# Letter to the Editor



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# Introducing Network Intervention Analysis to Investigate Sequential, Symptom-Specific Treatment Effects: A Demonstration in Co-Occurring Insomnia and Depression

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Dear Editor,

Insomnia and depression are among the most prevalent mental disorders and often co-occur [1]. Insomnia is a primary risk factor for the development of depression [2]. In people with depression, insomnia impedes remission [3, 4], while cognitive-behavioral therapy for insomnia (CBTI) accelerates it [5, 6]. However, overlap in the symptoms of insomnia and depression obscures the sequential development and order of treatment-induced changes: do CBTI-induced sleep improvements precede alleviation of depression symptoms, or does CBTI affect depression directly?

A possibility to investigate the effect of CBTI on insomnia and depression while taking symptom overlap into account is offered by network models that conceptualize mental disorders as networks of symptoms structured by their associations [7]. Here we introduce Network Intervention Analysis (NIA) as an extension of these models to follow the development of treatment-induced changes in symptoms and their association structure over time, while distinguishing direct and indirect effects.

We applied NIA to identify the sequential development and order of CBTI-induced effects on symptoms of insomnia and depression throughout the course of treatment. We used data from a randomized controlled trial [6] in which participants with symptoms of insomnia and depression received either 5 weeks of CBTI (n = 52) or no treatment (n = 52). Symptoms of insomnia (Insomnia Severity Index) and depression (Patient Health Questionnaire)

were assessed at the end of each week, for 10 weeks (before treatment T0–T1, during treatment T2–T6, after treatment T7–T9).

For each week, we estimated a LASSO-regularized network that included a treatment allocation variable in addition to all symptoms. This procedure allowed us not only to follow the sequential development of treatment-induced changes in the severity of individual symptoms, but also to distinguish, at each point in time, the specific symptoms that were most directly affected by treatment and those that were indirectly affected. NIA moreover revealed the sequential development of treatment-induced changes in the network structure of associations by estimating, for each symptom, the proportion of variance that is explained by the other symptoms in the network (called predictability). Details on sample characteristics and the NIA are provided in the online supplementary material (see www.karger.com/doi/10.1159/000495045 for all online suppl. material).

Figure 1 shows the sequential development of symptom-specific treatment effects throughout 10 weeks (animated version: https://github.com/tfblanken/NIA/blob/master/docs/animated Figure.gif). The edges visualize the unique associations between variables, controlling for their associations with all the other variables (i.e., conditional dependence relationships). Since the variable "treatment" can influence the symptom variables, but not vice versa, an edge between the treatment node and a symptom identifies the symptoms that are directly affected by treatment. The effect of treatment on each symptom relative to changes observed in controls is visualized in the size of the nodes.

First, the edges between the treatment and symptom variables revealed a direct effect of treatment primarily on insomnia symptoms. Effects commenced already after the first week of treatment and continued after completion of the intervention. The sequential development indicated that treatment first and most consistently improved the sleep problems "early morning awakening" (assessments T2, T4, T7, and T9) and "difficulty maintaining sleep" (first moderately during T2 and T3, and then increasingly so from T6 to T9). Only after 4 weeks of treatment did "dissatisfaction with sleep" improve (T5-T7 and T9), indicating a sequential effect of treatment on sleep problems first and dissatisfaction later. NIA moreover indicated that the direct effect of treatment on these specific symptoms secondarily "spread" through the network via their associations with other symptoms. For example, one of the strongest treatment effects was on "worry about sleep," even though it was directly affected by treatment on only one of the assessments (T9).

