

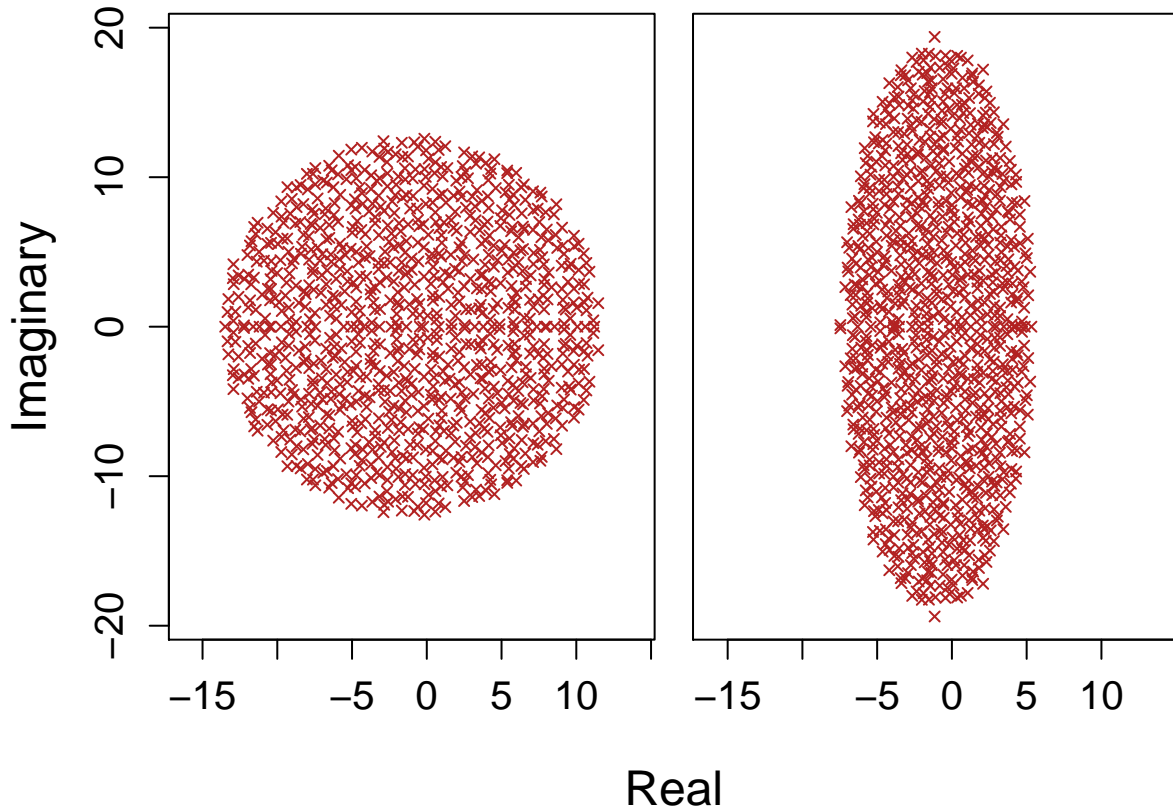
# Component response rate variation underlies the stability of complex systems (revision notes)

A. Bradley Duthie ( [alexander.duthie@stir.ac.uk](mailto:alexander.duthie@stir.ac.uk) )

Biological and Environmental Sciences, University of Stirling, Stirling, UK, FK9 4LA

