ModsNet: Performance-aware Top-k Model Search using Exemplar Dataset

Mengying Wang⁺, Hanchao Ma⁺, Sheng Guan⁺, Yiyang Bian⁺, Haolai Che⁺,

Abhishek Daundkar*, Alp Sehirlioglu*, Yinghui Wu⁺

⁺Depart. of Computer and Data Science, *Depart. of Materials Science and Engineering, CWRU, USA





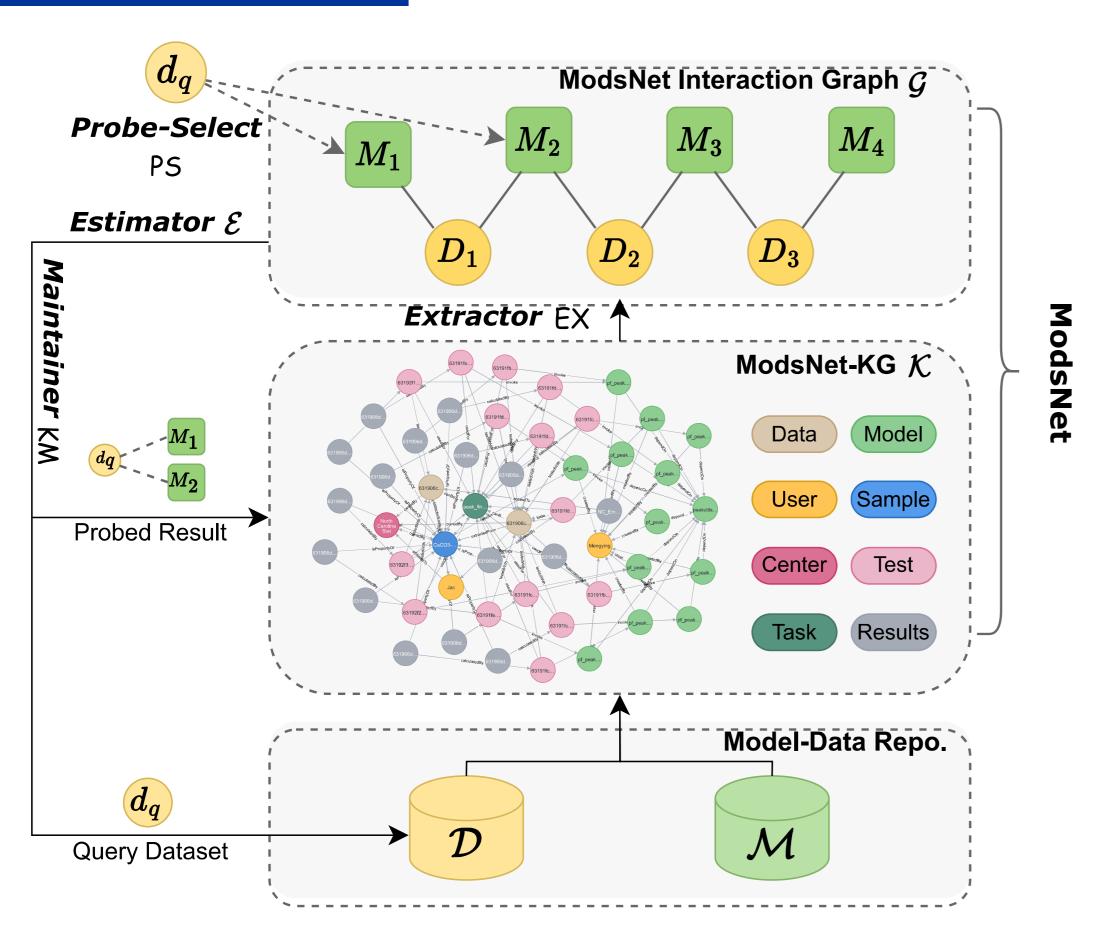
OVERVIEW

Pre-trained models are invaluable resources, enabling efficient reuse and fine-tuning without starting from **Platforms** like HuggingFace gathered abundant models, scripts and associated datasets.

- How can we make these models discoverable?
- Can we search for models using a 'query' dataset?

ModsNet: a performance-aware data engine designed to search for high-quality, underutilized pre-trained models for data-driven analysis.

FRAMEWORK



- Initialization:

- "Cold-starts" with a default knowledge graph K.
- Extract features and construct bipartite graph G.
- Trains an inductive GNN-based estimator ε.

