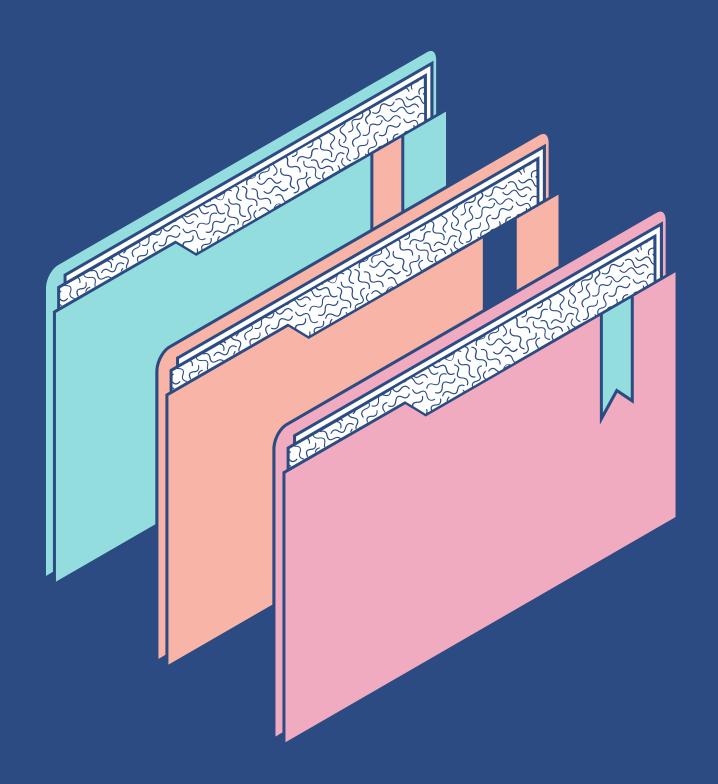


Neural Execution of Graph Algorithms

MohamedElfatih

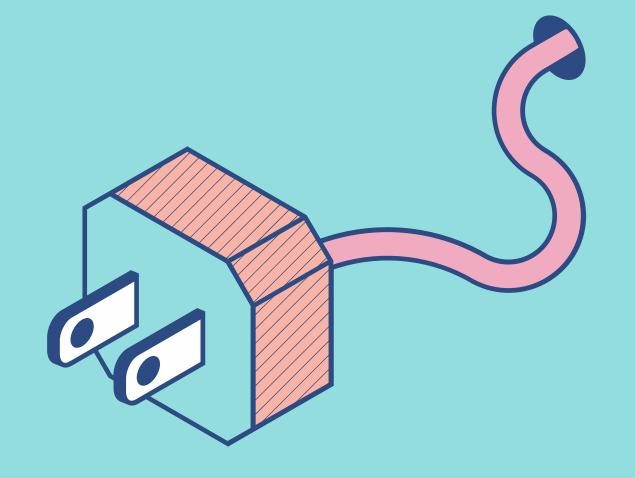


Agenda

KEY TOPICS DISCUSSED IN THIS PRESENTATION

- Introduction
- Implementation
- Experiments
- Results
- Future work

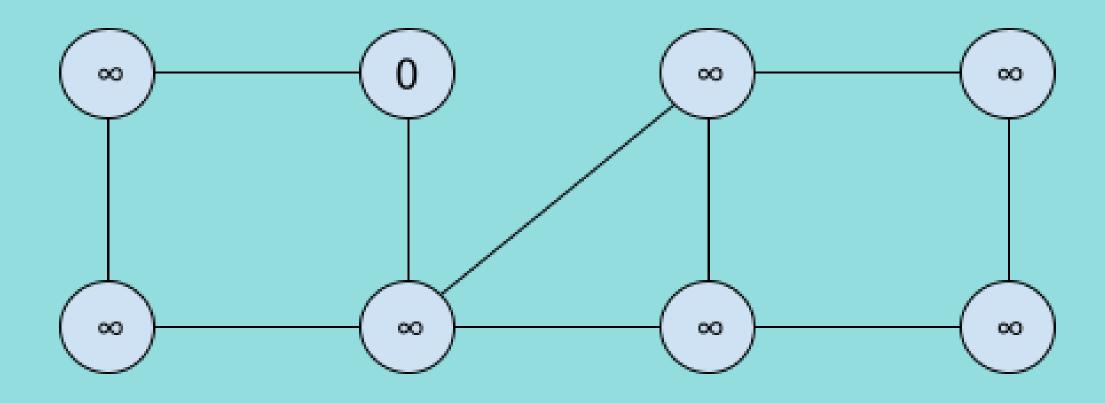
Can we make neural network reasons like an algorithms?