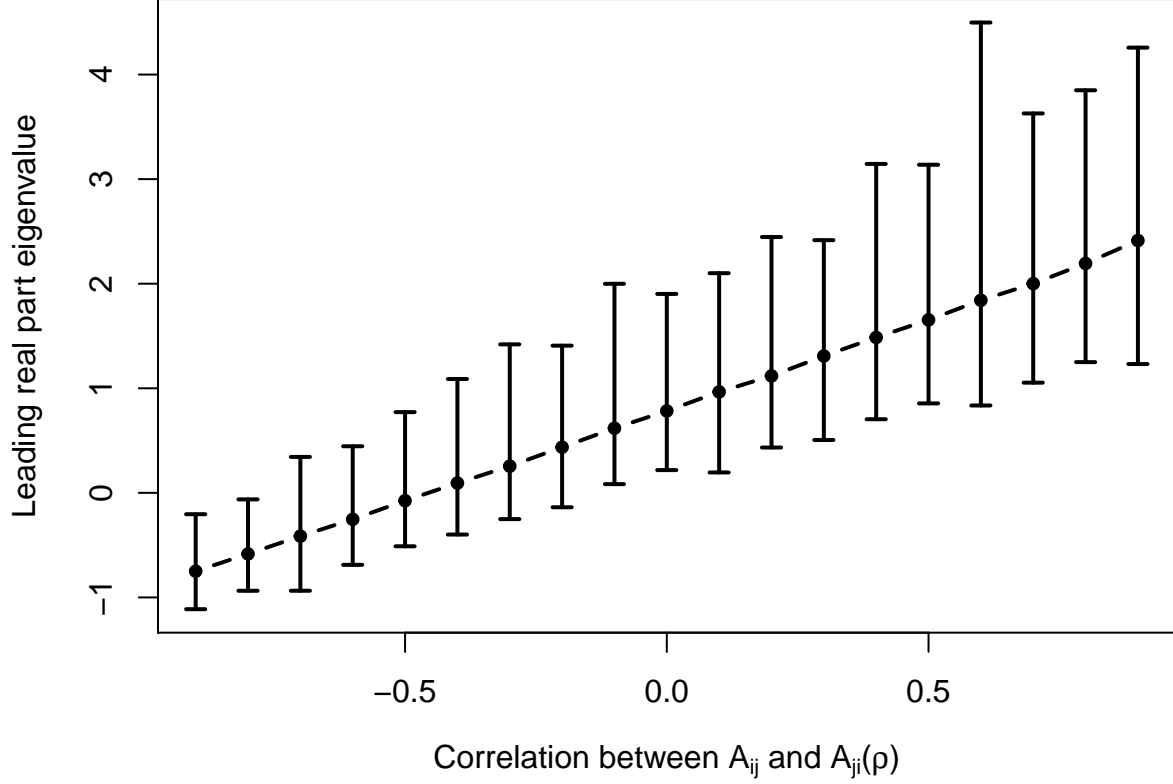
## Role of correlated matrices in stabilisation

In complex systems represented by large random matrices, correlation between matrix elements  $A_{ij}$  and  $A_{ji}$  affects the distribution of eigenvalues and therefore local stability. As the correlation between matrix elements ( $\rho$ ) decreases, the eigenvalue spectra changes such that more variation falls along the imaginary axis. The figure panels below compare a random matrix in which  $\rho = 0$  (left) to one in which  $\rho = -0.5$ . In both panels, complex systems include  $S = 1000$  components, with diagonal elements of  $-1$  and off-diagonal elements drawn from a normal distribution with a mean of 0 and  $\sigma = 0.4$ .



