

Automated SPR Data Processing and Sensogram Classification

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Abstract



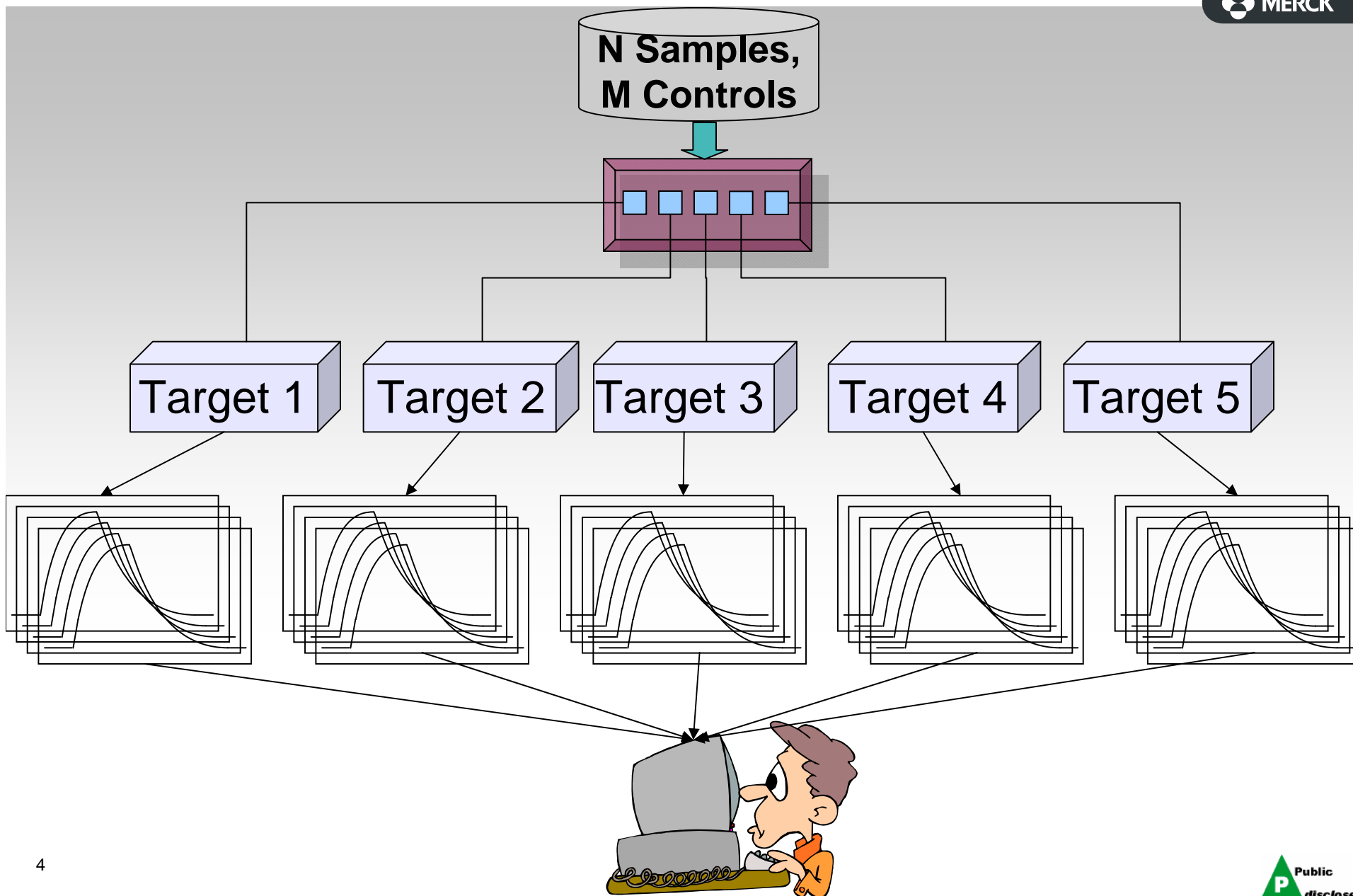
Fragment-based approaches have added to the arsenal of tools used to identify novel small molecule leads with high ligand efficiencies. A variety of label-free technologies have been developed and implemented throughout the industry for fragment screening. Using surface plasmon resonance (SPR) as a fragment screening platform is in its infancy. The miniaturization and automation of this technology has led to the associated problem of dealing with the large volume of raw data generated. The resources needed for the analysis, integration and prioritization of compounds screened makes it challenging to take the results of an SPR screen into the workflow of project teams engaged in the discovery process in a timely fashion. As such, several sets of equations were derived and implemented on Merck's intranet to score single sensograms to distinguish stable binders from weak or anomalous binders. This set of equations was optimized and validated on simulated data to both capture "fragment-like" behavior from SPR experiments and to filter out much of the anomalous behavior commonly observed. This talk will cover the creation of the theoretical data sets, the resultant equations, and the performance metrics used to evaluate them.

