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# Insights from a **Biotech Startup**: How **Computational Chemists** Shape the Design of **Proximity-Inducing Compounds**

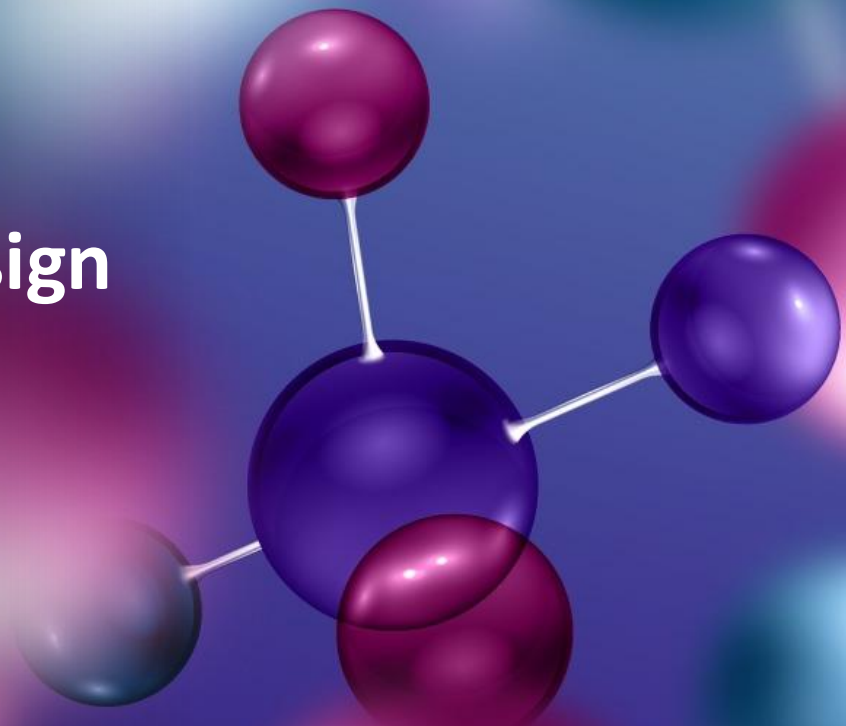
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**Alzbeta Tuerkova, Ph.D.**

Principal Computational Chemist

29<sup>th</sup> January 2024



# AGENDA

<b>Introduction to Targeted Protein Degradation</b>	<b>3</b>
<b>Ternary Complex Prediction (TCP)</b>	<b>9</b>
- Challenges	9
- Application of TCP in Drug Discovery Pipelines	13
- Methodology (AI-based vs Physics-based)	18
- Validation of the Approach	25
<b>Linker Generation</b>	<b>34</b>
<b>The Journey of Computational Chemists in Biotech</b>	<b>44</b>
- Work model in Academia vs. Industry	45
- From academia to industry: Transferable Skills	46
- Job Interview Process	47
- Two Types of Computational Chemists	48

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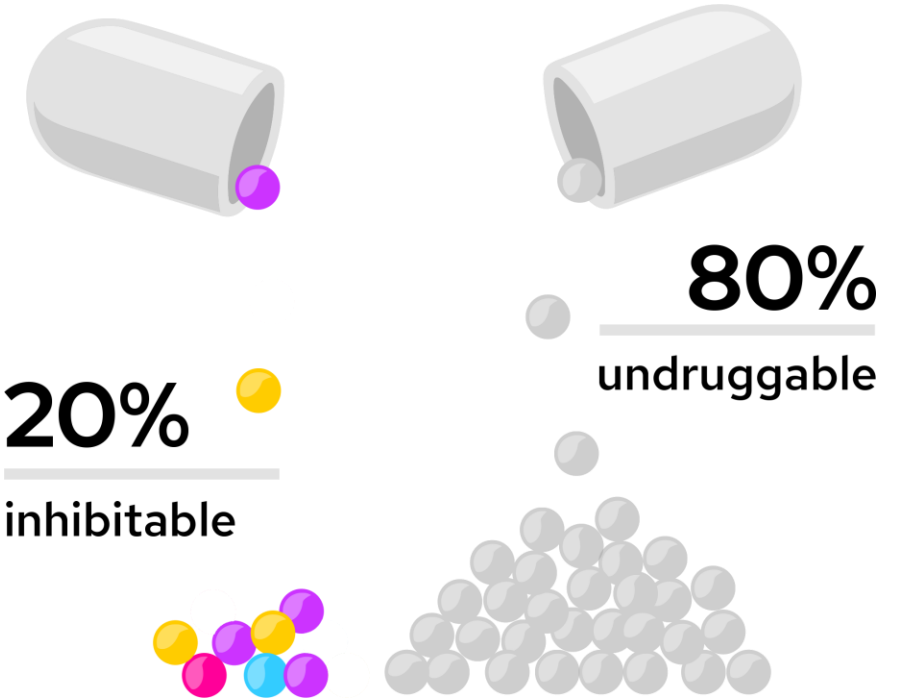
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# The Problem of Undruggable Targets

Most traditional small-molecule drugs deactivate functions of proteins

The typical mechanism of action is to target the active site, resulting in **inhibition**

Only **10-20 %**<sup>1</sup> of all pathogenic proteins in the human proteome possess such sites



[1] DANG, Chi V., et al. Drugging the 'undruggable' cancer targets. *Nature Reviews Cancer*, 2017, 17.8: 502-508.

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# From Protein Inhibition to Degradation



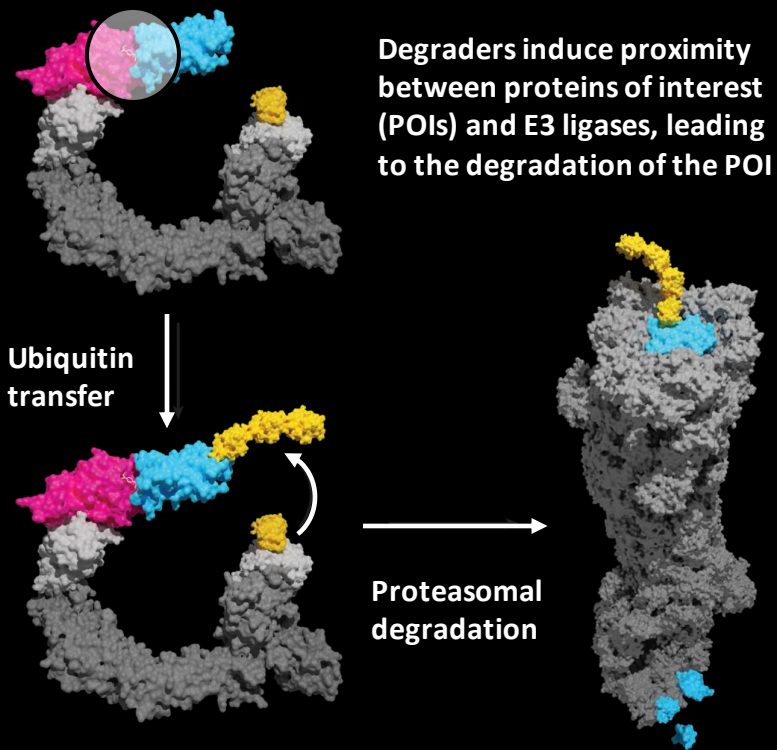
## Targeted Protein Degradation

- Humans have different systems to remove dysfunctional/misfolded proteins.
- One dominant system is the intracellular ubiquitin-proteasome complex.
- Degradors are drugs that take advantage of the ubiquitin-proteasome pathway.
- We can hijack these systems to tag specific proteins.

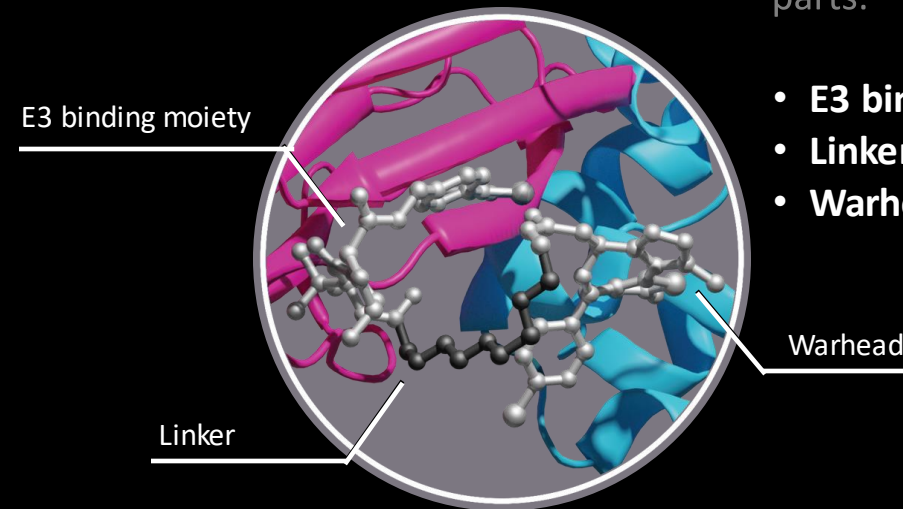
# What is a PIC™ Degradator?

**Proximity-inducing Compound (PIC™)** Degradators are the next generation of highly potent, rationally designed degradators that feature **E3 ligase selection** based on **protein-protein interfaces**.

## TARGETED PROTEIN DEGRADATION PATHWAY



## BIVALENT DEGRADER MOLECULES



A traditional bivalent degrader consists of three parts:

- E3 binding moiety
- Linker region
- Warhead (POI ligand)



# Advantages of Targeted Protein Degradation



	CRISPR-CAS9	RNAi	Traditional small molecules	Degraders
Stability	X	X	✓	✓
Going Beyond Druggables	✓	✓	X	✓
Working Directly on Proteins	X	X	✓	✓
Clinical Validation	✓	✓	✓	✓
Oral Application	X	X	✓	✓
Low Dosage / Tox	X	X	X	✓

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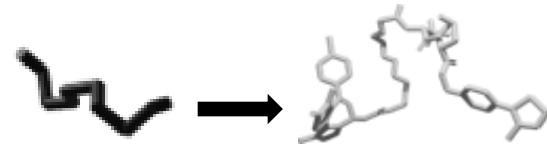
# The 'Success' in Designing PICs is Multi-factorial

**Degradation** = Solubility x PPB\* x Permeability x ([Complex] & Ubiquitination) x POI synthesis rate

\*) Plasma Protein Binding

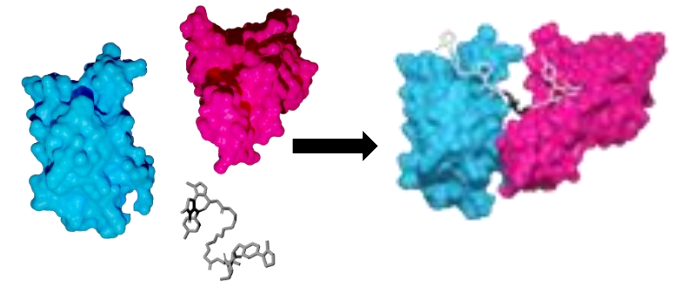


# The 'Success' in Designing PICs is Multi-factorial



Linker generation

Ternary  
Complex  
prediction



Degradation = Solubility x PPB x Permeability x ([Complex] & Ubiquitination) x POI synthesis rate

Distance  
measurements

# AGENDA

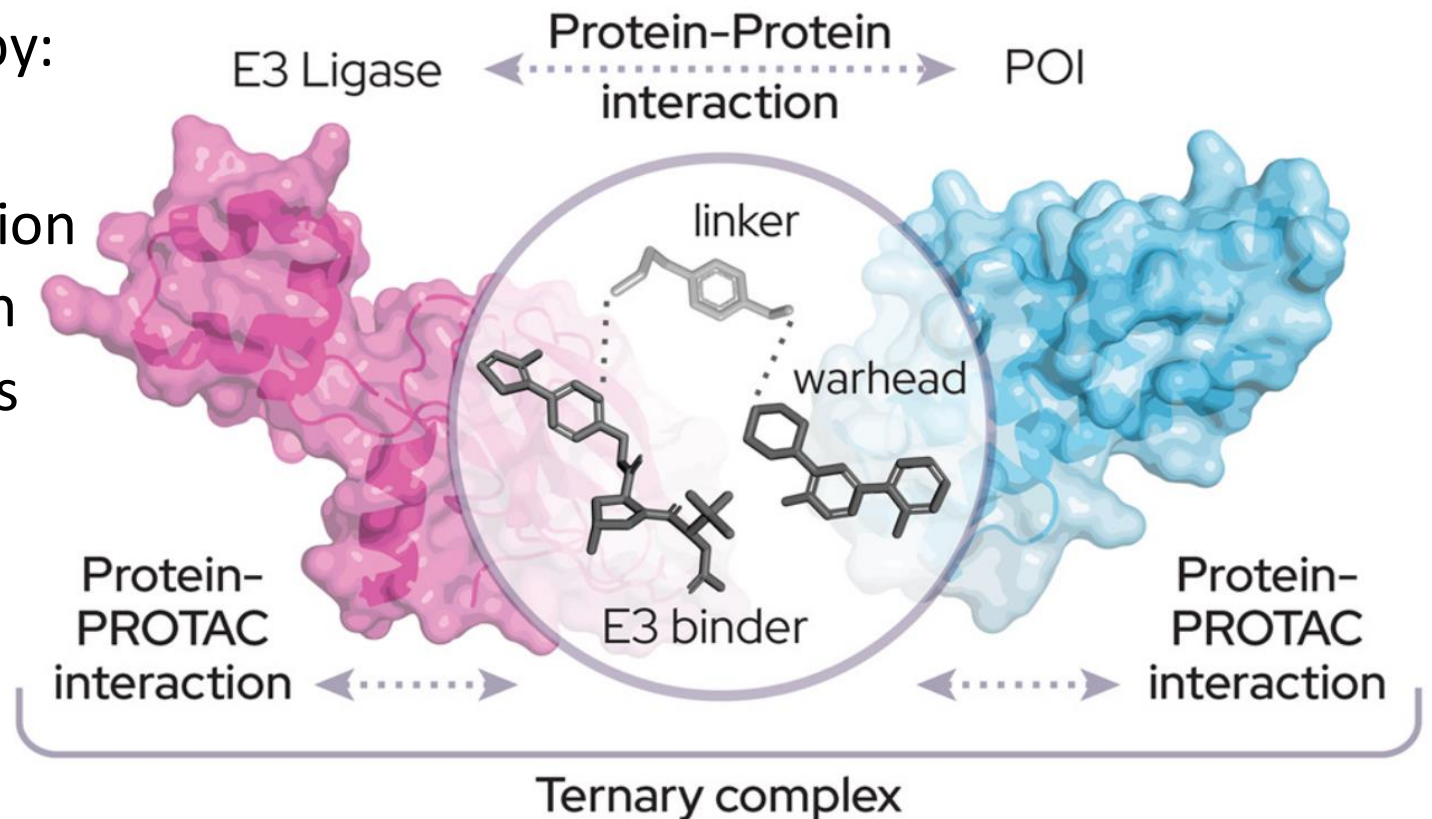
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# Ternary Complex Prediction: Challenges

## 1) Prediction of key interactions

Ternary complexes are mediated by:

