

Big Data Analytics and Text Mining Project

Testing GraphAny on a RelBench task

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01 The Context

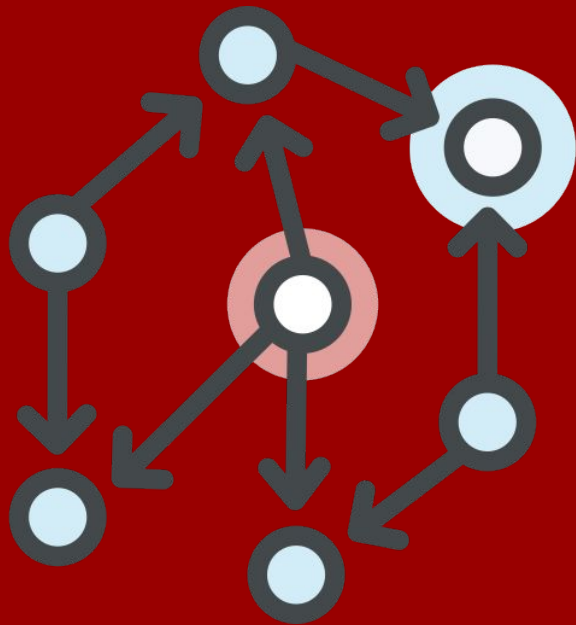


GraphAny

A fully-inductive Node Classification on Arbitrary Graphs

It is trained to perform node classification on:

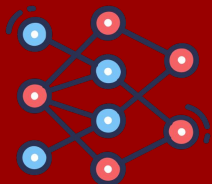
- Any graph.
- With any feature.
- With any number of labels.



Totally Inductive Node Classification!

Some other fancy aspects

GraphAny is a young model, (june 2024 on Arxiv)!



Built on only 5 LinearGNNs!

Tested on 27 datasets.



And the number of parameters?



4 different checkpoints available.



RelBench

Open benchmark for machine learning over relational databases

Full of interesting tasks to perform on relational graphs

More importantly: we have some Binary Node
Classification Tasks!

Project Idea

Import a task from RelBench and test GraphAny on it!

But How?

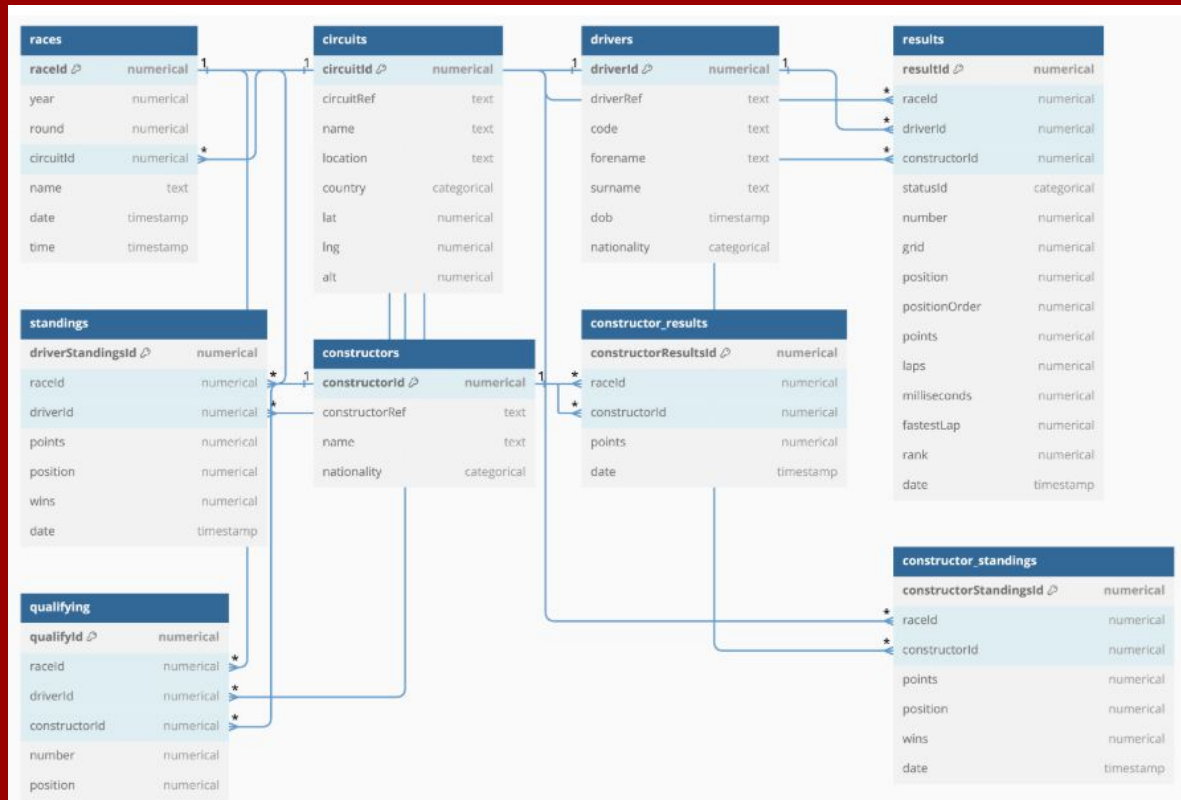
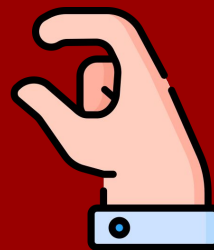


