SMILES to SMILES Morphing & Intermediate Extraction

Shreyansh Seth Shashwat Chandra June 2025

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1 Problem Statement

Given two valid SMILES strings \mathbf{A} (source) and \mathbf{B} (target), the script

- 1. identifies a common core between A and B,
- 2. prunes A down to that core,
- 3. grows the molecule atom by atom until it matches B logging every unique, chemically valid intermediate encountered,
- 4. repeats the process in the reverse direction (B to A), and
- 5. writes all intermediates indexed by exploration path to a user supplied CSV for downstream analytics.

The objective is to obtain a chemically interpretable trajectory while penalising synthetically unattractive detours.

2 Theoretical Foundations

2.1 Core Invariance

Why? Preserving a shared substructure constrains the search space and keeps each step chemically meaningful (e.g., scaffold hops rather than total synthesis). It minimises arbitrary bond rearrangements and encourages transformations that a synthetic chemist would recognise as feasible.

How? Three increasingly liberal strategies are attempted until one succeeds:

- 1. **SMARTS seeding** user supplied aromatic motifs (e.g., c1cccc1).
- 2. Murcko scaffold match retains ring systems that medicinal chemists typically consider immutable.
- 3. Strict MCS RDKit maximum common substructure with element/valence and ring closure constraints.

Precautions

- molvs.Standardizer normalises tautomers, charges, and stereochemistry, ensuring both molecules start from comparable baselines.
- Every sanitisation is followed by Chem.Kekulize to avoid mixed aromaticity states, which could otherwise mislead SMARTS matching or fingerprint generation.
- verify_core_invariance asserts that neither core atoms nor core bonds mutate, providing a hard guarantee that the reaction path stays chemically interpretable.

2.2 Search Strategy

2.2.1 Phase I – Beam Search

A stochastic best-first search with each state scored by

$$Composite = \alpha \cdot Sim_{target} - \beta \cdot SA + \gamma \cdot Sim_{source}, \tag{1}$$

where similarities are Tanimoto indices on Morgan fingerprints ($\times 100$) and SA is the synthetic accessibility score

We selected beam search in this case because it effectively searches many high-scoring routes in parallel, directing the search towards high potential partial structures without fully exploring low probability branches.

Table 1: Default scoring weights

Symbol	Meaning	Default
α	importance of reaching B	1.0
β	penalty for poor SA	0.1
$\underline{\gamma}$	encouragement to stay near A	0.5

Diversity vs. exploitation The algorithm has a beam size of BEAM_SIZE. We hold the top $\lfloor (1 - \varepsilon) \text{BEAM_SIZE} \rfloor$ states strictly by score while injecting a random sample from the next best RANK_CUTOFF candidates. This hybrid does not converge prematurely on suboptimal routes but targets computation on promising regions.

Termination criteria

- Similarity ≥ SIM_THRESHOLD (95 % by default), indicating that the target framework has successfully been attained.
- Frontier drained (no more children pass validity tests).
- MAX_STEPS attained an upper limit to avoid unbounded search.

2.2.2 Phase II – Breadth First Sweep

Once a high similarity molecule is found, a complete breadth first search over the remaining unmapped atoms ensures *exhaustive* enumeration of intermediates along that path. Because every addition is validated and canonicalised, duplicates are discarded via a global **seen** set.

We use BFS here, rather than other uninformed methods like DFS or uniform cost search, because in comparative trials, BFS consistently delivered the best coverage and shortest step enumeration, so we stuck with it for maximal reliability.

2.3 Why the Method Works

Fingerprint Guidance Morgan fingerprints record atom environments; maximizing their Tanimoto simsimilarity is geometrically reasonable since adding the appropriate atom in the appropriate context will generally enhance similarity to the target without sacrificing overlap with the source core.

Synthetic Plausibility Bias Executing the algorithm in both directions (A to B and B to A) captures asymmetries. Certain atoms that are easy to add may be difficult to remove, and vice versa. Logging in both directions thus maximizes the likelihood of finding a symmetric, interpretable midpoint set.

Bidirectional Coverage Running the algorithm in both directions (A to B and B to A) catches asymmetries. Some atoms that are easy to add might be hard to delete, and vice versa. Logging from both directions therefore increases the chance of discovering a symmetric, interpretable midpoint set.

Core Lock Guarantee By specifically forbidding any modification of the mapped core atoms and bonds, we prevent artefacts wherein the algorithm destroys some of the scaffold temporarily, only to rebuild later, a typical failure mode in simplistic graph searches. The core lock thereby grounds the trajectory in an informative chemical space.