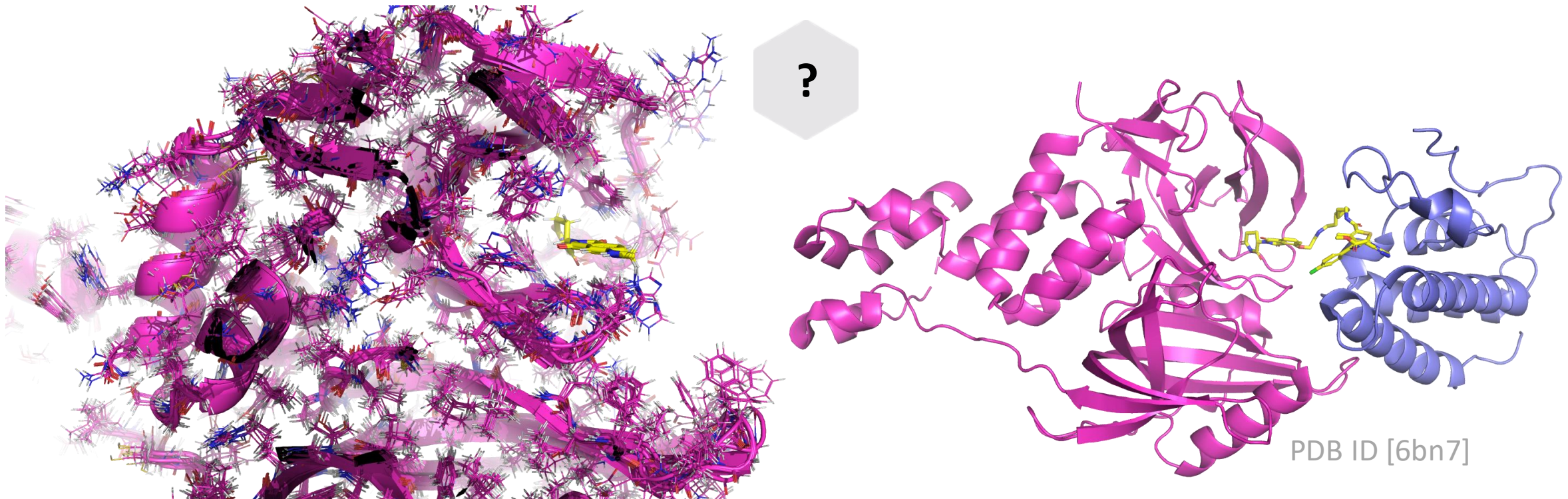
1. **Fragment-Protein** Interaction
2. **Protein-Protein** Interaction
3. **Linker** enforced constraints



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# Ternary Complex Prediction: Challenges

## 2) Accounting for protein conformational flexibility



"unbound" structures (CRBN)

"bound" structure

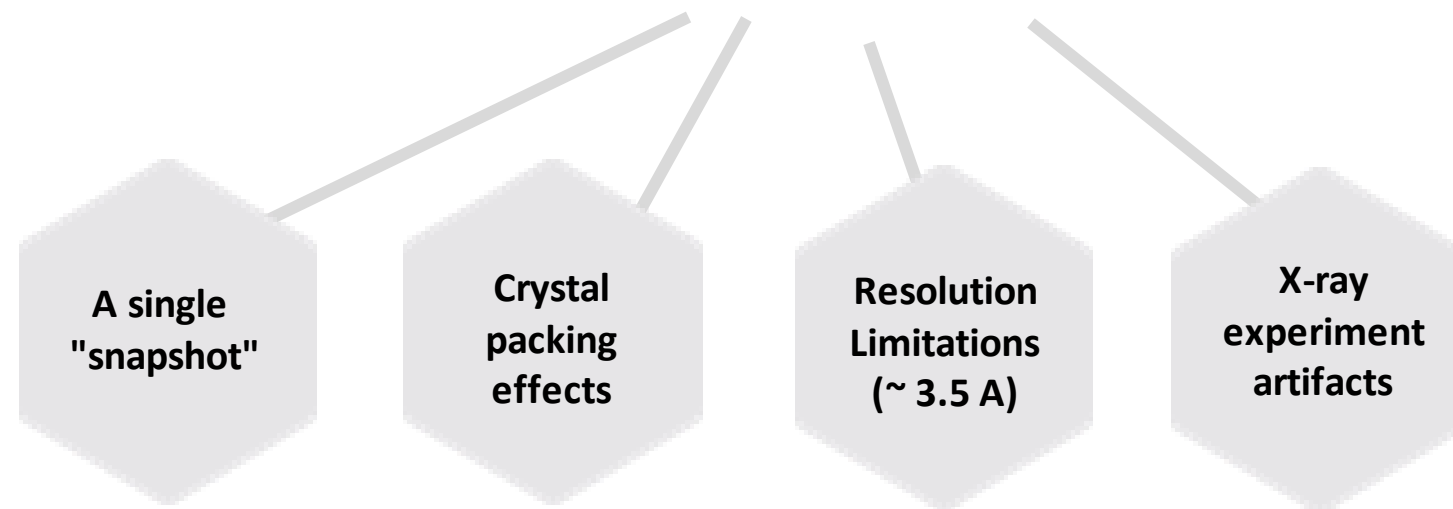
PDB ID [6bn7]

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# Ternary Complex Prediction: Challenges

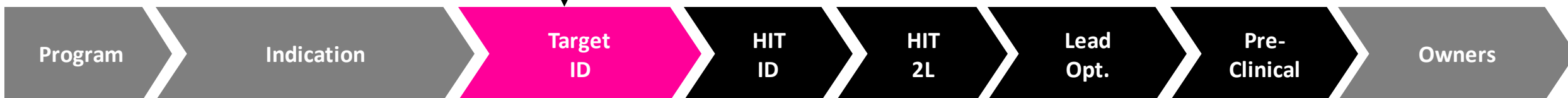
## 3) Validation of the approach

- **Limited number** of experimentally determined ternary complex structures (n=22)
- All available structures originate from **X-ray measurements**



