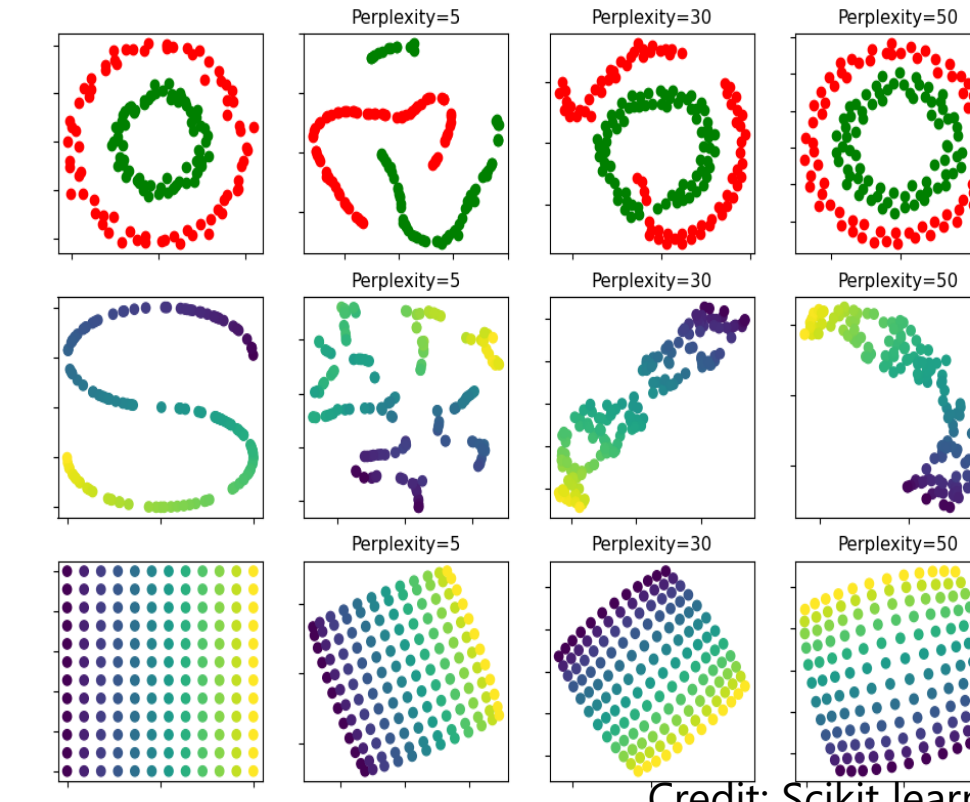
Affinity Propagation

- Affinity propagation is a clustering algorithm that measures the similarity between events[2].
- The center of a cluster is the exemplar.



t - Distributed Stochastic Neighbor Embedding (t-SNE)

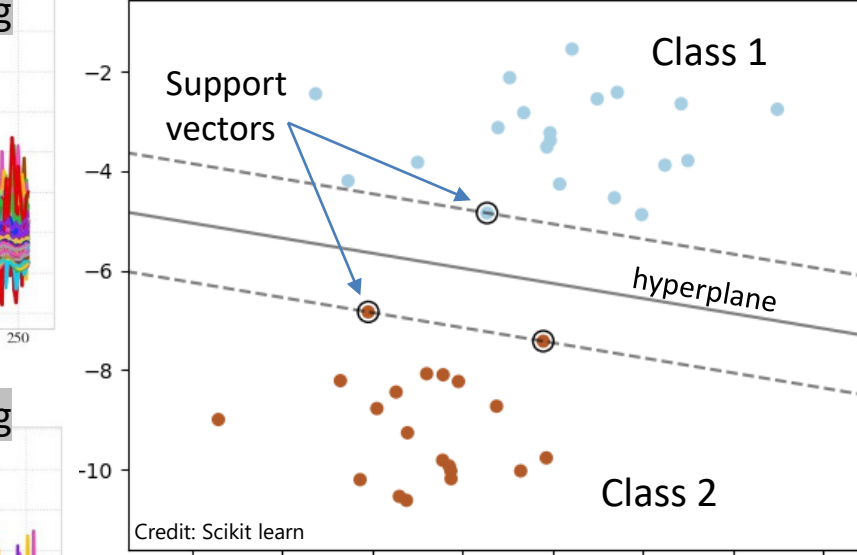
- t-SNE is used to visualize high-dimensional data[3].
- Hyperparameters:
- perplexity: number of neighbors considered for each point
 - learning rate: the step size of the algorithm



