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shaping futures

# CS343 Graph Data Science

## Graph Algorithms to Enhance Machine Learning

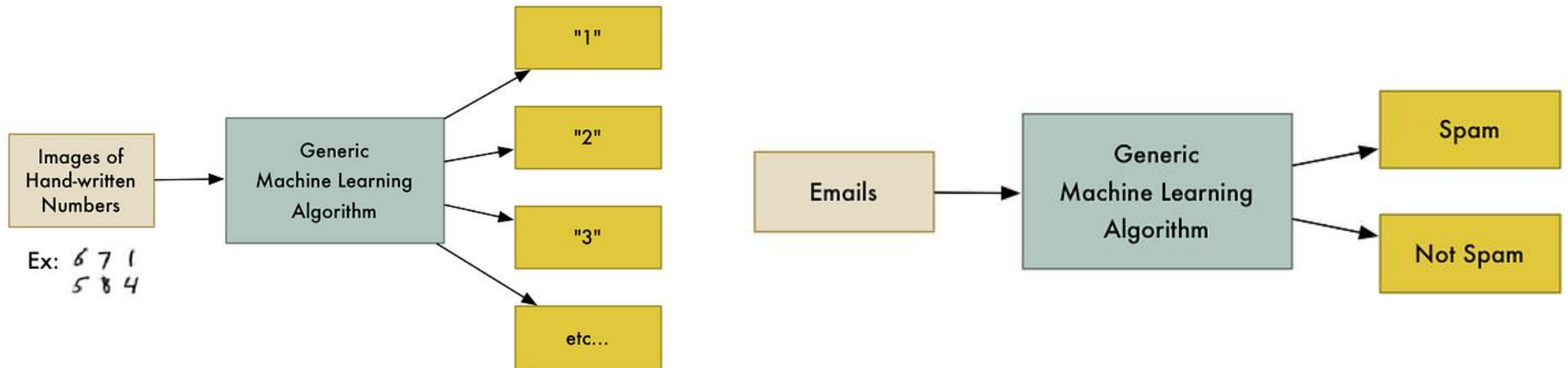
Chapter #8, Graph Algorithms, Mark

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# Machine Learning

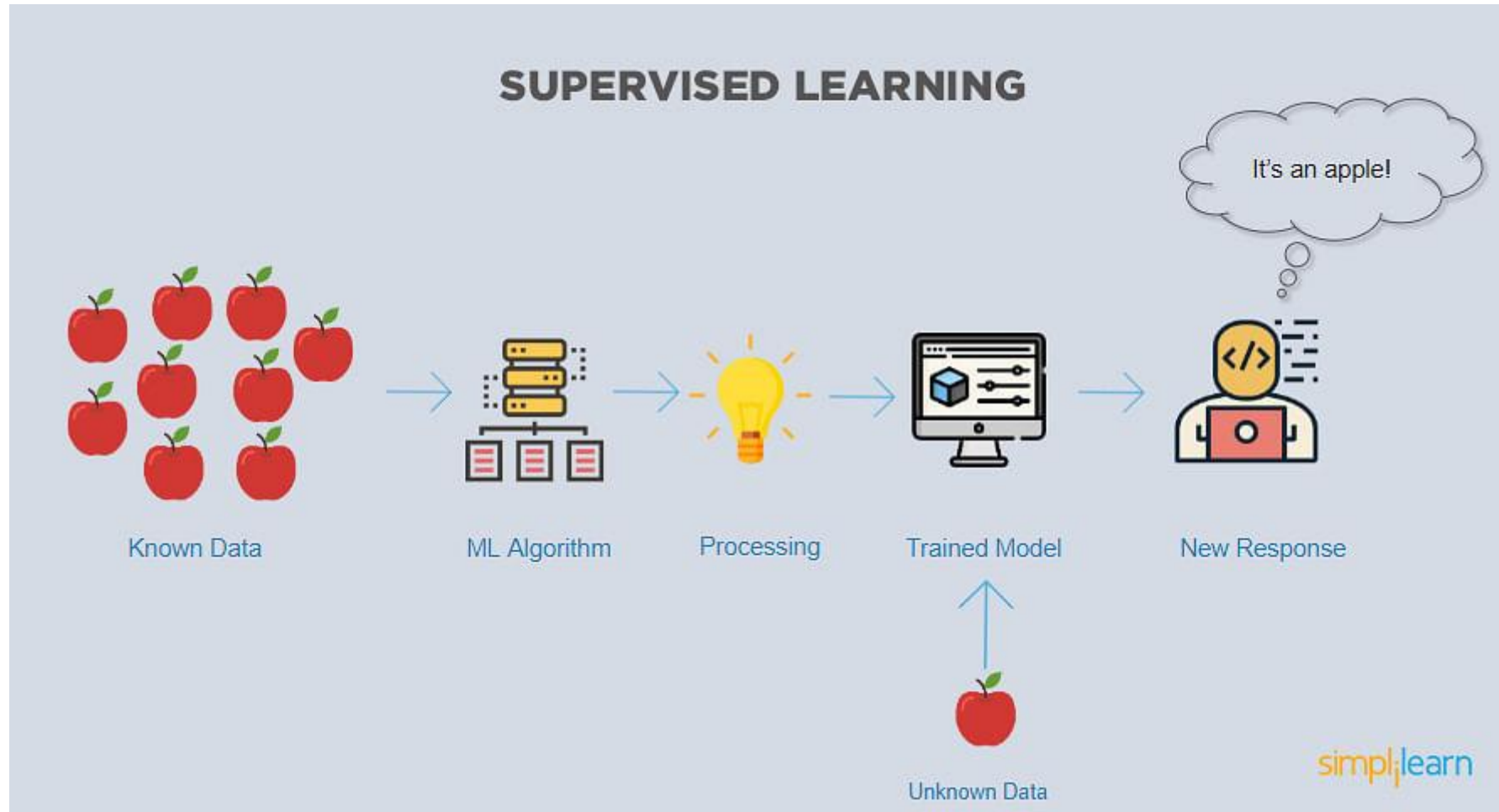
- Machine learning is a branch of artificial intelligence that enables computers to **learn from data** without being explicitly programmed
- Machine learning algorithms use **mathematical methods** to find patterns and make predictions or classifications based on data



