

TeachOpenCADD ******

Teaching Platform for Computer-Aided Drug Design Using Open Source Packages and Data

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Introduction

Open source programming packages for cheminformatics and structural bioinformatics are powerful tools to build modular, reproducible, and easy-to-share pipelines for computer-aided drug design (CADD). While documentation for such tools is available, only few freely accessible examples teach underlying concepts focused on CADD usage, such as the TDT initiative [1], addressing especially users new to the field. Here, we present TeachOpenCADD, a CADD teaching platform developed from students for students, using ChEMBL [6] and the PDB [7] as well as python tools such as RDKit [2], PyPDB [3], and PyMol [4]. Interactive Jupyter notebooks [5] were developed for central topics, integrating theoretical background and practical code, and are freely available on GitHub: https://github.com/volkamerlab/TeachOpenCADD.

T1. Data acquisition Chemble Chemble from ChEMBL

Get target ChEMBL ID(s) from UniProt ID

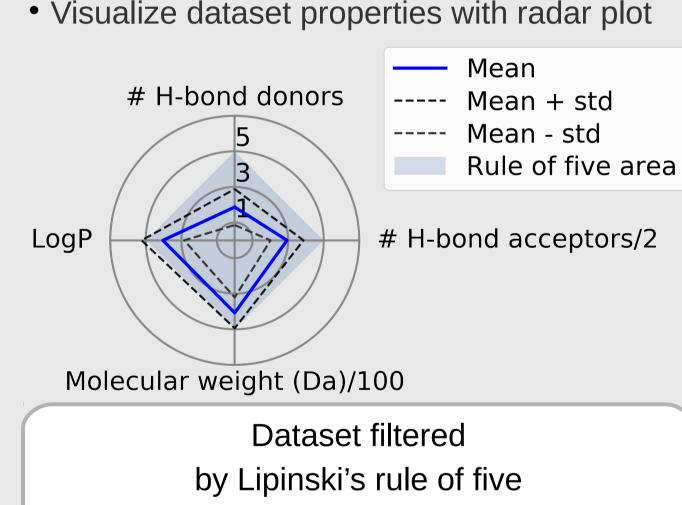
ChEMBL dataset for EGFR 6641 compounds

- Get bioactivity (IC50) & compound (SMILES) data from ChEMBL ID
- Filter out entries with missing values & duplicates (-1601)
- Filter for entries with IC₅₀ in molar units (-69)
- Convert bioactivity to unit nM and pIC₅₀

Dataset filtered & formatted by bioactivity & SMILES 4771 compounds

T2. Molecular filtering: **ADME** criteria

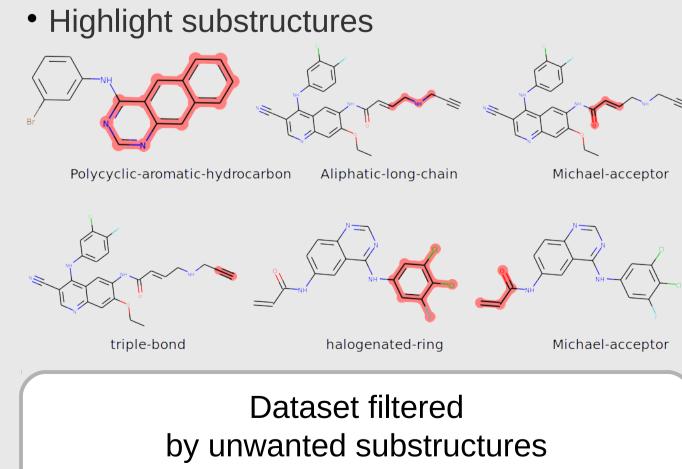
- Calculate 1D chemical properties
- Filter dataset by Lipinski's rule of five (-762)
- Visualize dataset properties with radar plot



4009 compounds

T3. Molecular filtering: **Unwanted substructures**

- Filter out PAINS [8] (-321)
- Filter out unwanted substructures by Brenk et al. [9] (-2058)



