

ADVANCED DIGITAL DESIGN OF PHARMACEUTICAL THERAPEUTICS



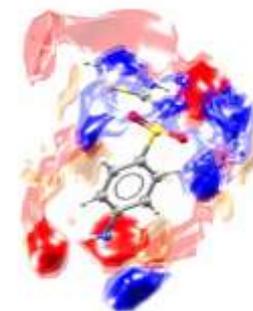
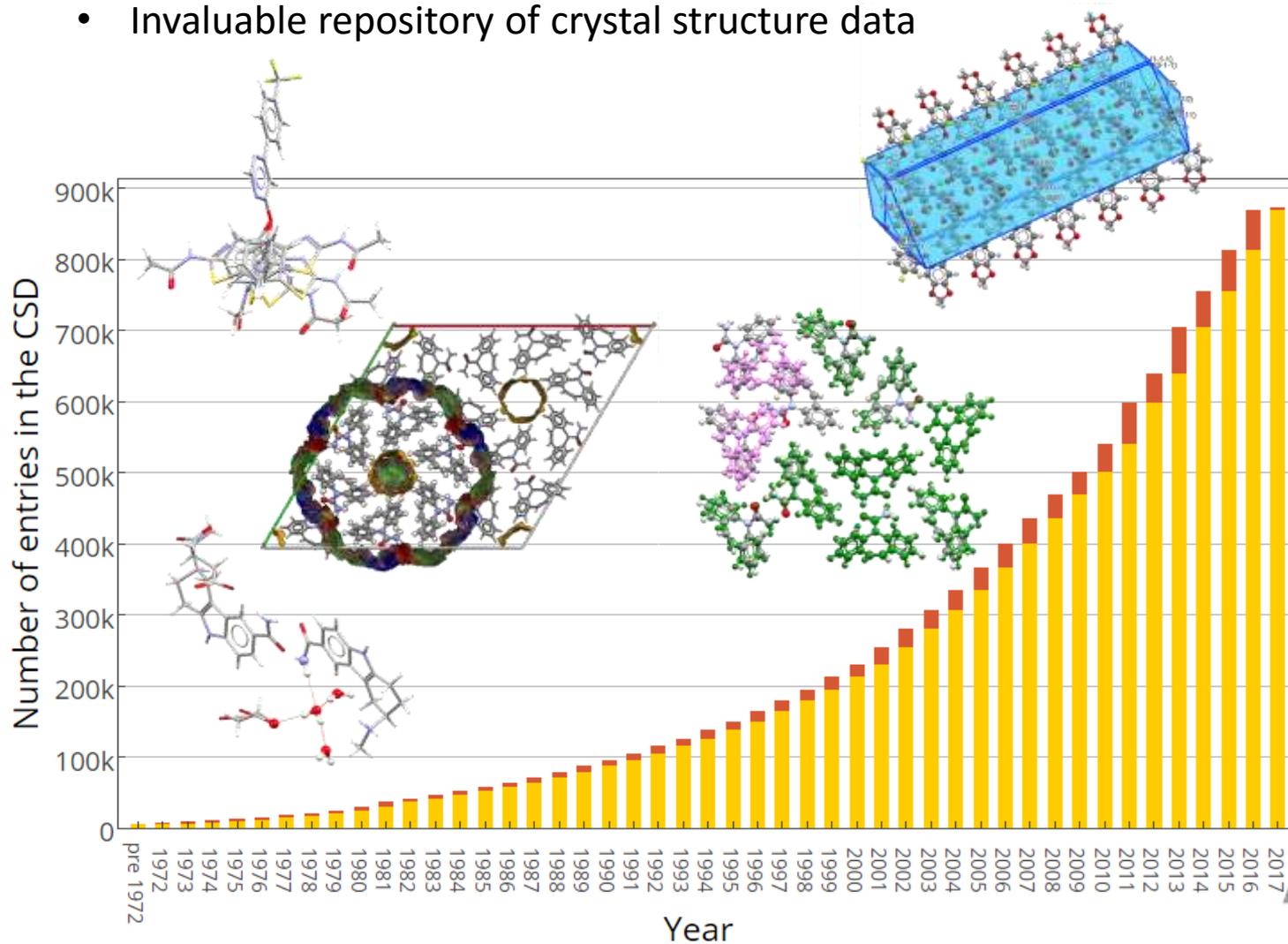
Streamlining pharmaceutical supply chain processes:

The emerging application of solid-state structural informatics

Mathew J. Bryant, Andrew G.P. Maloney and Neil Feeder

The Cambridge Structural Database

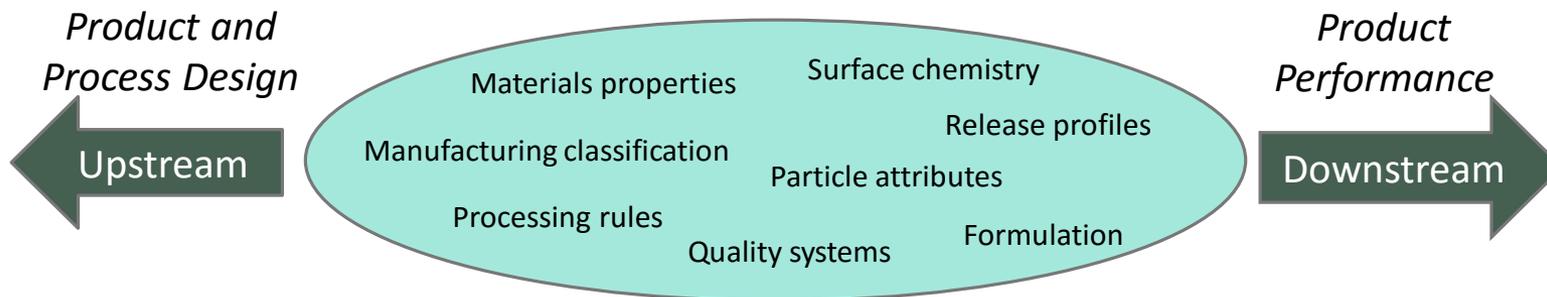
- All small-molecule organic & metal-organic crystal structures ever published.
- CSD contains over 875,000 crystal structures
- Invaluable repository of crystal structure data



- **Advanced Digital Design of Pharmaceutical Therapeutics**
- Four year collaboration between government, industry and academia
- Instigated by the Medicines Manufacturing Industry Partnership and part funded under the **Advanced Manufacturing Supply Chain Initiative**



Digital Design: Molecules to Medicine



Primary Manufacturing - Secondary Manufacturing



Processes

Products

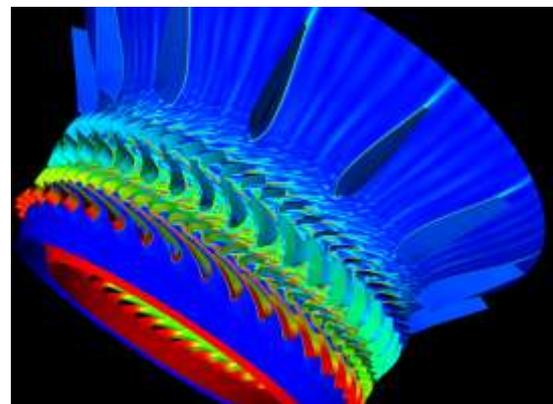
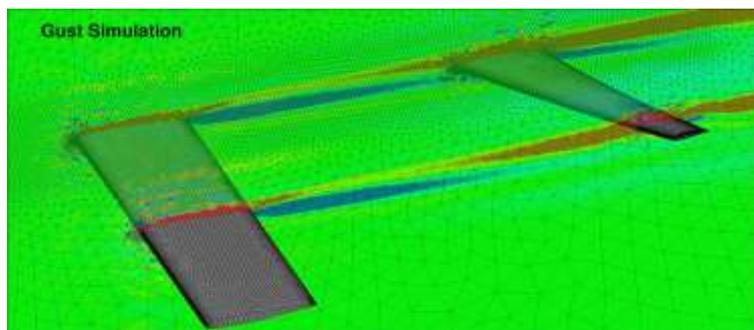
Performance

Design and control of optimised development & manufacturing processes through data analysis and first principle models



