

Applications of QSAR in drug discovery

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Computational Toxicology

Safety Assessment

AstraZeneca R&D

The Pharmaceutical Discovery Process

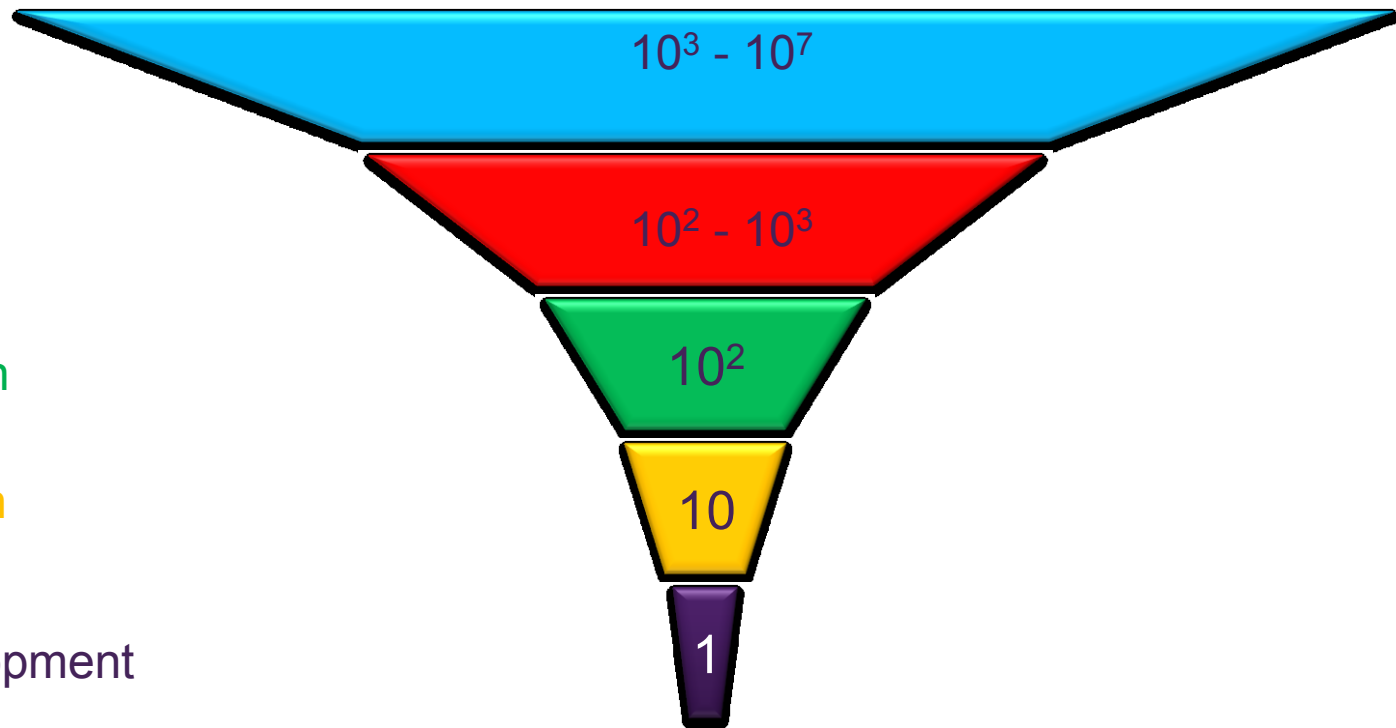
HTS
Virtual Screening
Library Design

Hit Identification

Lead Identification

Lead Optimization

Preclinical Development



Experimental Hierarchy

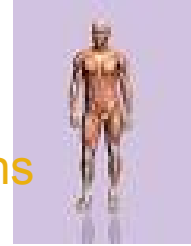
Method

Experimental Requirements

Clinical Trials



Humans



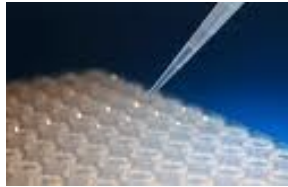
In Vivo



Animals



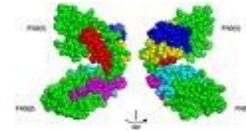
In Vitro



Cells



Biomolecules



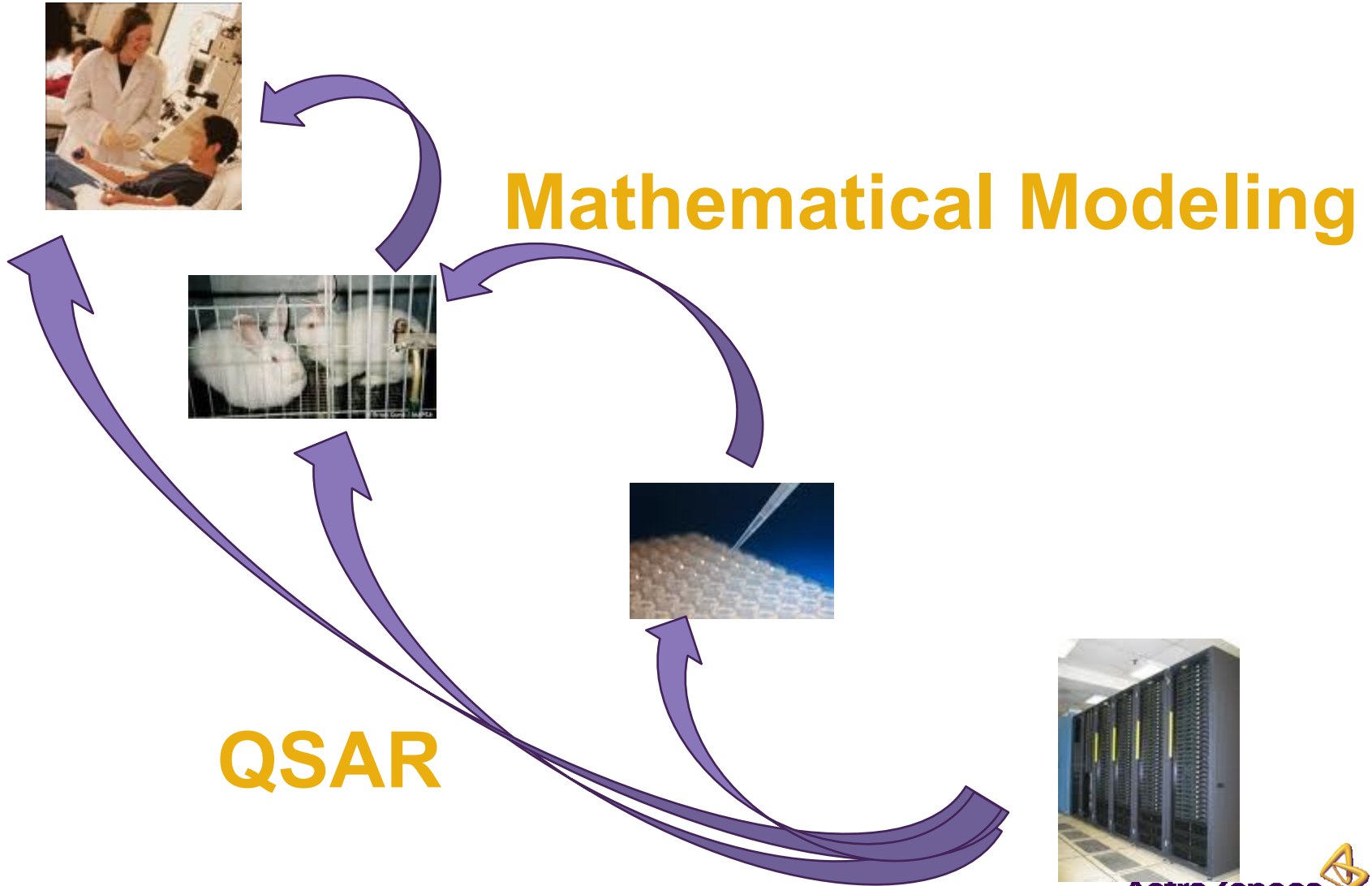
Molecular Structure



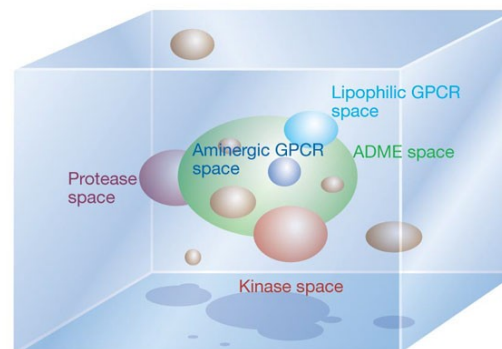
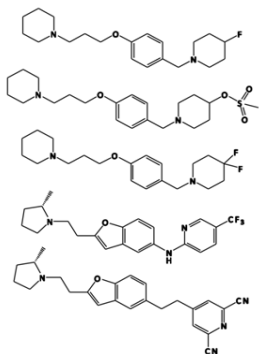
In Silico



Frontloading



Descriptive versus Predictive QSAR



**Descriptive QSAR/
Local QSAR/
SAR**

**Predictive QSAR/
Global QSAR/
QSAR**

Predicting congeneric chemical series

Predicting greatest possible part of chemical space

Impact on activity of small structural change

Which compounds are most likely to be active

Few compounds

As many compounds as possible

Free Wilson Analysis, PLS

Advanced statistical methods

Late discovery

Early and late discovery

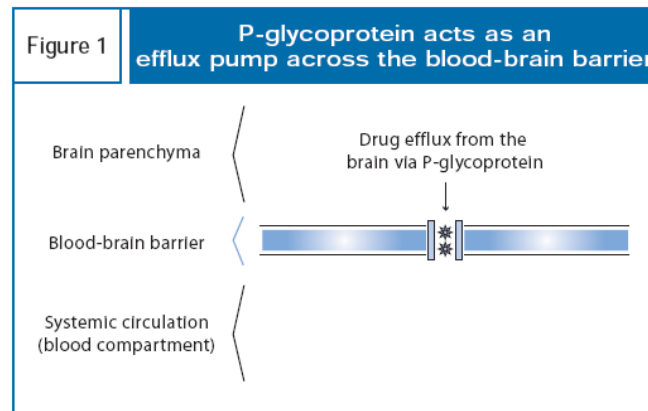
Predictive QSAR within Pharma

■ Physiochemical properties (QSPR)

- pKa
- Solubility
- logP

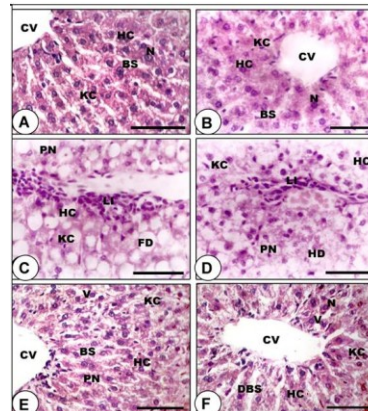
■ DMPK

- Permeability
- Plasma protein binding



■ Safety

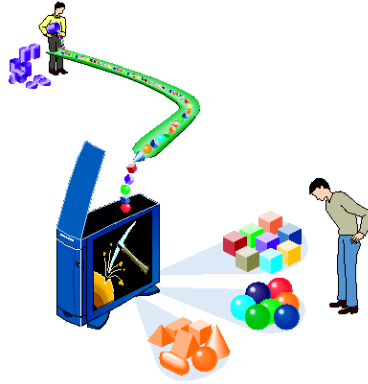
- Liver toxicity
- Carcinogenicity
- Seizure
- Off target activities



QSAR for Safety Risk Assessment

REMOVAL

Library Design



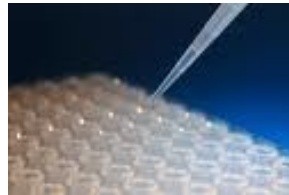
Positive prediction → Remove compound
Negative prediction → Keep in library

Few false positives

REMOVAL

EXPERIO

Early Discovery



Positive prediction → Remove compound/
Early in vitro testing
Negative prediction → Keep compound

Few false negatives

EXPERIO

Late Discovery



Positive prediction → Early in vivo testing
Negative prediction → Postponed in vivo test

