

Synchronization in populations of phase oscillators

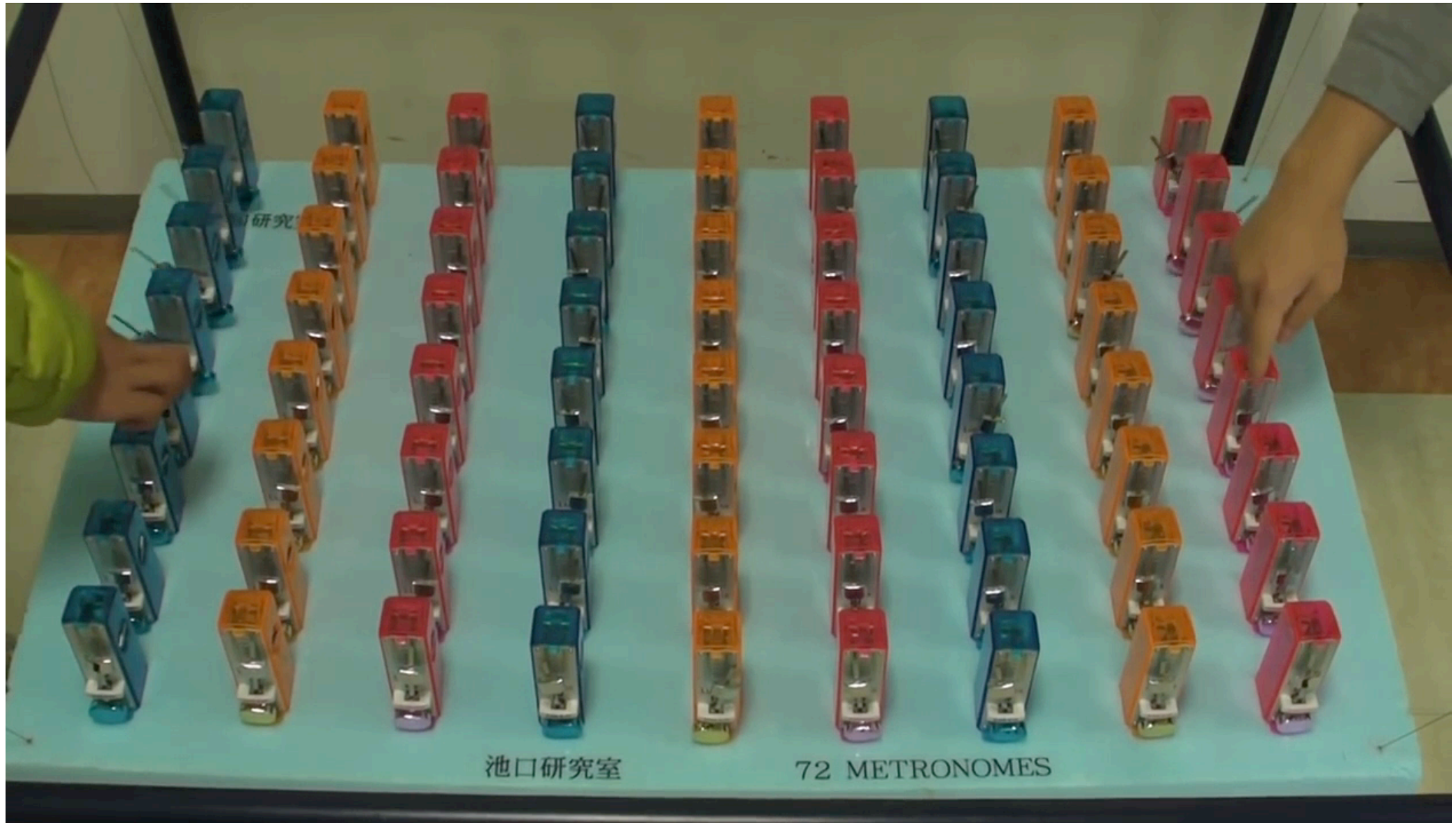
Kuramoto model
(Yoshiki Kuramoto, 1984)