Discovery Project Needs Drove Collaboration



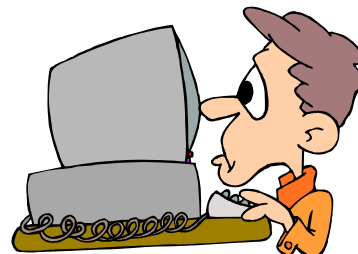
- Fragment Screening in the Discovery Process
 - Lead ID – time is critical!
 - Need rapid turnaround of results
 - Assay development takes time for novel targets
 - Biacore A100 assay takes little time (N fragments + M controls on 5 spots)
 - Visual analysis of each of the $5 \times (N + M)$ curves
 - ~1 week (Bottleneck)
- Selection criteria for follow up
 - Response relative to theoretical R_{\max}
 - Appropriate curve shape (box) for fragment-like kinetics
 - Selectivity over other targets

The Problem in a Nutshell

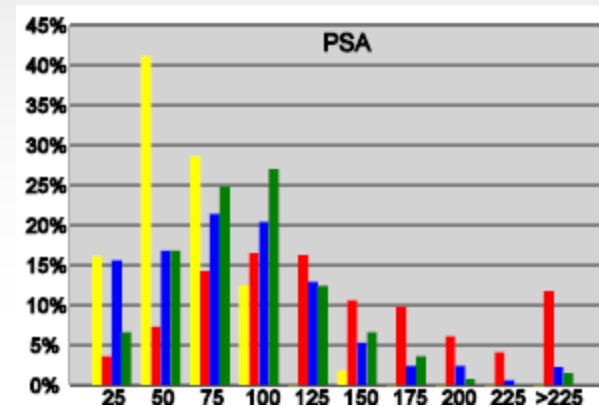
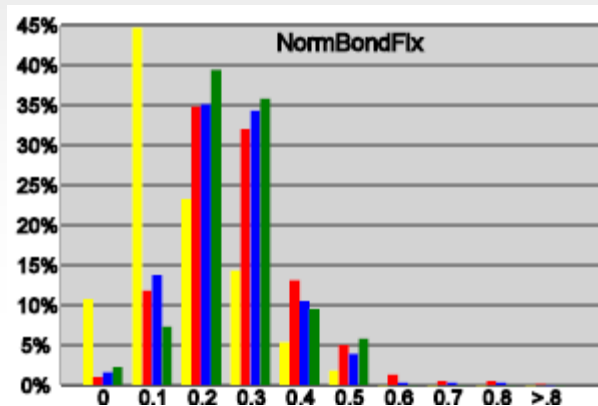
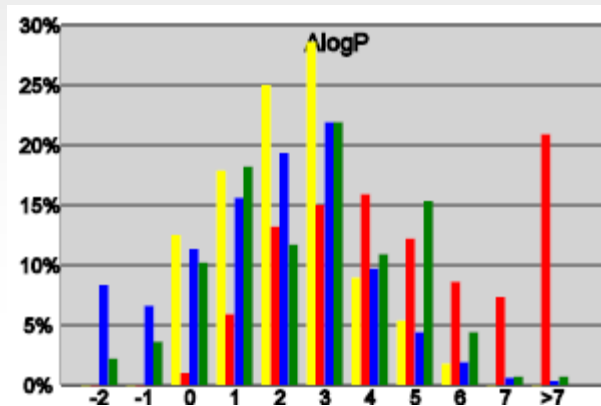
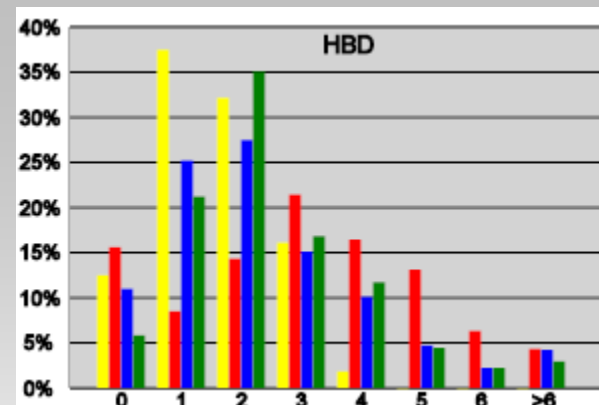
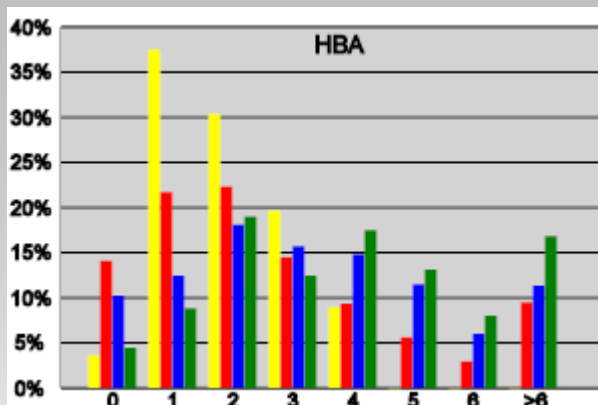
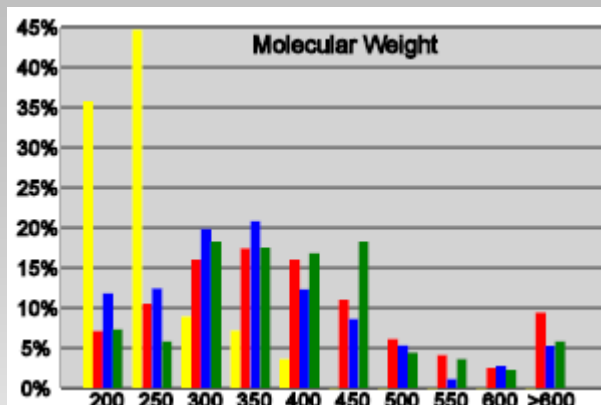


