

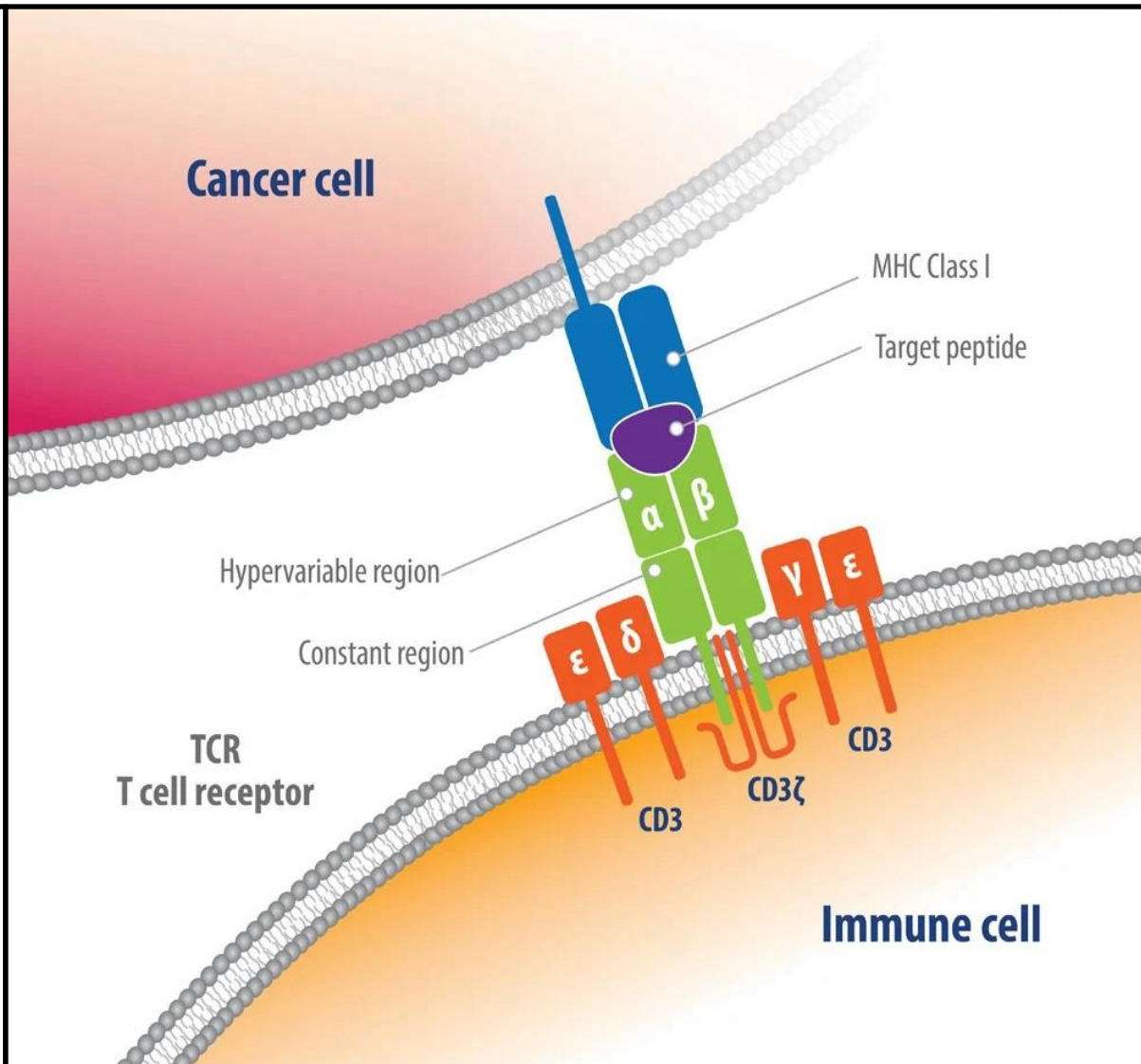


## Research Question:

This research investigates whether fine-tuning the RFdiffusion model on experimentally validated TCR-pMHC complex data can enable accurate, structure-conditioned generation of TCR CDR3β backbones with high binding complementarity to target epitopes.

