

Nonparametric MANOVA via Independence Testing

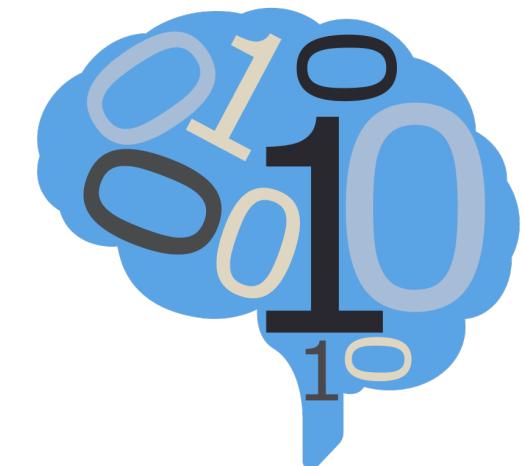
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Motivation

- Understand the relationship between k groups (*i.e.* control vs. disease)
- Question: Are they related? How?

One often desires to test groups

<i>U</i>	<i>V</i>
my grass	neighbor's grass
human brain connectivity	alien brain connectivity
control	disease
cancer risk group 1	cancer risk group 2

One often desires to test groups

<i>U</i>	<i>V</i>
my grass	neighbor's grass
human brain connectivity	alien brain connectivity
control	disease
cancer risk group 1	cancer risk group 2
any group	any other group

Statistics Background

- X is a random variable (some measurement)
- F_X is the distribution of X
- This means $F_X(a) = P(X \leq a)$
- This is denoted:

$$X \sim F_X$$

Statistics Background

- For two random variables X and Y , F_{XY} is called the joint distribution
- This means $F_{XY}(a, b) = P(X \leq a \text{ and } Y \leq b)$

or,

$$(X, Y) \sim F_{XY}$$

Informal Definition of Hypothesis Testing

- **Null Hypothesis:** The conventional belief about a phenomenon of interest, written H_0 .
- **Alternative Hypothesis:** An alternate belief about the same phenomenon, written H_A .
- **p-value:** The probability (under the null) of measurements more extreme than what was observed.

Formal Definition of K -Sample Testing

$$U_i^j \sim F_j, \quad j \in 1, \dots, k, \quad i \in 1, \dots, n_j$$

$$H_0 : F_1 = F_2 = \dots = F_k$$

$$H_A : \exists j \neq j' \text{ s.t. } F_j \neq F_{j'}$$

Note: These ideas and notation generalize for multivariate X and Y .

Outline

1. Intuition
2. Simulations
3. Multiway and Multilevel
4. Real Data
5. Conclusion

Intuition

Intuitive Desiderata of Testing Procedure

- Performant under *any* distribution
 - low- and high-dimensional
 - Euclidean and structured data (eg, sequences, images, networks, shapes)
 - linear and nonlinear relationships
- Is computational efficient

Provides a tractable algorithm that addresses the motivating question:

Are they related?

Analysis of Variance (ANOVA)

$$MST = \frac{\sum_{i=1}^k (T_i^2/n_i) - G^2/n}{k - 1}$$

$$MSE = \frac{\sum_{i=1}^k \sum_{j=1}^{n_i} Y_{ij}^2 - \sum_{i=1}^k (T_i^2/n_i)}{n - k}$$

$$ANOVA = \frac{MST}{MSE}$$

Multivariate ANOVA (MANOVA)

$$\mathbf{W} = \sum_{i=1}^k \sum_{j=1}^{n_i} (\mathbf{x}_{ij} - \bar{\mathbf{x}}_{i\cdot}) (\mathbf{x}_{ij} - \bar{\mathbf{x}}_{i\cdot})^T$$

$$\mathbf{B} = \sum_{i=1}^k n_i (\bar{\mathbf{x}}_{i\cdot} - \bar{\mathbf{x}}_{..}) (\bar{\mathbf{x}}_{i\cdot} - \bar{\mathbf{x}}_{..})^T$$

$$MANOVA = \sum_{i=1}^s \frac{\lambda_i}{1 + \lambda_i} = \text{tr}(\mathbf{B}(\mathbf{B} + \mathbf{W})^{-1})$$

MANOVA Assumptions

- Data is derived from a multivariate Gaussian distribution
- Each group has the same covariance matrix

There must be a better test out there

Slight Tangent - Independence Testing

- X and Y are independent if neither contains information about the other
- In other words,

$$F_{XY} = P(X \leq a \text{ and } Y \leq b) = P(X \leq a) \times P(Y \leq b) = F_X F_Y$$

$$F_{XY} = F_X F_Y$$

Note: These ideas and notation generalize for multivariate X and Y .

Distance Correlation (Dcorr)

$$\widehat{Dcov}_{xy} = \frac{1}{n^2} \text{tr}(\mathbf{H}\mathbf{D}^x\mathbf{H}\mathbf{D}^y\mathbf{H})$$

$$\widehat{Dcorr}_{xy} = \frac{\widehat{Dcov}_{xy}}{\sqrt{\widehat{Dcov}_{xx} \times \widehat{Dcov}_{yy}}}$$

$$\mathbf{C}_{ij}^x = \mathbb{I}_{i \neq j} \left(\mathbf{D}_{ij}^x - \frac{1}{n-2} \sum_{t=1}^n \mathbf{D}_{it}^x - \frac{1}{n-2} \sum_{t=1}^n \mathbf{D}_{tj}^x + \frac{1}{(n-1)(n-2)} \sum_{t=1}^n \mathbf{D}_{tt}^x \right)$$

$$Dcov_{xy} = \frac{1}{n(n-3)} \text{tr}(\mathbf{C}^x \mathbf{C}^y)$$

$$Dcorr_{xy} = \frac{Dcov_{xy}}{\sqrt{Dcov_{xx} \times Dcov_{yy}}}$$

Multiscale Graph Correlation (MGC)

- Compute local Dcorr **at all scales**
- Find scale with **max** smoothed test statistic
- Permutation test to determine p-value

Kernel Mean Embedding Random Forest (KMERF)

- Train random forest on X , compute kernel matrix
- Transform similarity kernel matrix to distance matrix
- Permutation test to determine p-value

Great, what now?

