

Data-driven pharmacogenomic classification and drug-response target prioritisation

Unsupervised reclassification of drug-ADR genetic associations,
with a latent (drug-triggered) PGx priority score – no expert input

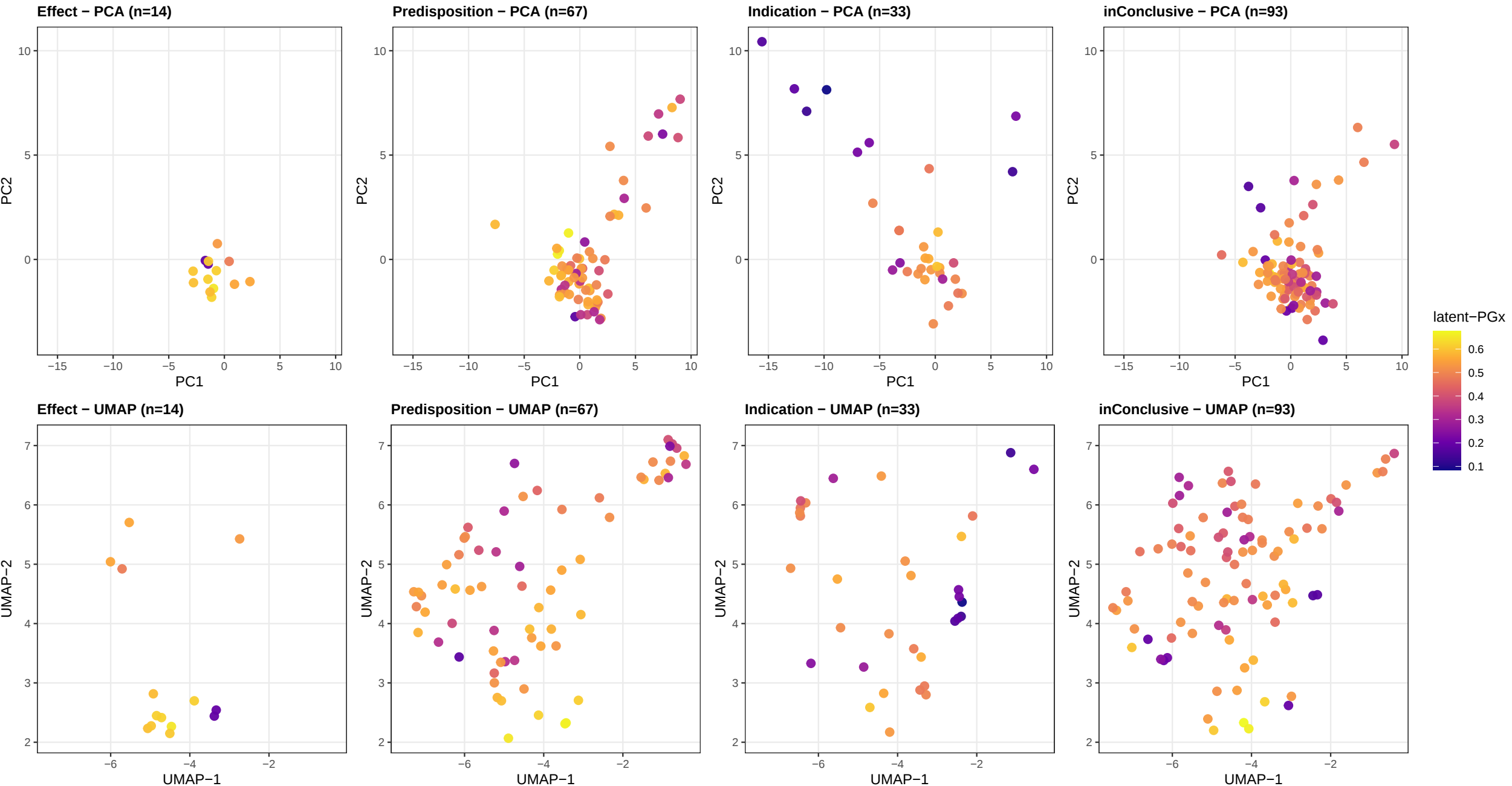
Model parameters

Only VARIANT descriptors and PHENOTYPE descriptors – no gene-specific parameters

- **GWAS backbone:** absolute effect size and $-\log_{10} P$ for four phenotype arms of each variant.
- **Cross-phenotype contrasts:** drug-ADR effect minus disease / dose / prescription effects.
- **Temporal:** counts, median years and onset spread of events BEFORE vs AFTER first prescription.
- **Held-out test set:** genotype counts, case-control frequency difference, effect calibration
(test-vs-discovery odds-ratio difference and ratio).
- **Variant function (VEP):** consequence severity, biotype, population allele frequency.
- **Pharmacovigilance:** FAERS disproportionality. **Internal pleiotropy:** variant specificity.
