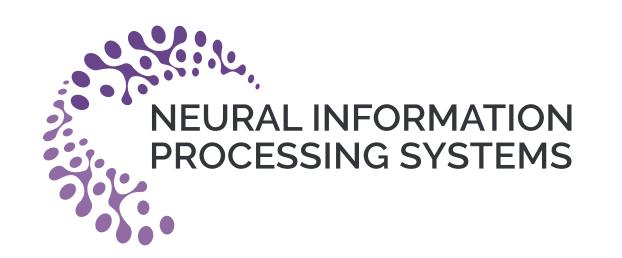


# BioBO:



### Biology-informed Bayesian Optimization for Perturbation Design

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### Motivation

- CRISPR screens enable gene perturbation studies, but exhaustive testing is infeasible due to large number of perturbations.
- Bayesian Optimization (BO) helps to select informative genes via a probabilistic surrogate and acquisition.
- Limitation of existing works: Current BO strategies do not generally leverage prior biological knowledge and multimodal gene representations.

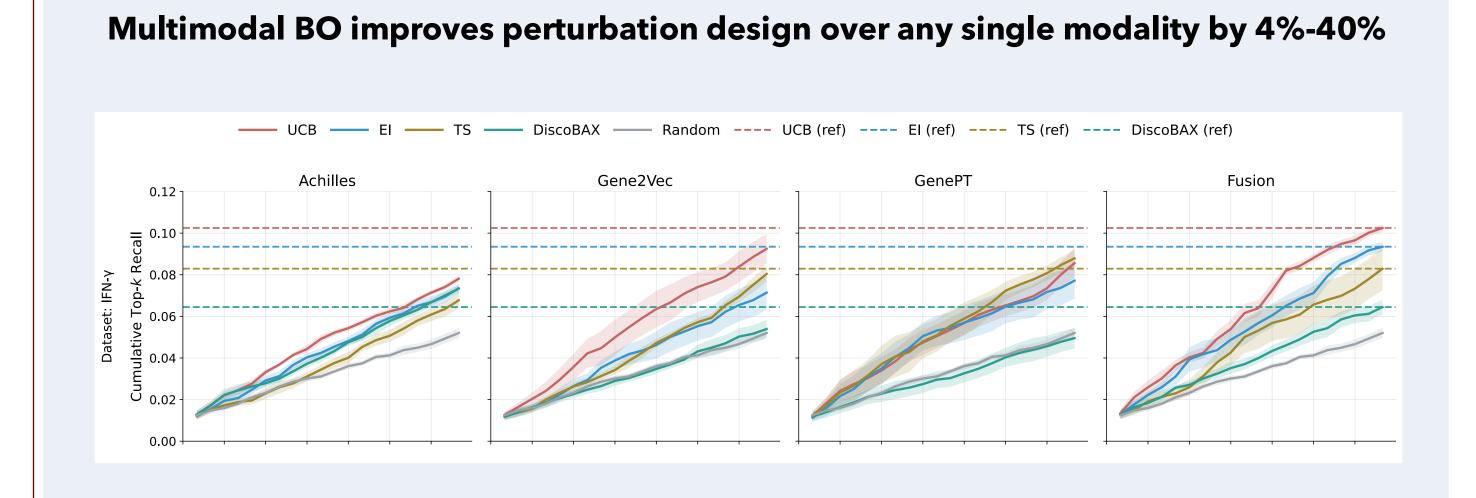
### Our contributions (BioBO)

- Using rich multimodal gene embeddings (Achilles, Gene2Vec, GenePT) for better surrogate modelling.
- **Biology inspired acquisition function**: novel Acquisition Functions (AFs) incorporating pathway enrichment analysis as a biological prior. Theoretically sound framework with no-harm guarantee under  $\pi$ -BO (Hvarfner et al., 2022).
- BioBO improves over conventional BO in labelling efficiency by 25-40% with interpretable biologically coherent pathways.

