

P. Behrouzi, S.M. Mahmoudi, F. Richter, M. Shafiee, M. Signorelli, E. Wit

p.behrouzi@rug.nl, s.m.mahmoudi@rug.nl, f.richter@rug.nl, m.shafiee.kamalabad@rug.nl, m.signorelli@rug.nl, e.c.wit@rug.nl

## Introduction

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## Box 1

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Nullam sollicitudin lobortis urna quis varius. Nullam sagittis blandit diam,  $DN = G_t(V_t, E_t)$ , risus  $E_t \subseteq V_t \times V_t$  ( $\forall t \geq 0$ ). vel tortor justo,  $G_0$ , quis malesuada lorem.

$$\cos^3 \theta = \frac{1}{4} \cos \theta + \frac{3}{4} \cos 3\theta \quad (1)$$

Vivamus porta lacus et lectus **porta lacus**. Pellentesque habitant morbi tristique senectus et netus et malesuada fames ac turpis. torte  $G_t$  hac millis **plates** Idk

## Causal effect in network

**Motivation:** Can we learn causal effects from observational data in high-dimensional systems?

### Nonparanormal distribution

- $f = F^{-1} \circ \Phi$  monotone univariate function
- $f(Y) = (f_1(Y_1), \dots, f_p(Y_p))^T \sim N(0, \Sigma)$  and  $Y = (Y_1, \dots, Y_p)^T$  has a nonparanormal distribution

### Causal Effect for Nonparanormal Graphical Models

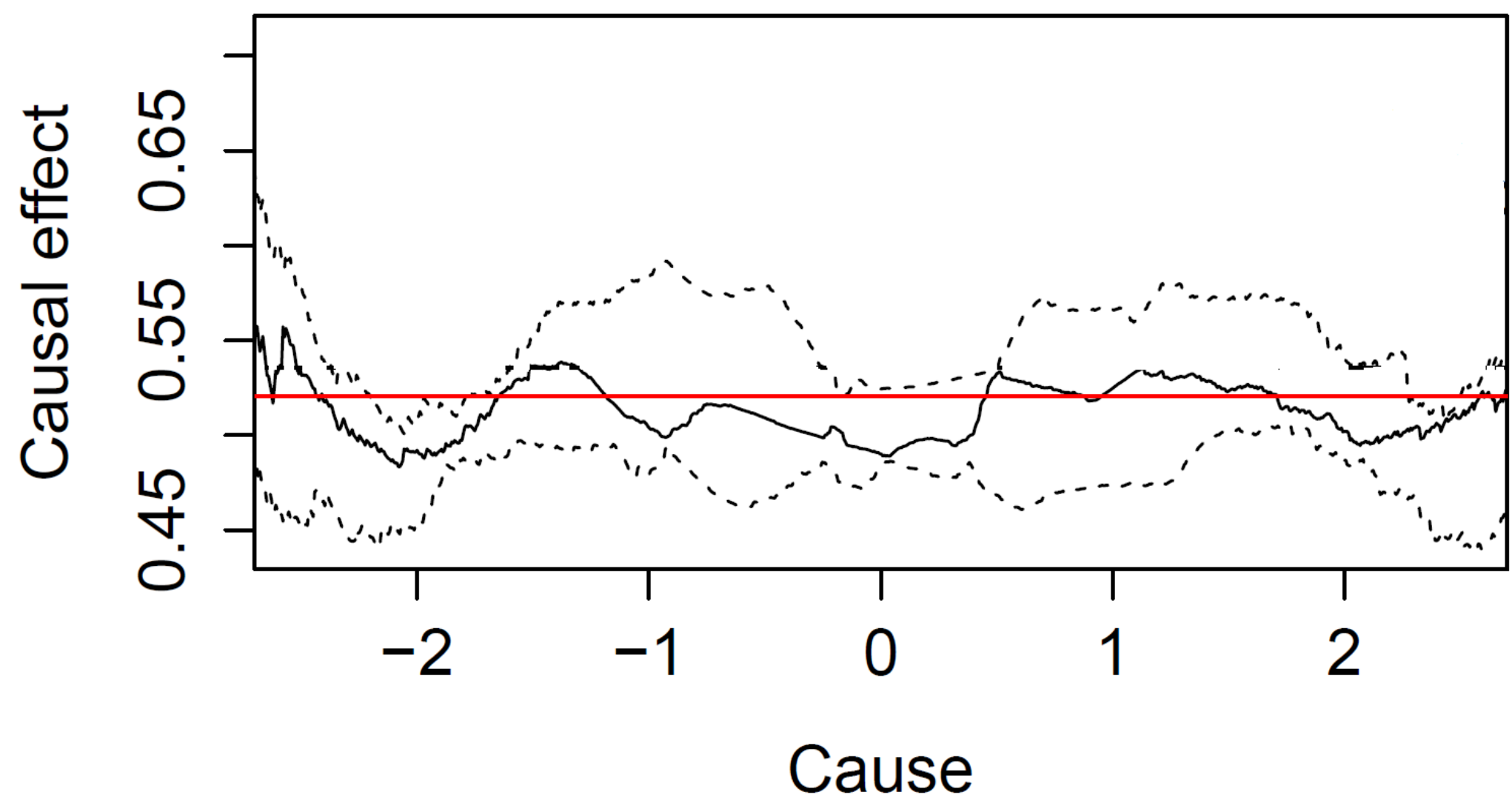
We are interested in the causal effect of  $Y_i$  on  $Y_p$  for  $i \in (1, \dots, p-1)$