- Can reduce the k -sample testing problem to the independence problem

$$\mathbf{x} = \begin{bmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \vdots \\ \mathbf{u}_k \end{bmatrix}, \quad \mathbf{y} = \begin{bmatrix} \mathbf{1}_{n_1 \times 1} & \mathbf{0}_{n_1 \times 1} & \cdots & \mathbf{0}_{n_1 \times 1} \\ \mathbf{0}_{n_2 \times 1} & \mathbf{1}_{n_2 \times 1} & \cdots & \mathbf{0}_{n_2 \times 1} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{0}_{n_k \times 1} & \mathbf{0}_{n_k \times 1} & \cdots & \mathbf{1}_{n_k \times 1} \end{bmatrix}$$

- Run any independence test
- ***Note: This process does not add any additional computational complexity to the independence testing algorithm***

Simulations

Definitions

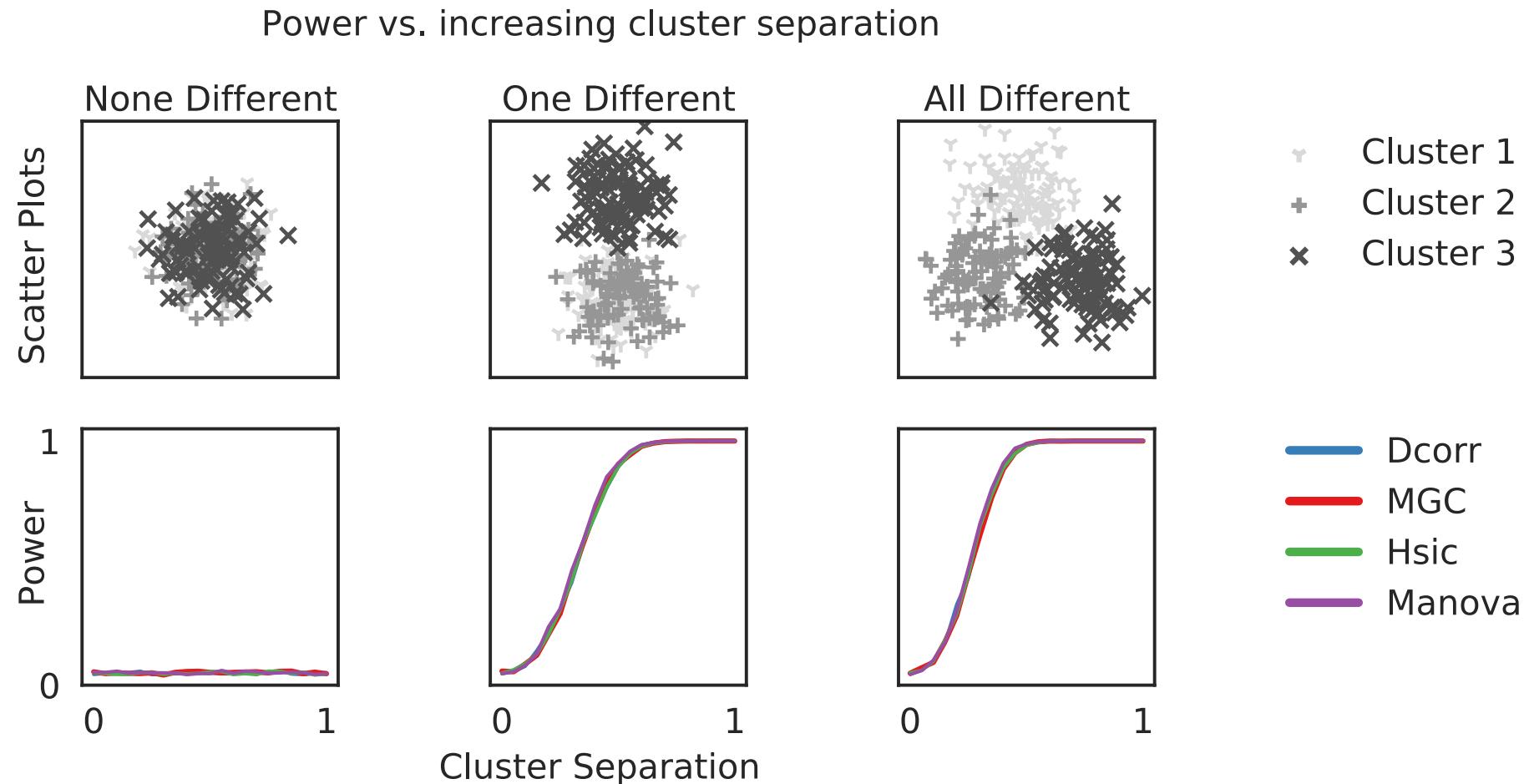
- **power** is the probability of rejecting the null when the alternative is true

