

# Time-aware Robustness of Temporal Graph Neural Networks

Marco Säbzer

Silvia Beddar-Viesing

University of Kassel, Germany

# Robustness in the Context of Neural Networks



# Robustness in the Context of Neural Networks

“pig”

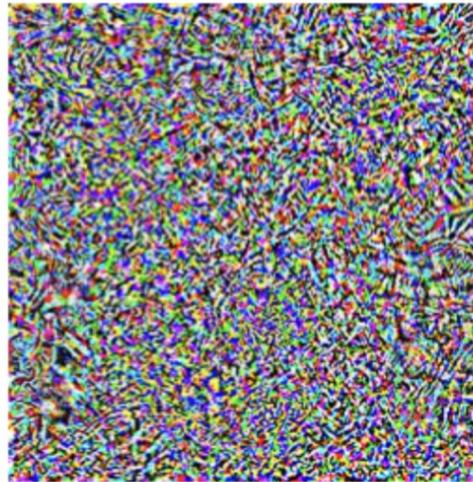


# Robustness in the Context of Neural Networks

“pig”



+ 0.005 x



=



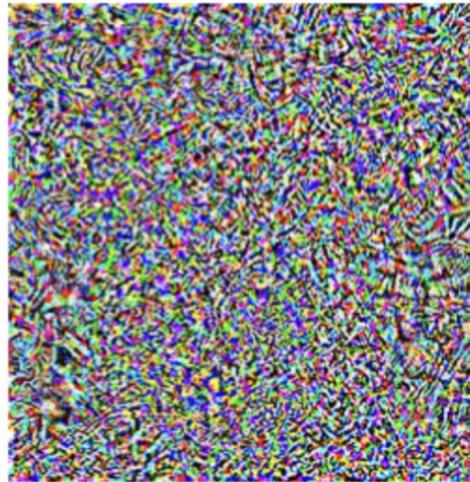
Madry & Schmidt, 2018

# Robustness in the Context of Neural Networks

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=

“airliner”



# Pointwise - Robustness over Real-Valued Data

We call  $P = (\bar{x}, B)$  a robustness property where

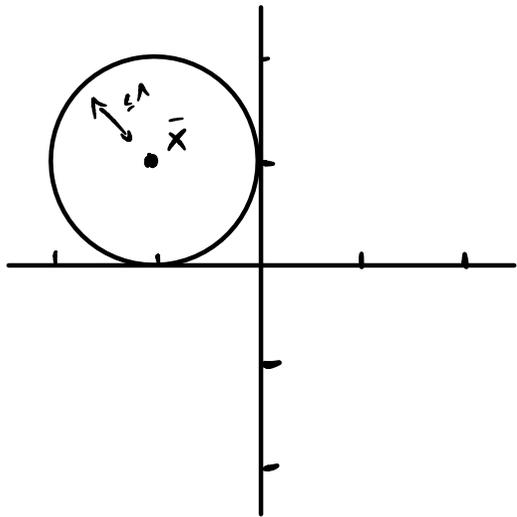
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Example:  $P = \left( \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \left\{ \bar{x}' \mid \left\| \begin{pmatrix} 1 \\ -1 \end{pmatrix} - \bar{x}' \right\| \leq 1 \right\} \right)$

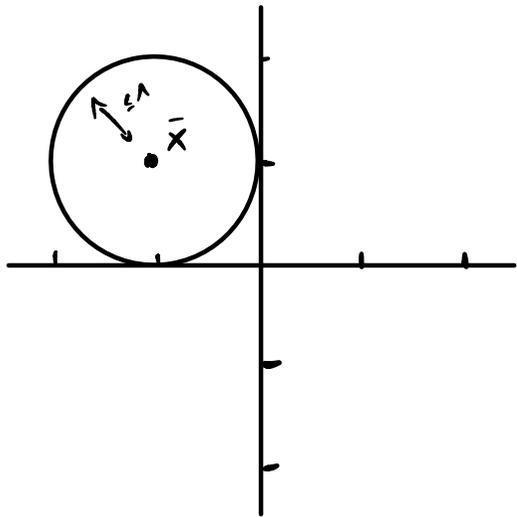


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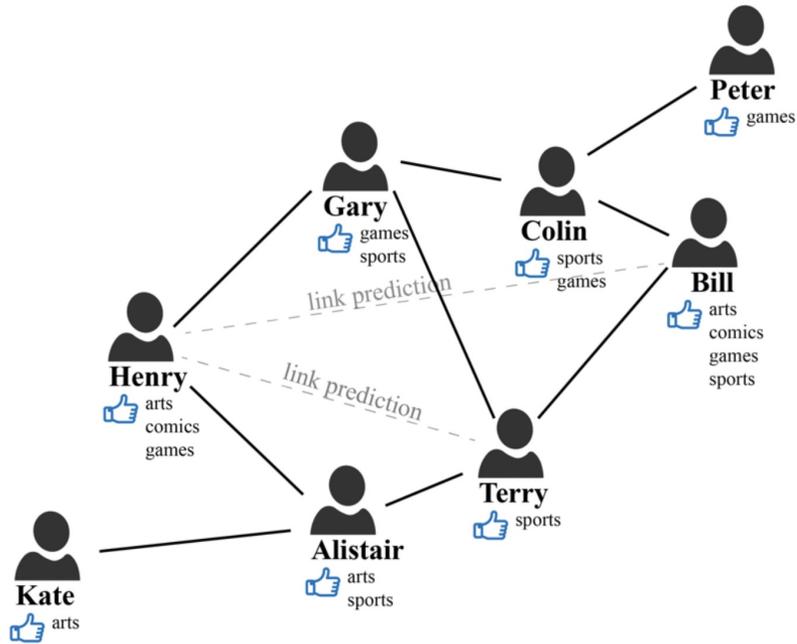
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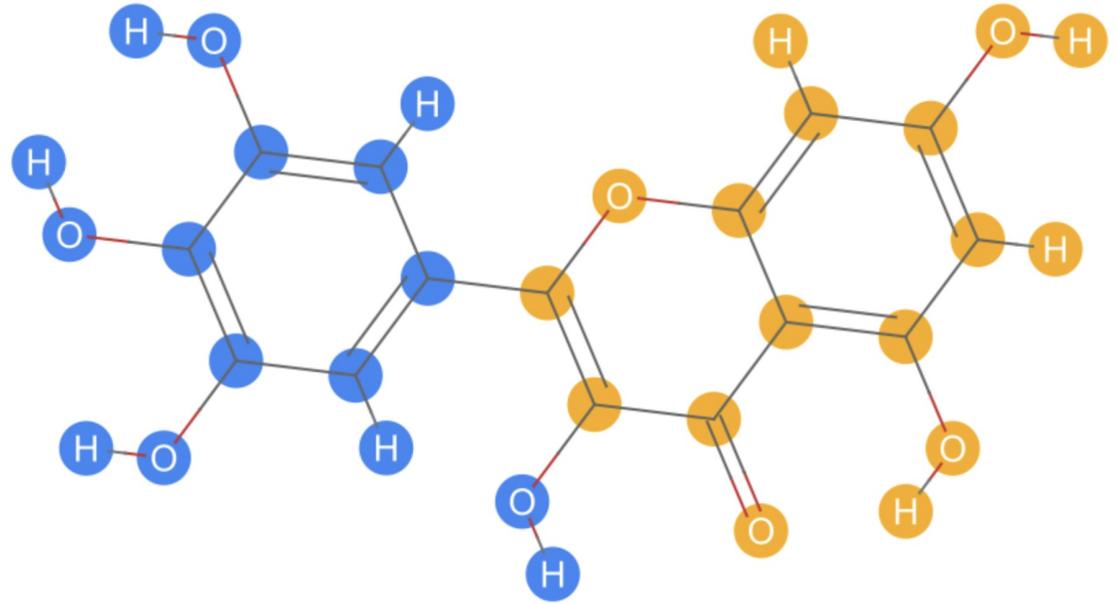