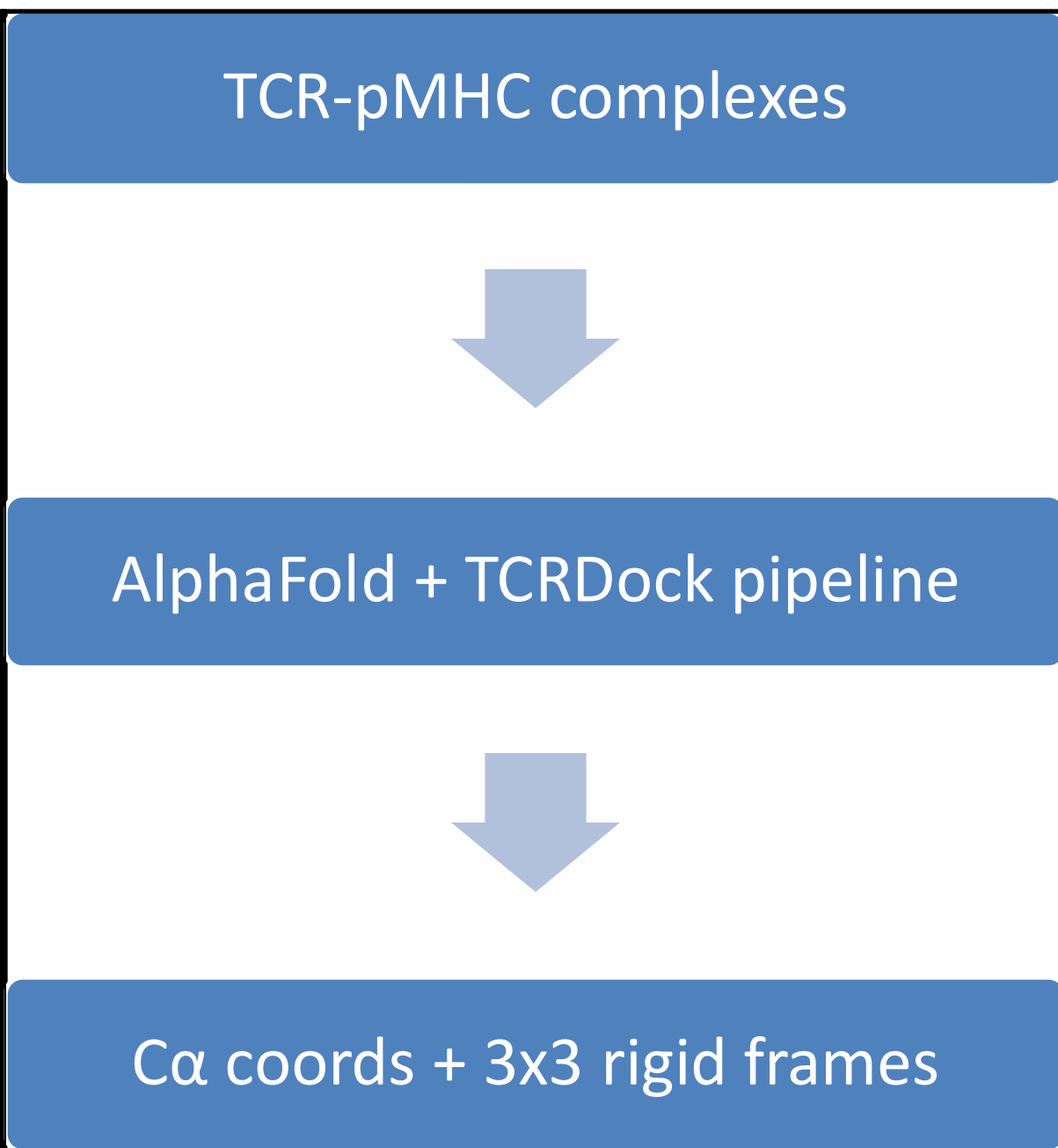
## Background:

- T cell receptors (TCRs) recognize small antigen peptides presented by MHC molecules on target cells.
- Binding specificity is driven by the α and β chains, especially the CDR3β region, which contacts the peptide directly.
- Proper recognition activates the immune system to destroy abnormal or infected cells.
- De novo TCR design could transform immunotherapy.
- Existing models lack 3D structural awareness, unlike RFdiffusion (Baker Lab, 2023), which generates protein backbones using rotationally equivariant diffusion.



