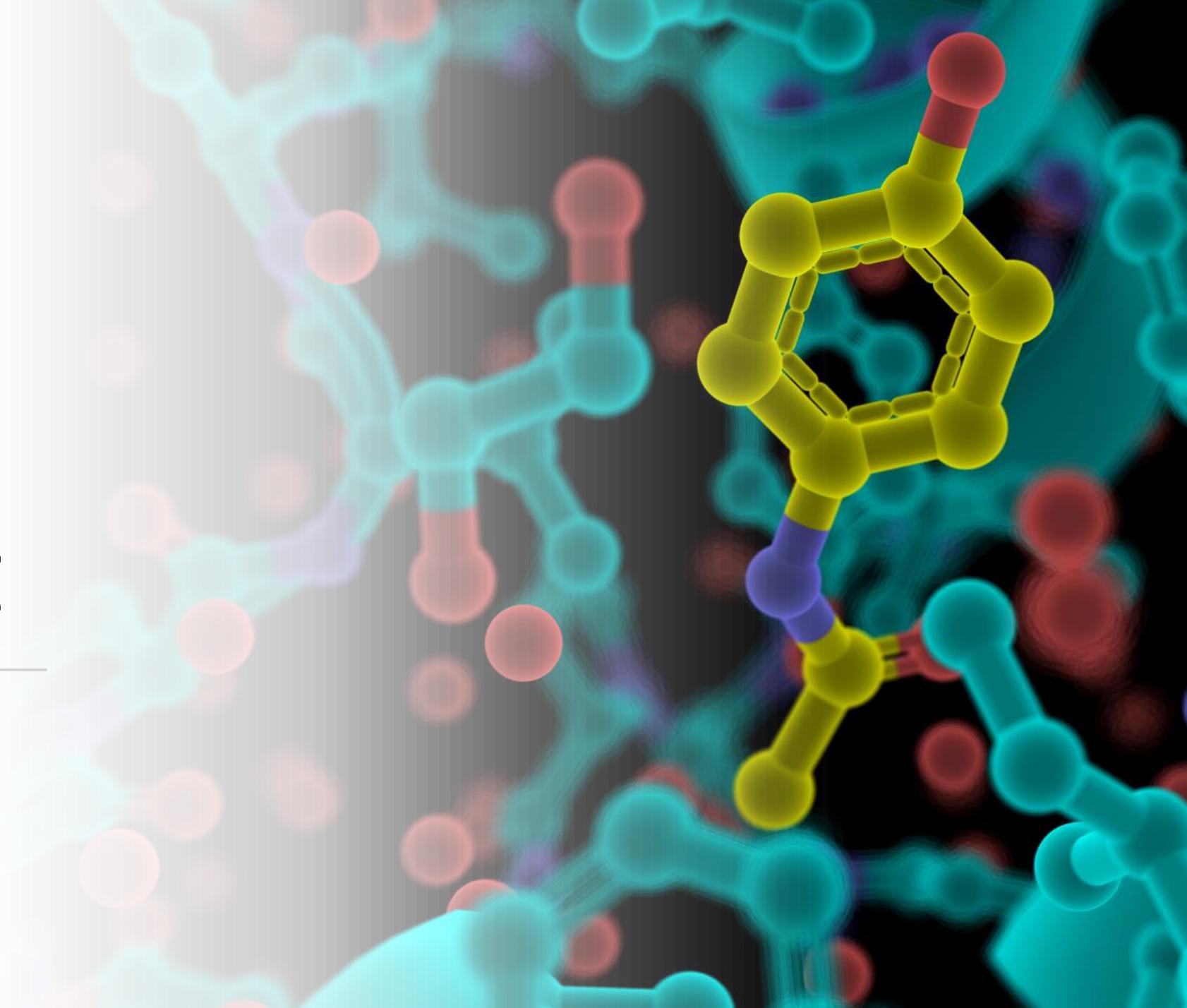




# Target-based drug screening

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University of Toronto



# Learning objectives

In this lecture you will:

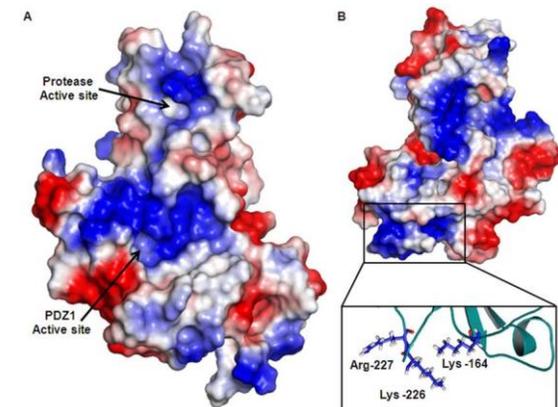
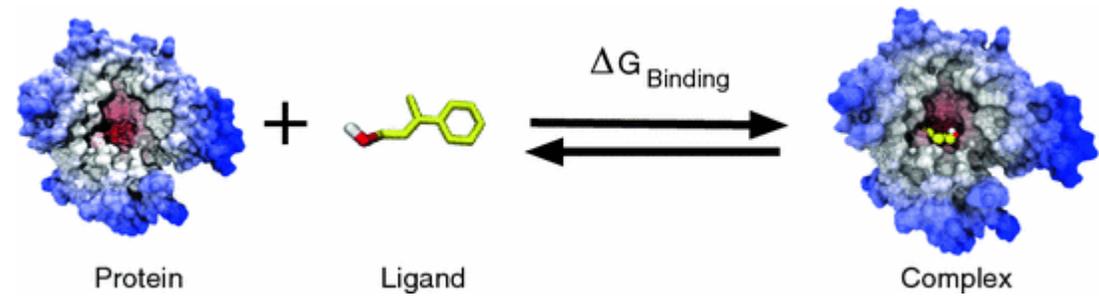
- Appreciate several major strategies for hit-finding in experimental drug screening
- Explore the contribution of chemical library choice to hit-finding
- Gain awareness of common artefacts of drug screens
- Be introduced to the complexity of drugs that intentionally target multiple proteins

# Early-stage drug discovery: **Hit-finding**

- In small-molecule (chemical) drug discovery, a **“hit” compound**:
- Binds to the protein target
- Modulates the function of the protein target
- Has relatively poor potency (eg 10  $\mu\text{M}$ ); the screen will often be conducted at a single concentration of 10  $\mu\text{M}$
- The “hits” will be improved by medicinal chemists to improve:
  - potency (10  $\mu\text{M}$  hit to 20 nM clinical drug)
  - Pharmacokinetics (chemical properties of drug determines how drug moves in people, eg crosses membranes etc)

# Molecular docking

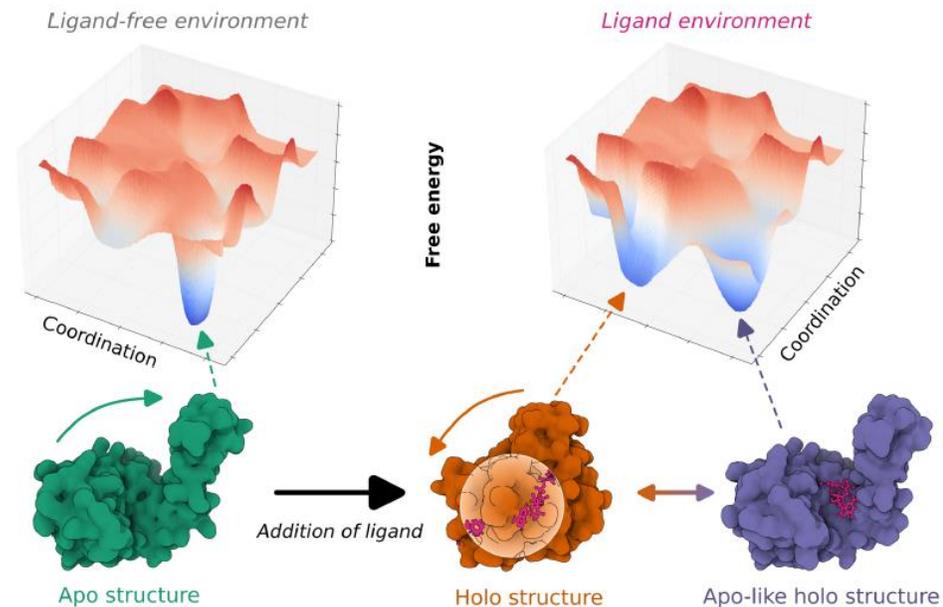
- The process of simulating the interaction between a small molecule (ligand or drug) and a macromolecule (eg a protein)
- Appropriate when a **protein's structure** is known (or well-predicted)
- Most intuitive:
  - Appropriate geometry (in 3D)
- Beyond just geometry,
- Multiple constraints:
  - Chemical interactions
    - Electrostatics and charge (see red/blue lower right)
    - Hydrogen bonding
    - Water – fat solubility (hydrophobicity, hydrophilicity)



# Protein-ligand complexes: distinct from the protein or ligand alone

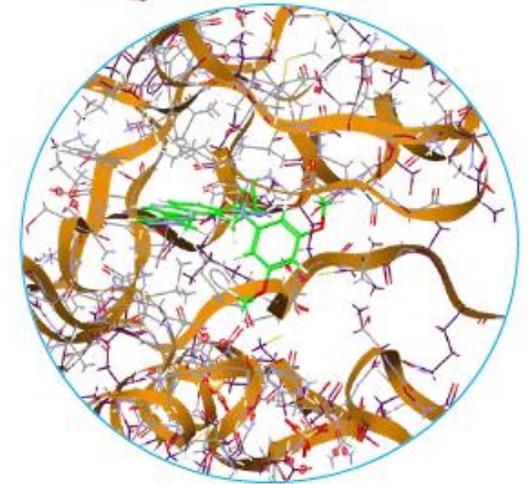
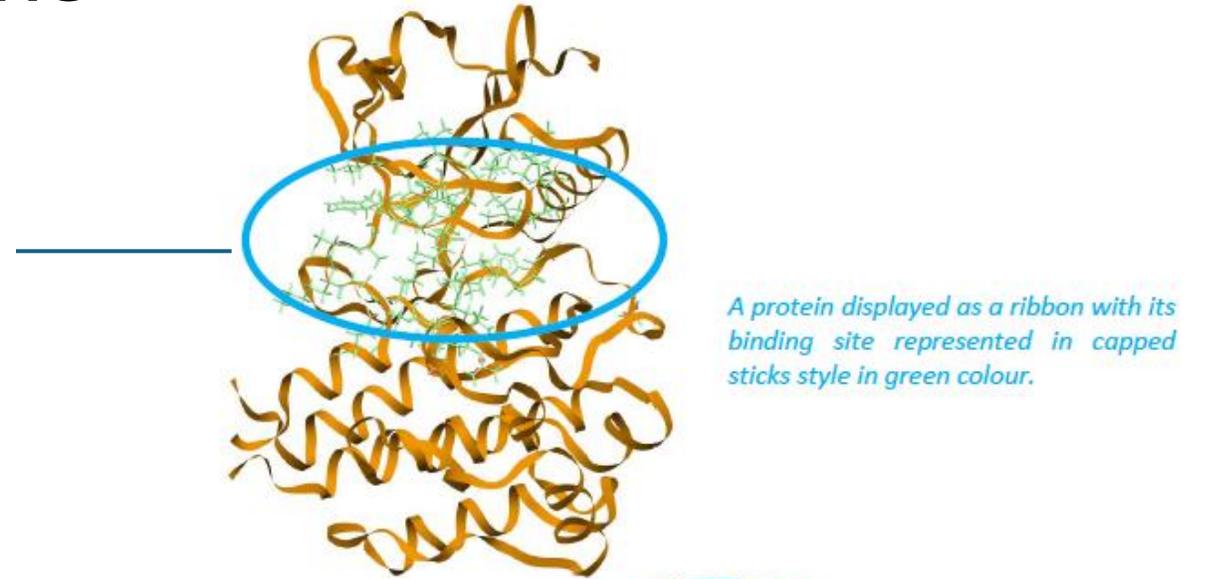
- Last session: idea of induced fit. Chemicals change the protein structure (“induce” a fit)

