

# Small Molecules, Big Opportunities:

# An in silico Platform for Off-target Profiling And Drug Repurposing

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#### Drug Development Pipeline: costs 1-2 Billion USD per Drug





# Why New Approach Methodologies ?

- every single compound entering production must be thoroughly tested and characterized:
  - drugs (human & veterinary)
  - cosmetics (UV filters, fragrances...)
  - additives (flame retardants, plasticizers...)
  - food contact materials and polymers
  - agrochemicals (pesti-, fungi-, herbi-cides)
  - colorants, dyes, pigments...
- 3R: reduction, replacement and refinement of animal testing
- regulatory needs: EC, EPA, BLV... (REACH)
- drug attrition rates getting worse and development more costly

Knowledge gathered can be used to rationally **explain** and **avoid toxic phenomena** !





# Aim at the Right Target!

"During preclinical testing, 75% of safety closures were compound-related – that is, they were due to '**off-target**' or other properties of the compound other than its action at the primary pharmacological target – as opposed to being due to the primary pharmacology of the target."

Waring M.J. et al. Nature Reviews: Drug Discovery (2015), 14, 475



#### **On-target binding**

(primary pharmacology)

- a **desired** interaction of a typically small molecule with biomacromolecular receptor(s) exhibiting preferably a **high** degree of specificity (e.g. astemizole at the histamine receptor)
- initiates a cascade of actions resulting in **curative** or at least illness **compensating** effect(s)



#### **Off-target binding**

(anti-target, promiscuous binding, secondary pharmacology)

- an **undesired** interaction of a typically small molecule with biomacromolecular receptor(s) exhibiting a **certain** degree of specificity (e.g. astemizole at the hERG ion channel)
- initiates a cascade of actions resulting in **adverse** effect(s) (astemizole  $\rightarrow$  cardiotoxicity)



#### Even approved drugs bind to multiple on/off-targets



\*Haupt J., Daminelli S., Schroeder M.: Drug Promiscuity in PDB: Protein Binding Site Similarity Is Key. PloS ONE 8(6):e65894 (2013)

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#### **On-target Design: Ligand-Protein Complementarity**



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nobelprize.org



muarchives.missouri.edu

**Emil Fischer** key-lock concept (1894)



oldlockandkeyco.co.uk



enantiomorphic.blogspot.com



"Um ein Bild zu gebrauchen, will ich sagen, dass Enzym und Glykosid zueinander passen müssen, wie Schloss und Schlüssel, um eine chemische Wirkung aufeinander ausüben zu können."

Emil Fischer (1894)

Lock: Carbonic anhydrase + Key: Acetazolamide



### **Off-target Effects: Induced-Fit**

Both, the biological lock and the key are flexible !





Daniel Koshland Induced-Fit (1958)

This mutual adaptation of protein and ligand is the origin of side effects.

bio1151.nicerweb.com



### **Current safety profiling strategy = start early !**



Valentin J.P., et al.: In vitro off-target profiling: impacting drug design and guiding quantitative translation risk assessment. *EuroTox 2018 Conference*, oral presentation on behalf of UCB, Novartis, Pfizer, Allergan & AstraZeneca (2018); *Tox Lett.* S1, 295, S21 (2018).



#### Simulation of off-target-mediated side effects



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### VirtualToxLab: an automated SBD platform for off-target profiling



http://www.virtualtoxlab.org / Toxicol. Appl. Pharmacol. 2012, 261, 142–153 / Tox. Lett. 2015, 232, 519–532

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/**irtualToxLal** technical flow diagram

\* Schrodinger Inc.\*\* University of Minnesota



#### **DOLINA: Docking with Local INduced-fit Algorithm**



Smieško M.: DOLINA – Docking Based on a Local Induced-Fit Algorithm: Application toward Small-Molecule Binding to Nuclear Receptors. *J Chem Inf Model*, 53, 6, 1415 – 1423 (2013).



#### **Molecular Docking - Accuracy**



**α-Zearalenol** (mycotoxin)

Binding at the Estrogen Receptor- $\!\alpha$ 

- grey carbons docked pose  $\rightarrow$  VirtualToxLab
- green carbons reference X-ray structure (4TUZ)





### VirtualToxLab – a few highlights

- used by more than 200 academic, non-profit and regulatory organizations worldwide
- recommended tool for identification of endocrine disruptors by EU/EC institutions

Guidance for the identification of endocrine disruptors in the context of Regulations (EU) No 528/2012 and (EC) No 1107/2009

- used by five global food and cosmetic companies
- in cooperation with the Swiss Federal Food Safety and Veterinary Office: screening study of 3.5k substances in inks for food packaging materials (3 months / 128 cores); Projects ToxOligo 1 & 2 ongoing (~500 plastic oligomers)





Smieško M, Don CG, Meuwly R, Kucsera S, Brüschweiler BJ: Large-scale in silico screening of compounds contained in printing inks for food packaging materials. *EuroTox* 2018, poster presentation; manuscript in preparation.