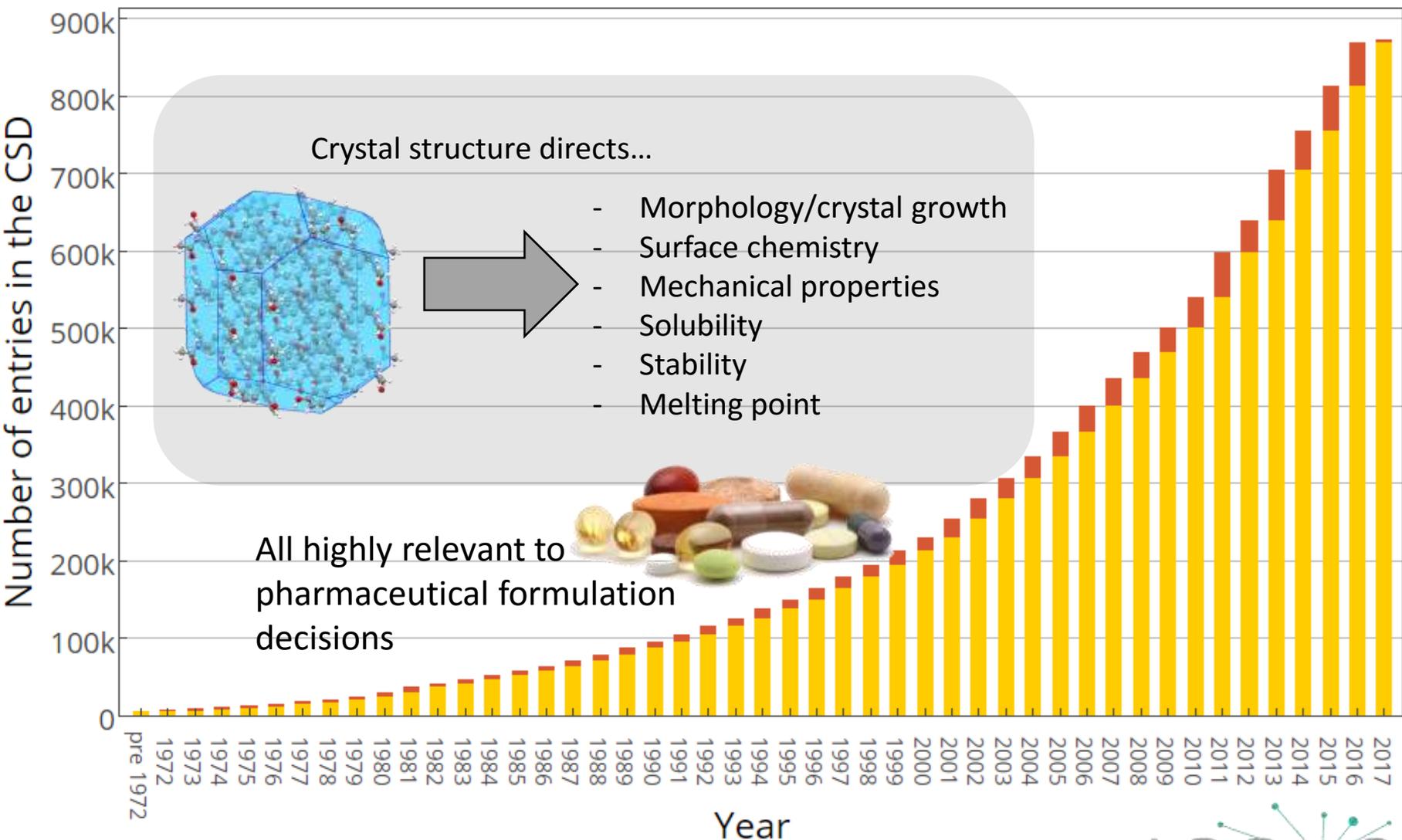
- If we designed airplanes like we design drugs...

Woltosz, W.S. J. Comput. Aided Mol. Des. (2012) 26: 159



“Why has pharmaceutical research and development lagged so far behind other industries in the development and application of simulation and modelling for research and development?”

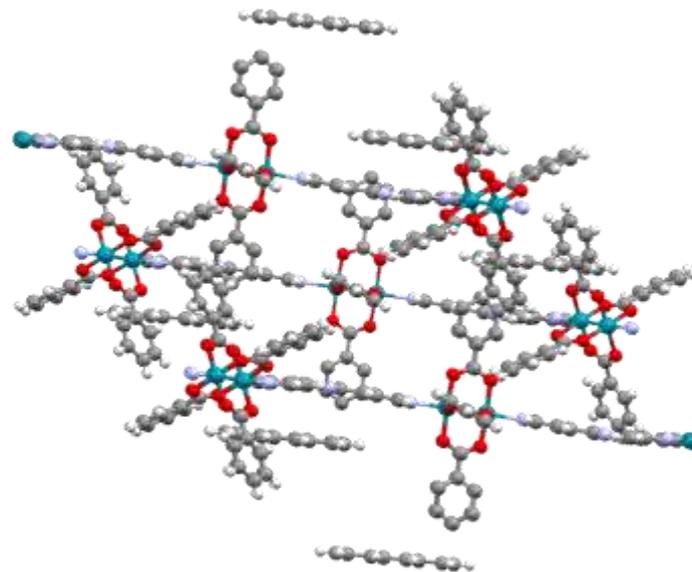
Digital design of pharmaceuticals using the CSD



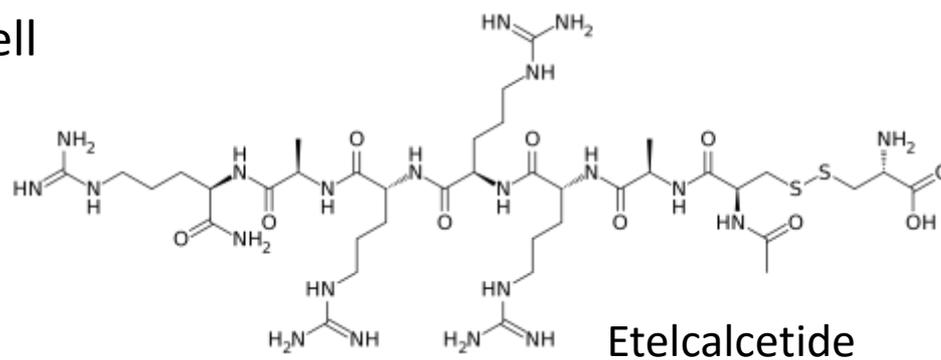
Making the CSD relevant to drugs

7

- 57% of structures in the CSD are metal-organic
- Currently working towards creating a 'Drug like' subset



- Lipinski's rule of 5 doesn't apply well to modern drugs



- Initially decided to look at actual approved drugs in the CSD

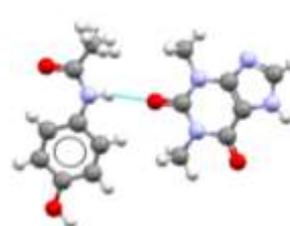
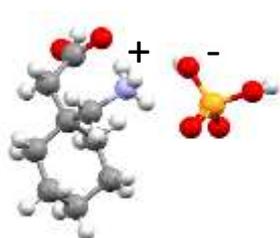
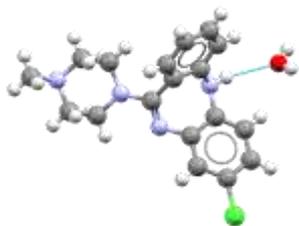


Making the CSD relevant to drugs

- CSD Drug Subset has now been established
- Drug definition taken from the approved drug database of Drugbank.ca