a) Conformational changes on free-energy landscapes



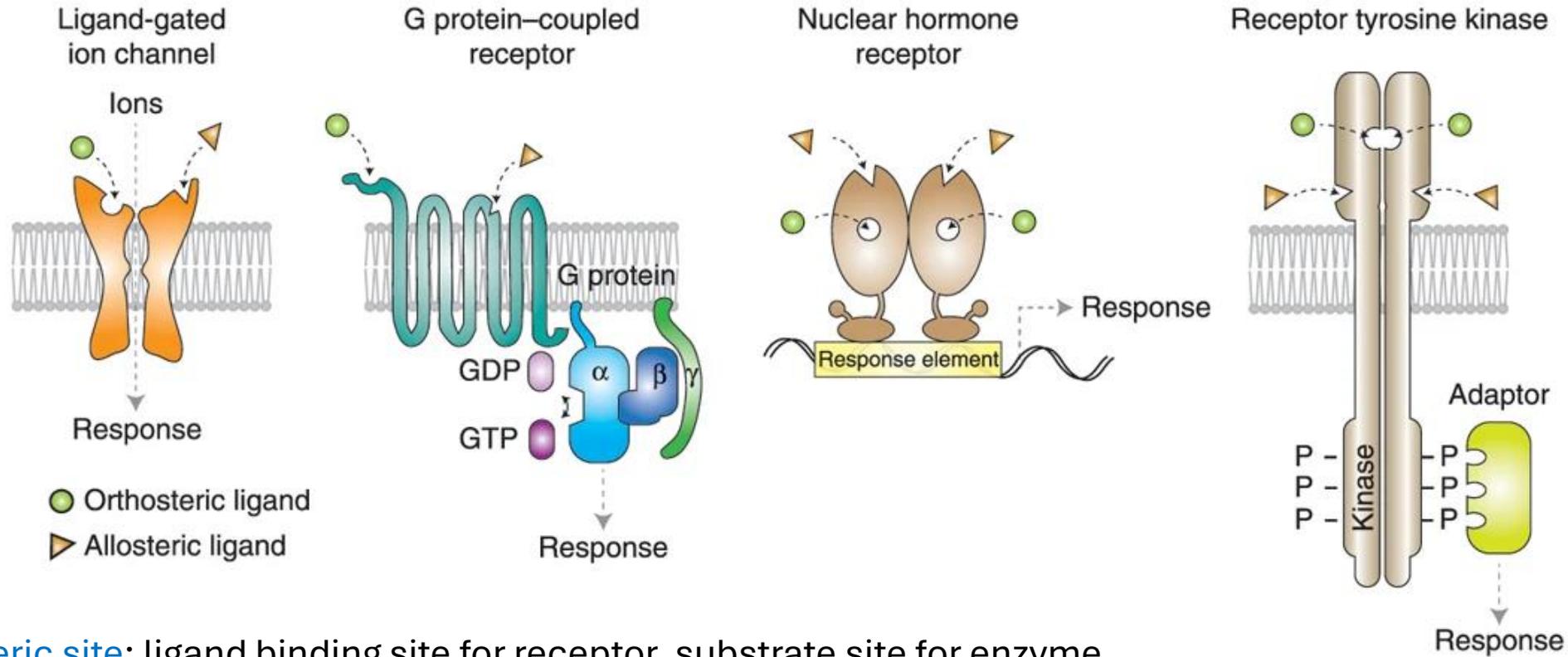
# Considerations

- **Binding site:** a region of the protein to which a drug or chemical binds
- **Pose:** a specific position (often a modeled position) of the **ligand-protein** complex
- **Affinity:** how tightly the chemical binds to the protein
- Affinity expressed as **Kd** (dissociation constant); low numbers = high affinity
- Screening assay:
  - Signal / no signal at single concentration of chemicals in a library
- **Potency:** drug's biological activity, expressed in units of concentration (IC50 = inhibitory concentration 50%)



Docking calculates a pose for the ligand and a score for the pose.

# All major classes of receptors can be allosterically modulated

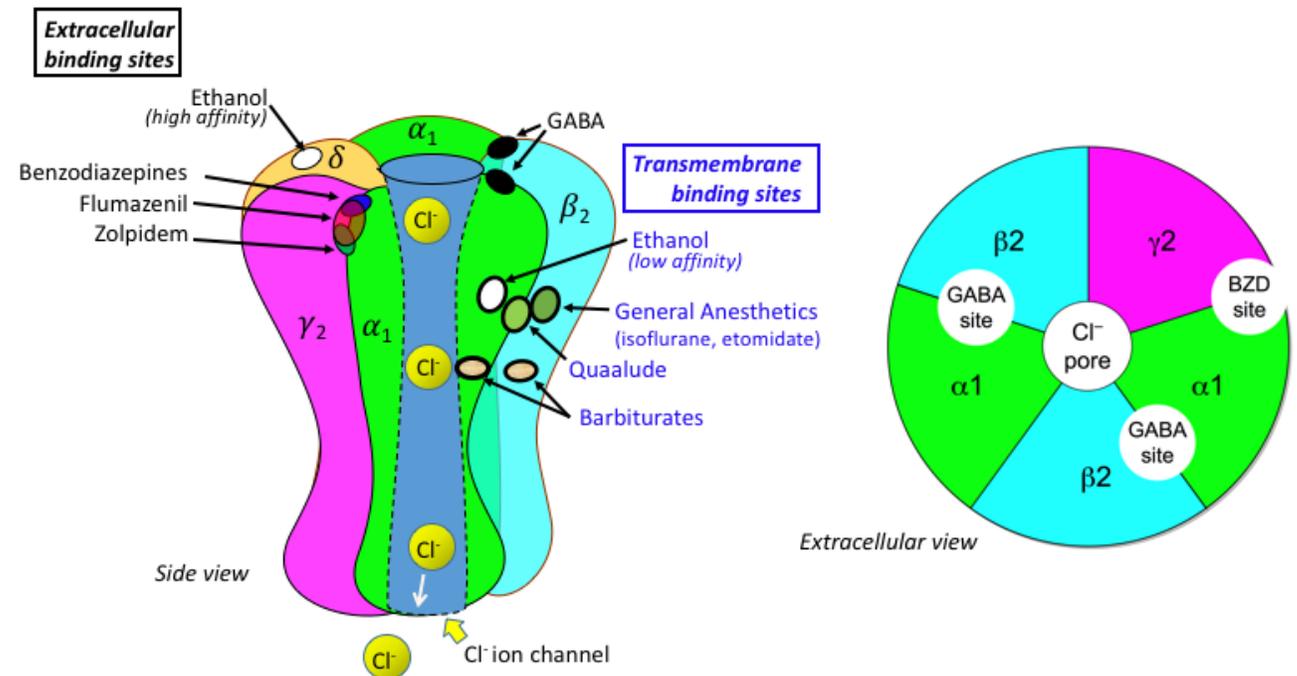
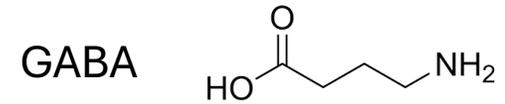


**Orthosteric site:** ligand binding site for receptor, substrate site for enzyme

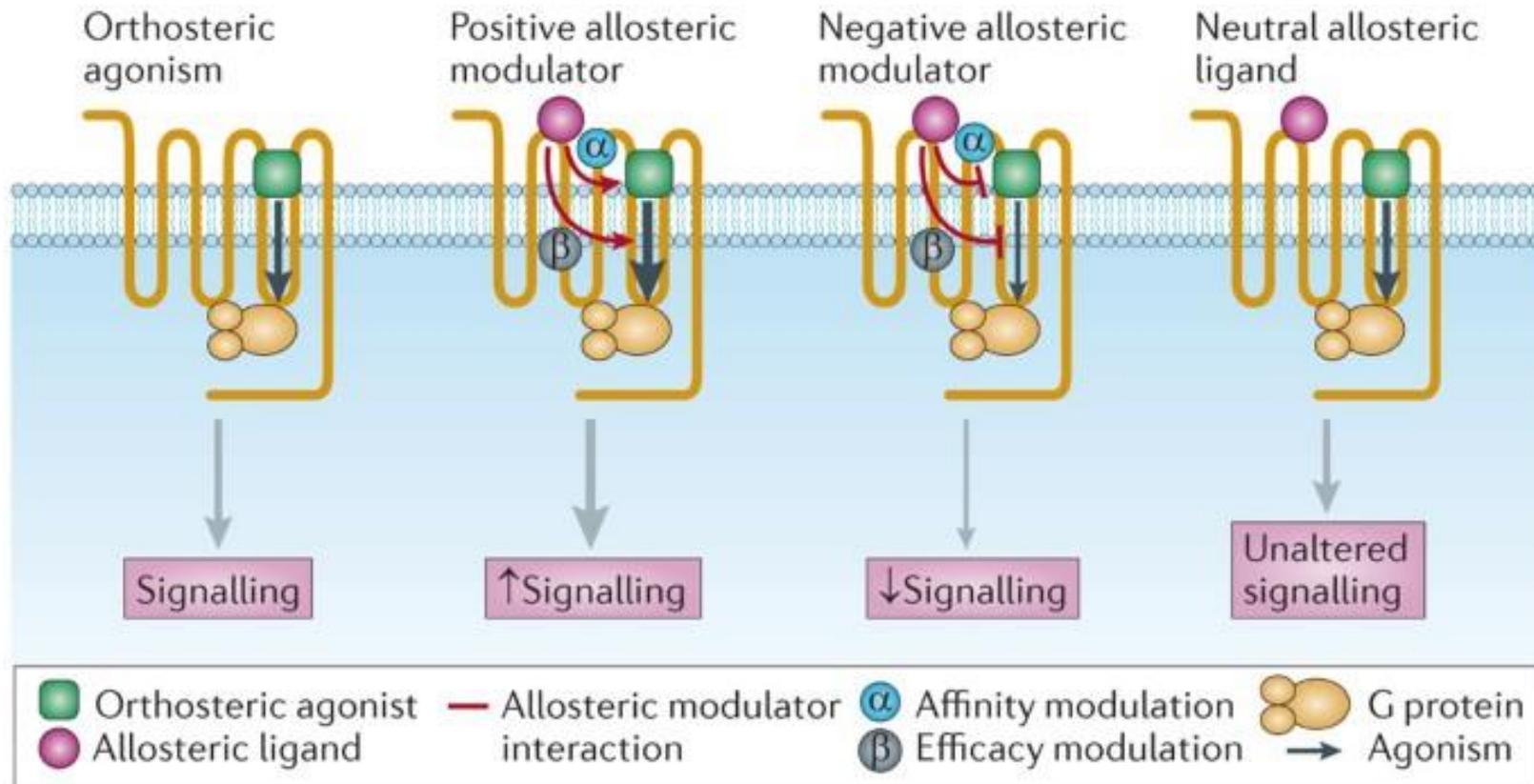
**Allosteric site:** site other than orthosteric site, causes shift in protein structure that indirectly influences protein function

