



# Studying Cell Migration (Random and Wound Healing) Parameters with Imaging and MATLAB Analysis

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#### **Abstract**

Cell migration is an essential biological process for organisms, in processes including embryonic development, immune response, and cancer metastasis. To elucidate the regulatory machinery of this vital process, methods that mimic in vivo migration, including in vitro wound healing assay and random migration assay, are widely used for cell behavior investigation. However, several concerns are raised with traditional cell migration experiment analysis. First, a manually scratched wound often presents irregular edges, causing the speed analysis difficult. Second, only the migration speed of leading cells is considered in the wound healing assay. Here, we provide a reliable analysis method to trace each cell in the time-lapse images, eliminating the concern about wound shape and creating a more comprehensive understanding of cell migration—not only of collective migration speed but also single-cell directionality and coordination between cells.

**Keywords:** Live-cell imaging, Single-cell tracing, Wound healing assay, Random migration

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## **Background**

Cell migration is an individual or collective cell movement that plays an important role in embryonic development (Weijer, 2009), immune response (Luster et al., 2005), and cancer metastasis (Friedl and Glimour, 2009). To investigate these essential physiological phenomena in vitro, wound healing assay or random migration assay are often applied (Jin et al., 2016).

Compared with traditional analyses of wound healing assay focusing on general migration speed, our script mainly focuses on precisely tracking each cell in each timeframe image, leading to speed, directionality, and coordination being analyzed simultaneously. Moreover, we can conquer the difficulties in analyzing the wound healing assay even when the edge of the wound is not smooth.

Since wound healing assay provides cell information like migration speed and directionality, and random migration provides cell coordination information, we applied both assays to elucidate the mechanisms of cell migration thoroughly. We found several pairs of genes and pathways that affect migration speed and cell coordination through these assays after performing *two-hit* inhibition: the knockdown of a migration-related gene using shRNA combined with the addition of an inhibitor of migration-related pathways at the same time. One of our best candidates, STK40 and MAPK, showed a synergistic effect on migration speed. By applying shSTK40 and MAPK inhibitor PD98059, we found that the absence of either of them has a mild effect on cell migration speed, while the speed significantly dropped when both of them were inhibited at the same time.

Our method is a half-automatic way to trace single cells in the time-lapse immunofluorescence images, which provides multiple parameters for a comprehensive analysis of cell migration, including speed, density, directionality, and cell–cell coordination. In sum, our method not only solves the aforementioned concern but provides a fast, cheap, and comprehensive analysis of cell migration.

## Materials and reagents

- 1. 96-well black/clear-bottom plate (Thermo Fisher Scientific, Nunc<sup>TM</sup>, catalog number: 165305)
- 2. Collagen I, rat tail (Thermo Fisher Scientific, Gibco<sup>TM</sup>, catalog number: A1048301)
- 3. Hoechst 33342 (Thermo Fisher Scientific, Invitrogen<sup>TM</sup>, catalog number: H3570)
- 4. HEPES (Thermo Fisher Scientific, Gibco<sup>TM</sup>, catalog number: 15630106)
- 5. Bovine serum albumin (BSA) (BioShop, catalog number: ALB001)
- 6. Recombinant human EGF (PeproTech, catalog number: AF-100-15)
- Human FGF-acidic recombinant protein (Thermo Fisher Scientific, Gibco<sup>TM</sup>, catalog number: PHG0014)
- 8. Heparin sodium salt from porcine intestinal mucosa (Merck, Sigma-Aldrich, catalog number: H3393)
- 9. Trametinib (LC laboratories, catalog number: T-8123)
- 10. PBS, pH 7.4 (Thermo Fisher Scientific, Gibco<sup>TM</sup>, catalog number: 10010023)
- 11. Cell migration supplements (see Recipes)

#### **Cell lines**

SAS cell line was used as indicated. Cells were grown in Dulbecco's modified Eagle medium (DMEM) with 10% fetal bovine serum (FBS), and 1% penicillin and streptomycin.

