

□ SCM:

We assume that the stochastic process $(\mathbf{X}_t)_{t \in \mathbb{Z}}$ admits a description by an SCM in which at most the past q values (for some q) of all variables occur:

$$X_t^j := f^j \left((\mathbf{PA}_q^j)_{t-q}, \dots, (\mathbf{PA}_1^j)_{t-1}, (\mathbf{PA}_0^j)_t, N_t^j \right), \quad (10.1)$$

where

$$\dots, N_{t-1}^1, \dots, N_{t-1}^d, N_t^1, \dots, N_t^d, N_{t+1}^1, \dots, N_{t+1}^d, \dots$$

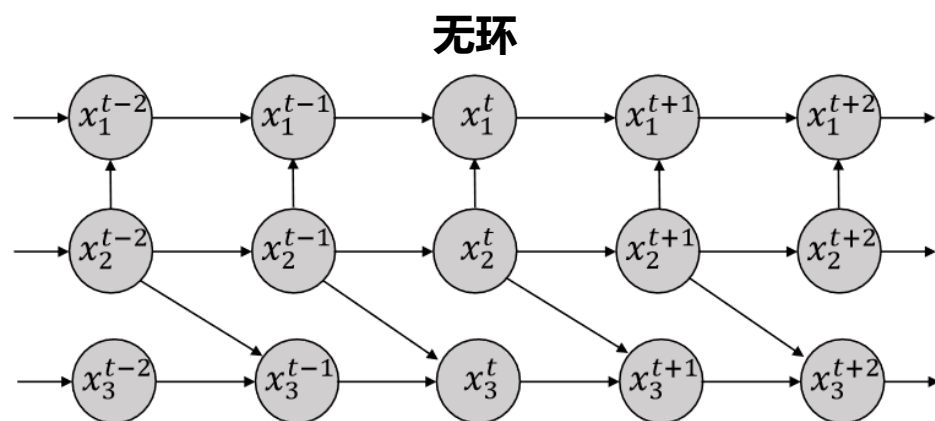
are jointly independent noise terms. Here, for each $s \in \mathbb{Z}$, the symbol $(\mathbf{PA}_s^j)_{t-s}$ denotes the set of variables X_{t-s}^k , $k = 1, \dots, d$, that influence X_t^j . Note that \mathbf{PA}_{t-s}^j may contain X_{t-s}^j for all $s > 0$, but not for $s = 0$. We assume the corresponding full time graph to be acyclic.

➤ 特例: VAR

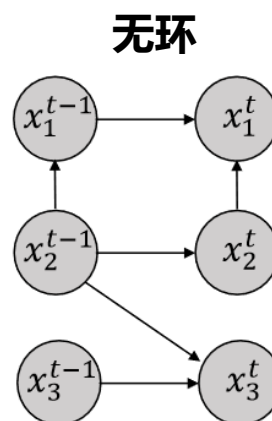
$$X_t^j := \sum_{i=1}^q A_i^j \mathbf{X}_{t-i} + N_t^j,$$

因果图：

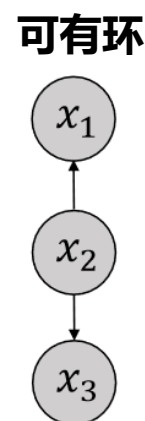
- **Full time causal graph**: 一次时序观测下，很难得到，几乎没有方法。
- **Window causal graph**: 少部分方法能得到。
- **Summary causal graph**: 大部分方法。