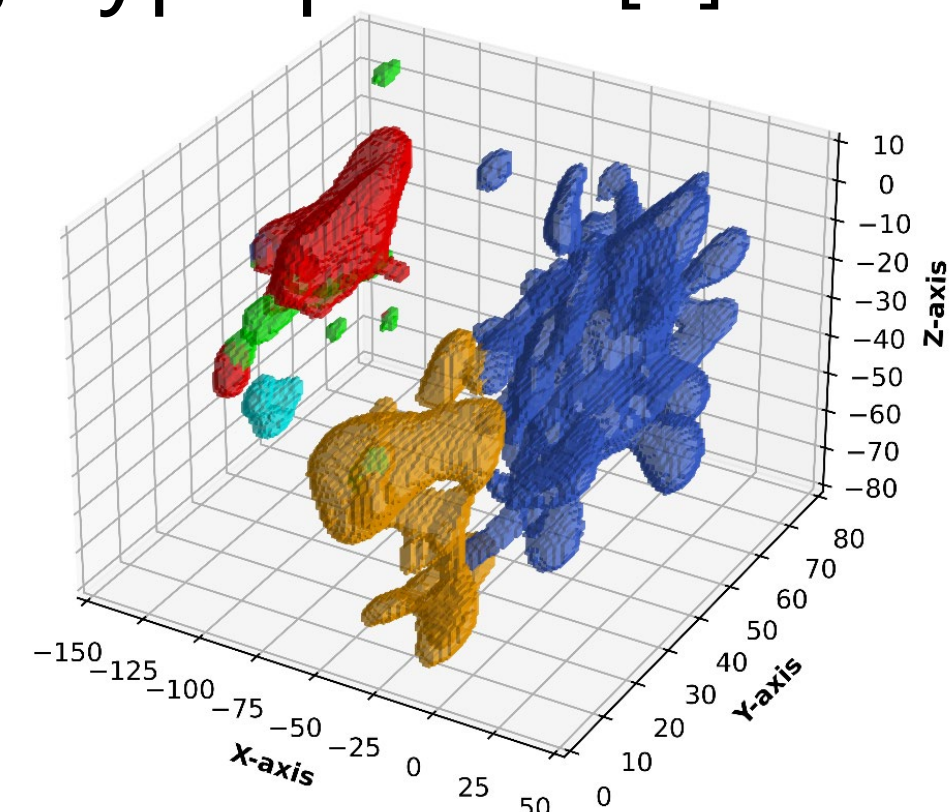
Perplexity may have dramatic impacts on the visualization

Support Vector Machine (SVM)

Support Vector Machines (SVM) classify objects by drawing hyperplanes [4].

