

# Understanding Compound Quality

## *Focus on Molecular Property Design*

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***A high level view  
Oral small molecules***

Success rates: Preclinical-Phase III **4.3%**; Phase II **23%**

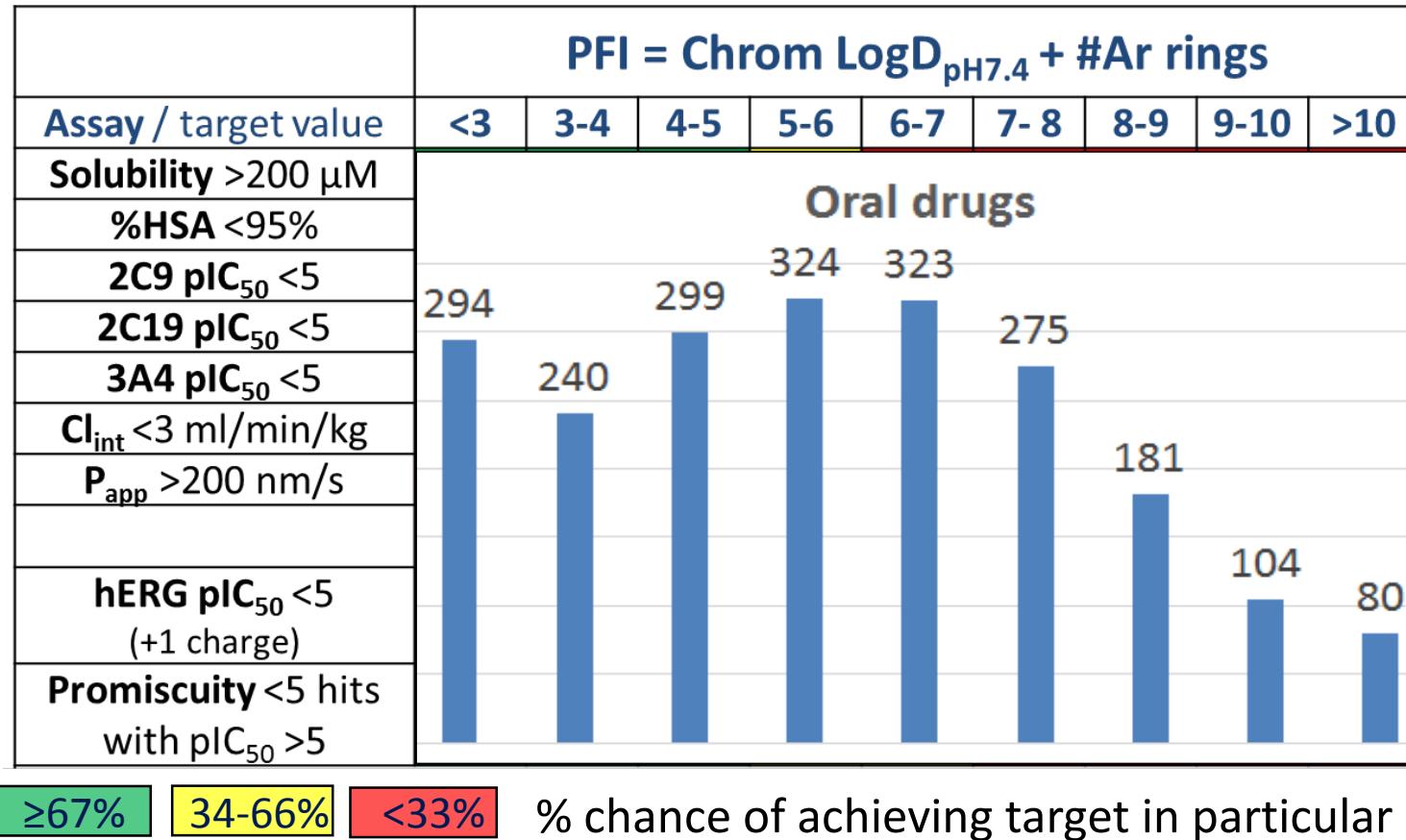
## *Evidence for progression of unoptimised compounds*

- **Pfizer: ‘4 Pillars’ for phase II success** (*44 phase II projects, 2005-9*)
  - **Exposure at target; Binding to target; Pharmacological response; Target linked clinically to disease modification**
  - Low confidence in *exposure* in 18/34 non-progressing molecules:  
**“cannot conclude mechanism tested adequately in 43% of cases”**
- **AstraZeneca: ‘5Rs’** (*>114 preclinical to phase II projects, 2005-10*)
  - **‘Right’: Target & Tissue (4Ps); Safety; Patient; Commercial potential**
  - 29% Clinical efficacy failures **“dose limited by compound characteristics or tissue exposure not established”**
  - **Decision making process:** eg, 38% projects advanced to clinic had *low confidence in safety* & 78% of these eventually failed due to toxicity
- **GSK: solubility-limited candidates** – *BCS II/DCS class IIb*
  - **Add 2 years to development:** “lack of efficacy owing to lack of exposure”
- **FDA submissions** (*302 NMEs, 2000-12; 151 (50%) unsuccessful 1<sup>st</sup> time*)
  - 29% Unsuccessful 1<sup>st</sup> submissions had **dose or clinical end point issues**

Success rates: Thomson Reuters, 2006-10; **4 Pillars**: Morgan et al, *Drug Discovery Today* 2012, **17**, 419; Bunnage, et al *Nat. Chem. Biol.* 2013, **9**, 195; **5Rs**: Cook et al, *Nat. Revs. Drug Disc.* 2014, **13**, 419; **Solubility**: Hann & Keserű, *Nat. Rev. Drug Disc.* 2012, **11**, 355; **FDA**: Sacks et al, *JAMA* 2014, **311**, 378; **Pharma’s problems**: Scannell et al, *Nat. Rev. Drug Discov.* 2012, **11**, 191

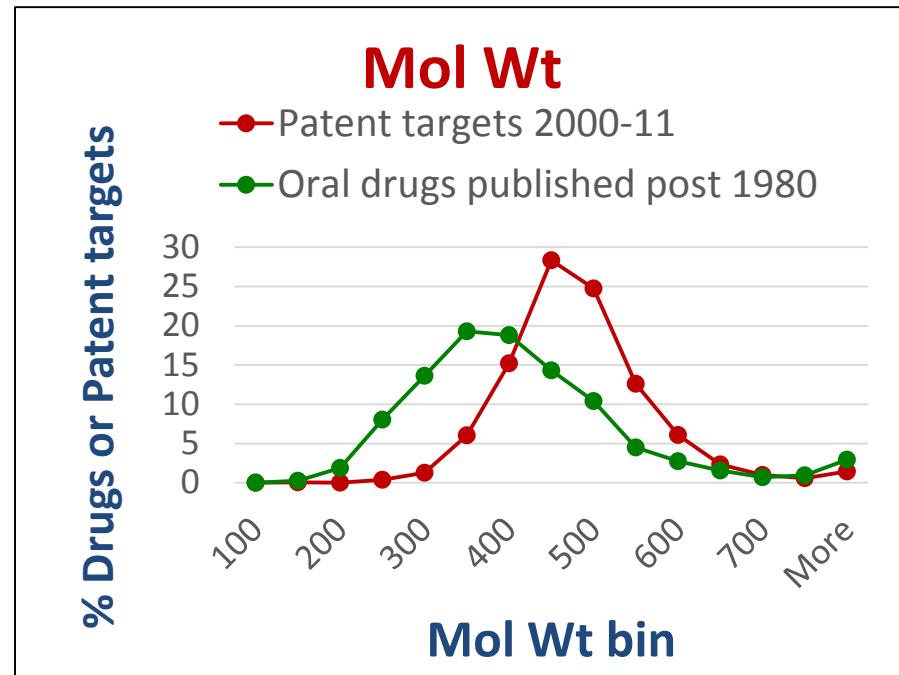
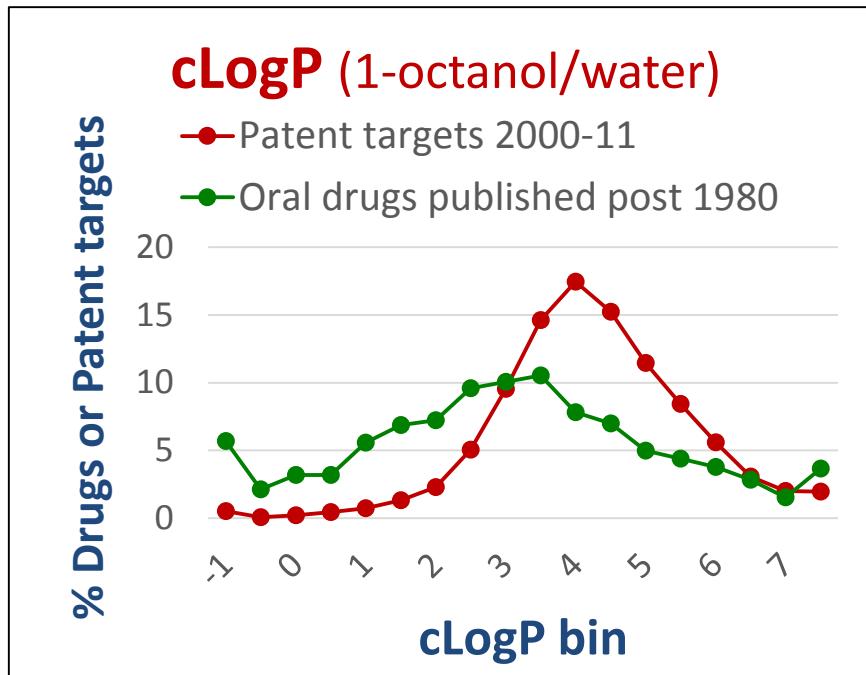
# A Significant Body of Evidence links Physical Properties to *Probability* of ADMET Risk

**Key properties:** lipophilicity + ionisation. *Property forecast index (PFI)*



**PFI:** Young et al, *Drug Disc. Today* 2011, **16**, 822; **Physical property reviews:** Meanwell, *Chem. Res. Toxicol.* 2011, **24**, 1420; Young, *Top Med. Chem.* 2015, **9**, 1; Gleeson et al, in *The Handbook of Medicinal Chemistry: Principles and Practice*, eds A.M. Davis and S. Ward, RSC, 2015, p1-31; Hann & Keserú, *Nat. Rev. Drug Disc.* 2012, **11**, 355; Gleeson et al. *Nat. Rev. Drug Disc.* 2011, **10**, 197; **Lipophilicity:** Waring, *Exp. Op. Drug Disc.* 2010, **5**, 235; **Ionisation:** Charifson & Walters, *J. Med. Chem.* 2014, **57**, 9701; **Ar rings review:** Ritchie & Macdonald, *J. Med. Chem.*, 2014, **57**, 7206; **Critique - statistics:** Kenny & Montanari, *J. Comp.-Aid. Mol. Des.* 2013, **27**, 1; **Critique - toxicity data:** Muthas et al, *MedChemCommun.* 2013, **4**, 1058