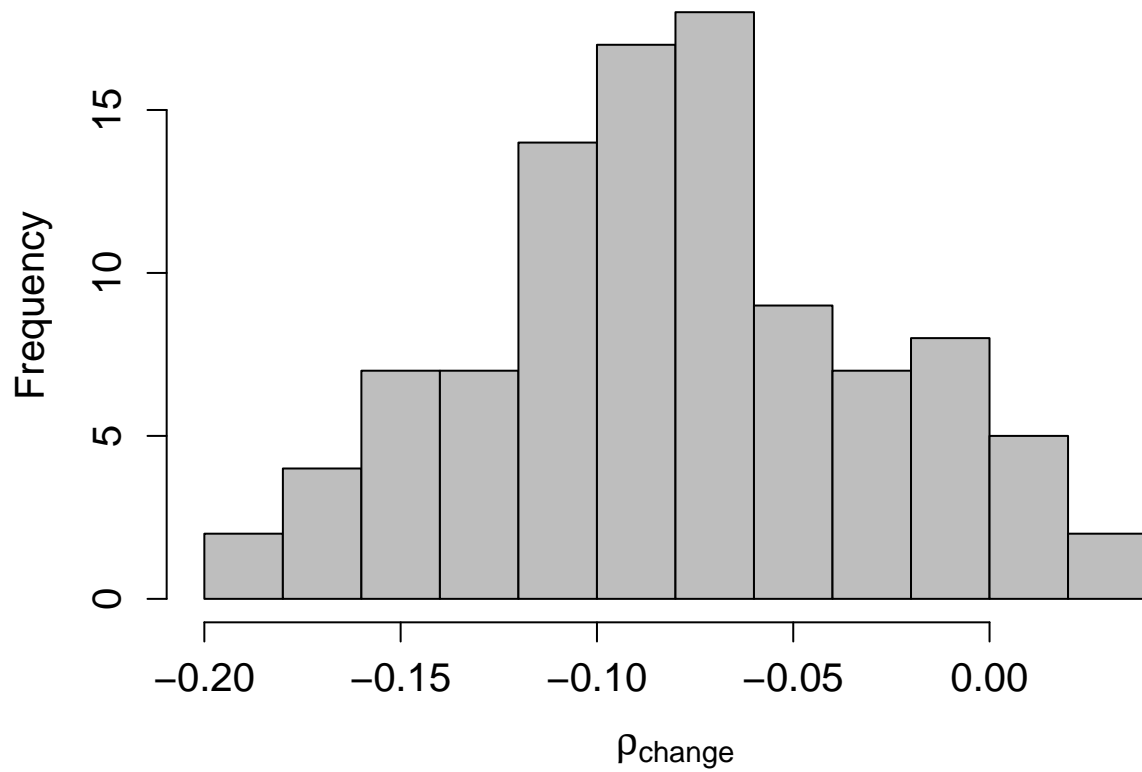
Because of this effect of  $\rho$  on the eigenvalue spectra, decreasing values of  $\rho$  will also tend to decrease the rightmost eigenvalue of the matrix  $\mathbf{A}$ . This makes it more likely that the complex system represented by  $\mathbf{A}$  is locally stable, as stability occurs when all real parts of eigenvalues are negative. Note that this elongation along the imaginary axis is also characteristic of predator-prey communities (in which, by definition  $A_{ij}$  and  $A_{ji}$  have opposing signs). Also note that as  $\rho$  increases such that  $\rho > 0$ , the same elongation happens along the real axis, making random complex systems less likely to be stable.

A simple numerical analysis illustrates the linear relationship between  $\rho$  and the expected value of the real part of the leading eigenvalue,  $\max(\Re(\lambda))$ . Below, I have run 1000 simulations across values of  $\rho$  from -0.9 to 0.9 for complex systems with  $S = 25$  components. Error bars show 95% bootstrapped confidence intervals.



22

23 In the main text, I demonstrated that when  $S$  is finite but system complexity  $\sigma\sqrt{SC}$  is high ( $C$  defines the  
 24 connectance of  $\mathbf{A}$ , or the proportion of non-zero off-diagonal elements), variation in component response rate  
 25  $\gamma$  often underlies system stability. In other words, highly complex systems that are observed to be stable  
 26 would not be if we removed the variation in their component response rates. Mathematically, this means  
 27 that by multiplying  $\mathbf{A}$  by a diagonal matrix  $\gamma$  with variable elements, the sign of  $\max(\Re(\lambda))$  is sometimes  
 28 flipped from positive to negative in these finite systems of high complexity. Interestingly, this increase in  
 29 stability given  $Var(\gamma) > 0$  is not necessarily caused by  $\gamma$  decreasing  $\rho$ . In fact, for  $S = 25$ ,  $\sigma = 0.4$ , and  
 30  $C = 1$ , random complex systems that are stabilised by  $\gamma$  typically have increased  $\rho$  values. Note that for  
 31 these parameter values, 1 million simulations found  $\mathbf{M} = \gamma\mathbf{A}$  to be stable for 36 systems when all  $\gamma_i = 1$ , but  
 32 383 systems when  $\gamma_i \sim \mathcal{U}(0, 2)$ . Below shows the distribution of the difference in  $\rho$  between systems with  
 33 versus without  $var(\gamma)$  for 100 stabilised systems; that is,  $\rho_{change} = \rho_{var(\gamma)} - \rho_{\gamma=1}$ .



34

35 For all cases in which  $\mathbf{A}$  was stabilised by  $Var(\gamma)$ , the resulting  $\mathbf{M} = \gamma\mathbf{A}$  had increased rather than decreased  
 36  $\rho$  ( $\rho_{var(\gamma)} > \rho_{\gamma=1}$ ). Hence, decreasing the correlation between  $\mathbf{A}_{ij}$  and  $\mathbf{A}_{ji}$  was not by itself the cause of  
 37 stability. For a clearer picture of the effect of  $\gamma$ , it is useful to show the relationship between  $\rho$  and  $\max(\Re(\lambda))$   
 38 again as above with, but this time with how the relationship changes given  $Var(\gamma)$ .