#### HelmholtzZentrum münchen German Research Center for Environmental Health



# Performance, Reliability and Robustness - A comparison of several experimental design strategies

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#### Motivation

• REACH legislation: Each chemical compound produced in or imported into the EU in an amount of more than one ton has to be registered according to a number of endpoints • In case of hazardous, dangerous or toxic compounds, these endpoints contain toxicity and bio-accumulation Experimental determination of all these values is not possible, as experiments consume a lot of time, money (estimated to €9.5 billion)

### **Selection approaches**

#### **Space filling designs**

- Usable only for few dimensions
- Chemical compounds are not

#### **Dissimilarity selections**

• Outlier detector in higher dimensional spaces

### Similarity search

- Bias towards the central region
- Disregarding the periphery

### **Required specifications**

#### Low prediction error

- Minimize average prediction error •
- Minimize the error of the worst sample
- Stability •
  - Low standard deviation in performance
  - Consistent development of performance
- Flexibility
  - Adaptability to small variations  $\bullet$
- Robustness •
  - Against small modifications in the dataset •
  - Against structural outliers  $\bullet$
- Reliability
  - Correlation between the number of selected  $\bullet$ compounds and the resulting performance



### **Chemical space representation**

#### k-Medoid approach

- Non-adaptive
- Using principal components
- **Based on space filling idea**

#### **PLS-Optimal**

- Stepwise execution
- Using PLS latent variables
- **Based on dissimilarity**

#### DescRep

Sel

- Stepwise execution
- Using selected descriptors
- **Based on similarity**



#### Datasets

**Boiling point** 

- 1198 compounds
- muted restrictions
- low complexity









#### LC<sub>50</sub>

- 535 compounds
- no restrictions
- high complexity



#### logK<sub>oc</sub>

- 648 compounds

- no restrictions
- average complexity

#### Validation

#### Bagging

## Performance # R2 g2 RMSE MAE Training set: k-Medoid logKOC 648 records 0.81 0.81 0.55 0.42





#### Results

Referring to a binomial test, models resulting from a selection based on

- adaptive approaches
- clustering approaches

perform significantly better concerning

- RMSE
- $\mathbf{Q}^2$
- correlation coefficient

#### In terms of

- all tested endpoints
- both external and internal validation sets
- each examined size of the dataset (250-5000 compounds)

#### Conclusion

- k-Medoid provides the best performance for all examined datasets
  - DescRep is robust against structural outliers
- Adaptive approaches help to stabilize the performance and to increases the reliability
- The major influence regarding the quality of resulting models is the