(a) Full time causal graph



(b) Window causal graph



(c) Summary causal graph

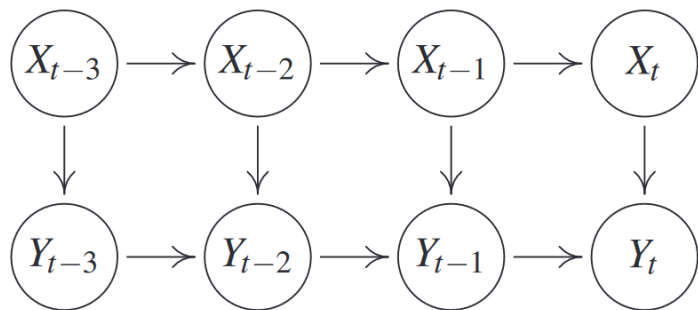


□ **时序因果可识别性：** 从观测数据中唯一确定因果方向。

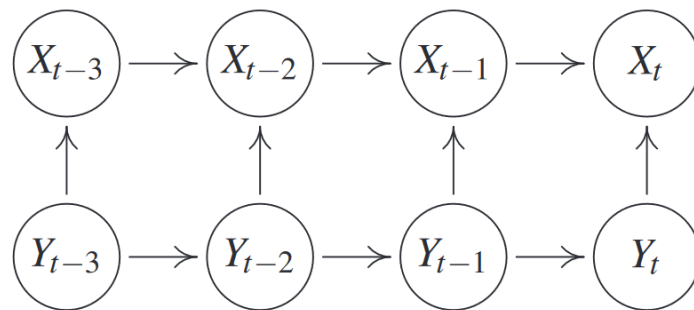
➤ **马尔科夫等价类的充要条件：**

Lemma 6.25 (Graphical criteria for Markov equivalence) *Two DAGs \mathcal{G}_1 and \mathcal{G}_2 are Markov equivalent if and only if they have the same skeleton and the same immoralities.*

➤ **时间优先性：** 原因必先于结果发生，只考虑时滞因果效应。



(a) There are v-structures at all nodes of $(Y_t)_{t \in \mathbb{Z}}$.



(b) There are v-structures at all nodes of $(X_t)_{t \in \mathbb{Z}}$.

Figure 10.5: Two DAGs that are not Markov equivalent although they coincide up to instantaneous effects.

时序因果发现



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➤ 时间优先性假设不成立的情况:

- 公鸡先于太阳升起打鸣，但太阳升起是公鸡打鸣的因:



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➤ 时间优先性假设不成立的情况:

- 欠采样/低时间分辨率数据:

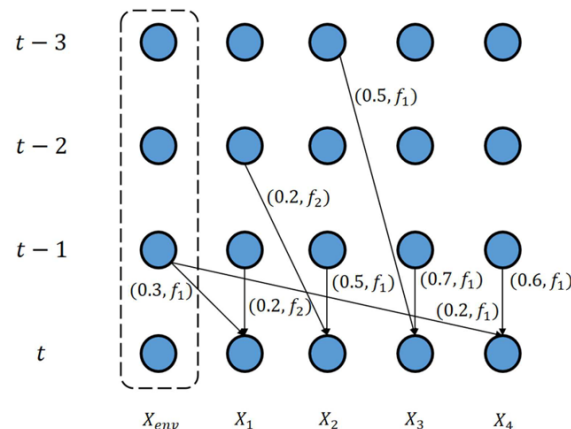
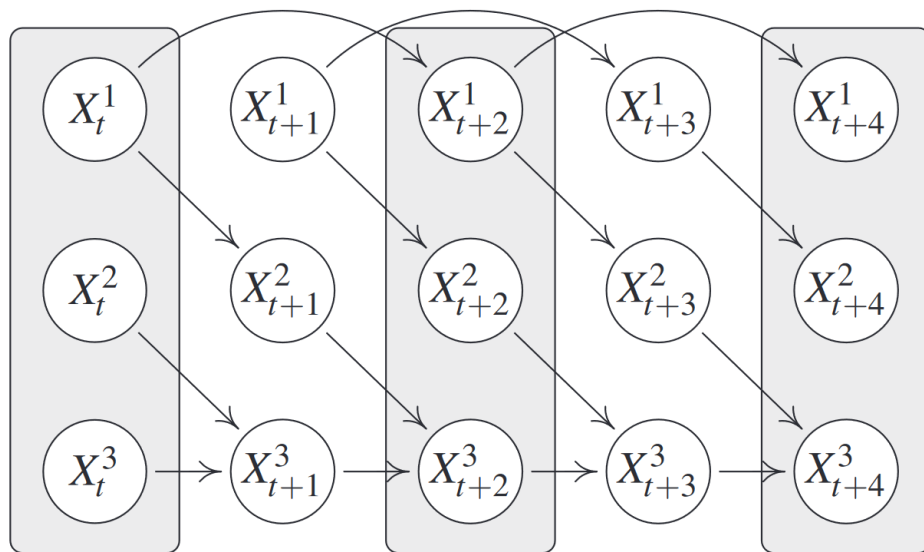


Figure 10.4: Example of a subsampled time series: only the variables in the shaded areas are observed.

在Real-World Data中，考虑瞬时+时滞因果效应更合理

[Peters J, et al. Elements of causal inference: foundations and learning algorithms[M]. The MIT Press, 2017.]

[Kang M, et al. Identifying Unique Causal Network from Nonstationary Time Series[J]. arXiv preprint arXiv:2211.10085, 2022.]