

Activity Mathematics:

Pattern Classification in Dense Sensor Fields

A brief summary of work to date

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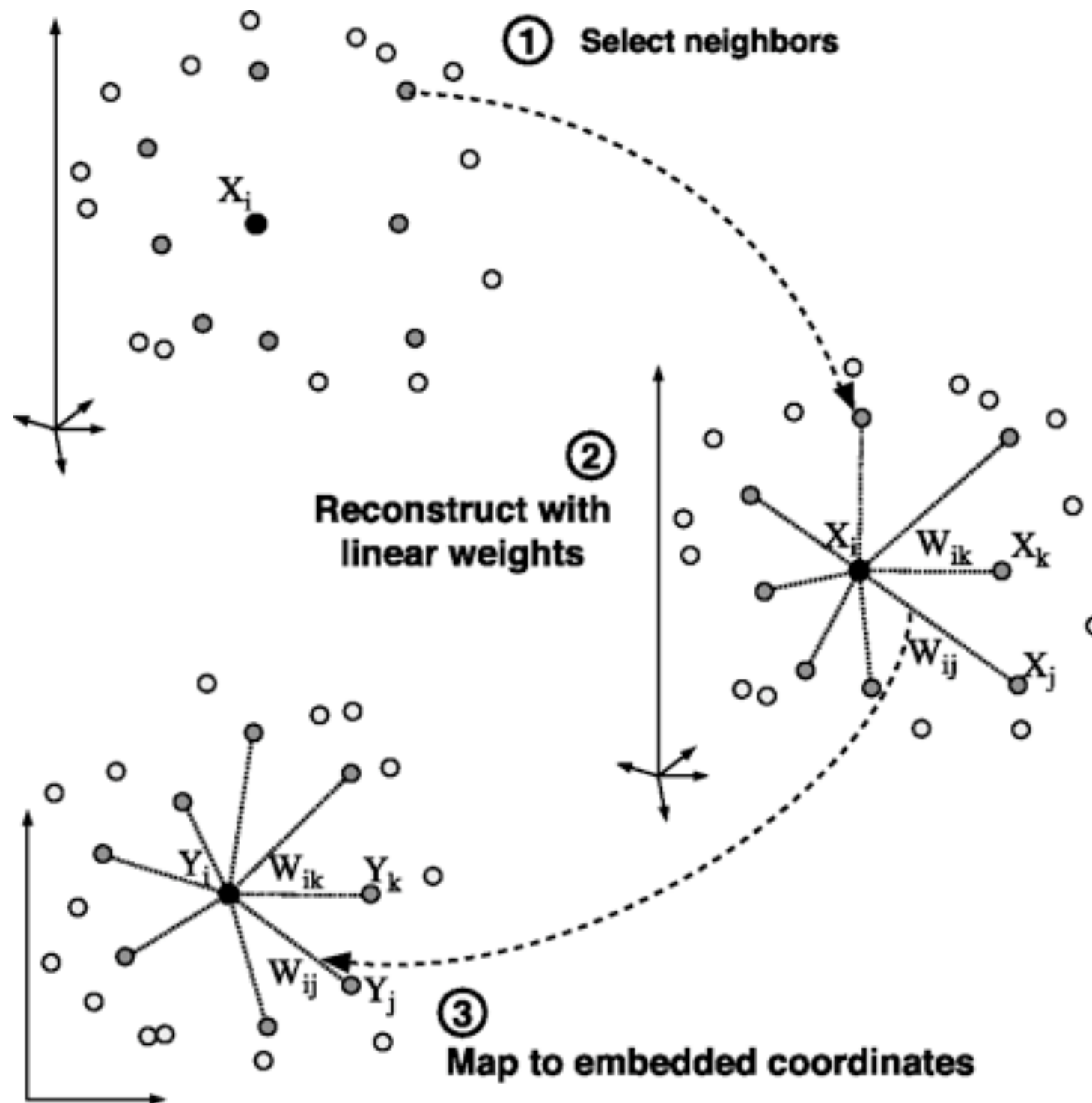
Major Ideas

- Dimensionality Reduction
- Clustering
- Pattern Matching
- Minimum Description Length
- Anomaly Detection

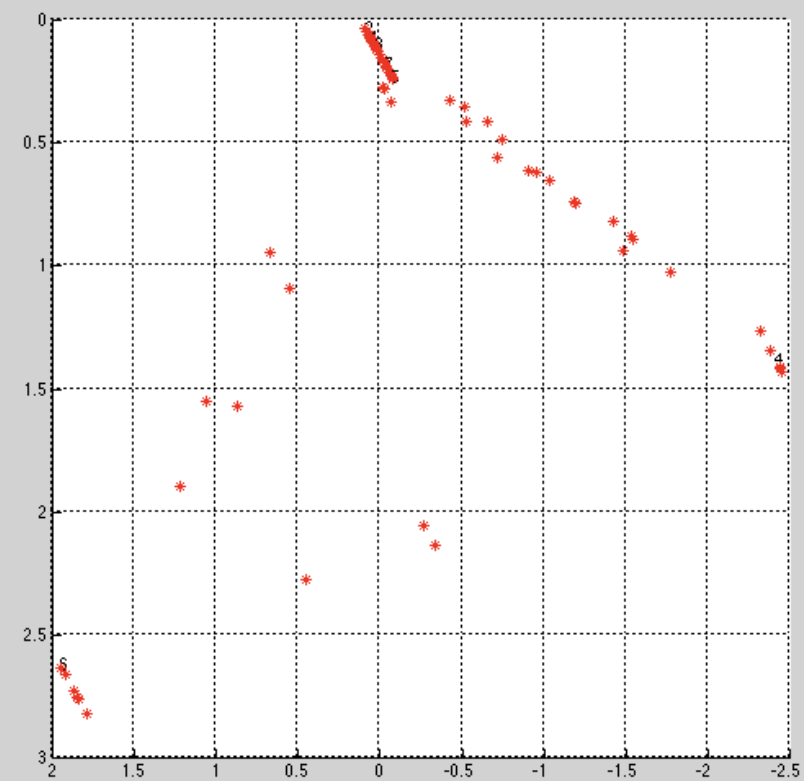
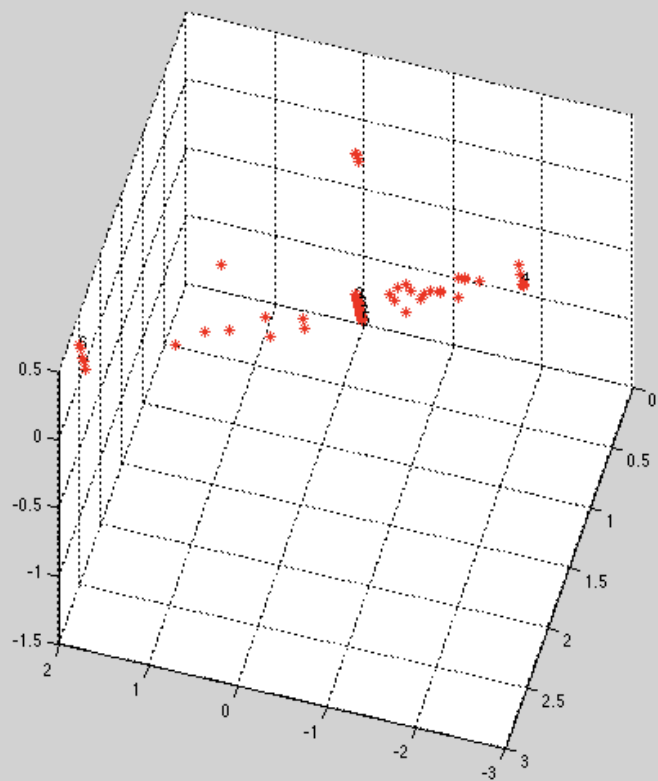
Major Ideas