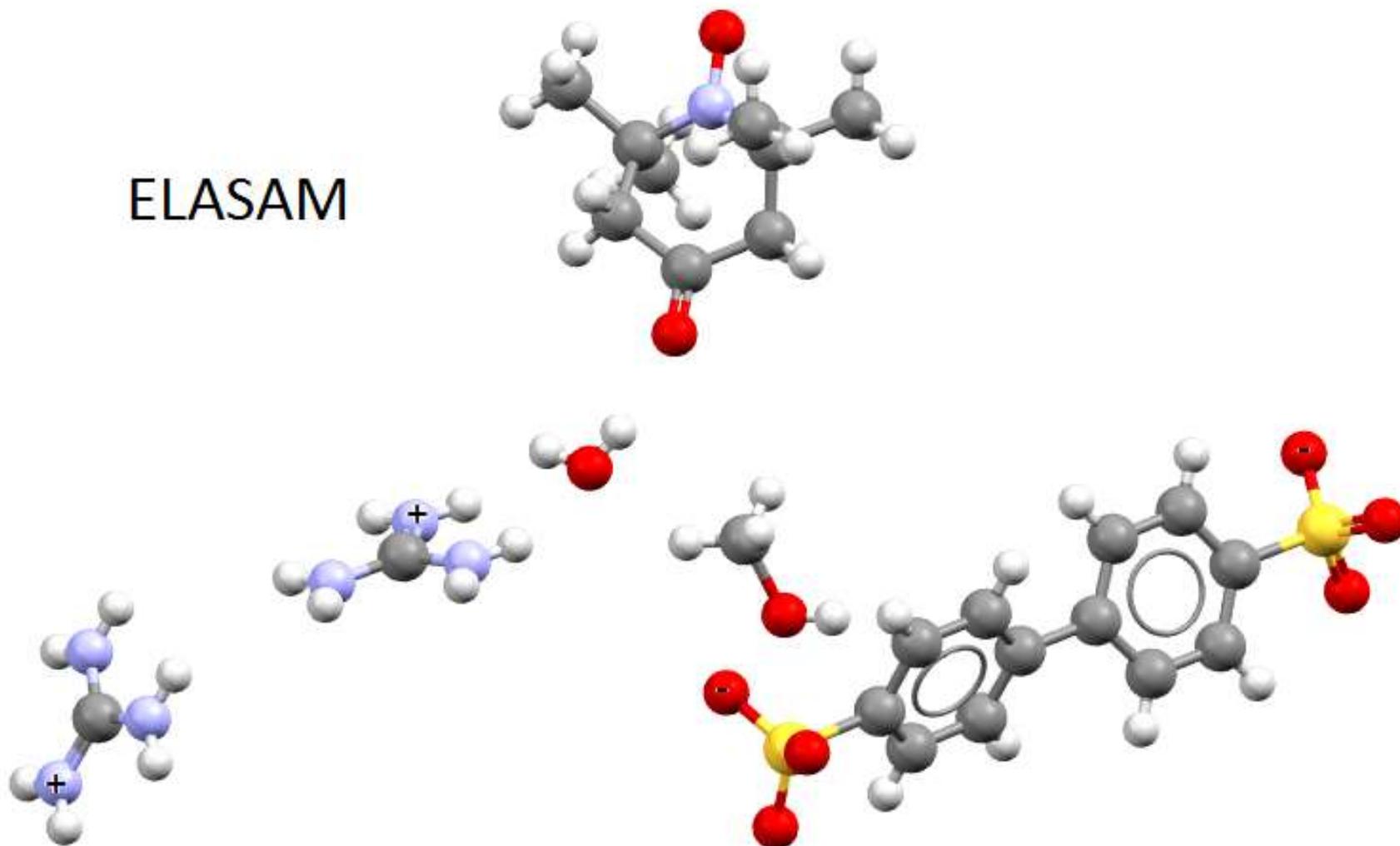


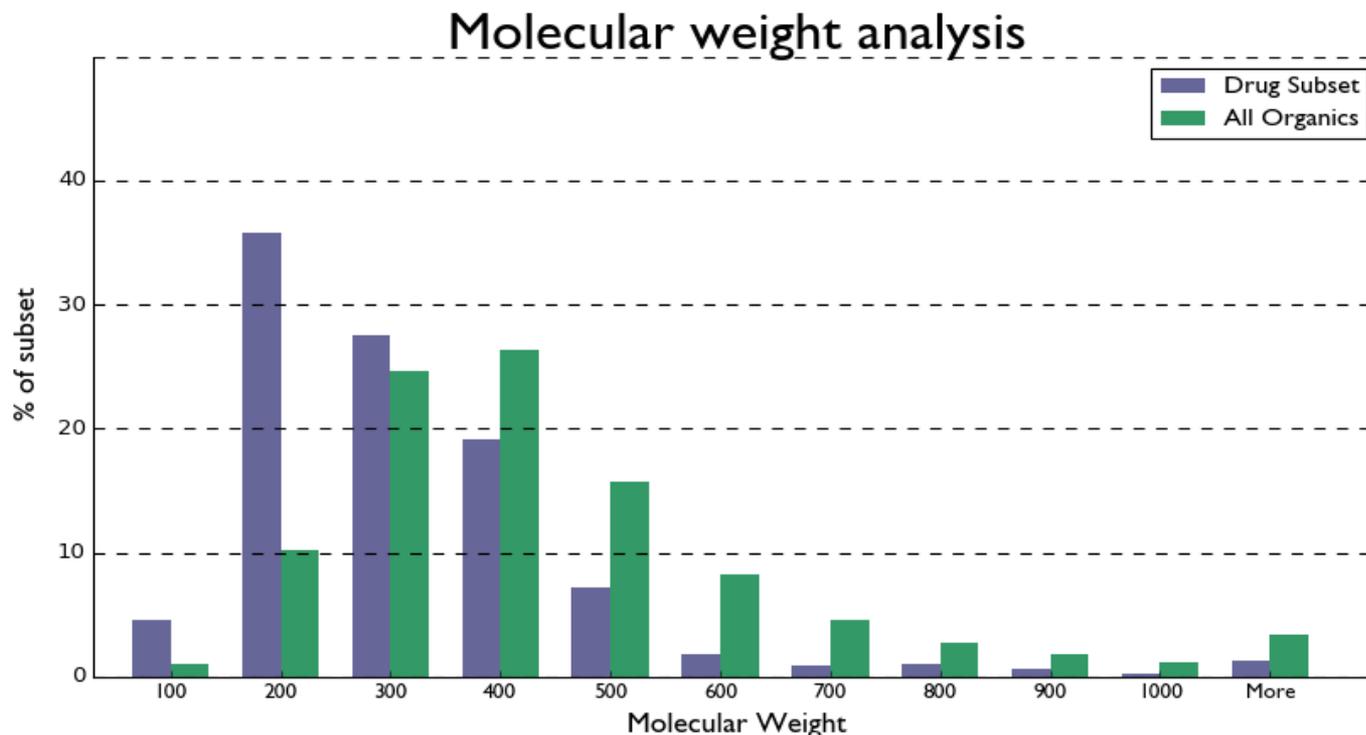
- Generated using InChI strings and the CSD Python API
- 8632 crystal structures representing 785 drug molecules
- Searchable and sortable by categories like hydrates, solvates, salts, co-crystals, pure drug (or any combination of these)



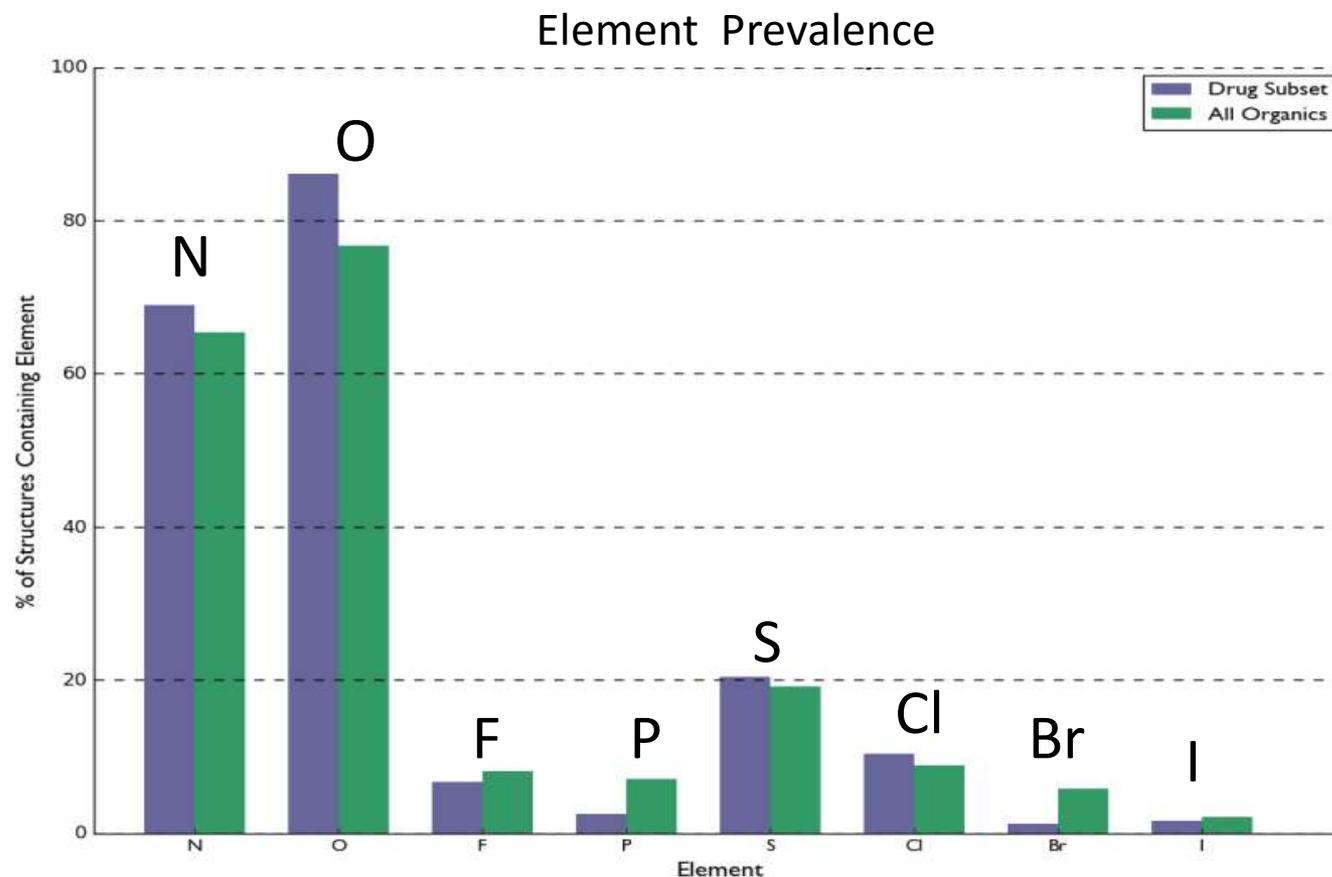
Proportions of each type in CSD Drug Subset