- Gaussian distribution:  $E(Y_p | Y_i = y_i, pa_i) = \beta_0 + \beta_i y_i + \beta_{pa_i}^T pa_i$

$$\frac{\partial}{\partial y_i} E[Y_p | do(Y_i = y_i)] \equiv \beta_i$$

- Non-Gaussian distribution

$$\begin{aligned} \frac{\partial}{\partial y_i} E[Y_p | do(Y_i = y_i)] &\cong f_p'(z_{0j}) \delta_i(f_i^{-1})'(y_i) \\ &= f_p'(z_{0i}) \beta_i(f_j^{-1})'(y_i) \end{aligned}$$



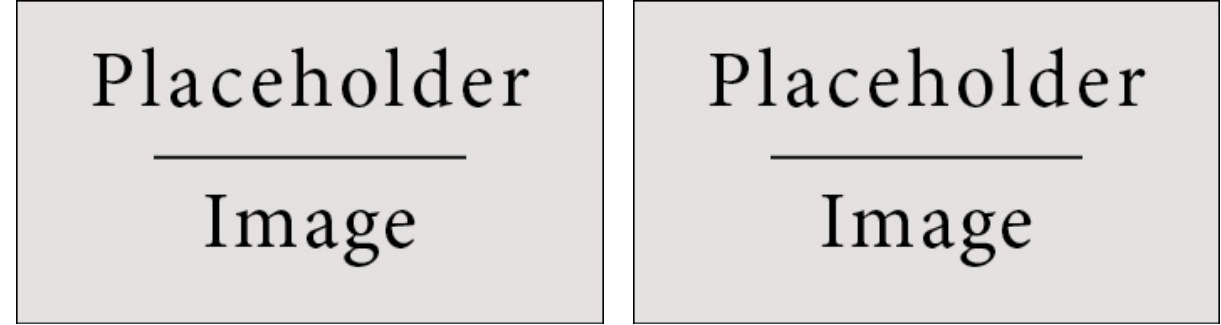
**Figure 2:** Dashed lines are quantile profile for the functional causal effect (shown here with solid black line) of  $Y_i$  on  $Y_p$ . Red line show true causal effect.

**Conclusion:** We have derived an explicit formula for describing a causal effect for Nonparanormal distribution.

**Contact:** Seyed Mahdi Mahmoudi, s.m.mahmoudi@rug.nl

## Box2

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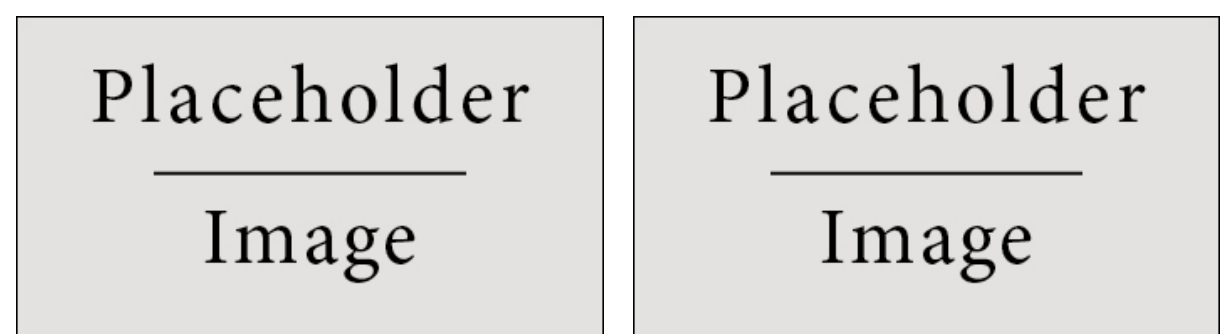


**Figure 1:** Figure caption 1 (left); Figure caption 2 (right)

1. Cras ac ipsum eu nisl imperdiet interdum nunc bibendum, est in pulvinar facilisis, mi purus fringilla tellus, eu varius ipsum ante laoreet ipsum
2. Sed cursus erat quis odio laoreet facilisis maecenas vehicula

## Box 4

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**Figure 3:** Figure caption 1 (left); Figure caption 2 (right)

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