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Locally Linear Embedding



LLE - Results



Warning: Matrix is close to singular or badly scaled.

Results may be inaccurate. RCOND = 3.700743e-17.

> In eigs>AminusSigmaBsolve at 1204

In eigs at 257

In lleMod at 119

In run at 69

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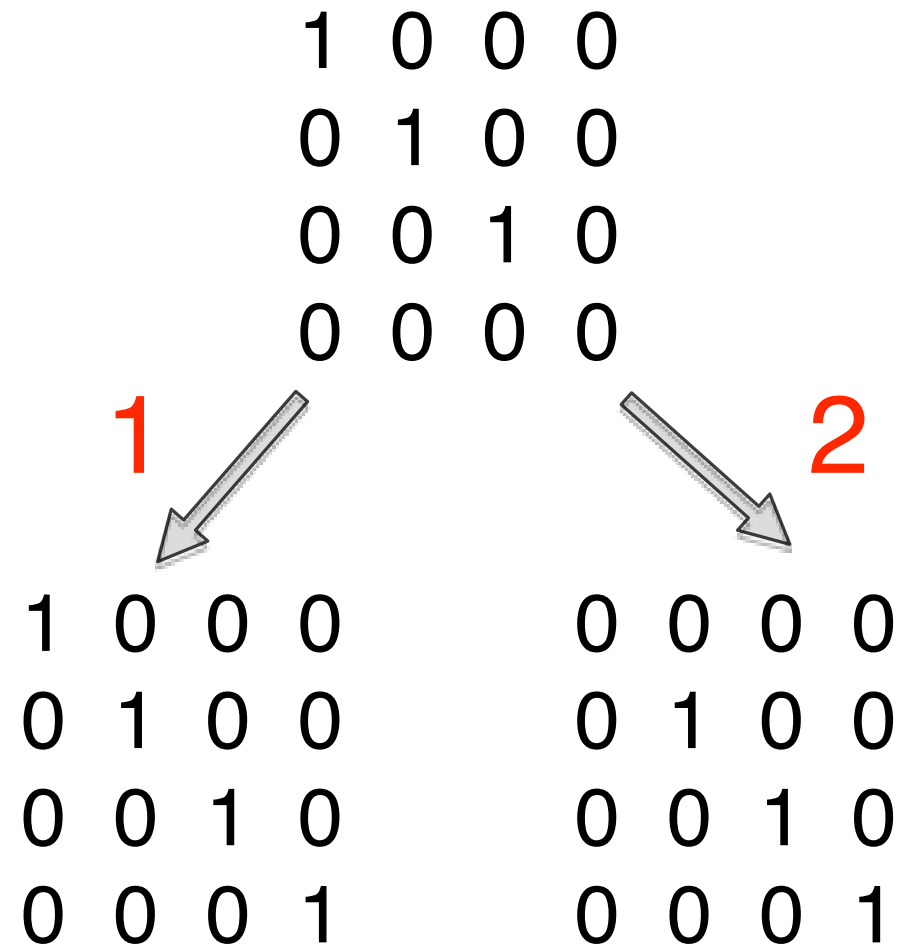
In lleMod at 119

In run at 69

Done.