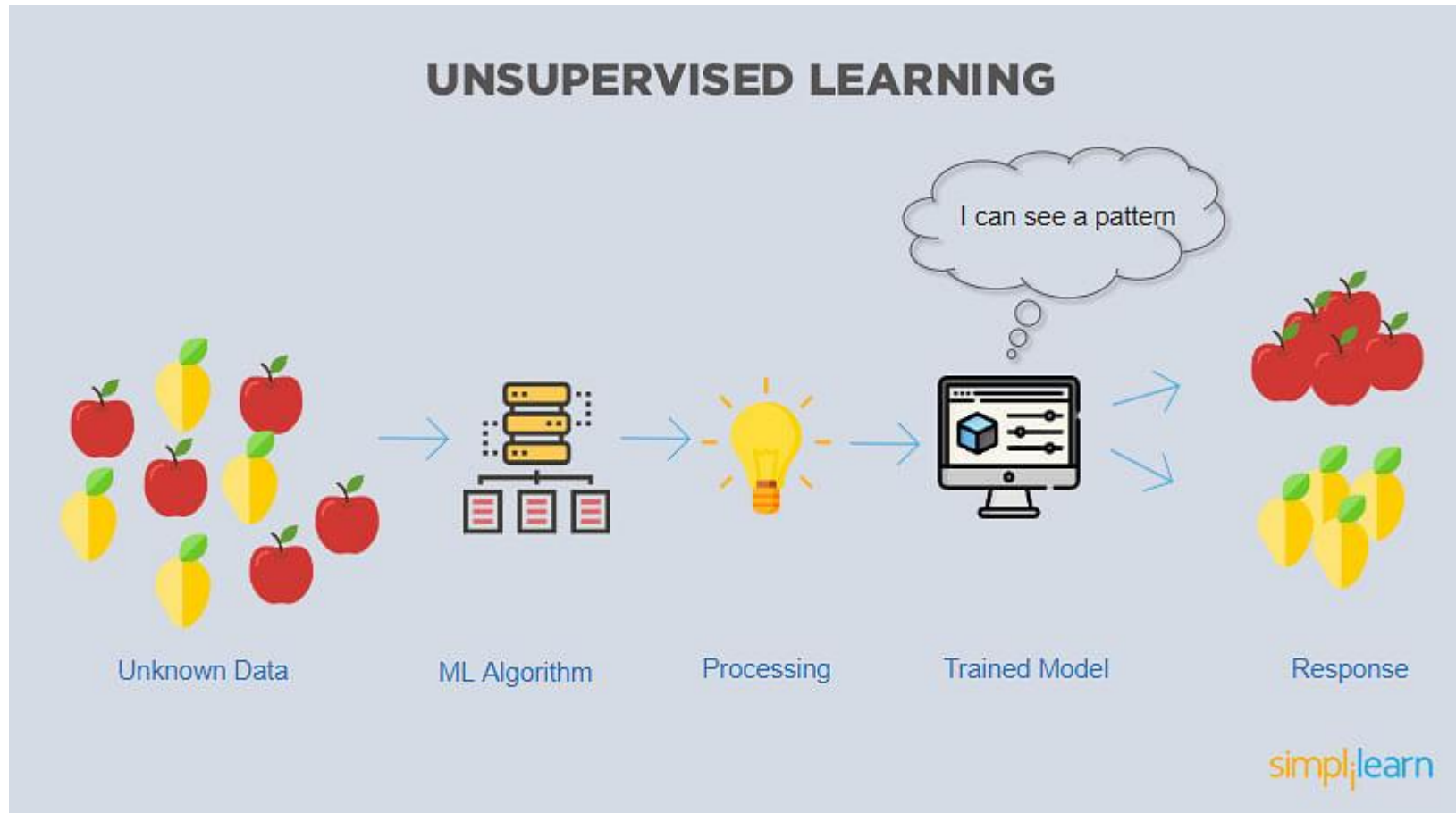
# Types of Machine Learning

- Supervised
- Unsupervised
- Reinforcement
- Self-supervised
- Semi-supervised

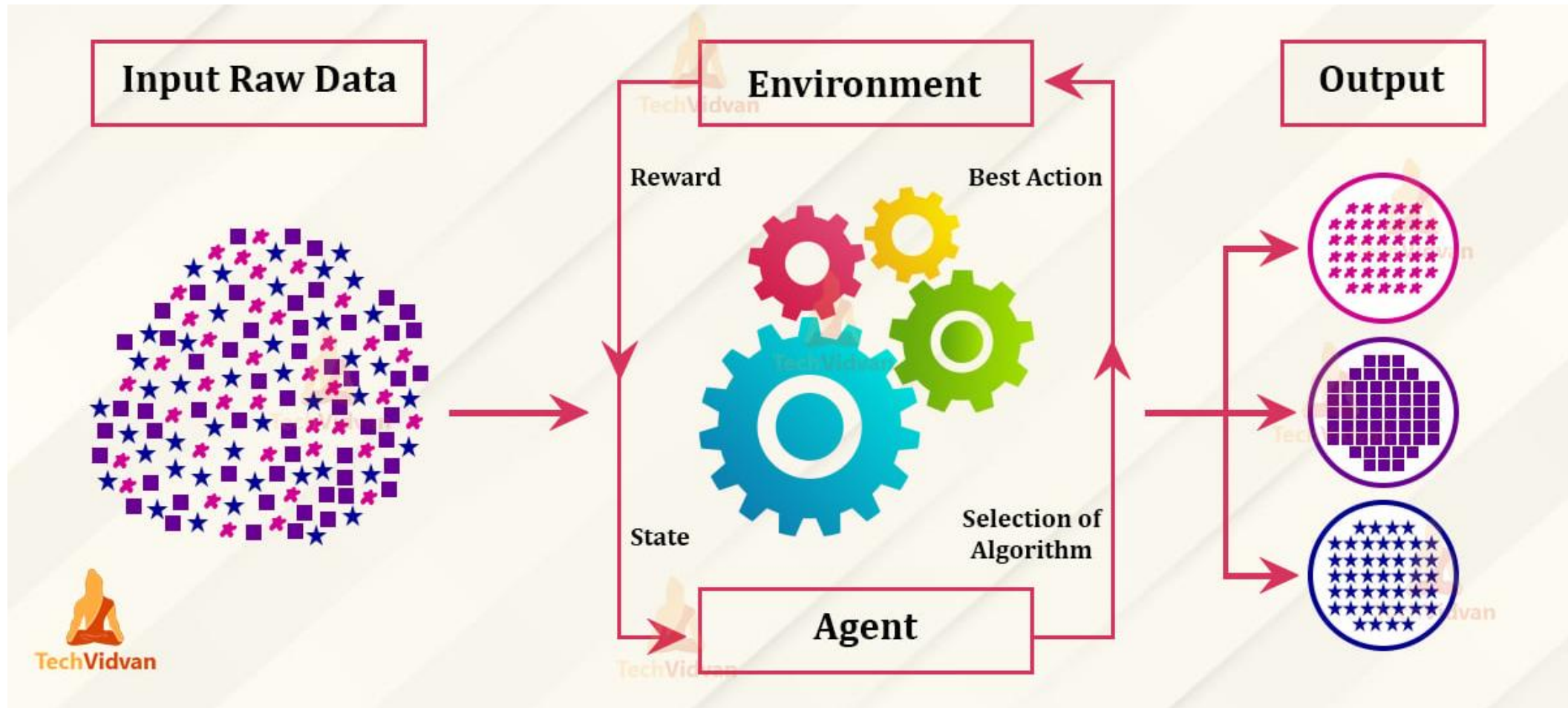
# Supervised Learning



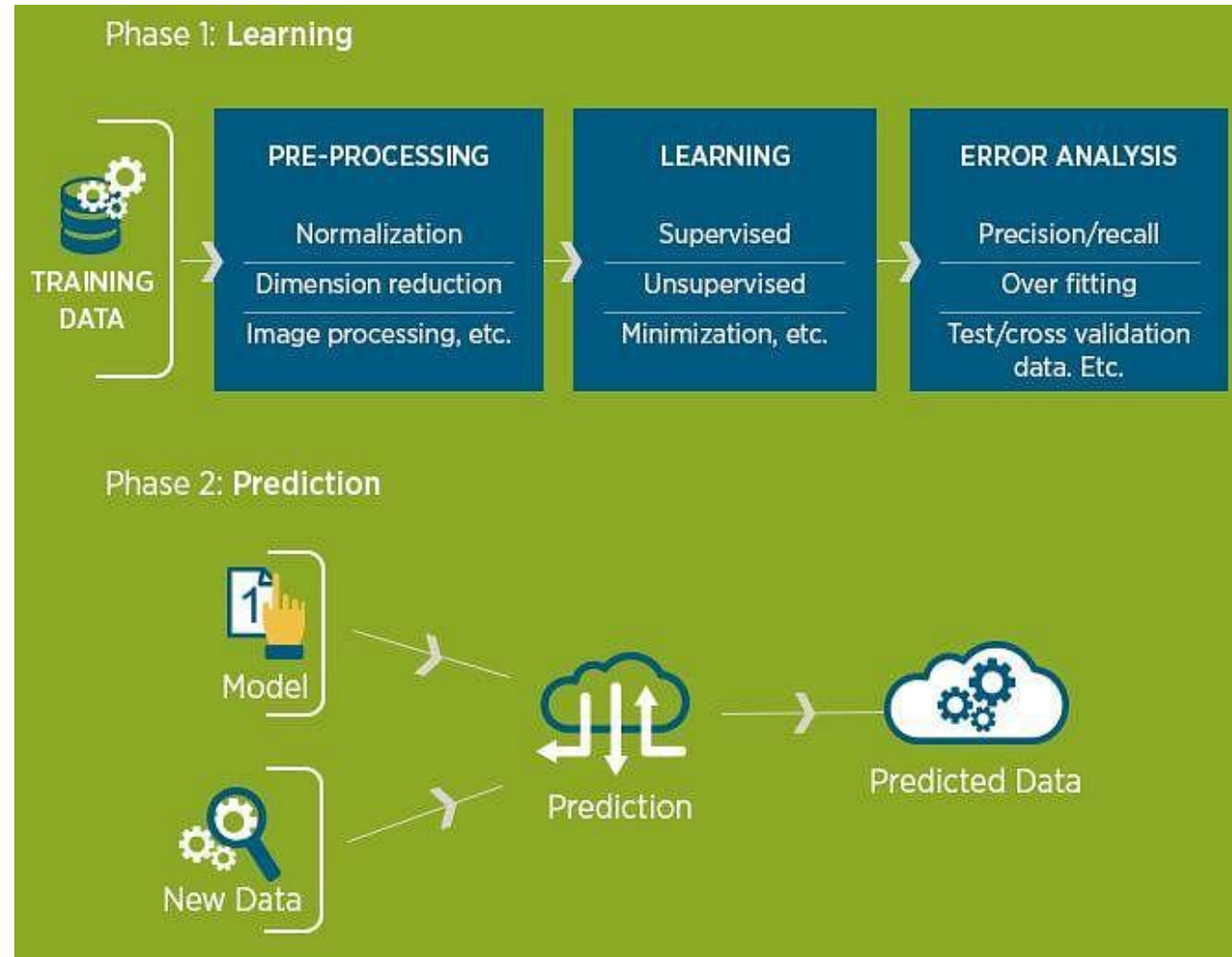
# Unsupervised Learning



# Reinforcement Learning



# Machine Learning Pipeline





# Feature Extraction

- Transforming raw data into a format that is suitable for analysis and model building.
- Simplifies complex data by extracting meaningful patterns.
- Reduces dimensionality, making the data more manageable and improving computational efficiency.
- Enhances model performance by focusing on important characteristics of the data.
- **Feature selection** is the process of determining the subset of extracted features that are most important or influential to a target goal.
  - Selects the most relevant features based on criteria such as correlation, significance, or importance scores.



# Connected Features

- Features extracted from the structure of the data.
- Graph-local queries based on parts of the graph surrounding a node
- Graph-global queries that use graph algorithms to identify predictive elements
- Graph for feature selection?
  - What if map features to nodes in a graph, create relationships based on similar features, and then compute the centrality of features.

# Node Embedding

- Maps each node in a graph to a low-dimensional vector while preserving structural relationships.