ELASAM

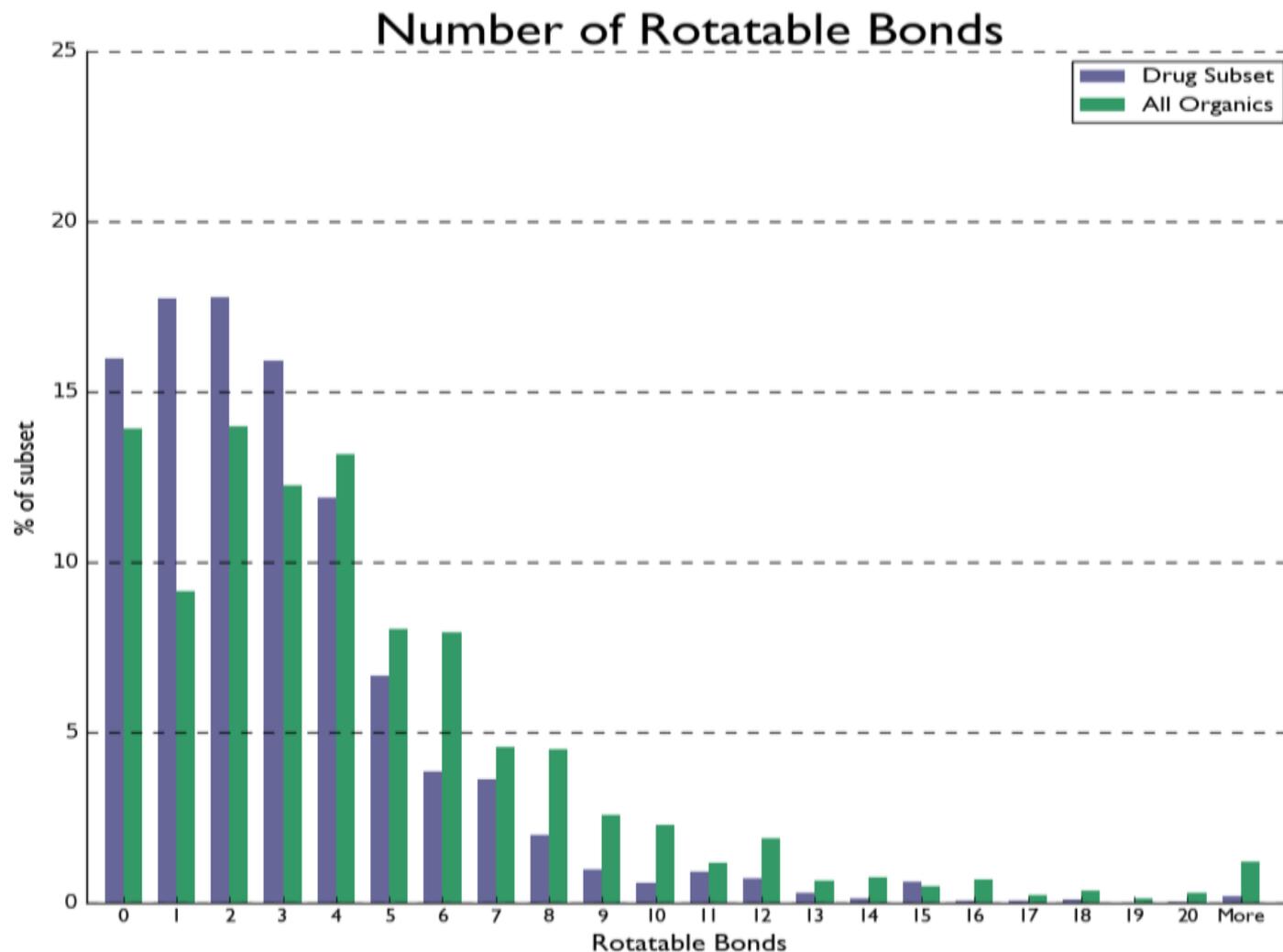




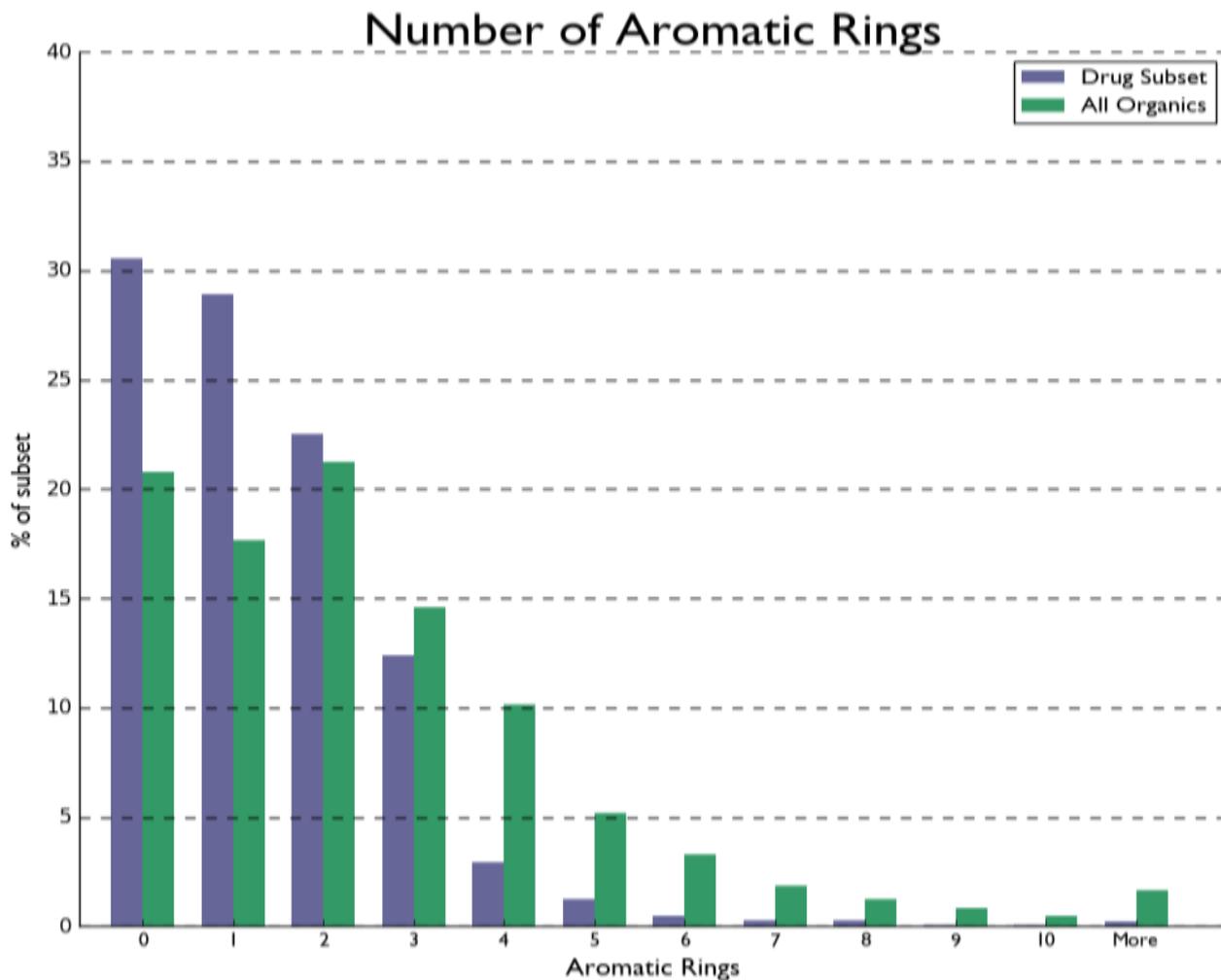
- Clear trend towards lower molecular weights for drug molecules
- Highest percentage of Drugs fall between 100 and 200 g/mol



