

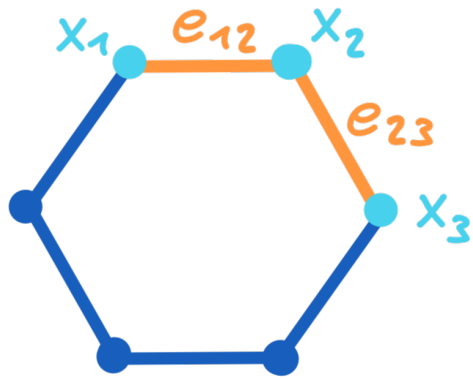
3D Infomax improves GNNs for Molecular Property Prediction

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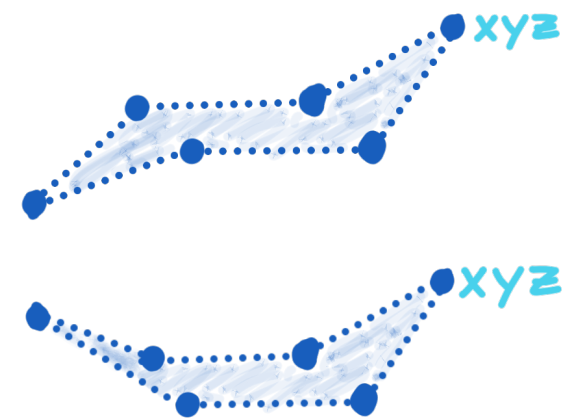
Prudencio Tosson, Christian Dallago,
(Valence Discovery) (Technical University of Munich)

Stephan Günnemann, Pietro Liò
(Technical University of Munich) (Cambridge University)