Teaching example (BIO-341). The Kuramoto model

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CHAPITRE 6. OSCILLATEURS DE PHASE ET ENTRAÎNEMENT

2 Modèle de Kuramoto

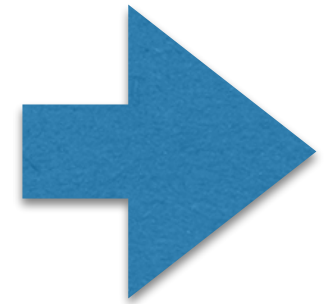
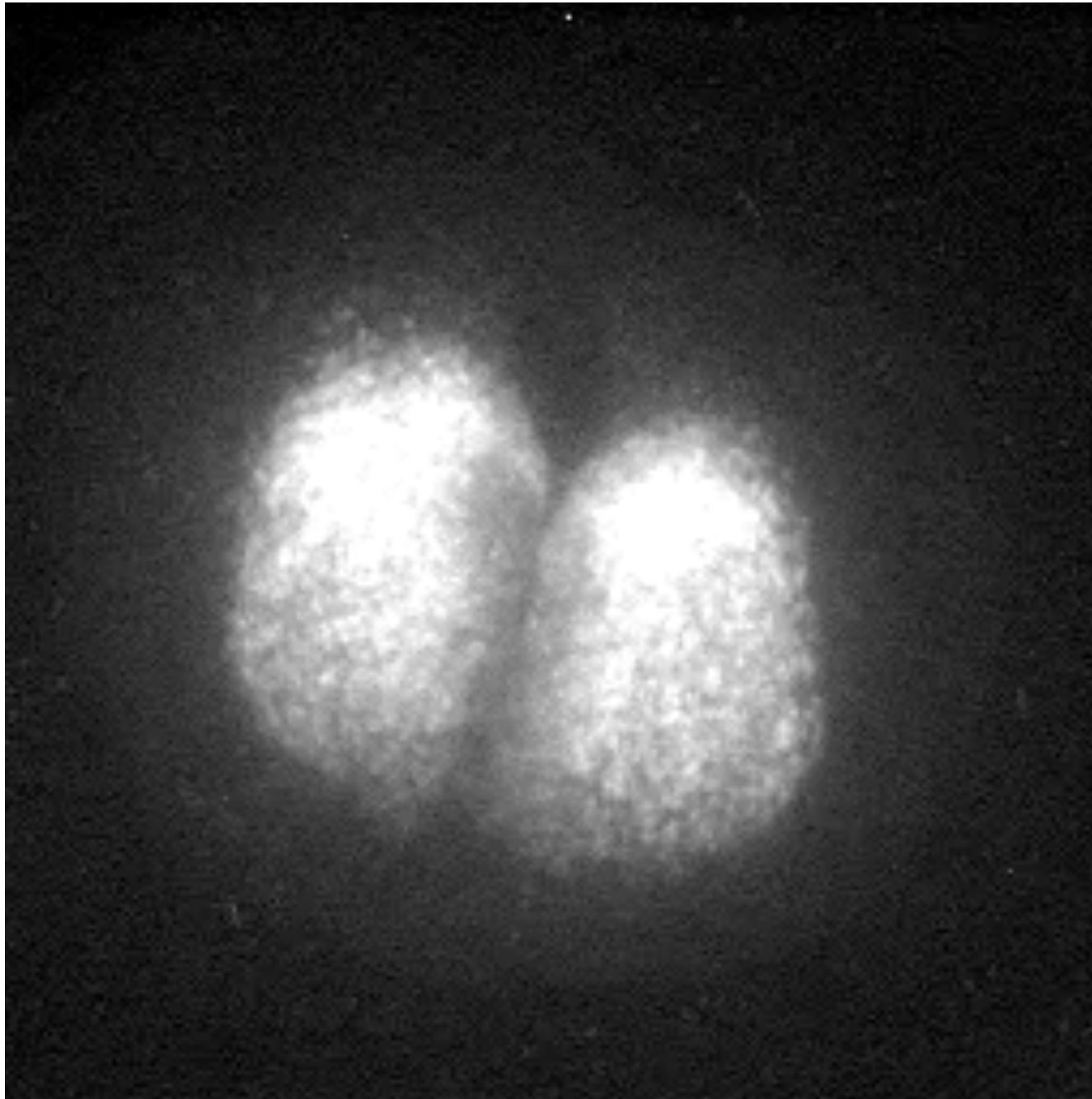