Few false negatives

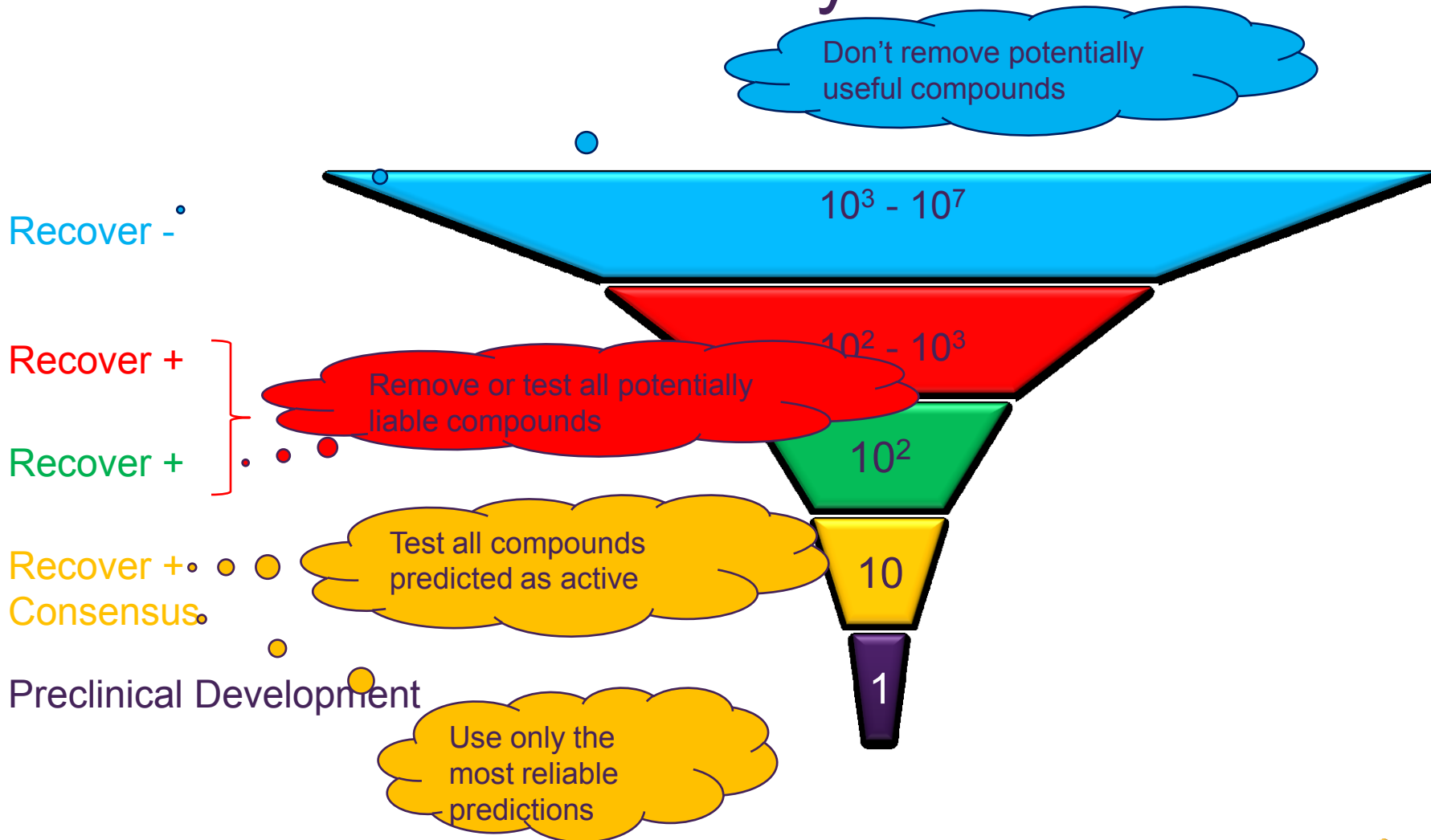
hERG Classifier – Pfizer 2005^a

- 60 000 in house compounds
- 12 000 selected as external test
- Balanced binary classifiers

Classifier	Concordance	Sensitivity	Specificity
ANN	85	86	83
Naïve Bayesian	82	84	80
Recover +	82	92	75
Recover -	85	79	89
Consensus (86% predicted)	89	91	87

