

Let us consider the calibrated noise at the 3 detectors  $h_1(t)$ ,  $h_2(t)$ ,  $e_1(t)$

Assumption 1: the 3 noise time series are (a) uncorrelated, (b) stationary stochastic processes characterized by their mean and autocorrelation functions:

$$\langle h_1(t) \rangle = \langle h_2(t) \rangle = \langle e_1(t) \rangle = 0 \quad (1)$$

$$\langle h_1(t) h_2(t) \rangle = \langle h_1(t) \rangle \langle h_2(t) \rangle = 0 \quad (2)$$

$$\langle h_1(t) e_1(t) \rangle = \langle h_1(t) \rangle \langle e_1(t) \rangle = 0 \quad (3)$$

$$\langle h_2(t) e_1(t) \rangle = \langle h_2(t) \rangle \langle e_1(t) \rangle = 0 \quad (4)$$

$$\langle h_1(t) h_1(t+\tau) \rangle = R_1(\tau) \quad (5)$$

$$\langle h_2(t) h_2(t+\tau) \rangle = R_2(\tau) \quad (6)$$

$$\langle e_1(t) e_1(t+\tau) \rangle = R_3(\tau) \quad (7)$$

the analysis here can be generalized to include  $\neq 0$  results in (2).

$$R_1(\tau) = \int_{-\infty}^{+\infty} e^{-i2\pi f t} S_{h_1}(f) df \quad (8)$$

$$R_2(\tau) = \int_{-\infty}^{+\infty} e^{-i2\pi f t} S_{h_2}(f) df \quad (9)$$

$$R_3(\tau) = \int_{-\infty}^{+\infty} e^{-i2\pi f t} S_{e_1}(f) df \quad (10)$$

where  $S_{h_1}(f)$ ,  $S_{h_2}(f)$ ,  $S_{e_1}(f)$  are the available noise power spectral densities.

(2)

Corr Power computes the  $r$ -statistics over 3 different coincidence windows  $T_1 T_2 T_3$  and for each of them evaluates:

$$r_T = \frac{\int_{-\frac{T}{2}}^{\frac{T}{2}} X_\alpha(t) X_\beta(t+\tau) dt}{\|X_\alpha(t)\| \|X_\beta(t+\tau)\|} \quad (11)$$

where  $\alpha, \beta = 1, 2, \dots, 3$  and (12)

$$N_\alpha = \|X_\alpha(t)\| = \sqrt{\int_{-\frac{T}{2}}^{+\frac{T}{2}} X_\alpha(t)^2 dt} \quad (13)$$

I assume that the energy in the noise does not depend on the exact location of the temporal window and that is simply proportional to the duration of the window.

$$N_\alpha = T \cdot m_\alpha \quad N_\beta = T \cdot m_\beta \quad (14)$$

I also assume that  $N_\alpha$  and  $N_\beta$  are not random variables.

A first ingredient to evaluate the probability distribution for the  $r$  evaluated by corr power is to evaluate the pdf of

$$\bar{r}_T = \int_{-\frac{T}{2}}^{\frac{T}{2}} X_\alpha(t) X_\beta(t+\tau) dt = \quad (15)$$

$$= \sum_{i=1}^{N_T} X_\alpha(t_i) X_\beta(t_i+\tau) \Delta t \quad (16)$$

ASSUMPTION:  $N$  is large enough that the central limit theorem applies to (16).

