# I) MOLECULAR TARGETS & THERAPEUTICS CENTER





# OCHEM consensus model wins Kaggle solubility challenge

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We outline our success in the EU-OPENSCREEN - a not-for-profit European Research Infrastructure
Consortium (ERIC) - and The Society for Laboratory
Automation and Screening (SLAS) solubility challenge.
The challenge was established to identify the state-of-the-art computational methods for reliable predictions of threshold solubility of compounds. Here, we present our consensus model which was the winning solution amid 100 contributing teams.



### https://ochem.eu

#### Imbalance of data Workflow with OCHEM Online chemical database → Model calculation thods Linear Regression Random forests Boosting Deep neural networks Convolutional NN Standardization Neutralize Remove salts Clean structure Public scorin in 28 single mode Evaluation 93% Metrics - ROC-AUC ■ high ■ medium ■ low (Balanced) accuracy RMSE mption of class freq

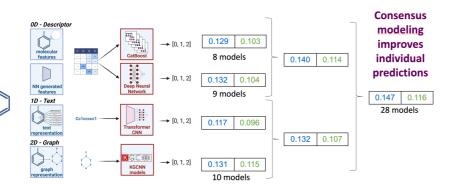
LSSVMG ASNN PLS KNN ALogPS, OEsta 0.68 0.61 0.64 0.74 0.75 0.71 CDDD 0.56 0.71 0.59 0.68 ors (pH 0 - 14:1) 3D:c 0.7 0.59 0.65 Dragon6 (2D blocks: 1 28) 0.64 6 (3D blocks: 1-29) 3D: 0.59 0.63 entor (length:2 - 4) 0.72 0.69 0.71 0.57 0.67 0.59 MAP4 0.65 0.67 0.55 0.67 Mera, Mersy 3D:corina OEstate 0.69 0.7 PyDescriptor 3D:corina 0.71 0.71 0.67 RDKIT (3D blocks: 1-11 15-16) 3D: 0.72 0.56 0.65 0.59 0.67 0.6 trophores (accuracy=20) 3D:corina 0.68 0.6 0.52 alvaDesc (3D blocks: (only) 1-30) 3D:corina 0.71 0.57 0.68

Challenge setup: classification of highly imbalanced data into three ordered classes

Workflow used for model development

**Example of models developed using OCHEM** 

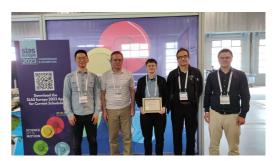
## Quadratic kappa metric scores $\rightarrow$ Public leaderboard Private leaderboard



Overview of molecular representations and models. Molecules were represented by descriptors, SMILES text, and graph representations. Subsequently, descriptor-based models including CatBoost and DNNs, TransformerCNN and KGCNN models were generated and tested. Combinations of models improved performance.

### Conclusions:

- Concensus modellling provided the best accuracy
- Different representations enhance the performance
- Reliable protocol is important to get best results
- Do not give up!



Challenge winners (Ms. A. Kopp is in the centre) with the challenge organizers during 2023 SLAS conference in Brussels

See pre-print at <a href="https://doi.org/10.26434/chemrxiv-2023-p8qcv">https://doi.org/10.26434/chemrxiv-2023-p8qcv</a>