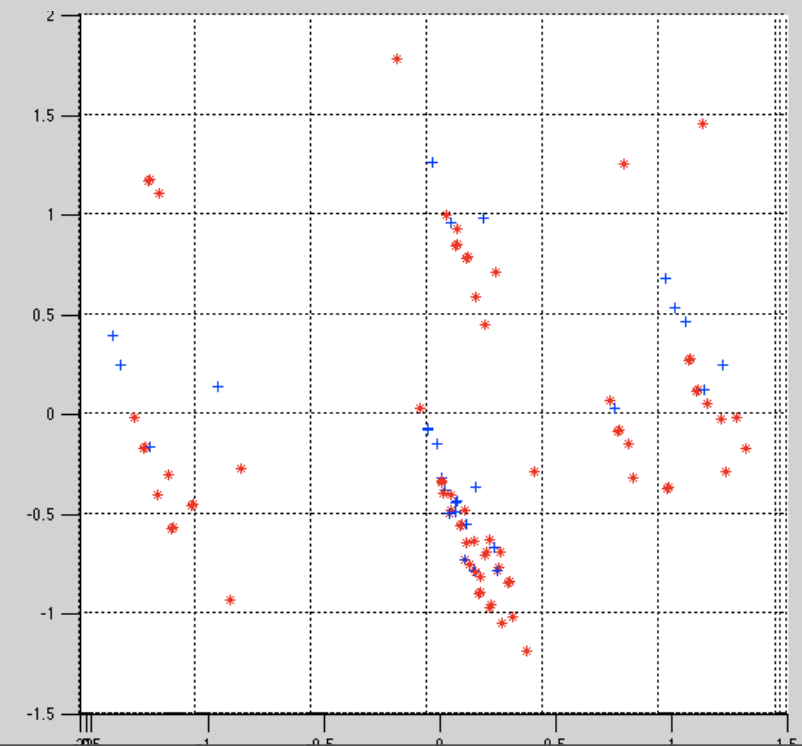
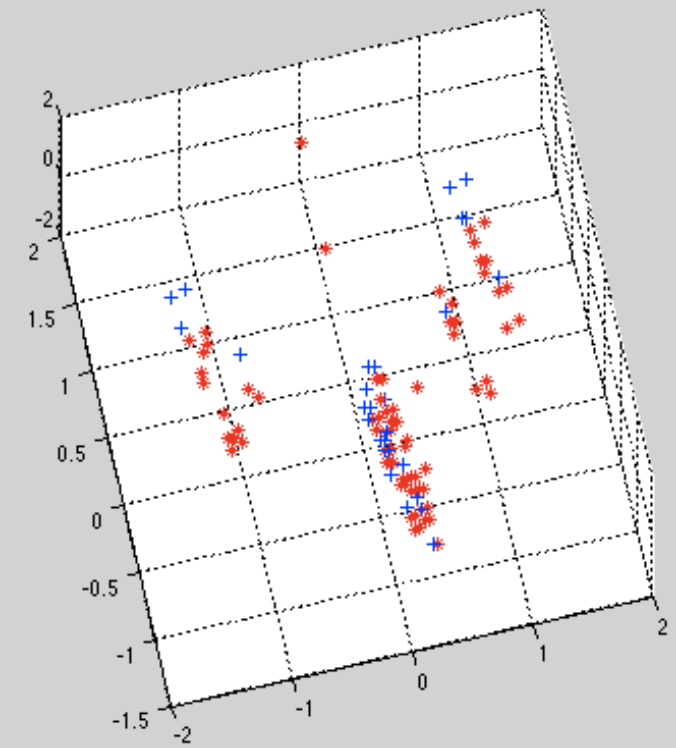
LLE - What's left?

- Assumption of locally linear data valid?
- AMPLE
- Distance metric definition
 - Walk and Loiter distance



Principal Component Analysis

$$S = \begin{bmatrix} 5.6482 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 5.4354 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 4.6176 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 4.5500 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 3.9379 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 3.5706 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 2.5820 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2.3930 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2.2864 \end{bmatrix}$$



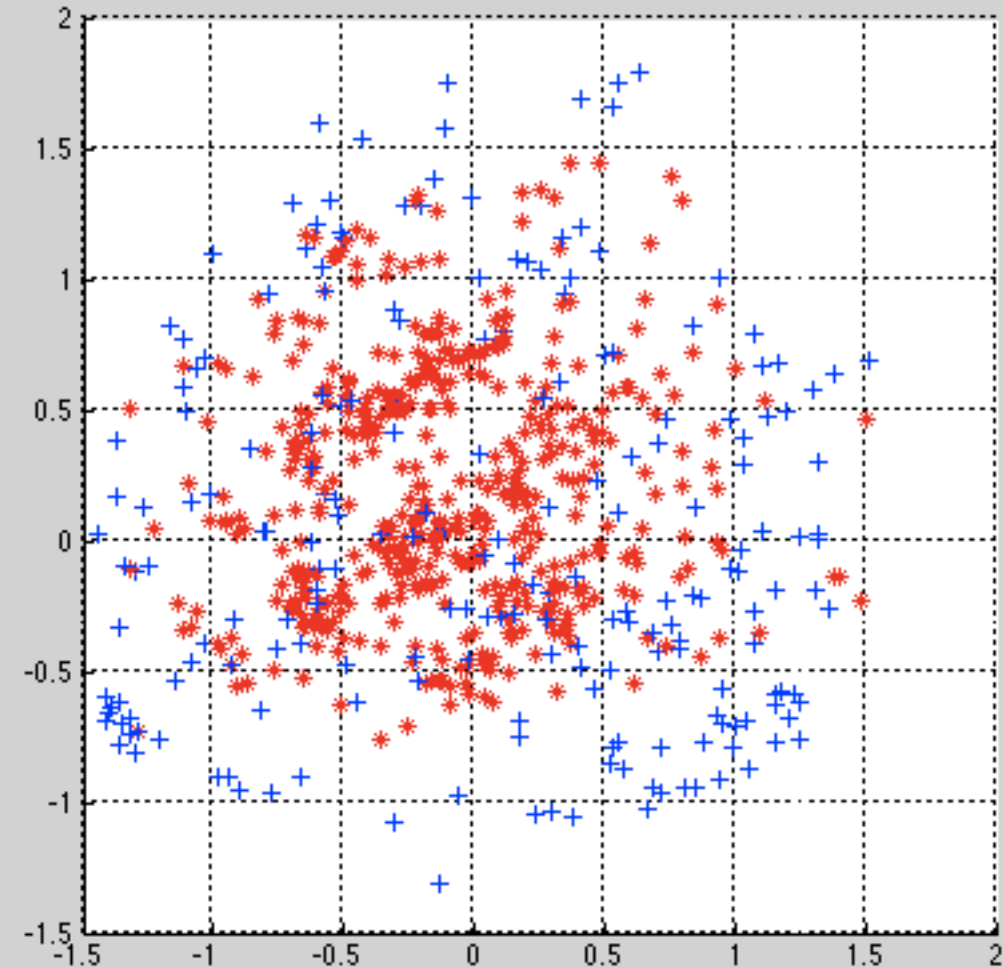
Dataset: (Dense)