# Allosteric sites are important in pharmacology

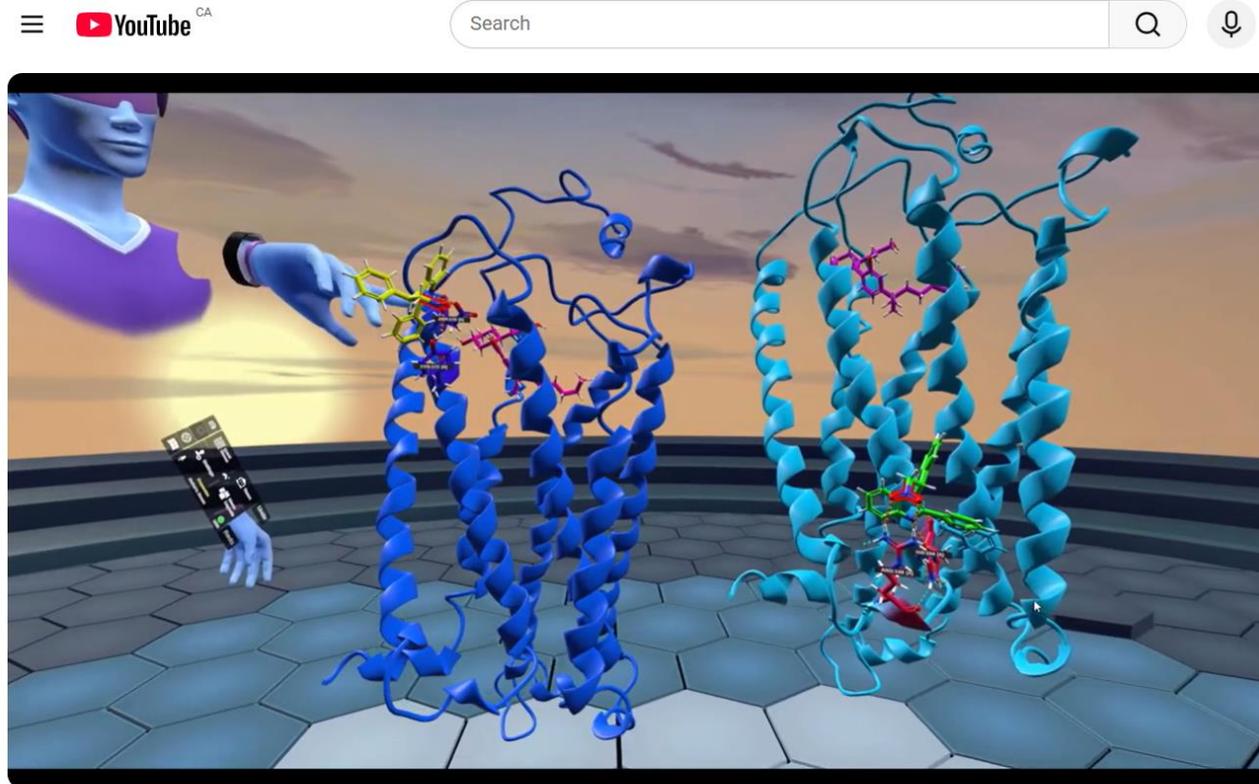
- Ligand-gated ion channel (GABA receptor)
- Endogenous ligand: GABA (neurotransmitter)
- Inhibitory neurotransmitter associated with mood
- Variety of allosteric sites with clinical impact
  - Benzodiazepines
  - Ethanol
  - Anesthetics



# Orthosteric and allosteric drugs: G protein-coupled receptors



# Exploring allosterics in Virtual Reality: Cannabinoid CB1 Receptor



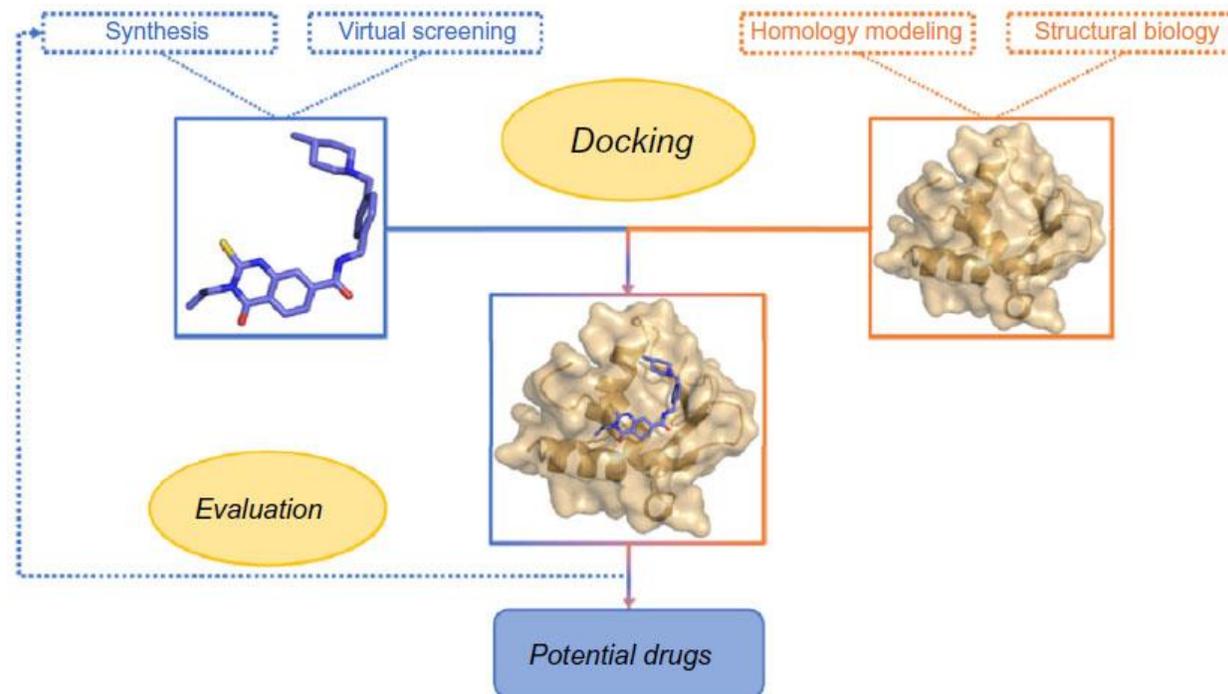
Nanome Virtual Reality - Allosteric Modulation of CB1 Receptor

- <https://www.youtube.com/watch?v=uVvDgSrvvYI>

# Modeling allosteric interactions

- Classical docking models assume stiff receptors
- Ensemble models: Implicit consideration of induced fit
  - Ensemble = set of all possible protein conformations
- Some models include partial consideration of flexibility
  - May consider flexibility of amino acid side chains but not the protein backbone
- “Blind Docking” location of the chemical’s binding site is not determined ahead of time
  - E.g. DiffDock
- Models that simulate molecular dynamics: true induced fit
  - Motion is accounted for
- Emerging models: jointly model the protein-ligand complex, allowing each component to move
- Computational cost ...

# Returning to drug screening: Iterative virtual – empirical cycles in drug discovery

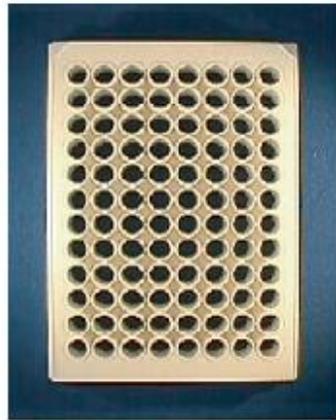


# Traditional High-Throughput Screening

Each well: a single chemical from the chemical library

Test compounds at a single concentration, usually 10  $\mu\text{M}$

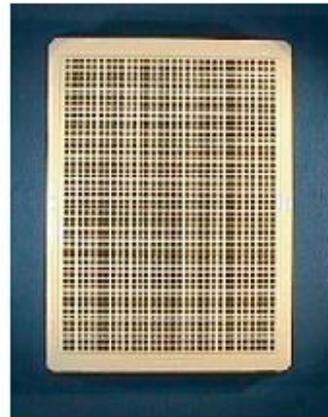
**96 wells**



**384 wells**



**1536 wells**



Compounds/day: 10.000  
Volume/well: 100-200  $\mu\text{l}$

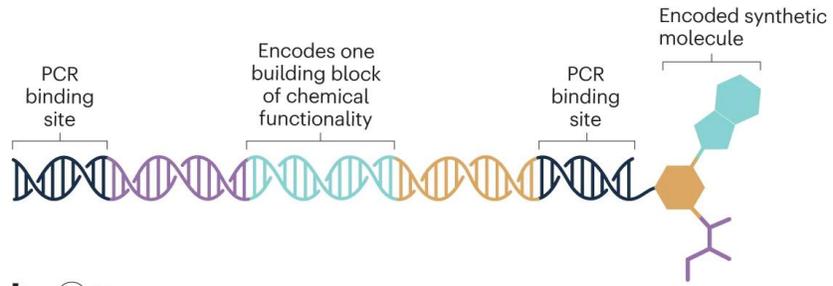
40.000  
30-100  $\mu\text{l}$

200.000  
2.5-10  $\mu\text{l}$

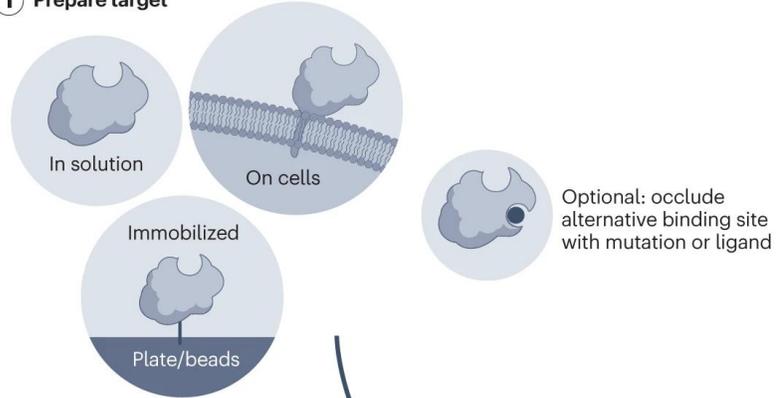
# Newer experimental libraries for hit-finding

