
Appendix for: Chemical Language Meets Geometric Graphs: A Multimodal Fusion Approach for Molecular Properties

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1 Appendix

2 Hyperparameter optimization

- 3 The tables below contain the hyperparameters which were optimized as the adapter layers for the LM (Table A1) and the LM+GNN approach (Table A2).

Hyperparameter	Search Range	Data Type
Dropout Rate	[0.1, 0.5]	Float
Learning Rate	$[10^{-6}, 10^{-1}]$	Float
Number of Layers	[1, 5]	Integer
ExponentialLR Gamma	[0.9, 0.999]	Float

Table A1: LM Adapter search space.

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Hyperparameter	Search Range	Data Type
Hidden Dimension	[50, 150] or [50, 300] (QM9)	Integer
Number of Convolutional Layers	[1, 5]	Integer
Number of Head Layers	[1, 2] or [1, 3] (QM9)	Integer
Number of Shared Layers	[1, 3] or [1, 5] (QM9)	Integer
Dimension of Shared Layers	[32, 100]	Integer
Dimension of Head Layers	[50, 100] (per layer)	Integer (multiple)
ExponentialLR Gamma	[0.9, 0.999]	Float
Learning Rate	$[10^{-6}, 10^{-2}]$ (log scale)	Float

Table A2: GNN and LM+GNN search space. On columns with another search range, the dataset which used the alternative search range is specified in parentheses.

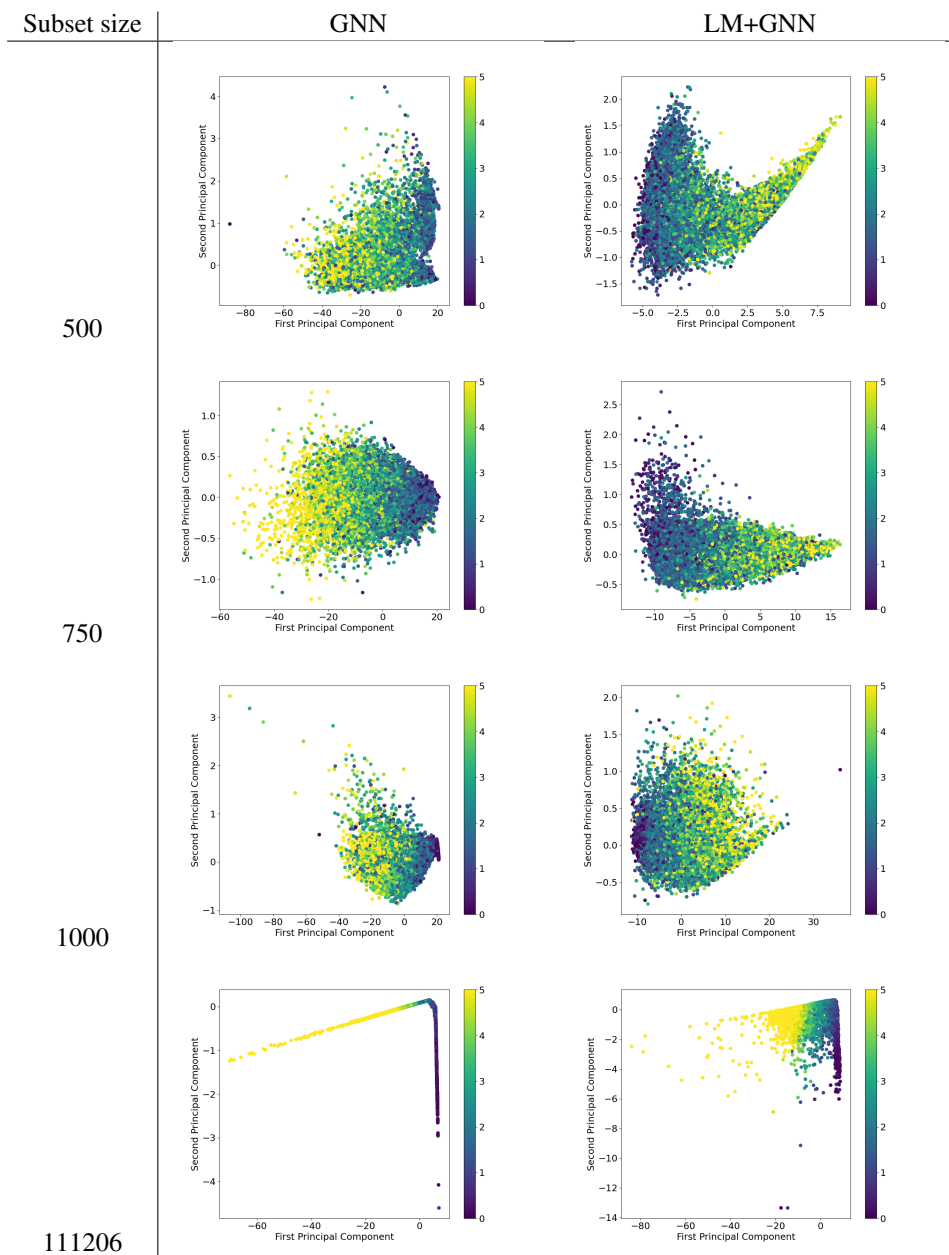


Table A3: Dimensionality reduction of embeddings for GNN and LM+GNN on the QM9 dipole moment task. Uses a randomly selected model from the 8 trials used in the scaling tests and runs PCA on test set embeddings. Colorbar corresponds to the dipole moment of the molecule of the test set.