



smarter chemistry | smarter decisions™

## Upcoming Features and New Horizons

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# What's new at Cresset



## > New features

- > XED FF3
- > Activity cliff and disparity analysis
- > GPU acceleration

## > New Horizons

- > Protein similarity
- > Off-target effects



# XED FF3

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# XED FF3



- > Evolution, not revolution
- > Huckel-based parameter scaling (since 1994!)
- > Nitrogens
  - > No separate types for Ntri and Nsp3
  - > Parameters are scaled based on calculated bond orders
- > Halogens
  - > Improved fields around Cl, Br, I
- > Additional elements
  - > Limited support for B, Si

# Disparity Analysis

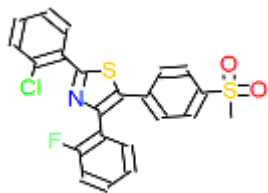
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# Disparity Analysis

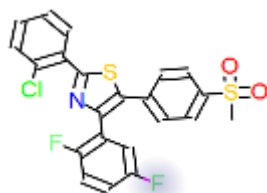
## > Useful tool to investigate SAR

> Which structural changes have had the biggest effect on activity?

> Disparity =  $(\Delta\text{activity}) / (1 - \text{similarity})$



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High similarity

High  $\Delta\text{activity}$

=> High disparity

# Disparity Analysis

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- > Useful tool to investigate SAR
  - > Which structural changes have had the biggest effect on activity?
  - > Disparity =  $(\Delta\text{activity}) / (1 - \text{similarity})$
- > Widely used technique
  - > Reinvented more recently as “activity cliffs”
- > Always been looked at from a 2D point of view
  - > MCS or 2D similarity

# 3D Disparity Analysis

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- > Issue – what is the 3D similarity of a pair of molecules?
  - > Context-specific – aligned to common reference
- > Calculate similarity matrix in field (or field+shape) space
  - >  $O(N^2)$ , but fast enough to be usable on 100's of mols
- > Demonstration this afternoon



# Disparity analysis

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- > Appearing in Forge and Torch in Q3
- > Still under development
- > More features to come

# GPU acceleration

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# GPU processing

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- > Graphics cards are now extremely powerful (but limited) parallel processors
- > Harness this to accelerate field comparisons
- > Using OpenCL
  - > Industry standard
  - > Multi-vendor
  - > Available on CPU as well

# Complicated algorithms...



$$S = \alpha \sum_{fields} w_f \frac{w_{A \rightarrow B} \sum_{fp_A} (\sqrt{scale(size(fp_A)) \times F_B(position(fp_A))} - constraint(fp_A)) + w_{B \rightarrow A} \sum_{fp_B} (\sqrt{scale(size)(fp_B) \times F_A(position(fp_B))} - constraint(fp_B))}{\sum_{fp_A} \sqrt{scale(size(fp_A)) \times F_A(position(fp_A))} + \sum_{fp_B} \sqrt{scale(size)(fp_B) \times F_B(position(fp_B))}} + (1 - \alpha) \frac{2V_{AB} - \tau \times constraint(AB)}{V_{AA} + V_{BB}}$$

(and I've left some bits out)

