



GNNforMD

Una GNN per la predizione delle forze per la dinamica molecolare

Generazione dati

units real
atom_style atomic
boundary p p p

pair_style lj/cut 10.0
mass 1 39.948
pair_coeff 1 1 0.2378 3.405 10

region box block -13.14 13.14 -13.14 13.14 -13.14 13.14
create_box 1 box

variable T equal 90
variable P equal 3.0
variable N equal 400

create_atoms 1 random \$N 42 NULL overlap 2.0 maxtry 50
velocity all create \$T 42 dist gaussian

timestep 1
thermo 1000
fix 2 all npt temp 1 \$T 1.0 iso \$P \$P 1000.0

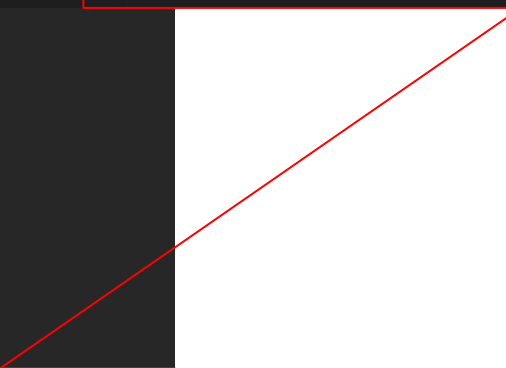
run 20000 # for equilibration
dump mydmp all custom 100 data\argon.lammpstrj id type x y z fx fy fz
run 9999

Input script

data > argon.lammpstrj

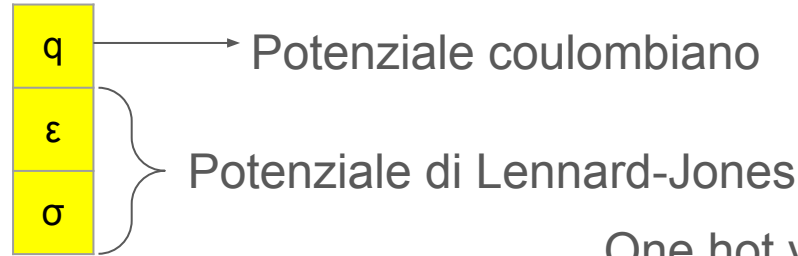
1 ITEM: TIMESTEP
2 20000
3 ITEM: NUMBER OF ATOMS
4 400
5 ITEM: BOX BOUNDS pp pp pp
6 -1.3559535422117442e+01 1.3559535422117442e+01
7 -1.3559535422117442e+01 1.3559535422117442e+01
8 -1.3559535422117442e+01 1.3559535422117442e+01
9 ITEM: ATOMS id type x y z fx fy fz
10 337 1 -10.8427 -10.4041 -9.6201 0.111551 -0.694874 0.816131
11 80 1 -11.4759 -6.92209 -8.1658 -1.2953 -1.40647 -0.520197
12 179 1 -12.6327 -12.299 -12.4528 -0.0491539 0.00615476 -0.348215
13 50 1 -10.5481 -7.81052 -11.9976 -0.829663 -0.00326355 -1.20126
14 1 1 -6.36677 -12.542 -10.6472 0.0330129 -0.151112 -0.0707339
15 89 1 -3.40216 -7.68648 -9.35792 -0.28323 -0.333796 0.509244
16 231 1 -6.15924 -10.0801 -7.77694 0.26985 0.0405483 -1.35934
17 175 1 -3.01399 -11.7082 -12.8122 0.328601 0.275039 0.126978