#### **Recipes**

#### 1. Cell migration supplements

20 mM HEPES 0.1% BSA 5 ng/mL EGF or 25 ng/mL FGF1 10 U heparin



## **Equipment**

- 1. Fluorescence microscope (Nikon Eclipse Ti)
- 2. 96-well scratcher (can be replaced by 200 μL tip or toothpick)

#### **Software**

- 1. MATLAB (MathWorks, <a href="https://www.mathworks.com/products/matlab.html">https://www.mathworks.com/products/matlab.html</a>) with the image processing toolbox (access date, 8/1/2017)
- Nikon NIS-Elements Viewer (Nikon image acquisition software) (access date, 9/21/2017)
- 3. Fiji from ImageJ (open resource software for image analysis) (access date, 8/1/2017)
- 4. Excel (Microsoft) (access date, 8/1/2017)

#### **Procedure**

#### Plate coating and cell seeding

- 1. Add 100  $\mu$ L of collagen I (30  $\mu$ g/mL, diluted in PBS) to the 96-well black/clear-bottom plate and put the plate in a 37 °C, 95% humidity incubator for 30–40 min.
- 2. Rinse the plate with 100 μL of PBS and aspirate the liquid in the plate thoroughly before use.
- 3. After that, plate SAS cells onto the plate. (The cell number should be controlled to 90%–100% confluency on the day performing cell migration.)
- 4. (Optional) Cells were infected with lentiviruses for 24 h and selected with antibiotics for 24–48 h.

#### **Nucleus staining**

Stain the cells with 1  $\mu$ g/mL of Hoechst 33342 at 37 °C for 1 h for SAS cells.

#### Scratching (for wound-healing assay only)

- 1. Change the medium to  $100 \mu L$  of pre-warmed PBS.
- 2. Scratch the cells with a scratcher (move the scratcher back and forth twice). This method is for a full 96-well plate and can be replaced by using toothpicks or 200 μL tips if only part of the 96 well plate is used.
- 3. After wound-scratching, wash the cells with 100 µL of PBS twice to remove cell fragments.

#### **Drug treatment**

Change the medium to serum-free DMEM medium (for SAS cells) containing cell migration supplements and drugs (we used trametinib 100 nM or DMSO).

#### **Imaging**

- Place the cells in an acrylic incubator at 32 °C for SAS cells. The acrylic incubator is a chamber with CO<sub>2</sub> circulation [the temperature is lower than normal incubation (37 °C) to optimize the migration speed for analysis].
- 2. Use Nikon Eclipse Ti microscope for live-cell imaging. Take images every 10 min for a duration of 6–10 h. Our script is designed for pictures taken by 4× objective.



## Data analysis

We process the image data through our MATLAB scripts. The schematic summary of how these images are processed is in Figure 1.

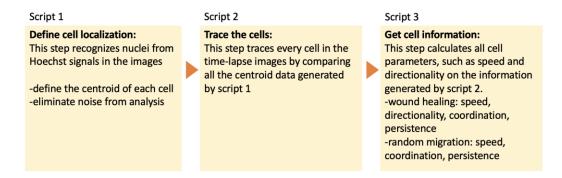
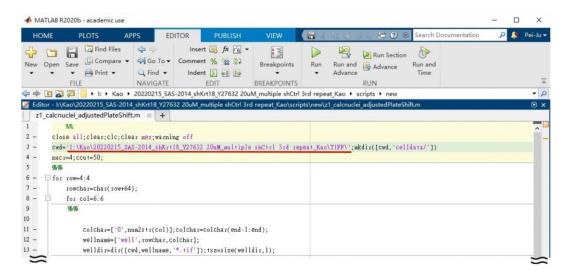


Figure 1. Schematic summary of data processing using MATLAB scripts

#### **Description of MATLAB scripts and analysis**

- Define nuclei location: Z1\_calcnuclei\_adjustedPlateShift
   This script defines the cell location in each image.
  - a. Open MATLAB, import data, and prepare the script.
  - b. Export images to working directory: an individual image is required for DAPI channel and time-lapse.

