

# Statistical Analysis of Phase Locking Value

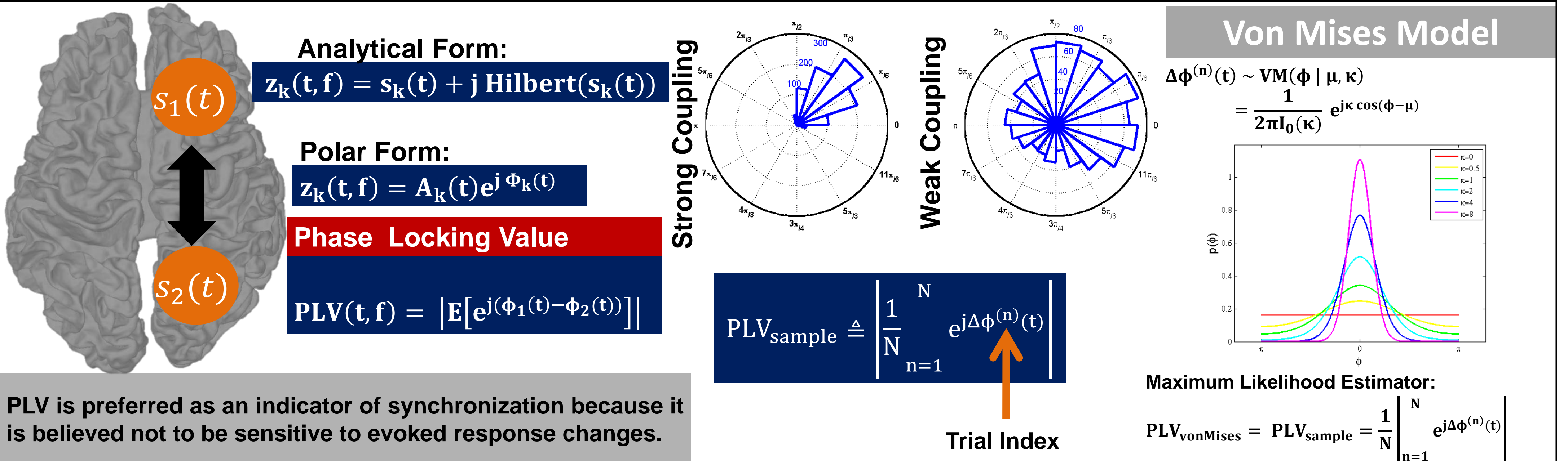
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## Synchronization in Brain Signals



## Circularly Symmetric Complex Gaussian Model

If  $s_1(t)$  and  $s_2(t) \sim N(0, 1)$ , the complex random vector  $z(t) = [z_1(t), z_2(t)]^T$  follows a circular complex Gaussian distribution:

$$p(z(t)) = \frac{1}{\pi^2 |K_z|} \exp\{-z(t) K_z^{-1} z(t)\}$$

where  $K_z^{-1} = \begin{bmatrix} \kappa_{11} & \kappa_{12} e^{j\mu_{12}} \\ \kappa_{21} e^{j\mu_{21}} & \kappa_{22} \end{bmatrix}$  and  $\kappa_{12} = \kappa_{21}, \mu_{12} = -\mu_{21}$

