

## Article

## High-Throughput Phenotyping of *Chlamydomonas* Swimming Mutants Based on Nanoscale Video Analysis

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**ABSTRACT** Studies on biflagellated algae *Chlamydomonas reinhardtii* mutants have resulted in significant contributions to our understanding of the functions of cilia/flagella components. However, visual inspection conducted under a microscope to screen and classify *Chlamydomonas* swimming requires considerable time, effort, and experience. In addition, it is likely that identification of mutants by this screening is biased toward individual cells with severe swimming defects, and mutants that swim slightly more slowly than wild-type cells may be missed by these screening methods. To systematically screen *Chlamydomonas* swimming mutants, we have here developed the cell-locating-with-nanoscale-accuracy (CLONA) method to identify the cell position to within 10-nm precision through the analysis of high-speed video images. Instead of analyzing the shape of the flagella, which is not always visible in images, we determine the position of *Chlamydomonas* cell bodies by determining the cross-correlation between a reference image and the image of the cell. From these positions, various parameters related to swimming, such as velocity and beat frequency, can be accurately estimated for each beat cycle. In the examination of wild-type and seven dynein arm mutants of *Chlamydomonas*, we found characteristic clustering on scatter plots of beat frequency versus swimming velocity. Using the CLONA method, we have screened 38 *Chlamydomonas* strains and detected believed-novel motility-deficient mutants that would be missed by visual screening. This CLONA method can automate the screening for mutants of *Chlamydomonas* and contribute to the elucidation of the functions of motility-associated proteins.

### INTRODUCTION

Cilia and flagella are cellular organelles that beat to move the surrounding fluid or to propel the cell itself. These movements play a fundamental role in various physiological functions, such as embryonic development (1), clearing mucus from airways (2), the immune system (3), and the reproductive system (4). Malfunctions of cilia/flagella cause various ciliopathies (5): hydrocephalus (6), juvenile myoclonic epilepsy (7), retinal pigmentosa (8), situs inversus (a left-right asymmetry defect) (1), and polycystic kidney disease (9). Therefore, understanding the mechanisms of cilia and flagella function are important for advancing our knowledge in biology as well as making advances in medicine.

At the core of the cilia and flagella is a cytoskeletal structure called an axoneme, which consists of nine doublet microtubules cross-bridged with dynein motors and a central pair of microtubules (9+2 structure). The bending of the axoneme is driven by dynein motors, which convert chemical energy derived from ATP hydrolysis to sliding movements between doublet microtubules. Despite its compact appearance, the axoneme is a highly complex structure composed of hundreds of proteins (10–12). Despite numerous studies that have been published on these

organelles, functions of many of the flagella components remain unknown.

The eukaryotic unicellular algae *Chlamydomonas reinhardtii* has been one of the most widely used model organisms to elucidate the function of flagella components for several reasons:

1. Flagella mutants can be easily screened by visual selection of slow swimming cells (13),
2. A wide variety of genetic tools are available, and
3. Biochemical and structural studies can be performed on the flagella.

However, identifying new flagella mutants is becoming difficult, partly because the genes involved in slow swimming mutants are saturated. Therefore, new methods for screening more subtle changes in swimming patterns are required.

In this study, we developed a method that can localize the position of each cell to within 10-nm precision from high-speed video images (1200 fps). The position of the cell in each frame was determined by the cross-correlation of the cell image in each frame and the reference image. By using this cell-locating-with-nanoscale-accuracy (CLONA) method, we obtained detailed information on the parameters related to swimming for wild-type *Chlamydomonas*, as well as for seven dynein mutants: four mutants lacking parts of the outer dynein arm, and three mutants lacking parts of the inner

Submitted April 17, 2014, and accepted for publication May 29, 2014.

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Editor: Charles Wolgemuth.

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0006-3495/14/07/0336/10 \$2.00

<http://dx.doi.org/10.1016/j.bpj.2014.05.033>



dynein arm. To show the effectiveness of the CLONA method in screening for motility-deficient mutants, we analyzed 38 *Chlamydomonas* strains and detected believed-novel mutants that showed subtle deficiency in the motility. These results demonstrate that precise analysis of seemingly simple movies of moving objects, such as flagella-driven cells, provides valuable information about the objects.

## MATERIALS AND METHODS

### Chlamydomonas mutants

*Chlamydomonas* flagella mutants *oda1*, *oda11*, *sup<sub>pp1</sub>*, *sup<sub>pp2</sub>*, *ida1*, *ida5*, and *ida9* were kindly provided by Dr. R. Kamiya (The University of Tokyo and Gakushuin University). The general and specific locations of the mutations and the affected genes in each mutant are listed in Table 1. Each *Chlamydomonas* strain was cultured in 250 mL Tris-acetate-phosphate medium with aeration at 25°C (14). To reduce the variations caused by different phases of the cell cycles, the circadian phases of cells were synchronized in a light-controlled room under a 12-h light: 12-h dark cycle and all the observations were carried out between 8 and 10 a.m.

The 38 *Chlamydomonas* strains used for the screening trial were prepared in the previous study (15). Each strain was cultured in a six-well plate using a shaking incubator at 25°C, under a 12-h light/12-h dark cycle. All videos in this study were recorded at 25°C.