# Properties of Patented Compounds & Oral Drugs



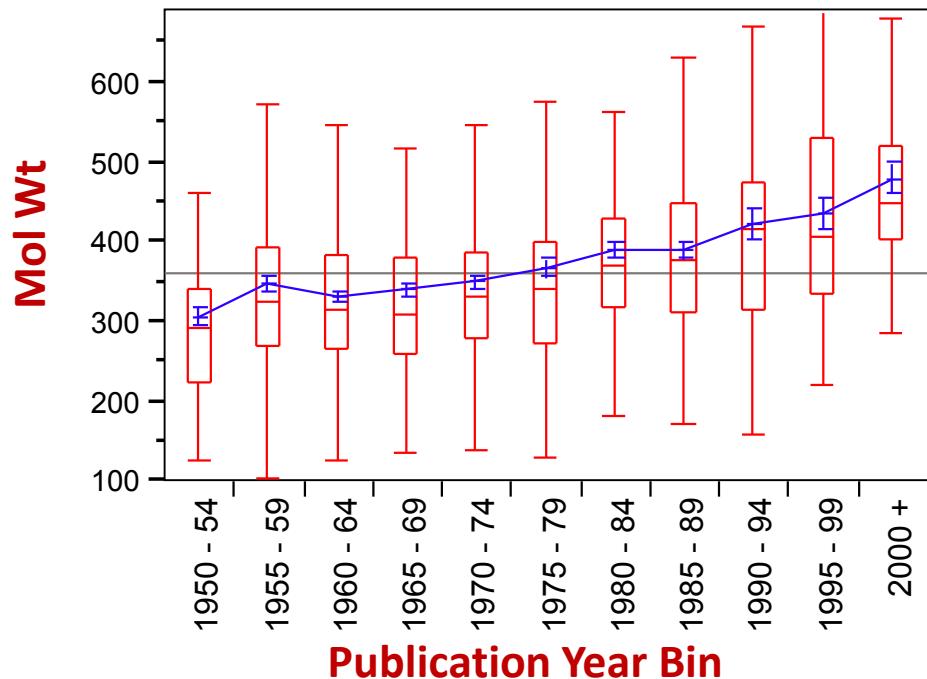
- ‘Inflated’ patented compounds are likely to possess increased ADMET risks vs recently marketed drugs → **pipeline attrition?**
- Will the **probability of success in a portfolio** of drug candidates increase as its balance of biological and physicochemical properties more closely resembles that of successful marketed drugs?
- What other viable strategies exist for medicinal chemists to improve productivity?
- **Compound quality is a medicinal accountability.** Fixed at the point of design, controllable in optimisation, **must not be the root cause of clinical attrition**

Drug data: Leeson et al, *Med. Chem. Comm.* 2011, **2**, 91, updated to 2014

Patent data: Leeson & St-Gallay, *Nature Revs. Drug Disc.* 2011, **10**, 749

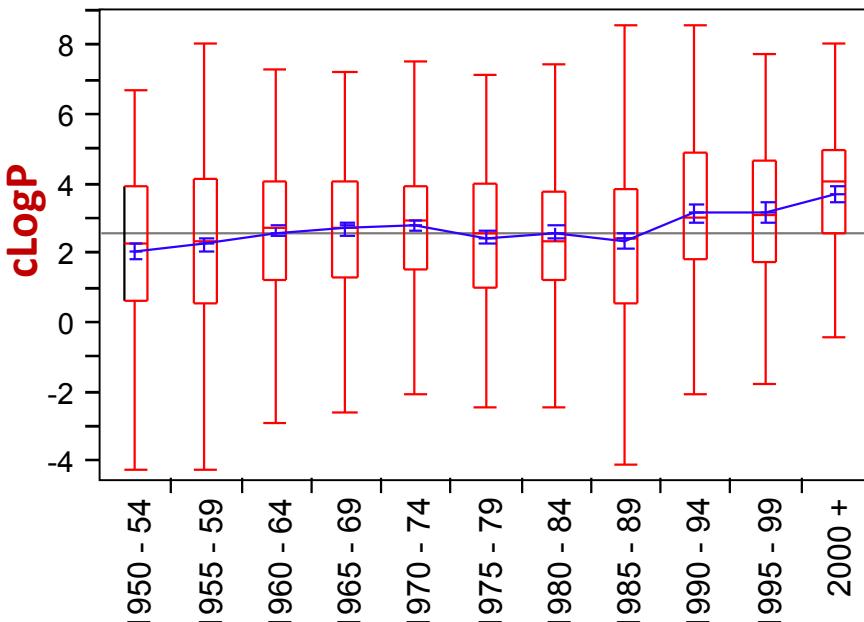
# Oral ‘Druglike’ Properties: Changes over Time

Median    291 324 313 308 331 339 371 376 416 409 451  
 n            144 223 302 236 217 164 141 107 78 53 85



*Increasing significantly ~10-20 years*

2.30 2.34 2.73 2.74 2.96 2.59 2.37 2.46 3.01 3.15 4.07  
 144 223 302 236 217 164 141 107 78 53 85



*No change until 2000 +*

- **Least change:** cLogP, HBD, %PSA, Fsp3 & chiral atoms
- **Most change:** Mol Wt, HBA, RotB, PSA & Ar; *all increasing*

**Hypothesis:** drug properties changing least are more important

# Does Size Matter?

neutral molecules	MWt < 400 and clogP < 4	MWt > 400 and/or clogP > 4
solubility	average	lower
permeability*	higher	average/higher
bioavailability	average	lower
volume of Dist. **	average	average
plasma protein binding	average	higher
CNS penetration***	higher/average	average/lower
brain tissue binding	lower	higher
P-gp efflux	average	higher/average
in-vivo clearance	average	average
hERG Inhibition	lower	lower
P450 inhibition****	lower 2C9, 2C19, 2D6 & 3A4 inhibition	higher 2C9, 2C19 & 3A4 inhibition
P450 inhibition****	higher 1A2 inhibition	lower 1A2 inhibition
P450 inhibition****		average 2D6 inhibition

GSK: ADME '4/400' rule

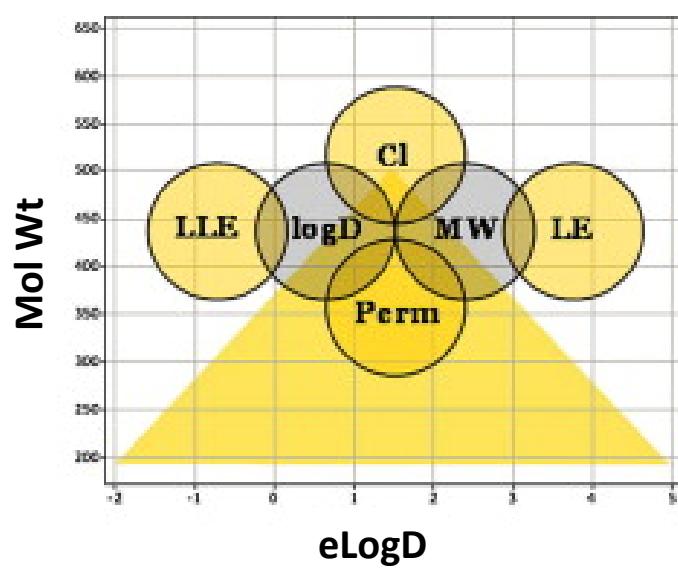
Gleeson, *J. Med. Chem.* 2008, **51**, 817

Mol Wt	AZLogD
<300	>0.5
300-350	>1.1
350-400	>1.7
400-450	>3.1
450-500	>3.4
>500	>4.5

AZLogD limits required to achieve >50% chance of high permeability for a given Mol Wt

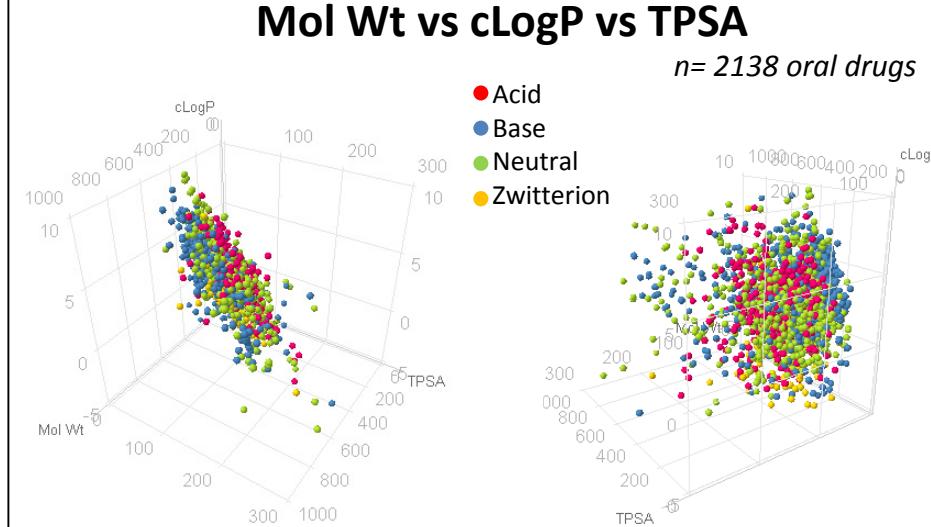
AZ: Mol Wt & LogD dependent permeability

