

# Large-scale collaborative assessment of binding free energy calculations for drug discovery using OpenFE

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## Introduction

Here we present the outcomes of a large-scale collaborative benchmark of relative alchemical binding free energies using the OpenFE toolkit (<https://openfree.energy/>).

This benchmark, carried out alongside **15 partner companies**, comprises of **1740 protein-ligand complexes** over **96 datasets**. It represents one of the largest evaluations of alchemical free energy calculations in a drug discovery context using **fully open source software**.

## Benchmark details

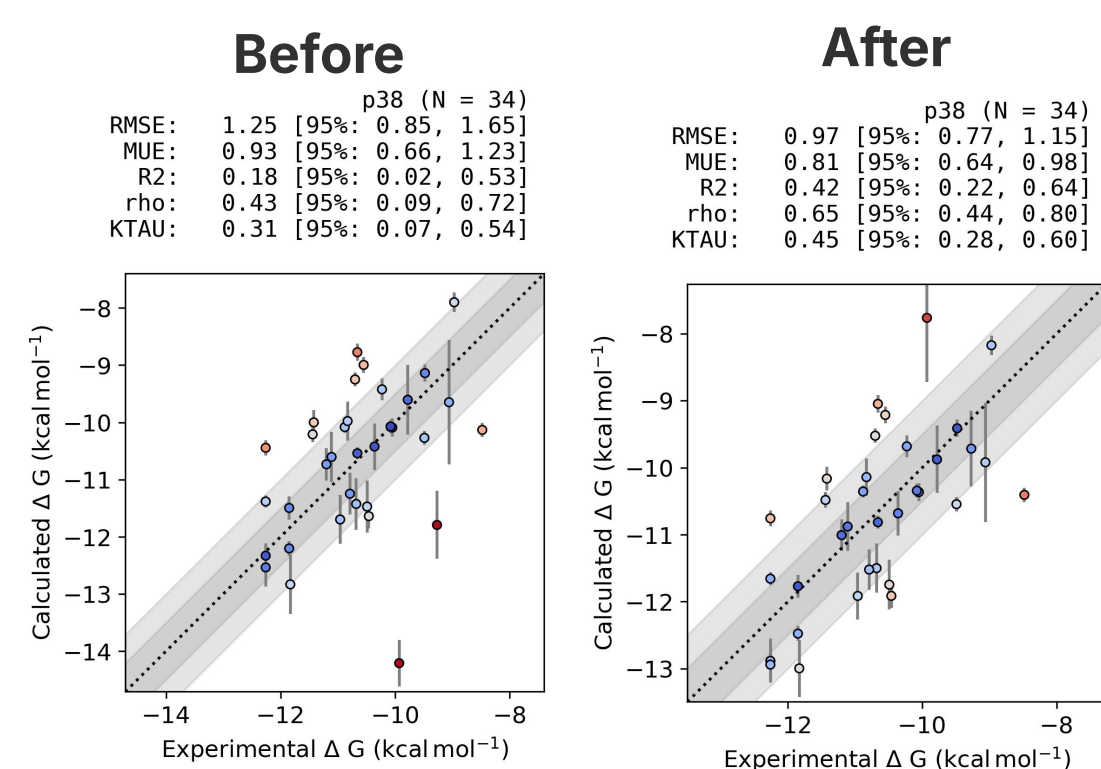
This benchmark comprises of two datasets:

1. **Public dataset:** 58 systems, totalling 876 ligands, from the Ross et al. (2023) Schrodinger public dataset.
2. **Private dataset:** 37 systems totalling 864 ligands, from in-house (blinded) datasets by 10 partner companies.

All simulations were carried out using the OpenFE hybrid topology Protocol using default settings:

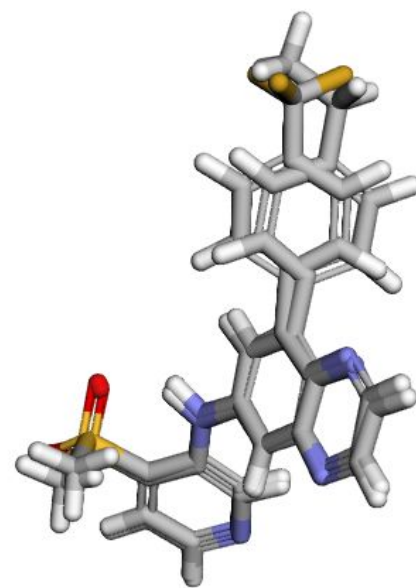
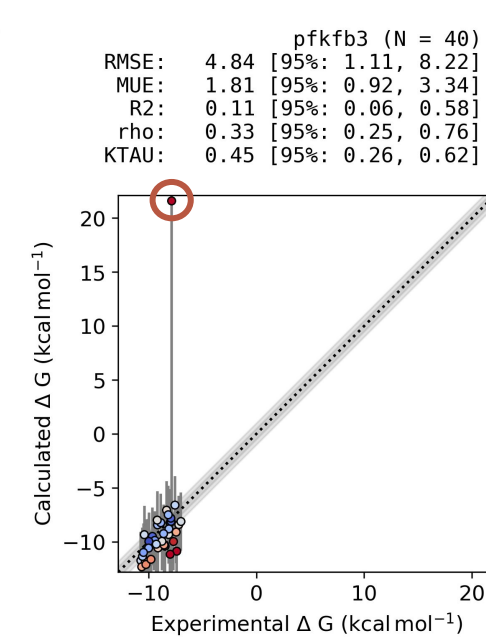
- OpenFF 2.2.0 & FF14SB
- Kartograf atom mapping
- HREX sampling
- 11 lambda windows (22 with net charges)
- 5 ns / window (20 with net charges)
- Triplicate repeats

## Some lessons learned



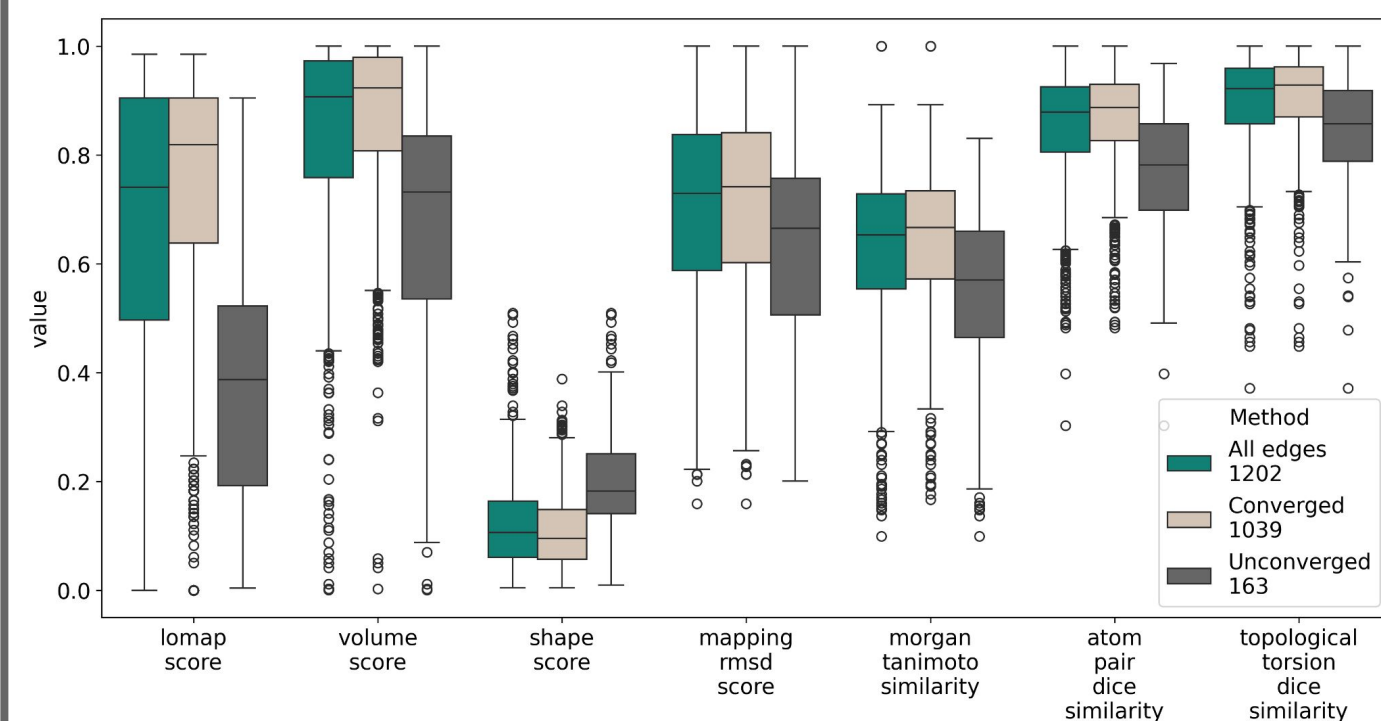
### Removing unconverged edges

Removing redundant unconverged (MBAR errors > 0.15 kcal/mol) edges can sometimes improve accuracy.