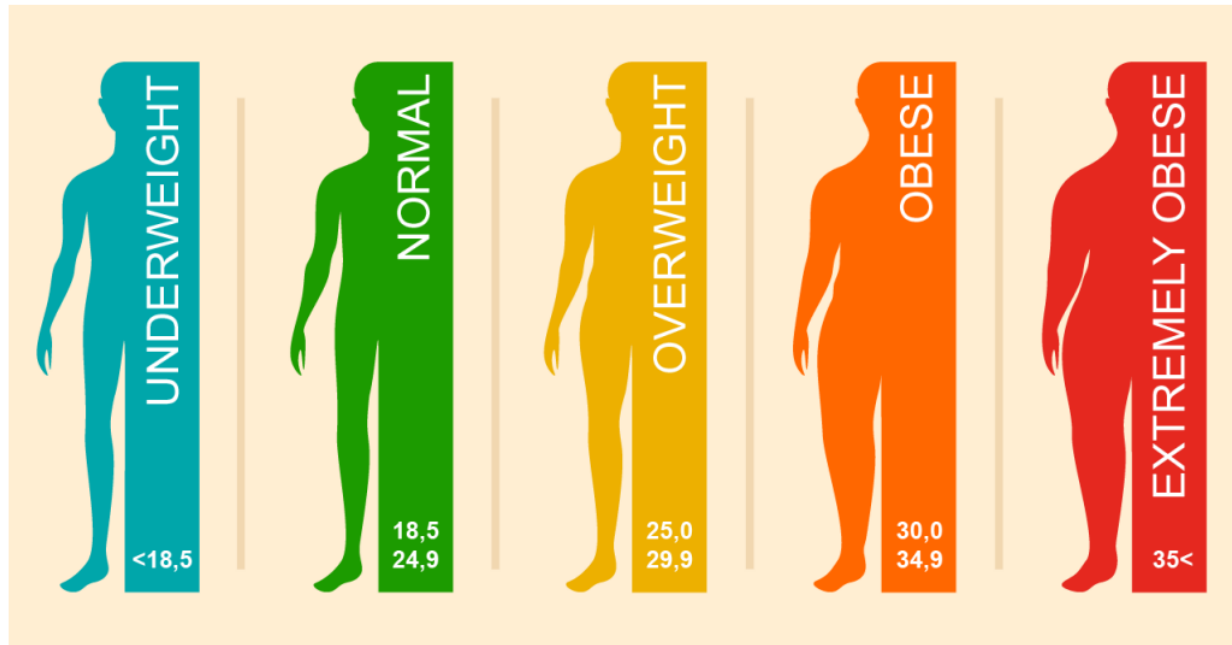
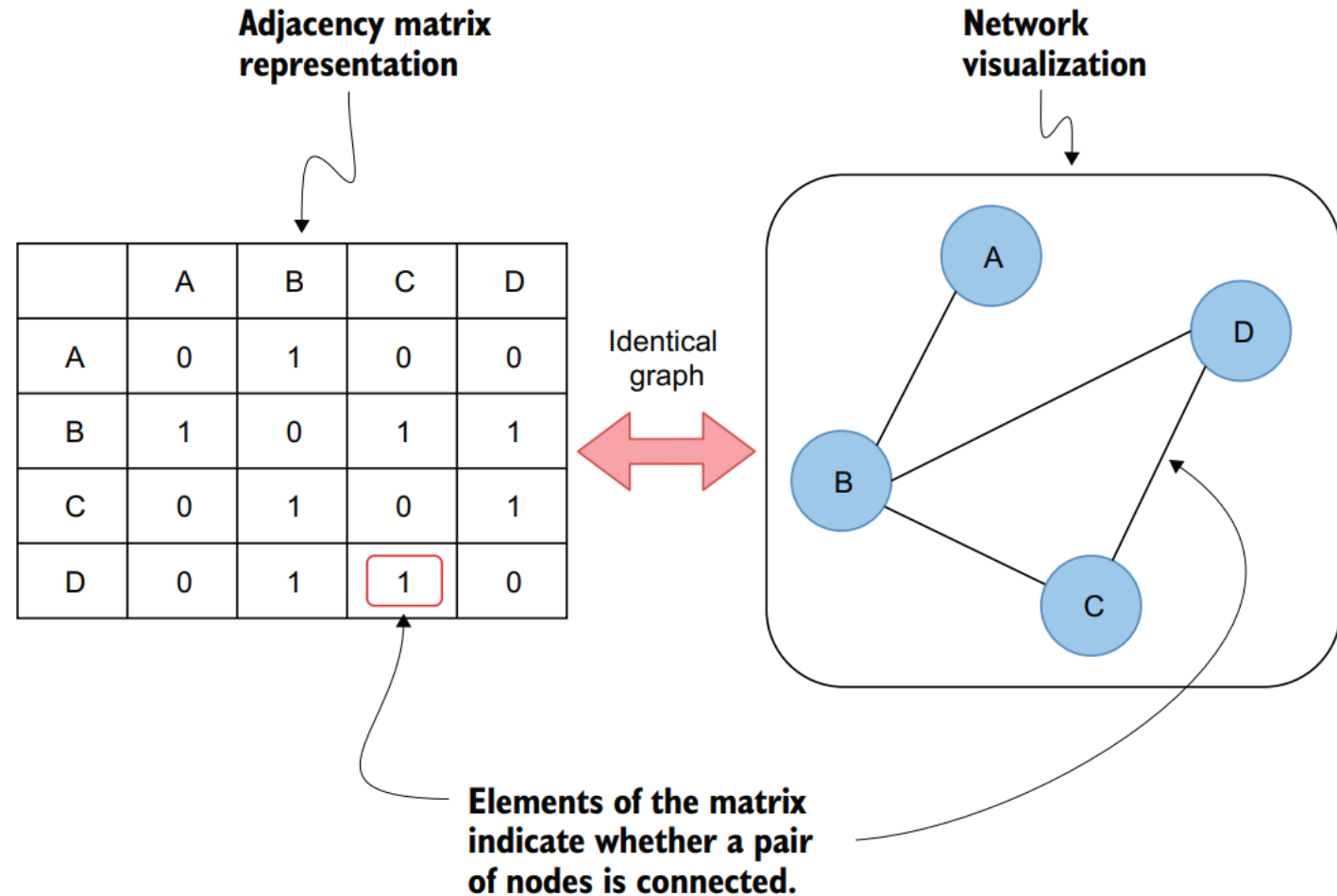