- DNA-encoded libraries (DELs) have a DNA tag attached to each chemical
- Why DNA as a tag?
  - DNA is highly **specific** (a sequence of ~20 bases maps uniquely to a position in the 3-billion-base human genome)
  - DNA is highly **sensitive** – we can detect very tiny amounts of it.
    - Forensics, polymerase chain reaction (PCR) to amplify
  - Consider DNA tag as the “Read” operation.
- Advantage of DNA-tagged chemicals: pool the chemicals together
  - Miniaturization, much higher throughput

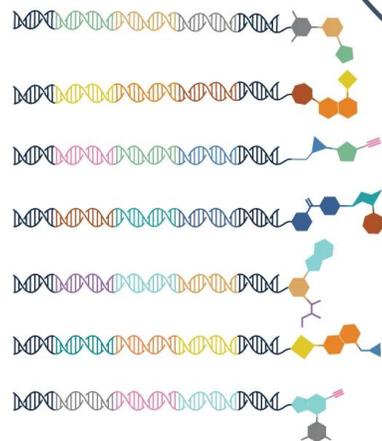
# DEL for pooled screening



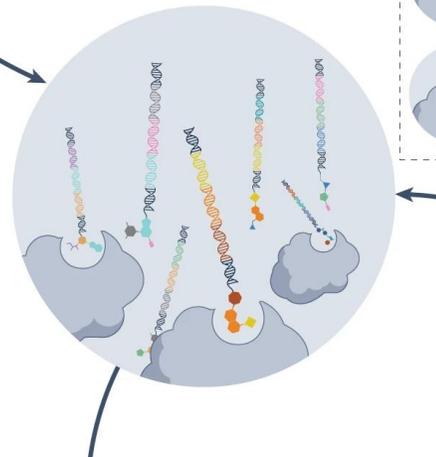
## b 1 Prepare target



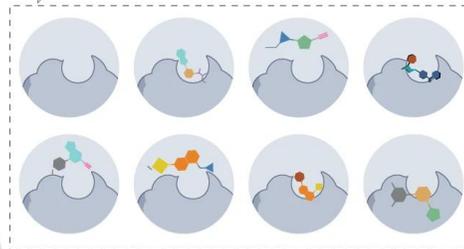
## 2 Add DEL



## 3 Perform pooled binding-based selection



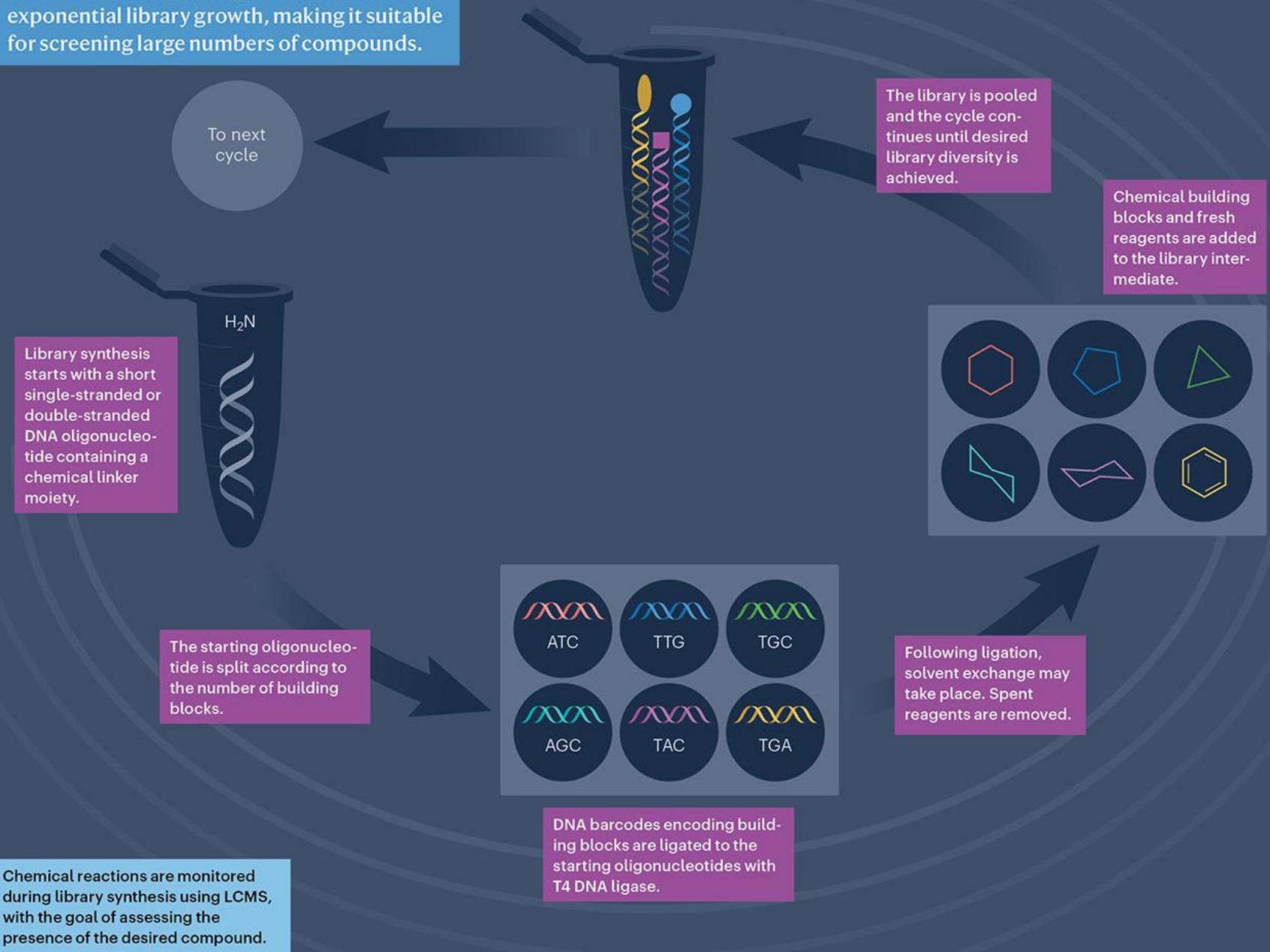
(Compare with traditional high-throughput screens — one well versus up to millions)



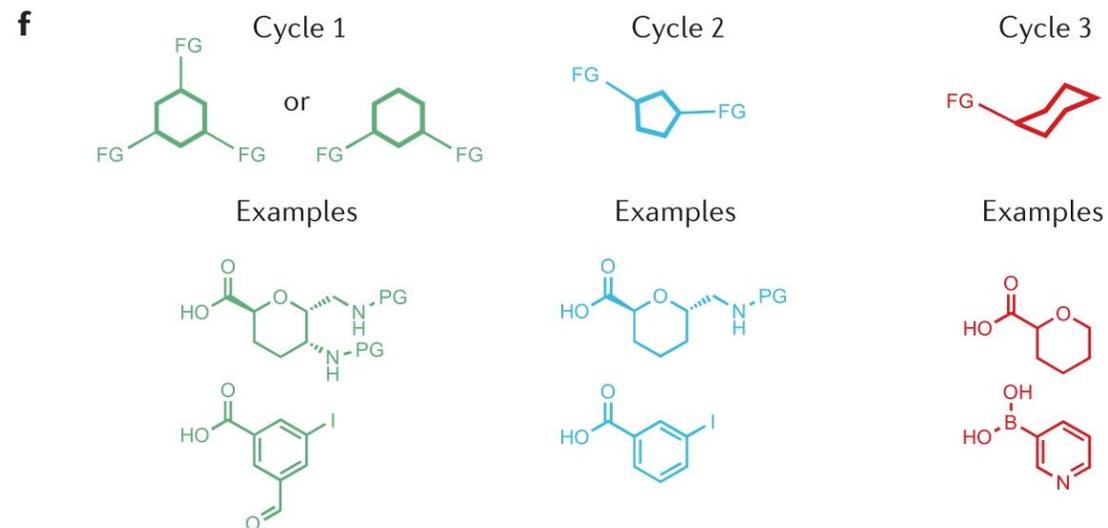
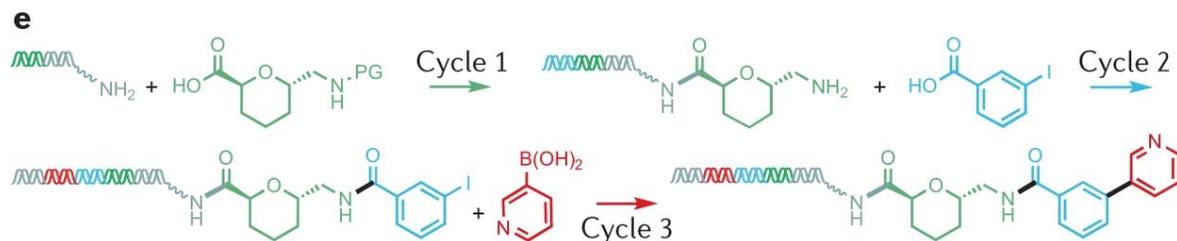
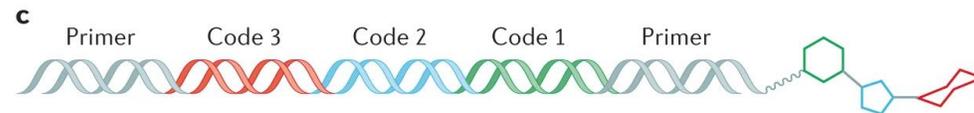
## Experimentation

DECLs are produced by split-and-pool combinatorial chemistry, which results in exponential library growth, making it suitable for screening large numbers of compounds.

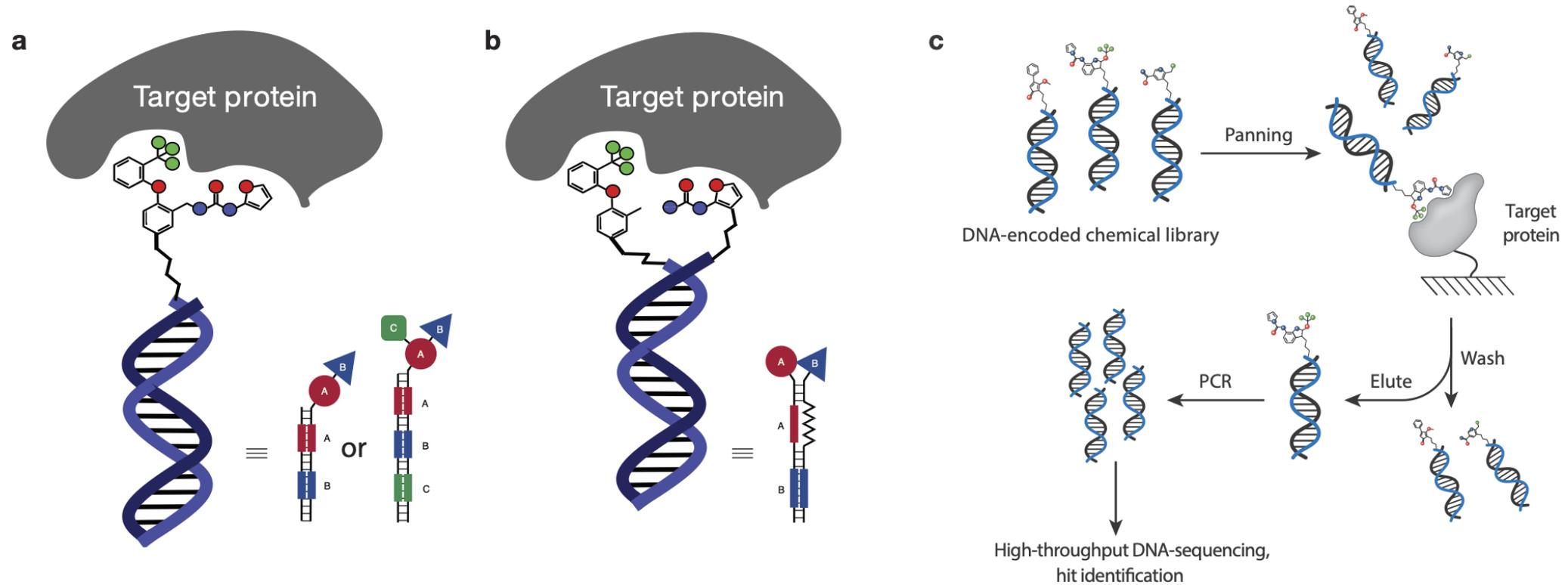
## Creating a DNA-encoded chemical library