(3)

$\bar{r}$  becomes Gaussian with mean & variances given by:

$$\langle \bar{r}_T \rangle = \sum_{i=1}^N \langle X_{\alpha}(t_i) \rangle \langle X_{\beta}(t_i + \tau) \rangle \Delta t = 0 \quad (17)$$

$$\begin{aligned} \sigma_{\bar{r}_T}^2 \langle \bar{r}_T^2 \rangle &= \sum_{i=1}^N \sum_{j=1}^N \langle X_{\alpha}(t_i) X_{\beta}(t_i + \tau) X_{\alpha}(t_j) X_{\beta}(t_j + \tau) \rangle = \\ &= \sum_{i=1}^N \sum_{j=1}^N \langle X_{\alpha}(t_i) X_{\alpha}(t_j) \rangle \langle X_{\beta}(t_i + \tau) X_{\beta}(t_j + \tau) \rangle = \\ &= \sum_{i=1}^N \sum_{j=1}^N R_{\alpha}(t_i - t_j) R_{\beta}(t_i - t_j) \quad (18) \end{aligned}$$

the probability of  $r_{T1}$  becomes:

$$P(r_{T1}) = \frac{1}{\sqrt{2\pi \sigma_{T1}^2}} \exp \left\{ -\frac{r_{T1}^2}{2 \sigma_{T1}^2} \right\} \quad (19)$$

the probability for the corr Power  $\Pi$

$$P(\Pi) = P(\Pi = \max \{ r_{T1}, r_{T2}, r_{T3} \}) = \quad (20)$$

$$P(r_{T1}, r_{T2}, r_{T3}) = \frac{1}{\sqrt{2\pi \det C}} \exp \left\{ -\vec{r} C^{-1} \vec{r} \right\} \quad (21)$$

$$\text{with } \vec{r} = (r_{T1}, r_{T2}, r_{T3}) \quad (22)$$

$$\text{and } C = \begin{pmatrix} \sigma_{T1}^2 & \sigma_{T1}^2 & \sigma_{T1}^2 \\ \sigma_{T1}^2 & \sigma_{T1+\Delta_1}^2 & \sigma_{T1+\Delta_1}^2 \\ \sigma_{T1}^2 & \sigma_{T1+\Delta_1}^2 & \sigma_{T1+\Delta_2}^2 \end{pmatrix} \quad (23)$$

$$P(I=k_M) = \frac{1}{3} \left( P(r_{T_1} = \Pi, r_{T_2} < \Pi, r_{T_3} < \Pi) + \right. \\
+ P(r_{T_1} < \Pi, r_{T_2} = \Pi, r_{T_3} < \Pi) + \\
\left. + P(r_{T_1} < \Pi, r_{T_2} < \Pi, r_{T_3} = \Pi) \right) \quad (24)$$

In deriving (23) it is assumed that (a) the most correlated  $T_1$  window is included in the most correlated  $T_2$  window and (b) the most correlated  $T_2$  window is included in the most correlated  $T_3$  window and (c) that

$$\langle r_{T_1} r_{T_2} \rangle = \sigma_{T_1}^2 \quad (25)$$

$$\langle r_{T_2} r_{T_3} \rangle = \sigma_{T_1}^2 + \Delta_1^2 \quad (26)$$

$$\langle r_{T_1} r_{T_3} \rangle = \sigma_{T_1}^2 \quad (27)$$

$$\langle r_{T_2} r_{T_2} \rangle = \sigma_{T_2}^2 + \Delta_1^2 \quad (28)$$

$$\langle r_{T_3} r_{T_3} \rangle = \sigma_{T_1}^2 + \Delta_2^2 \quad (29)$$

the assumptions above require a correlation time of few periods.

Eq 24 requires integrating (22) for 2 of the 3 random variables.

(5)

$$\Delta_1^2 = \langle K_{T_\alpha}^2 \rangle \quad \text{where} \quad T_\alpha = T_2 - T_1$$

$$\Delta_2^2 = \langle K_{T_\beta}^2 \rangle \quad \text{where} \quad T_\beta = T_3 - T_1$$

$p(T = k_M, \delta_{\text{NONOISE}})$  follow the same path  
 $\vec{j} = (\delta_1, \delta_2, \delta_3)$  from

$$P[\vec{r}, \delta] = \frac{1}{\sqrt{2\pi \det C}} \exp \left\{ -(\vec{r} - \vec{j}) C^{-1} (\vec{r} - \vec{j}) \right\}$$