

- Gauss-Legendre:

$$\int_{-1}^1 f(x)dx \approx \sum_i w_i f(x_i) \Rightarrow \int_a^b g(y)dy \approx \sum_i \omega_i g(y_i)$$

- Gauss-Laguerre:

$$\int_0^\infty e^{-x} f(x)dx \approx \sum_i w_i f(x_i) \Rightarrow \int_a^\infty g(y)dy \approx \sum_i \omega_i g(y_i)$$

- Gauss-Hermite: Often used for integrating over Normal  $(\mu, \sigma^2)$  expectations

$$\int_{-\infty}^\infty e^{-x^2} f(x)dx \approx \sum_i w_i f(x_i) \Rightarrow \int_{-\infty}^\infty \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left[-\frac{(y-\mu)^2}{2\sigma^2}\right] g(y)dy \approx \sum_i \omega_i g(y_i)$$

The idea of quadratures is to choose optimal weights and nodes  $(\omega_i, y_i)$  to improve accuracy with as few points as possible. The weights and nodes  $(w_i, x_i)$  are fixed for the standard cases, which we can use simple change of variables to integrate any integral of the above forms (for  $f(\cdot)$  that are  $C_1$ ). Standard weights and nodes for  $N$  points can be easily found online (of a reference textbook).

**Gauss-Legendre** For Gauss-Legendre, transform the weights and nodes  $(w_i, x_i)$  from the standard bounds of support  $(-1, 1)$  as follows:

$$\int_a^b g(y)dy = \frac{b-a}{2} \cdot \int_{-1}^1 \underbrace{g\left(\frac{b-a}{2} \cdot x + \frac{b+a}{2}\right)}_{\equiv f(x)} dx \approx \sum_i \left(\frac{b-a}{2} \cdot w_i\right) \cdot g(y_i)$$

so set the weights and nodes as

$$\omega_i = \frac{b-a}{2} \cdot w_i, \quad y_i = \frac{b-a}{2} \cdot x_i + \frac{b+a}{2}$$

**Gauss-Laguerre**

$$\int_a^\infty g(y)dy = \int_0^\infty \underbrace{g(x+a)}_{\equiv f(x)} dx = \int_0^\infty e^x e^{-x} f(x)dx \approx \sum_i w_i e^{x_i} f(x_i) = \sum_i w_i e^{y_i-a} g(y_i)$$

so set the weights and nodes as

$$\omega_i = w_i \exp(y_i - a), \quad y_i = x_i + a$$

**Gauss-Hermite** Clearly  $y = \sqrt{2}\sigma x + \mu$ , so

$$\omega_i = w_i / \sqrt{\pi}, \quad y_i = \sqrt{2}\sigma x_i + \mu$$