

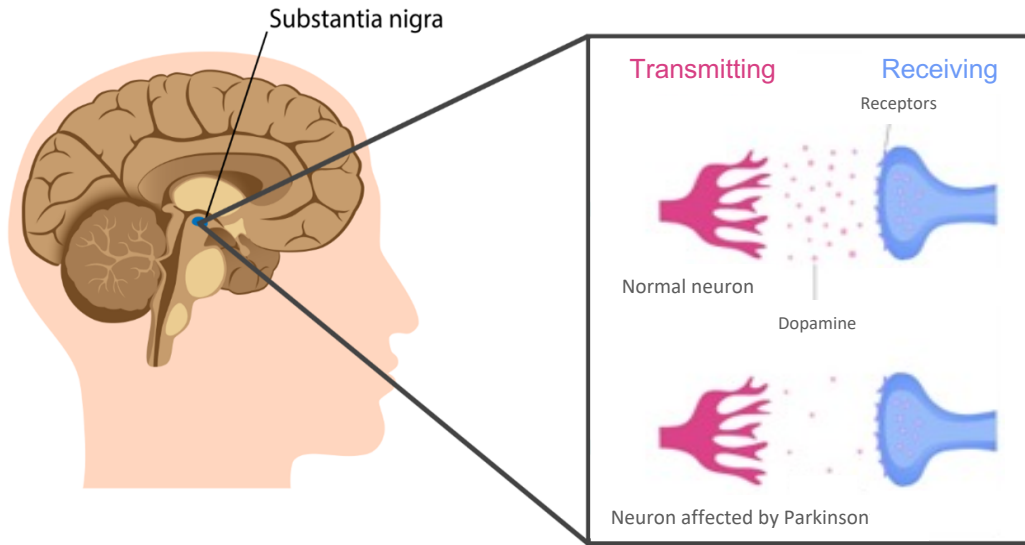
Machine learning for the study of Parkinson's Disease diagnosis and associated mechanisms

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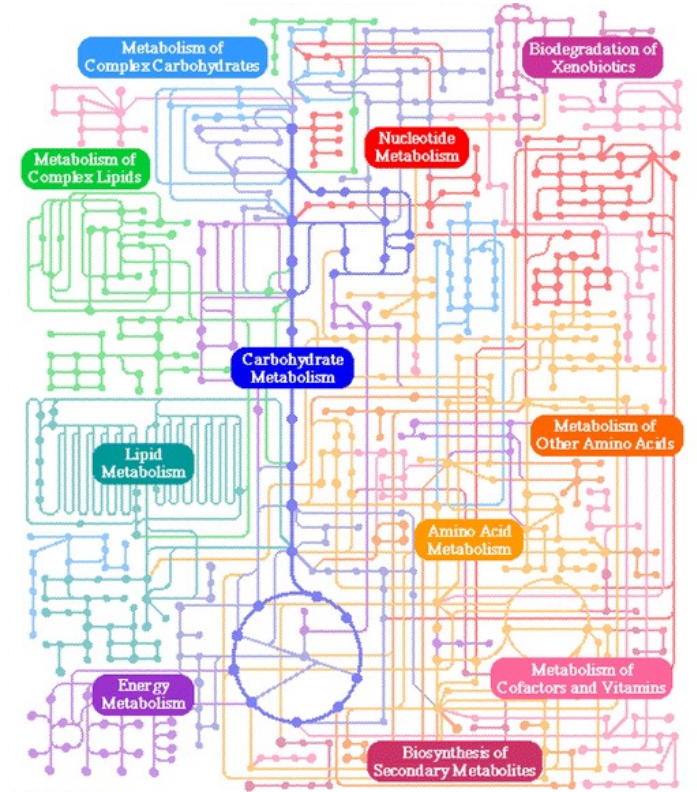
Parkinson's Disease & higher order functional representations