1000 Points

100 Loitering Events

100 Walk Left Events

100 Walk Right Events



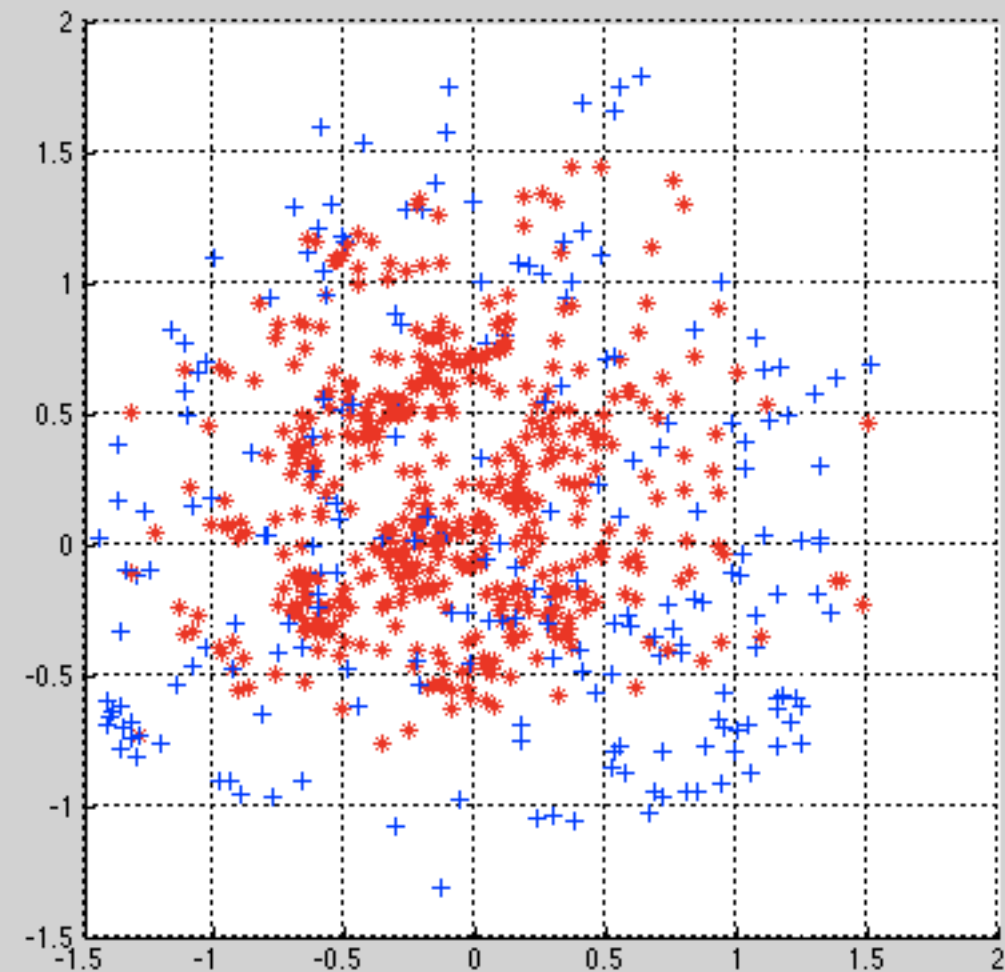
Dataset: (Sparse)

1000 Points

10 Loitering Events

10 Walk Left Events

10 Walk Right Events



Major Ideas

- Dimensionality Reduction
- Clustering
- **Pattern Matching**
- Minimum Description Length
- Anomaly Detection

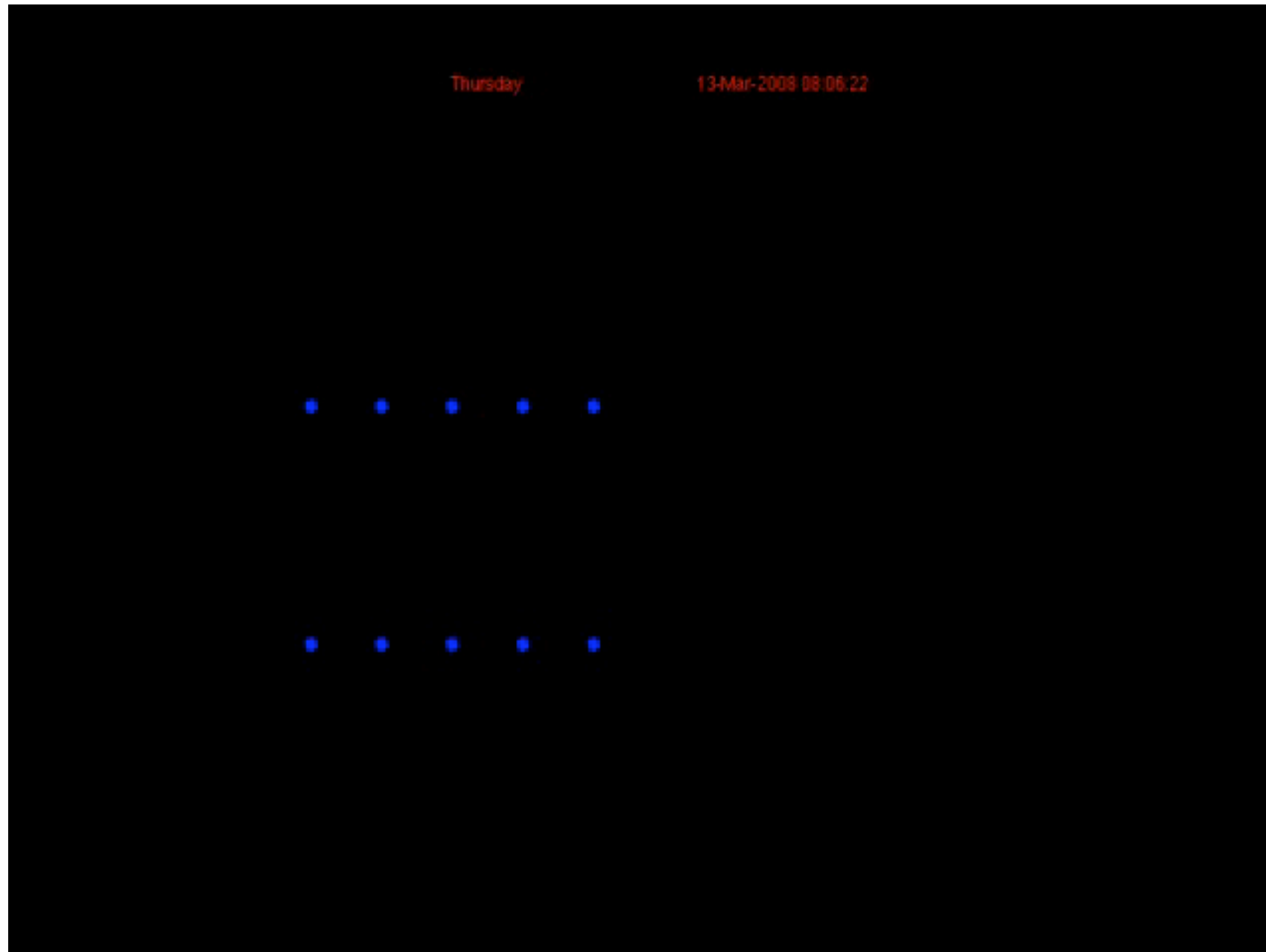
Closest Pattern Matching (CPM)