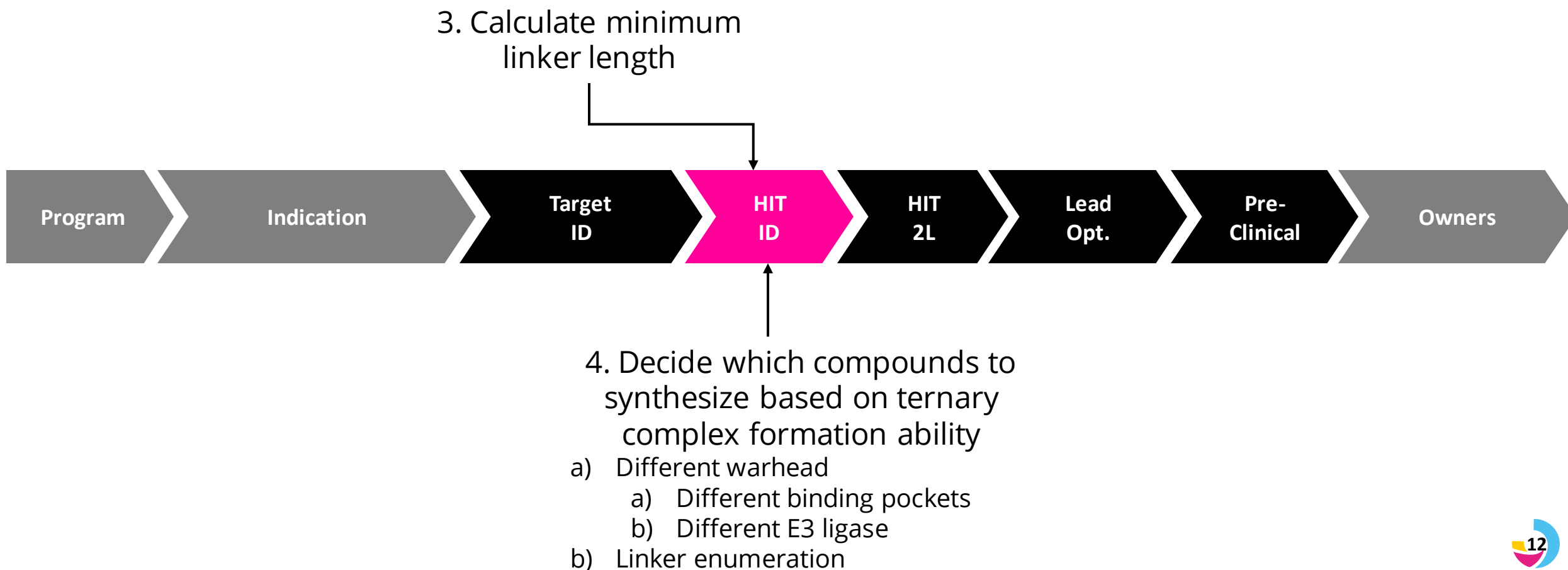
# Application of TCP in PIC design

2. Target selection based on selectivity prediction

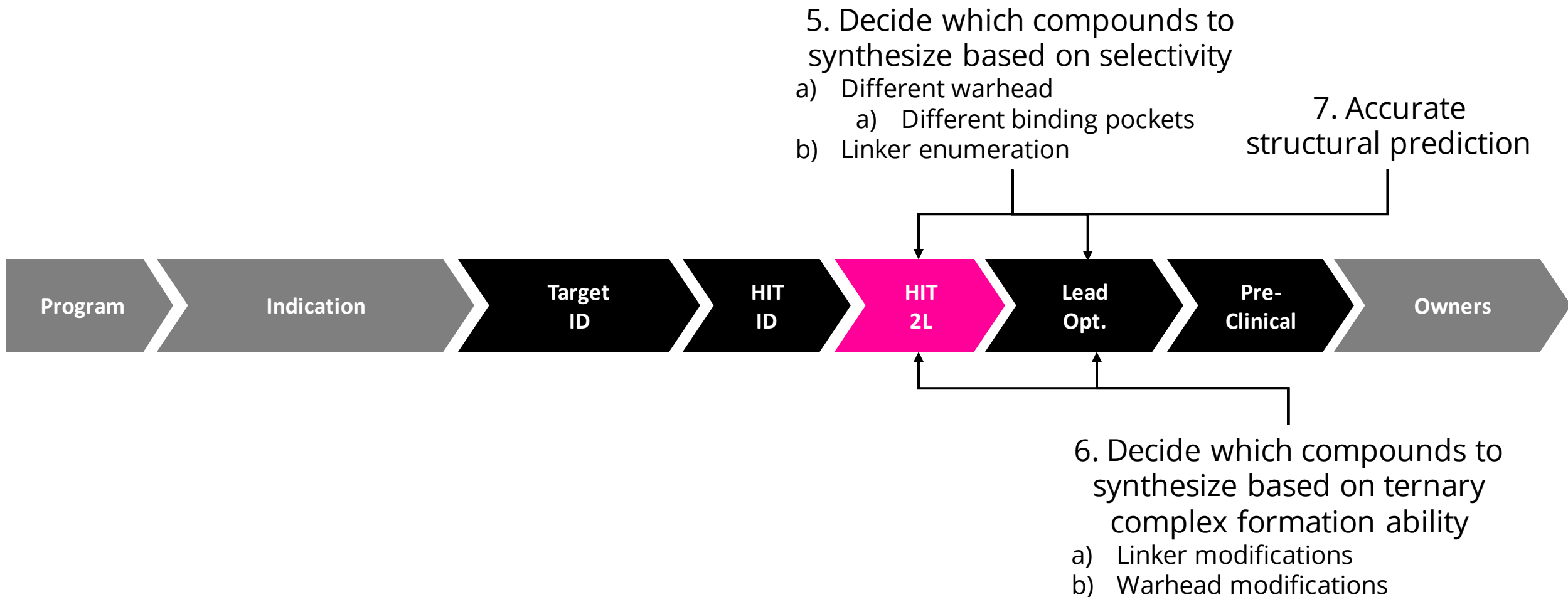


1. Target selection based on feasibility on developing ternary complex

# Application of TCP in PIC design

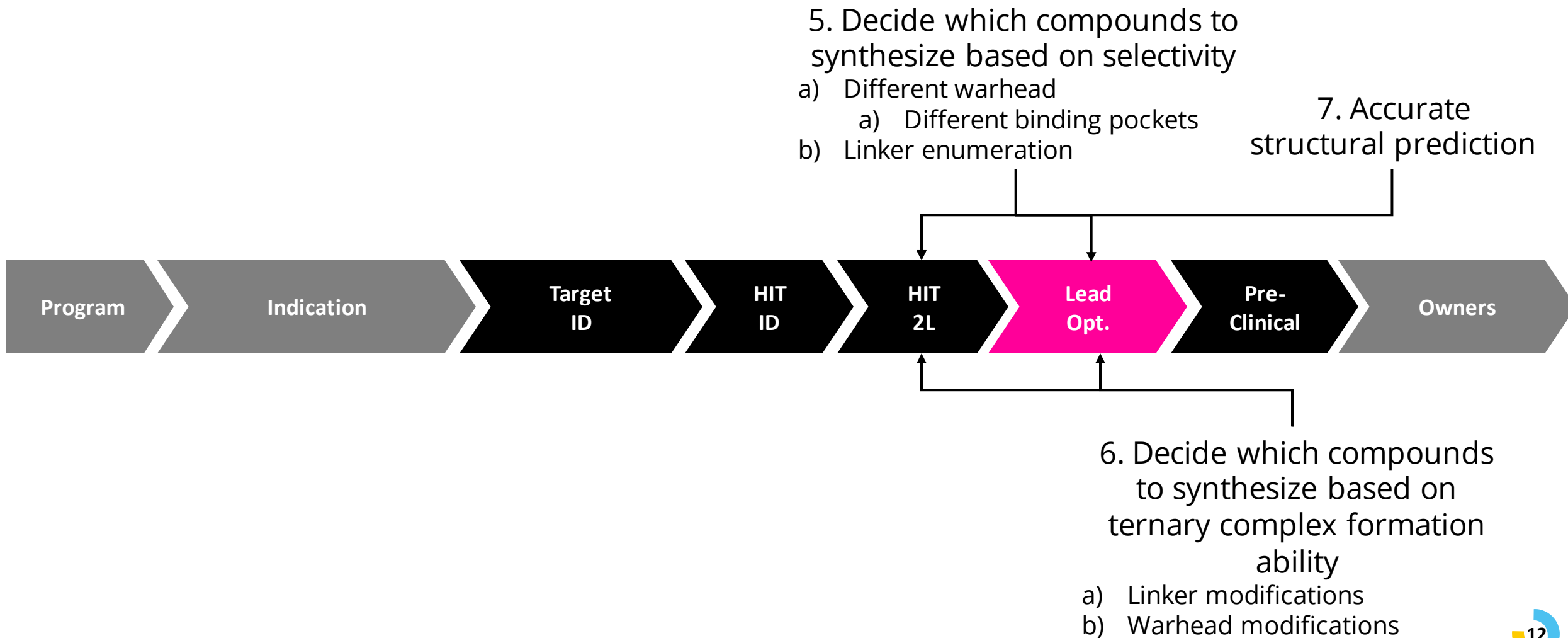


# Application of TCP in PIC design

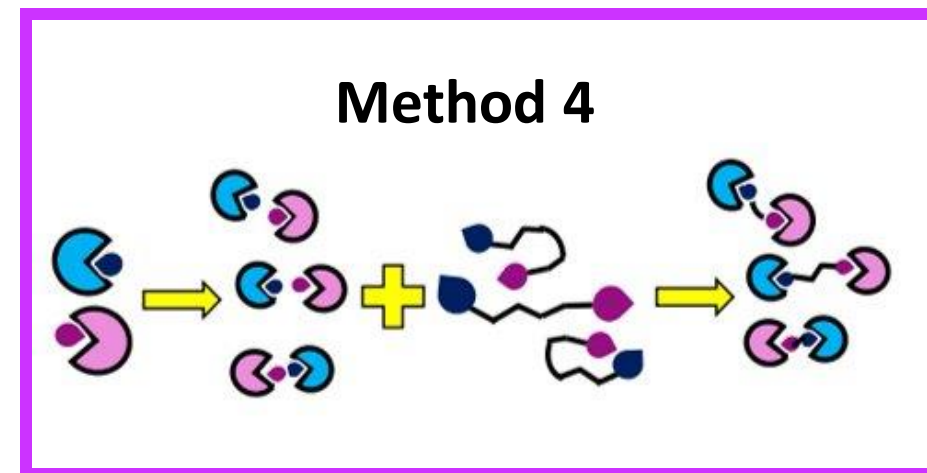
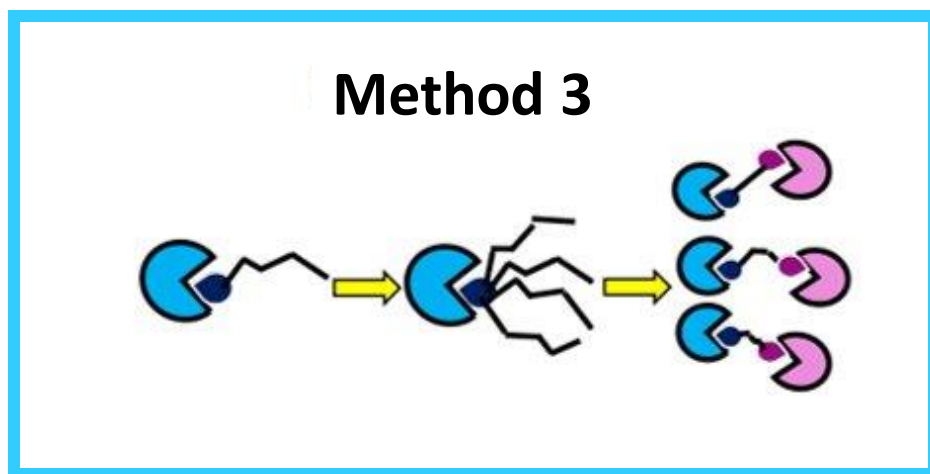
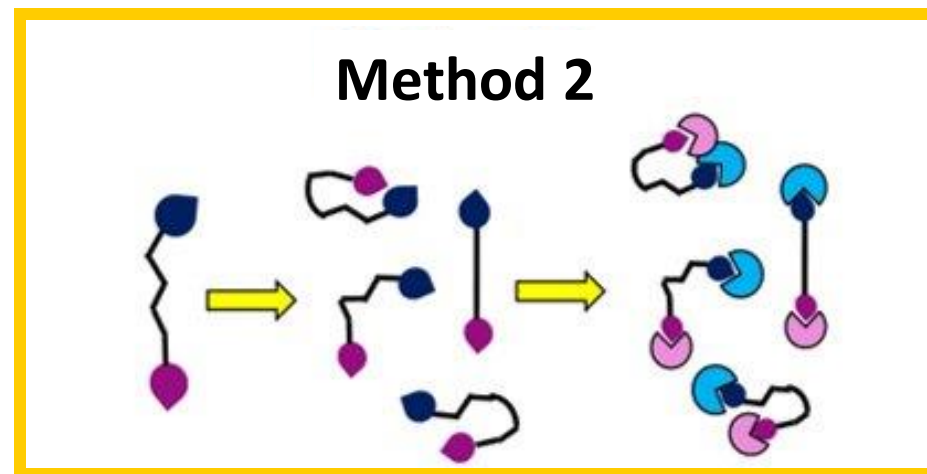
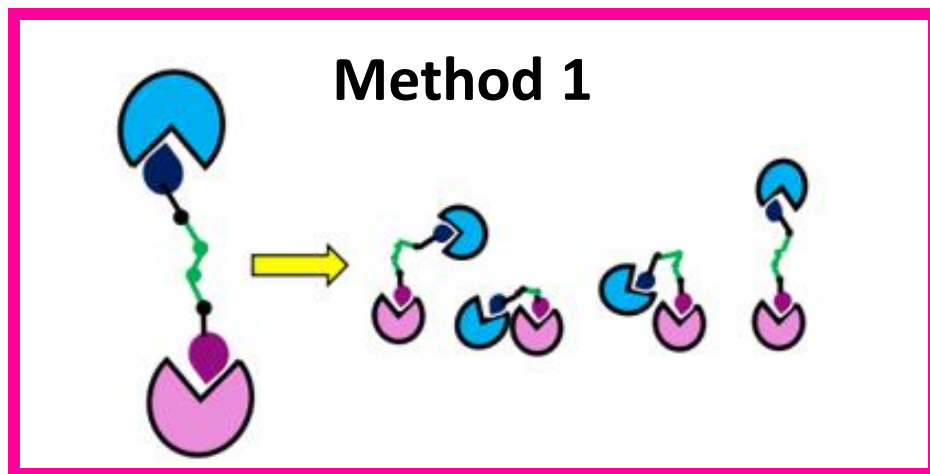




# Application of TCP in PIC design



# Four Different Approaches to TCP



DRUMMOND, Michael L.; WILLIAMS, Christopher I. In silico modeling of PROTAC-mediated ternary complexes: validation and application. *JCIM*, 2019, 59.4: 1634-1644.

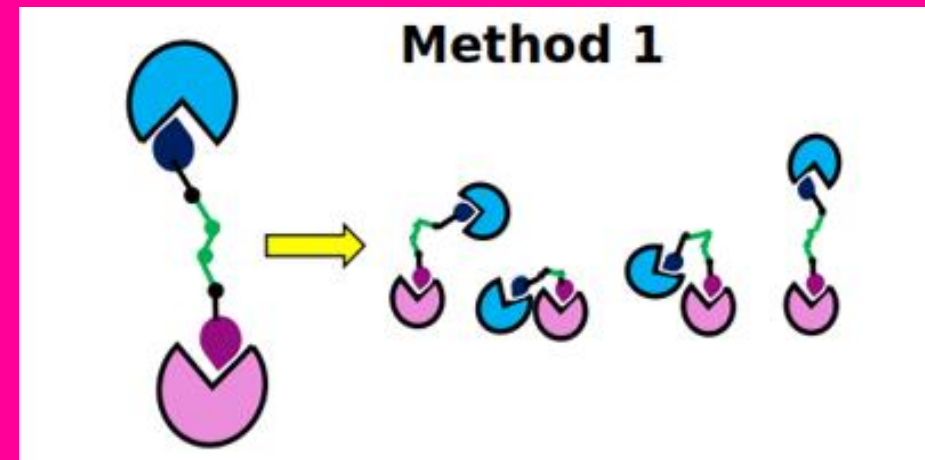
# Method 1

**Attachment of each fragment** (warhead/E3 binder) to the initial conformation of linker.

Positioning of **proteins** at around their respective fragments.

**Protein-ligand and protein-protein clashes** due to the separate starting environments (protein-fragment groups vs. linker conformation).

The PROTAC conformation is automatically adjusted to adopt an **extended conformation**.



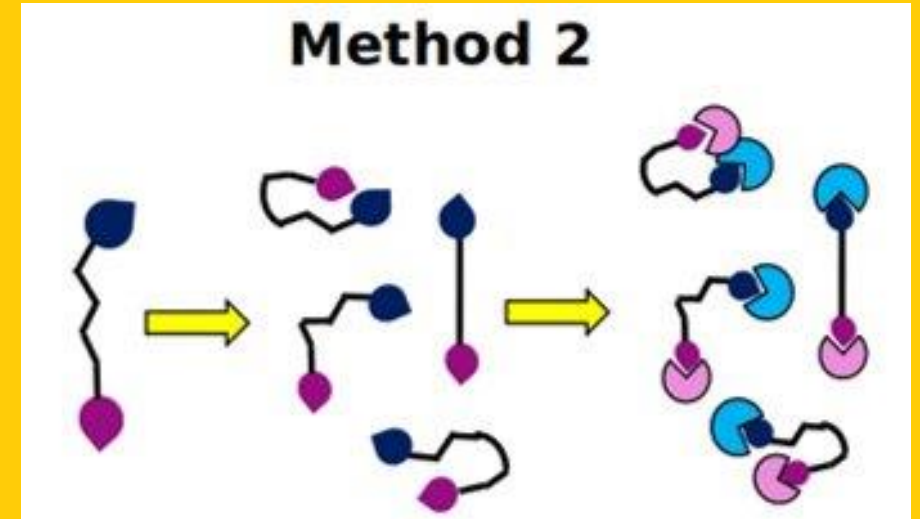
**The entire ternary complex is sampled at once.**

# Method 2

Two protein-fragment complexes and a full PROTAC are required as inputs.

Different PROTAC conformers are pre-generated/sampled on the fly (sampling may affect fragment conformer).

Because the PROTAC sampling occurs in the absence of any proteins, there is often significant overlap after the proteins are reintroduced.



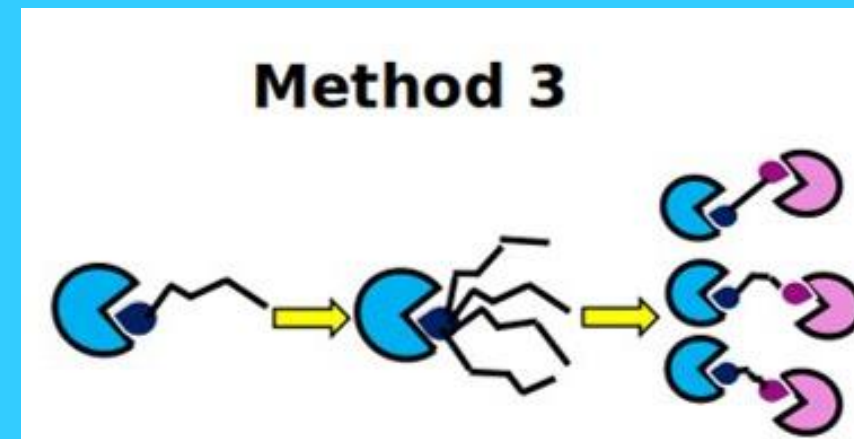
**PROTAC conformations are sampled independently, followed by post hoc addition of rigid body proteins.**

# Method 3

**Only one of the proteins is included** in the conformers sampling phase (the smaller of the two).

The fragment belonging to the binding moiety of the protein included in the sampling is tethered during the conformational sampling.

The second fragment is kept rigid to prevent deformations.

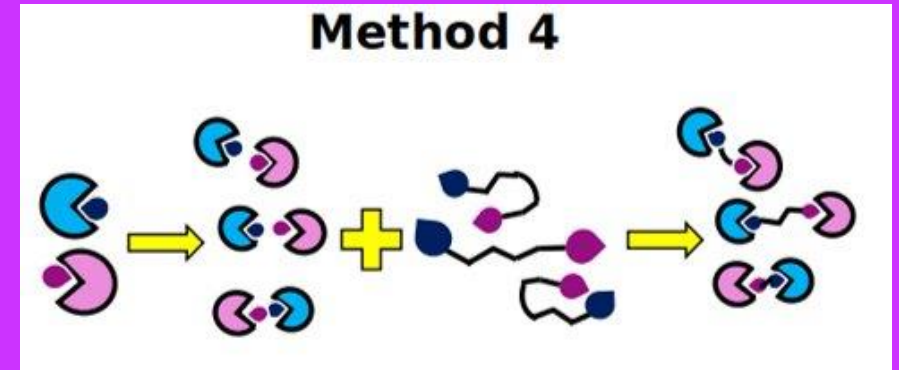


The PROTAC is sampled in the context of one of the proteins, with the second added afterwards.

# Method 4

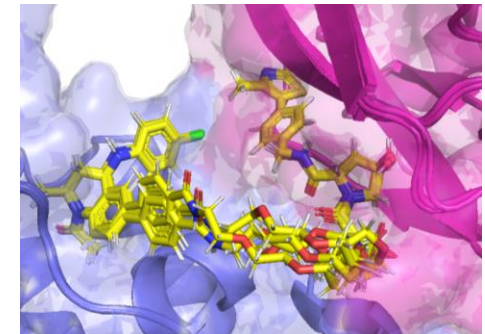
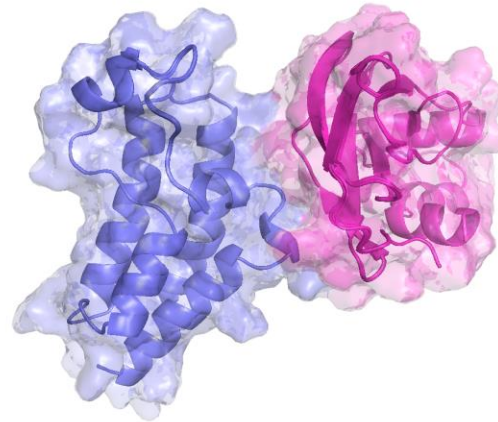
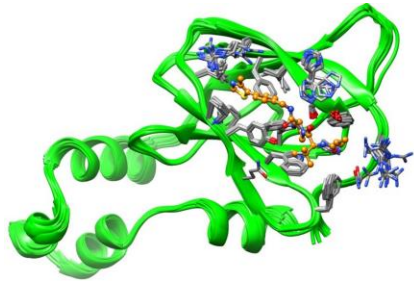
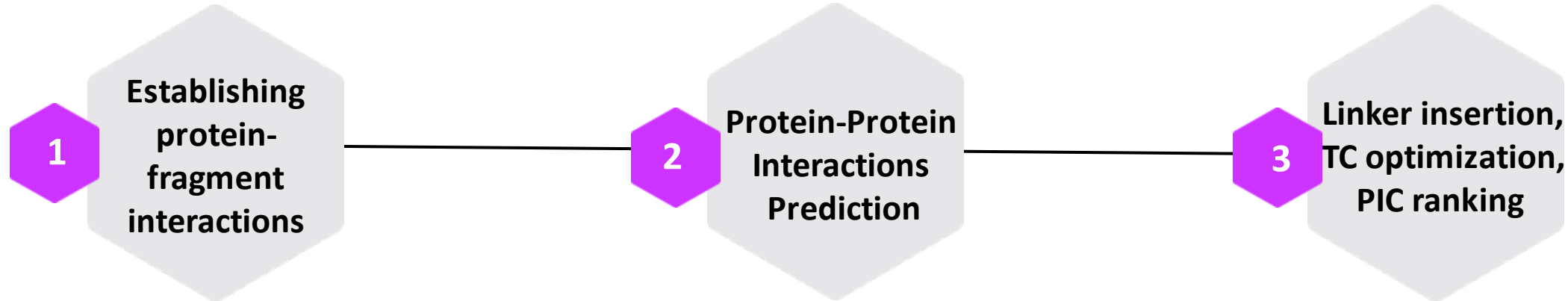
**Two protein-fragment complexes and different PROTAC conformations** are required as inputs.

- Phase 1: Protein-protein docking
- Postfiltering based on patch-based descriptors.
- Phase 2: PROTAC conformational ensemble.

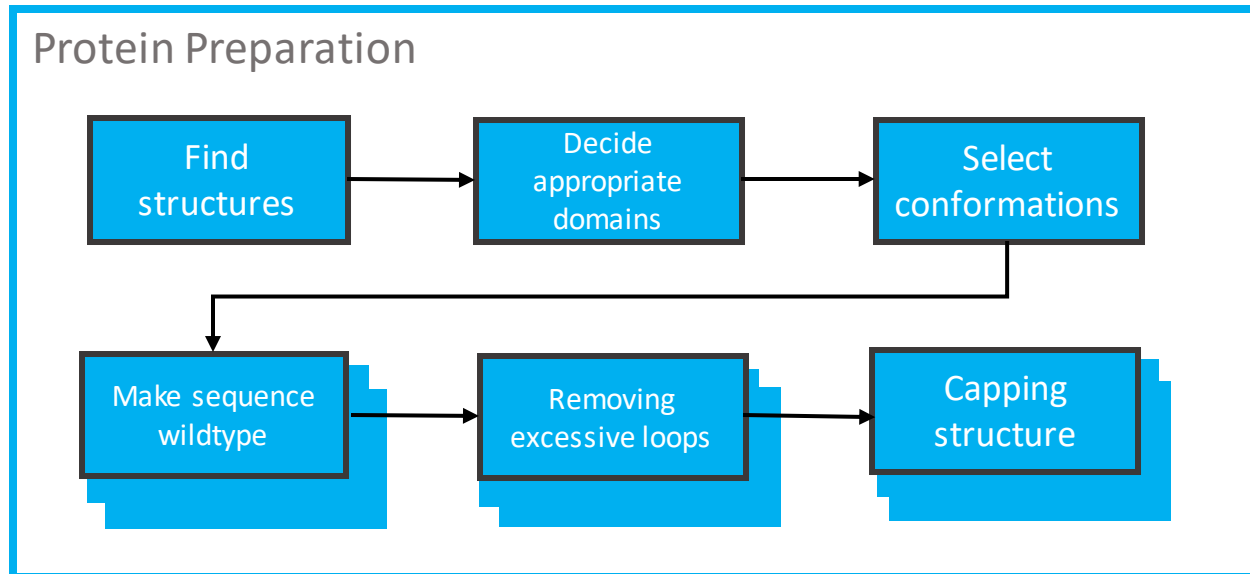


**PROTAC conformations are sampled independently of the proteins, but possible POI/E3 ligase are sampled via protein-protein docking.**