Source: Adapted from shutterstock

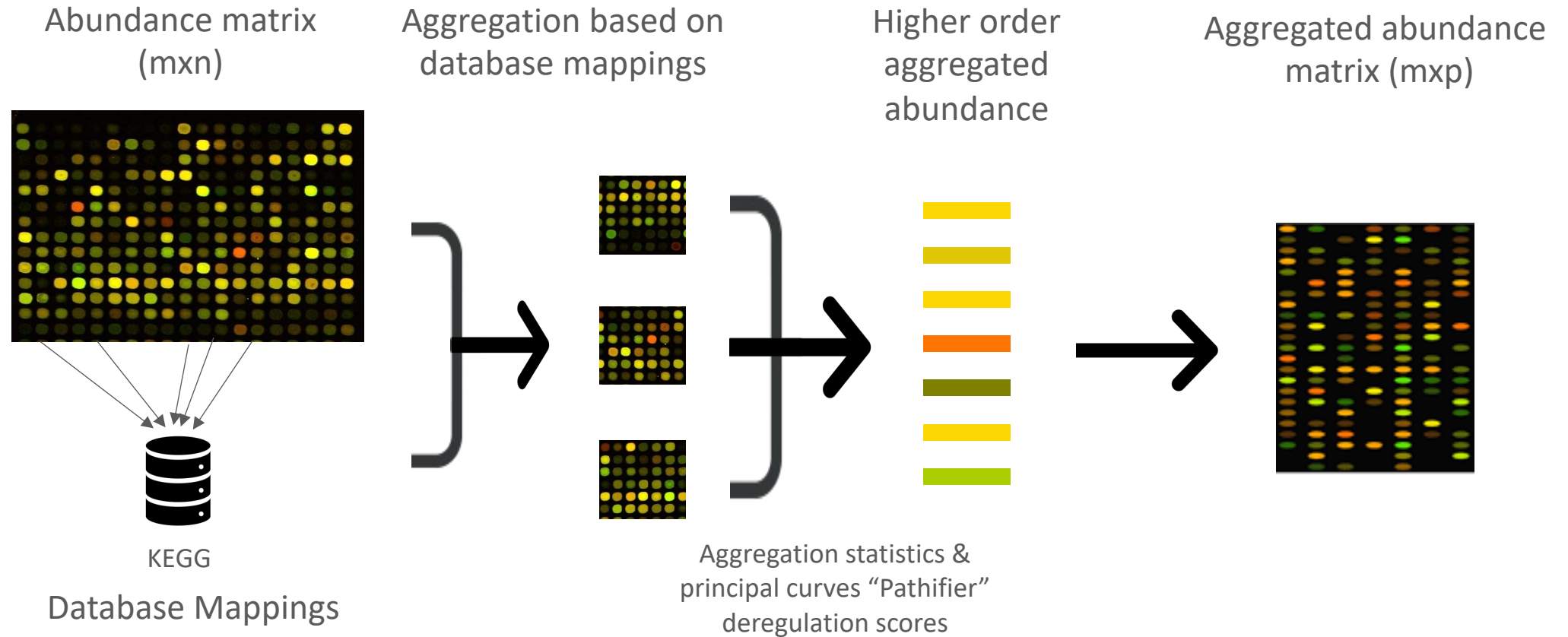
- Single gene mutations?
- Mitochondrial genetics?
- Environmental factors (toxins)?

Diagnosis is clinical & difficult



Schematic representation of metabolic networks
Source: *The Origin and Evolution of Metabolic Pathways: Why and How did Primordial Cells Construct Metabolic Routes?*

High throughput omics data into higher order functional features



KEGG = Kyoto Encyclopedia of Genes and Genomes

m = number of samples

n = number of single-level features (i.e. genes, metabolites, etc)