• **Conditional distribution of phase difference**

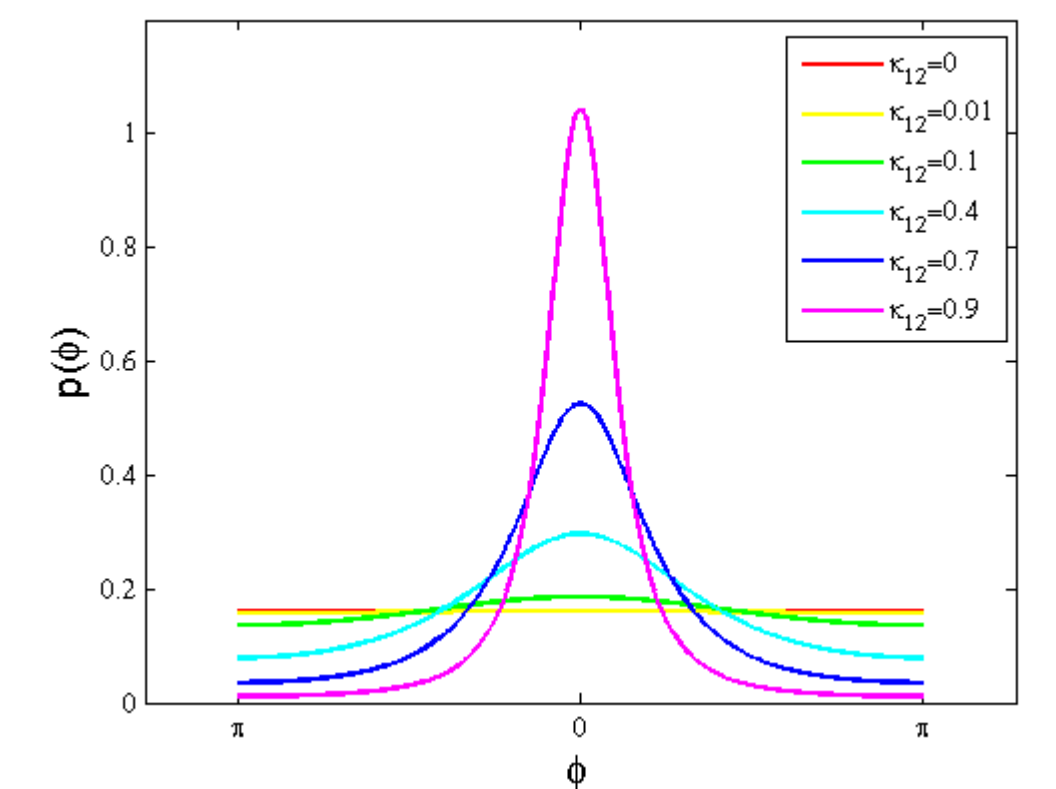
$$p(\Delta\Phi(t) | A(t)) = \frac{1}{2\pi I_0(-2\kappa_{12} A_1(t) A_2(t))} e^{-2\kappa_{12} A_1(t) A_2(t) \cos(\Delta\Phi(t) - \mu_{12})}$$

➤ Note: von Mises but amplitude and phase are NOT independent.

• **Marginal distribution of phase:**

$$p(\Delta\Phi(t)) = \frac{1}{2\kappa^2 \pi |K_z|} \left( \frac{1}{1-a^2} - \frac{a \cos^{-1}(a)}{\sqrt{(1-a^2)^3}} \right)$$

where  $a \triangleq \frac{\kappa_{12}}{\kappa_{11}} \cos(\Delta\Phi(t) - \mu_{12})$



•  $PLV_{\text{circgauss}} \triangleq |E[e^{j\Delta\Phi(t)}]| = \left| \frac{\pi}{\sqrt{2}} \left( \frac{3-2\omega^2}{2\omega+2} \right) \left[ \omega^3 {}_2F_1\left(\frac{3}{4}, \frac{5}{4}, 1, \omega^2\right) + \frac{3}{4} \omega^5 {}_2F_1\left(\frac{5}{4}, \frac{7}{4}, 2, \omega^2\right) \right] \right|$  where  $\omega = \frac{\kappa_{12}}{2\kappa_{11}\kappa_{22} - \kappa_{12}}$  which is a function of square coherence, namely  $\frac{\kappa_{12}}{\kappa_{11}\kappa_{22}}$ .

