

# Injecting a signal into a $\mathcal{M}$ -bit quantised data stream

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Assuming that the background noise is represented by a Gaussian distribution, the probability density function (pdf) of the noise  $\mathcal{N}$  is:

$$P(\mathcal{N}) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(\mathcal{N}-\mu)^2}{2\sigma^2}}$$

where the noise level  $\sigma$  is given by the **radiometer equation** for a single antenna:

$$\sigma = \frac{T_{sys}}{G\sqrt{N_{pol}}\delta t\Delta\nu}$$

The cumulative distribution function of noise is given by:

$$\phi(\mathcal{N}) = \frac{1}{2} \left[ 1 + \operatorname{erf} \left( \frac{\mathcal{N} - \mu}{\sigma\sqrt{2}} \right) \right]$$

For a given signal  $\mathcal{S}$  injected to data samples, we get digitized values ranging from 0 to  $2^{\mathcal{M}} - 1$  for the  $\mathcal{M}$  bit case. The probability matrix for digit changes upon injecting a signal would be:

$$P_{ij} = \begin{pmatrix} P_{00} & P_{01} & \cdots & P_{0(2^{\mathcal{M}}-1)} \\ 0 & P_{11} & \cdots & P_{1(2^{\mathcal{M}}-1)} \\ \vdots & & \ddots & \\ 0 & 0 & \cdots & P_{(2^{\mathcal{M}}-2)(2^{\mathcal{M}}-1)} \\ 0 & 0 & \cdots & 1 \end{pmatrix}$$

where  $P_{nm}$  is the probability to change the bit from state  $n$  to the state  $m$  and is given by:

$$P_{nm} = \frac{\text{Probability to go from state } n \text{ to state } m}{\text{Probability to remain in original state } n}$$

Say we have  $\mathcal{M}$  bit data; thus there are  $2^{\mathcal{M}}$  levels. Consider state  $n$  lying between  $\mu + i l \sigma$  and  $\mu + (i+1) l \sigma$ , where  $\mu$  is mean of Gaussian noise and  $l$  is level setting number for digitization. Here,  $i$  runs from  $-(\frac{2^{\mathcal{M}}}{2} - 1)$  to  $+(\frac{2^{\mathcal{M}}}{2} - 1)$  and  $n$  varies from 0 to  $2^{\mathcal{M}} - 1$ . As shown in the figure (1), the probability of bit shift can be calculated as,

$$P_{n \rightarrow n+m} = \frac{\text{Area under pdf between intersection of states } n \text{ and } n+m}{\text{Area under pdf up to state } n}$$

$$P_{n \rightarrow n+m} = \frac{\phi(\min(\mu + (i+m)l\sigma - \mathcal{S}, \mu + il\sigma)) - \phi(\max(\mu + (i+m-1)l\sigma - \mathcal{S}, \mu + (i-1)l\sigma))}{\phi(\mu + il\sigma) - \phi(\mu + (i-1)l\sigma)} \quad (1)$$

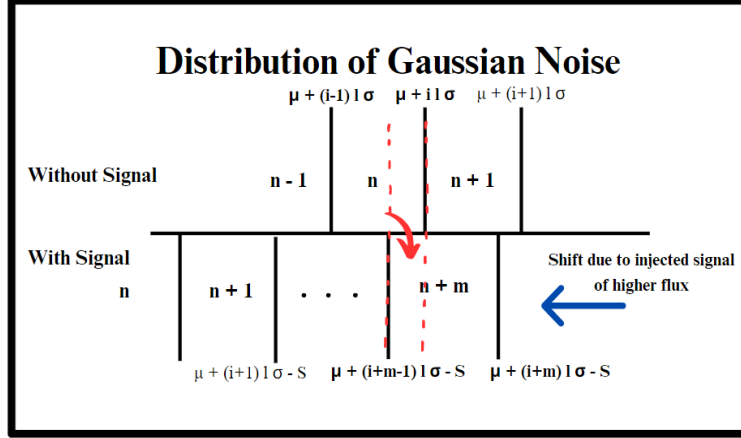


Figure 1: Distribution of Gaussian noise

where  $n$  and  $i$  are related as  $n - i = \frac{2^M}{2} - 1$ .

The first state (0) extends from  $-\infty$  to  $\mu - (\frac{2^M}{2} - 1)l\sigma$  and the last state ( $2^M - 1$ ) extends from  $\mu + (\frac{2^M}{2} - 1)l\sigma$  to  $+\infty$ . Therefore, we need to accommodate these variation at the boundary states. This can be done by modifying above formula as,

$$P_{0 \rightarrow 0} = \frac{\phi(\mu + (1 - \frac{2^M}{2})l\sigma - S)}{\phi(\mu + (1 - \frac{2^M}{2})l\sigma)} \quad (2)$$