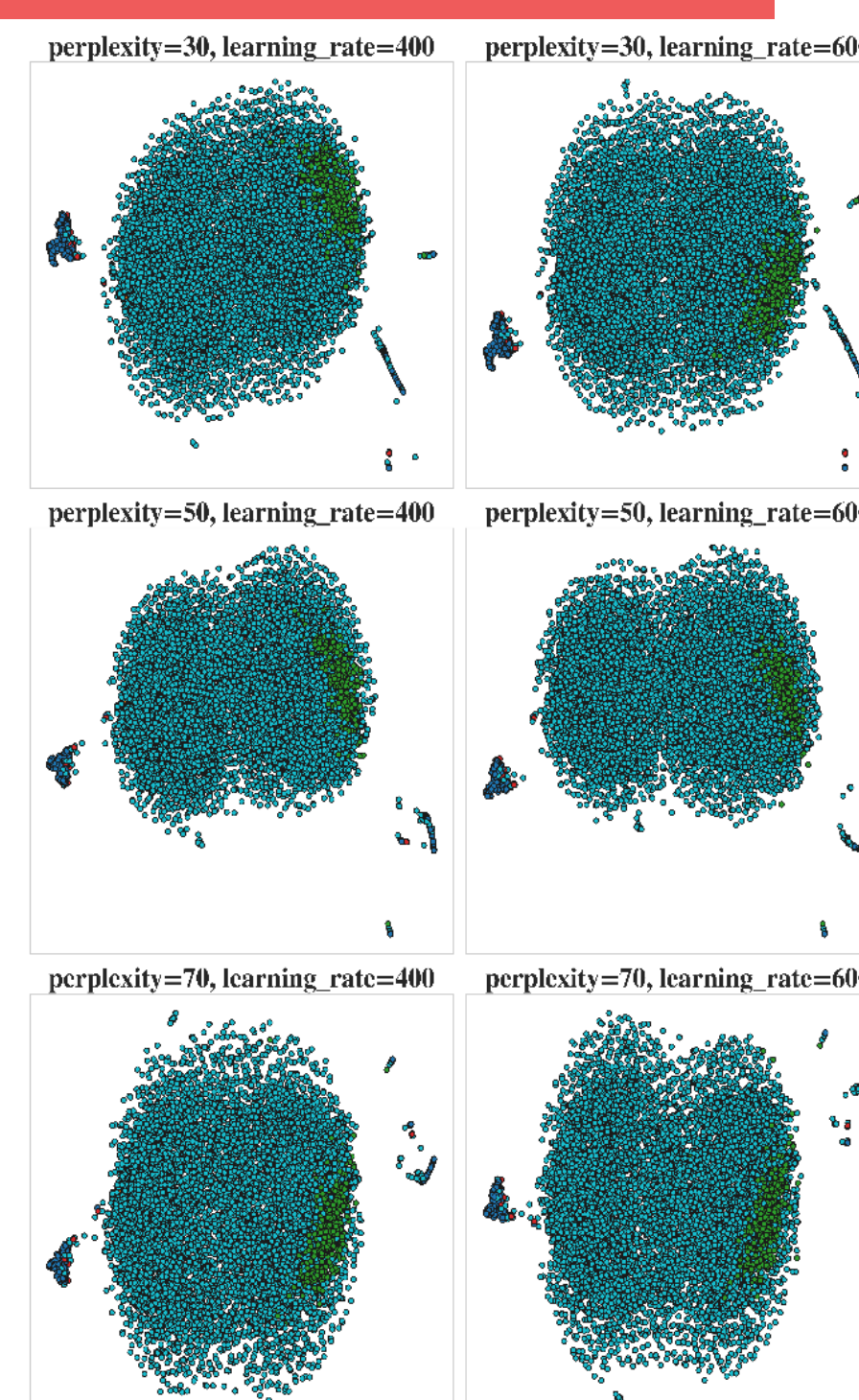


The SVM for LEGEND-200 data was trained from the 3-D t-SNE algorithm.