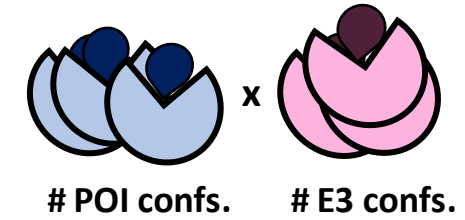
# CelerisTx: TCP Pipeline Introduction



# Conformational Ensemble Generation



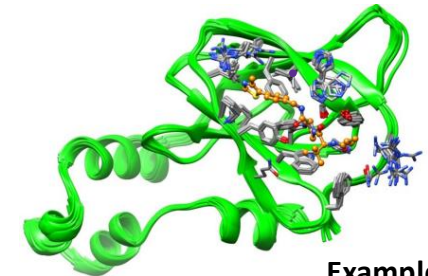
# PPI runs =



Typically 10-20 distinct structures per a given protein

- X-Ray
- NMR
- Cryo-EM
- *Ab initio*:

- o Molecular dynamics
- o Normal mode analysis
- o Homology modelling



Example:  
VHL ligase



# Binary Inputs Preparation

Data acquisition

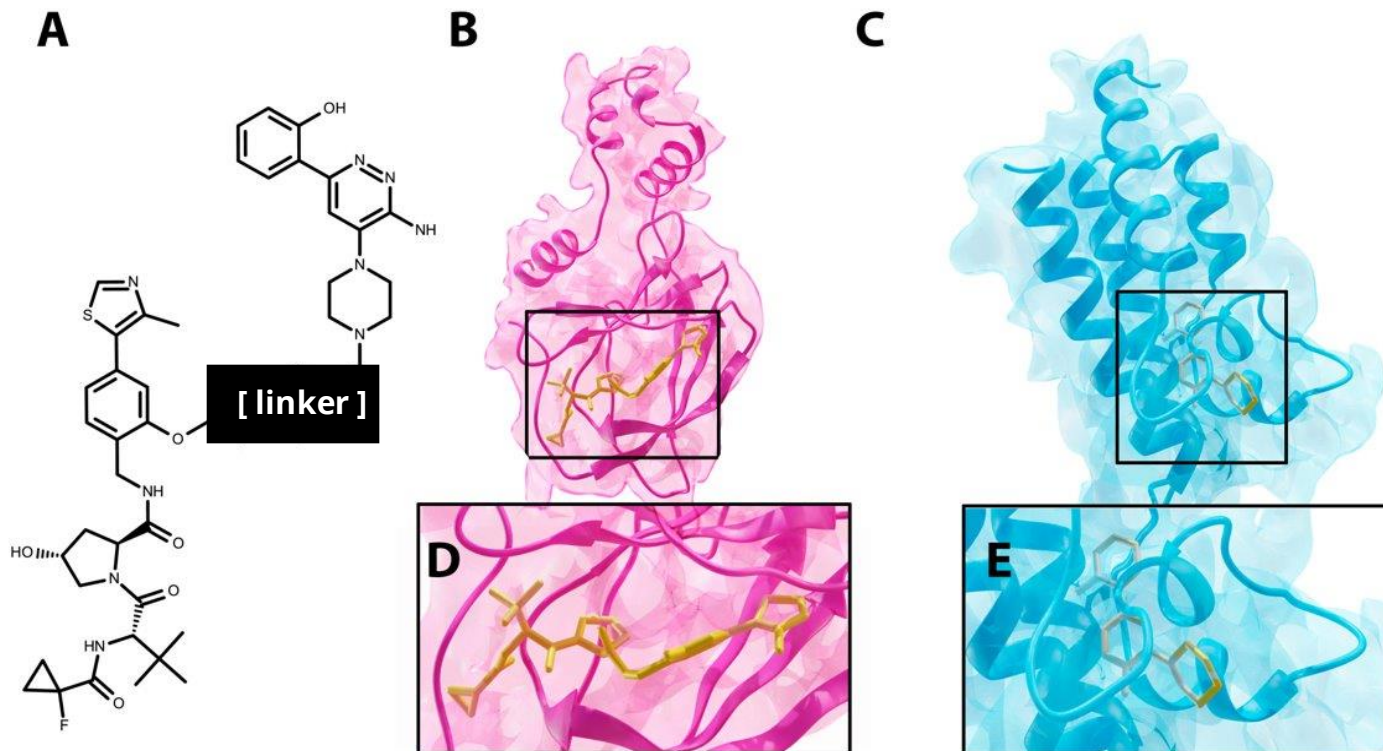
Protein/  
ligand  
structure  
preparation

Generating  
binary  
complexes

- **Retrieval of** protein structures
- Automated download of ligand **bioactivity data**

- Optimizing structure with co-resolved warhead
- Rigid body docking
- (Ensemble docking)

- **pKa predictions**
- Adding explicit hydrogens, fixing missing chains
- **Energy minimization**



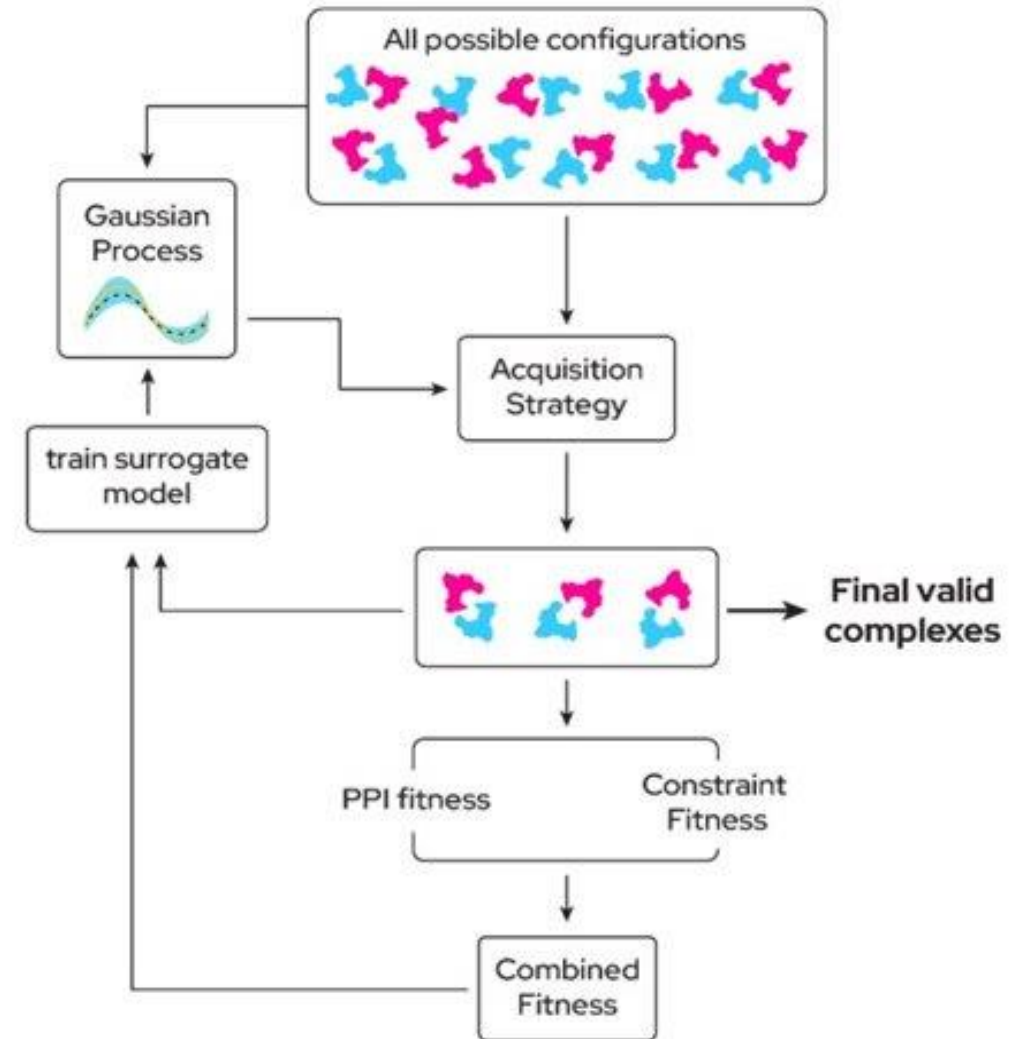
# Bayesian-optimization PPI Prediction

Generating different relative **orientations** and **translations** helps **sample alternate ternary complex poses**.

Our goal is to optimize a **score** that describes the **quality of a ternary complex pose**.

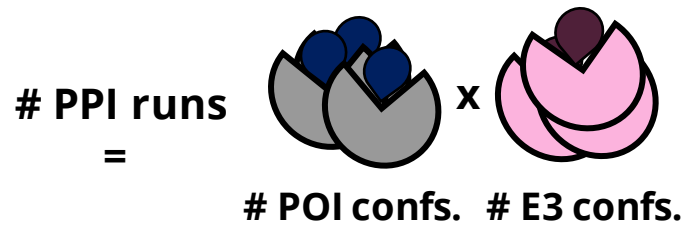
**Protein-protein interaction** score, **linker constraint** score, **PIC stability** score.

Clustering, reranking, and filtering final poses.

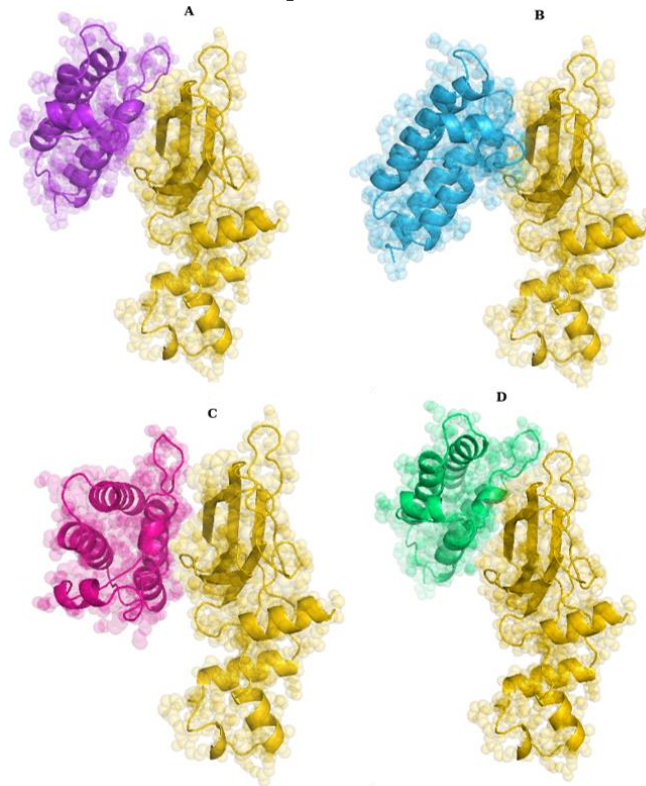


# Physics-based PPI Prediction

Conformational ensemble generation



Exploring PPI space



Inclusion of a "spacer" to mimic presence of a linker



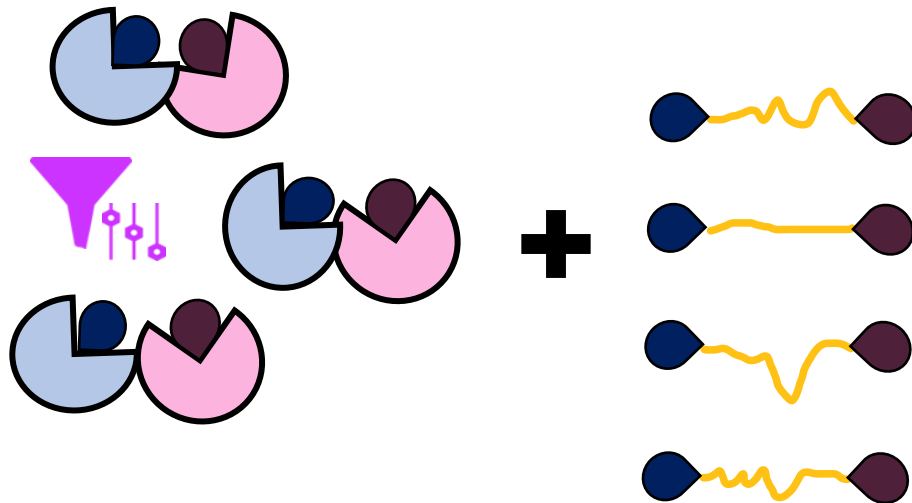
Pose filtering

- 1) Exit vector distance
- 2) PPI score
- 3) L-RMSD

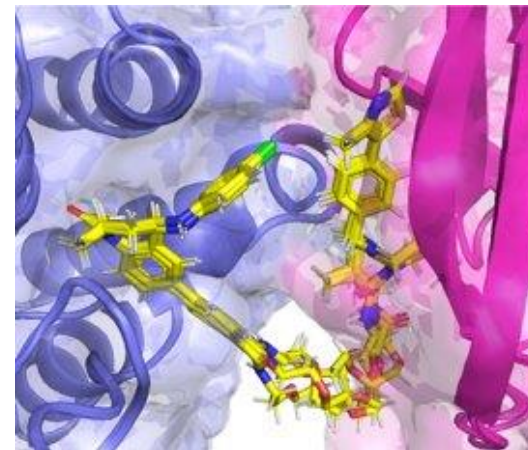


# Linker Insertion and Structure Refinement

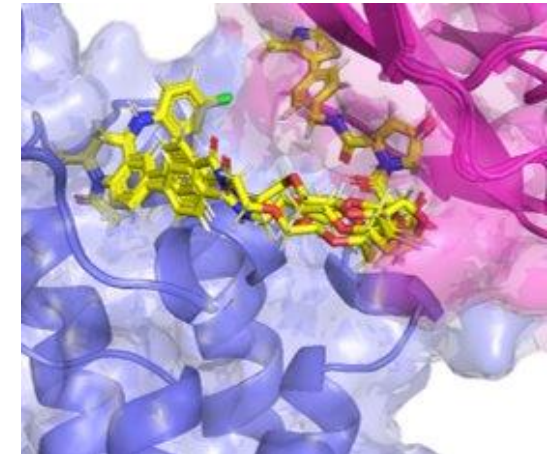
Input: Post-filtered poses +  
3D PIC structures



Modeling in a linker  
(\* fragments are constrained)