The Problem in a Nutshell

***How can we speed this up to
have an immediate project
impact?***

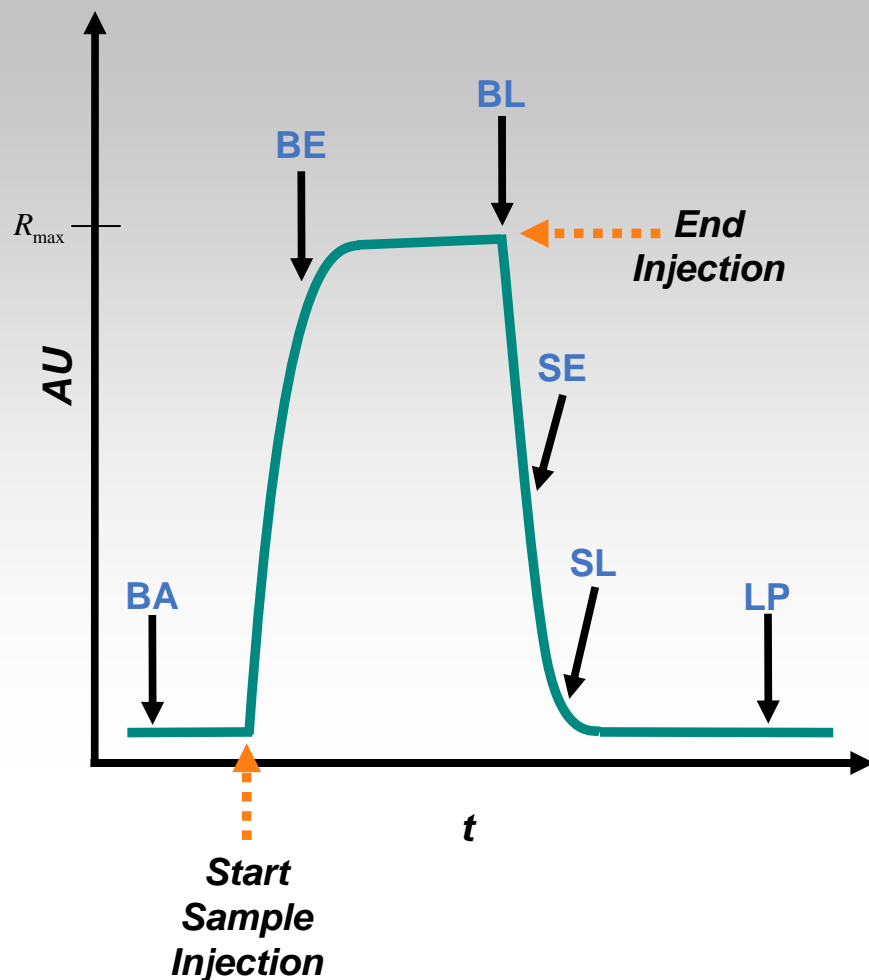


Aside: Summary of Merck Fragment Library Properties



Yellow: Fragments Assayed
Red: All Launched Drugs as of 2006 (MDDR)
Blue: Orally Delivered Drugs
Green: Top 200 Small Molecule Drugs of 2008