### Our Newest Solution to the Problem: PanScreen

User-submitted compound



- Java independent
- modular implementation of off-targets & computational tools for easy upgradeability
- developed as a **portable** python package, C++ utilities
- Back-end: easily **distributable** as Docker containers
- Front-end: user-friendly web interface for non-commercial users at https://www.panscreen.ch/





#### **PanScreen: Off-target Dataset Details**

Affinity data: ChEMBL repository (14 000 molecule/affinity pairs); Structural data: Protein Data Bank (up to 3 per off-target)

		Affinity range	Ratio	# size validation set
UniProt ID	Off-target Name	[kcal/mol]	active/inactive	(20% of total)
P10275	AR	[-4.8, -13.0]	2.8	169
P23458	JAK1	[-7.0, -13.6]	20.1	148
O60674	JAK2	[-5.7, -16.1]	9.0	180
P25103	Substance-P	[-6.2, -14.4]	12.7	137
P28222	5HT1B	[-6.7, -13.4]	6.6	145
P49286	Melatonin	[-6.6, -16.4]	15.2	178
Q9Y233	PDE10A	[-4.4, -16.4]	20.0	189
P04150	GR	[-6.3, -13.4]	8.4	150
P07550	b2-ADR	[-5.2, -14.6]	2.4	131
Q08499	PDE4D	[-5.2, -13.7]	1.6	55
P03372	ERa	[-4.2, -14.2]	3.3	78
Q92731	ERb	[-4.2, -14.2]	4.4	65
P37231	PPARg	[-1.2, -13.4]	2.1	87
P14416	Dopamine-2 R	[-6.7, -15.5]	4.1	1120



### **PanScreen:** Performance Metrics

Predictive performance of the implemented off-targets for the validation set

Name	PCC	MUE [kcal/mol]	RMSE [kcal/mol]	AUROC
Tyrosine-protein kinase JAK2	0.81	0.68	1.10	0.94
Estrogen receptor alpha	0.84	0.94	1.18	0.89
Glucocorticoid receptor	0.79	0.70	0.95	0.89
Beta-2 adrenergic receptor	0.79	0.92	1.17	0.91
Androgen receptor	0.81	0.79	1.03	0.88
Dopamine receptor D2	0.75	0.66	0.87	0.88
Tyrosine-protein kinase JAK1	0.81	0.55	0.85	0.94
Substance-P receptor	0.80	0.77	1.02	0.91
5HT receptor 1B	0.77	0.85	1.06	0.87
$\mathrm{PPAR}\gamma$	0.68	0.89	1.33	0.84
Melatonin receptor 1B	0.72	1.06	1.35	0.84
Phosphodiesterase 4D	0.80	0.95	1.36	0.82
Estrogen receptor beta	0.75	1.00	1.29	0.85
Phosphodiesterase 10A	0.80	0.88	1.18	0.93



- Correct predictions (83.1%; within +/- 1 log units of experimental)
- Slightly underestimated (6.2%; within -1 -2 log units of experimental)
- Slightly overestimated (8.8%; within +1 +2 log units of experimental)
- False negatives (0.7%)
- False positives (1.2%)



#### Search for xenobiotics interfering with the corticosteroid–androgen balance



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#### Search for xenobiotics interfering with the corticosteroid–androgen balance



Biochemical Pathways: An Atlas of Biochemistry and Molecular Biology, Second Edition



### Search for xenobiotics interfering with the corticosteroid–androgen balance

#### Preparation

- four anti-targets selected: CYP11B1, CYP17A1, 5-α reductase (SRD5A2), 5-β reductase (AKR1D1)
- ensembles of multiple protein structures created for each off-target
- several docking engines employed: Glide, smina, GOLD, Dolina
- docking protocols validated using known actives and inactives, decoys, re-docking & cross-docking

#### Virtual Screening

- screened ligands compiled from the DrugBank (FDA approved, experimental, investigational + metabolites)
- post-docking evaluation: MM-GB/SA, visual analysis, similarity searches within the compound group

