

¹ Chemical Recommender System: Replacement Suggestions for Small Molecules

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Software

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⁷ Summary

The Chemical Recommender System (CRS) is an open-source, high-performance toolkit that enables real-time similarity searches across the complete PubChem database (over 50 million molecules) using commodity hardware. The CRS addresses critical limitations in existing chemical informatics platforms through a novel vector database infrastructure, extensible model integration capabilities, and complete algorithmic transparency. The system implements a vector database deployment with partitioned indexing that achieves a ~60x speedup over traditional approaches. A containerized model integration framework allows researchers to seamlessly incorporate custom predictive models into the full-scale search and scoring pipeline, while complete configurability of search parameters, filtering logic, and scoring functions provides capabilities not available in existing black-box solutions. Beyond structural similarity, the CRS integrates OPERA QSAR models for thermophysical and toxicity predictions, RDKit synthetic accessibility scoring, and user-defined models to compute weighted final replacement scores. The complete system is accessible through an interactive web application supporting real-time progress monitoring, post-processing score re-weighting, automated PDF reporting, and batch processing capabilities.

²³ Statement of need

Small molecules frequently become ‘at risk’ or unavailable due to supply, legislative, or technical issues (e.g., the recent discontinuation of PFAS by 3M). In such cases, identifying replacements quickly becomes essential. The CRS has been created to provide a first step in the down-selection of possible replacements for a given target molecule. To use the system, users input a target molecule using SMILES notation or a PubChem identifier through an intuitive web interface, which then performs rapid similarity searches and ranks potential replacements based on multiple criteria.

Chemical recommender and down-selection workflows have been explored in various forms, for example in Recommender Systems for Organic Compounds ([Hayashi et al., 2022](#)), which uses machine learning techniques for candidate classification and ranking. In parallel, machine-learning toolkits such as DeepChem ([Ramsundar et al., 2017](#)) provide a wide range of predictive models for properties such as solubility and toxicity, but do not themselves constitute database-scale search or ranking systems. However, existing solutions face significant limitations: commercial platforms like PubChem and Sigma-Aldrich provide black-box similarity searches without technical transparency, configurability, or extensibility; academic tools typically operate on limited datasets and lack the infrastructure to handle full-scale chemical databases; and most systems do not offer seamless integration of custom predictive models. The CRS addresses these gaps by providing a completely open-source platform with full configurability

42 of search parameters, filtering logic, and scoring functions. The system's plug-and-play Docker
 43 architecture enables researchers to integrate their own models while leveraging the complete
 44 PubChem-scale search infrastructure, creating a unique combination of scale, transparency,
 45 and extensibility not available in existing solutions.

46 Software Overview

47 Architecture Overview

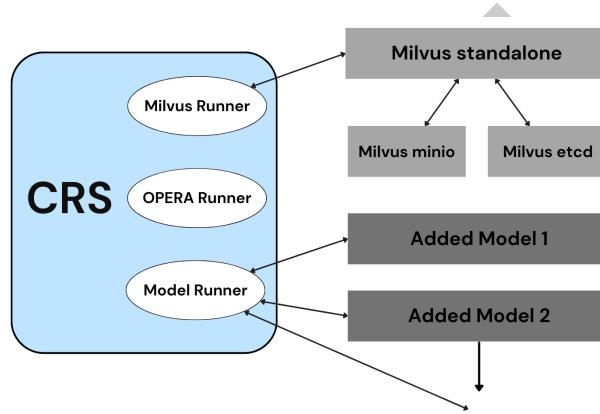


Figure 1: The CRS is a microservices solution that works as a series of containers. They all interact with one another stemming from the CRS using runners.

