

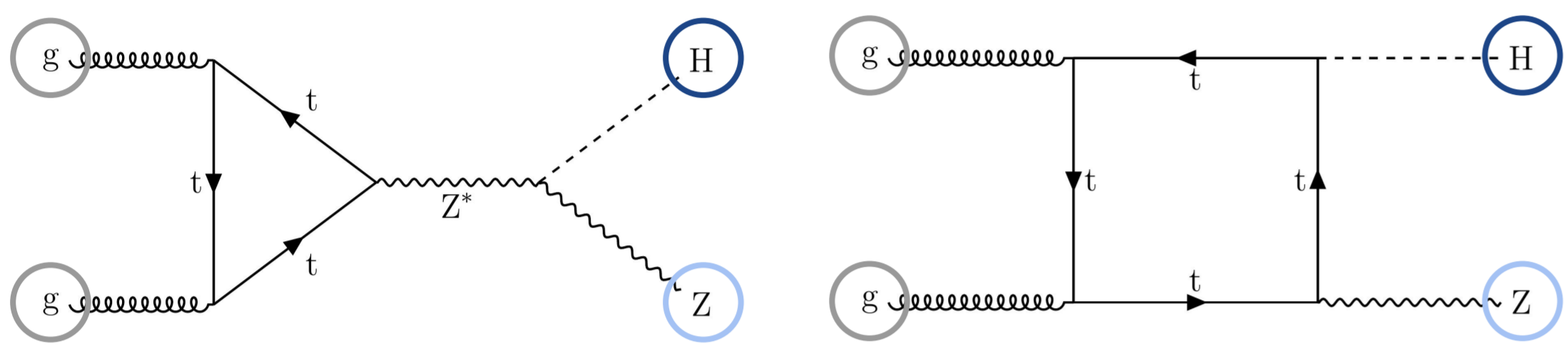
# Binning High-Dimensional Classifier Output for HEP Analyses through a Clustering Algorithm