Output file



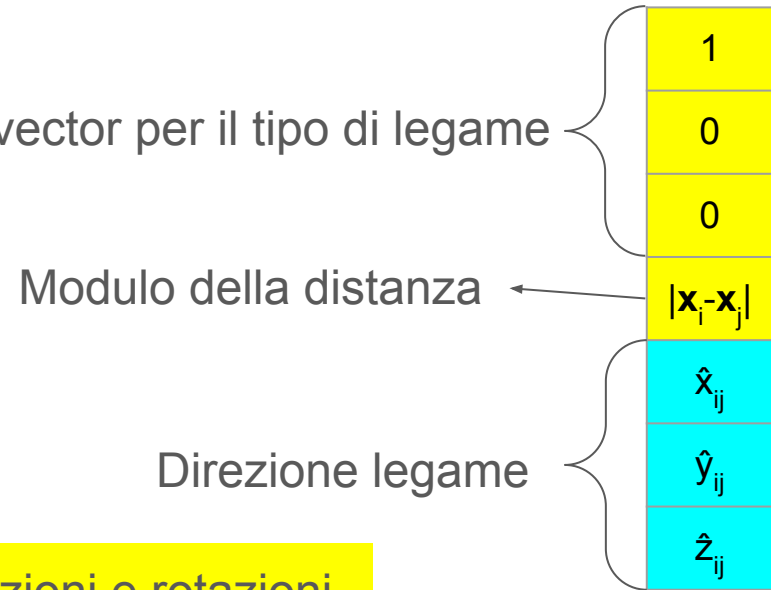
Elaborazione dati

Nodes features

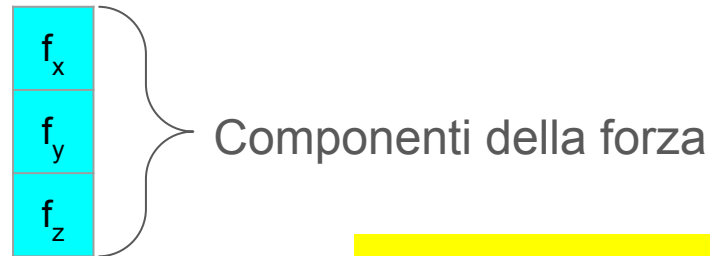


One hot vector per il tipo di legame

Edge features



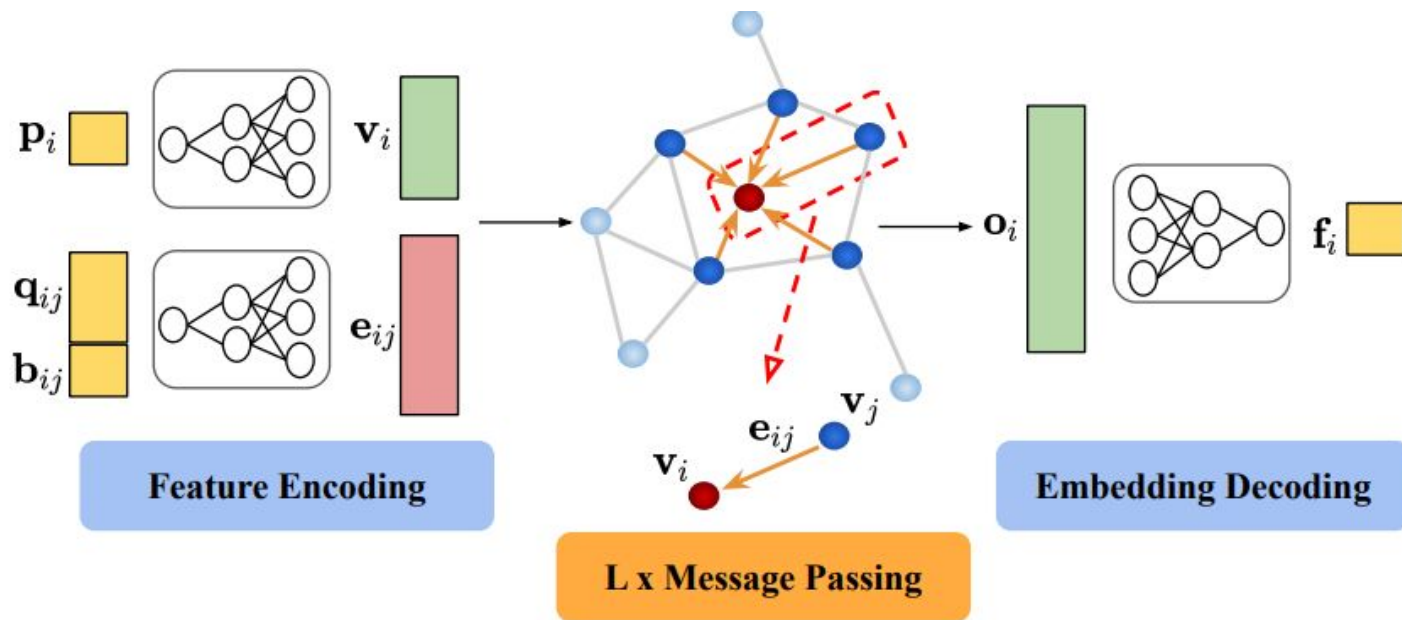
Target (node-level)



