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MAARS: a novel high-content acquisition software for the analysis of mitotic defects in fission yeast

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ABSTRACT Faithful segregation of chromosomes during cell division relies on multiple processes such as chromosome attachment and correct spindle positioning. Yet mitotic progression is defined by multiple parameters, which need to be quantitatively evaluated. To study the spatiotemporal control of mitotic progression, we developed a high-content analysis (HCA) approach that combines automated fluorescence microscopy with real-time quantitative image analysis and allows the unbiased acquisition of multiparametric data at the single-cell level for hundreds of cells simultaneously. The Mitotic Analysis and Recording System (MAARS) provides automatic and quantitative single-cell analysis of mitotic progression on an open-source platform. It can be used to analyze specific characteristics such as cell shape, cell size, metaphase/anaphase delays, and mitotic abnormalities including spindle mispositioning, spindle elongation defects, and chromosome segregation defects. Using this HCA approach, we were able to visualize rare and unexpected events of error correction during anaphase in wild-type or mutant cells. Our study illustrates that such an expert system of mitotic progression is able to highlight the complexity of the mechanisms required to prevent chromosome loss during cell division.

INTRODUCTION The correct partitioning of replicated chromosomes between two daughter cells at each division is essential to prevent genome instability. When this process is perturbed, aneuploid daughter cells (i.e., cells possessing an incorrect chromosome number) are generated. Aneuploidy is a well-known cause of severe genetic diseases, such as Down’s syndrome, and is an almost ubiquitous feature of human cancers (Holland and Cleveland, 2009; Compton, 2011). The two best-known mechanisms leading to aneuploidy involve chromosome segregation and spindle-positioning defects. Abnormal chromosome segregation is caused by defects in surveillance mechanisms (spindle assembly checkpoint or Aurora B kinase) or linked to increased centrosome copy number, kinetochore–microtubule attachment errors, or cell-cycle regulation defects (Vitre and Cleveland, 2012). On the other hand, the control of spindle positioning requires extranuclear players because many cell types orient their spindles according to preexisting polarity cues and use an astral microtubule contact with the cell cortex to position or align the mitotic apparatus (Carminati and Stearns, 1997; Shaw et al., 1997; Adames and Cooper, 2000; Yeh et al., 2000; Gachet et al., 2004; Kozlowski et al., 2007; Kotak and Gonczy, 2013). In addition, chromosome segregation, spindle positioning, and cytokinesis must be coordinated to prevent chromosome loss. Thus multiple mitotic parameters contribute to the spatiotemporal control of chromosome partitioning.

Studies in yeast have contributed to our present understanding of the general principles underlying the regulation of chromosome segregation and spindle positioning in dividing cells (Nurse and Thuriaux, 1980; Page and Snyder, 1993; Yanagida, 1998; Segal and
Bloom, 2001). The fission yeast Schizosaccharomyces pombe is a rod-shaped, symmetrically dividing eukaryote that splits by medial fission. *S. pombe* possesses three chromosomes (Kohli et al., 1977), which can be tracked by live imaging during mitotic progression (Tournier et al., 2004). In this organism, mitosis is divided into three phases (Nabeshima et al., 1998; Tatebe et al., 2001), and spindle formation begins in M phase, as in higher eukaryotic cells (Russell et al., 1989). During phase 1 (prophase), a short spindle of <2.0 μm is formed, and chromosomes are captured by microtubules at kinetochores, large protein complexes assembled at the centromeres of each chromosome. Kinetochores begin to align at the spindle center in phase 1 (Mary et al., 2015). In phase 2 (prometaphase/metaphase/anaphase A), the spindle maintains roughly the same length, and the kinetochores oscillate between the two spindle pole bodies (SPBs; Mary et al., 2015). At the end of phase 2, sister chromatids separate through the activity of an enzyme called separase, which degrades the cohesin link between sister chromatids, allowing their movement toward the SPBs (anaphase A; Yanagida, 2000). At this stage, it is generally accepted that the abrupt change in the rate of spindle elongation signals the beginning of anaphase B (phase 3; Nabeshima et al., 1998). Finally, during cytokinesis, the actomyosin ring contraction initiates the separation of the two daughter cells (Chang and Nurse, 1993).