- No gene-specific parameters (no tissue, druggability, gene-disease scores) and no arbitrary thresholds – continuous values and canonical ordinals only.

Each expert class, coloured by latent-PGx score

Only the named expert class is shown in each panel (top: PCA, bottom: UMAP); shared latent-PGx colour scale



What the parameters measure

Four GWAS views of one variant resolve drug-specificity

- drug-ADR arm: does the variant associate with the adverse event under the drug?
- disease-alone arm: does it associate with the disease in the general population (no drug)?
- dose arm: does it associate with the dose patients are titrated to?
- prescription arm: does it associate with who is prescribed the drug?
- **Drug-specificity = drug-ADR effect dominating ALL three alternatives -> a drug-conditional signal.**
- **Low internal pleiotropy + concordant direction + replication = a specific, real drug effect.**

Classification – unsupervised methods

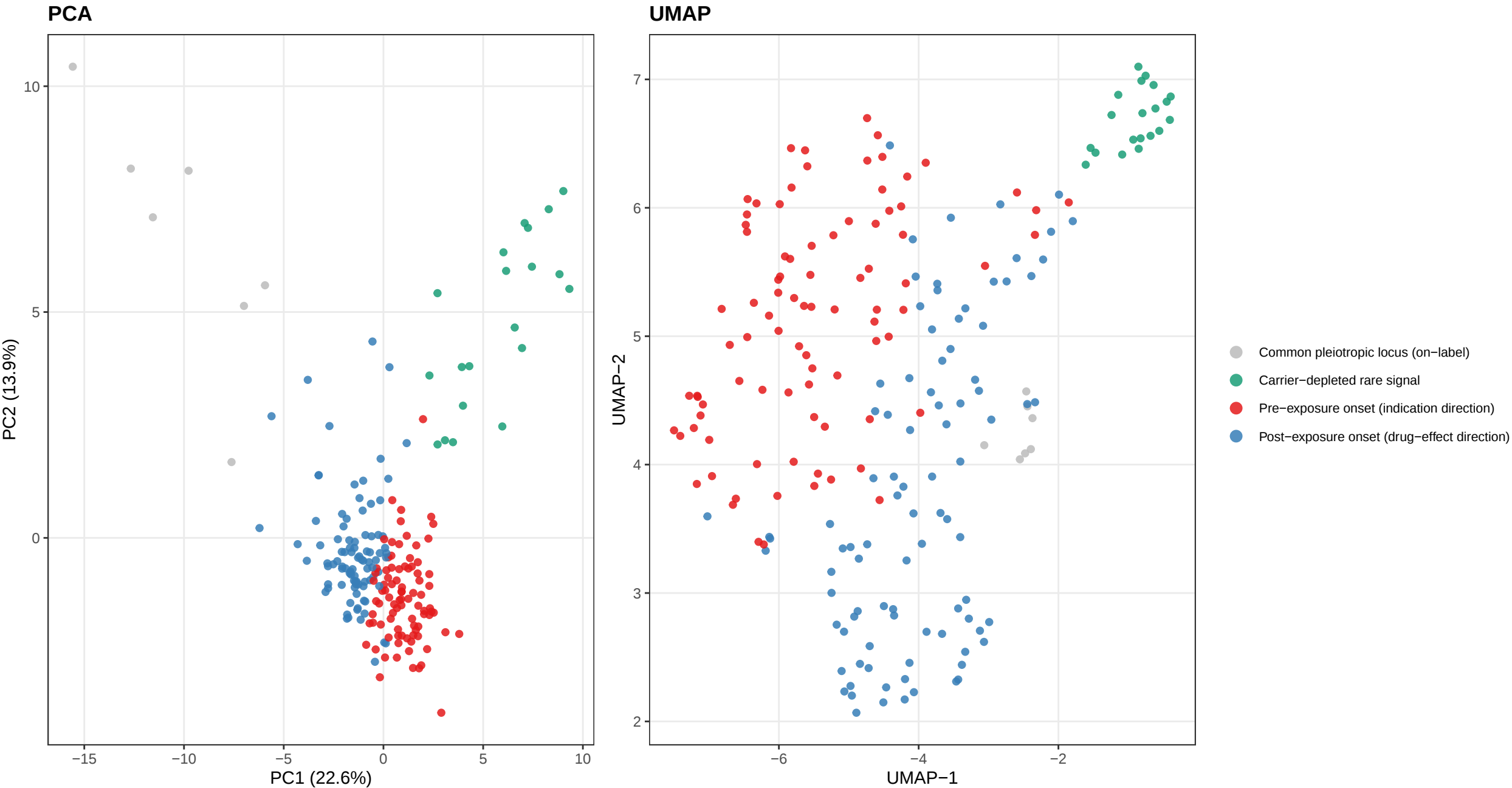
Four complementary algorithms; consensus over 100 seeds; cluster number chosen by the data

