Component response rate variation drives stability in large complex systems

Supporting information

Brad Duthie

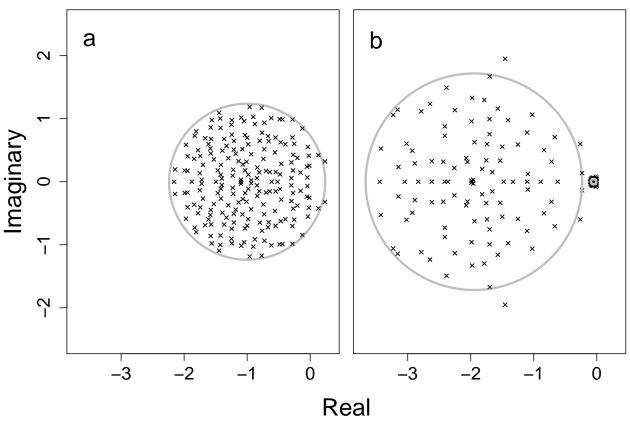
5 Code and simulations underlying Fig. 1

The sample M used for the eigenvalue distributions in Fig. 1 of the text is available on GitHub, and was produced with by running the following function.

```
find_bgamma <- function(S = 200, C = 0.05, Osd = 0.4, iters = 10000){
    while(iters > 0){
        A_dat <- rnorm(n = S * S, mean = 0, sd = Osd);
        A_mat <- matrix(data = A_dat, nrow = S);</pre>
        C_{dat} \leftarrow rbinom(n = S * S, size = 1, prob = C);
        C_mat <- matrix(data = C_dat, nrow = S, ncol = S);</pre>
        A_mat <- A_mat * C_mat;
        gammas \leftarrow c(rep(1.95, S/2), rep(0.05, S/2))
        mu_gam <- mean(gammas);</pre>
        diag(A_mat) <- -1;</pre>
                <- gammas * A_mat;
        ΑO
                <- mu_gam * A_mat;
        A0_e
                <- eigen(A0)$values;
        AO_r
                <- Re(A0_e);
        AO_i
                \leftarrow Im(A0_e);
        A1_e
                <- eigen(A1)$values;
                <- Re(A1_e);
        A1_r
        A1_i
               <- Im(A1_e);
        if(max(A0_r) >= 0 \& max(A1_r) < 0){
             return(list(A0 = A0, A1 = A1));
             break;
        print(iters);
        iters <- iters - 1;
    }
```

- The above function terminates when a matrix M is found that is not stable when all component response rates are set to $\gamma = 1$, but is stable when half of component response rates are 1.95 and half are 0.05. The function is used to illustrate the concept of how a fast versus slow component responses can cause a system to become stable. Simulations were run for iter = 1000000, but terminated once an acceptable A0 and A1 were found. The code below plots the eigenvalue distributions of A0 and A1 in panels a and b, respectively.
- The plot itself can be recreated with the code below.

```
text(x = -3.5, y = 2.25, labels = "a", cex = 2);
points(x = A0x0, y = A0y0, type = "1", 1wd = 3, col = "grey");
points(AO_r, AO_i, pch = 4, cex = 0.7);
plot(A1_r, A1_i, xlim = c(-3.7, 0.3), ylim = c(-2, 2), pch = 4, cex = 0.7,
     xlab = "", ylab = "", cex.lab = 1.5, cex.axis = 1.5, asp = 1,
     col = "black", yaxt = "n");
vl \leftarrow seq(from = 0, to = 2*pi, by = 0.001);
A0x1a \leftarrow sqrt(100) * sd(A1vec[fhalf]) * cos(vl) + mean(diag(A1)[1:100]);
A0y1a <- sqrt(100) * sd(A1vec[fhalf]) * sin(vl);
points(x = A0x1a, y = A0y1a, type = "1", lwd = 3, col = "grey");
A0x1b <- sqrt(100) * sd(A1vec[shalf]) * cos(vl) + mean(diag(A1)[101:200]);
A0y1b <- sqrt(100) * sd(A1vec[shalf]) * sin(vl);
points(x = A0x1b, y = A0y1b, type = "1", lwd = 3, col = "grey");
points(A1_r[1:100], A1_i[1:100], pch = 4, cex = 0.7);
text(x = -3.5, y = 2.25, labels = "b", cex = 2);
mtext(side = 1, "Real", outer = TRUE, line = 3, cex = 2);
mtext(side = 2, "Imaginary", outer = TRUE, line = 2.5, cex = 2);
```



To find out how frequently M was stable given that all $\gamma = 1$ versus $\gamma = \{1.95, 0.05\}$, the function below was

```
A1_count <- 0;
    while(iters > 0){
        A dat \leftarrow rnorm(n = S * S, mean = 0, sd = Osd);
        A_mat <- matrix(data = A_dat, nrow = S);
        C_{dat} \leftarrow rbinom(n = S * S, size = 1, prob = C);
        C_mat <- matrix(data = C_dat, nrow = S, ncol = S);</pre>
        A_mat <- A_mat * C_mat;
        gammas \leftarrow c(rep(1.95, S/2), rep(0.05, S/2))
        mu gam <- mean(gammas);</pre>
        diag(A_mat) < -1;
                <- gammas * A_mat;
        ΑO
                <- mu_gam * A_mat;
                <- eigen(A0)$values;
        A0_e
        AO_r
                \leftarrow Re(A0_e);
        AO_i
                \leftarrow Im(A0_e);
        A1 e
               <- eigen(A1)$values;
        A1_r
                <- Re(A1_e);
                <- Im(A1_e);
        A1 i
        if(max(A0_r) < 0){
             ress[iters, 1] <- 1;
             A0 count
                             <- A0_count + 1;
        if(max(A1_r) < 0){
             ress[iters, 2] <- 1;</pre>
             A1_count
                             <- A1_count + 1;
        print(c(iters, A0_count, A1_count));
        iters <- iters - 1;</pre>
    }
    return(ress);
}
```

The function above was run for iters = 1000000, and the resulting matrix ress was returned. Each row of ress represents a single M given $\gamma=1$ (column 1) versus $\gamma=\{1.95,0.05\}$ (column 2). Values of 0 indicate that M was found to be unstable (at least one real component of its eigenvalues greater than or equal to zero), whereas values of 1 indicate that M was found to be stable (all real components of eigenvalues are negative). The frequencies of stable M were 0 given $\gamma=1$ and 0 given $\gamma=\{1.95,0.05\}$, as reported in the main text and legend of Fig. 1.

23 Code and simulations underlying Fig. 2

- Figure 2 of the main text shows how eigenvalue distributions in a system where S = 1000, C = 1, and $\sigma = 0.4$.
- Eigenvalues can be reproduced using the code below for when $\gamma = 1$ (panel a) and $\gamma \sim \mathcal{U}(0,2)$ (panel b).
- The code below reproduces the figure itself.