Chromosome attachment to the mitotic spindle results from a stochastic search-and-capture process that is intrinsically prone to errors (Kirschner and Mitchison, 1986). In *S. pombe*, chromosome capture and biorientation successfully occur within minutes before sister chromatid separation (Grishchuk and McIntosh, 2006; Gachet et al., 2008). This process is particularly efficient due to the presence of surveillance mechanisms that prevent chromosome loss (less than one chromosome is lost in 10^4 divisions). This control is ensured by an error detection and correction system, which involves the Aurora B kinase, a master regulator of cell division (Petersen et al., 2001; Petersen and Hagan, 2003; Sampath et al., 2004; Kotwalwale et al., 2007; Mora-Bermudez et al., 2007; Steigemann et al., 2009; Tada et al., 2011) and the spindle assembly checkpoint (SAC), which delays anaphase onset in the presence of unattached chromosomes (Rieder et al., 1995; Rudner and Murray, 1996; Cleveland et al., 2003). Aurora B and SAC components operate at the level of kinetochores, where they promote the stabilization of proper attachment and the destabilization of erroneous ones. Of note, in metaphase, Aurora B destabilizes faulty kinetochore-to-microtubule attachments such as merotely (one chromosome attached to both poles) through the phosphorylation of key kinetochore substrates (Tanaka et al., 2002; Cimini et al., 2006). Merotelic attachment is also corrected after anaphase onset through spindle elongation forces (Courtheoux et al., 2009; Gay et al., 2012). Further work also suggests that metaphase spindle forces are also important to prevent the appearance of lagging chromosomes in anaphase (Choi and McCollum, 2012).

Quantitative analysis of mitotic fidelity in fission yeast is usually addressed using minichromosome-loss experiments (Niwa, 2016). However, this technique cannot be used to determine the precise cause of chromosome loss, and because the rate of chromosome loss is generally very low, the direct cytological examination of chromosome segregation errors using live-cell imaging is not feasible. Thus, to understand fully the mechanisms ensuring the accuracy of mitotic progression, we developed a high-content analysis approach, which combines automated fluorescence microscopy with quantitative image analysis of mitosis. This methodology allows the acquisition of unbiased multiparametric data at the single-cell level for high-content analysis of mitosis. We tested the robustness of this method on several *S. pombe* mutants and were able to identify unexpected phenotypes in wild-type and mutant cells. Our study illustrates the impressive benefits of using MAARS to analyze quantitatively mitotic fidelity in eukaryotic cells.

**RESULTS**

**MAARS: an automated, robust, and open-source software for high-content analysis of mitosis**

To analyze quantitatively the mechanisms controlling mitotic fidelity, we developed an automated open-source image acquisition and on-the-fly analysis pipeline named MAARS. We first built the connection between the computer and the microscope using open-source software called Micro-Manager (Edelstein et al., 2010). In contrast to all available commercial software, Micro-Manager supports multiple microscopy platforms and is compatible with the image-processing package ImageJ. Thus MAARS can also be installed on a variety of wide-field microscopes. MAARS delivers a graphic interface and controls image acquisitions via Micro-Manager. With its user-friendly window, it is possible to adjust microscope parameters such as exposure time, frequency of image acquisition, type of color channel, and so on. In addition, specific parameters for mitotic analysis (i.e., size of the spots, type of fluorescent markers) are also selected at this step.

We next implemented an accurate segmentation method to detect fission yeast or budding yeast cells (Figure 1 and Supplementary Figures S1 and S2). This segmentation technique is based on a mathematic algorithm (Jolou et al., 2013) that requires only bright field with multiple z-section acquisition (∼34 frames) to compute the correlation between cell edges (Supplemental Figure S1B, blue dots) and theoretical cell edges (Supplemental Figure S1B, green curve). Alternative methods consist of marking cell edges (Schmitter et al., 2013) or increasing the contrast between the cell and the background using fluorescent agents (Peng et al., 2013). However, our segmentation technique can be used on nonfluorescent native cell lines with no risk of phototoxicity (Supplemental Figure S1). The cell edges are drawn at the point of high correlation score using Otsu thresholding, as available in ImageJ (Supplemental Figure S1, A–C). Owing to the optimized parallelization described in Materials and Methods, MAARS can perform fast segmentation. The acquisition and segmentation of a single field of cells with multiple z-sections takes ∼15 s (1100 cells/field; see Materials and Methods).

To prevent the segmentation of false-positive cells, we implemented two types of filters—solidity and gray-level filters. The solidity value (area of an object divided by its convex area; Figure 1, bottom) can be adjusted to fit with several cell shapes (round, bent, or long cells). For example, the solidity parameter is 0.84 for *S. pombe* and 0.5 for *Saccharomyces cerevisiae*, as shown in Supplemental Figure S1E. In addition to this cell shape parameter, we implemented a second filter, called gray intensity filter, based on the pixel intensity of the inner cell on correlation image (see Materials and Methods). The pixel intensity of the cell is compared directly with the intercell space background; Supplemental Figure S1C). In
In contrast, when the distance between spindle poles remains constant, the cell is judged to be in telophase/cytokinesis. MAARS classifies cells in metaphase or anaphase as judged by the spindle elongation rates (Figure 2B and Supplemental Figures S3 and S4). Finally, MAARS identifies whether kinetochores (GFP signals) are detected using a line-scan function drawn between the two poles. If a kinetochore is located between the poles after anaphase onset, it is classified as a lagging kinetochore.

