

Discovering & deciphering relationships across disparate data modalities

Motivation:

↳ "Understand" the relationship between physical (brain) & mental properties

↳ Q1: are the 2 related at all?

↳ Q2: how are they related?

Stats Background:

↳ X is random variable (some measurement)

↳ F_X is distribution of X

$$\hookrightarrow F_X(a) = P(X \leq a)$$

$$\hookrightarrow X \sim F_X$$

↳ For 2 random variables X & Y , $F_{X,Y}$ is called the joint distribution

$$\hookrightarrow F_{X,Y}(a, b) = P(X \leq a \text{ \& \& } Y \leq b)$$

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

Independence

↳ X & Y are independent if neither contains info. about the other

$$\hookrightarrow F_{X,Y} = P(X \leq a \text{ \& \& } Y \leq b) = P(X \leq a) \times P(Y \leq b) = F_X F_Y$$

$$\hookrightarrow F_{XY} = F_X F_Y$$

②

Informal Definition of Hypothesis Testing

- \hookrightarrow Null Hypothesis: conventional belief about a phenomenon of interest (H_0)
- \hookrightarrow Alternative Hypothesis: alternate belief about the same phenomenon (H_A)
- \hookrightarrow p-value: probability (under the null) of measurements more extreme than what was observed

Formal Definition of Independence Testing:

$$\hookrightarrow (X_i, Y_i) \sim F_{XY} = F_{X|Y} F_Y, \quad i \in 1, \dots, n$$

$$\hookrightarrow H_0: F_{XY} = F_X F_Y$$

$$\hookrightarrow H_A: F_{XY} \neq F_X F_Y$$

Intuition

Intuitive Desiderata of Testing Procedure

1. Performant under any joint distribution

- \hookrightarrow low & high dimensional
- \hookrightarrow Euclidean & structured data (eg. sequence, images, networks, shapes)
- \hookrightarrow linear & nonlinear relationships

2. ~~Reveals~~ Reveals the "geometry" of dependence ^③

3. Is computationally efficiency

↳ Provides a tractable algorithm that addresses the 2 motivating questions

↳ Q1: are the two related at all?

↳ Q2: how are they related?

Correlation coefficient

$$\hookrightarrow r_{xy}^2 = \frac{\left(\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) \right)^2}{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2} \} *$$

Mantel correlation coefficient

↳ *

$$\hookrightarrow d_{xy}^2 = \frac{\left(\sum_{i,j=1}^n (x_i - x_j)(y_i - y_j) \right)^2}{\sum_{i,j=1}^n (x_i - x_j)^2 \sum_{i,j=1}^n (y_i - y_j)^2} \} **$$

Generalized correlation coefficient

↳ *

↳ **

$$\hookrightarrow c_{xy}^2 = \frac{\left(\sum_{i,j=1}^n \sigma_x(x_i, x_j) \sigma_y(y_i, y_j) \right)^2}{\sum_{i,j=1}^n \sigma_x(x_i, x_j)^2 \sum_{i,j=1}^n \sigma_y(y_i, y_j)^2}$$

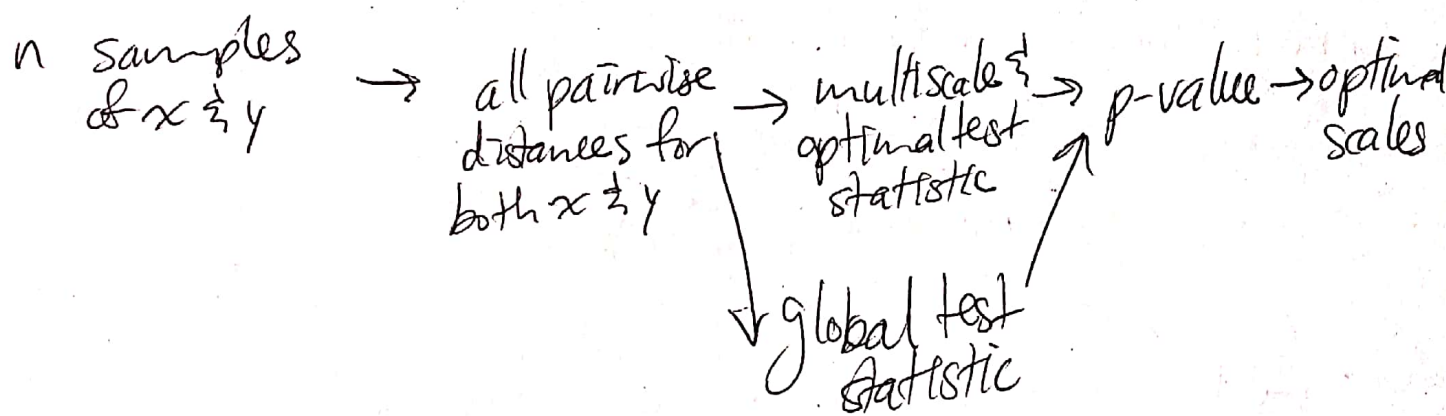
Local distance correlation

Multiscale distance correlation

- ↳ compute local dcorr @ all scales
- ↳ find scale w/ max smoothed test statistic
- ↳ permutation test to determine p-value

Multiscale Generalized Correlation (MGC)

↳



Simulations

- ↳ 20 diff. functions

Definitions

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

- ↳ $\beta_n(t)$ = power of ~~one approach~~ test statistic to given n samples

- ↳ relative power: power of one approach minus

- ↳ power of another
- ↳ $\beta_n(mgc) - \beta_n(t)$