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Title

Evaluation of Dynamic Cell Processes and Behavior Using Video Bioinformatics Tools

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<https://escholarship.org/uc/item/8kz4m4sc>

ISBN

978-3-319-23723-7

Authors

Lin, Sabrina C
Yip, Henry
Phandthong, Rattapol
[et al.](#)

Publication Date

2015

DOI

10.1007/978-3-319-23724-4_9

Peer reviewed

Metadata of the chapter that will be visualized in SpringerLink

Book Title	Video Bioinformatics	
Series Title		
Chapter Title	Evaluation of Dynamic Cell Processes and Behavior Using Video Bioinformatics Tools	
Copyright Year	2015	
Copyright HolderName	Springer International Publishing Switzerland	
Author	Family Name	Lin
	Particle	
	Given Name	Sabrina C.
	Prefix	
	Suffix	
	Division	UCR Stem Cell Center, Department of Cell Biology and Neuroscience
	Organization	University of California
	Address	Riverside, 92521, USA
	Email	
Author	Family Name	Yip
	Particle	
	Given Name	Henry
	Prefix	
	Suffix	
	Division	UCR Stem Cell Center, Department of Cell Biology and Neuroscience
	Organization	University of California
	Address	Riverside, 92521, USA
	Email	
Author	Family Name	Phandthong
	Particle	
	Given Name	Rattapol
	Prefix	
	Suffix	
	Division	UCR Stem Cell Center, Department of Cell Biology and Neuroscience
	Organization	University of California
	Address	Riverside, 92521, USA
	Email	
Author	Family Name	Davis
	Particle	
	Given Name	Barbara
	Prefix	
	Suffix	
	Division	UCR Stem Cell Center, Department of Cell Biology and Neuroscience
	Organization	University of California
	Address	Riverside, 92521, USA

Email

Corresponding Author	Family Name	Talbot
	Particle	
	Given Name	Prue
	Prefix	
	Suffix	
	Division	UCR Stem Cell Center, Department of Cell Biology and Neuroscience
	Organization	University of California
	Address	Riverside, 92521, USA
	Email	talbot@ucr.edu

Abstract

Just as body language can reveal a person's state of well-being, dynamic changes in cell behavior and morphology can be used to monitor processes in cultured cells. This chapter discusses how CL-Quant software, a commercially available video bioinformatics tool, can be used to extract quantitative data on: (1) growth/proliferation, (2) cell and colony migration, (3) reactive oxygen species (ROS) production, and (4) neural differentiation. Protocols created using CL-Quant were used to analyze both single cells and colonies. Time-lapse experiments in which different cell types were subjected to various chemical exposures were done using Nikon BioStations. Proliferation rate was measured in human embryonic stem cell colonies by quantifying colony area (pixels) and in single cells by measuring confluency (pixels). Colony and single cell migration were studied by measuring total displacement (distance between the starting and ending points) and total distance traveled by the colonies/cells. To quantify ROS production, cells were pre-loaded with MitoSOX Red™, a mitochondrial ROS (superoxide) indicator, treated with various chemicals, then total intensity of the red fluorescence was measured in each frame. Lastly, neural stem cells were incubated in differentiation medium for 12 days, and time lapse images were collected daily. Differentiation of neural stem cells was quantified using a protocol that detects young neurons. CL-Quant software can be used to evaluate biological processes in living cells, and the protocols developed in this project can be applied to basic research and toxicological studies, or to monitor quality control in culture facilities.



