



Building compound archives for the future

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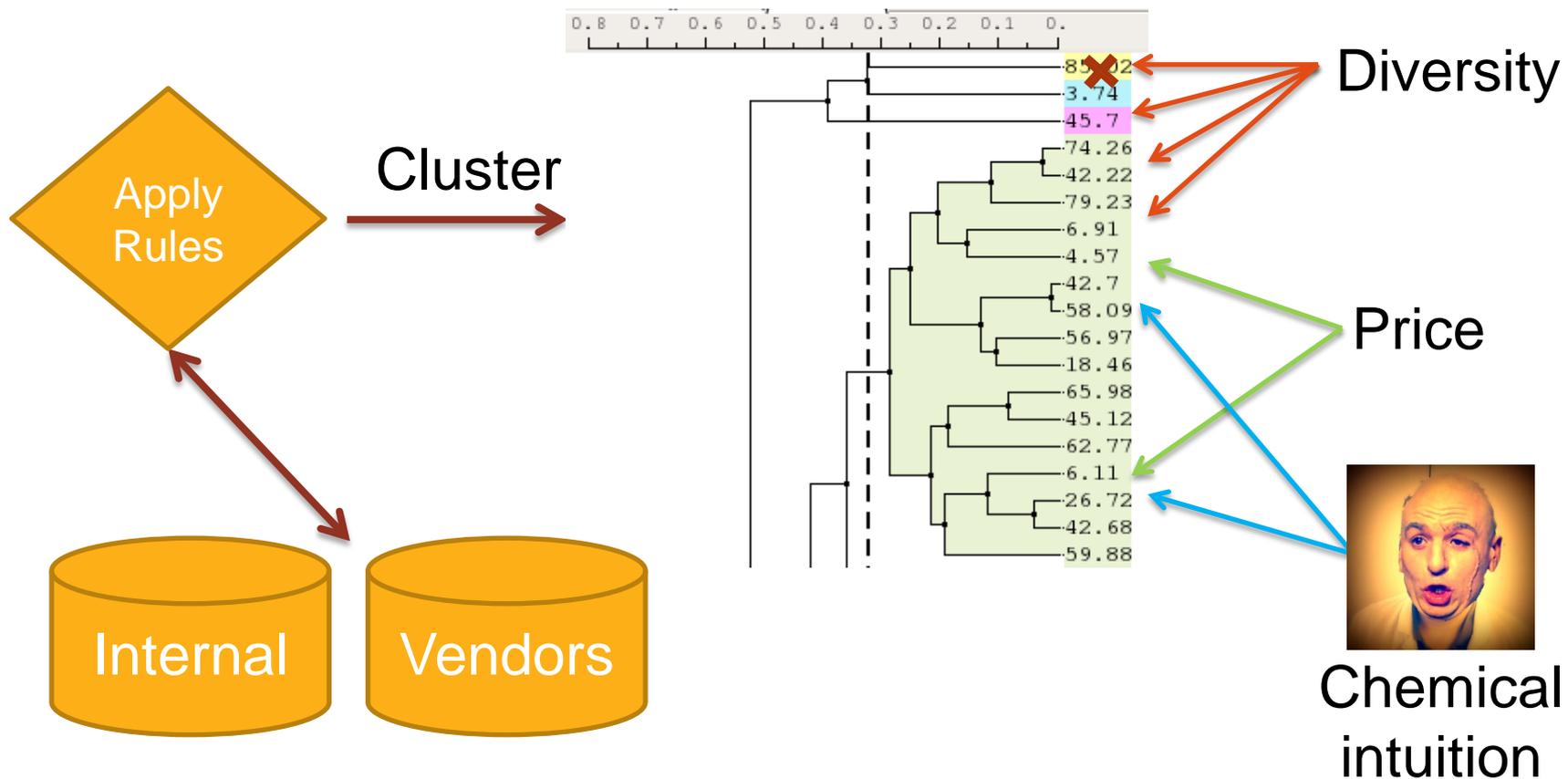
Strasbourg Summer School - Strasbourg, 2016

Overview

- Imagine you are building a new screening deck to provide hits for new biological assays
 - Horvath et al (2014), Design of a General-Purpose European Compound Screening Library for EU-OPENSREEN ChemMedChem, 9, 2309-2326
- Imagine you are not constrained to vendor catalogues
- Will the targets of the future be covered by the compound libraries of today?
 - A critical review of past strategies
- Understanding protein pockets
- Future directions

Archive selection process

General process



How we designed libraries or bought compounds in the past - Filters

Ugliness



Lilly Score

Rules for Identifying Potentially Reactive or Promiscuous Compounds

R. Bruns and I. Watson *J. Med.*

Chem., **2012**, 55 (22), pp 9763–9772

GW, AMGEN,
RPR filters

Aggregation
(Shoichet)

PAINS

Cousin filters

- Don't make/buy ugly compounds unless following a hit

Compound qualities and quantities

HTS –
screen
entire deck

Compounds >
assay slots
Hand-picked



uHTS, Comb Chem,
Peptide libraries

P'Chem profiles bad,
signal:noise bad,
costs high

Diversity sets

- What you ask for is what you get

Physicochemical space

DPD filters
for CMC –
like profiles

‘By eye’ filters
for cellular
assays



Rule-of-5 limits for oral
bioavailability

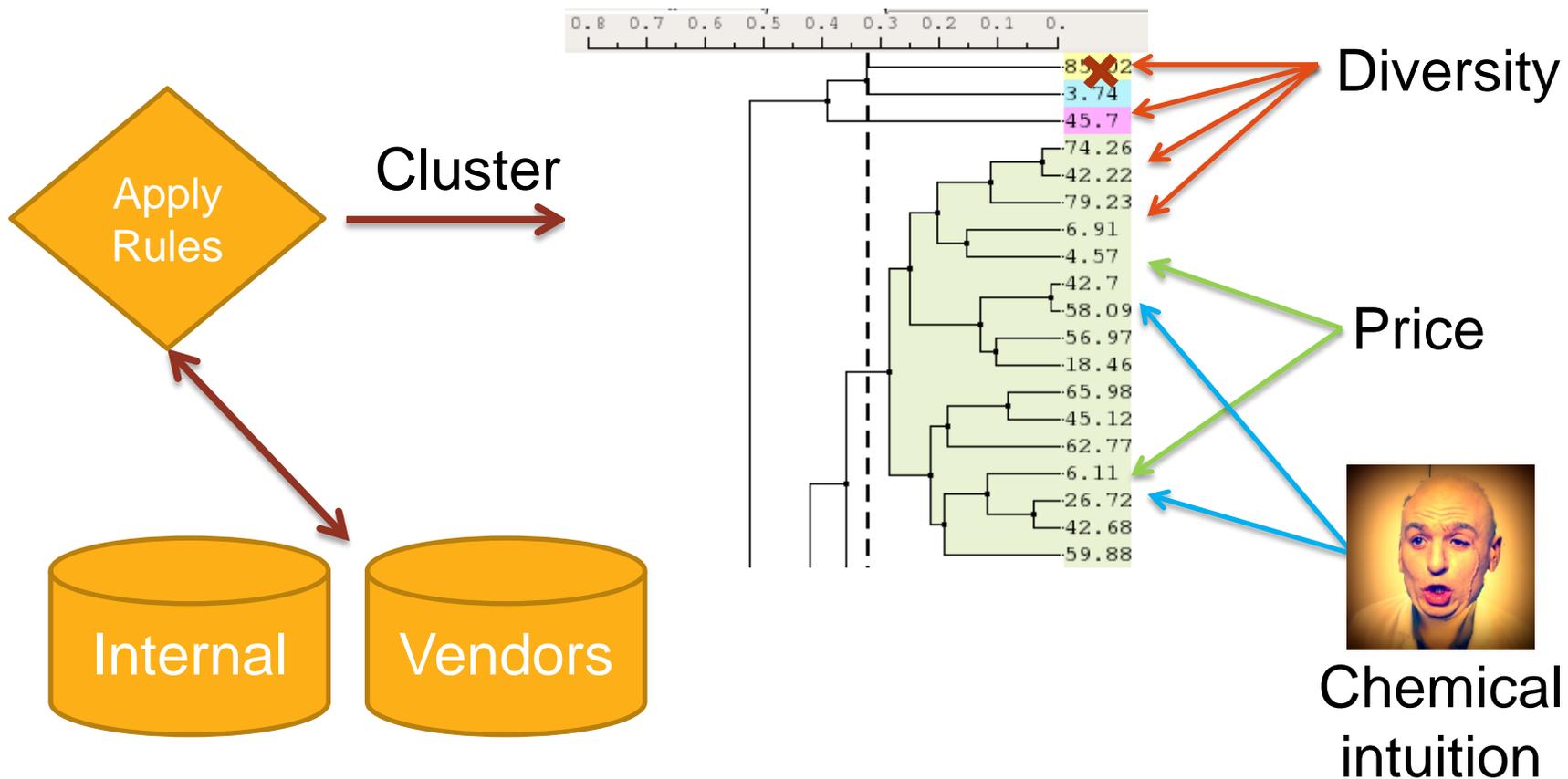
High LogP strongly
linked to promiscuity

High LogP not
strongly linked to
promiscuity

- Everything in moderation but limits can be broken

Archive selection process

General process



Why are we still failing to find hits?