- Drugs show a different spread of elements
- Much lower prevalence of bromine and phosphorus

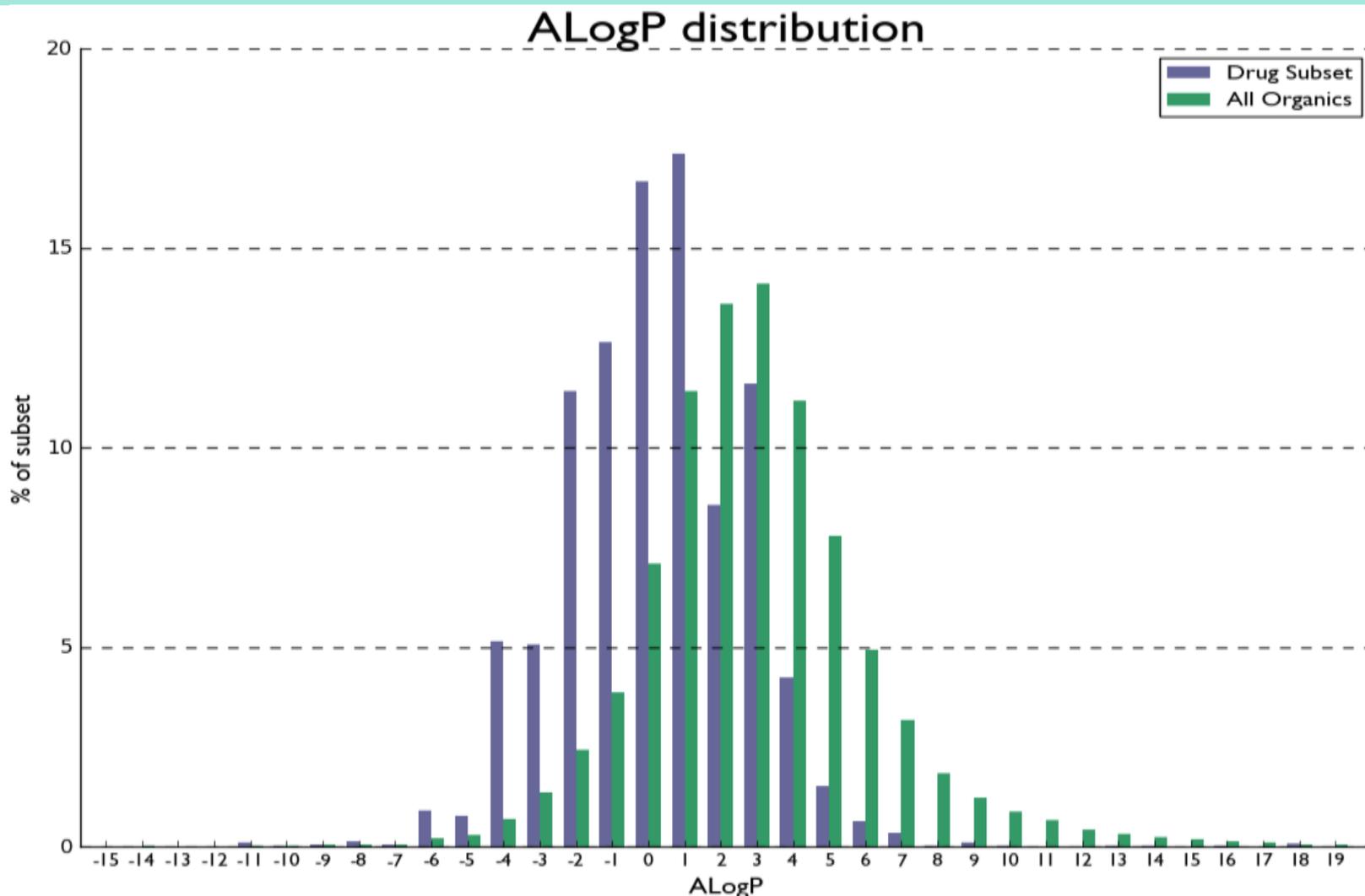


- Drugs have fewer rotatable bonds



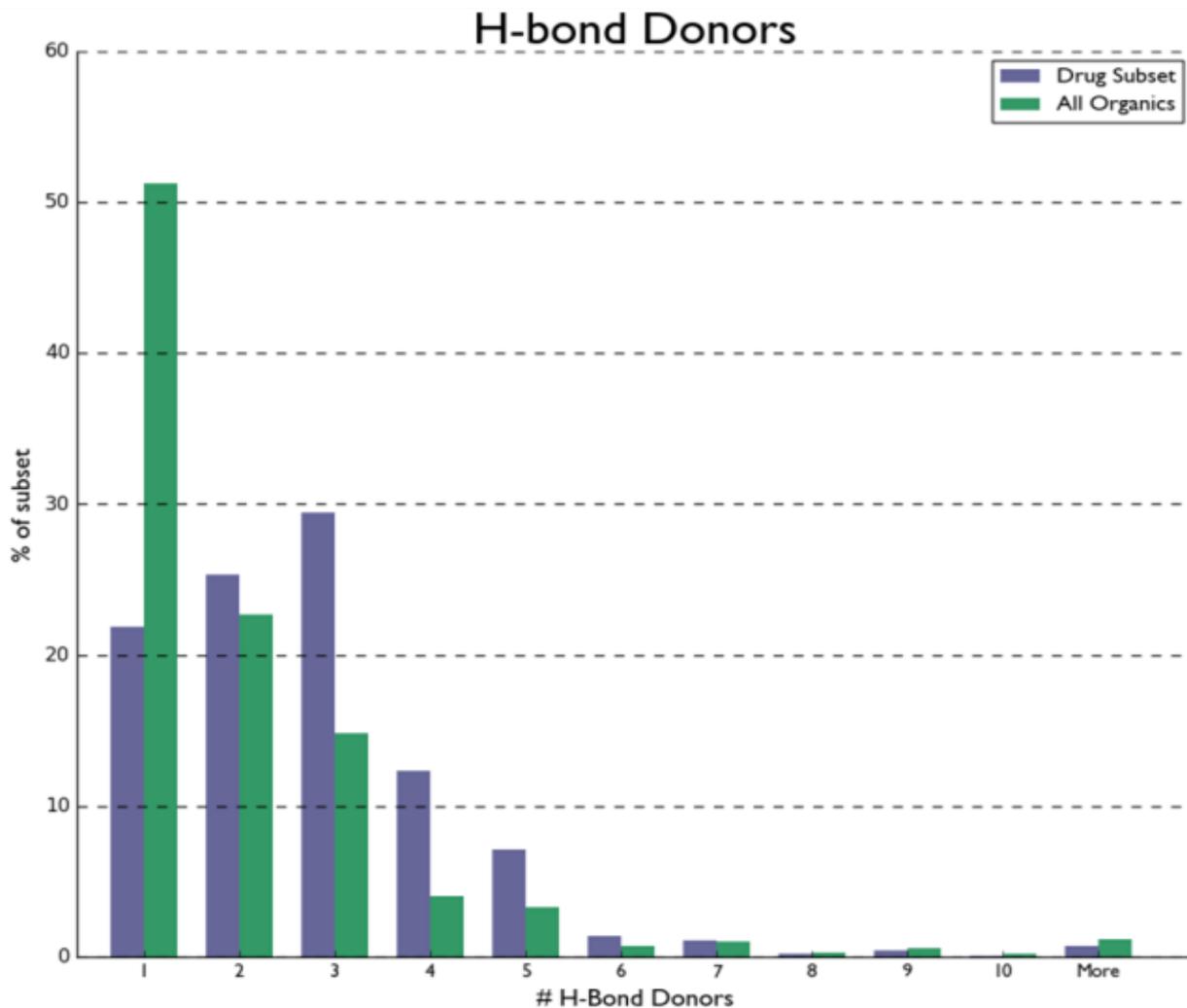
- Drugs have fewer aromatic rings

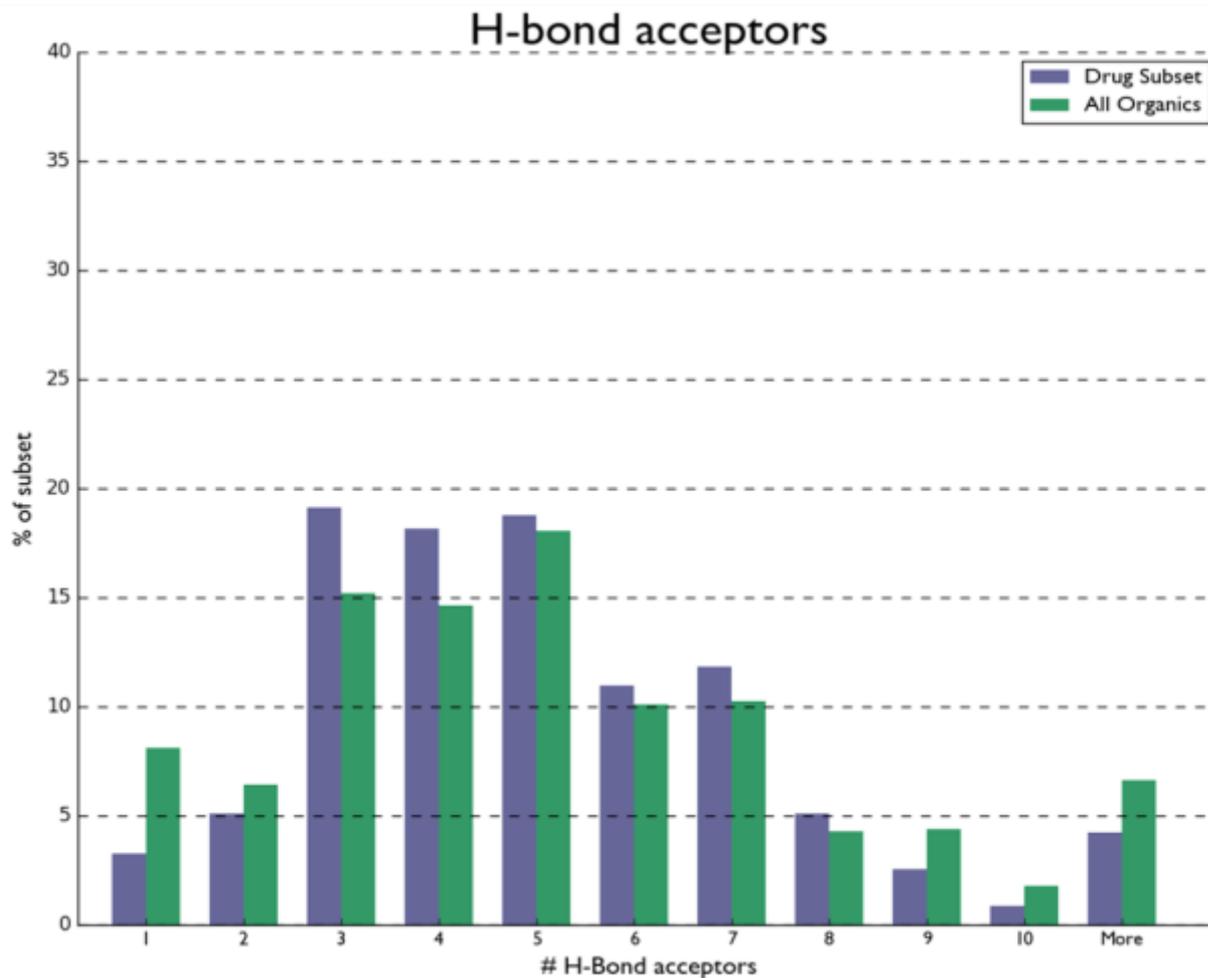
Comparison to organic molecules in the CSD - ALogP

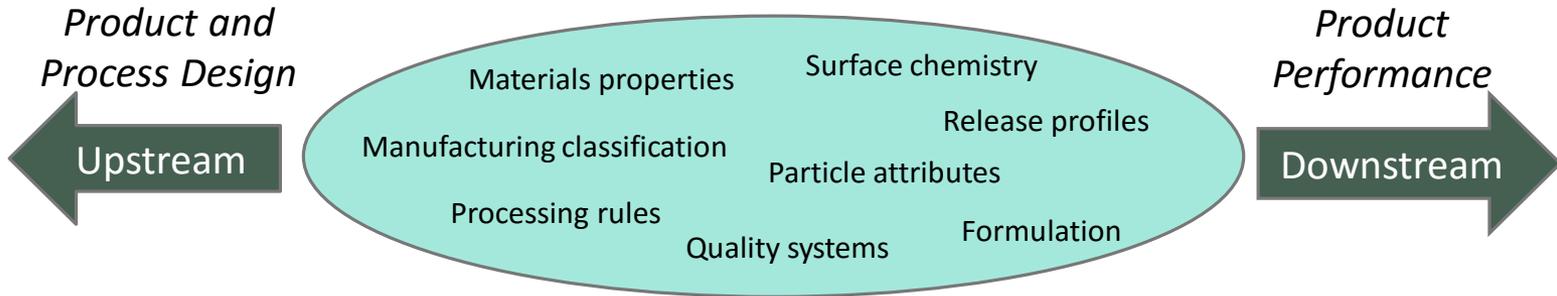


- Clear tendency towards greater water affinity for drug molecules

