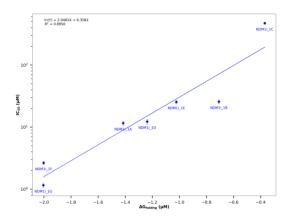
# Scientific test: peptide\_pnear\_vs\_ic50

# **FAILURES**

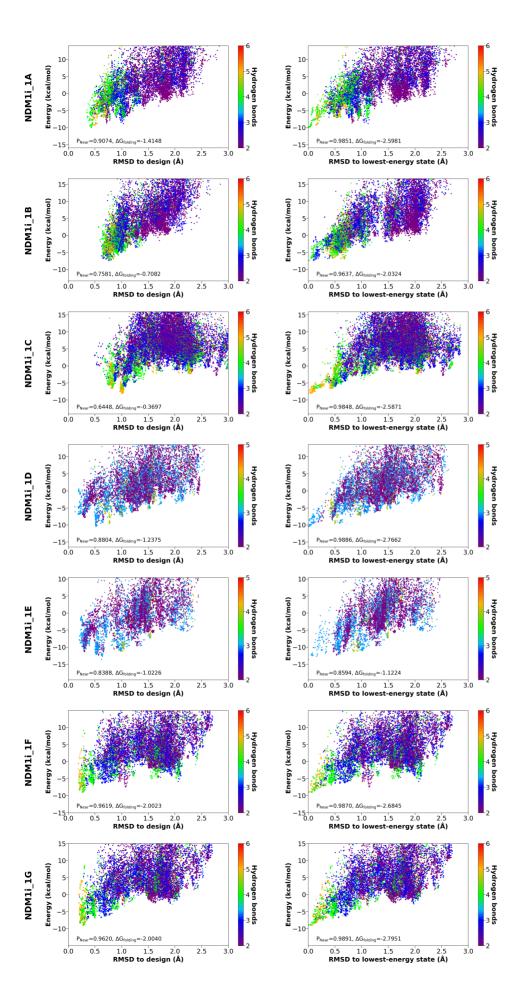
None.

# **RESULTS SUMMARY**

Analysis of correlation of predicted folding propensity with experimentally-measured inhibition values (which should be linear):



Folding funnels:



## Peptide NDM1i\_1A:

Total samples = 22869

Computed PNear = 0.9074

Computed PNear to lowest E = 0.9851

Computed DG\_folding = -1.4148

Computed DG\_folding to lowest = -2.5981

Lowest energy = -9.9745 kcal/mol

RMSD of lowest energy = 0.427 Angstroms

Lowest RMSD = 0.346 Angstroms

Highest RMSD = 2.830 Angstroms

Energy gap (minE>1.5A - minE) = 6.3262 kcal/mol

More than 18,000 samples? YES

PNear value over 0.83? YES

PNear value to lowest E over 0.83? YES

Lowest energy under 0.45 A RMSD? YES

Sampling below expected lower threshold RMSD(0.45 A)? YES

Sampling beyond 1.5 A RMSD? YES

Sampling beyond 2.5 A RMSD? YES

3+ kcal/mol energy gap? YES

**OVERALL PASS? YES** 

## Peptide NDM1i\_1B:

Total samples = 26888

Computed PNear = 0.7581

Computed PNear to lowest E = 0.9637

Computed DG\_folding = -0.7082

Computed DG folding to lowest = -2.0324

Lowest energy = -7.3829 kcal/mol

RMSD of lowest energy = 0.770 Angstroms

Lowest RMSD = 0.567 Angstroms

Highest RMSD = 3.332 Angstroms

Energy gap (minE>1.5A - minE) = 5.8710 kcal/mol

More than 18,000 samples? YES

PNear value over 0.68? YES

PNear value to lowest E over 0.68? YES

Lowest energy under 0.85 A RMSD? YES

Sampling below expected lower threshold RMSD(0.7 A)? YES

Sampling beyond 1.5 A RMSD? YES

Sampling beyond 2.4 A RMSD? YES

3+ kcal/mol energy gap? YES

**OVERALL PASS? YES** 

#### Peptide NDM1i 1C:

Total samples = 39786

Computed PNear = 0.6448

Computed PNear to lowest E = 0.9848

Computed DG\_folding = -0.3697

Computed DG folding to lowest = -2.5871

Lowest energy = -8.0769 kcal/mol

RMSD of lowest energy = 1.009 Angstroms

Lowest RMSD = 0.492 Angstroms

Highest RMSD = 3.138 Angstroms

Energy gap (minE>1.5A - minE) = 5.2921 kcal/mol

More than 18,000 samples? YES

PNear value over 0.57? YES

PNear value to lowest E over 0.57? YES

Lowest energy under 1.05 A RMSD? YES

Sampling below expected lower threshold RMSD(0.55 A)? YES

Sampling beyond 1.5 A RMSD? YES

Sampling beyond 2.9 A RMSD? YES

3+ kcal/mol energy gap? YES

**OVERALL PASS? YES** 

## Peptide NDM1i 1D:

Total samples = 22384

Computed PNear = 0.8804

Computed PNear to lowest E = 0.9886

Computed DG\_folding = -1.2375

Computed DG\_folding to lowest = -2.7662

Lowest energy = -10.4439 kcal/mol

RMSD of lowest energy = 0.520 Angstroms

Lowest RMSD = 0.122 Angstroms

Highest RMSD = 2.707 Angstroms

Energy gap (minE>1.5A - minE) = 5.2321 kcal/mol

More than 18,000 samples? YES

PNear value over 0.8? YES

PNear value to lowest E over 0.8? YES

Lowest energy under 0.55 A RMSD? YES

Sampling below expected lower threshold RMSD(0.25 A)? YES

Sampling beyond 1.5 A RMSD? YES

Sampling beyond 2.4 A RMSD? YES

3+ kcal/mol energy gap? YES

**OVERALL PASS? YES** 

# Peptide NDM1i\_1E:

Total samples = 26042

Computed PNear = 0.8388

Computed PNear to lowest E = 0.8594

Computed DG\_folding = -1.0226

Computed DG\_folding to lowest = -1.1224

Lowest energy = -13.5477 kcal/mol

RMSD of lowest energy = 0.622 Angstroms

Lowest RMSD = 0.205 Angstroms

Highest RMSD = 2.570 Angstroms

Energy gap (minE>1.5A - minE) = 6.2876 kcal/mol

More than 18,000 samples? YES

PNear value over 0.75? YES

PNear value to lowest E over 0.75? YES

Lowest energy under 0.65 A RMSD? YES

Sampling below expected lower threshold RMSD(0.3 A)? YES

Sampling beyond 1.5 A RMSD? YES

Sampling beyond 2.2 A RMSD? YES

3+ kcal/mol energy gap? YES

**OVERALL PASS? YES** 

# Peptide NDM1i\_1F:

Total samples = 23274

Computed PNear = 0.9619

Computed PNear to lowest E = 0.9870

Computed DG\_folding = -2.0023

Computed DG\_folding to lowest = -2.6845

Lowest energy = -9.0277 kcal/mol

RMSD of lowest energy = 0.260 Angstroms

Lowest RMSD = 0.189 Angstroms

Highest RMSD = 2.765 Angstroms

Energy gap (minE>1.5A - minE) = 3.2325 kcal/mol

More than 18,000 samples? YES

PNear value over 0.88? YES

PNear value to lowest E over 0.88? YES

Lowest energy under 0.32 A RMSD? YES

Sampling below expected lower threshold RMSD(0.22 A)? YES

Sampling beyond 1.5 A RMSD? YES

Sampling beyond 2.5 A RMSD? YES

3+ kcal/mol energy gap? YES

**OVERALL PASS? YES** 

#### Peptide NDM1i 1G:

Total samples = 23425