- Chemical Space is too vast
 - We cannot make everything we can design
 - We cannot store everything we could make
 - We cannot screen everything we could store
 - Current archives represent mostly what we know about (old LO series, known targets)



Dimensions - I

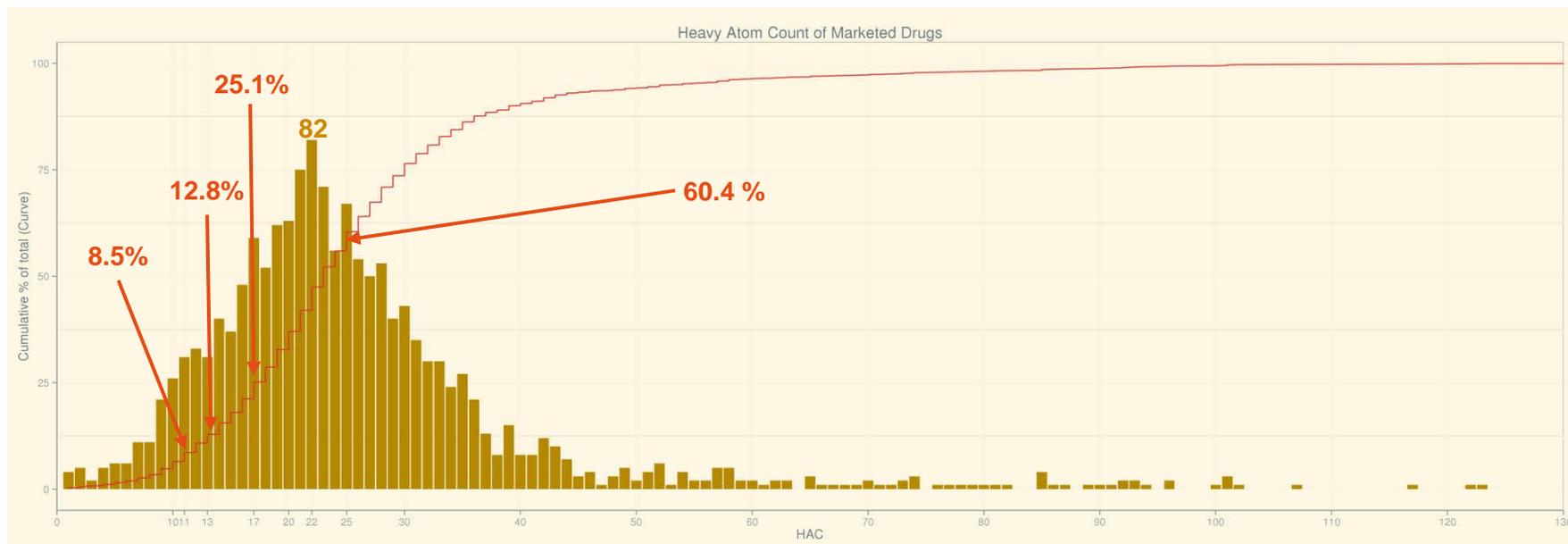
How many molecules can be made with X “heavy” atoms ?

GDB-11 : $2.6 \cdot 10^7$ molecules with 11 non-H atoms ¹ (MW ~ 160)

GDB-13 : $9.8 \cdot 10^8$ molecules with 13 non-H atoms ²

GDB-17 : $1.7 \cdot 10^{11}$ molecules with 17 non-H atoms ³

Extrapolation : $1.0 \cdot 10^{27}$ molecules with 25 non-H atoms (MW ~ 400)



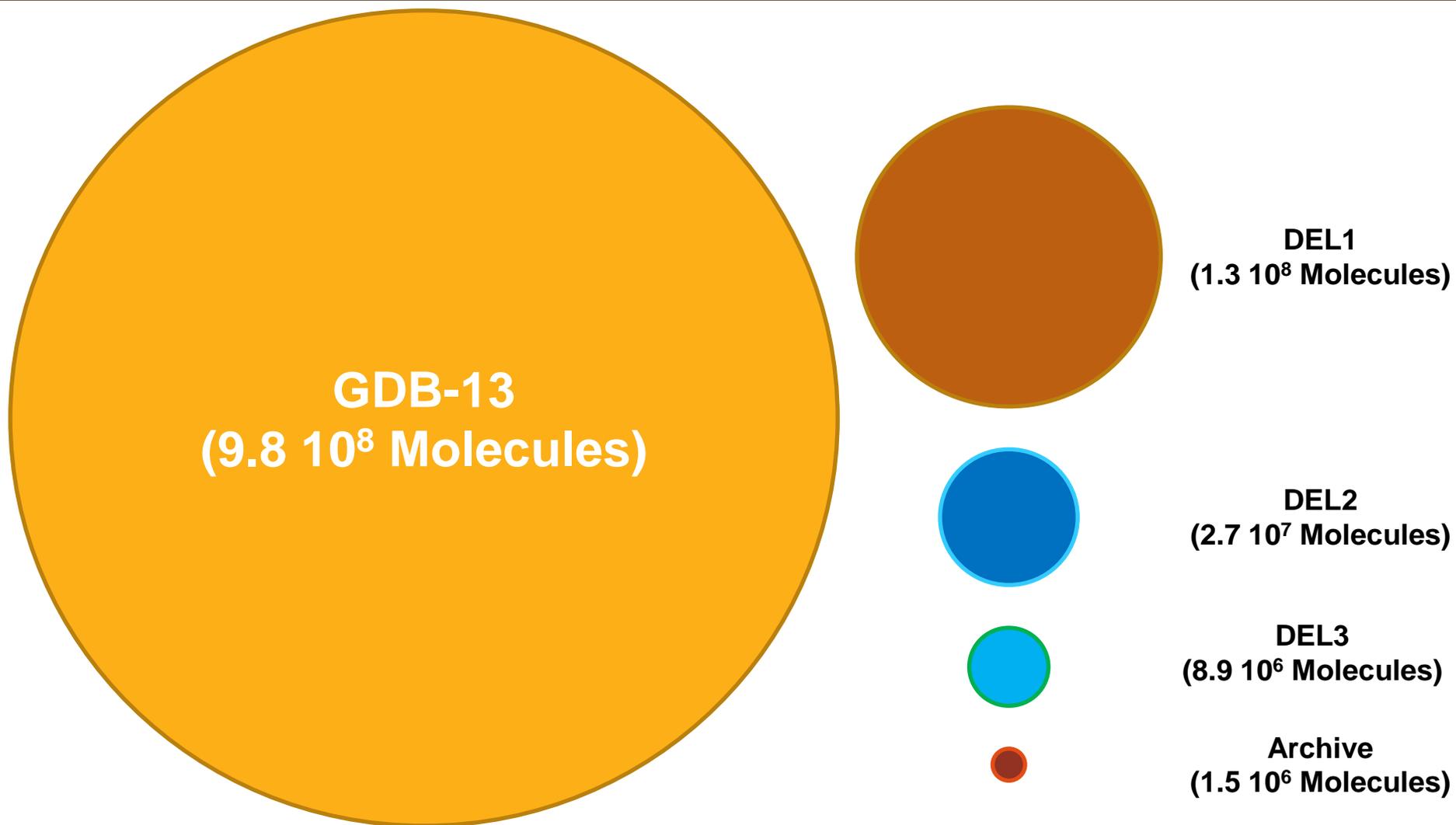
¹ Fink, T.; Reymond, J.-L. Virtual Exploration of the Chemical Universe up to 11 Atoms of C, N, O, F: Assembly of 26.4 Million Structures (110.9 Million Stereoisomers) and Analysis for New Ring Systems, Stereochemistry, Physicochemical Properties, Compound Classes, and Drug Discovery. *J. Chem. Inf. Model.* **2007**, *47* (2), 342–353.

