

Applying Causality Toolkit to Real-world Datasets





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Introduction

Everything in this universe happens for a reason and every action has a reaction.

Analyzing causality can help in medical diagnostic analysis, time series analysis, and strategic planning In real-world datasets, variables are inter-related, implying subtle correlations, which makes causal analysis difficult.

Causal Toolkits

- 1. HPCC Causality
- 2. Because
 - Visualization bundle
 - Dependence & Independence tests
 - **Causal Direction Tests**

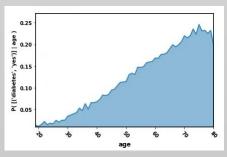
Analysis Steps

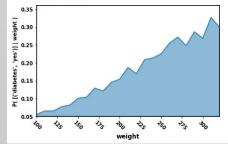
- 1. Finding & Analyzing Dataset
- Pre-processing the dataset
- Propose a Causal Hypothesis
- 4. Applying causal toolkits & analyzing
 - 5. Interpretation & Causal Model
- 6. Hypothesis & Model Verification

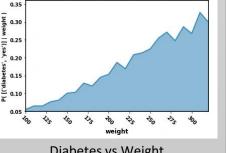
Causal Hypothesis Question

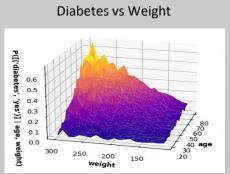
Proposed causal hypothesis question: "What factors can influence the likelihood of a person having Diabetes?"

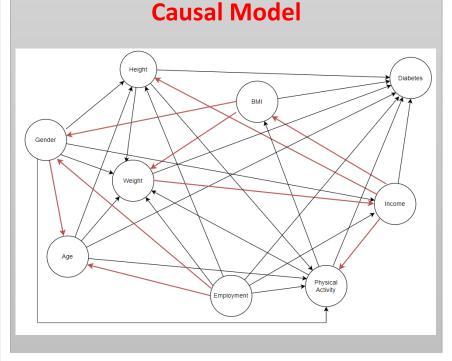
Analysis LLCP-CDC Dataset



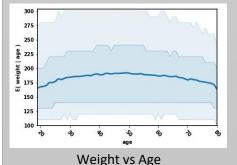


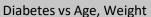


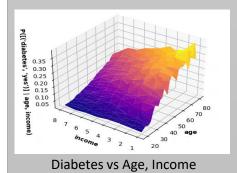


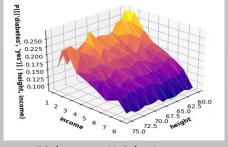


Diabetes vs Age









Diabetes vs Height, Income

Conclusion

Factors like Age, Height, Weight, Income, type of Employment, Physical Activity, and gender have their effect on Diabetes.

- Most of the relations are practically and analytically correct.
- Some relations are unexpected, but probable valid proof can be generated.
- For some other relations, any explanation is rationally invalid. Causality toolkit can be applied to analyze real-world datasets. But the cause-effect of latent variables cannot be incorporated into the causal model.