    Note: The file name should begin with "well + row number + column number" to fit our script. If not, this could be changed in line 13. All images should be exported as tiff files (Figure 2).

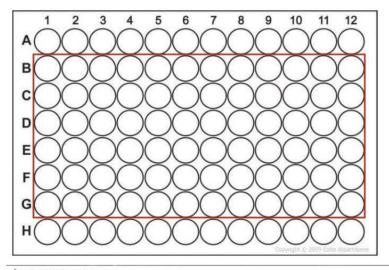


**Figure 2. Screenshot of Z1\_calcnuclei\_adjustedPlateShift user interface.** Underlined is the path of exported tiff files. Note that the image file name in the folder should follow the rule: well + row number + column number + other details, e.g., WellA01\_ChannelDAPI\_Seq0000\_Z01\_C1.tif (L12).

- c. Open Z1\_calcnuclei\_adjustedPlateShift.m.
- d. (Optional) Change the nuclear size in line 4. The nucleus size should be changed based on your cell line (in this case, we use 4 for SAS to best discriminate each cell).



e. Change the file path in line 3 and fill the row and column number based on your experimental setting in line 6 and 8. (In this case, we choose row B–G, and column 1–12, so we fill 2:7 for row and 1:12 for column, respectively.) (Figure 3)



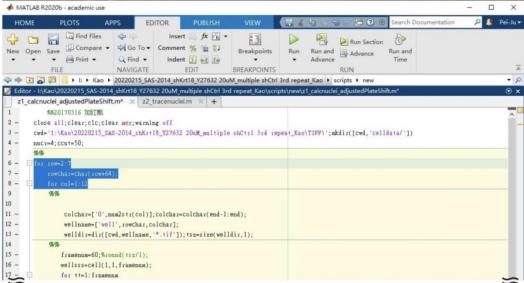


Figure 3. Example of a 96-well plate setting corresponding to the row and column number in Z1\_calcnuclei\_adjustedPlateShift

This script defines nucleus based on a modified threshold selection method. By measuring fluorescence signals (DAPI; Channel 1) to define centroid of every nucleus, we can acquire the location information of each cell from all the acquired images. This allows users to get all the information of cell location.

We first retrieved the centroid of each cell for further calculation of cell movement. The highest signal in the DAPI channel within an estimated distance would be seen as cell centroids, and we labeled these signals as red dots in the image (Figure 4, left panel). To confirm the nucleus definition correctness, we examine the overlap of red dots and green dots (nuclei mask) in the representative figure (Figure 4, right panel). The original Hoechst signal is labeled as red, and our nuclei mask is labeled as green. The nuclei mask is based on Hoechst signal with extra restrictions on size and shape to eliminate noise (such as cell debris) from our analysis. The higher the overlapped ratio of red and green dots, the higher the precision of recognizing the cells. Then, we compare the images between centroids and nuclei masks to double-



check if our centroids are in the right place.