- Model Selection:

- If $d_{\tilde{a}} \in D$, and $d_{\tilde{a}}$ has existing interactions in G, skip probing and use a transductive setting.
- If $d_{\alpha} \notin D$ or no interactions in G, Probe-Selector PS finds promising models, adds probe edges, and consults ε for selection.

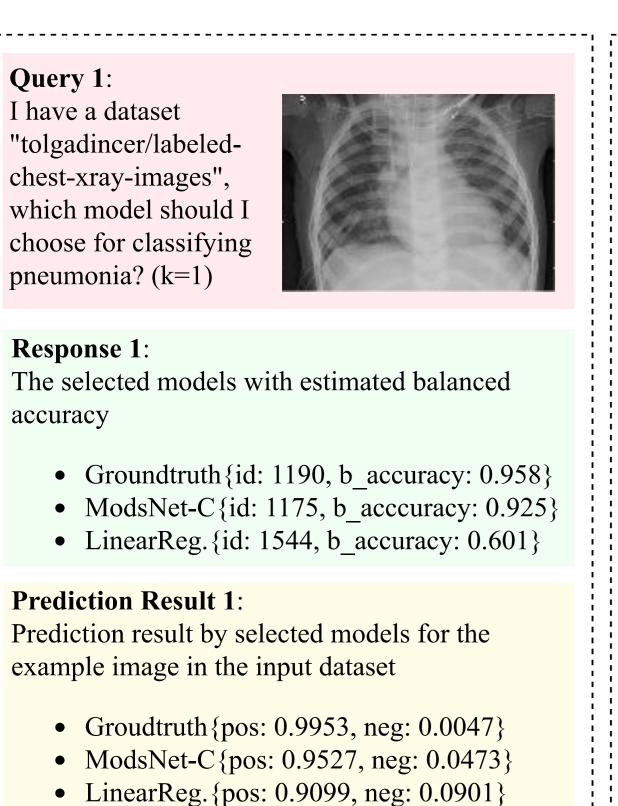
Query 2:

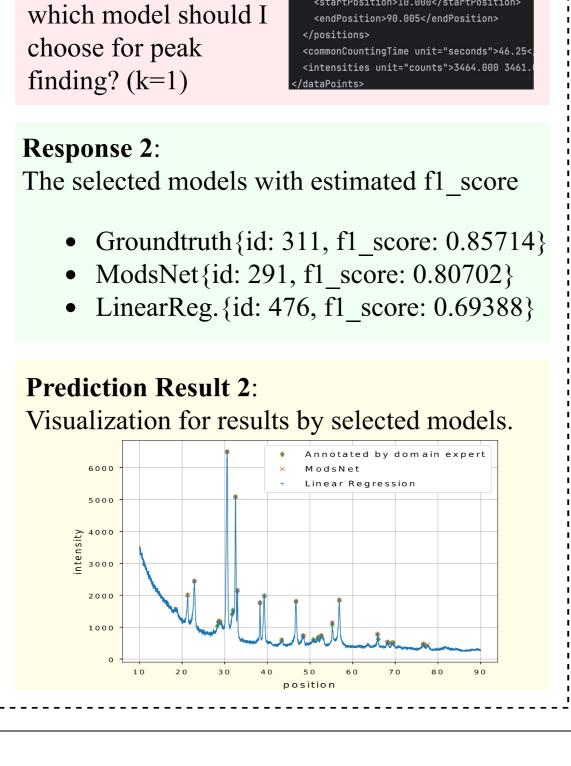
I have a XRDML file,

- Maintenance:

- Maintainer KM logs d_{α} and updates \mathcal{K} and \mathcal{G} .