### Resolving atom mapping issues

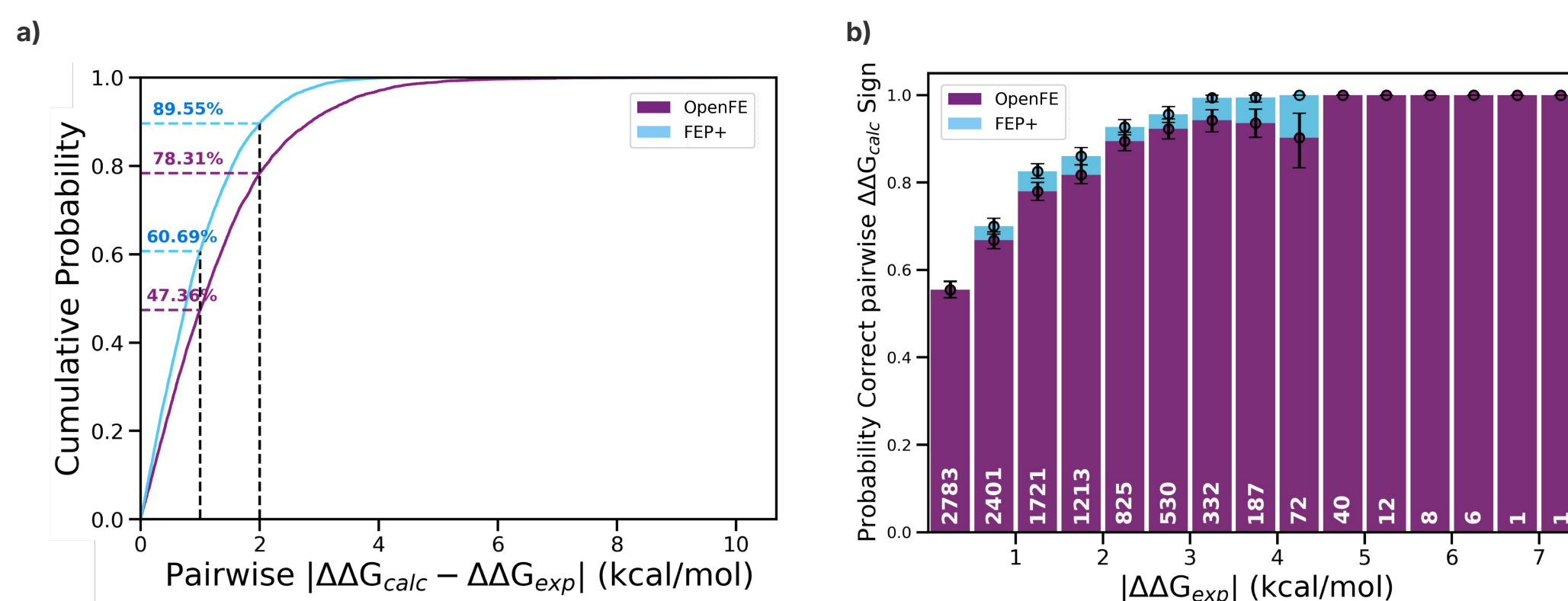
A large outlier in pfkfb3 led us to fix a bug in Kartograf that could lead to broken bond mappings.



### Mapping scores predict high difficulty edges

High MBAR error edges can be categorized by poor mapping scores, e.g. the LOMAP score.