Invariante per traslazioni e rotazioni

Invariante per traslazioni e equivariante per rotazioni

Architettura



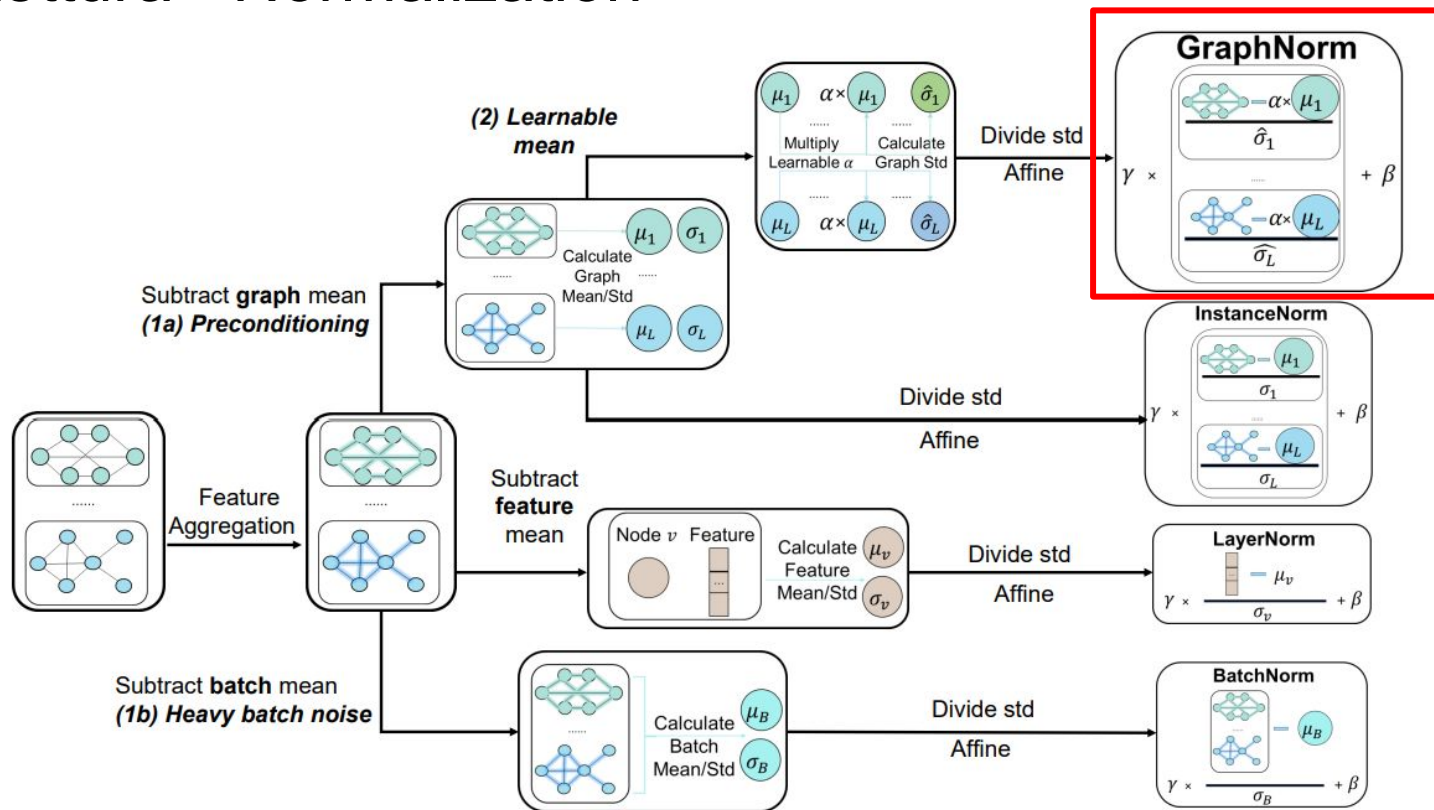
<https://arxiv.org/pdf/2112.03383>

Architettura - Message Passing

$$\mathbf{x}'_i = \text{MLP} \left(\sum_{j \in \mathcal{N}(i)} \{ \text{MLP} ((\mathbf{x}_i * \mathbf{x}_j) \parallel \mathbf{e}_{j,i}) \} \right)$$

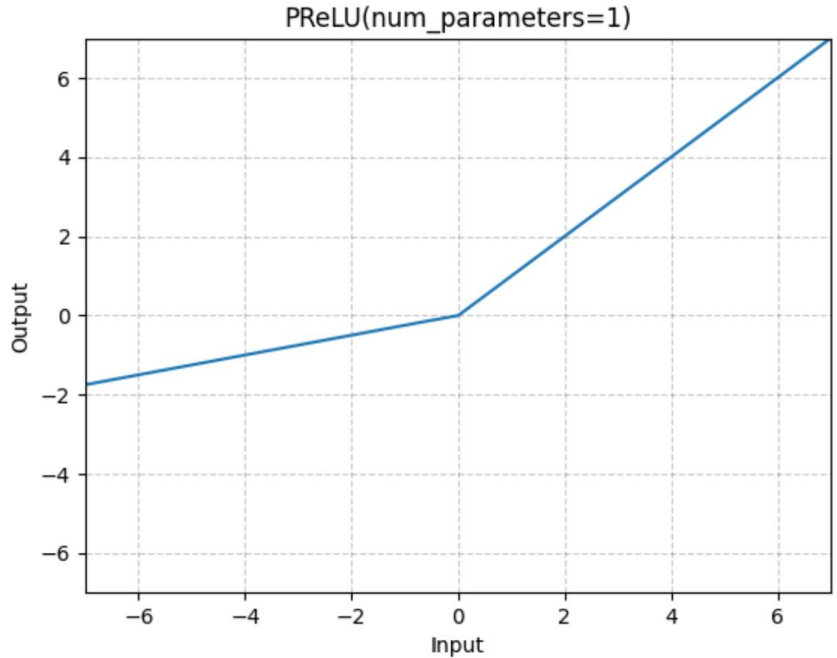
- Solo il primo layer usa informazioni sul legame
- Secondo e terzo layer usano residual connection

Architettura - Normalization



Architettura - Activation function

L'utilizzo della PReLU (Parametric Rectified Linear Unit), assieme ad alcune epoche di warm-up, ha risolto il problema dei *dead neurons* alla fine della rete



Training

- 5 epoche di warm-up con una *loss function* modificata per favorire output diversi da 0
- *Learning rate* iniziale uguale a 10^{-3} e poi decadimento esponenziale
- Salvataggio di un checkpoint ogni sei epoche
- *Data augmentation* con rotazioni casuali 3D durante il training
- Tracking dell'errore lungo ognuna delle tre coordinate
- Miglior modello scelto a posteriori

