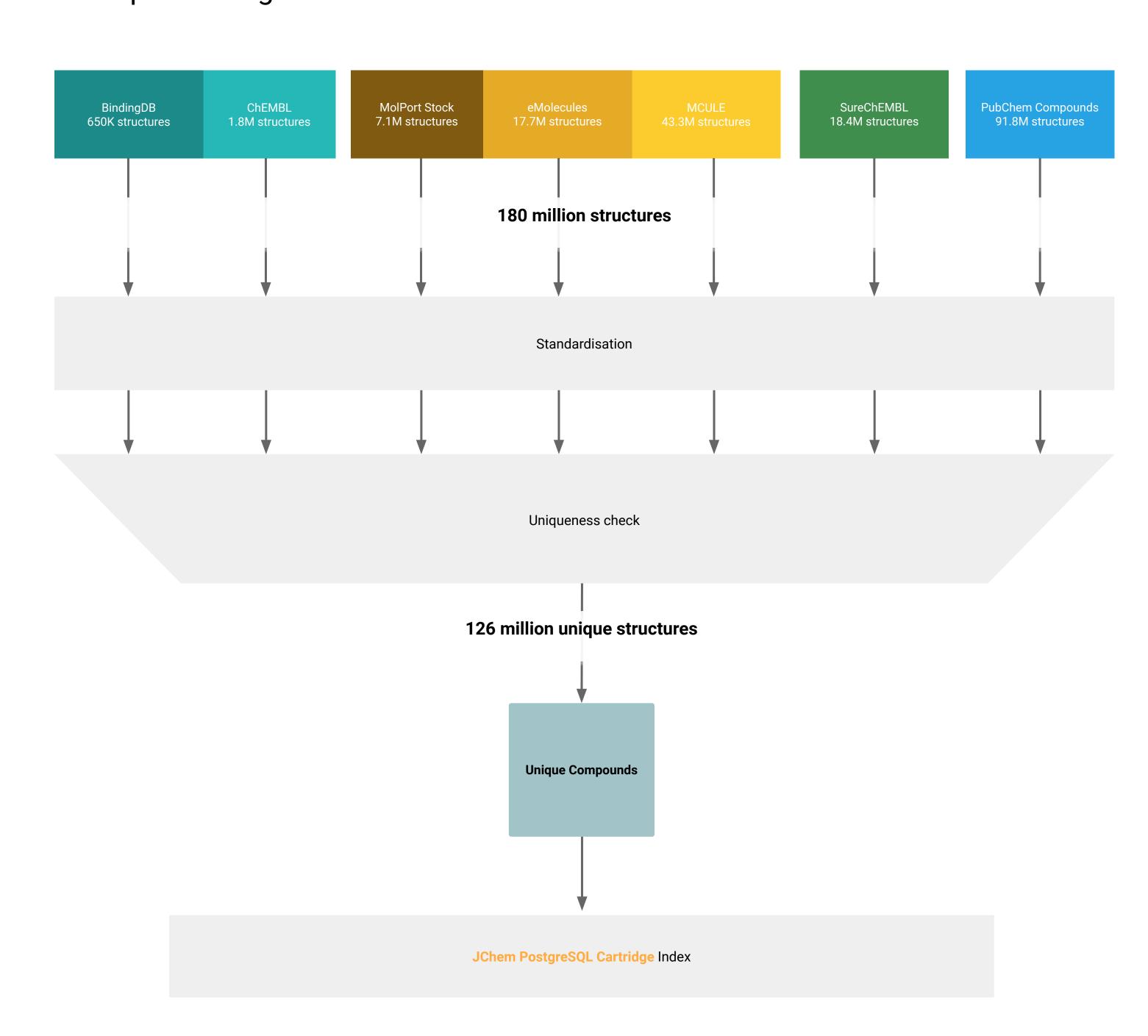
Shrinking the haystack: an overarching search in chemical databases

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Abstract Drug discovery is a knowledge-intensive process in which having the right information at hand can be critical in making the proper decision. With the exponentially growing amount of chemistry and biology-related data in public, commercial and corporate databases it becomes more and more challenging for chemists to find relevant information which helps them to move forward in the right direction with their research.

