A Dynamical Model for Activating and Silencing the Mitotic Checkpoint

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ABSTRACT

Supplement Material

Text S1 Turnover rate at unattached kinetochores

We want to calculate the average time that a particle needs in order to enter a sub-volume in the reaction vessel (in our case, this is the time it takes for an O-Mad2 particle to hit an unattached kinetochore). The radii of the nucleus and kinetochore are R and r, respectively, and C-Mad2 diffuses with a coefficient D. Applying a spherical Taylor expansion to an average hitting time model, and applying Neumann boundary conditions, leads to the following formula:

\[ \langle T \rangle \approx \frac{R^3}{3Dr} \]  

We can use this formula to calculate how long it takes for all O-Mad2 particles to diffuse to a kinetochore. In particular, we can calculate the average hitting time, which is the time needed for half the particles to diffuse to a kinetochore before that time. It is the half-time of the following reaction:

\[ \text{O-Mad2} \xrightarrow{k_3 \times [\text{KinU}]} \text{C-Mad2} \]  

Using the relationship between the half-time and the reaction rate

\[ k = \frac{\ln(2)}{t_{1/2}} \]  

we can estimate the maximum turnover rate of C-Mad at the kinetochores. Assuming a nucleus radius of 6µm, a kinetochore radius of 0.1µm and the Mad2 diffusion coefficient of 16.61µm²s⁻¹, we find a half-time of 43 s, which is consistent with experimental findings and results in a turnover rate of 0.016s⁻¹ at each kinetochore.

Text S2 Calculating diffusion coefficients

In addition to the reactions and their governing laws, a particle simulation also needs spatial information. It requires knowledge of the shape and size of every protein and their relative diffusion coefficients. For the sake of simplicity, we model all proteins as spheres with uniform densities. Given the mass, which is known from lab-experiments, and assuming a uniform spherical particle, we can calculate the radius of the sphere using

\[ r_i = 0.066m_i^{1/3} \]  

in which \( m_i \) refers to lab-measured mass of the molecule in Da and \( r_i \) is the radius of the molecule in nm. Following the Stokes-Einstein equation we can calculate the diffusion coefficient using

\[ D_i = \frac{k_BT}{6\pi\eta r_i} \]
where

\[ D_i = \text{diffusion coefficient in } m^2 s^{-1} \]

\[ k_B = \text{Boltzmann-Constant} \ (1.380 \times 10^{-23} J/K) \]

\[ T = \text{temperature in K} \]

\[ r_i = \text{radius of the assumed spherical-particle in m} \]

\[ \eta = \text{viscosity of the medium in Ns/m}^2 (0.891 \times 10^{-3} \text{ in water}; 6.75 \times 10^{-3} \text{ in nucleus}) \]

**Text S3 Coarse-graining**

Typical time-steps in particle simulations are \(10^{-9} s\), which would imply \(1.2 \times 10^{11}\) time-steps to realize the full metaphase (around 20 real-time minutes). Realistic particle numbers are around 1,000,000 per species, which leads to an infeasible large time requirement for simulating the full system. For reference, simulating 1,751 particles for a total of \(8 \times 10^7\) time-steps takes around one week. For this reason, we introduce two coarse-graining techniques: (1) pseudo particles and (2) scaling of the time.

One approach to reducing the amount of particles is to enlarge their reaction-surface and decrease their number. Every particle has a radius \(r_i\) and a surface where it interacts with other particles. As particles are assumed to be spherical, this surface is \(4\pi r_i^2\). A number of particles can be merged to one bigger one:

\[ N_i \to \hat{N}_i, \text{ with } N_i \gg \hat{N}_i \quad (6) \]

\[ r_i \to \hat{r}_i, \text{ with } r_i < \hat{r}_i \quad (7) \]

where \(N_i\) is the number of particles of type \(i\), \(\hat{N}_i\) is the reduced number and \(\hat{r}_i\) is the increased radius of pseudo-particle \(i\). Under the constraint that the reaction surface must be conserved, the formula for the new radius is given by

\[ N_i 4\pi r_i^2 = \hat{N}_i 4\pi \hat{r}_i^2 \quad (8) \]

\[ \Rightarrow \hat{r}_i = r_i \sqrt{\frac{N_i}{\hat{N}_i}} \quad (9) \]

It is also possible to reduce the time in the same manner. A process taking time \(T\) can be reduced to \(\hat{T}\) using a factor \(f_t\) so that

\[ T \to \hat{T} = f_t T, \text{ with } T \gg \hat{T} \quad (10) \]

To ensure that all reactions reach the steady state in time \(\hat{T}\) (if such a state exists), every reaction rate has to be multiplied by the factor \(f\):

\[ k_i \to \hat{k}_i = f k_i \quad (11) \]