a) O'Brien *et al*, J. Med. Chem., 2005, 48

Applications within Pharmaceutical discovery



Process for global QSAR model development and maintenance



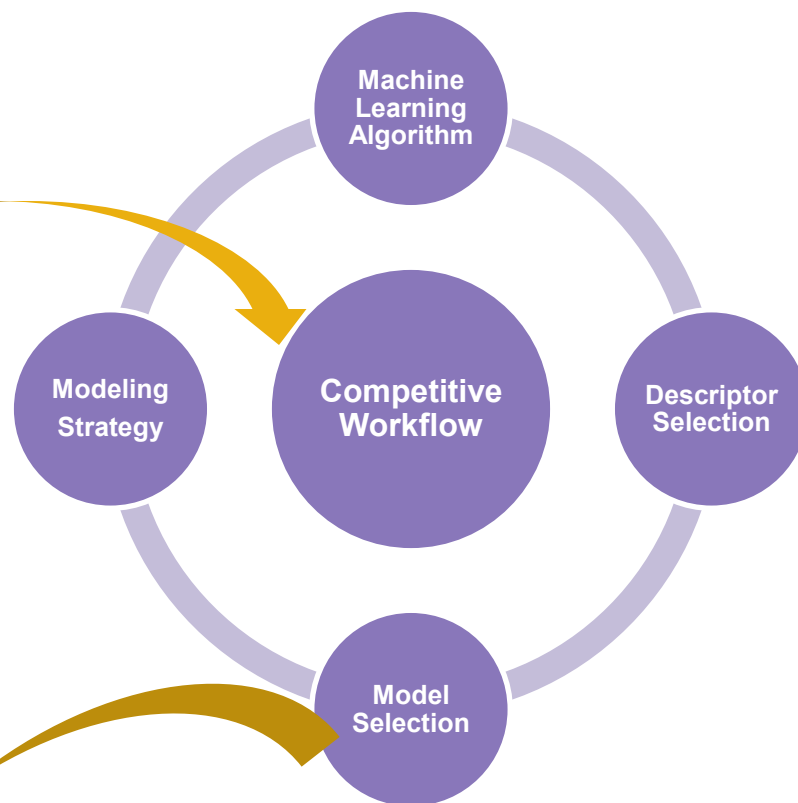
High throughput screens



Global databases

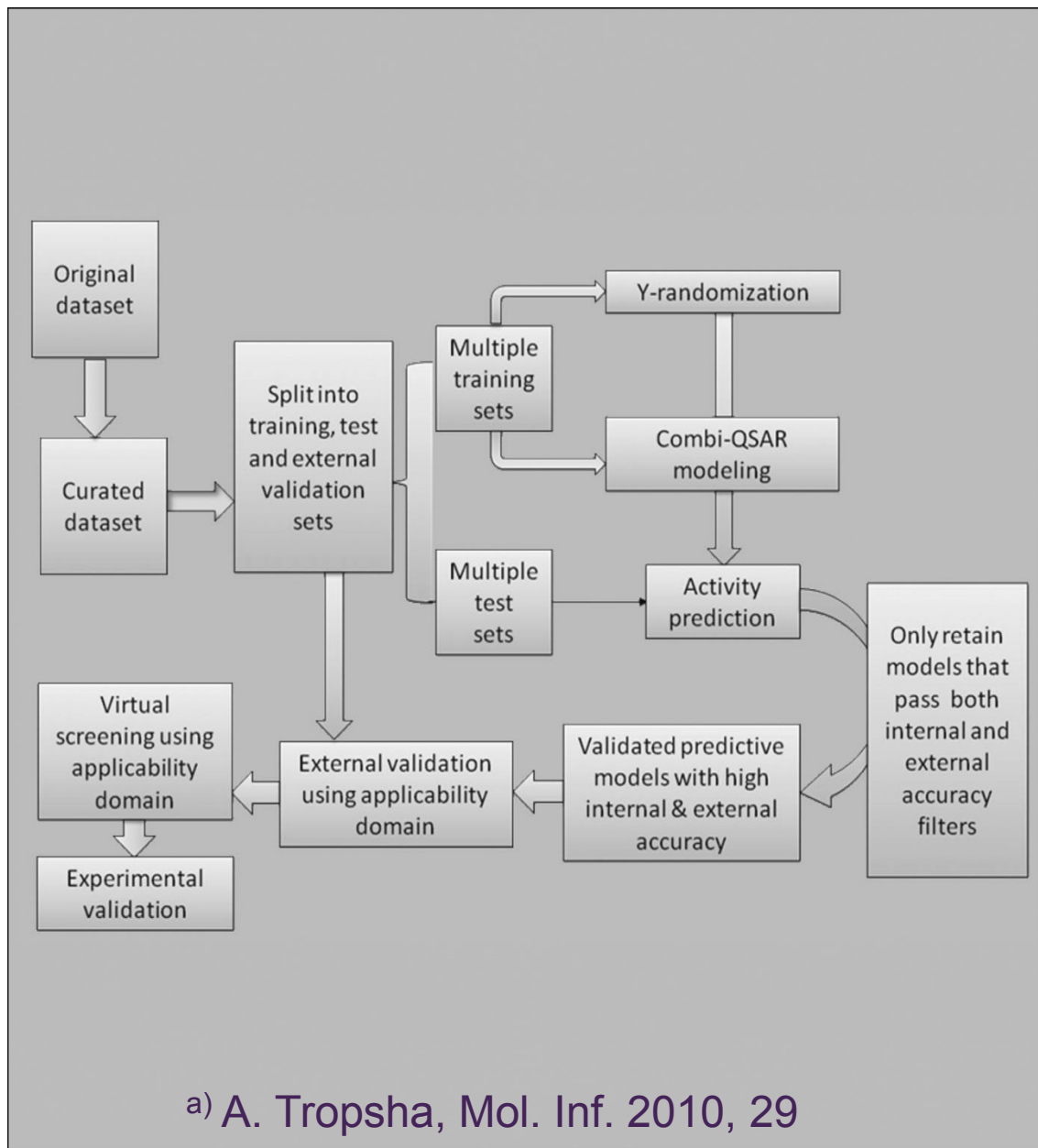
AZorange

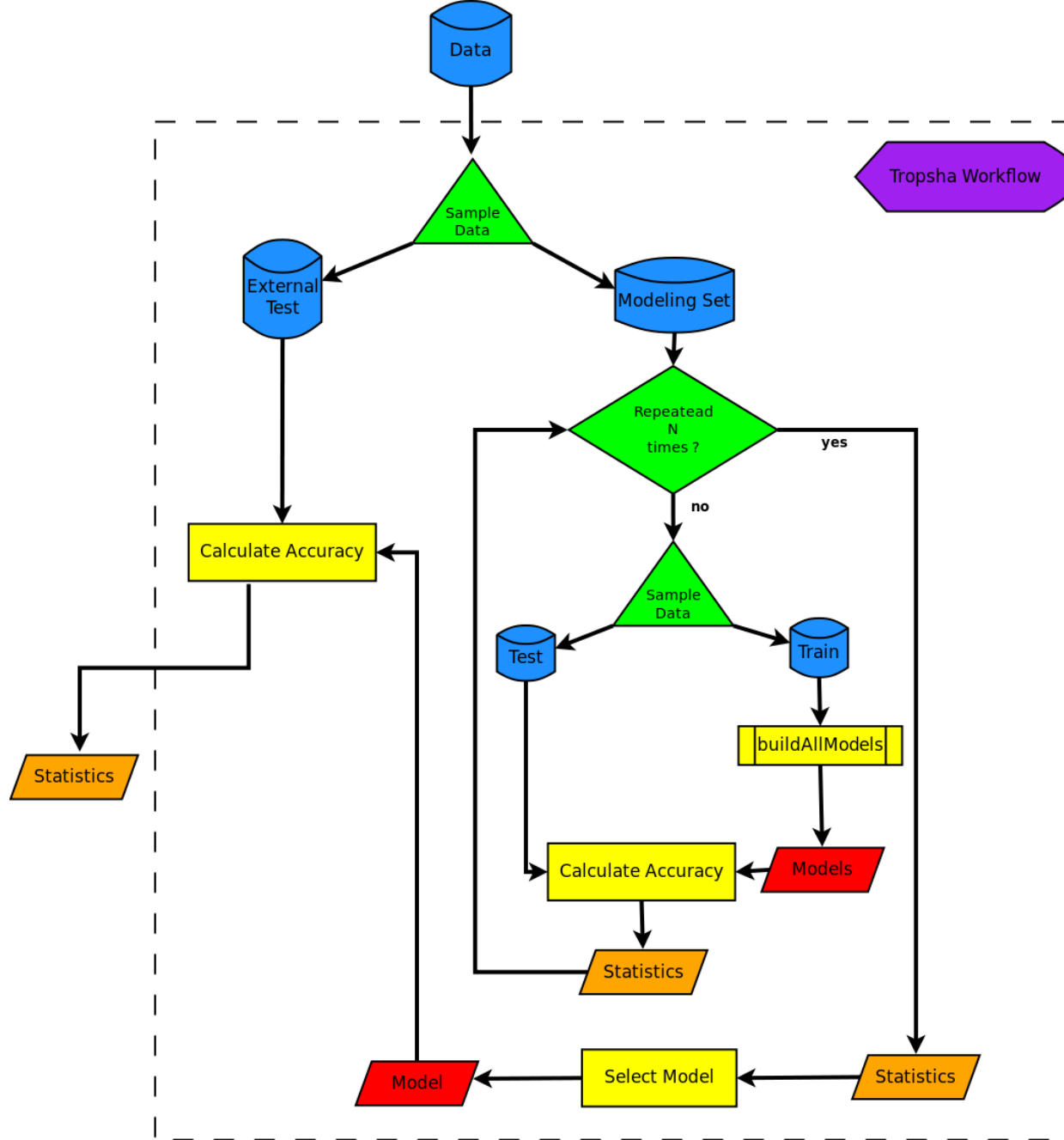
Statistical Modeling

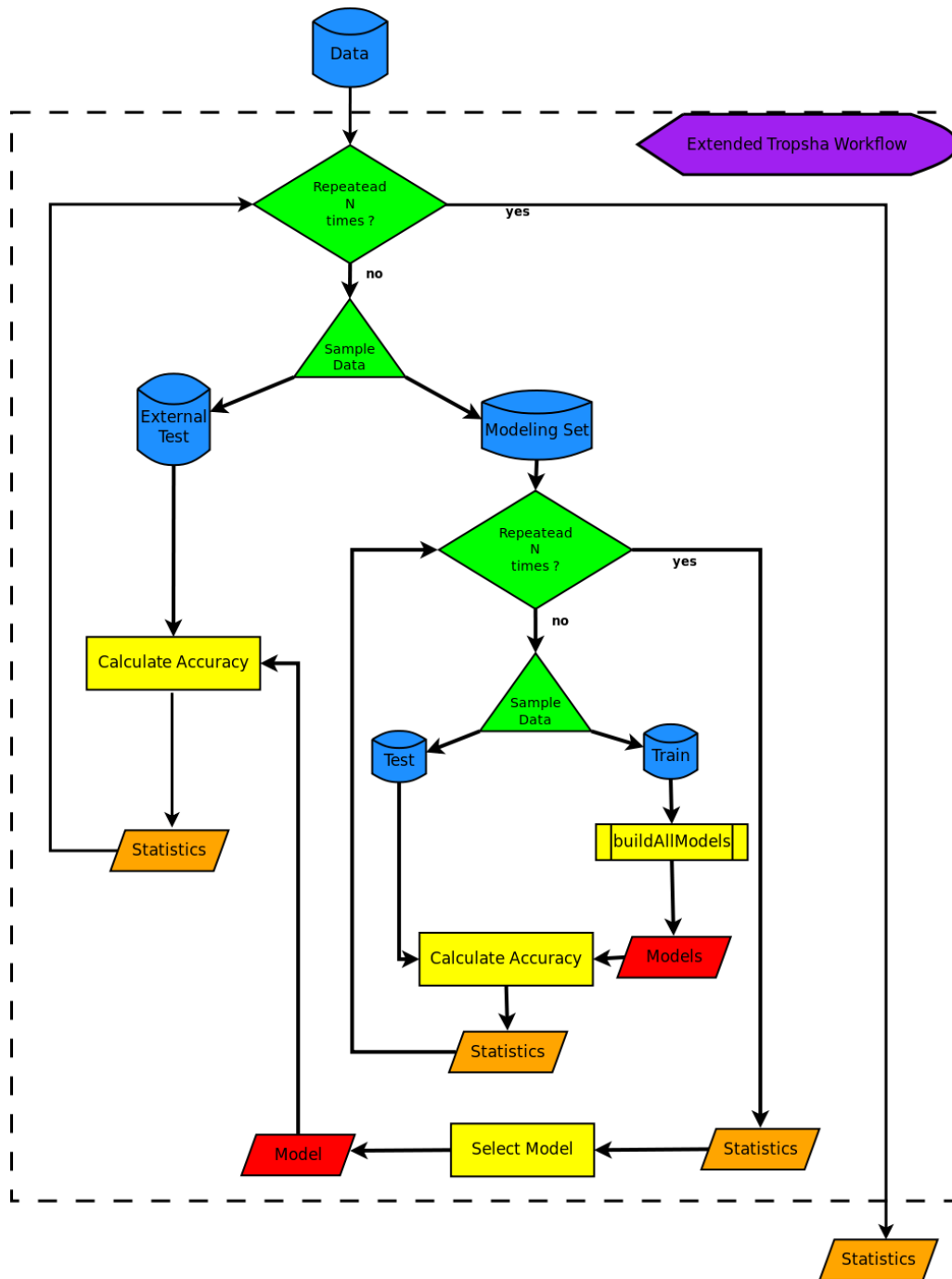


