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In particular, if  $\sigma_p = \sigma_f = \sigma$ , then

$$f(x_f | x_p, \sigma_p = \sigma_f = \sigma) = \frac{1}{\sqrt{2\pi} \sqrt{2}\sigma} \exp\left[-\frac{(x_f - x_p)^2}{2(\sqrt{2}\sigma)^2}\right]$$