Waring, *Bioorg. Med. Chem. Lett.*, 2009, **19**, 2844



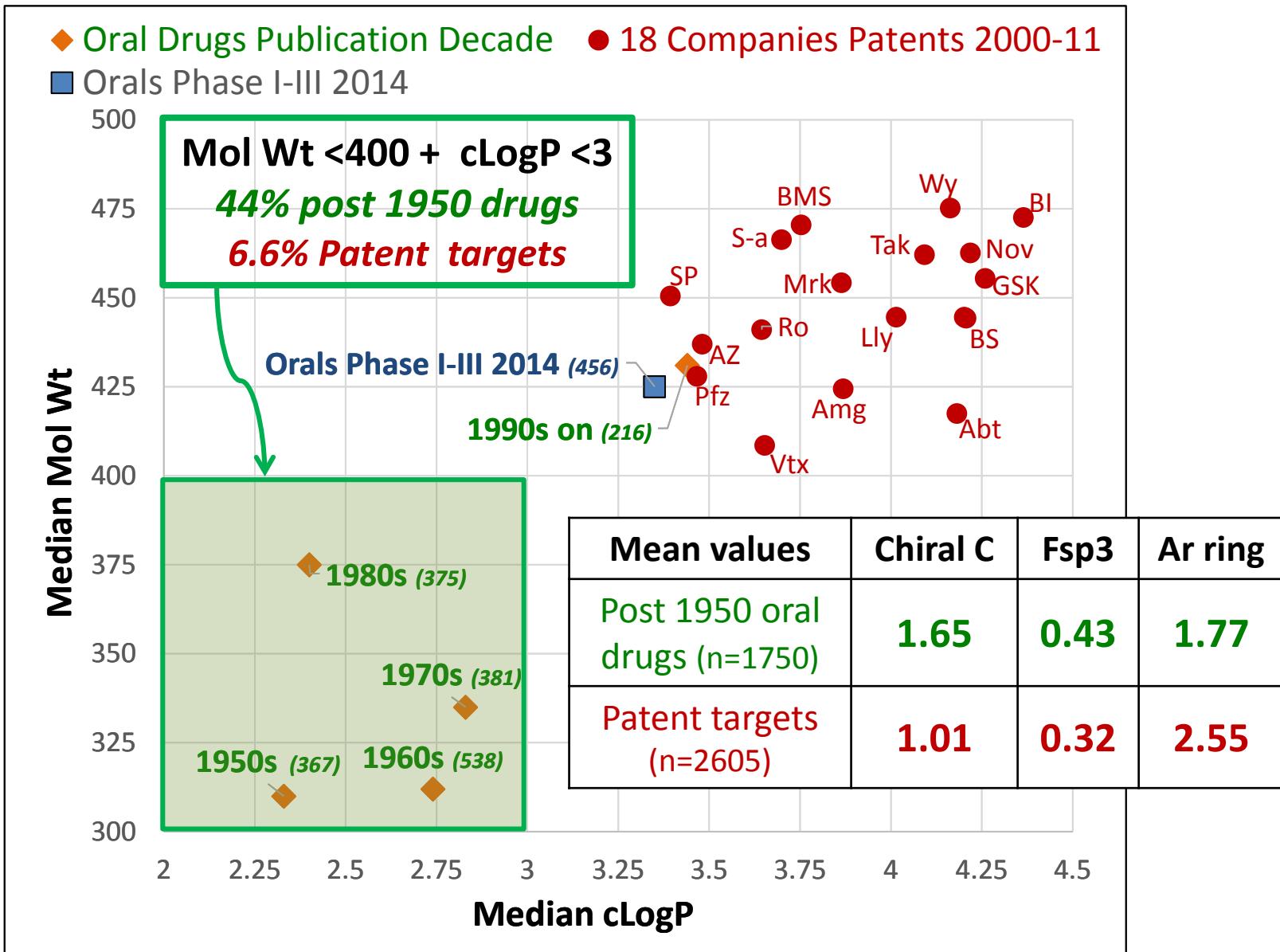
Pfizer: 'Golden triangle'

Johnson et al, *Bioorg. Med. Chem. Lett.*, 2009, **19**, 5560



Ro5 QSAR:  $c\text{LogP} = 0.0173 \text{ Mol Wt} - 0.564 \text{ O+N} - 0.439 \text{ OH+NH} + 0.246$   $n=2138, r^2 = 0.616$

# Inflation of 'Druglike' Physical Properties



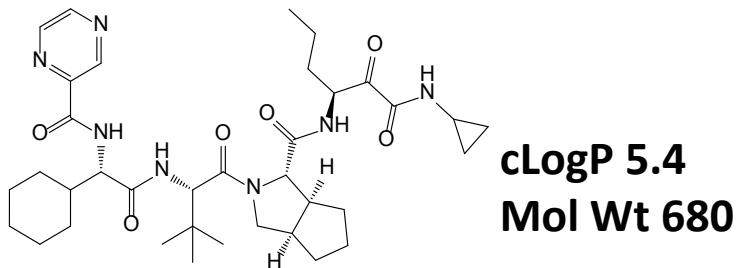
Drug data: Leeson et al, *Med. Chem. Comm.* 2011, **2**, 91, oral drugs updated to 2014; Patent targets 2000-11 from 18 companies: Leeson & St-Gallay, *NRDD* 2011, **10**, 749; Phase I-III orals: <http://www.citeline.com/>

# Disease Risk/Benefit & Property Inflation

**36% 2012-14 FDA approvals are orphan drugs**

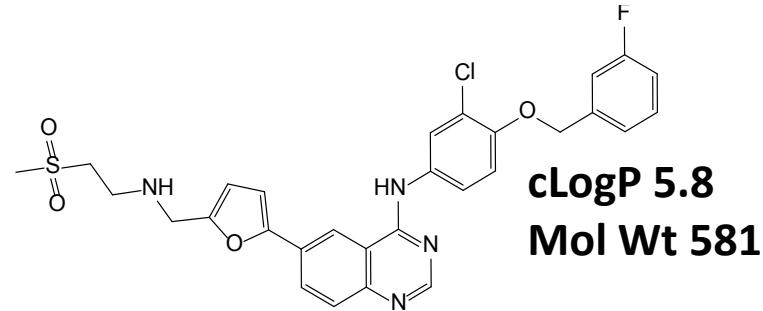
post-1990 Orals (n=216)	Median cLogP	Median Mol Wt	≥2 Ro5 unmet
Kinase, HIV prot., HCV (n=45)	4.64	556	40% (18)
Others (n=171)	3.07	420	12% (20)
			Pre-90: 6.5%

**Telaprevir:** HCV NS3 protease



Dose 750mg *tid*, high fat food; sol. 4.7 µg/ml, ‘less than marble;’ SDD formulation; **Black Box:** serious skin reactions; *efficacious, superceded*

**Lapatinib:** EGFR & ErbB2 kinases

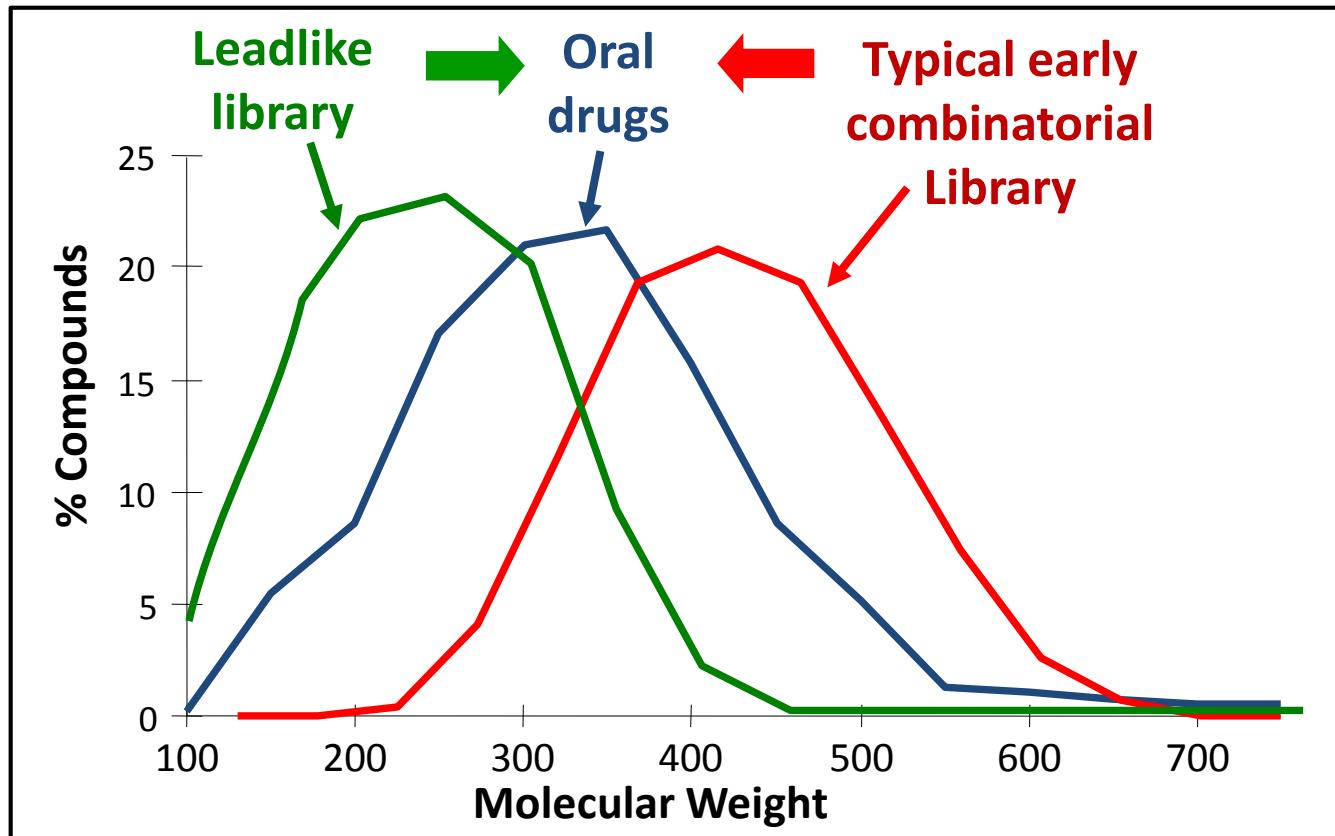


Dose 1500mg *uid*, 1hr before or after meal; sol. 7 µg/ml; hERG inhibitor; **Black Box:** hepatotoxic; *slow off-rate; standard treatment for breast cancer*

**Medical need & efficacy can overcome risk & dosing inconvenience**

**Telaprevir:** Kwong et al, *Nat. Biotech.* 2011, **29**, 993; **Lapatinib:** Lackey & Cockerell in *Kinase Inhibitor Drugs*, Wiley, 2009, p41; **Cancer drugs & food interaction:** Weitschies, *Clin. Pharm. & Therapeutics* 2013, **94**, 441