02 The Approach



The choice: F1 Database

The smallest dataset with a node classification task.



The task: For each driver predict if they will DNF (did not finish) a race in the next 1 month

The HeteroGraph problem

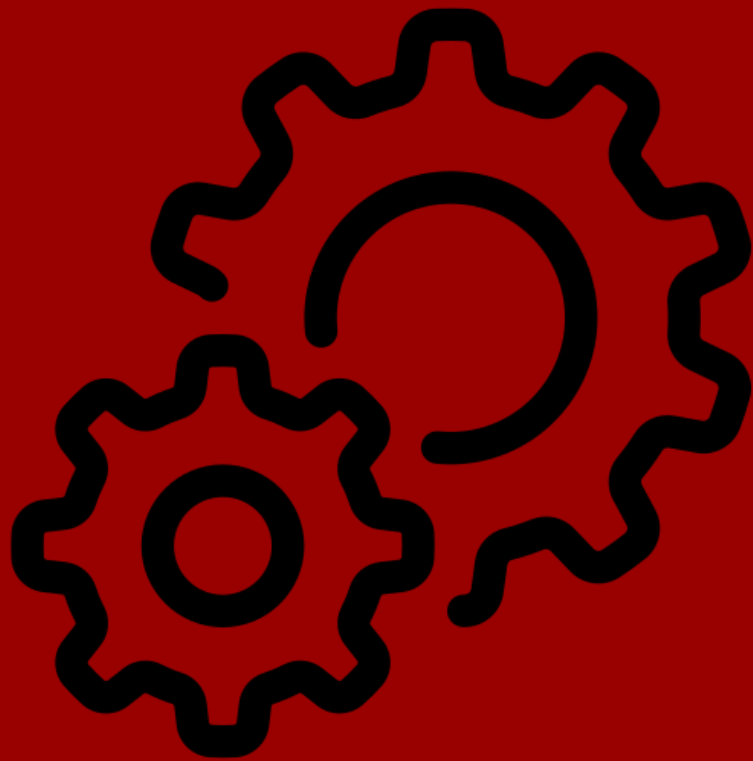
Different nodes with different features...