² Blum, L. C.; van Deursen, R.; Reymond, J.-L. Visualisation and Subsets of the Chemical Universe Database GDB-13 for Virtual Screening. *Journal of Computer-Aided Molecular Design* **2011**, *25* (7), 637–647

³ Ruddigkeit, L.; Blum, L. C.; Reymond, J.-L. Visualization and Virtual Screening of the Chemical Universe Database GDB-17. *J. Chem. Inf. Model.* **2013**, *53* (1), 56–65.

Dimensions - II

GDB-13 vs. Novartis Archive vs. In-House DELibraries



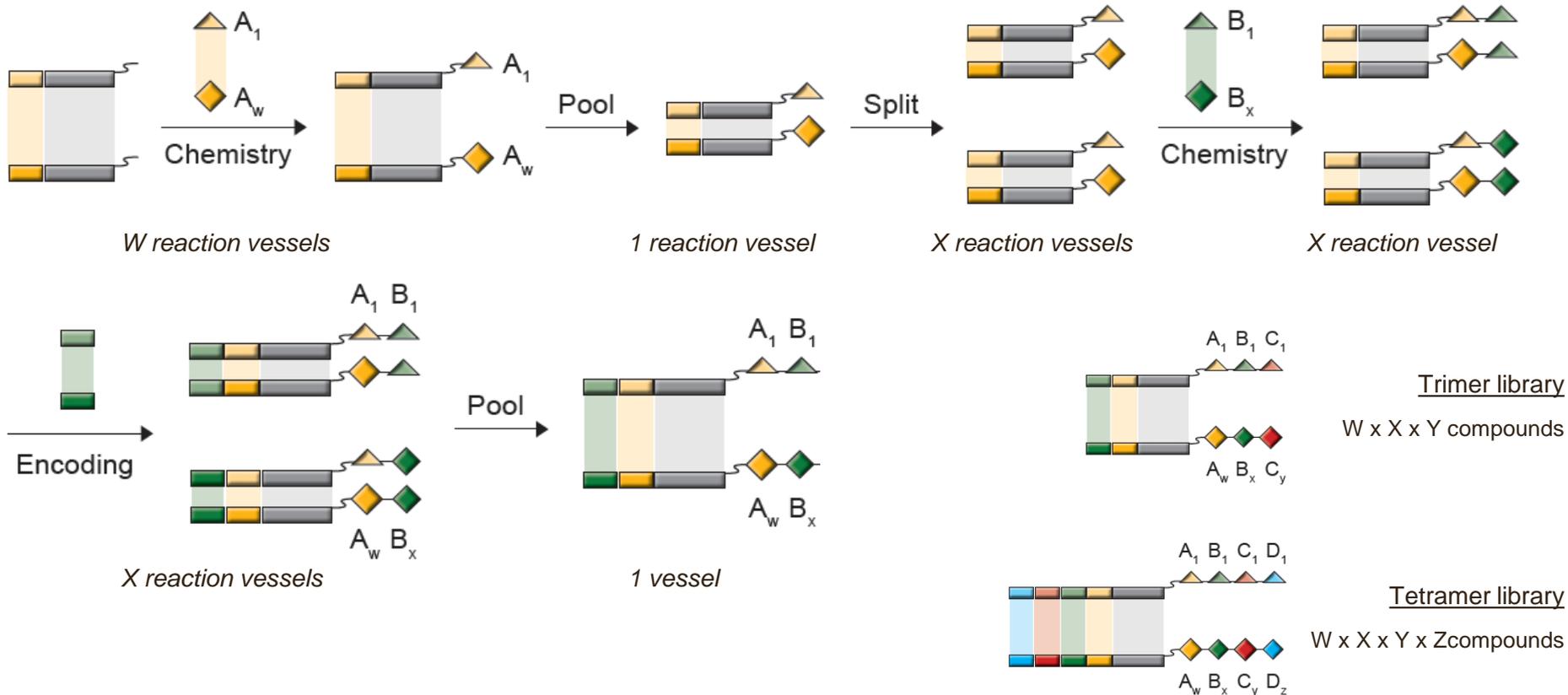
Chemical Space approaches

- 2D similarity metrics (based on substructural features) were designed to retrieve similar molecules quickly, not to assess dissimilarity
 - Manley PW, Stiefl N, Cowan-Jacob SW, Kaufman S, Mestan J, Wartmann M, Wiesmann M, Woodman R, Gallagher N., Bioorg Med Chem, 2010;18(19):6977-86. **Structural resemblances and comparisons of the relative pharmacological properties of imatinib and nilotinib.**
- Overwhelmed by the granularity of chemical space
- Driven by what we know about current structures, not future targets
- No real guide to novelty – libraries driven by synthetic accessibility

DEL technology uses DNA oligonucleotides to record the combinatorial synthesis of organic molecules...

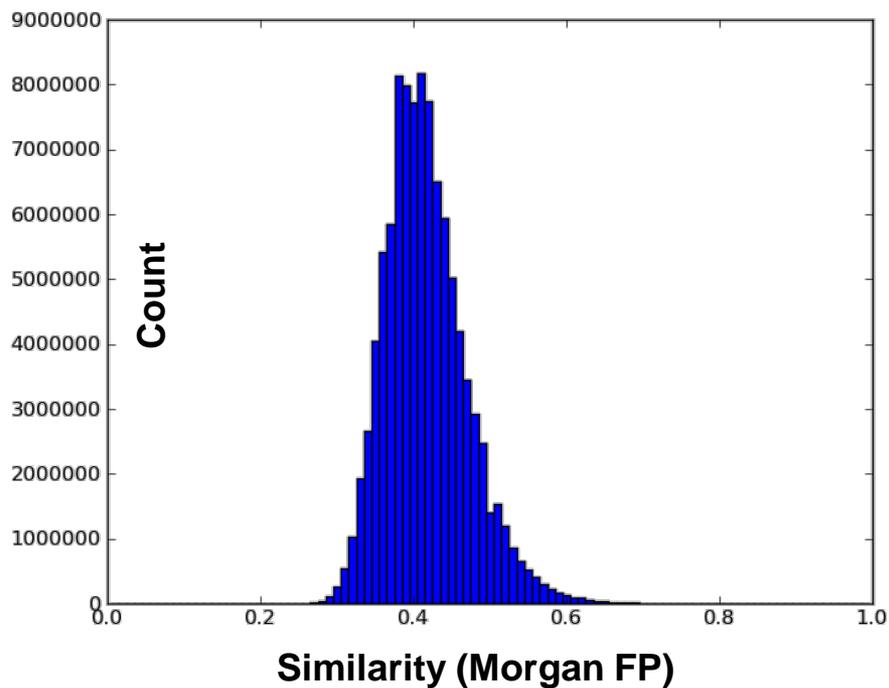
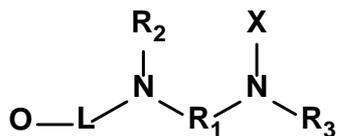
■ Dimer library, $W \times X$ compounds

- Pos 1: W building blocks, A_1 to A_W
- Pos 2: X building blocks, B_1 to B_X



How much does a DEL library contribute?