Best Practices for QSAR Model Development, Validation and Exploitation

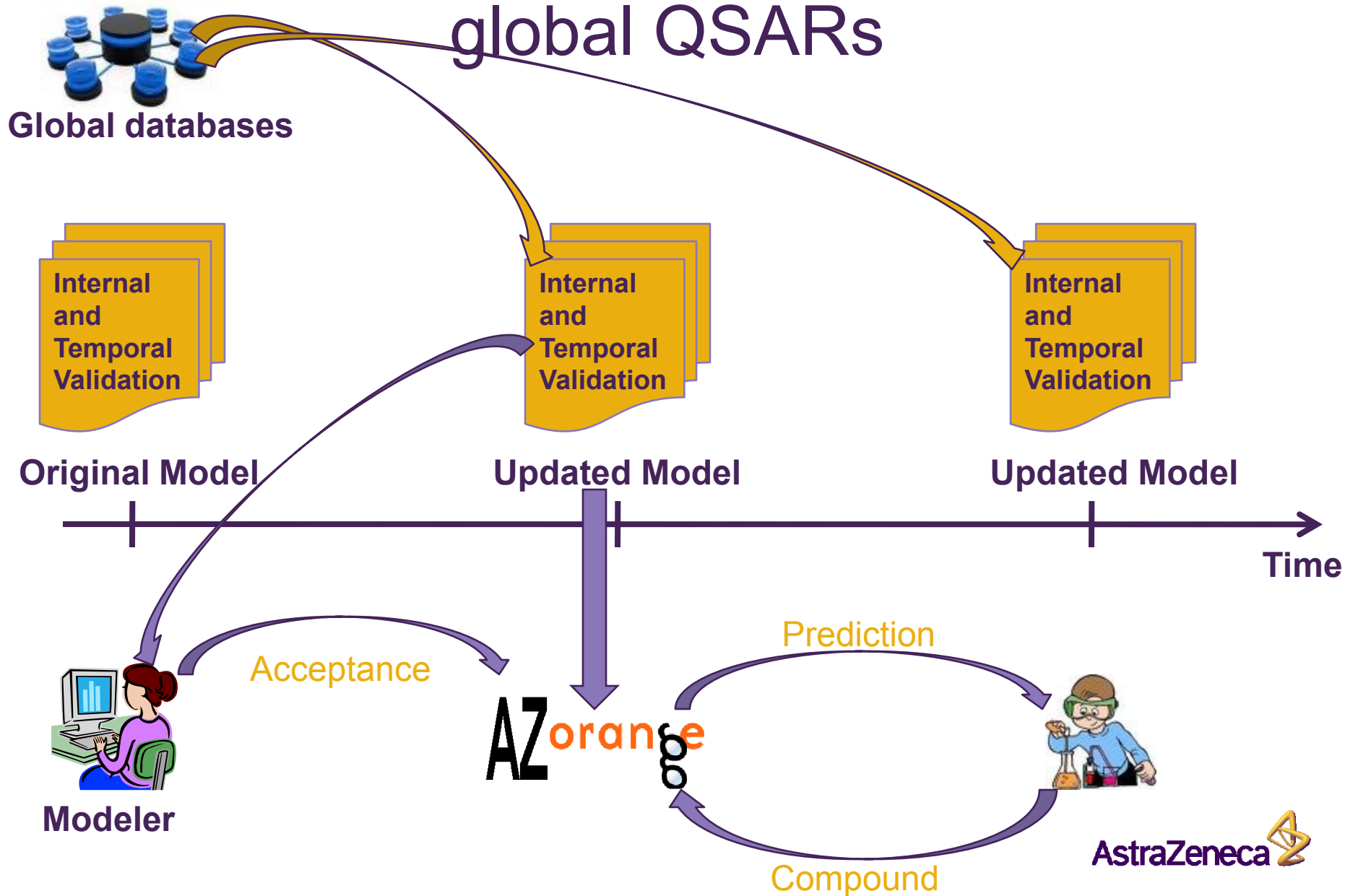
Review by Alexander Tropsha 2010 ^a







Automated Updating and Validation of global QSARs



Public Safety Models for Drug Discovery

Public

QSAR TOOLBOX

The OECD QSAR Toolbox
for Grouping Chemicals
into Categories

Open Q S A R



OpenTox

eTOX



- Method Development
- Small Public Data Sets

Pharmaceutical



- Standardized experiments
