## High-Sensitivity Pattern Discovery in High-Dimensional Heterogeneous Datasets

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## Outline

- Background
- Proposed Method
- Validation
- Evaluation
- Application
- Software

**Fundamental Problem** 

How to discover quickly and confidently latent associations within high-dimensional, heterogeneous datasets in the presence of noise and collinearity?



Image is from: http://freshgypsy.com/post/21824134701/rainbowpointelismart

### **Example: Multi-omics Integration**





### Proposed method HIERARCHICAL ALL-AGAINST-ALL ASSOCIATION TESTING

## Discretizing, Clustering, and Coupling



## Hypothesis Testing, FDR Correction





- Benjamini-Hochberg for FDR correction
  - Family of hypotheses
  - Level of hypotheses
  - All tested hypotheses

### Associations



### Method Overview





## Validation SYNTHETIC DATA

# Simulated datasets

- Two synthetic datasets:
  - 12 features and 200 samples
  - 3 block of features with each datasets
  - Features within blocks are linearly correlated
  - The second dataset is spiked linearly from the first dataset
  - Each block from the first dataset is linearly correlated with one block from the second dataset



## Associations



A cluster from the second dataset



# Evaluation **PERFORMANCE**

# Simulated Datasets

- 10 different simulated datasets are used:
  - Datasets scaled from 4 features and 100 samples to 24 features and 500 samples.
  - These datasets have simulated clusters in their features with linear association within and between them.

## Naïve AllA vs. HAllA



- Higher statistical power in HAIIA
- Benjamini-Hochberg has controlled the FDR in the naïve All-against-All approach

# HAIIA vs. AllA: ROC Curve





## Application GENERA VS. METABOLITES

## The Multi-Country DIABIMMUNE Study





### Two datasets for 104 samples :

- 16S rRNA gene sequencing to profile 20 genera
- 284 labeled metabolites





Aleksandar Kostic Tommi Vatanen

# Serum lipids and metabolites are associated with T1D-related microbial taxa



- Positive correlation
  - Blautia and branched-chain amino acids
  - Ruminococcus and branched-chain amino acids
- Negative correlation
  - Veillonella and branched-chain amino acids

### Tommi Vatanen





# Software FEATURES

## Extensibility

### Testing Approaches

- Naïve AllA
- Hierarchical AllA

Similarity Metrics NMI MIC Pearson AMI Decomposition Methods PCA ICA KPCA PLS CCA

# Packaging

- Python package
  - Software
  - Unit tests
  - Evaluation Modules
- Online References
  - Hierarchical All-against-All(HAllA)
    - http://huttenhower.sph.harvard.edu/halla
  - The Huttenhower Lab
    - http://huttenhower.sph.harvard.edu

# Installation and Running

- Download
- python setup.py install
- Command line
  - halla –X dataset1.txt –Y datsest2.txt –o output --plotting-results

## Summary Advantages

- Works with heterogeneous data (continuous+ categorical)
- Able to detect individually weak, but collectively strong signals
- Interpretable approach to data dimension reduction
- Assigns q-values to associations with high sensitivity
- Extensible to the use of different clustering and association methods

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# **FDR Controlling**

## Benjamini-Hochberg-Family FDR correction

### 1.0 + + 0.8 + +0.6 Recal -FDR **—** 0.4 +++ +0.2 +Ţ 0.0 1 0.01 0.05 0.2 0.1 Method

## Benjamini-Hochberg-Level FDR correction