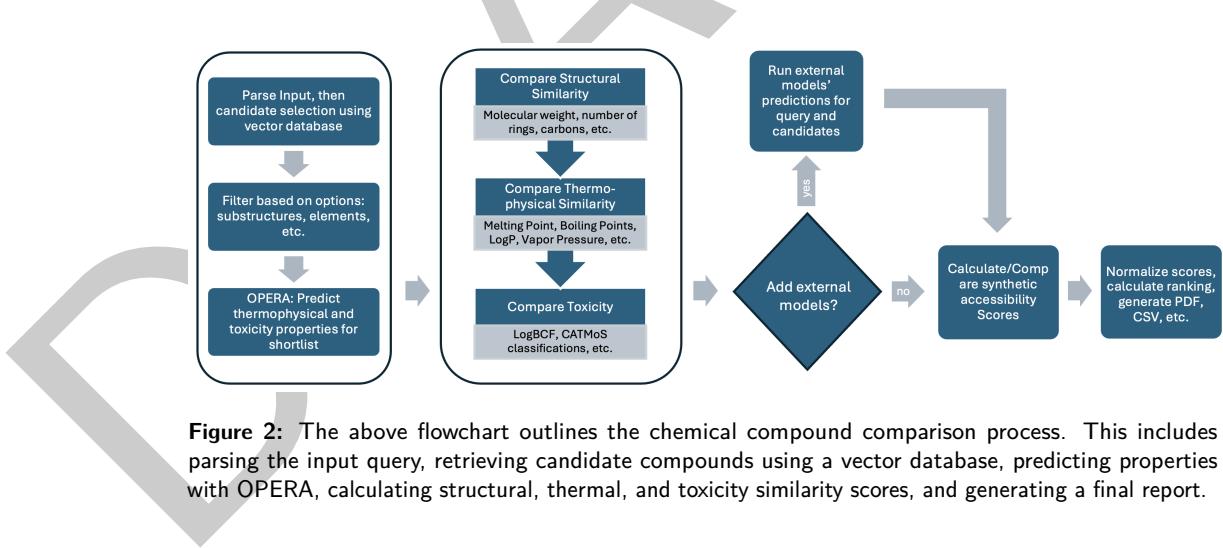


Figure 2: The above flowchart outlines the chemical compound comparison process. This includes parsing the input query, retrieving candidate compounds using a vector database, predicting properties with OPERA, calculating structural, thermal, and toxicity similarity scores, and generating a final report.

48 The CRS is architected as a microservices system deployed via Docker Compose, enabling
 49 users to initialize the complete infrastructure with a single command.

50 Vector Database Infrastructure

51 The CRS implements a sophisticated vector database infrastructure using Milvus ([Wang et al., 2021](#)) to enable real-time similarity searches across the complete PubChem database (over
 52 50 million molecules) on off-the-shelf machines. The system preprocessing pipeline computes
 53 2048-bit Morgan fingerprints ([Cereto-Massagué et al., 2015](#)) of radius 2 for all PubChem CIDs
 54 using RDKit, creating a comprehensive chemical space representation.
 55

56 The production deployment utilizes Milvus in standalone mode, supported by dedicated etcd and
57 MinIO containers for metadata management and object storage, respectively. The PubChem
58 dataset is strategically partitioned into 120 indexed segments, with each partition containing
59 approximately 800,000 molecular fingerprints. Precomputed inverted files for indexing with
60 1024 cluster centroids computed via k-means clustering of centroids in the JACCARD distance
61 space are prebuilt into the CRS image, and is the key to allowing faster similarity searches.
62 Moreover, the partitioning strategy enables parallel processing and memory optimization while
63 maintaining search performance, using a pseudo sliding window of the data in RAM.

