

In the given PyMC code example, `pm.logp(mu, 0)` represents the log-probability of the random variable `mu` taking the value 0. Mathematically, this can be expressed as:  $\log(p(\mu = 0))$

where:

- `mu` is the random variable representing the mean of the Gaussian distribution.
- `p(mu = 0)` is the probability density function (PDF) of the Gaussian distribution evaluated at `mu = 0`.
- `log()` is the natural logarithm function.

$$p.m.logp(\mu, 0) \Big|_{@x=0}$$

The PDF of a Gaussian distribution with mean  $\mu$  and standard deviation  $\sigma$  is given by:

$$p(x) = (1 / (\sigma * \sqrt{2\pi})) * \exp(-(x - \mu)^2 / (2\sigma^2))$$

where:

- `x` is the value at which the PDF is evaluated.
- `μ` is the mean of the Gaussian distribution.
- `σ` is the standard deviation of the Gaussian distribution.
- `exp()` is the exponential function.
- `sqrt()` is the square root function.

$$\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \Big|_{@x=0, \mu=0, \sigma=1} \left| \frac{1}{\sqrt{2\pi}} \times e^{(0-0)} \right|_{\mu=0} = \frac{1}{\sqrt{2\pi}}$$

In the case of `pm.logp(mu, 0)`, we are evaluating the log-probability of `mu` at `x = 0`. So, plugging in `x = 0` and the specified values of `μ = 0` and `σ = 1` into the PDF equation, we get:  $p(\mu = 0) = (1 / (1 * \sqrt{2\pi})) * \exp(-(0 - 0)^2 / (2 * 1^2)) = (1 / \sqrt{2\pi}) * \exp(0) = (1 / \sqrt{2\pi})$

Taking the natural logarithm of both sides, we get:

$$\log(p(\mu = 0)) = \log((1 / \sqrt{2\pi})) = -0.5 * \log(2\pi) = -0.9189385332046727$$

Therefore, the mathematical equation for `pm.logp(mu, 0)` in this specific case is:

$$\log(p(\mu = 0)) = -0.5 * \log(2\pi)$$

$$\left| \log \left( \frac{1}{\sqrt{2\pi}} \right) = -0.5 * \log 2\pi \right. \\ \left. = -0.91 \right|$$