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A Tutorial on Hypergraph Neural Networks: An In-Depth and Step-by-Step Guide

Part 5. Applications



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Part 5. Applications

Part 1.
Introduction

Part 2.
Inputs

Part 3.
Message
Passing

Part 4.
Training
Strategies

Part 5.
Applications

Part 6.
Discussions

The slides are available at <https://sites.google.com/view/hnn-tutorial>



Presenter



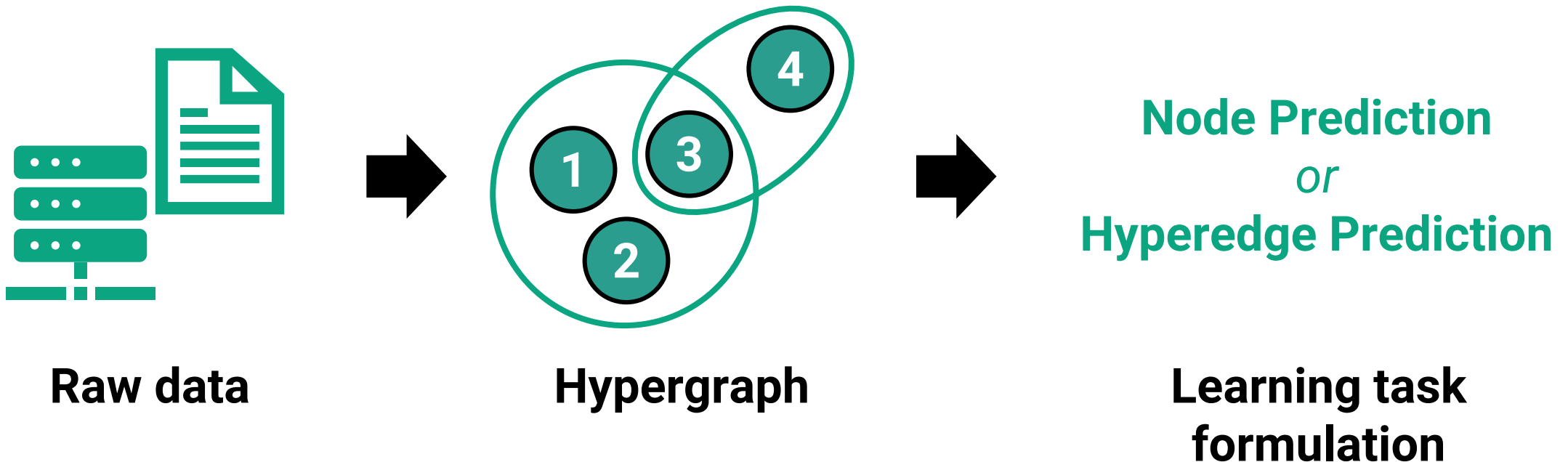
Soo Yong Lee.

Ph.D. Student @

KAIST

What are Some Issues in Applying HNNs?

- Two major issues include:
 - Q1) How to **express the raw data** as hypergraphs?
 - Q2) How to **formulate the learning task** for HNN?



What are Notable Applications of HNNs?

- In the following domains, we will cover how the issues have been addressed:



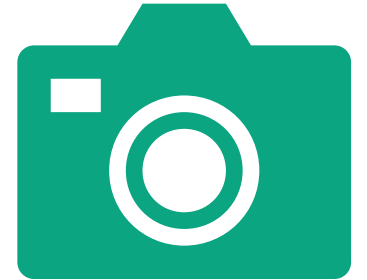
Recommender system



Bioinformatics & Medical science



Time series analysis



Computer vision

What are Notable Applications of HNNs? (cont.)

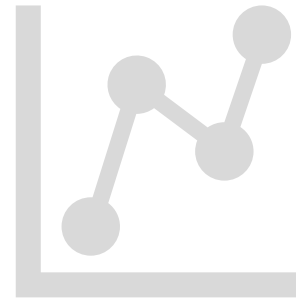
- In the following domains, we will cover how the issues have been addressed:



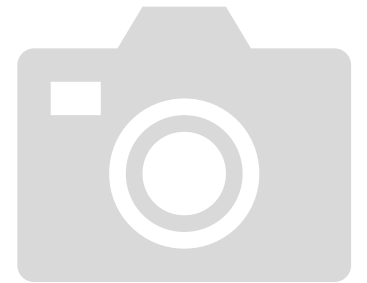
Recommender system



Bioinformatics & Medical science



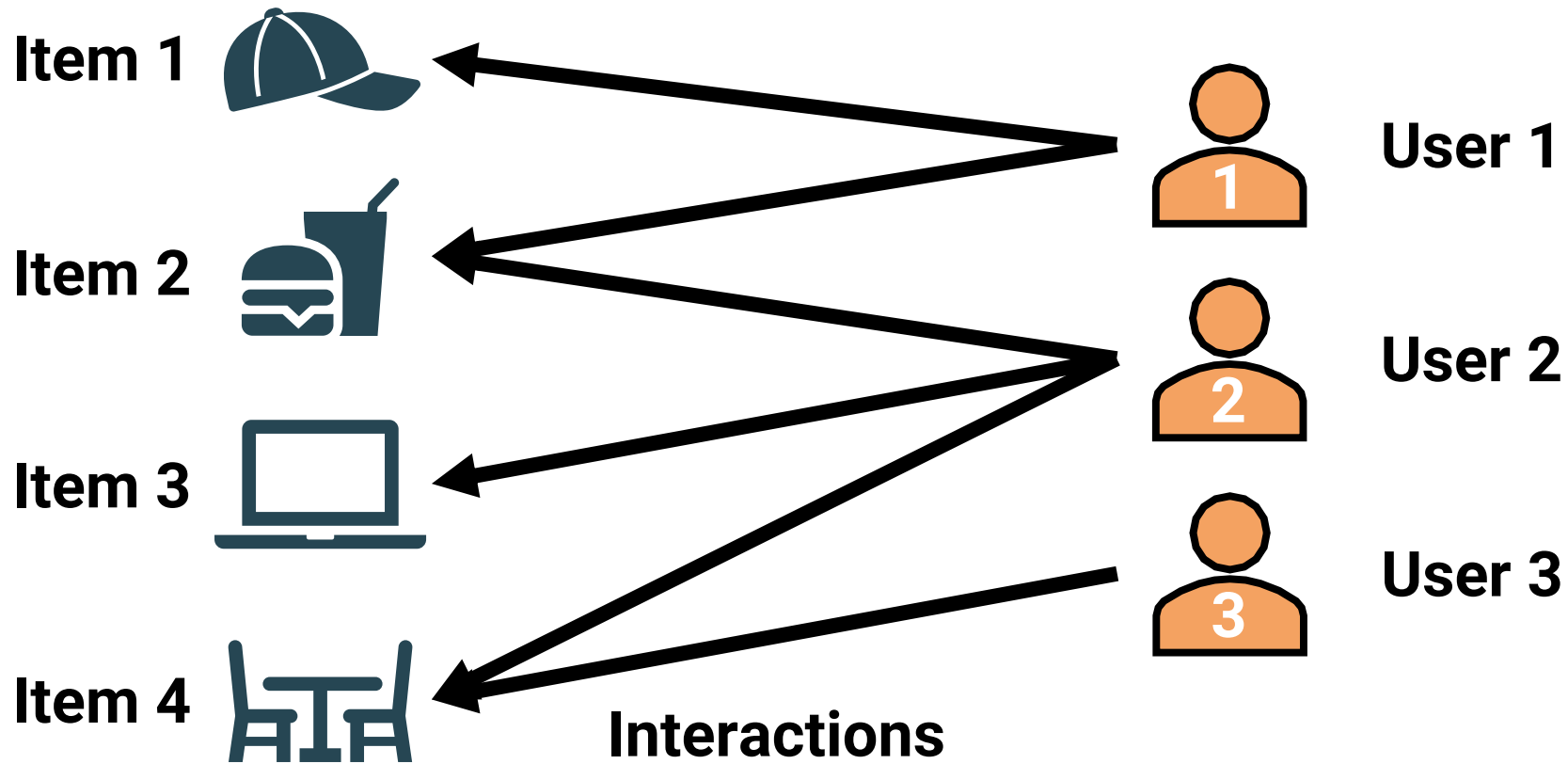
Time series analysis



Computer vision

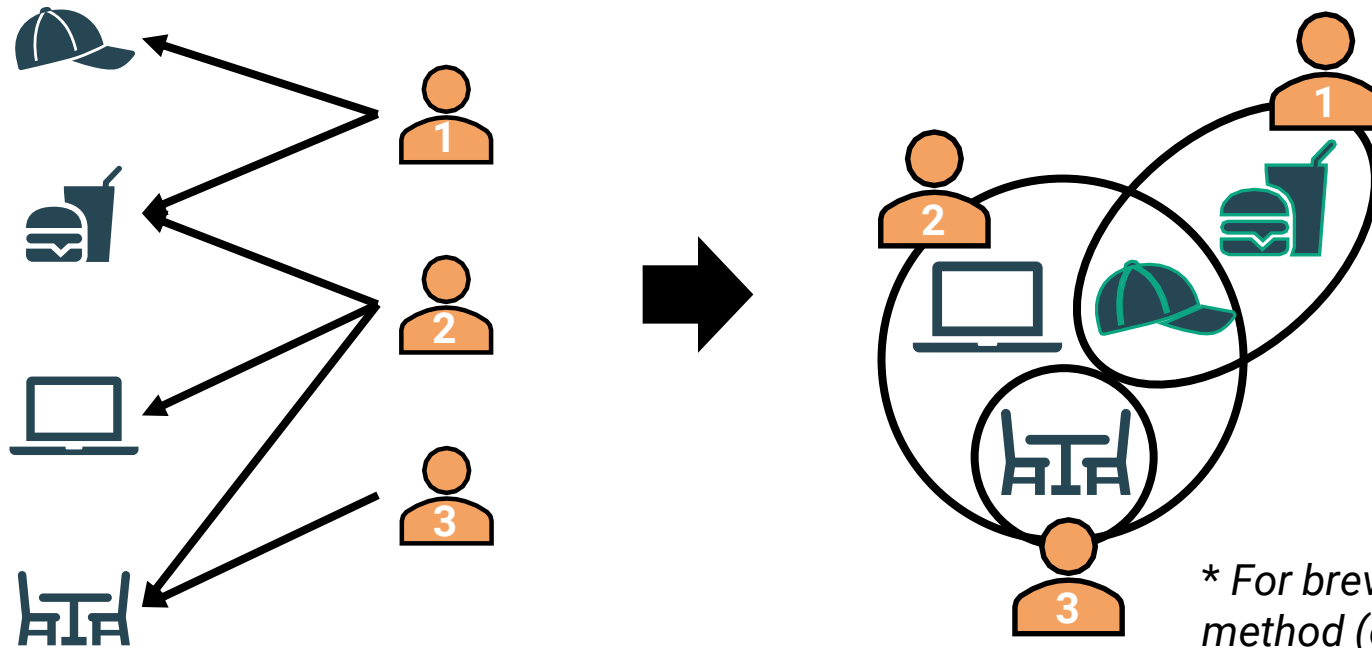
Recommendation: Hypergraph Construction

- **Raw data** typically include users & items and their interactions.



Recommendation: Hypergraph Construction (cont.)

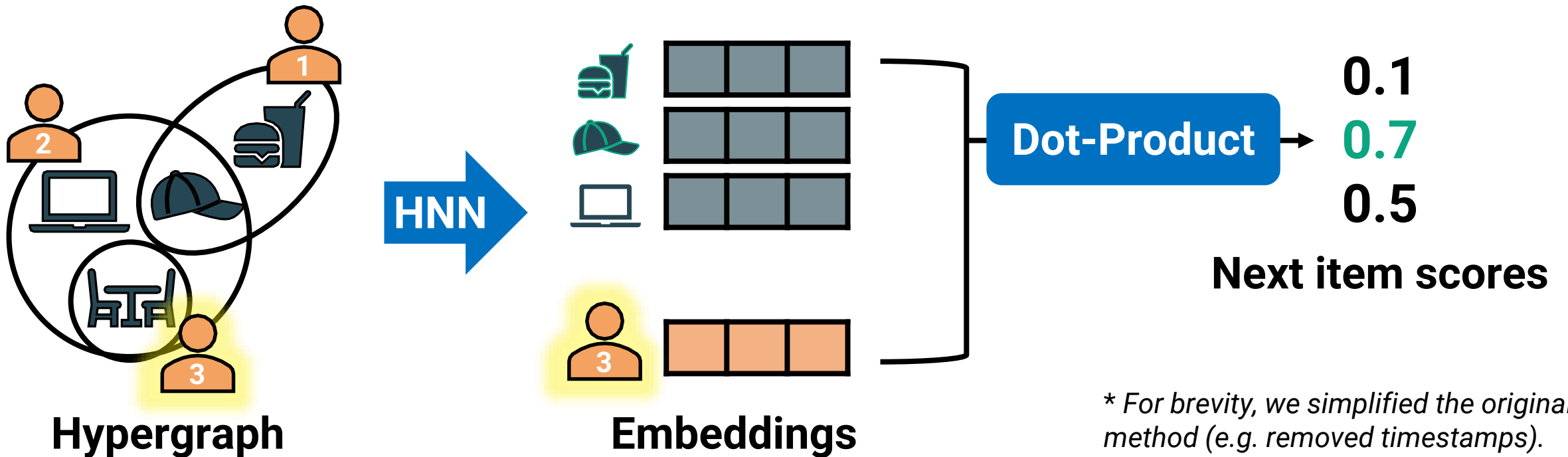
- Wang et al. (2020) converted the **interactions** into a hypergraph
 - A node is an item; a hyperedge is a user.
 - A user hyperedge connects all item nodes that it interacted with.



* For brevity, we simplified the original method (e.g. removed timestamps).