MAARS is very efficient because it can detect \( \sim 100 \) mitotic cells per field containing \( \sim 1000 \) unsynchronized cells, and it enables easy retrieval of the films and analyses of interest. To test the performance of MAARS software, we performed high-content acquisition of spindle elongation (spindle pole–spindle pole distance), cell shape, spindle orientation/positioning, rate of spindle elongation, and presence of lagging chromosomes in different genetic backgrounds.

Quantitative analysis of spindle elongation and cell shape distribution to identify cell cycle delays

To evaluate MAARS pipeline, we analyzed several \( S\). \( pombe \) strains with established cell cycle deficiencies and compared them with wild-type cells. We hypothesized that the distribution of spindle length and cell shape in a population of cells could reveal abnormal cell cycle progression, such as G2 delays or mitotic delays. We chose to measure pole-to-pole distance as an approximation for spindle length because in most situations, spindles are not curved before telophase. \( S\). \( pombe \) growth occurs during G2, and the integrity of the genome is controlled at this stage before mitotic entry. In the presence of DNA damage or unreplicated DNA, cells are delayed in G2 and become elongated compared with wild type. Thus cell shape measurement can serve as a good marker to investigate cell cycle defects such as G2 delays. We decided to determine the ratio of width to length (W/L) as a characteristic of cell shape. Long cells (cells delayed in G2: W/L small) are characterized with a very low ratio of vertical axis to longitudinal axis (Figure 3A). In contrast, short cells (short G2 cells: W/L high) are characterized by a high ratio (Figure 3A). For example, in \( S\). \( pombe \), cell cycle–deficient cells, also known as cdc mutants (Nurse and Thuriaux, 1980), display a G2 cell cycle delay and become elongated before entering mitosis compared with wild-type cells (Figure 3A). We also compared spindle size in cdc25-22 mutants (Figure 3B, red) versus wild type (Figure 3B, black). MAARS detected two main populations of spindle sizes corresponding to metaphase (maximum peak \( \sim 2–3 \) \( \mu m \)) and anaphase (maximum peak \( \sim 6–9 \) \( \mu m \); Figure 3B). As expected, the size of metaphase spindles in the cdc25-22 mutant was comparable to that in the absence of filters, MAARS can detect \( \sim 85\% \) of cells per field (Supplemental Figure S1D). The rejected 15\% of cells includes cells located on the edge of the image or overlapping cells. Using proper gray filter intensity and solidity, only 1\% of the cells (\( n = 7434 \)) were false positives (Supplemental Figure S1D).

After segmentation, fluorescent images are automatically analyzed on the fly with MAARS (Figure 2A and Supplemental Figure S3). Cells are illuminated with the appropriate channel, and fluorescent spots (e.g., spindle poles in cyan fluorescent protein [CFP] or kinetochores in green fluorescent protein [GFP]) are detected using TrackMate (Tinevez et al., 2017). MAARS (Kernel in Java) first determines whether the fluorescent spots are within the cell (Figure 2A and Supplemental Figure S3) before analyzing their position (Figure 2B and Supplemental Figure S3). The recognition of mitotic cells is achieved in several postanalysis steps (extension in Python). MAARS detects cells with two spindle poles that move apart for at least 3 min (spindle elongation). In contrast, when the cell is judged to be in telophase/cytokinesis, MAARS classifies cells in metaphase or anaphase as judged by the spindle elongation rates (Figure 2B and Supplemental Figures S3 and S4). Finally, MAARS identifies whether kinetochores (GFP signals) are detected using a line-scan function drawn between the two poles. If a kinetochore is located between the poles after anaphase onset, it is classified as a lagging kinetochore.
Quantitative analysis of spindle orientation and positioning during mitotic progression

