Neural Graph Matching Improves Retrieval Augmented Generation in Molecular Machine Learning

Runzhong Wang*, Rui-Xi Wang*, Mrunali Manjrekar, Connor W. Coley ICML 2025 https://github.com/coleygroup/ms-pred

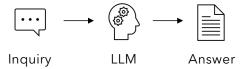




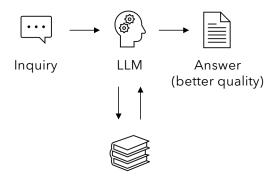


Retrieval Augmented Generation (RAG)

Text generation (LLMs)



with RAG:

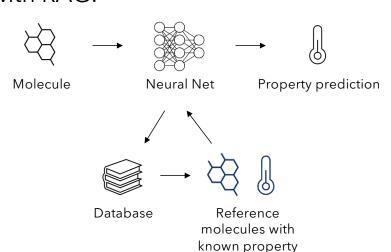


Database (e.g., Bing's search, Wikipedia)

Molecular machine learning (focus of this paper)



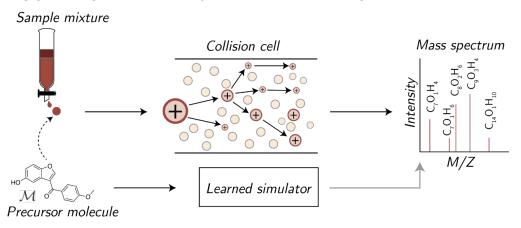
with RAG:



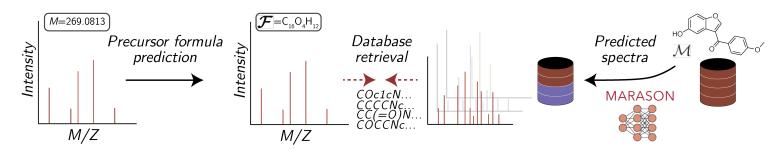


We took mass spectra simulation as the case study

Task: mimic what's happening in the analytical tool, mass spectrometer (MS)



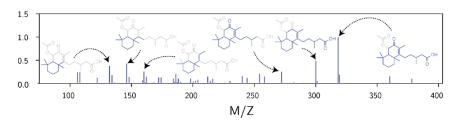
Scientific value: MS simulation models could help structural elucidation



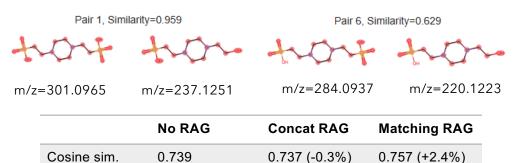


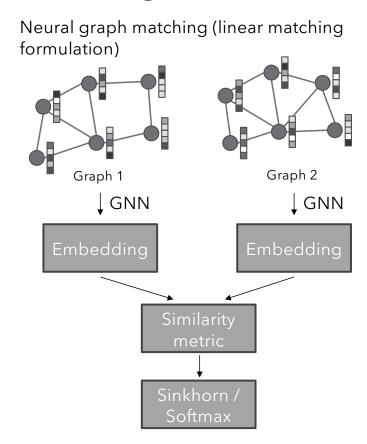
Key insight: matching reference and target molecuels

In mass spectrometry, peaks are attributed to molecular fragments



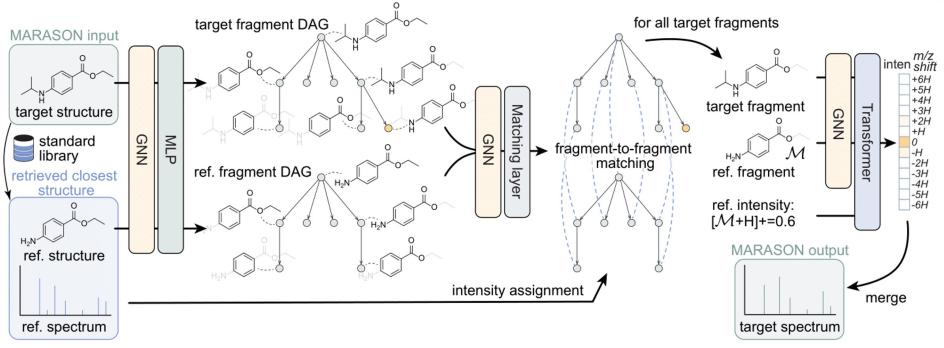
The reference structure might have fragments with distinct masses, and the mapping is not obvious until the molecular graphs are matched







MARASON: RAG with graph matching for MS/MS simulation



MARASON

- RAG query by Tanimoto similarity on the training dataset (could further extend)
- Once target and reference fragments are generated, MARASON uses a nested neural graph matching layer to match them

