

31.1

Advanced Machine Learning

$$\begin{aligned}
 \text{a.) we have } p(\vec{x}) &= \frac{1}{Z(\beta)} \exp(-\beta \tilde{E}(\beta)) \\
 \ln(1/p(\vec{x})) &= \ln(Z(\beta)) + \ln(\exp(\beta \tilde{E}(\beta))) \\
 &= \ln(Z(\beta)) + \beta \tilde{E}(\beta)
 \end{aligned}$$

$$\Rightarrow S = \sum p(\vec{x}) (\ln(1/p(\vec{x})))$$

$$= \ln(Z(\beta)) + \beta \tilde{E}(\beta) \left(\sum_x p(\vec{x}) \right)$$

This is normalized, and thus equal to 1

$$\Rightarrow S = \ln(Z(\beta)) + \beta \tilde{E}(\beta) \quad \square$$

$$\begin{aligned}
 \text{b.) } -\frac{\partial F}{\partial T} &= -\frac{\partial F}{\partial \beta} \frac{\partial \beta}{\partial T} = -\frac{\partial F}{\partial \beta} \left(\frac{\partial}{\partial T} \left(\frac{1}{T} \right) \right) = \frac{1}{T^2} \frac{\partial F}{\partial \beta} \\
 &= \beta^2 \frac{\partial F}{\partial \beta} \quad \text{where } F = -\frac{1}{\beta} \ln(Z(\beta))
 \end{aligned}$$

$$\Rightarrow \frac{\partial F}{\partial \beta} = \frac{1}{\beta^2} \ln(Z(\beta)) - \frac{1}{\beta} \frac{\partial}{\partial \beta} \ln(Z(\beta))$$

$$\rightarrow -\partial F / \partial T = \ln(Z(\beta)) + -\beta \frac{\partial}{\partial \beta} \ln(Z(\beta))$$

$$\begin{aligned}
 (\text{eq 31.8 McK}) &= \ln(Z(\beta)) + \beta \tilde{E}(\beta) = \boxed{-\partial F / \partial T} \\
 &= \boxed{S}
 \end{aligned}$$

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