# Combinatorial chemistry to create DNA-encoded chemical library



# Screening with DNA-encoded libraries (DELs)



# Screening with DNA-encoded libraries

- Tagging of small molecules (chemicals) with DNA “bar tags” allows for miniaturization of high-throughput screens
- Millions of DNA-tagged chemicals in a single 1 mL tube
- Sequence the DNA to identify the “hit” chemicals
- Requires computational approach based on **DNA sequences**, not small molecule chemistry

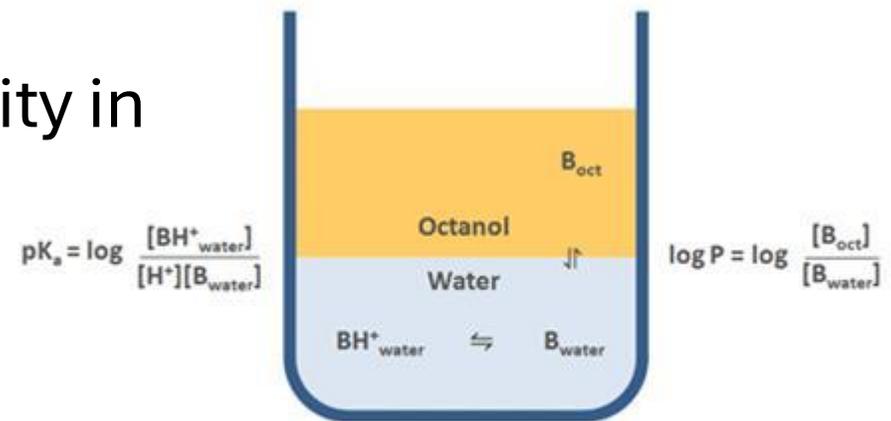
# Advantages of DNA-encoded libraries for drug screening

1. enormous library size (trillions of compounds) → much better coverage of chemical space
2. small physical space for compound storage
3. miniaturization of assays
4. low-cost tools for academic institutions and small pharmaceutical companies
5. track record of success in eventually producing drug-like compounds

Pursuing challenging drug targets:  
antibiotic drug discovery

# Chemical characteristics of drugs and compounds in chemical libraries

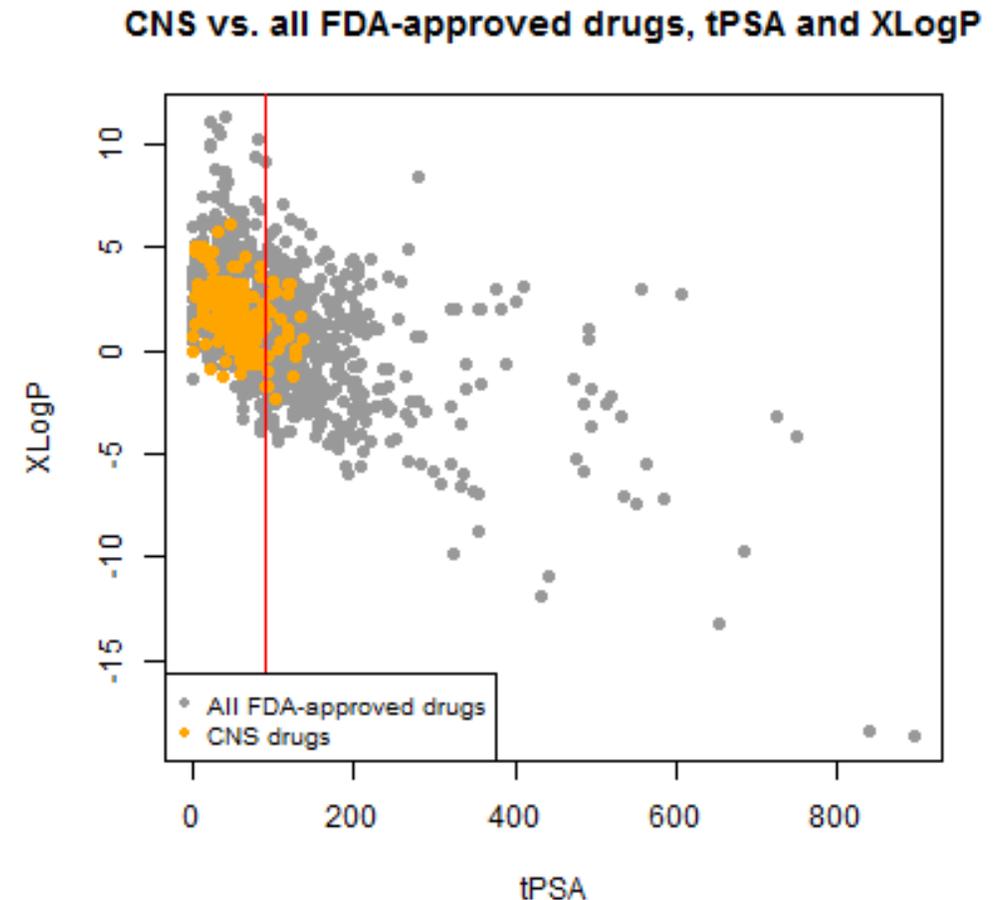
- Since human cells have membranes and hydrophilic regions, drugs need chemical characteristics of both fat solubility (lipophilicity=hydrophobicity) and water solubility (hydrophilicity)
- Partition coefficient: ratio of solubility in solvent vs water



$$P = \text{Partition Coefficient} = \frac{\text{Concentration of neutral species dissolved in partition solvent}}{\text{Concentration of neutral species dissolved in water}}$$

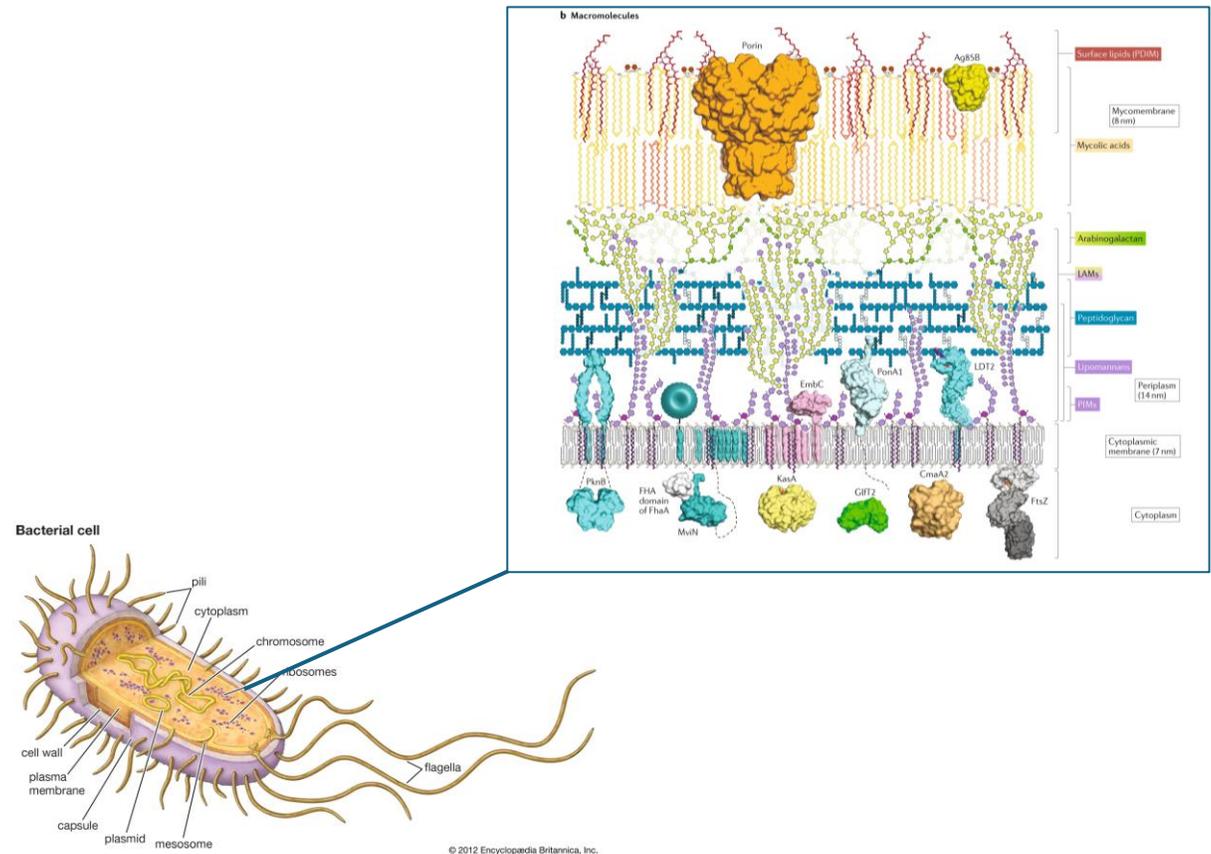
# Chemical libraries – choosing an appropriate subsection of chemical space

- Are the chemical characteristics of your library relevant to your clinical goal?
- CNS drugs occupy a subset of all FDA-approved drugs
- CNS drugs tend to be:
  - Fat-soluble (higher logP)
  - Smaller (lower polar surface area)