# Physical Properties Tend to Increase in Optimisation: *the ‘Leadlike’ Hypothesis*

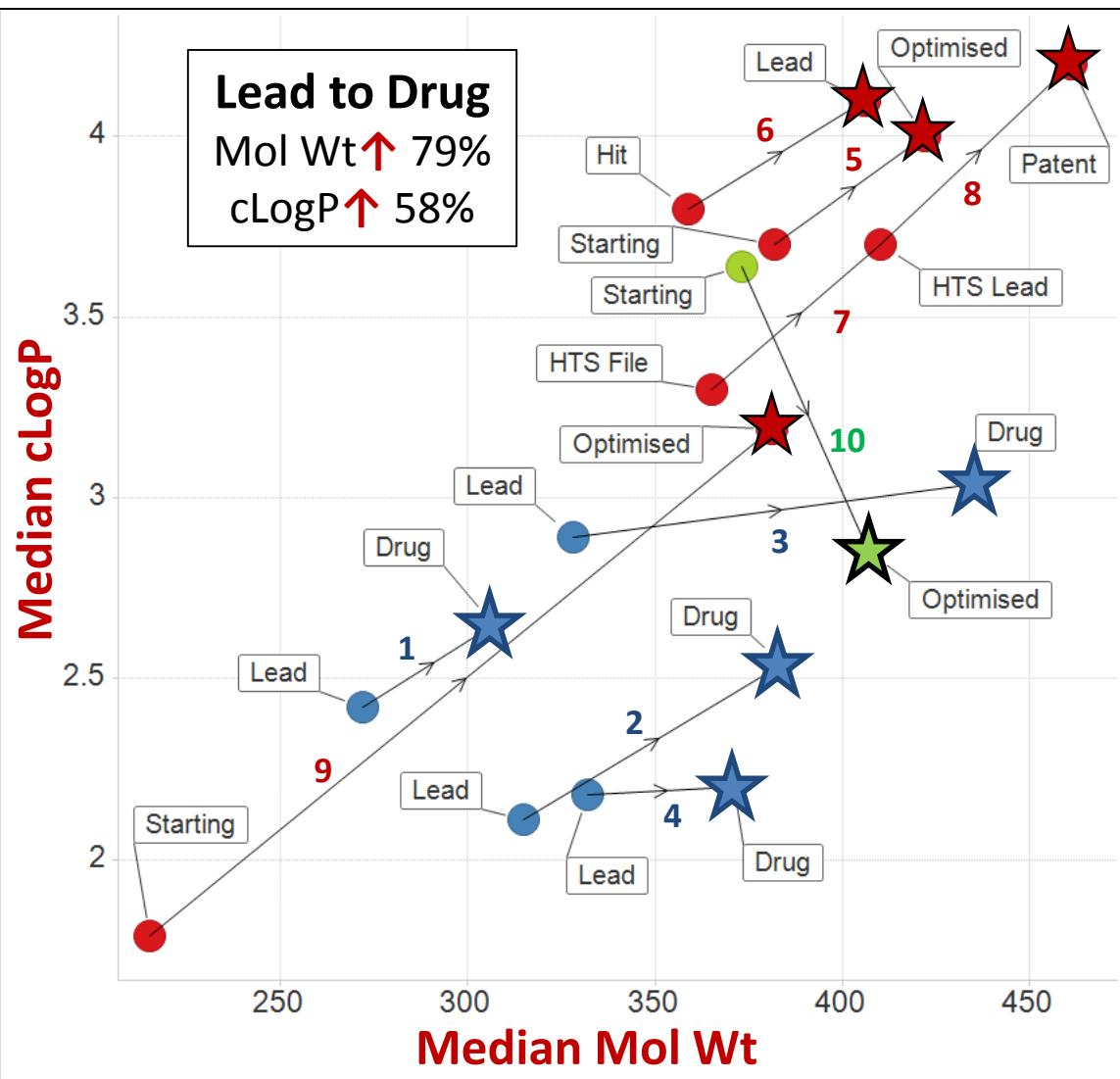


‘Leadlike’ lead: Affinity >0.1 $\mu$ M; Mol Wt 100-350; cLogP 1-3

Leadlikeness: Teague, Davis, Leeson & Oprea, *Angew. Chem. Int. Ed.* 1999, **38**, 3743; Oprea et al, *J. Chem. Inf. Comput. Sci.* 2001, **41**, 1308; Hann et al, *J. Chem. Inf. Comput. Sci.* 2001, **41**, 856; Synthetic challenges: Doveston et al., *Org. Biomol. Chem.* 2015, **13**, 859

# Property Inflation in Optimisation

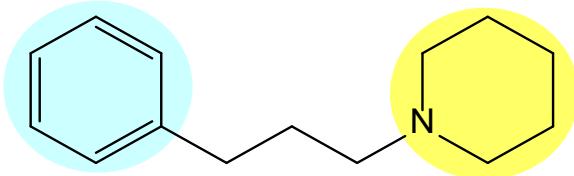
*Leadlike hypothesis:* Teague et al, Angew. Chem. Int. Ed. 1999, 38, 3743



1. Hann, J. Chem. Inf. Comput. Sci. 2001, **41**, 856; 2. Oprea, J. Chem. Inf. Comput. Sci. 2001, **41**, 1308; 3. Perola, J. Med. Chem. 2010, **53**, 2986;
4. Giordanetto, DDT 2011, **16**, 722; 5. Morphy, J. Med. Chem. 2006, **49**, 2969; 6. Keseru, NRDD 2009, **8**, 203; 7. Macarron, NRDD 2011, **10**, 188; 8. Leeson, NRDD 2011, **10**, 749; 9. Ferenczy, J. Med. Chem. 2013, **56**, 2478; 10. Hopkins, NRDD, 2014, **13**, 105

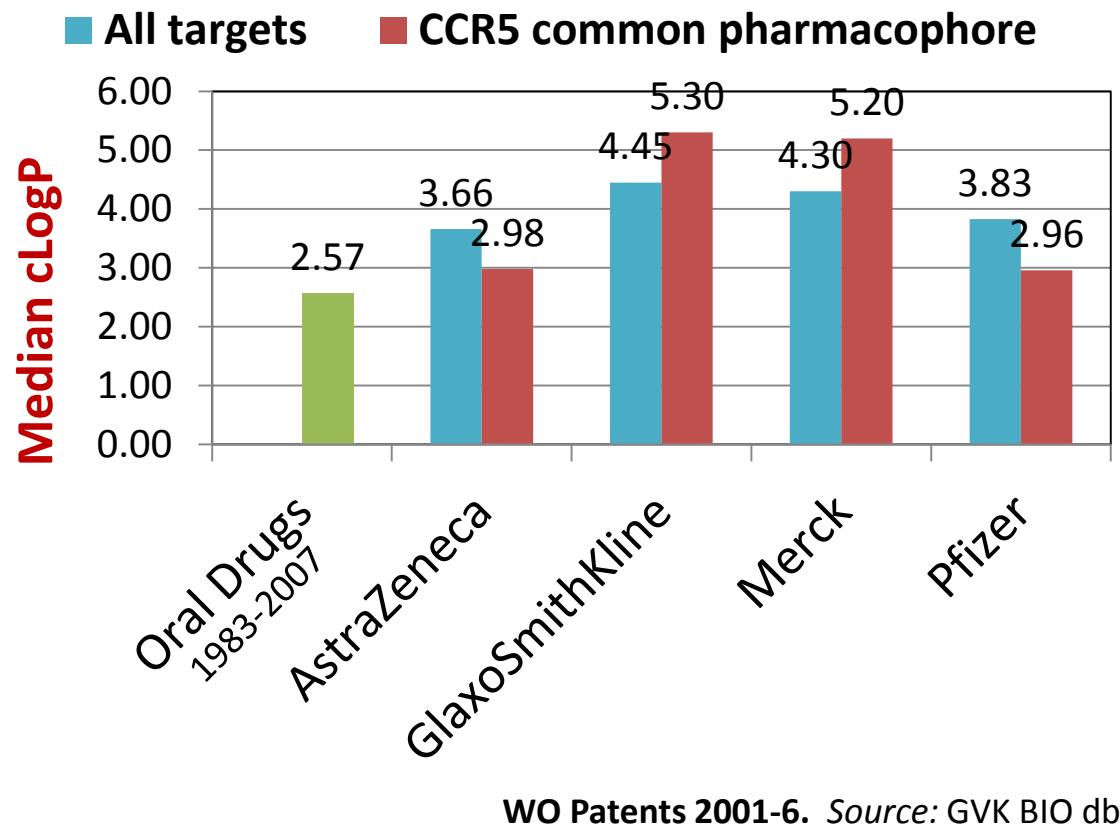
# Divergent Company Design Practices

*eg CCR5 Antagonists with a Common Pharmacophore*



**Phenylpropyl-piperidine**

CCR5 antagonist pharmacophore pursued by all 4 Companies:  
AstraZeneca & Pfizer reached the clinic

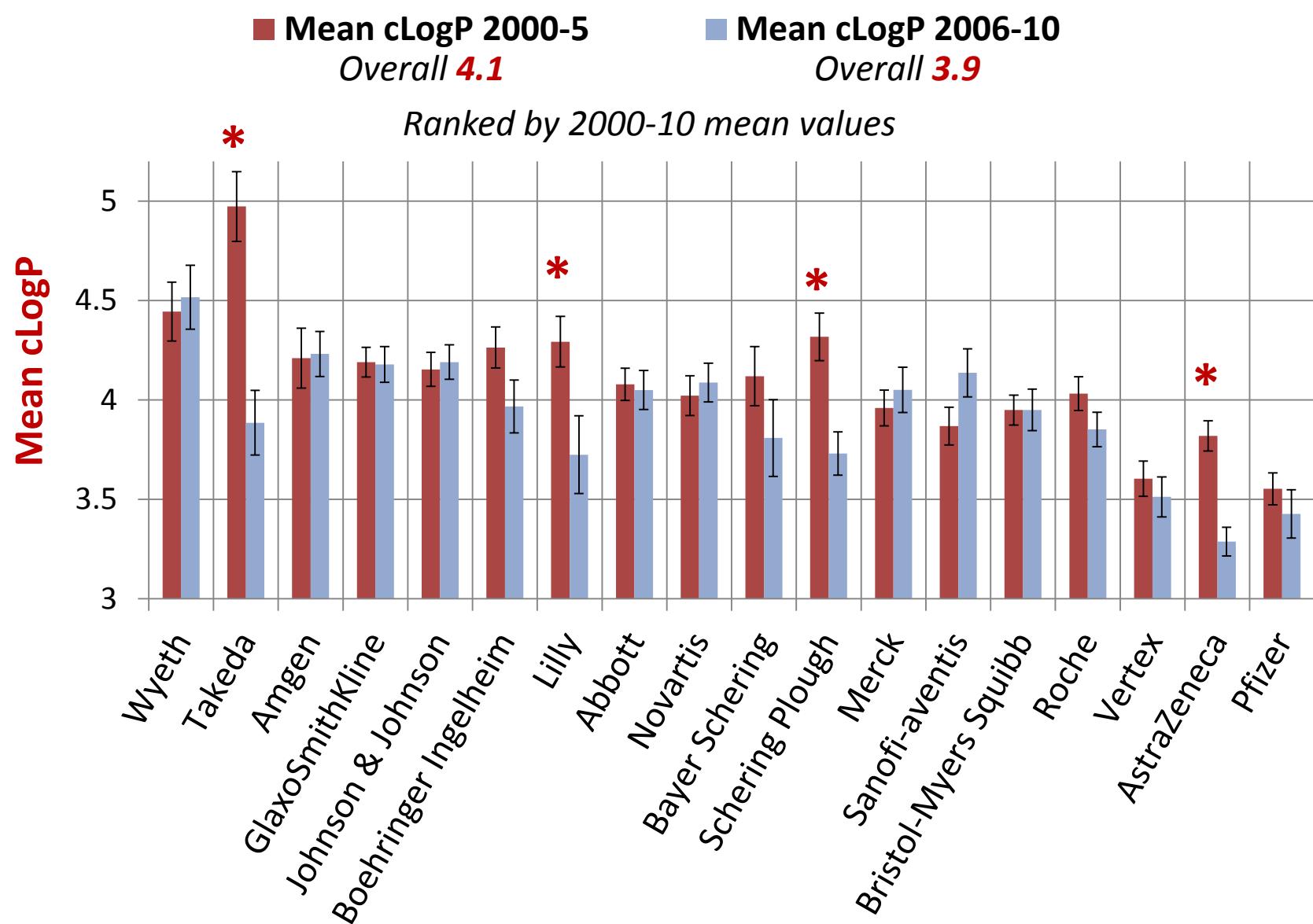


- Company differences: **comparable to target class differences**
- Companies' design strategies: **powerful impact of culture, history, experience, expertise; slow to change**

Leeson & Springthorpe, *Nat. Rev. Drug Disc.* 2007, **6**, 881;

18 Company target-unbiased 2000-11 analysis: Leeson & St-Gallay, *Nat. Rev. Drug Disc.* 2011, **10**, 749

# Some Companies are Changing, Many are Not



# Some Causes of Molecular Inflation

- **Increasing potency as the primary goal**
  - Often leads to increased cLogP & Mol Wt in a series
  - Medicinal ‘obsession’?
- **Misinterpreting the ‘rule of 5’**
  - Ro5 uses 90 percentile values
  - cLogP 4.5-5 + Mol Wt 450-500 is Ro5 compliant, *but occurs in only 1% of oral drugs*
- **Hit selection**
  - Hit validation / selection is a critical step
  - Mean literature HTS hit: pAct 6.1 & **cLogP 3.7**
- **Synthetic feasibility**
  - Parallel chemistry mostly adds Mol Wt
  - Complex molecules & ‘difficult’ chemistry sometimes avoided?

