

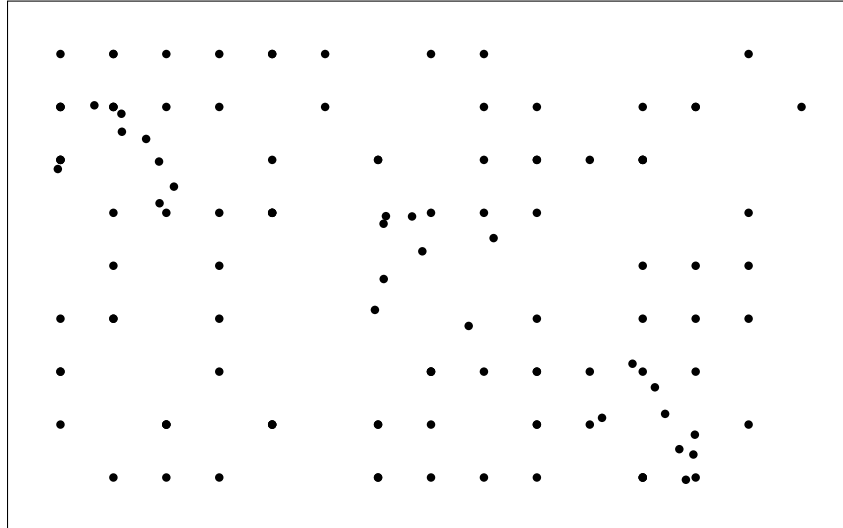
Pattern Discovery Worksheet

Pre-Session Activity

Machine Learning for Innovation - Week 1

Exercise 1: The Sorting Challenge

Task: Look at the dots below. Group them into categories by drawing circles around dots that belong together.



Questions:

1. How many groups did you find? _____
2. How long did this take you? _____
3. What if there were 10,000 dots instead of 100? How long would it take? _____
4. What rule did you use to decide which dots belong together?

Exercise 2: The Innovation Challenge

Task: Below are 30 innovation concepts. Organize them into meaningful categories. Write the category names and list which innovations belong in each.

Electric transportation	Photo sharing
Music streaming	Instant messaging
Video communication	Online marketplaces
Food delivery automation	Video streaming
Digital payments	Virtual reality
Remote healthcare	3D printing
Social networking	Drone delivery
Cloud storage	Artificial intelligence
Voice assistants	Gene editing
Renewable energy	Autonomous vehicles
Online education	Blockchain
Fitness tracking	Augmented reality
Smart home devices	Quantum computing
Cryptocurrency	Internet of things
Ride sharing	Sustainable packaging

Your Categories:

1. Category 1: _____

Items: _____

2. Category 2: _____

Items: _____

3. Category 3: _____

Items: _____

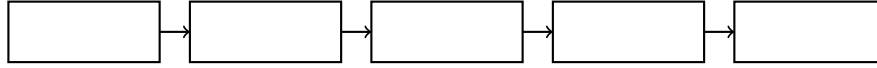
4. Category 4: _____
Items: _____

Questions:

1. Were some innovations hard to categorize? Which ones? _____
2. Could the same innovation fit multiple categories? _____
3. What if you had 5000 innovations to sort? _____

Exercise 3: Technical Process Sequencing

Task: Below are five stages of a technical analysis process, listed alphabetically. Place them in the correct operational sequence.



Stages (alphabetical):

- **Deploy:** Put solution into production
- **Data:** Collect raw information
- **Evaluate:** Test solution performance
- **Model:** Build computational solution
- **Preprocess:** Clean and prepare information

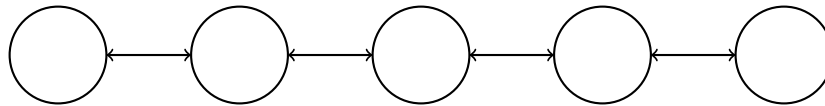
Questions:

1. Why must this sequence be followed strictly? _____

2. What happens if you skip the second step? _____
3. At which step do errors become most costly? _____
4. Which step requires the most computational resources? _____
5. Can any two steps be performed simultaneously? _____

Exercise 4: Human-Centered Process Sequencing

Task: Below are five stages of a human-centered problem-solving process, listed alphabetically. Arrange them in logical sequence.



Stages (alphabetical):

- **Define:** Articulate the problem
- **Empathize:** Understand user needs
- **Ideate:** Generate potential solutions
- **Prototype:** Build preliminary solution
- **Test:** Validate with users

Questions:

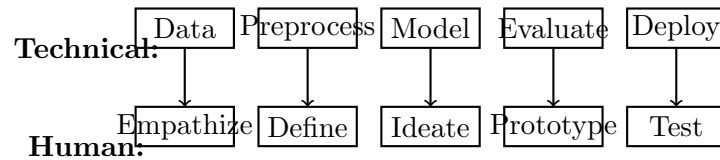
1. Unlike the technical process, why are the arrows bidirectional? _____

2. Which stage requires the most human interaction? _____
3. Can you begin at stage 3? Why or why not? _____

4. How does stage 1 inform stage 2? _____
5. Why might you return to stage 1 after stage 5? _____

Exercise 5: Methodological Integration

Task: Identify how each technical stage could inform the corresponding human-centered stage.



Analysis Questions:

1. How does "Data" inform "Empathize"? _____

2. How does "Define" constrain "Preprocess"? _____

3. Where is the strongest connection? Why? _____

4. Where is the weakest connection? Why? _____

5. What is lost if these pipelines operate independently? _____

6. Identify one feedback loop between pipelines: _____

Reflection: Methodological Synthesis

Synthesize your observations from all exercises:

1. Compare the rigidity of technical versus human-centered processes:

2. Explain why iteration is essential in one pipeline but problematic in the other:

3. Identify the primary constraint in each pipeline:

4. Describe how quantitative and qualitative methods complement each other:

5. What is the epistemological difference between "Model" and "Ideate"?

6. Propose how automation could enhance human-centered design:

In our first session, we will explore how these two pipelines converge to create innovation opportunities through the systematic application of machine learning to design thinking challenges.