$\beta_n(t)$: power of test statistic t given n samples

- **relative power** power of one approach minus power of another

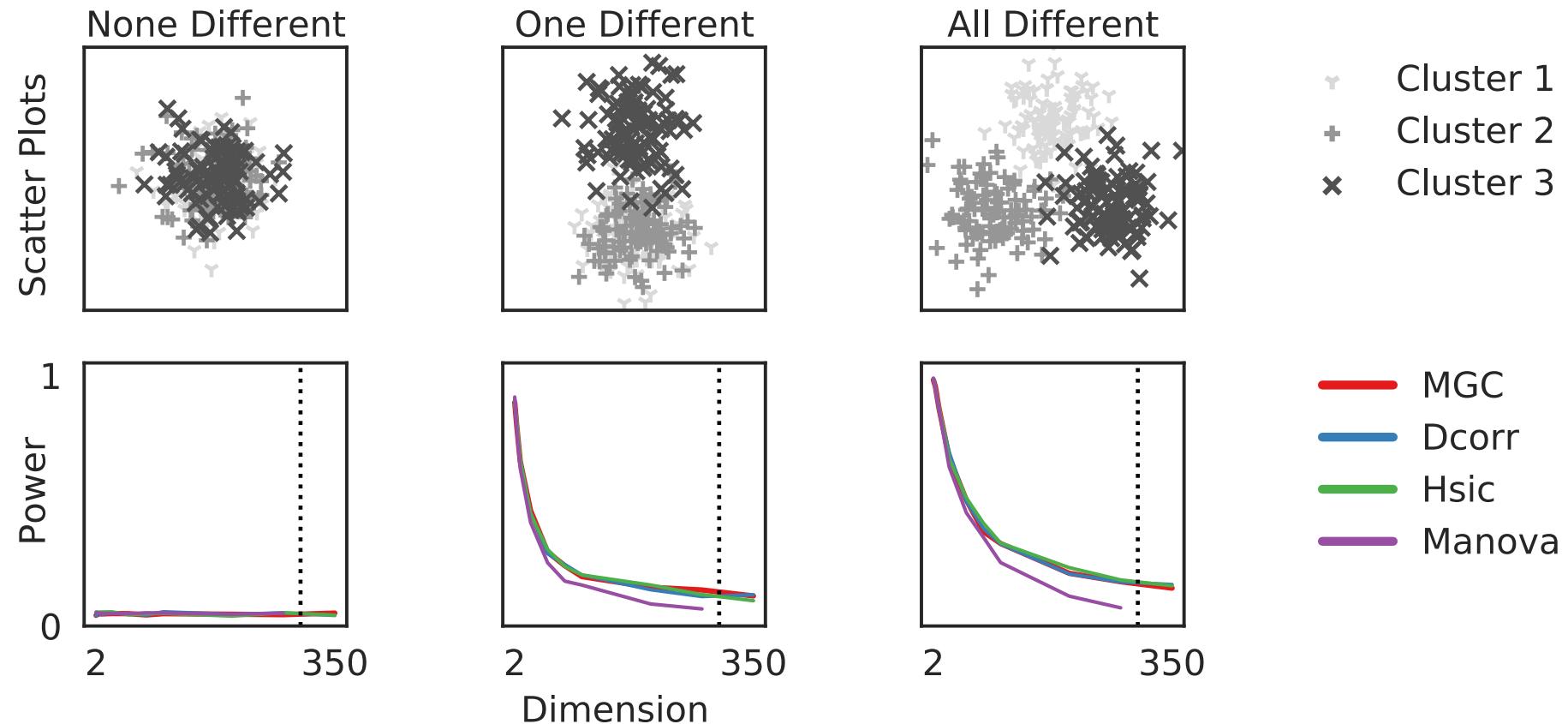
$$\beta_n(t) - \beta_n(\text{manova})$$

Optimal Settings for MANOVA (1D)

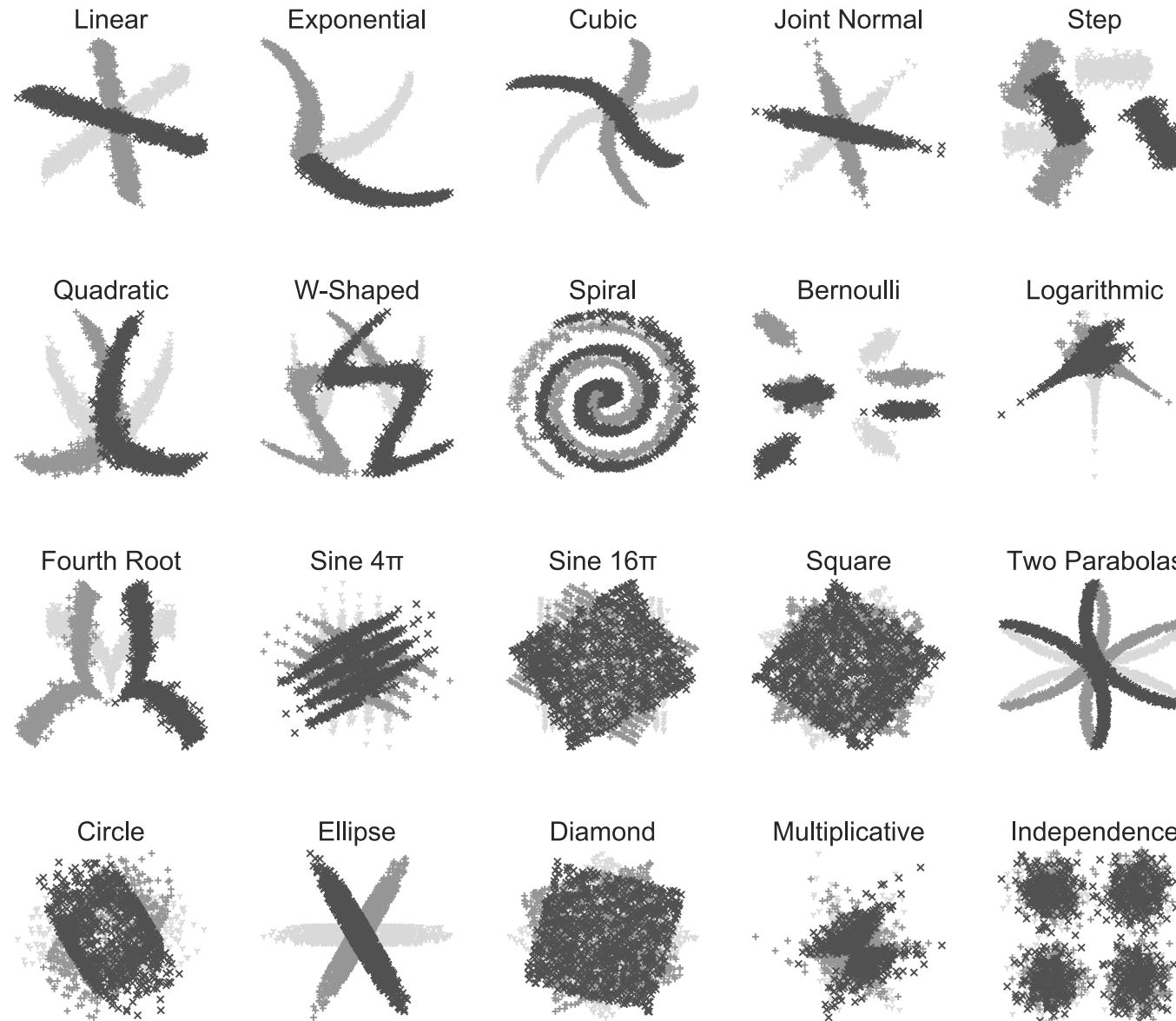


Optimal Settings for MANOVA (HD)