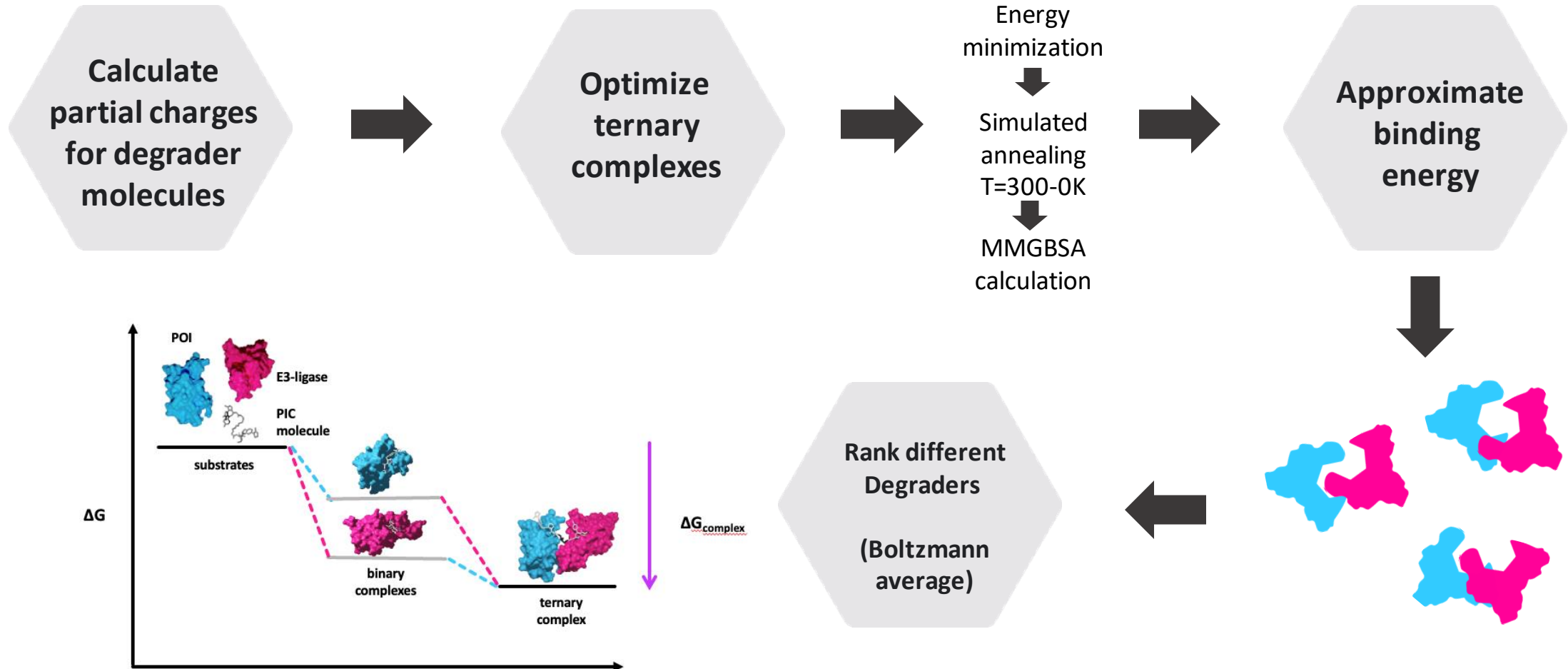
Top view



Side view

# Linker Insertion & Structure Refinement

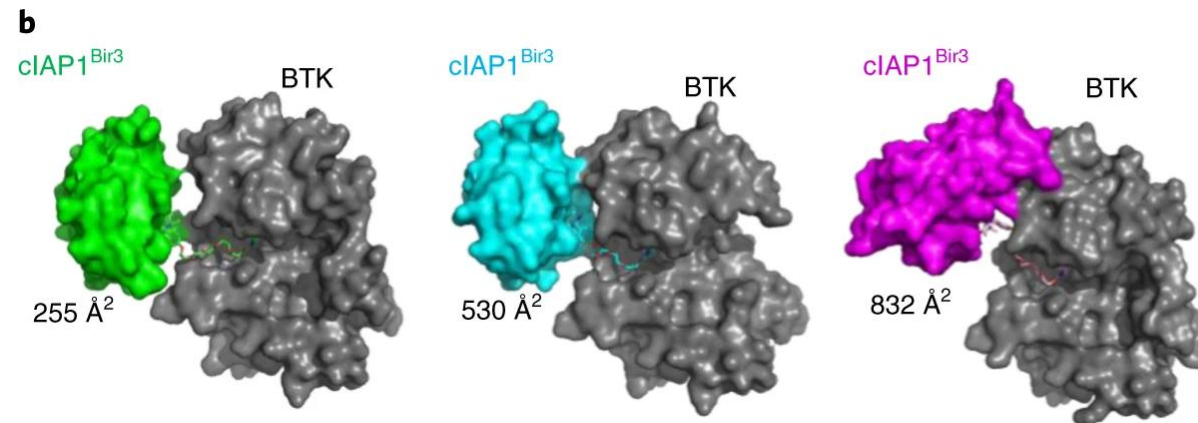
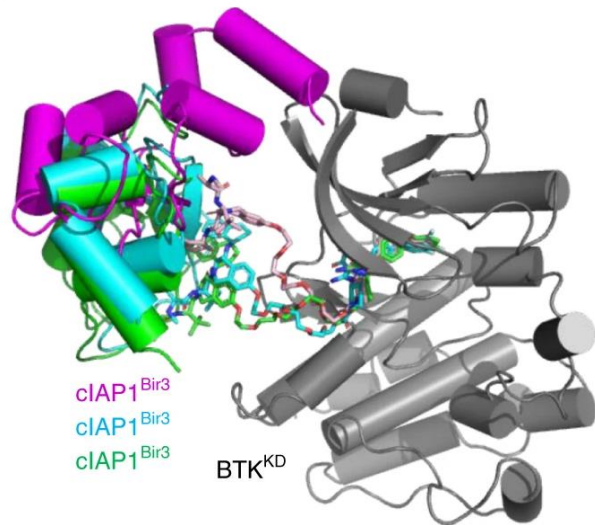
Molecular Mechanics with Generalized Born and Surface Area Solvation



# Validation

Moving away from structure to affinity-based comparisons

- 22 experimentally determined structures available in PDB
- All available structures originate from **X-ray measurements** (a single "snapshot")



**PDB ID: 6w8I**

# Validation

## Use case

1. Target selection based on feasibility on developing ternary complex

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2. Target selection based on selectivity prediction

---

3. Calculate minimum linker length

---

4. & 6. Decide which compounds to synthesize based on ternary complex formation ability

---

5. Decide which compounds to synthesize based on selectivity

---

7. Accurate structural prediction

## Validation

Known degraded targets vs random targets as negatives

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Known example pairs of proteins where selectivity has been achieved vs random pairs

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Datasets with same warhead + E3 binder, different linker size

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Compound series needed

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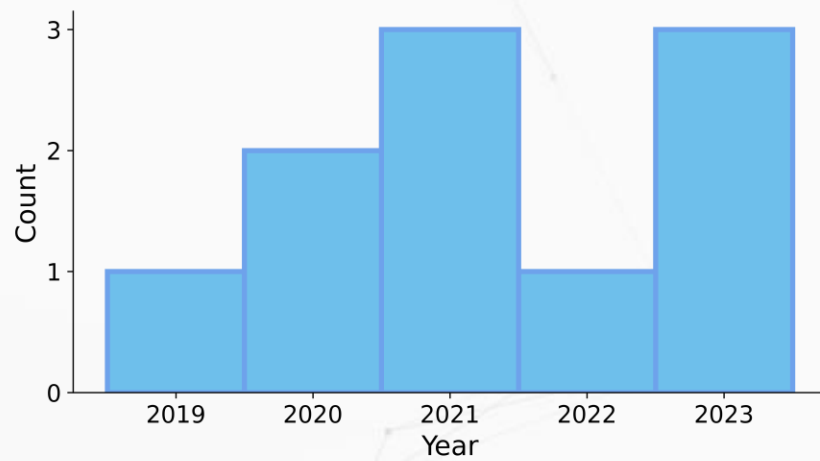
Compound pair data with the same E3 ligase but different POI tested

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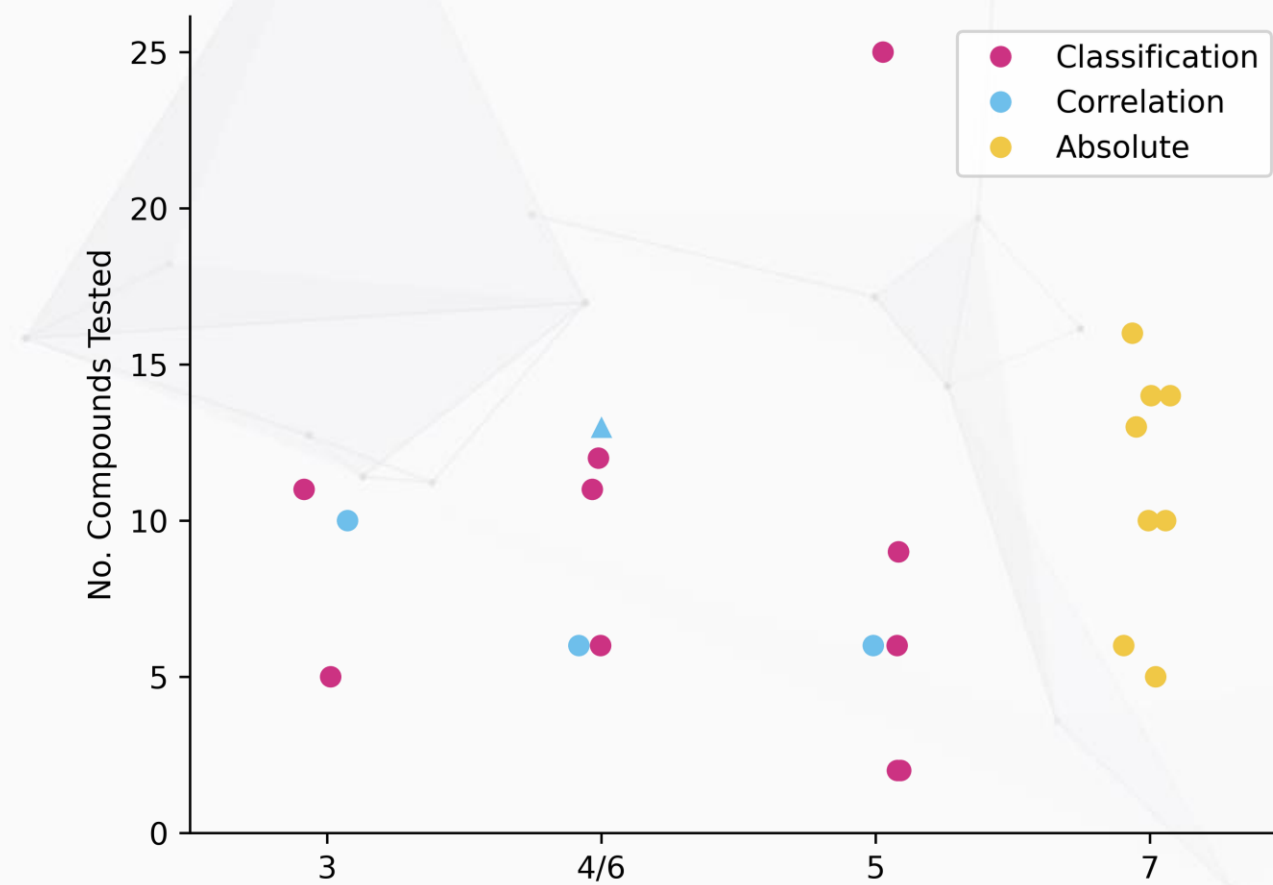
Experimentally-determined structures

# Validation: Examples from Literature

3. Calculate minimum linker length
4. / 6. Decide which compounds to synthesize based on ternary complex formation ability
5. Decide which compounds to synthesize based on selectivity
7. Accurate structural prediction



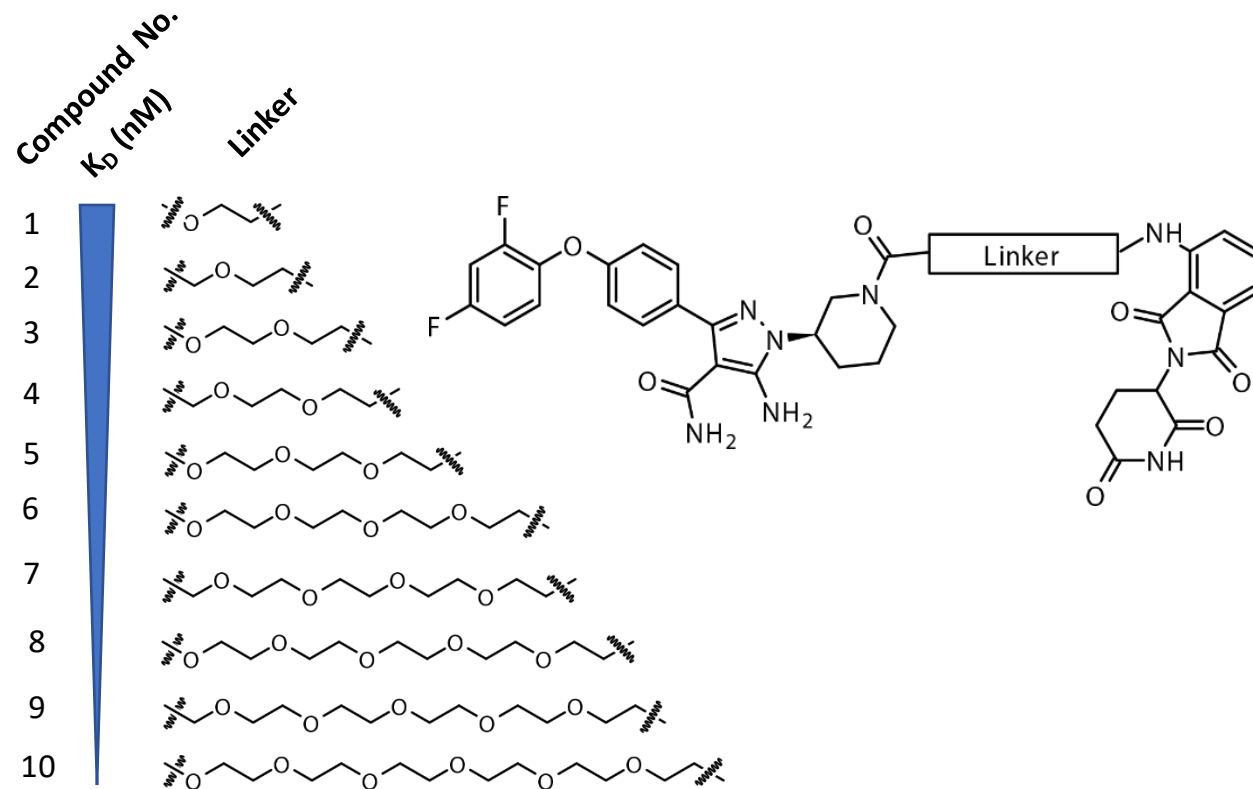
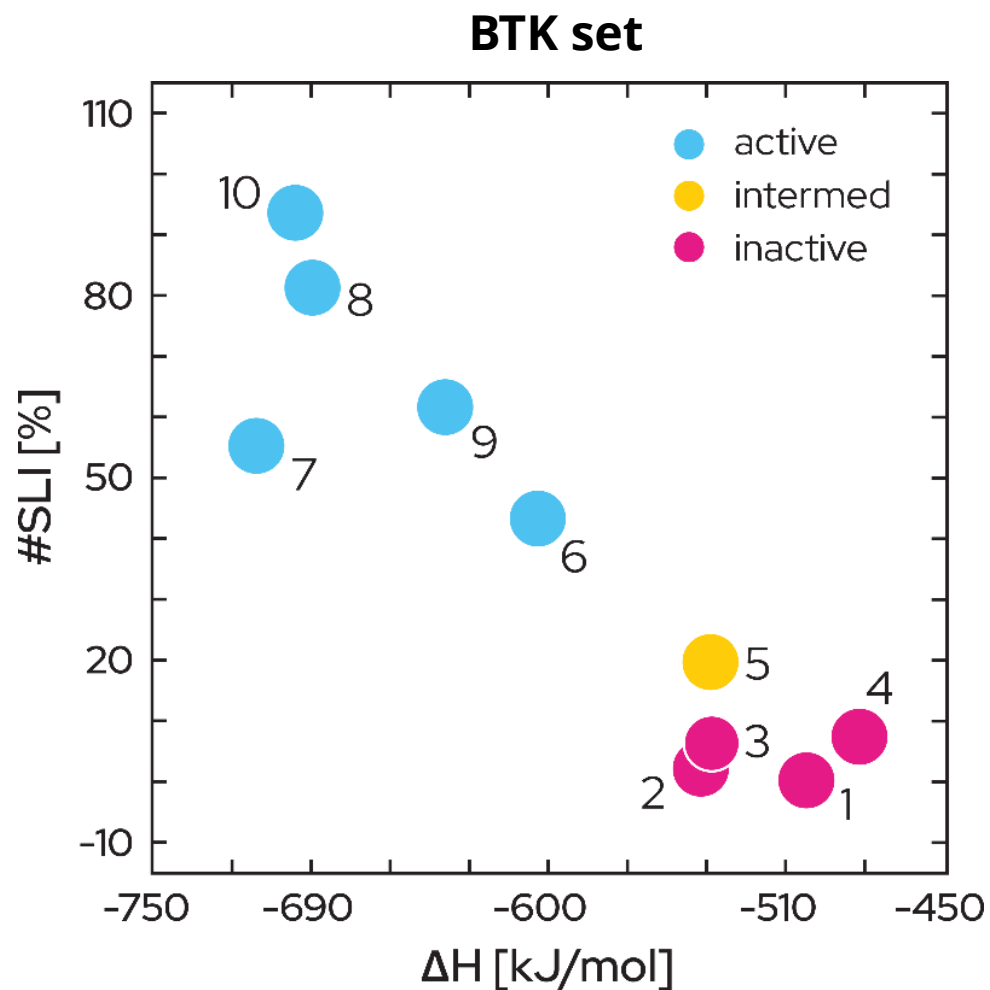
Histogram showing number of ternary complex pipeline method papers vs time



Plot showing number of compounds validated against in different type of ternary complex pipeline validation settings.