Another Aside: A Single Sensogram



Single-point experiment

- *Not* titrated
- > 1500 compounds/screen

$$R_{\max} = \frac{n \cdot R_L \cdot MW_{\text{analyte}}}{MW_{\text{ligand}}}$$

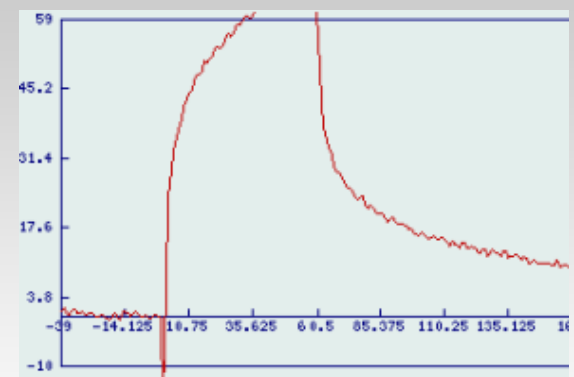
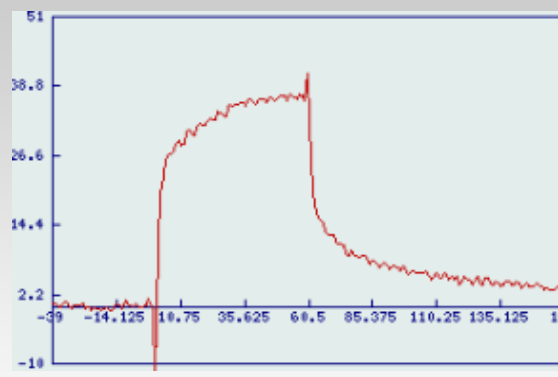
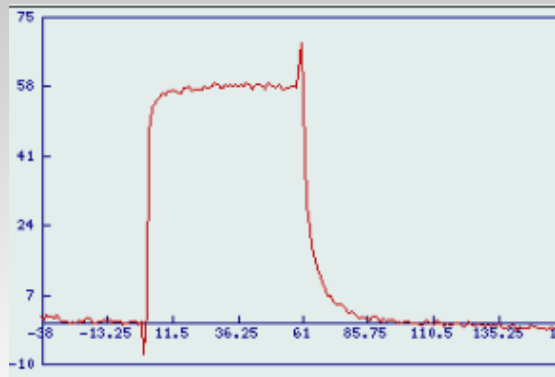
Reporter Points

- BA: baseline (15 sec window)
- BE: binding early (10 sec)
- BL: binding late (10 sec)
- SE: stability early (10 sec)
- SL: stability late (10 sec)
- LP: late points (10 sec)

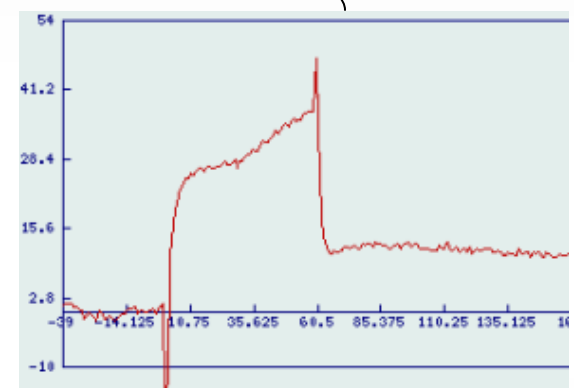
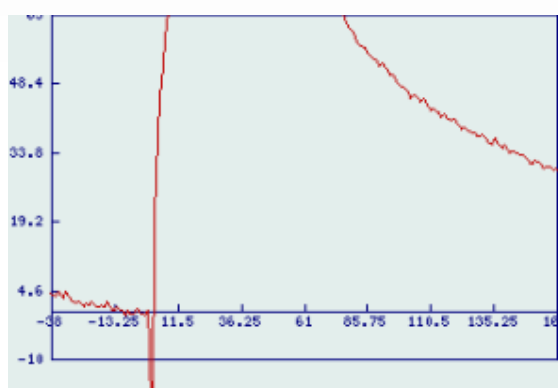
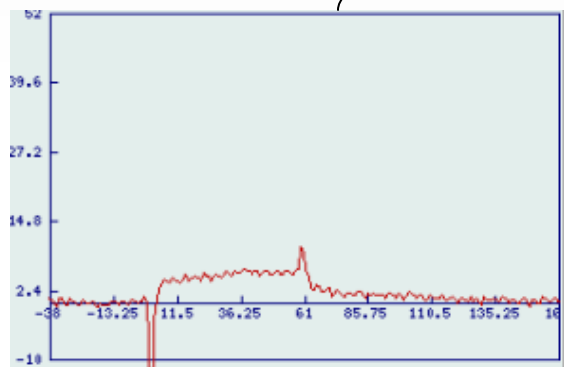
Theory vs. Experiment

Many curve types, need to account for them all:

The Good



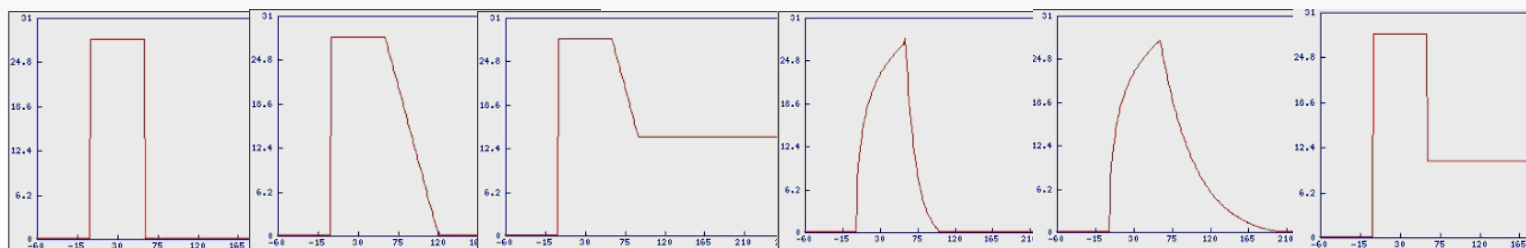
The Bad and Ugly



Theoretical Data Sets

Used to mimic experimental data and benchmark ranking equations

| <i>Data Set</i> | <i>IS1</i> | <i>IS2</i> | <i>IS3</i> | <i>IS4</i> | <i>IS5</i> | <i>IS6</i> |
|------------------------|---------------------------------------------------|-----------------------------------|---------------------------------------------------|-----------------------------------------------------|-----------------------------------------------|------------------------------------------------------------|
| Design Criteria | <i>Fast-on, fast-off kinetics (fragment-like)</i> | <i>Fast-on, slow-off kinetics</i> | <i>Fast-on, slow-off; incomplete dissociation</i> | <i>Slow-on, moderately fast-off (high affinity)</i> | <i>Slow-on, slow-off kinetics (drug-like)</i> | <i>Fast-on, fast-off kinetics; incomplete dissociation</i> |