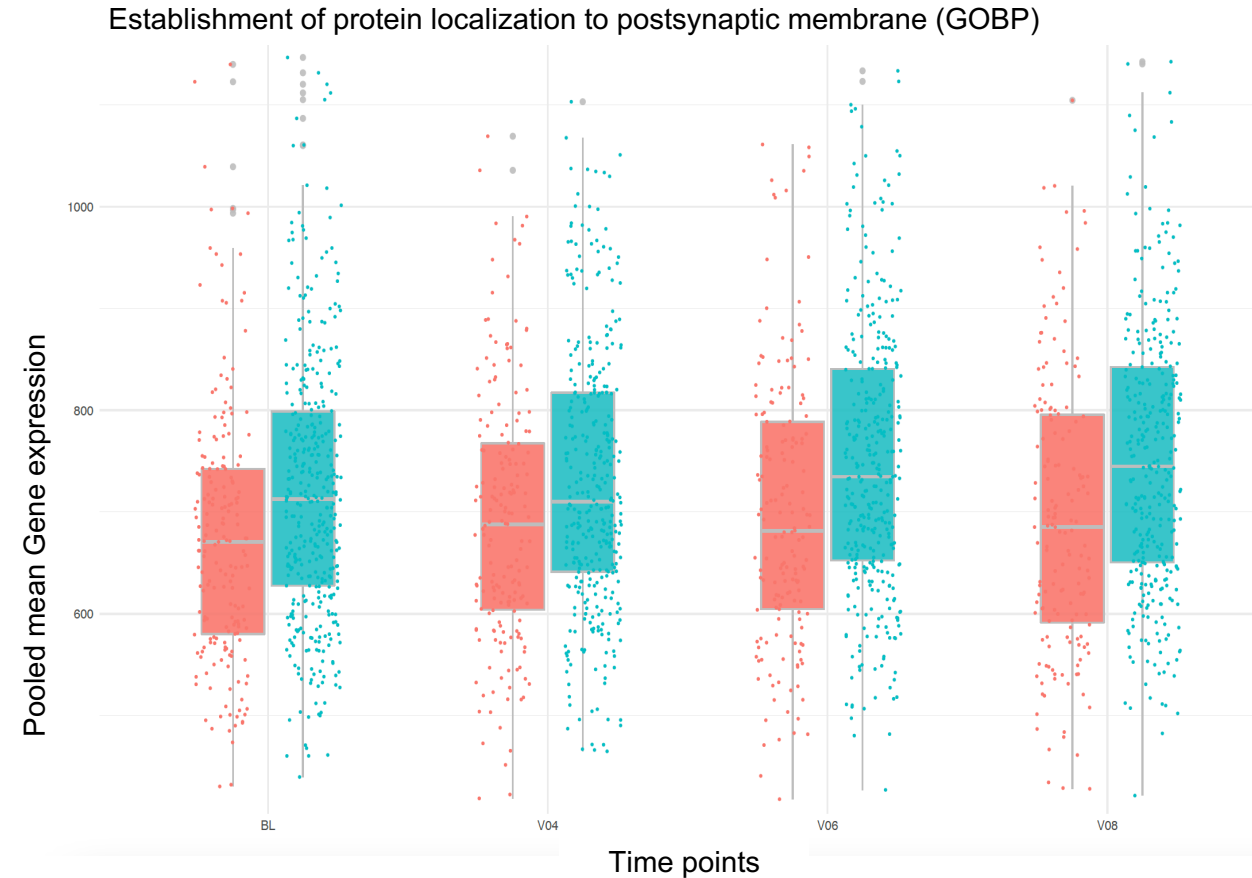
p = number of higher order functional aggregates (e.g. number of pathways)