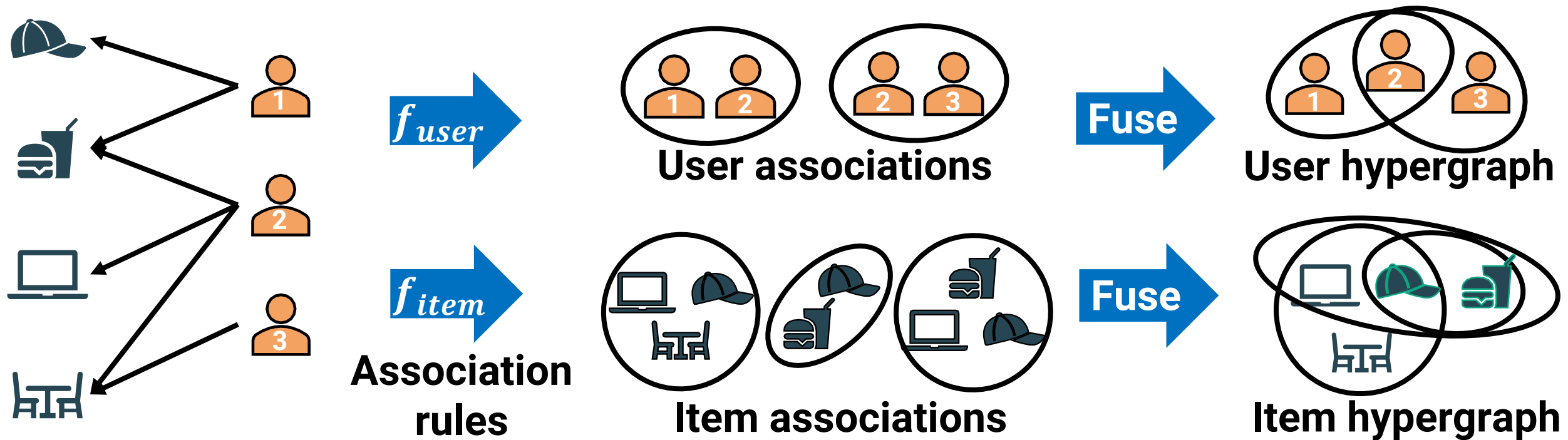
Recommendation: Task Formulation

- Wang et al. (2020) used an HNN for **next-item recommendation**.
 - An HNN encodes nodes and hyperedges.
 - The embeddings are used to predict items that users will interact next.



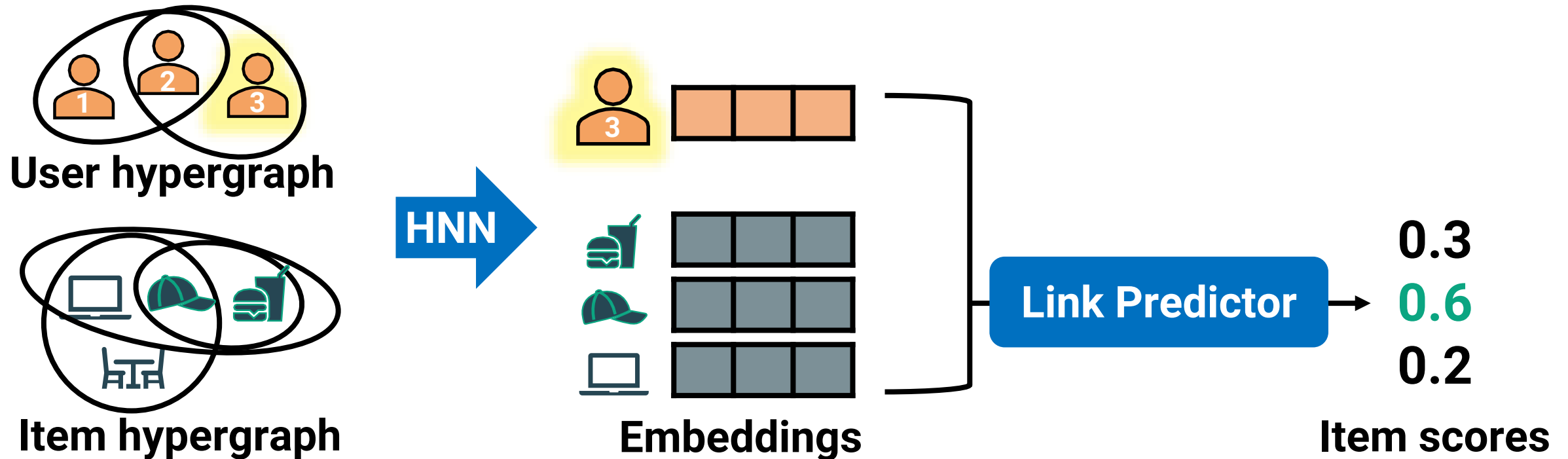
Recommendation: Hypergraph Construction (cont.)

- Ji et al. (2020) used **association rules** to construct two hypergraphs.
 - The pre-determined association rules aim to capture high-order information among users and among items, returning two hypergraphs.



Recommendation: Task Formulation (cont.)

- Ji et al. (2020) used an HNN for **collaborative filtering**.
 - An HNN encodes nodes and hyperedges.
 - The embeddings are used to recommend items for users.



What are Notable Applications of HNNs? (cont.)

- In the following domains, we will cover how the issues have been addressed:



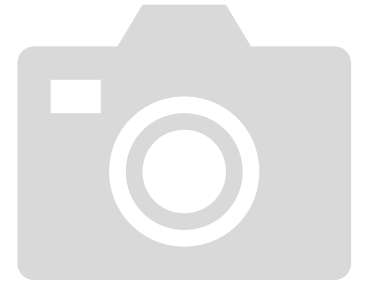
Recommender
system



**Bioinformatics &
Medical science**



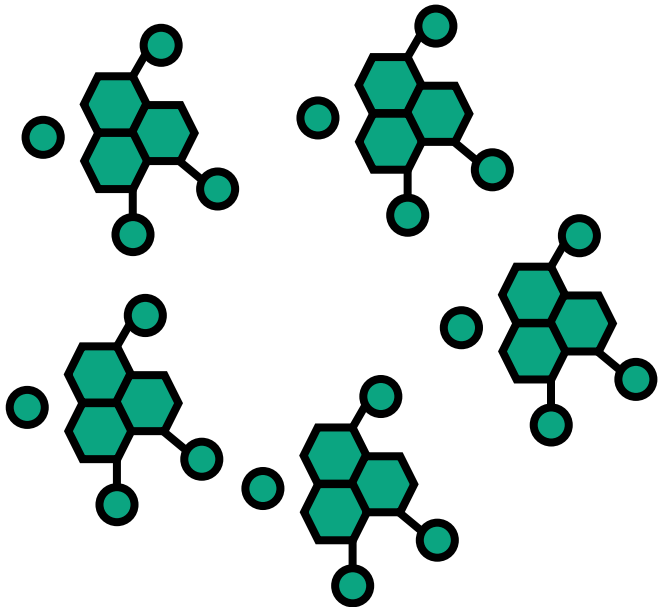
Time series
analysis



Computer
vision

Bioinformatics: Hypergraph Construction

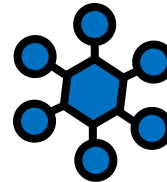
- **Raw data** typically include molecular-level structures and their relations.



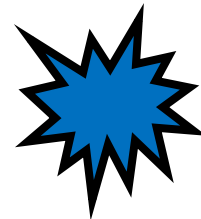
Molecular-level structures



**Presence within
each drug**



**Association with
disease**

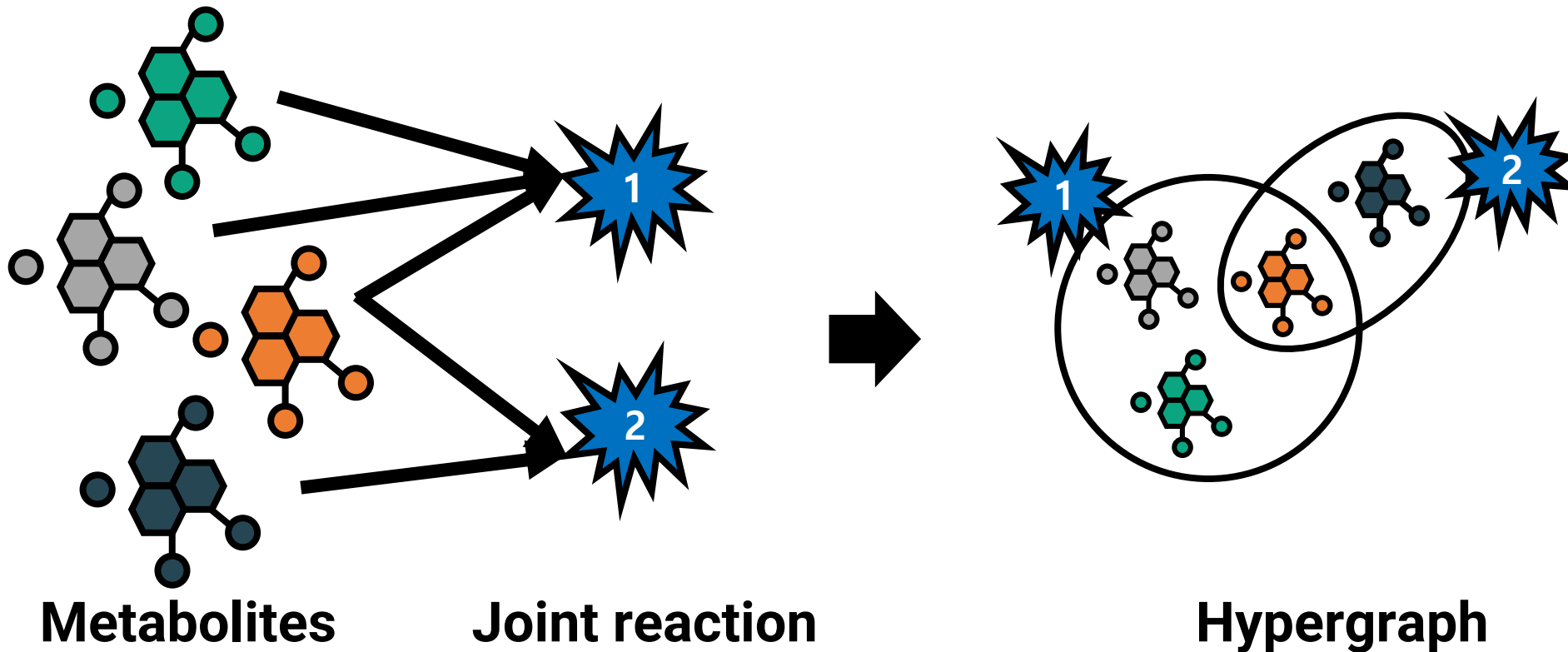


Joint reaction

Joint relations examples

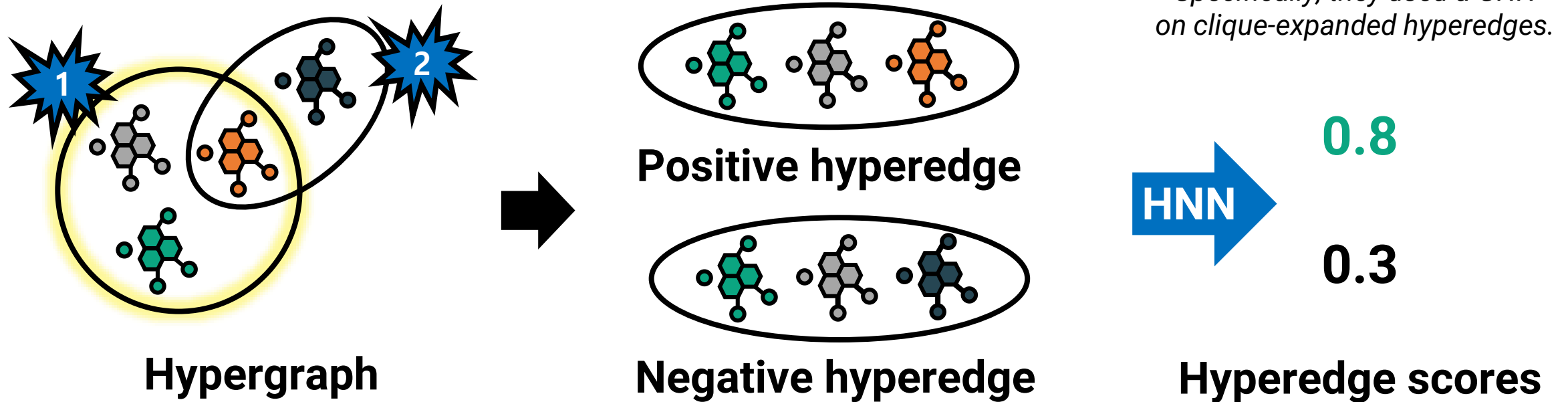
Bioinformatics: Hypergraph Construction (cont.)

- Chen et al. (2023) transforms a **metabolic network** into a hypergraph.
 - A node is metabolite; a hyperedge connects those with joint reaction.