Svenja Diekmann, Niclas Eich & Martin Erdmann - RWTH Aachen University



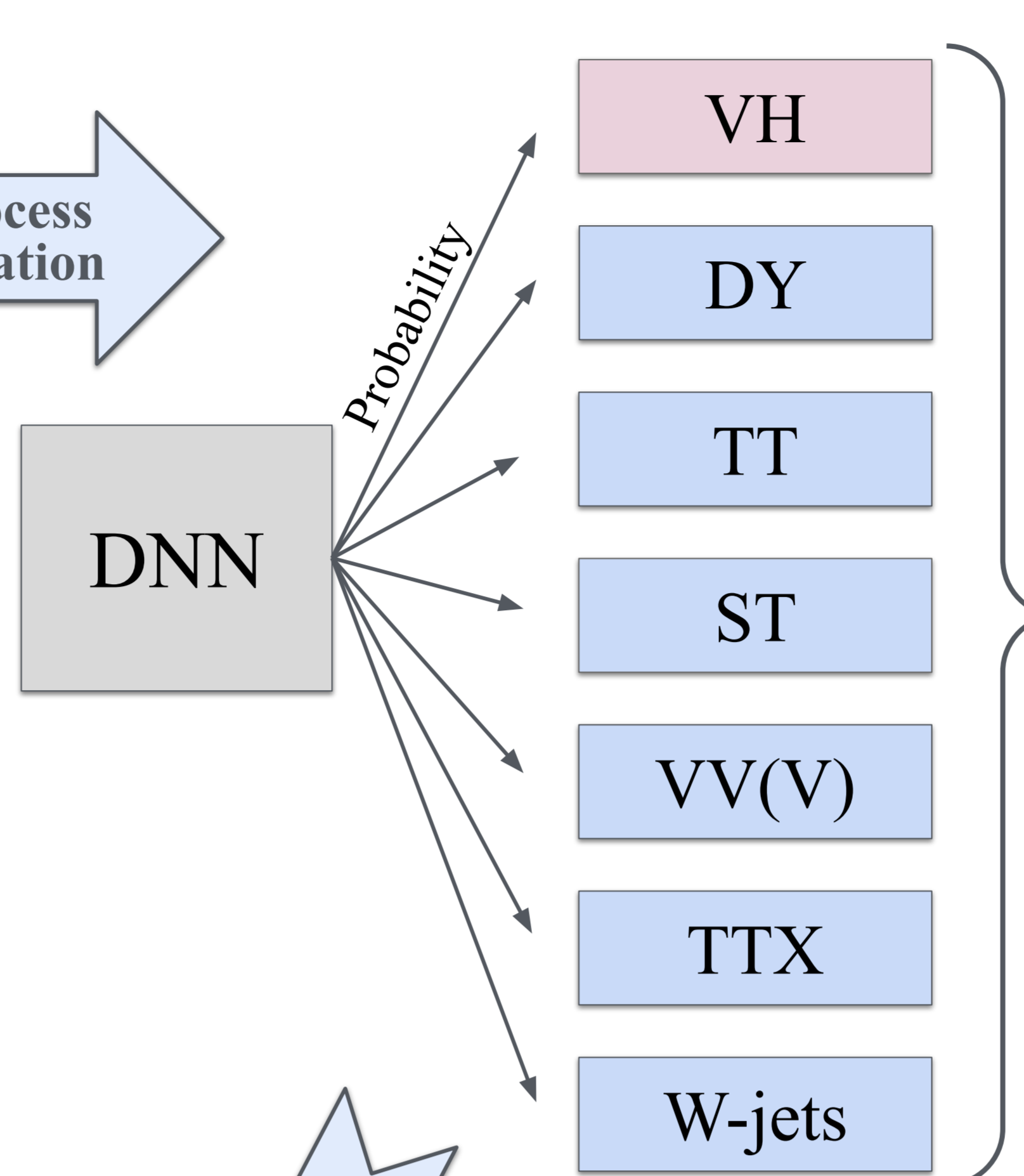
## Motivation & Context

- ♦ search for **VH**
- ♦ specialised analysis for **gg → ZH** production
- ♦ yet unmeasured
- ♦ small cross section:
  - ◊ need for sensitivity enhancement



## Process Assignment by DNN

Multiprocess Classification



Standard Approach:

- ♦ projection to 1 dimension
- ♦ loss of information

High dimensional output:

$$E_1 = (p_{VH}, p_{DY}, p_{TT}, \dots)$$

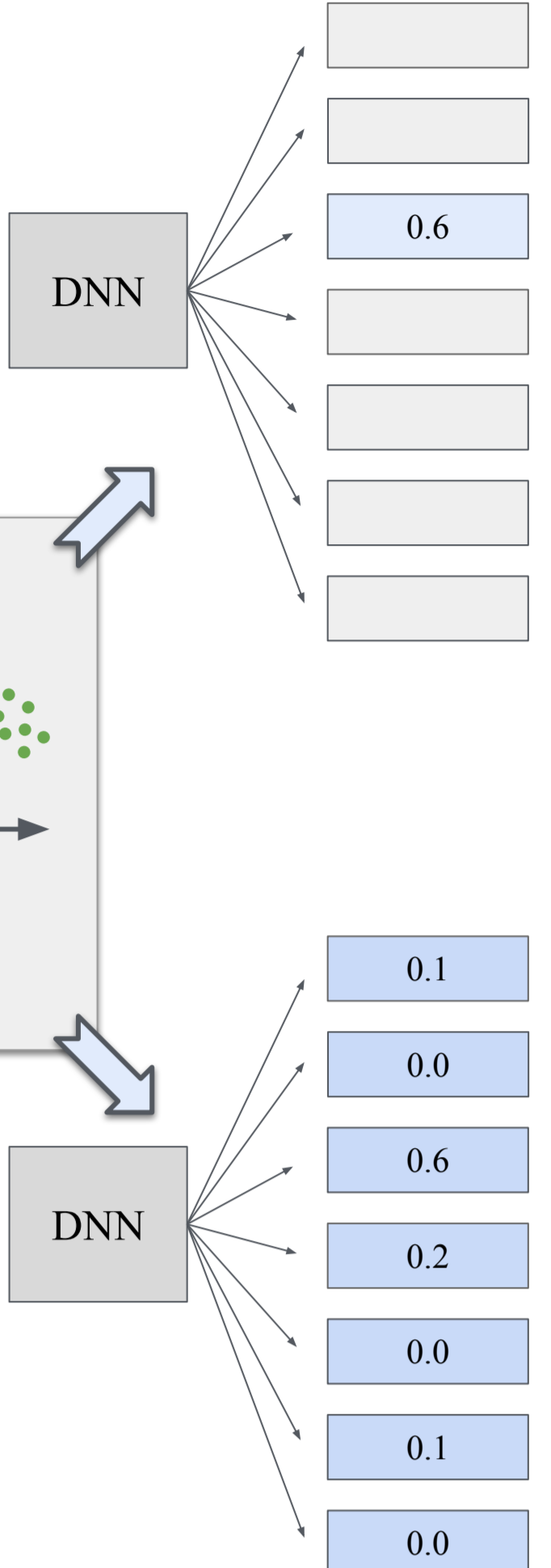
$$E_2 = (p_{VH}, p_{DY}, p_{TT}, \dots)$$

