

#### smarter chemistry | smarter decisions

Analysing selectivity through multi-dimensional activity cliff analysis Tim Cheeseright

#### > Growing and profitable company

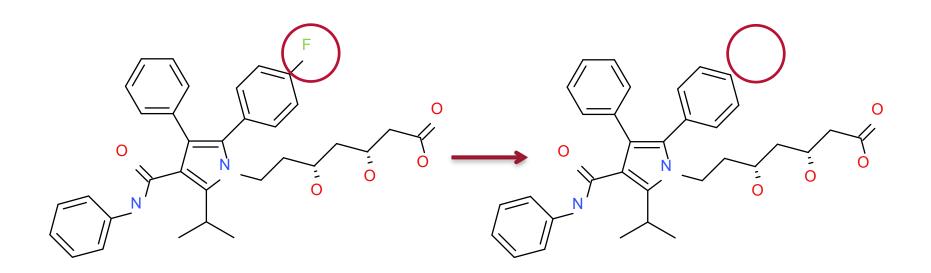
- > 20% year on year growth since 2009
- > 20 People, 12 with PhDs

#### > Primary market pharmaceutical and biotech R&D

- > Software:
  - > 14 of the top 20 pharmaceutical companies use Cresset's technology in their research programmes
- > Consultancy Services:
  - > ~200 collaborative projects delivered to global clients
- > Secondary markets: agrochemicals, flavours and fragrances, consumer health and fine chemicals



## Drug discovery's similarity hypothesis



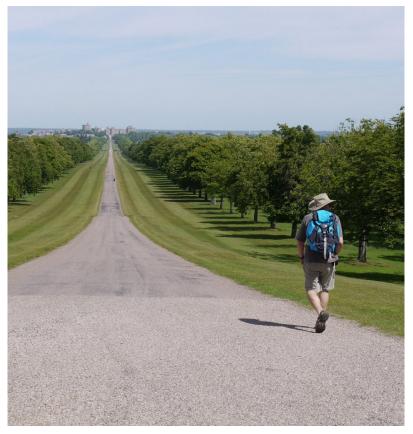
- > Similar molecules have similar activities
- > Small changes lead to small changes
- $\rightarrow$  QSAR, virtual screening, lead optimization



## (Un)Interesting SAR

What about the bits where the similarity hypothesis breaks down?

Nothing happens



Something dramatic happens





## Activity Cliffs – interesting regions of SAR

#### > Many names:

- > Disparity (Merck 1990s)
- > SALI (Guha/Drie 2008)
- > Activity Landscapes
- > Activity Cliffs

## > Definition:

> For each pair of molecules

$$\kappa = \frac{Act_1 - Act_2}{Distance_{12}}$$

> Usually distance = 1 – similarity

> Similarity from 2D fingerprints, tanimoto etc

> Large K indicates an activity cliff



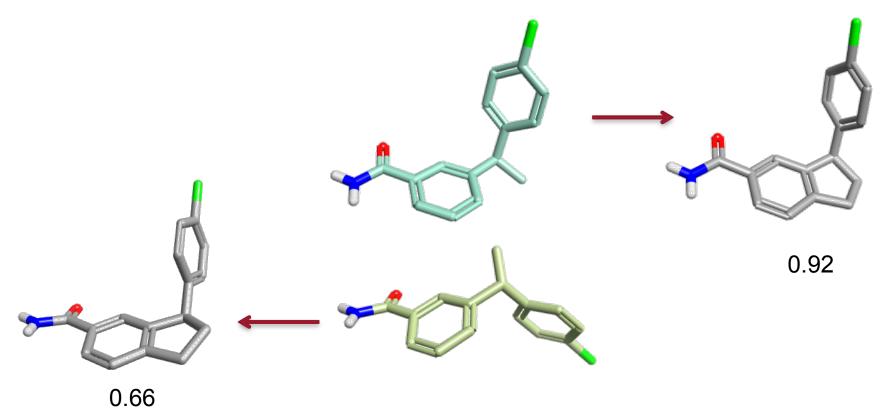
## Gaining understanding of Activity Cliffs

- > Activity cliffs from 2D similarity highly valuable
- > But no explanation for why the cliff is present
- > Without an explanation we cannot use the cliff to design new compounds with confidence
- > True understanding can come from 3D metrics
  - > Shape
  - > Electrostatics
- > What about using 3D similarity from the outset?



