Linear Stability Analysis of the Kuramoto model with noise

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July 2020

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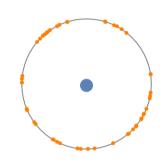
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Kuramoto model

$$\theta'_i(t) = \omega_i + Kr\sin(\psi - \theta_i)$$

Adding noise to the model gives

$$\theta'_{i}(t) = \omega_{i} + \zeta_{i}(t) + Kr\sin(\psi - \theta_{i})$$

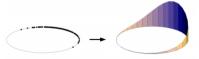


Where ζ is the white noise defined as

$$<\zeta_{i}(t)>=0$$

 $<\zeta_{i}(t)\zeta_{j}(t)>=2D\delta_{ij}\delta(s-t)$

$$\frac{\partial \rho}{\partial t} = D \frac{\partial^2 \rho}{\partial \theta^2} - \frac{\partial}{\partial \theta} [\rho v]$$



We add a small perturbation to he incoherent state

$$\rho(\theta,t,\omega) \; = \; \epsilon \, \eta(\theta,t,\omega) + \frac{1}{2 \, \pi} \label{eq:rho}$$

Where the strength is set $\epsilon << 1$, the arbitrary function must obey for normalization of ρ

$$\int_0^{2\pi}\! \eta(\theta,t,\omega)\,d\theta\,=\,0$$

$$\epsilon \, \frac{\partial \, \eta(\theta,\,t,\,\omega)}{\partial \, t} = \epsilon \, D \, \frac{\partial^2 \, \eta(\theta,\,t,\,\omega)}{\partial \, \theta^2} \, - \, \frac{\partial}{\partial \, \theta} \, \Big[\left(\, \frac{1}{2 \, \pi} \, + \, \epsilon \, \eta \, \right) v \, \Big]$$

$$r e^{i\psi} = \epsilon \int_0^{2\pi} \int_{-\infty}^{\infty} e^{i\theta} \eta(\theta, t, \omega) g(\omega) d\omega d\theta$$
 $r = \epsilon r_1$

$$v = \omega_i + \epsilon K r_1 \sin(\psi - \theta_i)$$



Evolution of the perturbation η

$$\frac{\partial \eta}{\partial t} = D \frac{\partial^2 \eta}{\partial \theta^2} - \omega \frac{\partial \eta}{\partial \theta} + \frac{K}{2\pi} r_1 \cos(\psi - \theta_i)$$

We use Fourier method to analyze the solutions, solutions of the form

$$\eta(\theta,t,\omega) \; = \; c(t,\omega) \, e^{i\,\theta} + c^*(t,\omega) \, e^{-i\,\theta} + \; \eta^\perp(\theta,t,\omega)$$

$$r_1 \cos(\psi - \theta) = \pi \left[\left(\int_{-\infty}^{\infty} c^*(t, \omega) g(\omega) d\omega \right) e^{-i\theta} + \left(\int_{-\infty}^{\infty} c(t, \omega) g(\omega) d\omega \right) e^{i\theta} \right]$$

$$\begin{split} e^{i\theta} \left(\frac{\partial c}{\partial t} + Dc + i\omega c - \frac{K}{2} \int_{-\infty}^{\infty} c \, g(v) \, dv \right) + \, e^{-i\theta} \left(\frac{\partial \, c^*}{\partial \, t} + D\, c^* - i\omega \, c^* - \frac{K}{2} \int_{-\infty}^{\infty} c^* \, g(v) \, dv \right) + \\ \left(\frac{\partial \, \eta^{\perp}}{\partial \, t} - \, D \, \frac{\partial^2 \, \eta^{\perp}}{\partial \, \theta^2} + \omega \, \frac{\partial \, \eta^{\perp}}{\partial \, \theta} \right) = \, 0 \end{split}$$

$$\frac{\partial c}{\partial t} = -(D + i\omega) + \frac{K}{2} \int_{-\infty}^{\infty} c g(v) dv$$
$$\frac{\partial \eta^{\perp}}{\partial t} = D \frac{\partial^2 \eta^{\perp}}{\partial \theta^2} - \omega \frac{\partial \eta^{\perp}}{\partial \theta}$$

To study the evolution of the fundamental node

$$\begin{split} c(t,\,\omega) &= b\left(\omega\right)e^{\lambda\,t} &\qquad (L-\lambda\,\mathrm{I})\,c = (L-\lambda\,\mathrm{I})\,b\,\,e^{\lambda\,t} = 0 \\ L\,c &= -(D+i\,\omega)\,c + \frac{K}{2}\,\int_{-\infty}^{\infty}\!c\,g(v)\,d\,v \end{split}$$

$$\lambda b = L b = -(D + i \omega) b + \frac{K}{2} \int_{-\infty}^{\infty} b(v) g(v) dv$$

$$1 = \frac{K}{2} \int_{-\infty}^{\infty} \frac{\lambda + D}{(\lambda + D)^2 + v^2} g(v) dv \qquad \lambda > -D$$



