# Approximate Counting with Deterministic Guarantees for Binding Affinity Computation

Clément Viricel<sup>1,2</sup>, David Simoncini<sup>1</sup>, David Allouche<sup>1</sup>, Simon de Givry<sup>1</sup>, Sophie Barbe<sup>2</sup> and Thomas Schiex<sup>1</sup>

<sup>1</sup>Unité de Mathématiques et Informatiques Appliquées UR 875, INRA, F-31320 Castanet Tolosan, France, <sup>2</sup>Laboratoire d'Ingénierie des Systèmes Biologiques et des Procédés, INSA, UMR INRA 792/CNRS 5504, F-31400 Toulouse, France





## What is a protein ?

Protein: amino acids (AA) sequence

Protein: Backbone + side-chains





Side-chain have different conformations





![](_page_2_Picture_1.jpeg)

![](_page_3_Figure_0.jpeg)

- Protein Design Objective
- Sequence  $\rightarrow$  structure  $\rightarrow$  function so new function requires new sequence
- Identify sequences that adopt 3D structure with suitable function (enhances activity, control recognition of partners)

![](_page_3_Picture_4.jpeg)

![](_page_4_Figure_0.jpeg)

# **Protein Design**

Issue : Combinatorial Explosion

For a *n* amino acids protein, 20 natural amino acid types  $\Rightarrow 20^n$  sequences

 $\Rightarrow 20^n$  sequences

For a 50 amino acids protein :  $20^{50} \approx 10^{65}$  sequences.

For 1  $\mu$ g/prot  $\Rightarrow \sim 10^{21}$  times the Earth's mass.

The Computational Protein Design (CPD) Goal:

• Increase the odd of finding hits

• Reduce cost and time development How:

- Mathematical model of proteins
- Criteria and algorithms for finding suitable sequences.

![](_page_4_Picture_12.jpeg)

![](_page_5_Figure_0.jpeg)

![](_page_5_Picture_1.jpeg)

# Modeling Protein Flexibility

Usual modeling assumptions:

- Rigid backbone
- Discrete side-chain orientations (rotamers : most frequent conformations)

![](_page_6_Picture_4.jpeg)

Search space = Sequence space x conformation space

![](_page_6_Picture_6.jpeg)

![](_page_7_Figure_0.jpeg)

![](_page_7_Picture_1.jpeg)

![](_page_8_Figure_0.jpeg)

![](_page_8_Picture_1.jpeg)

# Computing Z

- Limited Guarantees
  - Monte Carlo(sampling), mean field, message passing (TRW).
- Exact
  - Cachet, #SAT (SAT solver, caching)
- $(\delta, \varepsilon)$ -guarantees:
  - WISH(+optimisation), MIS: XOR hashing based
  - Gumbel perturbations (+optimization)
- *ɛ*-guarantees
  - OSPREY-K\*

![](_page_9_Picture_10.jpeg)

# Binding Affinity Constant

The binding constant  $K_A$  represents the affinity between two proteins (for each sequence)

![](_page_10_Figure_2.jpeg)

![](_page_10_Picture_3.jpeg)

![](_page_11_Figure_0.jpeg)

![](_page_11_Picture_1.jpeg)

![](_page_12_Figure_0.jpeg)

# K\* algorithm (OSPREY)

DEE removes strongly dominated rotamers

A\* enumeration produces conformations in decreasing order of probability mass

A Novel Ensemble-Based Scoring and Search Algorithm for Protein Redesign, and its Application to Modify the Substrate Specificity of the Gramicidin Synthetase A Phenylalanine Adenylation Enzyme. [Ryan H. Lilien] RECOMB'04

![](_page_12_Picture_5.jpeg)

#### **Tree Exploration**

Enforce invariant:  $U < \varepsilon \hat{Z}$ U : Amount of pruned probability mass  $\hat{Z}$  : Current Z approximation

Initially:  $U \leftarrow 0$ if  $U_b + U < \varepsilon \hat{Z}$ Prune and  $U \leftarrow U + U_b$ else Branch

 $< (1+\epsilon)\hat{Z}$ 

At a leaf : 
$$\hat{Z} \leftarrow \hat{Z} + e^{-\frac{E}{RT}}$$

![](_page_13_Picture_5.jpeg)

# Cost Function Networks (Toulbar2)

- $X = (x_1, ..., x_n)$  set of variables
- $D_i$  set of domains over  $x_i$ ,  $|D_i| \leq d$
- W set of non-negative cost functions  $w_S$  each with a scope S
- Goal: Minimize  $\sum w_S \rightarrow NP$ -hard
- $w_{\emptyset}$  : constant function  $\rightarrow$  Lower bound

![](_page_14_Figure_6.jpeg)

 $w_{\emptyset} = 0$ 

![](_page_14_Picture_8.jpeg)

![](_page_15_Figure_0.jpeg)

#### Local Consistency

Local consistencies transform the problem into an equivalent one, increasing the upper bound  $w_{\emptyset}$ .