## Using 3D similarity

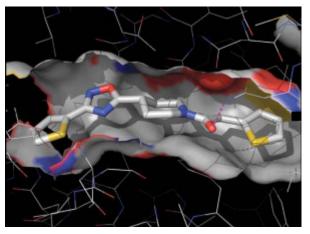
- > 2D metrics are easy: 1:1 map to topology
- > 3D is defined for **conformers**, not for **molecules**





## Context is everything

- > Don't need/want **generic** 3D similarity
  - > Have activity context bound to the protein



- > Align all molecules to known bioactive reference conformer
- > Provides a conformation context to each molecule



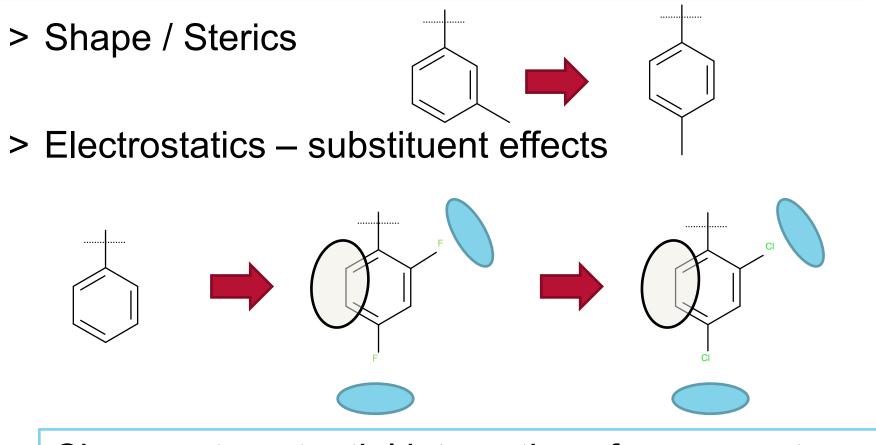
## 3D disparity

- 1. Generate conformers
- 2. Align to reference(s)
- 3. Calculate 3D similarity matrix on aligned conformations

What 3D properties do we want to capture?



## Properties of a 3D similarity



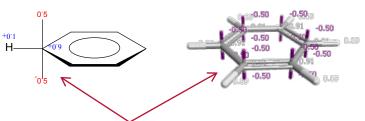
Changes to potential interactions from new atoms

Changes induced in retained portions



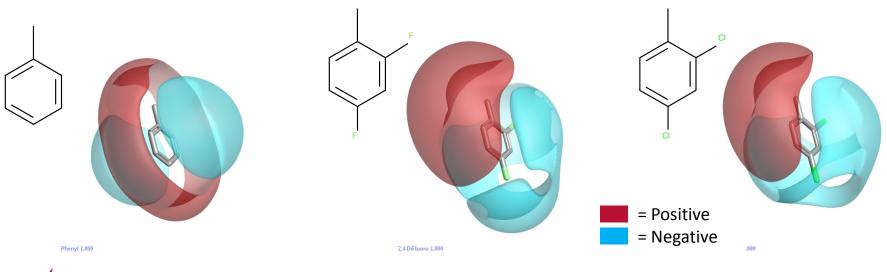
## Detailed electrostatics from XED

> eXtended Electron Distribution gives detailed electrostatic interaction patterns



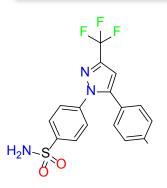
Separation of  $\pi$ - and  $\sigma$ - charges enables modelling of substituent effects

XED adds p-orbitals to get detailed representation of atoms



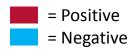


## Field points



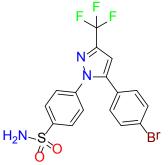
MIP contains too much information to use computationally in a reasonable time

3D Molecular Electrostatic Interaction Potential (MIP)