	0	0	0	0	0	0
	1	0	0	0	0	0
	0	1	1	0	1	1
	1	0	0	1	0	0
	1	0	0	1	1	1
	0	1	1	0	0	1
Predicted Row	0	1	1	0	0	0

CPM - Continued

	0	0	0	0	0	0
	1	0	0	0	0	0
	0	1	1	0	1	1
	1	0	0	1	0	0
	1	0	0	1	1	1
	0	1	1	0	0	1
Predicted Row	0	1	1	0	0	0

CPM - Continued



Growing Pattern CPM

0	0	0	0	0	0
1	0	0	0	0	0
0	1	1	0	1	1
1	0	0	1	0	0
1	0	0	1	1	1
0	1	1	0	0	1
0	1	1	0	0	0

0	0	0	0	0	0
1	0	0	0	0	0
0	1	1	0	1	1
1	0	0	1	0	0
1	0	0	1	1	1
0	1	1	0	0	1
0	1	1	0	0	0

0	0	0	0	0	0
1	0	0	0	0	0
0	1	1	0	1	1
1	0	0	1	0	0
1	0	0	1	1	1
0	1	1	0	0	1
0	1	1	0	0	0

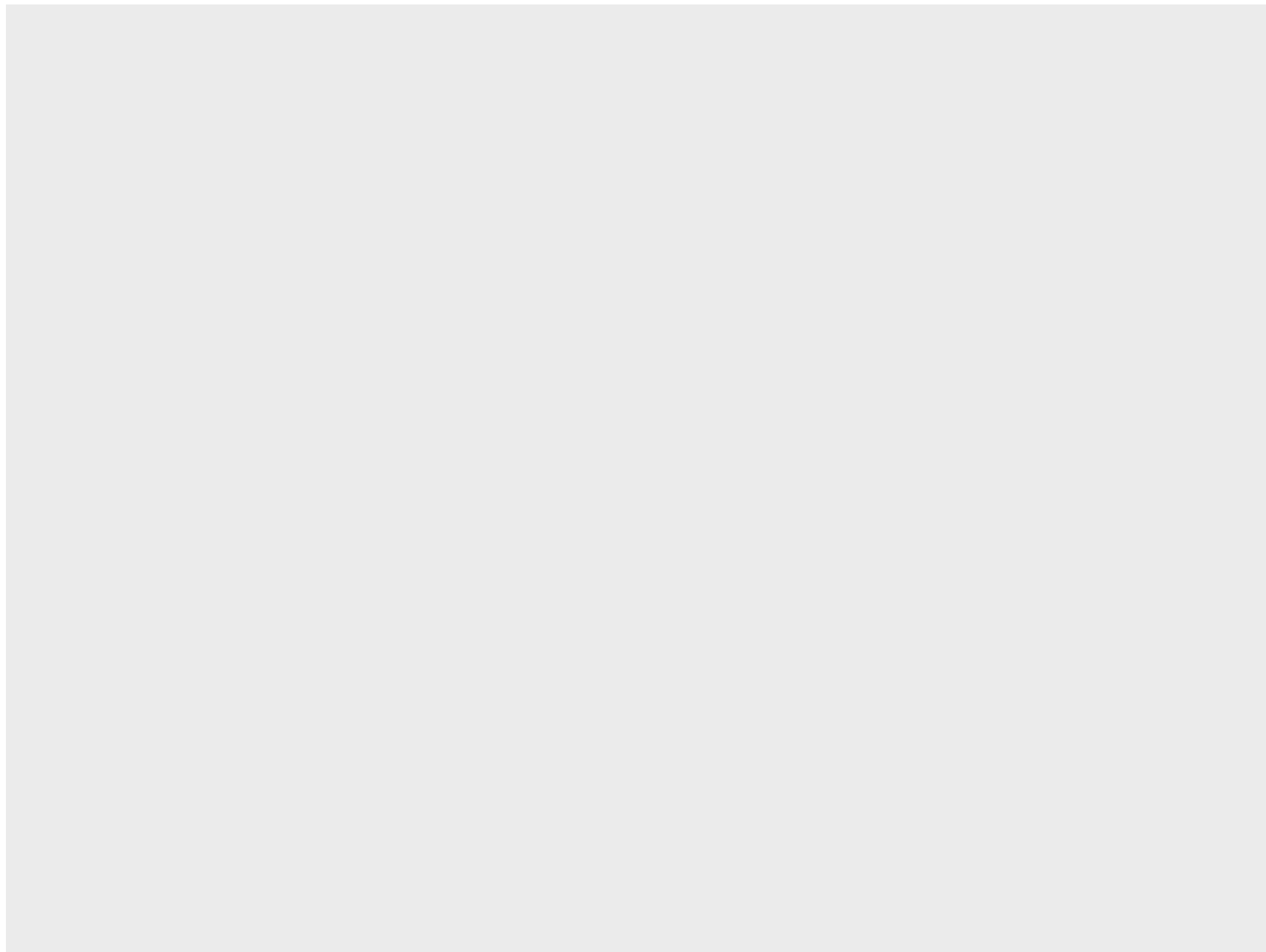
0	0	0	0	0	0
1	0	0	0	0	0
0	1	1	0	1	1
1	0	0	1	0	0
1	0	0	1	1	1
0	1	1	0	0	1
0	1	1	0	0	0