- Large Proprietary Data Sets
- Established Methods

Collaboration

Applications of QSAR in drug discovery - Summary

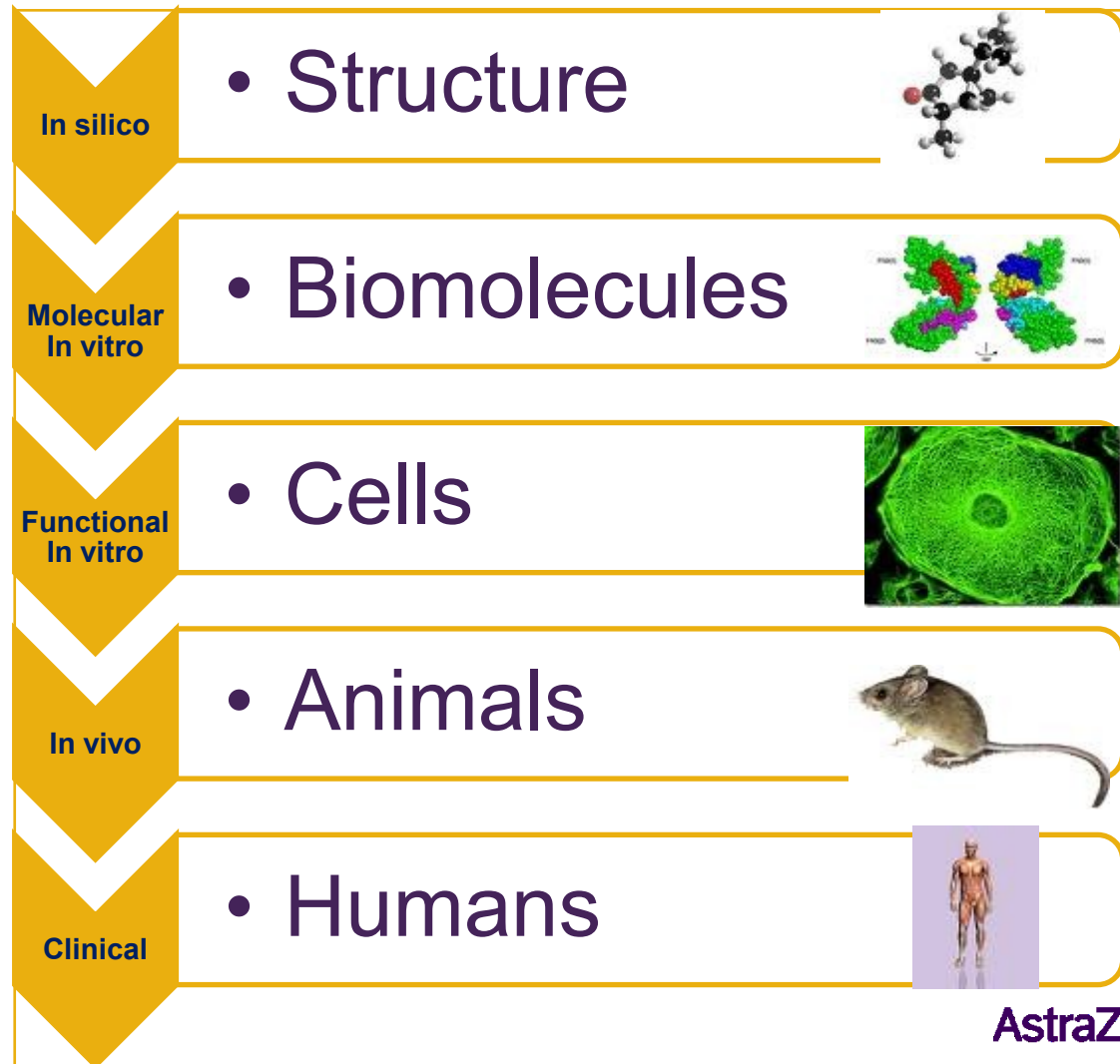
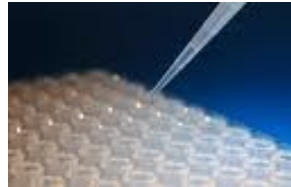
- Drive towards frontloading in pharma
- In silico and in vitro replace in vivo
- Global QSAR mainly in ADMET modeling
- Competitive workflow
- Automatic updating and temporal validation
- Public safety models and methods