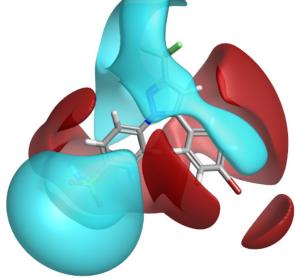




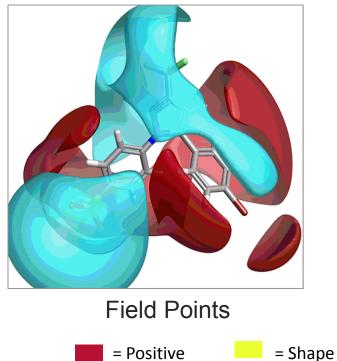
## Field points



MIP contains too much information to use computationally in a reasonable time



3D Molecular Electrostatic Interaction Potential (MIP) Field Points provide computationally tractable framework for electrostatic similarity

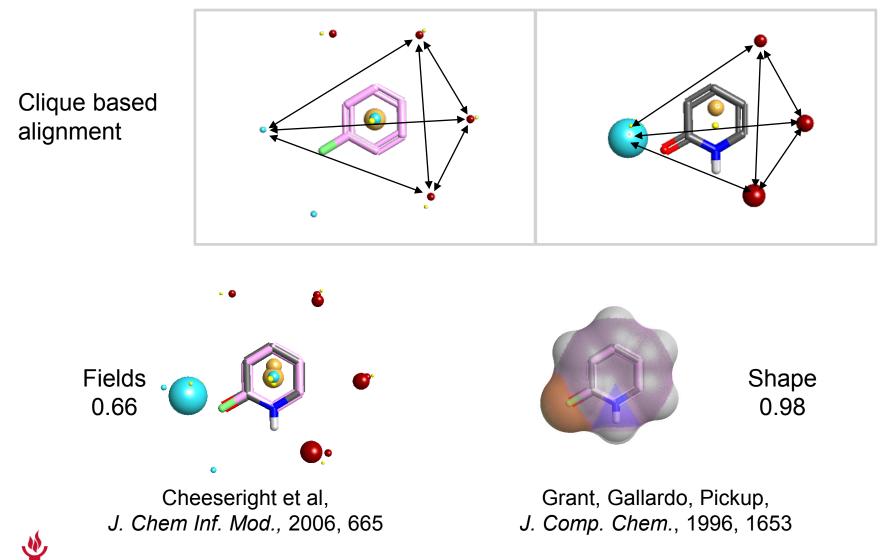


= Negative

= Hydrophobic

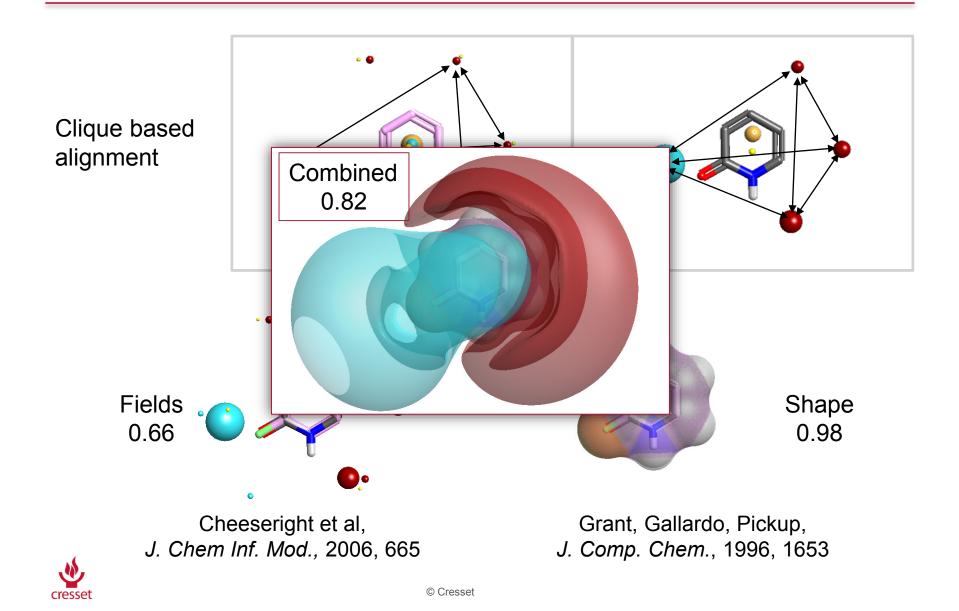


## Alignment, scoring and comparisons



cresset

## Alignment, scoring and comparisons



## 3D disparity workflow

- 1. Generate conformers
- 2. Align to reference(s)
- 3. Calculate 3D shape & electrostatic similarity matrix
  - > Allow small movements
- Calculate disparity matrix from similarity numbers
  - > Similarity cutoff of 0.95 (Distance cutoff of 0.05)