0	0	0	0	0	0
1	0	0	0	0	0
0	1	1	0	1	1
1	0	0	1	0	0
1	0	0	1	1	1
0	1	1	0	0	1
0	1	1	0	0	0

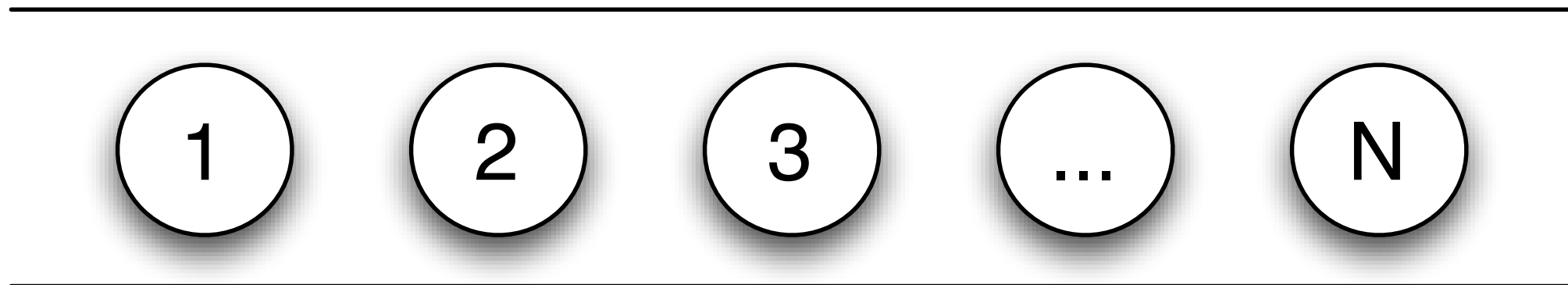
Growing Pattern CPM

0	0	0	0	0	0
1	0	0	0	0	0
0	1	1	0	1	1
1	0	0	1	0	0
1	0	0	1	1	1
0	1	1	0	0	1
0	1	1	0	0	0