Solution: Linearization and Glove embedding!

```
HeteroData(  
  constructor_standings={  
    tf=TensorFrame([13051, 4]),  
    time=[13051],  
  },  
  results={  
    tf=TensorFrame([26080, 11]),  
    time=[26080],  
  },  
  circuits={ tf=TensorFrame([77, 7]) },  
  drivers={ tf=TensorFrame([857, 6]) },  
  races={  
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    time=[1101],  
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  standings={  
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  },  
  qualifying={  
    tf=TensorFrame([9815, 3]),  
    time=[9815],  
  },  
  constructors={  
    tf=TensorFrame([211, 3]) },  
  constructor_results={  
    tf=TensorFrame([12290, 2]),  
    time=[12290],  
  }  
)
```

03

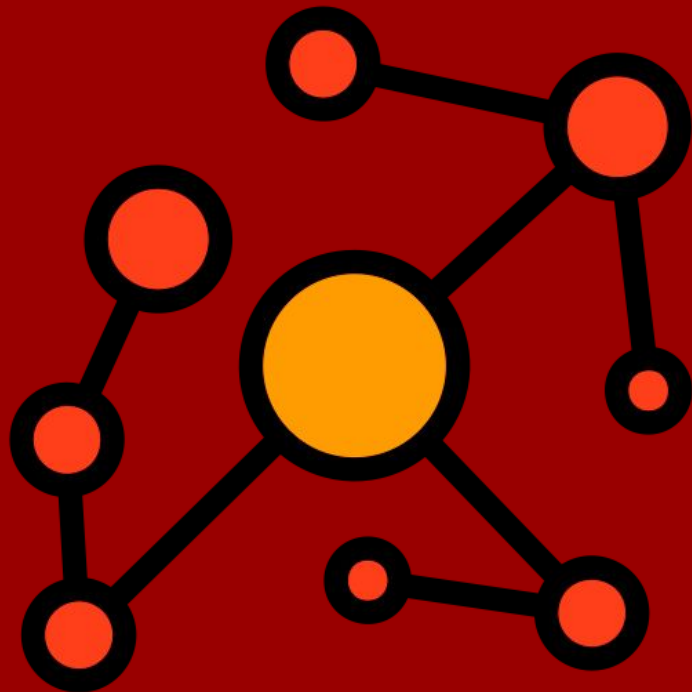
Experimental Setup



Our Dataset

A HUGE Knowledge Graph:

- 97605 nodes.
- 455432 edges.
- but only 3 labels.



