

# Can domain knowledge help in handling latent variables in causal models

Yes, domain knowledge plays a crucial role in handling latent variables in causal models by providing prior information that helps reconstruct or infer the effects of unobserved variables, improving causal inference accuracy and robustness. Specifically:

- Domain knowledge can identify when variables are latent in one dataset but observed or partially known in others, enabling data reconstruction for unobserved variables and reducing confounding bias<sup>[1]</sup>.
- Prior knowledge about expected causal relationships helps guide the modeling of latent variables, even if that knowledge is imperfect, by allowing methods to detect inconsistencies and selectively apply relevant information<sup>[1]</sup>.
- Expert insights help specify plausible causal structures involving latent variables, improving model design and interpretability, and reducing false discoveries<sup>[2] [3]</sup>.
- Incorporating domain knowledge supports overcoming challenges in temporal and complex data where latent variables vary over time or are difficult to measure directly<sup>[1] [4]</sup>.
- This approach leads to more reliable causal effect estimation and better decision-making despite incomplete observations<sup>[1] [4]</sup>.

In summary, domain knowledge acts as a powerful tool to mitigate the impact of latent variables by informing causal model structure, guiding data reconstruction, and validating assumptions, thereby enhancing the effectiveness of causal AI models in real-world, imperfect data settings.

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1. <https://pmc.ncbi.nlm.nih.gov/articles/PMC7050445/>
2. <https://bg.copernicus.org/articles/19/2095/2022/>
3. [https://causalnex.readthedocs.io/en/latest/03\\_tutorial/05\\_latent\\_variable\\_tutorial.html](https://causalnex.readthedocs.io/en/latest/03_tutorial/05_latent_variable_tutorial.html)
4. <https://arxiv.org/html/2406.13966v1>