

## Theory Discovery Tools

### Building Your Own Clustering Theory - Week 1

BSc Machine Learning for Innovation  
*Discovery Cards & Worksheets*

#### How to Use These Tools

These cards and worksheets guide you through discovering clustering theory yourself. Cut out the cards, use them during exercises, and document your discoveries in the worksheets.

## Part 1: Discovery Cards

### Observation Cards

Cut along dotted lines. Use these to guide your observations.

#### **OBSERVATION CARD 1**

*the spacing between points?*

What do you notice about...

Look for: gaps, clusters, density

Document: Where are points close?

Far?

Count: How many groups do you see?

#### **OBSERVATION CARD 2**

*the shapes of groups?*

What do you notice about...

Look for: circles, lines, blobs

Document: Are groups round?

Stretched?

Question: Why these shapes?

#### **OBSERVATION CARD 3**

*points that don't fit?*

What do you notice about...

Look for: isolated points, bridges

Document: Which points are alone?

Consider: Are they noise or important?

#### **OBSERVATION CARD 4**

*the center of groups?*

What do you notice about...

Look for: dense areas, empty centers

Document: Where is the "middle"?

Measure: Distance from center to edges

## Hypothesis Cards

Use these to form theories about clustering.

If... <b>HYPOTHESIS CARD 1</b>
points are within distance D of each other
<b>Then...</b> _____
<b>Because...</b> _____
Test this with: Different D values

If... <b>HYPOTHESIS CARD 2</b>
we use the average position
<b>Then...</b> _____
<b>Because...</b> _____
Test this with: Calculate group averages

If... <b>HYPOTHESIS CARD 3</b>
groups have similar sizes
<b>Then...</b> _____
<b>Because...</b> _____
Test this with: Count points per group

If... <b>HYPOTHESIS CARD 4</b>
we know the number of groups
<b>Then...</b> _____
<b>Because...</b> _____
Test this with: Try different K values

## Test Cards

Validate your theories with these experiments.

### TEST CARD 1

**Try your theory on...**

A dataset with clear, separated groups

*Does it work?* YES / NO

*What happened?* \_\_\_\_\_

*Adjust theory:* \_\_\_\_\_

### TEST CARD 2

**Try your theory on...**

A dataset with overlapping groups

*Does it work?* YES / NO

*What happened?* \_\_\_\_\_

*Adjust theory:* \_\_\_\_\_

### TEST CARD 3

**Try your theory on...**

A dataset with different densities

*Does it work?* YES / NO

*What happened?* \_\_\_\_\_

*Adjust theory:* \_\_\_\_\_

### TEST CARD 4

**Try your theory on...**

A dataset with outliers

*Does it work?* YES / NO

*What happened?* \_\_\_\_\_

*Adjust theory:* \_\_\_\_\_

## Reflection Cards

Understand why your discoveries work.

<del>This works because</del> REFLECTION CARD 1
My clustering rule succeeds when: _____
It fails when: _____
The        key        insight        is:

<del>This works because</del> REFLECTION CARD 2
Distance matters because: _____
Different distances give: _____
The        best        distance        is:

<del>This works because</del> REFLECTION CARD 3
Groups form naturally when: _____
The pattern I see is: _____
This    relates    to    innovation    by:

<del>This works because</del> REFLECTION CARD 4
Outliers are important because: _____
They represent: _____
In    innovation,    outliers    are:

## Part 2: Theory Building Worksheets

### Worksheet 1: Rule Creator

*Template for writing clustering rules*

#### Create Your Clustering Algorithm

##### Step 1: Define “Belongs Together”

Two points belong in the same group if:

\_\_\_\_\_

\_\_\_\_\_

##### Step 2: Write Your Algorithm

1. Start with: \_\_\_\_\_

2. For each point: \_\_\_\_\_

3. Calculate: \_\_\_\_\_

4. Assign to group if: \_\_\_\_\_

5. Repeat until: \_\_\_\_\_

6. Stop when: \_\_\_\_\_

##### Step 3: Handle Special Cases

If a point doesn't fit any group: \_\_\_\_\_

If groups overlap: \_\_\_\_\_

If I don't know K: \_\_\_\_\_

##### Step 4: Name Your Algorithm

My clustering algorithm is called: \_\_\_\_\_

Because it: \_\_\_\_\_

**Worksheet 2: Pattern Finder***Guide for identifying regularities***Finding Patterns in Data****Visual Pattern Recognition**

Pattern Type	I See This	It Means	Algorithm Needed
Circular groups	YES / NO		
Elongated groups	YES / NO		
Nested groups	YES / NO		
Chain connections	YES / NO		
Varying density	YES / NO		
Clear outliers	YES / NO		

**Numerical Pattern Recognition**

When I count groups at different scales:

- Fine scale (small distance): \_\_\_\_\_ groups
- Medium scale: \_\_\_\_\_ groups
- Coarse scale (large distance): \_\_\_\_\_ groups

The pattern is: \_\_\_\_\_

**Innovation Pattern Recognition**

In my field, natural groups form around:

- \_\_\_\_\_
- \_\_\_\_\_
- \_\_\_\_\_

These patterns suggest the clustering approach: \_\_\_\_\_

**Worksheet 3: Theory Tester***Framework for validation***Testing Your Clustering Theory****Theory Statement**

My theory: \_\_\_\_\_

\_\_\_\_\_

**Prediction Table**

If my theory is correct...	I expect to see...	Actually saw...
On dataset A:		
On dataset B:		
On dataset C:		

**Success Metrics**

My clustering is good when:

- Within-cluster distance is: \_\_\_\_\_
- Between-cluster distance is: \_\_\_\_\_
- Number of outliers is: \_\_\_\_\_
- Groups are balanced: YES / NO / SOMETIMES

**Theory Refinement**

What worked: \_\_\_\_\_

What failed: \_\_\_\_\_

Revised theory: \_\_\_\_\_

\_\_\_\_\_

## Worksheet 4: Concept Connector

*Linking discoveries to formal terms*

### From Your Words to Formal Theory

#### Translation Table

Connect what you discovered to the formal terminology:

What I Called It	Formal Term	Why It's Important
"Group middle point"	Centroid	Represents the cluster
"How spread out"		
"Doesn't fit anywhere"		
"How tight the group is"		
"Space between groups"		
"Best number of groups"		

#### Algorithm Matching

My approach is most similar to:

- ☐ K-means (uses centers, equal groups)
- ☐ DBSCAN (uses density, finds shapes)
- ☐ Hierarchical (builds tree, multiple scales)
- ☐ GMM (uses probability, overlapping groups)
- ☐ Something new: \_\_\_\_\_

#### Innovation Applications

My clustering discovery applies to innovation by:

1. \_\_\_\_\_
2. \_\_\_\_\_
3. \_\_\_\_\_

#### Key Insight

The most important thing I learned about clustering is:

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