Chapter 9

Evaluation of Dynamic Cell Processes and Behavior Using Video Bioinformatics Tools

Sabrina C. Lin, Henry Yip, Rattapol Phandthong, Barbara Davis and Prue Talbot

Abstract Just as body language can reveal a person's state of well-being, dynamic changes in cell behavior and morphology can be used to monitor processes in cultured cells. This chapter discusses how CL-Quant software, a commercially available video bioinformatics tool, can be used to extract quantitative data on: (1) growth/proliferation, (2) cell and colony migration, (3) reactive oxygen species (ROS) production, and (4) neural differentiation. Protocols created using CL-Quant were used to analyze both single cells and colonies. Time-lapse experiments in which different cell types were subjected to various chemical exposures were done using Nikon BioStations. Proliferation rate was measured in human embryonic stem cell colonies by quantifying colony area (pixels) and in single cells by measuring confluency (pixels). Colony and single cell migration were studied by measuring total displacement (distance between the starting and ending points) and total distance traveled by the colonies/cells. To quantify ROS production, cells were pre-loaded with MitoSOX Red™, a mitochondrial ROS (superoxide) indicator, treated with various chemicals, then total intensity of the red fluorescence was measured in each frame. Lastly, neural stem cells were incubated in differentiation medium for 12 days, and time lapse images were collected daily. Differentiation of neural stem cells was quantified using a protocol that detects young neurons. CL-Quant software can be used to evaluate biological processes in living cells, and the protocols developed in this project can be applied to basic research and toxicological studies, or to monitor quality control in culture facilities.

Electronic supplementary material The online version of this article (doi:[10.1007/978-3-319-23724-4_9](https://doi.org/10.1007/978-3-319-23724-4_9)) contains supplementary material, which is available to authorized users.

S.C. Lin · H. Yip · R. Phandthong · B. Davis · P. Talbot (✉)
UCR Stem Cell Center, Department of Cell Biology and Neuroscience,
University of California, Riverside 92521, USA
e-mail: talbot@ucr.edu



9.1 Introduction

Evaluation of dynamic cell processes and behavior is important in basic research [11, 40, 43, 46], in the application of stem cell biology to regenerative medicine [29, 41], and in studies involving the toxicity of drug candidates and environmental chemicals [13, 24, 25, 31, 32, 35–38, 42]. Prior work in basic and toxicological research has often involved microscopic observation of cells or assays that evaluate single endpoints after chemical exposure (e.g., [4, 6–8, 25, 33]). However, much additional insight can be learned about a cells response to its environment by comparing dynamic processes, such as cell growth and motility, in treated and control cells [30, 31, 44]. Just as human body language can reveal information about human mood and well-being, cellular dynamics can often reveal information about the mode of action and the cellular targets of chemical exposure. For example, impairment of cell motility would likely be correlated with an adverse effect on the cytoskeleton. Such an effect can be quantified in video data without using any labels or genetic transformation of the cells [25, 31, 36, 49]. In addition, fluorescent labels can be used to report the condition of cells in time-lapse data thereby revealing more information about a treatment than a single endpoint assay [27]. Finally, multiple endpoints can be multiplexed and mined from video data to gain additional insight from a single experiment [2, 34].

The interest and importance of video data in cellular studies has led to the commercialization of a number of instruments (e.g., BioStation CT/IM, Cell IQ, Tokai Hit) optimized for collecting live cell images over time [10, 44]. Videos can now be made for hours, days, or even months using conditions that support in vitro cell culture and experimentation. However, while dynamic video data are rich with information about cell health and cell processes, they are often difficult to analyze quantitatively. This is due to the complexity of the data and the generally large size of the data sets. Moreover, video analysis can be very time-consuming and is error-prone due to subjectivity of the human(s) performing the analysis. The recent interest in live cell imaging has been accompanied by a need for software tools for extracting information from video data. This field of study has been termed “video bioinformatics” (www.cris.ucr.edu/IGERT/Index.php). Video bioinformatics includes the development and application of software tools for extraction and mining of information and knowledge from video data. The advantages of using video bioinformatics tools are enormous. Tremendous amounts of time can be saved, and when properly applied, video bioinformatics tools will extract more accurate reproducible data than would generally be the case for a human performing the same task. Video bioinformatics tools are available commercially [3] and are also being developed in research laboratories to solve specific problems such as quantification of cells in colonies, cell identification, and prediction of successful development of human embryos to the blastocyst stage [14–17, 21, 51].

