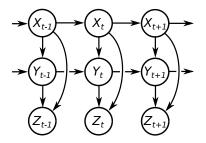
Fundamentals of Probabilistic Data Mining Chapter II - Erratum on class exercise

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Ensimag/Inria

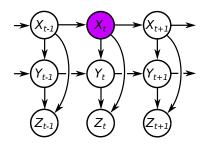
The model and the question

Given the following model,

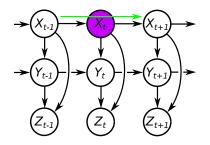


the question in class was weather or not $\{X_{t-1}\}$ was D-separated from $\{X_{t+1}\}$ by X_t . I wrongly answered that is was not, when it actually is.

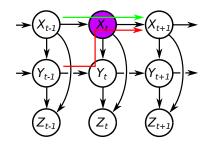
In this erratum we will see why, both graphically and algebraically.



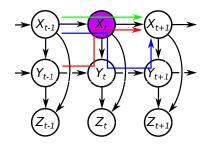
Let us consider all paths from $\{X_{t-1}\}$ to $\{X_{t+1}\}$.



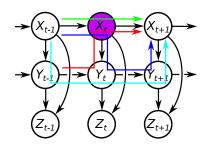
Let us consider all paths from $\{X_{t-1}\}$ to $\{X_{t+1}\}$. Blocked by $\{X_t\}$ because head-to-tail at X_t .



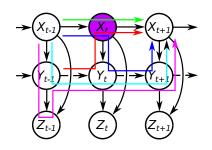
Let us consider all paths from $\{X_{t-1}\}$ to $\{X_{t+1}\}$. Blocked by $\{X_t\}$ because head-to-tail at X_t . Blocked by $\{X_t\}$ because tail-to-tail at X_t .



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Let us consider all paths from \{X_{t-1}\} to \{X_{t+1}\}. Blocked by \{X_t\} because head-to-tail at X_t. Blocked by \{X_t\} because tail-to-tail at X_t. Blocked by \{X_t\} because head-to-tail at X_t.
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Let us consider all paths from \{X_{t-1}\} to \{X_{t+1}\}.
Blocked by \{X_t\} because head-to-tail at X_t.
Blocked by \{X_t\} because tail-to-tail at X_t.
Blocked by \{X_t\} because head-to-tail at X_t.
Blocked by \{X_t\} because head-to-head at Y_{t+1} (and Y_{t+1} is not in \{X_t\}).
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Let us consider all paths from \{X_{t-1}\} to \{X_{t+1}\}. Blocked by \{X_t\} because head-to-tail at X_t. Blocked by \{X_t\} because head-to-tail at X_t. Blocked by \{X_t\} because head-to-tail at X_t. Blocked by \{X_t\} because head-to-head at Y_{t+1} (and Y_{t+1} is not in \{X_t\}). Blocked by \{X_t\} because head-to-head at Z_{t-1} (and Z_{t-1} is not in \{X_t\}). The same rule applies to any path going through any "Z".
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Algebraic answer

The model writes:

$$p(x_{1:T}, y_{1:T}, z_{1:T}) = p(z_1|y_1, x_1)p(y_1|x_1)p(x_1) \times \prod_{t \ge 2} p(z_t|y_t, x_t)p(y_t|y_{t-1}, x_t)p(x_t|x_{t-1})$$

We seek for $p(x_{t+1}, x_{t-1}|x_t)$ and use the Bayes theorem for that

$$p(x_{t+1},x_{t-1}|x_t) = \frac{p(x_{t+1},x_{t-1},x_t)}{p(x_t)}.$$

Now we "just" need to compute $p(x_{t+1}, x_{t-1}, x_t)$.

Albegraic answer (II)

Let's compute by marginalizing (I skipped the "d"):

$$\rho(x_{t+1}, x_{t-1}, x_t) = \int_{x_{1:t-2}, x_{t+2:T}, y_{1:T}, z_{1:T}} \rho(z_1|y_1, x_1) \rho(y_1|x_1) \rho(x_1) \times \prod_{t \ge 2} \rho(z_t|y_t, x_t) \rho(y_t|y_{t-1}, x_t) \rho(x_t|x_{t-1})$$

The z's integrate to one. Similarly for the y's.

$$p(x_{t+1}, x_{t-1}, x_t) = \int_{x_{1:t-2}, x_{t+2:T}} p(x_1) \prod_{t \ge 2} p(x_t | x_{t-1})$$

The x's from t+2 to T integrate to one as well, and the previous ones, I can write them as:

$$p(x_{t+1},x_{t-1},x_t)=p(x_{t+1}|x_t)p(x_t|x_{t-1})\int_{x_{1,t-2}}p(x_1,\ldots,x_{t-1})$$

Albegraic answer (III)

The integral is easy to write:

$$p(x_{t+1}, x_{t-1}, x_t) = p(x_{t+1}|x_t)p(x_t|x_{t-1})p(x_t)$$

Going back to Bayes:

$$p(x_{t+1}, x_{t-1}|x_t) = \frac{p(x_{t+1}, x_{t-1}, x_t)}{p(x_t)} = \frac{p(x_{t+1}|x_t)p(x_t|x_{t-1})p(x_{t-1})}{p(x_t)}$$

which is:

$$p(x_{t+1},x_{t-1}|x_t)=p(x_{t+1}|x_t)p(x_{t-1}|x_t),$$

which obviously aligns with the graphical result.