In this poster we present an ongoing development aiming at providing chemists and other scientists in the pharmaceutical and biotech industries with relevant hits from vendor catalogs, virtual libraries, corporate inventories, in house and publicly available bioassay, metabolic and toxicology databases, as well as patent collections matching the chemical series standing in the focus of their research. Our novel chemical search technologies utilized in this development allow for an instant feedback from very extensive data collections, opening new perspectives in data driven molecule design.

Data processing

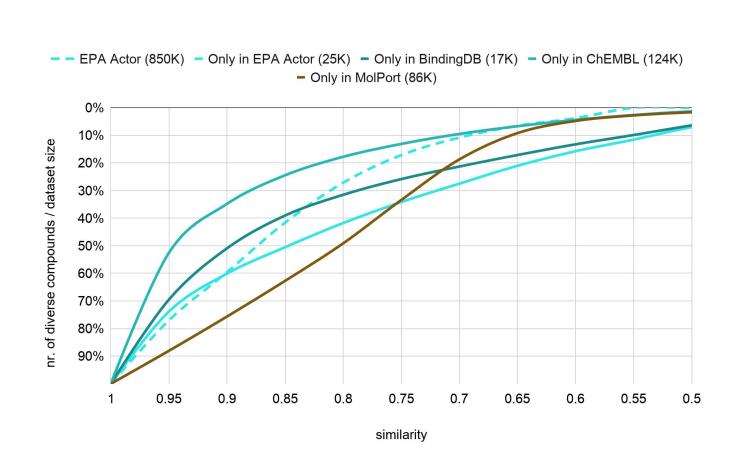


Overlap and diversity of databases

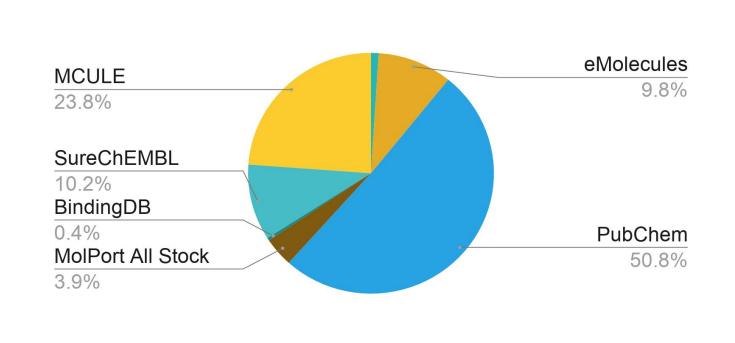
In order to understand the value and the amount of novelty compared to the rest of the data of a newly added database, we have investigated three main aspects: i) the amount of compounds which are only present in a given database, ii) the diversity of the database to be added and iii) the distribution of molecules in a "Sweet Spot" 11 chart.



▲ HEATMAP. Pairwise overlap ratio of databases. Example: 2.80% of eMolecules is found within ChEMBL, and 25.39% of ChEMBL is found in eMolecules. The total amount of compounds are displayed in the header cells



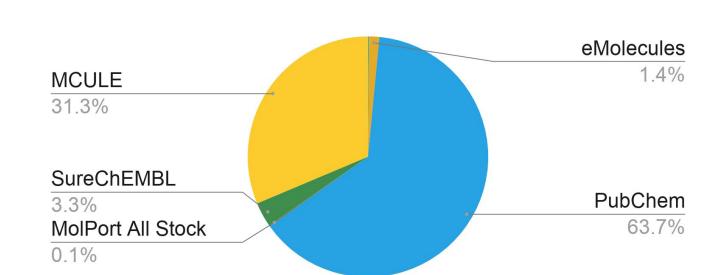
▲ CHART. Number of diverse picks required - normalized to each database's size - to cover the database at different similarity thresholds. Example: at 0.95 ChEMBL² shows low diversity, while Molport is highly diverse. Some databases have been omitted from this chart.