In this chapter, four applications of video bioinformatics tools to toxicological problems are presented. First, cell colony and individual cell growth were

72 monitored using time lapse data. Second, single cell and colony migration were
73 analyzed to provide information on rate of migration, distance traveled, and total
74 displacement. Third, a method is presented for direct observation and quantification
75 of ROS production in cultured cells. Finally, quantification of differentiating neu-
76 rons was accomplished by evaluating time-lapse videos collected over a period of
77 10 days. Video data were collected in either a BioStation CT or BioStation IM, both
78 available from Nikon. Analyses were done using protocols created using a com-
79 mercial software package (CL-Quant). Each application can be used with either
80 single cells or colonies.

81 9.2 Collection of Time-Lapse Data

82 The BioStation IM is a fully motorized, automated, environmentally controlled
83 microscope and imaging system that captures images using a cooled monochrome
84 CCD camera. It was designed to enable live cell imaging using optimal in vitro
85 conditions. It can accommodate 35 mm culture dishes including HiQ4 dishes
86 (Nikon Instruments Inc., Melville, NY) that allow four different treatments to be
87 monitored in a single experiment. Cells are incubated at 37 °C in a CO₂ controllable
88 atmosphere with a high relative humidity. Multiple magnifications are possible for
89 capturing phase contrast and/or fluorescence images using software that controls
90 point selection and collection of data. Perfusion is an option to allow for real-time
91 addition or subtraction of cell culture media and to enable longer-term observation.
92 The BioStation IM robotics are capable of precise cell registration so the resultant
93 movies can be analyzed quantitatively.

94 The BioStation CT is a much larger incubation unit that can perform high
95 content work ideal for live cell screening. The culture conditions inside the
96 BioStation CT can be regulated. While our unit is usually operated at 5 % CO₂,
97 85 % relative humidity and 37 °C, hypoxic conditions are also possible if needed.
98 The BioStation CT is especially suitable for data collection in long-term experi-
99 ments, in which cells are studied over weeks or months. It has a robotic arm for
100 transfer of plates to and from a microscope stage which enables complete auto-
101 mation of the time-lapse experiment. The BioStation CT holds up to 30 experi-
102 mental samples in various plate formats (6, 12, 24, 48, 96 well plate formats, 35, 60
103 and 100 mm dish formats, and 25 and 75 cm² flask formats). A cooled monochrome
104 CCD camera collects phase and/or fluorescence images at defined intervals and
105 points of interests. Large montages of the entire well area can be taken over the
106 magnification range of 2×–40× which allows for complete cell characterization over
107 the life of the cell culturing period.



9.3 CL-Quant Software

All video analyses were performed using CL-Quant software, a live-cell image analysis program produced for Nikon by DRVision Technologies (Bellevue, Washington). It can either be purchased from Nikon as CL-Quant or from DRVision under the name SVCCell. The current version of the software is user friendly, features an intuitive GUI to manage high content imaging experiments, and comes with webinar instruction. All ground-truth evaluations of CL-Quant were done using either ImageJ or Photoshop.

CL-Quant comes with several modules professionally developed by DRVision for basic processing of videos. For example, bioinformatics tools for measuring cell confluency, cell migration, and cell counting can be obtained from Nikon and applied to users' videos. CL-Quant also provides tools that end users can work with to develop protocols for recognition and quantitative analysis of microscopic video data [3]. CL-Quant protocols can be applied with user directed learning and do not require image processing knowledge. Although the software has great depth, basic analyses can be done with relatively little training. CL-Quant can be used to detect, segment, measure, classify, analyze, and discover cellular phenotypes in video data. Preconfigured modules are available for some applications such as cell counting, confluency, cell division, wound healing, cell motility, cell tracking, and measuring neurite outgrowths. Moreover, the software has significant depth and can be configured for other more complex applications by the user.