24 total “training” sets:

- Curves modulated from 100%, 75%, 50% and 25% of R_{max}
- Two modulated to 150% of R_{max}

How to Evaluate the Resultant Rankings?

Mathematically:

AUC = Area Under Accumulation Curve

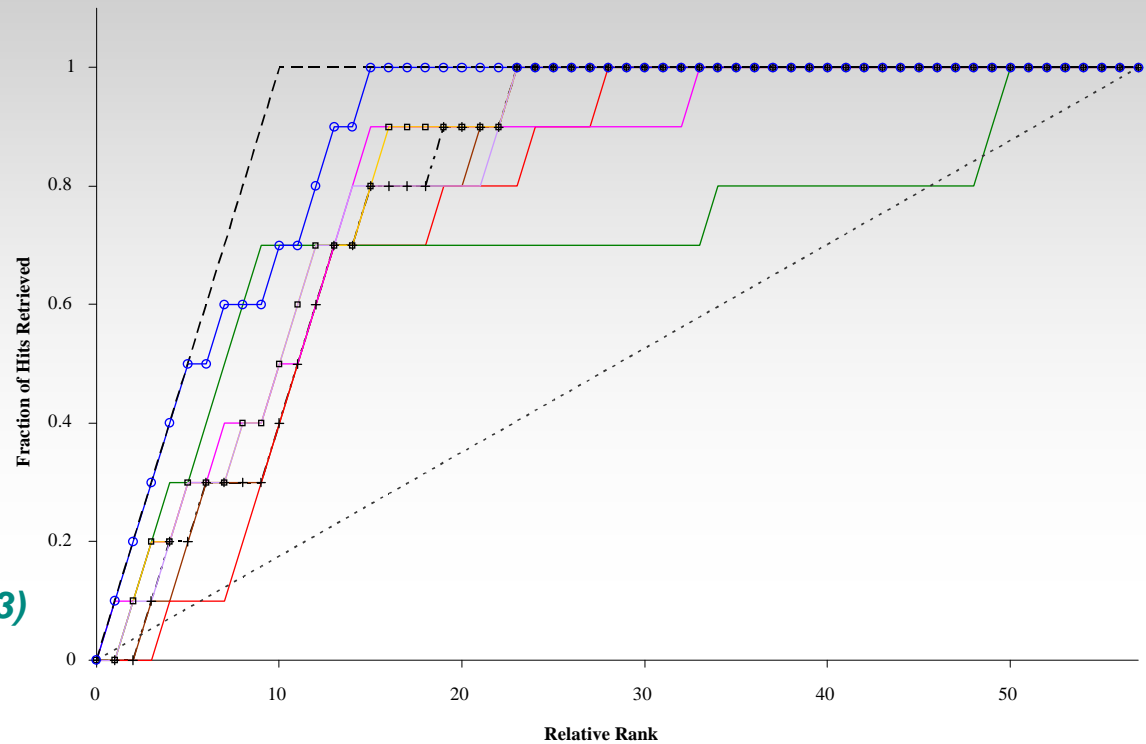
$$ROC = \frac{2AUC - R_a}{2R_i} \quad (1)$$

$$RIE = \frac{\sum_{i=1}^{hit} e^{-\alpha(r_i/N)}}{R_a \left(\frac{1 - e^{-\alpha}}{e^{\alpha/N} - 1} \right)} \quad (2)$$

$$BEDROC = \frac{RIE - \left(\frac{1 - e^{-aR_a}}{R_a(1 - e^{-\alpha})} \right)}{\left(\frac{1 - e^{-aR_a}}{R_a(1 - e^{-\alpha})} \right) - \left(\frac{1 - e^{-aR_a}}{R_a(1 - e^{-\alpha})} \right)} \quad (3)$$

Graphically:

Accumulation Curve



Possible Ranking Metrics

Simple, widely used:

$$S = \frac{B_L - B_A}{R_{\max}} \quad (4)$$

$$S_{Occ} = 100 \cdot \frac{B_L}{R_{\max}} \quad (5)$$

Penalize slow on-rate:

$$S = \frac{2B_L - (B_E + B_A)}{R_{\max}} \quad (6)$$

Penalize slow off-rate:

$$S = \frac{B_L - S_E}{R_{Max}} \quad (7)$$

$$S = \frac{B_L - (2S_E - S_L)}{R_{\max}} \quad (8)$$

Penalize aggregators:

$$S = \frac{B_L - L_P}{R_{Max}} \quad (9)$$

$$S = \frac{B_L - (2S_L - L_P)}{R_{\max}} \quad (10)$$

or combine them...