Bioinformatics: Task Formulation

- Chen et al. (2023) used HNN to **predict missing metabolic reaction**.
 - An HNN learns to classify negative and positive hyperedges, where negative hyperedges represent false joint reactions.



fMRI Data: Hypergraph Construction

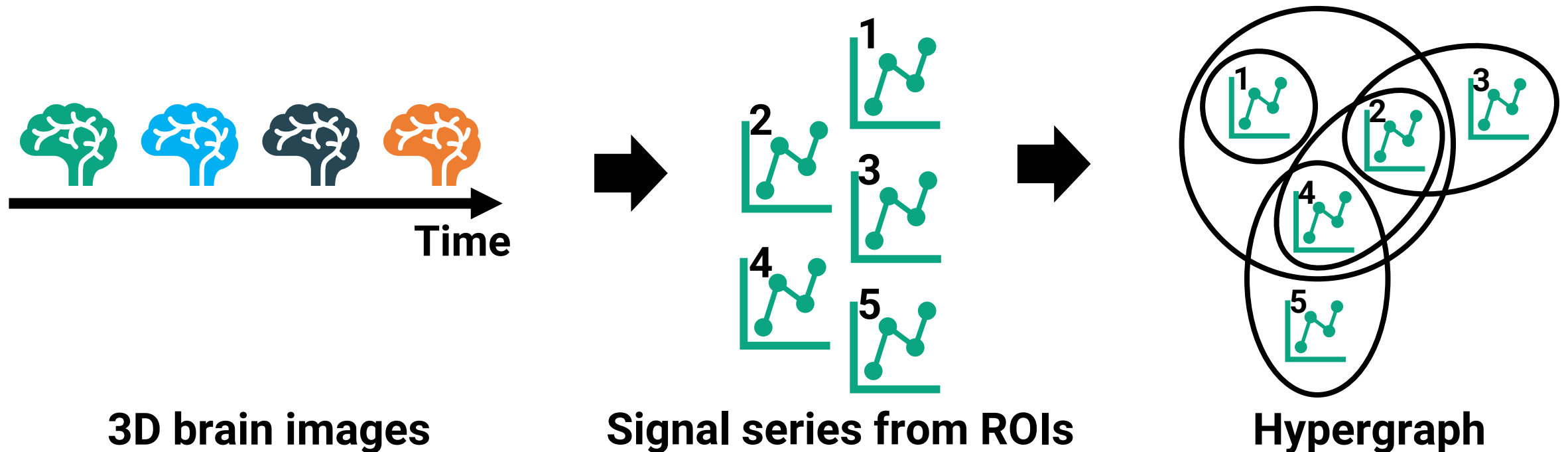
- **Raw data** typically is series of brain images with brain signals.
 - The images can be parcellated into signals from region-of-interests (ROIs).



3D brain images in time series

fMRI Data: Hypergraph Construction (cont.)

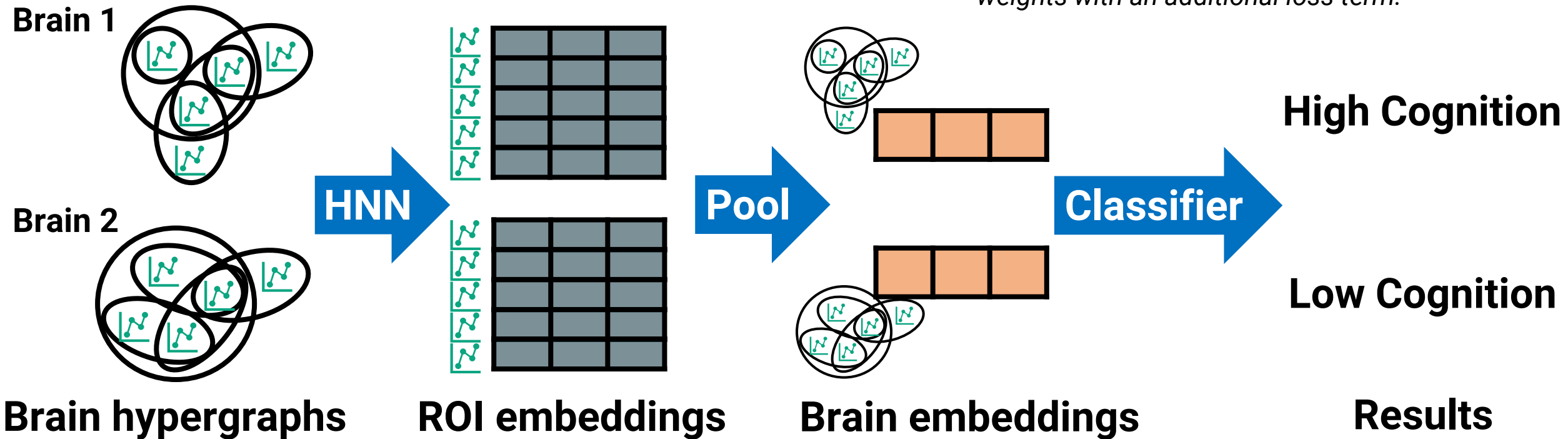
- Wang et al. (2023) converted **fMRI** data into a hypergraph.
 - A node is a ROI; for each ROI, its hyperedge connects the most similar ROIs (estimated by a learning algorithm).



fMRI Data: Task Formulation

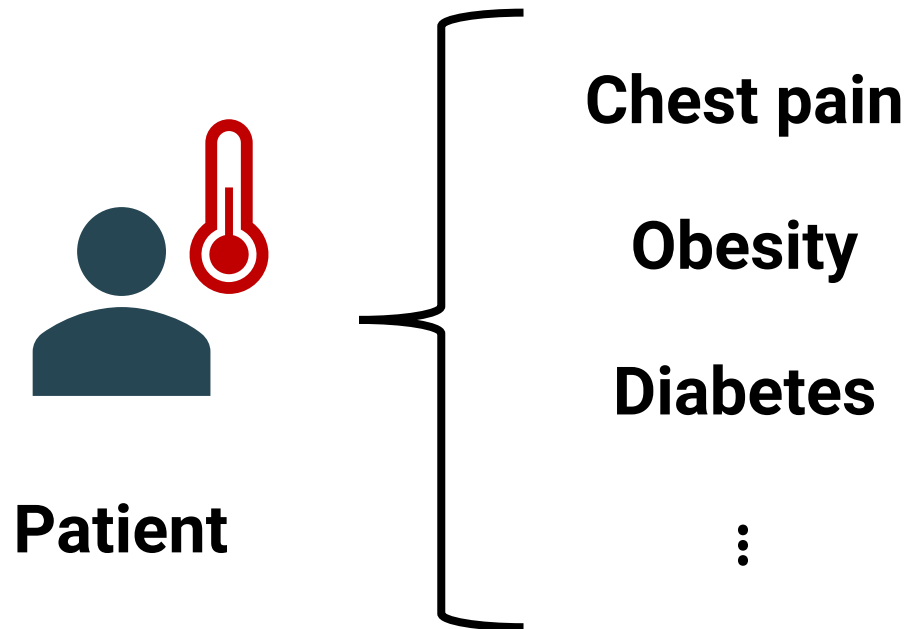
- Wang et al. (2023) used an HNN to **predict ages & cognitive functions**.
 - An HNN encodes brain hypergraphs for classification.

** Specifically, they also estimated hyperedge weights with an additional loss term.*



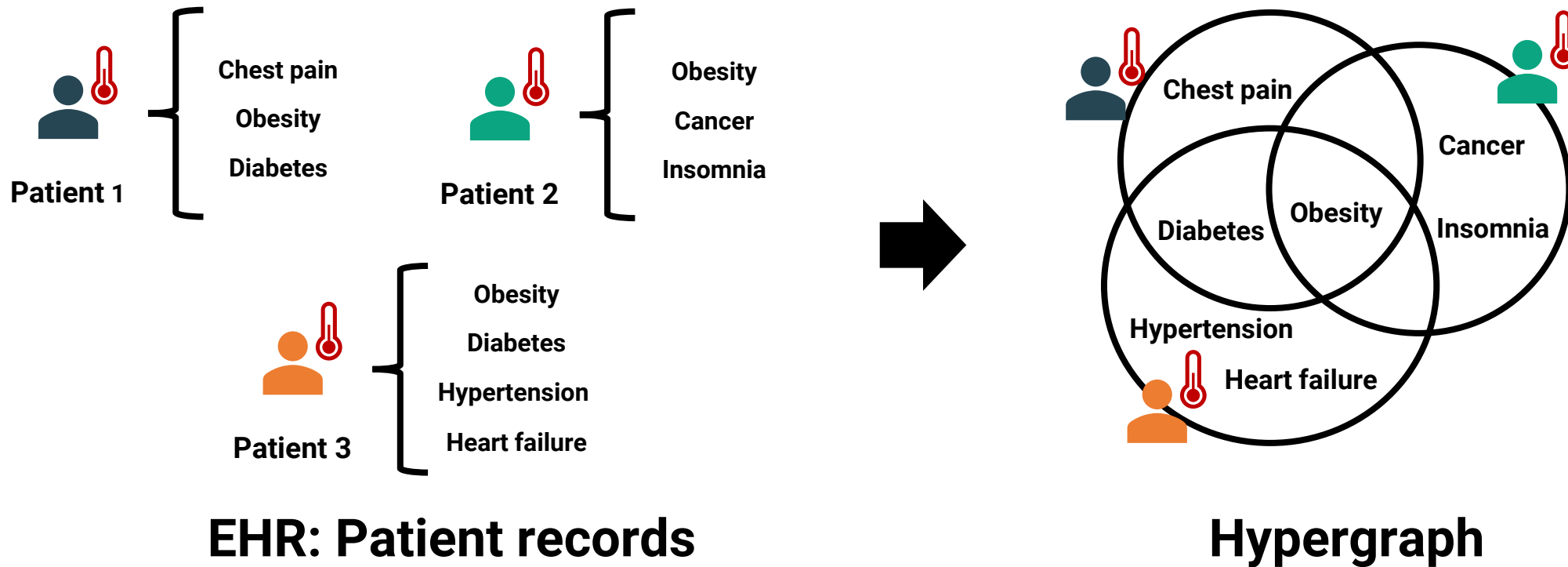
EHR: Hypergraph Construction

- **Raw data** typically include medical codes and records of each patient.



EHR: Hypergraph Construction (cont.)

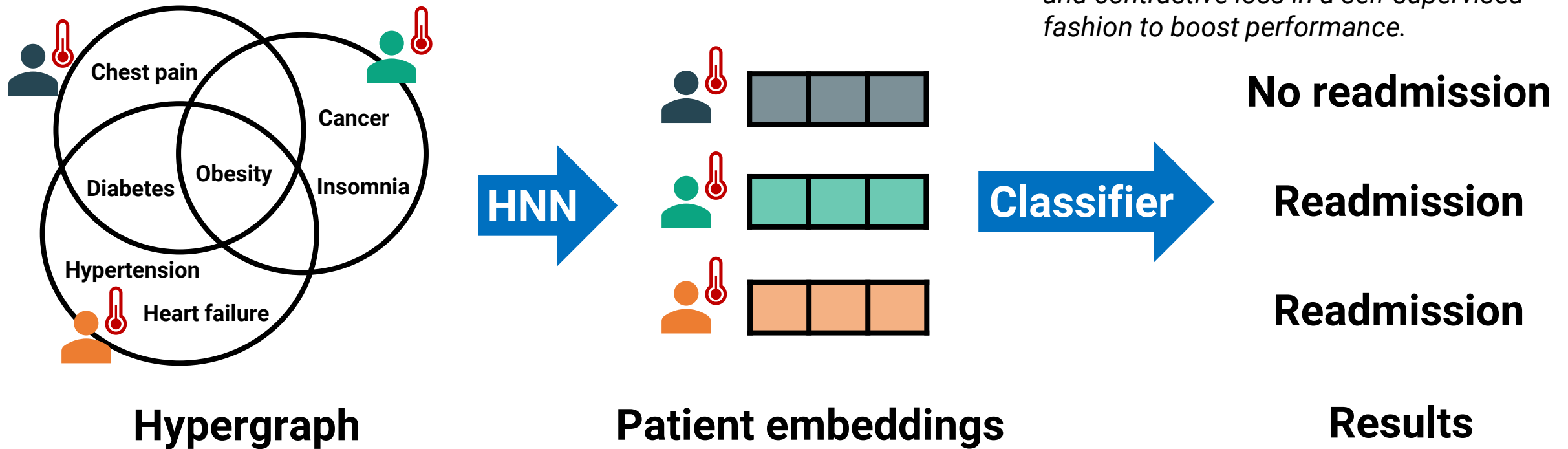
- Cai et al. (2022) converted **patient records** into a hypergraph.
 - A node is a medical code; a hyperedge is each patient visit.



EHR: Task Formulation

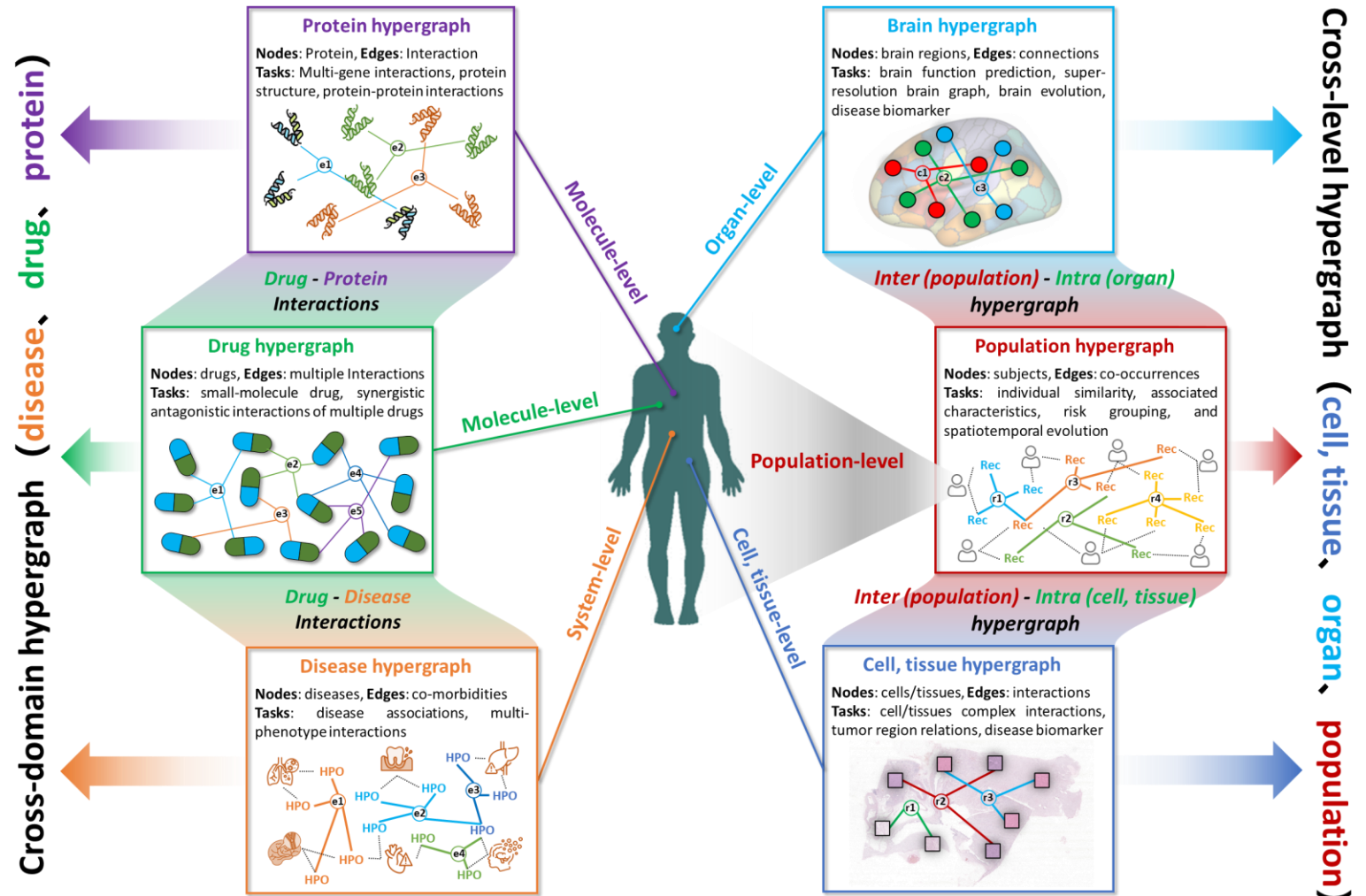
- Cai et al. (2022) used an HNN to **predict mortality & readmission**.
 - The HNN encodes hyperedges for their classification.