• Critical point $\lambda = 0$

$$K_c = 2 \left[\int_{-\infty}^{\infty} \frac{D}{D^2 + v^2} g(v) dv \right]^{-1}$$

For a Lorentzian distribution,

$$g(\omega) = \frac{\gamma}{\pi(\gamma^2 + \omega^2)}$$

$$\lambda = \frac{K}{2} - D - \gamma$$

$$K_c = 2(D + \gamma) = \beta^2 + 2\gamma$$

• Adding the second term of perturbation $[O(\epsilon^2)]$

$$\rho(\theta,\,t,\,\omega) \; = \frac{1}{2\,\pi} \; + \; \epsilon\,\eta(\theta,\,t,\,\omega) \;\; + \; \epsilon^2\,\mu(\theta,\,t,\,\omega) \label{eq:rho}$$

$$\epsilon \frac{\partial \eta}{\partial t} + \epsilon^2 \frac{\partial \mu}{\partial t} = \epsilon D \frac{\partial^2 \eta}{\partial \theta^2} + \epsilon^2 D \frac{\partial^2 \mu}{\partial \theta^2} - \frac{\partial}{\partial \theta} \left[\left(\frac{1}{2\pi} + \epsilon \eta + \epsilon^2 \mu \right) v \right]$$

$$r = \epsilon r_1 + \epsilon^2 r_2$$
 $v = \omega_i + K(\epsilon r_1 + \epsilon^2 r_2) \sin(\psi - \theta_i)$

$$\frac{\partial \mu}{\partial t} = D \frac{\partial^2 \mu}{\partial \theta^2} + K \left[\frac{r_2}{2\pi} + \eta r_1 \right] \cos(\psi - \theta_i) - K r_1 \sin(\psi - \theta) \frac{\partial \eta}{\partial \theta} - \omega \frac{\partial \mu}{\partial \theta}$$



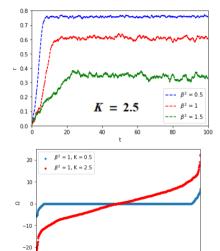
$$\begin{split} e^{i\theta} \left[\frac{\partial s}{\partial t} + D s + i \omega s - \frac{K}{2} \int_{-\infty}^{\infty} s \, g(v) \, dv - \pi \, K \left(\eta^{\perp} - i \, \frac{\partial \eta^{\perp}}{\partial t} \right) \int_{-\infty}^{\infty} c \, g(v) \, dv \right] + \\ e^{-i\theta} \left[\frac{\partial s^{*}}{\partial t} + D \, s^{*} - i \, \omega \, s^{*} - \frac{K}{2} \int_{-\infty}^{\infty} s^{*} \, g(v) \, dv - \pi \, K \left(\eta^{\perp} - i \, \frac{\partial \eta^{\perp}}{\partial t} \right) \int_{-\infty}^{\infty} c^{*} \, g(v) \, dv \right] + \\ \left(\frac{\partial \mu^{\perp}}{\partial t} - D \, \frac{\partial^{2} \mu^{\perp}}{\partial \theta^{2}} + \omega \, \frac{\partial \mu^{\perp}}{\partial \theta} \right) - 2 \, \pi \, K \, e^{2i\theta} \left[c \int_{-\infty}^{\infty} c \, g(v) \, dv \right] - 2 \, \pi \, K \, e^{-2i\theta} \left[c \int_{-\infty}^{\infty} c^{*} \, g(v) \, dv \right] = 0 \end{split}$$

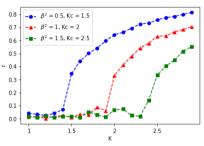
$$\frac{\partial s}{\partial t} = -(D + i\omega) + \frac{K}{2} \int_{-\infty}^{\infty} s \, g(v) \, dv$$
$$\frac{\partial \mu^{\perp}}{\partial t} = D \frac{\partial^{2} \mu^{\perp}}{\partial \theta^{2}} - \omega \frac{\partial \mu^{\perp}}{\partial \theta}$$

Test results

•
$$K_c = 2(D + \gamma) = \beta^2 + 2\gamma$$
 $\gamma = 0.5$ $N = 5000$

$$y = 0.5 \qquad N = 5000$$





Conclusions

- Powerful way to study synchronization mathematically
- Random noise can be easily dealt with in the Kuramoto model
- Numerical results show that the simulations agree with the model's predictions