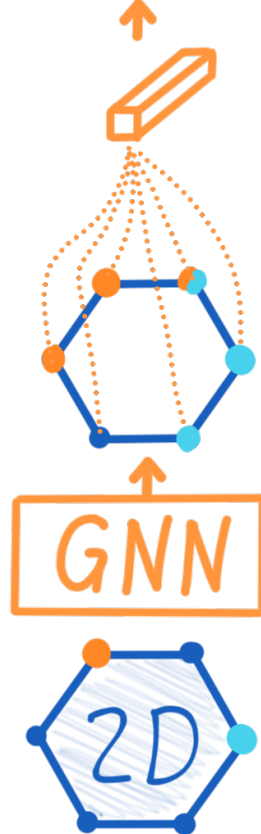
2D
Graph



3D
Conformers

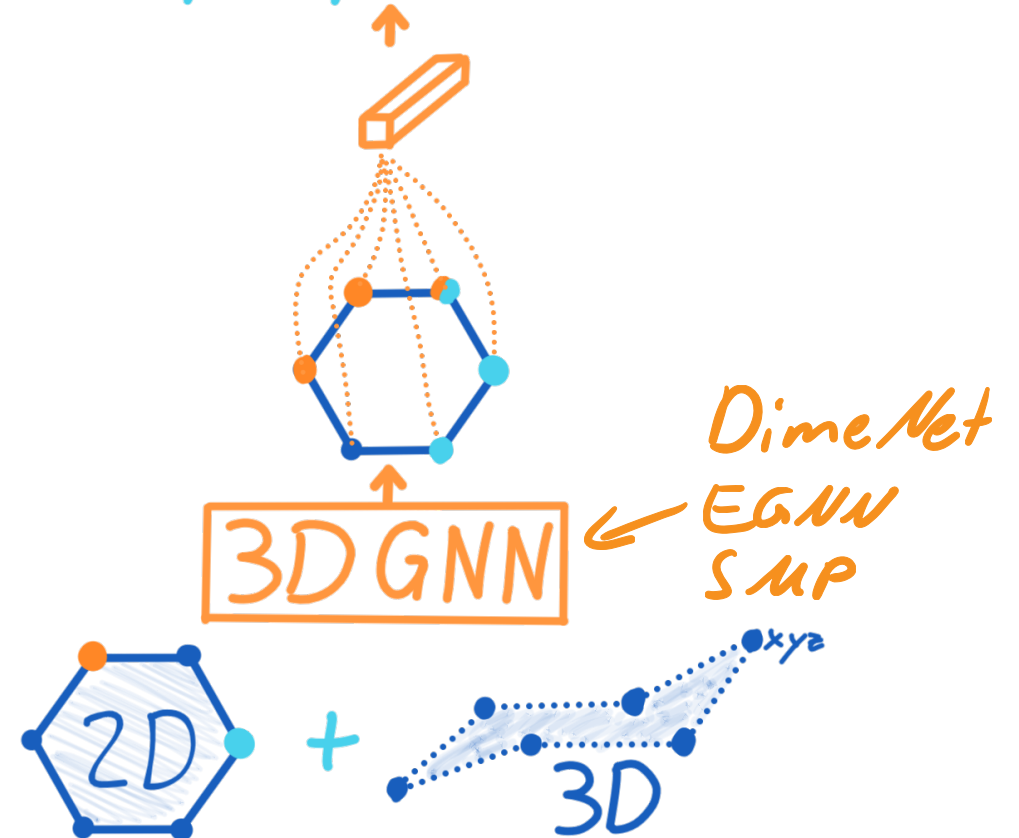


property

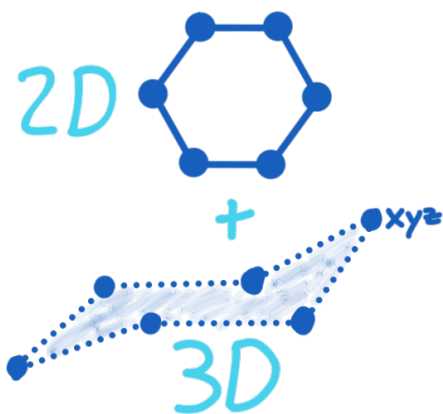
