# **Discovering Potential Correlations via Hypercontractivity**

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## INTRODUCTION

#### Discovering associations in large datasets

Example: Data for 300 indicators for 200 countries

Which pairs of indicators are associated?

~ 900,000 pairs of indicators!

Associations are used to make policy decisions

Important both in industry and scientific research

? ?				
	Population	Energy Use		CO <sub>2</sub> Emissions
Afghanistan	26088	470		0.02
Albania	3172	761		0.98
Zambia	11696	620		0.21
Zimbabwe	13228	741		0.94

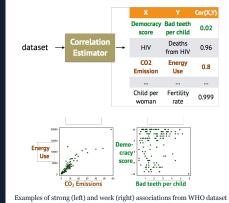
World Health Organization (WHO)

## **BACKGROUND**

# Correlation analysis to discover associations

Estimate correlation coefficients for all pairs of indicators

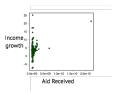
Pairs w. large corr coeff: candidates for strong association



# **PROBLEM STATEMENT**

#### **Motivation:**

Existing correlation estimators discover average correlations but fail to discover *potential* correlations



Discovering potential correlations can affect policy decisions and lead to scientific findings

Goal: discover potential correlations

# **METHOD**

1. Propose 7 axioms for a measure of potential correlation

 $0 \le \rho(X, Y) \le 1$ 

 $\rho(X,Y) = 0 \text{ iff } X \text{ and } Y \text{ are independent}$   $\rho(X,Y) = 1 \text{ if } Y = f(X) \text{ for } (X,Y) \in \mathcal{X}_r \times \mathcal{Y} \text{ for some } \mathcal{X}_r \subseteq \mathcal{X}$ 





2. Show hypercontractivity coefficient satisfies all axioms

$$s(X;Y) \equiv \sup_{U-X-Y} \frac{I(U;Y)}{I(U;X)}$$

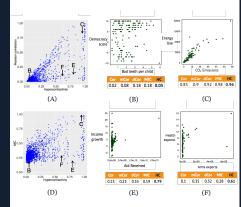
- 3. Propose a novel estimator
  - based on an alternative definition

$$\begin{split} s(X;Y) &= \sup_{r(x) \neq p(x)} \frac{D_{\mathrm{KL}}(r(y) \| p(y))}{D_{\mathrm{KL}}(r(x) \| p(x))} \\ &\quad \text{where} \quad r(y) = \sum r(x) p(y|x) \end{split}$$

via joint optimization and estimation

## **EXPERIMENTS**

#### 1. WHO dataset



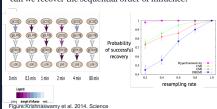
- (A): Scatter plot of Pearson correlation vs. HC
- (D): Scatter plot of Maximal Info. Coefficient vs. HC
- (B): All correlations are small
- (C): All correlations are large
- (E) and (F): Only HC discovers potential correlations

# 2. Genetic Pathway Recovery

Gene expression time series data for four genes

Biological fact:

can we recover the sequential order of influence?



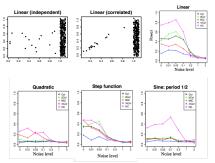
Hypercontractivity: robust measure strength of influence

# **EXPERIMENTS**

#### 3. Power test

Binary hypothesis testing of potential correlations

Power: true positive rate for a fixed false positive rate



HC is more powerful than others in hypothesis testing of canonical examples of potential correlations

## CONCLUSION

- We postulate a set of natural axioms that we expect a measure of potential correlation to satisfy
- We show that rate of information bottleneck, i.e., the hypercontractivity coefficient (HC), satisfies all the proposed axioms
- 3. We provide a novel estimator for HC
- 4. Experimental results: WHO datasets, genetic pathway recovery, power tests

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