DEL3 sample similarity distribution vs. NCA

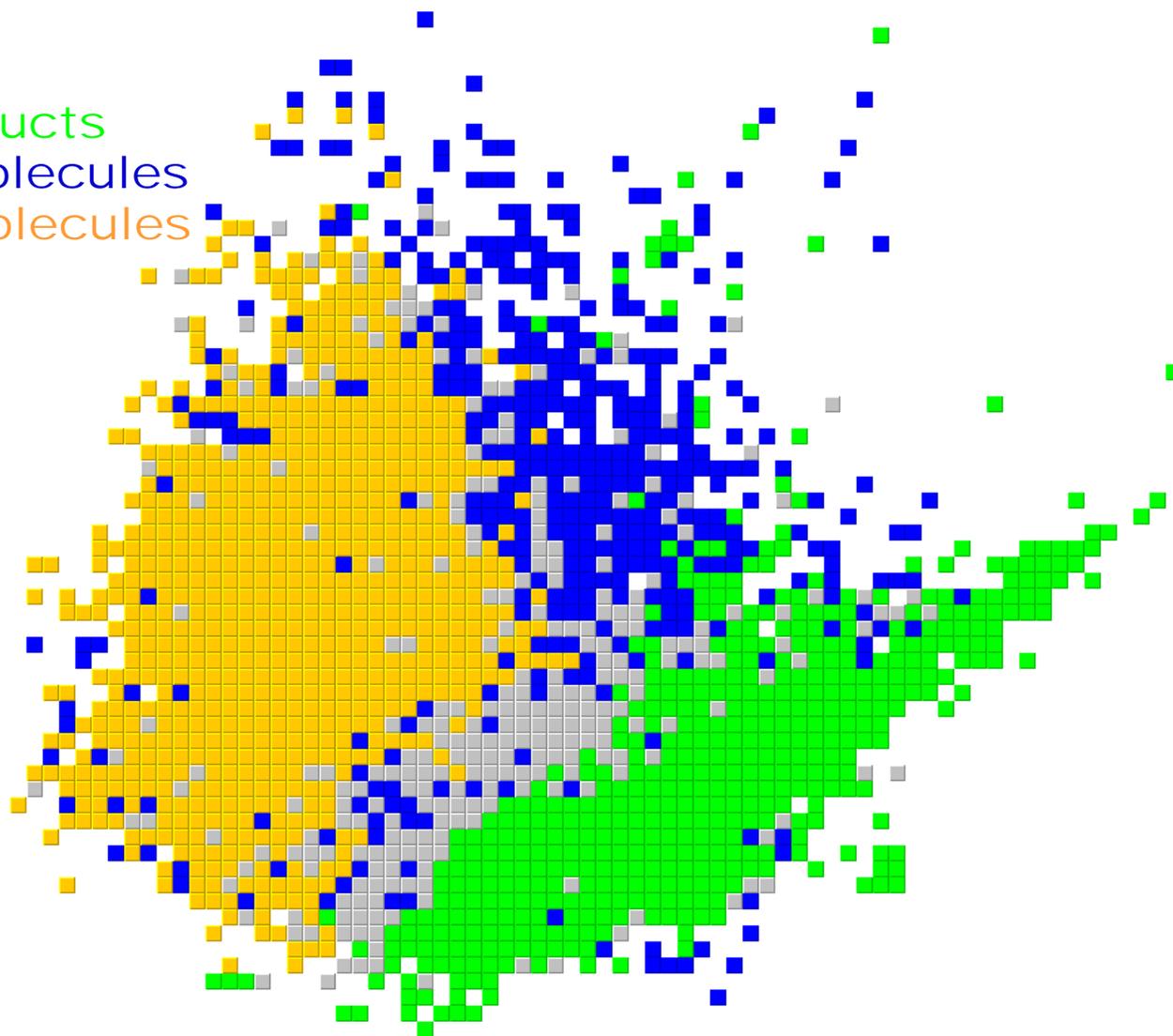


Biological Space - an experimental signpost

- We can classify molecules by their biological fingerprints
 - Petrone, Paula M., Benjamin Simms, Florian Nigsch, Eugen Lounkine, Peter Kutchukian, Allen Cornett, Zhan Deng, John W. Davies, Jeremy L. Jenkins, and Meir Glick. "Rethinking molecular similarity: comparing compounds on the basis of biological activity." *ACS chemical biology* 7, no. 8 (2012): 1399-1409.
 - Can identify which targets/combinations/profiles are well populated
 - And which are not
- Build Bayesian activity models to score new compounds
- Use de novo methods to refine promising proposals
 - Ertl, P., & Lewis, R. (2012). IADE: a system for intelligent automatic design of bioisosteric analogs. *Journal of computer-aided molecular design*, 26(11), 1207-1215.
- But again is limited to what we have assayed
 - Models score for 'kinase-like' not 'new-target like'

Map of Chemistry “Structural Space”

- Natural Products
- Bioactive molecules
- Synthetic molecules

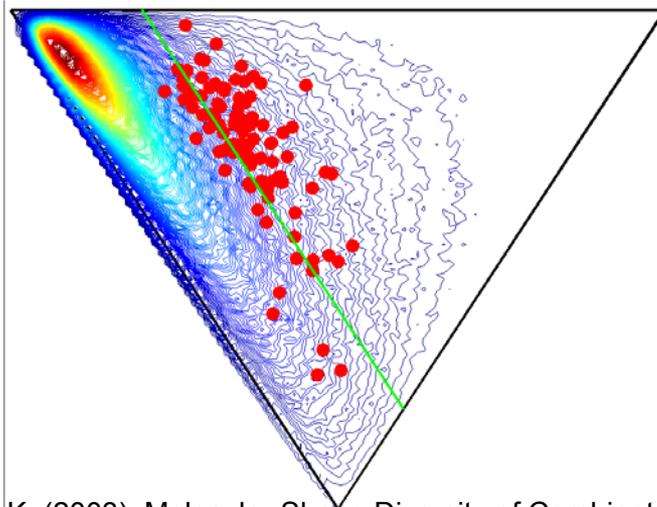


Dark Matter

- Molecules that have shown no activity across many screens
- Useful Pharmacophore but physicochemical properties prevent assay?
 - Signal in SAR model
 - Mark as extrema in physicochemical space
- Pharmacophore that is not sensible in biological space
- Is dark matter clustered with active space?

Shape

- Moments of inertia → shape descriptors



Sauer, W. H. B., & Schwarz, M. K. (2003). Molecular Shape Diversity of Combinatorial Libraries: A Prerequisite for Broad Bioactivity. *Journal of Chemical Information and Computer Sciences*, 43(3), 987-1003.

- But what evidence do we have that diversity in this shape space has any relevance to the future composition of libraries
 - Other than to avoid areas already very heavily sampled?
- This only make sense if one follows the distribution of the space of binding sites – work in progress

Shape 2

- Fraction sp3 has become a popular war cry
 - No more flat heterocycles, long live DOS
 - Even though the analysis was flawed
 - PW Kenny, CA Montanari (2013) Inflation of correlation in the pursuit of drug-likeness JCAMD 27 (1), 1-13
- An in-house analysis did indicate one area where the archive was deficient
- It was also an area where no drugs were found
- Compounds with long linear alkyl chains

Enrichment Drivers: Design vs Accessibility

- There is a slow change from synthetic accessibility being the main driver
 - Metric of cost/compound
 - Acceptance that new chemistry will not be cheap/simple
- Goal of design
 - Reduce the number of targets for which no chemical matter is found
 - Do it in the most compound-efficient way
 - Not by creating mega-archives

So how do we perform such a design?

Filling the holes

- We have illustrated various metrics
 - Ugliness, 2D fingerprints, P'Chem space, Shape, bioFPs
- You can identify holes in the space with any current compound library
- Does it make sense to fill the holes?
- Or are the holes there for a reason?

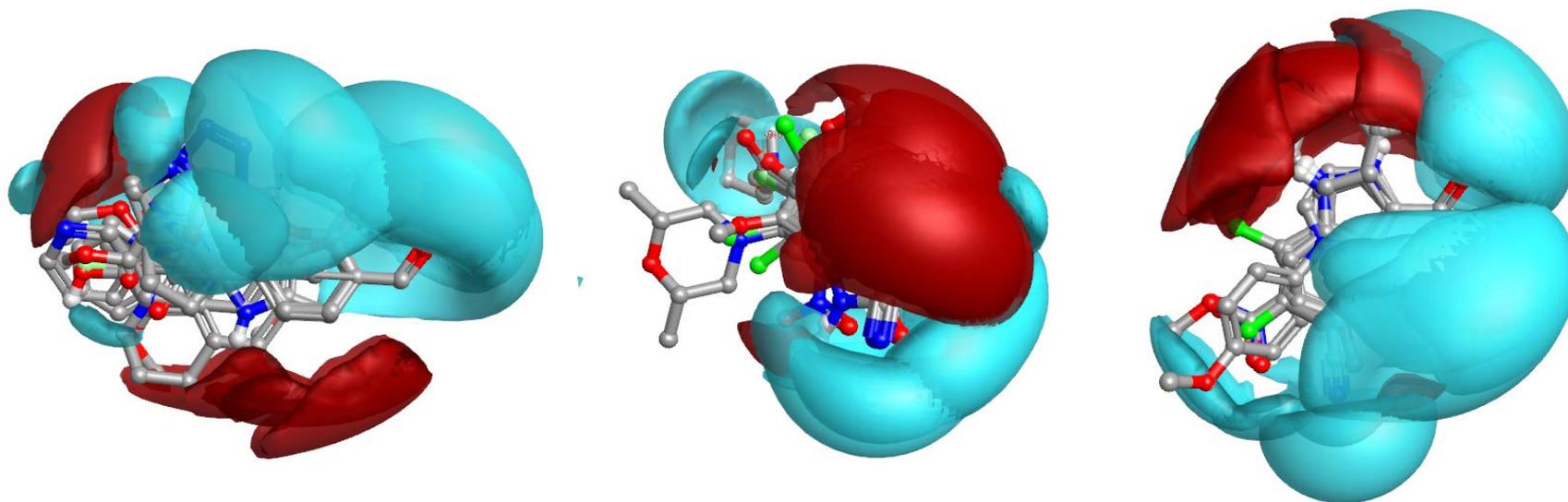
- How do you go about it algorithmically?