Figure 9.1 Body mass index chart

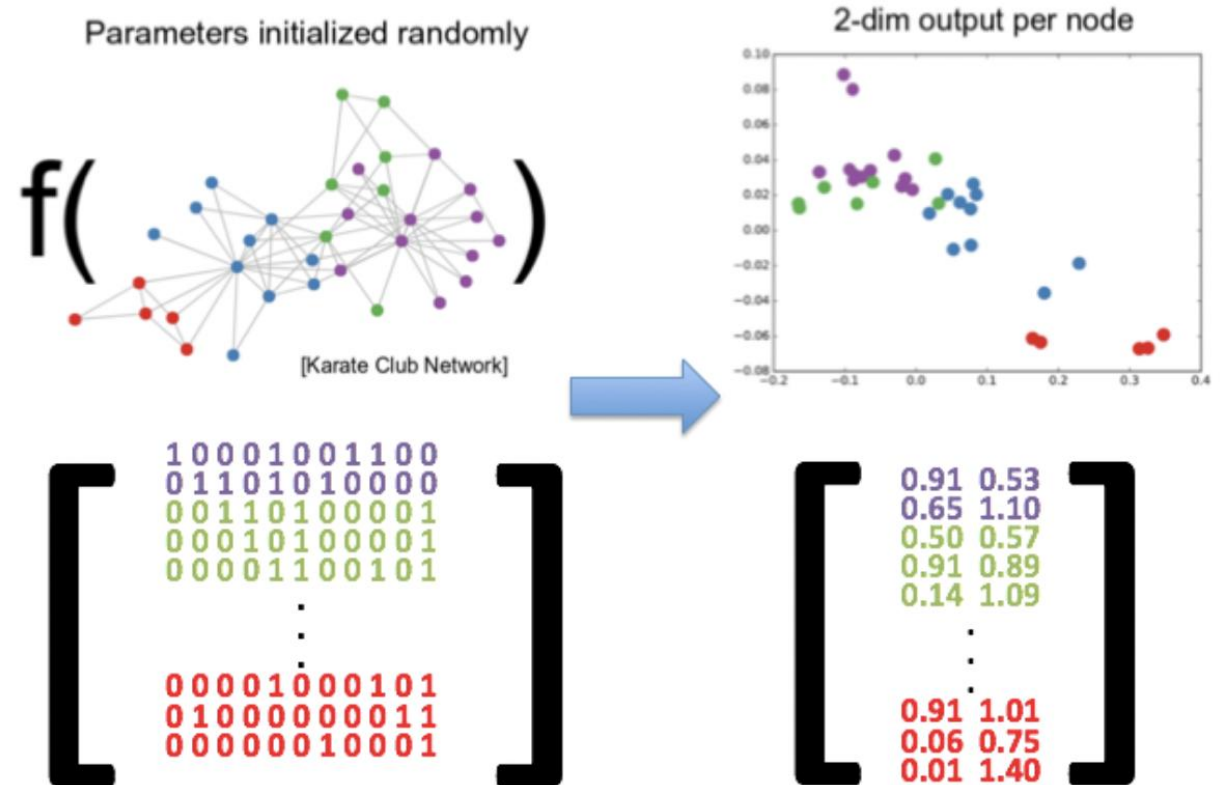
# Adjacency Matrix

- High-dimensional network representation



# Node Embedding

- Lower-dimensional vectors can be used in a downstream machine learning workflow, or they can be used to infer a similarity network using the nearest neighbor graph algorithm.