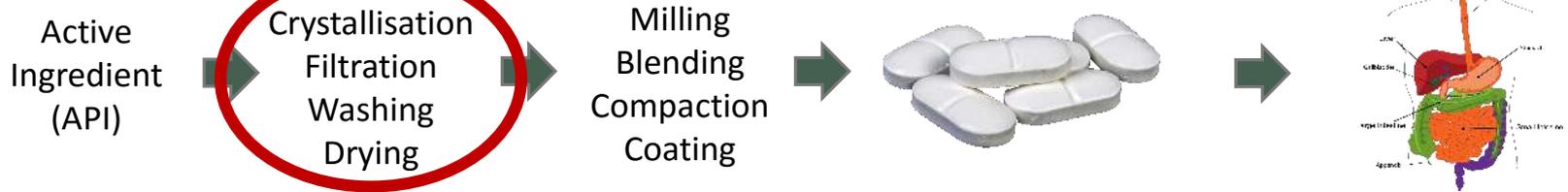
Comparison to organic molecules in the CSD – H-Bond Donors







Primary Manufacturing - Secondary Manufacturing



Processes

Products

Performance

Design and control of optimised development & manufacturing processes through data analysis and first principle models

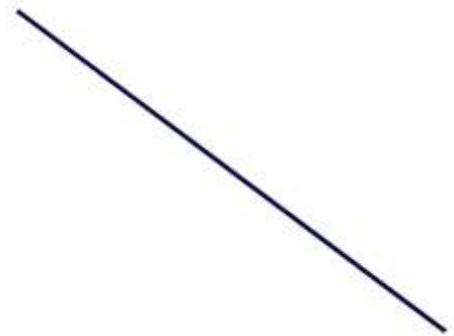
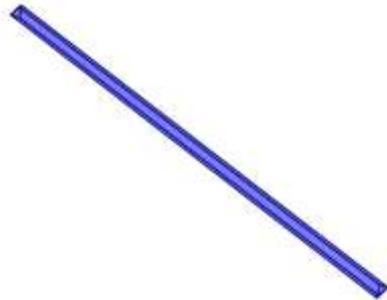
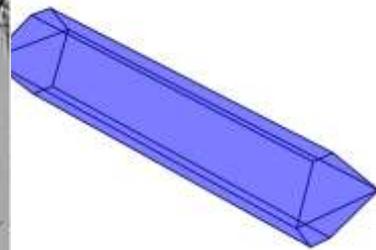
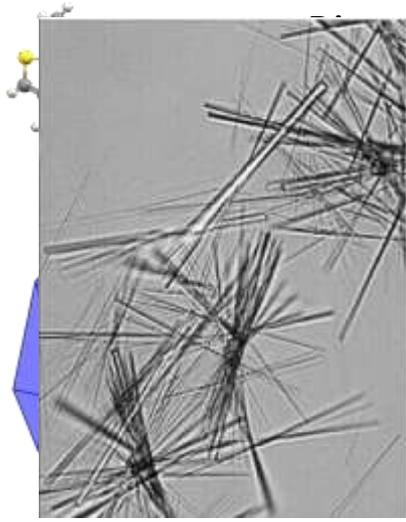