DELs: Maximise pharmacophore coverage and attractiveness?

Clustering capping groups that distribute similarly PH4 in surrounding space

- Pharmacophore clustering of all candidates using an algorithm developed in collaboration with Cresset

3 example clusters from the processing of 1832 aldehydes currently in stock



Frequency Fingerprints

von Korff, M.; Freyss, J.; Sander, T. *Flexophore, a New Versatile 3D Pharmacophore Descriptor That Considers Molecular Flexibility*. *J. Chem. Inf. Model.* 2008, 48 (4), 797–810.

- Capture the feature, its geometry and its relative abundance in bins
- Allows operations on large datasets where $N \times M$ comparisons would be infeasible
 - Union (for library proposal)
 - Difference (to exclude populated areas/dark matter)
 - Intersection (to compare designs)
 - Histogram comparisons for distance metrics

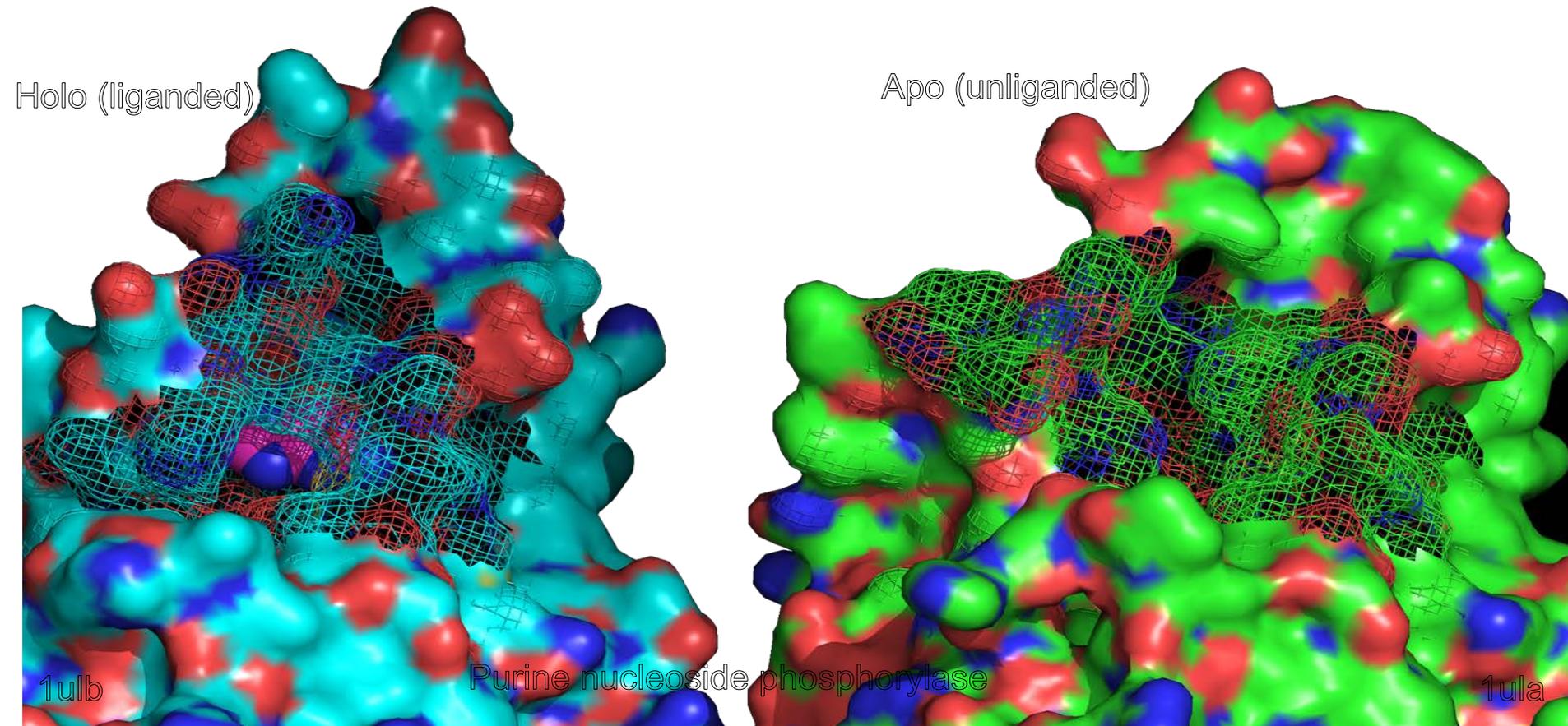
Basic Assumptions

- Ligands exert their effects by
 - Binding to targets
 - (Assuming that they can get to their targets)
- Ligand sites (binding pockets) can be identified
 - Assume that there is a finite repertoire of pockets
 - Druggability vs ligandability of pockets
 - We concentrate on ligandability
 - How much DG_{binding} is possible
 - Build in druglikeness during optimisation
- Build a database of ligand pockets
- Identify which pocket families have cognate ligands, which do not.

Protein pockets

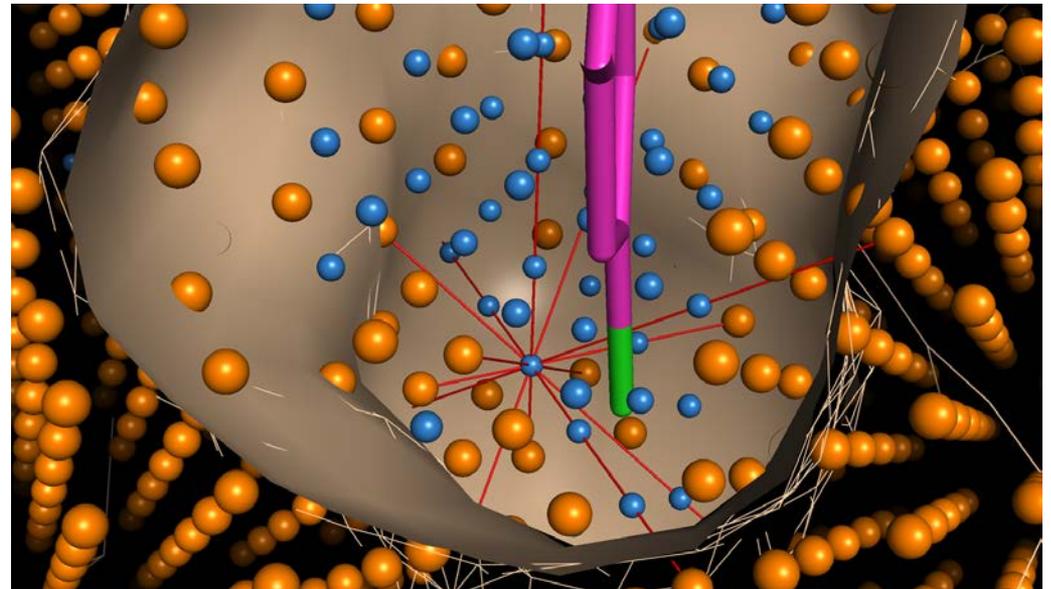
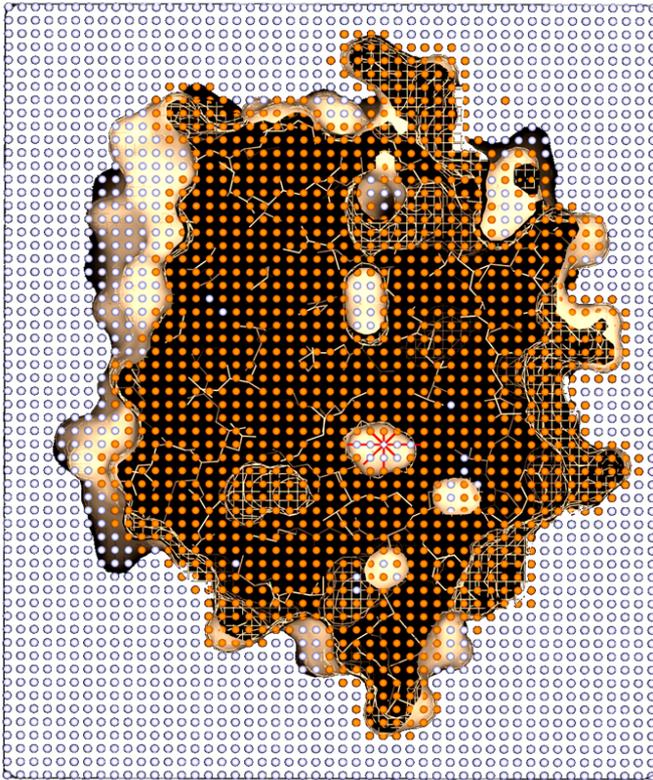
- A less granular universe?
 - Are the number of protein pockets/pharmacopores more limited in scope than chemical space?
 - Abdullah Kahraman, Richard J. Morris, Roman A. Laskowski, Janet M. Thornton (2007) Shape Variation in Protein Binding Pockets and their Ligands *J. Mol Bio*, 368, 283-301
 - Mason, J. S., Morize, I., Menard, P. R., Cheney, D. L., Hulme, C., & Labaudiniere, R. F. (1999). New 4-point pharmacophore method for molecular similarity and diversity applications: overview of the method and applications, including a novel approach to the design of combinatorial libraries containing privileged substructures. *Journal of medicinal chemistry*, 42(17), 3251-3264.
 - What architectures are well-covered?
 - What pockets/cryptic pockets are poorly covered?
 - Can we distinguish pocket space from dark matter space?