# Homophily vs Structure Holes

## FastRP

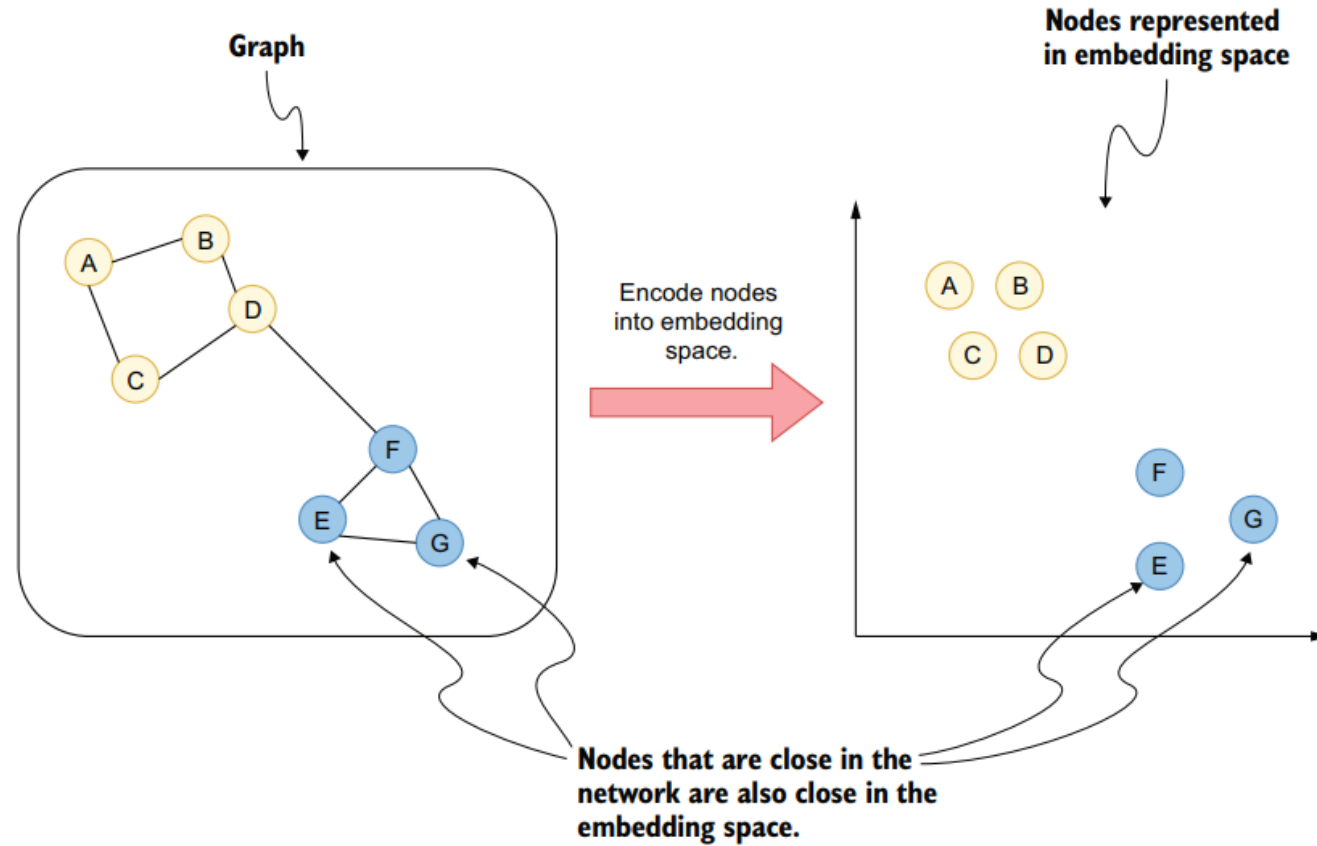


Figure 9.4 Homophily approach to node embedding

# Homophily vs Structure Holes

node2vec

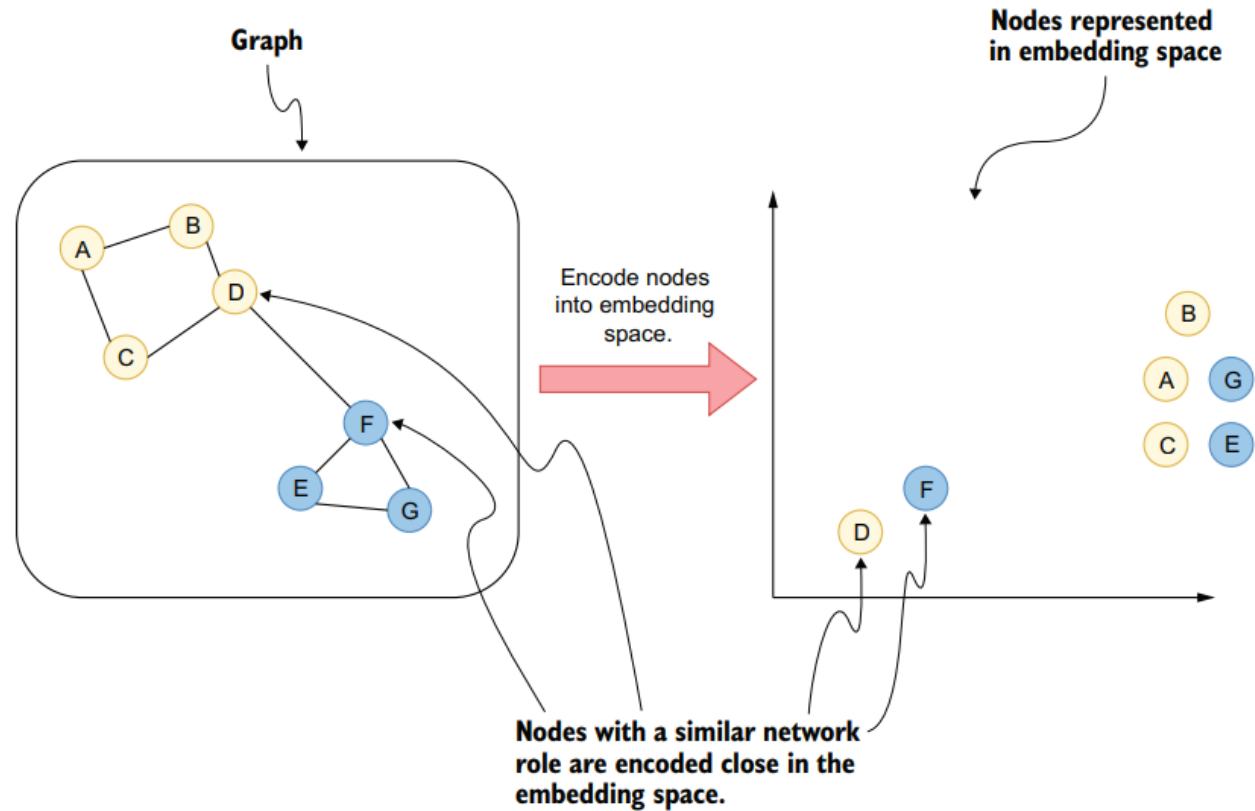


Figure 9.5 Structural roles approach to node embedding

# Transductive vs Inductive

- Transductive:
  - cannot calculate embeddings for nodes not seen during the initial embedding calculation
  - cannot simply retrieve the embeddings for the new unseen nodes