1951 compounds

Query target

EGFR (UniProt ID: P00533)

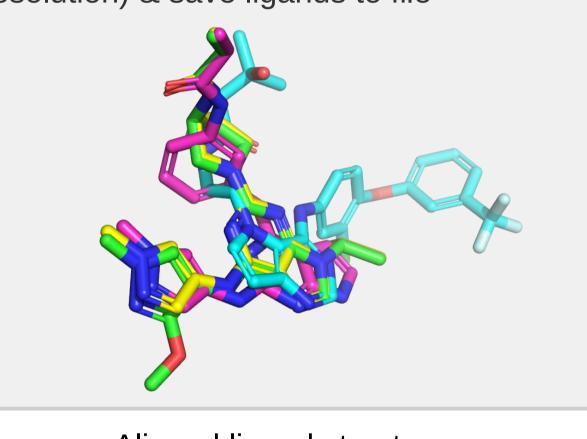
T8. Data acquisition PDB from PDB

Get PDB IDs from UniProt ID of query target

PDB dataset

179 PDB structures

- Filter PDB structures: X-ray, ≤ 3Å resolution & presence of ligand
- Align top 4 protein-ligand structures (by resolution) & save ligands to file



Aligned ligand structures

4 ligands

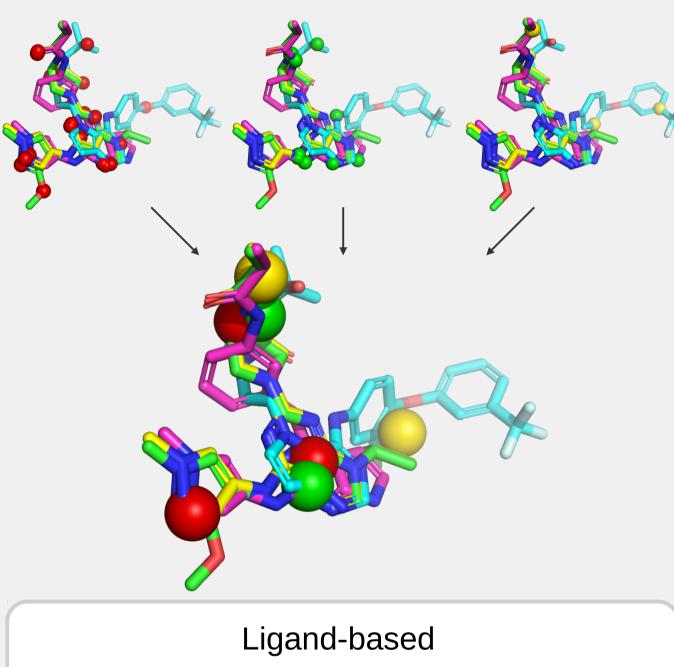
Topics

Illustrated with the example of the epidermal growth factor receptor (EGFR) kinase, we discuss how to acquire data from ChEMBL (T1), filter compounds for drug-likeness (T2), and identify unwanted substructures (T3). Furthermore, we introduce measures for compound similarity, applied for virtual screening (VS) of Gefitinib based on a similarity search (T4) and for compound clustering (T5), including the use of maximum common substructures (MCS) (T6). We also employ machine learning (ML) approaches to build models for predicting active compounds (T7). Lastly, structures are fetched from the PDB (T8), used to generate ligandbased ensemble pharmacophores (T9) and to conduct RMSD-based binding site comparison of Imatinib-binding proteins for off-target prediction (T10).