### Video microscopy

High-speed cameras, EX-F1 (Casio, Tokyo, Japan) or EoSens MC1362 (Mikrotron, Unterschleissheim, Germany), were connected to a dark-field light microscope (BX51; Olympus, Tokyo, Japan) equipped with a 20× objective lens (UPLFLN 20×; Olympus, Tokyo, Japan) and a mercury lamp (BH2-RFL-T3; Olympus). For each video, a 30-μL quantity from a cell suspension was pipetted in a chamber closed on either side by a glass slide (S1214; Matsunami, Osaka, Japan) and a 18 × 18 mm cover glass (C218181; Matsunami), and sealed with Vaseline (Unilever, London, UK). The distance between the glass slide and cover glass was estimated to be ~180 μm (30 μL volume/154 mm<sup>2</sup> area). To avoid overlaps between cell images and cell-cell collisions, the concentration of the cells was adjusted to <10<sup>5</sup> cells/mL. Videos were recorded at 1200 fps with 336 × 96 pixel field size or at 600 fps with 432 × 192 pixel field size. The pixel size is 0.56 μm at the specimen. The recorded videos were transferred to an iMac computer (Apple, Cupertino, CA), and cells were tracked using a multiple-particle tracker plug-in (MTrack2) for the software ImageJ (16). The cells in each frame were detected, and then tracked by determining which cells in the successive frames were the closest. The location for each cell was roughly calculated using the centroid of the binary image. Further analysis was done using custom Ruby-Helix scripts (17). The software R (18) was used for plotting the results and statistical analysis. Statistical significances between different strains (see Fig. 4, Fig. 5 C, and those mutants categorized into the third group) were tested using the *F* test of Wilks's lambda.

**TABLE 1** *Chlamydomonas* flagella mutants used in this study

Type	Strain	Affected components	References
Outer dynein arm mutants	<i>oda1</i>	Outer arm, ODA-DC (deleted)	(27–29)
	<i>oda11</i>	αHC+LC5+LIS1 (deleted)	(30,31)
	<i>sup<sub>pp1</sub></i>	βHC (truncated)	(32,33)
	<i>sup<sub>pp2</sub></i>	γHC (truncated)	(32,34)
Inner dynein arm mutants	<i>ida1</i>	<i>f</i> (deleted)	(35)
	<i>ida5</i>	<i>a</i> , <i>c</i> , <i>d</i> , <i>e</i> (deleted)	(36,37)
	<i>ida9</i>	<i>c</i> (deleted)	(38)

### Selecting target cells

High-speed video of swimming *Chlamydomonas* cells was recorded using a dark-field light microscope and the recorded videos were analyzed to select cells that were in focus and swimming parallel to the glass slide (Fig. 1 A).

The cells were classified based on their swimming path (Fig. 1 B). We selected cells that swam long distance in the *XY* plane (Fig. 1 B, *a* and *c*) and excluded the cells that swam in the perpendicular direction, wiggled, or were stationary (Fig. 1 B, *b* and *d*, respectively). Next, cells that were in the in-focus zone, which has a depth of ~4 μm (Fig. 1 B, *a*), were identified based on the focus index, *F*, which reflects high-frequency components in the image (Fig. 1 C).

We introduced the index to measure the quality of focus of each image. The index was calculated as follows: from each frame, a 32 × 32 pixel region centered on the centroid of the cell binary image was selected. The selected image was transformed to a frequency space by fast Fourier transform (19). For each image, a score *S<sub>i,j</sub>* (score of frame *j* of cell *i*; see Fig. 1 C) was calculated based on the amplitude of high frequency components in the Fourier transform of the image. In-focus images have higher scores, due to the amplitude of the high frequency component being relatively higher than that of poorly focused images (20). This scoring was applied to all frames, and the focus index of cell *i* or (*F<sub>i</sub>*) was determined as

$$F_i = \max_j \left\{ \min_{k=j \dots j+t} \{S_{i,k}\} \right\}, \quad (1)$$

where *t* is a given time window and *S<sub>i,j</sub>* is the score of frame *j* of cell *i*. Here, the theory of Minimax was used to guarantee a certain level of focusing within the given time window *t* (*t* = 50 frames in this study). Selecting image sequences with a threshold of *F* > 90 excludes cells that swam outside the in-focus zone (Fig. 1, B and C).

### Procedures for position determination

The position of the cell in each frame was determined by maximizing cross-correlation between the image of cell in each frame and the reference image constructed from the cell itself (Fig. 1 A). The procedure consists of two steps:

1. Construction of the reference image (see Fig. S1 in the Supporting Material), and
2. Taking cross-correlation between the image of the cell in each frame and the reference (lower-half of Fig. 1 A).

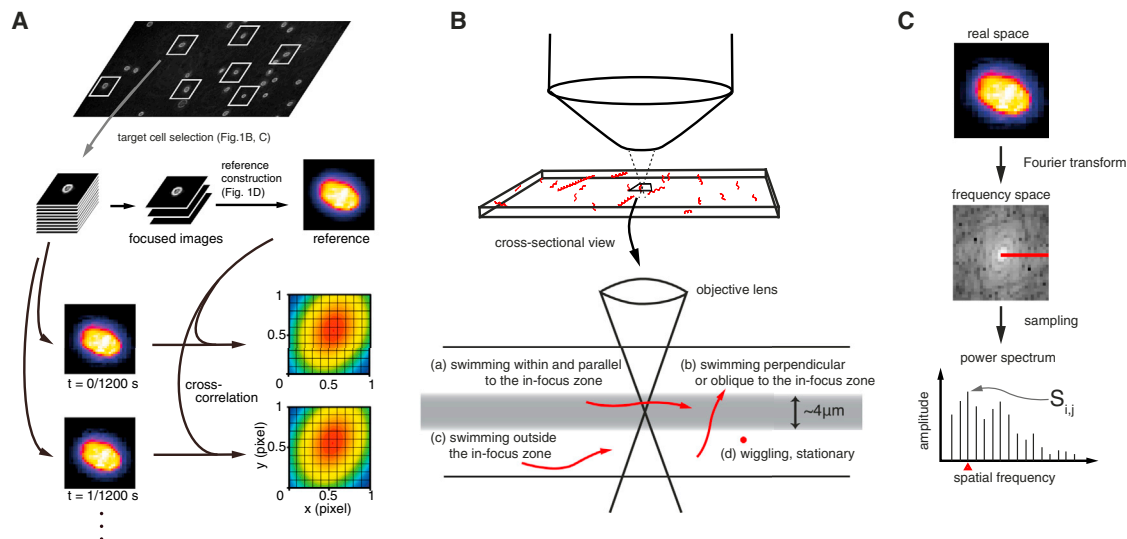
To construct a reference, the best 30 in-focus images were selected. Each image was recentered using a two-dimensional Gaussian fit (21), and these recentered images were averaged using subpixel location and rotation angle based on bicubic interpolation (22) to produce the initial reference image (see Fig. S1). Taking this initial reference image as the reference, each cropped image was again translationally and rotationally aligned using cross-correlation (19). The final reference image was obtained by iterating this procedure two-to-three times using the preceding reference image as a reference.