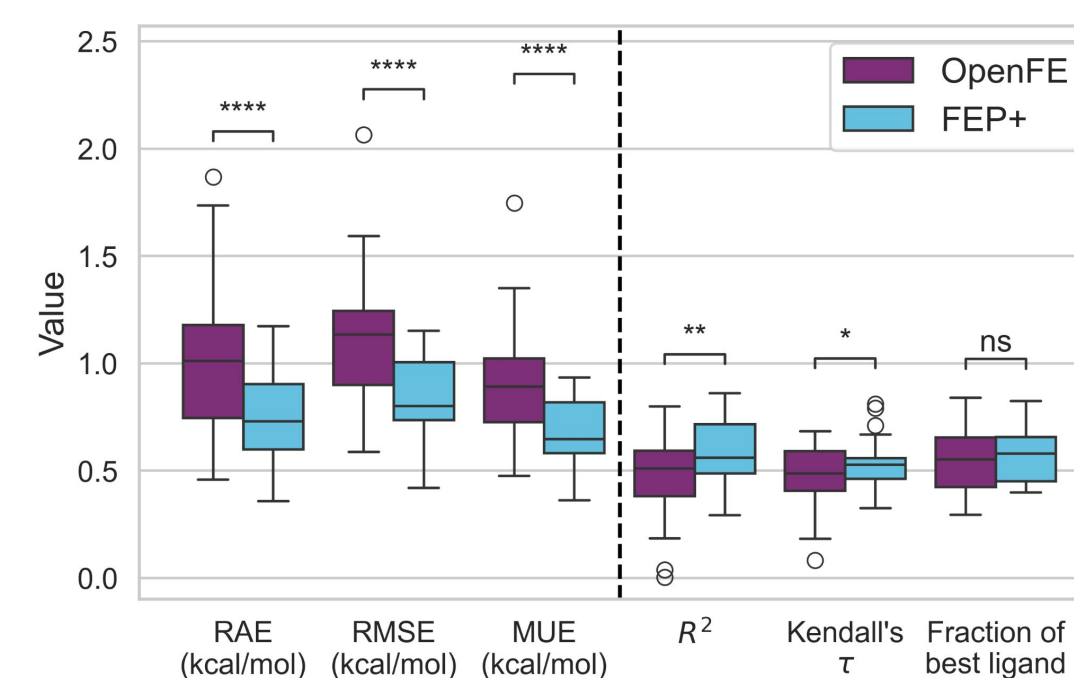
## Public dataset



a) Absolute pairwise  $\Delta\Delta G$  errors and b) sign prediction accuracy as a function of experimental  $\Delta\Delta G$  magnitude, for both OpenFE and FEP+.

OpenFE shows a good agreement with experiment, with an RMSE across all systems of 1.7 kcal/mol. However, it falls slightly short of the current state-of-the-art tooling (FEP+, 1.2 kcal/mol, Ross et al. 2023). Looking at absolute pairwise  $\Delta\Delta G$  errors, 78.3% (vs 89.6% FEP+) of predictions fall within 2 kcal/mol of experiment.

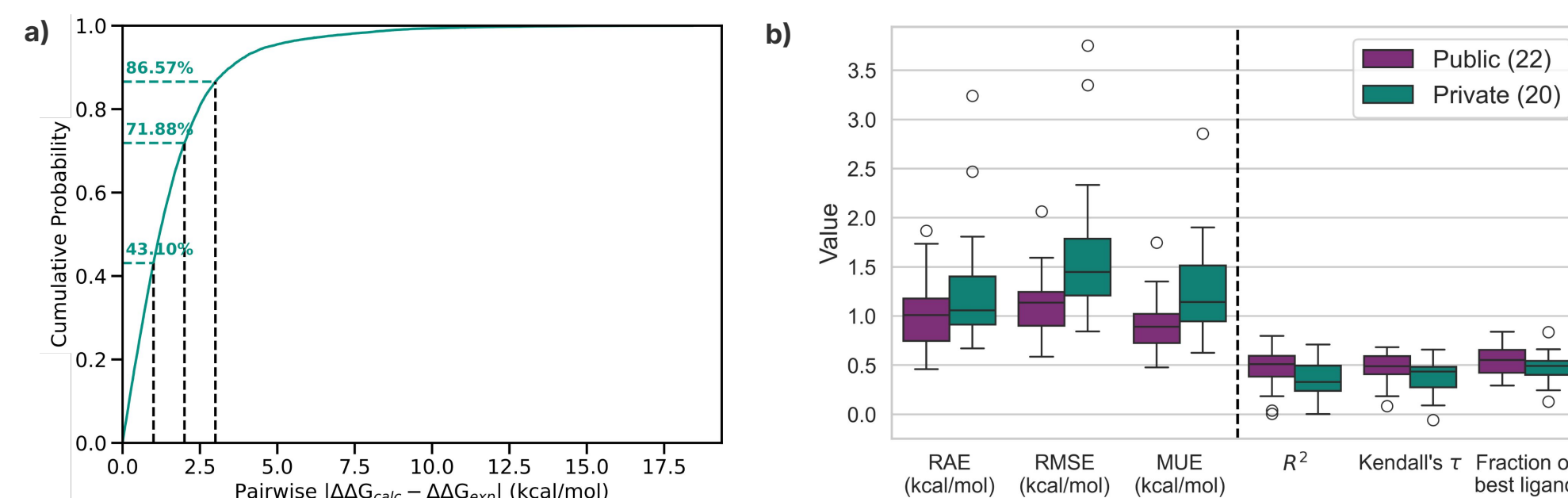
Larger errors partly stem from the use of a naive benchmark setup & sampling scheme, e.g. lack of water placement optimization and poor ligand alignments.



Ranking metrics for OpenFE and FEP+ across all public set systems.

Despite larger errors, OpenFE demonstrates a good predictive power with an average fraction of best ligand score of 0.61.