Protein cavities to Fingerprints



Ligsite

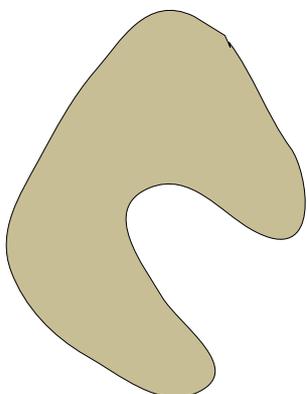
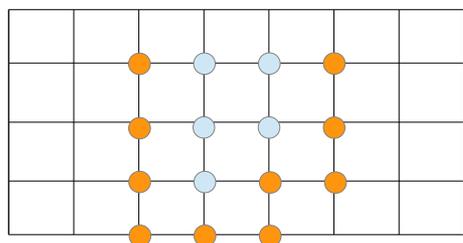
7 axes: X, Y, Z + 4 cubic diagonals



Protein – Solvent - Protein

Identification : Modified LIGSITE algorithm

Grid orientation

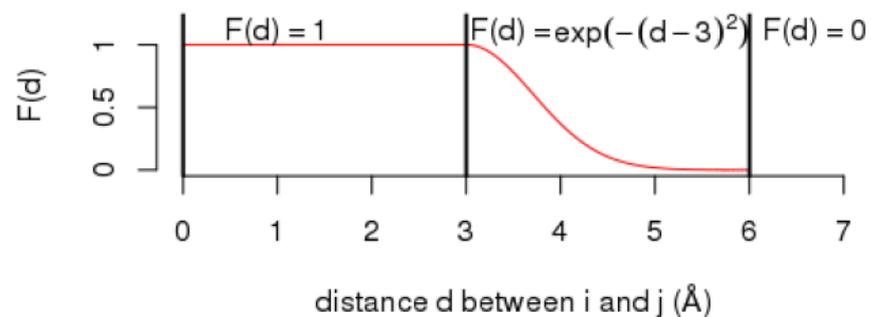


P-S-P ==> P-S < 5.5 Å

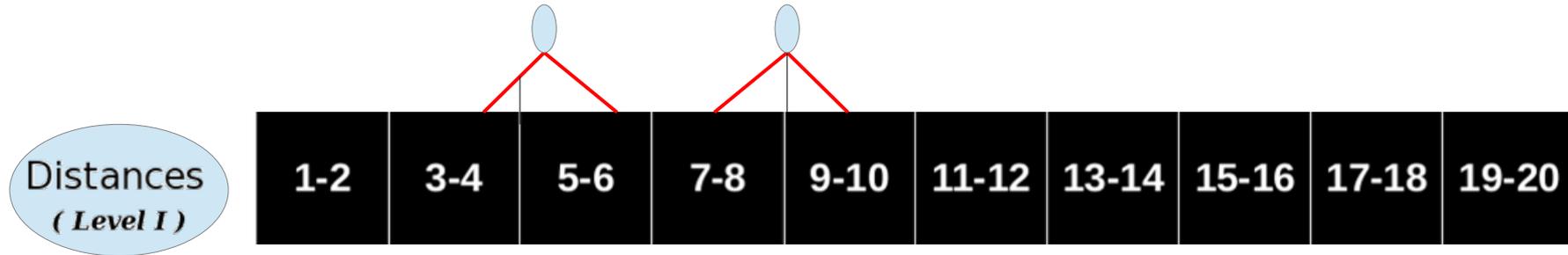
14 levels of buriedness