$$E_3 = (p_{VH}, p_{DY}, p_{TT}, \dots)$$

$$\dots$$

New Approach:

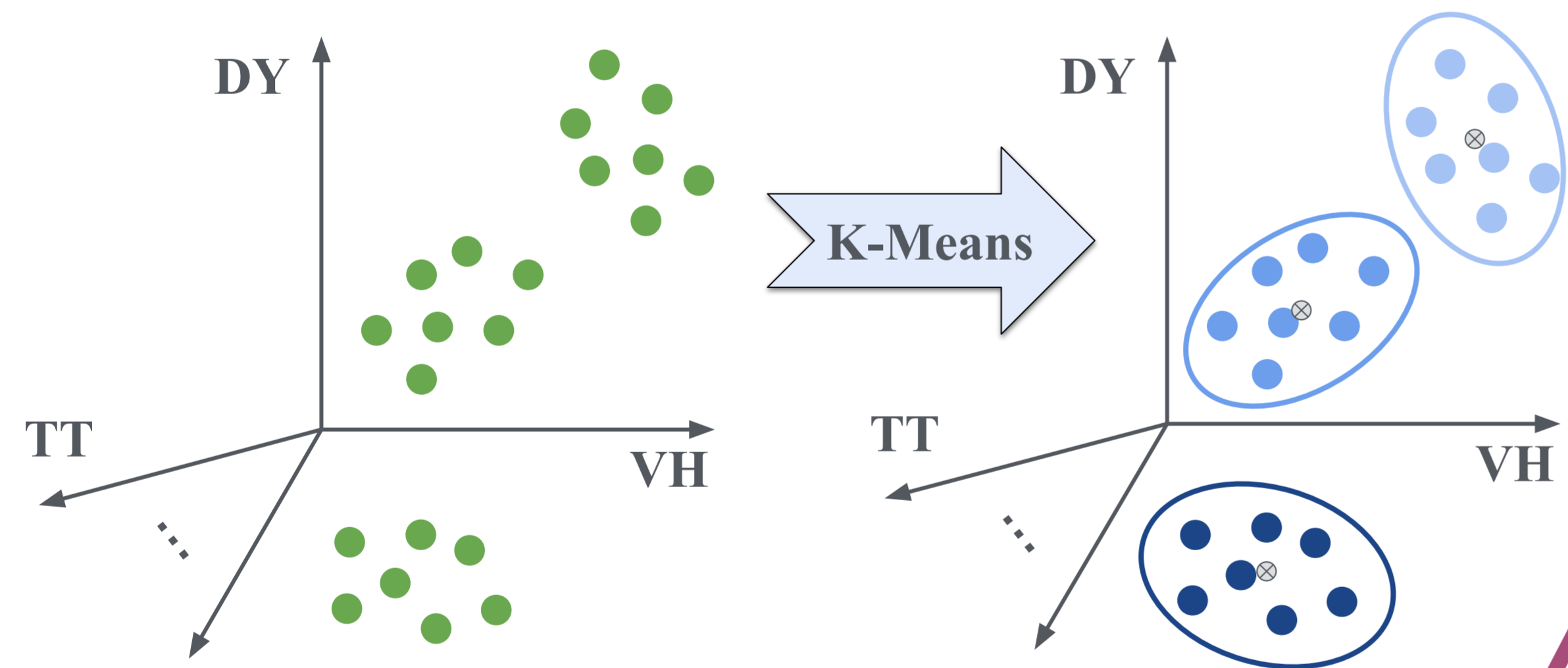
- ♦ keep **all** dimensions
- ♦ **no** loss of information