The cross-correlation between the reference image and the cell image in each frame was calculated using a fast-Fourier transform-based procedure, and the position and rotation angle that give the maximum correlation coefficient were obtained (lower-half of Fig. 1 A).

## RESULTS

### Precision of cell localization

To estimate the precision of localizing the cell, we applied the CLONA method to stationary cells. Wild-type *Chlamydomonas* cells were fixed using 5% formaldehyde and



**FIGURE 1** Processes of locating cellular size objects at high spatial and temporal resolution. (A) Overview of the localizing method. The recorded videos were first analyzed to identify target cells. The position of the cell in each frame was determined by cross-correlation between the cell image in each frame and the reference image. (B) Schematic representation of the four types of swimming paths. (Red solid lines) The possible swimming paths of *Chlamydomonas* cells: (a) swimming within and parallel to the in-focus zone, (b) swimming perpendicular or oblique to the  $XY$  plane, (c) swimming outside of the in-focus zone, and (d) wiggling within the trajectory or stationary. Cells that swim parallel to the glass slide and that have an in-focus image were selected for further analysis. (C) The definition of the score  $S_{i,j}$ . The image of cell  $i$  in frame  $j$  was transformed into frequency space, and the values at the horizontal axis were sampled to produce a power spectrum. The amplitude of the frequency component at  $32/3$  pixel  $\approx 6 \mu\text{m}$  gives  $S_{i,j}$ .

mounted to the slide. Video of these nonmotile cells was recorded at 1200 fps with a pixel size of  $0.56 \mu\text{m}$ , and the position of each cell was calculated by the CLONA method. Fig. 2 A shows a scatter plot with marginal histograms of the stationary cell over 2000 frames. The red solid line is a fitted normal distribution, with  $\sigma_x = 6.5 \text{ nm}$  and  $\sigma_y = 7.4 \text{ nm}$ . The mean  $\pm$  standard deviation of the positions of stationary cells was  $8.0 \text{ nm}$  ( $n = 16$ ). This standard deviation of the obtained coordinate values indicates that the location can be obtained with precision of  $\approx 10 \text{ nm}$  using this method.

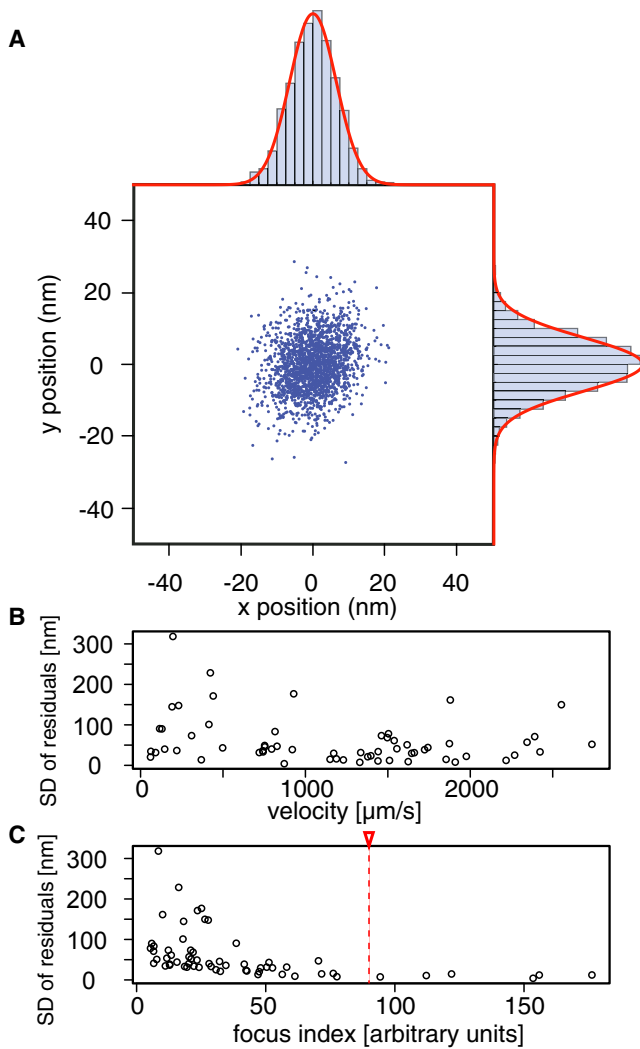
We next evaluated the precision of the CLONA method when applied to moving cells. Because the images of moving cells can be blurred, we were concerned that the precision would be lowered. To test the effect of this movement on image clarity, wild-type *Chlamydomonas* cells were formalin-fixed and placed in a flow chamber (well width,  $2 \text{ mm}$ ; well depth,  $0.080 \text{ mm}$ ). We assumed that the velocity of the cells was constant over  $5 \text{ ms}$ , and the residuals between the linear regression and the position determined by the CLONA method were used to evaluate the precision. The standard deviation of the residuals, which indicates the precision, was plotted against the velocity of the cells (Fig. 2 B). The precision was found to vary among cells but did not depend on the velocity. On the other hand, plotting the standard deviation of the residuals against the focus index showed a clear inverse relation (Fig. 2 C): when the focus index is low, the standard deviation of the residuals varies greatly, showing that some cells were localized with high precision whereas others were not. When the

focus index was high, all the cells were localized with high precision. These results show that the precision of the CLONA method is dependent mostly on whether the images are well focused. Therefore, the CLONA method can locate the cells within  $10\text{-nm}$  precision, irrespective of the velocity, if cells with a focus index  $>90$  are selected (Fig. 2 C). As shown below, the maximum velocity of *Chlamydomonas* swimming is  $\sim 1000 \mu\text{m/s}$  and, thus, error associated with image blur can be neglected for analysis of the swimming of *Chlamydomonas*.