Buriedness > 8 => pocket

$$PW(i) = \sum_{j=1}^{N_{probes}} Buriedness_j * F(d_{ij})$$

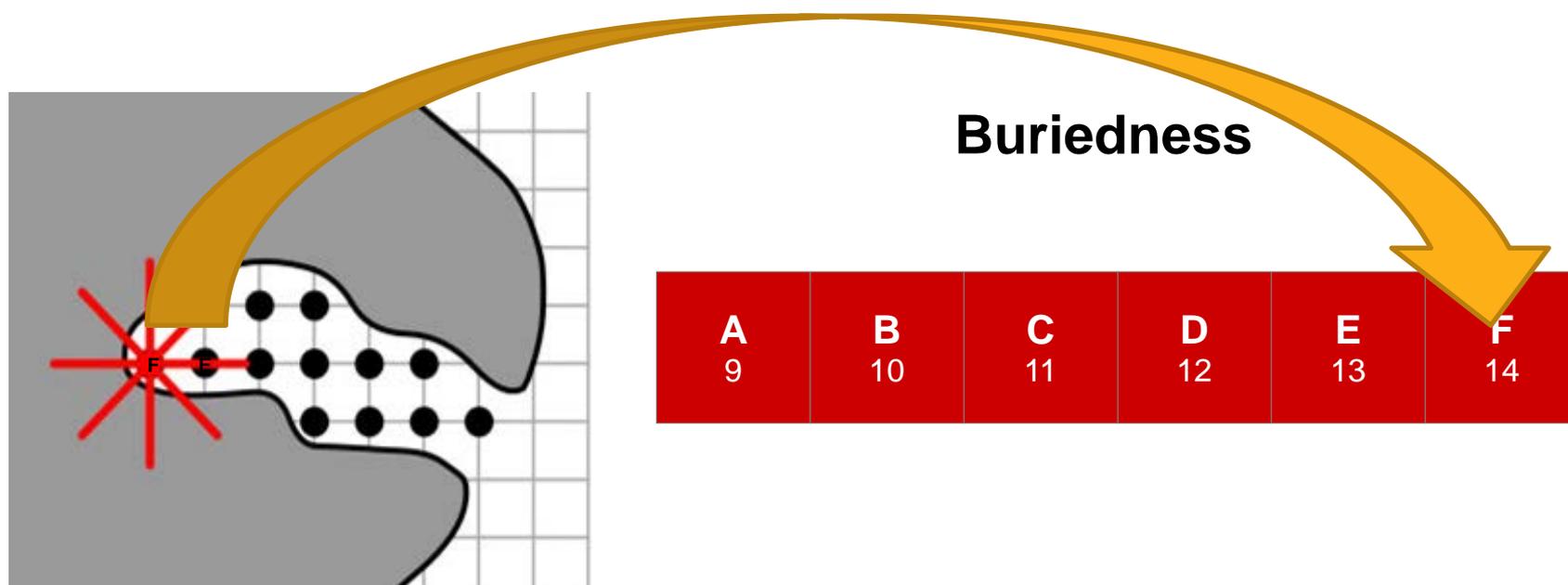


Fingerprint



10 distance bins with fuzzy membership

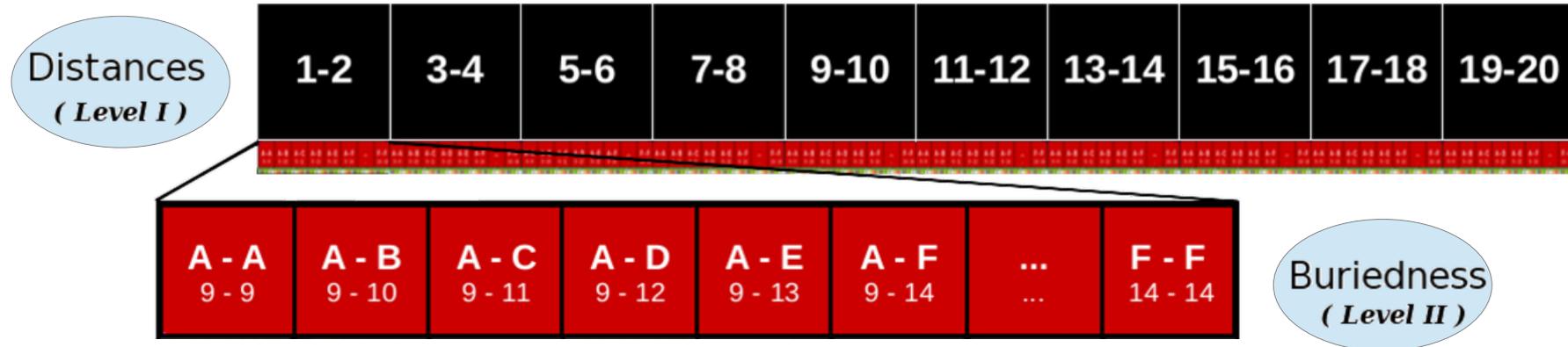
Description : Geometrical features



Weisel et al. Chemistry Central Journal 2007 1:7

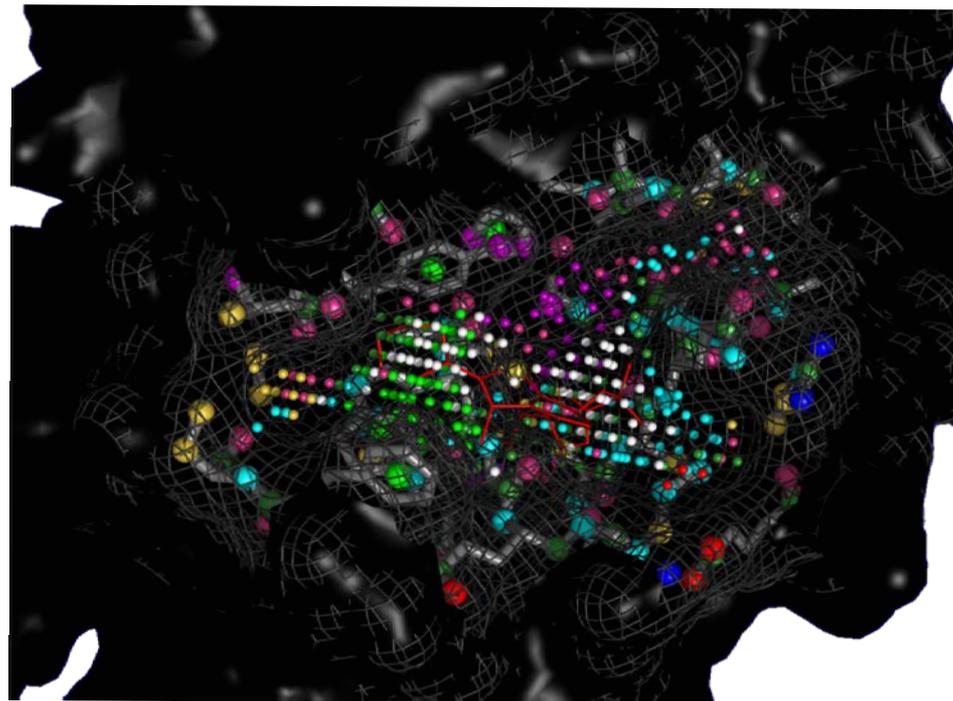
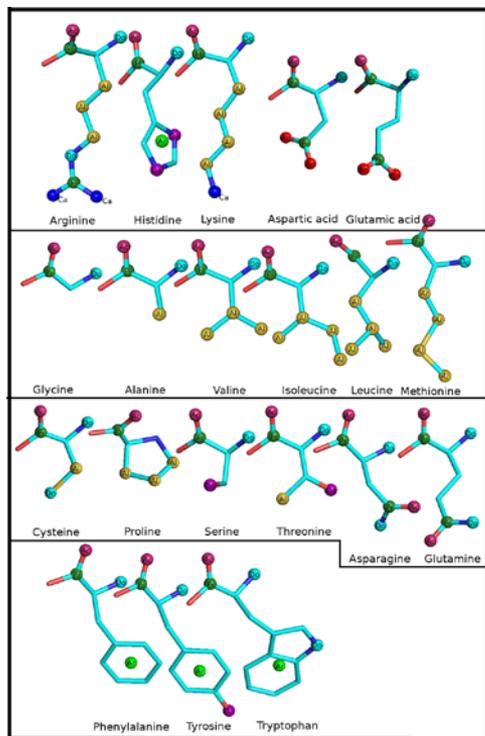
Fingerprint