Extra Slides

Experimental Hierarchy

Method

Experimental Requirements



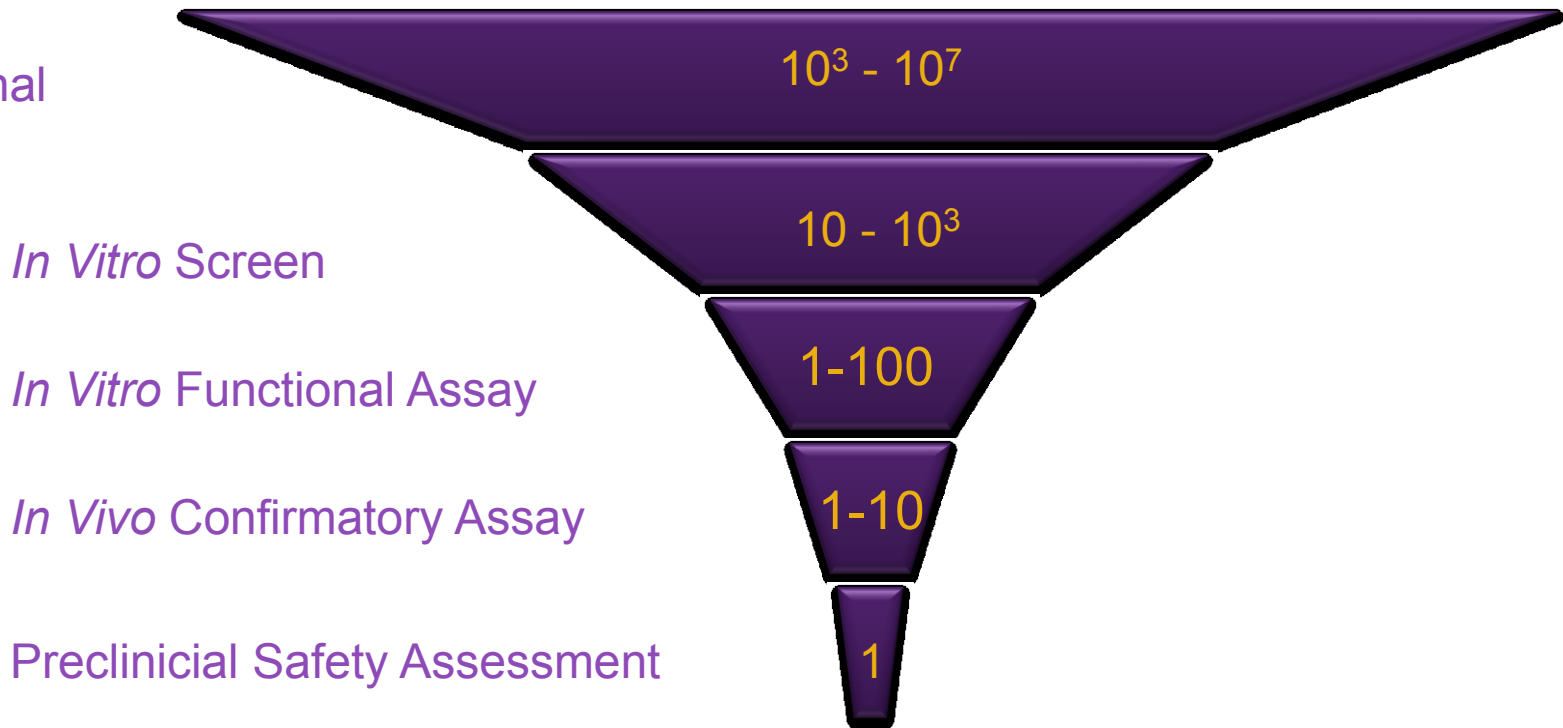
Multi-mechanistic responses

- Examples – Genotoxicity
- Ames assay, strains
- Local/global modeling
- MTC

General Safety Strategy

Computational

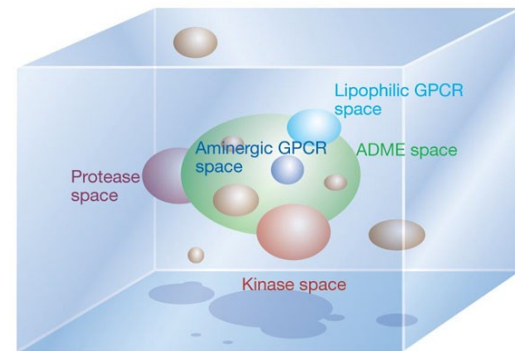
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Challenges in global toxicological QSAR modeling in pharma