# blazeV10 GPU benchmark

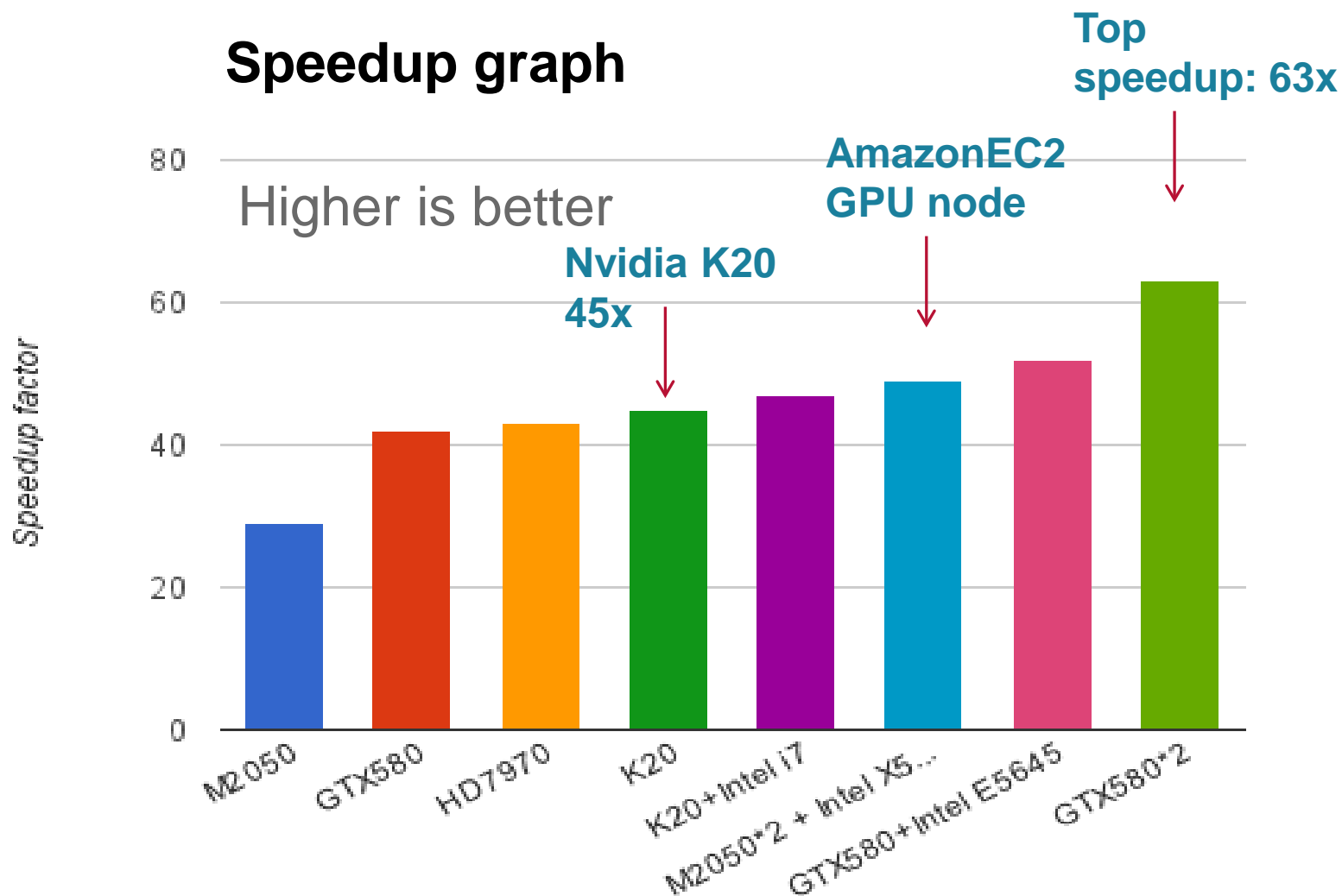
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- > ~1000 molecules – 80k conformations
- > Standard instance: 12 conformations processed per second on a single core of Intel® Core i7-3770 CPU @ 3.40GHz
- > CPUs:
  - > Intel® Core i7-3770 CPU @ 3.40GHz (4 cores - 4 threads)
  - > Intel® Xeon CPU X5570 @ 2.93GHz (4 cores – 8 threads)
  - > Intel® Xeon CPU E5645 @ 2.40GHz (6 cores - 12 threads)
- > GPGPUs:
  - > NVIDIA GTX580
  - > AMD HD7970
- > HPC GPUs:
  - > NVIDIA M2050
  - > NVIDIA K20
- > CPUs and GPUs will work together, it's a heterogeneous world!