** Specifically, they leveraged multiple views and contrastive loss in a self-supervised fashion to boost performance.*



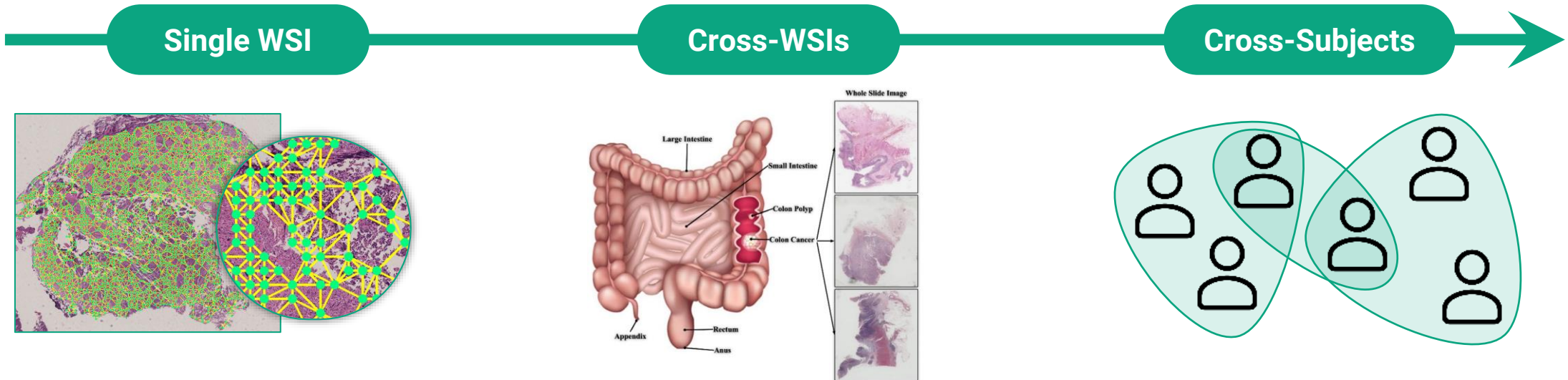
Hypergraph in Medicine Science

- Cell and tissue hypergraphs
- Organ hypergraphs
- Biological system hypergraphs
- Population-level hypergraphs



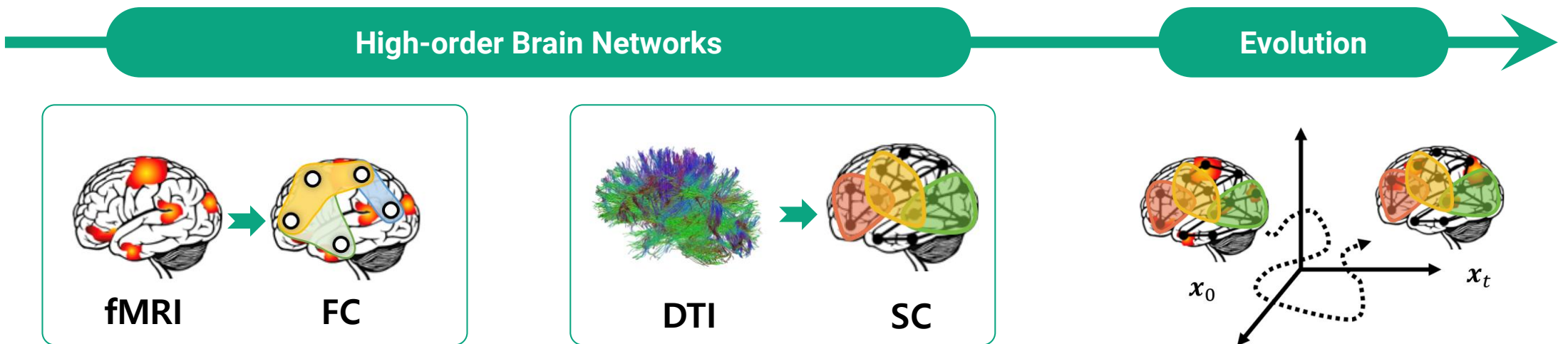
Pathology Analysis

- Gigapixel WSIs contains multiple high-order correlations
 - Within a single WSI
 - Cross-WSIs for a single subject
 - Cross-subjects



Brain Network

- Functional and Structural Brain Network contains high-order correlations
 - Within individual functional network or structural network
 - Functional and structural network coupling
 - Brain network evolution



What are Notable Applications of HNNs? (cont.)

- In the following domains, we will cover how the issues have been addressed:



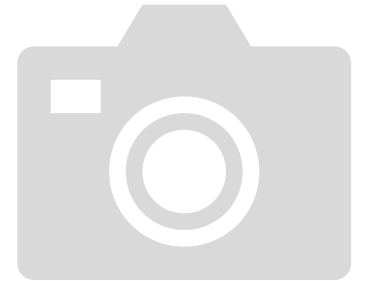
Recommender
system



Bioinformatics &
Medical science



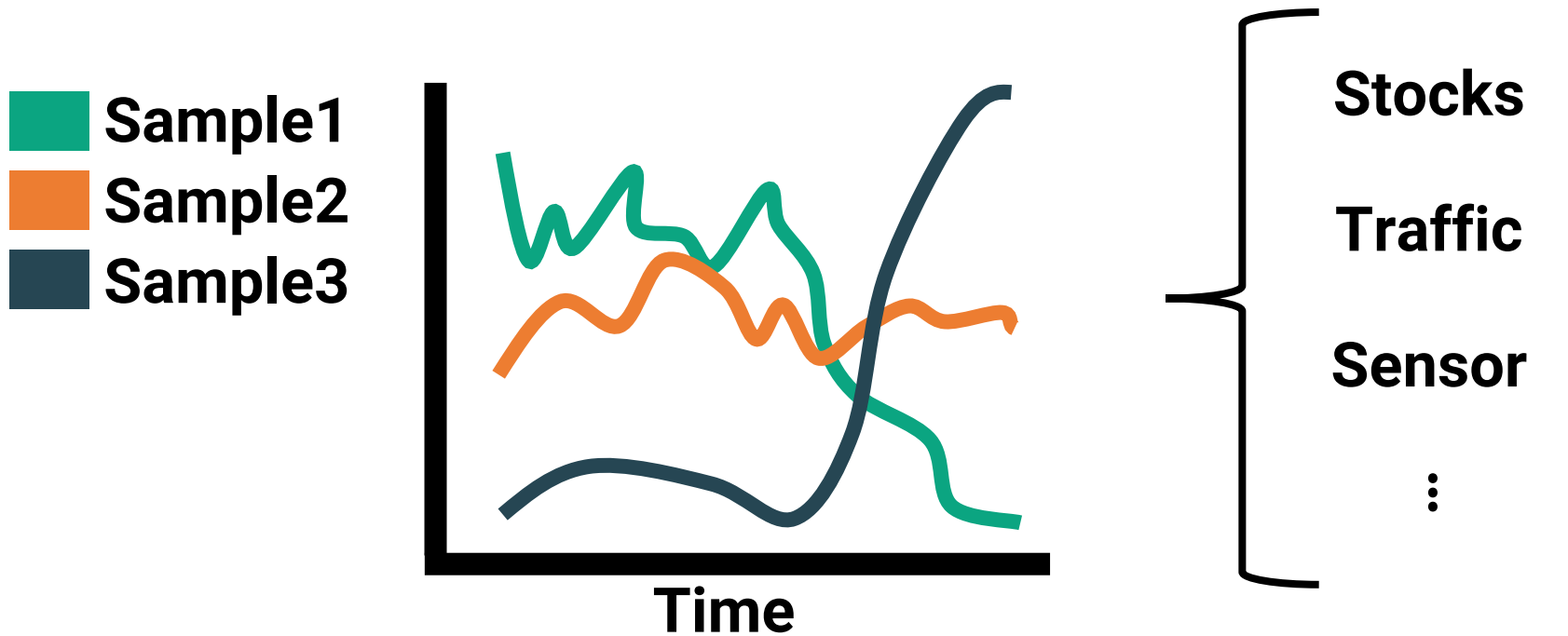
**Time series
analysis**



Computer
vision

Time Series: Hypergraph Construction

- **Raw data** typically are periodically recorded data over time.



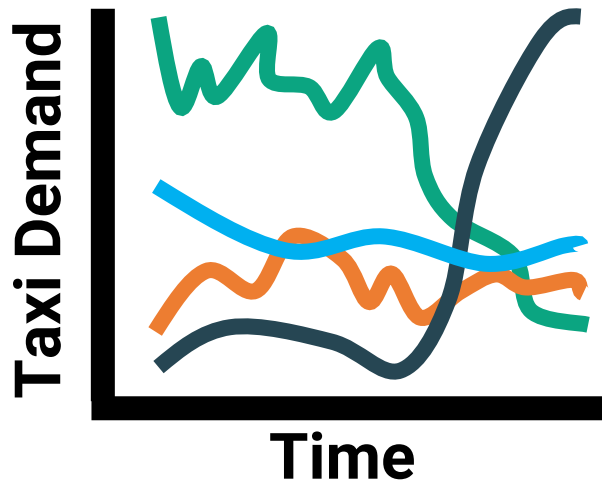
Time series data

Examples

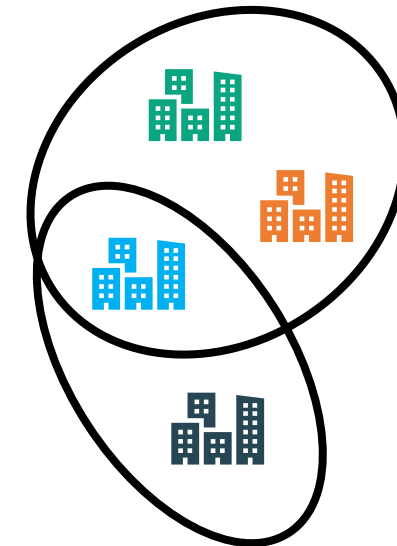
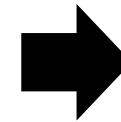
Time Series: Hypergraph Construction (cont.)

- Yi et al. (2020) converted **taxi demand** data into a hypergraph.
 - A node is a region; a hyperedge connects regions with similar taxi demand patterns (found by an algorithm).

 **Region1**
 **Region2**
 **Region3**
 **Region4**



**Taxi demand series
across regions**

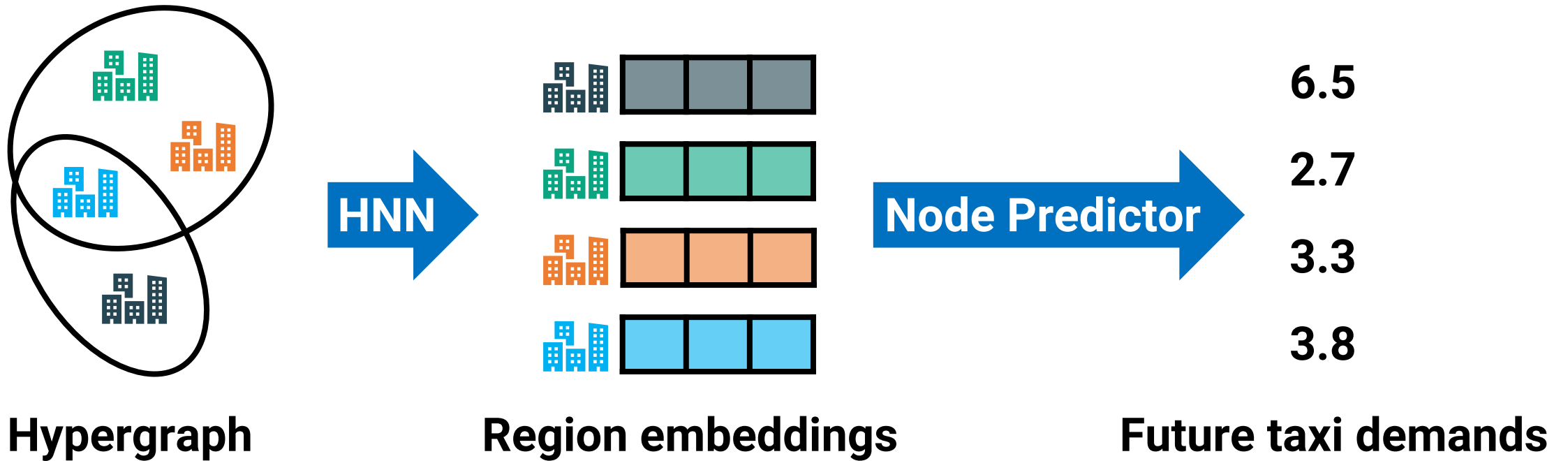


Hypergraph

Time Series: Task Formulation

- Yi et al. (2020) used an HNN to predict **future taxi demands** in each region.
 - The HNN encodes each city to predict its future taxi demand.

