# Pair-wise analysis of structure solubility relationships Application to Drug Design to Enhance the Solubility of NCEs

# **PhysChem FORUM 10**

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### Agenda

- Introduction
- Solubility and Discovery
- Data Quality and Quantity for Development of Solubility Prediction
- Pairwise Analysis
- Conclusion
- Acknowledgement



### Introduction

#### Solubility is an import factor through every stage of drug discovery

#### (1) Low soluble compounds may show

- inappropriate PK (absorption) profile
- inaccurate results in *in vitro* biological/toxicological tests, which provides inaccurate SAR to medicinal chemists (ex. 10 uM = 5 ug/mL: MW 500)

#### (2) Low solubility compounds may cause significant challenges in

- *in vivo* pharmacological evaluation
- achieving target exposure in preclinical/clinical studies
- formulation development (enabling formulation is great but not a magic)
- *in vitro* pharmacological/safety profiling

For drug discovery and development effectiveness  $\rightarrow$  medicinal chemist need tools to help design compounds potent as well as soluble.

#### Robust tools for solubility prediction is strongly required.



## **Solubility Screening**

	Data Assur.	Oral Absorption/	Formulation Feasibi	ility Assessment
	In vitro broad assay panel	SAR	Biorelevant	Extended
Purpose	Underwrite in vitro assay	SAR for oral exposure	SAR/in-vivo interpretation for oral administration	Basic data for formulation development
Timing	Broad Ligand Profiling	ADME Screening	In-vivo pharmacology/PK	Compound Selection
Crystallinity	N/A	PLM (residue only)	PLM (residue)	PLM (initial and residue) (will move to PXRD)
Equilibration Time (hour)	24	>20	24	24, 48
Buffer	PBS (pH 7.4)	50 mM PBS (pH 6.5)	SGF (pH1.2), 50 mM PBS, Fassif (pH 6.5)	SGF, PBS (pH 6.5), FaSSIF, FeSSIF, 0.1 N NaOH (opt'l)
Replicates (sample)	N = 1 (from DMSO)	N = 2 (from DMSO)	N = 1 (from powder)	N = 1 – 3 (from powder)
Requirement	10 mM	20 μL of 30 mM	15 mg	50 mg
Dynamic Range	<b>0.5 − 200 μM</b> (0.25-100 μg/mL, if MW: 500)	0.1 – 300 μM	0.3 – 300 μg/mL	0.3 – 300 μg/mL (– 3 mg/mL for SGN)
Throughput	-	~96 cpd/week	~32 cpd/week	~2 – 4 cpd/month
Timelines	-	1 week from compound receipt	1 week from compound receipt	2 weeks from compound receipt



PLM: Polarised Light Microscopy; PXRD; Powder X-Ray Difraction; PBS: Phosphate Buffered Saline; SGF: Simulated Gastric Fluid (w/o enzyme); FaSSIF: Fasted State Simulated Intestinal Fluid; FeSSIF: Fed State Simulated Intestinal Fluid

#### Solubility needed in in vitro assay





#### **Discovery** loop

#### **Design – Synthesis – Characterisation – Analysis**



Solubility Prediction Models: tool to help medicinal chemists design molecules having higher solubility



## **Training Data Issue?**

- ADME In Silico Modeling: Towards
  Prediction Paradise?
  - H. Van De Waterbeemd, E. Gifford. Drug Discovery 2003, 2, 192–204.
    - ••• However, at present, no approaches are robust enough to accurately predict low solubility.
    - Many current predictive solubility programs use training data from different experimental conditions. Hopefully, by measuring many compounds under standardized conditions, current predictive models can be improved.

What needs to be measured under which condition?