### Application of the CLONA method to wild-type *Chlamydomonas* cells

High-speed video images of swimming wild-type *Chlamydomonas* cells, CC125<sup>+</sup>, were analyzed using the CLONA method. Fig. 3, A and B, shows the typical swimming paths of a wild-type cell (121 frames, each time step is  $1/1200 \text{ s} \approx 0.83 \text{ ms}$ ; see Movie S1 in the Supporting Material) determined by the CLONA method (Fig. 3 A), and by the conventional binary-centroid method (Fig. 3 B). Compared to the path determined by binary-centroid method (Fig. 3 B), the smooth curve in Fig. 3 A indicates the accuracy of the CLONA method in tracing the swimming path of the *Chlamydomonas* cell.

Using the accurate positions of the cells, we determined the swimming velocity based on the movement between two frames (Fig. 3, C and D). The shape of the swimming velocity curves was surprisingly constant among strokes (Fig. 3 D) and among cells (Fig. 3 E). Note that the velocity



**FIGURE 2** Precision of determining the cell position using the CLONA method. (A) Position of stationary cells for 2000 frames calculated using the CLONA method. The average position of the 2000 frames was used as the origin of coordinate system. (Red solid line) Fitted normal distribution, with  $\sigma_x = 6.5$  and  $\sigma_y = 7.4$  nm. Pixel size of the original images: 560 nm. (B) The standard deviation of residuals plotted against the velocity of moving cells. The precision varies among cells but does not depend on velocity. (C) The standard deviation of residuals plotted against the focus index. The two variables show a clear inverse relation. For a high focus index, all the targets were located in high-precision region, independent of the velocity. The precision of the CLONA method can be  $<10$  nm, independent of the velocity if cells have a focus index  $>90$  (red dotted line).

curves were obtained from a single beat without applying any filters. Although movements perpendicular to the XY plane were not determined, we estimated the contribution of perpendicular movements to be  $<10\%$  because we selected cells that swam parallel to the XY plane using the focus index criterion (see Fig. S2). The average time and distance traveled through the in-focus zone was 58 ms and  $12 \mu\text{m}$ .

Fig. 3 C illustrates the relation between the swimming velocity parallel to the swimming direction and the position of the flagellum (the red dotted line represents the mean

swimming velocity; see also Movie S2). The *Chlamydomonas* cell accelerates as the flagellum bends at the base and the distal part straightens (Fig. 3 C, index = 1–5). The velocity reaches a maximum ( $1.05 \times 10^3 \mu\text{m/s}$ ) when the flagellum straightens vertically in the swimming direction (Fig. 3 C, index = 6). The cell gradually decelerates as the bending of the flagellum slows down (Fig. 3 C, index 7–10). When the flagellum reverses the bending direction at the base (Fig. 3 C, index = 10–13), the cell went backward despite that the distal half of the flagellum was still moving backward. The velocity curve of the recovery stroke has a dull phase (index = 10–17); that is, when the base of the flagellum starts to bend forward while the distal half is still moving backward, the forward propulsive force is canceled.

To quantitatively describe the characteristics of the swimming velocity curve, i.e., the period of the effective stroke and the recovery stroke ( $t_E$  and  $t_R$ , respectively) and the distance through which each stroke travels ( $d_E$  and  $d_R$ , respectively), we calculated the mean swimming velocity,

$$\bar{V} = \frac{d_E + d_R}{t_E + t_R},$$

and flagella beat frequency,

$$f = \frac{1}{t_E + t_R}$$

for each beat cycle (Fig. 3 D). Scatter plots that were generated using these parameters (665 strokes, 167 wild-type cells, 600 fps) showed a cluster in each plot (Fig. 3 E). Each point represents one beat cycle in each two-dimensional space. The red points represent the beat cycles of a single typical wild-type cell. The spread of the clusters, which show the variance of the swimming parameters, is due to inter- and intracell variances. Intercell variance is the variance among different cells, whereas intracell variance is the variance among different beat cycles by the same cell. As shown in Fig. 3 E, the cluster of the red points was more dense than the actual black clouds in the plot, indicating that variances among cells contribute greatly to the overall variance. Regarding beat frequency,  $f$ , the variance among cells (intercell variance) was 8.9%, whereas that within cells for different beat cycles (intracell variance) was 2.9%. Brokaw and Luck (23) showed the variance of the beat frequency of the entire population to be 15.4% ( $65 \pm 10$  Hz,  $20^\circ\text{C}$ ), whereas our study showed that beat cycles are fairly uniform for individual cells and that the variance among cells dominates the overall variance.

### Application of the CLONA method to flagella mutants

To demonstrate the ability of CLONA method to distinguish motility-deficient mutants, we have analyzed high-speed video images of seven flagella mutants (Table 1). The scatter