source: <https://www.youtube.com/watch?v=suxu1bmPm2g>

Fireflies resonate/synchronize too



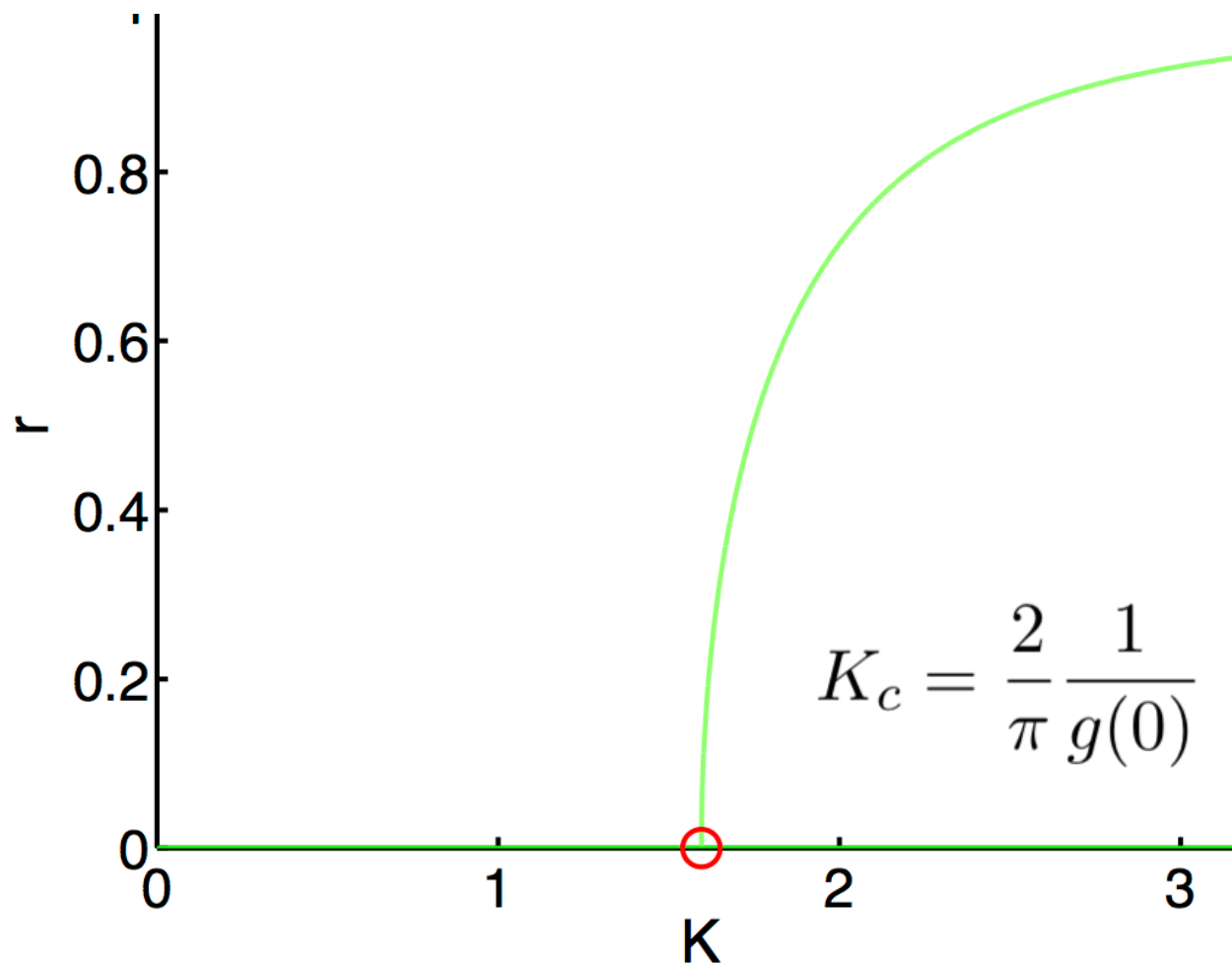
The mammalian master clock in the brain (SCN slice in mouse)



courtesy of Mick Hastings

The famous Kuramoto model

$$\dot{\theta}_i = \omega_i + \frac{K}{N} \sum_{j=1}^N \sin(\theta_j - \theta_i), \quad i \in \{1, \dots, N\}$$



The famous Kuramoto model (summary of key steps)

