

# How to Teach Unsupervised Machine Learning with Analogies

## A Study on the Effectiveness of Analogies in Teaching Unsupervised Machine Learning

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### ① Introduction

- The importance and use of ML is growing across fields [1]
- There is a growing need for ML teaching methods [2]
- Unsupervised learning is a relevant field, and some concepts struggle with interpretability [3]
- Analogies often provide effective teaching, since they are good for making mental model [4]

### Research Question

*How does the use of analogies in teaching unsupervised learning affect the knowledge gain measured by the answering of theoretical questions?*

### ② Background

- An analogy is “a comparison between things that have similar features” [5]
- They have a mapping between the target concept (what is explained) and the source concept (what the target is compared to). A mapping consists of a set of relations between features of the target and source [6]
- There is quite some research on analogies in education, but not on ML education. Only one research exists, which lists analogies without evaluating them [7]

### ③ Analogy Generation

- Lecture notes from the ML course in the Computer Science bachelor at Delft University of Technology were used for the concept list
- LLMs were used to generate analogies. They are good for that purpose, if their output is verified by experts.
- Results: 10 analogies for 10 concepts in unsupervised ML with a full mapping.

#### Concept list

- Unsupervised learning
- Clustering
- Intra-cluster cohesion
- Inter-cluster cohesion
- K-means clustering
- Single linkage
- Complete
- Average linkage
- Agglomerative dendrogram
- Divisive dendrogram

#### Example Analogy

**Complete Linkage:** If two cities want to ensure their farthest apart homes are still within reach, complete linkage measures the distance between the two farthest houses before connecting the cities.

- Clusters = Cities
- Data points = Houses
- Linkage distance = Max distance between two houses across cities
- Result = clusters only merge when elements are close

### ④ Expert Evaluation

- 15 experts (students who passed ML course, ML course TA's, ML lecturer) evaluated the analogies on target-concept coverage, mapping strength and metaphoricity
- Results: See Figure 1 and 2. Analogies 1-3 were rated best with high means and some agreement among experts. These continue to the student evaluation

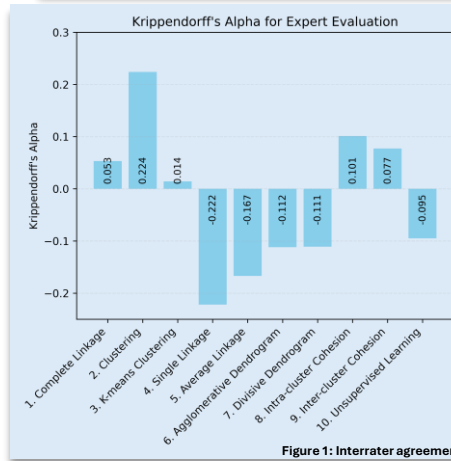


Figure 1: Interrater agreement

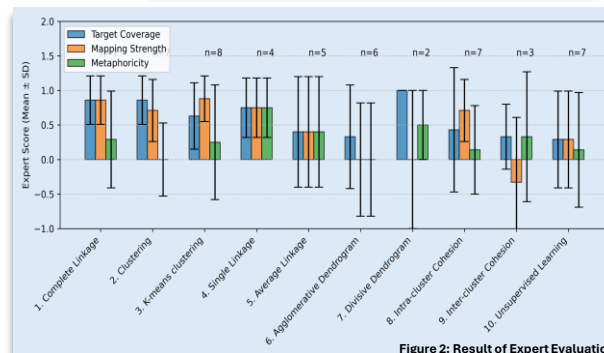


Figure 2: Result of Expert Evaluation

### ⑤ Student Survey

- A/B test to evaluate effect of analogies on knowledge gain of first-year CSE bachelor students. Group A got explanation with analogies. Group B got generic explanations.
- Both groups did a pre-test and post-test with questions on the level of ‘Understanding’ on Bloom’s taxonomy. Students also did non-cognitive evaluation: the Reduced Instructional Materials Motivation Survey.
- Results: See Figure 3 and 4. There was a higher knowledge gain for 2 of 3 analogies and post-test scores for Group A were higher/similar than B. Students with analogies were more motivated and confident. No results were statistically significant.

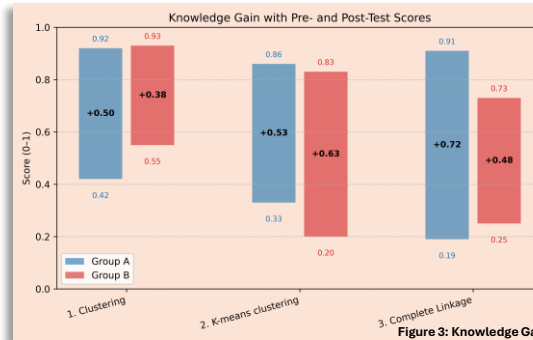


Figure 3: Knowledge Gain

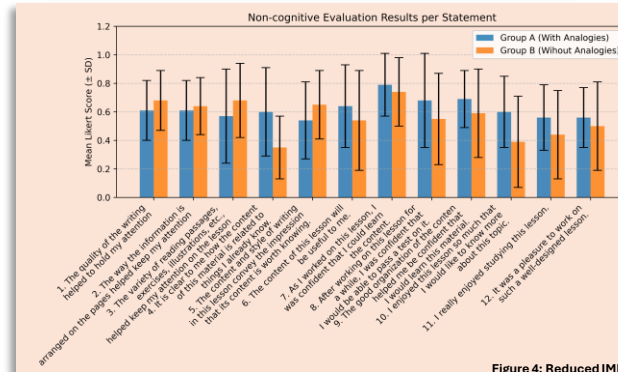


Figure 4: Reduced IMMS

### ⑥ Takeaways

- 3 generated analogies had a positive effect on students’ learning
- There is a great potential for more analogy-use in unsupervised ML education.
- All analogies and evaluation results can be found on [ml-teaching-analogies.github.io](https://ml-teaching-analogies.github.io)

### ⑦ Future work

- Develop more analogies for these concepts or for different concepts
- Evaluate analogies further to conclusively prove their effectiveness
- Investigate the long-term effect or effect on different levels on Bloom’s taxonomy

### References

- [1] Shapiro & Fiebrink (2019). Launching Agenda for Learning ML, ACM TOCE 19(4).
- [2] Hazzan & Mike (2023). Pedagogical Challenge of ML Education, Springer.
- [3] Almuqati et al. (2024). Challenges in Super- & Unsupervised Learning, IJASEIT 14(4).
- [4] Dagher (1995). Instructional Analogies in Science Ed, Sci Educ 79(3).
- [5] Analogy – Def.: “a comparison” (Cambridge Dictionary).
- [6] Bhavya et al. (2024). Long-Form Analogy Evaluation, INLG Gen Challenges.
- [7] Pendyala (2022). ML Real-World Analogies, Springer.