**Potency:** Hann, *MedChemComm.* 2011, **2**, 349; **HTS hit selection:** Keserű & Makara, *Nat. Rev. Drug Disc.* 2009, **8**, 203; Dahlin & Walters, *Future Med. Chem.* 2014, **6**, 1265; **Synthetic pragmatism:** Keserű et al, *Chem. Soc. Rev.*, 2014, **43**, 5387; **PPI:** Kuenemann et al, *J. Chem. Inf. Model.* 2014, **54**, 3067; **Company practice:** Leeson & St-Gallay, *Nat. Rev. Drug Disc.* 2011, **10**, 749; **Chemist behaviour:** Kutchukian, et al, *PLoS ONE*, 2012, **7**, e48476; **MPO:** Wager et al, *J. Med. Chem.* 2013, **56**, 9771

# Some Causes of Molecular Inflation, contd.

- **Increase in less ‘druggable’ targets**
  - ‘Low-hanging fruit’ at the centre of drug-like space has been picked?
  - New, tougher targets – eg protein-protein interactions with large hydrophobic interfaces?
- **Disease risk/benefit**
  - Increased acceptance of safety risk & dosing inconvenience
- **Divergent design practices**
  - Search for new intellectual property; most targets are pursued by >1 organisation
  - Multiparameter optimisation used? Influence on medicinal chemists’ decisions from **computational & ADMET** scientists?
- **It does not matter**
  - ‘There are already highly lipophilic drugs on the market’

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# Ligand Efficiency (LE) & Lipophilic LE (LLE or LipE)

*'Bang for your buck' guidelines*  $p(\text{Activity}) = pK_d, pK_i, \text{pIC}_{50}, \text{pEC}_{50}$

$$\text{LE} = p(\text{Activity}) * 1.37/\text{HA}$$

Units: kcal/mol/heavy atom

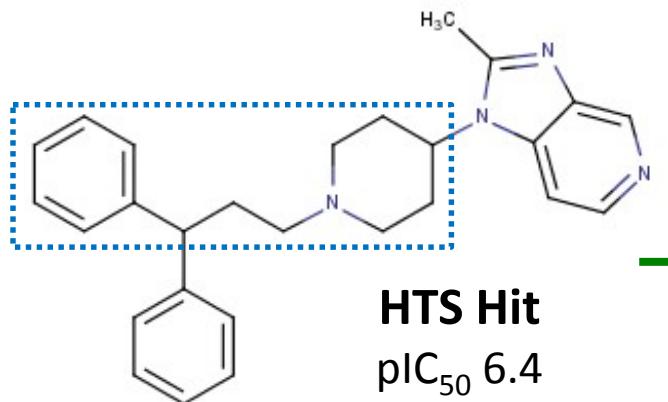
Mean oral drug LE = 0.45

$$\text{LLE} = p(\text{Activity}) - \text{LogP/D}$$

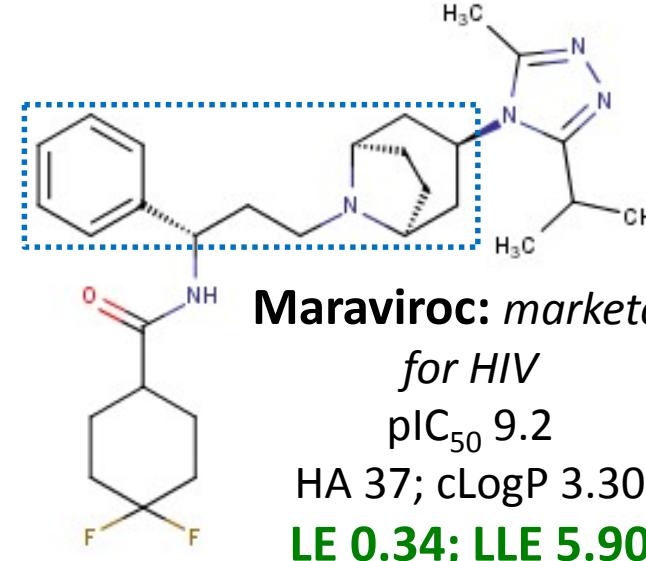
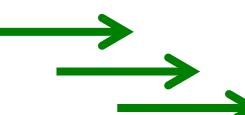


Mean oral drug LLE (cLogP) = 4.4

## Pfizer CCR5 Receptor Antagonist Optimisation



$\Delta \text{LLE} 4.1$



**Employed LE; issues addressed:** antiviral activity, P450 & hERG inhibition, permeability

Wood & Armour, *Prog. Med. Chem.* 2005, **43**, 239; Price et al, *Bioorg. Med Chem Lett* 2006, **16**, 4633; Armour et al, *ChemMedChem* 2006, **1**, 706

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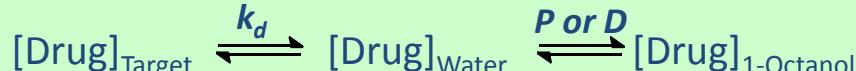
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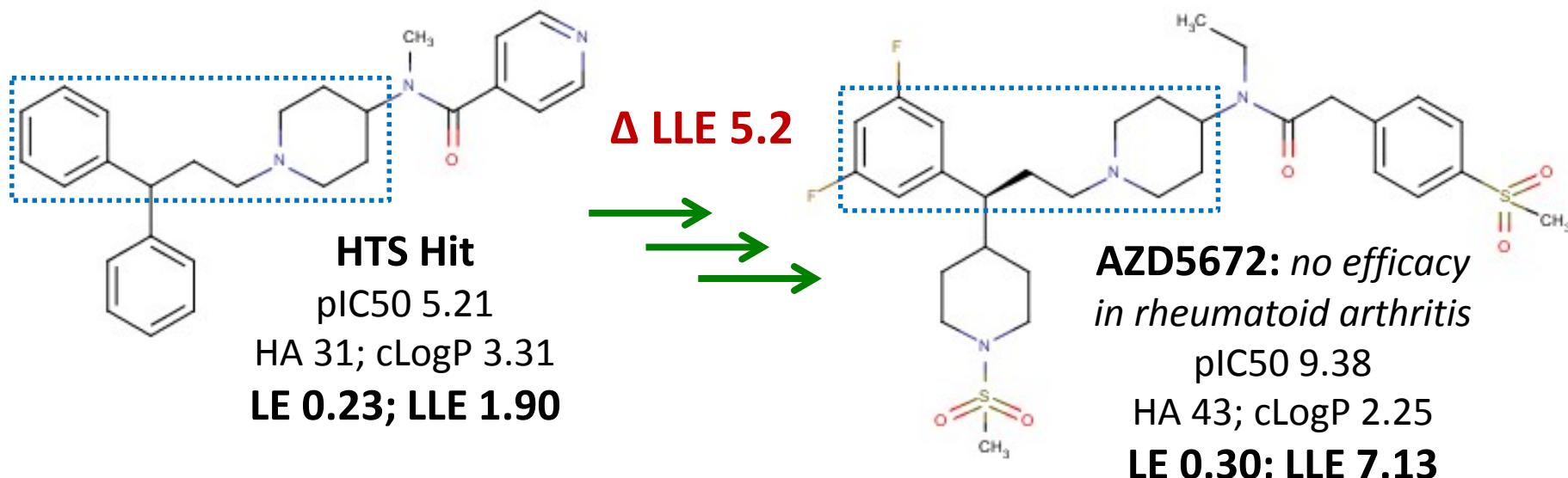
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## AstraZeneca CCR5 Receptor Antagonist Optimisation



**Issues addressed:** affinity, hERG inhibition, absorption

Cumming et al, *Bioorg. Med. Chem. Lett.* 2012, **22**, 1655; 2005, **15**, 5012; 2006, **16**, 3533

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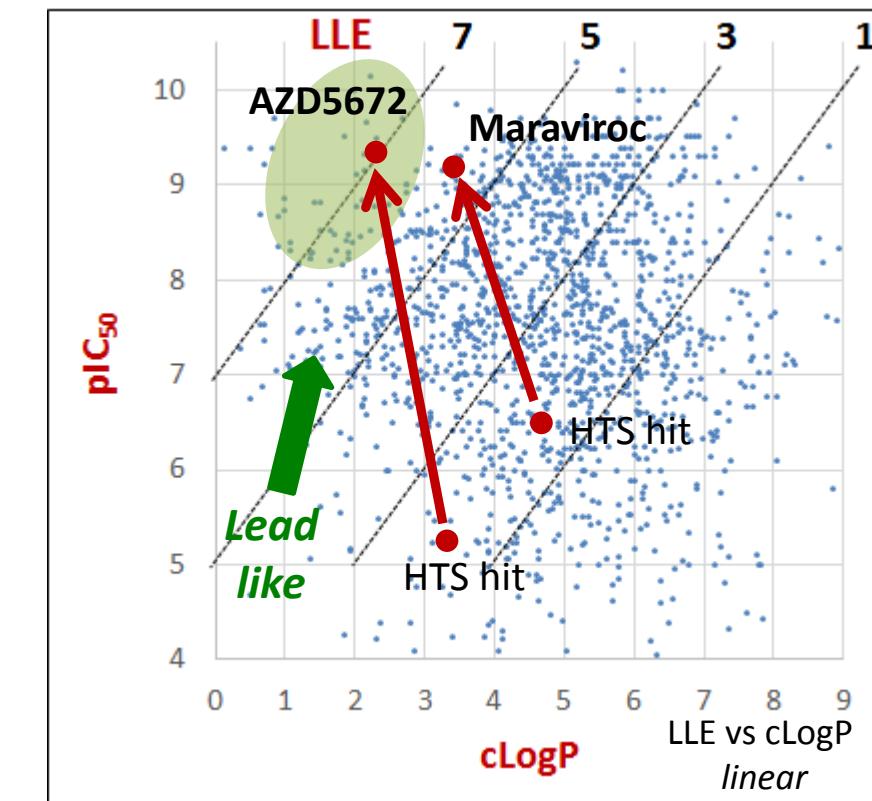
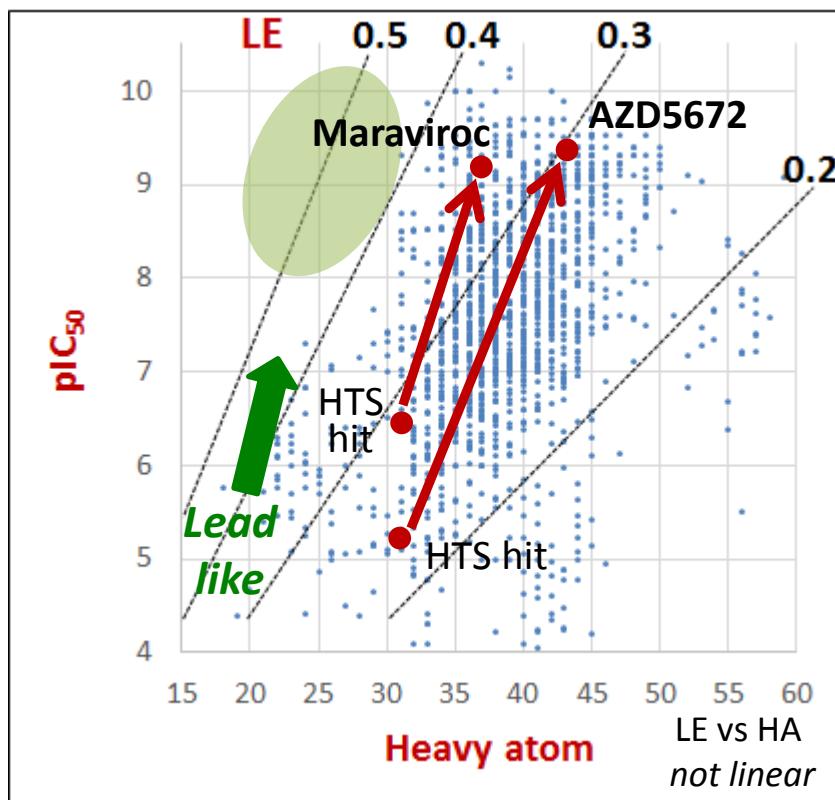
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CCR5 Receptor Ligands:  $\text{pIC}_{50}$  values ex CHEMBL (n=1513)