## Simulations: ROC Analysis of Roessler Oscillators

2 Roessler oscillators  $\xi_j$  where  $j \in \{1, 2\}$

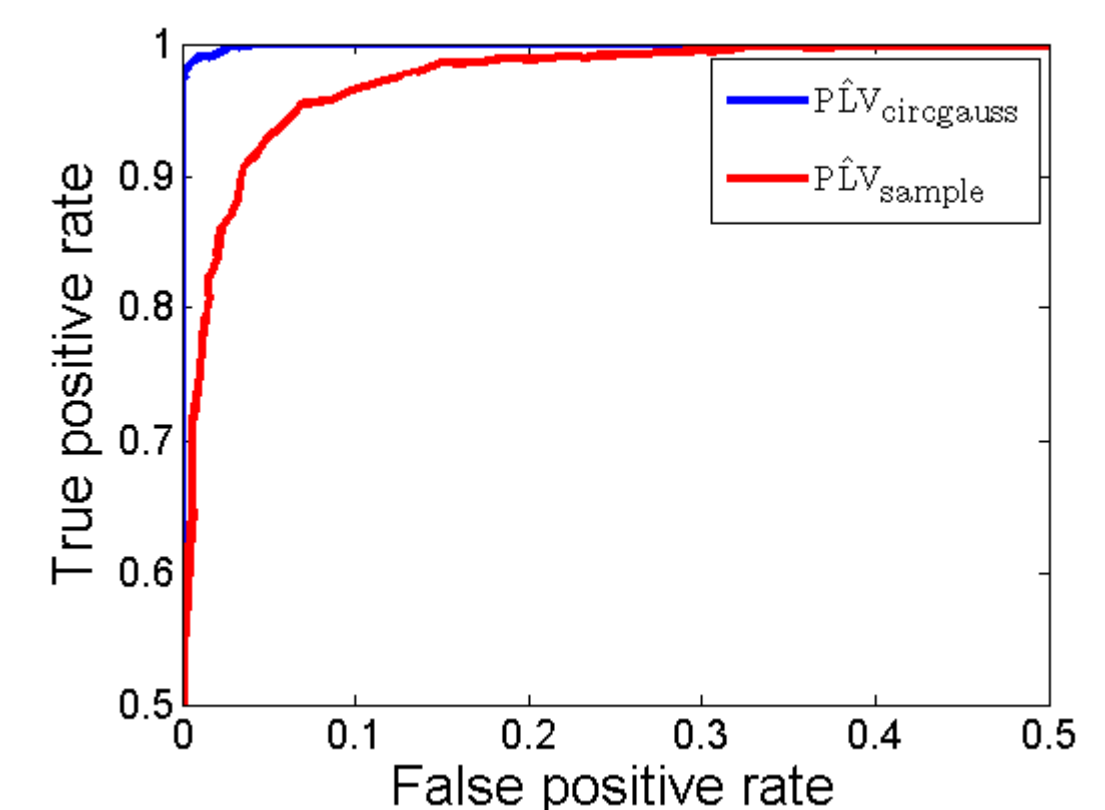
$$\xi_j = \begin{pmatrix} X_j \\ Y_j \\ Z_j \end{pmatrix} = \begin{pmatrix} -\omega_j Y_j - Z_j + \left[ \begin{matrix} \epsilon(X_i - X_j) \\ i \neq j \end{matrix} \right] + \sigma_j \eta_j \\ \omega_j X_j + a Y_j \\ b + (X_j - c) Z_j \end{pmatrix}$$

Controls the amount of coupling



**True Positive** : if  $PLV > \tau$  and  $\epsilon > 0$

**False Positive** : if  $PLV > \tau$  and  $\epsilon = 0$

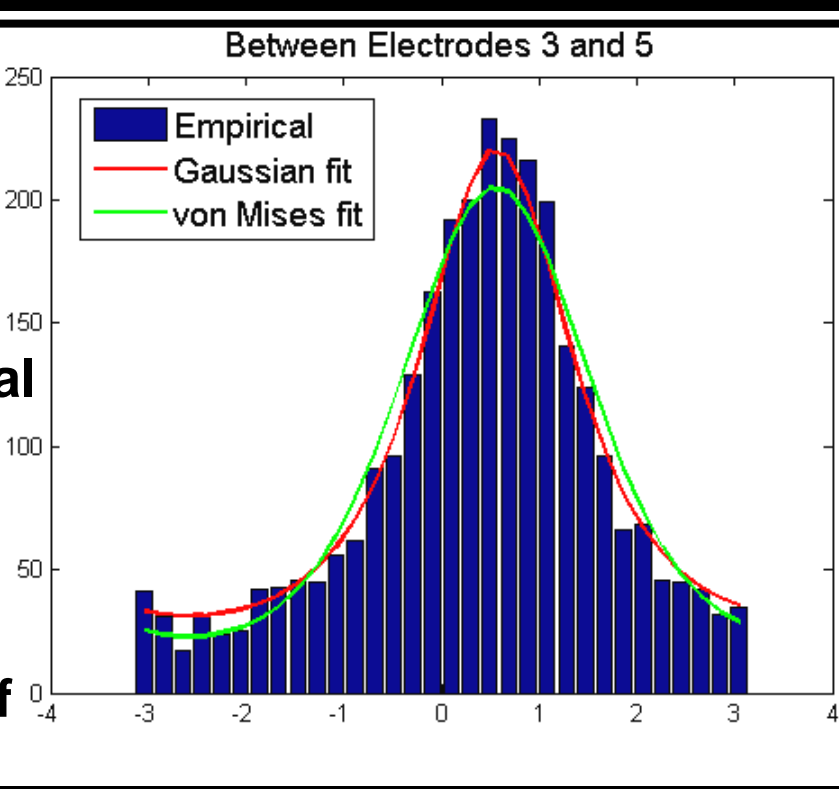


## Analysis of LFP Data

• Local field potential (LFP) time series, sampled at 200 Hz, from a macaque monkey implanted with transcortical bipolar electrodes at 15 sites in the right hemisphere.