Combined Ranking Metric (aka Eq. 17)



Fast On &
Occupancy



Fast Off



Drug-Like



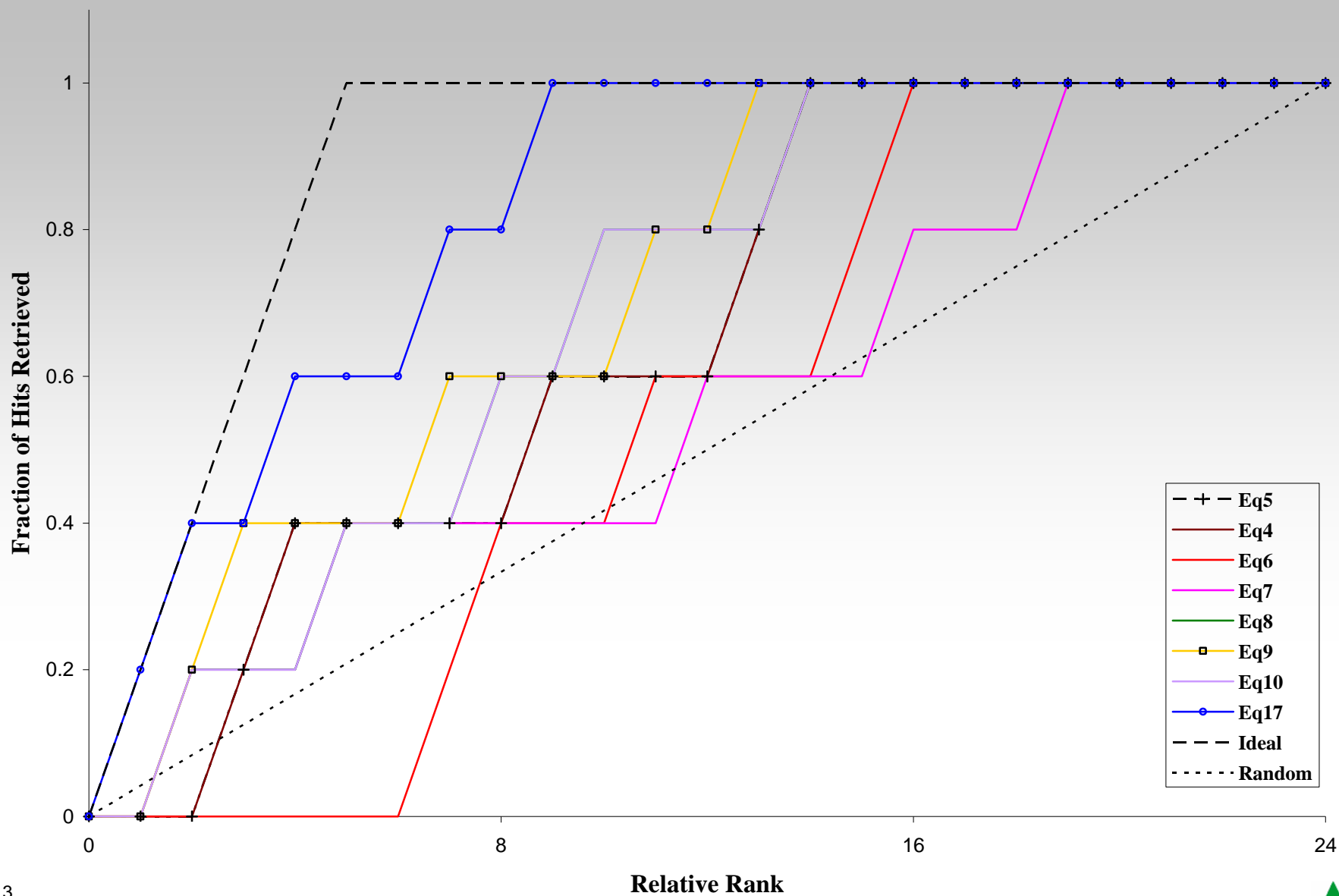
Irreversible



$$S = \frac{w_1 \left\{ 1 - \left| 1 - \frac{\| (BE, BL) - BA \|}{R_{\max}} \right| \right\} + w_2 \left\{ \left[\frac{\| BL - SE \|}{\| BL - BA \|} \right]^3 \right\} + w_3 \left\{ \left[\frac{\| BL - SL \|}{\| BL - BA \|} \right]^3 \right\} + w_4 \left\{ \left[1 - \frac{\| LP - BA \|}{R_{\max}} \right]^3 \right\}}{\sum_{i=1}^4 w_i}$$

Weights, w_i , allow for a single equation to be used for fragment and small molecule ranking.