Drug data (n=261) calcd. from: Gleeson et al. *Nat. Revs. Drug Disc.* 2011, **10**, 197; LE: Hopkins et al, *Drug Disc. Today* 2004 **9**, 430; LLE: Leeson & Springthorpe, *Nat. Rev. Drug Disc.* 2007, **6**, 881; Review: Hopkins et al, *Nat. Rev. Drug Disc.*, 2014, **13**, 105; Debate: Shultz, *ACS Med. Chem. Lett.* 2014, **5**, 2; Murray et al, *ACS Med. Chem. Lett.* 2014, **5**, 616; Kenny et al, *J. Comput. Aided Mol. Des* 2014, **28**, 699

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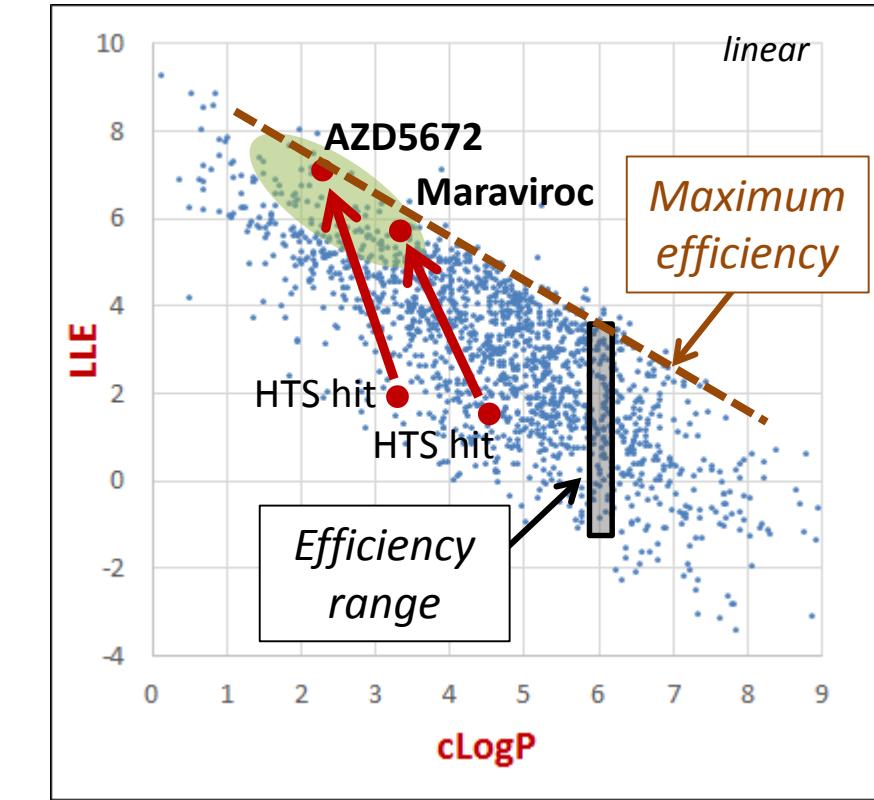
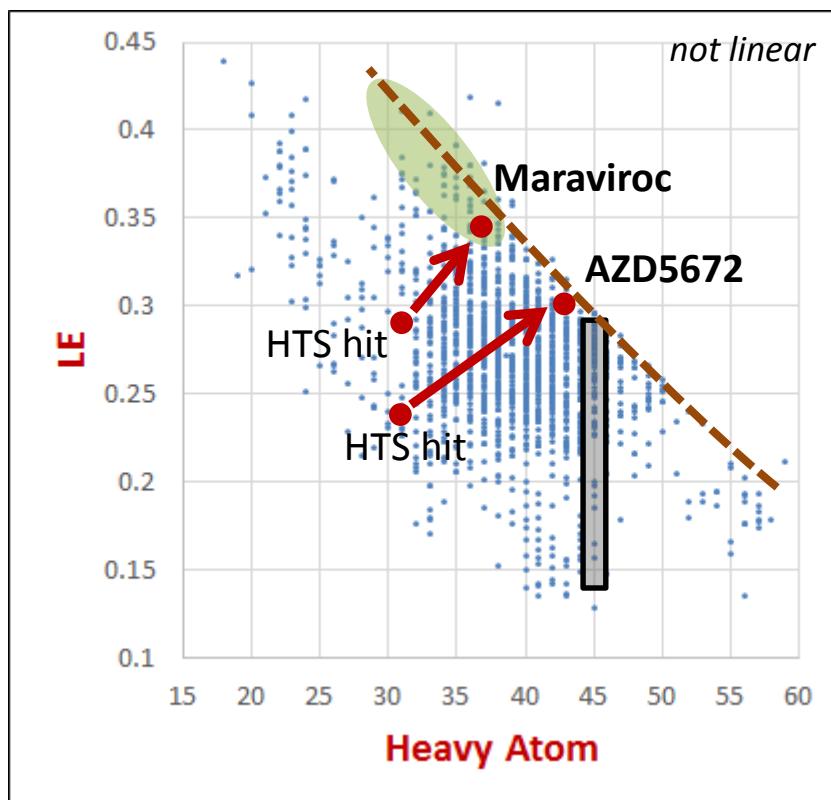
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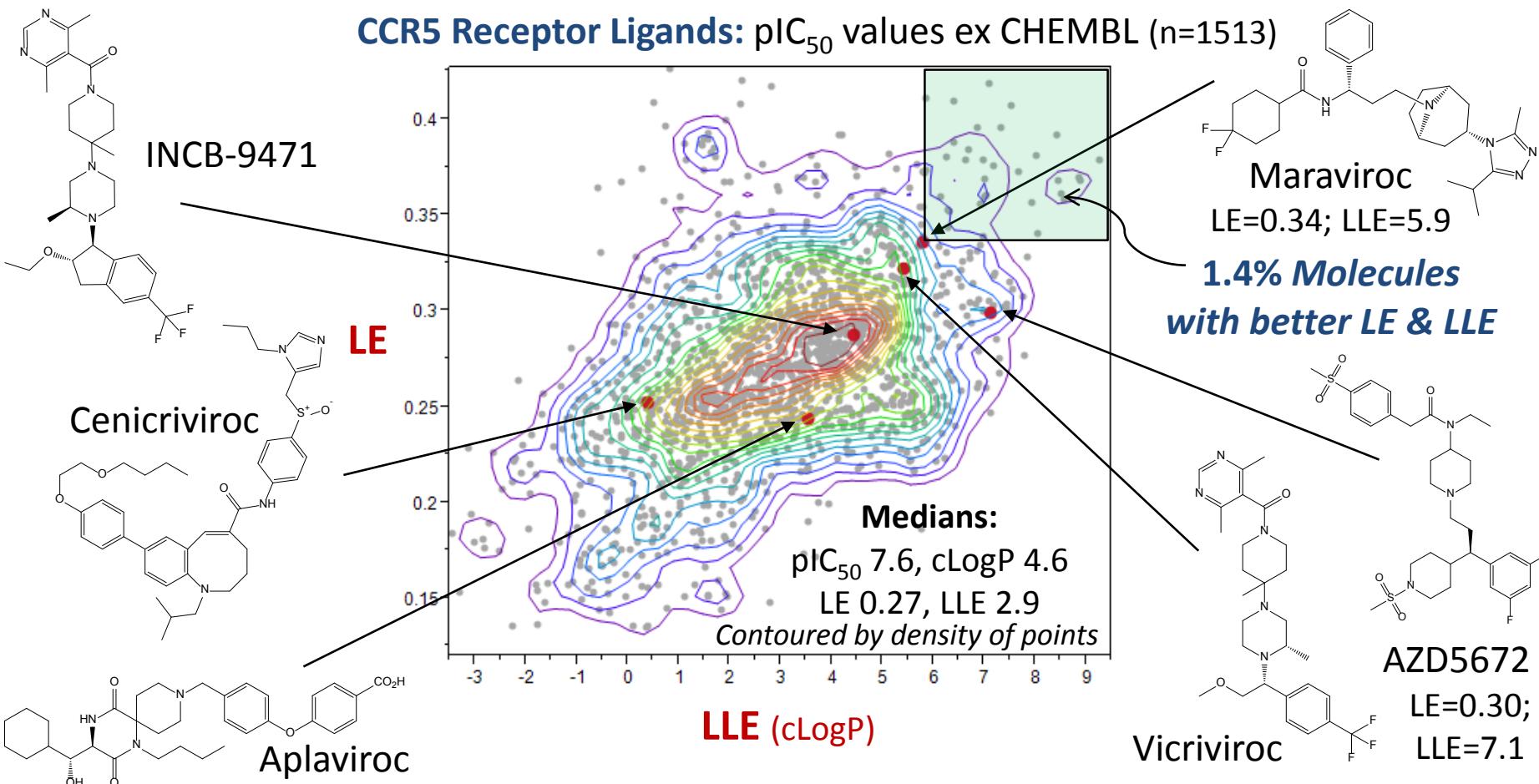
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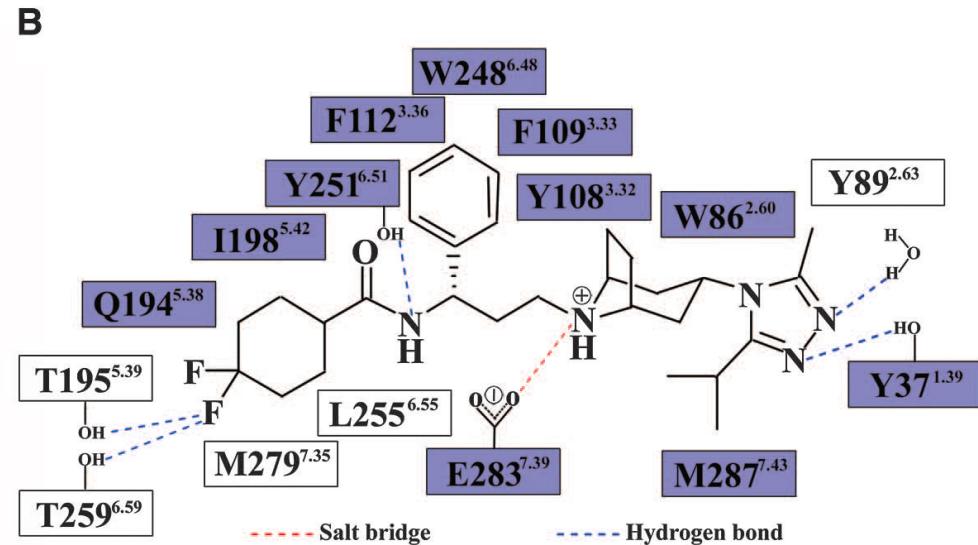
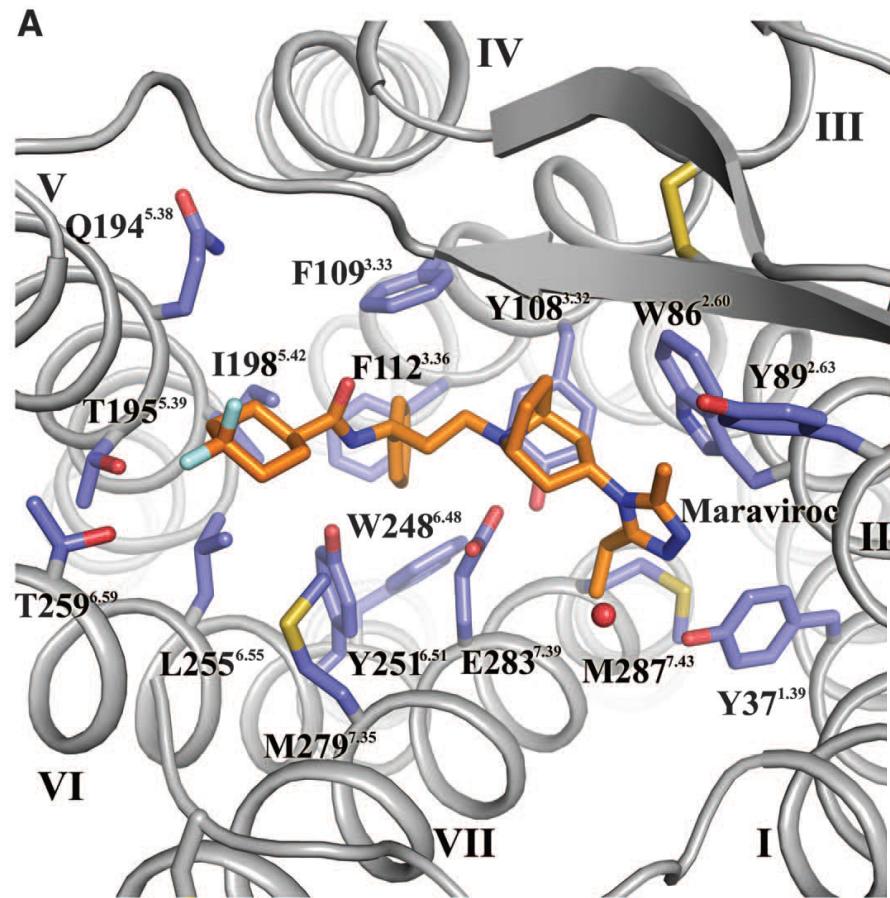
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# Structure of Maraviroc Bound to CCR5