 $Act_1 - Act_2$ 

 $Distance_{12}$ 

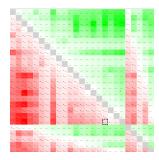
- 5. Visualize
  - > Difficult 100 molecules gives 4950 pairs!



#### Visualization

## > Existing ways to visualize

> Table & Matrix views



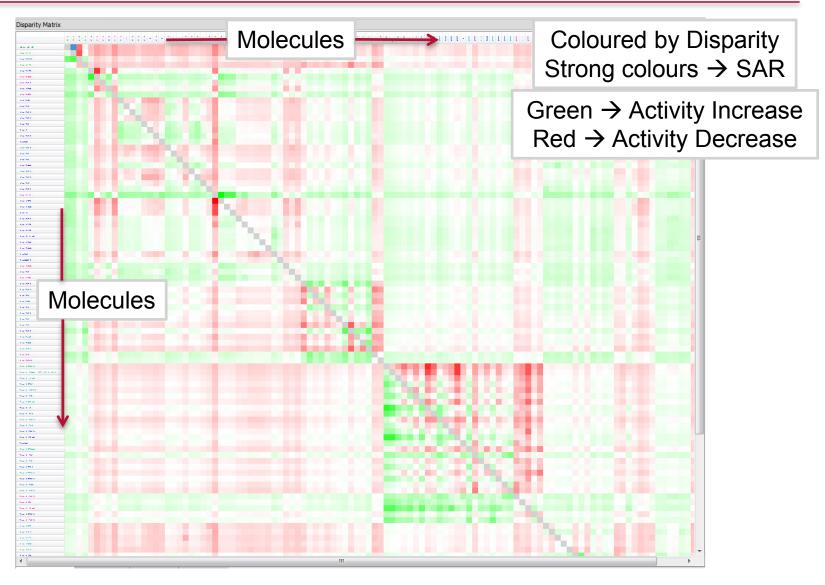


### Top pairs table

Good 6	Good Activity	Bad 2-27	Bad Activity	Disparity	Similarity	Fav	∆ Activity	ΔLE	ΔLLE	Δ TPSA	∆ SlogP	2D Sim
BA <sub>EO</sub> S	6.84	Of Contraction	4.39	-49	0.951	*	-2.45	-0.094	-0.094	0	0	0.776
in Altonomical Alt	8.7	3-17	6.52	-41.2	0.947	*	-2.18	-0.067	-0.07	3.2	-0.1	0.633
5 MD~ <sup>8</sup> KD	6.39	2-34	4.42	-39.4	0.959	*	-1.97	-0.055	-0.026	0	-1.2	0.77
iep	8.7	3-15a	6.32	-38.7	0.939	*	-2.38	-0.074	-0.057	0	-0.7	0.701
I3e_E2020	8.24	3-15a	6.32	-38.4	0.952	*	-1.92	-0.069	-0.056	0	-0.3	0.835
	6.77	2-27	4.39	-34.5	0.931	*	-2.38	-0.102	-0, 108	0	0.3	0.791
13e_E2020		3-17										
or Whee	8.24	STULLER	6.52	-34.4	0.963	${\leftrightarrow}$	-1.72	-0.061	-0.07	3.2	0.2	0.762



## **Disparity matrix**



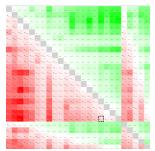


## Visualization

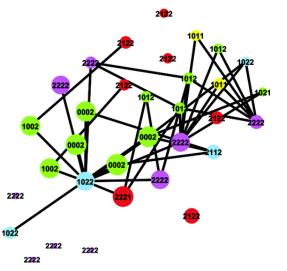
- > Existing ways to visualize
  - > Table & Matrix views
  - > Graph view (Guha/van Drie 2008)