Distance + buriedness



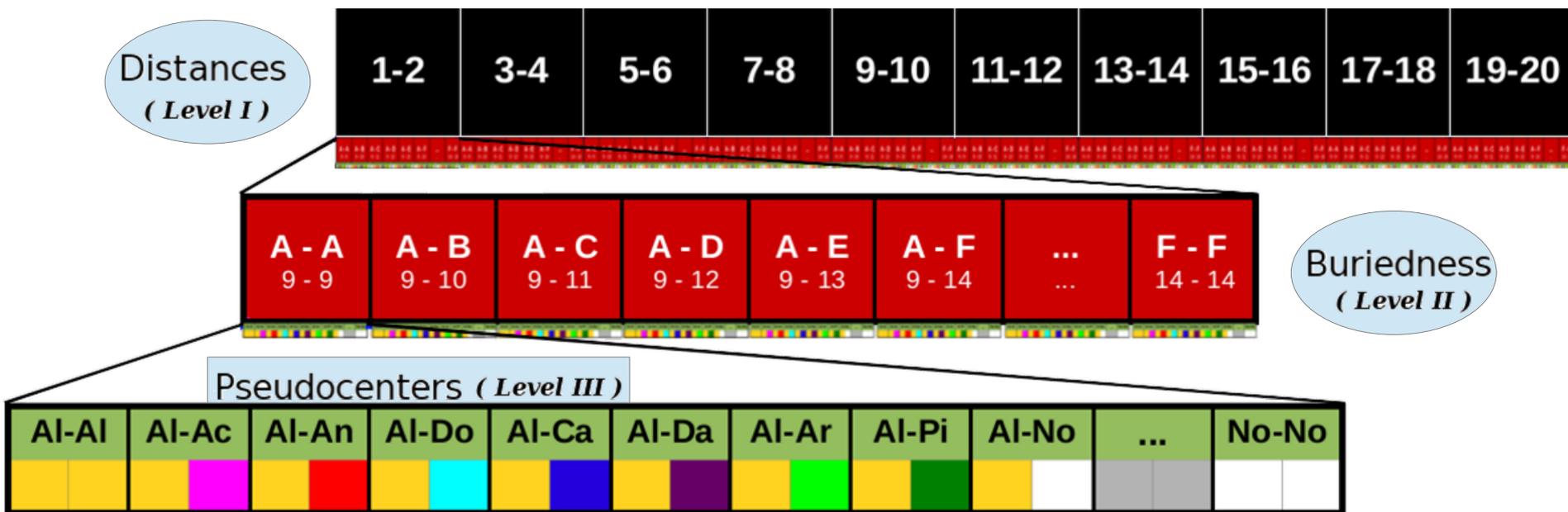
10 X 21 bins

Description : Physico-chemical features



Pseudocenter types for the 20 standard amino acids

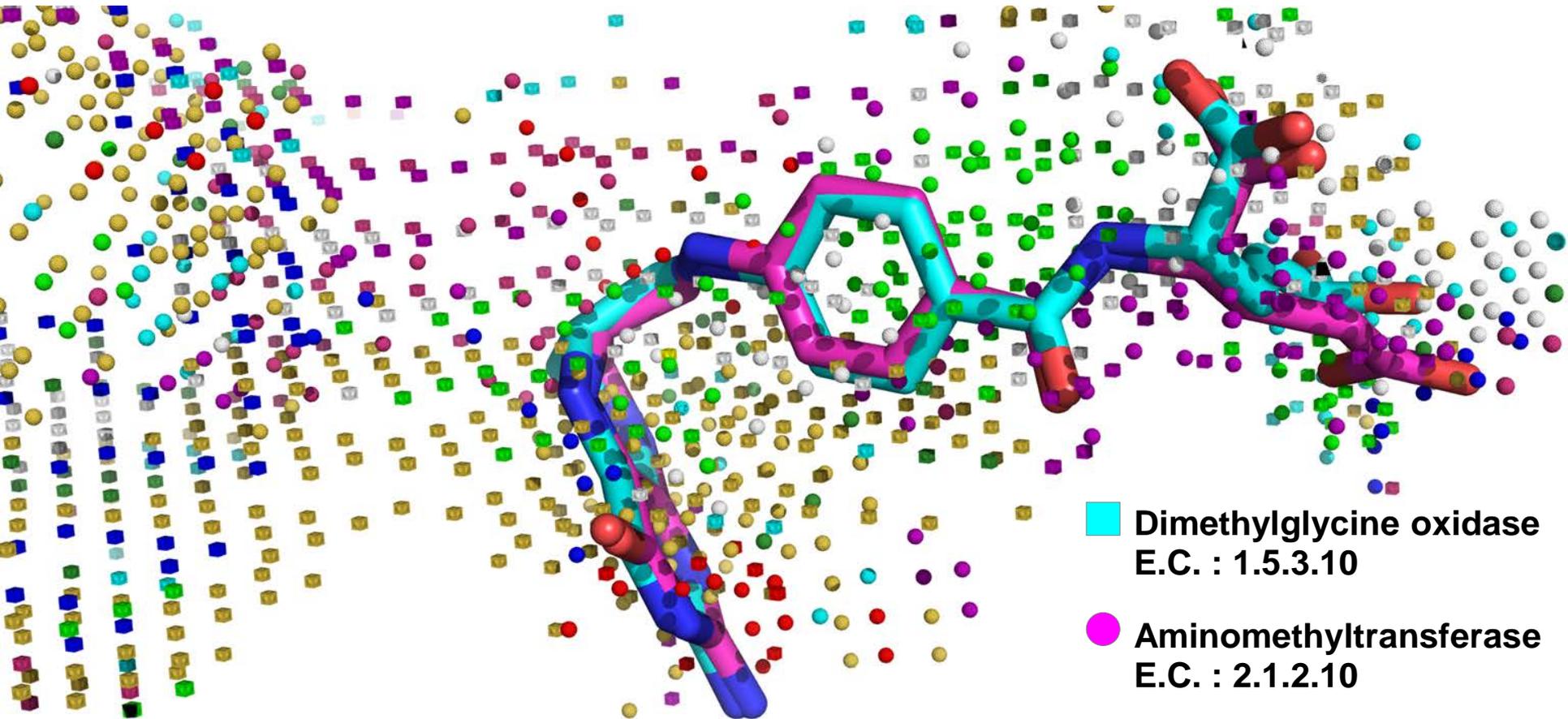
Complete pocket Fingerprint



$$10 \times 21 \times 45 = 9450 \text{ bits}$$

Similar to the 4-centre pharmacophore key but less sparse

Are pockets conserved between unrelated proteins?



Pocket database statistics



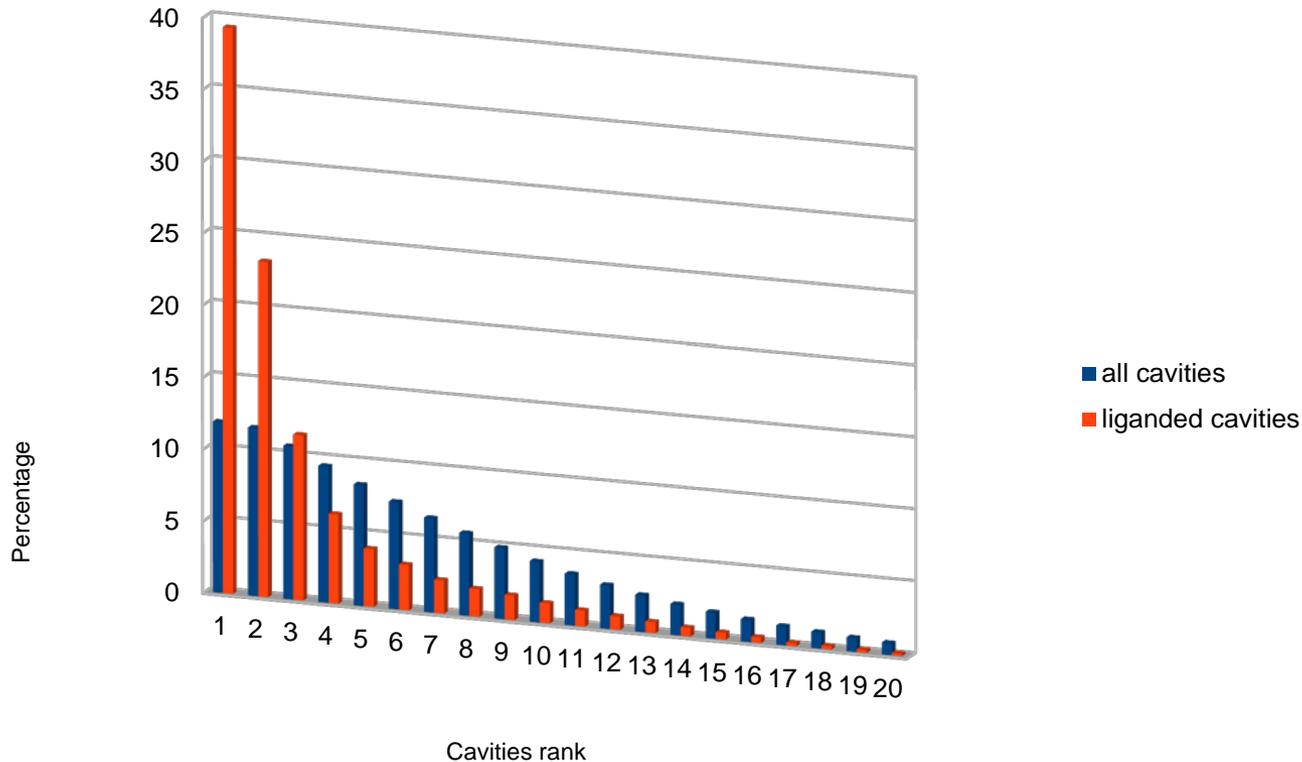
- 86'279 structures
- 811'922 cavities
- 2 hours to return best hits from fingerprint screening

Database : statistics

Average pocket size (number of grid points):

all 175

liganded 350



Work in progress

- We have extended the shape analysis map
 - There are regions of our map that are populated by new chemistries (macrocycles)
 - And regions where we have nothing
 - Looking at the inertial axes of pockets for Sauer shape plot
- Looking to map pocket and ligand space together on a single common fingerprint
 - Weighting for partial matches
 - Allowance for pharmacophore colouring
 - Already using the shapes of empty pockets to search for ligands
- Use the pocket fingerprint to drive the search for poorly populated pockets
 - Correlate with hit rates in HTS?
- Sensitivity to conformation
 - Initial work suggests reasonable robustness

Conclusions

- Early enrichment efforts were driven by abstract considerations of diversity
 - Anchored in what we knew, not in what we were trying to discover
- Diversity through Chemical space too vast to cover in any archive
- Pocket space offers a more tractable and relevant space
- We have succeeded in the first steps of producing a viable FP with enough granularity to be useful
- Each new x-ray structure will give us a map of terra incognita for pocket-based measures of enrichment and diversity

Acknowledgements

- CADD colleagues at Novartis Basel
- Structural biologists everywhere