$$P_{0 \rightarrow m} = \frac{\phi(\min(\mu + (m + 1 - \frac{2^M}{2})l\sigma - S, \mu + (1 - \frac{2^M}{2})l\sigma)) - \phi(\mu + (m - \frac{2^M}{2})l\sigma - S)}{\phi(\mu + (1 - \frac{2^M}{2})l\sigma)} \quad (3)$$

$$P_{n \rightarrow 2^M - 1} = \frac{\phi(\mu + (n + 1 - \frac{2^M}{2})l\sigma) - \phi(\max(\mu + (\frac{2^M}{2} - 1)l\sigma - S, \mu + (n - \frac{2^M}{2})l\sigma))}{\phi(\mu + (n + 1 - \frac{2^M}{2})l\sigma) - \phi(\mu + (n - \frac{2^M}{2})l\sigma)} \quad (4)$$

$$P_{0 \rightarrow 2^M - 1} = \frac{\phi(\mu + (1 - \frac{2^M}{2})l\sigma) - \phi(\mu + (\frac{2^M}{2} - 1)l\sigma - S)}{\phi(\mu + (1 - \frac{2^M}{2})l\sigma)} = 1 - \frac{\phi(\mu + (\frac{2^M}{2} - 1)l\sigma - S)}{\phi(\mu + (1 - \frac{2^M}{2})l\sigma)} \quad (5)$$

and,

$$P_{2^M - 1 \rightarrow 2^M - 1} = 1 \quad (6)$$

Equation (1) to (6) are general formulas to calculate probability of bit change from one state to other when a signal  $S$  is injected into the quantized data stream.

Now, **for the 8-bit case**, we have  $n - i = \frac{2^8}{2} - 1 = 127$ . Thus,

$$P_{n \rightarrow n+m} = \frac{\phi(\min(\mu + (n + m - 127)l\sigma - S, \mu + (n - 127)l\sigma)) - \phi(\max(\mu + (n + m - 128)l\sigma - S, \mu + (n - 128)l\sigma))}{\phi(\mu + (n - 127)l\sigma) - \phi(\mu + (n - 128)l\sigma)}$$

The formula is appropriate given we are away from boundary states (the first and the last state). Probability of changing bit from 0 to 0 and 255 to 255, is

$$P_{0 \rightarrow 0} = \frac{\phi(\mu - 127l\sigma - S)}{\phi(\mu - 127l\sigma)} \text{ and } P_{255 \rightarrow 255} = 1, \text{ respectively.} \quad (7)$$

For bit change from state 0 to any higher state  $m$ , the formula modifies as

$$P_{0 \rightarrow m} = \frac{\phi(\min(\mu + (m - 127)l\sigma - S, \mu - 127l\sigma)) - \phi(\mu + (m - 128)l\sigma - S)}{\phi(\mu - 127l\sigma)}$$

Similarly, for bit change from any non-zero state  $n$  to the highest state 255, the probability is

$$P_{n \rightarrow 255} = \frac{\phi(\mu + (n - 127)l\sigma) - \phi(\max(\mu + 127l\sigma - S, \mu + (n - 128)l\sigma))}{\phi(\mu + (n - 127)l\sigma) - \phi(\mu + (n - 128)l\sigma)}$$

Lastly, the probability for bit change from the lowest state 0 to the highest state 255 is

$$P_{0 \rightarrow 255} = \frac{\phi(\mu - 127l\sigma) - \phi(\mu + 127l\sigma - S)}{\phi(\mu - 127l\sigma)} = 1 - \frac{\phi(\mu + 127l\sigma - S)}{\phi(\mu - 127l\sigma)}$$

**For a 1-bit case**,  $\mathcal{M}$  is 1 and thus there are only 2 levels 0 and 1. So,  $n$  and  $i$  are equal for 1-bit data. Only possible transition between these two states is 0 to 1, and the resulting probability from equation 5 is

$$P_{0 \rightarrow 1} = \frac{\phi(\mu + (1 - \frac{2^1}{2})l\sigma) - \phi(\mu + (\frac{2^1}{2} - 1)l\sigma - S)}{\phi(\mu + (1 - \frac{2^1}{2})l\sigma)} = 1 - \frac{\phi(\mu - S)}{\phi(\mu)} = 1 - \frac{\phi(\mu - S)}{0.5} \quad (8)$$

$$P_{1 \rightarrow 0} = 1 - P_{0 \rightarrow 1} = \frac{\phi(\mu - S)}{0.5} \quad (9)$$

**For a 2-bit case**,  $\mathcal{M}$  is 2 and thus there are 4 levels namely 0, 1, 2, and 3. In this case,  $n$  and  $i$  are related as  $n - i = \frac{2^2}{2} - 1 = 1$ .

