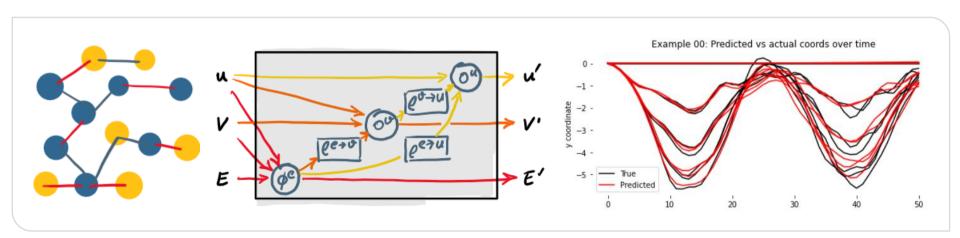




Data Driven Engineering II: Advaced Topics

Graph Neural Networks II

Institute of Thermal Turbomachinery Prof. Dr.-Ing. Hans-Jörg Bauer



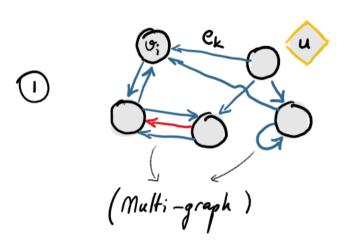
Graph Neural Networks:

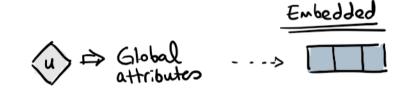


- 1) GNN Basics
- 2) How GNN works
- 3) Basic architectures
- 4) Coding: Graph Nets library











- u is for the whole graph \Rightarrow label, soraneter (\vec{q}) ...
- $V = \{ v_i \}_{i=1,N^o}$

•
$$\epsilon = \begin{cases} e_k, r_k, s_k \end{cases}$$

$$v_i \Rightarrow \text{"particle } i, \Rightarrow C \times, y, z$$
 u, v, w
 m

$$e_k \Rightarrow Edge attribute$$
 $e_k \Rightarrow receiver index$

Understanding Graph Network:



*
$$h_{i}^{(k \uparrow l)} = P_{update} \begin{pmatrix} h_{A}, P_{aggregate} \begin{pmatrix} k \\ k \end{pmatrix}, \forall k \in \mathcal{N}(i) \end{pmatrix}$$

arbitrary

differentiable

functions

message from

functions

Understanding Graph Network:

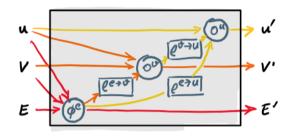


- * Algorithm of a graph network

- edges
 nodes
 nodes
 global
 (i) 3 update functions = edge
 nodes
 global
 (ii) 3 aggregate functions = globalo
- Update edge affributes $e'_{k} = \phi^{e}(e_{k}, V_{r_{k}}, V_{s_{k}}, u)$
- Aggregate edge att. per node $\overline{e}_i' = \ell^{e \rightarrow v} \left\{ (e_k', r_k, s_k) \right\}_{k=i, k=1: N^e}$

Graph Neural Networks





U whid.

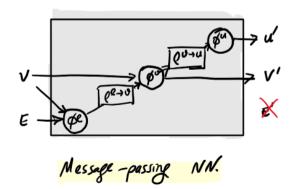
V Vhid.

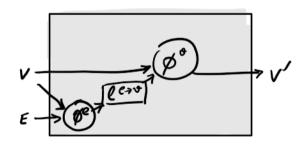
E Ehid

E Lid

Full GN block

Independent recurrent block





Non-local N.N.



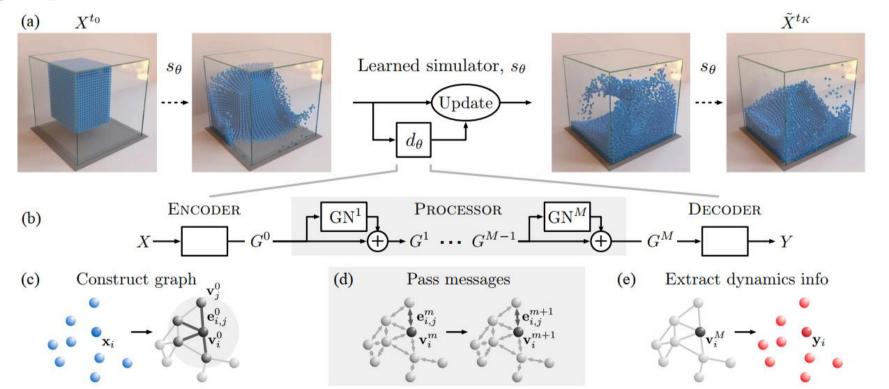
Learning to Simulate Complex Physics with Graph Networks

Alvaro Sanchez-Gonzalez * 1 Jonathan Godwin * 1 Tobias Pfaff * 1 Rex Ying * 1 2 Jure Leskovec 2 Peter W. Battaglia 1

Ground truth	Prediction	Ground truth	Prediction

Graph Neural Networks





Data for training



WaterDrop Water Sand Goop MultiMaterial RandomFloor WaterRamps SandRamps FluidShake FluidShakeBox Continuous WaterDrop-XL Water-3D Sand-3D Goop-3D

Check
download_dataset.sh

* "WaterDrop"





colab