▲ CHART. The total contribution of

filtered dataset of 126M molecules.

each dataset to the duplicate



▲ CHART. The proportion of exclusive contributions (*i.e.* the unique molecules which are present in only one dataset). Actual values in spreadsheet. Example: PubChem ⁷ and MCULE have an order of magnitude larger exclusive compound collection than others. TABLE ▼

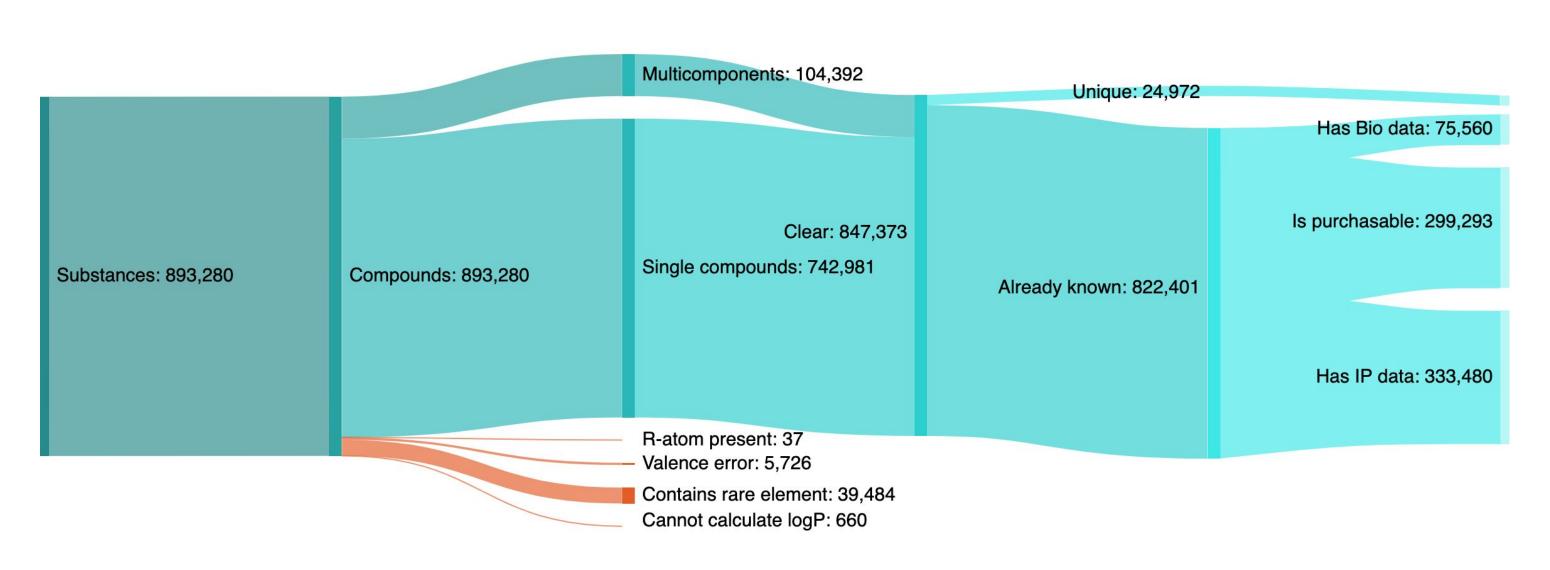
Database	Nr. of exclusives
ChEMBL	122 913
eMolecules	1 021 050
PubChem	44 932 492
MolPort Stock	85 039
BindingDB	16 919
SureChEMBL	2 979 143
MCULE	28 221 313
EPA Actor	24 890

MolPort MTO

Cost and benefit of adding a new dataset

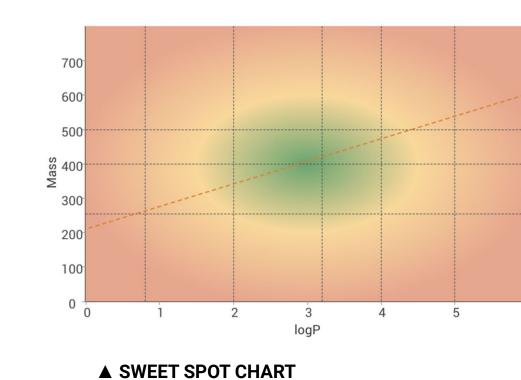
EPA Actor database compiles toxicology data from thousands of public sources ⁹. Over 44M assay results of 506 534 assays are made available for 893 280 structures. Integrating it into this content database could provide useful new data for compounds previously known, or surface purchasability and IP information for the compounds. The process below is used to analyze the quality of the content and the size of the most valuable portions. Quality checks were performed with ChemAxon Structure Checker.