The control of spindle positioning according to the cell center during mitosis is important to prevent anucleated cells. During G2, interphasic microtubules are required to maintain the nucleus at the cell center (Tran et al., 2001), ensuring correct spindle positioning at the beginning of mitosis. To confirm the robustness of MAARS pipeline, we analyzed spindle orientation and positioning during mitotic progression in a population of synchronized cells. A cdc25-22 thermosensitive mutant expressing SPB markers was blocked for 4 h at 36°C (G2 block) and released synchronously into mitosis at the permissive temperature of 25°C. Mitotic progression was recorded in fixed cells by following the presence of anaphase cells (4′,6-diamidino-2-phenylindole [DAPI] staining) or septated cells, characteristic of cytokinesis (calcofluor staining; Figure 4A). In parallel, live-cell analysis using MAARS was conducted every 20 min to analyze the distance between SPBs during mitotic progression (Figure 4B). As expected, we observed an increase in spindle length corresponding to anaphase and telophase cells (Figure 4, B and C). A virtual cell representation shows the average position and length of the spindle (pole-to-pole distance) in this synchronized population of cells (Figure 4D).

Combining histograms of pole-to-pole distance, orientation, and positioning as a virtual cell representation proves once again that our analysis pipeline is powerful. We next performed a similar type of experiment to analyze spindle orientation as a function of spindle length (pole-to-pole distance; Figure 4E). In exponentially growing wild-type or cdc25-22 cells, changes in spindle angle are observed from metaphase to anaphase onset (from 2 to 5 μm). The angle remains low throughout anaphase B. In contrast and as previously reported, an increased number of cells with misoriented spindles is observed in mal3Δ or mod20Δ cells from the onset to the end of mitosis (Figure 4E). Our data suggest that the MAARS pipeline is sufficiently sensitive and robust to investigate the mechanisms controlling spindle positioning in a single- or whole-cell population.

Quantitative analysis of spindle elongation during mitotic progression

At the metaphase-to-anaphase transition, an abrupt change in the rate of spindle elongation is observed in fission yeast cells. This change in the slope of the curve can be used as a tool to automatically predict the timing of anaphase and analyze the distribution in spindle elongation rates in single-cell analysis (see Materials and
Quantitative detection and phenotypic analysis of lagging chromosomes in anaphase

The fidelity of chromosome attachment is ensured by an error detection and correction system that operates in metaphase. However, merotelic attachment can also be corrected at anaphase through spindle elongation forces, and the presence of merotelic attachment in anaphase antagonizes spindle elongation rate (Courtheoux et al., 2009). Thus simultaneous detection of lagging chromosomes with the analysis of spindle elongation in single cells may provide a valuable tool to predict merotelic attachment and follow its correction in live cells.

A frequency analysis of spindle elongation rate in various mutants revealed the presence of two distinct peaks, one corresponding to prophase/metaphase (Figure 6A) and the second to anaphase cells (Figure 6B). From prophase to metaphase, the average rate of spindle elongation was similar between mutants and wild-type cells (0.7950 μm/min in wild type; n = 75). In contrast, in anaphase, mod20Δ and rad21-K1 mutants exhibited shifted or non-Gaussian distributions of spindle elongation rates compared with wild type (either lower or higher than wild type), although the average elongation rate was similar to wild type (0.7936 μm/min in wild type, n = 75; 0.7950 μm/min in mod20Δ, n = 33; and 0.8241 μm/min in rad21-K1, n = 88). To test whether some of these defects could be attributed to merotelic attachment, we upgraded the MAARS pipeline with a Python extension to automatically detect the presence of lagging kinetochores at anaphase in cells expressing spindle poles and kinetochore markers (see Materials and Methods and Supplemental Figure S3). The analysis of the films revealed several types of phenotypes. The cohesin mutant rad21-K1 displayed a large number of single, stretched kinetochores during anaphase (Figure 6, C and D; 14.45% ± 4.67; three independent experiments performed for a total number of mitoses of 254; Supplemental Movie S1), coinciding with changes in the rate of spindle elongation (Figure 6E, yellow) compared with rad21-K1 cells without lagging kinetochores (Figure 6E, pink). These stretched lagging kinetochores were not observed in wild type (Figure 6F, yellow; n = 187) or mal3Δ cells (Figure 6F, yellow; n = 203) and only occasionally detected in mod20Δ cells (Figure 6F, yellow; 2.02% ± 1.58; three independent experiments performed for a total number of mitoses of 148). In wild-type and mutant cells, we also recorded the presence of unstretched lagging kinetochores (Figure 6F, red; defined as a lagging kinetochore that regain the poles 40 s to 1 min later than other kinetochores). This type of lagging kinetochore did not affect the rate of spindle elongation (unpublished data). These results prompted us to perform a quantitative analysis of kinetochore segregation in wild-type cells to evaluate the efficiency of elongation mechanisms in wild-type background. We extensively filmed wild-type cells (n > 8000) expressing spindle pole and kinetochore markers and identified the presence of two additional types of error correction. In the first type, a pair of kinetochores (as opposed to a single chromatid) separated later than the other pairs but regained the poles soon after (Supplemental Figure S5, a single event in >8000 mitotic cells; and Supplemental Movie S2),

