

Temporal Graph Neural Networks with GraphNeuralNetworks.jl

Aurora Rossi

▶ To cite this version:

Aurora Rossi. Temporal Graph Neural Networks with GraphNeuralNetworks.jl. Julia and Optimization Days 2023, Oct 2023, Paris, France. hal-04230797

HAL Id: hal-04230797 https://hal.science/hal-04230797v1

Submitted on 6 Oct 2023

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



Temporal Graph Neural Networks with GraphNeuralNetworks.jl



Aurora Rossi ¹

¹ Université Côte d'Azur, COATI, INRIA, CNRS, I3S, France

GraphNeuralNetworks.jl

A graph neural network Julia package based on the Flux.jl.

Among its features:

- Standard graph convolutions
- Batched graph support
- Custom layer creation
- GPU computing via CUDA.jl
- Graphs.jl integration
- Various ML task examples
- Heterogeneous and temporal graphs

Installation pkg> add GraphNeuralNetworks

Gender classification in brain networks

Gender classification is the problem of determining the category of the graph, in this case female and male.

Temporal Brain Networks Dataset

The networks are obtained from resting-state fMRI data.

Nodes: 102 brain regions

Edge weights: region activity correlations

Node features: concatenation of one-hot encoding and

mean activation of a node

Graph features: one-hot encoding of the gender

Model

- Graph Isomorphism Network for every snapshot $X'=f_{\theta}([(1+\epsilon)I_N+A]X)$ where f_{θ} is a Multi Layer Perceptron
- Global pooling (mean) on node dimension
- Global pooling (mean) on time dimension
- A fully connected layer



The authors are grateful to the OPAL infrastructure from Université Côte d'Azur for providing resources and support. This work has been supported by the French government, through the UCA DS4H Investments in the Future project managed by the National Research Agency (ANR) with the reference number ANR-17-EURE-0004.



SYSTÈMES NUMÉRIQUES
POUR L'HUMAIN



Temporal graphs

Real-world dynamic networks (social, transportation, brain) are modeled as sequences of static graphs.

Traffic prediction example

Traffic forecasting is the problem of predicting future **traffic trends** on a road network given historical traffic data.

METR-LA Dataset

It contains traffic data from 207 sensors in highways of Los Angeles County.

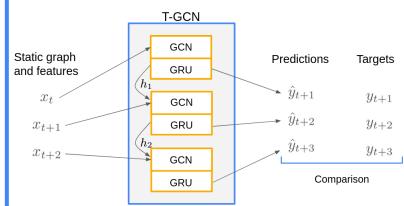
Nodes: sensors

Edge weights: distances between the sensors **Node features:** traffic speed and time of the day

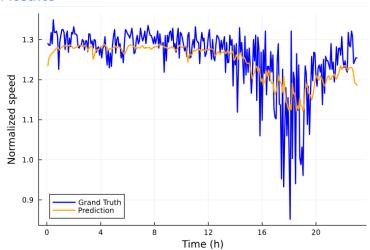
T-GCN Model

It is a recurrent model and a cell is composed of:

- Graph Convolutional Network to model spatial dependence
- Gated Recurrent Unit to model temporal dependence



Results



References

"T-GCN: A Temporal Graph Convolutional Network for Traffic Prediction" Zhao et al.

"Diffusion Convolutional Recurrent Neural Network: Data-driven Traffic Forecasting" Li et al.

"How Powerful are Graph Neural Networks?", Xu et al. https://summerofcode.withgoogle.com/archive/2023/projects/vbQaqNYWhttps://github.com/CarloLucibello/GraphNeuralNetworks.jl