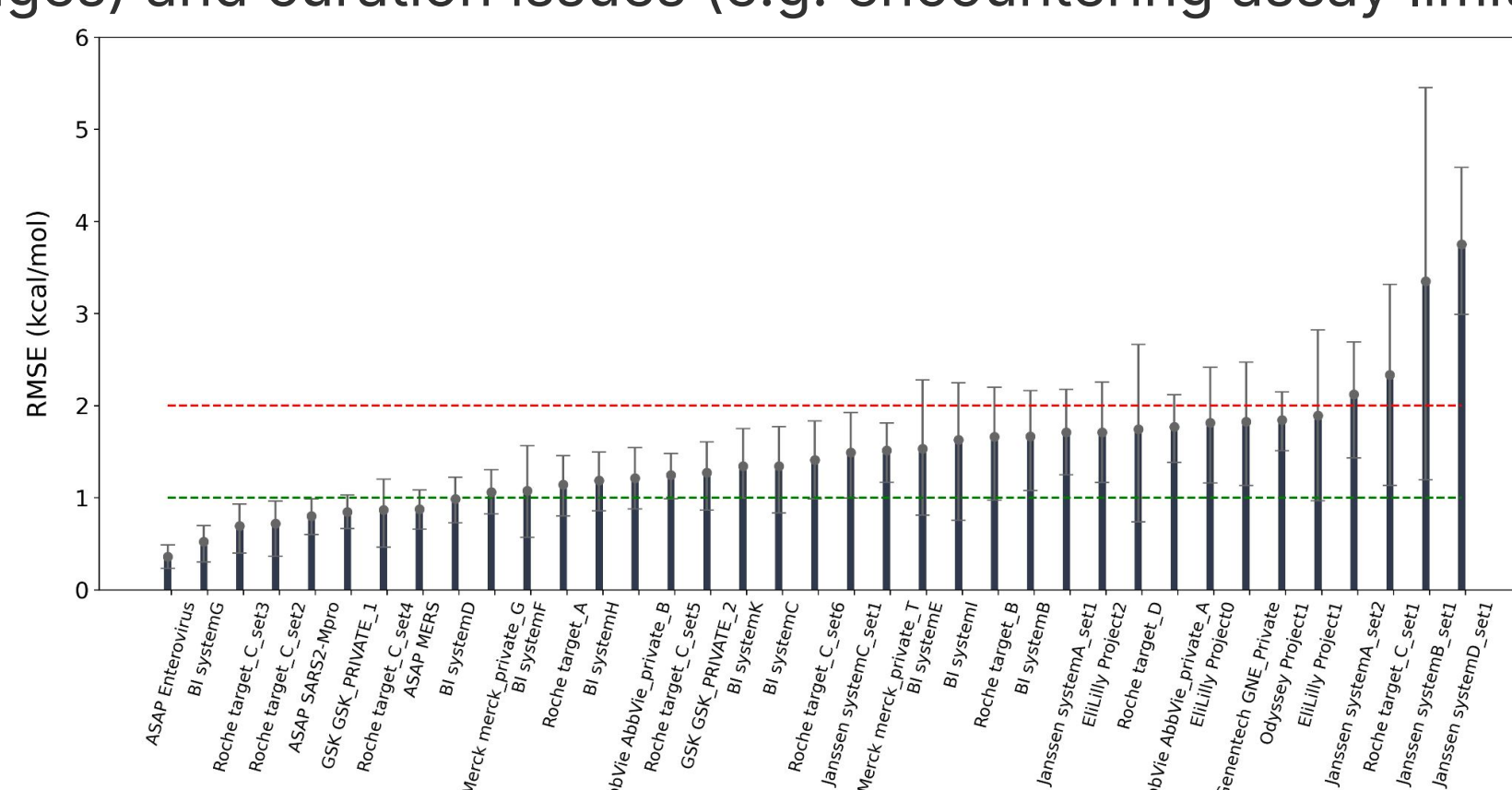
## Private dataset



a) Absolute pairwise  $\Delta\Delta G$ . Dashed lines indicate data filtered to include edges with minimum off-diagonal MBAR overlap > 0.03 (excludes outliers with large inter-repeat variability). b) Ranking statistics for public and private set results.

The private dataset, consisting of in-house benchmarks, offers insights into the likely “real world” effectiveness of OpenFE. While still providing reasonable estimates, we see reduced accuracy and predictive power with only 71% of predictions within 2 kcal/mol of experiment.

High variability in accuracy between systems is partly explained by more challenging transformations (e.g. large alchemical changes) and curation issues (e.g. encountering assay limits).



## Acknowledgements

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- Merck KGaA
- Odyssey Therapeutics
- Johnson & Johnson