In a reaction system with relevant spatial characteristics, the time-scaling means that a given particle may no longer be able to cover the distance \(d\) between point \(P = (p_x, p_y, p_z)\) and \(Q = (q_x, q_y, q_z)\). Thus, the diffusion coefficients for each type of particle must also be modified. In \(n\)-dimensional space the mean-square-displacement (MSD) is an estimate for the average distance a particle has moved and is calculated using

\[ MSD_i = 2dT D_i \quad (12) \]

where

\[ d = \text{dimension of the simulation space} \]

\[ T = \text{simulated time} \]

\[ D_i = \text{diffusion-coefficient of particle } i \]
The square root of the MSD is used as the mean distance that the particle has moved from its origin

\[ \text{dist}_i = \sqrt{\text{MSD}_i} \]  

(13)

and this \( \text{dist}_i \) must also conserved through the time-scaling process:

\[
\sqrt{2d\hat{T}D_i} = \text{dist}_i = \sqrt{2d\hat{T}\hat{D}_i} \\
\hat{D}_i\hat{T} = TD_i \\
\hat{D}_i\hat{T} = TD_i \\
\hat{D}_i = f_iD_i
\]  

(14)  

(15)  

(16)  

(17)

This theoretical result is valid given the limited range of diffusion coefficients of \( \approx 0.5 - 30\mu m^2 s^{-1} \). If the compression-factor \( f \) exceeds 100 (which is quite a low factor) then the diffusion coefficient can no longer be adapted in this way, because the displacements of particles in each time-step would become large and the system would become unstable.

An alternative way to make the particle move from P to Q in the scaled time \( \hat{T} \) is to decrease the distance between the points by scaling the space (the reaction volume). A space of dimensions X, Y, Z can be transformed using

\[
S = (X,Y,Z) \rightarrow (\hat{X}, \hat{Y}, \hat{Z}) = \hat{S}
\]  

(18)

where every individual point is transformed according to

\[
p = (x, y, z) \rightarrow f(p) = \left( \frac{x}{f_s}, \frac{y}{f_s}, \frac{z}{f_s} \right)
\]  

(19)

This transformation represents a consistent spatial dilation of all axis by an as yet unknown factor \( f_s \). Given that dilatations are length-conserving, the distance in the new space is scaled directly by the factor \( f_s \) according to

\[
d = \frac{\hat{d}}{f_s}
\]  

(20)

One distance of interest is the MSD (described above), which gives the average displacement of a particle by diffusion. The relation between this distance and the time-compression-factor is

\[
\text{MSD} = \sqrt{6D\hat{T}} \\
\hat{\text{MSD}} = \sqrt{6D\hat{T}} \\
\Rightarrow \text{MSD} = f_i\hat{\text{MSD}} \\
\Rightarrow d^2 = f_i\hat{d}^2
\]

which can be incorporated into Equation (20) to give

\[
f_s = \sqrt{f_i}
\]  

(21)

Those two scalings in time and space ensure that a particle which took \( \delta \) seconds to move a distance \( \eta \) in the original system, needs \( f_i\delta \) seconds to cover a distance from \( f_i\eta \) in the scaled system.

References

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Figure S1. Parameter study for the most crucial reaction rates, that are $k_3$ and $k_8$. The colorbars on the right of each panel present the half-time of Securin degradation in seconds. The x-axis varies the rate of the labeled reaction ($k_3$) while the y-axis alters reaction rate $k_8$. A: Normal mitosis, where all kinetochores attach in an average time of 20 – 30 minutes, corresponding to $\approx 1500$ seconds. Looking at the presented heatmap this corresponds to a value of $> 0.015$ for $k_3$ and $< 0.015$ for $k_8$. B: Disturbed mitosis, where one chromosome is unable to attach. It shows that single kinetochores are able to extend metaphase for several hours (the shown values are only half-times of Securin). The rates for a reliable arrest in mitosis coincide with the ones in panel A. Both panels together suggest that $k_3$ value should be around 0.015 as it was theoretically determined to be a maximum of 0.016 (cf. Text S1) and $k_8$ should not exceed 0.015. In other terms have both kinetochores a strong enough influence to maintain the SAC as their reaction rates are similar.

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Figure S2. Shown is the schematic reaction volume used for the spatially-stochastic simulation. The whole nuclear space with a radius of 6µm is modeled, but all species are hold in the green area. This space exclusion is necessary to guarantee fast turnover from O-Mad2 to C-Mad2. All our model only take place in the green area, which can be seen as a well-mixed soup.

