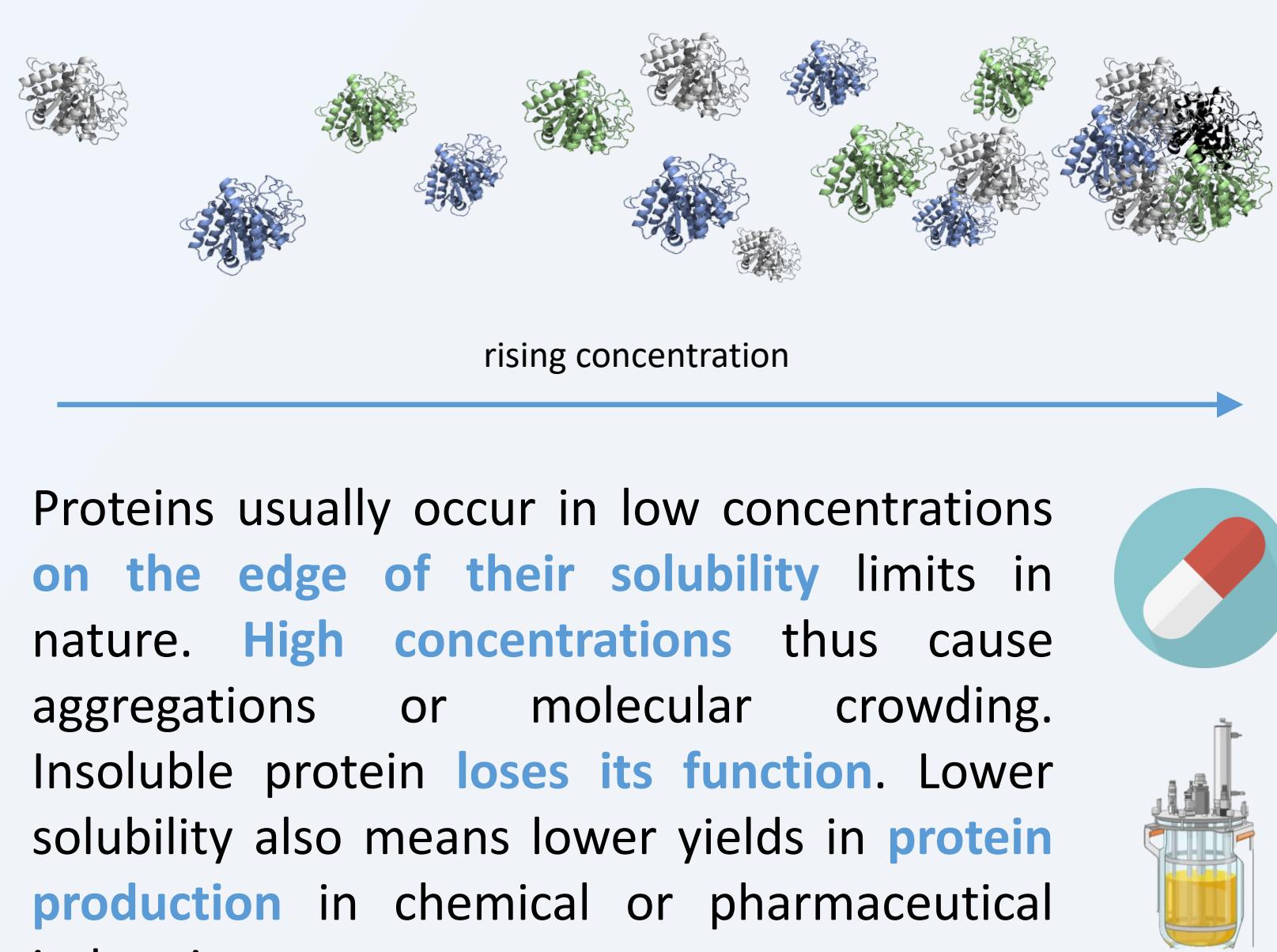


# Protein Learning for Solubility Design

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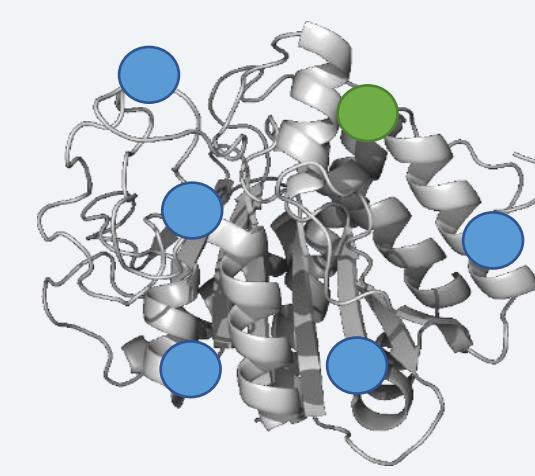


## Motivation



## Solubility & its importance

- Ability to freely dissolve in a solution
- Evolutionary stabilized solubility of natural proteins
- Design of protein properties = **choice of mutations**



## Protein design



## Solubility predictors

Solubility-change predictors:

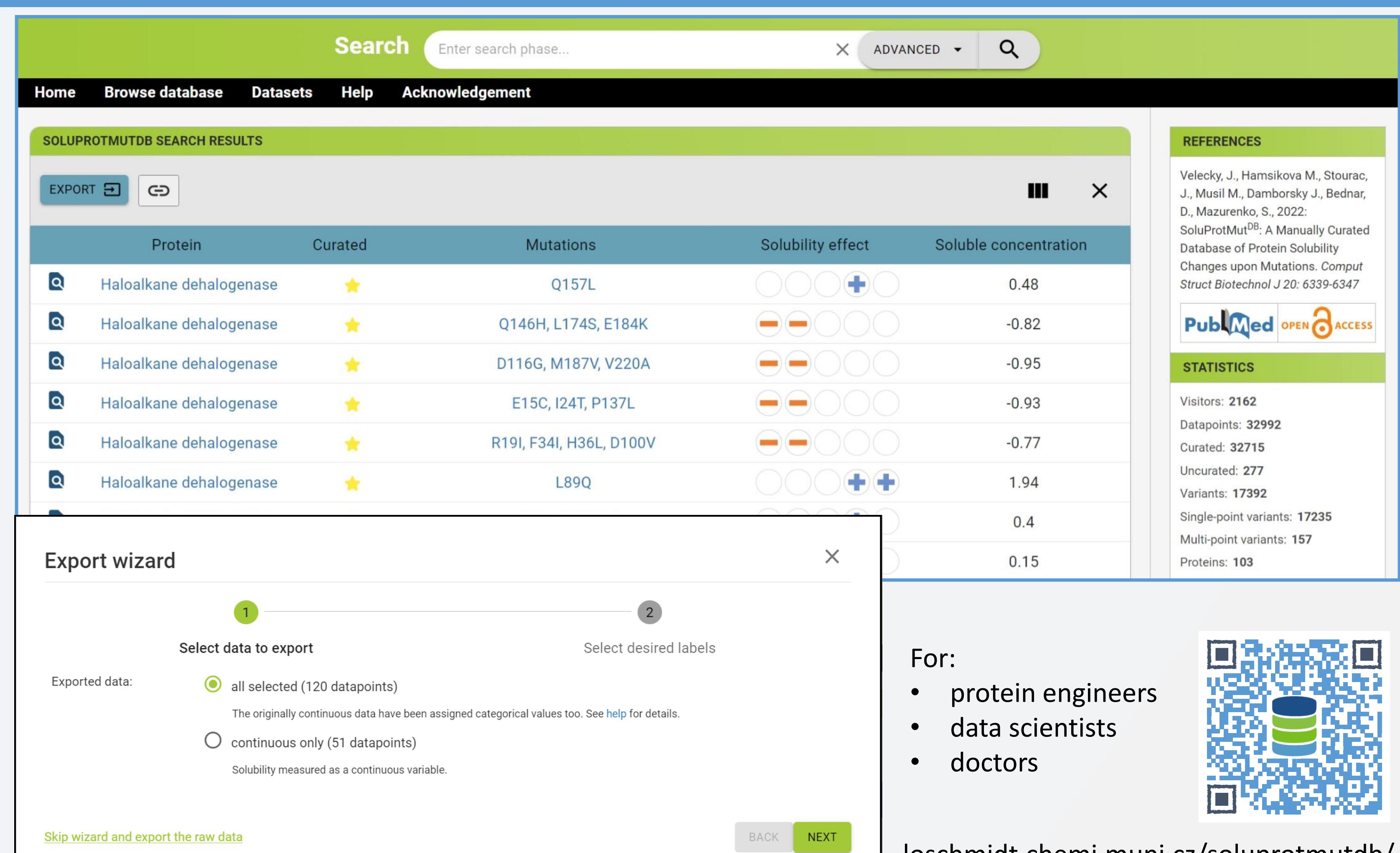
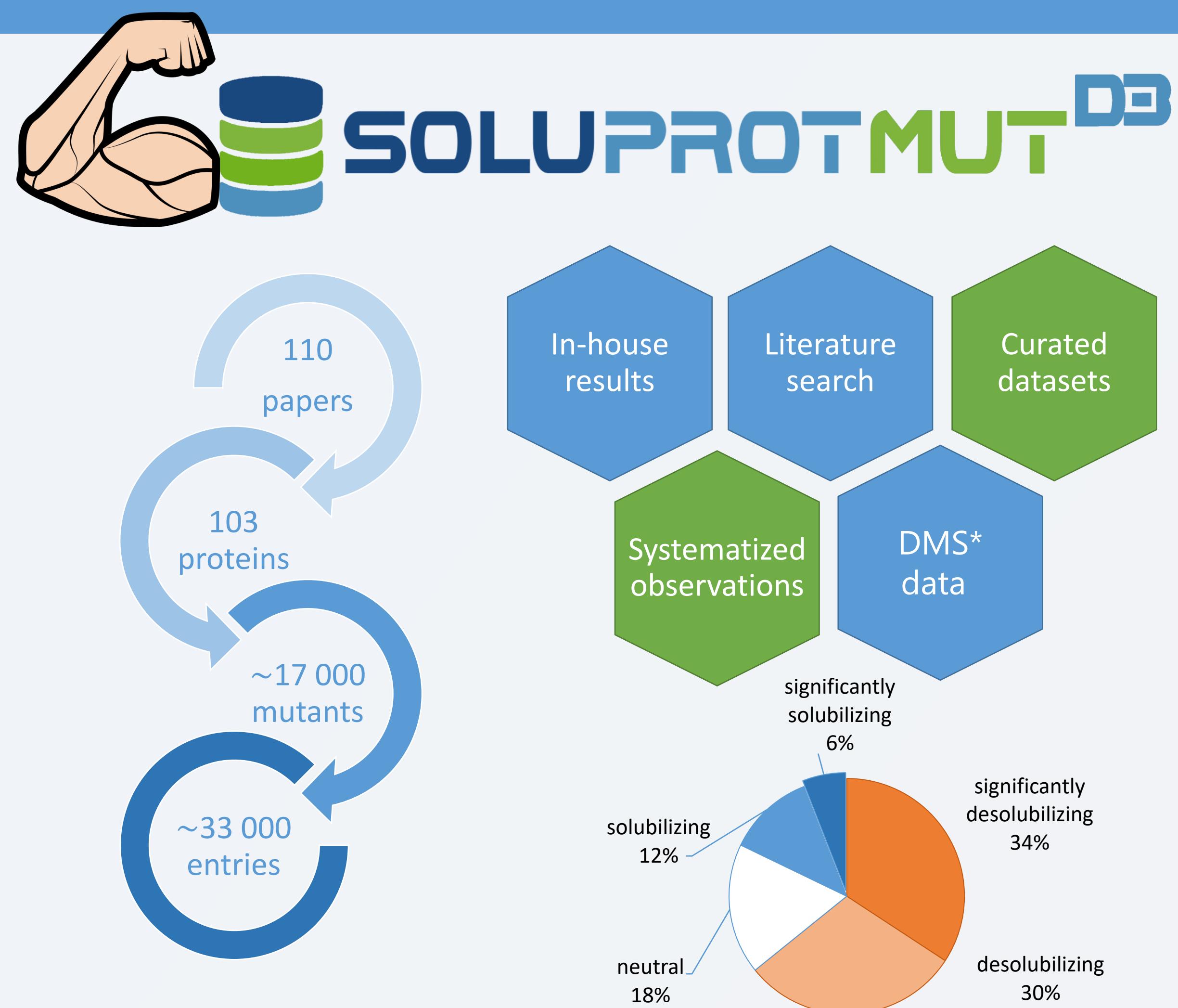
- OptSolMut (2010)
- CamSol (2014)
- PON-Sol (2016)
- SODA (2017)
- PON-Sol2 (2021)

Reported **accuracy of 60–80 %**. Even lower solubilizing precision.