Computed PNear = 0.9620

Computed PNear to lowest E = 0.9891

Computed DG\_folding = -2.0040

Computed DG folding to lowest = -2.7951

Lowest energy = -8.9837 kcal/mol

RMSD of lowest energy = 0.284 Angstroms

Lowest RMSD = 0.198 Angstroms

Highest RMSD = 2.713 Angstroms

Energy gap (minE>1.5A - minE) = 6.1895 kcal/mol

More than 18,000 samples? YES

PNear value over 0.9? YES

PNear value to lowest E over 0.9? YES

Lowest energy under 0.32 A RMSD? YES

Sampling below expected lower threshold RMSD(0.25 A)? YES

Sampling beyond 1.5 A RMSD? YES

Sampling beyond 2.6 A RMSD? YES 5+ kcal/mol energy gap? YES OVERALL PASS? YES

Correlation analysis:

R-squared value from fitting greater than 0.85? YES

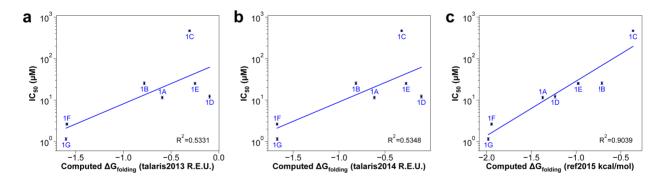
#### **AUTHOR**

Vikram K. Mulligan (vmulligan@flatironinstitute.org), Center for Computational Biology, Flatiron Institute, 5 June 2020.

# PURPOSE OF THE TEST

When designing peptides to bind to targets, the flexibility of the peptide is a major impediment to binding. This is because there is an entropic cost associated with ordering a disordered molecule on binding. A rigidly-structured peptide that is pre-organized in the binding-competent conformation has a lower entropic cost and can bind more tightly. When interactions between peptide and target have been optimized with Rosetta, peptide rigidity becomes the major determininant of binding affinity. Rosetta predictions of peptide folding propensity, carried out with the simple\\_cycpep\\_predict application, correlate strongly with experimentally-measured binding affinity, at least with the ref2015 energy function. This correlation has improved from talaris2013 through talaris2014 to ref2015 due to improved training of the energy function against physical properties of fluids and against fluid MD simulation (see image below). This test exists to ensure that this correlation between prediction and experiment is maintained with future versions of the energy function and with future releases of Rosetta.

Improvements from talaris 2013 through ref 2015:



What does the benchmark test and why?

This test benchmarks seven peptides that were previously designed to bind to and inhibit the New Delhi metallo-beta-lactamase 1 (NDM-1). (For our purposes, IC50 values for these competitive inhibitors, measured at a constant concentration of the substrate of the reaction being inhibited, are proportional to inhibitor binding affinity.) The test carries out the following steps:

1. Large-scale conformational sampling for each of the seven peptides.

- 2. From the energy landscape sampled, calculation of the metrics PNear and estimated Delta-G of folding (which are closely-related measures of propensity to be rigid in the binding-competent conformation).
- 3. Plotting of the logarithm of experimentally-measured IC50 values against the computed Delta-G of folding values computed in the previous step.
- 4. Fitting of these data to a simple linear model, and extraction of the R-squared value for the fit.

The R-squared value should not drop below 0.85.

#### BENCHMARK DATASET

How many proteins are in the set?

- Seven peptides, named NDM1i-1A through NDM1i-1G. (These were previously called 8res1B, 8res2, 8res3, Holmes, Holmes\_18p, Adler, and Moriarty, respectively). All have been synthesized, the IC50 value for each binding to and inhibiting the NDM-1 enzyme's hydrolysis of nitrocefin have been measured.

What dataset are you using? Is it published? If yes, please add a citation.

- These peptides are described in Mulligan \_et al.\_ (2020). Computationally-designed peptide macrocycle inhibitors of New Delhi metallo-beta-lactamase 1. Manuscript in preparation.

What are the input files? How were the they created?

- The input for each structure prediction run is the peptide, designed with Rosetta, in PDB format, along with its sequence in ASCII text format.
- The input for the correlation analysis is the set of output Delta-G of folding values from the structure prediction runs, plus an experimentally-measured set of IC50 values.