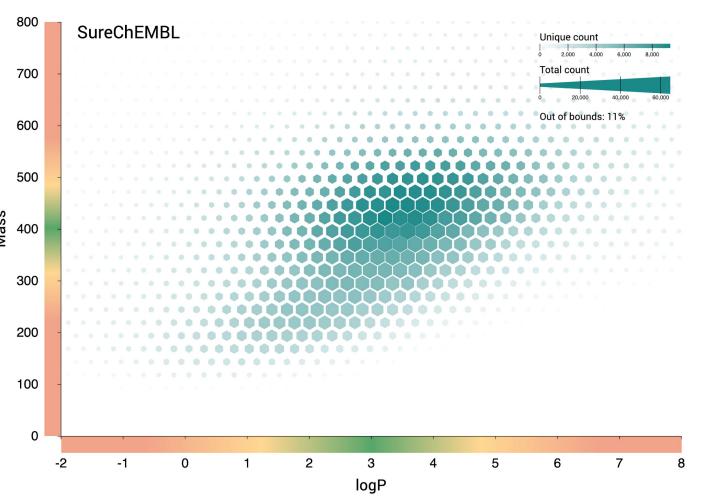
FLOW CHART ▼

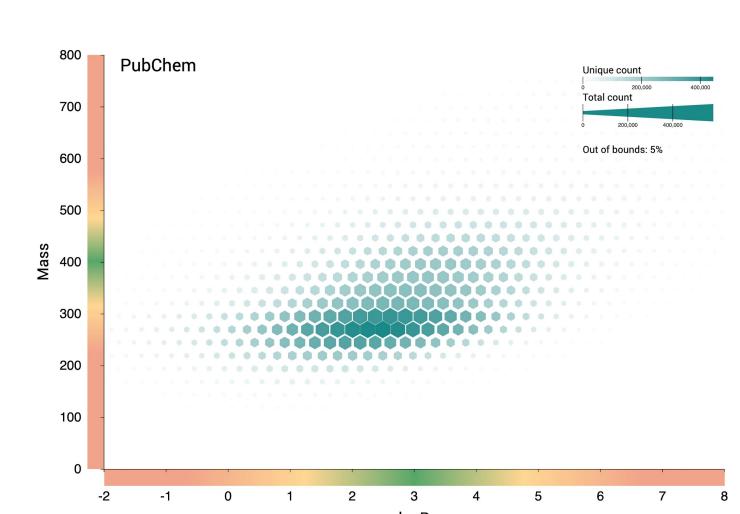


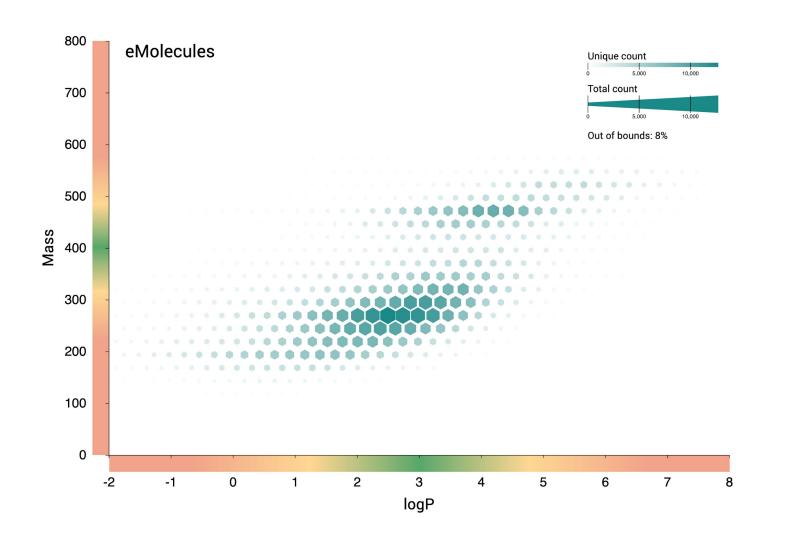
Density plot of databases

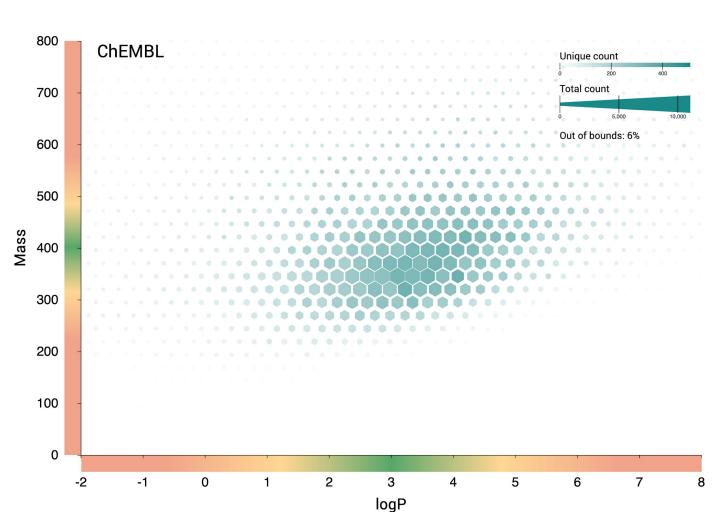
While measuring uniqueness and different data types may provide much needed insight into the value of a database, understanding the drug-likeness of chemical space it covers is crucial. Furthermore, as the cleanliness of these databases vary, understanding where the single source compounds spread out is critical. Traditional scatter plots would misrepresent the density of overlapping points in sets of millions of points, so we selected hexbin plots, and used the coloring suggestions of *Hann*, *et al.*'s "Sweet Spot" visualization ¹¹ to highlight where in mass and logP chemists might want to work. logP values were predicted using ChemAxon Partitioning Plugin. **HEXBIN PLOTS** ▼