## High-Dimensional Clustering

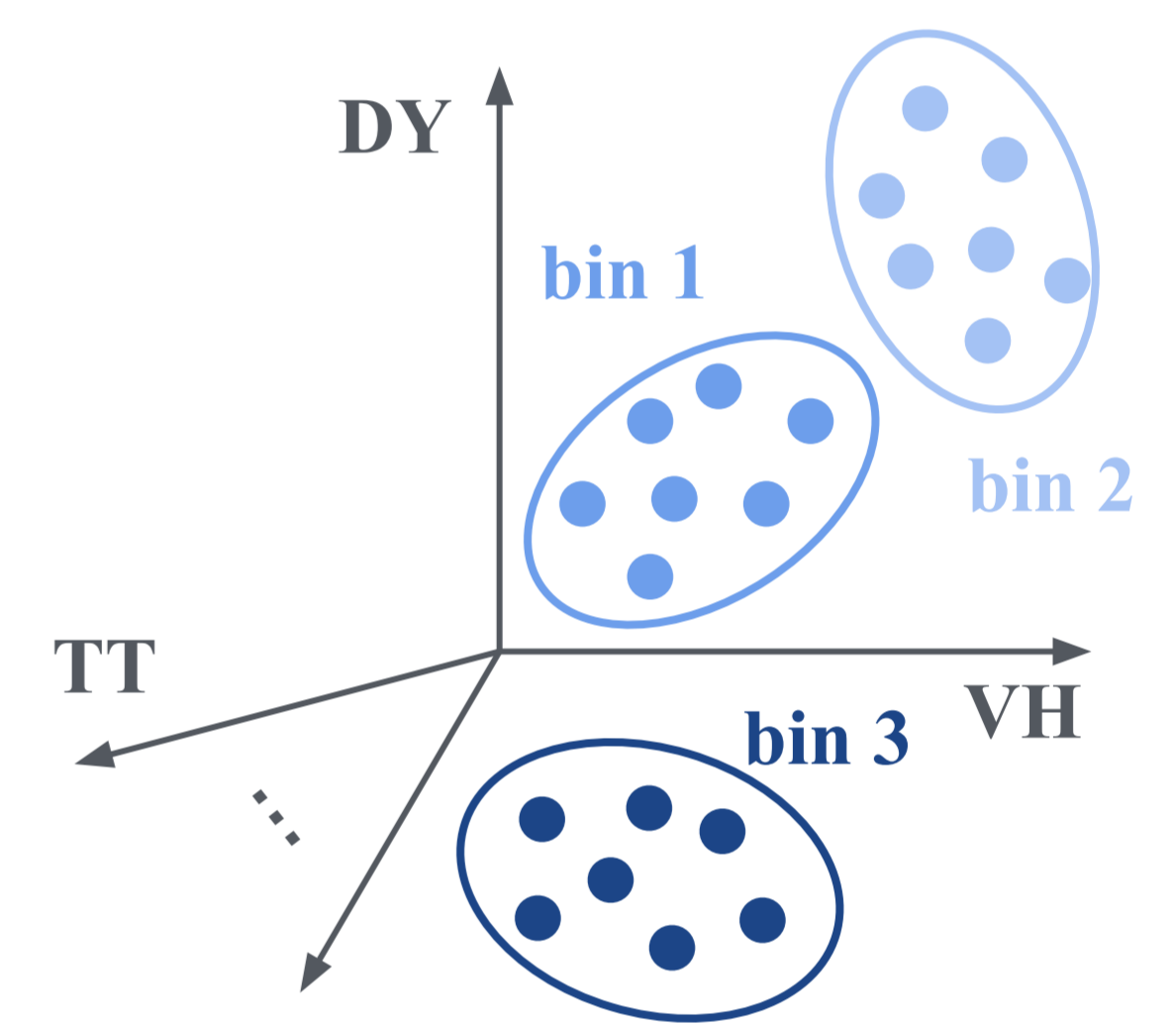
**K-Means** Clustering Algorithm:

- ♦ random initialisation of **k** cluster centers
- ♦ eventwise assignment to closest cluster
- ♦ iterative update of cluster centers by **mean** of assigned events



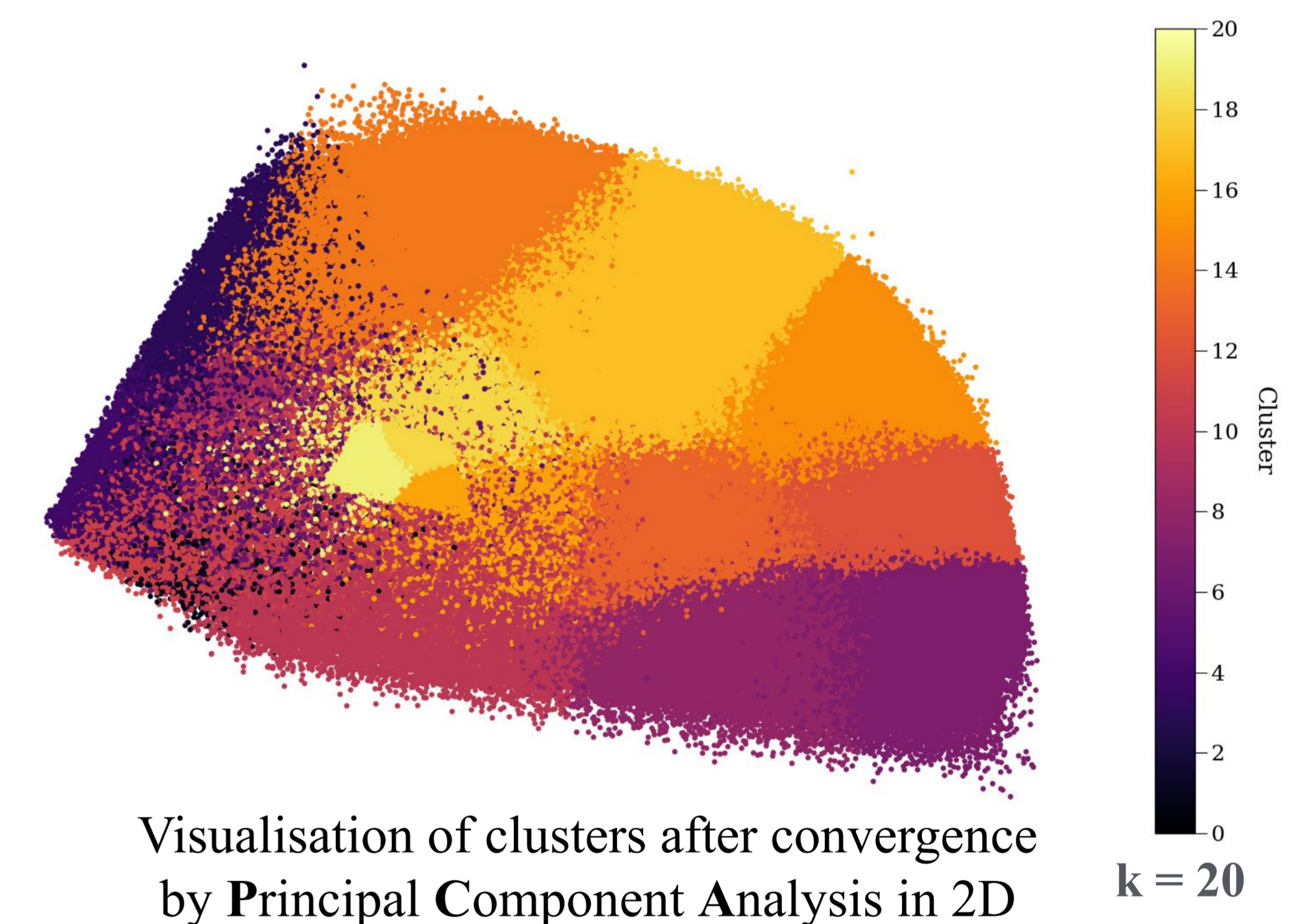
## Binning from Clusters

Identify clusters as bins for likelihood fit