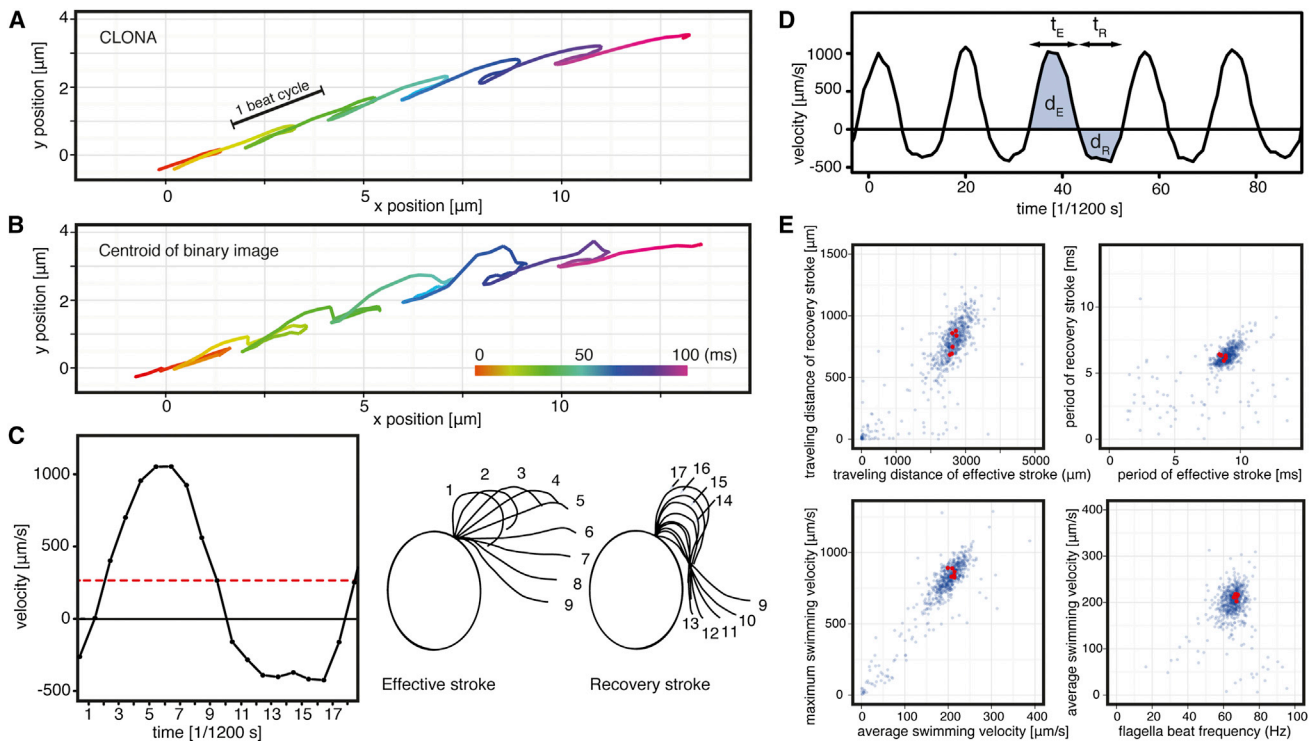


FIGURE 3 Application of the CLONA method to high-speed video images of wild-type *Chlamydomonas* cells (see [Movie S1](#) in the [Supporting Material](#)). Swimming paths of a typical wild-type *Chlamydomonas* cell determined by (A) the CLONA methods and (B) conventional binary-centroid method. The position in each time frame is indicated by a dot and the frame interval is  $1/1200\text{ s} \approx 0.83\text{ ms}$ . The smooth curve indicates the better accuracy of the CLONA method in tracing the swimming path of *Chlamydomonas* cells. (C) The relation between the instantaneous swimming velocity parallel to the swimming direction and the flagellar beat stage obtained from [Movie S2](#). The curve was obtained from a single beat without applying any filters. The mean swimming velocity is indicated by the red dotted line. (D) A typical plot of instantaneous swimming velocity. The following swimming parameters obtained from each beat cycle are shown: period of the effective and recovery strokes ( $t_E$  and  $t_R$ , respectively); travel distance of each stroke ( $d_E$  and  $d_R$ , respectively). (E) Scatter plots of wild-type beat cycles. Each point represents one beat cycle in each two-dimensional space. (Red points) Beat cycles of a single typical wild-type cell.

plots of beat frequency,  $f$ , versus mean swimming velocity,  $\bar{V}$ , of the beat cycles of four outer dynein arm mutants (*oda1*, *oda11*, *sup<sub>pf</sub>1*, and *sup<sub>pf</sub>2*) are shown in [Fig. 4 A](#), and those for the inner dynein arm mutants (*ida1*, *ida5*, and *ida9*) are shown in [Fig. 4 B](#) (24). The beat cycles of each strain formed a characteristic cluster in the scatter plot and the difference between the strains are statistically significant (assuming Wilks's lambda distribution,  $p > 0.001$ ). The swimming speeds of these outer dynein arm mutants are proportional to the beat frequencies, consistent with the idea that the waveform of flagella is similar to that of wild-type (25,26). On the other hand, *ida1* and *ida9* cells swam slower, although the beat frequencies are in the same range as those of wild-type cells. This was also consistent with the idea that the beat frequencies of the inner dynein arm mutants are at almost normal value, but with reduced bend angle (25,26). Note that the swimming patterns of *oda11* (missing  $\alpha$ -heavy chain of outer dynein arm) are known to be similar to that of wild-type and difficult to distinguish by visual screening. Using the CLONA method, these mutants are distinguished in the scatter plot, and subtle changes in swimming behavior can be detected. Details about the mutant strains in the table are described in the literature (27–38).

### Application of the CLONA method to flagella mutant screening

To test the effectiveness of the CLONA method in screening for motility-deficient mutants, we have examined 38 *Chlamydomonas* strains that had been previously isolated as circadian mutants (15). These mutants are insertional mutants with a CBR genetic background and screened by defective circadian rhythms. The responsible gene for each strain has been identified, but their swimming phenotypes had not been examined. For each strain, 150–200 s of high-speed video was taken and its swimming was phenotyped using the CLONA method ([Fig. 5](#), and see [Fig. S3](#)). Note that the swimming phenotype of CBR<sup>+</sup> (39) is different from CC125<sup>+</sup> in [Figs. 3](#) and [4](#). Roughly, the swimming phenotypes of these strains were divided into three groups:

1. The first group shows no or little deficiency in motility ([Fig. 5 A](#)).
2. This group (*roc30*, *roc34*, *roc54*, *roc63*, and *roc106*) shows severe deficiency in motility ([Fig. 5 B](#)).
3. This group (*roc23*, *roc77*, *roc78*, *roc88*, *roc105*, and *roc112*) shows mild deficiency in motility ([Fig. 5 C](#)).