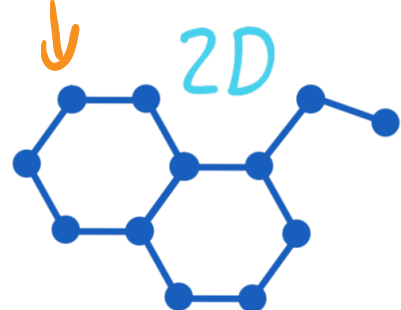


vs.

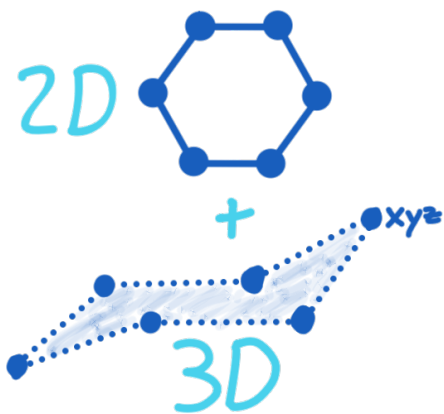
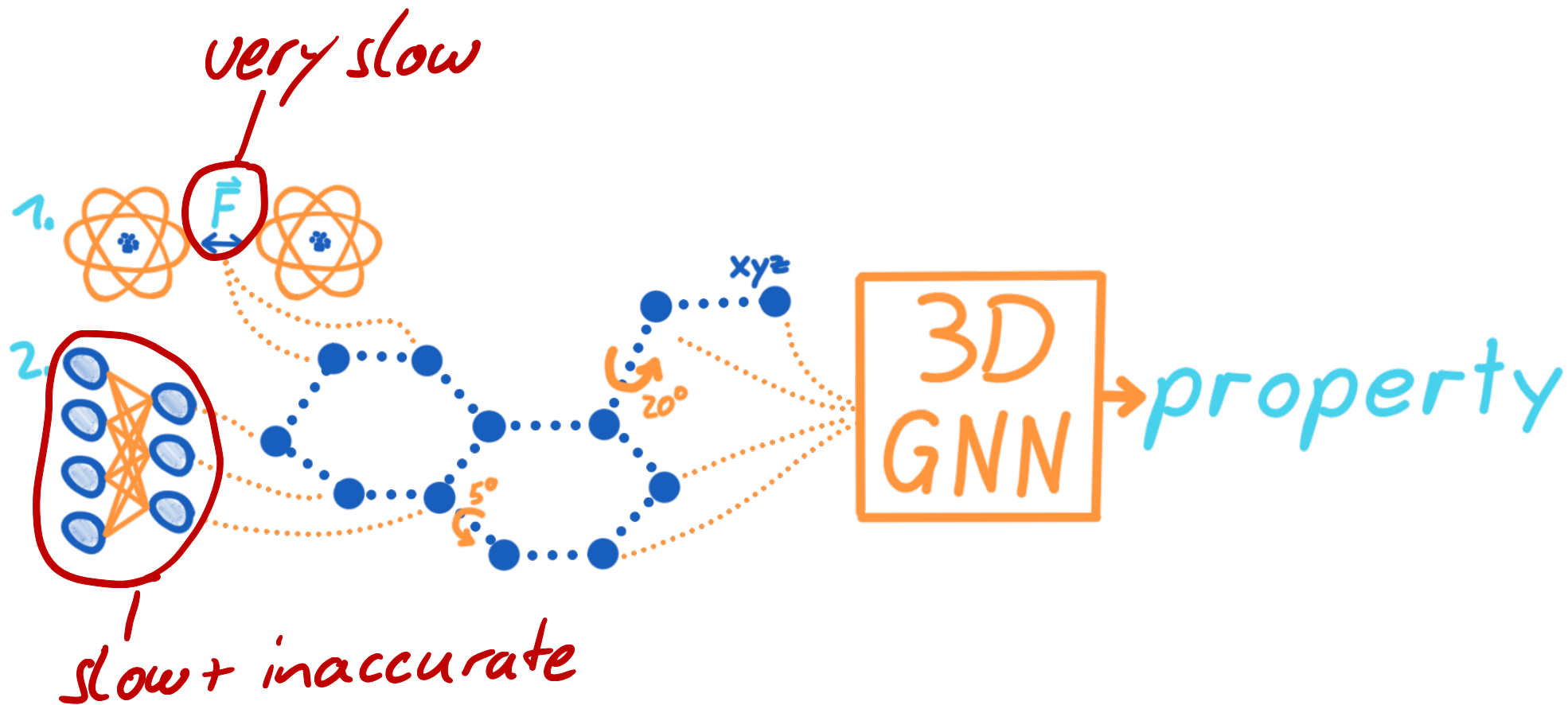
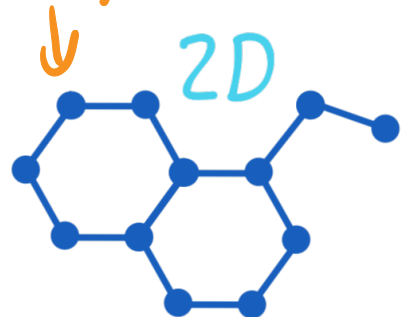
property