> Activity landscapes (Bajorath)



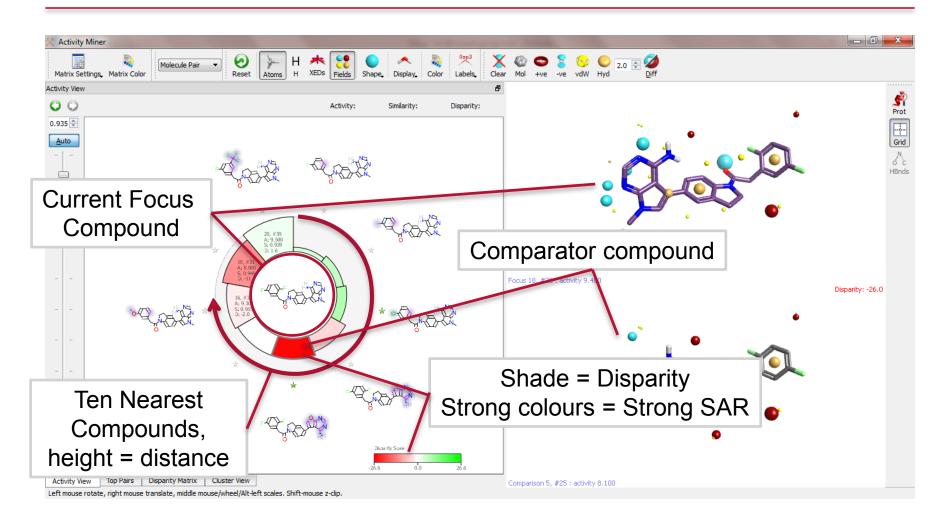


4969999



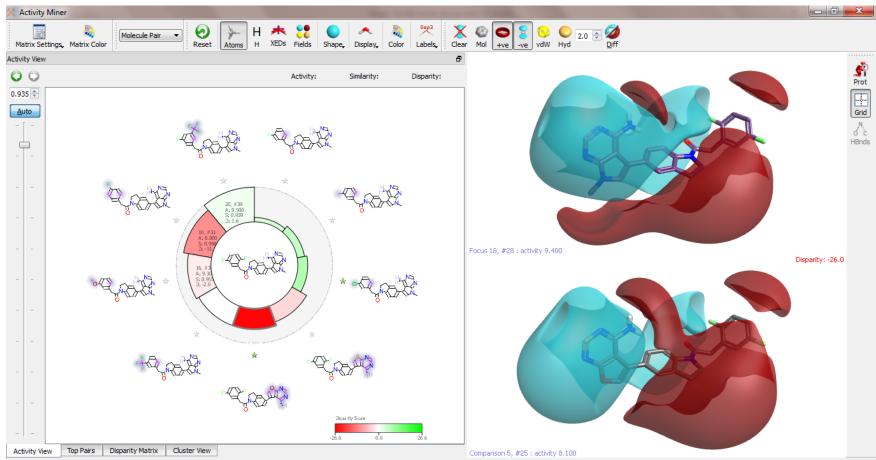
22 29 35 33

## Activity View





#### Electrostatic comparison

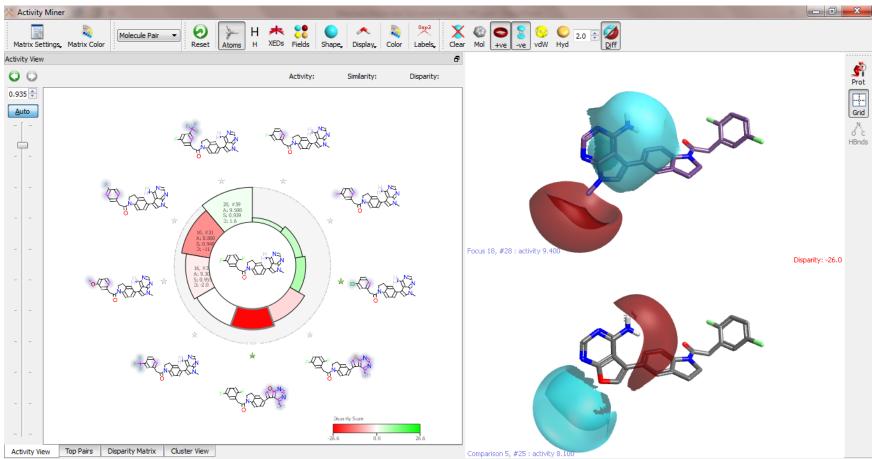


Left mouse rotate, right mouse translate, middle mouse/wheel/Alt-left scales. Shift-mouse z-clip.