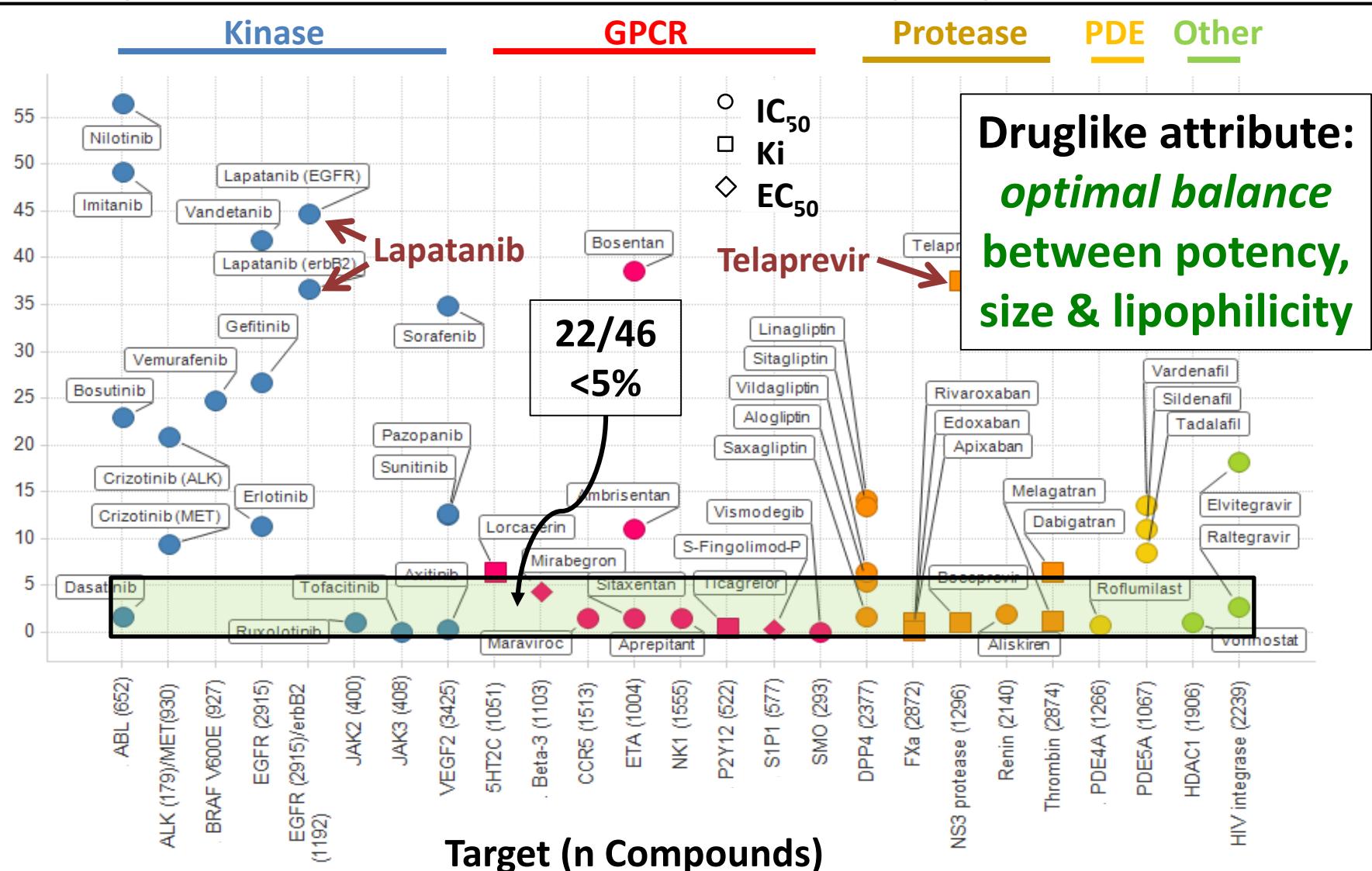
- Efficient use of H-bonding atoms
  - 7 Polar atoms make 6 polar interactions: enthalpy ↑
- Efficient local hydrophobic interactions
  - Phenyl, isopropyl, tropane & cyclohexyl binding pockets

Maraviroc dissociation:  
 $t_{1/2}$  6.4hrs

# Oral Drug Ligand Efficiencies: 46 Drugs, 25 Targets

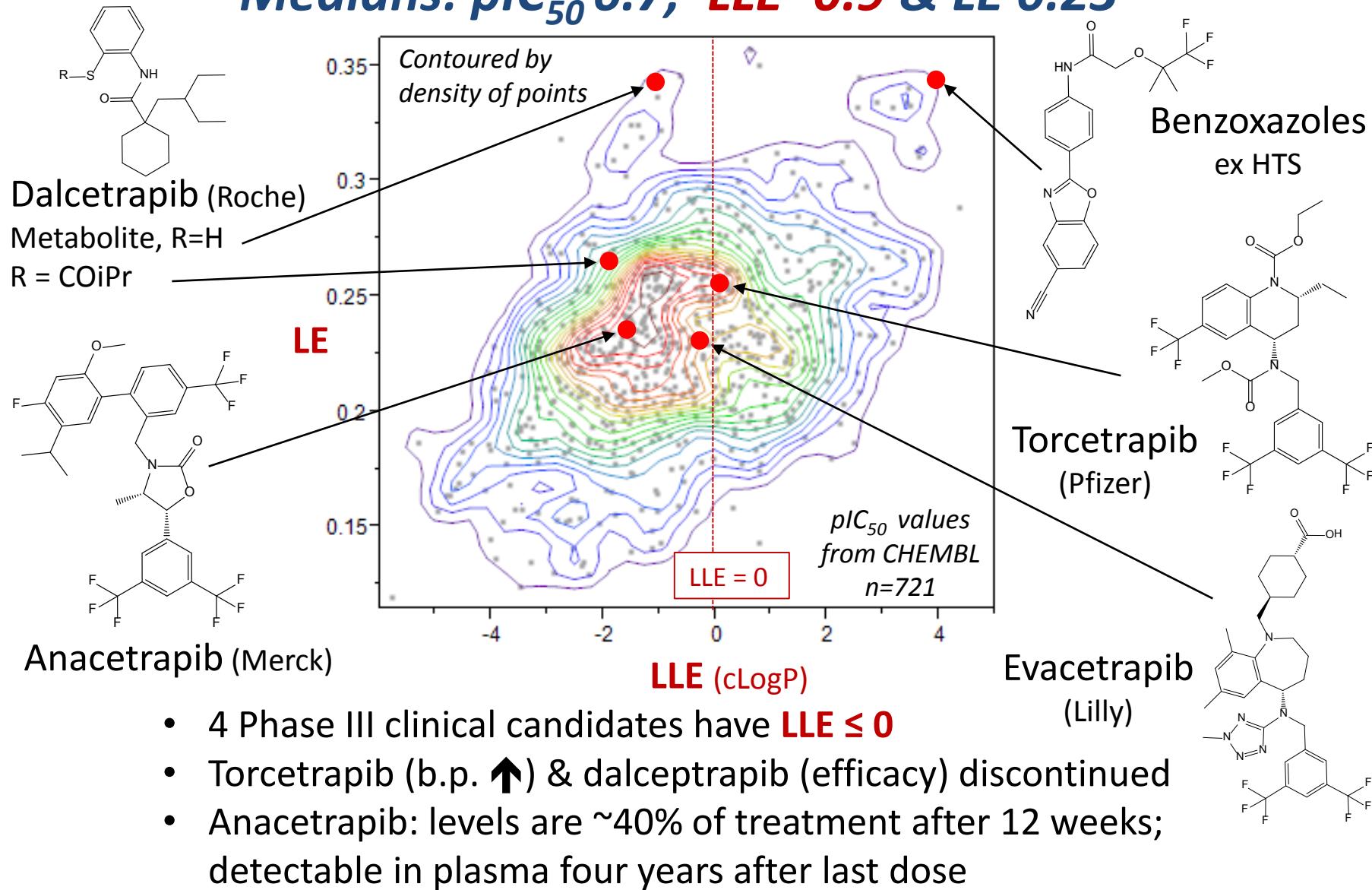
% LE + LLE better vs drug: kinases 22%; other targets 2.7%;  
only in class 1.5%. LE & LLE contribute equally to % score

