Freedom Space - Design of Ultra-Large Synthetically **Accessible Chemical Spaces Using Machine Learning**based Reagent Filtering

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Introduction

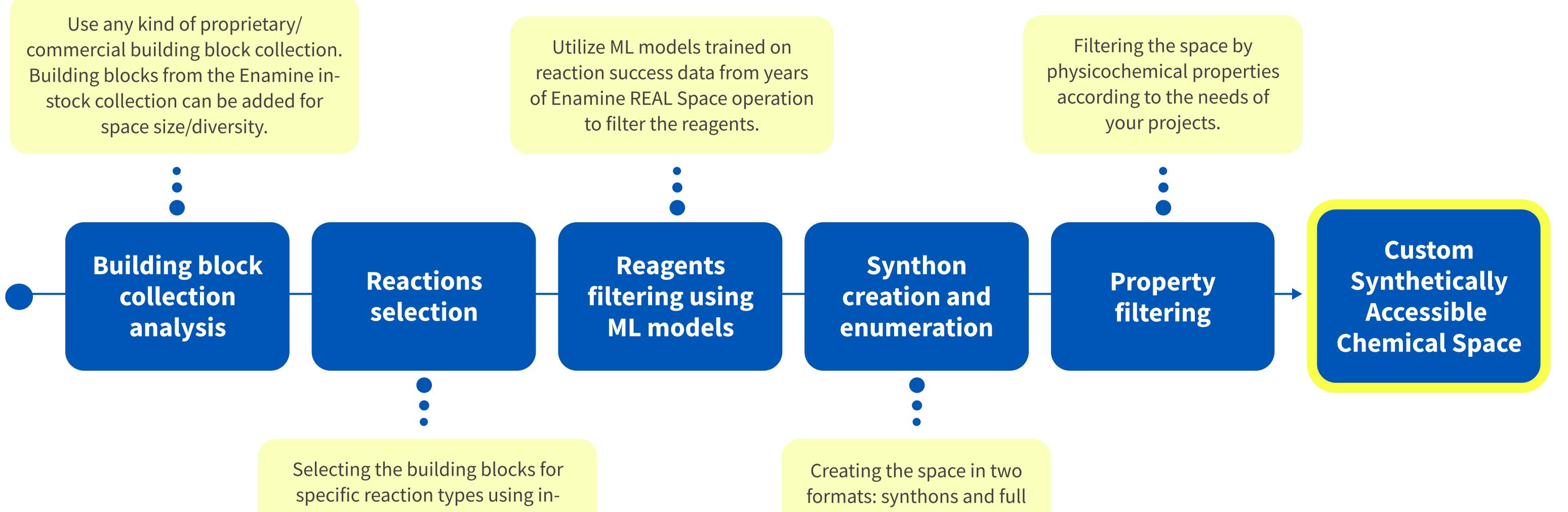
Ultra-large chemical spaces are transforming drug discovery by increasing the chances of identifying novel drug candidates. Efficient use of these spaces requires that compounds be synthesizable, a challenge addressed through methods like synthesizability scores and rule-based filtering. However, improvements are still needed. One of the leading examples is Enamine REAL Space (48B compounds), which has a confirmed 80% synthesis success rate and is created from well-validated reactions and building blocks. In this work, we present Freedom Space 3.0, a 5 billion compound space that is REAL-like and complementary to Enamine REAL and has a confirmed synthesis success rate of 75%+. Using this space as an example, we showcase a novel workflow for creating synthetically accessible chemical spaces from proprietary building block collections using machine learning-based reagent filtering.

Freedom Space 3.0 **5B compounds**

- 10 chemical transformations
- 75+ synthesis success rate
- **Complementary to Enamine REAL**



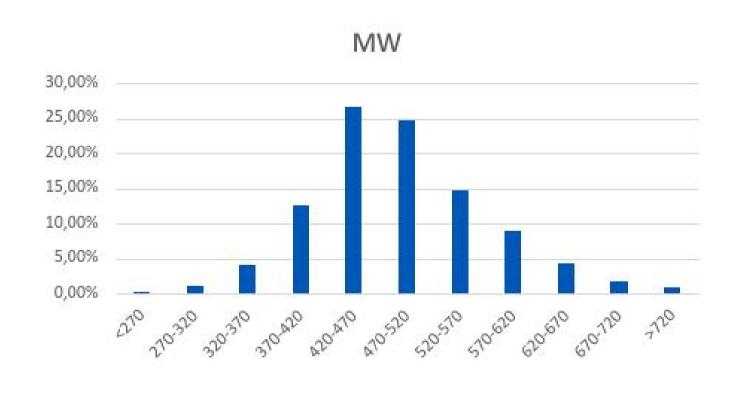
General workflow for the ML-assisted space generation

