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so many fake sites. this is the first one which worked! Many thanks

But  $X, Y$  are independent, so:  $p(x, y) = p(x)p(y)$ ,  $p(x|y) = p(x)$ . Hence:

$$\begin{aligned} H(X|Y) &= - \int_{-\infty}^{\infty} p(y) \int_{-\infty}^{\infty} p(x) \log p(x) dx dy \\ &= - \int_{-\infty}^{\infty} p(y) \log p(y) dy \\ &= H(Y) = \frac{1}{2} \log(2\pi\sigma_y^2) \end{aligned}$$

where the last equality stems from the Gaussian pdf of  $X$ .

(b)

$$H(X|Y) = H(Y) - H(Y|X)$$

Since  $Y$  is the sum of two independent, zero-mean Gaussian  $x$ 's, it is also a zero-mean Gaussian  $r.v.$  with variance:  $\sigma_y^2 = \sigma_x^2 + \sigma_x^2$ . Hence:  $H(Y) = \frac{1}{2} \log(2\pi(\sigma_x^2 + \sigma_x^2))$ . Also, since  $y = x + y$ :

$$p(y) = p(y-x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} e^{-\frac{y^2}{2\sigma_x^2}}$$

Hence:

$$\begin{aligned} H(Y|X) &= - \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p(x, y) \log p(x, y) dx dy \\ &= - \int_{-\infty}^{\infty} p(x) \log p(x) \int_{-\infty}^{\infty} p(y|x) \log \left( \frac{1}{\sqrt{2\pi\sigma_x^2}} \exp\left(-\frac{(y-x)^2}{2\sigma_x^2}\right) \right) dy dx \\ &= \int_{-\infty}^{\infty} p(x) \log p(x) \left[ \int_{-\infty}^{\infty} p(y|x) \left( \ln(\sqrt{2\pi\sigma_x^2}) + \frac{(y-x)^2}{2\sigma_x^2} \right) dy \right] dx \\ &= \int_{-\infty}^{\infty} p(x) \log p(x) \left[ \ln(\sqrt{2\pi\sigma_x^2}) + \frac{1}{2\sigma_x^2} \int_{-\infty}^{\infty} (y-x)^2 p(y|x) dy \right] dx \\ &= \left[ \ln(\sqrt{2\pi\sigma_x^2}) + \frac{1}{2} \log(2\pi\sigma_x^2) \right] \int_{-\infty}^{\infty} p(x) dx \\ &= \frac{1}{2} \log(2\pi\sigma_x^2) \quad (= H(X)) \end{aligned}$$

where we have used the fact that:  $\int_{-\infty}^{\infty} p(y) dy = 1$ ,  $\int_{-\infty}^{\infty} (y-x)^2 p(y|x) dy = E\{Y^2|X\} = \sigma_x^2$ .

From  $H(Y) = H(Y|X) + H(X|Y)$ :

$$H(X|Y) = H(Y) - H(Y|X) = \frac{1}{2} \log(2\pi(\sigma_x^2 + \sigma_x^2)) - \frac{1}{2} \log(2\pi\sigma_x^2) = \frac{1}{2} \log\left(1 + \frac{\sigma_x^2}{\sigma_x^2}\right)$$

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