In this chapter, examples will be shown for adapting CL-Quant to measure cell/colony growth rate, cell/colony migration, ROS production, and neural differentiation. Protocols, developed by DRVision and Nikon software engineers and those created by novices learning to use the CL-Quant software, will be compared and used to study cell behavior. The above parameters can be useful in toxicological studies, in work that requires knowledge of cell health, in clinical applications of stem cells to regenerative medicine, or in basic studies of cell biology.

9.4 Cell and Colony Growth

9.4.1 Growth of Human Induced Pluripotent Stem Cells (hiPSC)

Human-induced pluripotent stem cells (hiPSC; RivCAG-GFP), created in the UCR Stem Cell Core Facility and grown in 12-well plates as described previously [28] were either incubated in control medium (mTeSR, Stem Cell Technologies) or in mTeSR containing 0.1 puff equivalents (PE) of sidestream cigarette smoke (PE = the amount of smoke in one puff that dissolves in 1 ml). This concentration was shown previously to inhibit human embryonic stem cells (hESC) colony growth [31]. Cells were imaged at 10 \times magnification for 48 at 6 h intervals in a

BioStation CT maintained at 37 °C, 5 % CO₂ and a relative humidity of 85–90 %. Phase contrast images show the growth of a single control and treated colony over 48 h (Fig. 9.1a–c and g–i). All colonies analyzed were selected to be relatively close in size before treatment. During incubation, the sizes of the treated colonies appeared to be smaller than the control colonies. Some treated colonies did not grow, and some eventually died due to treatment (as shown in Fig. 9.1i). To obtain quantitative information on colony growth rates, a protocol, which was developed in our lab with the CL-Quant software (version 3.0) and used previously with hESC [31], was applied to these iPSC video data. The protocol first segmented images of control and treated colonies and overlaid each colony with a mask (Fig. 9.1d–f and j–l). The fidelity of the mask was excellent for both control and treated groups. After images were segmented, small objects, dead cells, and debris were removed with an enhancement module, and finally the size of each colony was measured in pixels in each frame of each video. Because each colony is slightly different at the start of an experiment, resulting data for each set of videos were normalized to the size of the colony in frame 1, then data were averaged and growth curves were graphed (Fig. 9.1m). Results showed a clear difference in growth rates between the control and treated colonies. In fact, the treated colonies decreased in size and appeared not to grow over the 48 h incubation period. Treatment was significantly different than the control (2-way ANOVA, $p \leq 0.05$), and the iPSC were more sensitive to sidestream smoke treatment than the hESC studied previously [32]. The protocol used for this analysis had previously been compared to ground-truth derived using Adobe Photoshop, and excellent agreement was found between the data obtained with CL-Quant analysis and the ground-truth [31].

The data shown in Fig. 9.1 involved analysis of 60 images. To perform this analysis by hand would require approximately 3–4 h. CL-Quant was able to perform this analysis in about 1 h, and it can be run in a large batch so that the users' time is not occupied during processing. With a larger experiment having more frames, the difference between CL-Quant and manual analysis would be much greater.

Video examples of iPSC colony growth and CL-Quant masking can be viewed by scanning the bar code.

9.4.2 Growth of Mouse Neural Stem Cells (mNSC)

Monitoring the growth of single cells can be more challenging than monitoring hiPSC or hESC colony growth. Some single adherent cells grow very flat and do not differ much in contrast from the background making segmentation difficult. However, images can be enhanced by adjusting the brightness, the contrast, and/or the gamma parameters using CL-Quant software or other image processing software (e.g., Photoshop and ImageJ). CL-Quant comes with some professionally developed modules for use with some types of single cells. Investigators can try these protocols to see if one works well with their cell type or, alternatively, they