≡ training_log.txt

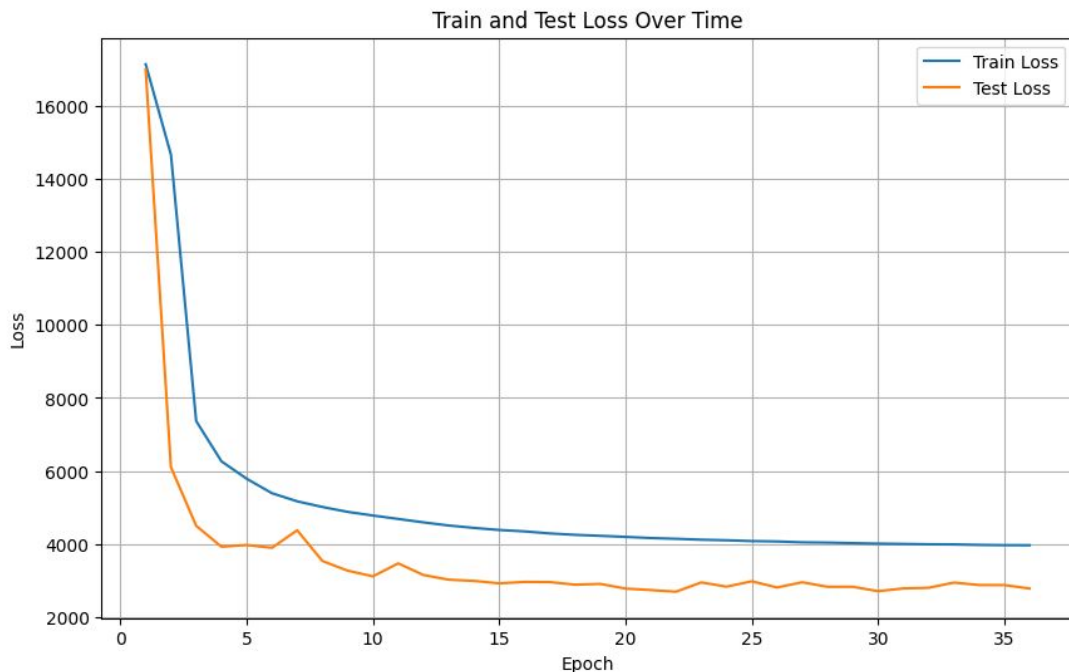
```
1 Epoch: 01, Train Loss: 17136.9064, Test Loss: 16993.3735, LR: 2000.00*10^(-7)
2 Losses: x: 5668.9822, y: 5674.6931, z: 5650.0226
3 Epoch: 02, Train Loss: 14662.1705, Test Loss: 6105.3166, LR: 4000.00*10^(-7)
4 Losses: x: 2052.6152, y: 2063.3387, z: 1989.3783
5 Epoch: 03, Train Loss: 7368.2516, Test Loss: 4500.1855, LR: 6000.00*10^(-7)
6 Losses: x: 1485.8003, y: 1517.2870, z: 1497.1134
7 Epoch: 04, Train Loss: 6264.9336, Test Loss: 3926.5310, LR: 8000.00*10^(-7)
8 Losses: x: 1255.5848, y: 1333.2477, z: 1337.7135
```


Risultati

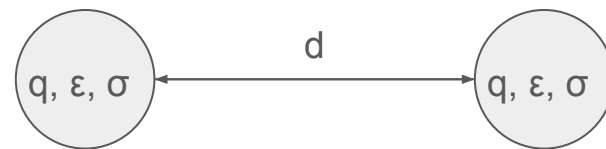
Validation error = 4,18 u/atomo ($\approx 20\%$)

Test error (sistema grande di molecole d'acqua) = 4,24 u/atomo

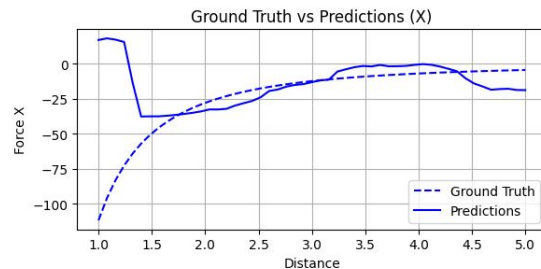
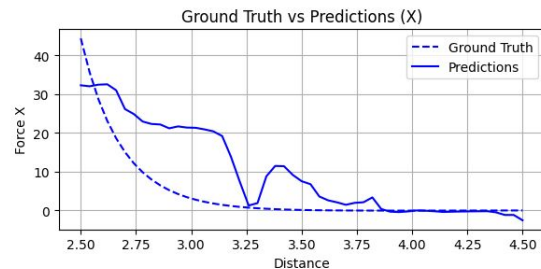
Test error (argon) maggiore della somma dei moduli delle forze!



