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Q1. Given g(X1-T|X0) = 1 g(Xt | Xt-1)
                       Show g(X1-T | X0) = g(XT | X0) 1 g(Xt1 | Xt, X0)
                                     since X1, X2...., XT from a Markov chain when conditioned on Xo.:
                                                        => & (X1-7/X0) = TT & (Xt) Xt-1, X0
                           \Rightarrow g(X_{1-T}|X_{0}) = \frac{g(X_{T-1}, X_{0})}{g(X_{T-1}, X_{0})} \times \frac{g(X_{T-1}, X_{T-2}, X_{0})}{(X_{T-2}, X_{0})}
                                                                                                          2(X2, X0) X 2(X1, X0) X 2(X1, X0) X 2(X0)
                           \Rightarrow 2(X_{1-1}(X_{0})) = \frac{2(X_{1}, X_{1-1}, X_{0})}{2(X_{0})} \times \frac{2(X_{1-1}, X_{1-2}, X_{0})}{2(X_{1-1}, X_{0})} \times \cdots \times \frac{2(X_{2}, X_{1}, X_{0})}{2(X_{2}, X_{0})}
                                = \frac{2(X_{7}, X_{7-1}, X_{0})}{2(X_{0})} \times \left[2(X_{7-2} | X_{7-1}, X_{0}) \times \cdots \times 2(X_{1} | X_{2}, X_{0})\right]
                                                                                                      2 \(\frac{g(\times_{\tau}, \times_{\to})}{g(\times_{\tau}, \times_{\to})} \times \(\frac{\frac{g(\times_{\tau}, \times_{\to})}{g(\times_{\tau}, \times_{\to})}} \times \(\frac{\frac{1}{7}}{7} \\ \frac{g(\times_{\tau}, \times_{\to})}{g(\times_{\tau}, \times_{\to})} \times \(\frac{1}{7} \\ \frac{1}{7} \\ \frac{1} \\ \frac{
                                                                              = Z(XT/X0) x [Z(XT-1/XT, X0) x TT Z(Xt-1/Xt, X0)]
                                                                                                                = 2 (XT/X0) 1 2 (Xt-1/Xt, X0) *
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eq(b). We know that 2(Xt|Xt-1,Xo) = 2(Xt|Xt-1) since Markov Chain. 2(Xt|Xt-1,Xo) = 2(Xt|Xt-1) = N(II-Bt|Xt-1, BtI)And we also know that 2(Xt|Xo) = N(IZt|Xo, (I-Zt)I) from eq(4).

C(Xt, Xo) is an irrelevant part of Xt+, so we can omit it.

Thus, q(Xt-1|Xt, Xo) is Gaussian Distribution.

And 
$$\widetilde{\beta}_{t} = \frac{1}{\left(\frac{\partial L}{\beta_{t}} + \frac{1}{1 - \overline{\partial L}_{-1}}\right)} = \frac{1 - \overline{\partial L}_{1}}{\left(\frac{\partial L}{\beta_{t}} + \overline{\partial L}_{-1}}\right)} = \frac{1 - \overline{\partial L}_{1}}{1 - \overline{\partial L}_{1}} \cdot \beta_{t}$$

$$\widetilde{\mu}_{t}(X_{t}, X_{0}) = \frac{\left(\frac{\overline{\partial L}}{\beta_{t}} + \frac{\overline{\partial L}_{-1}}{1 - \overline{\partial L}_{1}} X_{0}\right)}{\left(\frac{\partial L}{\beta_{t}} + \frac{\overline{\partial L}_{-1}}{1 - \overline{\partial L}_{1}} X_{0}\right)} = \left(\frac{\overline{\partial L}}{\beta_{t}} + \frac{\overline{\partial L}_{-1}}{1 - \overline{\partial L}_{1}} X_{0}\right) \cdot \frac{1 - \overline{\partial L}_{-1}}{1 - \overline{\partial L}_{1}} \cdot \beta_{t}$$

$$= \frac{\overline{\partial L}(1 - \overline{\partial L}_{1})}{1 - \overline{\partial L}} \times L + \frac{\overline{\partial L}_{-1}}{1 - \overline{\partial L}} \times L$$