- A large aspect of the filtering, drying and flow properties of a solid crystalline form is influenced by the crystal habit or morphology
- Various ways to predict this, but very few take the crystallisation conditions into account, or tell you how dynamic this property is likely to be
- A tool we are currently working on, uses forcefield potentials to investigate the effects of supersaturation on the growth rates of crystal faces

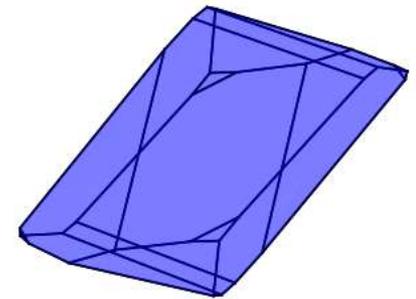
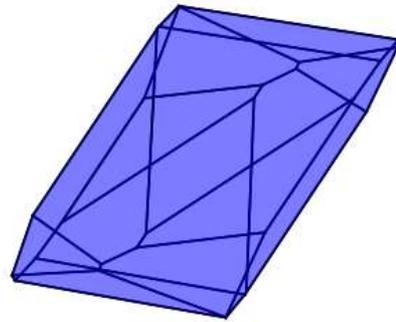
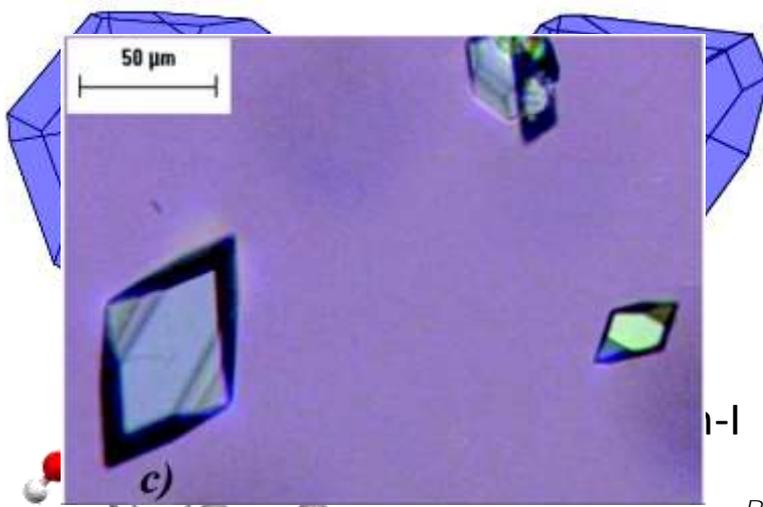


Simulated impact of supersaturation

Avir Form-II



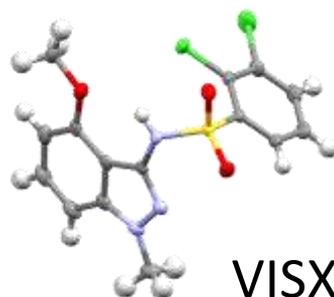
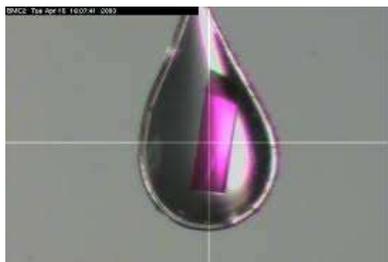
Supersaturation



- Bauer J.F., *Pharmaceutical Solids*, *Journal of Validation Technology* [Winter 2009]
- *Cryst. Growth Des.*, 2008, 8 (9), pp 3316–3322



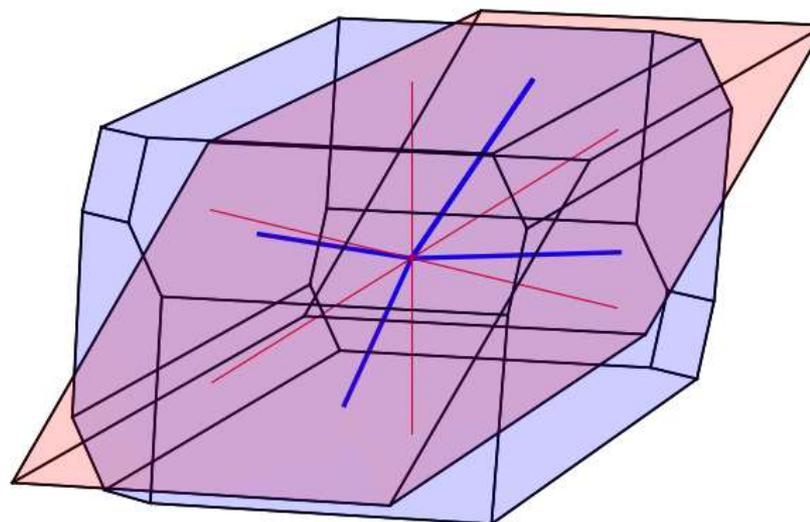
Linking experimental and predicted crystal morphology



VISXUS

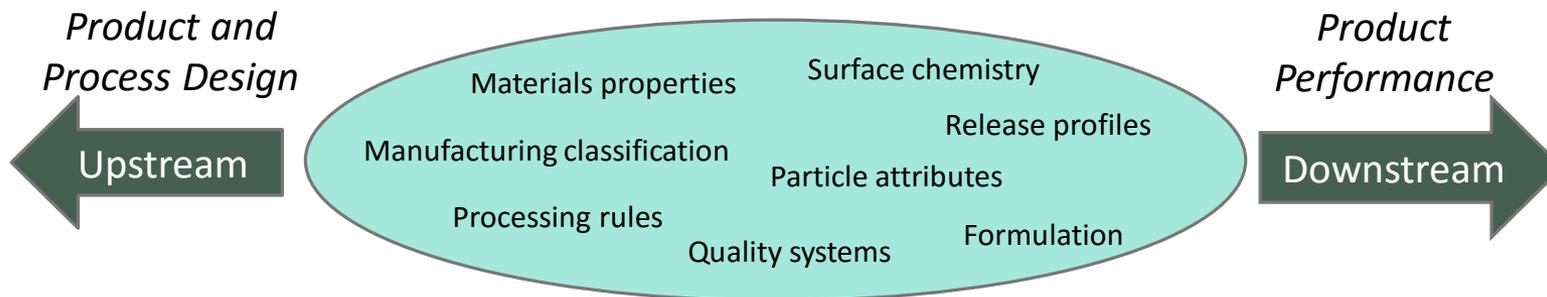
Crystal Face 1	Crystal Face 2	Dihedral angle
(-1, 1, 0)	(1, -1, -1)	155.8158
(-1, 1, 0)	(-1, 1, 1)	155.8158
(-1, 1, 0)	(1, 0, 0)	146.1096
(-1, 1, 0)	(1, -1, 0)	180
(-1, 1, 0)	(0, 0, 1)	95.45741
(-1, 1, 0)	(1, 0, -1)	140.2617
(-1, 1, 0)	(1, -1, -1)	155.8158

```
loop_  
_exptl_crystal_face_index_h  
_exptl_crystal_face_index_k  
_exptl_crystal_face_index_l  
_exptl_crystal_face_perp_dist  
0.00 1.00 1.00 0.239  
0.00 -1.00 2.00 0.451  
0.00 -2.00 -1.00 0.216  
0.00 0.00 -1.00 0.201  
1.00 0.00 -1.00 0.574  
-1.00 0.00 1.00 0.383
```



RMSD: 10.14





Primary Manufacturing - Secondary Manufacturing



Processes

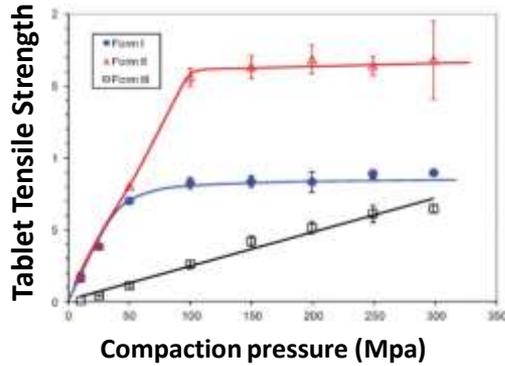
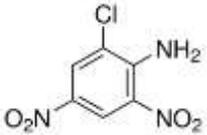
Products

Performance

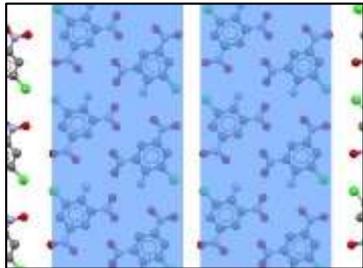
Design and control of optimised development & manufacturing processes through data analysis and first principle models