- **k-means: centroid / Euclidean partitioning; spherical clusters; k chosen by silhouette.**
- **GMM: Gaussian mixture (tied covariance); soft posterior membership; k chosen by BIC.**
- **LCA: latent class analysis; handles mixed-type features; k chosen by BIC.**
- **HDBSCAN: density-based; discovers k; isolates outliers as a separate group.**
- 100-seed consensus gives a soft membership and a per-association stability score,

so borderline associations are reported as borderline rather than forced into one class.

Data-driven classification of drug-response associations

Each point = one association; colour = unsupervised class



The discovered classes (by parameter signature)

Groups defined by GWAS parameters alone (self-contained model)

- **On-label indication: high prescription-arm signal; disease precedes the drug (3 sub-groups).**
- **Rare-variant predisposition: large effect size and SE, low allele frequency, drug-amplified.**
- **Drug-amplified mechanism (effect-enriched): drug-ADR dominates other arms, replicated –**
the group holding the candidate drug-triggered effects.
- **Low/moderate-evidence mass + weak-evidence residual: the large, low-signal background.**
- Self-contained features classify more coarsely than annotation-rich models,
but the latent-PGx prioritisation (next) is independent of the clustering.

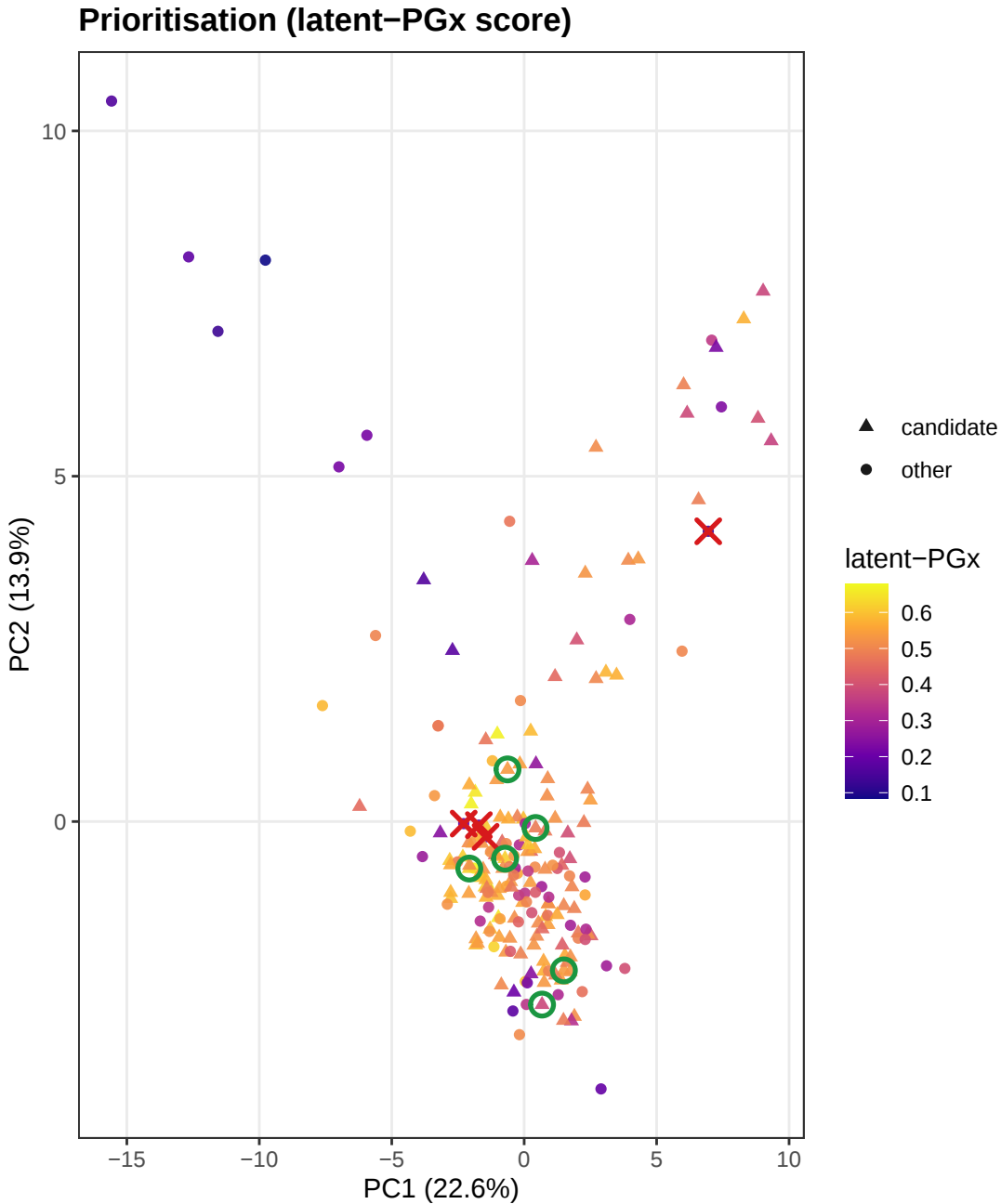
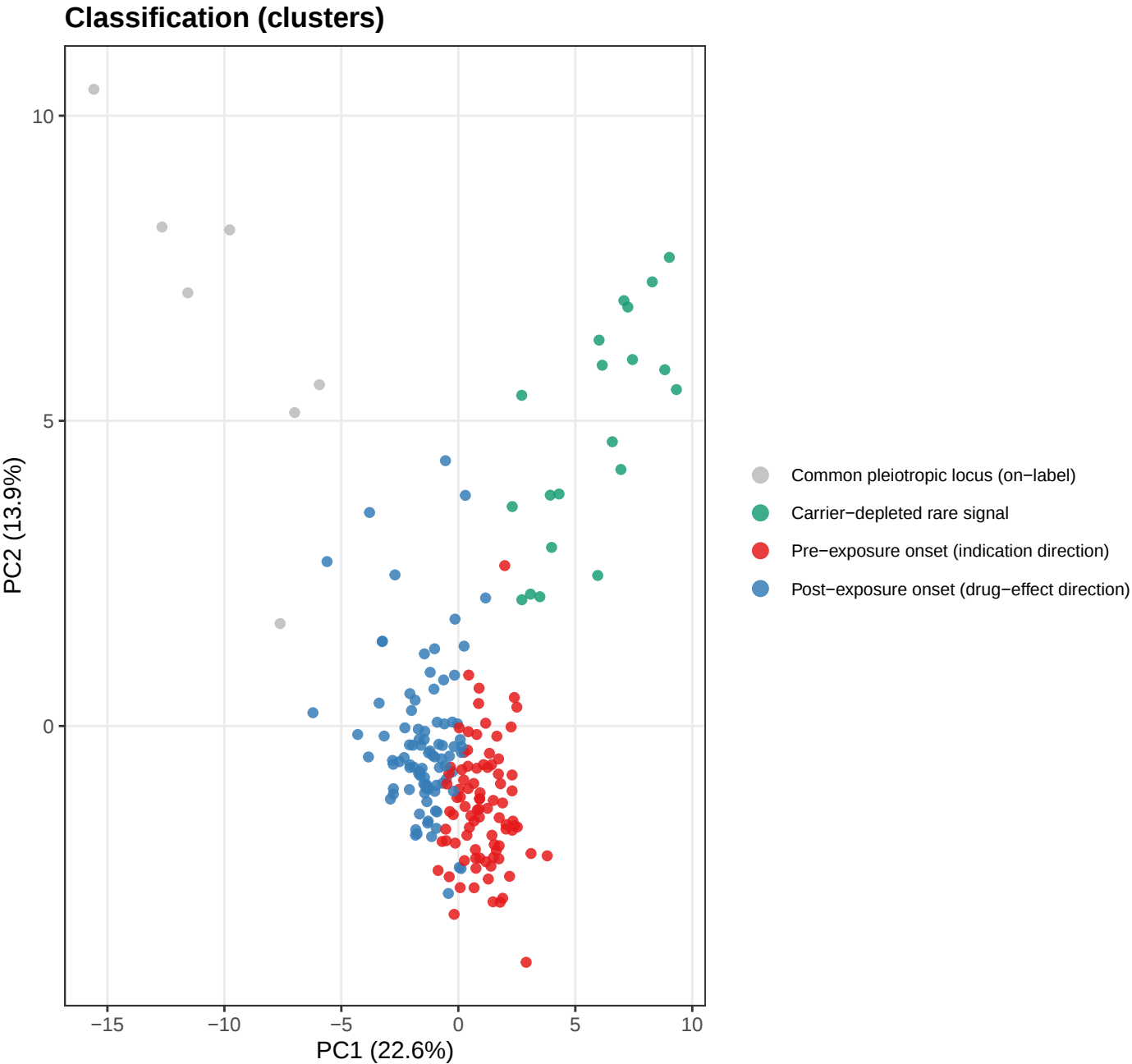
The pharmacogenomically interesting class