64 The CRS allows for the number of partitions searched per batch to be configurable via
65 environment variables, allowing fine-tuning for different hardware configurations. The system
66 uses the JACCARD metric for Tanimoto similarity computation [Equation 1](#), with optimized
67 search parameters including configurable probe values for index traversal efficiency.

$$Tani(F_i, F_j) = \frac{F_i \cdot F_j}{\sum_k F_{i_k} + \sum_k F_{j_k} - F_i \cdot F_j} \quad (1)$$

68 The search algorithm employs a two-stage approximate nearest neighbor approach. First,
69 it identifies the closest cluster centroids to the query vector in JACCARD space. Then, it
70 performs exhaustive similarity calculations within those selected clusters, reducing computational
71 complexity from $O(N)$ to $O(nprobe \times \frac{N}{nlist} + nlist)$ for large-scale searches. An LRU cache
72 stores recent search results keyed by fingerprint hash and result count to minimize redundant
73 database queries. Our experiments demonstrate that this infrastructure enables searches over
74 the whole dataset that complete in minutes rather than hours on standard hardware.

75 Extensible Model Integration

76 The CRS implements a containerized model integration framework using Docker Compose,
77 enabling any SMILES-to-numeric model to plug into the search pipeline with minimal setup.
78 Researchers wrap their model in a Docker container exposing a Flask API, add it to the
79 Compose file with networking and resource constraints, and supply its name to the CRS via
80 CLI. At runtime, the CRS invokes models over HTTP for each candidate, and integrates the
81 predictions—weighted by user-configurable parameters—into the overall ranking.

82 Despite the proliferation of bespoke predictive models in cheminformatics, there remains no
83 simple standardized environment for rigorous testing and benchmarking within full-scale search
84 workflows. Many researchers must invest significant effort to validate container deployments,
85 configure execution pipelines, and harmonize output metrics, hindering reproducibility and
86 comparability. The CRS addresses this gap by providing a transparent framework that unifies
87 model deployment, invocation, and performance evaluation against a live PubChem similarity
88 search pipeline, enabling domain experts to concentrate on model development rather than
89 infrastructure.

90 Open Architecture and Configurability

91 Unlike many current black-box commercial solutions, the CRS provides complete transparency
92 and control over search parameters, filtering criteria, and scoring algorithms, significantly
93 benefiting researchers. The open-source nature of the CRS allows researchers to modify the
94 source code, customize scoring algorithms, and implement domain-specific filters, ensuring that
95 the system can be tailored to meet specific application needs. The modular architecture allows
96 for targeted modifications without affecting other components, while comprehensive logging
97 and monitoring capabilities enhance usability, empowering researchers with the flexibility and
98 control that black-box solutions lack.

99 Multi-Property Scoring and Integration

100 The CRS computes final replacement scores by integrating multiple property-based similarity
 101 metrics through a sophisticated comparison pipeline. The system uses OPERA ([Mansouri et al., 2018](#)) QSAR models to predict five categories of thermophysical properties (melting point,
 102 boiling point, logP, vapor pressure, Henry's law constant), structural descriptors (molecular
 103 weight, ring count, Lipinski failures, carbon count, topological polar surface area), and toxicity
 104 endpoints (BCF, CATMoS EPA categories, LD50). Additionally, RDKit synthetic accessibility
 105 scoring ([Skoraczyński et al., 2023](#)) (ranging from 1=easy to 10=difficult) estimates synthesis
 106 complexity using fragment contribution models trained on PubChem fingerprints. The CRS
 107 normalizes these scores to a 1-10 scale where higher values indicate easier synthesis by applying
 108 the transformation: $SA_{final} = 10 - SA_{raw}$.

110 The scoring algorithm computes similarity metrics by comparing candidate properties against
 111 query molecule properties using normalized difference calculations. Each property category
 112 contributes a similarity score (C_1 through C_5 for structural, molecular weight, thermophysical,
 113 toxicity, and synthetic accessibility respectively), with additional categories (C_{5+n}) generated by
 114 user-provided models. The final replacement score is computed using a weighted multiplicative
 115 model:

$$FRS = \frac{C_1^{W_1} \times C_2^{W_2} \times C_3^{W_3} \times C_5^{W_5} \times \prod_{i=6}^n C_i^{W_i}}{C_4^{W_4}} \quad (2)$$

116 where C_1 represents structural (Tanimoto) similarity, C_2 molecular weight similarity, C_3
 117 thermophysical similarity, C_4 toxicity score (inversely weighted), C_5 synthetic accessibility, and
 118 C_{5+n} represent external model contributions. The weights W_i are user-configurable parameters
 119 that enable domain-specific prioritization of different molecular properties.