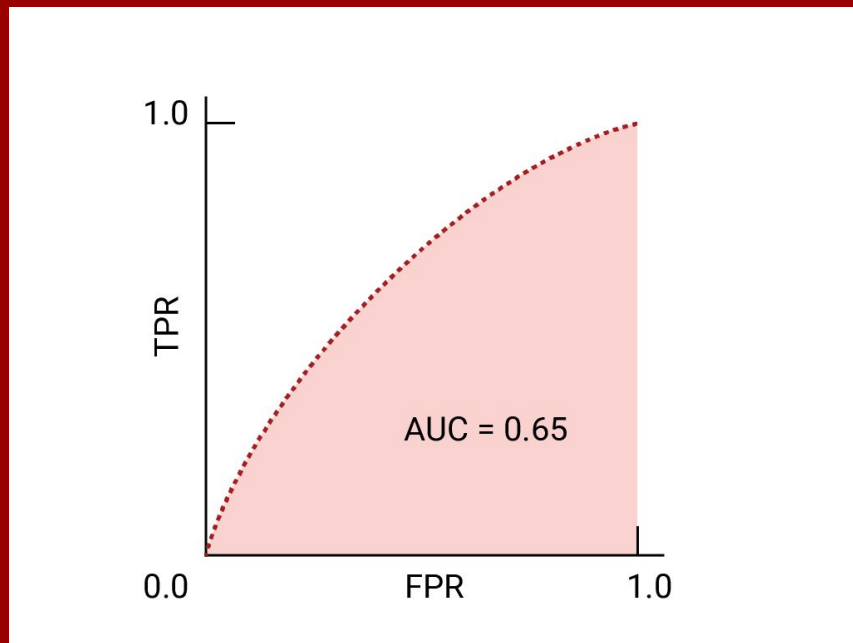
The Metrics

1. F1-Score

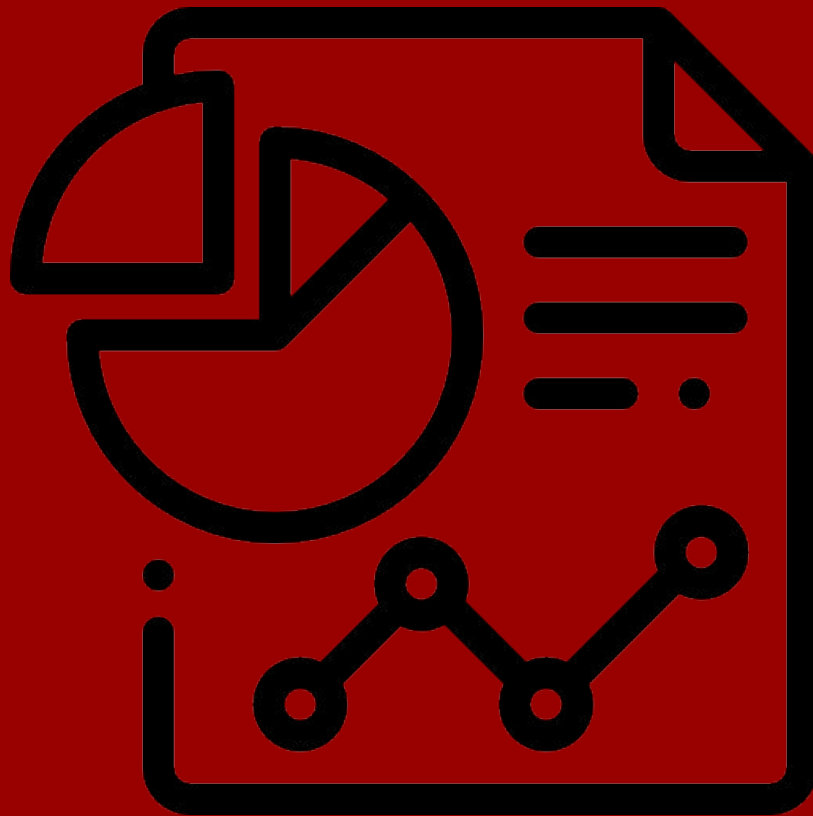


2. Binary ROC AUC

3. One-vs-rest ROC AUC



04 The Results



GraphAny performances on Task

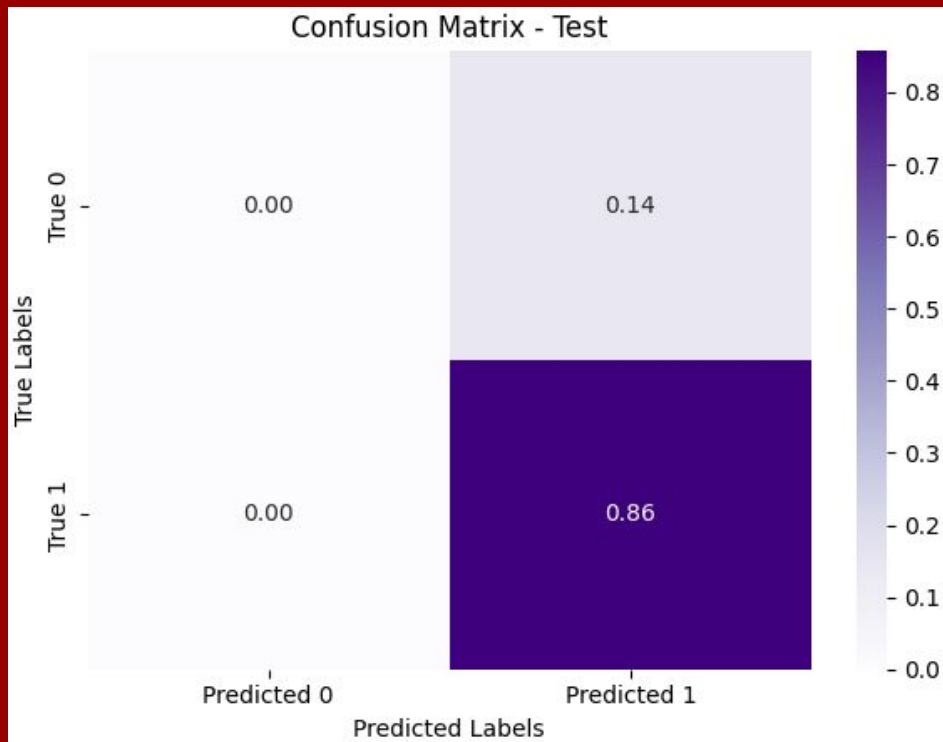
Precision, Recall and F1 per class.