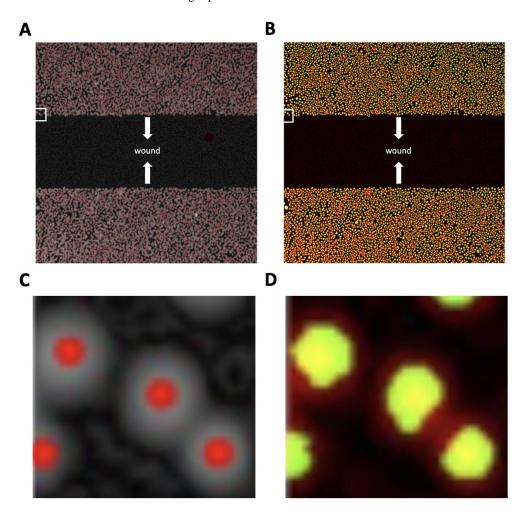


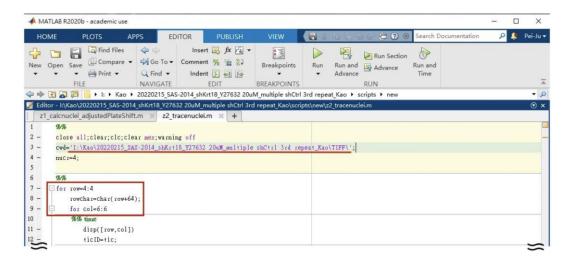
Figure 4. Example of capturing cell nuclei after running Z1\_calcnuclei\_adjustedPlateShift. (A) The red dots are the highest Hoechst signal compared with its surrounding signal as our centroids; this labeled centroid area is magnified in (C). (B) Overlap of the image in the panel with the nuclei mask (green) generated by the processed Hoechst signal and the original Hoechst signal (red), which is magnified in (D). The white arrows in (A) and (B) indicate the direction of cell migration, and the white squares are the cells chosen for magnification.

#### 2. Monitoring cell movement: Z2\_trace nuclei

This script calculates changes of cell location between each time frame.

- a. Open **Z2\_trace nuclei.m**.
- b. Change the file path in line 3 and fill in the row and column number based on your experimental setting in line 7 and 9 (Figure 5).
- c. (Optional) Change the nuclear size (line 4). The nucleus size should be changed based on your cell line (in this case, we use 4 for SAS to best discriminate each cell).
- d. Click Run.





**Figure 5. Screenshot of the Z2\_trace nuclei user interface.** The underlined and circled areas need to be adjusted in each experiment.

3. Get cell information: Z3\_getcellparameters062945

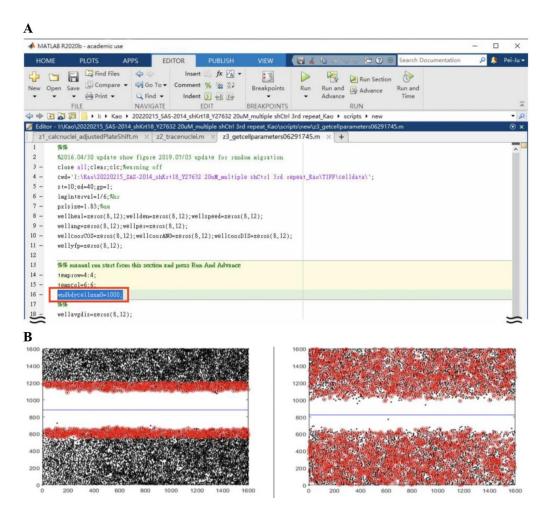
Parameters such as speed, coordination, and directionality were analyzed using MATLAB.

- a. Open **Z3\_getcellparameters062945.m**.
- b. Change the file path in line 3 and fill the row and column number based on your experimental setting in line 7 and 9 (Figure 6A).
- c. (Optional) Change the nuclear size in line 4. The nucleus size should be changed based on your cell line (in this case, we use 4 for SAS to best discriminate each cell).
- d. Click Run.

Define the number of analyzed cells:

a. Fill out the cell number in line 16 that you would like to analyze (Figure 6A). Selected cells should be marked as red. Below is an example of analyzing 1,000 and 10,000 cells (Figure 6B). Note that for random migration, we select all the cells in the frame.





**Figure 6.** Example of different parameter settings for cell numbers used for migration analysis. (A) User interface of Z3\_getcellparameters062945. The circled area defines the number of cells being analyzed in tracking. (B) The left panel shows the analysis of 1,000 cells, and the right panel shows the analysis of 10,000 cells. The black dots represent the cells in the entire image, and the red dots represent the cells going to be analyzed. We also show the XY axis from 0 to 1,600 to better spot the exact cell location from the figure and exported data.