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:. g(Xt, | Xt, Xo) = N(Xt-1; M+(Xt, Xo), B+I)

where 
$$\widetilde{\mathcal{U}}_{t}(X_{t},X_{0}):=\frac{\sqrt{\lambda_{t-1}}\beta_{t}}{1-\lambda_{t}}X_{0}+\frac{\sqrt{\lambda_{t}}(1-\overline{\lambda_{t-1}})}{1-\overline{\lambda_{t}}}X_{t}$$
 and  $\widetilde{\beta}_{t}:=\frac{1-\overline{\lambda_{t-1}}}{1-\overline{\lambda_{t}}}\beta_{t}$ 

$$\begin{split} & = \mathbb{E}_{q} \left[ -\log \frac{p_{\theta}(X_{0:T})}{q(X_{1:T}|X_{0})} \right] \\ & = \mathbb{E}_{q} \left[ -\log p(X_{T}) - \sum_{t \geq 1} \log \frac{p_{\theta}(X_{t-1}|X_{t})}{q(X_{t}|X_{t-1})} \right] \\ & = \mathbb{E}_{q} \left[ -\log p(X_{T}) - \sum_{t \geq 1} \log \frac{p_{\theta}(X_{t-1}|X_{t})}{q(X_{t}|X_{t-1})} - \log \frac{p_{\theta}(X_{0}|X_{1})}{q(X_{1}|X_{0})} \right] \\ & = \mathbb{E}_{q} \left[ -\log p(X_{T}) - \sum_{t \geq 1} \log \frac{p_{\theta}(X_{t-1}|X_{t})}{q(X_{t-1}|X_{t},X_{0})} - \frac{q(X_{t-1}|X_{0})}{q(X_{t}|X_{0})} - \log \frac{p_{\theta}(X_{0}|X_{1})}{q(X_{1}|X_{0})} \right] \\ & = \mathbb{E}_{q} \left[ -\log \frac{p(X_{T})}{q(X_{T}|X_{0})} - \sum_{t \geq 1} \log \frac{p_{\theta}(X_{t-1}|X_{t})}{q(X_{t-1}|X_{t},X_{0})} - \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{t-1}|X_{t},X_{0})||p_{\theta}(X_{t-1}|X_{t})) - \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{t-1}|X_{t},X_{0})||p_{\theta}(X_{t-1}|X_{t})) - \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{t-1}|X_{t},X_{0})||p_{\theta}(X_{t-1}|X_{t})) - \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{t-1}|X_{t},X_{0})||p_{\theta}(X_{t-1}|X_{t})) - \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{t-1}|X_{t},X_{0})||p_{\theta}(X_{t-1}|X_{t})) - \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{t-1}|X_{t}),X_{0}) + \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{t-1}|X_{t}),X_{0}) + \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \sum_{t \geq 1} D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \log p_{\theta}(X_{0}|X_{1}) \right] \\ & = \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + \mathbb{E}_{q} \left[ D_{KL}(q(X_{T}|X_{0})||p(X_{T})) + D_{KL}(q(X_{T}|X_{0})||p(X_{T})) \right] \right]$$

Then PKL(q(Xt1 | Xt, Xo) | Po(Xt1 | Xt))

$$= D_{KL}(N(X_{t-1}; \widetilde{\mu}(X_{t}, X_{0}), \sigma_{t}^{2} I) \| N(X_{t-1}; M_{0}(X_{t}, t), \sigma_{t}^{2} I))$$

$$= \frac{1}{2} (n + \frac{1}{\sigma_{t}^{2}} \| \widetilde{\mu}_{t}(X_{t}, X_{0}) - M_{0}(X_{t}, t) \|^{2} - n + \log I)$$

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$$= \frac{1}{2\sigma_{t}^{2}} \| \widetilde{\mu}_{t}(X_{t}, X_{0}) - M_{0}(X_{t}, t) \|^{2}$$

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$$= \frac{1}{2\sigma_{t}^{2}} \| \widetilde{\mu}_{t}(X_{t}, X_{0}) - M_{0}(X_{t}, t) \|^{2} + C.$$