| Checkpoint | Precision | Recall | F1 Class 0 | F1 Class 1 |
|------------|-----------|--------|------------|------------|
| Wisconsin | 0.86 | 1.00 | 0.00 | 0.92 |
| Cora | 0.86 | 1.00 | 0.00 | 0.92 |
| Arxiv | 0.86 | 1.00 | 0.00 | 0.92 |

Table 1: GraphAny Performances on Formula 1 Test set RelBench Knowledge Graph in terms of Precision, Recall and F1-Score on both classes.

GraphAny performances on Task

Confusion Matrix.



GraphAny performances on Task

Cohen's Kappa Score, Binary ROC-AUC and One-vs-Rest ROC-AUC.

| Checkpoint | Cohen's kappa | Bin-AUROC | OvR-AUROC |
|------------|---------------|-----------|-----------|
| Wisconsin | 0.00 | 58.80 | 53.70 |
| Cora | 0.00 | 60.19 | 56.25 |
| Arxiv | 0.00 | 56.94 | 53.47 |

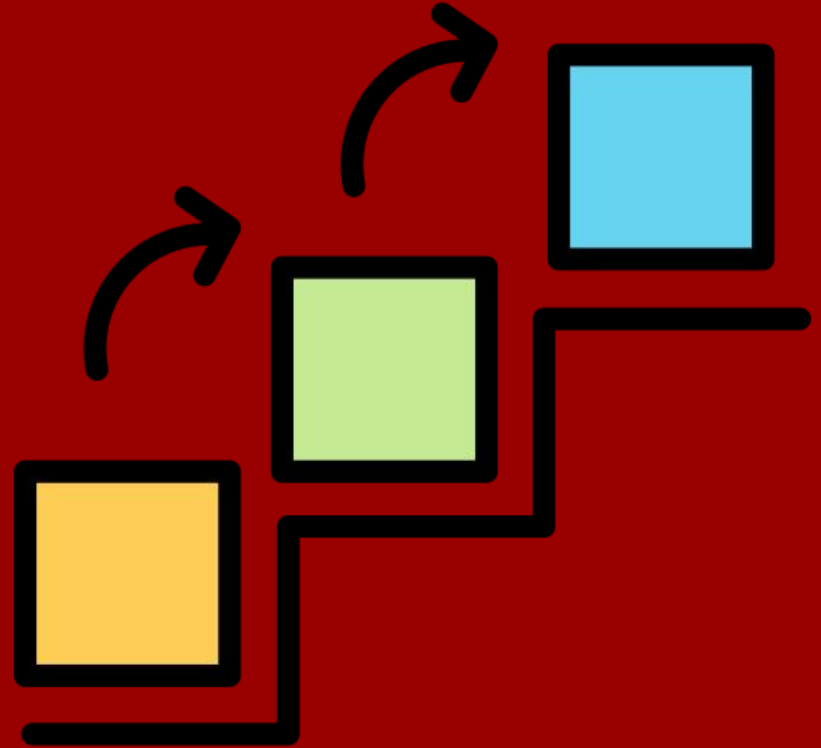
Table 2: GraphAny Performances on Formula 1 Test set RelBench Knowledge Graph in terms of Cohen's Kappa (computed between a random classifier and GA classifier), Binary ROC-AUC and One-vs-Rest ROC-AUC.

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Future

Project

Improvements



Some critical aspects

- We tested the model on only one task.
- We tested the model on only one dataset.



Our metrics could be biased by the task choice!

Thank you for your attention!