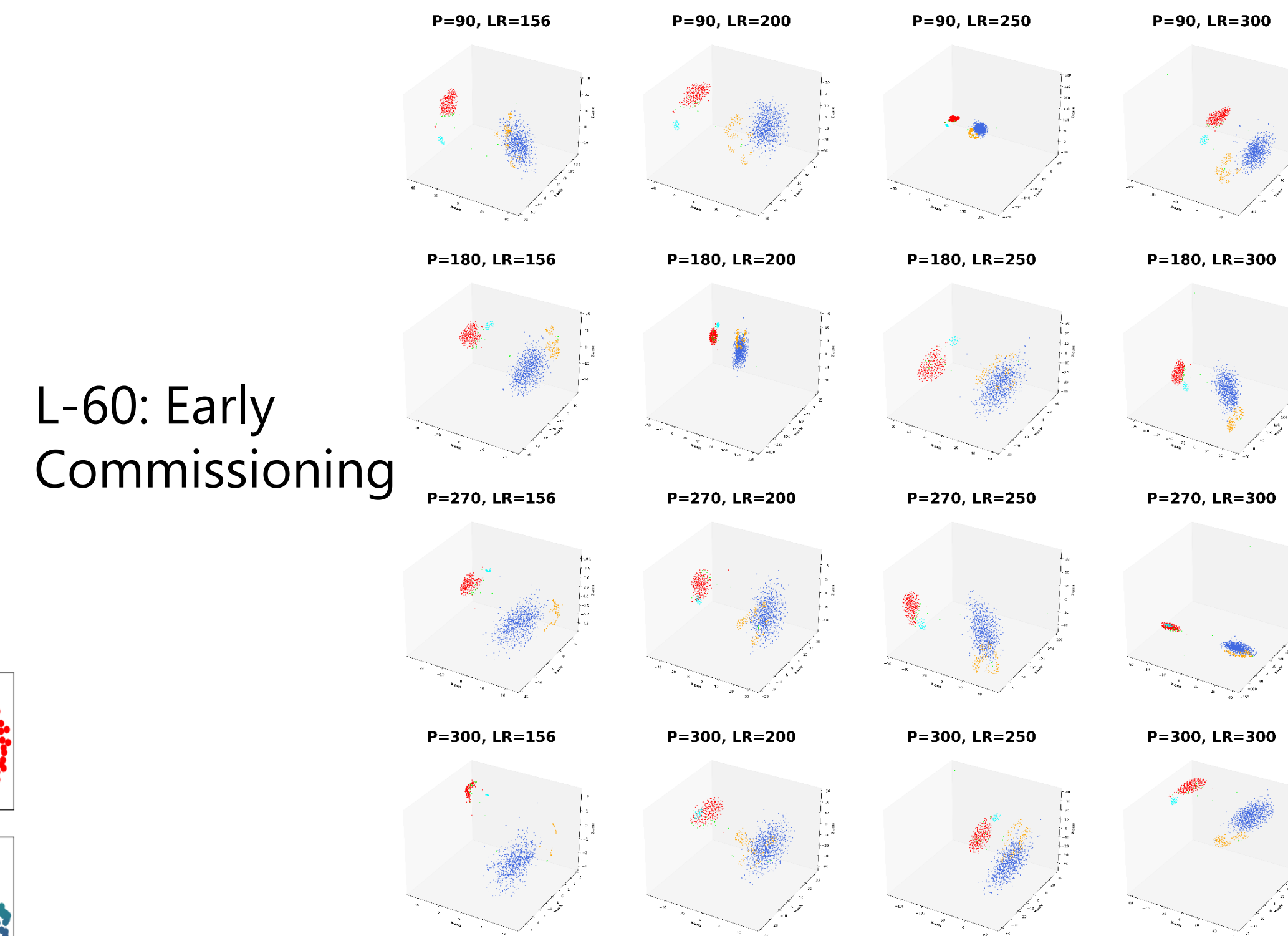


Methodology

- Different combinations of hyperparameters for t-SNE are made.
- Hyperparameters are optimized
- A 3-D SVM is trained on the selected t-SNE.
- The SVM is rendered in voxels