Soluble protein predictors:

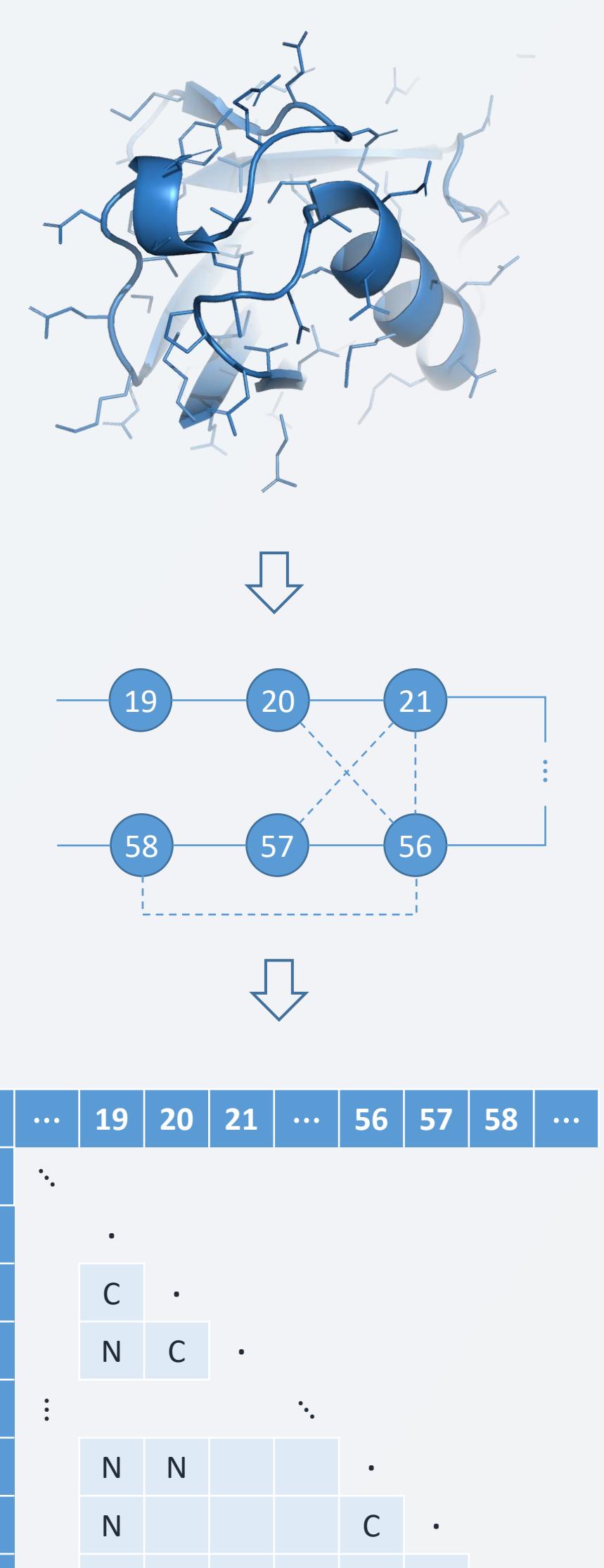
- Either soluble, or insoluble
- SoA\* balanced accuracy up to 59 %
- Mutational prediction up to 57 %