All-to-all phase coupling model

$$\dot{\theta}_i = \omega_i + \frac{K}{N} \sum_{j=1}^N \sin(\theta_j - \theta_i), \quad i \in \{1, \dots, N\}$$

Distribution of frequencies

$\omega_i \sim g(\omega)$, $g(-\omega) = g(\omega)$
probability distribution
for example $g(\omega) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{\omega^2}{2\sigma^2}}$

Rewrite in terms of the 'mean amplitude' r .

$$r = \frac{1}{N} \sum_i e^{i\theta_i} \rightarrow \int e^{i\theta(\omega)} g(\omega) d\omega$$

$$\dot{\theta}_i = \omega_i - rK \sin(\theta_i)$$

What is the value of r in function of K and ? Let's just calculate it

$$r = \underbrace{\langle e^{i\theta} \rangle_s}_{\text{interesting part}} + \underbrace{\langle e^{i\theta} \rangle_d}_{= 0 \text{ due to symmetry (cf Notes)}} \quad \text{e: entrained, d: drifting}$$

That's the result

$$r = \langle \cos(\theta) \rangle_s = rK \int_{-\pi/2}^{\pi/2} \cos^2(\theta) g(Kr \sin(\theta)) d\theta$$

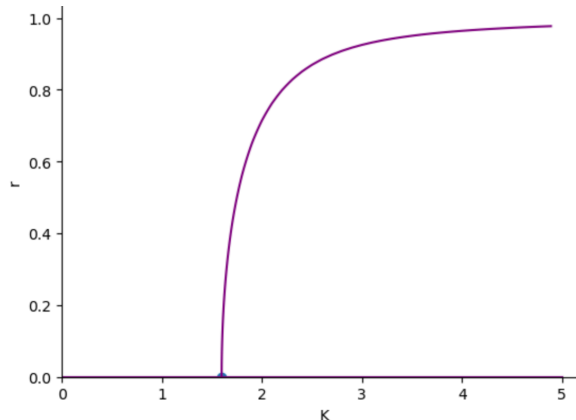
- Implicit relation between r , K , and g
- $r=0$ is always a solution (branch 1)
- The equation has a 2nd branch $r(K)$. What can we say about this second branch?