#### Electrostatic comparison

Difference plot – Regions where each molecule has stronger electrostatics



Left mouse rotate, right mouse translate, middle mouse/wheel/Alt-left scales. Shift-mouse z-clip.



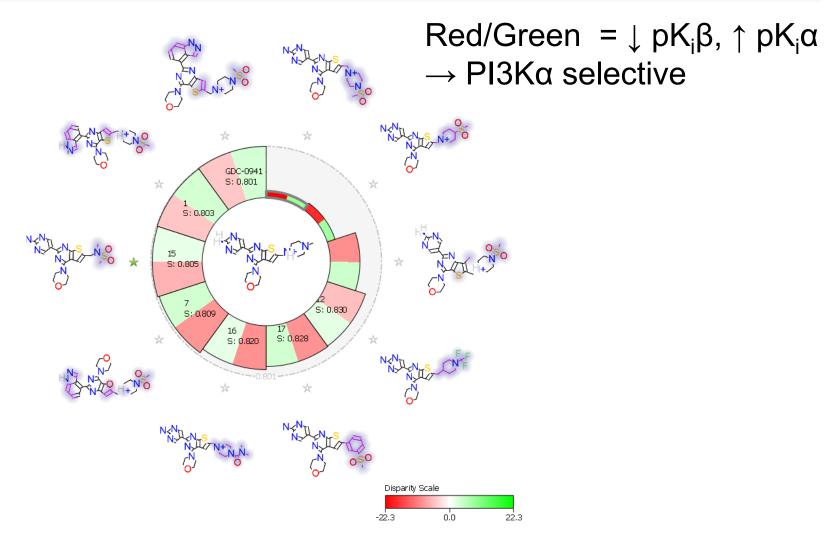
- > Selectivity often as important as potency
- > Look at what structural changes caused large changes in selectivity
- > Use Selectivity Endpoint as Activity?

$$\kappa \approx \frac{\Delta Selectivity}{\Delta Structure} = \frac{\left(\frac{Activity_{\beta}}{Activity_{a}}\right)_{A} - \left(\frac{Activity_{\beta}}{Activity_{a}}\right)_{B}}{(1 - Similarity)}$$

- > What about 3 activities?
- > How would we visualize that?

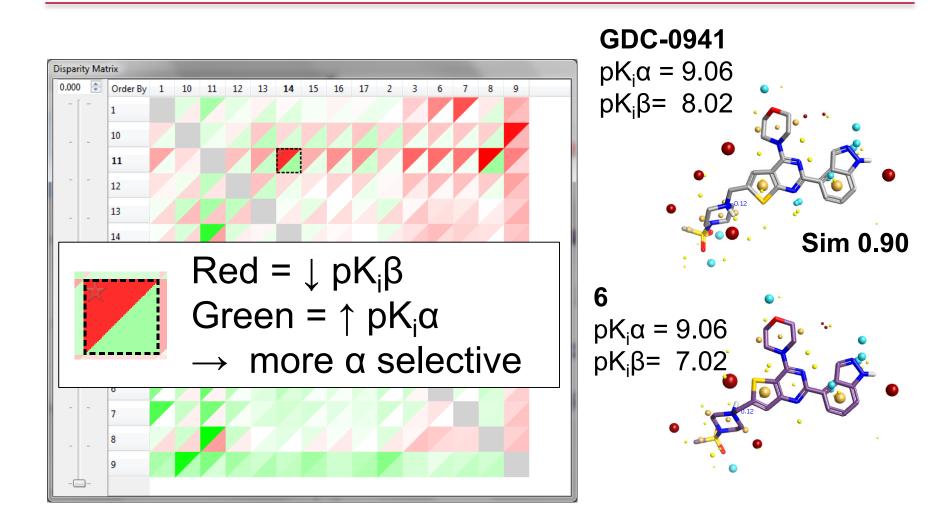


### Activity View – 2 activities





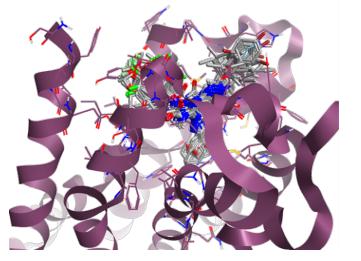
#### Selectivity matrices – 2 activities