Imatinib (STI)

T9. Ligand-based ensemble pharmacophores

- Extract pharmacophore features:
- H-bond donors/acceptors & hydrophobic
- Cluster (k-means) each feature type
- Generate ensemble pharmacophore



T10. Binding site comparison

Query ligand

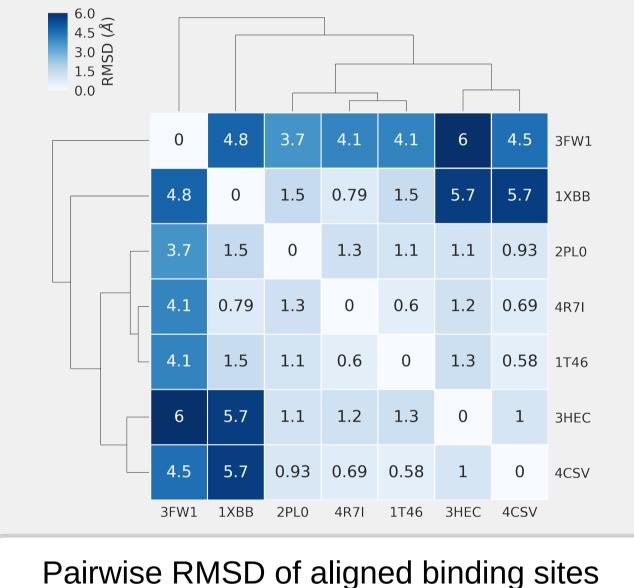
• Get all protein-STI complexes from the PDB

PDB dataset

53 PDB structures

- Filter PDB data set: X-ray & ≤ 3Å resolution
- Align PDB structures & get pairwise RMSD

for whole protein & binding site



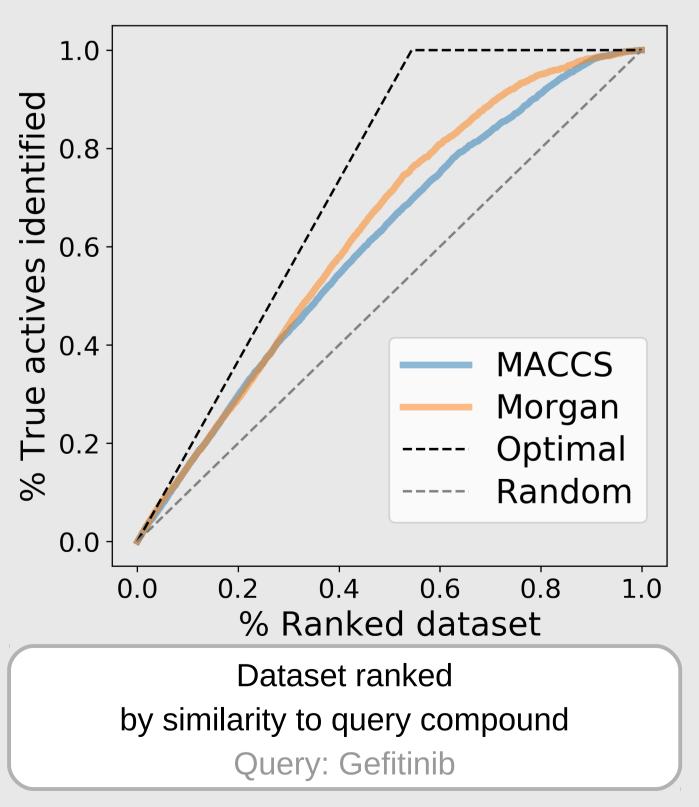
Pairwise RMSD of aligned binding sites

8 protein structures

ensemble pharmacophore

T4. Ligand-based screening: **Compound similarity**

- Calculate MACCS/Morgan fingerprints and Tanimoto/Dice similarity for dataset
- Split dataset into active & inactive compounds (pIC₅₀ cutoff > 6.3)
- VS of query Gefitinib against dataset based on a similarity search
- Evaluate screening with enrichment plots



T5. Compound clustering

- Cluster dataset with the Butina algorithm
- Pick diverse subset based on clusters Threshold: 0.2 Cluster 2 Cluster 1