- b. Data will be saved under the file called *cell data*. File name: Well name + bestsp.
- c. Parameters will be saved into several spreadsheets (Figures 7 and 8, Table 1); each spreadsheet is a parameter of migration behavior. An example of exported data is shown in Figure 8. Note that the angle data (wellang) need to be further calculated through the formula:  $\pi/2 \theta$ .  $\theta$  is the original value shown on the spreadsheet.



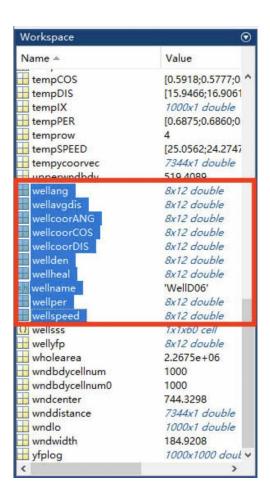
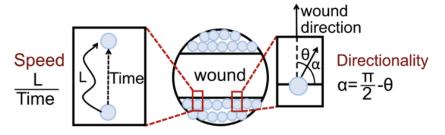


Figure 7. Screenshot of Z3\_getcellparameters062945 in workspace after running this script. MATLAB files have been generated after running Z3\_getcellparameters062945, and the circled files are exported to Excel for further analysis.

# Wound healing



**Figure 8.** Graphical schematic of variables used in cell migration analysis. The directionality is the angle of cell movement relative to the wound direction. Retrieved from Yu et al. (2021).

**Table 1. Migration statistics** 

Variable	Calculation	Description				
Wellang	π/2-α	Average angle of cell movement relative to the wound direction in the well  Average cell–cell coordination in the well				
wellcoorCos	$\cos \theta$					
wellden	Total cell number in each frame	Cell number of this frame. Note that cell				



		migration speed could be influenced by cell
		density, so this data has to be quite consistent
		among each well
wellspeed	Distance/time	Average cell migration speed in the well. The
		unit is μm/h.
wellper	Displacement/distance	Average consistency of directionality of the cell
		movement in the well

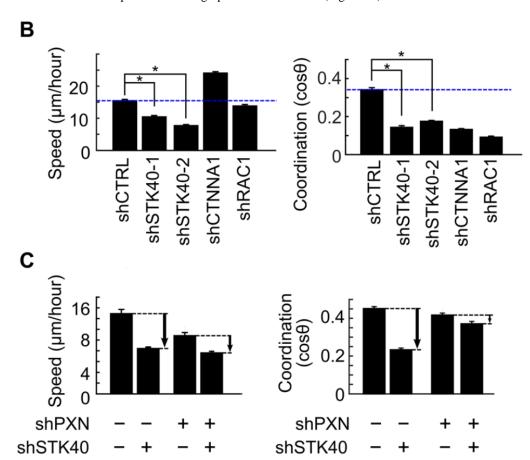
To make the definition clear, Figure 7 explains the variables being used.

Although our scripts generate all variables, we use different parameters for analyzing different cell migrations. For wound healing assay, we analyzed the speed, angle, coordination, and persistence of cell migration. However, we do not analyze cell angles for random migration because we define cell angle as how cells deviate from the direction of the wound. Here, we provide an example of the data we used for analyzing the wound healing assay result (Figure 9).