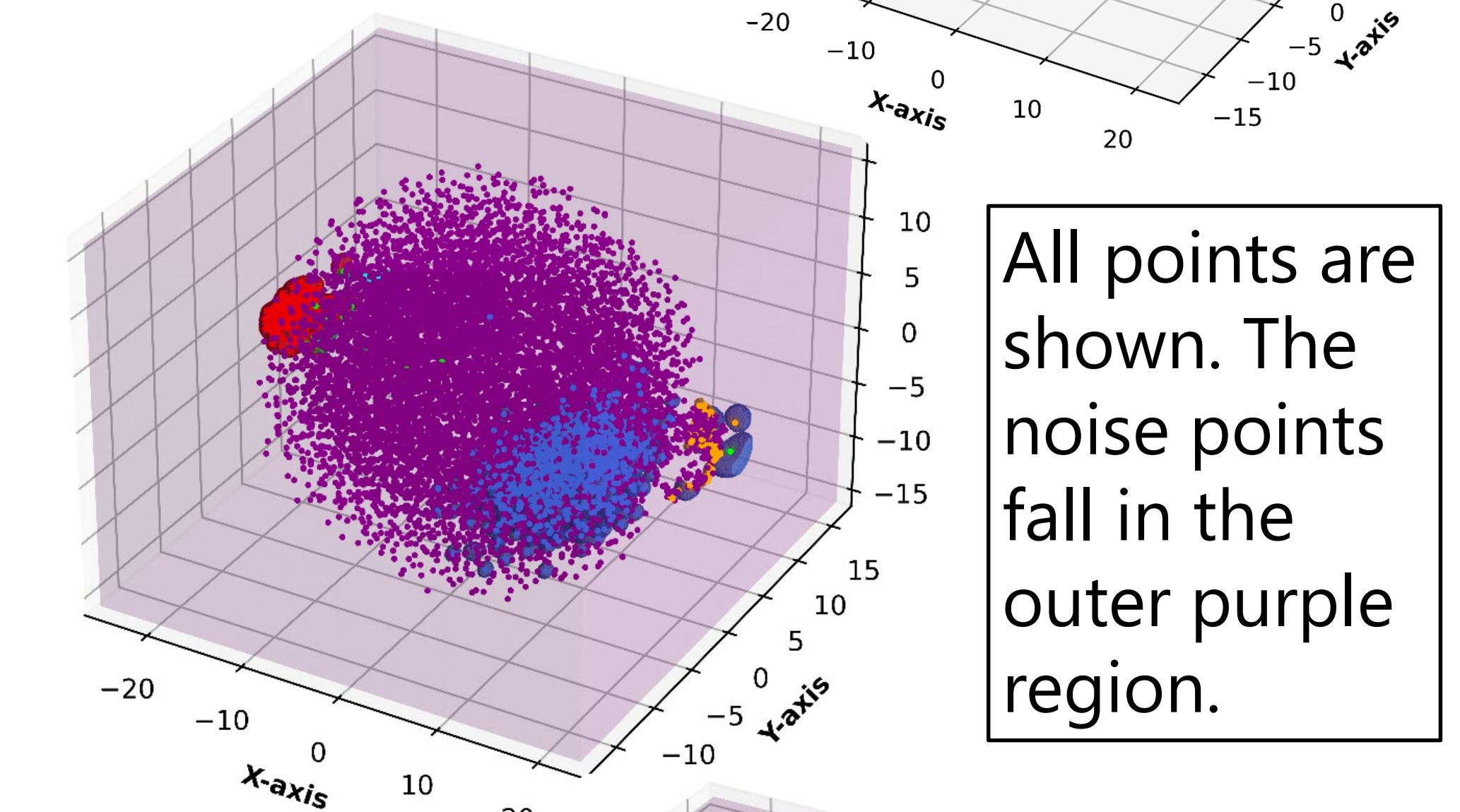
LEGEND - 60 Run 14 Data



L-60: Early Commissioning

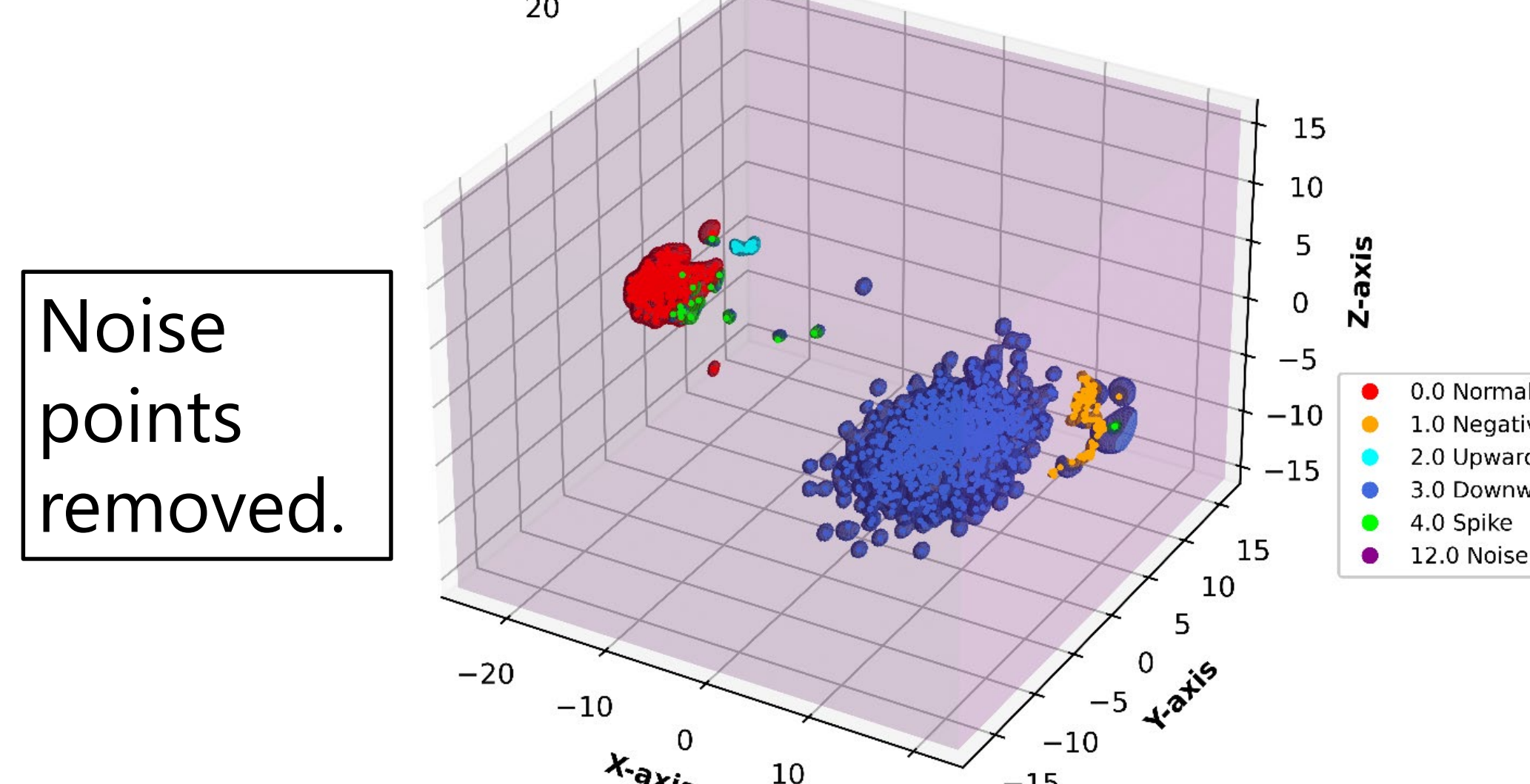
Non-noise points
perplexity = 270,
learning rate = 156,
with the *Chebyshev* metric.

