

Advances in Molecular Docking for Predicting Protein Ligand Interactions

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DESCRIPTION

Molecular docking has emerged as one of the most influential computational techniques in modern drug discovery and structural biology, a powerful framework for understanding how small molecules interact with biological macromolecules. At its core, molecular docking seeks to predict the preferred orientation of a ligand when it binds to a target protein, as well as estimate the strength and stability of this interaction. By simulating molecular recognition events *in silico*, docking reduces the need for exhaustive experimental screening and accelerates the early stages of therapeutic development. The conceptual foundation of molecular docking is rooted in the principles of structural complementarity and binding energetics. Proteins possess complex three-dimensional structures with distinct binding pockets formed by amino acid residues arranged in specific spatial configurations. Docking algorithms attempt to explore the vast conformational space of ligand orientations and conformations to identify poses that minimize the free energy of binding. This optimization problem is computationally challenging due to the flexibility of both ligands and protein receptors. Early docking approaches treated both the ligand and receptor as rigid bodies, significantly simplifying the search space. While this rigid-body assumption allowed for faster computations, it often failed to capture biologically relevant conformational changes that occur upon binding. Modern docking methodologies have evolved to incorporate flexibility, particularly in ligand structures and, increasingly, in receptor side chains. This advancement has improved the accuracy of predictions, enabling more realistic modeling of induced-fit effects and conformational selection mechanisms. However, incorporating full protein flexibility remains computationally expensive, and balancing accuracy with efficiency continues to be a central challenge in the field.

Scoring functions play a crucial role in molecular docking by estimating the binding affinity of predicted ligand poses. These functions are mathematical models designed to approximate the free energy of binding using physical, empirical, or knowledge-based terms. Physics-based scoring functions rely on molecular mechanics force fields, while empirical scoring functions are derived from regression models trained on experimental binding

data. Knowledge-based approaches, on the other hand, use statistical potentials derived from known protein ligand complexes. Molecular docking has become an essential tool in virtual screening, where large libraries of compounds are computationally evaluated for potential biological activity. This approach has significantly reduced the cost and time associated with experimental high-throughput screening. By prioritizing a subset of compounds with favorable predicted binding affinities, researchers can focus laboratory efforts on the most promising candidates. Docking is also widely used in lead optimization, where chemical modifications are introduced to improve binding affinity, selectivity, and pharmacokinetic properties of drug candidates. Beyond drug discovery, molecular docking plays an important role in understanding fundamental biological processes. Protein-protein interactions, enzyme-substrate recognition, and receptor-ligand signaling events can all be studied using docking simulations. These applications provide valuable insights into molecular mechanisms underlying cellular functions and disease pathways. In particular, docking has contributed to the identification of critical interaction hotspots in protein interfaces, which can serve as targets for therapeutic intervention.

Advances in computational power and algorithm design have significantly expanded the scope of molecular docking. High-performance computing and cloud-based platforms now enable large-scale virtual screening campaigns involving millions of compounds. Machine learning techniques are also increasingly being integrated into docking workflows to improve scoring accuracy and predict binding affinities more effectively. These data-driven approaches are helping to overcome some of the limitations of traditional scoring functions by learning complex nonlinear relationships from experimental datasets. Another emerging direction is the application of docking in systems biology and network pharmacology. Instead of focusing on single targets, researchers are now exploring multi-target interactions to design drugs that modulate entire biological pathways. Docking can help map the interaction landscape of small molecules across multiple proteins, providing insights into polypharmacology and potential side effects. This systems-level perspective is particularly important for complex diseases such as cancer, neurodegeneration, and metabolic disorders. From an

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evolutionary perspective, molecular docking also provides insights into the principles of molecular recognition that have shaped biological systems. The specificity of protein ligand interactions reflects millions of years of evolutionary

optimization, where subtle changes in amino acid sequences can dramatically alter binding behavior. Studying these interactions computationally helps uncover conserved binding motifs and evolutionary constraints that govern molecular recognition.