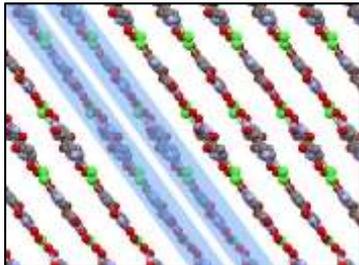
6-chloro-2,4-dinitroaniline



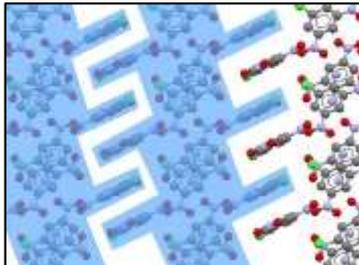
Form-I



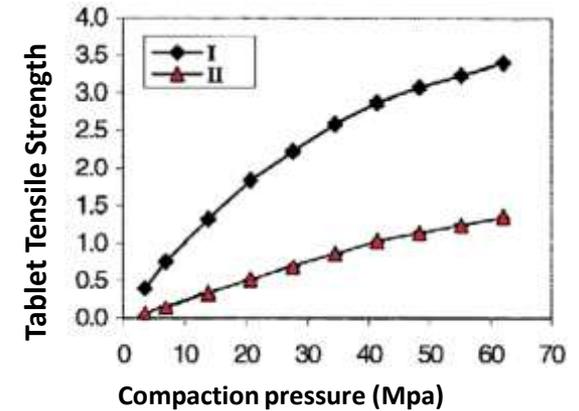
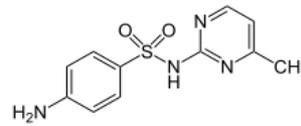
Form-II



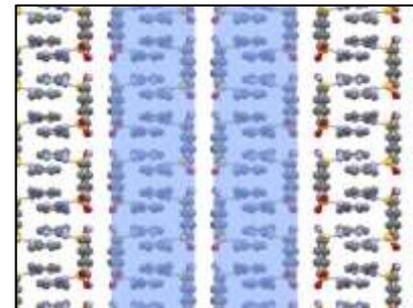
Form-III



Sulfamerazine



Form-I



Form-II

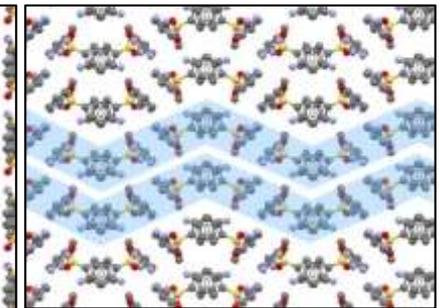
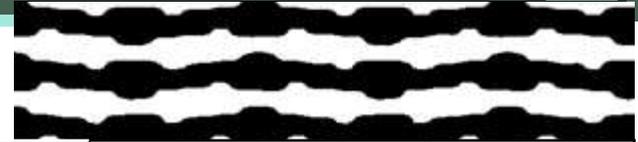


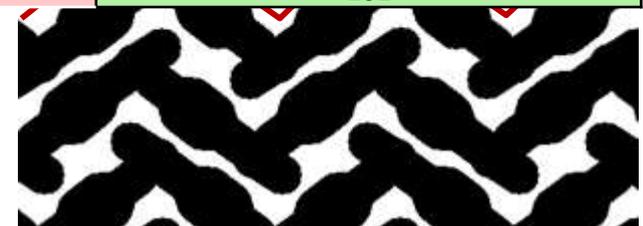
Image analysis Approach: OpenCV



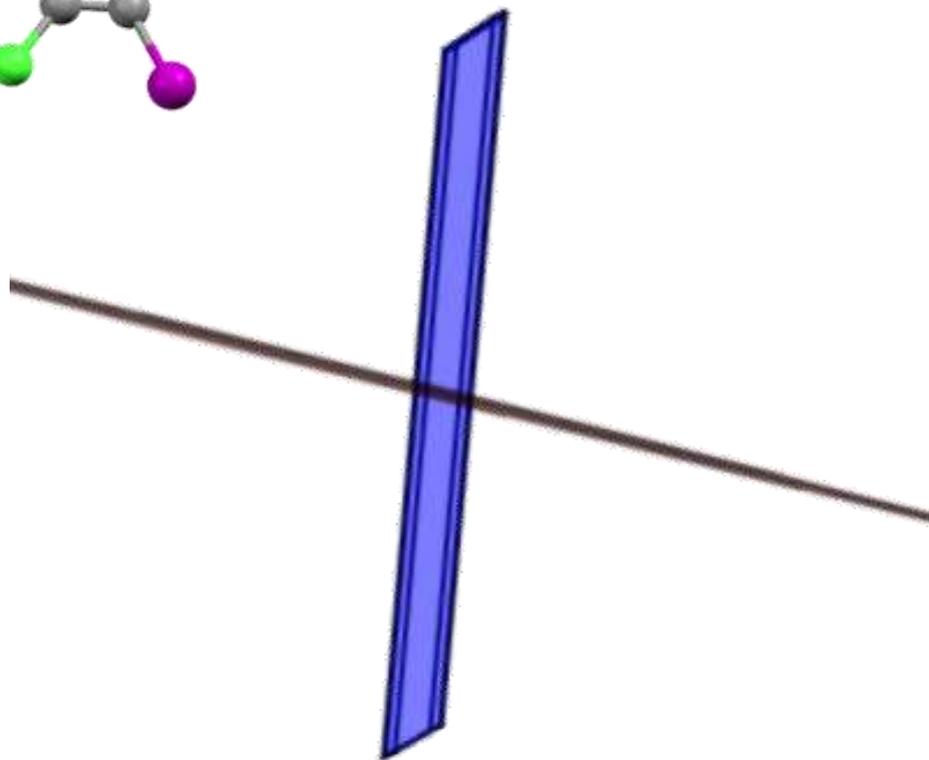
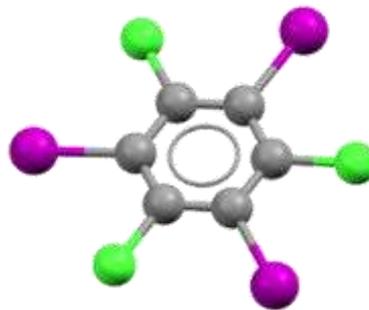
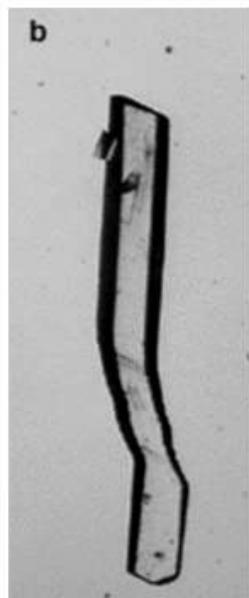
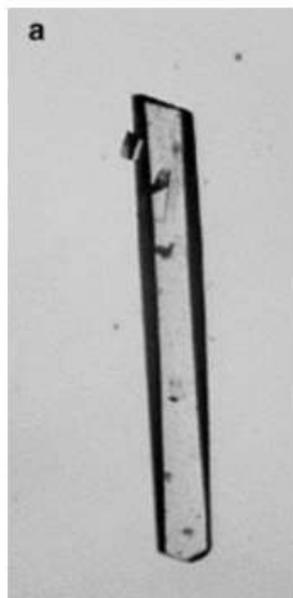
Crystal	Observed	Dreiding	cvff	compass	CCDC Rugosity tool
SLFNMA01	020	002	002	002	010
SLFNMA02	020	020	020	020	010
260457 (UCECAG03)	001	001	001	001	001
CITRAC10	002	002	002	002	001
260456 (UCECAG02)	001	001	001	001	001
PUPBAD01	10-2	10-2	020	011	10-2
PUPBAD02	101	10-1	011	020	101
HXACAN	002	002	200	200	001
HXACAN01	010	110	110	110	010
DIJVOH	002	200	200	200	001
260455 (UCECAG01)	10-1	100	100	100	10-1
ethyl paraben (FEGLEI)	101	100	100	100	101
propyl paraben (DUPKAB)	101	100	100	100	101

calculate the miller plane the line follows using the CSD Python API

Plane [0 1 0]
Rugosity = 1.30



Combining tools: Plane of weakness



- Lots more areas the CCDC are looking into developing tools for
- Looking at generating broad landscapes of hypothetical crystal structures to quantify the potential variance that might be encountered in the solid forms of a drug
- Working closely with ADDoPT partners to validate work done so far