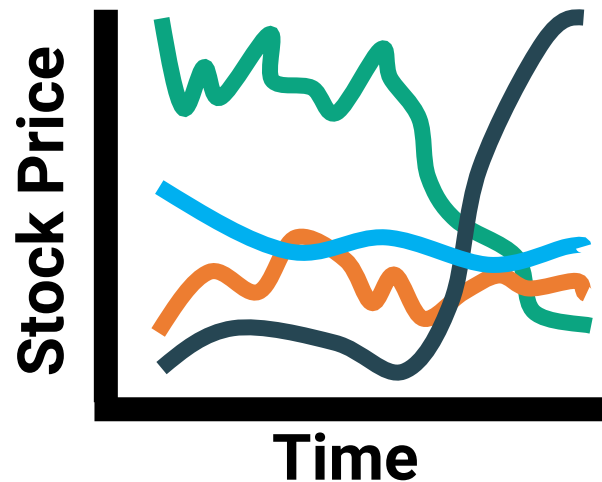
** Specifically, they further used an RNN to model the temporal aspects of the series.*



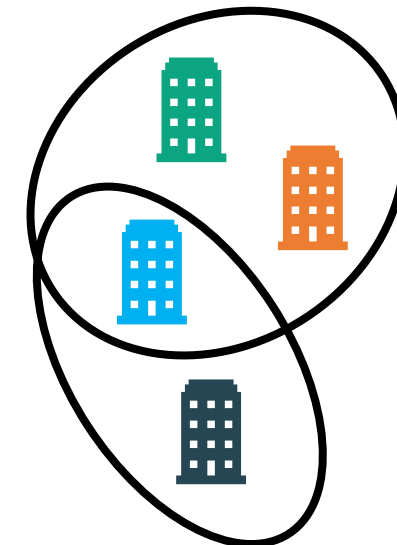
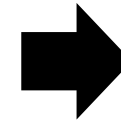
Time Series: Hypergraph Construction (cont.)

- Sawhney et al. (2021) converted **stock price** data into a hypergraph.
 - A node is a stock; a hyperedge connects stocks (i.e., companies) in the same industry or with 1st or 2nd order relations (heuristically determined).

 Stock1
 Stock2
 Stock3
 Stock4



**Stock price series across
different stocks**

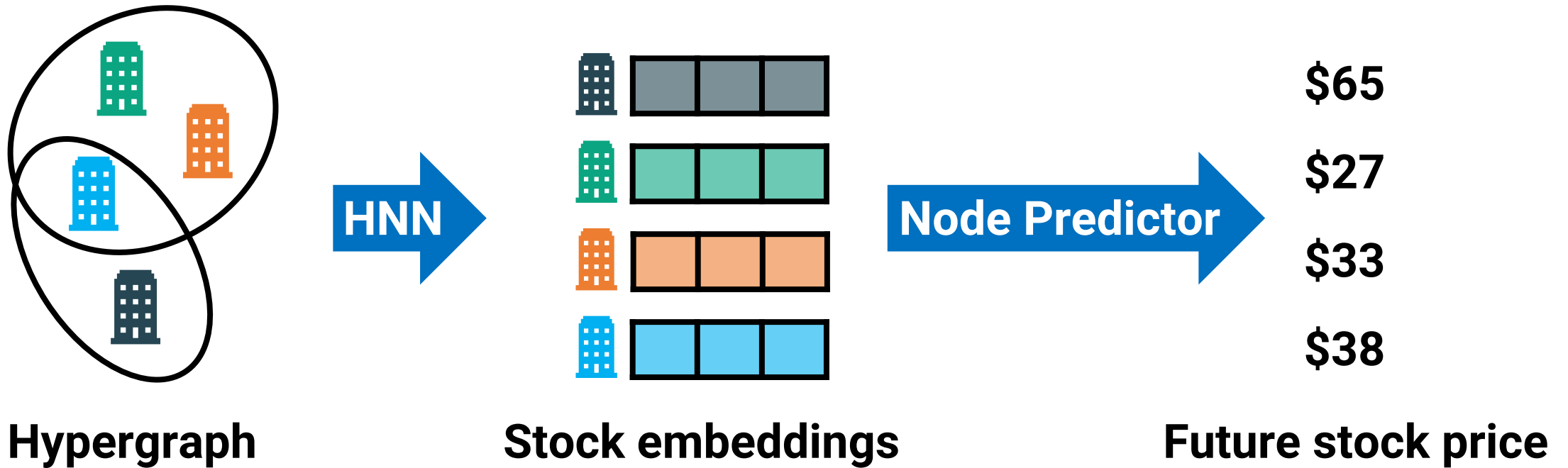


Hypergraph

Time Series: Task Formulation (cont.)

- Sawhney et al. (2021) used an HNN to **predict future stock price**.
 - The HNN encodes each stock to predict its future price.

** Specifically, they further used an RNN and attention to model the temporal aspects of the series.*



What are Notable Applications of HNNs? (cont.)

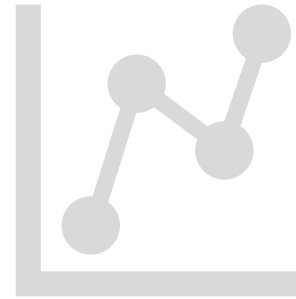
- In the following domains, we will cover how the issues have been addressed:



Recommender system



Bioinformatics & Medical science



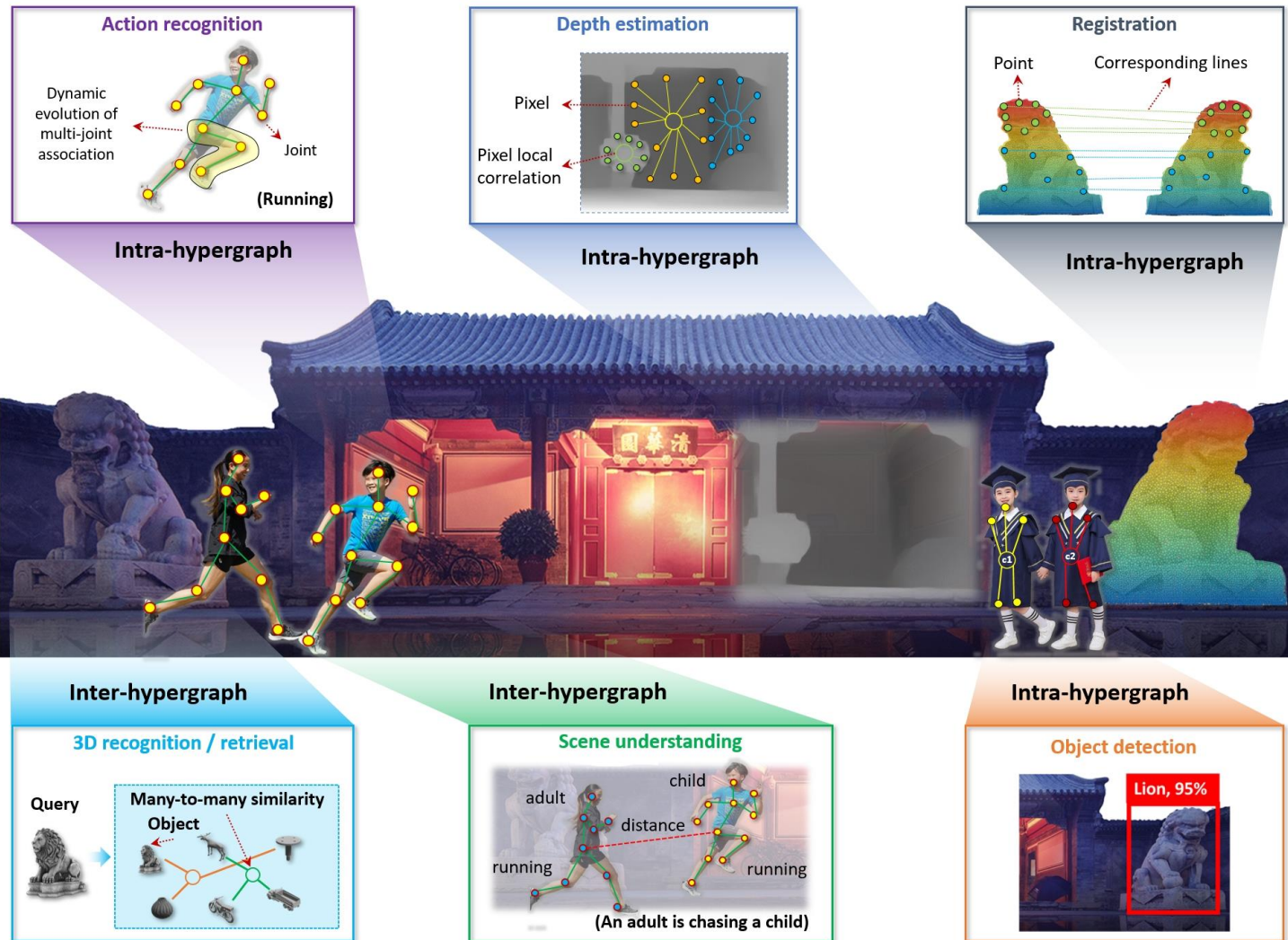
Time series analysis



Computer vision

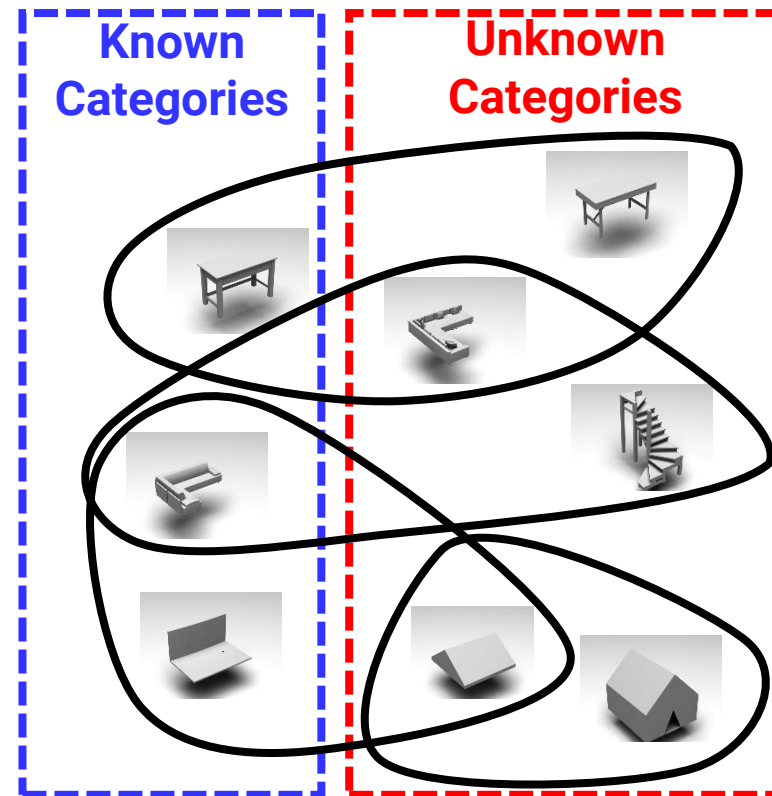
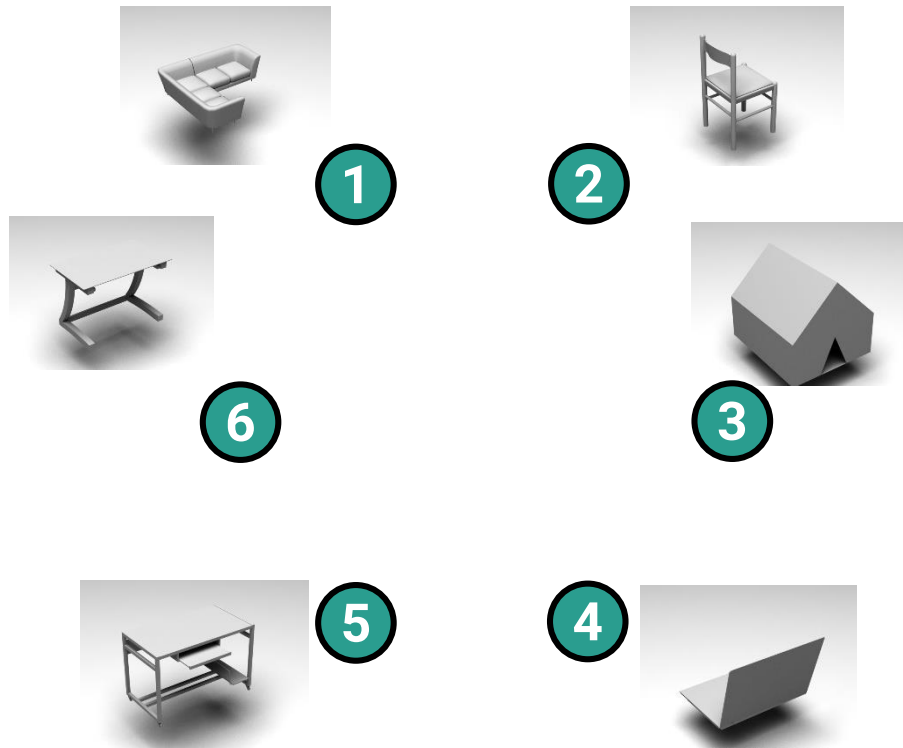
Hypergraph in Computer Vision

- Action Recognition
- Depth Estimation
- Registration
- 3D Recognition/Retrieval
- Scene Understanding
- Object Detection