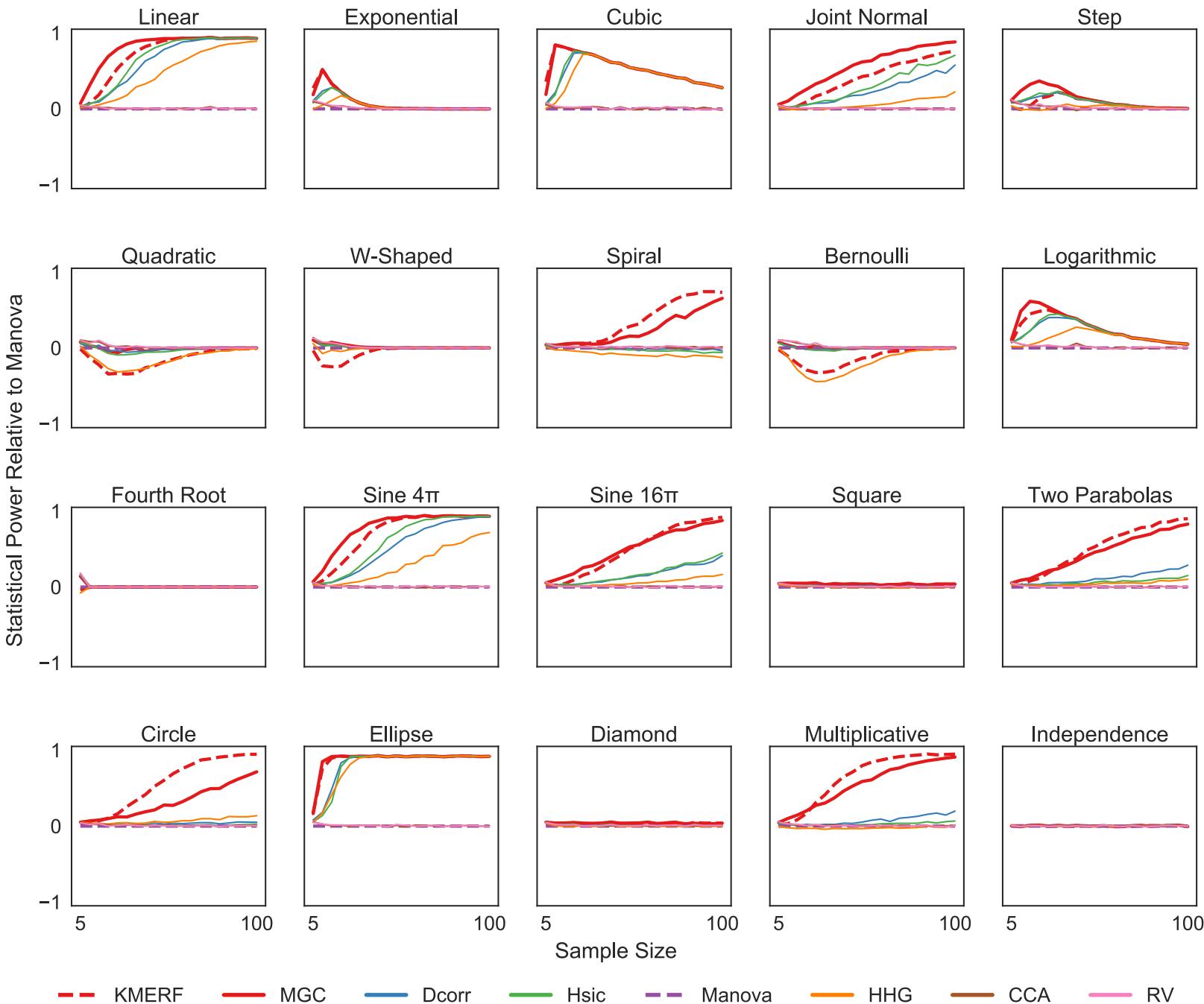
Power vs. increasing Gaussian dimension



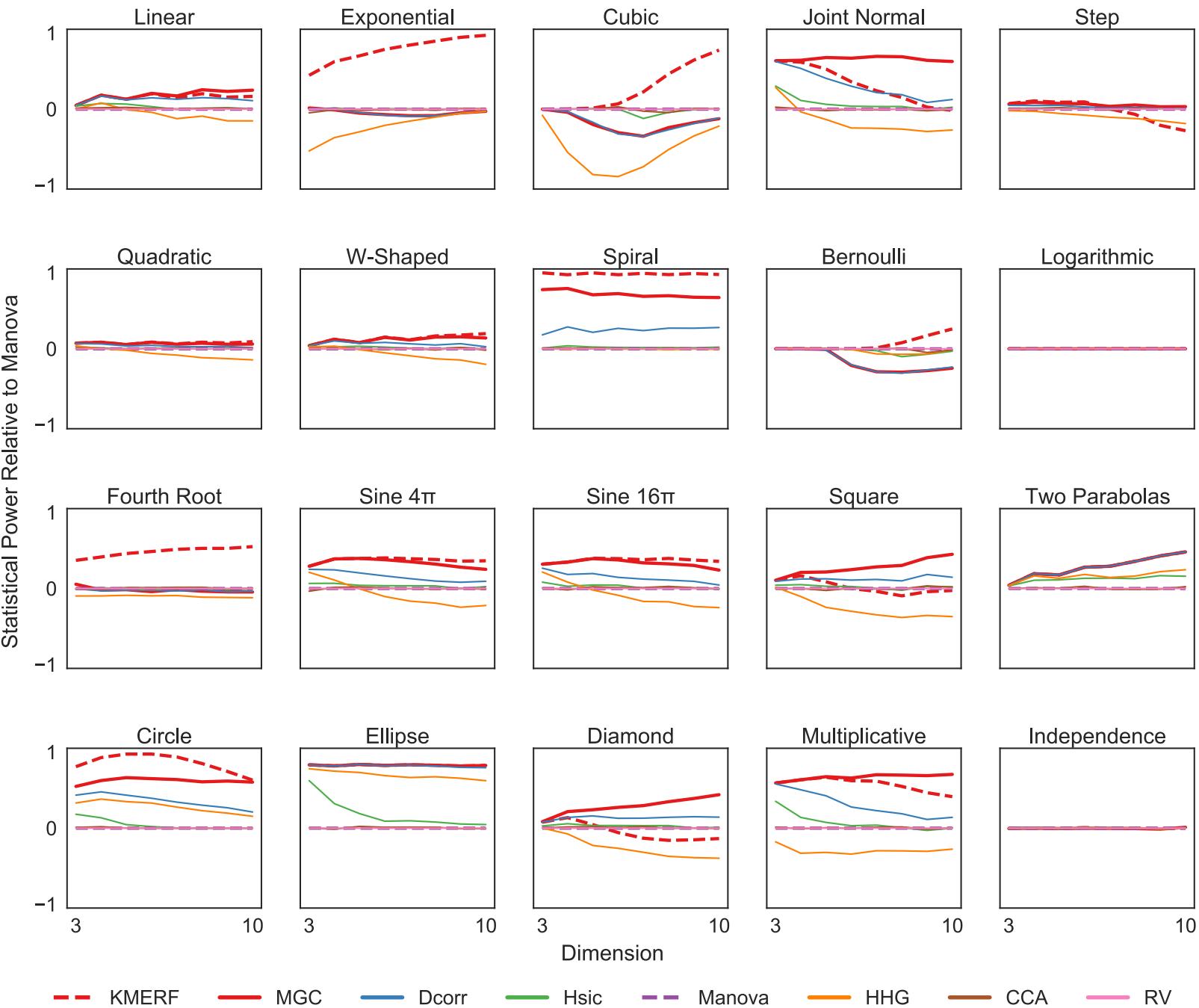
20 Different Functions (2D version)