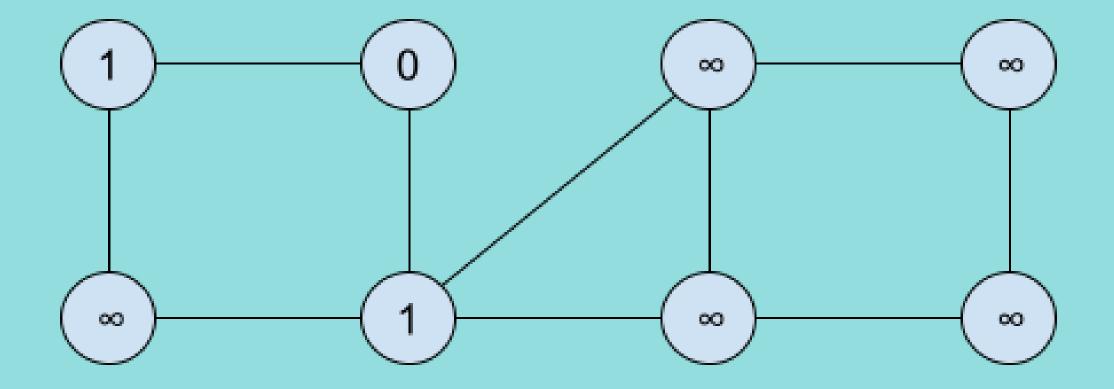
But.... What is an Algorithm?

Breadth-First Search

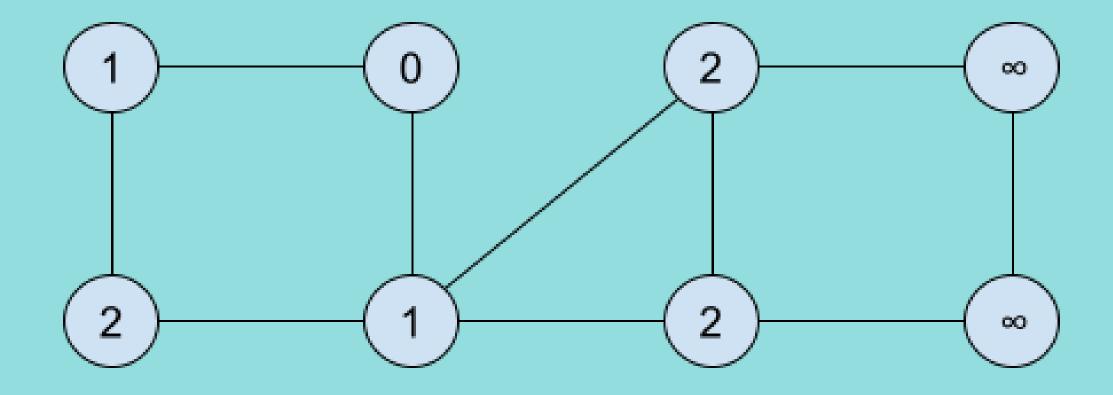
Initially



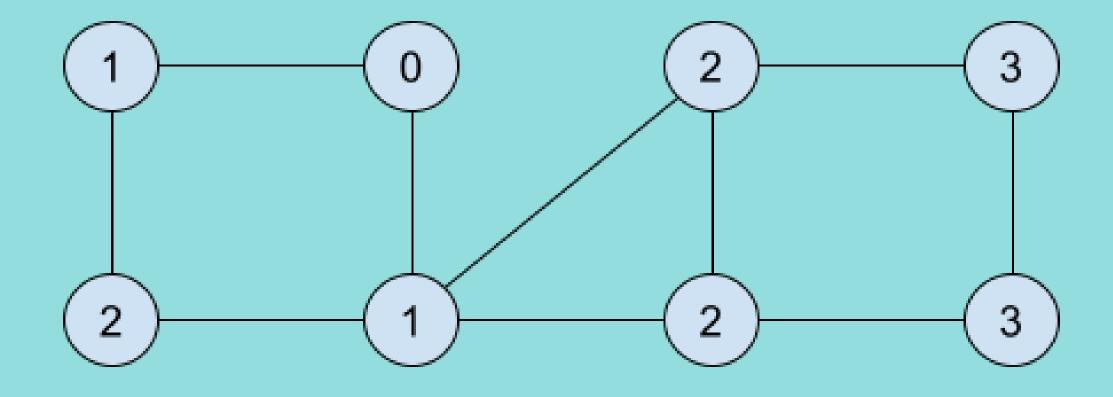
First step



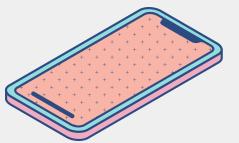
Second step

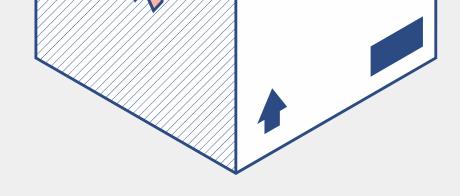


Third step









Neural Network

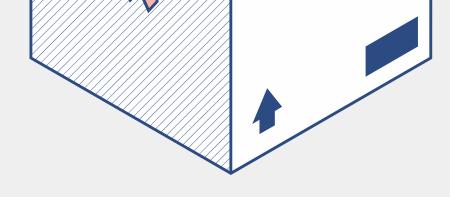
- Operate on raw input data.
- Generalize on noisy conditions.
- Reusable across tasks.
- More data.
- Not quite interpretable