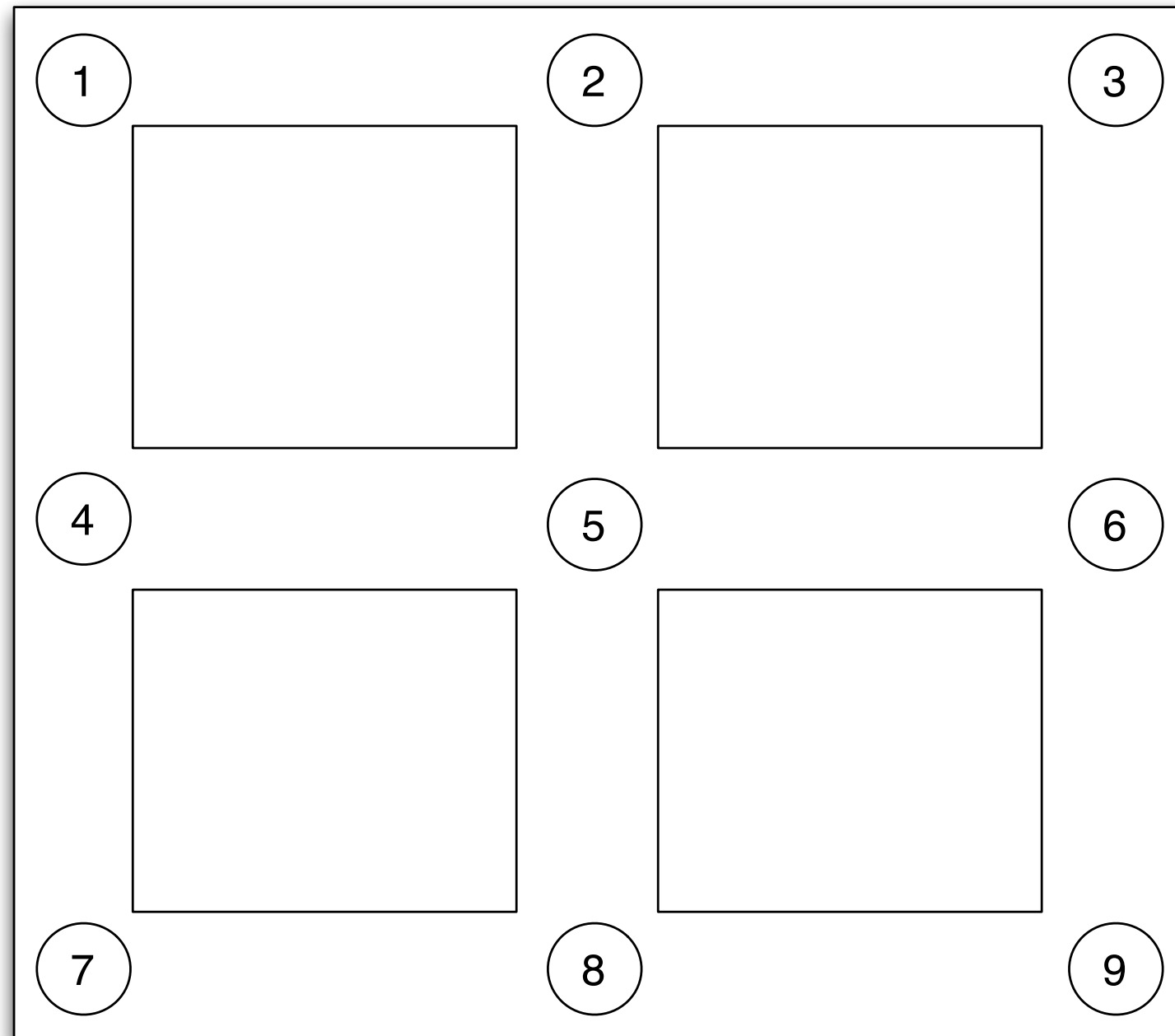
Growing Pattern CPM



CPM Problem

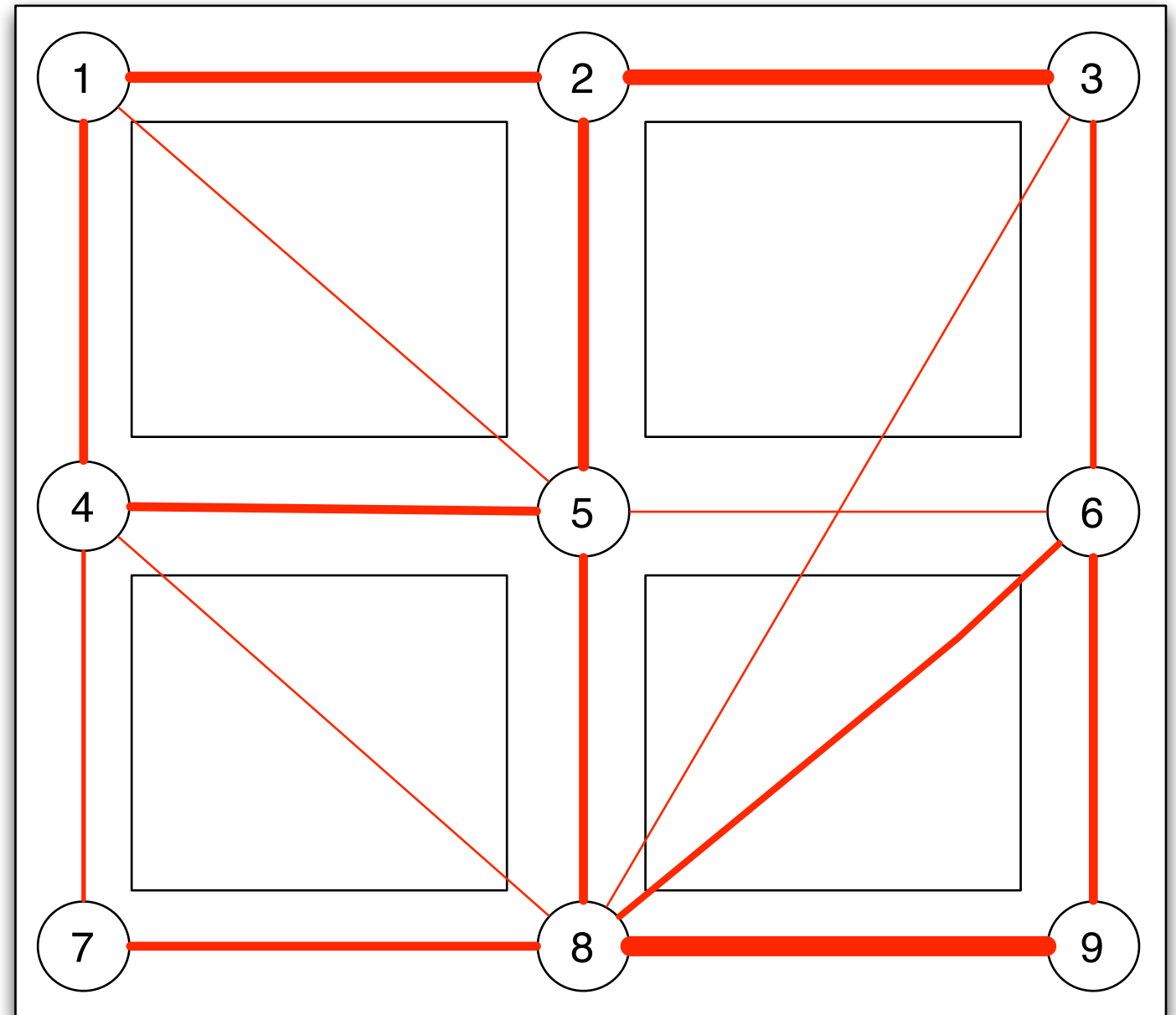


CPM - Problem



CPM - Current

Dynamically create
patterns based on
correlation score.