- For $r=0$, $K = K_c$ $K_c = \frac{2}{\pi} \frac{1}{g(0)}$

Gaussian $K_c = \frac{2\sqrt{2}}{\sqrt{\pi}}, \sigma$

Let's check numerically for the Gaussian $g()$

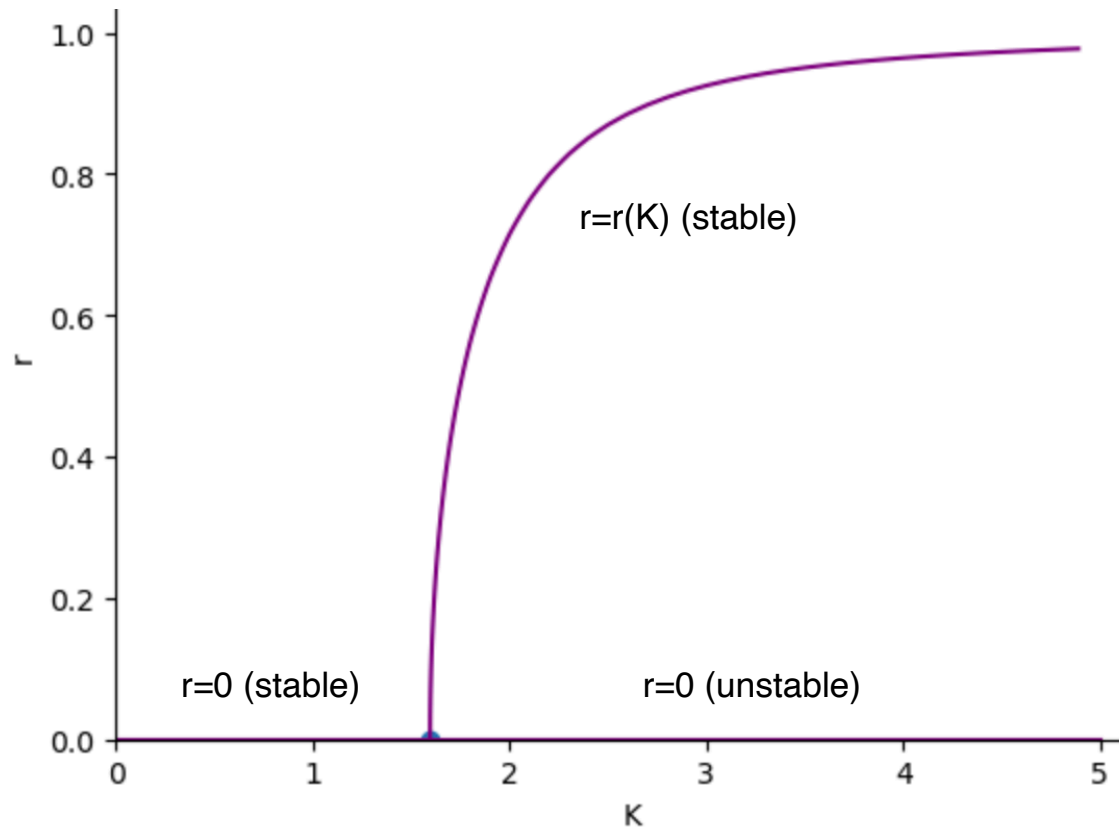
```
1 def Gaussian(x, mu=0, sig=1):
2     return np.exp(-(x-mu)*(x-mu) / (2*sig*sig)) / (np.sqrt(2*np.pi)*sig)
3
4 def integrand(x,K,r,sig):
5     return np.cos(x)**2 * Gaussian(K*r*np.sin(x),0,sig)
6
7 def integral(K,r,sig):
8     return quad(integrand, -np.pi/2, np.pi/2, args=(K,r,sig))
9
10 def rel(K,r, sig):
11     return K*integral(K,r,sig)[0]-1
```



```
1 sigma_w = 1
2 Kc = 2 / np.pi * np.sqrt(2*np.pi) * sigma_w
3
4 grid_r=np.arange(0,1.1,0.01)
5 grid_K=np.arange(0,5,0.1)
6 KK,rr=np.meshgrid(grid_K,grid_r)
7 V=np.array([0.])
8 Z=np.zeros_like(KK)
9 for i in range(KK.shape[0]):
10     for j in range(KK.shape[1]):
11         Z[i,j]=rel(KK[i,j],rr[i,j], sigma_w)
12 plt.contour(KK,rr,Z,V, colors='purple')
13 plt.plot([0,5],[0,0], color='purple')
14 plt.scatter(Kc,0)
15 plt.xlabel('K')
16 plt.ylabel('r')
17 plt.show()
```

Try with other distributions (see Notebook on Moodle)!

The two branches



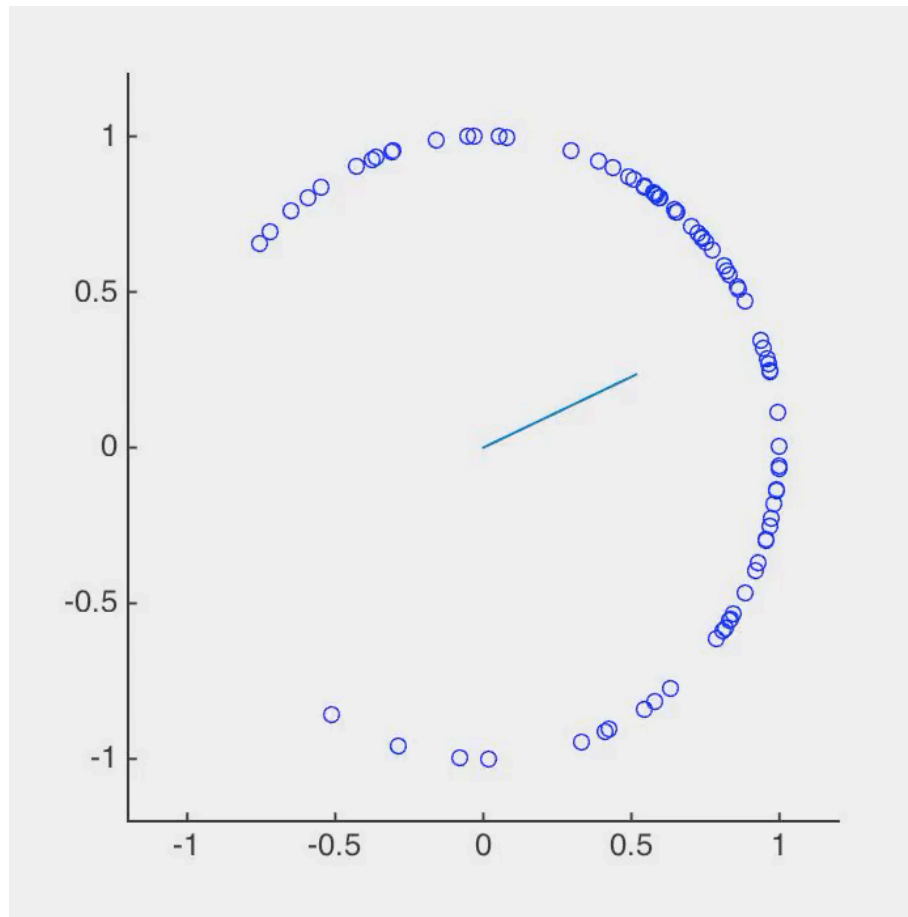
Think about what happens if you increase K from 0?