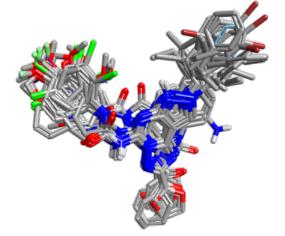




## Application to Adenosine Receptor Antagonists

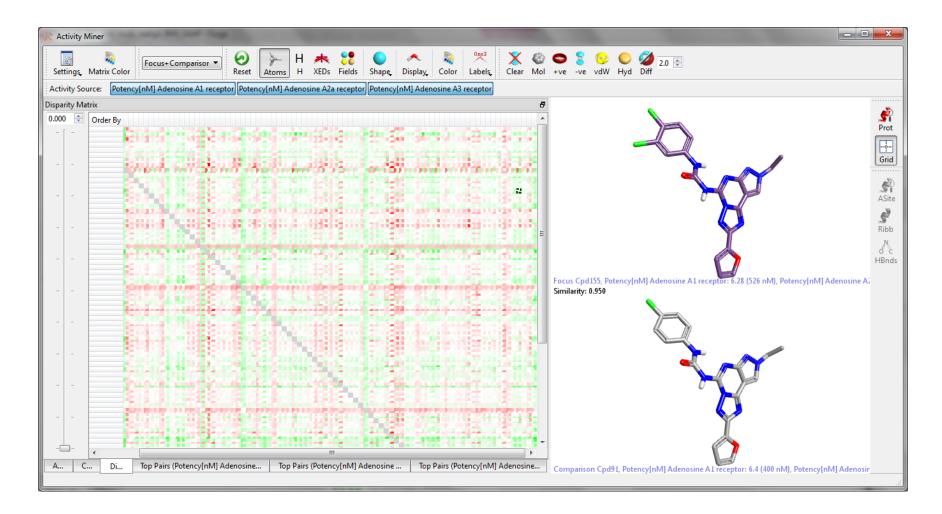
- > Data set from Bajorath J. Chem. Inf. Model 51 258-266 2011
- > 3 Activities A1, A2a, A3 receptors
- > Ligands aligned to x-ray structures 3PWH, 3EML
- > 89 cmpd sub-set with high 3D similarity (>0.7)