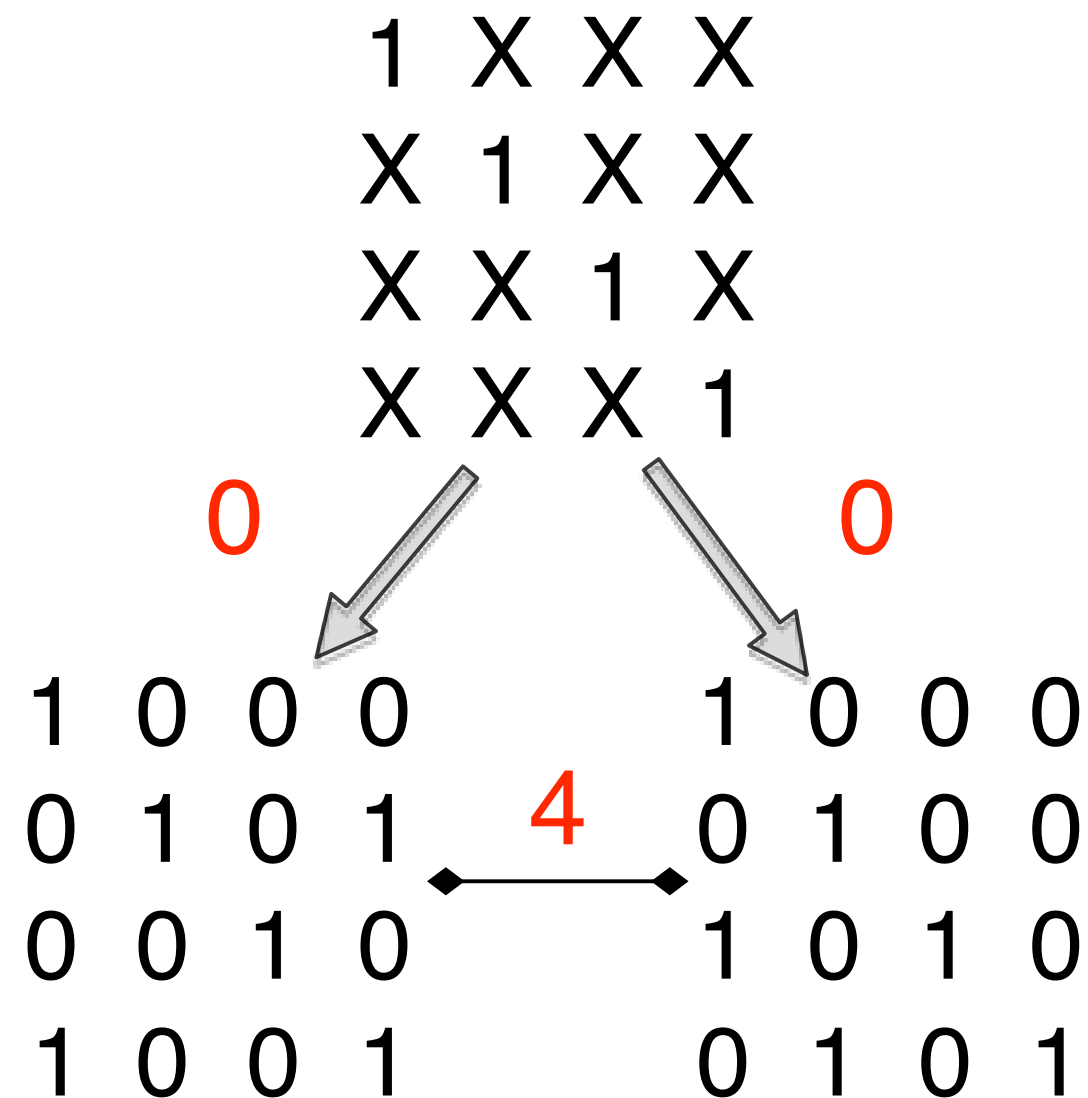
CPM - Current

- Construct a model per sensor
- Preprocess the data
 - Compress over time

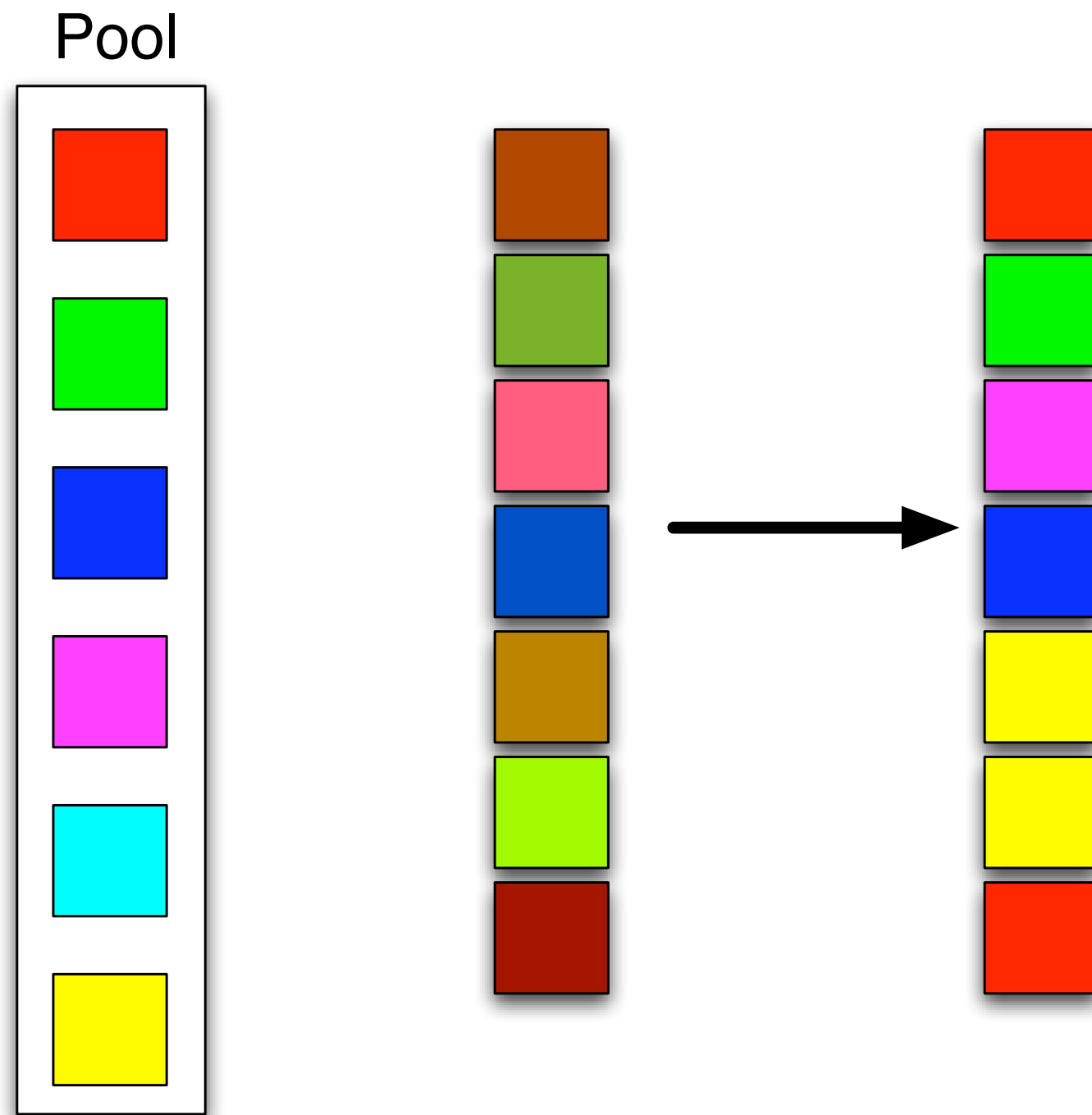
Major Ideas

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- **Minimum Description Length**
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Minimum Description Length



Minimum Description Length



Minimum Description Length

- Demo by Bill

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What is an Anomaly



Context Discovery and Model Linking

- How can we define contexts using only state information?
 - Prior definition
 - Use a technique to dynamically create them.
 - Simulated Annealing with Latent Semantic Analysis as a move approximation metric

Non Pursued Paths

- Match previously defined activities to events in dataset
- Attempt to classify nodes dynamically if they exhibit behavior outside the defined realm.
- Utilizing some of the more standard Machine Learning algorithms