## Data: the database of mutational solubility data

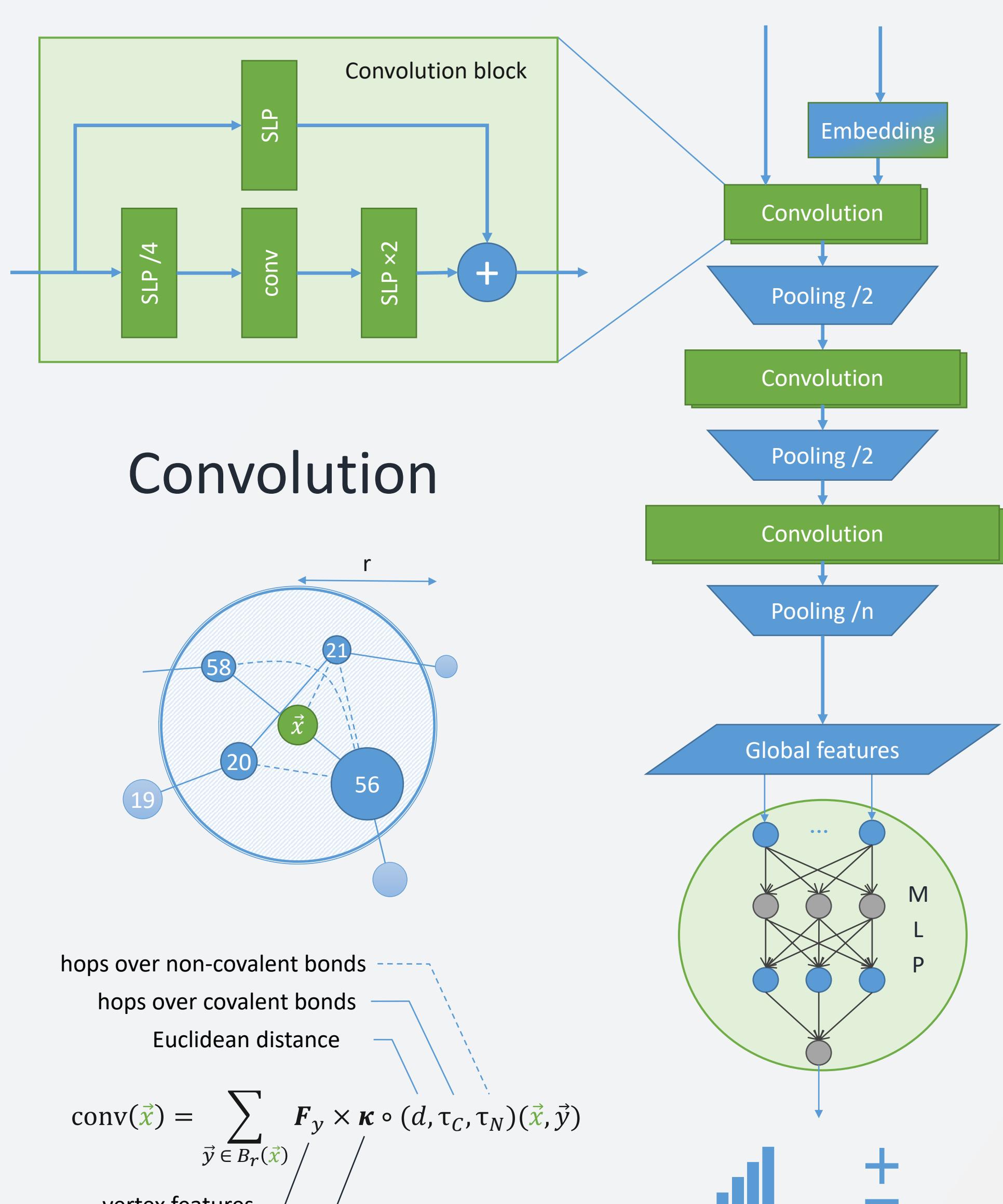


## Model: protein convolutional neural network

### Representation



### Architecture



- Graph-based convolutional network for **protein learning**

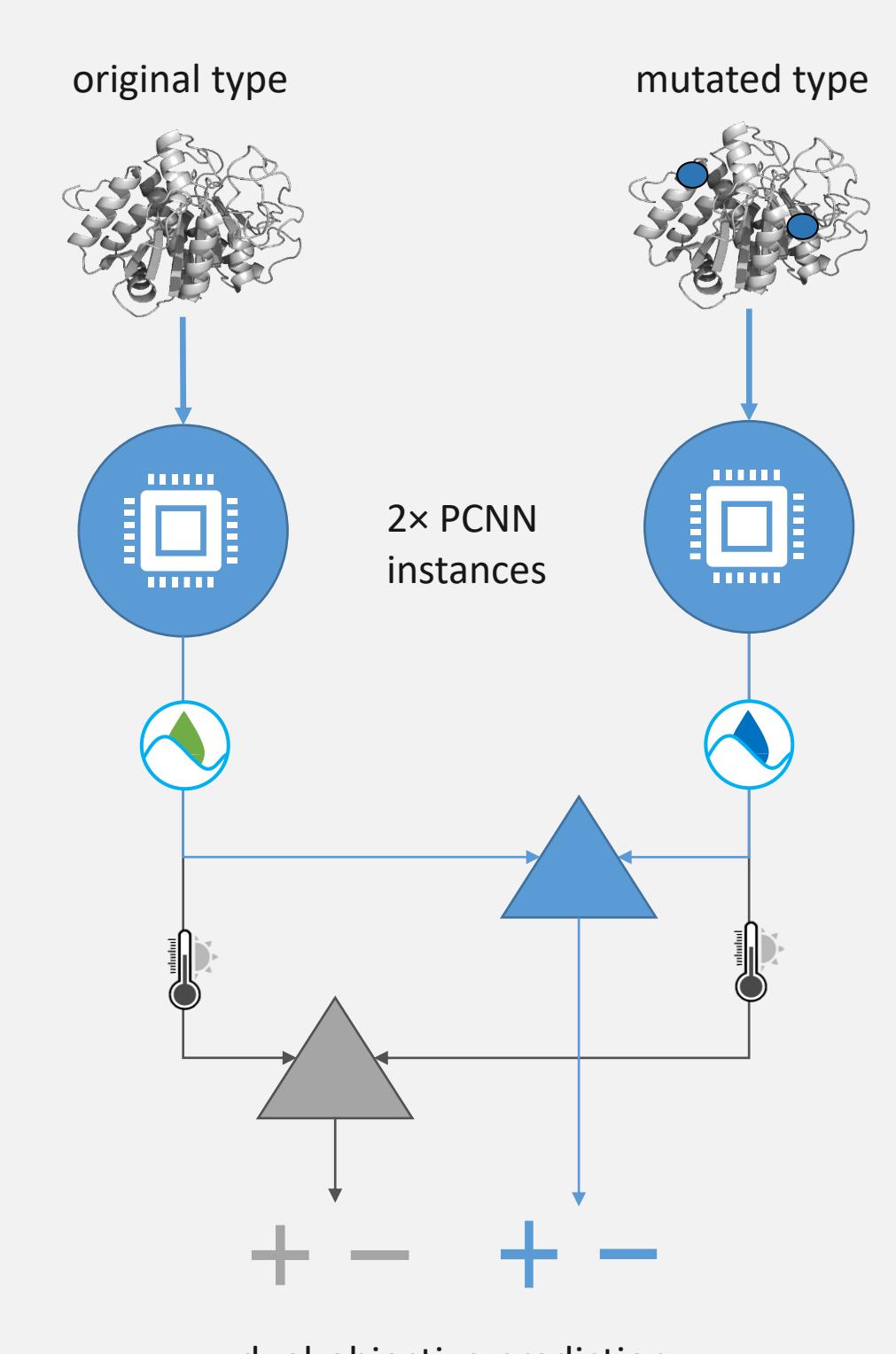
- Vertex-weighted multi-graph representation of protein structure
- SoA\*** prediction of EC\* numbers
- Rewritten, simplified and optimized under the author's supervision
- Expertise brought from Dr. Pedro Hermosilla from Visual Computing Group, Ulm University, Germany



- Protein Convolutional Neural Network (PCNN) displays perceptivity to **local spatial features**
- Meanwhile, is **invariant to rotations & translations** of a protein structure
- Appears to be a perfect model for mutational predictions: stability, solubility, **dual-objective**

### Action plan

#### Model training



#### Augmenting data

Reverse mutations	two-fold
Stability changes	two-fold
Soluble/insoluble	two-fold
PDBs*	30-fold
EC* numbers	three-fold