Methods and Supplemental Figure S4). With this method, we analyzed quantitatively the rate of spindle elongation in wild-type (Figure 5A) and mutant cells (Figure 5, B–D) expressing Cdc11-cfp (poles) and Ndc80-GFP (kinetochores; Figure 5, top). To automatically detect anaphase onset, we initially considered looking at colocalization of kinetochores and SPB fluorescence. However, this could not accurately detect anaphase onset in cells defective for the poleward movement of kinetochores at anaphase, as previously observed in the absence of Dam1 (Gao et al., 2010). We thus chose to use the slope-change method, which requires acquisition of a single fluorescent channel, allowing higher-frequency imaging and less photobleaching. Based on spindle elongation slope changes, the MAARS pipeline automatically detected anaphase onset in single cells (time zero in Figure 5, bottom, corresponds to the onset of anaphase, and all the curves are aligned according to time zero). We checked the robustness of this automatic detection of anaphase by directly visualizing kinetochore dynamics in each film (ndc80-GFP signal in green). We accurately detected anaphase B onset in 81% (n = 75) of analyzed cells. The incorrect detection can be explained by low sample quality (i.e., one SPB moves out of focus during acquisition or either metaphase or anaphase length is too short). This quantitative analysis reveals some heterogeneity in metaphase length in mal3Δ (Figure 5C, from −20 to 0 min) and in the cohesin mutant rad21-K1 (Figure 5B, from −27 to 0 min) as opposed to wild type (Figure 5A, from −15 to 0 min). Our automated analysis also shows some heterogeneity in spindle elongation rates at anaphase, especially in the case of rad21-K1 mutant (Figure 5B, bottom), possibly due to the presence of merotelic attachment, as previously reported (Courtheoux et al., 2009).

Thus MAARS provides an efficient tool to analyze spindle elongation and to highlight metaphase delays in individual cells within a population.

FIGURE 3: Automated analysis of cell shape and pole-to-pole distance distribution in several genetic backgrounds. (A) Normalized density profiles of cell shape in wild-type (black), cdc25-22 (red), mal3Δ (green), and mod20Δ (blue) cells. (B) Normalized density profiles of distance between SPBs in wild-type (black), cdc25-22 (red), mal3Δ (green), and mod20Δ (blue) cells.
our software is able to record cell division, specifically find mitotic cells, and extract on-the-fly multiple mitotic parameters. Our results demonstrate that MAARS is practical, time-efficient, and robust software for the multiparametric analysis of mitosis. This software provides statistical analysis at the cell population or single-cell level. Within the entire cell population, MAARS determines the percentage of cells in metaphase, anaphase, or cytokinesis and identifies cell size, spindle pole position, and so on. The results are displayed as a spots distribution, but it is possible to return to the cell image to confirm unexpected results. MAARS can adapt to different cell types (including the budding yeast by adjusting the solidity parameter) and different fluorescent markers (GFP, red fluorescent protein [RFP], or CFP). By simply adjusting the solidity parameter, it is possible to use MAARS for the analysis of cells displaying abnormal cell shapes, such as a pear or T shape. We tested MAARS with long cells (**cdc25**-22) and short cells (**wee1**-50; Nurse, 1975) and obtained very similar data qualities, suggesting that our software is insensitive to cell shape variations (unpublished data).

In this study, we specifically used MAARS to detect spindle poles and kinetochores, but other types of organelles could be detected by simply adjusting MAARS parameters such as number, size, or shape of spots. Of importance, MAARS is delivered with the software Micro-Manager and is freely available (Edelstein et al., 2010). By characterizing a population of 10,000 cells within 5 min, MAARS provides statistical significance comparable to flow cytometry combined with a spatiotemporal analysis of mitotic events. MAARS can extract several parameters from single cells, such as spindle size, orientation, positioning, and elongation, as well as mitotic phase duration. MAARS was tested in various conditions and challenged with established genetic mutants. Our quantitative single-cell analysis revealed an unexpected heterogeneity in cell size, spindle elongation rate, and spindle positioning in **mal3**Δ or **mod20**Δ mutants, which are defective for cytoplasmic microtubule dynamics. Of importance, most of these differences were not detected when average values were measured. These observations illustrate the strength of using MAARS for quantitative analysis of mitosis as opposed to a classic manual phenotypic analysis.