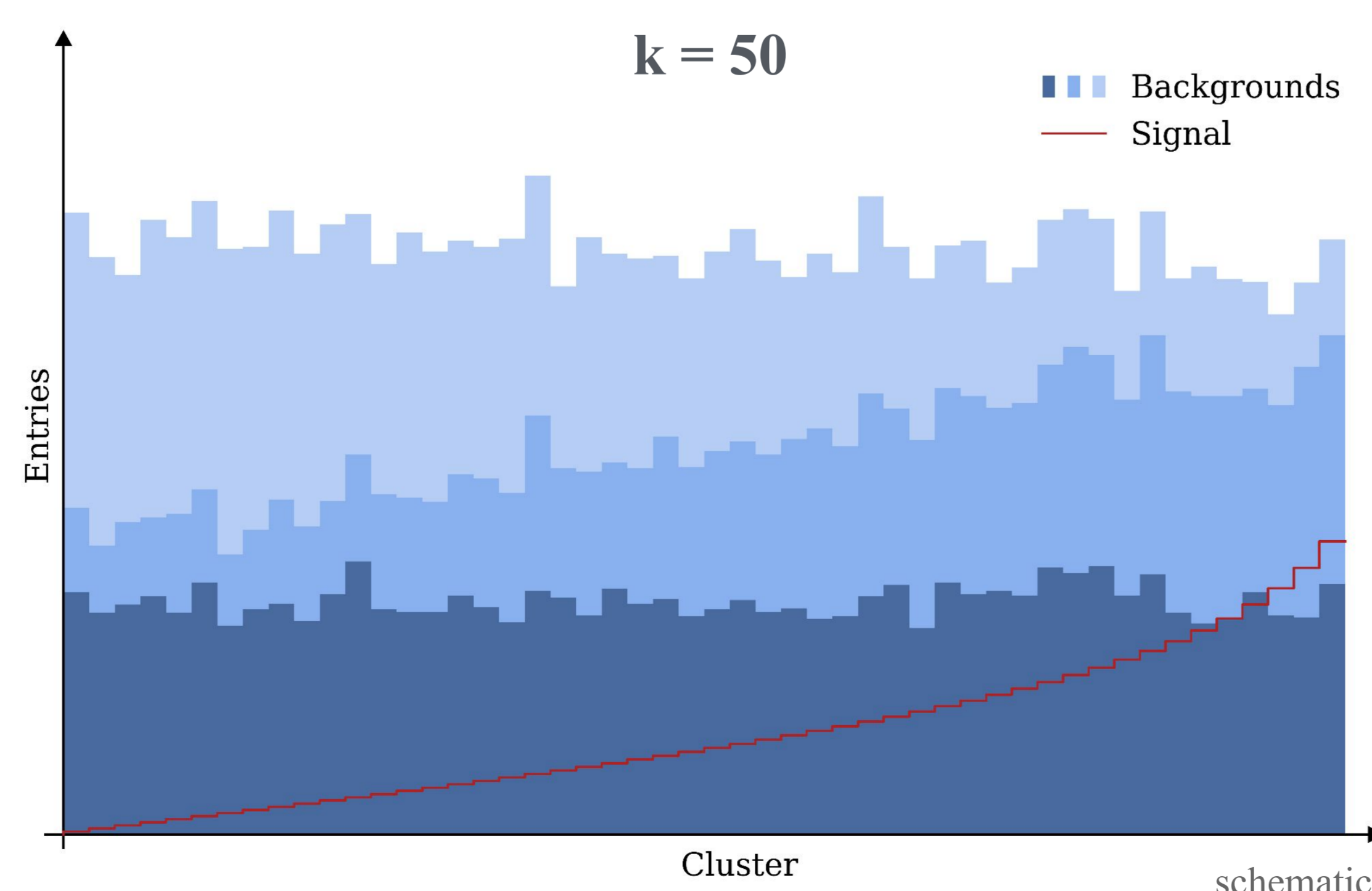
Convergence

Agnostic binning algorithm in high-dimensional space



## Resulting Histograms

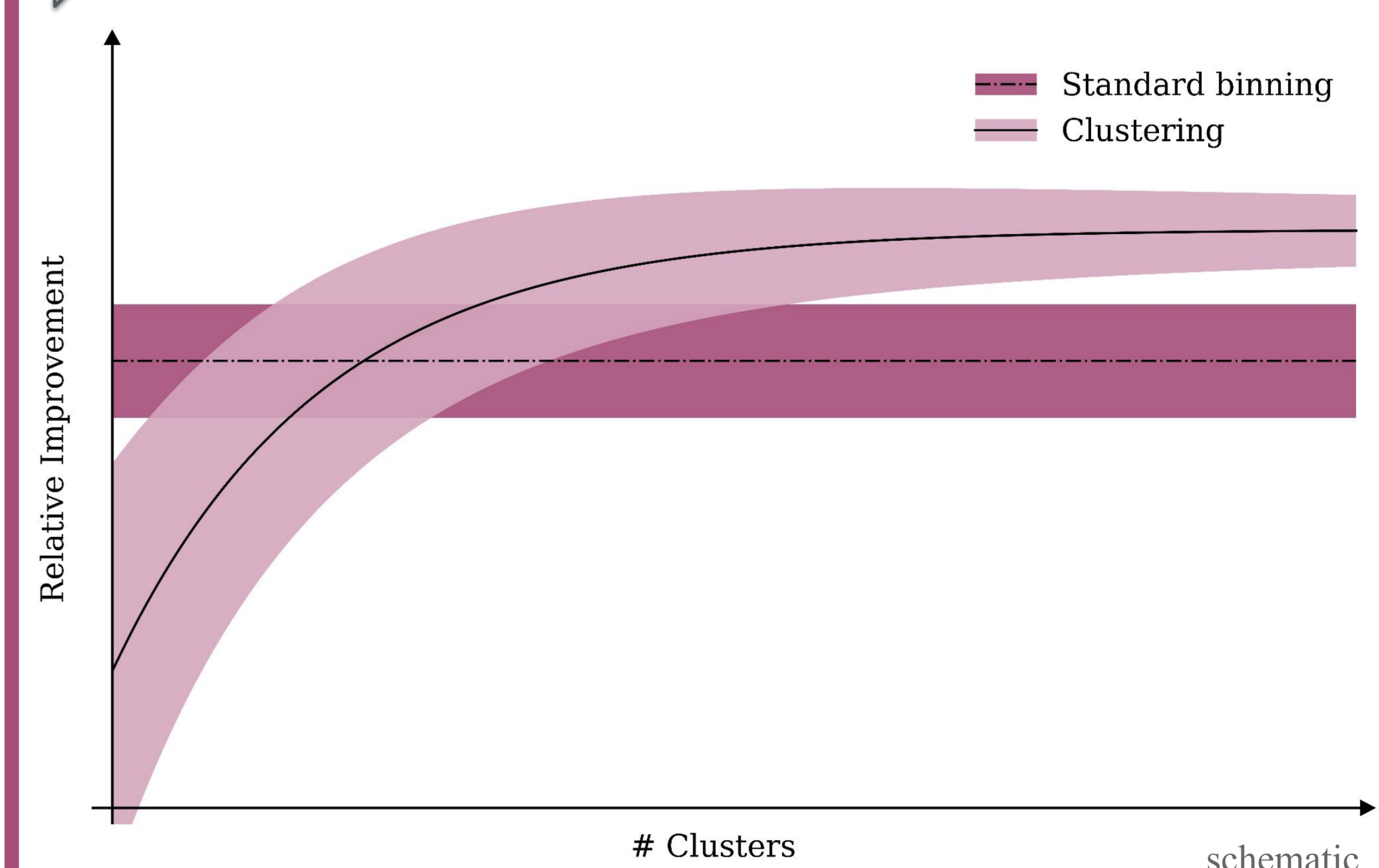
- ♦ clustering creates high signal and high background regions
- ♦ cluster index has no physical meaning
  - ◊ sorted by the total signal contribution (red line)



## Sensitivity Enhancement

Sensitivity increase for high numbers of clusters:

- ◊ improvement compared to the standard method including statistical uncertainties



Physical Interpretation

Inference