## Data Collection and Filtering:

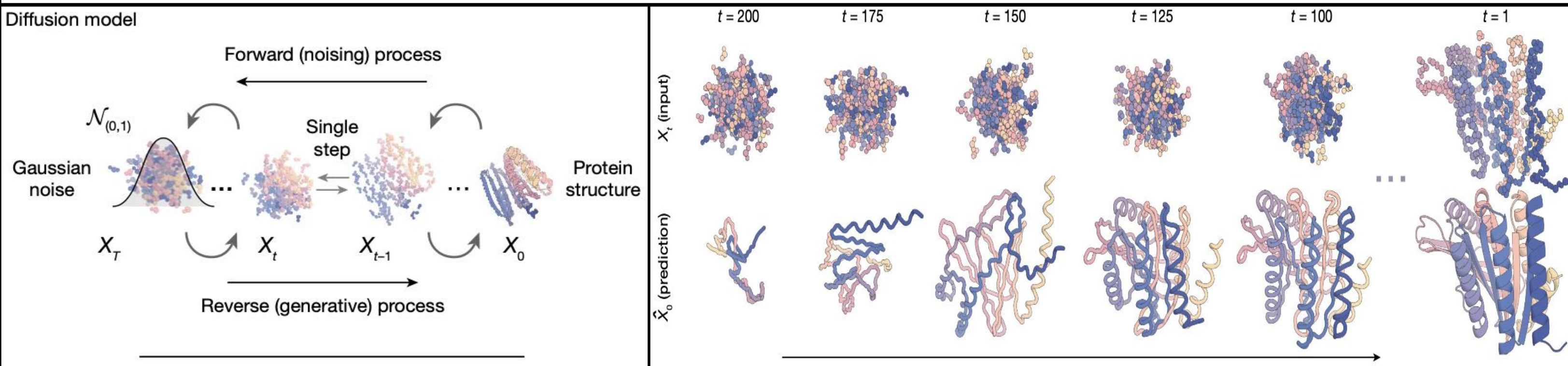
- Source:** Experimentally solved TCR-pMHC complexes from the TCR3d database (~300 observations).
- Fields:** {*Peptide*, *TRBV*, *TRBJ*, *CDR3β*, *PDB ID*}
- Annotation:** V/J gene calls and numbering via ANARCI.
- Processing:** TCRDock + AlphaFold refinement
- Filtering:**
  - α RMSD ≤ 2 Å vs experimental
  - Human only (restrict to HLA-A\*02:01 allele)
  - Binders (positive pairs) only
- Output:** Extract α coords + rigid frames for pMHC (condition) and TCR/CDR3β (target) → inputs for RFdiffusion fine-tuning.



LLFGYPVYV (peptide)	TRBV6-5 (TRBV)	TRBJ2-1 (TRBJ)	CASRPGLAGGRPEQYF (CDR3β)	1ao7 (PDB ID)
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## Model Framework and Training Plan:

- Fine-tuning RFdiffusion, a denoising diffusion probabilistic model that generates 3D protein backbones by iteratively denoising residue coordinates and orientations.
- Each residue represented as a rigid frame (Cα coordinate + 3×3 rotation matrix).
- Conditioned on epitope (pMHC) structure and interface hotspot residues to guide TCR-epitope complementarity.
- Trained to minimize MSE loss between predicted and native residue frames, learning to reverse the noise process.
- Uses self-conditioning to stabilize denoising and improve sample consistency.
- Generated backbones passed to ProteinMPNN for sequence design compatible with predicted structures.
- Evaluated using α RMSD, interface TM-score, and AlphaFold/TCRDock revalidation for structural plausibility.



## Future Work:

- Implement and fine-tune the proposed epitope-conditioned RFdiffusion model using curated TCR pMHC structural data.

## References:

- Watson, J.L., Juergens, D., Bennett, N.R. *et al.* De novo design of protein structure and function with RFdiffusion. *Nature* **620**, 1089–1100 (2023). <https://doi.org/10.1038/s41586-023-06415-8>
- J. Dauparas *et al.*, Robust deep learning-based protein sequence design using ProteinMPNN. *Science* **378**, 49-56 (2022). DOI: 10.1126/science.add2187
- <https://tcr3d.ibbr.umd.edu/>