120 Prior to final score computation, all similarity metrics undergo Min-Max normalization to
 121 ensure comparable scales across different property types: This transformation maps all scores
 122 to a 1-10 range while preserving relative differences between candidates. The system supports
 123 real-time re-weighting through the web interface, enabling interactive exploration of results
 124 with different prioritization schemes without requiring database re-queries.

125 Advanced Workflow Capabilities

126 The CRS implements comprehensive batch processing functionality that enables high-throughput
 127 chemical analysis workflows. The system provides sophisticated filtering capabilities including
 128 configurable elemental restrictions beyond the default set (H, C, N, O, F, P, S, Cl, Se, Br,
 129 I), SMARTS-based substructure matching with occurrence count requirements, and isotope
 130 handling options. Advanced search parameters support both PubChem CIDs, IUPAC names, and
 131 arbitrary SMILES strings, including hypothetical molecules not present in existing databases.

132 The CRS implements intelligent progress monitoring with real-time status updates via
 133 Server-Sent Events, enabling users to track job progress, identify processing bottlenecks, and
 134 receive immediate feedback on search completion or error conditions. The system generates
 135 comprehensive PDF reports automatically for each query, combining molecular visualizations,
 136 property predictions, similarity scores, and ranking justifications into publication-ready
 137 documentation. Batch jobs produce consolidated reports with cross-query analysis and
 138 statistical summaries.

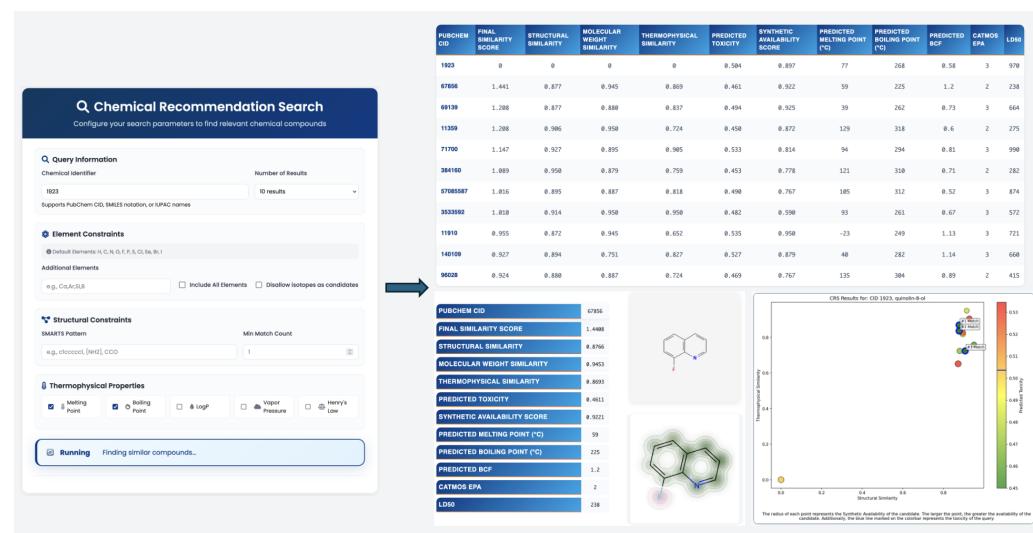


Figure 3: Visualization of CRS output from the web app searching interface, including the query molecule, possible replacements, and (C_1, C_2, C_3, C_4, C_5) for top CRS candidates. Notably, all retrieved candidates preserve the quinoline scaffold and differ only by localized substitutions at the same functional position.

Code Availability

The CRS source code is available for free on Github under the BSD-3-Clause license (<https://github.com/sandialabs/chemical-recommender-system>), and can be used to install and run the CRS.

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