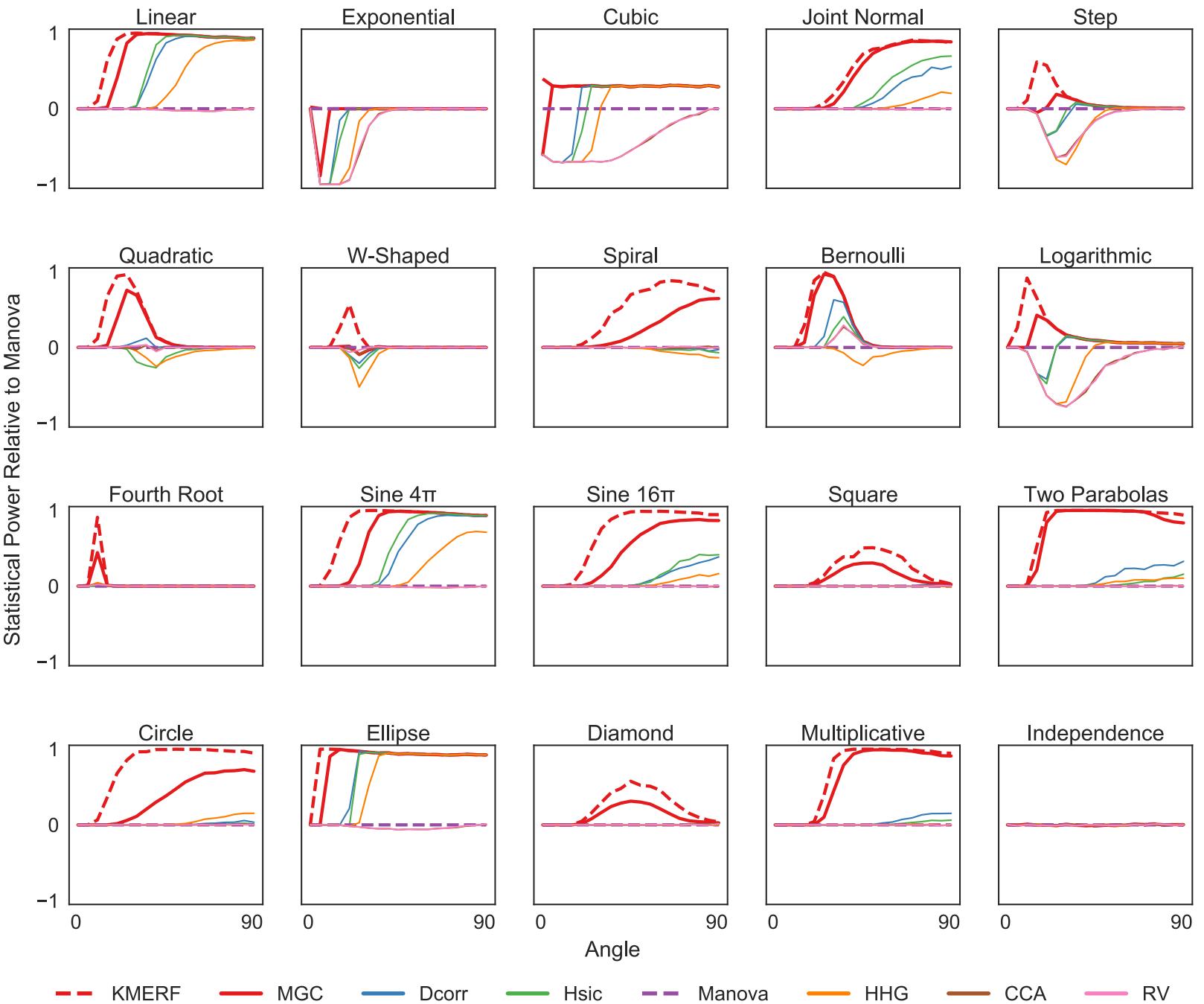
Multivariate Three-Sample Testing Increasing Sample Size



Multivariate Three-Sample Testing Increasing Dimension



Multivariate Three-Sample Testing Increasing Angle



Multiway and Multilevel

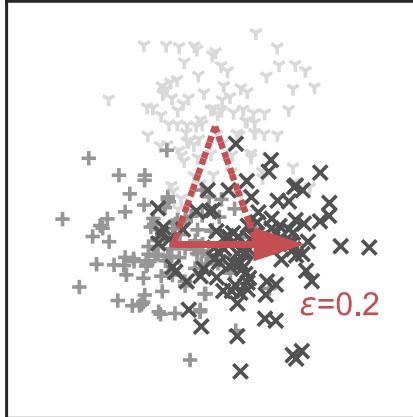
Multiway Tests

- **Multiway:** More than one treatment group
- Instead of one-hot encoding, add 1's columns of label matrix