We say that  $N$  is robust regarding  $P$  if  $\forall \bar{x}' \in B$  it holds  $N(\bar{x}) \approx N(\bar{x}')$ .

# Neural Models for Graph-Data



(1) social networks



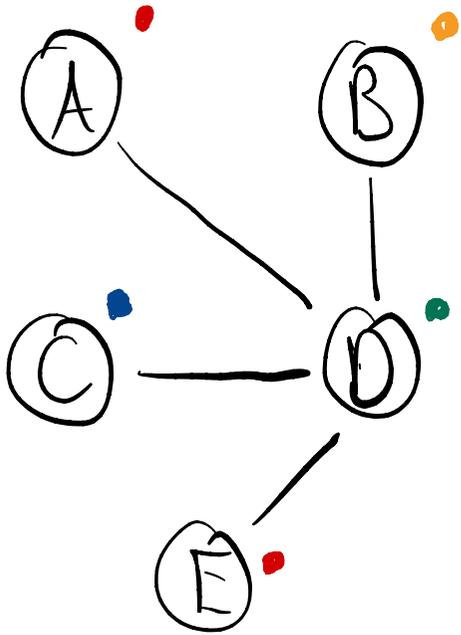
(2) molecule classification

(1) Cui & Beech, 2022

(2) wolfram.com

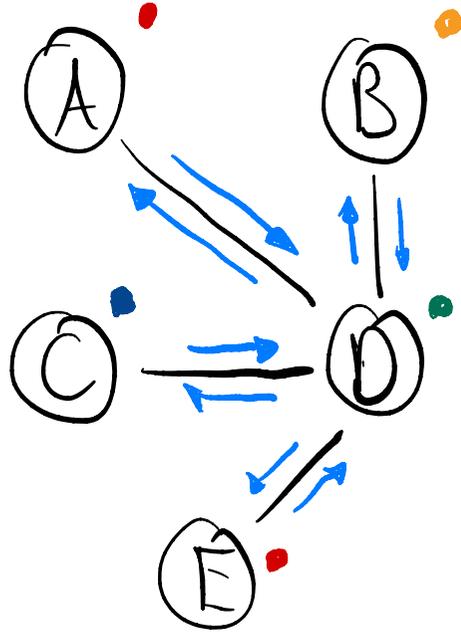
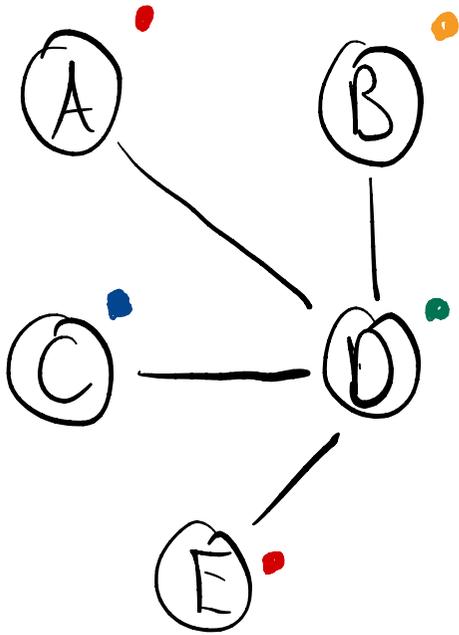
# Graph Neural Networks (GNN)