Statistical differential analyses & time course analyses

Metabolomics aggregated data on 'de novo' patients

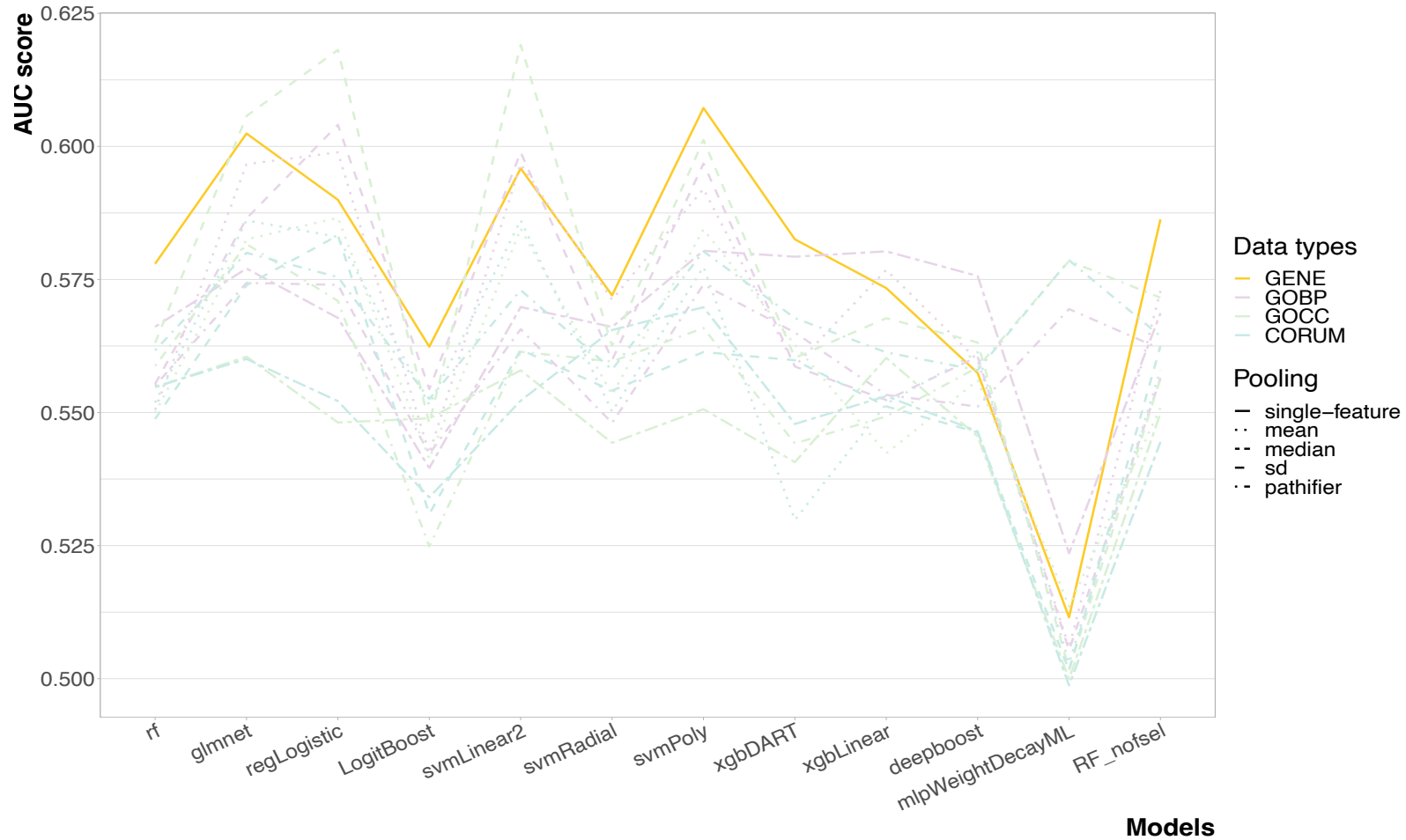


Transcriptomics aggregated temporal profiles



Predictive PD diagnosis with ML models on transcriptomics