Performance on Theoretical Data Set

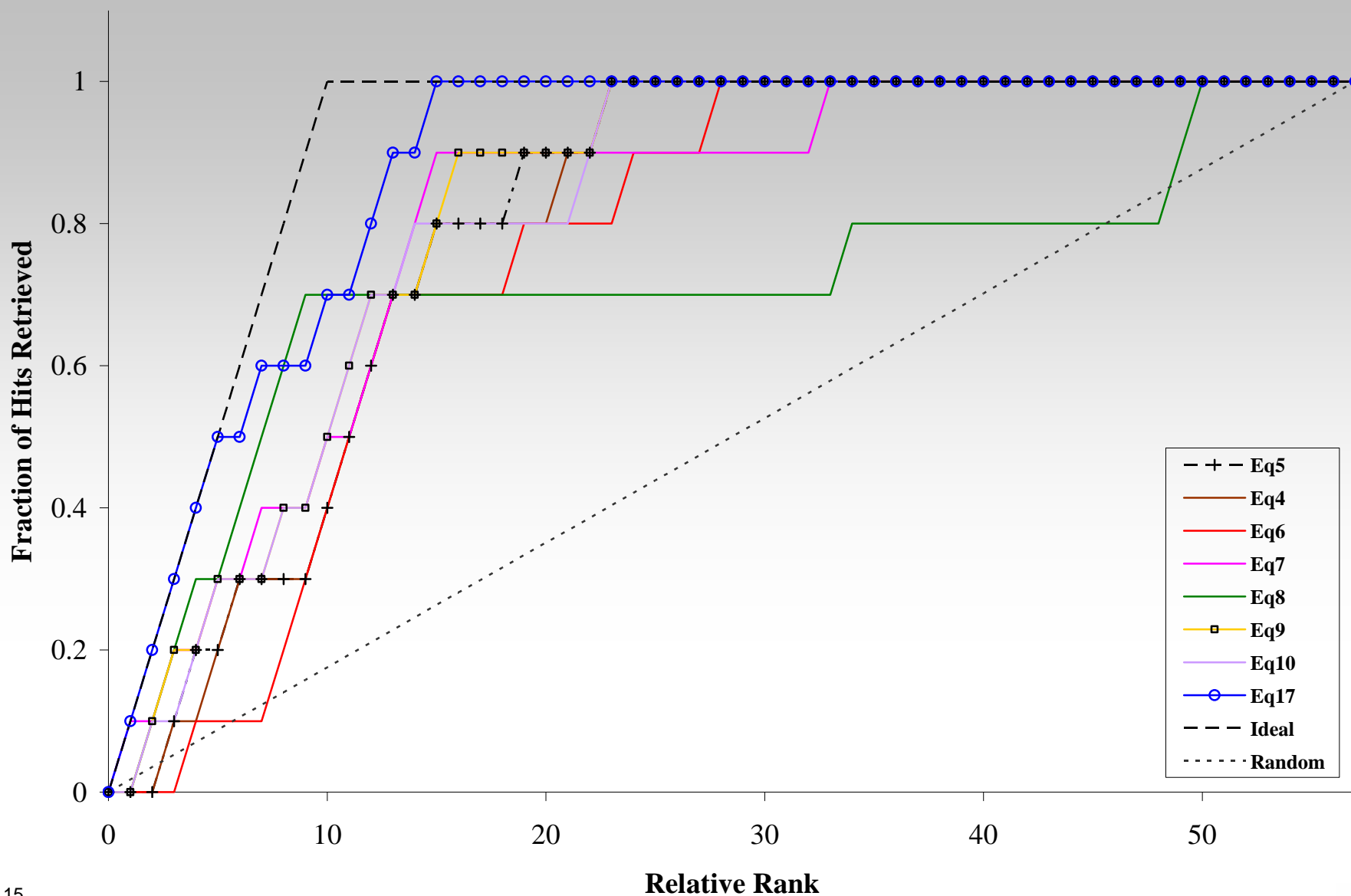


Performance on Theoretical Data Set



| | <i>AUC</i> | <i>ROC</i> | <i>RIE</i> | <i>BEDROC</i> |
|------------------------|------------|------------|------------|---------------|
| <i>Eq. (4)</i> | 0.65 | 0.71 | 2.34 | 0.49 |
| <i>Eq. (5)</i> | 0.65 | 0.71 | 2.34 | 0.49 |
| <i>Eq. (6)</i> | 0.53 | 0.56 | 1.15 | 0.24 |
| <i>Eq. (7)</i> | 0.55 | 0.59 | 1.93 | 0.41 |
| <i>Eq. (8)</i> | 0.68 | 0.76 | 2.36 | 0.50 |
| <i>Eq. (9)</i> | 0.71 | 0.79 | 2.76 | 0.58 |
| <i>Eq. (10)</i> | 0.68 | 0.76 | 2.36 | 0.50 |
| <i>Eq. (17)</i> | 0.82 | 0.93 | 3.55 | 0.75 |
| <i>Ideal</i> | 0.89 | 1.02 | 4.48 | 0.95 |
| <i>Random</i> | 0.49 | 0.52 | 1.76 | 0.37 |