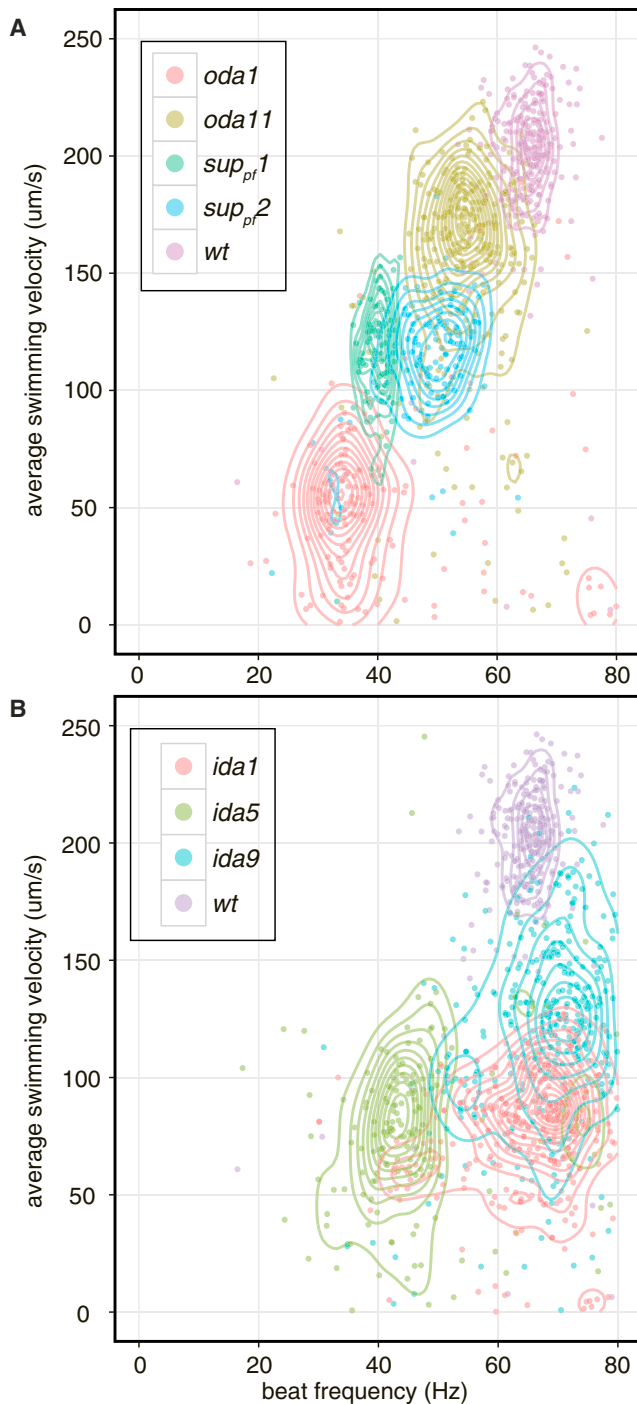


FIGURE 4 Phenotyping *Chlamydomonas* mutants based on the CLONA method. Scatter plots of beat frequency  $f$  versus average swimming velocity  $\bar{V}$  of the beat cycles, generated using the software R (18). Each dot represents one beat cycle, and mutant and wild-type strains are shown with distinct colors. Descriptions of the flagella mutants used here are shown in Table 1. The contour lines were generated to aid with interpreting the distribution of the dots using STAT\_DENSITY2D of the GGLOT2 software package (24), which estimates two-dimensional kernel densities. The difference between the strains are statistically significant ( $p < 0.001$ ). (A) Scatter plot of beat cycles of four outer dynein arm mutants (*oda1*, *oda11*, *sup<sub>pt1</sub>*, and *sup<sub>pt2</sub>*) and wild-type. The beat cycles of wild-type and mutants formed a cluster around each characteristic area in the scatter

See also Movie S3 and Movie S4 for high-speed videos of the control strain and a representative strain of the third group. All the responsible genes of the second group have been previously reported to be flagella-related, whereas those of the third group were newly detected as motility-deficient. We also applied statistical analysis to the third group and showed that the difference between these mutant strains and wild-type were statistically significant ( $p < 0.001$ ).

## DISCUSSION

Our CLONA method can determine the position of cell-sized objects (diameter,  $10 \mu\text{m}$ ; pixel size,  $0.56 \mu\text{m}$ ) with an accuracy of  $\sim 10 \text{ nm}$ , and this method provides valuable information about flagella motility. By applying the CLONA method to high-speed videos (600 or 1200 fps) of *Chlamydomonas* cells exhibiting swimming behavior, parameters related to swimming can be obtained with high spatial and temporal resolution for each individual cell. This application provides new ways to quantify aspects of *Chlamydomonas* motility that were difficult to observe without high spatiotemporal analysis: maximum swimming velocity, ratio of mean and maximum swimming velocity, variance among cells, and variance among beat cycles of the same cell. Analysis of seven *Chlamydomonas* flagella strains demonstrated that subtle changes in motility can be detected using the CLONA method, and the screening of 38 *Chlamydomonas* strains demonstrated that selection of mutants can be quickly and easily accomplished.

### Precision of cell localization

For superresolution methods, such as stochastic optical reconstruction microscopy (40), one of the key factors in achieving high spatial resolution is to collect enough photons from the target object. Our CLONA method collects many photons diffracted from large objects, and achieves both high temporal (1 ms) and spatial ( $\sim 10 \text{ nm}$ ) resolution. One of the main factors for achieving temporal resolution is to collect many photons diffracted by a large object. According to our estimate,  $> 10^5$  photons were collected from a *Chlamydomonas* cell in 1 ms and were used in the localization of the cell. Therefore, if the same number of photons were collected from a molecule, precision of a few nanometers can be achieved according to the following equation (21),

$$\sigma_{u_i} = \sqrt{\frac{s^2}{N} + \frac{a^2/12}{N} + \frac{8\pi s^2 b^2}{a^2 N^2}}, \quad (2)$$

where  $s$  is the standard deviation of the point spread function,  $a$  is the pixel size of the detector,  $N$  is the number

plot. (B) Scatter plot of the beat cycles of three inner dynein arm mutants (*ida1*, *ida5*, and *ida9*) and the beat cycles of wild-type.