# Validation: Tyrosine-protein kinase BTK

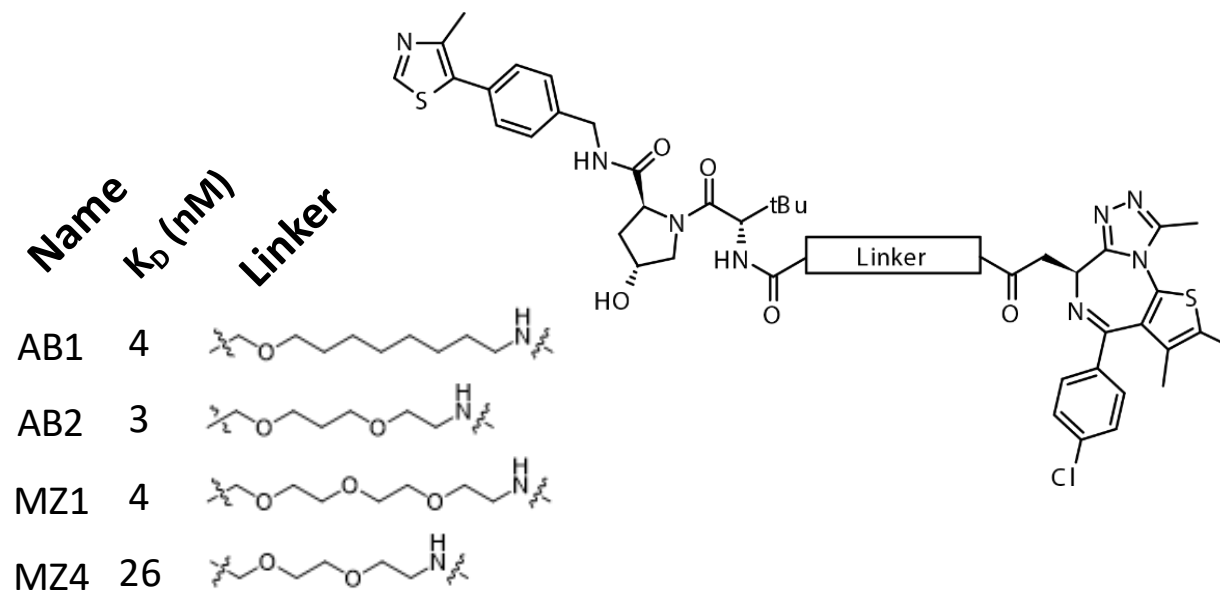
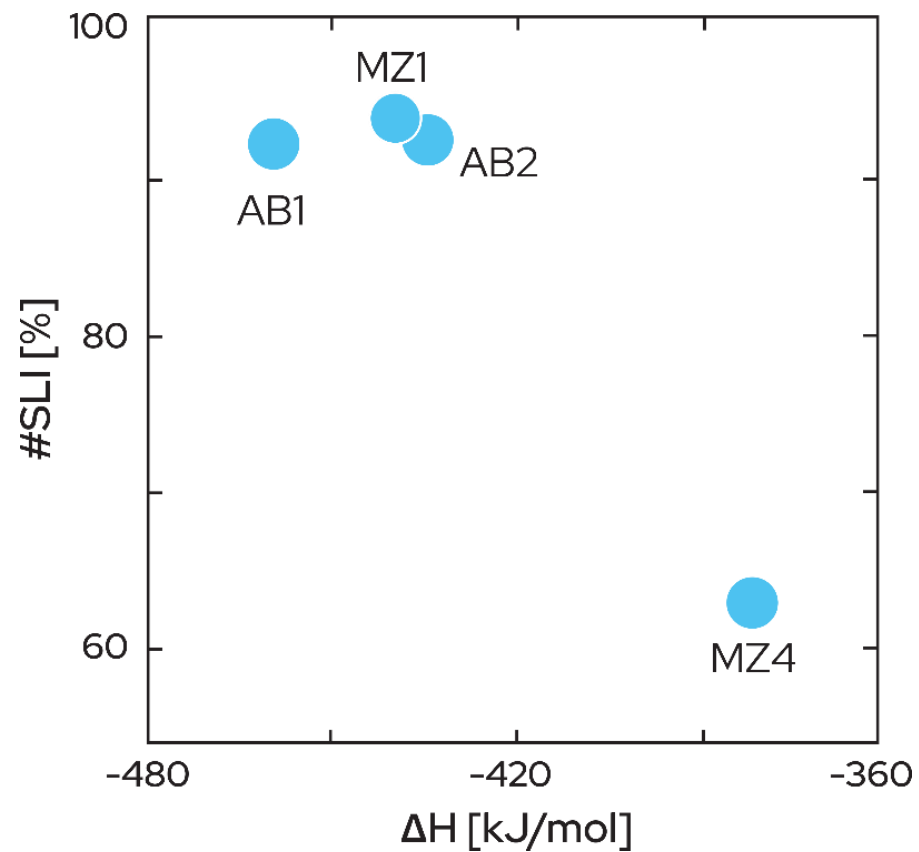


Validates “What is the shortest linker length?” + HitID virtual screen

Experimentally-determined data  
Zorba, A. et al. (2018). Delineating the role of cooperativity in the design of potent PROTACs for BTK. PNAS, 115(31), E7285–E7292.

# Validation: BRD4

BRD4 set



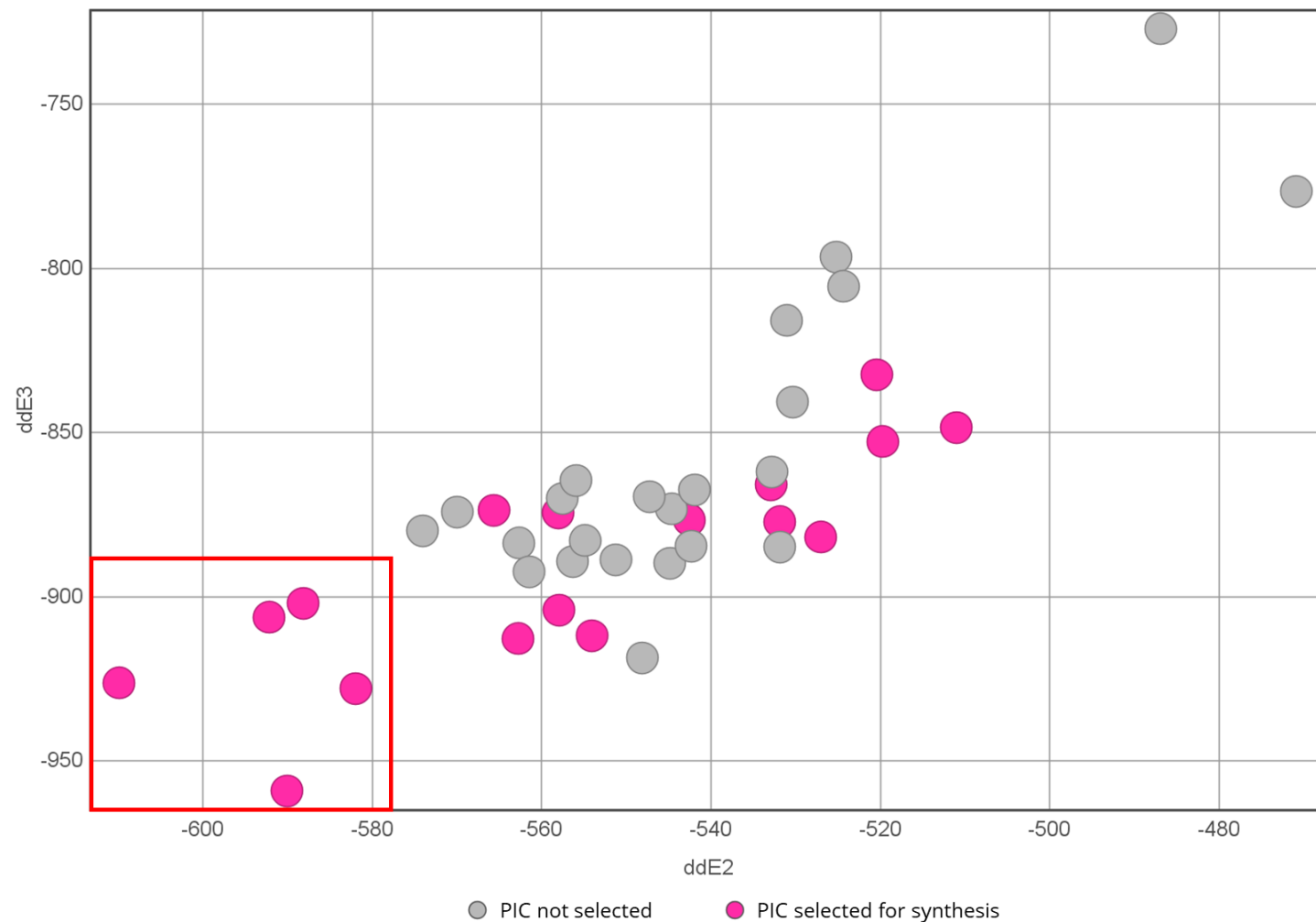
Validates “What is the shortest linker length?” + HitID virtual screen

# Using TCP Pipeline For Hit Identification

Compound	#SLI (%)	# rotatable bonds	$\Delta H$ (kcal/mol)
Domain-X control (+ve)	33.4	21	-832
Domain-X control 1 (-ve)	8	7	-710
Domain-X control 2 (-ve)	2.4	9	-691
Domain-Y control (+ve)	27.2	11	-815
Domain-Y control (-ve)	12.6	7	-739