Usual model: Message-Passing or Aggregation-Combine



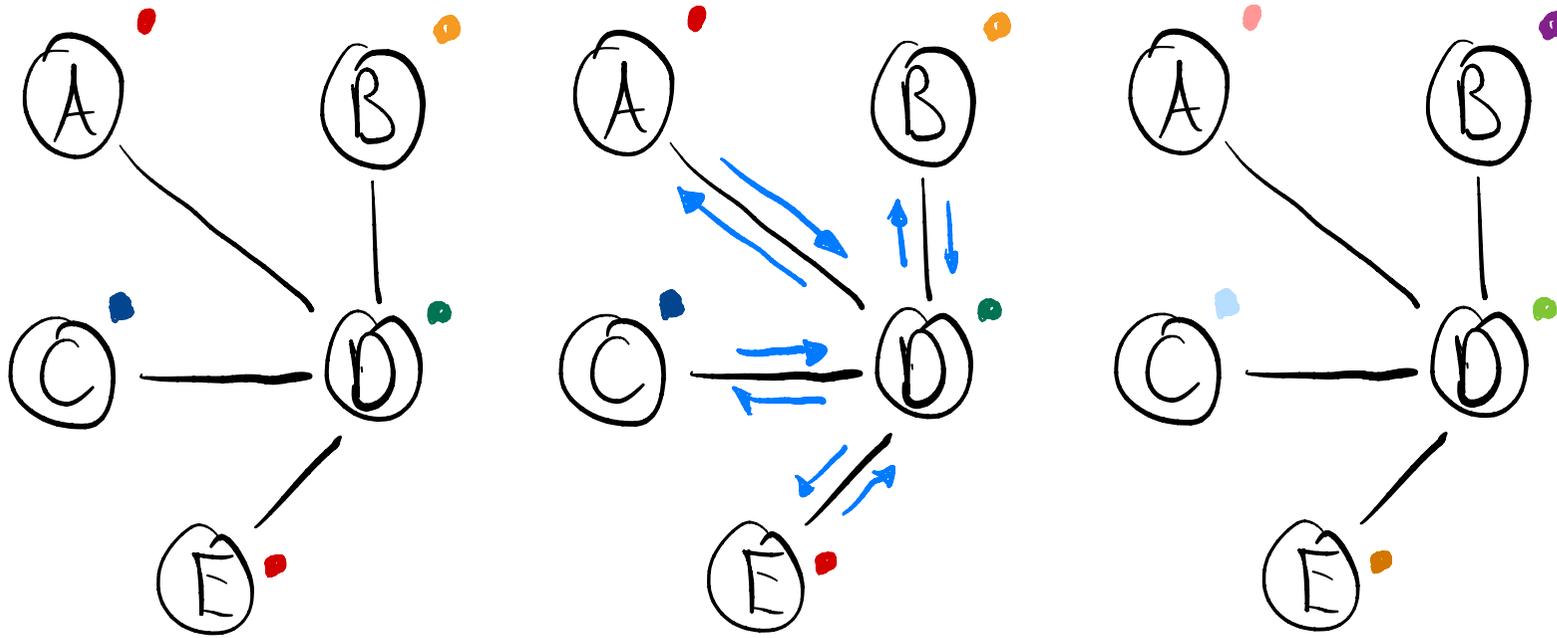
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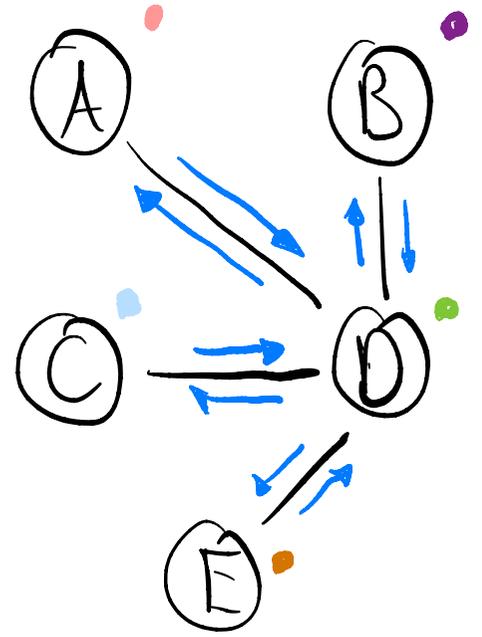
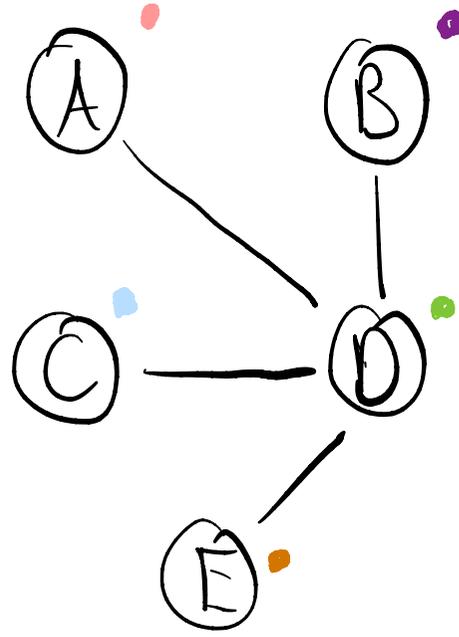
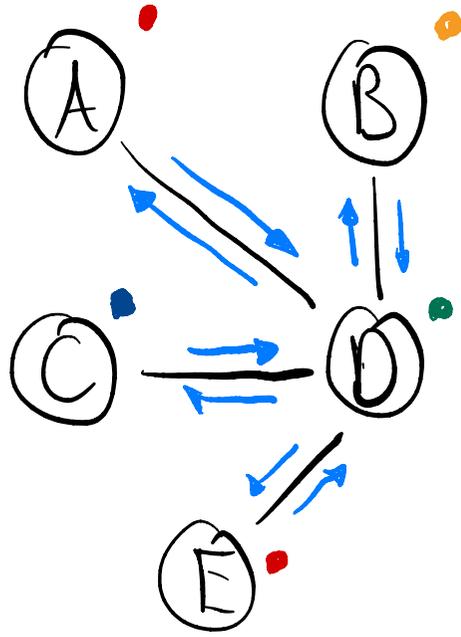
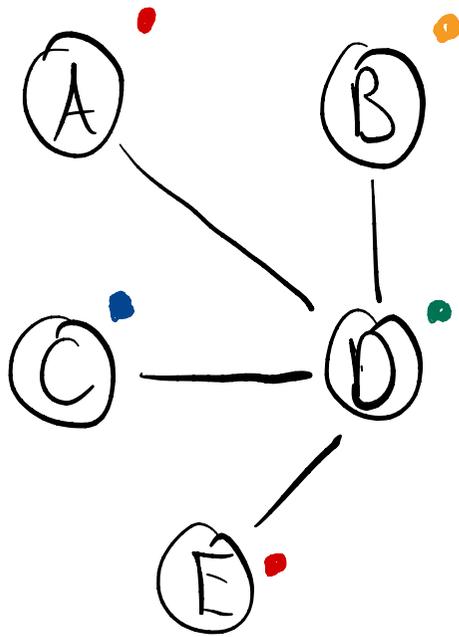
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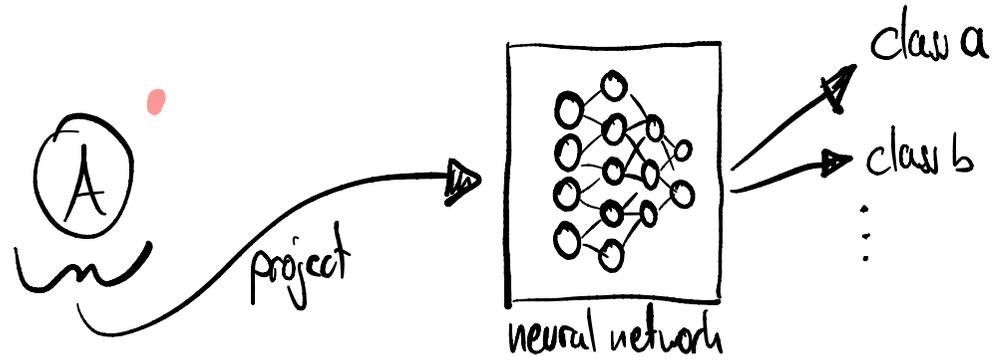


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Usual tasks:

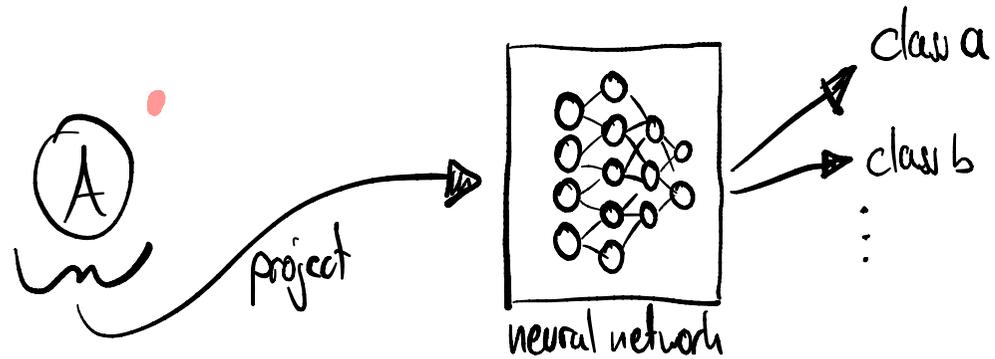
node classification



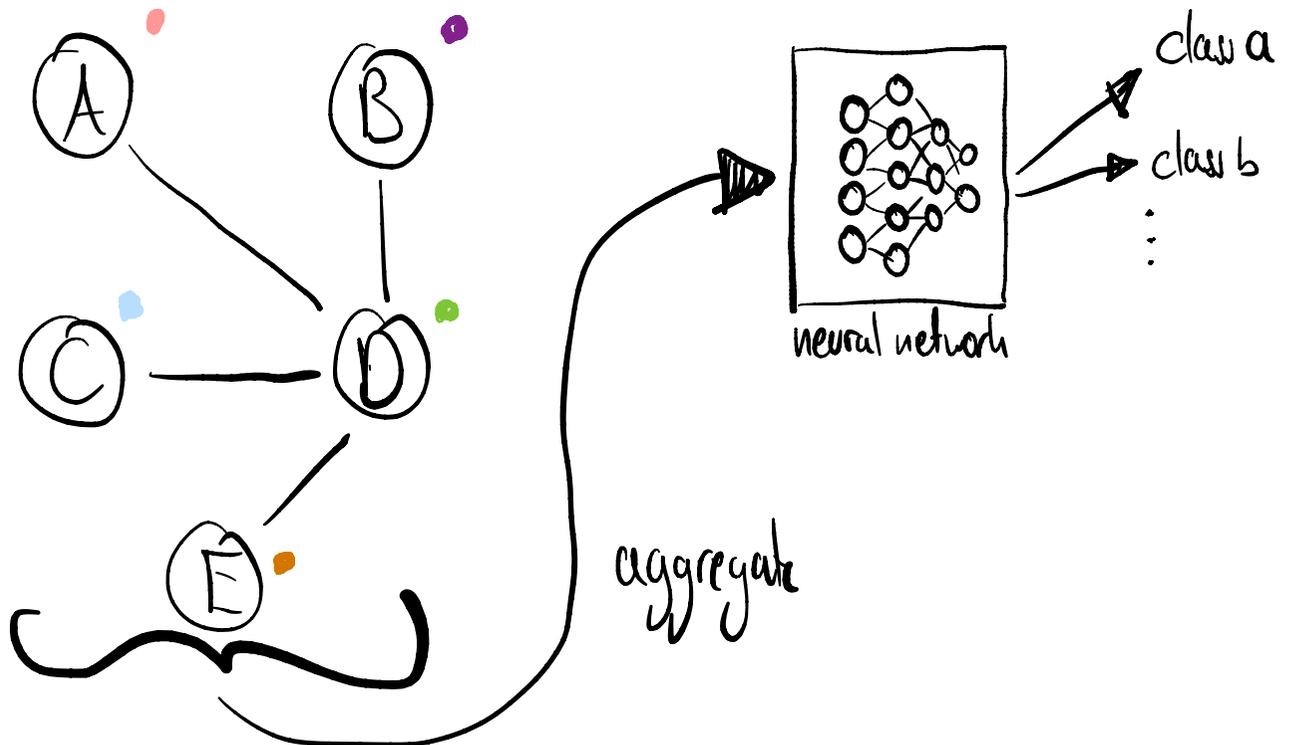
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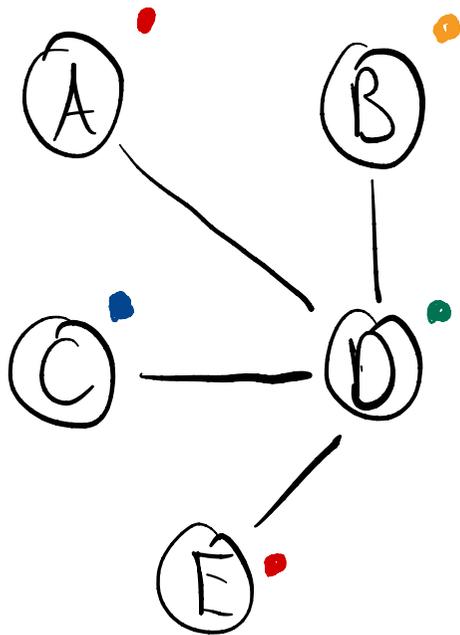
whole graph classification



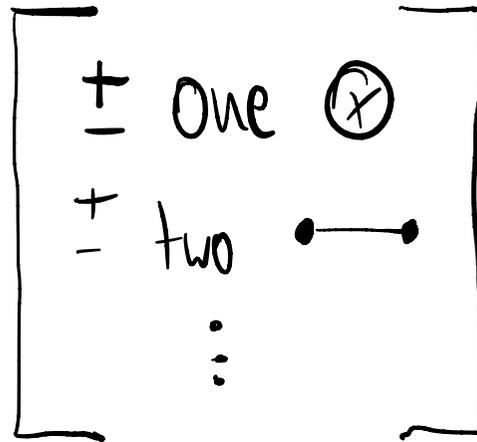
# Point-wise Robustness over Graph Data

Analogous to real-valued data, Center point is some graph  $G$  and budget are allowed amount of perturbations.

Example:



center point

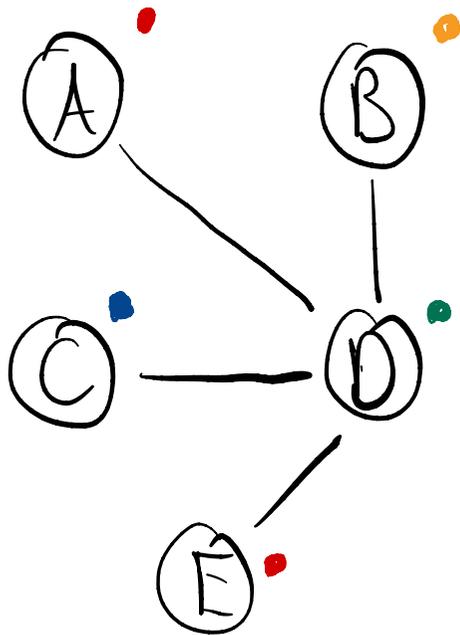


budget

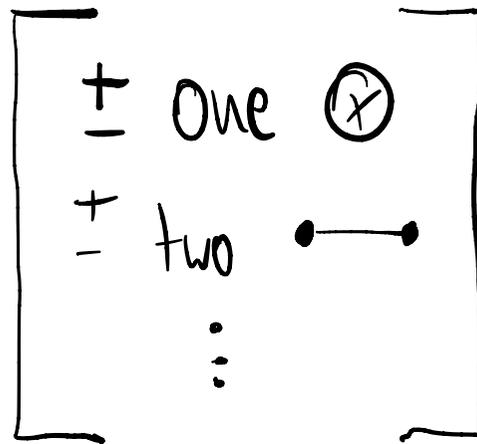
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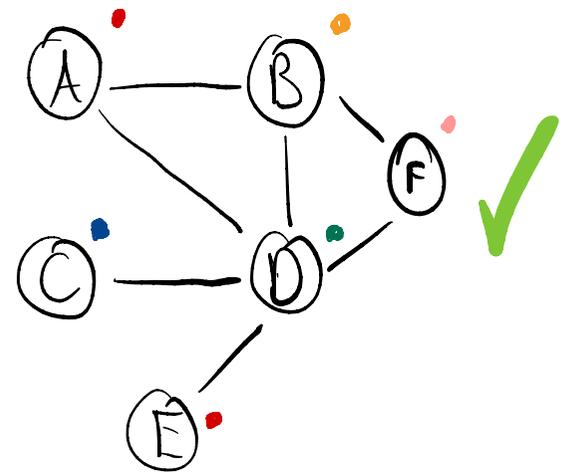
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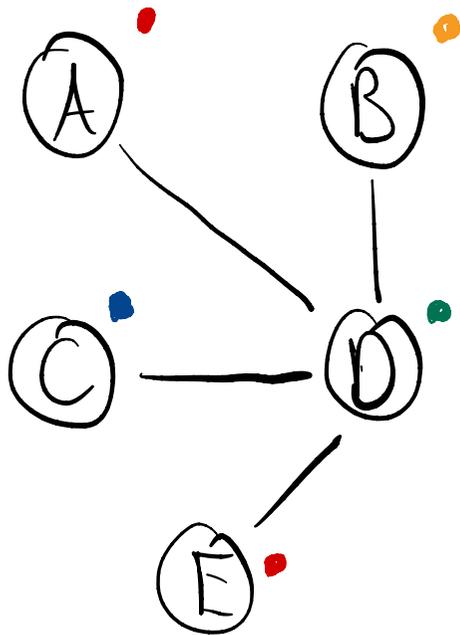
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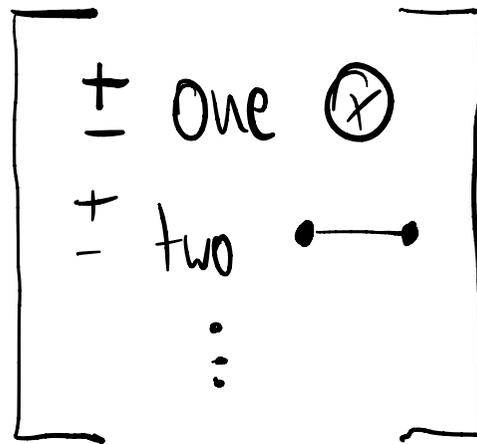
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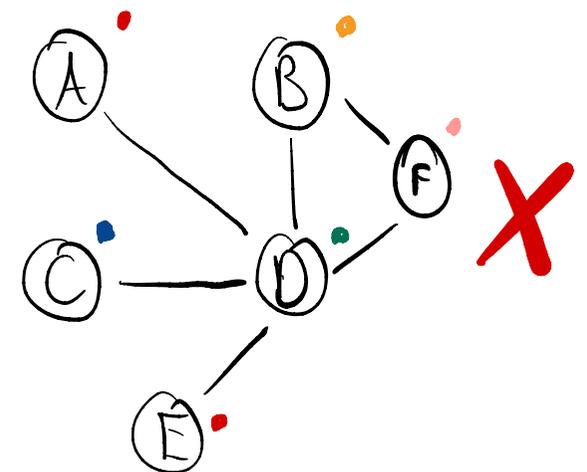
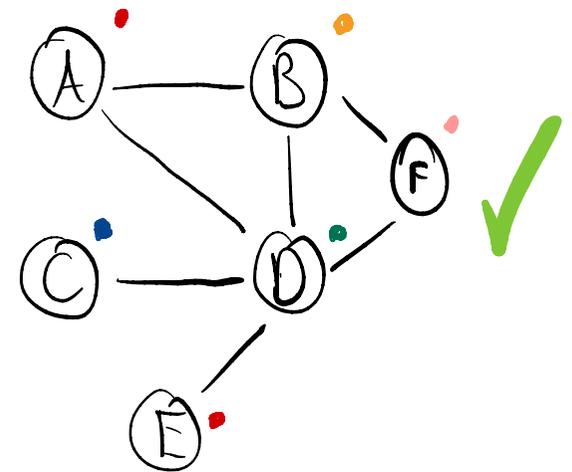
Example:



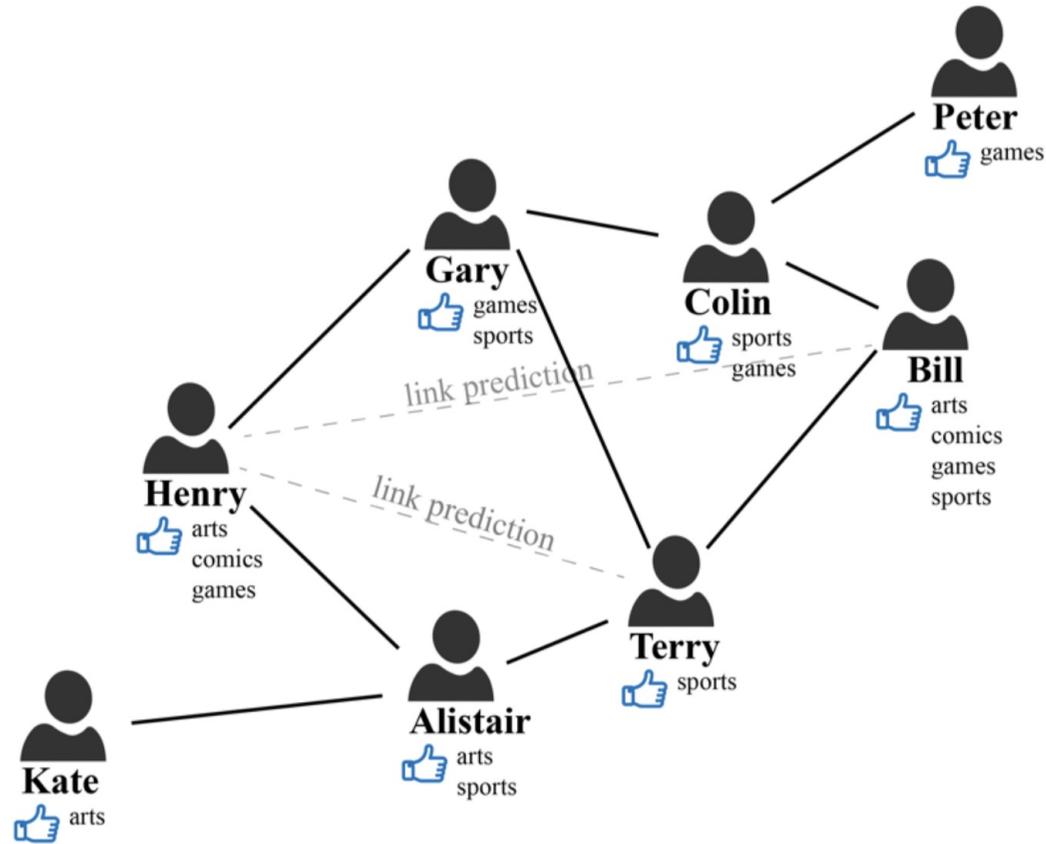
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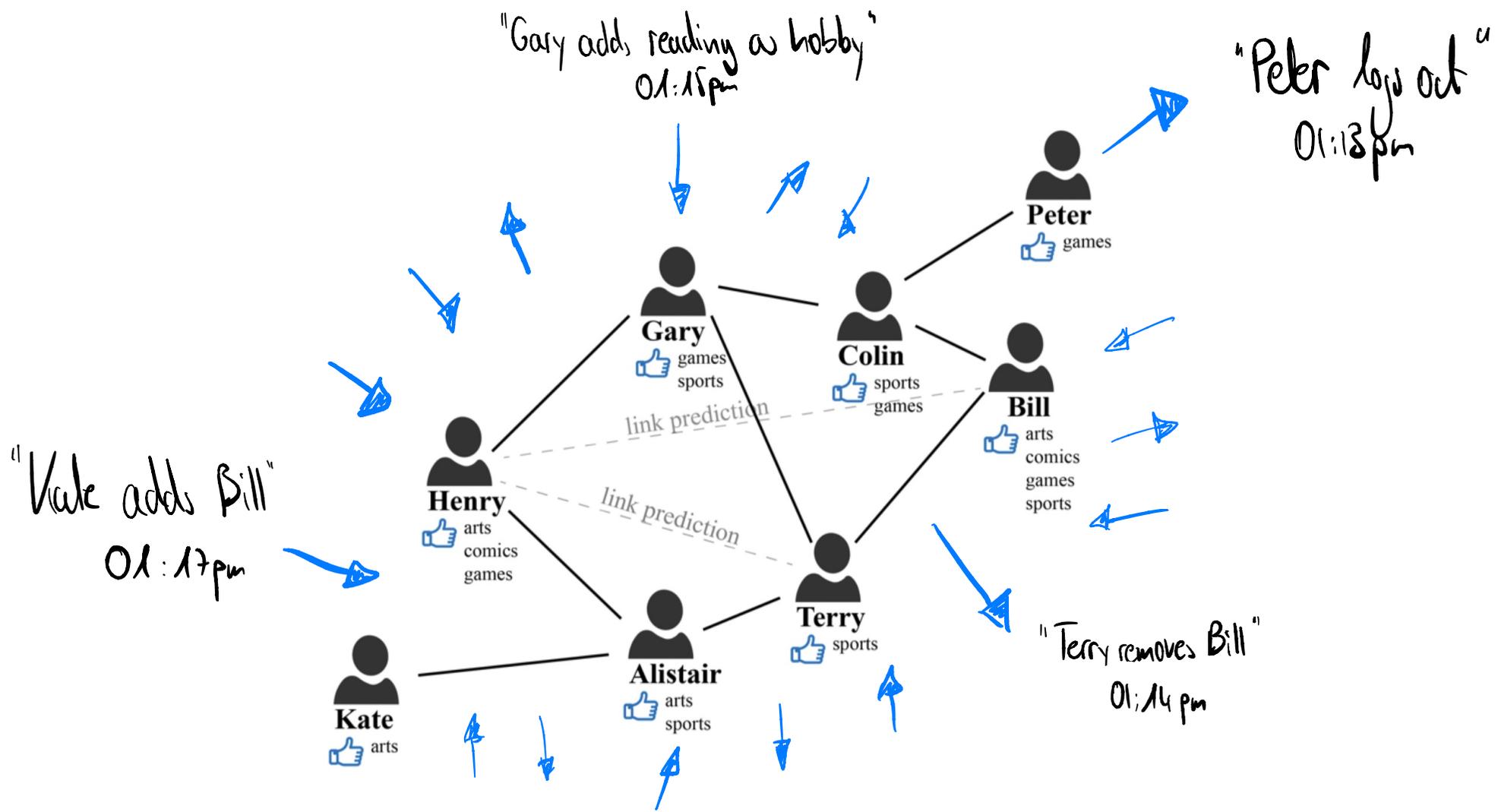
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# Limitations of the Usual GNN Setting



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# Temporal Graph Neural Networks (TGNN)

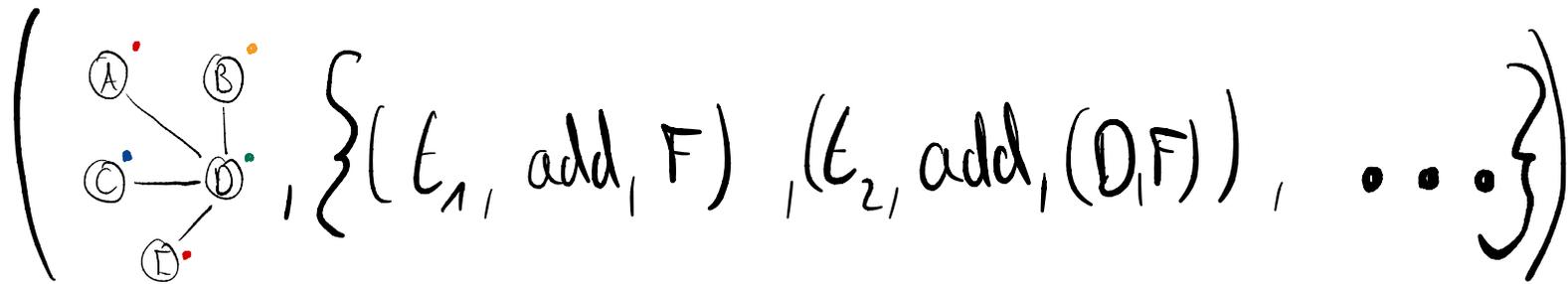
A Continuous-Time Temporal Graph (CTG) is a tuple  $(G, O)$

- $G$  is a (usual) graph, called start graph
- $O$  is a (finite) set of time-stamped observations