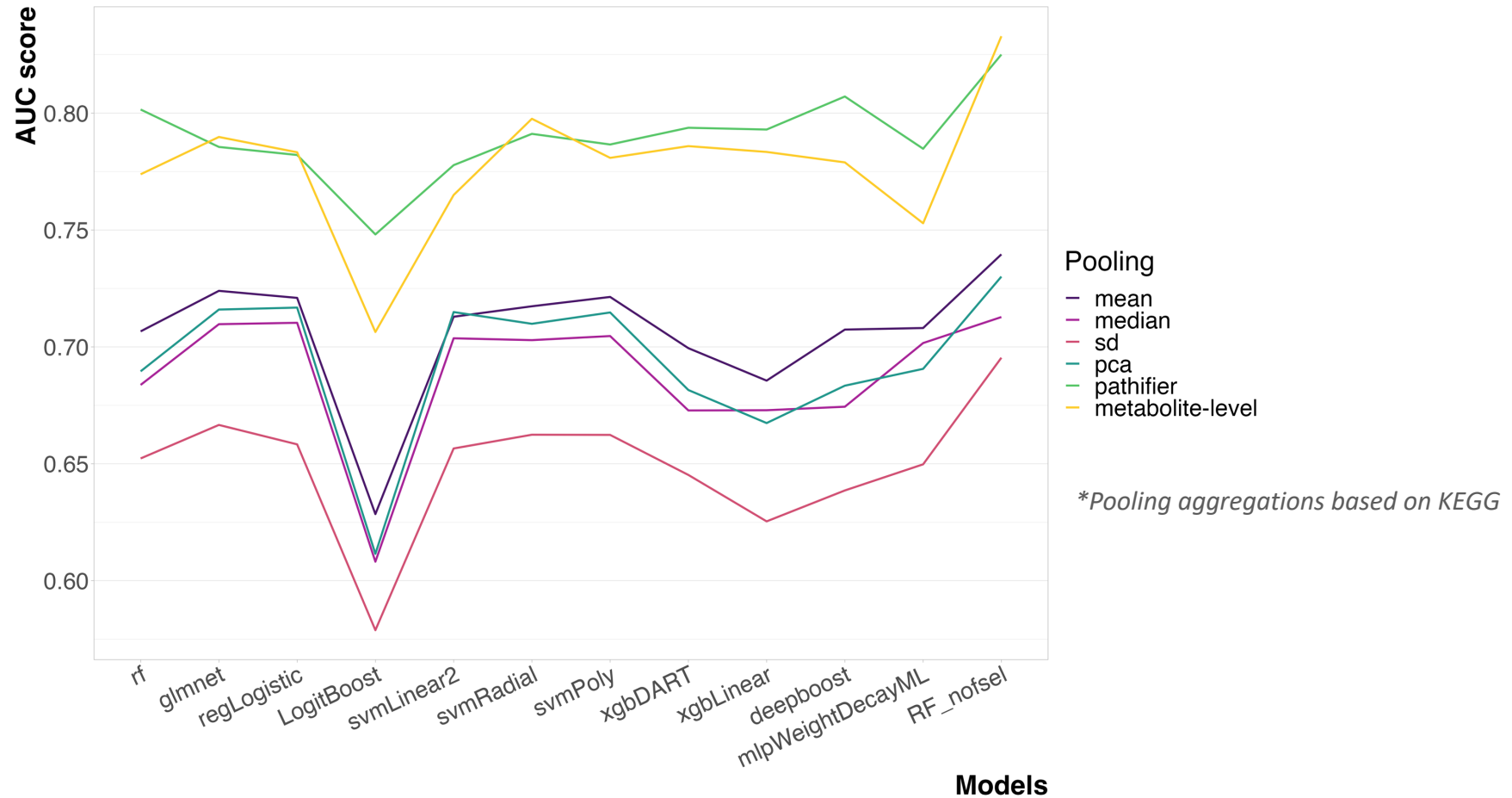
Line chart of crossvalidated AUC scores from models & pooling types on transcriptomics data



External two-level cross-validation was used (including nested feature selection)