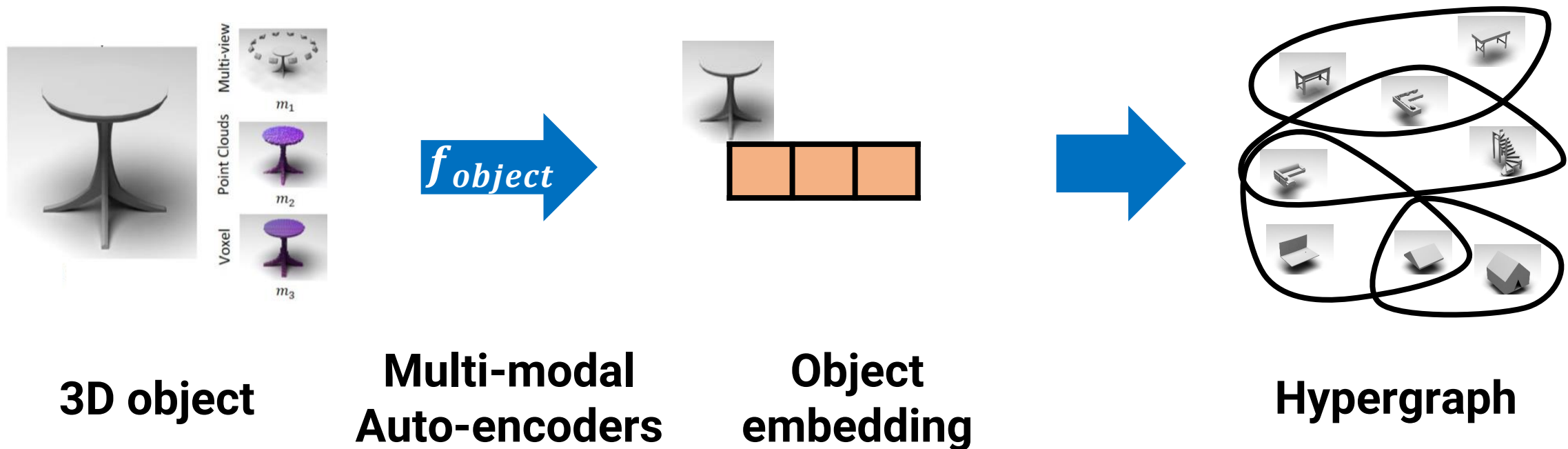
Vision: Hypergraph Construction (cont.)

- Feng et al. (2023) converted **3D objects** into a hypergraph
 - A node is a 3D object; a hyperedge connects the similar 3D objects



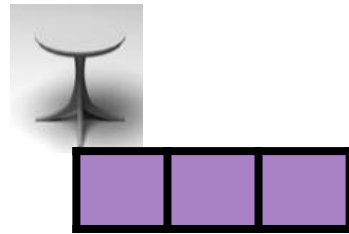
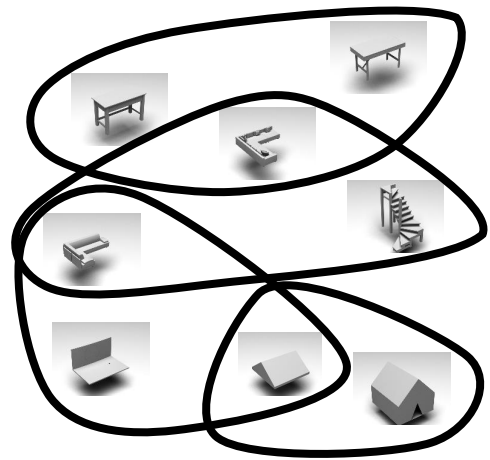
Vision: Task Formulation

- Feng et al. (2023) used an HNN for **3D object open set retrieval**
 - The hypergraph models relationships among multimodal data



Vision: Task Formulation

- Feng et al. (2023) used an HNN for **3D object open set retrieval**
 - The HNN exploits the inherent connections in multimodal data



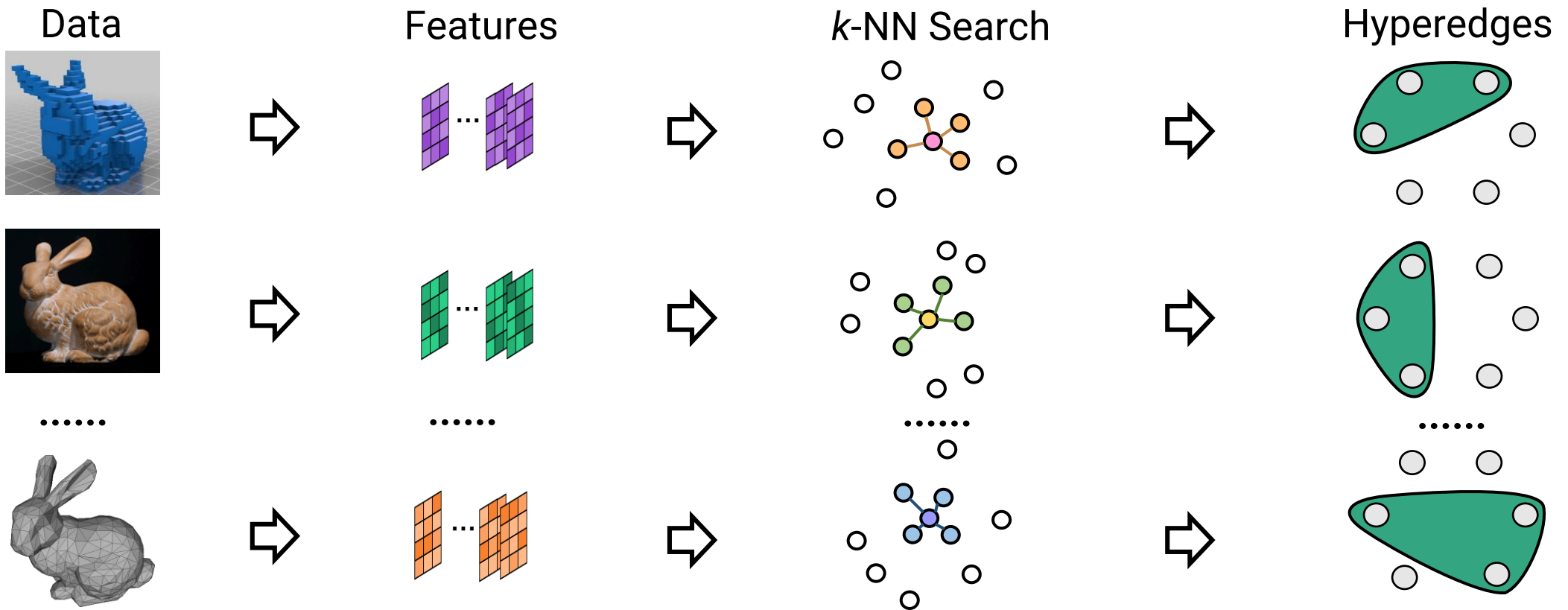
Aligned embedding



Similar objects

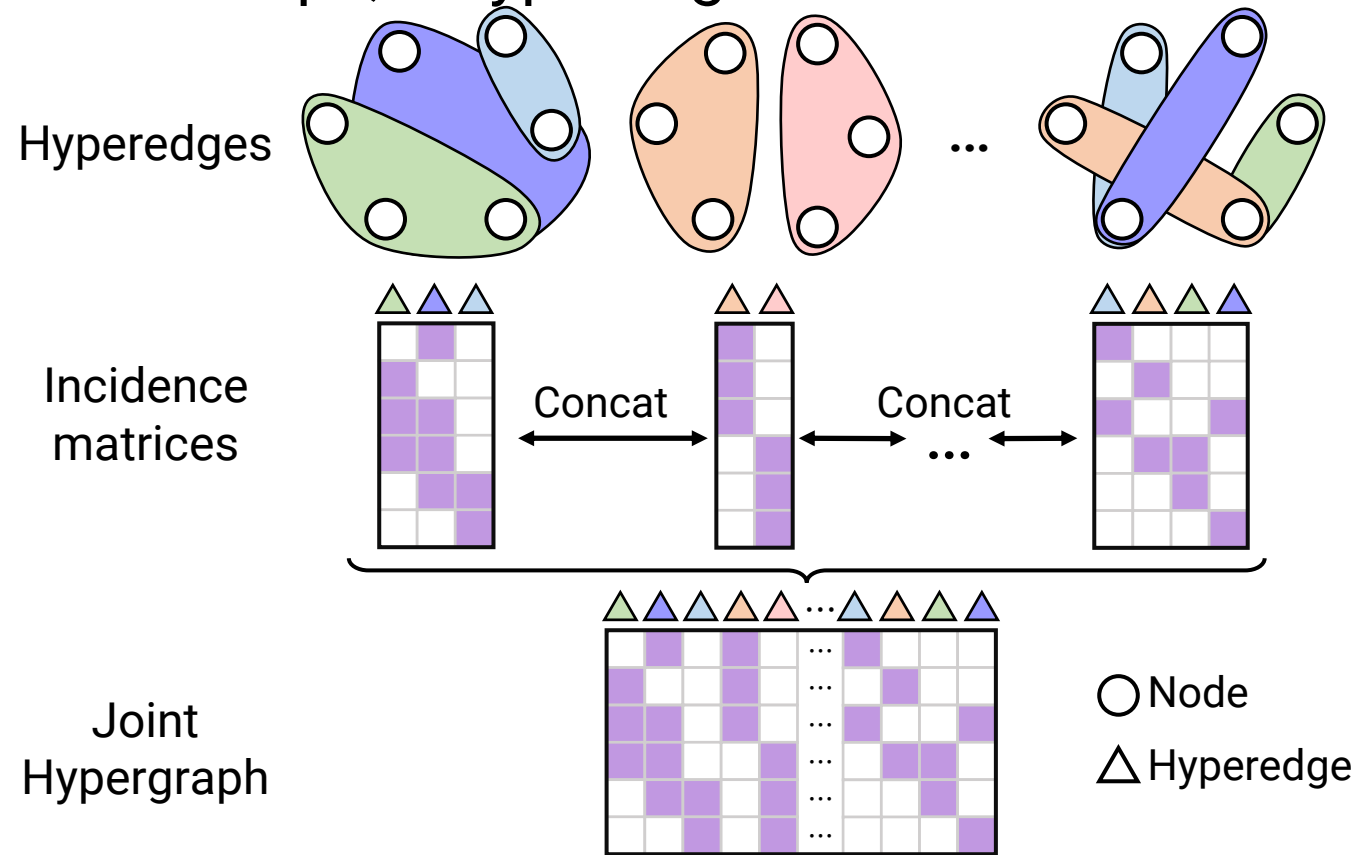
Vision: Hypergraph Construction (cont.)

- Bai et al. (2021) joined multi-scale hyperedges to the hypergraph
 - A node is a 3D shape; a hyperedge connects the similar shapes



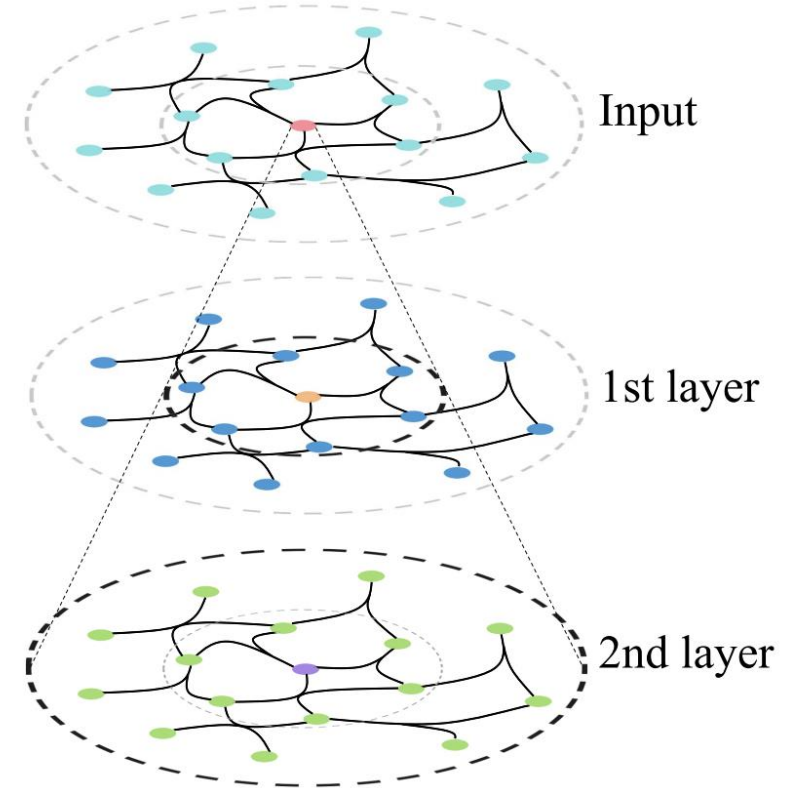
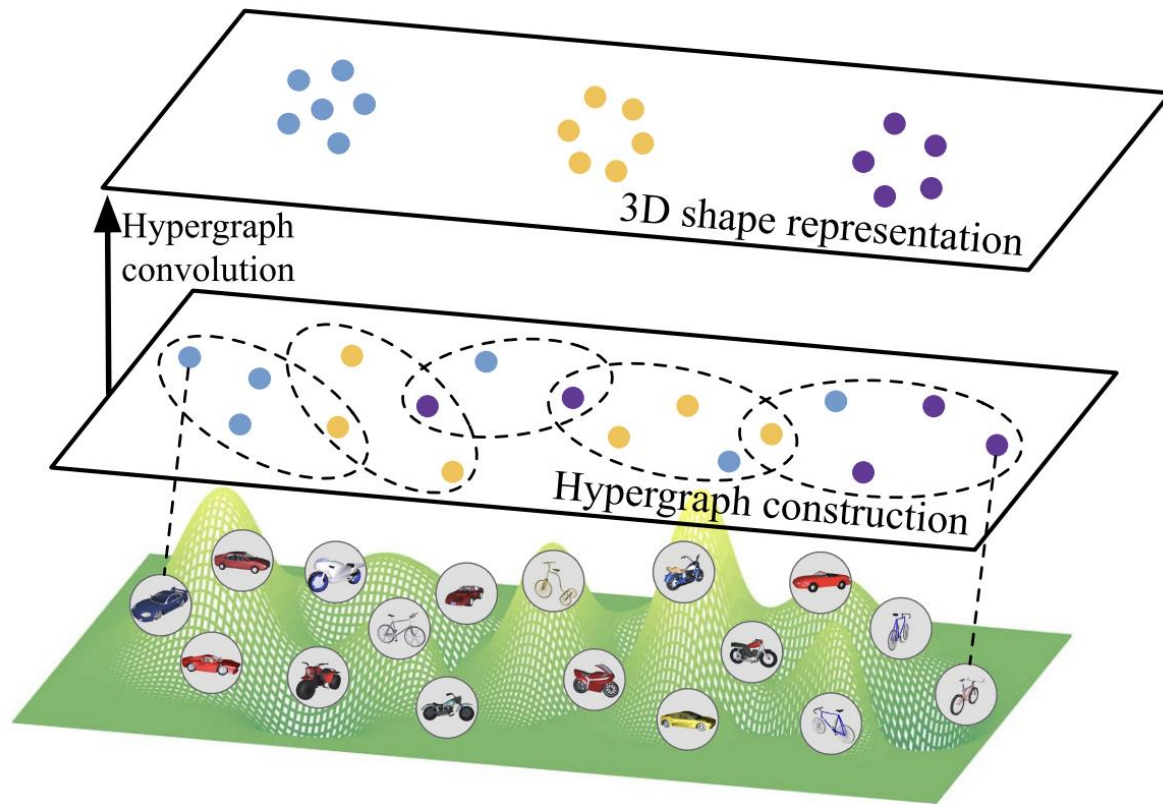
Vision: Hypergraph Construction (cont.)