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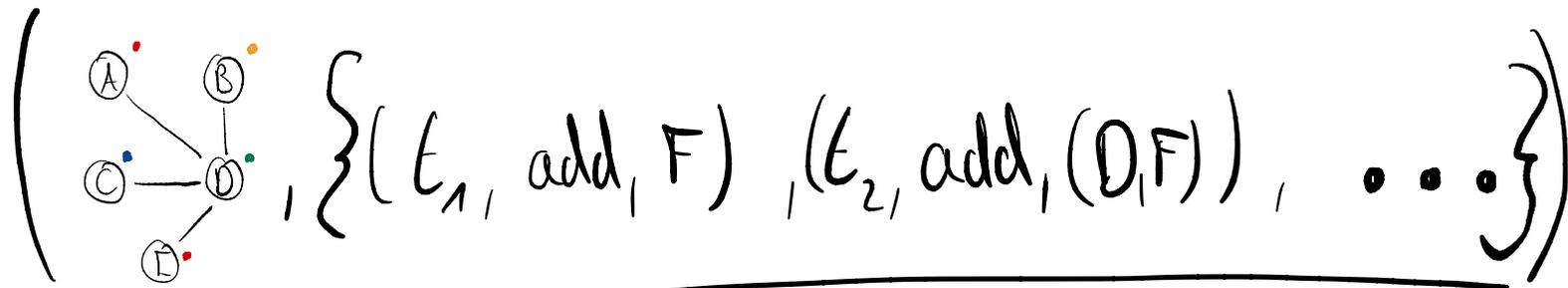
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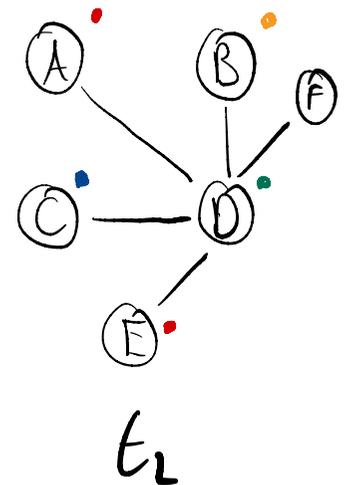
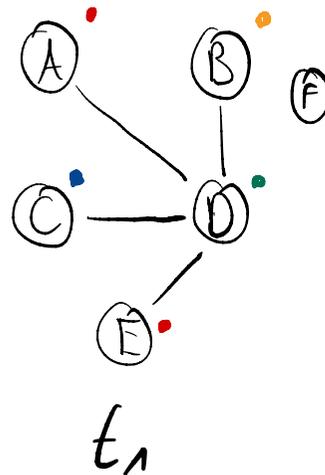
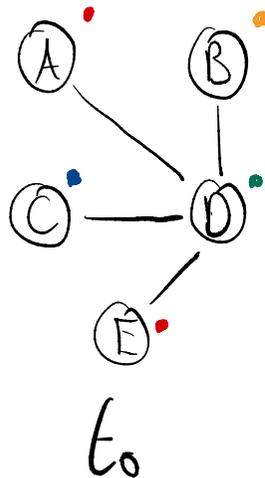
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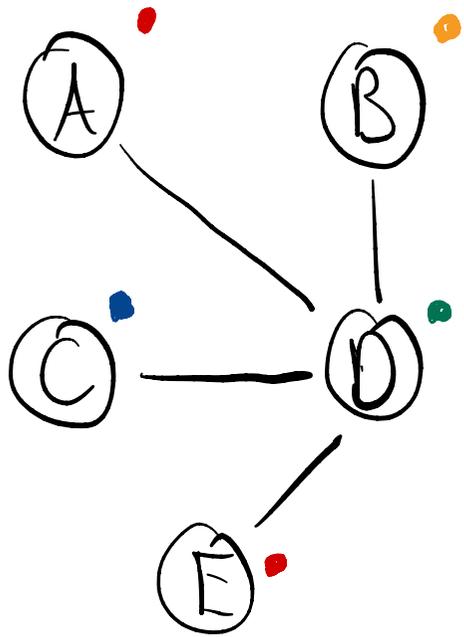
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Unfold



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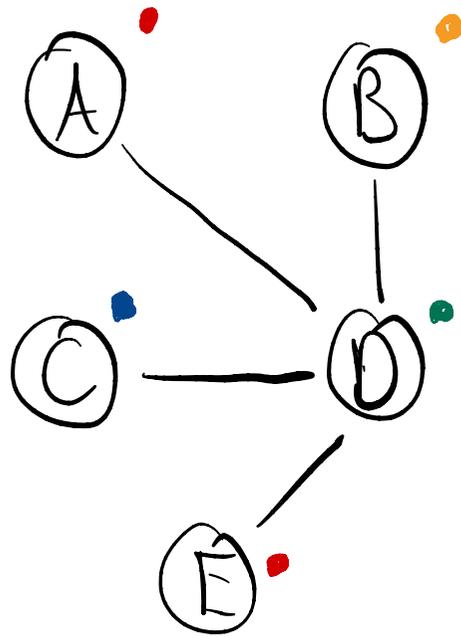


initial (message passing)

A	:	●
B	:	●
C	:	●
D	:	●
E	:	●

memory

# Temporal Graph Neural Networks (TGNN)



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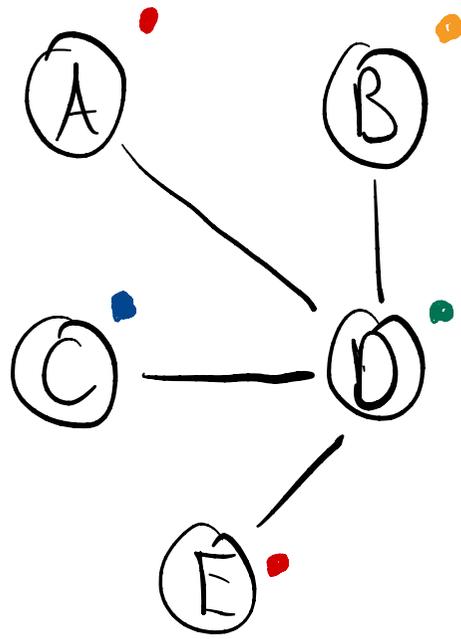
update

$[t_0, \text{add}, (C, E)]$

A	:	•
B	:	•
C	:	• → •
D	:	•
E	:	• → •

memory

# Temporal Graph Neural Networks (TGNN)



initial (message passing)

update

init

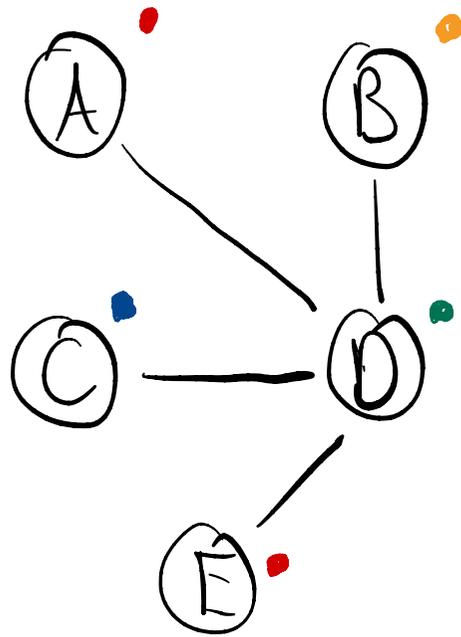
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$[t_1, \text{add}, F]$

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B	:	●
C	:	●
D	:	●
E	:	●
F	:	●

memory

# Temporal Graph Neural Networks (TGNN)



initial (message passing)



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B	:	•
C	:	•
D	:	•
E	:	•
F	:	•

update

$[t_0, \text{add}, (C, E)]$

init

$[t_1, \text{add}, F]$

memory

continuous updates

$[t_2, \text{add}, (B, F)]$

...