Predictive PD diagnosis with ML models on metabolomics

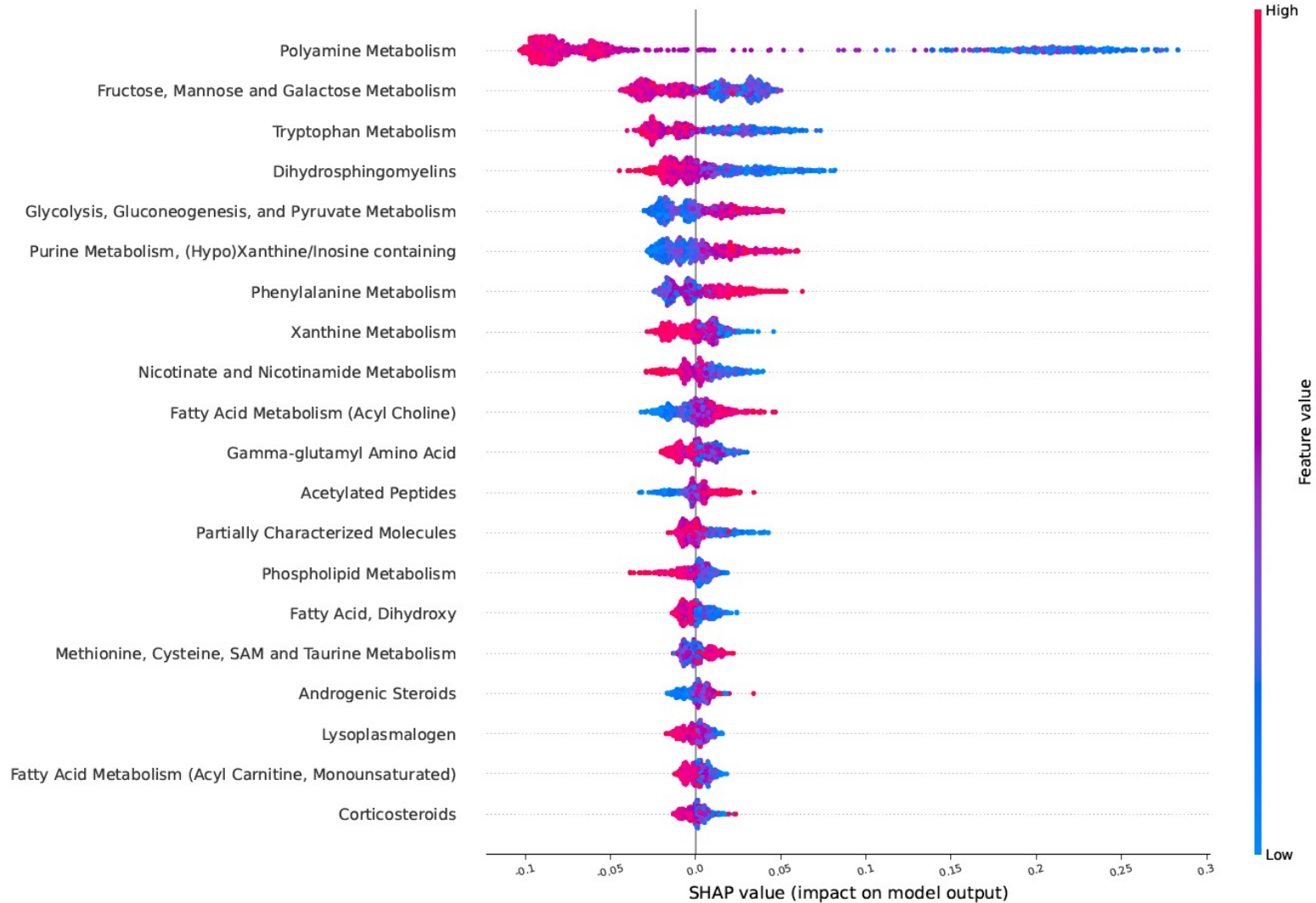
Line chart of crossvalidated AUC scores from models on metabolomics data*



External two-level cross-validation was used (including nested feature selection)

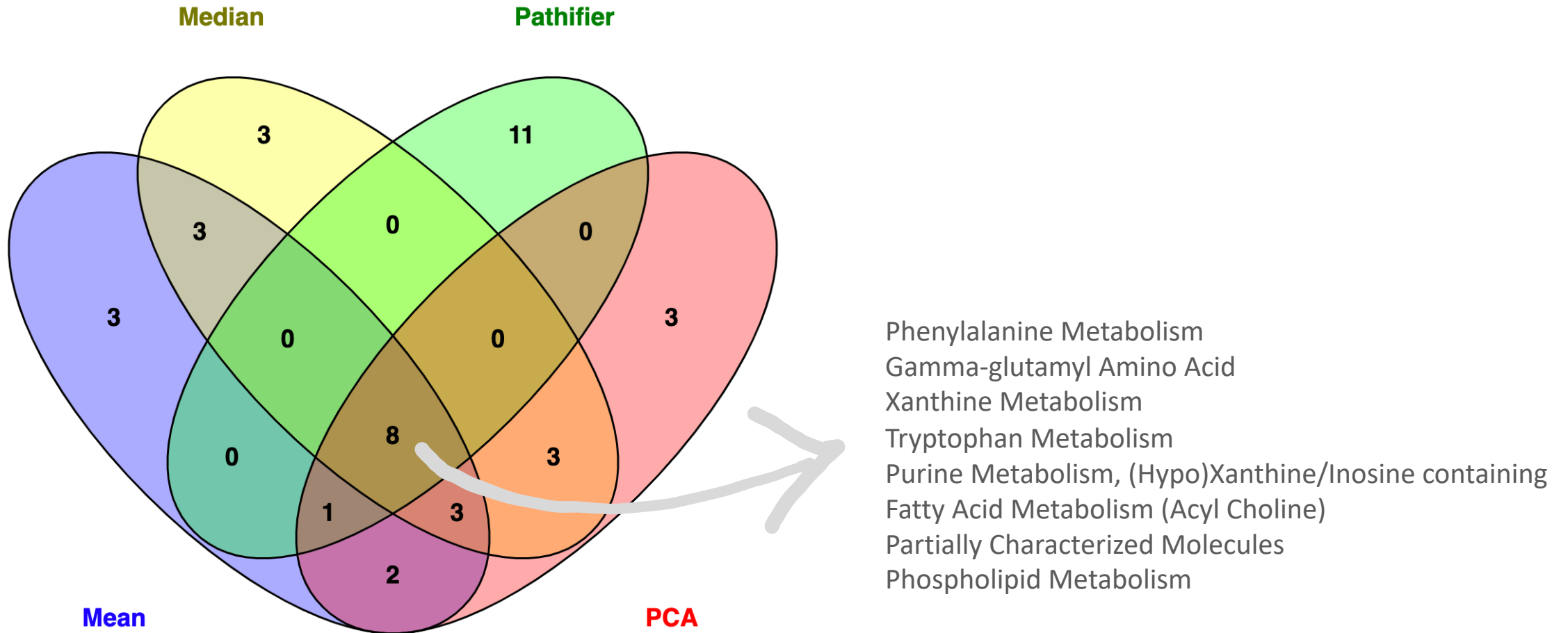
Relevant features from predictive PD diagnosis met/transc