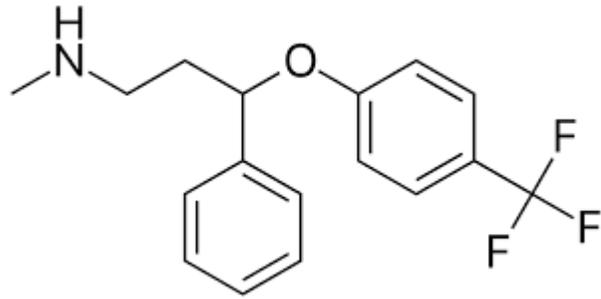


# Chemical libraries for Antibiotic drug discovery: different region of chemical space

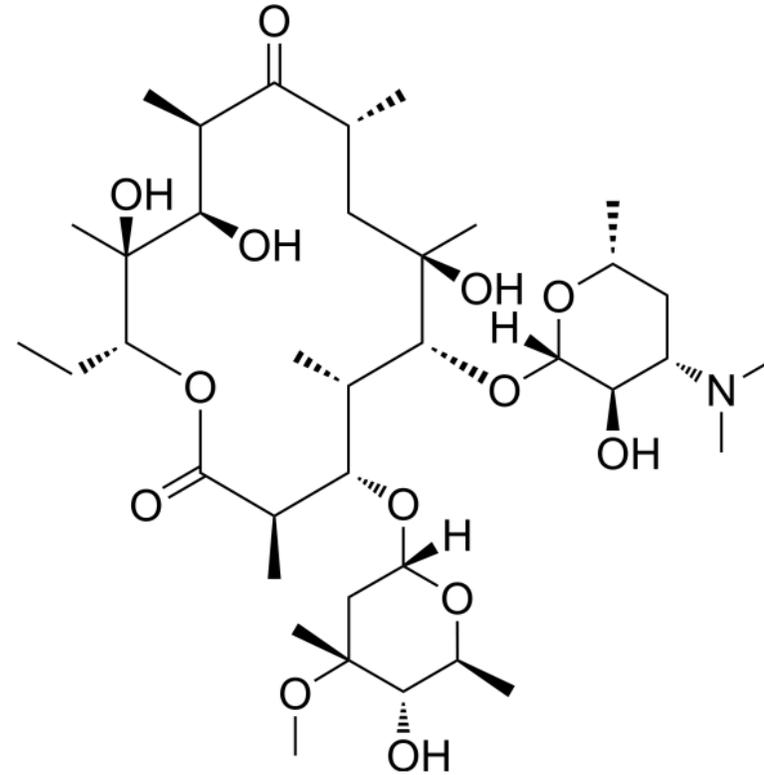
- Bacterial cell wall:
  - highly impermeable
- Relevant drug target may be inside the bacteria, so drug needs to cross the cell wall
- Tuberculosis (MTb)
  - Bacterial cell wall has the permeability of candle wax
  - Challenging for protein targets inside the bacteria



# Representative chemical structures

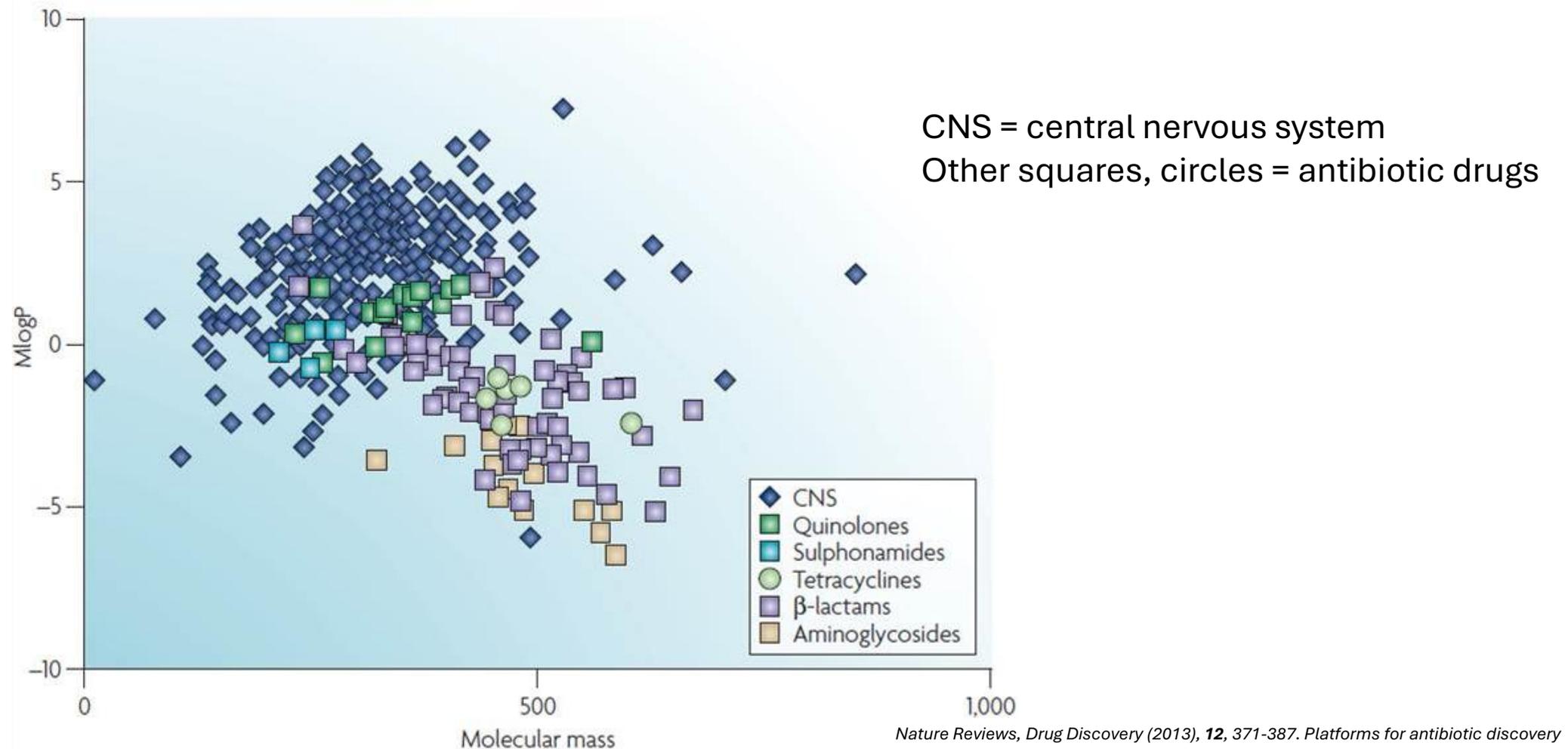


Antidepressant: fluoxetine

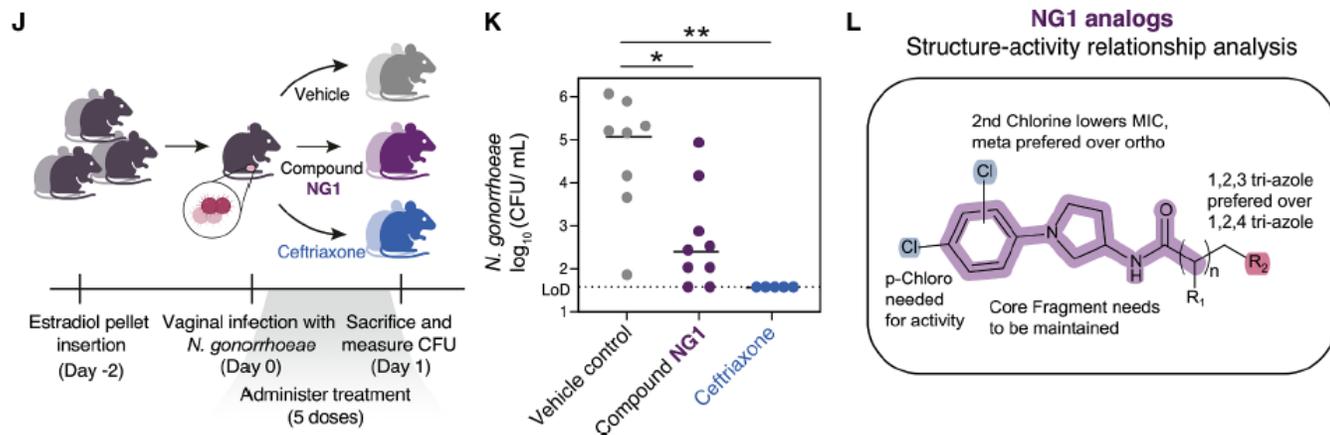
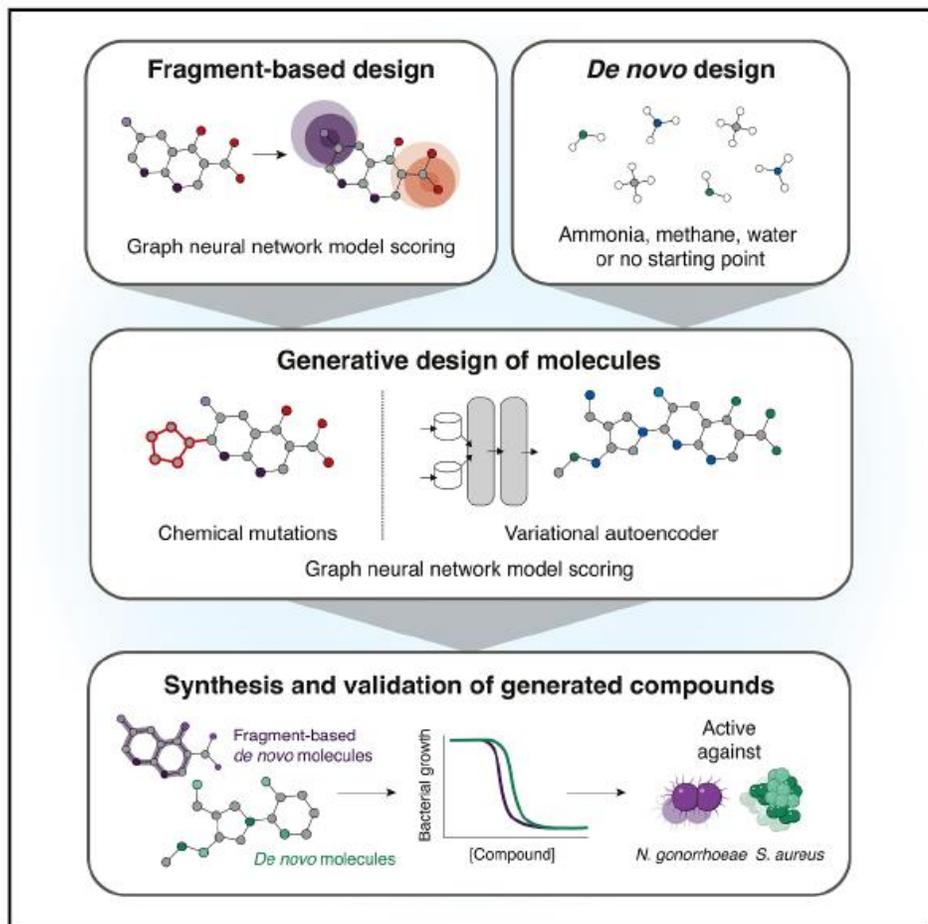


Antibiotic: erythromycin

# Depending on your goal, your compound likely inhabits a particular region of chemical space



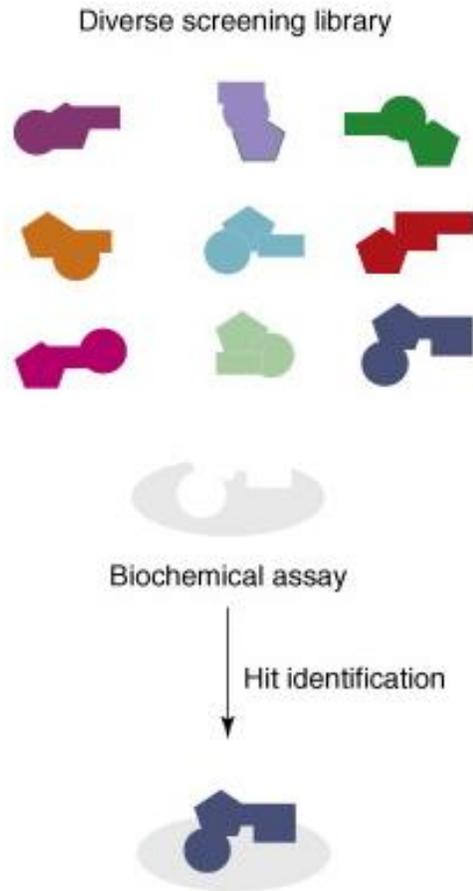
# Antibiotic drug design: generative deep learning



