



Award #: IIS-1909096

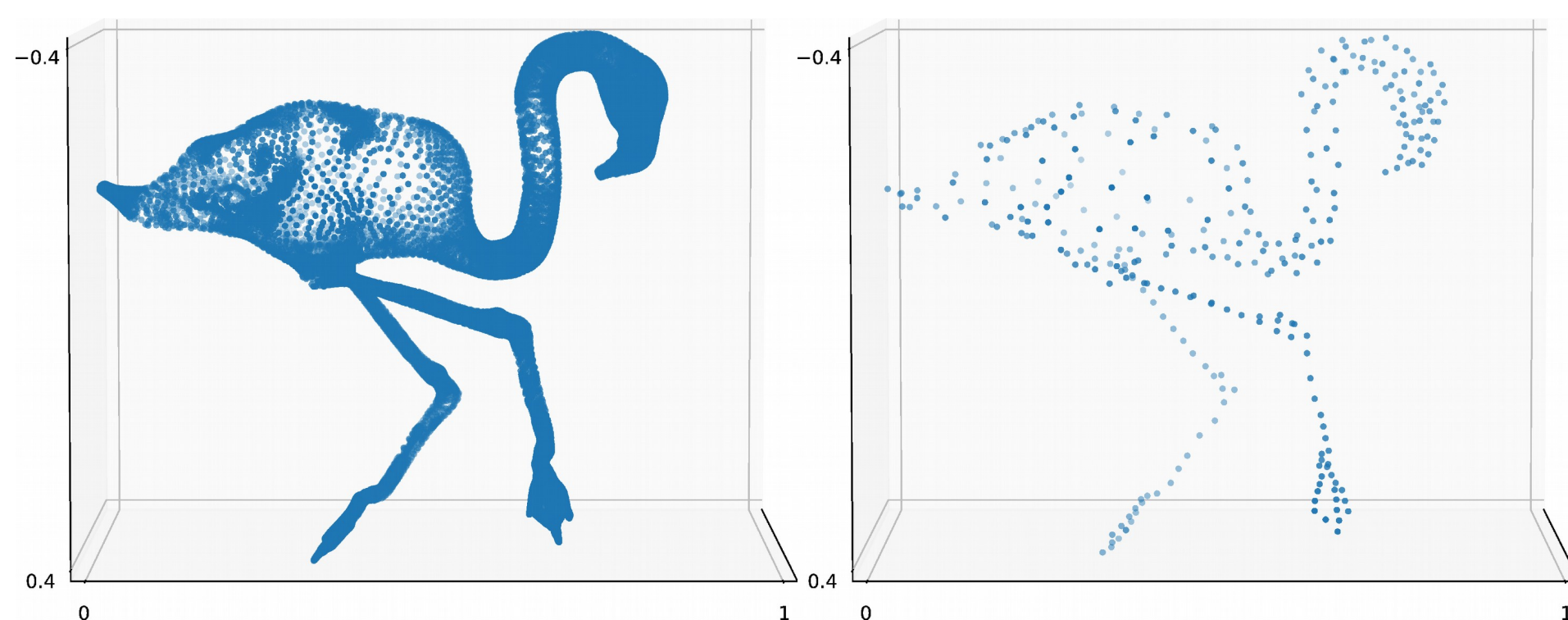
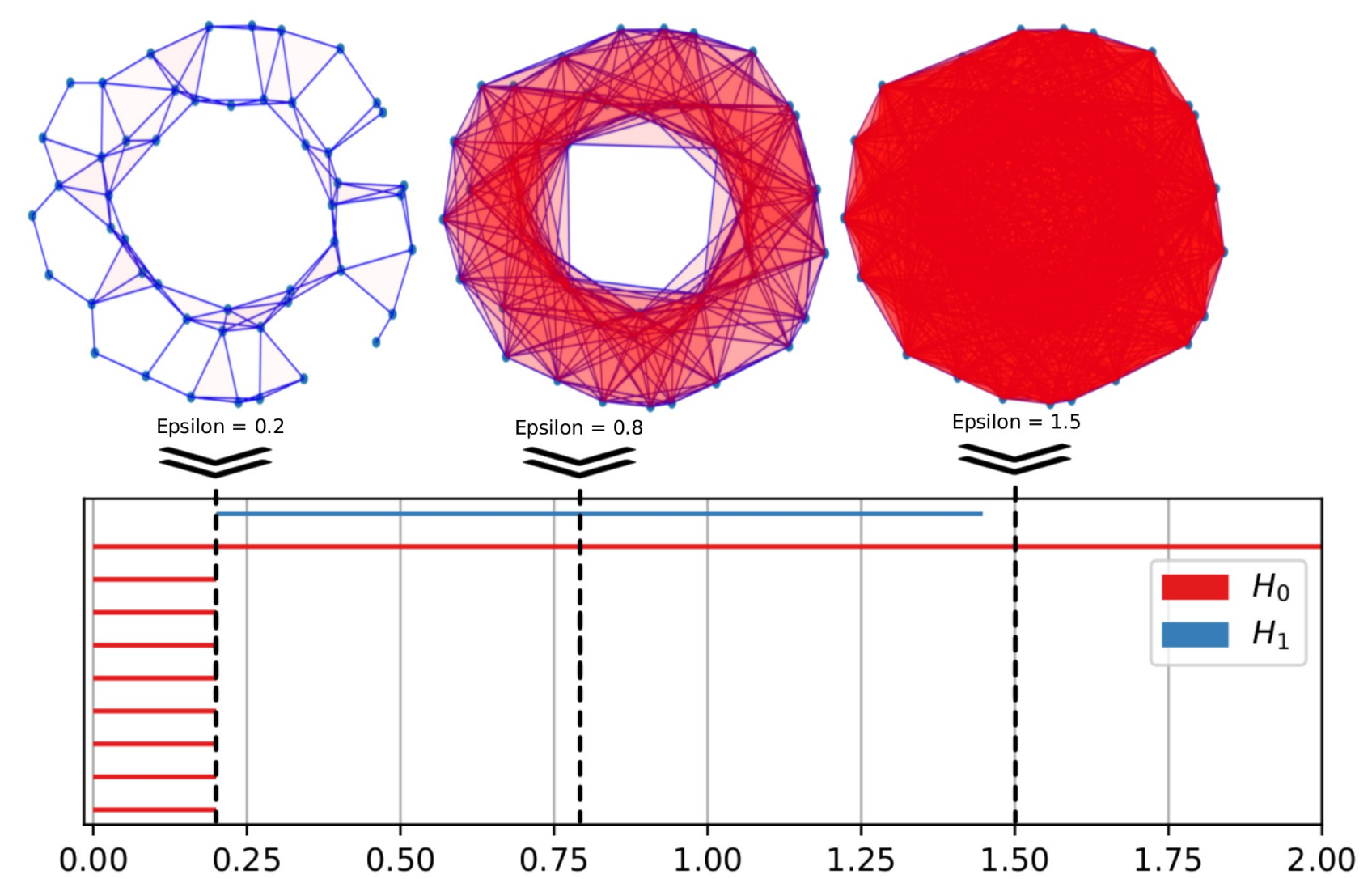
# III:Small: Partitioning Big Data for High Performance Computation of Persistent Homology