Validates HitID virtual screen

# Using TCP Pipeline for Hit Expansion



**80 design ideas**  
from MedChem

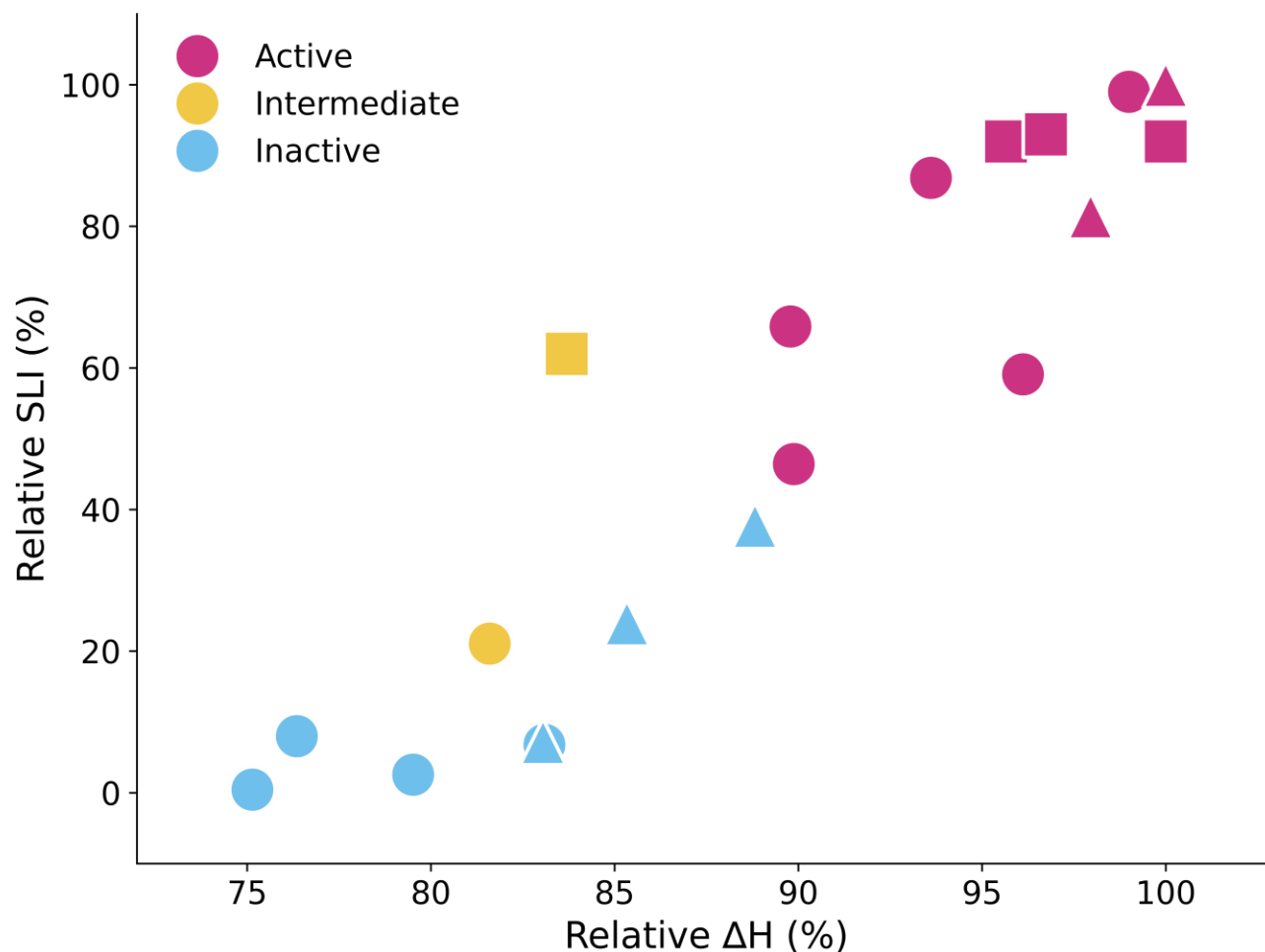


**40 compounds**  
for PBTCP runs



**17 compounds**  
selected for synthesis

# In-house TCP pipeline: Overall Validation Results



**Validation of our PB-TCP method** for different PROTACs. Different marker shapes represent different targets.

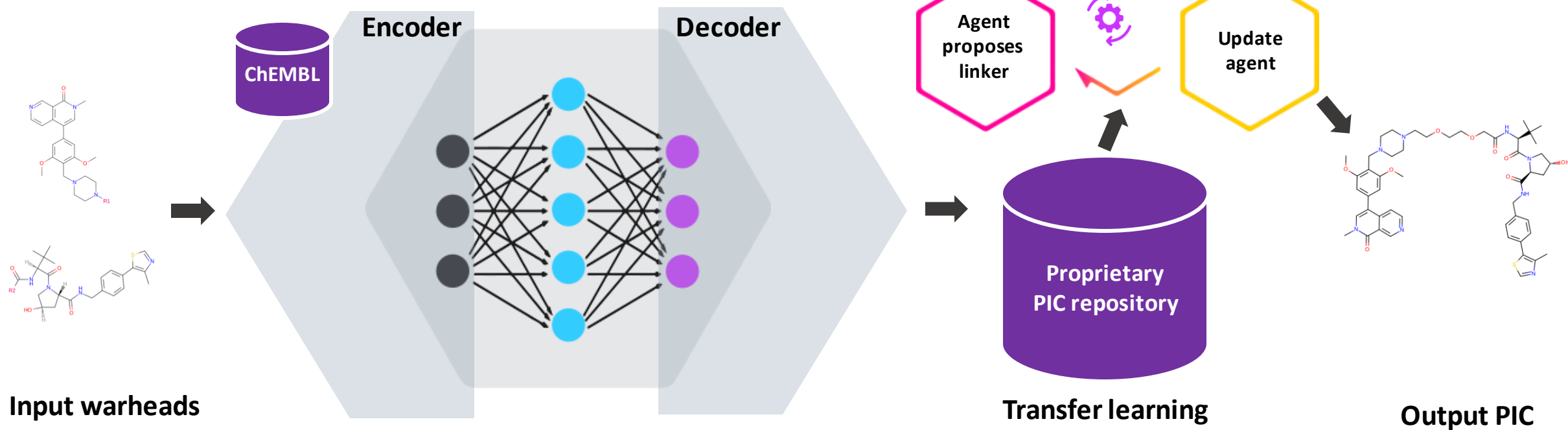
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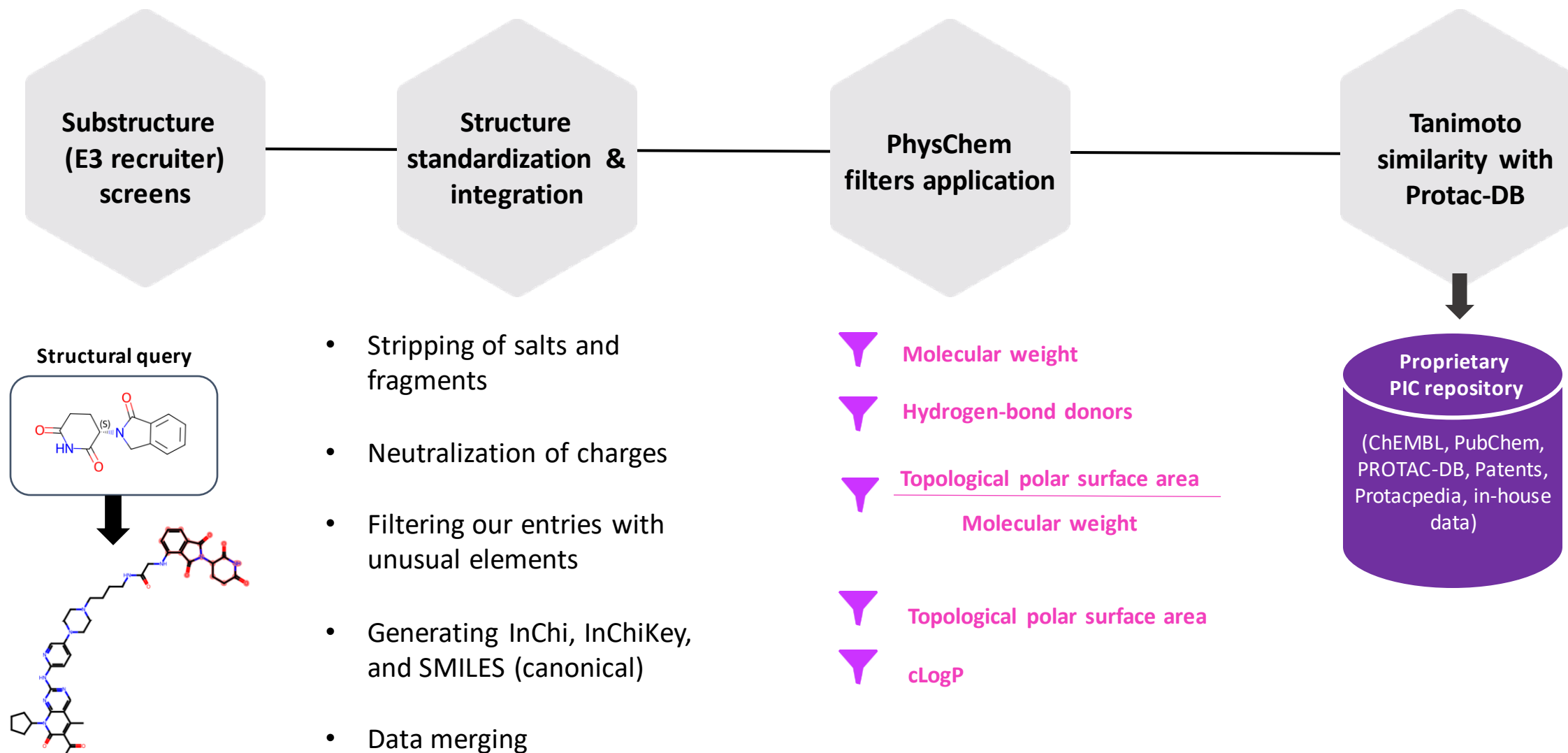
# Generative AI for Molecular Linker Design

## Input linker criteria

- Number of hydrogen bond donors
- Desired linker length
- Number of rotatable bonds
- .....

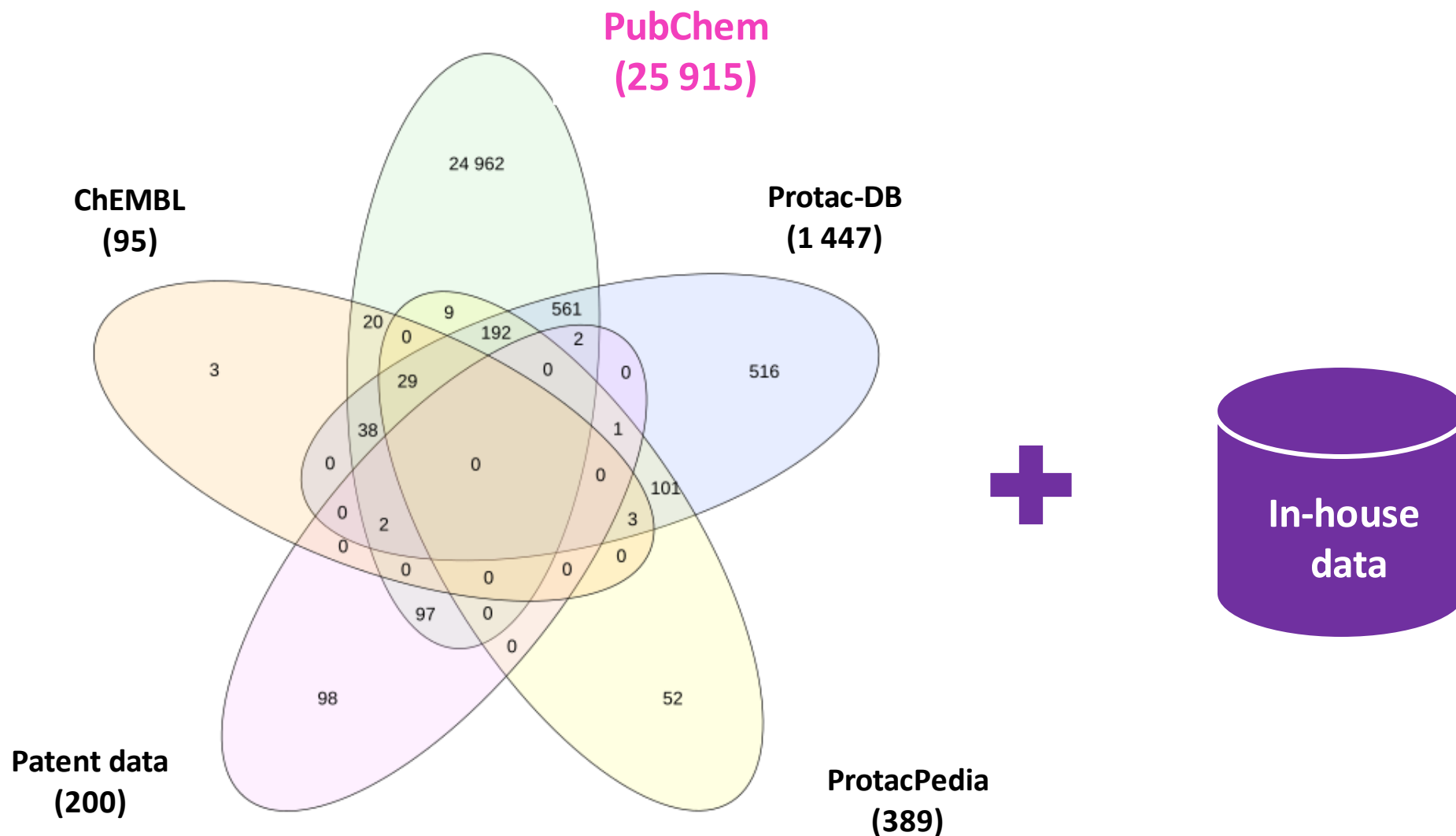


# Compilation of Training Data Set: Strategy



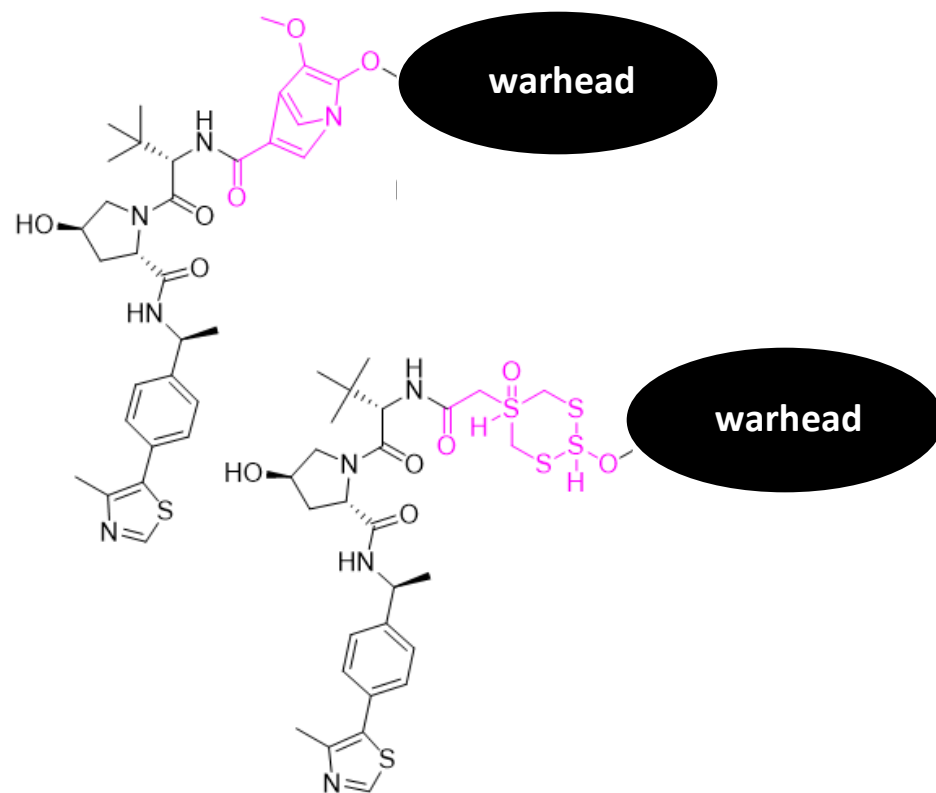


# Compilation of Training Data Set: Data Overlaps

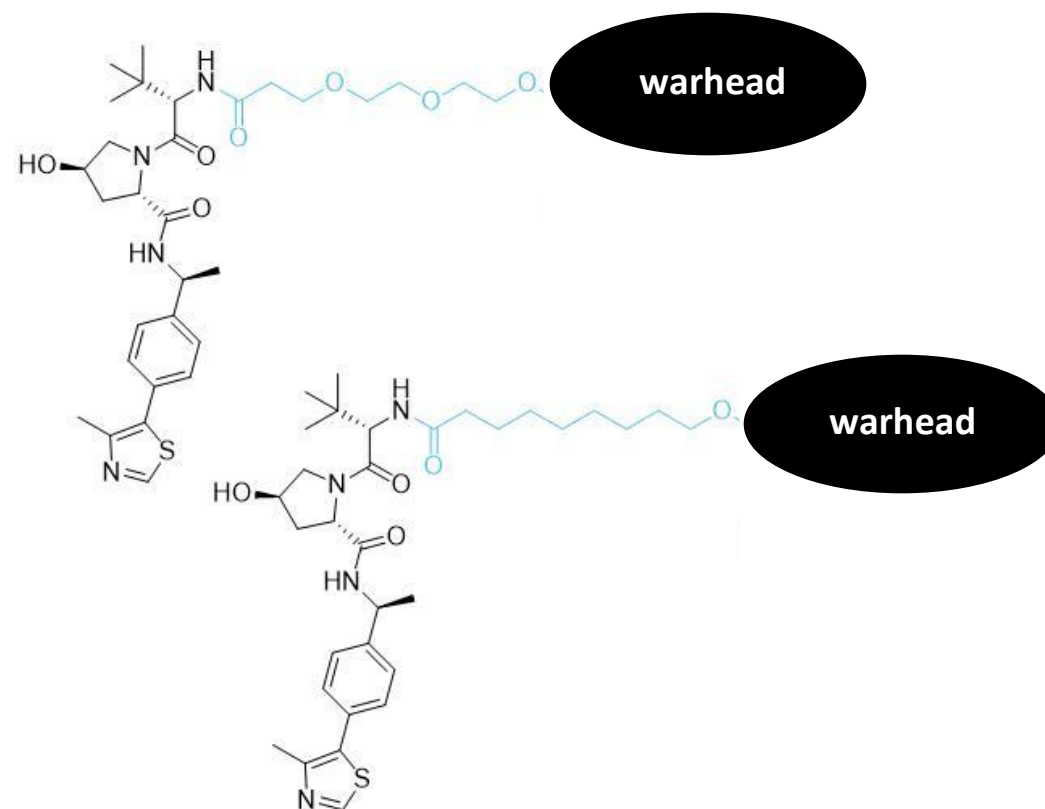


# Linker Generation: Examples

creative & novel

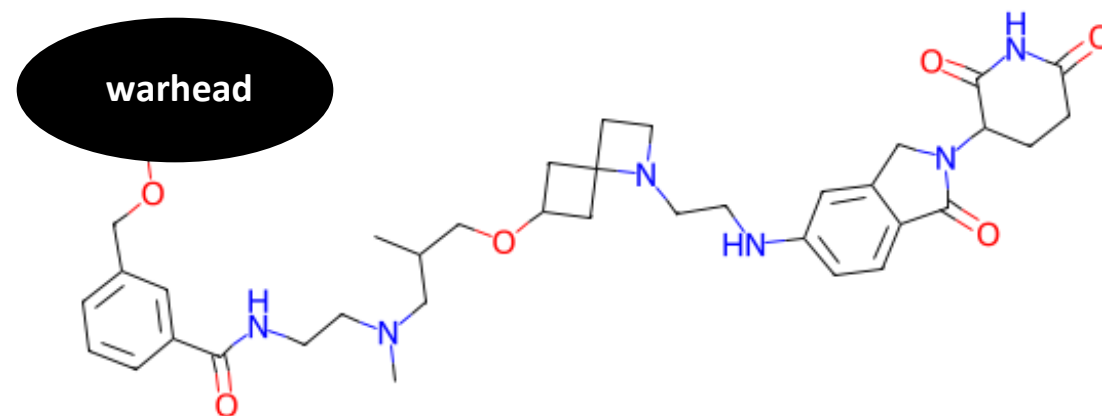
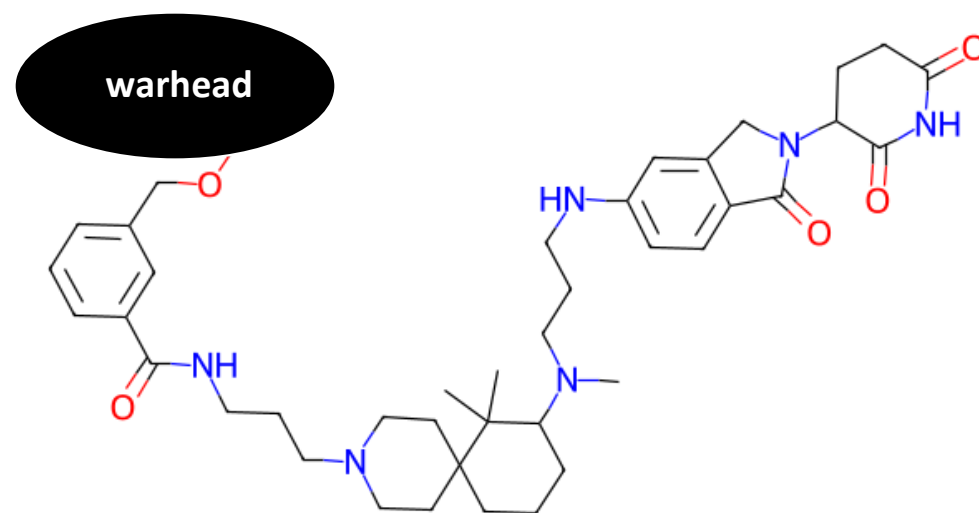


stable & synthesizable



VS.

# Linker Generation: Examples



Linker generation model augmented  
by a subset of commercially  
available PIC building blocks

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# Conclusions

- ◆ The development of PICs can greatly benefit from the application of in silico approaches:
  - ◆ **Predicting ternary complexes** can be used for both **structure-based design**, and for **ranking compounds** during selection.
  - ◆ **Validation** of TCP approaches is crucial to ensure its applicability (by comparison of calculated TC stabilities with **ternary dissociation constants** or by obtaining additional experimental structures, preferably using **Cryo-EM or NMR** methods, as these methods capture protein conformational dynamics).