Shap values of (pathifier -aggregated) KEGG metabolic pathways predictors on random forest model



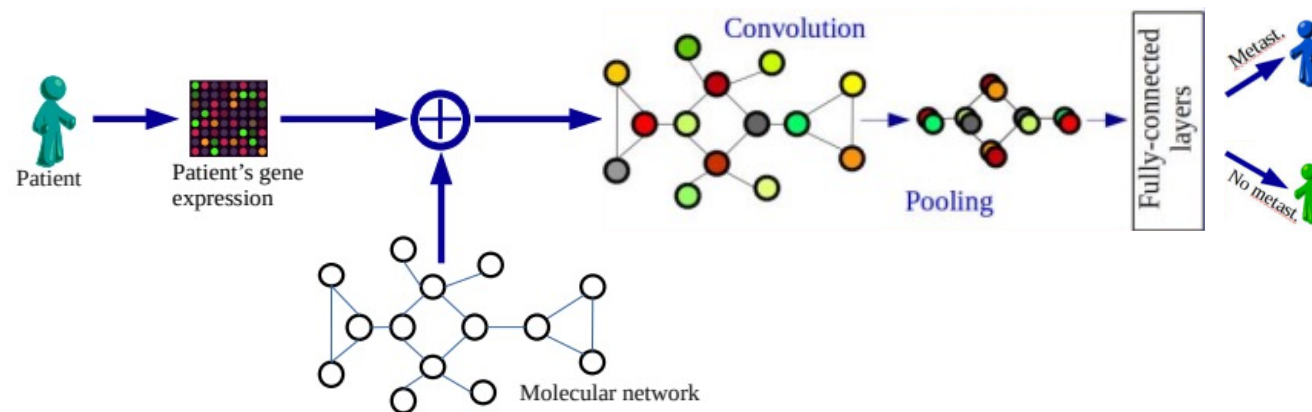
Relevant features from predictive PD diagnosis on metabolomics

Top 20 most relevant features per pooled aggregation based on shapley values



Limitations & outlook for future analyses

- ✗ Unknown confounders
- ✗ Large variability among PD patients makes identifying common trends difficult
- ✗ Data represents late stages of the disease
- ➔ Modelling other PD prognostic outcomes (e.g. motor dysfunction scores)
- ➔ Use a graph representation of the data via protein-protein interactions and metabolic networks



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Gene expression profile as a graph signal of the molecular network

Source: Chereda, H., 2022. Explaining decisions of graph convolutional neural networks for analyses of molecular subnetworks in cancer [Doctoral thesis, Georg-August-Universität Göttingen]

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Find my poster here!

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