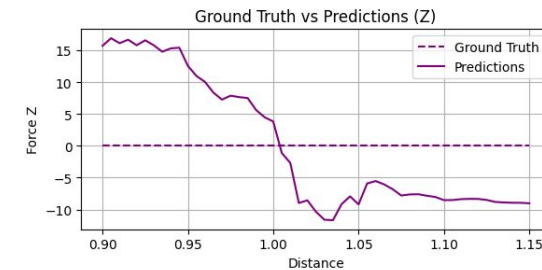
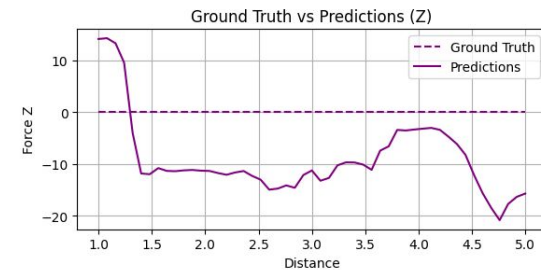
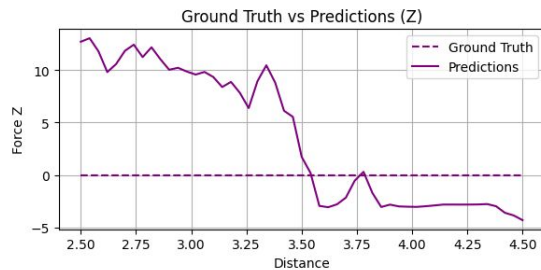
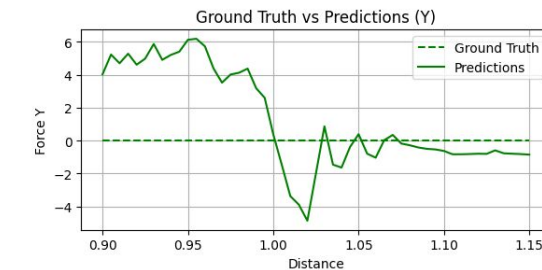
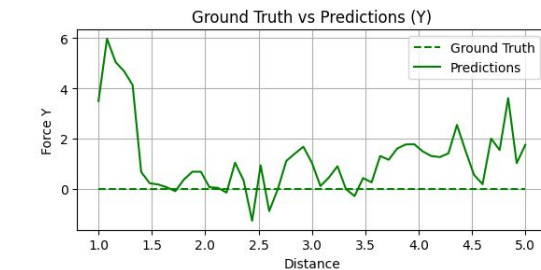
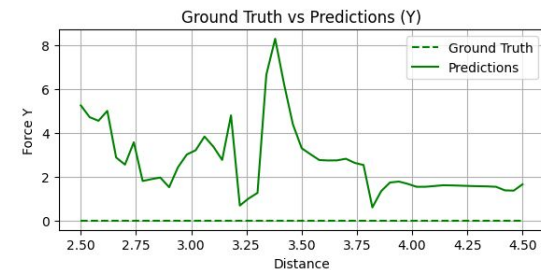
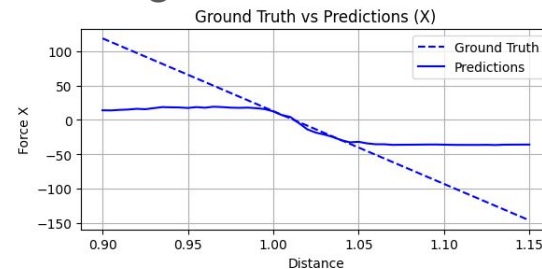
Risultati - Fit delle forze



Lennard-Jones



Legame armonico



Fine-tuning

Un modello davvero generalizzabile richiede dati più vari, ma almeno alcune parti della rete dovrebbe essere simili anche con input diversi (per esempio il decoder e secondo e terzo layer di Message Passing) → Il modello pre allenato sull'acqua “impara” più facilmente l'argon?