Using equation (1),

$$P_{1 \rightarrow 1} = \frac{\phi(\min(\mu + (i + 1)l\sigma - S, \mu + il\sigma)) - \phi(\max(\mu + (i + 1 - 1)l\sigma - S, \mu + (i - 1)l\sigma))}{\phi(\mu + il\sigma) - \phi(\mu + (i - 1)l\sigma)} \quad (10)$$

$$= \frac{\phi(\mu - S) - \phi(\mu - l\sigma)}{\phi(\mu) - \phi(\mu - l\sigma)} \quad (11)$$

$$= \frac{\phi(\mu - S) - \phi(\mu - l\sigma)}{0.5 - \phi(\mu - l\sigma)} \quad (12)$$

$$P_{1 \rightarrow 2} = \frac{\phi(\min(\mu + (i + 1)l\sigma - S, \mu + il\sigma)) - \phi(\max(\mu + (i + 1 - 1)l\sigma - S, \mu + (i - 1)l\sigma))}{\phi(\mu + il\sigma) - \phi(\mu + (i - 1)l\sigma)} \quad (13)$$

$$= \frac{\phi(\min(\mu + l\sigma - S, \mu)) - \phi(\max(\mu - S, \mu - l\sigma))}{\phi(\mu) - \phi(\mu - l\sigma)} \quad (14)$$

$$= \frac{\phi(\min(\mu + l\sigma - S, \mu)) - \phi(\max(\mu - S, \mu - l\sigma))}{0.5 - \phi(\mu - l\sigma)} \quad (15)$$

and,

$$P_{2 \rightarrow 2} = \frac{\phi(\min(\mu + (i+1)l\sigma - S, \mu + il\sigma)) - \phi(\max(\mu + (i+1-1)l\sigma - S, \mu + (i-1)l\sigma))}{\phi(\mu + il\sigma) - \phi(\mu + (i-1)l\sigma)} \quad (16)$$

$$= \frac{\phi(\mu + l\sigma - S) - \phi(\mu)}{\phi(\mu + l\sigma) - \phi(\mu)} \quad (17)$$

$$= \frac{\phi(\mu + l\sigma - S) - 0.5}{\phi(\mu + l\sigma) - 0.5} \quad (18)$$

Using equation (2),

$$P_{0 \rightarrow 0} = \frac{\phi(\mu + (1 - \frac{2^M}{2})l\sigma - S)}{\phi(\mu + (1 - \frac{2^M}{2})l\sigma)} = \frac{\phi(\mu - l\sigma - S)}{\phi(\mu - l\sigma)} \quad (19)$$

From equation (3),

$$P_{0 \rightarrow m} = \frac{\phi(\min(\mu + (m-1)l\sigma - S, \mu - 1l\sigma)) - \phi(\mu + (m-2)l\sigma - S)}{\phi(\mu - 1l\sigma)} \quad (20)$$

$$P_{0 \rightarrow 1} = \frac{\phi(\min(\mu - S, \mu - l\sigma)) - \phi(\mu - l\sigma - S)}{\phi(\mu - l\sigma)} \quad (20)$$

$$P_{0 \rightarrow 2} = \frac{\phi(\min(\mu + l\sigma - S, \mu - l\sigma)) - \phi(\mu - S)}{\phi(\mu - l\sigma)} \quad (21)$$

from equation (4).

$$P_{n \rightarrow 3} = \frac{\phi(\mu + (n-1)l\sigma) - \phi(\max(\mu + l\sigma - S, \mu + (n-2)l\sigma))}{\phi(\mu + (n-1)l\sigma) - \phi(\mu + (n-2)l\sigma)}$$

$$P_{1 \rightarrow 3} = \frac{\phi(\mu) - \phi(\max(\mu + l\sigma - S, \mu - l\sigma))}{\phi(\mu) - \phi(\mu - l\sigma)} = \frac{0.5 - \phi(\max(\mu + l\sigma - S, \mu - l\sigma))}{0.5 - \phi(\mu - l\sigma)} \quad (22)$$

$$P_{2 \rightarrow 3} = \frac{\phi(\mu + l\sigma) - \phi(\max(\mu + l\sigma - S, \mu))}{\phi(\mu + l\sigma) - \phi(\mu)} = \frac{\phi(\mu + l\sigma) - \phi(\max(\mu + l\sigma - S, \mu))}{\phi(\mu + l\sigma) - 0.5} \quad (23)$$

from equation (5)

$$P_{0 \rightarrow 3} = \frac{\phi(\mu - l\sigma) - \phi(\mu + l\sigma - S)}{\phi(\mu - l\sigma)} \quad (24)$$

and lastly from equation (6)

$$P_{3 \rightarrow 3} = 1 \quad (25)$$

Equations (8) to (25) match with the probability formulas given in the paper [Luo et al. \[2022\]](#), Appendix B.

## References

- R. Luo, G. Hobbs, S. Y. Yong, A. Zic, L. Toomey, S. Dai, A. Dunning, D. Li, T. Marshman, C. Wang, P. Wang, S. Wang, and S. Zhang. Simulating high-time resolution radio-telescope observations. *Monthly Notices of the Royal Astronomical Society*, 513(4):5881–5891, 04 2022. ISSN 0035-8711. doi: 10.1093/mnras/stac1168. URL <https://doi.org/10.1093/mnras/stac1168>.