#### **Solubility Difference Between Crystalline & Amorphous States**



High Throughput Solubility Measurement With Automated Polarized Light Microscopy Analysis. K. Sugano, T. Kato, K. Suzuki, K. Kako, T. Sujaku, T. Mano. *J. Pharm. Sci.* **2006**, *95*, 2115–2122. Also see: Discovery Pharmaceutics—Challenges and Opportunities. X.-Q. Chen, M. D. Antman, C. Gesenberg, O. S. Gudmundsson. *The AAPS Journal* **2006**, *8*, E402–E408; What is the True Solubility Advantage for Amorphous Pharmaceuticals? B. C. Hancock, M. Parks. *Pharm. Res.* **2000**, *17*, 397–404. Predicted: 10–1600 times; Experimental:1–25 times; Impact of Solid State Properties on Developability Assessment of Drug Candidates L.-F. Huang, W.-Q. Tong. *Adv. Drug Delivery Reviews* **2004**, *56*, 321–334. Experimental: 1.1–500-1000 times

# **Balancing Quality and Quantity**

- The quality and quantity of data to develop solubility prediction tools in this presentation
  - Quality
    - Solubility data with crystalline residue
       By PLM
    - Dynamic range: 0.3–300  $\mu$ g/mL
    - Vehicle/pH: Phosphate buffered saline/pH 6.5
  - Quantity
    - Approx. 1,000 in-house powder compounds
    - Selected based on similarity



### **Prediction Approaches**

#### What guides design of soluble molecule?

- Empirical rules/principles, personal experiences through project work→ Would be a first option to rely on but not so versatile enough to cover large chemical space of pharmaceutical entities.
- Classification model → Useful for rough prediction of solubility of a large number of compounds (e.g., from combinatorial virtual library or compound subset). *However, not useful for singleton molecule design* (helpful in HTS triage or lead seeking stage)
- Regression model → Only valid within local chemical space covered by training set molecules
- Pairwise analysis → What minor modification improve solubility? This analysis provides a table of some examples for pairs of two molecules whose solubility difference is high despite high similarity to each other. (applicable in lead development stage)



### **Pairwise Analysis (Matched-Pair Analysis)**

• To find useful relationships between (minor) changes in structure and changes in properties.



- The analysis on solubility would help medicinal chemists to design compounds with higher solubility.
- Pfizer has a huge database of solubility values for discovery compounds.

Matched Molecular Pairs as a Guide in the Optimization of Pharmaceutical Properties; a Study of Aqueous Solubility, Plasma Protein Binding and Oral Exposure A. G. Leach, H. D. Jones, D. A. Cosgrove, P. W. Kenny, L. Ruston, P. MacFaul, J. M. Wood, N. Colclough, B. Law. J. Med. Chem. 2006, 49, 6672-6682



#### **Pairwise Analysis, MCS similarity**



#### Similarity Calculation → MCS (maximum common substructure) similarity

Vmcs	atom	number	of	mcs
Emcs	bond	number	of	mcs
Va	atom	number	of	molecule_A
Vb	atom	number	of	molecule_B
Ea	bond	number	of	molecule_A
Eb	bond	number	of	molecule B

Similarity = (Vmcs+Emcs) \* (Vmcs+Emcs) / (Va+Ea) \* (Vb+Eb)



From dataset, pick up pair of two molecules whose similarity is > 0.8 while the ratio of solubility is larger than a threshold value.

### **Results**



Indole – imidazole transformation decreases solubility → Remarkable contrast with general trend (clogp)



Insertion of one methylene group increases solubility by order of  $10^3 \rightarrow$ (1) destroy planarity (2) larger decrease in



7.5 μg/ml



2551.5 μg/ml



Low planarity can be detrimental for effective  $\pi - \pi$  stacking

The number of accessible conformations would increase (significant entropy increase in dissolution process) 13



### Results





## **Usage from medicinal chemistry viewpoint (1)**



#### Compound\_A

Lead molecule in a project but very insoluble and easy to crystallise  $\rightarrow$  How we can make this molecule soluble?

SAR (Structure-Activity Relationship) for activity against the project target → The oxazole ring can replaced by diverse groups without large disruption in activity



## **Usage from medicinal chemistry viewpoint (2)**



#### Compound\_B

Lead molecule in a project but very insoluble and easy to crystallise  $\rightarrow$  How we can make this molecule soluble?

**SAR for activity against the project target** → The biphenyl group can be replaced by different aromatic-aromatic framework without loss in activity against the project target.



### Conclusion

- 1. A novel tool for pairwise analysis was successfully developed by selecting proper dataset on solubility and crystallinity.
- 2. Pairwise analysis on solubility is useful to help medicinal chemists to design high-solubility compounds.



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