  - New node: must calculate the embeddings for the whole graph
  - must also retrain the classification model.
- Inductive
  - can calculate embeddings for unseen nodes during the initial computation
  - can calculate the embedding for the new node without recalculating embeddings for the whole graph
  - don't need to retrain the classification model for every new node.



# Node Classification Example

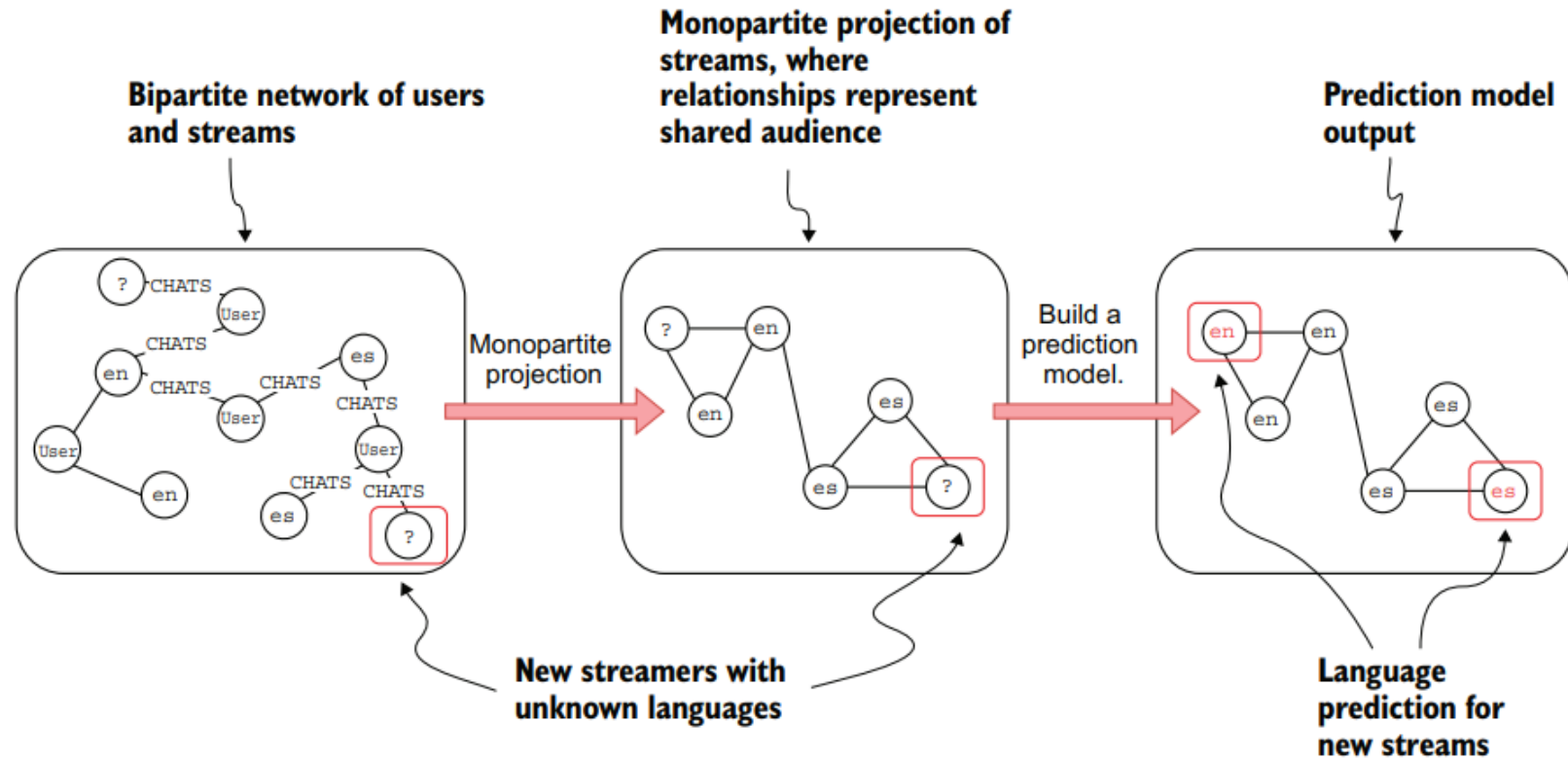


Figure 9.6 The process of predicting the language for new streams