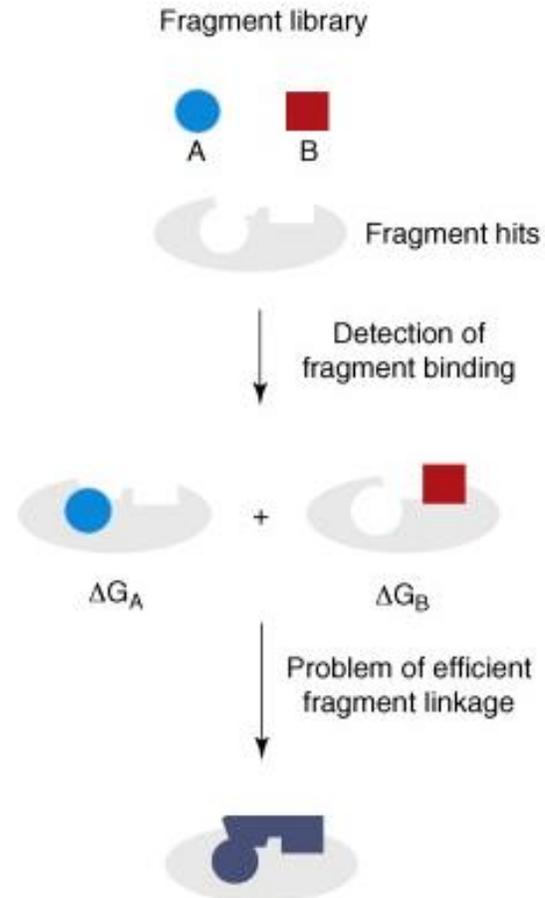
- Numerous research groups working on AI for antibiotics
- This particular paper leveraged a **fragment-based** approach
- Instead of “large” chemicals, this project worked with much smaller chemical fragments

# HTS vs Fragment-Based Drug Discovery

(a) High-throughput screening



(b) Fragment-based drug discovery



# Fragment-Based Drug Discovery

## Disadvantages

- Fragment-sized chemicals are weaker binders due to their small size
  - Fewer possible interactions possible with protein target
- Fragment binding is often non-specific
  - Bind to off-targets as well as desired targets
  - Fragments have the potential for being useless if they are too non-specific (too small)
- Need a biophysical assay for weak binding (instead of traditional HTS assays)
- Need to “build out” or join the fragments
  - After identifying fragment, there are more steps before you arrive at a testable compound

# Advantages of Fragment-Based Drug Discovery

- An option for challenging drug targets, especially when you don't have an appropriate chemical library for meaningful high-throughput screening (eg antibiotic drug discovery)
- FBDD libraries are smaller (fewer possibilities)
  - Estimated 166 billion possibilities for compounds with 17 heavy atoms
  - Compare to  $10^{63}$  possibilities for “drug-like small molecules”
- Screening a fragment library covers a much greater fraction of the relevant chemical space

# Advantages of Fragment-Based Drug Discovery

- Chances of **novelty** are much greater
  - Even from the same hit against the same target, no two companies are ever likely to end up with the same compound
  - Patents require intellectual property – require novelty
- Optimal properties can be built in at each stage of expansion
  - Reduced lipophilicity giving fewer PK/tox issues
- Selectivity can be built in from an early stage
- Potentially allows far greater chance of finding the rare compound which will hit difficult targets
- Avoidance of complexity reduces chances of unfavourable interactions
  - As complexity increases probability of matching a binding site actually decreases
- Commercially-available building blocks and availability of analogues can allow for rapid elucidation of SAR and hit expansion

Drug screening: Beware of artefacts

# PAINS compounds: false positive “hits” in numerous drug screens

- If you identify these chemical classes as “hits” in your screen and announce them as true hits → naive

## WORST OFFENDERS

Pan-assay interference compounds (PAINS) fall into hundreds of chemical classes, but some groups occur much more frequently than others. Among the most insidious are the eight shown here (reactive portions shown in red and purple). These and related compounds should set off alarm bells if they show up as ‘hits’ in drug screens.

**TOXOFLAVIN**  
Redox cycler: can produce hydrogen peroxide, which can activate or inactivate different proteins.

**ISOTHIAZOLONES**  
Covalent modifier: reacts chemically with proteins in non-specific, non-drug-like ways.

**CURCUMIN**  
Covalent modifier, membrane disruptor: muddles response of membrane receptors.

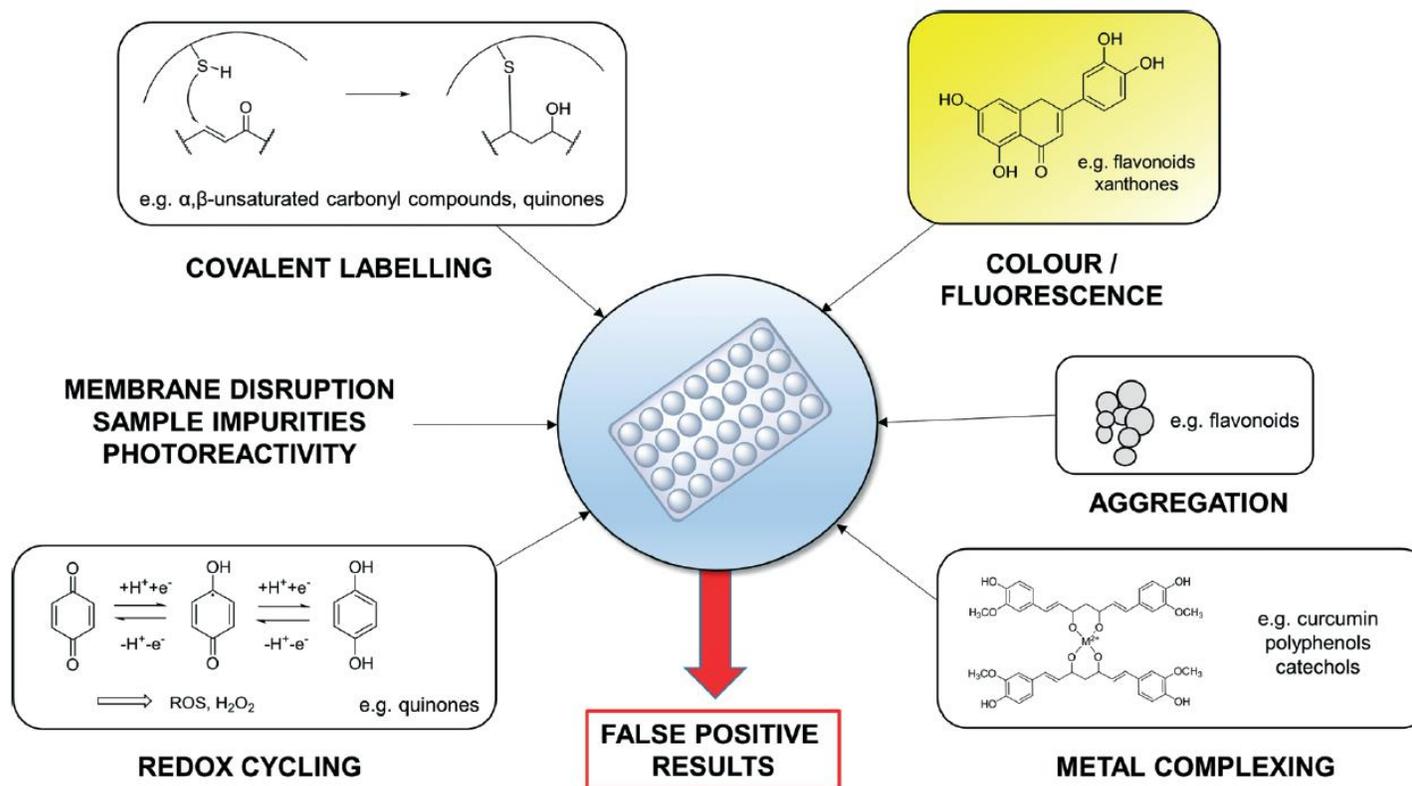
**HYDROXYPHENYL HYDRAZONES**  
Covalent modifier, metal complexer: sequesters metal ions that inactivate proteins.

**ENE-RHODANINE**  
Covalent modifier, metal complexer.

**PHENOL-SULPHONAMIDES**  
Redox cycler, covalent modifier, unstable compound: breaks down into molecules that give false signals.

# Pan-Assay interference compounds – variety of mechanisms of non-specificity

- Covalent modifiers – a frequent mode of artefacts, others possible



# How to deal with PAINS compounds?

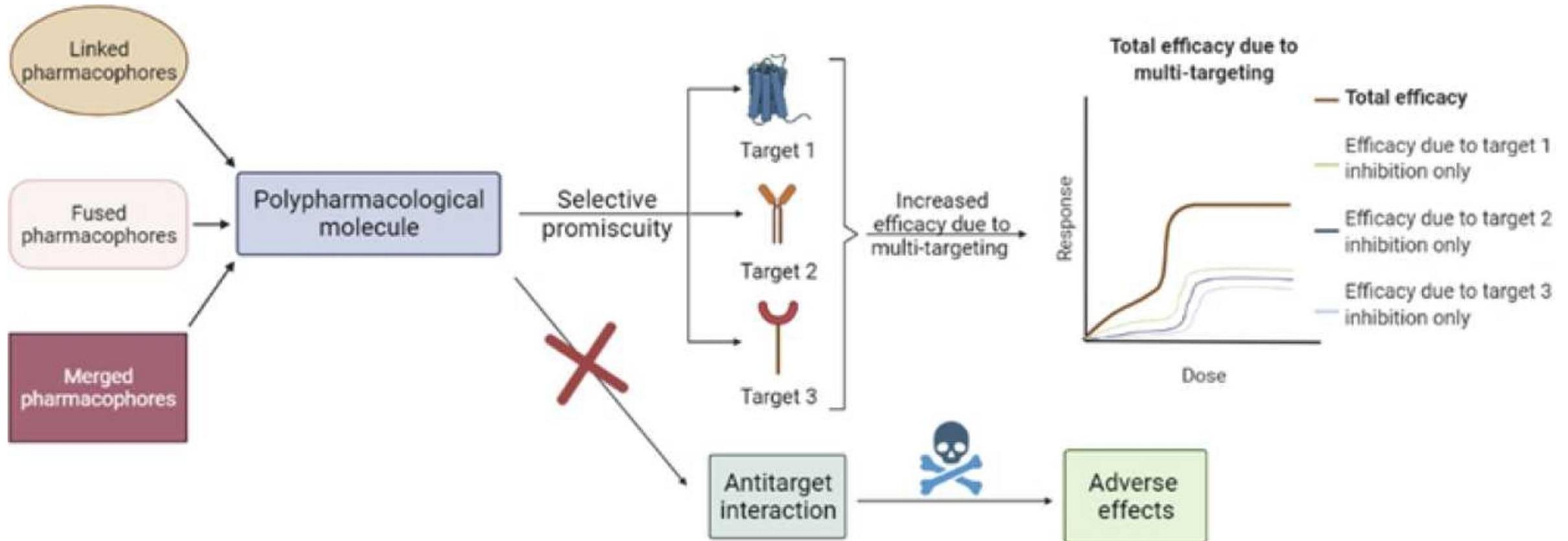
- Remove the PAINS “hits” from a screen data prior to moving forward.
  - May require cross-comparison from other screens since by definition these compounds show up in many different assays
- Don’t train your model on PAINS...
- Model may take problematic shortcuts when identifying “binders”
- Analogy: neural networks trained to find tanks – or clouds?