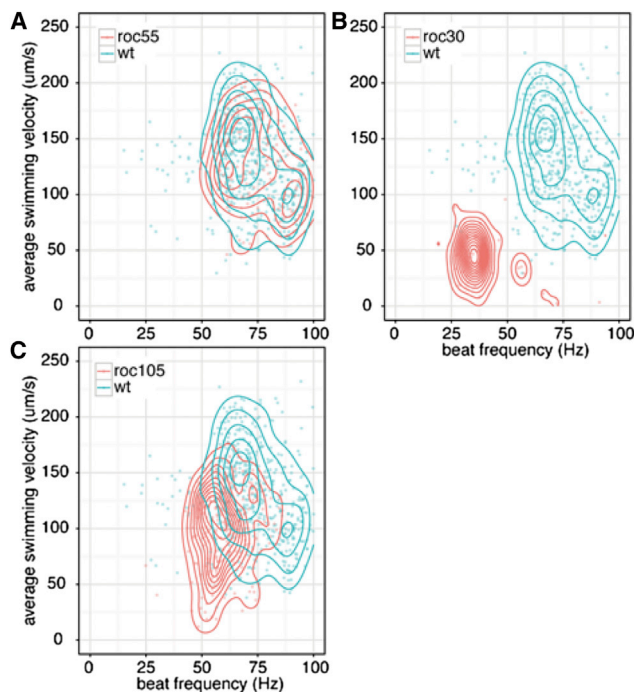


FIGURE 5 Screening of 38 *Chlamydomonas* strains for motility-deficient mutants using the CLONA method. Scatter plots of beat frequency  $f$  versus average swimming velocity  $\bar{V}$  of the beat cycles. (Blue contour lines) The characteristic cluster of the control. The difference between *roc30* and wild-type and the difference between *roc105* and wild-type are statistically significant (assuming Wilks's lambda distribution,  $p < 0.001$ ). (Each dot) One beat cycle. Focus index  $> 80$ . (A) A scatter plot of *roc55*, a representative strain showing no deficiency in motility. (B) A scatter plot of *roc30* (*ODA4*), a representative strain showing severe defects in motility. (C) A scatter plot of *roc105*, a representative strain of newly detected motility deficient mutants, showing mild deficiency in average swimming speed and beat frequency.

of total collected photons, and  $b$  is background noise. We assumed  $s = 300$  nm,  $a = 1000$  nm, and  $N = 10^5$ . This equation is often used to estimate the resolution of superresolution methods, such as fluorescence imaging with 1-nm accuracy (FIONA) (41) or stochastic optical reconstruction microscopy, and based on the assumption that the point-spread function can be approximated by the Gaussian distribution and that the photon shot noise is the limiting factor for determining precision.

The image of a *Chlamydomonas* cell, of course, cannot be approximated by Gaussian distribution, because the cell is larger than the point-spread function of the microscope. To overcome this problem, another key technique of our CLONA method is that the reference images used for cross-correlation are generated from experimental images by averaging well-focused images of the target cell. For the determination of the position of the cell in Fig. 2 A using the Gaussian distribution as a reference, the standard deviation is 2.5 times larger than for the CLONA method, showing the effectiveness of using the averaged image. Although *Chlamydomonas* cells rotate while swimming, the shape of

the cells seems unchanged probably because the cell shape is symmetric around the swimming axis. Therefore, the rotation of the cell would not largely affect the accuracy of the CLONA method. It is noteworthy that the standard deviation of the cell position doubles when the image in the neighboring frame is used as the reference, indicating that having a high signal/noise of the reference image is important. In conclusion, the use of large cell images and having a high signal/noise image on average are the two key factors for the success of the CLONA method.

### Analysis of propulsive force generated by free-swimming *Chlamydomonas* cells

Because we determined the velocity of swimming cells to a high degree of accuracy, we can now translate the velocity into the drag force acting on the cell body, which balances the propulsive force produced by the two flagella. The viscous force on the cell body can be estimated from the formulas describing the motion of an ellipsoid in Stokes' flow. *Chlamydomonas* cells can be considered to be well described by Stokes' Law because the Reynolds number ( $Re$ ) is very small ( $Re \approx 10^{-3}$  to  $10^{-2}$ ) (42)). Chwang and Wu (43) presented the relationship between velocity and viscous drag force on a prolate spheroid moving in the direction of the major axis with a very low Reynolds numbers to be

$$F = 6\pi\mu avC_F, \quad (3)$$

$$C_F = \frac{8}{3}e^3 \left[ -2e + (1 + e^2) \ln \frac{1+e}{1-e} \right]^{-1}, \quad (4)$$

$$e = \sqrt{1 - b^2/a^2}, \quad (5)$$

where  $F$  is the drag force,  $\mu$  is the viscosity,  $e$  is the eccentricity,  $a$  is the major axis, and  $b$  is the minor axis. From the equations above, the average of maximum instantaneous propulsive force of *Chlamydomonas* cells was estimated to be  $46.7 \pm 4.8$  pN (assuming that  $v = 802.0 \pm 93.0$   $\mu\text{m/s}$ ,  $\mu = 0.893 \times 10^{-3}$  Pa s,  $a = 4$   $\mu\text{m}$ ,  $b = 3$   $\mu\text{m}$ ).

There are several ways to estimate the force generated by *Chlamydomonas* flagella. One way is to use velocity and the Stokes' formula as described above. Minoura and Kamiya (44) have calculated the propulsive force of a free-swimming *Chlamydomonas*, using the mean swimming velocity, to be 10–12 pN. The values are consistent with our results, because the propulsive force estimated from the mean velocity determined from our measurement was 11.8 pN. Our results show that the instantaneous maximum swimming velocity can be approximately four times the mean velocity, which is consistent with results obtained from measuring flow field around the swimming cell (45). Therefore, the propulsive force estimated from the

instantaneous maximum swimming velocity in this study was 40–48 pN. Racey et al. (46) also examined the velocity of *Chlamydomonas* cells using quasielastic light scattering and cinematographical techniques at 100–200 fps. Because the temporal resolution is limited to 10 ms in their analysis, they obtained the mean forward velocity of  $379 \pm 108 \mu\text{m/s}$ , which is approximately one-half that of the maximum velocity found in our study.

