

Derivation of the mean and variance with the posterior probability

$$\pi(\mu|D) \propto \underbrace{\exp\left[-\frac{(\bar{x}-\mu)^2}{2\frac{\sigma^2}{n}}\right]}_{\text{likelihood}} \times \underbrace{\exp\left[-\frac{(\mu_0-\mu)^2}{2\sigma_0^2}\right]}_{\text{prior}}$$

posterior

The exponential part

$$= -\left[\frac{(\bar{x}-\mu)^2}{2\frac{\sigma^2}{n}} + \frac{(\mu_0-\mu)^2}{2\sigma_0^2}\right]$$

$$= -\frac{1}{2\sigma_0^2} \left[ \frac{n\sigma_0^2}{\sigma^2} (\mu-\bar{x})^2 + (\mu-\mu_0)^2 \right]$$

$$= -\frac{1}{2\sigma_0^2} \left[ \frac{n\sigma_0^2}{\sigma^2} (\mu^2 - 2\mu\bar{x} + \bar{x}^2) + (\mu^2 - 2\mu_0\mu + \mu_0^2) \right]$$

$$= -\frac{1}{2\sigma_0^2} \left[ \frac{n\sigma_0^2}{\sigma^2} \mu^2 - \frac{n\sigma_0^2}{\sigma^2} 2\mu\bar{x} + \frac{n\sigma_0^2}{\sigma^2} \bar{x}^2 + \mu^2 - 2\mu_0\mu + \mu_0^2 \right]$$

$$= -\frac{1}{2\sigma_0^2} \left[ \left(\frac{n\sigma_0^2}{\sigma^2} + 1\right) \mu^2 - \left(\frac{n\sigma_0^2}{\sigma^2} \bar{x} + \mu_0\right) 2\mu + \left(\frac{n\sigma_0^2}{\sigma^2} \bar{x}^2 + \mu_0^2\right) \right]$$

Express the inside bracket in a perfect square.

$$= -\left[ \frac{1}{2\sigma_0^2} \left(\frac{n\sigma_0^2}{\sigma^2} + 1\right) \left( \mu - \frac{\frac{n\sigma_0^2}{\sigma^2} \bar{x} + \mu_0}{\frac{n\sigma_0^2}{\sigma^2} + 1} \right)^2 + \{\text{constant}\} \right]$$

Therefore, the posterior  $\mu_1$  and  $\sigma_1$  will be

$$\mu_1 = \frac{\frac{n\sigma_0^2}{\sigma^2} \bar{x} + \mu_0}{\frac{n\sigma_0^2}{\sigma^2} + 1} = \frac{\frac{n\bar{x}}{\sigma^2} + \frac{\mu_0}{\sigma_0^2}}{\frac{n}{\sigma^2} + \frac{1}{\sigma_0^2}}$$

$$\sigma_1^2 = \frac{\sigma_0^2}{\frac{n\sigma_0^2}{\sigma^2} + 1} = \frac{1}{\frac{n}{\sigma^2} + \frac{1}{\sigma_0^2}}$$