- Diversity of chemical space
- Multi-mechanistic biological processes
- Lack of structural information

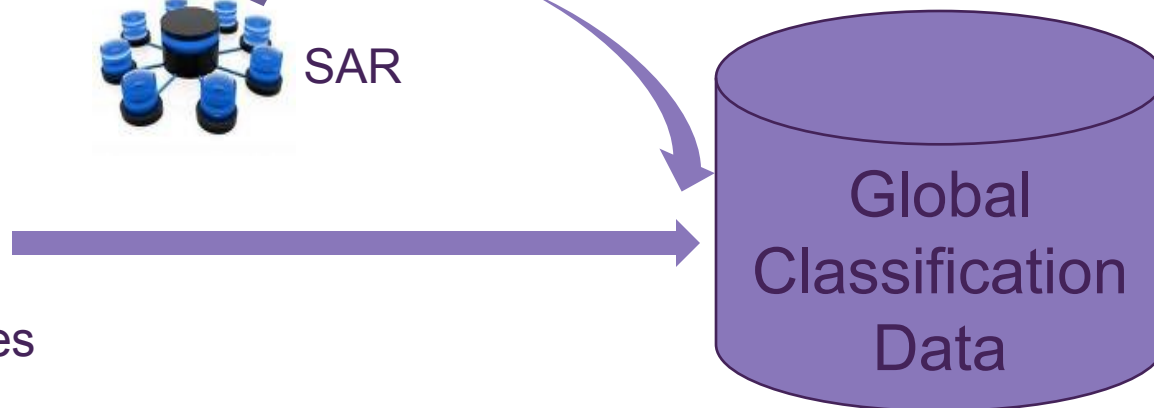
Diversity of Chemical Space



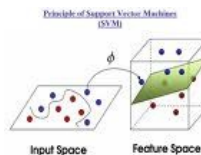
- Merging data from various sources



NIH Molecular Libraries

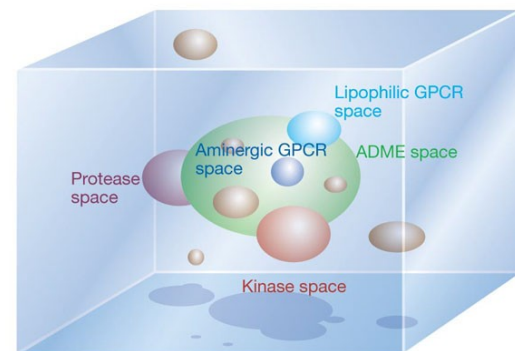


- Automated updating and validation



Predictions & Validation

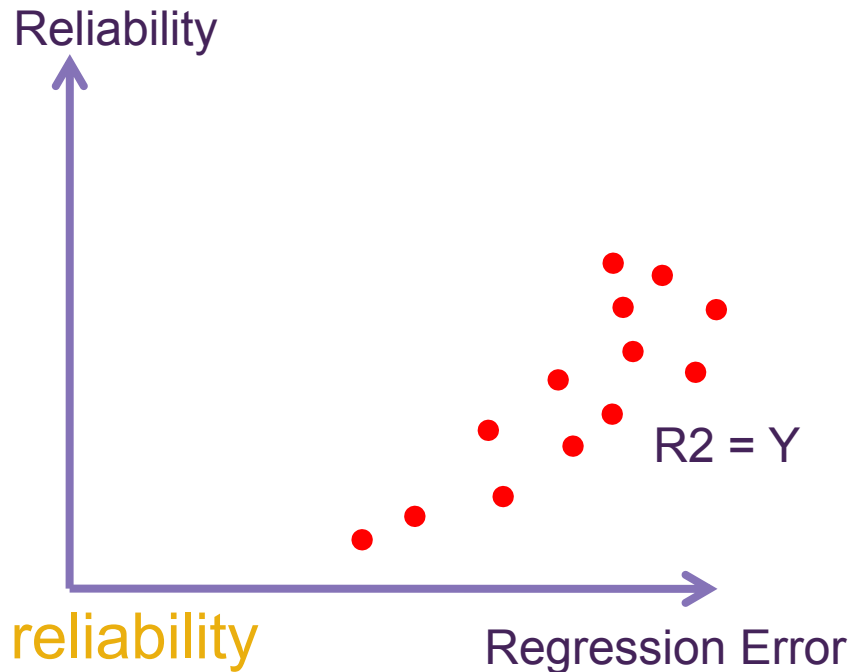
Diversity of Chemical Space



- Developing the concept of the applicability domain

“The applicability domain of a QSAR model is the response and chemical structure space in which the model makes predictions with a given reliability”

- Half of the data sets show a positive correlation between reliability metrics and prediction error ^a