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Usual tasks:

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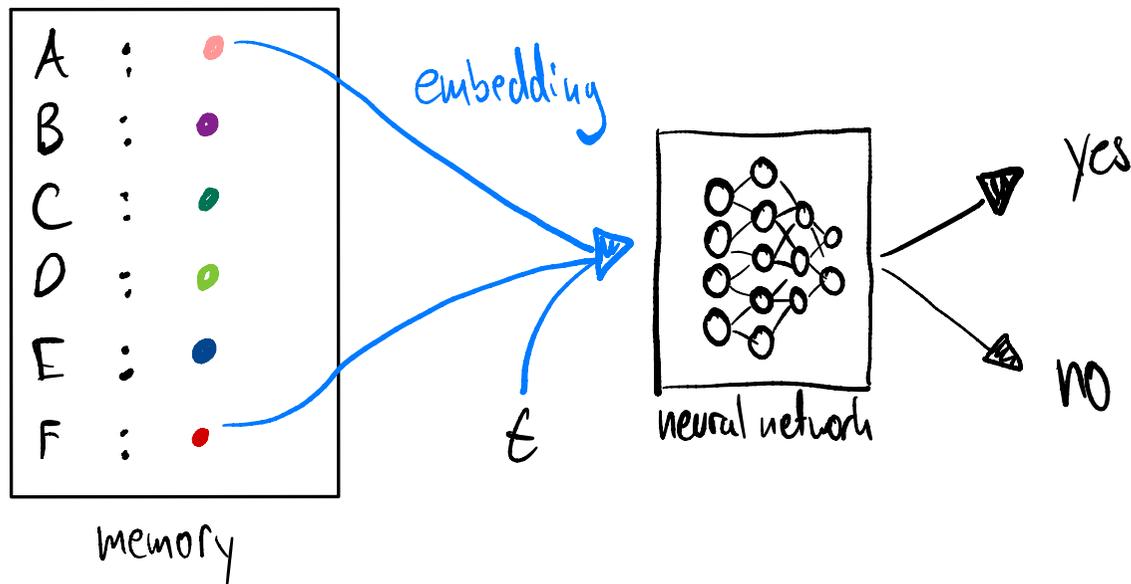
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- future link/edge prediction:  
given CTG  $(G, O)$  and nodes  $u, v$  present at time  $t_0$ :  
decide whether  $(u, v)$  edge is present at time  $t \geq t_0$

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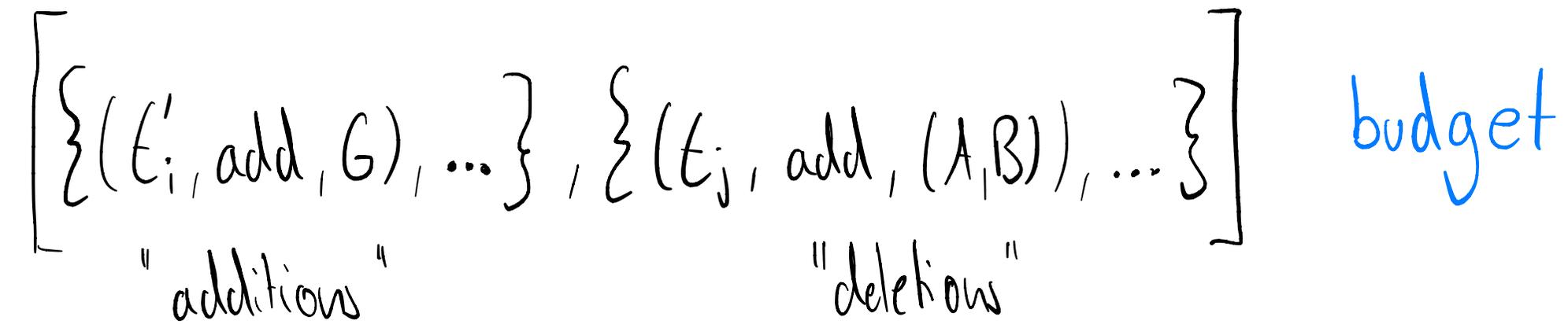
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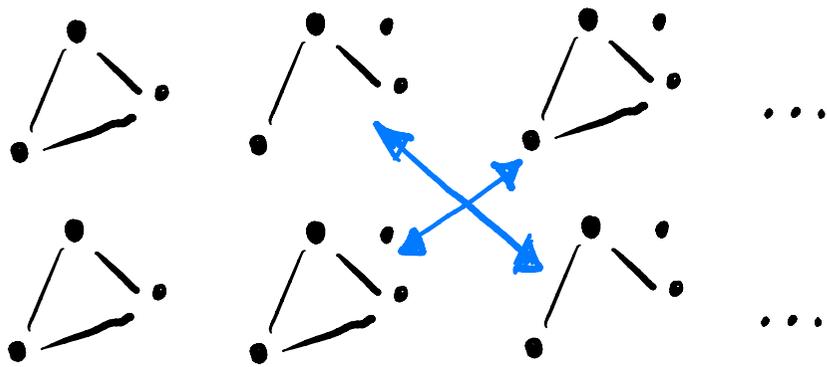
What about Robustness of TGM?

# Point-wise Robustness for Link Prediction over CTG



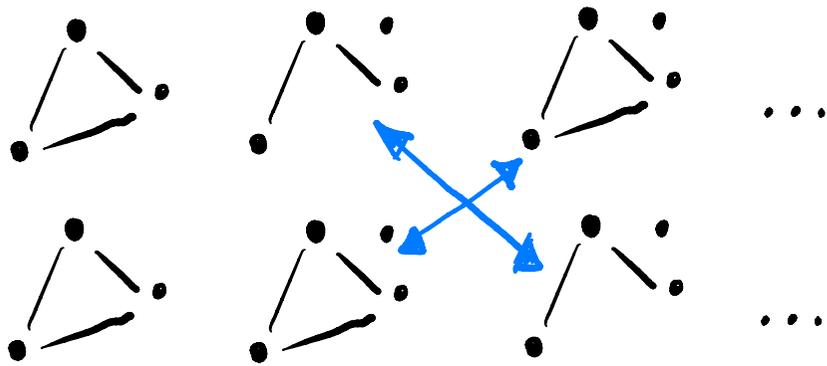
We say that TGMN  $\mathcal{N}$  is robust for nodes  $u, v$  at time  $t$  regarding  $P$  if for all  $C' \in P \dots$

# Some Thoughts on Similar CTG



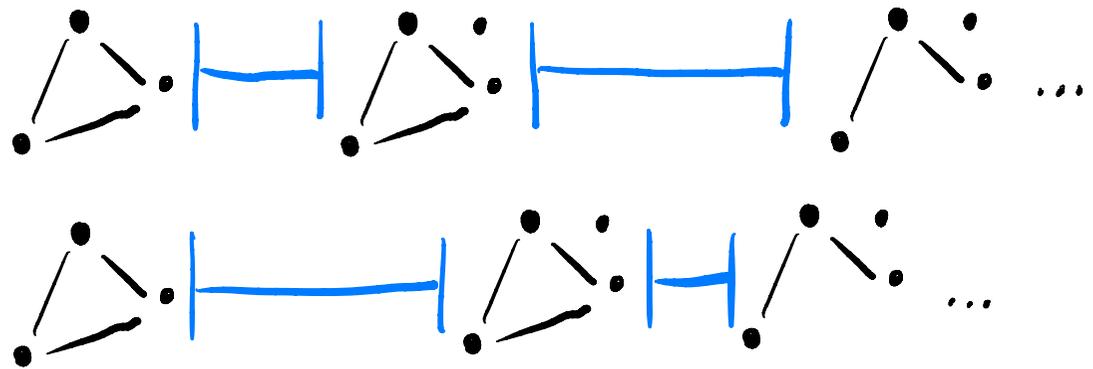
rotations

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rotations

time shifts



⋮

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Thanks!