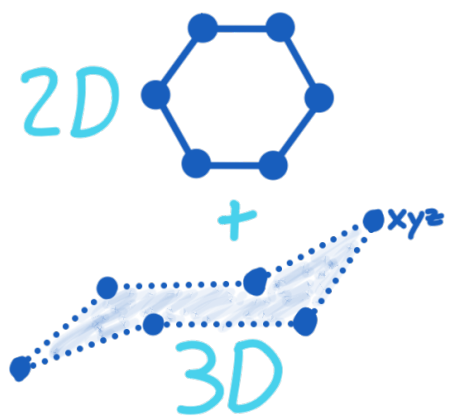
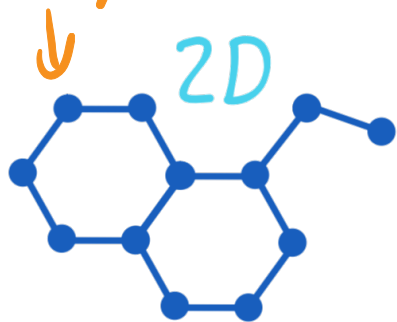
property?



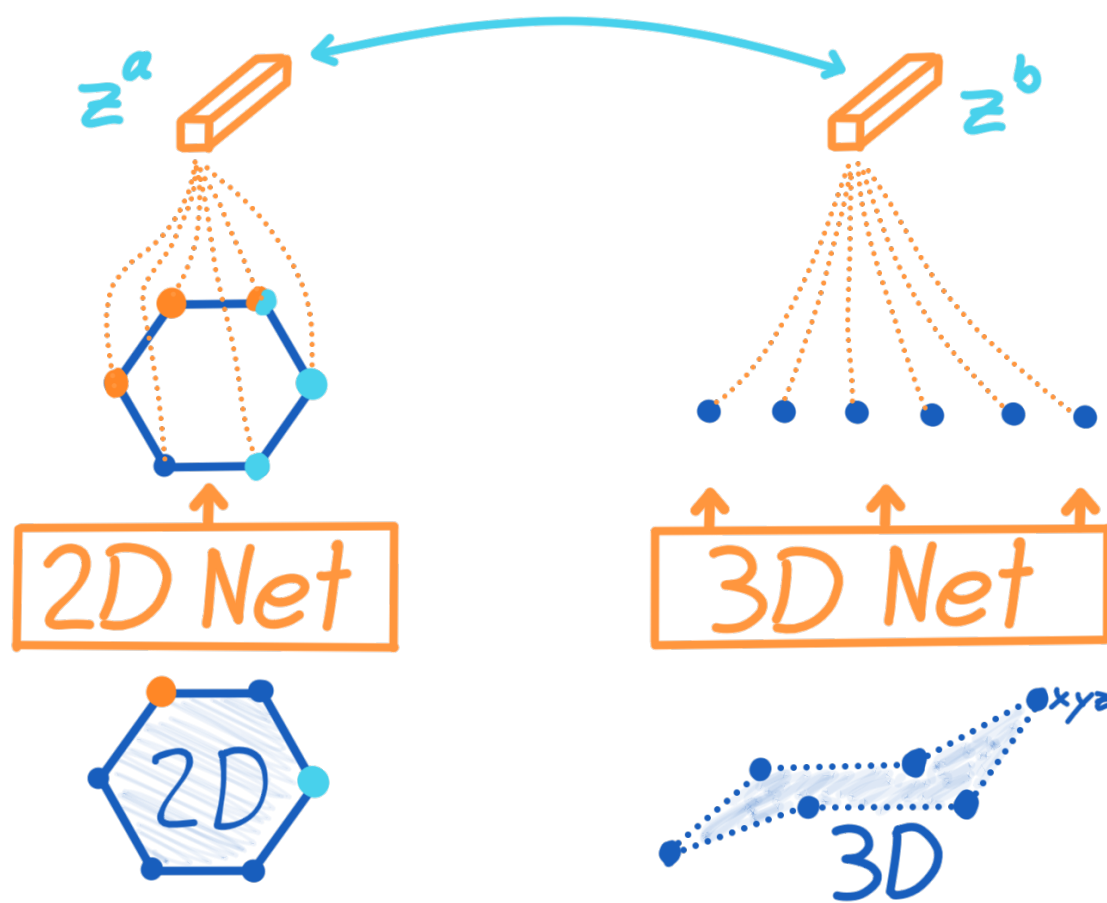
property?



property?