![](_page_15_Figure_3.jpeg)

![](_page_15_Picture_4.jpeg)

# Computational Protein Design as Cost Function Network

#### CPD as CFN

- *n* AA positions,  $X = \{P_1, P_2, \dots P_n\}$
- $D_i$  set of rotamers of position  $P_i$
- W pairwise energy functions  $W = \{E(i), \dots, E(i, j)\}$

![](_page_16_Figure_5.jpeg)

[Allouche et al. CP2012] [Allouche et al. Al 2014] [Traoré et al. Bioinformatics 2013]

![](_page_16_Picture_7.jpeg)

# Upper Bound on The Partition Function $Z_0^*$ algorithm:

$$U_b = N \times exp\left(\frac{-c_{\emptyset}}{RT}\right)$$

Takes in account the number of leaves N below the current node

 $Z_1^*$  algorithm:

$$U_b = exp\left(\frac{-c_{\emptyset}}{RT}\right) \prod_{i \in X} \sum_{a \in d_i} exp\left(\frac{-E_i(a)}{RT}\right)$$

Takes in account unary costs

 $Z_2^*$  algorithm:

$$U_b = Z_{STP}$$

Takes in account unary costs + binary costs on a spanning tree

![](_page_17_Picture_9.jpeg)

# Comparison $Z^*_{0,1,2}$ and $K^*$

$\varepsilon = 10^{-3}$	$Z_0^*$		$Z_1^*  \boldsymbol{\nu s}  Z_0^*$		$(Z_1^* + VAC) vs Z_1^*$		$Z_2^* vs Z_1^*$		<i>K</i> *	
PDB ID (#Seq.)	Nodes	Times	Nodes	Times	Nodes	Times	Nodes	Times	Nodes	Times
1ACB (6)	129	0.2 sec	≈ 0%	≈ 0%	pprox 0%	pprox 0%	$\approx -2\%$	pprox 0%	$\propto 10^5$	4,859 min
1AMU (1584)	$8.45 \times 10^{4}$	$\frac{1}{2}$ min	$\approx -23\%$	$\approx -10\%$	≈ +13%	$\approx -21\%$	$\approx -3\%$	≈ +13%	$6.45 \times 10^{6}$	1,278 min
3SGB (173)	$2.2 \times 10^{6}$	30 min	pprox 0%	pprox 0%	pprox 0%	$\approx -5\%$	$\approx -10\%$	≈ +35%	$\infty$	ω
1TP5 (1121)	$3.19 \times 10^{6}$	31 min	$\approx -51\%$	$\approx -47\%$	pprox 0%	≈ -75%	≈ -36%	≈ +11%	$\infty$	$\infty$
1B74 (1809)	$5.64 \times 10^{6}$	85 min	$\approx -41\%$	≈ -35%	$\approx +1\%$	$\approx -70\%$	$\approx -9\%$	≈ 17%	$\infty$	$\infty$
2Q2A (4716)	$39.9 \times 10^{6}$	590 min	≈ -56%	$\approx -45\%$	$\approx -1\%$	≈ -72%	$\approx -5\%$	$\approx +4\%$	00	00

Limit time out: 250 h 64 GB RAM & 1 proc

![](_page_18_Picture_3.jpeg)

![](_page_19_Figure_0.jpeg)

![](_page_19_Picture_1.jpeg)

![](_page_20_Picture_0.jpeg)

#### Acknowledgements

 Biometrics & Artificial Intelligence Unit

![](_page_20_Picture_3.jpeg)

- George Katsirelos
- Simon de Givry
- Thomas Schiex
- Catalysis & Enzyme Molecular Engineering Team
- Sophie Barbe

![](_page_20_Picture_9.jpeg)

![](_page_20_Picture_10.jpeg)

![](_page_20_Picture_11.jpeg)

![](_page_21_Figure_0.jpeg)

![](_page_21_Picture_1.jpeg)