# **Data Science Meets Biology**

Boosting phenotypical profiles for early drug discovery: team Air



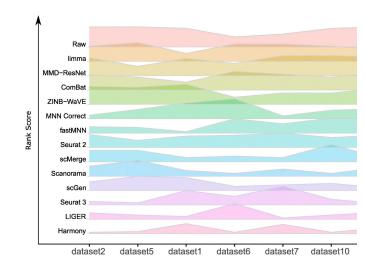


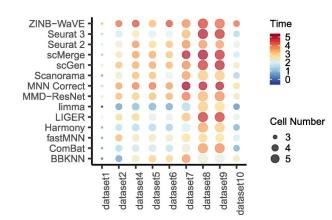
# Batch effects correction: background

Batch effects are data variations due to unintended technical differences in reagents, processing times, equipment, or experimental platforms.

Batch effect correction is necessary to detect true biological differences.

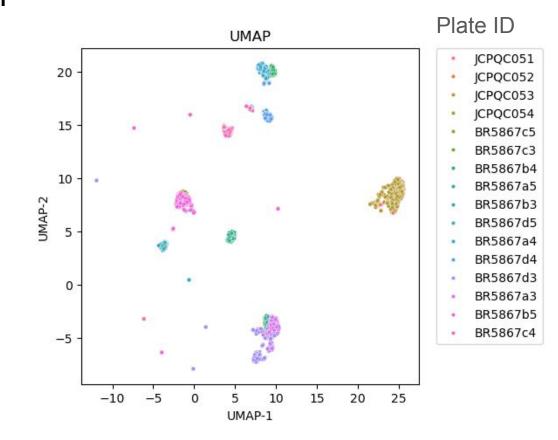
- Linear methods (Combat and Sphering)
- Neural-network based methods (scVI and DESC)
- Mixture-model based method (Harmony)
- Nearest neighbor-based methods (MNN, fastMNN, Scanorama, Seurat-CCA, and Seurat-RPCA) (compared by Arevalo et al., 2024; Tran et al., 2020)



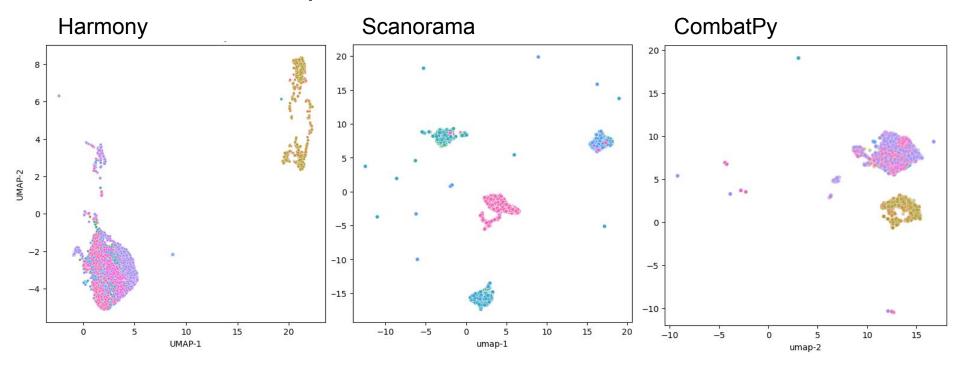


## Batch effects correction

- Harmony
- Scanorama
- CombatPy



## Performance comparison



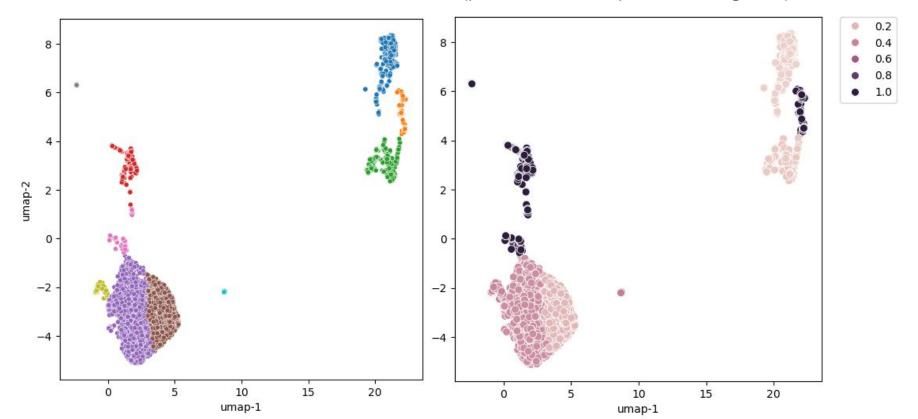
Recommended for all types of batch effects, is faster and Python-based (Arevalo et al., 2024; Korsunsky, 2019; Tran et al., 2020)