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# Conclusions

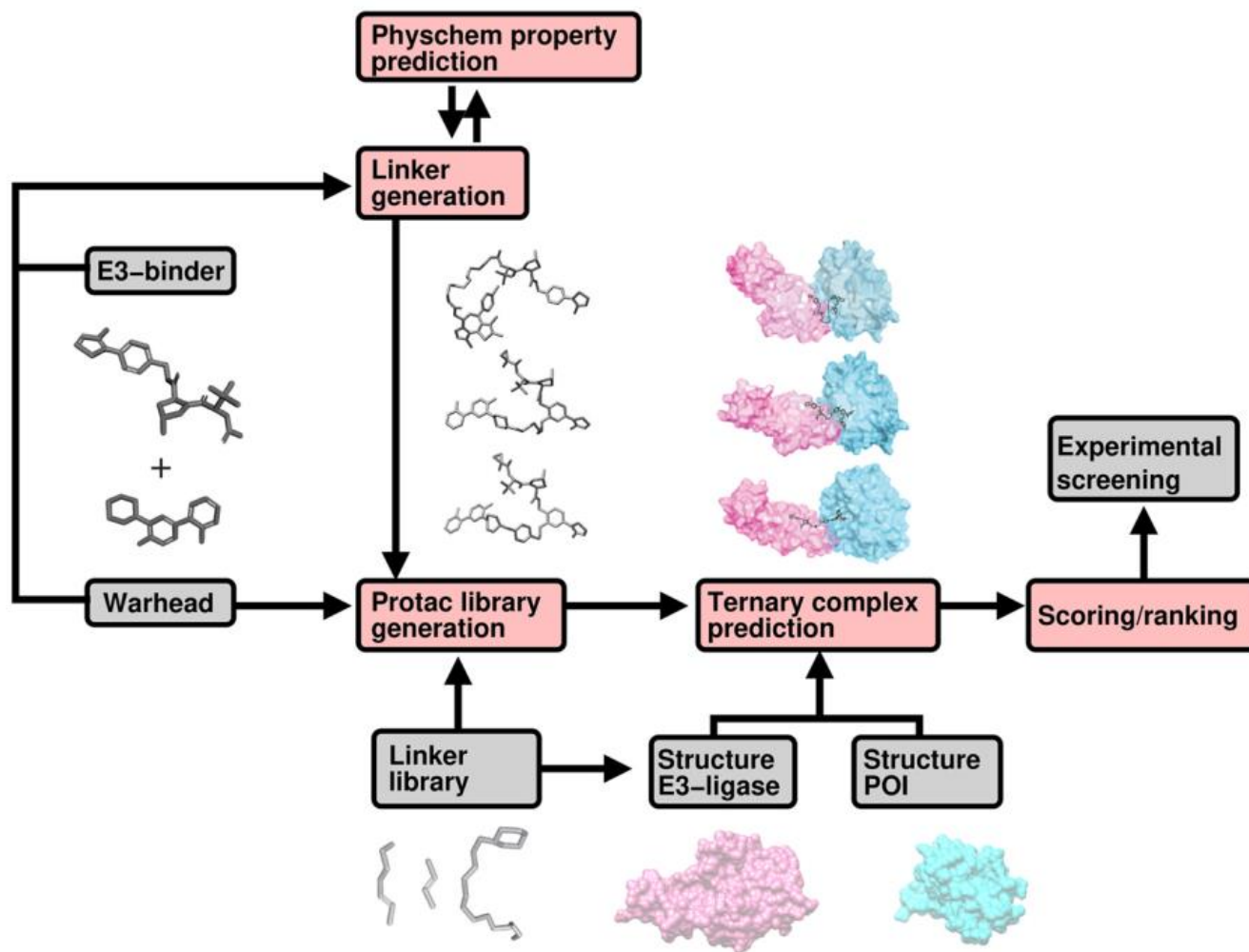
- **Linker generation tools** can provide valuable assistance to medicinal chemists in designing novel linkers by systematically optimizing different compound properties.
- Applying **data mining** approaches to augment current PIC datasets with **novel** and **standardized** data points is essential to enhance efforts in linker generation approaches.

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# Conclusions

- ◆ Coupling **ternary complex prediction** with **linker generation** has the potential to strengthen the current degrader discovery pipeline

# Example of a PROTAC Screening Pipeline



Gray boxes:

- experimental steps
- input data

Pink boxes:

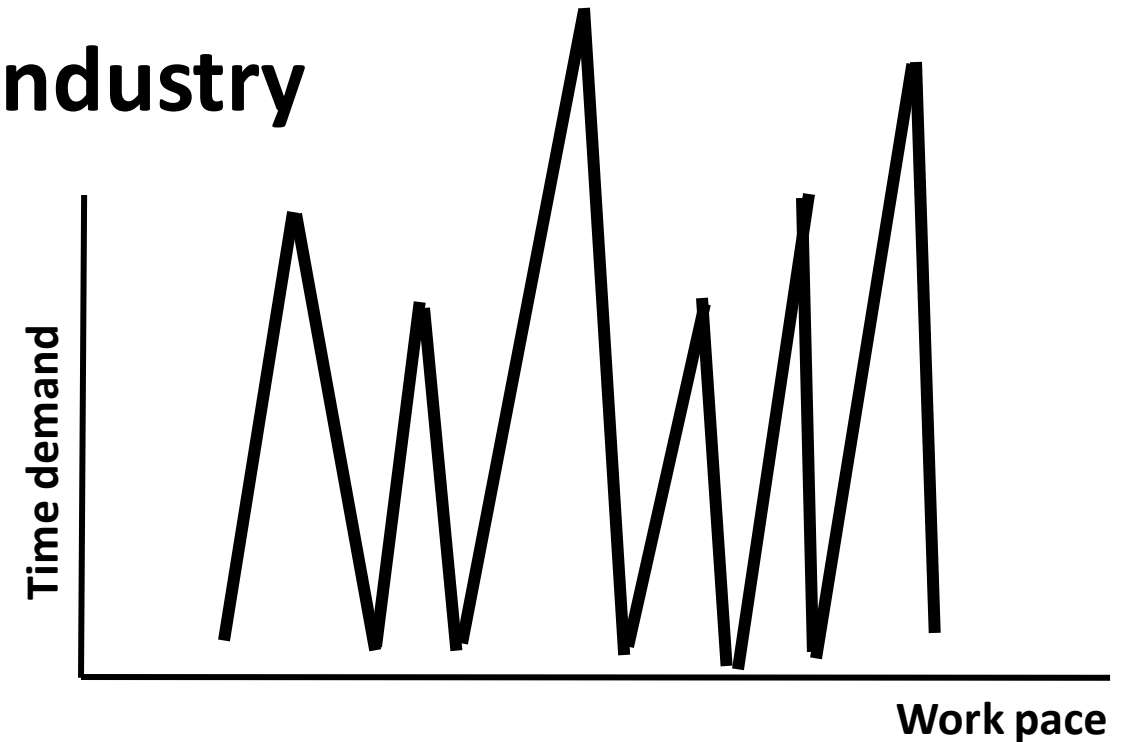
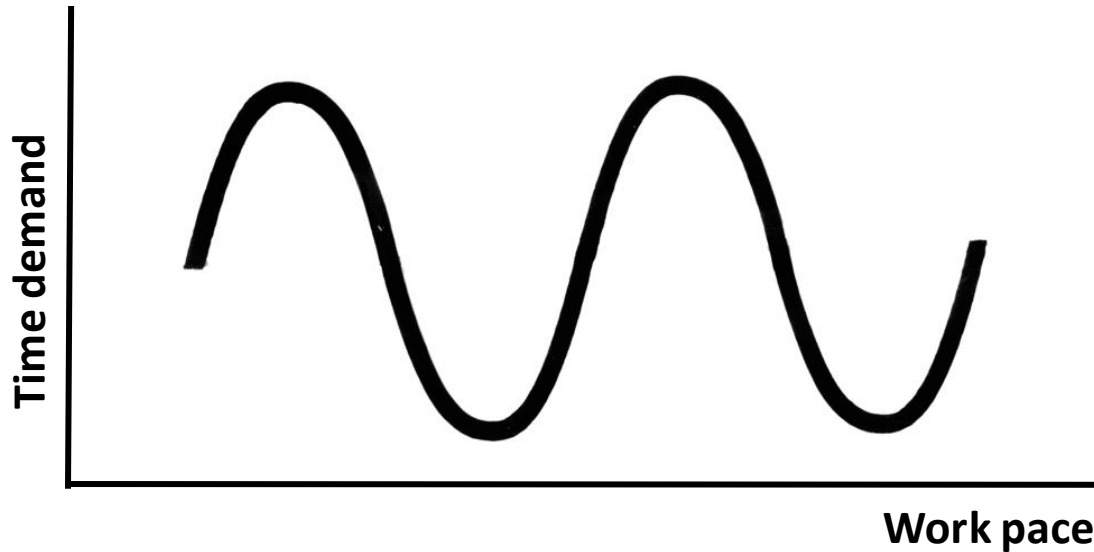
- in-silico steps

# AGENDA

<b>Introduction to Targeted Protein Degradation</b>	<b>3</b>
<b>Ternary Complex Prediction (TCP)</b>	<b>9</b>
- Challenges	9
- Application of TCP in Drug Discovery Pipelines	13
- Methodology (AI-based vs Physics-based)	18
- Validation of the Approach	25
<b>Linker Generation</b>	<b>34</b>
<b>The Journey of Computational Chemists in Biotech</b>	<b>44</b>
- Work model in Academia vs. Industry	45
- From academia to industry: Transferable Skills	46
- Job Interview Process	47
- Two Types of Computational Chemists	48



# Work model in Academia vs. Industry



## Academia

## Industry

<b>Work pace</b>	Cyclic	Gas-Brake-Gas-Brake
<b>Deliverable</b>	Publication (patent)	Marketed product/service
<b>Time frame</b>	Long-term projects	Tight deadlines and project timelines
<b>Team culture</b>	Individualism	Cross-functional teams, frequent <b>coordination</b>
<b>Decision-making structure</b>	High degree of autonomy	Hierarchical

# From Academia to Industry:

# Transferable Skills

Publishing research papers

**Data Visualization and Communication**

Public outreach (Researcher's night)

**Communication with non-experts** (investors, collaboration partners, etc.)

Customized code

**Documented code/ data pipelining tools** (KNIME, LiveDesign, etc.)

Teaching activities

**Cross-functional project team communication** (matrix organization )

Student (co-)supervision

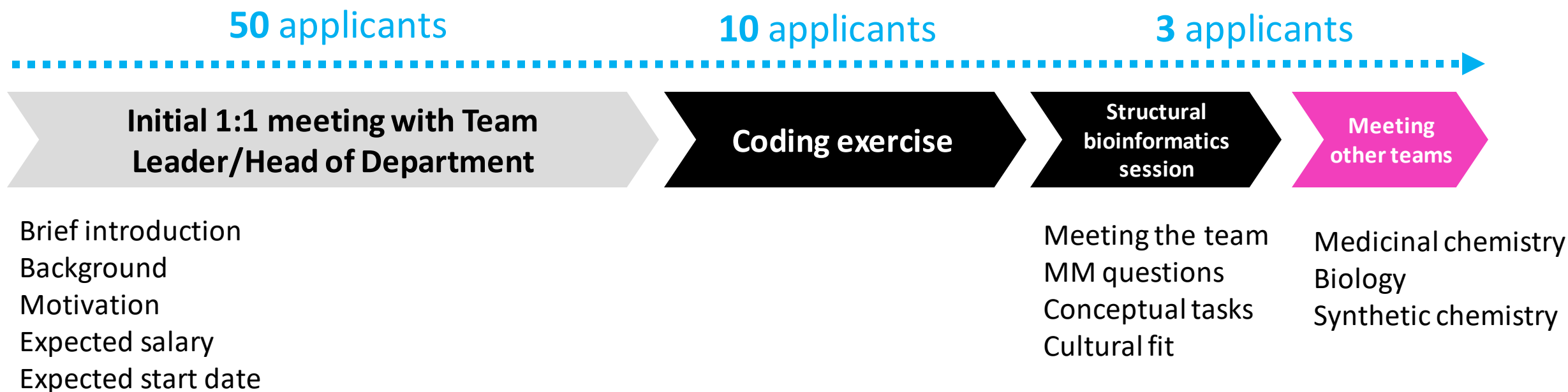
**Managing a team, supervising junior team members/research interns**

Critical thinking, problem solving

**Task decomposition to manage workload**

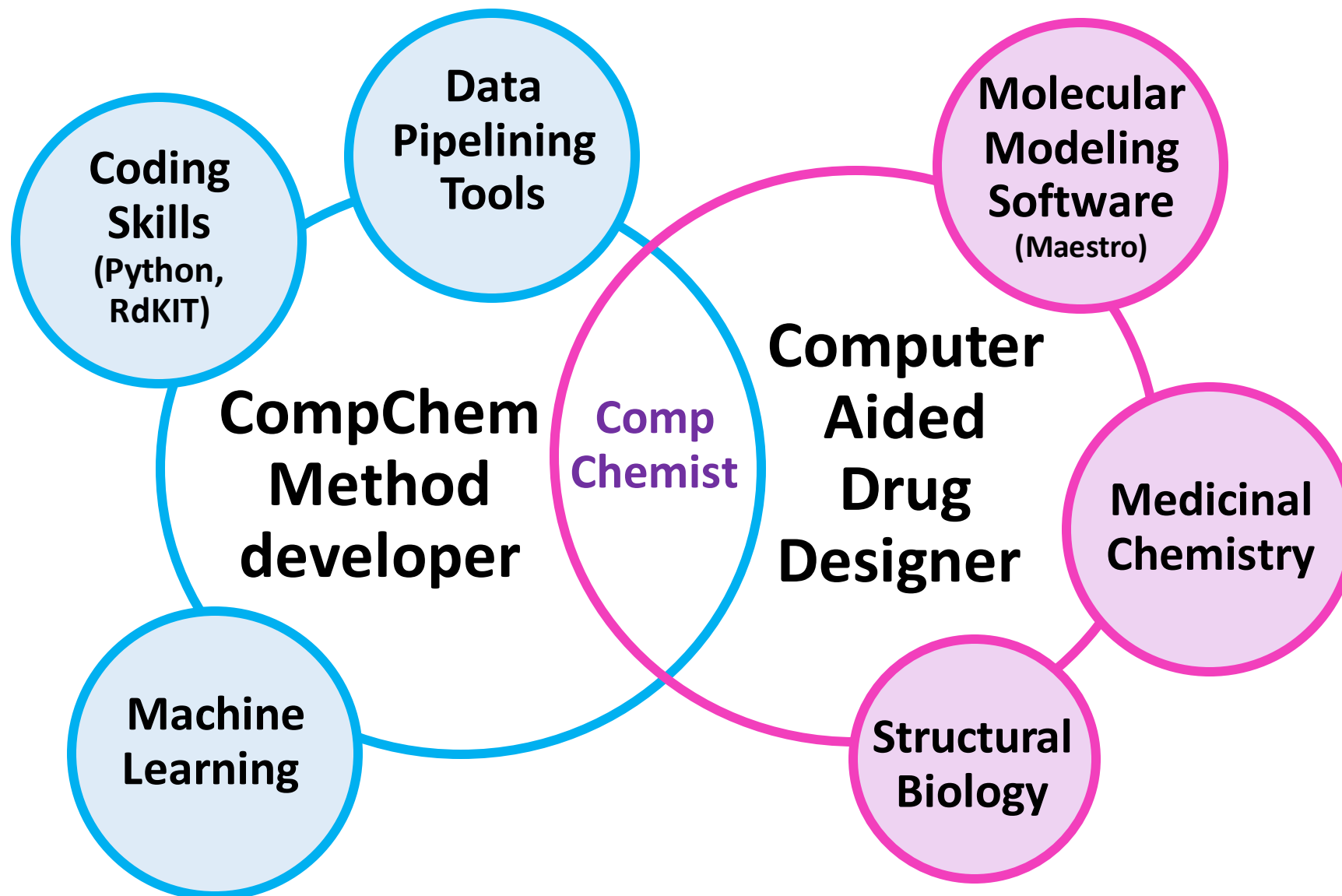
# Job interview process

Role: **Senior/Principal Computational Chemist** (London, Graz)



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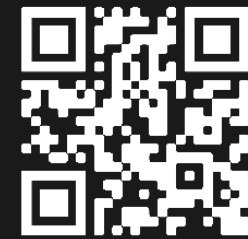
# Different roles as a Computational Chemist



# Thank you for attention!



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