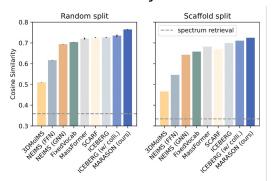


MARASON shows state-of-the-art accuracy on benchmarks

Retrieval accuracy, NIST'20:

| Accuracy @ Top-k | 1 | 2 | 3 | 4 | 5 | 8 | 10 |
|----------------------------------|-----------------|-------------------|-------------------|-------------------|-------------------|-------------------|---------------------|
| Random | 0.026±0.001 | 0.052±0.001 | 0.076±0.002 | 0.098±0.001 | 0.120±0.001 | 0.189±0.003 | 0.233±0.004 |
| 3DMolMS (Hong et al., 2023) | 0.055 ± 0.003 | 0.105 ± 0.000 | 0.146 ± 0.005 | 0.185 ± 0.007 | 0.225 ± 0.009 | 0.332 ± 0.005 | 0.394 ± 0.008 |
| FixedVocab (Murphy et al., 2023) | 0.172 ± 0.004 | 0.304 ± 0.004 | 0.399 ± 0.002 | 0.466 ± 0.007 | 0.522 ± 0.012 | 0.638 ± 0.009 | 0.688 ± 0.006 |
| NEIMS (FFN) (Wei et al., 2019) | 0.105 ± 0.003 | 0.243 ± 0.012 | 0.324 ± 0.013 | 0.387 ± 0.011 | 0.440 ± 0.014 | 0.549 ± 0.010 | 0.607 ± 0.005 |
| NEIMS (GNN) (Zhu et al., 2020) | 0.175 ± 0.005 | 0.305 ± 0.003 | 0.398 ± 0.002 | 0.462 ± 0.004 | 0.515 ± 0.005 | 0.632 ± 0.007 | 0.687 ± 0.005 |
| MassFormer (Young et al., 2024a) | 0.191 ± 0.008 | 0.328 ± 0.006 | 0.422 ± 0.004 | 0.491 ± 0.002 | 0.550 ± 0.005 | 0.662 ± 0.005 | 0.716 ± 0.003 |
| SCARF (Goldman et al., 2023) | 0.187±0.008 | 0.321 ± 0.006 | 0.417 ± 0.007 | 0.486 ± 0.008 | 0.541 ± 0.009 | 0.652 ± 0.008 | 0.708 ± 0.009 |
| ICEBERG (Goldman et al., 2024) | 0.189 ± 0.012 | 0.375 ± 0.005 | 0.489 ± 0.007 | 0.567 ± 0.005 | 0.623 ± 0.004 | 0.725 ± 0.003 | 0.770 ± 0.002 |
| ICEBERG (w/ collion energy) | 0.202±0.009 | 0.399 ± 0.008 | 0.513 ± 0.008 | 0.585 ± 0.008 | 0.639 ± 0.010 | 0.749 ± 0.006 | 0.793 ± 0.007 |
| MARASON (ours) | 0.278±0.002 | $0.455{\pm}0.004$ | 0.562 ± 0.009 | $0.636{\pm}0.006$ | $0.685{\pm}0.004$ | $0.784{\pm}0.002$ | $0.827 {\pm} 0.004$ |

Cosine similarity, NIST'20:



Retrieval accuracy, MassSpecGym:

| Accuracy @ Top-k | 1 | 5 | 20 |
|--------------------------------|------------------------|------------------------|------------------------|
| Precursor m/z | 0.0209 (0.0166-0.0259) | 0.0852 (0.0765-0.0953) | 0.2265 (0.2126-0.2401) |
| NEIMS (FFN) (Wei et al., 2019) | 0.0762 (0.0677-0.0854) | 0.2270 (0.2132-0.2412) | 0.4412 (0.4251-0.4575) |
| NEIMS (GNN) (Zhu et al., 2020) | 0.0363 (0.0305-0.0429) | 0.1355 (0.1246-0.1468) | 0.3377 (0.3226-0.3537) |
| FraGNNet (Young et al., 2024b) | 0.3193 (0.3040-0.3350) | 0.6320 (0.6164-0.6476) | 0.8270 (0.8145-0.8393) |
| MARASON (ours) | 0.3403 (0.3286-0.3520) | 0.6404 (0.6277-0.6519) | 0.8539 (0.8448-0.8624) |

Ablation study, NIST'20:

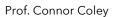
| Base model | RAG strategy | Match layer | Cosine sim. |
|------------------|--------------|-------------|---------------|
| | No RAG | - | 0.739 |
| | Concat | - | 0.737 (-0.3%) |
| MARASON | Hungarian | - | 0.746 (+0.9%) |
| (shared GNN) | RRWM | - | 0.742 (+0.4%) |
| | NGM | Sinkhorn | 0.749 (+1.4%) |
| | NGM | Softmax | 0.753 (+1.9%) |
| MARASON | NGM | Sinkhorn | 0.753 (+1.9%) |
| (not shared GNN) | NGM | Softmax | 0.757 (+2.4%) |
| | | | _ |



Thank you!

MIT Coley Group MS/MS Team







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Code

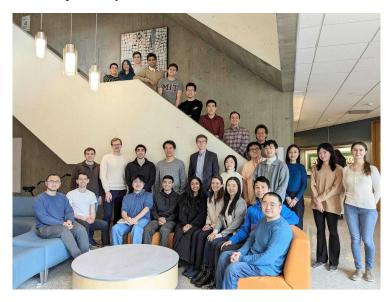




https://github.com/coleygroup/ms-pred

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MIT Coley Group



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Machine Learning for Pharmaceutical Discovery and Synthesis