Dataset clustered by RDK fingerprint similarity

Cluster index

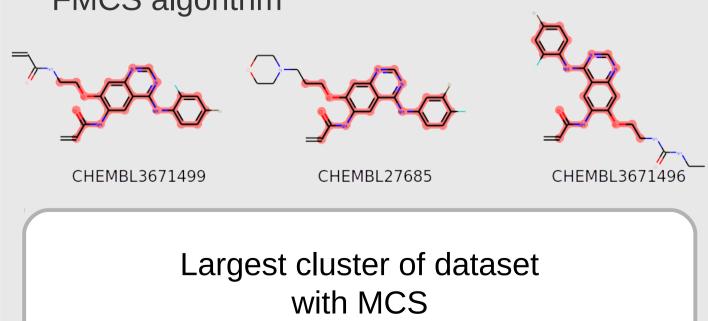
600

800

T6. Maximum common substructures

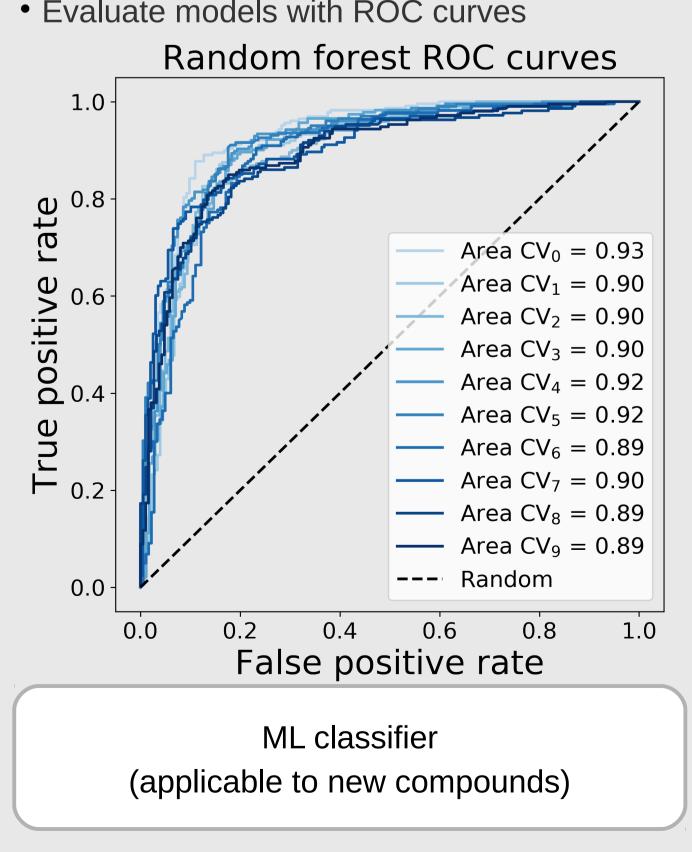
200

 Show MCS for largest cluster using the FMCS algorithm



T7. Ligand-based screening: **Machine learning**

- Split dataset into active & inactive compounds (pIC50 cutoff > 6.3)
- Train ML classifiers using random forest, support vector machines & artificial neural networks
- Apply K-fold cross validation (CV)
- Evaluate models with ROC curves



Conclusion

With this platform, we aim at introducing interested researches and students to the ease and benefit of using open source cheminformatics tools. Topics will be continuously expanded and are open for contributions from the community. Beyond their teaching purpose, the notebooks can serve as starting point for users' project-directed modifications and extensions.

References

[1] S. Riniker et al., F1000Research, 2017, 6, 1136. [2] http://www.rdkit.org. [3] W. Gilpin, Bioinformatics, 2016, 32, 156-60. [4] The PyMOL Molecular Graphics System, Version 1.8. [5] T. Kluyver et al., IOS Press, 2016, 87-90. [6] A. Gaulton et al., Nucleic Acid Res., 2017, 40, D1100-7. [7] H. Berman et al., Nucleic Acid Res., 2000, 28, 235-42. [8] Brenk et al., Chem. Med. Chem., 2008, 3, 435-44. [9] Baell et al., J. Med. Chem., 2010, 53, 2719-40.