The force generated by flagella can also be measured by an optical trap. McCord et al. (47) determined that a force of  $26.5 \pm 10.4 \text{ pN}$  was required for *Chlamydomonas* to escape from the optical trap. Although the value is smaller than the maximum propulsive force determined in our study ( $46.7 \pm 4.8 \text{ pN}$ ), it seems reasonable because the force required to escape the optical trap needs to be sustained for several milliseconds for a successful escape. Note that the force exerted by optical traps was determined by applying Stokes' equation to dead trapped cells in a medium with a constant flow rate (47). Therefore, our method is a more direct measurement of the propulsion force than these trapping methods.

### New approach toward the elucidation of flagella mechanisms

Our CLONA method provides a quantitative approach for screening flagella mutants. As described by Kamiya (13), motility-deficient *Chlamydomonas* cells have been identified by visual inspection under a microscope, which requires considerable time and effort. This visual inspection method may be biased toward selecting mutants that show substantial decreases in mean swimming velocity and many mutants that exhibit only small changes in swimming velocity might be missed. The CLONA method can select mutants that have subtle changes in mean velocity, or changes in other parameters, such as maximum velocity or the proportion of effective stroke time and recovery stroke time. Selection of mutants using the CLONA method could provide expanded screening capabilities and represents a labor-saving method because of its accurate velocity analysis and automated procedures.

Although several methods have been developed for the detailed analysis of flagella movement, they are not suitable for the screening of new flagella mutants. For example, Brokaw and Luck (23) established waveform analysis using the *uni1* mutant strain. The *uni1* mutant is suitable for conducting extensive waveform analysis because the strain has only one *trans*-flagellum (48) and rotates in place. Recently, a computed flagella waveform analysis was developed for *uni1* cells (49) and used for the high-resolution estimation of the force generated by the flagellum (50). However, this waveform analysis requires selection of double mutant strains, which carry both *uni1* mutation and a target flagella mutation. Therefore, this *uni1*-based waveform analysis is not suitable for screening for new mutants. Micropipettes

have been used to hold *Chlamydomonas* cells and observe the detailed waveform (51,52). Because the micropipette analysis acquires the precise waveform by preventing movement and rotation, it can reveal the complex dynamics exhibited by flagella (53,54). However, capturing cells by micropipette requires a certain amount of time, making it unsuitable for screening a large number of mutants.

One of the important findings of our study is that the position and velocity of *Chlamydomonas* cells accurately reflect the flagella movement. Therefore, even though the shape of the flagella is not always visible in the images due to the rotation of cell, the CLONA method is able to analyze the flagellar phenotypes of the free swimming cells in their natural, unrestrained state.

In the screening using 38 *Chlamydomonas* strains, CLONA detected motility-deficient mutants that would have been difficult to detect previously, just by taking a 150–200 s high-speed video for each strain. We have chosen a set of circadian mutants for this screening because the mutated genes include previously reported flagella genes, which could be used as a positive control (*ODA4*, *PF9*, *DLC7a*, and *FLA14*; see Table S1 in the Supporting Material). CLONA detected all of these mutants in the screening, ensuring the validity of the screening. Within the 38 strains, we have actually detected six mutants that were not known to play a role in motility (*roc23*, *roc77*, *roc78*, *roc88*, *roc105*, and *roc112*; see Table S1 and Fig. S3).

One of these mutants, *roc78* is a kinase, which may modify both circadian rhythm and motility via the signaling pathway. In fact, several studies have shown that there are multiple phosphorylation sites on outer and inner dynein arms (55,56), which suggests that kinases such as *roc78* could modify the activity of dynein and thus affect motility. It is interesting that all of the mutants that showed severe deficiency in motility were previously characterized mutants, whereas all of the mutant that showed mild defects were genes that were not known to play a role in motility. This may suggest that more unidentified motility related gene can be found by screening and examining mutants that show mild defects in motility. Although future studies are required to find how the newly detected genes caused the motility defects, this screening test have demonstrated the strength of CLONA as a mutant screening tool.

### CONCLUSIONS

This study demonstrated an observational method, CLONA, which localizes cells and similarly sized objects with high spatial (<10 nm) and temporal (millisecond scale) resolution. The features of CLONA that distinguish it from other superresolution methods:

1. CLONA generates the reference image from experimental images of the target itself.



2. CLONA can locate target whose size is significantly larger than the precision.
3. It does not require special instruments.

One suitable application of CLONA is analyzing the fluid flow driven by bacterial flagella: fluid flows are often estimated by monitoring the movements of small plastic beads (57). The positions of beads can be determined more accurately by CLONA than by commonly-used binary-centroid procedures. Similarly, the CLONA method may contribute analyzing behavior analysis of model organisms including sperms and evaluation of cilia movement in fallopian tubes or ventricles of brain (58). Hence, the CLONA method can be applicable also to various systems that require high spatiotemporal resolution.

## SUPPORTING MATERIAL

Three figures, one table, and four movies are available at [http://www.biophysj.org/biophysj/supplemental/S0006-3495\(14\)00571-2](http://www.biophysj.org/biophysj/supplemental/S0006-3495(14)00571-2).

We thank all of the members of the Kikkawa Lab, especially Dr. T. Yagi, for his valuable suggestions and critical reading of the manuscript.

This work was supported by the Funding Program for Next Generation World-Leading Researchers (grant No. LS027), the Uehara Memorial Foundation, and the Takeda Science Foundation (all to M.K.).

## SUPPORTING CITATIONS

Ref. (59) appears in the [Supporting Material](#).

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