In conclusion, our study illustrates the power of MAARS software to identify new and rare events of chromosome segregation defects in wild-type or mutant conditions.

**DISCUSSION**

In this study, we developed an expert system to automatically analyze mitotic progression. As illustrated in the video tutorial (https://github.com/bnoi/MAARS/tree/master/doc/demos), our software is able to record cell division, specifically find mitotic cells, and extract on-the-fly multiple mitotic parameters. Our results demonstrate that MAARS is practical, time-efficient, and robust software for the multiparametric analysis of mitosis. This software provides statistical analysis at the cell population or single-cell level. Within the entire cell population, MAARS determines the percentage of cells in metaphase, anaphase, or cytokinesis and identifies cell size, spindle pole position, and so on. The results are displayed as a spots distribution, but it is possible to return to the cell image to confirm unexpected results. MAARS can adapt to different cell types (including the budding yeast by adjusting the solidity parameter) and different fluorescent markers (GFP, red fluorescent protein [RFP], or CFP). By simply adjusting the solidity parameter, it is possible to use MAARS for the analysis of cells displaying abnormal cell shapes, such as a pear or T shape. We tested MAARS with long cells (**cdc25**-22) and short cells (**wee1**-50; Nurse, 1975) and obtained very similar data qualities, suggesting that our software is insensitive to cell shape variations (unpublished data).

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In addition to these mitotic parameters, MAARS can also detect automatically the presence of unattached kinetochores (kinetochores that are not present on a line drawn between the two SPBs; unpublished data). This may be extremely useful for studying the mechanisms of kinetochore retrieval in fission yeast (Grishchuk and McIntosh, 2006; Franco et al., 2007; Gachet et al., 2008; Kalinina et al., 2013). Another important defect occurring in mitosis is the...
two distinct spots. Finally, MAARS automatically detects the presence of lagging chromosomes, including stretched merotelic attachment during mitotic progression. Merotelic attachment is the most frequent error during metaphase in eukaryotic cells. Several studies reported the presence of merotelic attachment in mammalian cells released from nocodazole treatment (Cimini et al., 2004) or in S. pombe cohesin mutants (Gregan et al., 2007; Courtheoux et al., 2009; Rumpf et al., 2010), although this attachment defect has not been observed in wild-type conditions. Merotelic attachment leads to intrakinetochore stretching followed by correction in anaphase through spindle elongation mechanical forces (Cimini et al., 2004; Courtheoux et al., 2009). Of note, the presence of merotelic chromosomes linearly antagonizes spindle elongation rate, and its correction in anaphase prevents spindle collapse and cell death (Courtheoux et al., 2009). However, quantitative analysis of merotelic attachment and its correction in live cells is poorly documented. MAARS provides a tool to understand the molecular mechanisms of correction.

Quantitative analysis by MAARS of lagging chromosomes has enabled us to detect rare events of correction during anaphase in wild-type and mutant conditions. For example, the presence of merotelic attachment has never been reported in mod20Δ cells. Similarly, we were also able to visualize rare events of correction of a missegregated kinetochore in late anaphase in wild-type cells, although it is generally admitted that correction only occurs in metaphase or early anaphase. By following single cells during mitotic progression, it will be possible to identify the timing of correction of merotelic attachment. MAARS will thus be a perfect expert system to decipher the mechanisms of merotelic attachment signaling and correction.

With future developments, MAARS software will deliver high-content information for other mitotic parameters such as kinetochore congression at the metaphase plate or kinetochore oscillation (Mary et al., 2015). A combination analysis of these parameters will create new insights into of mitosis. For example, because mitosis in fission yeast implicates almost 500 proteins, a large-data combinational analysis could unravel new functions for known proteins or assign mitotic functions for unknown proteins. By adapting MAARS’s existing segmentation parameters, we will be able to detect and analyze prokaryotic cell division with either single or multiple chromosomes. In the future, it will even be possible to analyze metazoan mitotic progression using spindle poles

appearance of a monopolar spindle, which corresponds to defects in spindle formation (Vardy and Toda, 2000). MAARS can also detect this type of defect as long as SPBs are duplicated and detected as...
MATERIALS AND METHODS

Cell culture
Media, growth, maintenance of strains, and genetic methods were as reported (Moreno et al., 1991). Cells were grown at 25°C in yeast extract before being mounted on an imaging chamber.

All fluorescently tagged strains (GFP, RFP, and CFP) are integrated at their native chromosomal loci and under the control of their endogenous promoters. The different strains used in this study are listed in Supplemental Table S1.