4 wellcoorCOS									
5		2	3	4	5	6	7	8	9
6	В	0.44532	0.45554	0.53648	0.64353	0.46759	0.53212	0.59526	0.53939
7	С	0.45476	0.41184	0.51486	0.5776	0.4495	0.62345	0.58638	0.60804
8	D	0.48727	0.4295	0.52541	0.6443	0.4446	0.55123	0.50883	0.50575
9	Е	0.49534	0.49482	0.5291	0.57067	0.50489	0.55568	0.54707	0.49715
10	F	0.60199	0.48747	0.5669	0.61712	0.49353	0.54968	0.46094	0.43857
11	G	0.57621	0.50913	0.59633	0.67727	0.45138	0.56975	0.56765	0.598
12									
13 wellspeed									
14									
15	В	17.8872	26.7484	25.1572	26.9069	21.1367	17.359	19.3748	19.3298
16	С	28.1117	30.5382	22.9752	26.9723	22.4562	17.9675	21.3322	16.4062
17	D	24.4717	32.8122	28.2629	26.7423	23.19	24.4721	23.5052	17.9359
18	Ε	21.8745	31.3301	21.766	23.9614	22.6538	21.1184	27.8036	19.0055
19	F	21.0974	31.2436	21.8575	24.8107	22.7746	21.6974	28.9194	16.8903
20	G	21.0034	26.9531	21.433	27.9122	22.5588	20.9455	32.8397	17.9732
21 wellang									
22									
23	В	0.89812	0.88345	0.86313	0.65636	0.84943	0.75449	0.673	0.78039
24	С	0.86862	1.00863	0.83792	0.72486	0.91573	0.6124	0.66337	0.67338
25	D	0.84075	1.02754	0.82688	0.62266	0.91406	0.70506	0.78484	0.78753
26	Е	0.80925	1.03515	0.79092	0.71132	0.78975	0.72982	0.76581	0.78193
27	F	0.65627	0.92745	0.76507	0.68379	0.84638	0.7642	0.88898	0.94601
28	G	0.71801	0.82323	0.69144	0.56031	0.93477	0.70653	0.70814	0.61901
yellper wellper									
80									
31	В	0.65905	0.65915	0.72248	0.7896	0.65055	0.71242	0.75518	0.72612
32	С	0.65686	0.61902	0.6989	0.73003	0.64127	0.79772	0.76799	0.76001
33	D	0.69276	0.62734	0.71555	0.79272	0.64967	0.73728	0.69818	0.70197
84	Ε	0.68426	0.66579	0.72078	0.74011	0.71624	0.72864	0.72638	0.70013
35	F	0.74195	0.68496	0.73871	0.76454	0.67236	0.73089	0.65187	0.65728
36	G	0.73947	0.69089	0.75895	0.81098	0.63957	0.75307	0.7384	0.76305
wellden									
38									
39	В	7292.72	7277.53	6621.52	7939.44	7895.97	9057.24	9312.91	6945.06
40	C	6867.19	6445.38	6574.23	8050.17	7371.02	8716.46	8728.9	8196.36
41	D	6809.7	6643.91	6980.07	7991.3	7018.17	9732.28	8098.3	8310.24
12	Е	8695.24	6296.19	6266.51	7958.08	6937.88	7770.42	7282.29	7387.36
13 14	F	8435.18	6475.98	6458.26	8059.08	7008.47	8775.05	6905.97	6973.18
14	G	8967.65	7362.28	7088.44	7793.67	7827.03	8458.76	7949.64	10331.4
45									

Figure 9. Example of the population of cell data. Each data point corresponds to the mean value of each well. These data are from row B–G, column 2–9 of a 96-well plate.





These data could be plotted into bar graphs for visualization (Figure 10).

Figure 10. Example of the bar graph for cell migration analysis. Retrieved from Yu et al. (2021).

# Validation of protocol

This protocol or parts of it has been used and validated in the following research article(s):

• Yu et al. (2021). Synthetic dysmobility screen unveils an integrated STK40-YAP-MAPK system driving cell migration. *Sci. Adv.* 7(31): eabg2106 (Figure 1, panel C and G, and Figure 2, panel B and C).

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# **Competing interests**

The authors declare that they have no competing interests.

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# **Supplementary information**

The following supporting information can be downloaded here:

- 1. Z1\_calcnuclei\_adjustedPlateShift.m
- 2. Z2\_trace nuclei.m
- 3. Z3\_getcellparameters062945.m