% Compounds with both LE & LLE better than drug



# CETP: A High Value ‘Lipophilic’ Target

*Medians:  $pIC_{50}$  6.7; LLE -0.9 & LE 0.23*

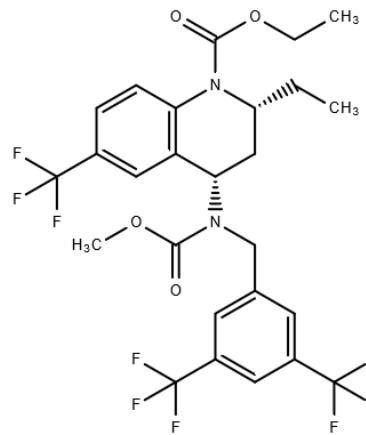


LE & LLE data: Hopkins et al, *Nat. Rev. Drug Disc.*, 2014, **13**, 105; CETP review: Mantlo & Escribano. *J. Med. Chem.* 2014, **57**, 1; Anacetrapib: Gotto et al, *Am. J. Cardiol.* 2014, **113**, 76; Benzoxazoles, eg *Bioorg. Med. Chem. Lett.* 2010, **20**, 1019

# CETP: Designing Less Lipophilic Inhibitors

*C → N & O, hydrophilic substituents, control HA*

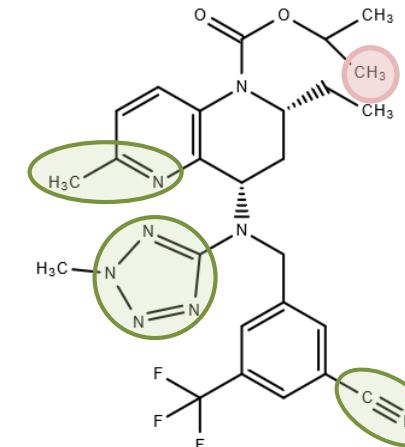
Fernandez et al (Lilly), *Bioorg. Med. Chem. Lett.* 2012, **22**, 3056



Torcetrapib (Pfizer)  $\text{pIC}_{50}$  7.7

cLogP 7.6; HA 41; **LLE 0.1; LE 0.26**

**'Mitigate lipophilicity'**  
*LogP values not cited*  
 $\Delta \text{LLE} = 3.8$   
 $\Delta \text{LE} = 0.01$

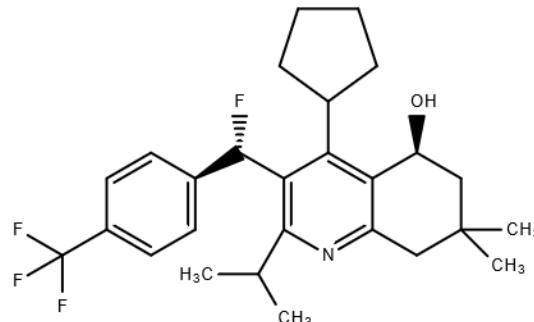


Lilly lead  $\text{pIC}_{50}$  7.7

cLogP 3.8; HA 39; **LLE 3.9 LE 0.27**

LE + LLE % better  
**1.4%**

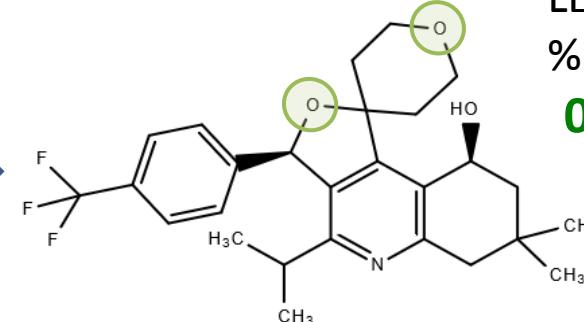
Trieselmann et al (BI), *J. Med. Chem.* 2014, **57**, 8766



BI hit  $\text{pIC}_{50}$  6.6

cLogP 7.6; HA 33; **LLE -1.0; LE 0.27**

**'Reduce lipophilicity'**  
*LogP values tracked*  
 $\Delta \text{LLE} = 4.1$   
 $\Delta \text{LE} = 0.04$



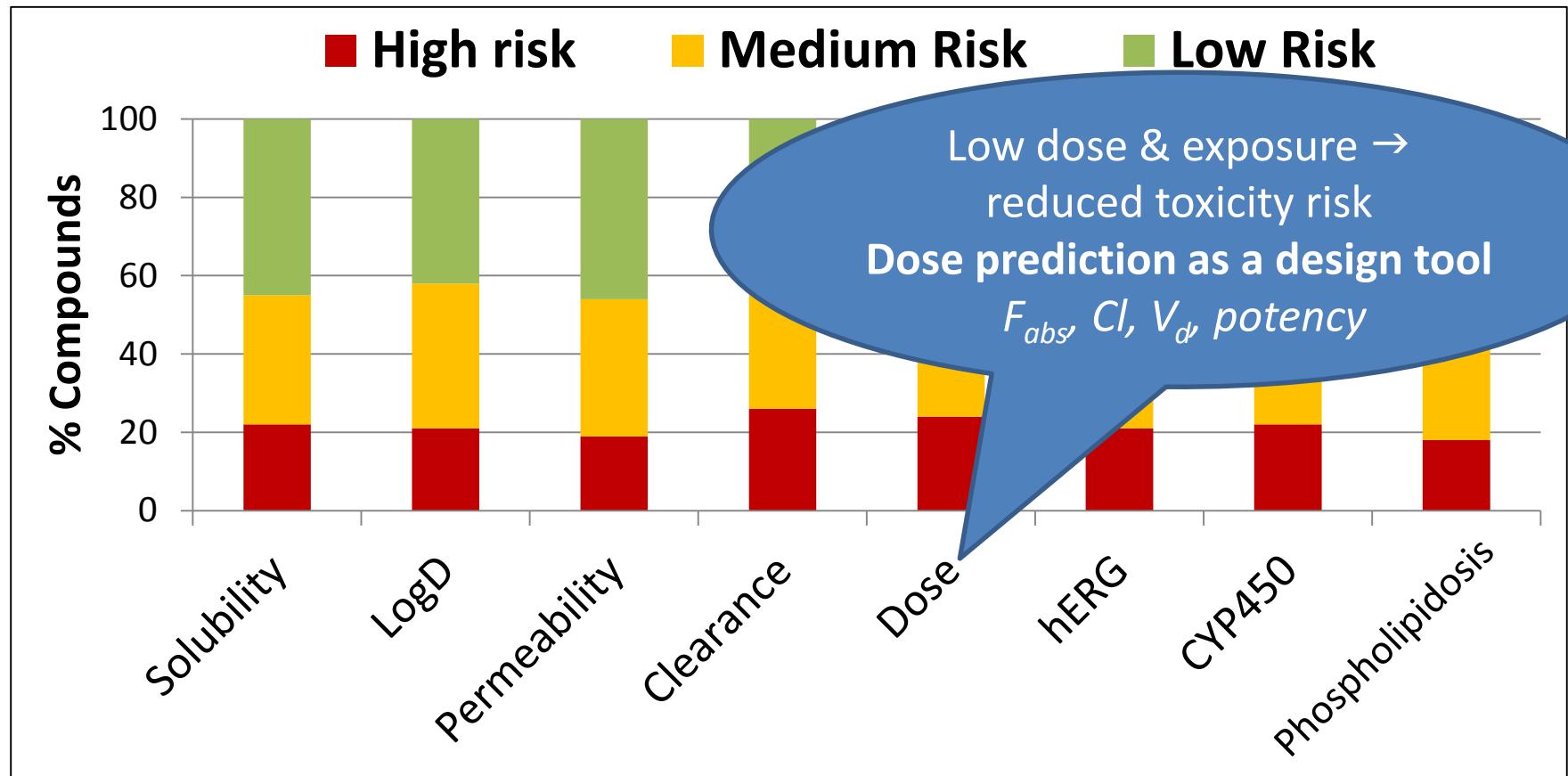
BI lead  $\text{pIC}_{50}$  7.7

cLogP 4.6; HA 34; **LLE 3.1; LE 0.31**

LE + LLE % better  
**0.28%**

# Controlling Risk: Compound Quality Guidance

*Global & project predictive ADMET models: use pre-synthesis*



- If prediction is poor, why synthesise?
- Using predictive models, AZ improved candidate drug **solubility**
- '**Virtual medicinal chemist**' – Σ tools using existing knowledge

# Designing Better Compounds

- Conduct multi parameter biology/ADMET optimisation
- Engage computational chemists & ADMET experts in design decision-making
- Seek leadlike starting points. Drop unpromising series; have back-up hit & lead generation plans
- Control physicochemical properties, especially lipophilicity; optimise ligand efficiencies & solubility
- Employ advanced computational chemistry tools; don't make compounds with poor predicted properties
- Reduce reliance on 'easy' synthesis & catalogue building blocks
- Learn constantly from past experience, avoid bias, consult others, challenge dogma
- **Never compromise on candidate compound quality**
- *...and persist*

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**'Minimal hydrophobicity' in drug design.** "Without convincing evidence to the contrary, drugs should be made as hydrophilic as possible without loss of efficacy."

C. Hansch, J.P. Björkroth & A. Leo, *J. Pharm. Sci.* 1987, **76**, 663-687