Cell fixation
To determine the percentage of binucleated and septated cells in cdc25-22 experiments, cells were fixed in 3.7% formaldehyde for 7 min at room temperature, washed once in phosphate-buffered saline (PBS), and observed in presence of DAPI/calcofluor.

Live-cell imaging
Live-cell microscopy was performed on an imaging chamber (CoverWell PCI-2.5; Grace Bio-Labs) filled with 1 ml of 2% agarose in minimal medium and sealed with a 22 × 22 mm glass coverslip. The temperature was maintained at 25°C during acquisitions. Images were acquired from an inverted wide-field microscope (Nikon Eclipse TI) equipped with a Neo scientific complementary metal oxide semiconductor camera (2560 × 2160 pixel resolution; Andor Technology), an LED light source (Lumencor Spectra), and either a 60x objective (1.40 numerical aperture [NA]) or a 100x objective (1.45 NA).

Mitotic synchronization
The temperature-sensitive mutant cdc25-22 was synchronized in G2 by incubation at 36°C for 4 h and then released synchronously into mitosis by shifting cells at the permissive temperature of 25°C (>80% of cells were in prometaphase).

Description of MAARS software
MAARS is a Micro-Manager plug-in designed to perform on-the-fly analysis and real-time microscope control or postacquisition analysis. MAARS was written using the Java programming language and uses numerous image-processing libraries, including Fiji (Schindelin et al., 2012), TrackMate (Tinevez et al., 2017), LOCI bioformats (Lichtert et al., 2010), and Micro-Manager (Edelstein et al., 2010). MAARS is licensed under open-source license BSD (Berkeley Software Distribution License). Documentation and source code are available in a public repository (https://github.com/bnoi/MAARS/releases/tag/v1.0.0). The plug-in as a Jar file can be used in ImageJ. We recommend following

and kinetochore markers as long as a suitable segmentation algorithm is implemented. In conclusion, mitosis research will largely benefit from a quantitative analysis and an expert system approach to understand the ballet of chromosome dynamics in time and space.
S. pombe changes from a dark to a bright color at various focal positions. This dark-to-bright intensity variation is very similar to the invert derivative of a Gaussian distribution.

We implemented this method to segment fission yeast cells. This segmentation computes a correlation image from a stack of 34 bright-field images taken with a Z-step of 0.3 μm; 17 images are taken below and the other 17 above the longitudinal axis of the cell. In this three-dimensional stack, every pixel has X-, Y-, and Z-coordinates. For each stack of pixels having the same X- and Y-coordinates, MAARS determines a correlation score between observed and theoretical Gaussian gray value distributions and provides a correlation image transformed in a binary image. Finally, the Particle Analyzer plug-in from Fiji returns polygonal regions of interest (ROIs) as potential “cells.”

**Generation of automated ROIs around cells.** Previously detected ROIs corresponds to single cells but also to false-positive regions (fused cells, intercell space). Before proceeding to mitotic analysis, MAARS rejects unwanted regions based on decision tree filters such as minimum cell size, maximum cell size, ROI solidity, and ROI mean gray value.

**Optimized parallelization.** Parallelization in MAARS ensures correct acquisition timing and fast computation. One thread is dedicated to image acquisition, and other threads are allocated for heavy computation such as segmentation or single-cell analysis.

**Spot detection.** All of the analyses are done in two dimensions after a maximum projection of the stacks. Kinetochore or SPB epitope-tagged strains used in this study appear as bright spots inside cells. MAARS applies a maximum Z-projection of images before spot detection. Then a Laplacian of Gaussian (LoG) detector (from TrackMate Fiji plug-in) detects the spots. To guarantee on-the-fly spot detection with a very low rate of false positives, it is crucial to optimize two parameters: the spot radius and the quality of the spots of interest. Owing to the high variability between biological samples, multiple parameters, including spot intensity, signal/noise ratio, and so on, could vary slightly. Thus users have to test and modify the parameters of spot detection before launching the MAARS analysis process, by using the test button in the Graphical User Interface. These parameters can be adjusted for each experiment to provide high-quality detection for kinetochores and SPBs.