Figure S3. Simulation outcome of all mutation experiments. Rows correspond to the species BubR1:Bub3, Mad2 and Cdc20, respectively. Columns present an initial 10-fold over-expression, 40% depletion and 5% depletion, respectively. Axis and labels of the curves coincides with the ones in figure 3 and 4. Last kinetochores attachment is after 1500 seconds. Outcome of the classification can be found in the manuscript.
### Table S1. Kinetic Parameters of the full SAC model

| Classification | Parameter | Value     | Remark                                                        |
|----------------|-----------|-----------|---------------------------------------------------------------|
| Attachment     | $k_{\text{attach}}$ | $0.0035 s^{-1}$ | Metaphase around 20min                                       |
| Activation     | $k_2$     | $100 \mu M^{-1} s^{-1}$ | in-silico parameter study$^1$                                |
|                | $k_{-2}$  | $0.08 s^{-1}$ | in-silico parameter study$^1$                                |
|                | $k_3$     | $> 0.015 s^{-1}$ | per kinetochore; taken from Howell et. al.$^2$ and theoretical calculation |
|                | $k_{-3}$  | $0.2 s^{-1}$ | taken from Howell et. al.$^2$                                |
|                | $k_4$     | $10.0 \mu M^{-1} s^{-1}$ | in-silico parameter study$^3$                                |
|                | $k_5$     | $10.0 \mu M^{-1} s^{-1}$ | in-silico parameter study$^4$                                |
|                | $k_{-5}$  | $0.02 s^{-1}$ | in-silico parameter study$^4$                                |
|                | $k_6$     | $0.001 \mu M^{-1} s^{-1}$ | taken from Musacchio$^5$                                     |
|                | $k_{-6}$  | $0.01 s^{-1}$ | in-silico parameter study$^3$                                |
|                | $k_7$     | $0.01 \mu M^{-1} s^{-1}$ | in-silico parameter study$^4$                                |
|                | $k_{-7}$  | $0.2 s^{-1}$ | in-silico parameter study$^4$                                |
|                | $k_{T1}$  | $0.01 \mu M^{-1} s^{-1}$ | taken from DeAntoni et. al.$^6$                             |
|                | $k_{T2}$  | $0.02 s^{-1}$ | taken from DeAntoni et. al.$^6$                             |

Silencing

| Classification | Parameter | Value     | Remark                                                        |
|----------------|-----------|-----------|---------------------------------------------------------------|
| $k_1$          | $5 \mu M^{-1} s^{-1}$ | in-silico parameter study$^1$                                |
| $k_{-1}$       | $0.08 s^{-1}$ | in-silico parameter study$^1$                                |
| $k_8$          | $< 0.015 s^{-1}$ | per kinetochore; this study                                  |
| $k_9$          | $< 0.015 s^{-1}$ | per kinetochore; this study                                  |
| $k_{10}$       | $0.2 s^{-1}$ | in-silico parameter study$^4$                                |
| $k_D$          | $0.05 s^{-1}$ | taken from Dick et. al.$^7$                                  |

### Table S2. Spatial Parameters of the SAC Model

| Species         | Mass in kDa | Diffusion in $\mu m^2 s^{-1}$ | Initial Concentration in $\mu M$ | Source Concentration |
|-----------------|-------------|-------------------------------|----------------------------------|----------------------|
| O-Mad2          | 26.06       | 16.61                         | 0.15                             | 2, 8, 9              |
| C-Mad2          | 26.06       | 16.61                         | 0.01875                          | 2, 8, 9              |
| Cdc20           | 54.72       | 12.97                         | 0.13                             | 8, 9                 |
| BubR1:Bub3      | 242.00      | 7.92                          | 0.22                             | 8, 9                 |
| APC/C           | 836.5       | 5.23                          | 0.09                             | 9                    |

Kinetochores initial amount is 92 and they do not diffuse. Other species particles start from zero. Their mass and diffusion coefficient combine from the basic blocks.

### Table S3. Kinetic Parameters of the coarse-grained SAC model

| Classification | Parameter | Value     | Remark                                                        |
|----------------|-----------|-----------|---------------------------------------------------------------|
| Attachment     | $k_{\text{attach}}$ | $0.0035 s^{-1}$ | Metaphase around 20min                                       |

SAC Activation

| Classification | Parameter | Value     | Remark                                                        |
|----------------|-----------|-----------|---------------------------------------------------------------|
| $k_{M1}$       | $> 0.015 s^{-1}$ | per kinetochore; taken from Howell et. al.$^2$ and theoretical calculation |
| $k_{M2}$       | $100 \mu M^{-1} s^{-1}$ | in-silico parameter study$^1$                                |
| $k_{-M2}$      | $0.08 s^{-1}$ | in-silico parameter study$^4$                                |

SAC Silencing

| Classification | Parameter | Value     | Remark                                                        |
|----------------|-----------|-----------|---------------------------------------------------------------|
| $k_{M3}$       | $< 0.015 s^{-1}$ | per kinetochore; this study                                  |
| $k_D$          | $0.05 s^{-1}$ | taken from Dick et. al.$^7$                                  |