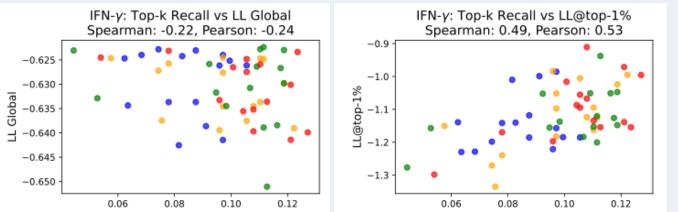
# BioBO pipeline Our methodological contributions Enrichment analysis prior Pathway scores Modality fusion BNN Acquisition from large unlabeled gene set Wet lab perturbations Multi modal gene descriptors

#### Method

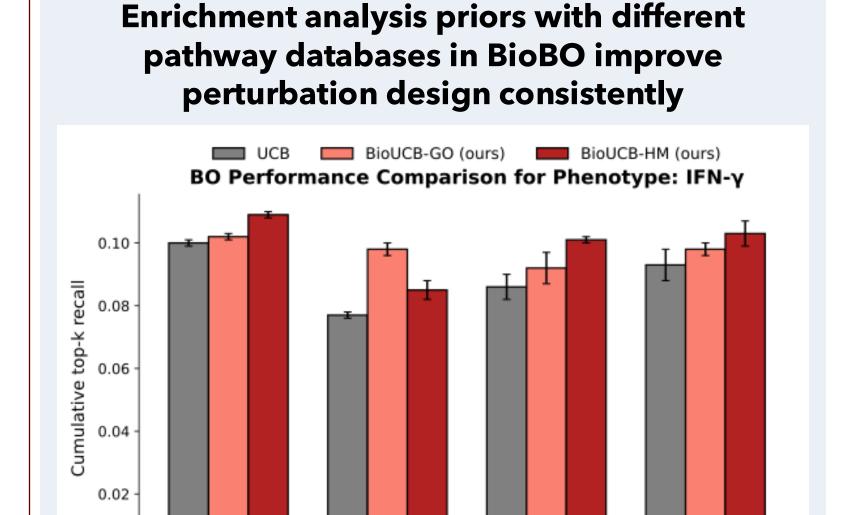
- Surrogate: Two-layer Bayesian Neural Networks (BNN) over fused/single embeddings
- EA prior: Combine odds ratio & p-value into a pathway score; map to gene-wise probabilities
- AFs: Baselines: UCB / EI / TS / DiscoBax; Ours: BioUCB/ BioEl / BioTS (multiply UCB / EI / TS by EA-derived prior  $\pi(x)$  with no-harm guarantee under  $\pi$ -B0 (Hvarfner et al., 2022) )
- Datasets from GeneDisco and DiscoBax benchmark (Lyle et al., 2023)



# Multimodal BO improves design by improving the surrogate near optimum IFN-y: Top-k Recall vs LL Global IFN-y: Top-k Recall vs LL@top-1%



BO design accuracy (top-k recall) correlates more with the surrogate model's loglikelihood near the optimum (top-1%, right) than across the whole space (global, left).

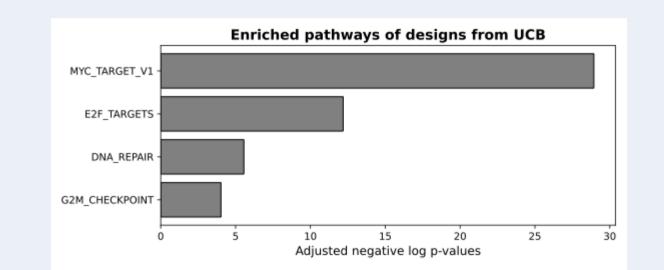


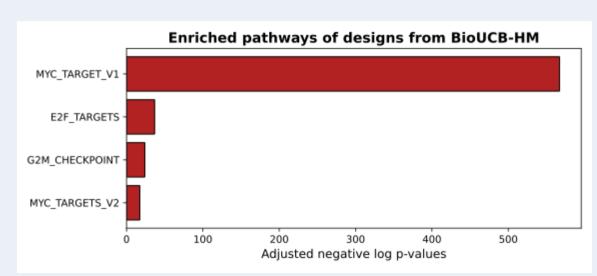
Achilles

0.00

Fusion

## BioBO produces stronger enrichment signals in pathways closely tied to IFN- $\gamma$ regulation in T cells than BO





Designs from BioUCB-HM (AF with enrichment based prior) are more significantly enriched (red) in pathways closely tied to IFN- $\gamma$  regulation in T cells compared with vanilla UCB (grey).

1. Hvarfner, Carl, et al. " $\pi$ BO: Augmenting Acquisition Functions with User Beliefs for Bayesian Optimization" ICLR 2022.

Gene2Vec

2. Mehrjou, Arash et al. "GeneDisco: A Benchmark for Experimental Design in Drug Discovery" ICLR 2022.

GenePT

3. Lyle Clare, et al. "Discobax: Discovery of optimal intervention sets in genomic experiment design." ICML 2023.