Bad performance

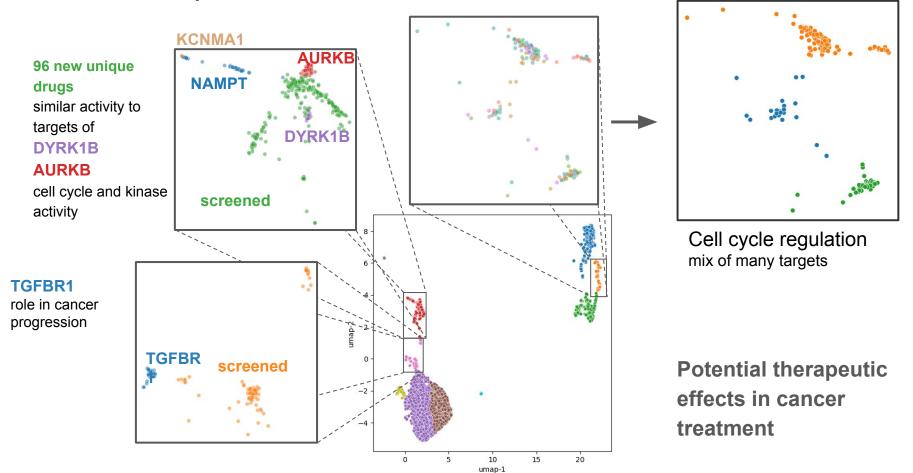
Positive controls did not cluster with any treatment

# Spectral clustering

Cluster purity (positive control / positive + negative)



# Which compounds show effects?



## Activity of tested drugs

## Group 1:

### Effect similar to targets of:

### 2 important protein kinases

**AURKB** - key regulator of mitosis, inhibitors are known to suppress tumor growth **DYRK1B** -promoting resistance to apoptosis

#### And less similar to:

**(NAMPT) -** overexpressed in cancer cells regulating NAD+ levels  $\rightarrow$  DNA repair, gene expression, and stress response.

## Group 2:

### Effect similar to targets of:

**TGFBR1** - tumor suppression (early), promotion (late)

Potential therapeutic effects in cancer and immune disorders

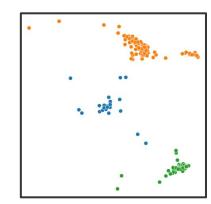
## Common pathways of target genes clustered together

# Subgroup 1: cancer, signaling, apoptosis, development, and immune responses

- 1. Cell cycle regulation: ATM, BRD4, CCND1, CDK2, CDK7, CDK9
- 2. Apoptosis: BAX
- 3. Immune response: BTK, CSF1R, FPR1, HCK, FLT3, ITGB2, LCK, LYN
- 4. Signal transduction: BTK, DDR2, FPR1, KRAS, LCK, PAK4, PDPK1, PIK3CG, PLD1, PRKCE, RGS4, TGFBR1
- 5. Neurotransmission: GRIN2A, OPRM1, RGS4
- 6. Stress response: HSP90AA1, HSP90AB1
- 7. Calcium signaling: CATSPER4, TNNC1

# Subgroup 2: signaling, development, metabolism, and immune responses

- 1. Cell growth and differentiation: RET, CDK7
- 2. Inflammatory response: TNF
- 3. Vascular development and homeostasis: KDR, PTGIR, S1PR1
- 4. Sugar metabolism: AKR1B1



# Subgroup 3: cell proliferation, differentiation, and responses to external stimuli

- 1. Cell cycle regulation: CCND1, FOXM1, PLK1
- 2. Signal transduction: CHRM2, FLT3, OPRL1, RGS4, S1PR1
- 3. Stress response and protein folding: HSP90AA1, HSP90AB1
- 4. Metabolic pathways: HSD11B1, PPARD, IMPDH1
- 5. Cytoskeletal dynamics: TUBB, TUBB3, TUBB4B

# Thank you!

Organizers

Bogdan Avanesy Ekaterina Vasileva

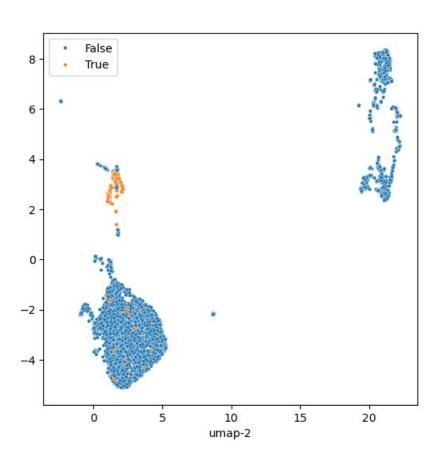
Mentors

Bayer & MDC





## Class0 drug specificity





JCP2020\_53074 JCP2020\_51978 JCP2020\_51993 JCP2020 53037 JCP2020 51979 JCP2020\_51940 JCP2020\_54031 JCP2020\_53827 JCP2020\_53215 JCP2020\_51813 JCP2020\_53888 JCP2020\_53918 JCP2020\_51710 JCP2020\_51780 JCP2020\_51801 JCP2020\_51931 JCP2020\_52638 JCP2020 52219 JCP2020\_53225 JCP2020\_53208 JCP2020\_52789 JCP2020\_54161 JCP2020\_52235 JCP2020\_44320 JCP2020\_52633 JCP2020\_53218 JCP2020\_52624 JCP2020\_52259 JCP2020 52677 JCP2020 52658 JCP2020 53103 JCP2020\_53526 JCP2020\_52217 JCP2020\_52650 JCP2020\_53271 JCP2020\_53443 JCP2020\_52207 JCP2020\_52695 JCP2020\_51770 JCP2020\_51708 JCP2020 51706 JCP2020 53849 JCP2020 53001 JCP2020 54033