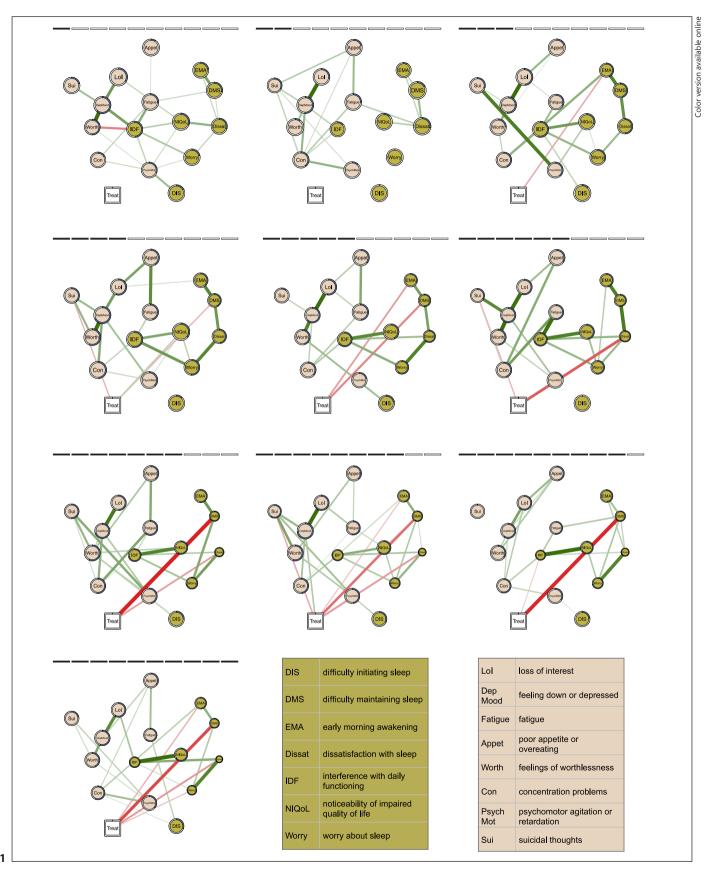
Second, NIA allowed us to observe how treatment ignited changes in the associations among the symptoms. For each symptom we inspected whether the amount of variance explained by the other symptoms (i.e., predictability) changed during treatment. The predictability systematically and robustly increased over time from, on average, 30% at baseline to, on average, 53% after assess-

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ment. There was no systematic increase or decrease in symptom variance over the course of treatment that drove the increase in predictability (r = 0.10). The finding suggests that treatment alters the structure of the network, which is in line with the network theory of mental disorders [7] and earlier findings [8].

We demonstrated NIA as a method to investigate the sequential process of symptom-specific direct and indirect effects of treatment. The use was illustrated by elucidating direct and indirect effects of CBTI on symptoms of insomnia and depression. Across all assessments, CBTI most strongly and directly affected specific sleep complaints. Accordingly, in the current sample of people suffering from insomnia and subclinical depression, the likely route for CBTI to improve symptoms of depression is indirect, through its effects on two dominant sleep complaints.

One previous study incorporated a treatment-allocation variable in network analysis [9], but only to determine the relative efficacy of two treatments on individual symptoms of depression in a pre-post comparison. A particular strength of the current NIA approach is that it additionally elucidates the sequential order of direct and indirect treatment-induced symptom improvements. Some limitations merit consideration. As in any multivariate approach, "directness" of effects should be interpreted cautiously if unmeasured variables are likely to be involved. Specific to our study, small treatment effects and weak associations may have gone undetected due to the moderate sample size and the use of LASSO regularization to ensure high specificity [10]. The absence of a direct edge should not be interpreted as an absence of any effect, but rather as an indication that an indirect effect is more likely given the available data. NIA can be considered as a statistical "searchlight" to identify plausible trajectories of treatment-induced effects.

NIA offers a technique to investigate the sequential development and order of treatment-induced changes of specific symp-

Fig. 1. The estimated regularized networks correspond to the 10 assessment weeks (2 before treatment (T0, T1), 5 during treatment (T2–T6), 3 after treatment (T7–T9); left to right, top to bottom). The networks include the Insomnia Severity Index and Patient Health Questionnaire items (circles) and treatment (square). The edges represent the conditional dependence relations among the variables that capture the unique associations among the variables, while controlling for all the other variables in the network. Green edges represent positive associations, red edges represent negative associations, and the thickness and color saturation of the edge is proportional to the strength of the association. The size of the node is proportional to the difference in symptom severity between the treatment and control group, where smaller node sizes represent greater differences in favor of the treatment group. For each node, the proportion of explained variance by the other nodes in the network, i.e., the predictability, is visualized by a ring around each node: a completely filled ring indicates that 100% of the variance is explained, a completely empty ring corresponds to an explained variance of 0%. Treatment is directly related to "early morning awakenings" (T2, T4, T7, T9), "difficulty maintaining sleep" (T3, T4, T6-T9), "suicidal thoughts" (T3, T5), "dissatisfaction with sleep" (T5-T7, T9), "psychomotor agitation" (T7)\*, or "worry about sleep" (T9). \* Note that this edge is partly obscured by the edge between treatment and "dissatisfaction with sleep." However, a small edge was retrieved in the network.

toms, highlighting likely pathways through which treatment effects evolve. Use of this technique can aid to understand treatment mechanisms and reveal clues to their optimization.

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The authors have no conflicts of interest to declare.

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### Author Contributions

T. Van Der Zweerde and J. Lancee carried out recruitment and data collection and coordinated the randomized controlled trial from which the data stem. T.F. Blanken, T. Van Der Zweerde, and J. Lancee designed the current study. T.F. Blanken performed the network analyses and interpretation of the results, supervised by D. Borsboom. T.F. Blanken and T. Van Der Zweerde drafted the manuscript. All authors contributed to and approved the manuscript.

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