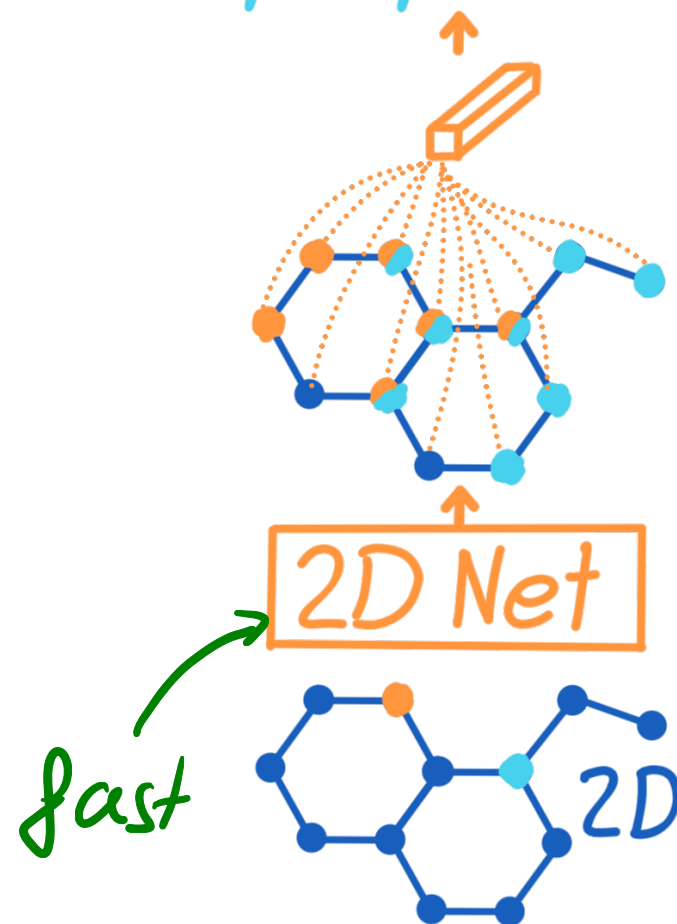


1. maximize MI

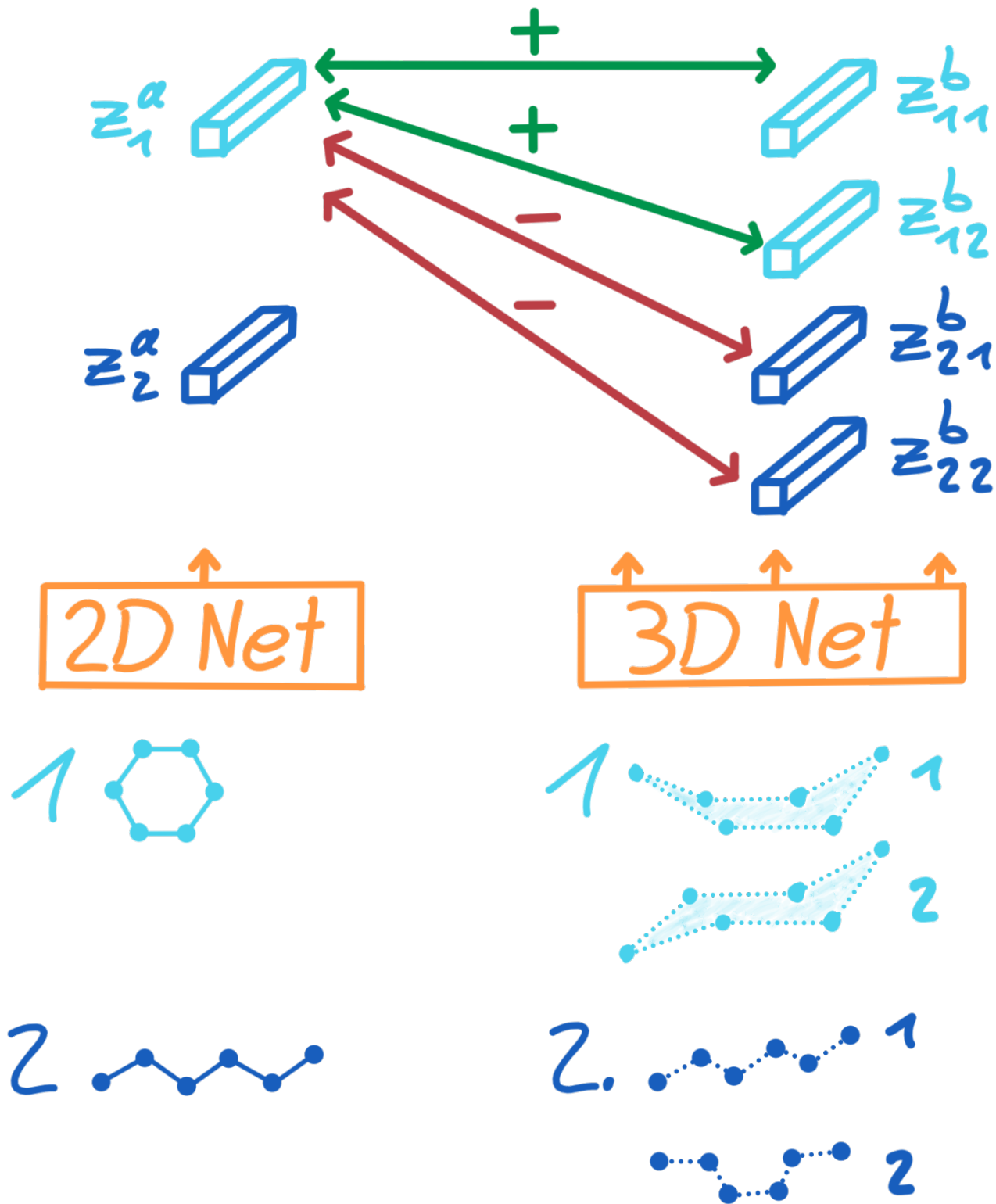


more accurate

2. property



fast



for batchsize N
and c conformers

$$L_i = -\log \frac{\sum_{j=1}^c e^{\text{sim}(z_i^a, z_{i,j}^b)}}{\sum_{k=1}^N \sum_{j=1}^c e^{\text{sim}(z_i^a, z_{k,j}^b)}}$$

pre-training datasets

Target	Rand Init	Pre-training baselines		Our 3D			RDKit SMP	<i>True 3D SMP</i>
		GraphCL	PropPred	QM9	Drugs	QMugs		
μ	0.4133 ± 0.003	0.3937	0.3975	0.3507	0.3512	0.3668	0.4344	0.0726
α	0.3972 ± 0.014	0.3295	0.3732	0.3268	0.2959	0.2807	0.3020	0.1542
homo	82.10 ± 0.33	79.57	93.11	68.96	70.78	70.77	82.51	56.19
lumo	85.72 ± 1.62	80.81	99.84	69.51	71.38	78.10	80.36	43.58
gap	123.08 ± 3.98	120.08	131.99	101.71	102.59	103.85	114.24	85.10
r2	22.14 ± 0.21	21.84	29.21	17.39	18.96	18.00	22.63	1.51
ZPVE	15.08 ± 2.83	12.39	11.17	7.966	9.677	12.06	5.18	2.69
c_v	0.1670 ± 0.004	0.1422	0.1795	0.1306	0.1409	0.1208	0.1419	0.0498