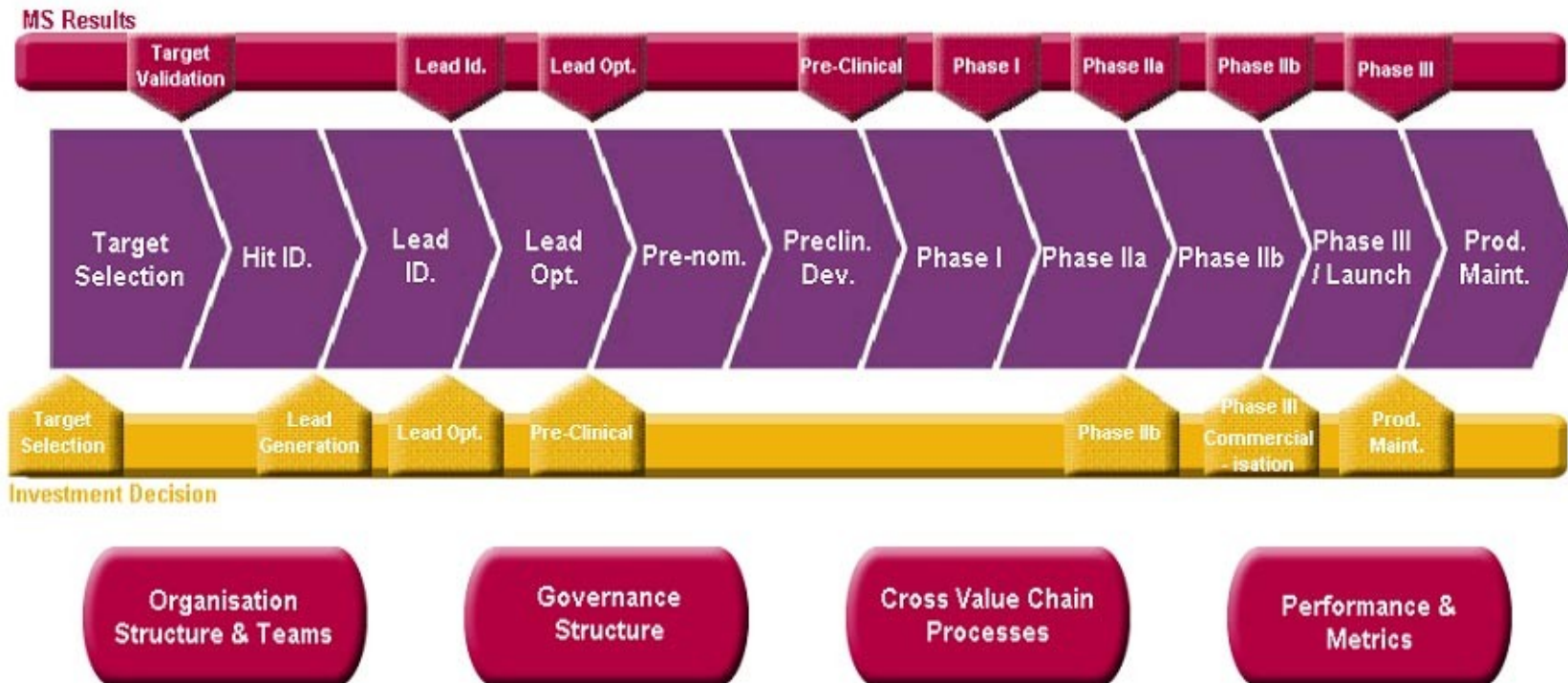
^a) Bosnic, The Knowledge Engineering Review (2010), 25: 27-47

Multimechanistic Endpoints

- Local/Global Modeling
- Structural class similarity as descriptor
- MTC

PLD example

Pharmaceutical discovery and development process

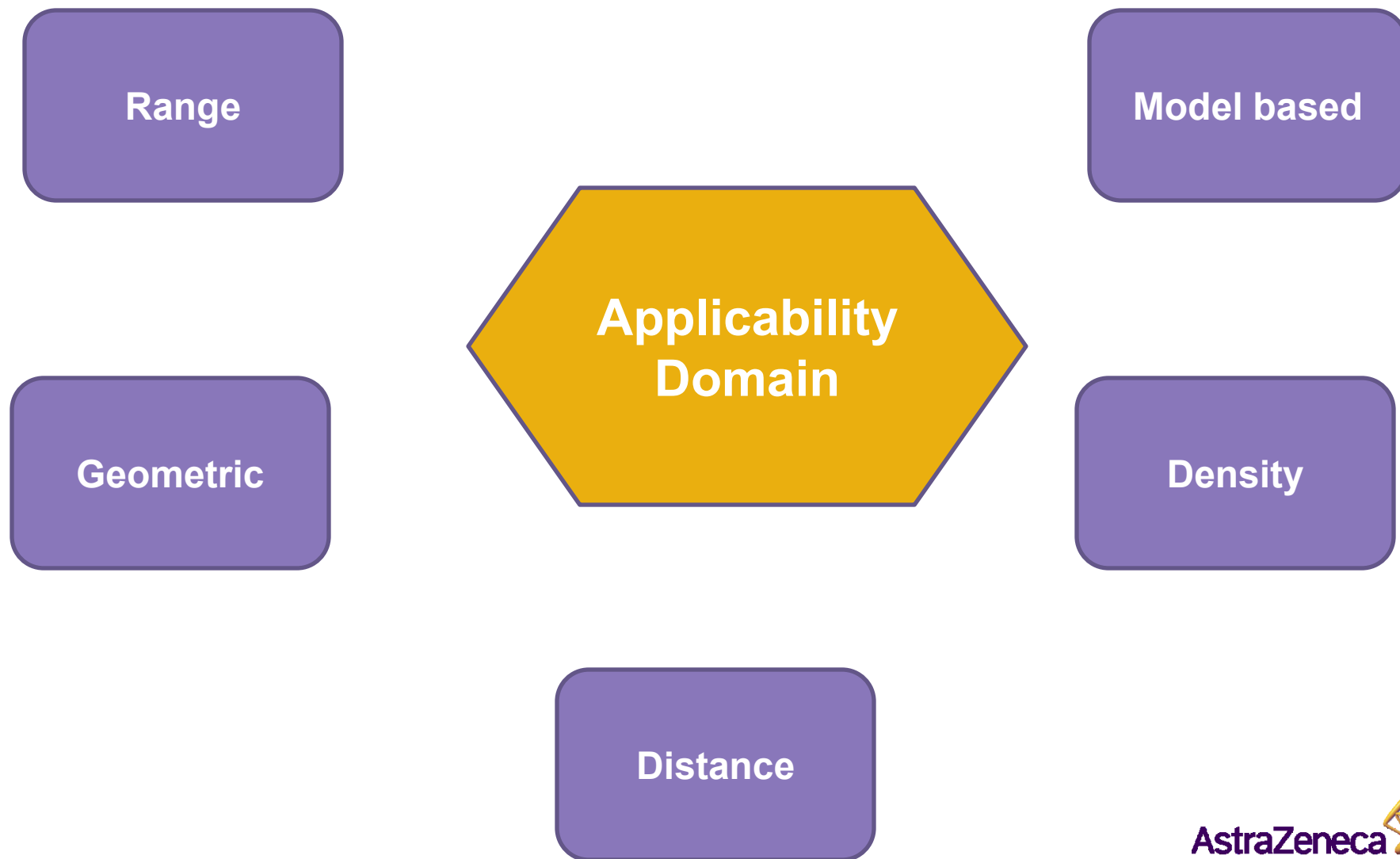


The Applicability Domain

- The third principal of “The OECD principals for QSAR validation”[1] (2004) requires “a defined domain of applicability”
- The ECVAM Workshop [2] in 2005 agreed on the definition:

“The applicability domain of a QSAR model is the response and chemical structure space in which the model makes predictions with a given reliability”




Applicability Domain Methods in QSAR literature



Distance – “Distance to Model” [3, 5]

- Distance between the test compound and the train set
- Distances;
 - Distance to mean
 - Average distance to all
 - Average distance to near neighbors
 - Count of neighbors
- Molecular representation;
 - Chemical similarity
 - Descriptor space
 - Property space (includes response)
- Metrics; Euclidean, Mahalanobis, City block

Model Independent Reliability – Proposed by Orange group

- Sensitivity Analysis - 45% positive correlation with prediction error
- Bagging Variance – 53% 
- Local Cross-Validation – 44%
- Density Estimation – 34%
- Local Error Modeling – 48% 
- Meta classifier selecting reliability method – 55%
- Cross validation selection – 78% 

AZOrange implementation