Classical Algorithms

- Strongly generalize.
- Interpretable.
- Enables a more personalized kind of learning for students.
- Input must match specifications
- tasks variations

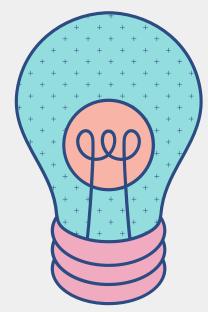






Neural Network

- Operate on raw input data.
- Generalize on noisy conditions.
- Reusable across tasks.
- More data.
- Not quite interpretable



Combine Both!

Classical Algorithms

- Strongly generalize.
- Interpretable.
- Input must match specifications
- tasks variations



- Operate on raw input data.
- Generalize on noisy conditions.
- Reusable across tasks.
- More data.
- Not quite interpretable



Combine Both!



- Strongly generalize.
- Interpretable.
- Input must match specifications
- tasks variations

Can we make neural network reasons like algorithms?

Why?



Strong Generalization

Learning reasoning not a mapping

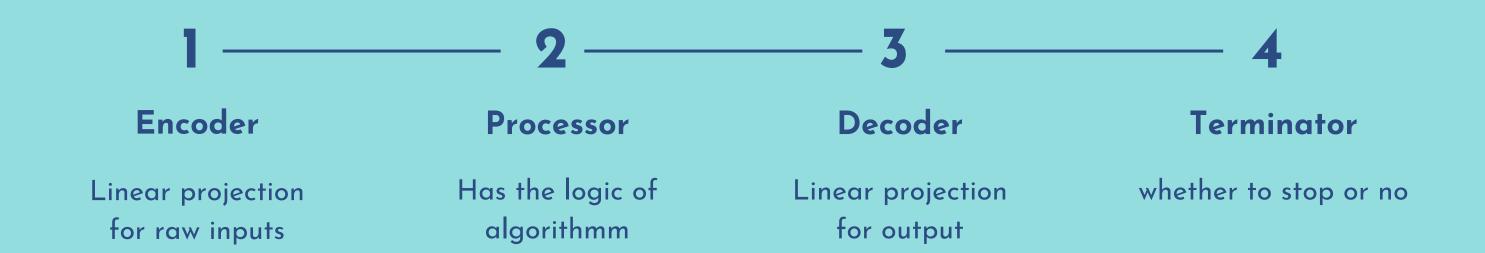
Multi-task learning

resuse algorithms across tasks

Discovering Novel Algorithms

Improved heuristics for interactable problems

Executor Framework



Results

Reachability Results

	My Results	Paper Results
GAT	99/100	92.34/99.97
MPNN-max	100/100	99.92/99.8

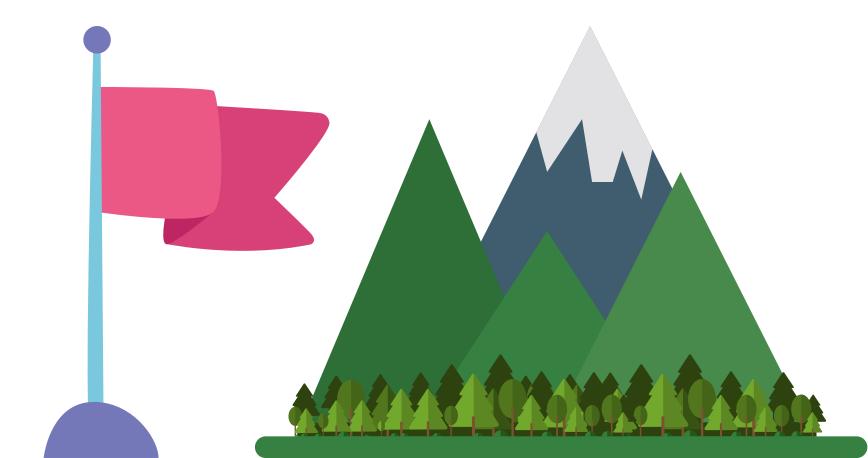
Bellman Ford meansquared error

Method	My Results	Paper Result
MPNN-max (no reach)	4.24	2.628

Future Work







Do you have any questions?

Feel free to ask any question