# Beyond single drug targets

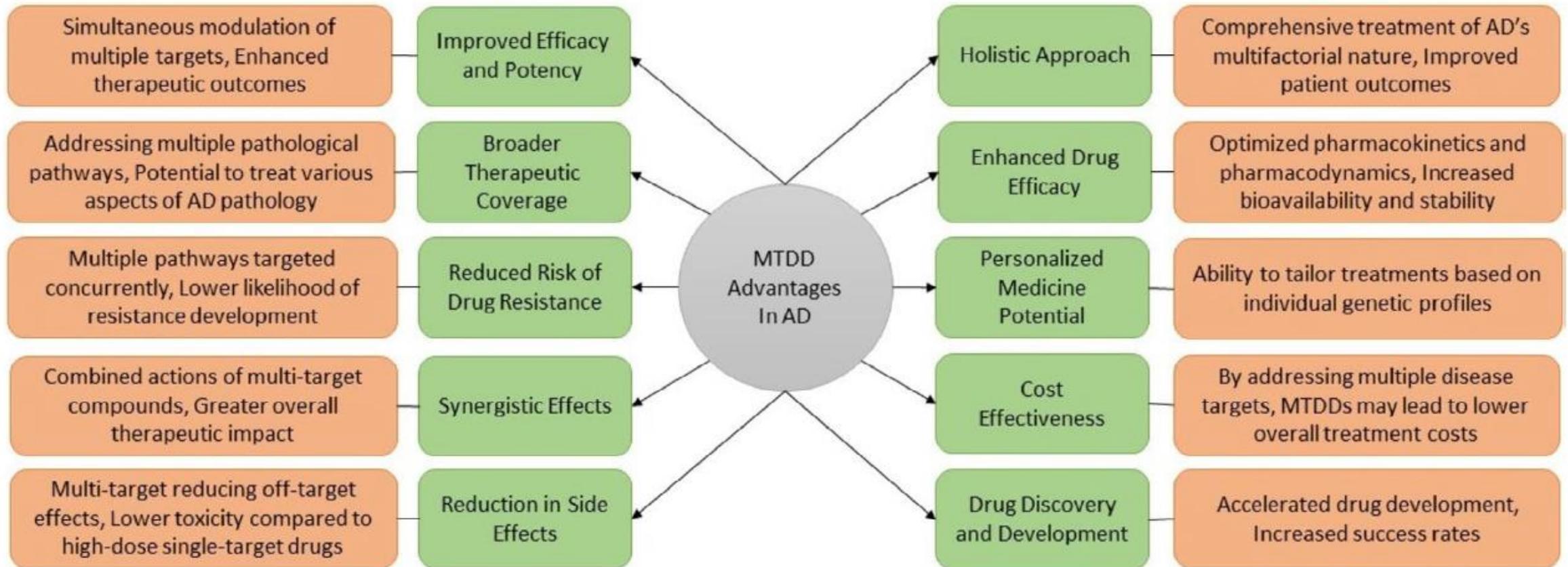
One drug → one target → clinical efficacy is an over-simplification

# Drugs with multiple drug targets

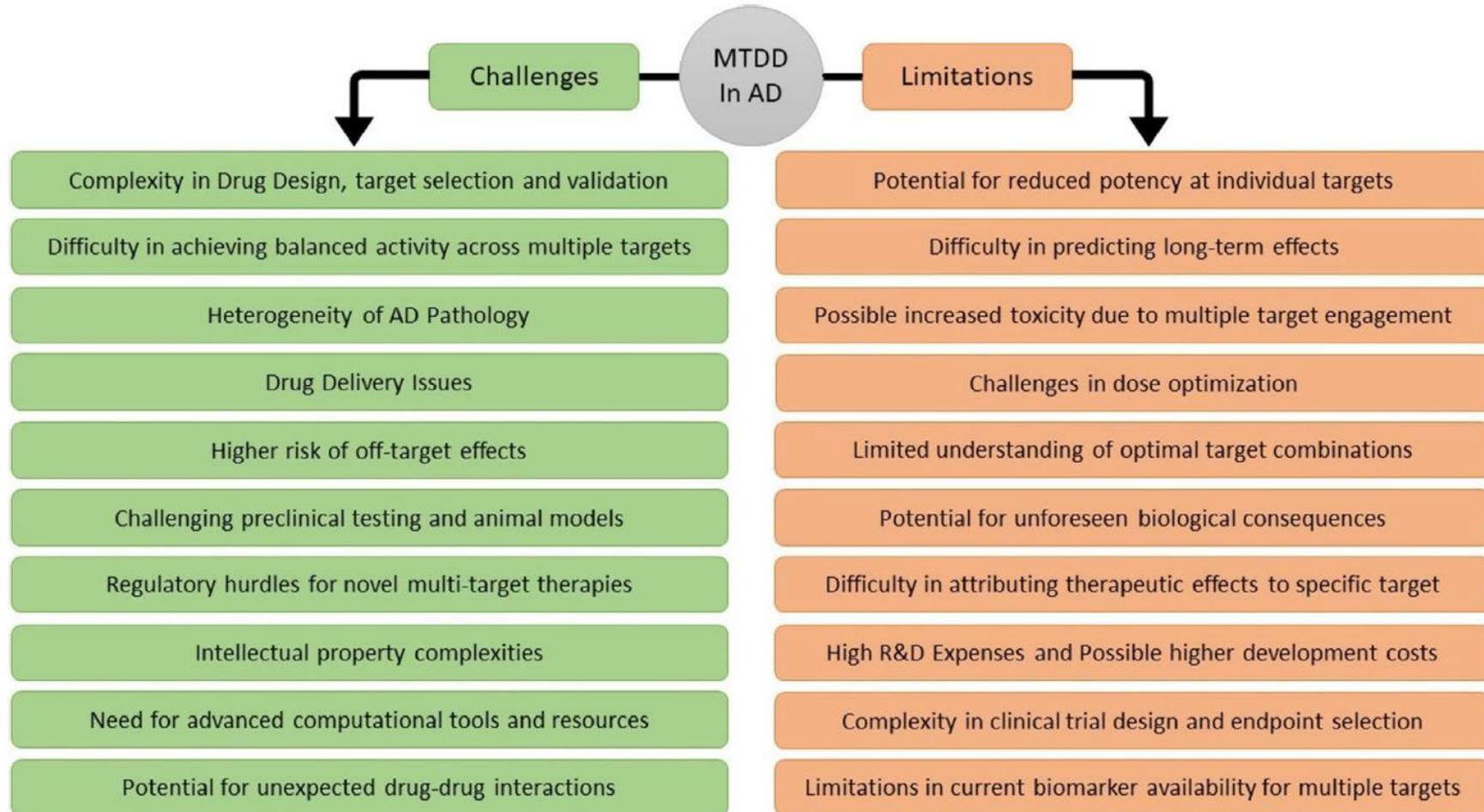
- Clinical utility and efficacy



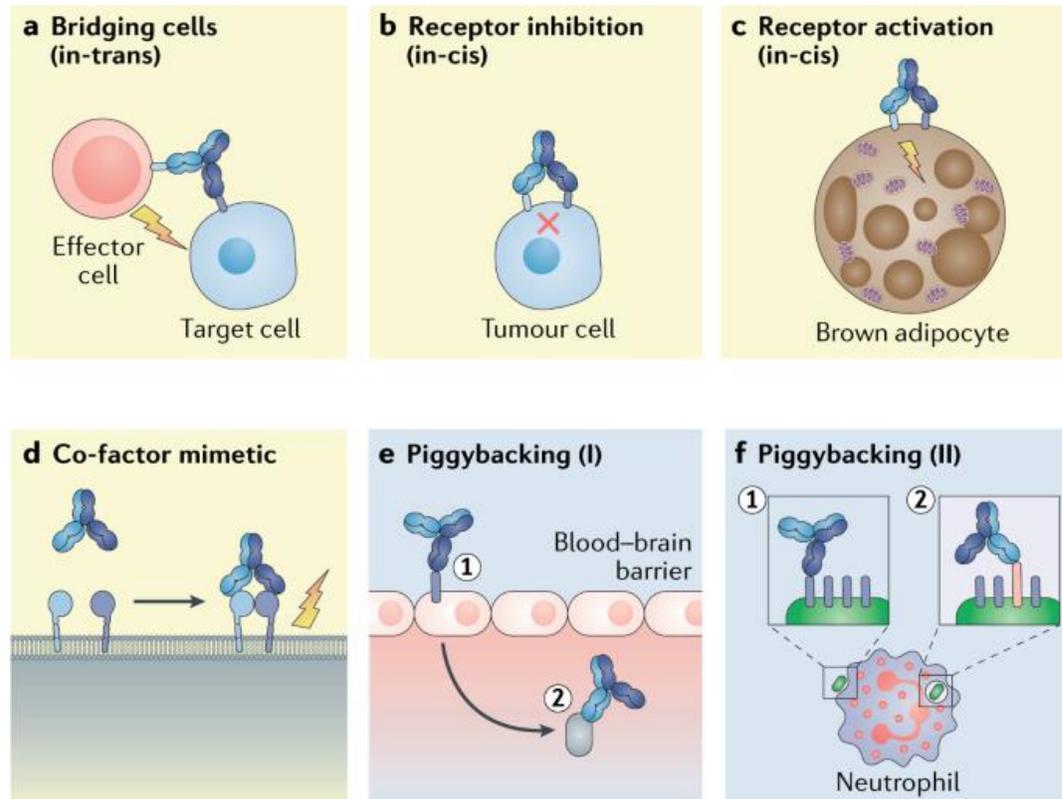
# Advantages of Multi-target drug development in Alzheimer's Disease



# Challenges and Limitations of Multi-target drug development in Alzheimer's Disease



# Bispecific antibody drugs: two intended targets



# Takeaways

- High-performance AI/ML models should address both orthosteric and allosteric interactions
- There is an opportunity for subject matter-informed AI/ML in the antibiotic space. Interrogate appropriate regions of chemical space
- Strategies to improve coverage of chemical space:
  - DNA encoded libraries
  - fragment-based drug discovery or DNA-encoded libraries
- Beware of Pan-assay interference compounds. Don't be naïve and call these “hits”; remove them from your models
- Beyond single-target drugs. Clinical efficacy of multi-target drug development (intentionally targeting  $> 1$  protein)