**Mitotic cells and phase detection.** The spindle is defined by the recognition of the two SPBs in a single cell. Metrics related to the distance between the two SPBs or their position within the cell can

![FIGURE 7: Automated detection of rare events of chromosome segregation defects in wild-type cells.](image)

(A) Example of a rare event of segregation defect detected by MAARS in a wild-type cell (n > 8000 mitotic cells filmed). SPBs in red and kinetochores in green. (B) Top, kymograph representation of the cell shown above. Bottom, automated tracking analysis of the cell shown in A. The number of kinetochores at each SPB is indicated for each time point on both sides of the kymograph; zero (yellow), two (purple), three (green), and four (red). Note the presence of four kinetochores at the upper pole and the correction of the defect during anaphase.

precisely the installation instructions (https://github.com/bnoi/MAARS/blob/master/doc/install_MAARS.md). Several tutorials are also provided (https://github.com/bnoi/MAARS/tree/master/doc/demos) to illustrate the simplicity and efficiency of MAARS software.

**Bright-field segmentation.** Segmentation of S. pombe was performed using a correlation imaging technique based on bright-field images taken at various focal positions, as previously described (Julou et al., 2013). The principle of this correlation imaging technique is that under bright-field illumination, the cell wall of
be easily calculated (e.g., spindle length, spindle orientation). Because the spindle forms in prophase and elongates to reach cell ends in late anaphase B, we determine in each cell the mitotic period as the period of time between the minimum and maximum spindle length as judged by the distance between the SPBs.

A first step in the detection of mitosis consists of finding cells with at least two fluorescent signals (e.g., presence of two spindle poles inside a ROI for a minimum of 3 min). However, fused cells presenting a high solidity value can contaminate this detection. Thus a second step consists of a more precise description of the directed movement of the two spindle poles from prophase to anaphase B (i.e., corresponding to spindle poles moving away from each other during spindle elongation). Although the rate of spindle elongation differs from prophase to anaphase B, we assume that the entire elongation process is roughly linear. Building a model of spindle elongation and finding the best parameters for each strain to detect mitosis is not the optimum way to identify every mitotic cell. Indeed, the detection process may ignore mitotic cells with an abnormal fluorescent signal. In addition, the way in which the spindle elongates may vary from strain to strain. To circumvent this problem, we chose to use a minimum number of parameters to detect mitotic cells by testing the linearity of the elongation curves. For this, we segmented spindle elongation trajectories into a region between the minimum distance between the SPBs and the maximum distance between the SPBs. Then, the program applies a line-fit function on the elongation curve and determines its $p$ value. A threshold of $p$ value resulting in fewer mistakes is used to specifically select mitotic cells. In practice, we chose to use a higher threshold than the optimum $p$ value (i.e., $10^{-11}$ vs. $10^{-13}$) to decrease false-negative detection and accept more false-positive cells, which can be easily eliminated afterward (Supplemental Figure S4). Supplemental Figure S4 shows the total number of errors measured as a function of this threshold.

Virtual cell representation. For the virtual cell representation, the positions of the SPBs are projected into a single virtual cell in order to describe spindle positioning and orientation in a population of cells. However, cells are filmed in random positions, and the coordinates of the cell centers are different. It is thus required to perform both a rotation and a translation of the SPB coordinates. Cell rotation is performed by multiplying all spot coordinates with the corresponding rotation matrix using the angle of the cell major axis. Then cell translation is obtained by subtracting the cell center's coordinates. Thus the reference of each set of cell coordinates is zero.

Determination of slope changes during spindle elongation. Slope changes in the rate of spindle elongation can be used to detect prophase or anaphase B onset. It also can be used to detect the presence of merotelic attachment in anaphase because spindle elongation is reduced in their presence. To locate such changes, we analyze spindle elongation in sliding 3-min windows throughout mitotic progression. Anaphase B onset corresponds to the time point having the maximum slope change among those 3-min windows. This method can also be used to find prophase onset. However, in mutant cells, especially in the case of the rad21-K1 mutant, for which the slope change at anaphase onset is less brutal, anaphase B onset is determined with less accuracy ($\pm 1$ min). In such situations, manually annotating anaphase onset is highly recommended before proceeding to further analysis. Once the annotated elongation curves are aligned according to the onset of anaphase, the program calculates spindle elongation rates before or after this time-point (corresponds to pro/metaphase or anaphase cells). The speed in each phase is shown as a mean value.

Manual annotation. The entire automated detection is not error-free. Thus several semiautomated manual efforts are required. The first type of error is the presence of telophase cells within the analyzed population. Indeed, the two SPBs of cells in telophase can move apart by chance or due to problems of spot detection. Because MAARS saves the cropped images of each cell, the user can get back to the cropped films and reject unwanted telophase cells.

Lagging-chromosome detection. MAARS searches for cells with spots (kinetochores) located between the two spindle poles during anaphase. It confirms an event of anaphase chromosome lagging when a kinetochore between SPBs is detected after anaphase B onset. The corresponding cell appears in a pop-up window at the end of the analysis to allow user confirmation of this event.

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