- Bai et al. (2021) joined multi-scale hyperedges to the hypergraph
 - A node is a 3D shape; a hyperedge connects the similar shapes



Vision: Task Formulation

- Bai et al. (2021) used an HNN for **3D shape retrieval and recognition**.
 - The HNN captures relationships among **3D shapes in multi-scale**



Vision: Hypergraph Construction

- **Raw data** typically are images in pixels or voxels.



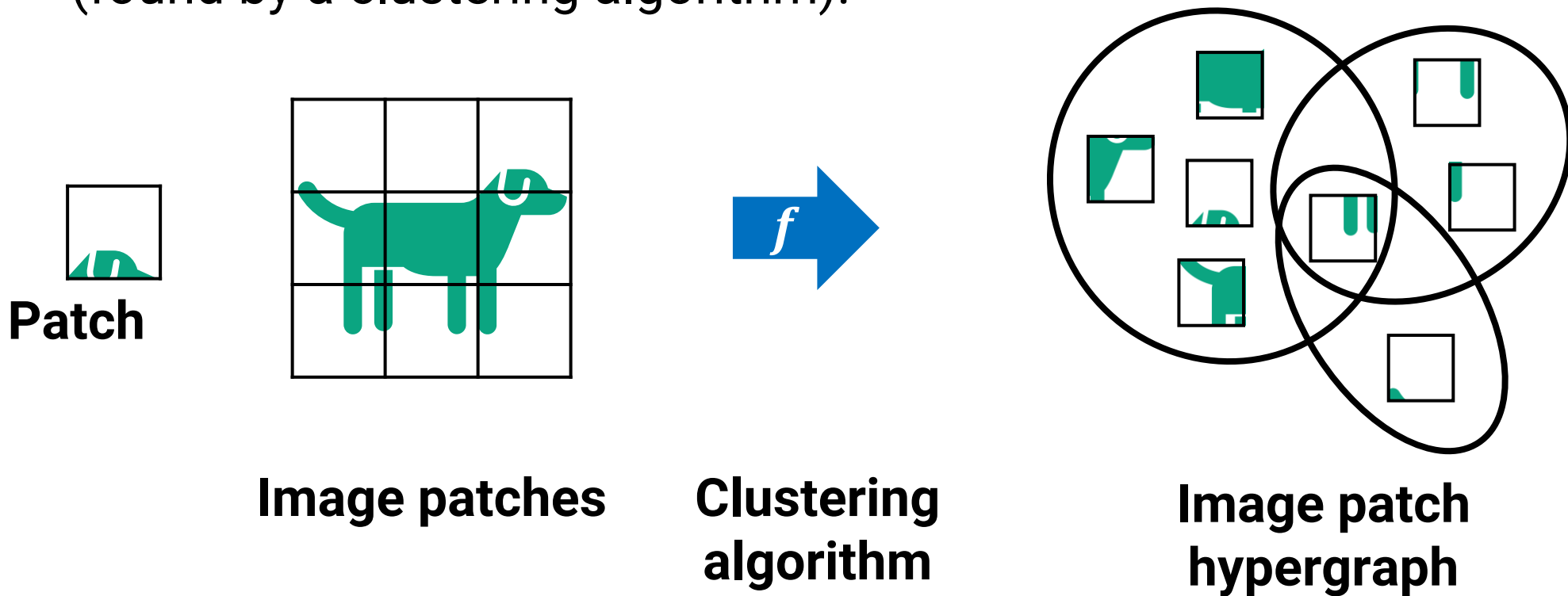
Pixel images



Voxel images

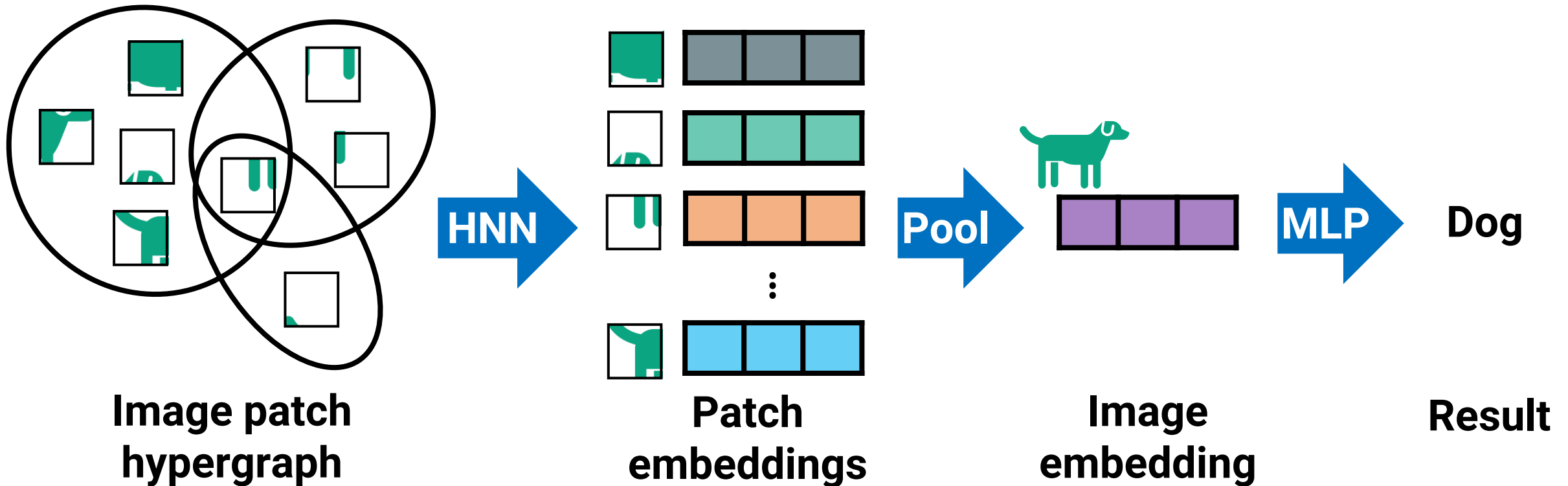
Vision: Hypergraph Construction (cont.)

- Han et al. (2023) converted **pixel image** data into a hypergraph.
 - A node is a image patch; a hyperedge connects similar patches (found by a clustering algorithm).



Vision: Task Formulation

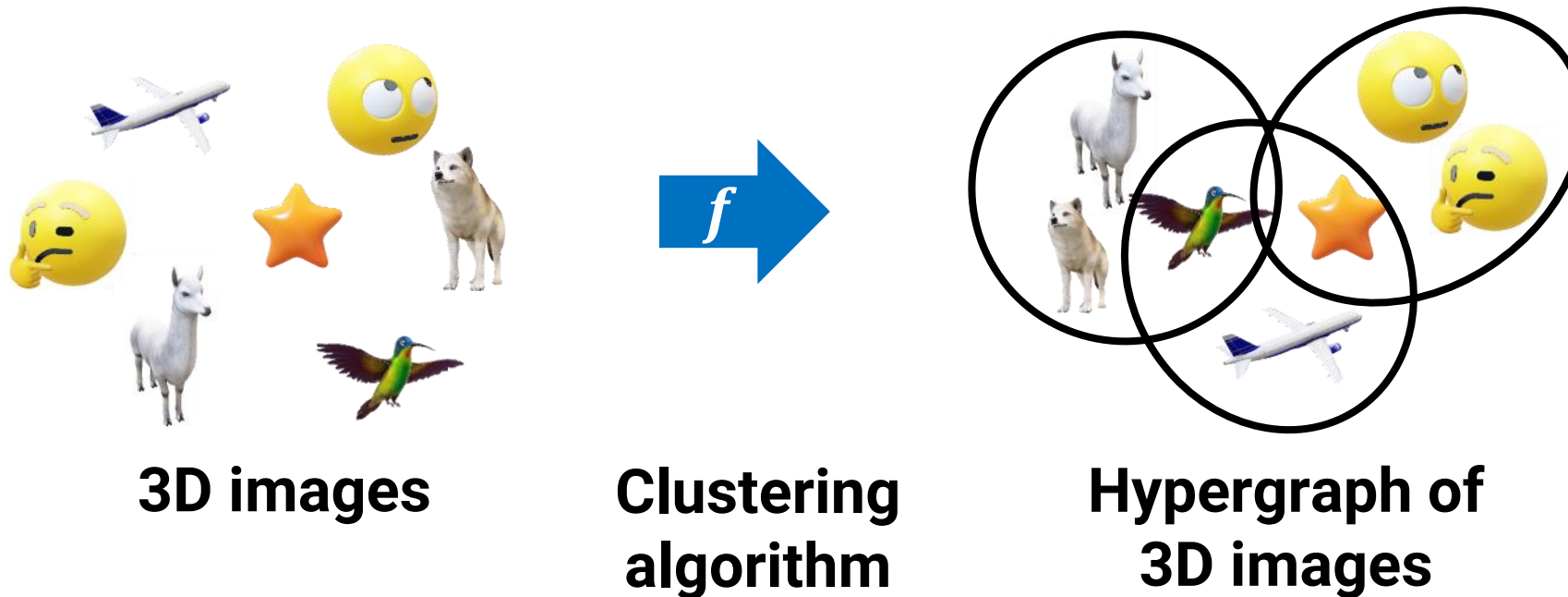
- Yi et al. (2020) used an HNN to **classify images** or **detect objects**.
 - The HNN encodes each city to predict its future taxi demand.



Vision: Hypergraph Construction (cont.)

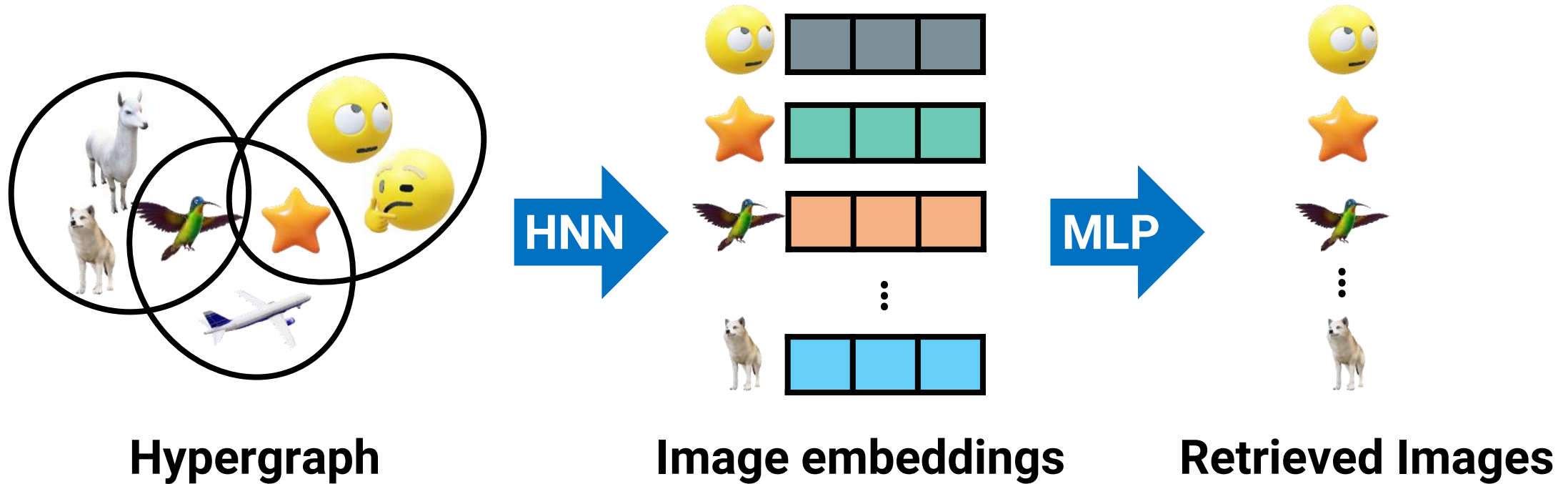
- Bai et al. (2021) converted **3D images** into a hypergraph
 - A node is a 3D image; a hyperedge connects the similar 3D images (found by a clustering algorithm).

