

## Estimation of log-normal distribution from data

- Assume a known value for the mean ( $\mu_x$ ) of the sample and the coefficient of variation ( $cv$ ). Use VALE values and the CV obtained by data conciliation
- Assume that the coefficient of variation of the random variable  $x$  is constant after the transformation into a random variable with a log-normal distribution  $w = \ln(x)$ . So,  $cv_x = cv_w$
- Estimate the mean from the data or assume VALE mean as the arithmetic mean for the batch

$$\mu_x = \frac{1}{n} \sum_{i=1}^n x_i$$

- Estimate the standard deviation from the coefficient of variation of the data or assume a known cv for the batch

$$\sigma_x = cv_x \cdot \mu_x$$

- Estimate the mean of the log-normal distribution by using the equation:

$$\exp(\mu_w) = \frac{\mu_x^2}{\sqrt{\mu_x^2 + \sigma_x^2}}$$

- Estimate the standard deviation of the log-normal distribution by using the equation:

$$cv_w = \sqrt{\exp(\sigma_w^2) - 1} \rightarrow \sigma_w = \sqrt{\ln(1 + cv^2)}$$

- Compute the distribution:

$$f(x) = \frac{\exp\left[-\frac{1}{2} \left(\frac{\ln x - \mu_w}{\sigma_w}\right)^2\right]}{x \sigma_w \sqrt{2\pi}}$$

- Compute the other statistics as:

- mean =  $\exp(\mu_w + \frac{1}{2}\sigma_w^2)$
- median =  $\exp(\mu_w)$
- mode =  $\exp(\mu_w - \sigma_w^2)$
- variance =  $[\exp(\sigma_w^2) - 1] \cdot \exp(2\mu_w + \sigma_w^2)$