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## Topological Data Analysis/ Persistent Homology

- **Exponential Complexity** (time & space)
  - Limited to ~10K points in  $R^3$  (64GB RAM)
- **Data reduction & Partitioning**
  - Use cluster centroids (samples)
  - Use clusters  $+\delta$  as partitions
- **Parallelism and concurrency**



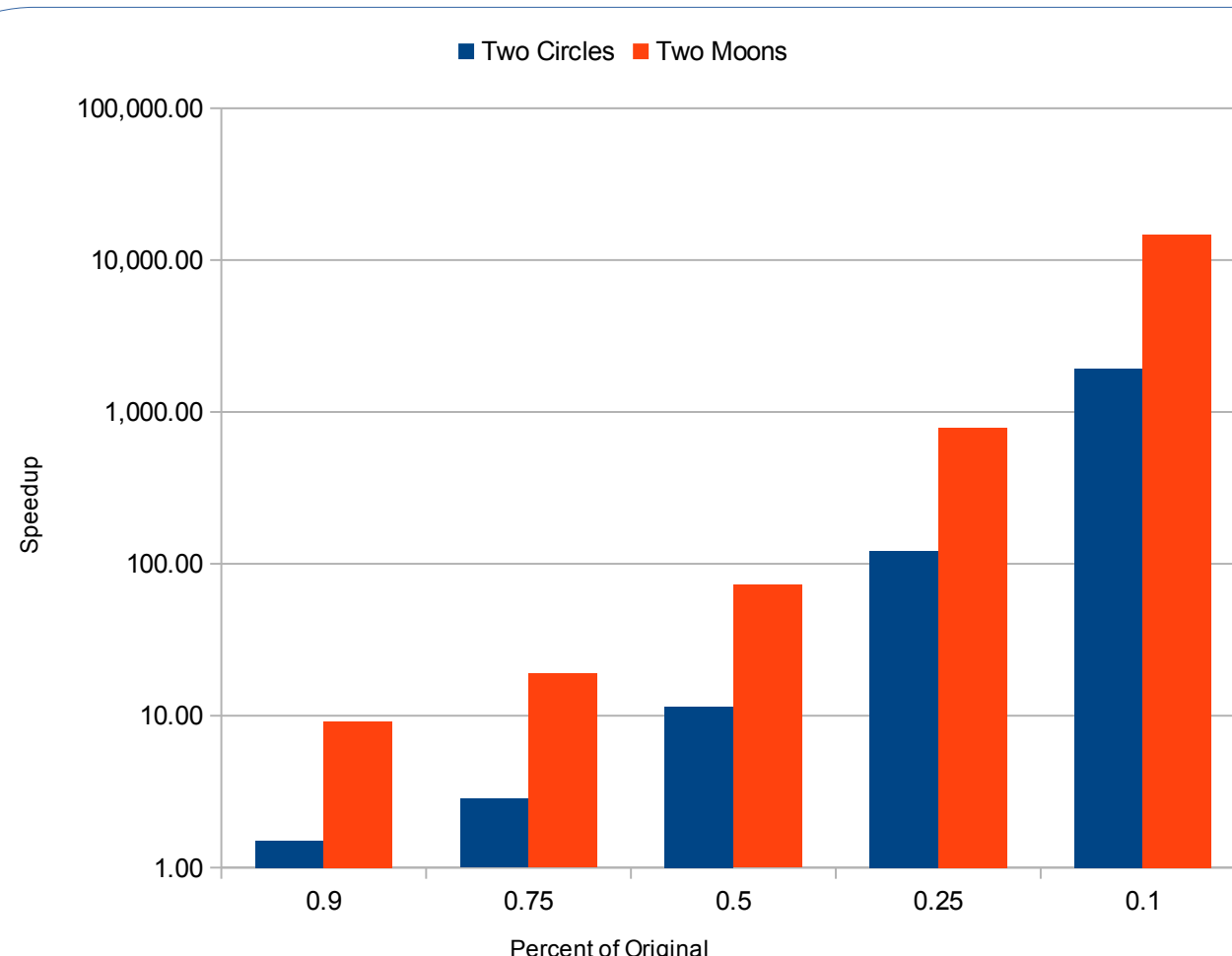
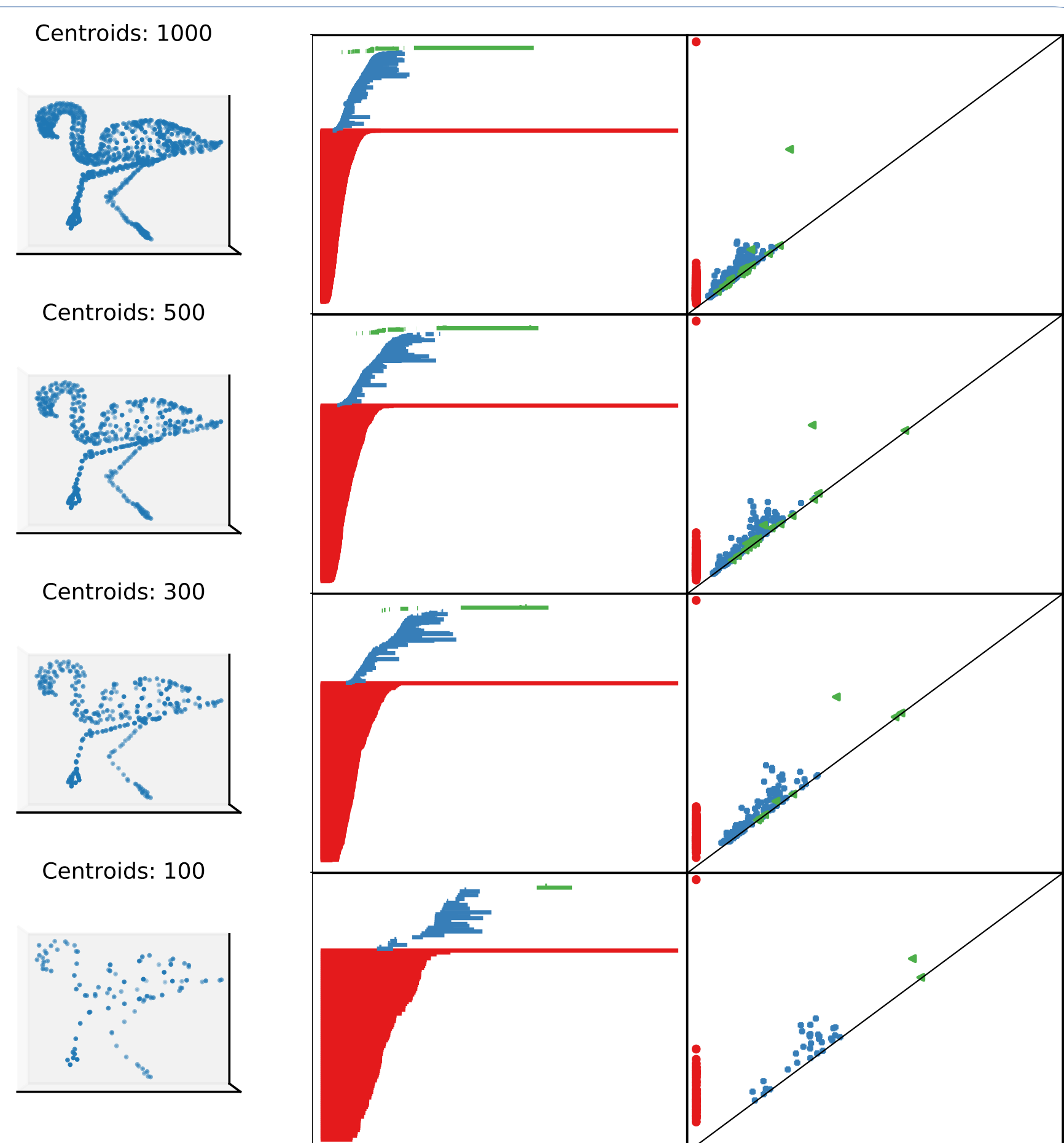
Data reduction: 27K  $\rightarrow$  300 points  
Utilizing *k-means++*

## Data Sampling and Partitioning

- **Witness complex**
- **Random Sampling**
- **Clustering (enables upscaling)**
  - Density-based: DBScan
  - Grid-based (distance-independent)
  - Partition-based: k-Means++
  - Hierarchical-based: Agglomerative
- **3-4 orders of magnitude perf gains**

## Output Analysis

- **Quantitative Analysis**
  - **Persistence Interval Comparison:**
    - Bottleneck, Wasserstein, Heat Kernel Distances
  - **Performance:**
    - Runtime, Memory use, Scalability
  - **Data Sampling & Partitioning:**
    - Persistent Homology preserving sampling
    - Persistence interval preservation
    - Topological feature preservation
- **Qualitative Analysis**
  - Barcodes
  - Persistence Diagrams
  - Landscape Diagrams
  - Persistence Images
  - Feature Boundary Extraction



## Data Reduction performance Improvements

- **Reduced input points results in:**
  - Reduced Memory Footprint
  - Faster CPU and Wall Time
  - Approximations of large features with bounded error
  - Upscaling to improve boundary identification of topological feature

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