# blazeV10 GPU benchmark

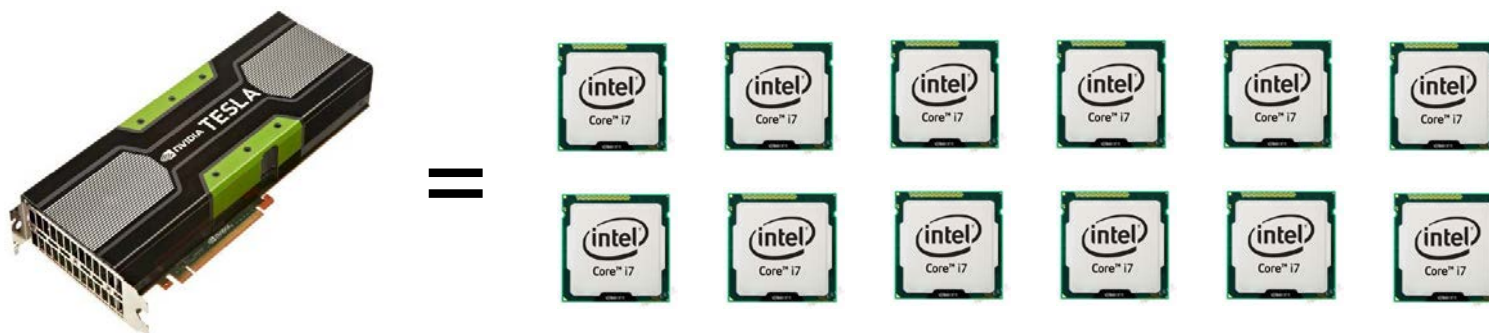


## Speedup graph



# Saving you space and money

- > **Less hardware:** blazeV10 GPU is ~45 times faster on a K20 than blazeV10 on a single core of Intel i7 CPU!



- > **It's cheaper:** for a \$2.10/hour GPU instance on AmazonEC2 you can process 2m conformations, whereas you can only process 1.3m conformations with 14 dual-core High-CPU Medium instances.
- > **It's greener:** one GPU consumes ~400watts, one quad-core CPU workstation consumes ~200watts: we achieve 5x performance per watt = 5x less gCO2 consumed per answer.



# Plus new science!

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- > Diversity analysis in 3D with fields
  - >  $O(N^2)$  calculation – not currently cost-effective on large data sets
- > Screening virtual databases
- > Flexible overlays



# Protein Similarity

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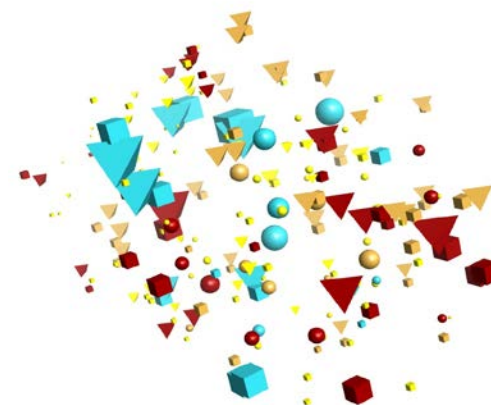
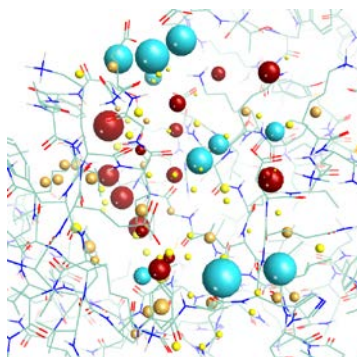
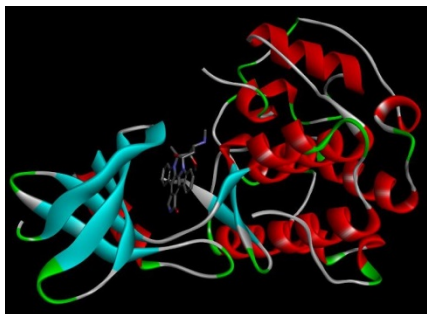
# Fields on Proteins

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- > Can we do (sensible) analysis of proteins in field space?
  - > Issues of charge, solvation, calculation time, etc
- > Pilot project: active site similarity metrics
  - > How similar are two proteins?
  - > Which proteins should you be routinely counter-screening against?