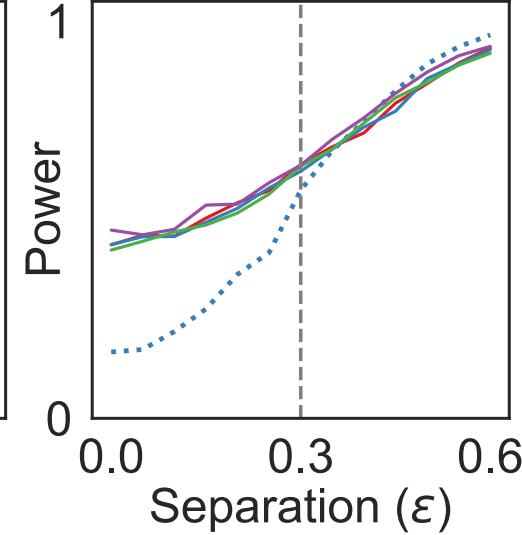
(A) Default distances

0	1	1
1	0	1
1	1	0

(B) Weak multiway



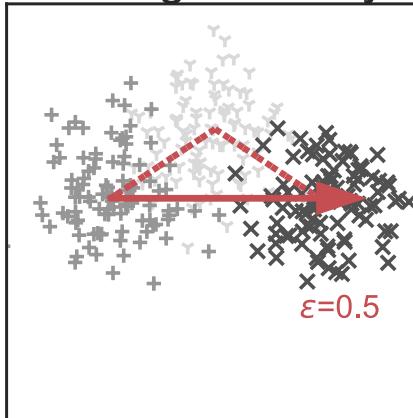
(C) Cluster separation



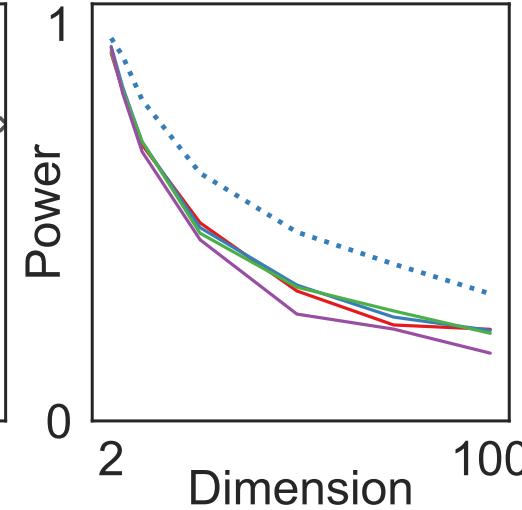
Multiway distances

0	1	1
1	0	2
1	2	0

Strong multiway



Added noise dimensions



- Cluster 1
- Cluster 2
- Cluster 3

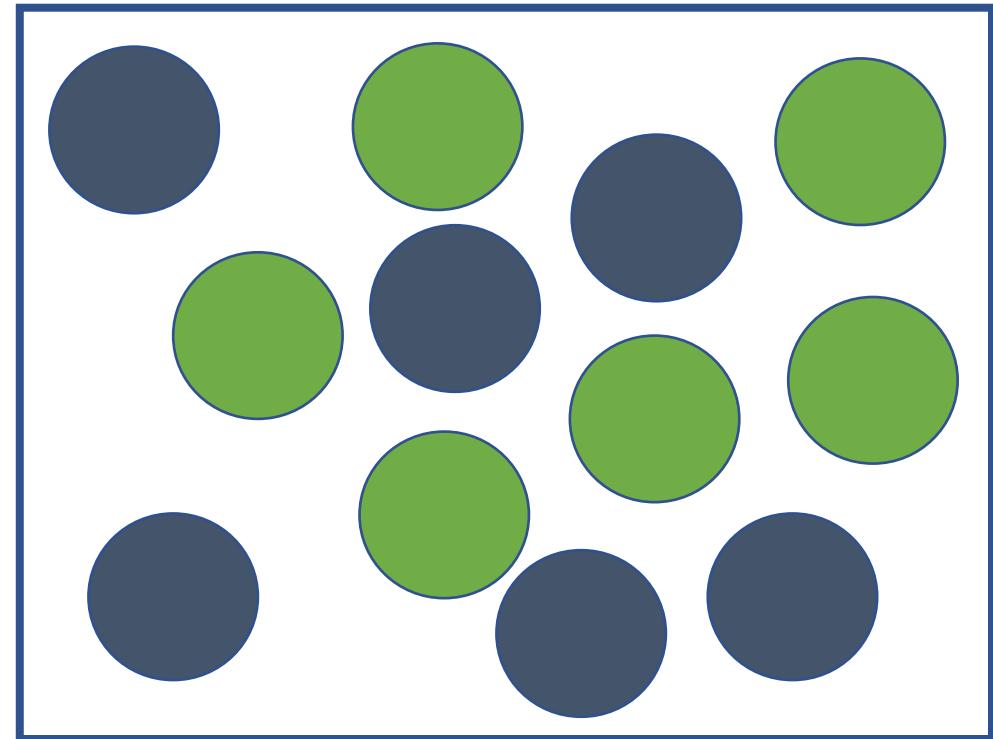
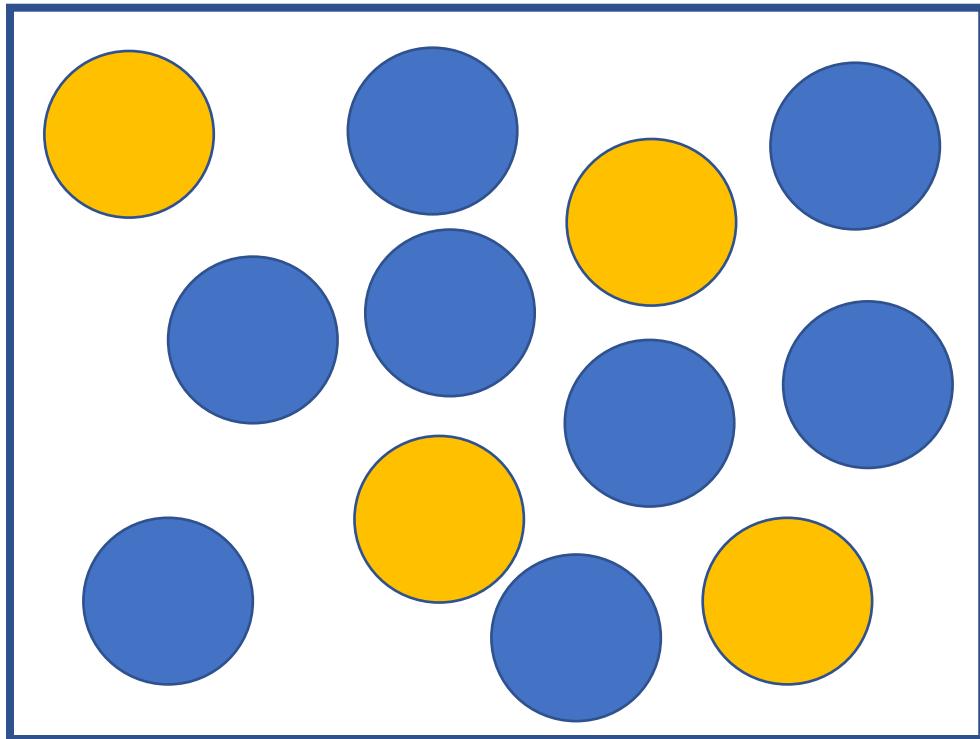
- MGC
- Dcorr
- Hsic
- Manova
- Multiway Dcorr

Multilevel Tests

- **Multilevel:** Samples are not always exchangeable with one another
- Need block permutation