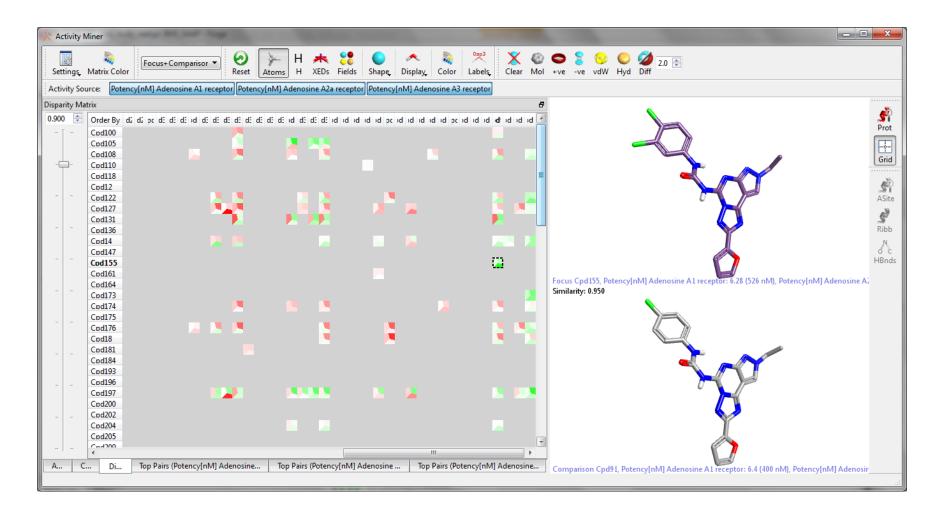


## Disparity Matrix – 11,748 data points



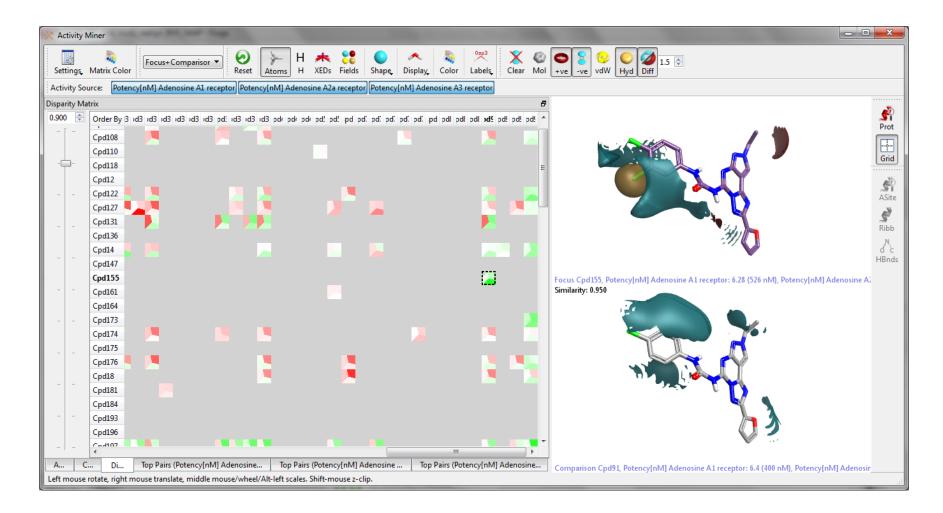


### Disparity Matrix – focus on highly similar pairs





## Why?



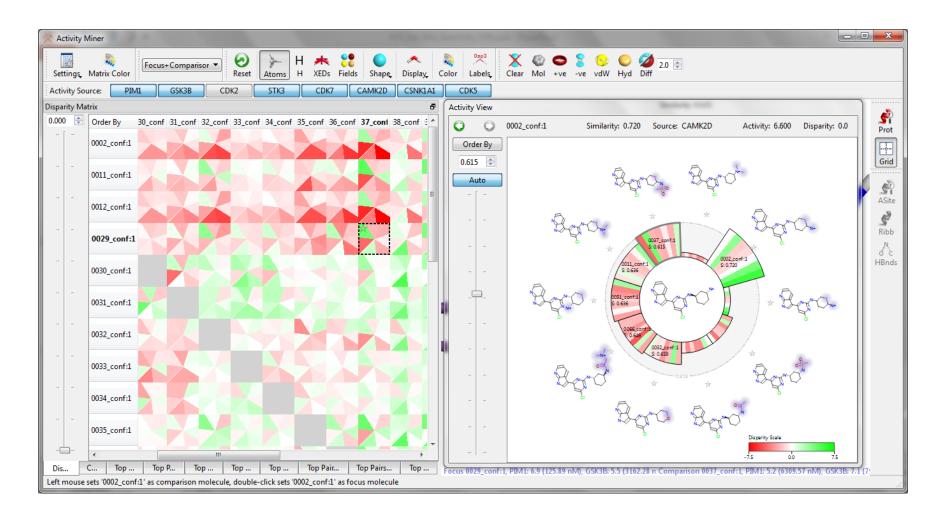


## Limitations

- > 2 Activities work well
- > 3 is OK
- > 7 is too many!



## Limitations





## Conclusions

- > Activity Cliff/Disparity analysis provides quick insights into SAR
  - > Focus on understanding the reason for a cliff
  - > Drive design decisions

#### > Multiple ways to navigate the data

- > Compound focus
- > Most significant changes
- > Global overview
- > Cluster analysis

#### > 2D and 3D both useful

- > 2D provides insights into conformational changes
- > 3D provides insights into electrostatic effects
- > Visualizing multiple activities simultaneously allows selectivity analysis
  - > Large amounts of data difficult to visualize



## Acknowledgements

- > Mark Mackey
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# Thank you!

#### **Questions Welcomed**



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