How about stability of the two solutions?

- Stability calculations are much harder since we have infinite-dimensional SD (more advanced course)
- The easiest is to proceed numerically (simulations)

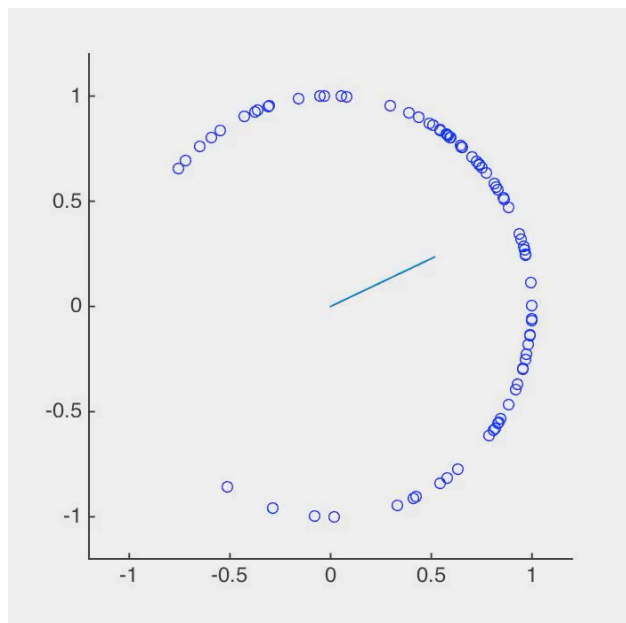
Simulating the Kuramoto model

$$\dot{\theta}_i = \omega_i + \frac{K}{N} \sum_{j=1}^N \sin(\theta_j - \theta_i), \quad i \in \{1, \dots, N\}$$

