Population-based de novo Molecule Generation, **Using Grammatical Evolution**

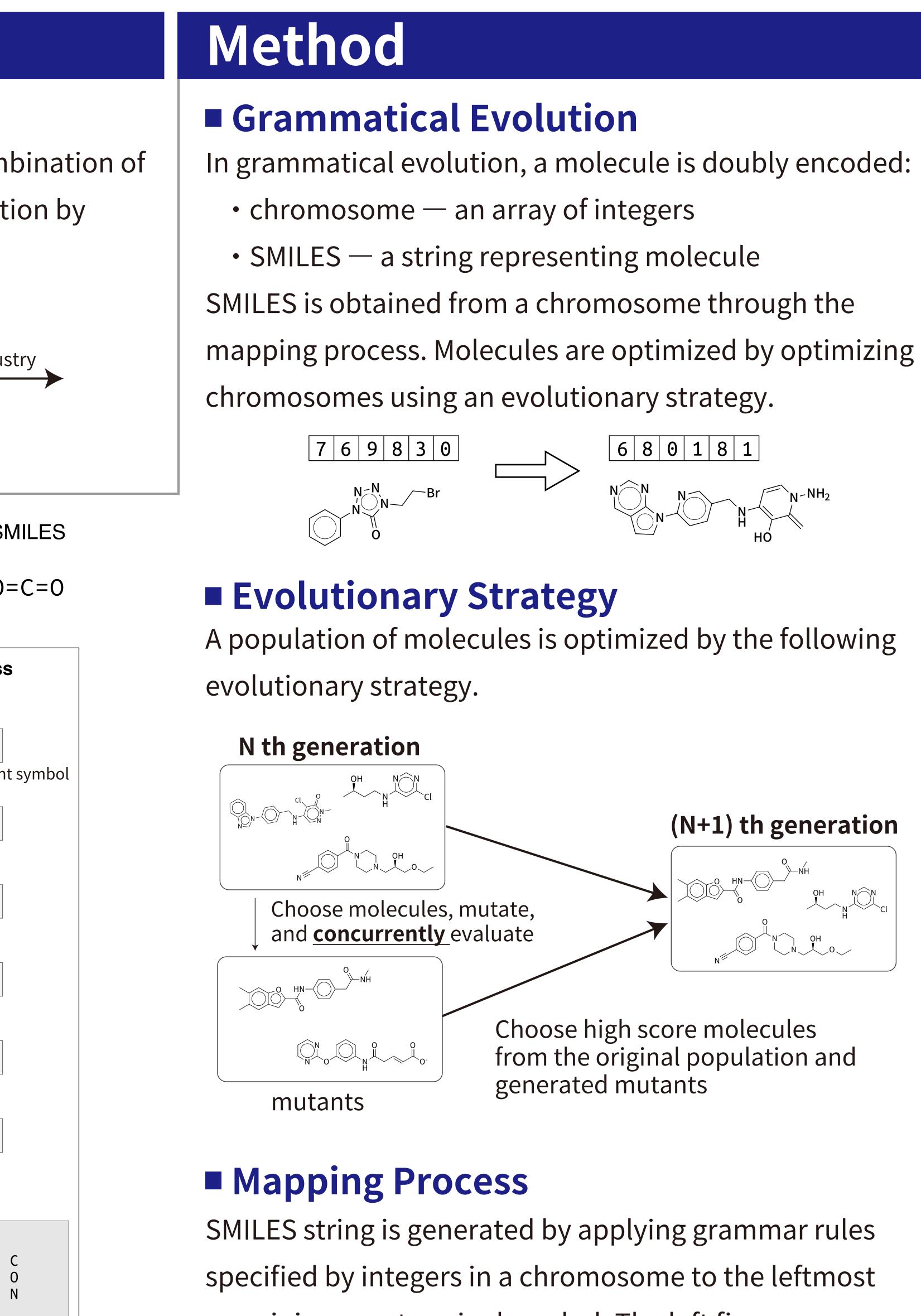
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Introduction

Automatic Molecular Design

Automatic molecular design is conducted by a combination of molecule design by generative process and evaluation by simulators or machine learning models.

| initial molcules | |
|---|---|
| Chromosome SM | |
| k 1 2 3 C[k] 2 5 7 | $\begin{array}{c ccccccccccccccccccccccccccccccccccc$ |
| Translation from chromosome to SMILES in Mapping process | |
| <smiles></smiles> | The starting symbol is <smiles></smiles> |
| | $0 (= \frac{2}{5} \% \frac{1}{5}) < smiles > \Rightarrow < chain > 1$ |
| <u><chain></chain></u> | \The number of rules available for the current s |
| rule | $1 (= 5 \% 2) < chain> \rightarrow < chain> < bond> < atom>$ |
| <u><chain></chain></u> <bor< td=""><td>nd> <atom></atom></td></bor<> | nd> <atom></atom> |
| rule | $1 (= 7 \% 2) < chain> \rightarrow < chain> < bond> < atom>$ |
| <u><chain></chain></u> <box< td=""><td>nd> <atom> <bond> <atom></atom></bond></atom></td></box<> | nd> <atom> <bond> <atom></atom></bond></atom> |
| rule | $0 (= 6 \% 2) < chain> \rightarrow < atom>$ |
| <u><atom></atom></u> <bond< td=""><td>d> <atom> <bond> <atom></atom></bond></atom></td></bond<> | d> <atom> <bond> <atom></atom></bond></atom> |
| rule | $1 (= 1 \% 3) < atom> \rightarrow 0$ |
| 0 <u><bond></bond></u> <a< td=""><td>tom> <bond> <atom></atom></bond></td></a<> | tom> <bond> <atom></atom></bond> |
| | 1 (= 4 % 3) <bond> → =</bond> |
| O = <u><atom></atom></u> ↓ | <body> atom></body> |
| • | Grammar <smiles> \rightarrow (0) <chain> <atom> \rightarrow (0) C</atom></chain></smiles> |
| | <pre><chain> \rightarrow(0) <atom> \rightarrow(1) 0 <(2) N</atom></chain></pre> |
| • 0=C=0 | \rightarrow (1) <chain><bond><atom> <bond> \rightarrow (0) - \rightarrow (1) =</bond></atom></bond></chain> |
| | |
| | |