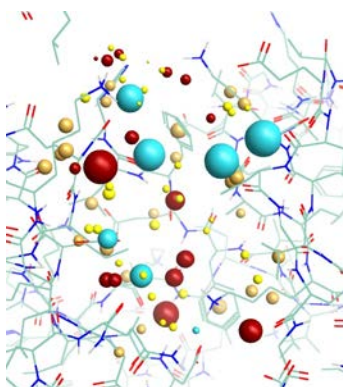
# Protein Fields

Reference Protein 1AQ1

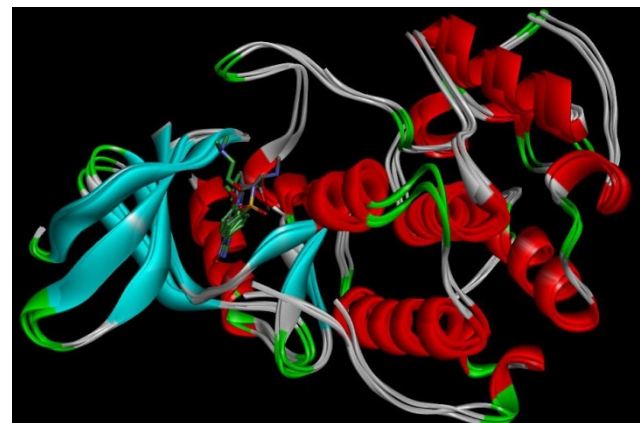


Generate Protein Fields

Target Proteins 1JSV & 1OIR



Align Field Points



# Active site similarity



Can we align the active sites of proteins bound to the same ligand?



Can we align the active sites of the same protein bound to different ligands?



Can we distinguish functionally-related proteins from the rest?



Can we differentiate proteins involved in known OTEs from the mass?



# Off-target effects

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# Can we predict OTEs via field similarity?



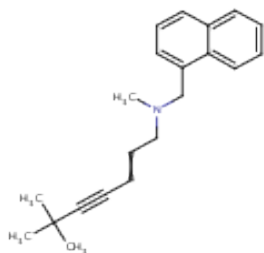
- > Several methods looking at this in 2D (eg SEA)
- > Field similarity might find ones that 2D similarity doesn't

# Initial study

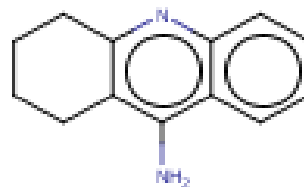
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- > Compare new ligand to ligands from PDB
  - > Inverse Blaze
- > Raw scores not useful
  - > Phe gets a decent score to everything
- > Convert to Z-scores

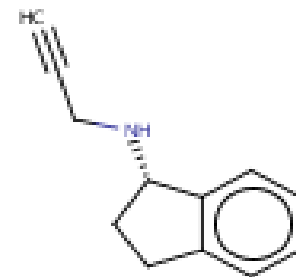
# Initial results



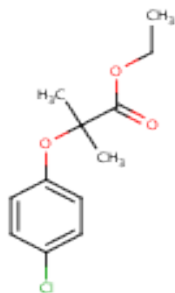
Terbinafine  
OTE: MAO & AChE



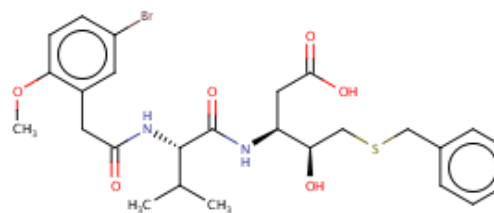
Imipramine  
AChE



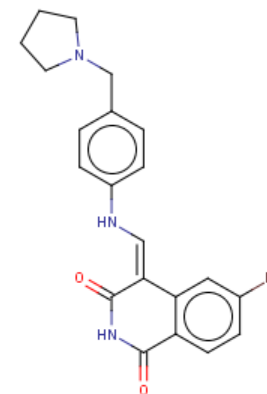
MAO B



Clofibrate  
Caspase-1 & tyrosine kinses



Caspases



IGF-1R



# OTEs



- > Currently extending the method to conformer populations
  - > More computationally expensive
  - > Harder

# What next?

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# Coming soon...



- > More protein fields
  - > Protein-protein comparisons
  - > Protein-ligand comparisons
  - > Using protein info to guide ligand alignments
- > Qualitative models
- > Extend off-target models
- > ADMET models