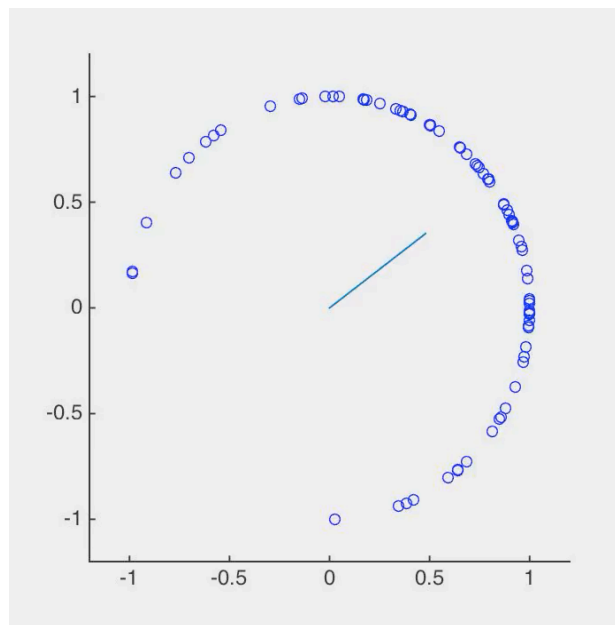


$$K=0.2 \cdot K_c$$

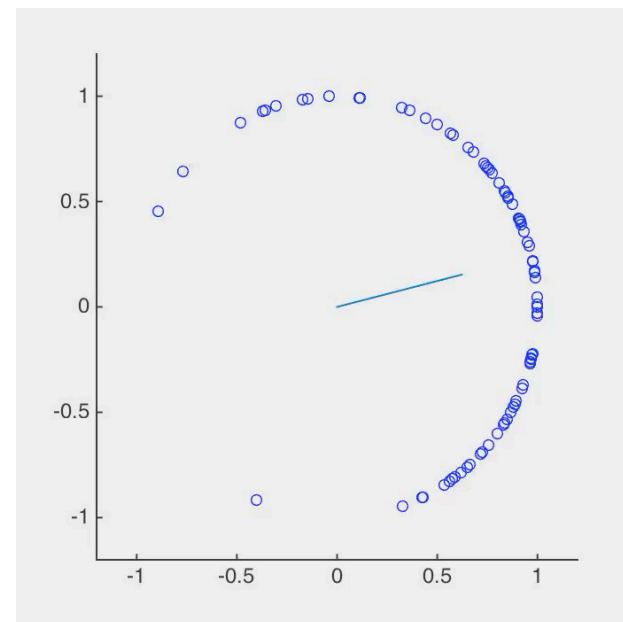
Kuramoto model



$$K=0.2 \cdot K_c$$



$$K=2 \cdot K_c$$



$$K=3 \cdot K_c$$

Kuramoto-like models

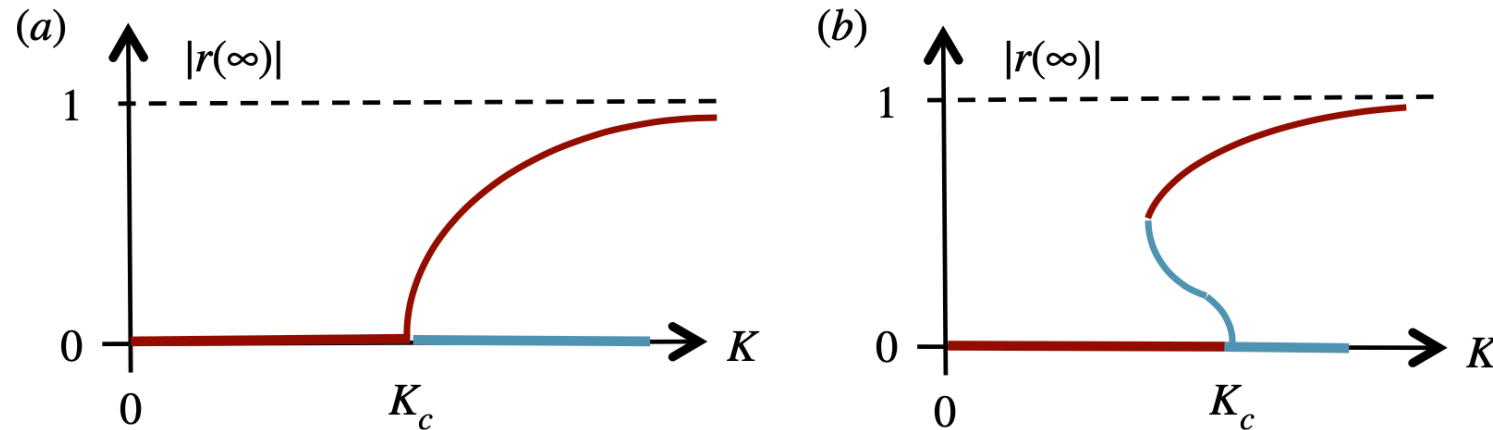


Figure 1. Schematic bifurcation diagram (a) for a symmetric and unimodal frequency distribution g and (b) for the bi-Cauchy distribution $g_{\Delta,\Omega}$ when bimodal (see §5). Red (resp. blue) lines indicate stable (resp. unstable) stationary solutions, see text for details.

Unimodal $g(\omega)$

$$g_{\Delta,\Omega}(\omega) = \frac{\Delta}{2\pi} \left(\frac{1}{(\omega - \Omega)^2 + \Delta^2} + \frac{1}{(\omega + \Omega)^2 + \Delta^2} \right)$$

Experiment with the notebook to obtain those diagrams