• Monkey performed a GO, NO-GO visual pattern discrimination task.

• We used 10,178 trials taken from 18 sessions focusing on  $120 \pm 25$  msec and  $260 \pm 25$  msec after stimulus presentation in the frequency range of beta [13-30] Hz.



## Conclusion

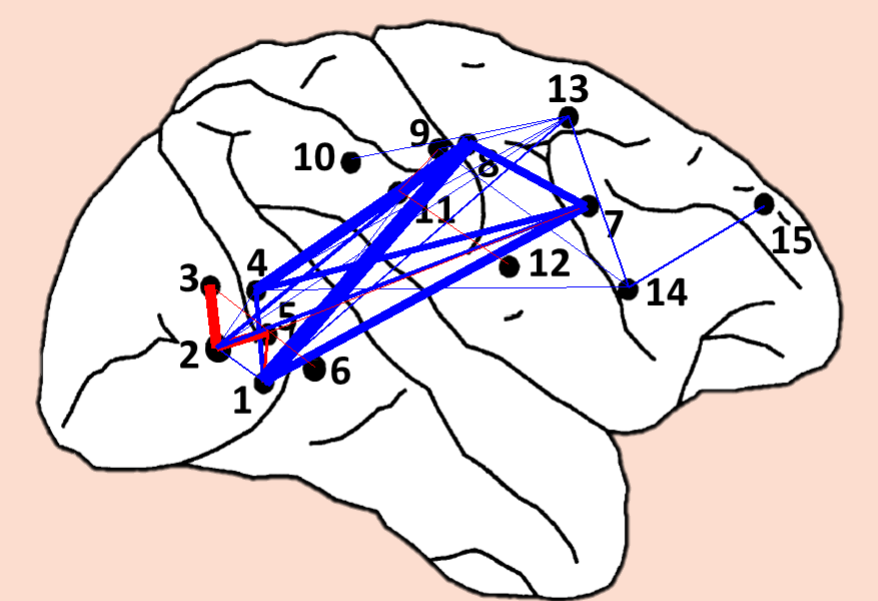
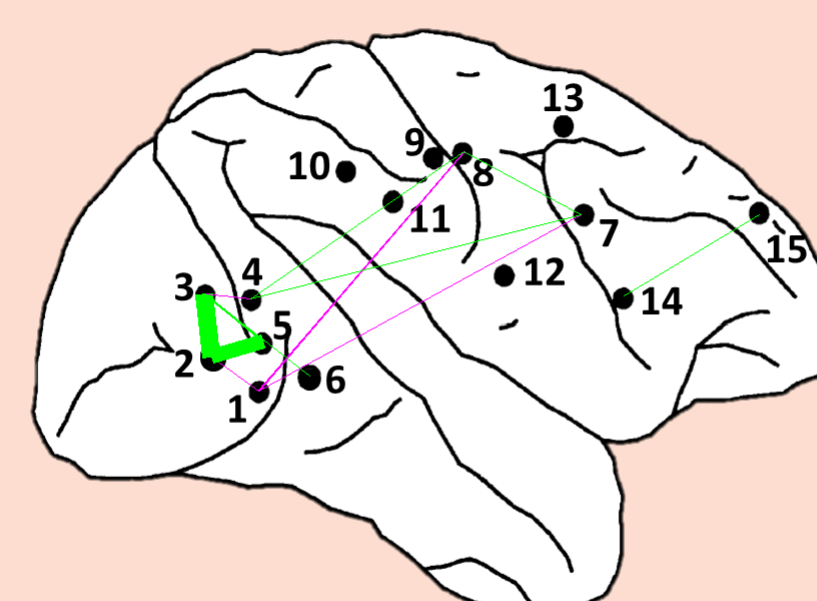
- For Gaussian signals, phase and amplitude are NOT statistically independent. Phase synchronization is a function of coherence.
- LFP networks verify strong relationship between coherence and PLV.
- For approximate Gaussian data, this result clarifies the relationship between coherence and PLV.
- We observe very little difference between Gaussian and Von Mises based methods. Gaussian based method has some advantages as seen from ROC curves.

PLV Type

**Diamond vs Line**  
(Early Response at  $120 \pm 25$  msec)

**GO vs NO-GO**  
(Late Response at  $260 \pm 25$  msec)

PLV<sub>sample</sub>



PLV<sub>circgauss</sub>