- All points are used to train a 3-D SVM

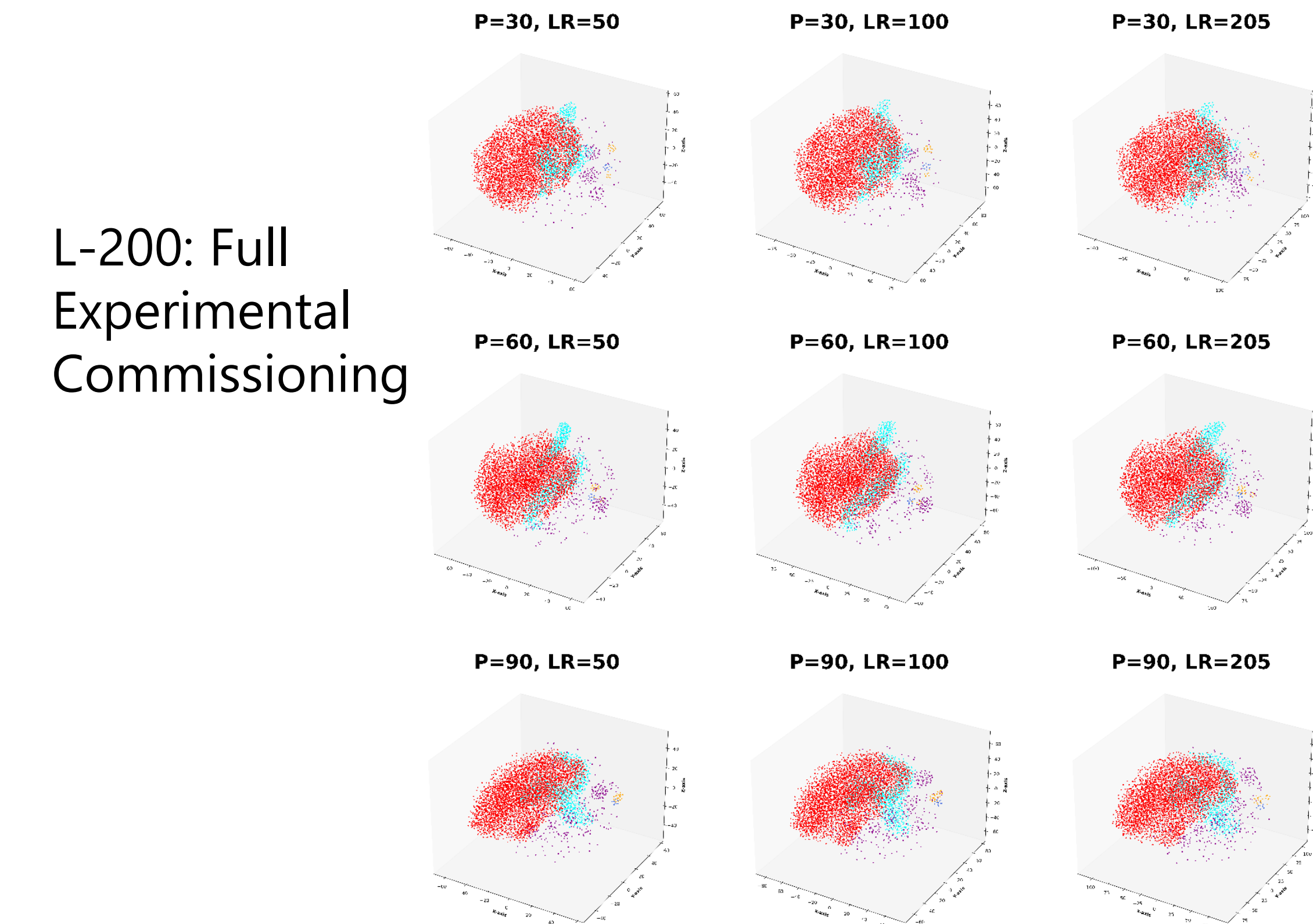


All points are shown. The noise points fall in the outer purple region.