Performance on a Fragment Screen



Performance on a Fragment Screen



| | <i>AUC</i> | <i>ROC</i> | <i>RIE</i> | <i>BEDROC</i> |
|------------------------|------------|------------|------------|---------------|
| <i>Eq. (4)</i> | 0.80 | 0.87 | 1.91 | 0.35 |
| <i>Eq. (5)</i> | 0.81 | 0.87 | 1.92 | 0.35 |
| <i>Eq. (6)</i> | 0.78 | 0.83 | 1.13 | 0.20 |
| <i>Eq. (7)</i> | 0.81 | 0.88 | 3.31 | 0.60 |
| <i>Eq. (8)</i> | 0.81 | 0.88 | 2.51 | 0.45 |
| <i>Eq. (9)</i> | 0.82 | 0.89 | 2.76 | 0.50 |
| <i>Eq. (10)</i> | 0.81 | 0.88 | 2.51 | 0.45 |
| <i>Eq. (17)</i> | 0.89 | 0.97 | 5.18 | 0.94 |
| <i>Ideal</i> | 0.91 | 1.00 | 5.53 | 1.00 |
| <i>Random</i> | 0.49 | 0.49 | 1.85 | 0.33 |

Deployed via Intranet: Modify on the Fly



myMerck portal: home | test
 MedSys >

Sensogram File Upload

Tab Separated file to upload:

Ligand Mwt File Upload

Comma or tab separated file to upload:

Set Parameters

Analysis Control:

R_max only used when there is no Ligand mwt file given.

Drug-like
 Fragment-like
 User-defined:

W1: W2:
 W3: W4:

R_L(RU):
 R_max(RU):
 mwt Prot(D):

BA start:
 BE start:
 BL start:
 BA stop:
 BE stop:
 BL stop:

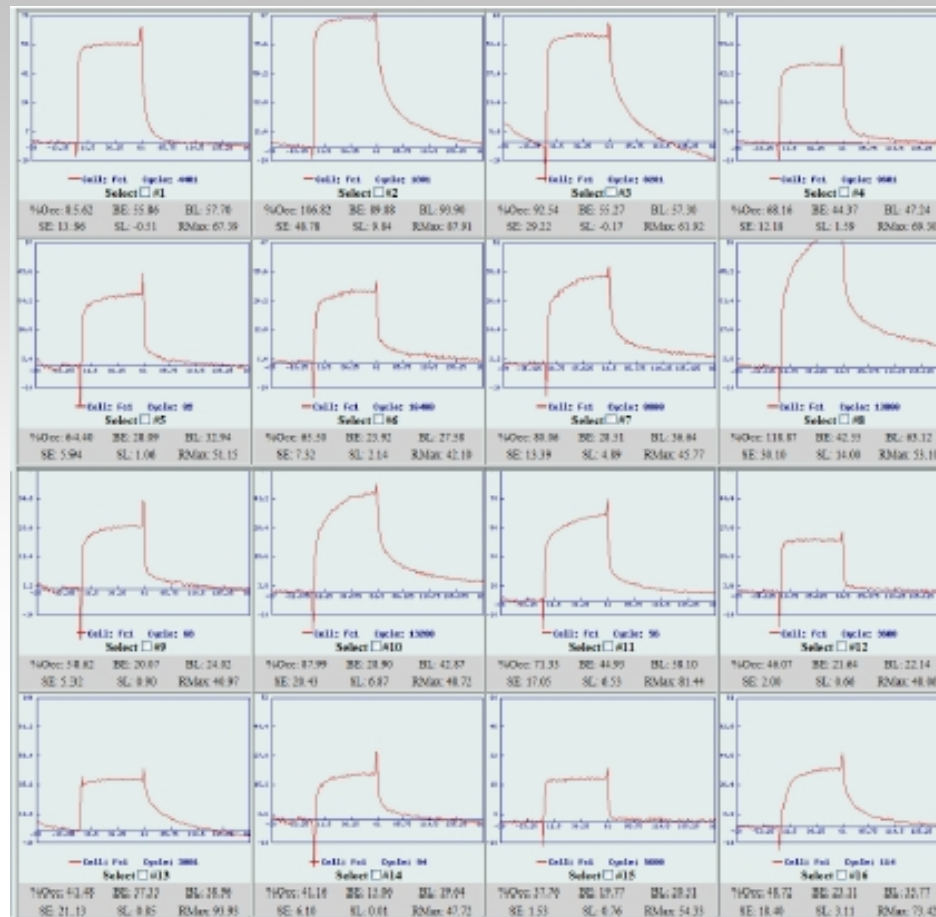
SE start:
 SL start:
 LP start:
 SE stop:
 SL stop:
 LP stop:

Plot Control:

Leave as "Calc" to let the software calculate the values from the input.

Sort by:
 Y Min:
 Y Max:
 X Min:
 X Max:

If nothing happens, press the button again.



Impact

- Shortened SPR screen analysis time > 50x
- Rapid selection of true positives
- Outperforms usual methods of fragment selection
- Can use for fragments or fully elaborated molecules

Next Steps

- Use of machine learning to select best w_i from a series of screens and users' rankings
- Optimization of w_i for drug-like molecules
- Would love to hear how this works in others' hands
- Not meant to replace titrations!

Acknowledgements



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- *Kartik Narayan*
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Thank you!

Reference:

Kreatsoulas & Narayan, *Analytical Biochemistry*, 402 (2010) 179-184.