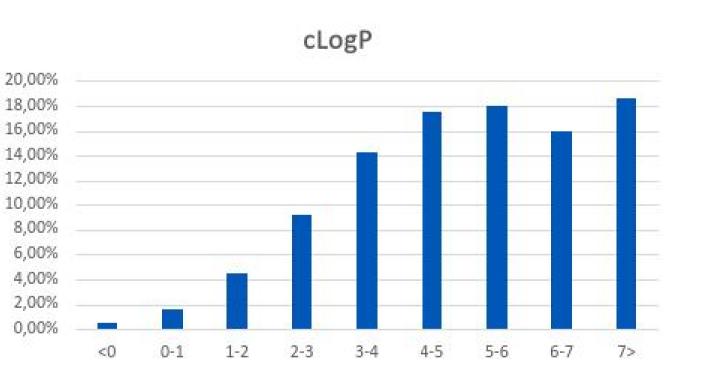


house designed SMARTS patterns.

Freedom Space Overview

Physicochemical properties





Reaction Schemes

Reaction diversity

7 ML-supported (6 two-component and 1 three-component) and 2 human-filtered reactions were used in the space creation.

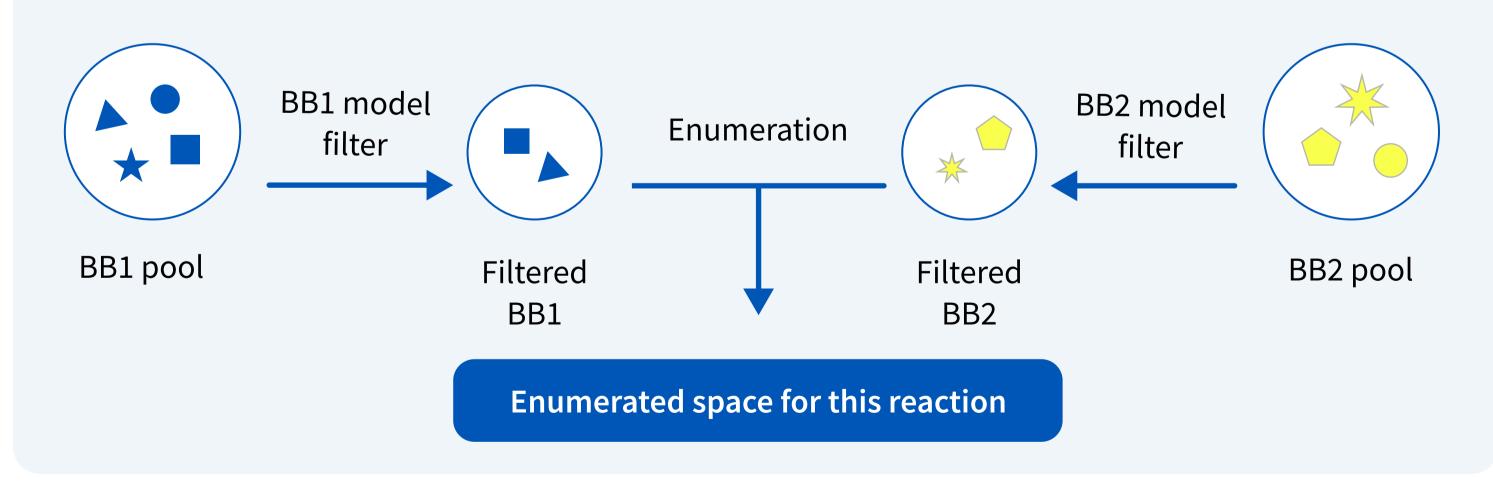
Theoretical synthesisability evaluation

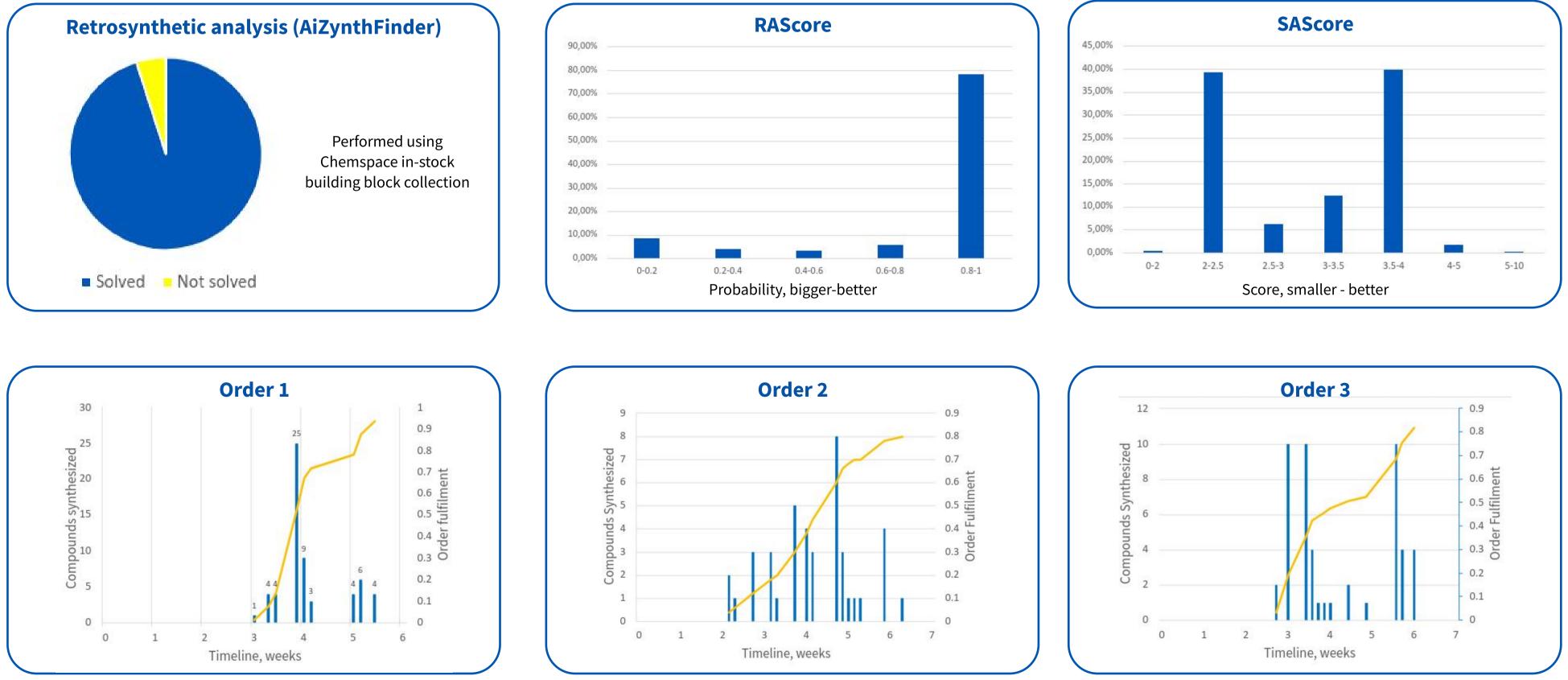
To evaluate the synthesizability of the molecules in the space we utilized commonly used synthesizability scores: RAScore¹, SAscore² and performed retrosynthetic analysis using AiZynthFinder³.

enumerated molecules.

ML-based reagent filtering

The reagents were filtered using machine-learning-based filters, trained on historical reaction success data from the operation of Enamine REAL. For each reagent, a neural network model was trained using a combined synthon-building block representation.





RAScore	
	45,00%
	40,00%
	35,00%
	30,00%
	25,00%

	SASco	ore		
5,00%				
0,00%				
5,00%				
0,00%				
5,00%				

Practical synthesisability evaluation

Freedom Space 3.0 has been commercially available for over a year and has been used in a number of commercial and academic projects. To showcase the synthesisability of the molecules in the space, we are showing some of the timelines of the projects it was used in.

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References:

1.Thakkar, A., Chadimová, V., Bjerrum, E. J., Engkvist, O., and Reymond, J.-L., 2021, Retrosynthetic accessibility score (RAscore) – rapid machine learned synthesizability classification from AI driven retrosynthetic planning: Chemical Science, v. 12, p. 3339–3349.

2.Ertl, P., and Schuffenhauer, A., 2009, Estimation of synthetic accessibility score of drug-like molecules based on molecular complexity and fragment contributions: Journal of Cheminformatics, v. 1. 3.Genheden, S., Thakkar, A., Chadimová, V., Reymond, J.-L., Engkvist, O., and Bjerrum, E., 2020, AiZynthFinder: a fast, robust and flexible open-source software for retrosynthetic planning: Journal of Cheminformatics, v. 12.