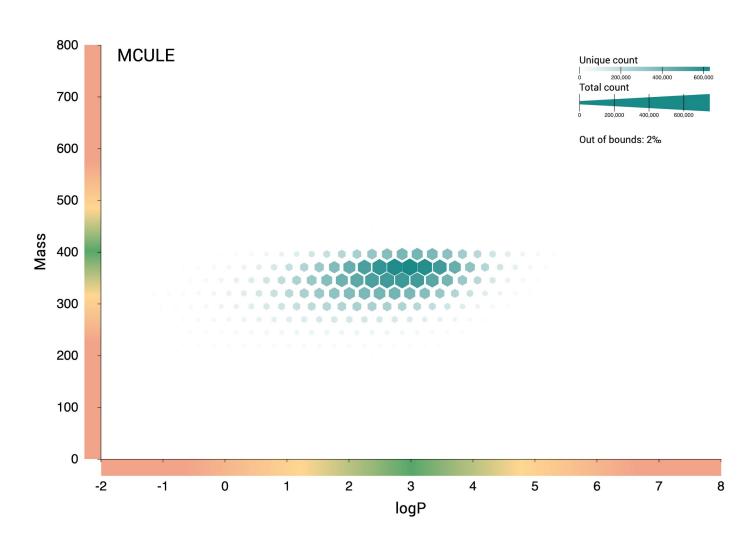


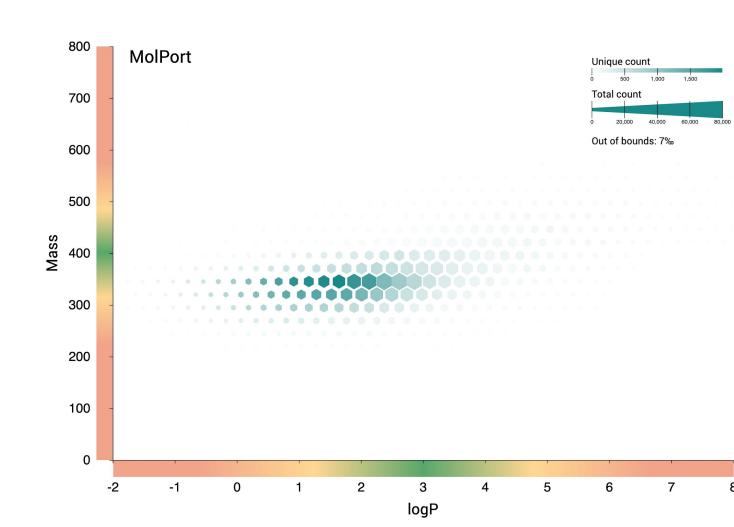












Conclusions

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A new compound database has been created with a collection rivaling the size of leading content services. Each investigated database appears to contribute novel compounds, data. A robust workflow has been developed for adding more databases where the limitation will be comprehension of analysis results and the availability of tools and algorithms that deal with hundreds of millions of records. Operating the service requires modest hardware but great performance is already achievable.

Work should continue on the understanding and exploitation of collected biological/IP/purchasability data.

Citations

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