#### **PROTOCOL**

State and briefly describe the protocol.

The simple\\_cycpep\\_predict application uses the generalized kinematic closure algorithm (GenKIC) to rapidly sample closed conformations of a heteropolymer macrocycle built from any combination of alpha-amino acids, peptoids, or other related building-blocks. Each closure attempt is relaxed using the FastRelax protocol. For small (~8 to ~10 residue) peptide macrocycles, the application can usually sample close to the native state with less expense than protein \_ab initio\_.

The simple\\_cycpep\\_predict application supports hierarchical MPI-based job distribution and data reduction, as well as multi-threaded parallel job execution within a node. In MPI mode, statistics about the full sampled ensemble, including the funnel quality metric PNear, are computed automatically during the data collection and reduction phase.

*Is there a publication that describes the protocol?* 

The simple\\_cycpep\\_predict application is described in the following publications:

- 1. Bhardwaj G, Mulligan VK, Bahl CD, \_et al.\_ (2016). Accurate de novo design of hyperstable constrained peptides. \_Nature\_ 538(7625):329-35.
- 2. Hosseinzadeh P, Bhardwaj G, Mulligan VK, \_et al.\_ (2018). Comprehensive computational design of ordered peptide macrocycles. \_Science\_ 358(6369):1461-6.
- 3. Mulligan \_et al.\_ (2020). Computationally-designed peptide macrocycle inhibitors of New Delhi metallo-beta-lactamase 1. Manuscript in preparation.

How many CPU hours does this benchmark take approximately?

Approximately 400 CPU-hours. The test runs on 4 nodes, 20 cores per node, for roughly 5 wall hours.

In debug mode, this test takes approximately 10 CPU-hours (1 node, 20 cores, for 0.5 wall hours).

#### PERFORMANCE METRICS

What are the performance metrics used and why were they chosen?

All of the following must be true for the test to pass:

- More than 18,000 successful samples (200 in debug mode).
- Lowest-RMSD sample within a threshold RMSD of the design model. (Varies by peptide.)
- Highest-RMSD sample outside of a thrshold RMSD of the design model. (Varies by peptide.)
- Lowest-energy sample within a threshold RMSD of the design model. (Varies by peptide.)
- Energy gap (gap between lowest-energy sample > 1.5 A and overall lowest) bigger than a threshold value. (Varies by peptide.)
- PNear greater than a threshold value. (Varies by peptide.)
- Most importantly, R-squared value for the linear relationship between ln(IC50) and Delta-G of folding is greater than 0.85.

How do you define a pass/fail for this test?

Failure of any of the above results in an overall failure.

How were any cutoffs defined?

These are based on Rosetta performance as of 5 June 2020. The observed R-squared value reported in Mulligan \_et al.\_ (2020) is 0.9.

#### **KEY RESULTS**

What is the baseline to compare things to - experimental data or a previous Rosetta protocol?

Past iterations of this test and performance reported in Mulligan \_et al.\_ (2020). (See Figure 3 in that paper.)

Describe outliers in the dataset.

NDM1i-1C should have the lowest PNear value and highest Delta-G of folding; NDM1i-1F and NDM1i-1G should have the highest PNear value and lowest Delta-G of folding. These are the worst and best binders, respectively.

#### **DEFINITIONS AND COMMENTS**

State anything you think is important for someone else to replicate your results.

For full reproducibility, see the 92-page supplement to Mulligan \_et al.\_ (2020).

# **LIMITATIONS**

What are the limitations of the benchmark? Consider dataset, quality measures, protocol etc.

The parameters of the correlation are expected to vary from target to target, and for any given target, there are very few data. (Chemically synthesizing a candidate peptide binder and testing its affinity are inherently low-throughput experimental techniques.)

How could the benchmark be improved?

More peptides, and more targets.

What goals should be hit to make this a "good" benchmark?

Well, I think it's pretty good as it is.