Interdisciplinary Research Classification Based on A Combined **Conceptual-Empirical Framework**

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1. Introduction

• The failure of universal indicators of interdisciplinarity

The concept of IDR (Interdisciplinary research) is multi-faced and complex. Various indicators have been presented to measure one or more aspects of this concept. However, recent studies have uncovered that most of the interdisciplinarity indicators are unsatisfying, with deviant results based on supposedly similar indicators.

These studies attest to the pronouncement posted by Ismael Rafols that the universal indicators of interdisciplinarity have failed. In order to confront this failure, a feasible solution with great potential is to construct a theoretical framework about IDR and link it to measurements.



• Classifying interdisciplinarity instead of measuring it directly

Classifying or categorizing is chosen as the guiding idea for the framework construction of IDR, because classifying or categorizing is an effective means to comprehend "beings" in essence, offering resistance against cognitive ambiguity and uncertainty. This study starts from the deconstruction of the concept of IDR, and then propose a classification framework.

• Aims

- Design an IDR classification framework, spanning from the theoretical to the empirical levels.
- 2. Combine qualitative and quantitative methods to conduct the classification task.
- 3. As a typical interdisciplinary field, the COVID-19 research field is selected to prove the validation of our proposed classification framework.

Fig 1. Research design of the proposed IDR classification framework spanning from the theoretical and empirical levels



Fig 2. Distribution of PLOS journals by the number of COVID-19 articles

2 **BERT** is a pre-trained language representation model standing for **B**idirectional Encoder **R**epresentations from Transformers.

In this study, a pre-trained BERT to do feature model used **1**S representation work, and the then obtain converted feature vectors for the next classification step.

DEEP LEARNING

Convolutional neural network (CNN) is a form of feed-forward neural network that achieves automatic feature engineering through the optimization of filters. In this study, MLP and CNN are used for the IDR classification task, feeding converted feature vectors obtained from the pre-trained BERT model.





Fig 5. Illustration of the proposed IDR classification model applied in the field of COVID-19

0.0 0.2 0.4 0.6 0.8 1.0 0.0 0.2 0.4 0.6 0.8 False Positive Rate False Positive Rate Fig 7. The ROC results of the four types based on the proposed IDR classification model

— ROC curve (area = 0.95)

Type 1

57%

ROC (Type 2)

ROC curve (area = 0.95)

0.4 0.6 0.8 False Positive Rate

ROC (Type 4)

—— ROC curve (area = 0.95)

4. Conclusions

This study proposes a well-designed classification framework about IDR and link it to a mixedmethod approach for implementation to better understand the concept of IDR under its contextualization and thus attempt to overcome the failure of universal interdisciplinarity indicators.

DEEP LEARNING

The coding method is used to manually categorize COVID-19 articles based on the 'why' and 'how' dimensions, classifying them into one of four types of IDR. Then the labeled articles are utilized to train a deep learning model, comprising pre-trained language model and deep learning classifier to derive the results of the resting unlabeled articles. The final results show that the number of articles of Synthetic type that well-satisfy the definition of IDR accounted for the largest proportion of the total.

5. References

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