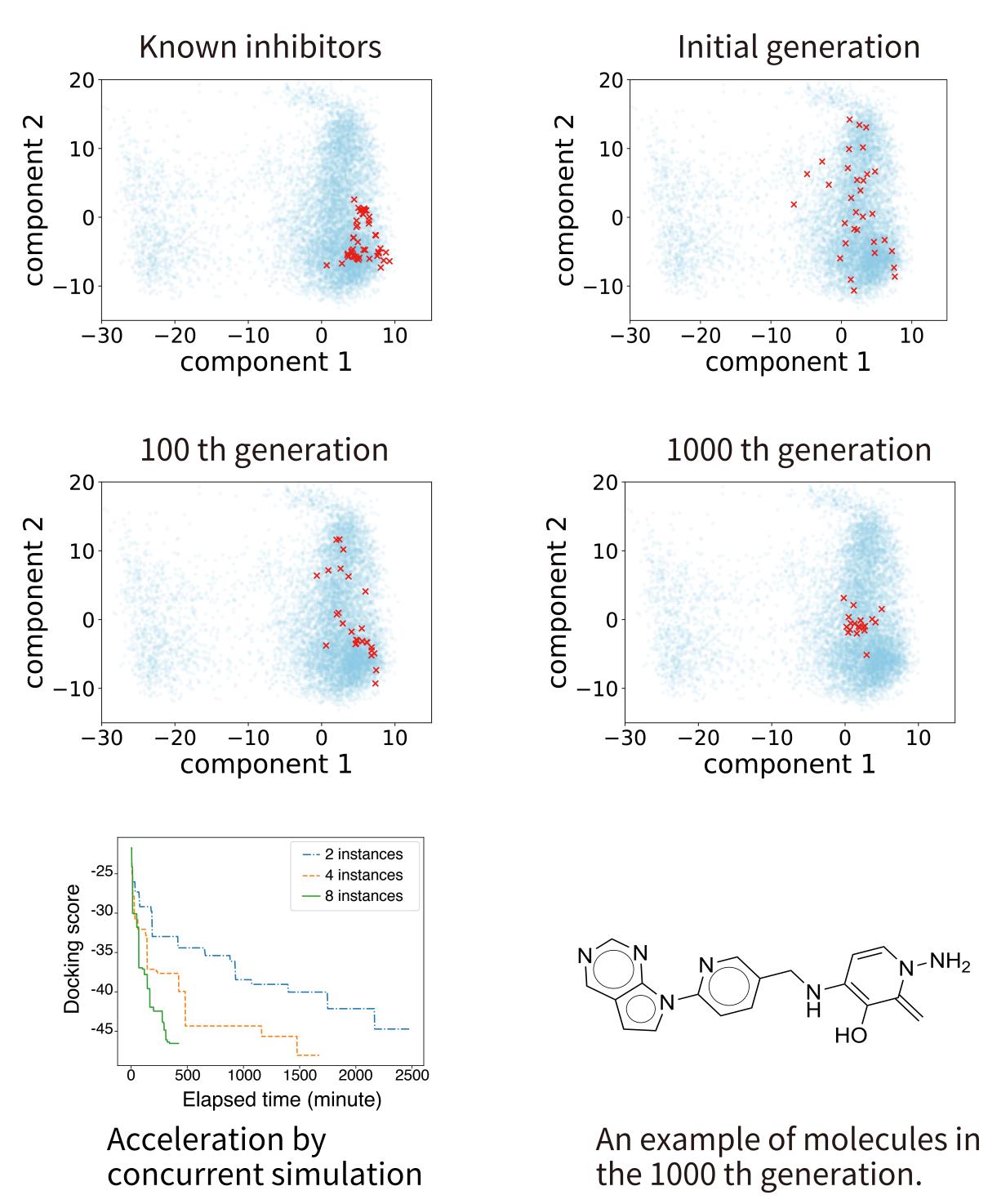
remaining non-terminal symbol. The left figure illustrates how this translation is conducted.

- 6 8 0 1 8 1

(N+1) th generation

Results

- accessibility score.



Conclusion

We developed a new molecular generator using grammatical evolution. This work demonstrated that molecule generation is possible without costly deep learning and showed a new direction for research. Our paper is available at https://arxiv.org/abs/1804.02134

• We optimized the sum of docking score (representing interaction with thymidine kinase, calculated by rDock) and the synthetic

• We found new molecules whose scores are better than known inhibitors. We used isomap to visualize molecules in 2D space.