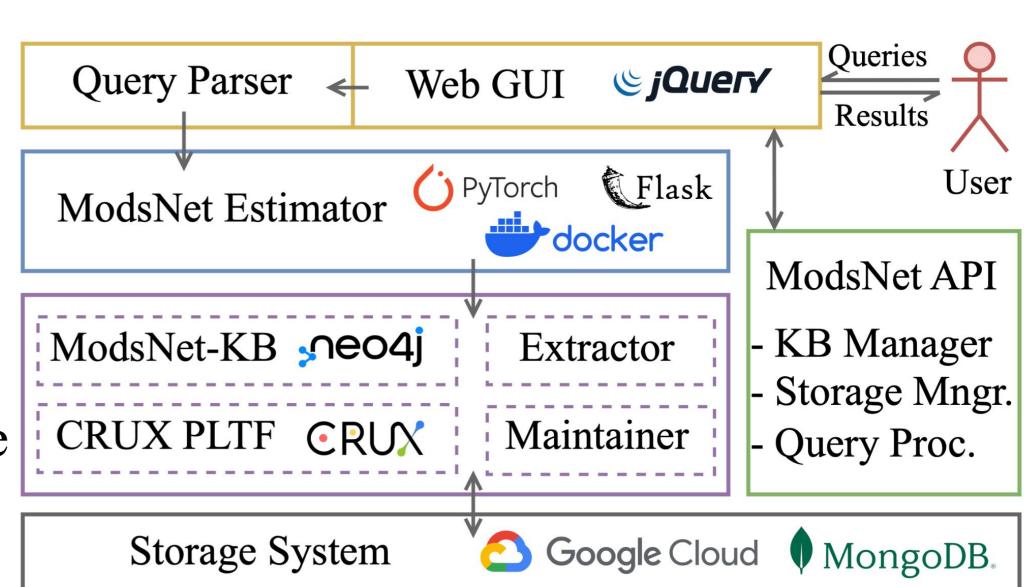
APPLICATIONS





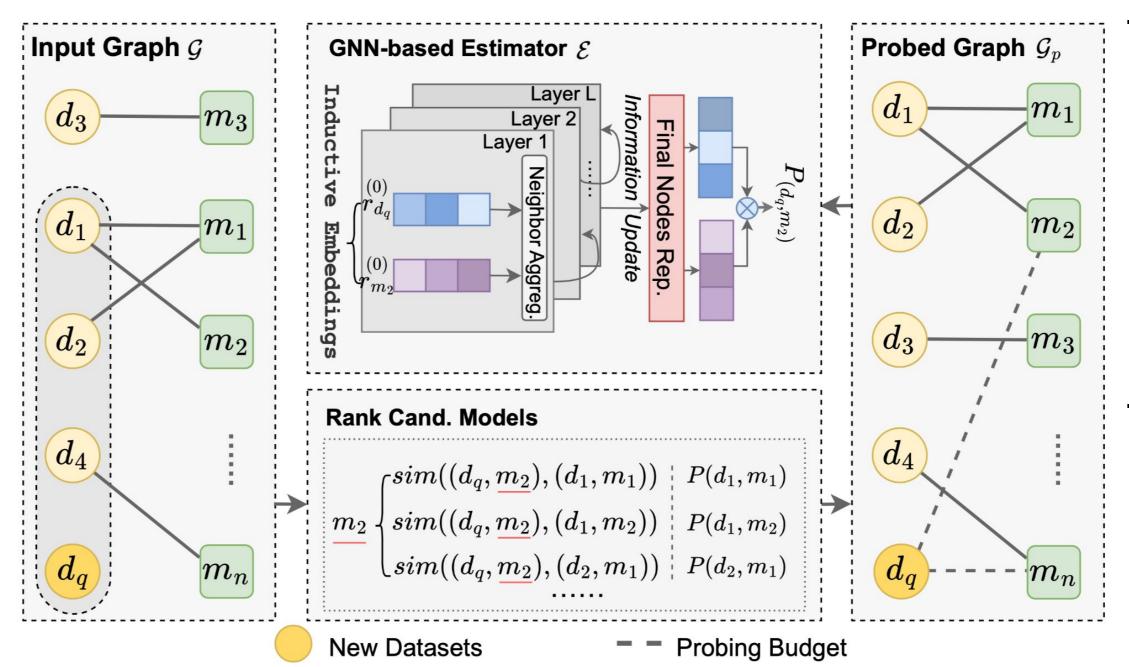
ARCHITECTURE

- **Query Layer:**
- Collects and processes inputs.
- **Recommendation Layer:** Estimates performances and selects top-k models by GNN.
- Knowledge Manag. Layer: Manages the matains knowledge graph and extract features.
- Storage Layer: Stores data and models for quick access.