Noise points removed.

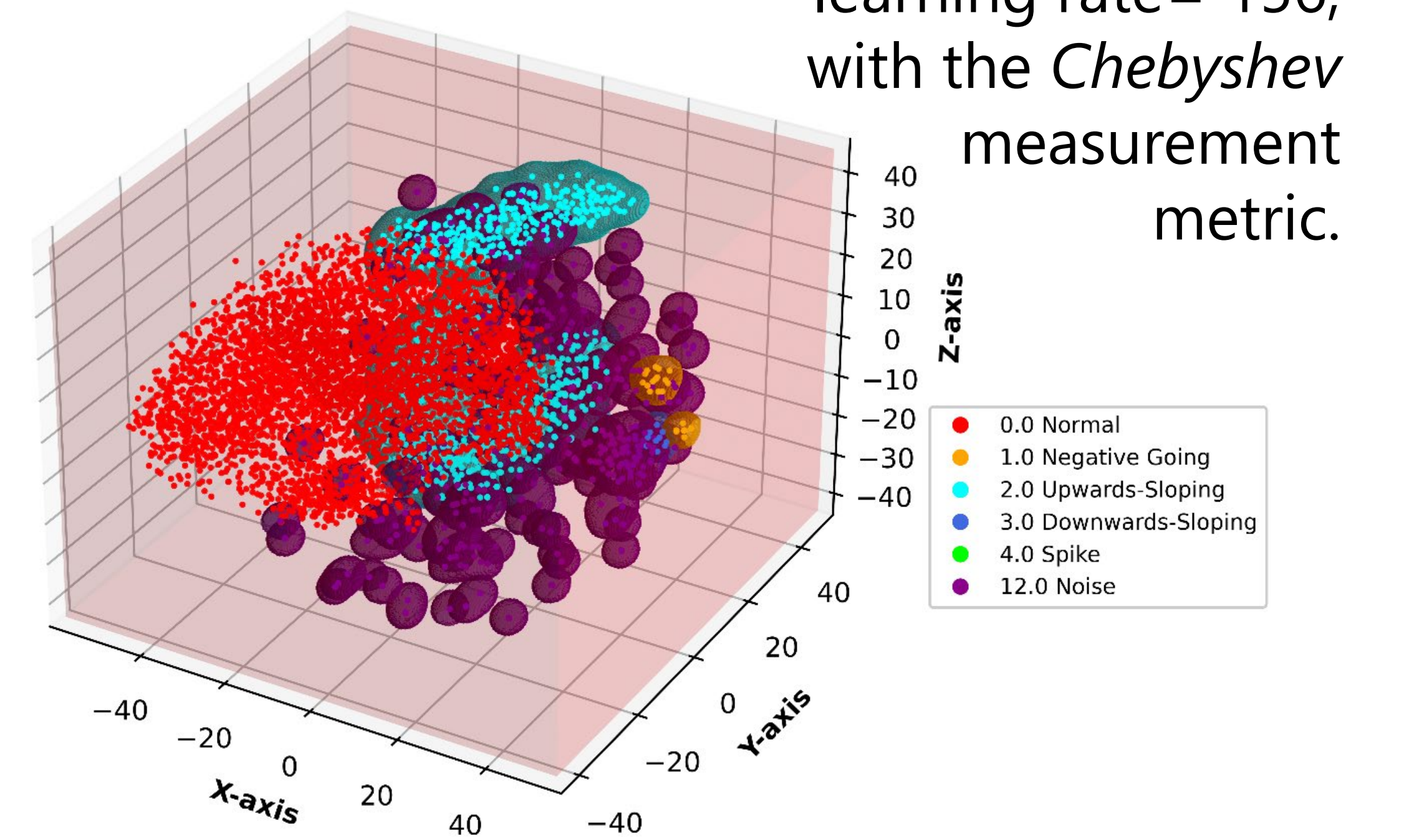


LEGEND - 200 Run 10 Data

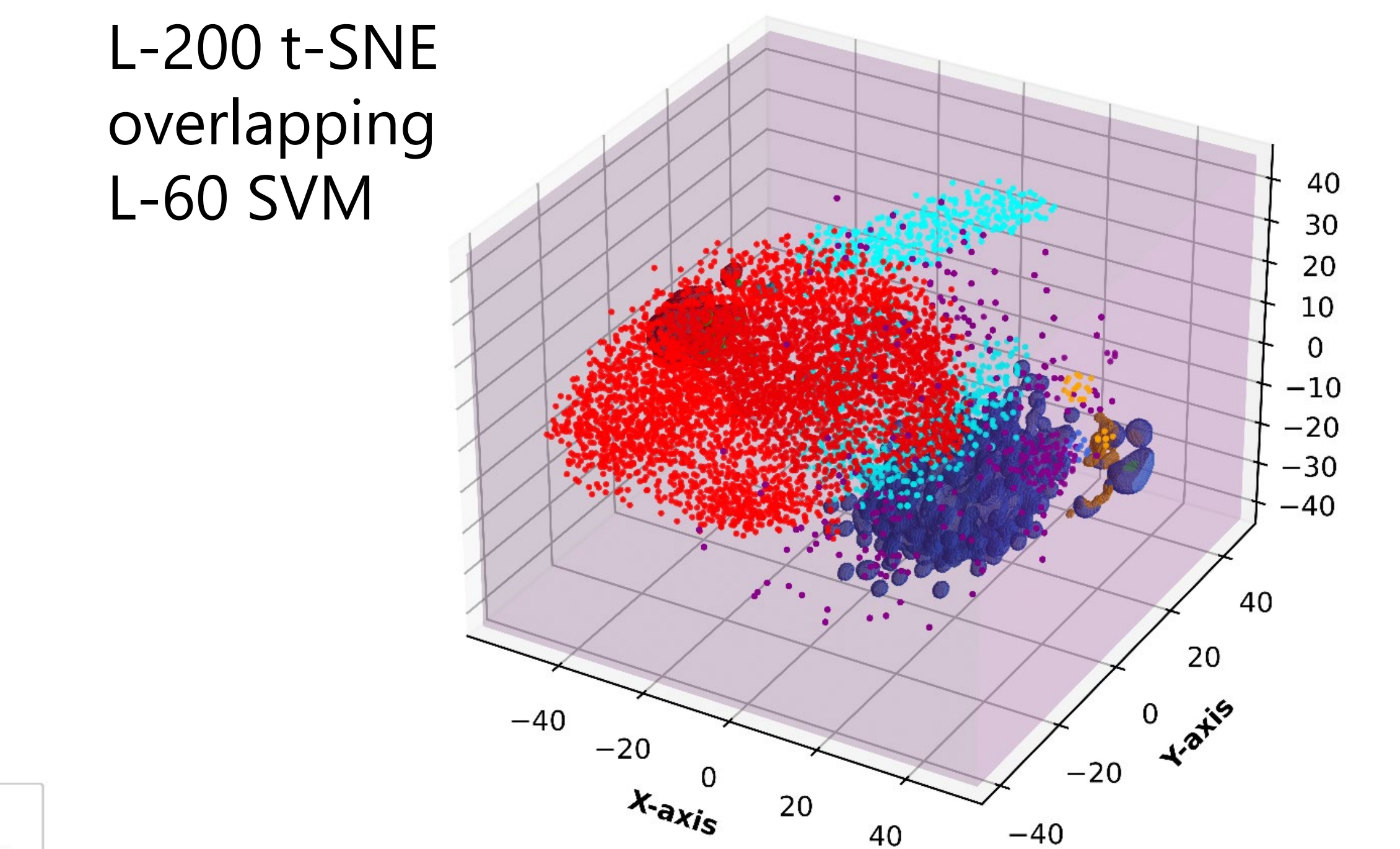


L-200: Full Experimental Commissioning

perplexity = 270,
learning rate= 156,
with the *Chebyshev* measurement metric.



L-200 t-SNE overlapping L-60 SVM



References

[1] LEGEND Collaboration, LEGEND-200 First Stage, Online. <https://legend-exp.org/science/legend-pathway/legend-200>
 [2] Frey BJ, Dueck D. Clustering by passing messages between data points. Science. 2007 Feb 16;315(5814):972-6. doi: 10.1126/science.1136800. Epub 2007 Jan 11. PMID: 17218491.
 [3] van der Maaten, L.J.P.; Hinton, G.E. Visualizing High-Dimensional Data Using t-SNE. Journal of Machine Learning Research 9:2579-2605, 2008.
 [4] "A Tutorial on Support Vector Regression" Alex J. Smola, Bernhard Schölkopf - Statistics and Computing archive Volume 14 Issue 3, August 2004, p. 199-222.