Latent predisposition realised only by the drug (gene x drug interaction)

- **Parameter signature: drug-ADR effect AND $-\log_{10} P$ dominate disease, dose and prescription arms,**
concordant direction, replicated in validation, and low internal pleiotropy.
- Silent in the general population (low disease-alone signal); unmasked by the drug.
- Distinct from disease-driven variants (which associate with the disease without the drug).
- Captured by a continuous latent-PGx score, layered on top – not a clustering input.
- **The most interesting and the rarest class.**

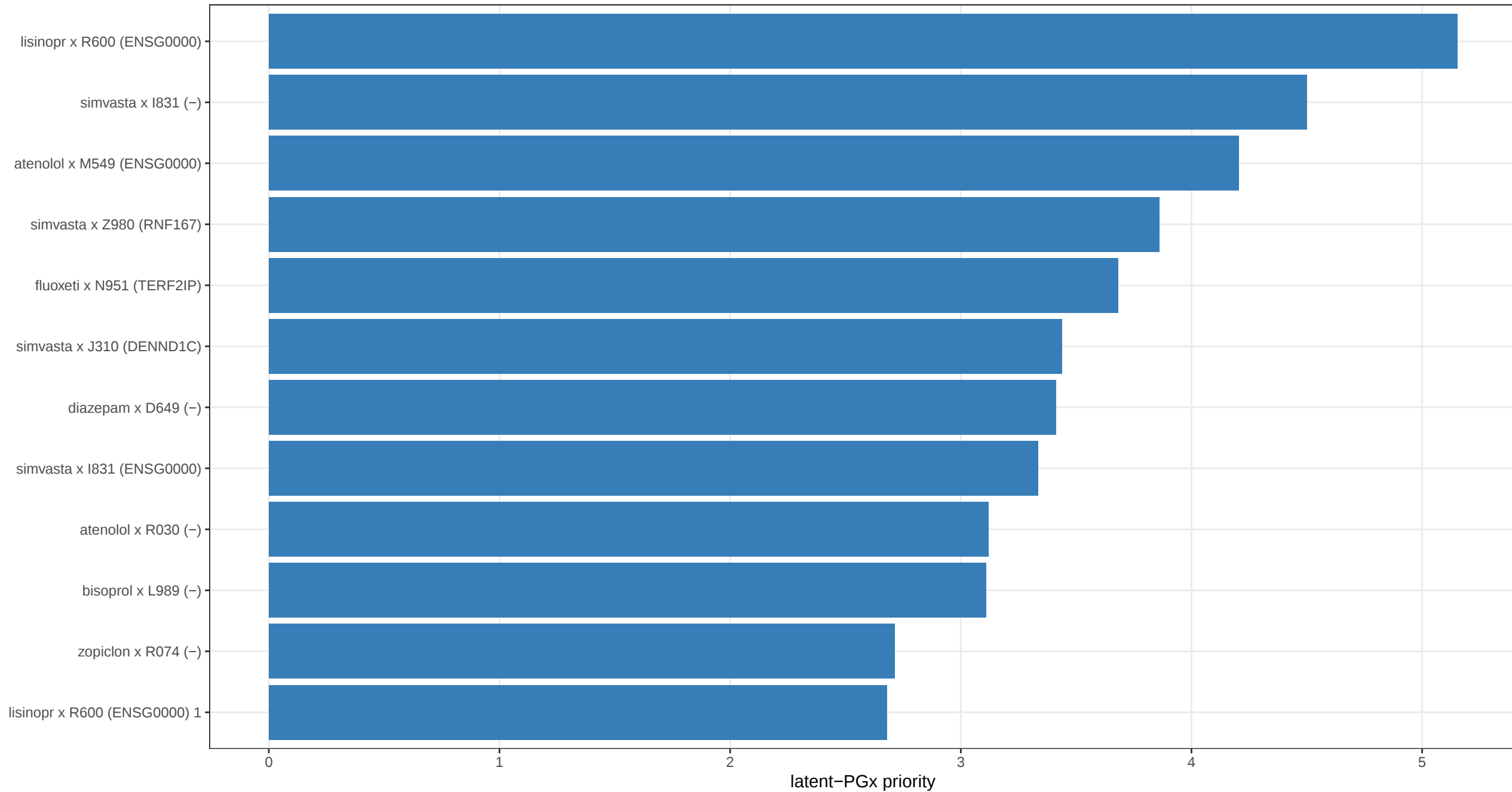
Same PCA space: class membership vs latent-PGx priority

Left: unsupervised class. Right: latent-PGx score (triangles = gated candidates)



Top latent-PGx priority candidates

Ranked drug-gene-ADR triples for fine-mapping and replication



Example 1 – a latent-PGx target

lisinopril x angioedema (R60.0)

- ACE-inhibitor angioedema is a known drug-triggered reaction (bradykinin pathway).
- **Latent-PGx score high (+2.7): drug-ADR effect dominates disease, dose and prescription arms.**
- Internal pleiotropy = 3 (variant-specific); concordant direction; replicated.
- Lead variant is a tag SNP -> flagged FINE-MAPPING NEEDED.
- Archetype: a real gene x drug pair where the causal variant must be resolved.

Example 2 – correctly excluded

Factor V Leiden (F5) x thrombophlebitis (I80.2)

- F5 is a strong genetic risk for thrombosis – but it causes disease WITHOUT any drug.
- **Latent-PGx score deeply negative (–7.0): flagged MANIFEST RISK.**
- High disease-alone signal; drug-specificity near zero; internal pleiotropy elevated.
- Disease-driven / genetic comorbidity – not pharmacogenomics.
- Shows the score separates drug-triggered effects from plain genetics.

Example 3 – deprioritised

ramipril x acute myocardial infarction (I21.9)

- Ramipril is prescribed after MI – confounding by indication, not an adverse reaction.
- **Latent-PGx score negative: the drug-ADR signal does not exceed the prescription arm.**
- High prescription-arm association (the variant tracks who receives the drug).
- Correctly placed in the indication class and excluded from latent candidates.
- Demonstrates indication confounding is separated by the prescription-arm parameter.