2.4 Implementation Highlights

This subsection links each theoretical element above to the concrete Python code in morph.py.

- Module Setup Global constants (BEAM_SIZE, EPSILON, etc.), RDKit log suppression, Morgan fingerprint generator, and SA-Score resource path are declared at the top of the script.
- Core Detection Functions

seed_core_by_smarts Implements SMARTS seeding.
try_murcko_core Extracts Murcko scaffolds using RDKit's ring info utilities.
strict_mcs Wraps rdFMCS.FindMCS with strict ring and valence parameters.

- Sanitisation Pipeline validate_smiles → MolVS standardiser → RDKit sanitise → Kekulise, guaranteeing consistent input.
- Beam Search Engine beam_search embodies Phase I, computing composite scores, maintaining a priority ordered list, and applying ε greedy diversification.

• BFS Enumeration – add_atoms_bfs performs Phase II, traversing the target adjacency map once the similarity threshold is satisfied.

• Integrity Guards

- verify_core_invariance halts execution if any core atom symbol or bond order mutates.
- Fragmentation checks ensure molecules stay single component.
- Data Logging Novel intermediates are streamed to explored_molecules.csv via csv.DictWriter; canonical + isomeric SMILES compose the uniqueness key used by the global seen set.
- Bidirectional Driver morph_bidirectional sequentially calls morph_direction for $A \to B$ and then $B \to A$, passing along an incrementing path ID counter to preserve lineage across both directions.
- Post processing Utility The optional statistics block at the end of the script reopens the CSV to compute min, mean, median and std for SA and similarity metrics and prints the five highest scoring intermediates.

3 File Outputs

- explored_molecules.csv each intermediate with SA, SimSource, SimTarget, CompositeScore, PathID.
- Console summary descriptive statistics and **Top 5** intermediates by composite score.
- Inline molecule grids (Jupyter / VS Code) for visual inspection.

4 Usage Guide

4.1 Prerequisites

Ensure that you have Python 3.8 or newer installed. Then run this single Jupyter cell to install all dependencies via pip:

```
# ------ In Cell 1: Install Dependencies ------
%pip install -q rdkit molvs
%pip install -q ipython notebook matplotlib
%pip install -q pillow urllib3
```

Note: Also put sascore.py & fpscores.pkl.gz in the same directory as the jupyter notebook.

4.2 Editing the Script

```
# ------ In Cell 2: USER PARAMETERS ------
CSV_PATH = r"PATH_TO_explored_molecules.csv"  # <- CSV Path
SOURCE = "SMILES_A_HERE"  # <- Source (A)
TARGET = "SMILES_B_HERE"  # <- Target (B)
#OR
if __name__ == "__main__":
    morph_bidirectional(
        "SMILES_A_HERE",  # <- Source (A)
        "SMILES_B_HERE",  # <- Target (B)
        r"PATH_TO_explored_molecules.csv"  # <- CSV Path
)</pre>
```

```
# ------ In Cell 4(Optional): Visualization -----
smile_to_show = "ANY_SMILES_TO_PLOT"
```

4.3 Running in Jupyter Notebook

Execute the notebook cells in order:

- 1. Cell 1: Run the first cell once to download all the needed libraries and dependencies.
- 2. Cell 2: Run the second cell to load all imports, set parameters, and define the core functions and search routines.
- 3. Cell 3: Run the third cell to post-process the CSV (summary statistics for SA and similarity values).
- 4. Cell 4: Run the fourth cell to visualize any single SMILES structure using the PlotSmiles function.

5 Parameter Tuning Tips

Knob	Effect	Typical Range
BEAM_SIZE	search breadth vs. RAM	200 - 5000
arepsilon (EPSILON)	exploration rate	0.05 - 0.25
SIM_THRESHOLD	early stop criterion	90-99~%
α, β, γ	scoring weights	project-specific

6 Conclusion

This pipeline offers a chemically informed, solid collection of conversions between any two SMILES strings by:

- Anchoring to a validated core to maintain interpret ability,
- Utilizing a hybrid beam/BFS search to distribute search and achieve full coverage,
- Steer the search according to the similarity and synthetic-accessibility scores to favour probable intermediates,
- Tracking each unique, legitimate intermediate for downstream analysis or visualization.

In general, this approach offers a defined, reproducible sequence of molecular checkpoints linking source and target structures, both for theoretical reasons and for practical route planning.