** Specifically, they used multi-view representation of 3D images.*



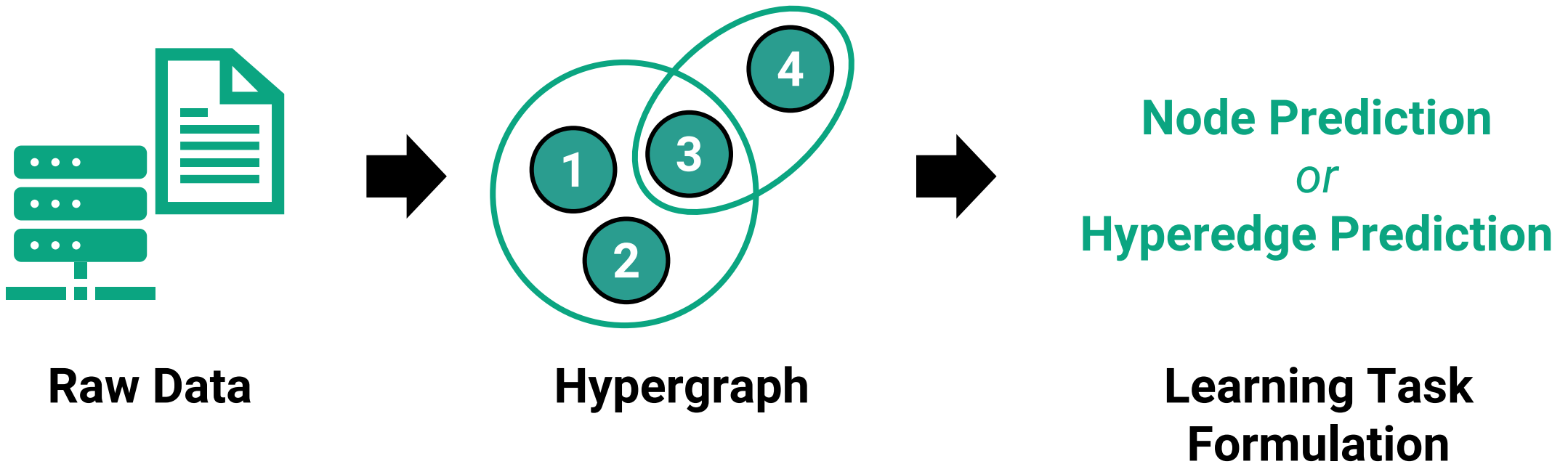
Vision: Task Formulation (cont.)

- Bai et al. (2021) used an HNN to retrieve and classify 3D images.
 - The HNN encodes each image to retrieve and classify it.



Part 5 Summary

- Two key issues include:
 - Q1) How to **express the raw data** as hypergraphs?
 - Q2) How to **formulate the learning task** for HNN?



Part 5 Summary (cont.)

- We covered how the issues have been addressed in:



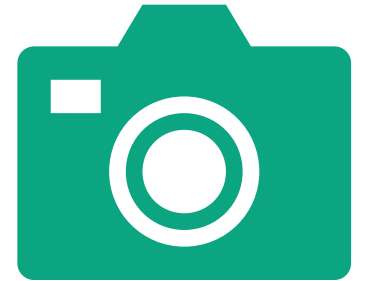
Recommender system



Bioinformatics & Medical science



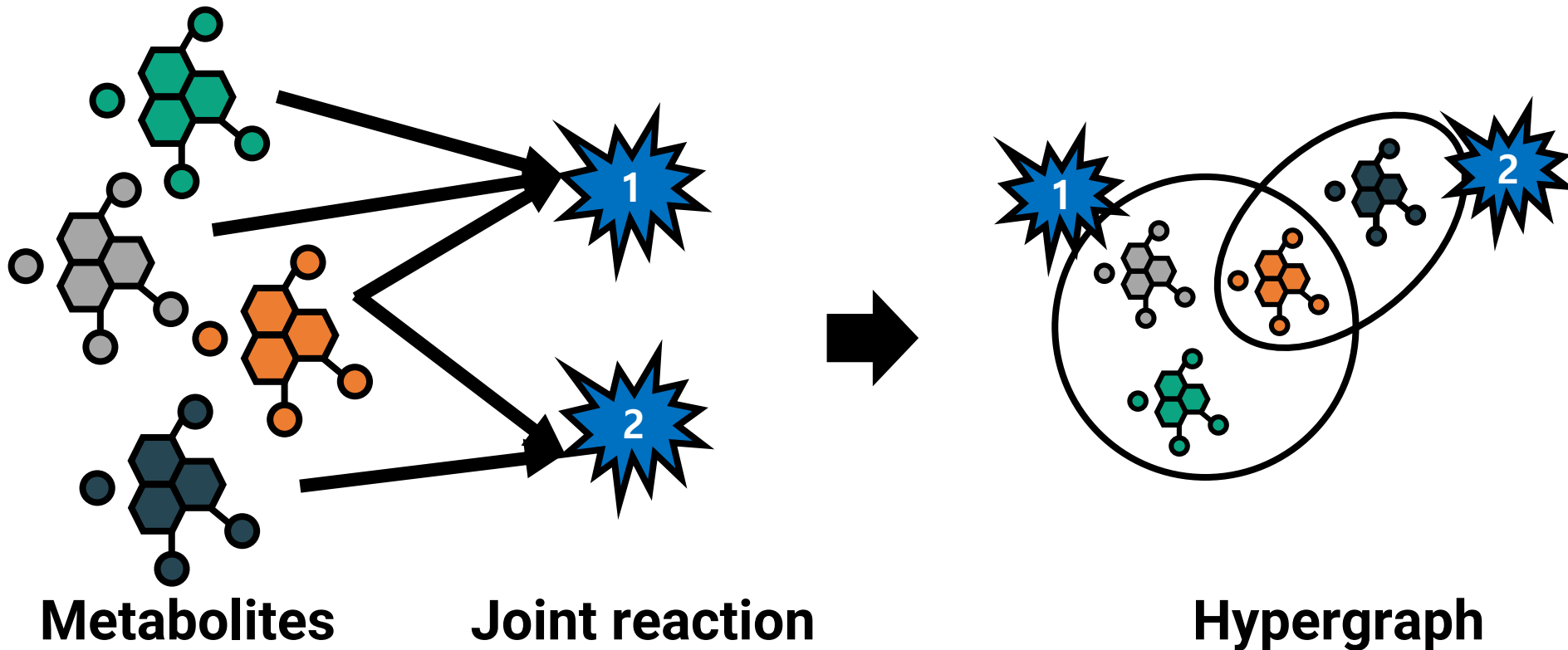
Time series analysis



Computer vision

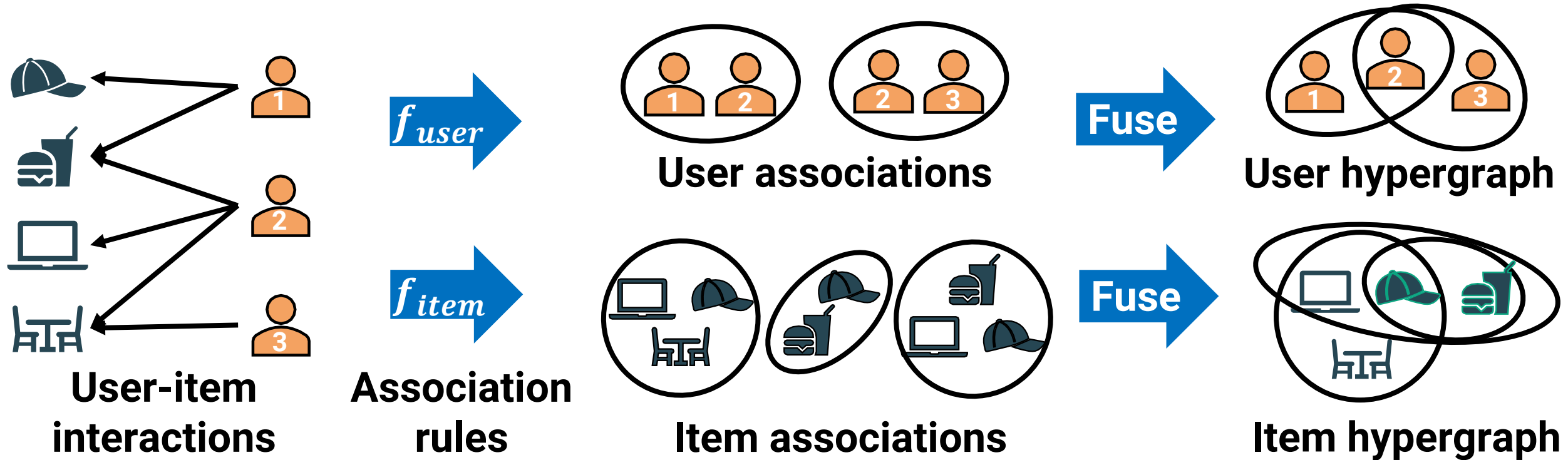
Part 5 Summary (cont.)

- After determining the nodes, hyperedges were constructed often by
 - 1) **Domain knowledge** or 2) similarity search



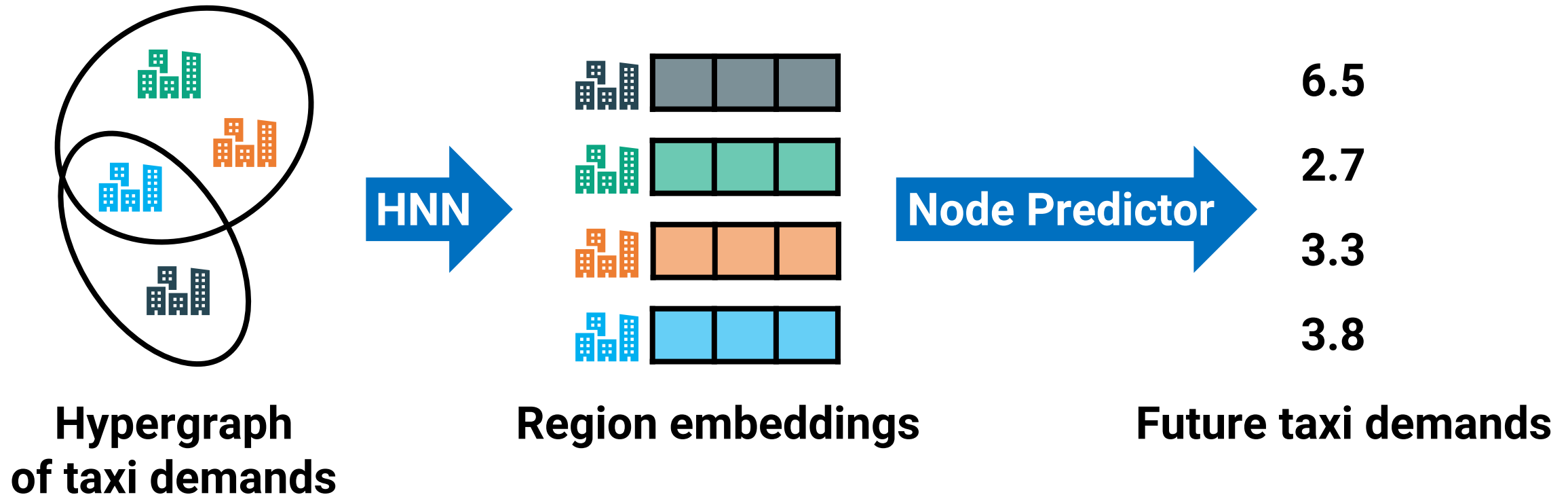
Part 5 Summary (cont.)

- After determining the nodes, hyperedges were constructed often by
 - 1) Domain knowledge or 2) **similarity search**



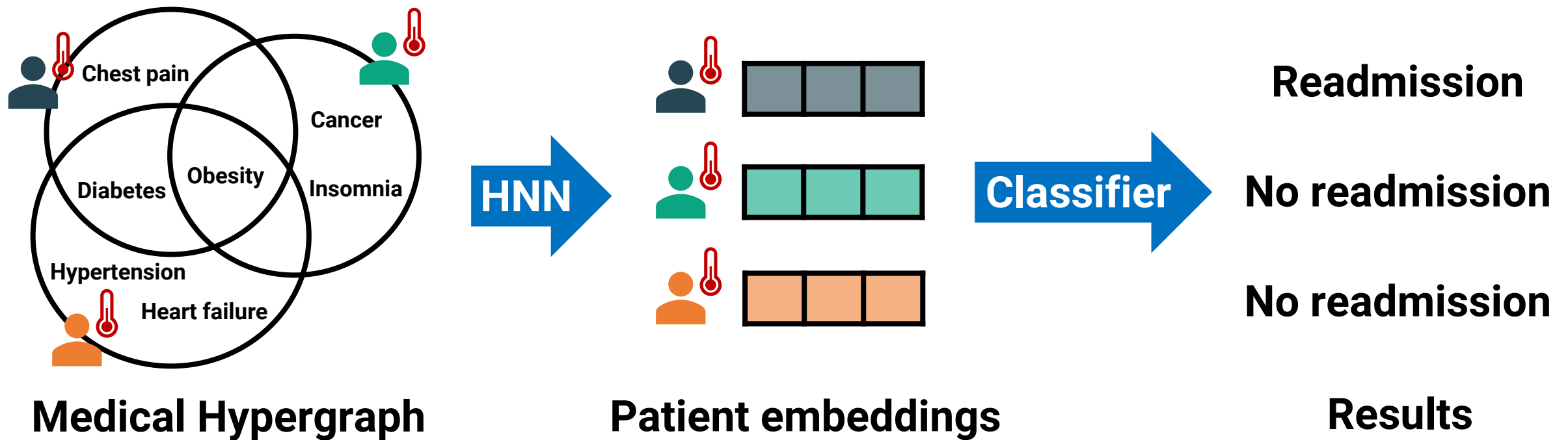
Part 5 Summary (cont.)

- After constructing the hypergraphs, HNNs often were used to predict
 - 1) **nodes**, 2) hyperedges, or 3) hypergraphs



Part 5 Summary (cont.)

- After constructing the hypergraphs, HNNs often were used to predict
 - 1) nodes, 2) **hyperedges**, or 3) hypergraphs



Part 5 Summary (cont.)

- After constructing the hypergraphs, HNNs often were used to predict
 - 1) nodes, 2) hyperedges, or 3) **hypergraphs**