#### Results

- CYP11B1 (39): 19 novel inhibitors, several < 10 nM, 3 < 1 nM!!!  $\rightarrow$  10.1016/j.taap.2023.116638
- CYP17A1 (11): 2 novel hits found (IC50=1.47  $\mu M$  & IC50=2.2  $\mu M) \rightarrow$  in preparation
- SRD5A2 (11): found 1 inhibitor (IC50 to be determined)
- AKR1D1 (46): found 1 inhibitor (IC50=2.3  $\mu$ M), 2 weak inhibitors (60% activity at 10 $\mu$ M); 6 anabolic steroids  $\rightarrow$  10.1016/j.toxlet.2023.07.006



# (Phase 1) Metabolism – Enzyme (CYP450) based

1) Identification of poses by molecular docking and molecular dynamics that expose "vulnerable" ligand regions for the oxidation from the oxyferryl group (Fe=O) at the active site of CYP2D6 wild type / variant  $*53 \rightarrow$  UM phenotype



Don C.G., Smieško M.: Deciphering Reaction Determinants of Altered-Activity CYP2D6 Variants by Well-Tempered Metadynamics Simulation and QM/MM Calculations. *J Chem Inf Model* 60 (12), 6642-6653, (2020)

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# (Phase 1) Metabolism – Enzyme (CYP450) based

2) calculation of the activation barrier for oxidation reactions at these sites using quantum mechanics (DFT)



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Table 1. QM/MM Activation Barriers (kcal/mol, M06-2X/LACV3P\*) of the Hydrogen Abstraction of Bufuralol's SoMS Studied in CYP2D6 WT and CYP2D6\*53<sup>a</sup>

acti	Δrmsd (Å)		
SoM	CYP2D6 WT	CYP2D6*53	bufuralol
S1_H1	13.3	12.4	1.8
S1_H2	22.8	31.9	2.7
S2	22.9	21.2	6.4
<b>S</b> 3	28.8	61.8	4.0
C1	57.2	42.4	4.7

#### Calculated trends in agreement with experiment → personalized metabolism prediction is possible!

Don C.G., Smieško M.: Deciphering Reaction Determinants of Altered-Activity CYP2D6 Variants by Well-Tempered Metadynamics Simulation and QM/MM Calculations. *J Chem Inf Model* 60 (12), 6642-6653, (2020)

#### Rare event MD simulation: *Paracetamol* binding to CYP2D6\*53

#### A rare event in atomic detail:

- ligand interaction with the membrane
- recognition at the protein surface
- ligand entry to the tunnel
- tunnel bottleneck/gating residues
- passage toward the active site
- intermolecular interactions and dynamics of the binding mode within the active site
- residence time

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- solvent effects/displacement
- protein fluctuations/ rearrangement/ local & long range induced-fit effects
- membrane influence
- allosteric binding sites



periodic boundary system of CYP2D6\*53 within the membrane (80 x 100 x 130 Å), simulation software: Desmond, hardware accelerator: nVidia Titan X



# Conclusions

- compound safety is still an open issue in pharma, food, agro/chemical, cosmetic industries
- even some approved (!) drugs show off-target binding

- structure-based design methods are effective in screening for the off-target binding
- if properly applied, docking & scoring-based methods can identify off-target binding
  - molecular dynamics offer a comprehensive view of the on/off-target binding



- QM/MM approaches can be used to accurately model reactivity of ligands in CYPs ( $\rightarrow$  metabolism), even for SNP variants  $\rightarrow$  precision toxicology
- offer mechanistic interpretation with direct hints to avoid (design out) the off-target binding





# The Future – Near and Far

- more accurate force fields and parameters (polarized, AI-based)
- growing toxicological evidence, new in vitro data
- faster performance CPUs / GPUs
- **smarter algorithms:** docking and scoring + machine learning approaches
- **complex simulations**  $\rightarrow$  larger and more realistic molecular systems
- **new targets** cover all relevant human macromolecules and their complexes available
- complete adverse outcome pathways (AOPs)
- physiology-based **pharmacokinetic modeling** (PBPK)
- **personalised medicine** genetic information explicitly included in the computer simulation
- **linking** to "omics", clinical ( $\rightarrow$  drug repurposing) and toxicological data
- tox-enabled medicinal chemistry optimization  $\rightarrow$  drugs safe by design

#### More accurate simulations $\rightarrow$ Reliable predictions $\rightarrow$ Safer compounds !





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