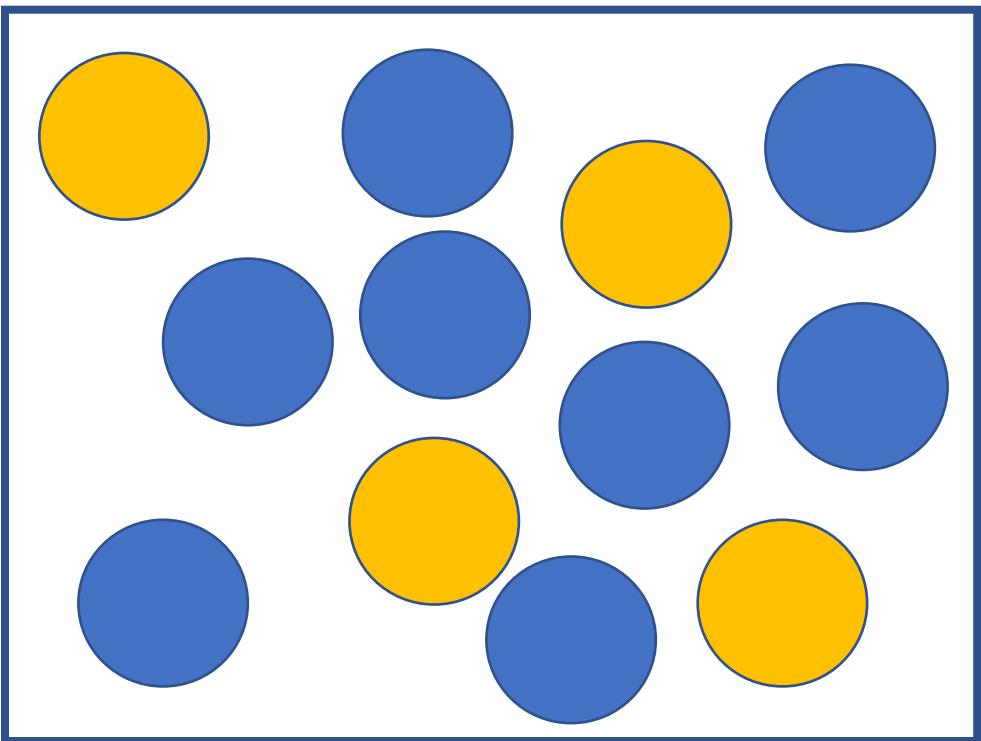
Multilevel Tests

- **Multilevel:** Samples are not always exchangeable with one another

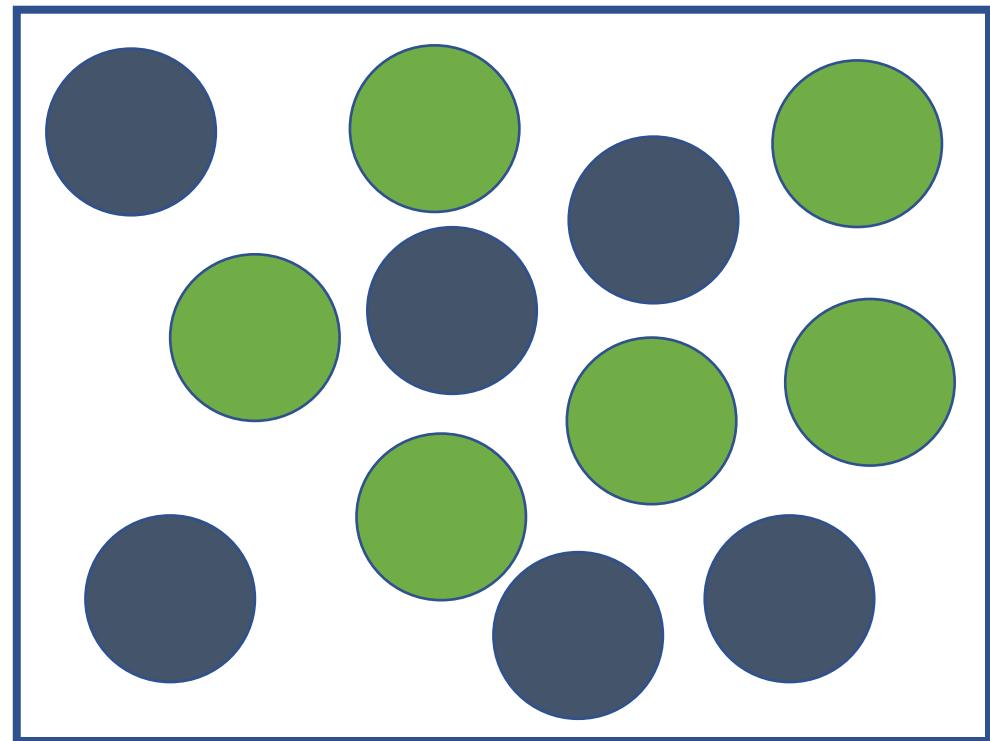


Multilevel Tests

- **Multilevel:** Samples are not always exchangeable with one another

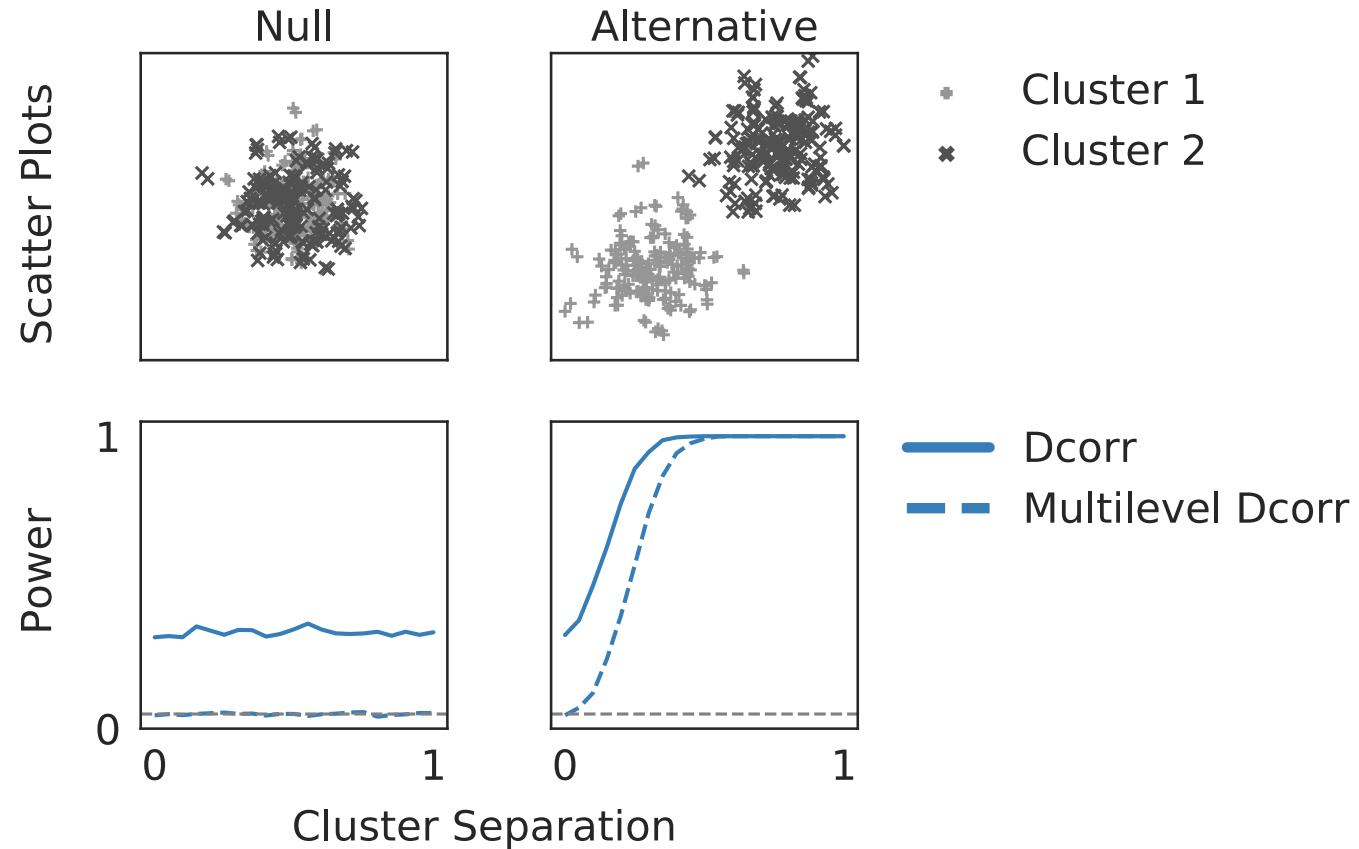


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Multilevel Tests

Multilevel Dcorr: Power vs. Cluster separation

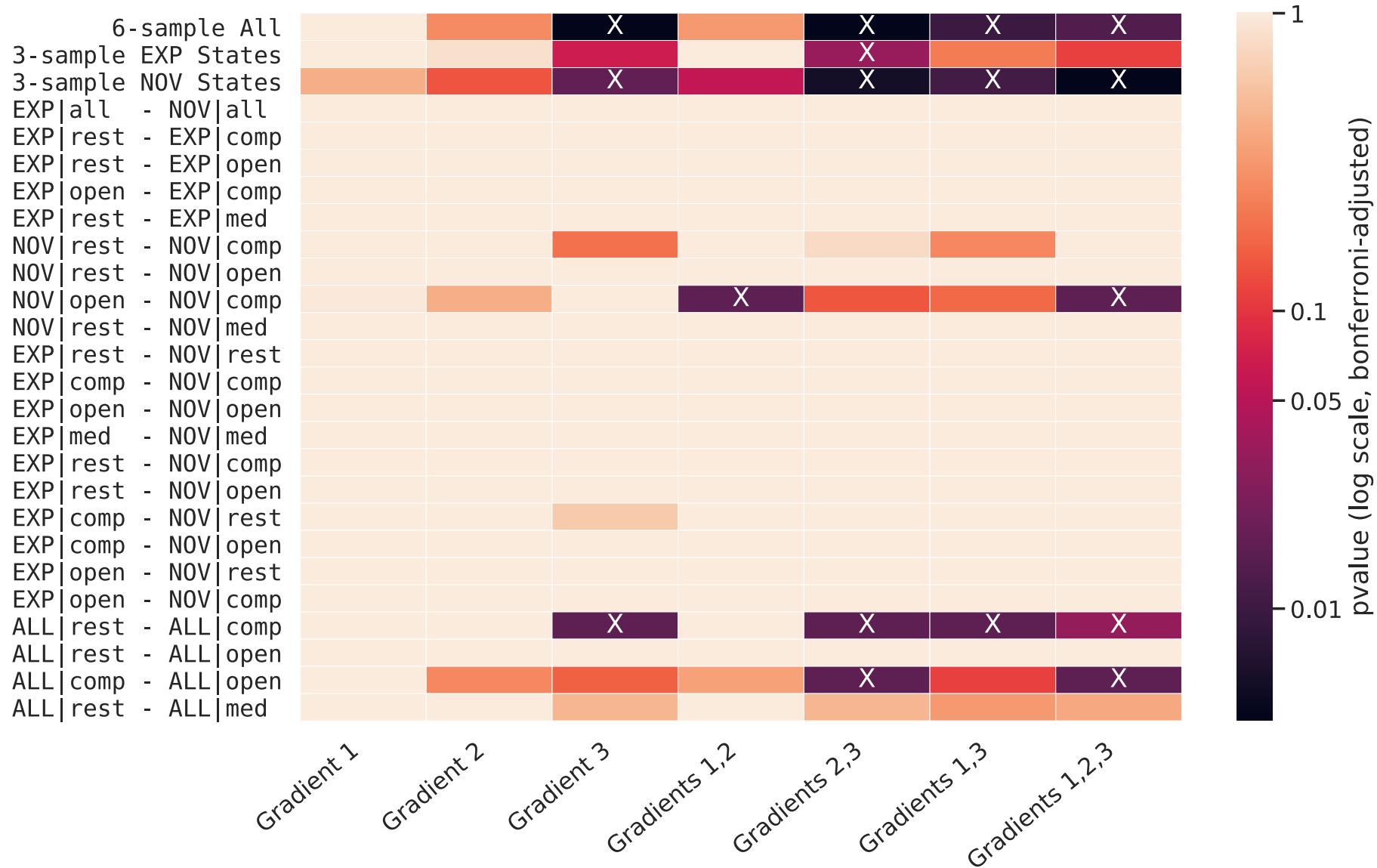


Real Data

The Procedure

- Data: 75 subjects – 28 experienced and 47 novice meditators
- 3 Recording sessions for each meditator
- Computed gradients and tested for difference between traits and novice
- This is a **multilevel and multiway test**

3rd Embedding shows significance



Conclusions

- Presented several new k -sample tests using our framework
- At a simulation setting that fulfills MANOVA assumptions, our implementation performs as well or better
- Multiway tests give additional power when strong multiway effect is suspected
- Multilevel tests can now be performed

Next Steps

- All algorithms can be found in the [hippo](#) package
 - [Documentation](#)
 - [Install](#)
 - [Tutorials](#)
- [Paper](#)

[Email](#) | [Website](#) | [Twitter](#)

Acknowledgements

- **Joshua Vogelstein, Cencheng Shen:** Theory, and paper writing
- **Ronan Perry:** Multiway, Multilevel, and Real Data
- **Jelle Zorn, Antoine Lutz:** Raw real data
- **Carey E. Priebe:** Theory
- **Russell Lyons, Minh Tang, Ronak Mehta, Eric Bridgeford:** Review
- ...and the rest of the NeuroData Lab



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Questions?