Four-layers Architecture

PROBE-SELECTOR

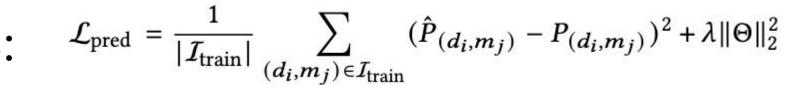


Probe: Samples promising models and inserts virtual probe edges into \mathcal{G} for verification. Scored by:

$$Score(m) = \frac{1}{|I_s|} \sum_{e \in I_s} lsim((d_q, m), e) \cdot P(e)$$

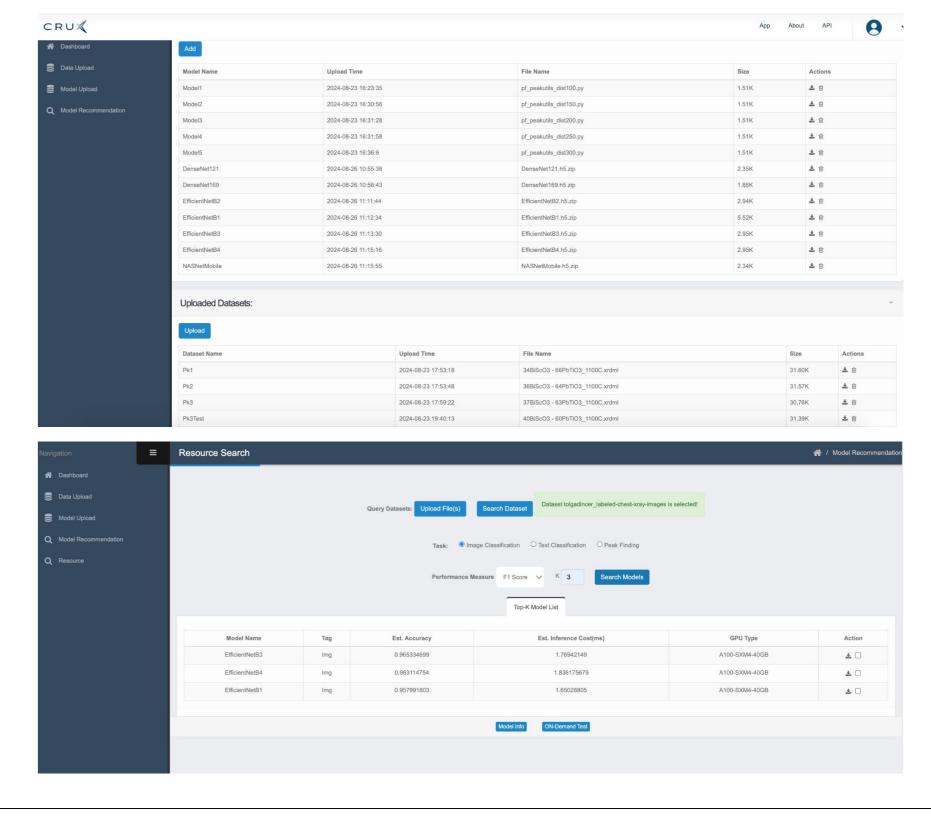
- **Select**: Uses the estimator to predict performance, ranks the models, and returns the top-k models.

Learning objective of ε that minimizes MSE: $\mathcal{L}_{pred} = \frac{1}{|I_{train}|} \sum_{(d_i, m_j) \in I_{train}} (\hat{P}_{(d_i, m_j)} - P_{(d_i, m_j)})^2 + \lambda \|\Theta\|_2^2$



DEMONSTRATION

- Model Libraries. Three built-in libraries (KIZoo, HFZoo, PKZoo) with the option to upload new ones.
- Query by Exemplar Dataset. Users can find promising top-k models without writing complex queries.
- Performance-aware. selects models that outperform baselines in effectiveness and efficiency.
- Visual Performance Analysis. Offer tools for interactive evaluation.



BUILD-IN MODEL LIBRARIES

Dataset	# Models	# Datasets	# Interaction	# Features	Density
PKZoo	462	289	98257	21	0.73591
KIZoo	1800	72	9304	41	0.07179
HFZoo	932	66	974	13	0.01583



kaggle

- Peak-finding models, XRD datasets.
- Crowdsourced from material science community, keep growing.
- Image datasets are collected from Kaggle.
- Self-curated, over 1,000 GPU hours, various CNN architectures.
- Recorded detailed training and testing information.



- Text classifiers, text datasets.
- Crowdsourced from a fast-growing AI community.

CKNOWLEDGEMENT



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ModsNet TEAM

