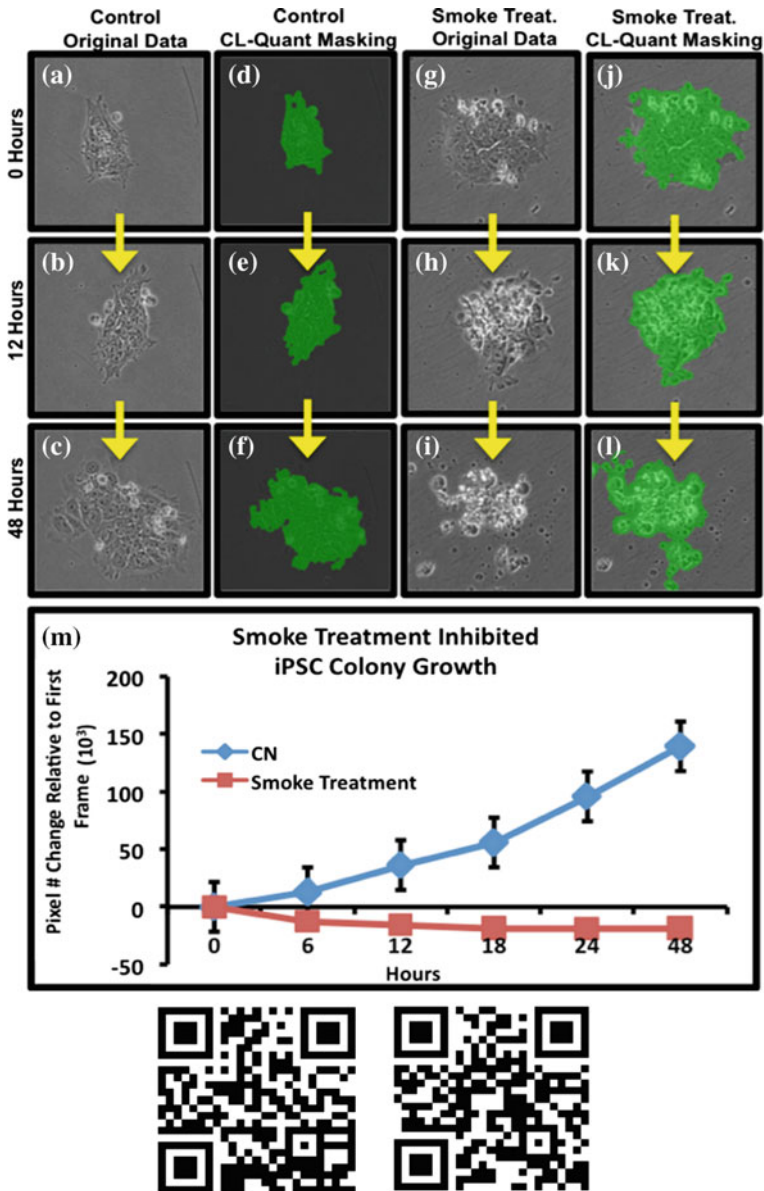


Fig. 9.1 Growth of hiPSC colonies over 48 h. **a–c** Phase contrast images of control iPSC colonies at various times during growth. **d–f** The same control images segmented using a CL-Quant protocol developed in our lab. **g–i** Phase contrast images of smoke treated iPSC colonies at various times during growth. **j–l** The same treatment group images masked using the same CL-Quant protocol as applied to control colonies. **m** Graph of control and treated cells showing growth rate. Data are means and standard errors of three experiments. *CN* control

187 can use the CL-Quant software to develop their own protocol and tailor it to the
 188 specific requirements of the cells they are using. However, one should not assume
 189 that the protocols are accurate and should check them against ground-truth.

190 In our experiments, the effect of cigarette smoke treatment on mNSC proliferation
 191 was examined. mNSC were plated in 12-well plates at 2,500 cells/well, and
 192 cells were allowed to attach for 24 h before treatment, incubation, and imaging.
 193 Various fields of interests were imaged over 48 h in the BioStation CT in 5% CO₂
 194 and 37 °C. The collected video data (Fig. 9.2a–c) were then processed and analyzed

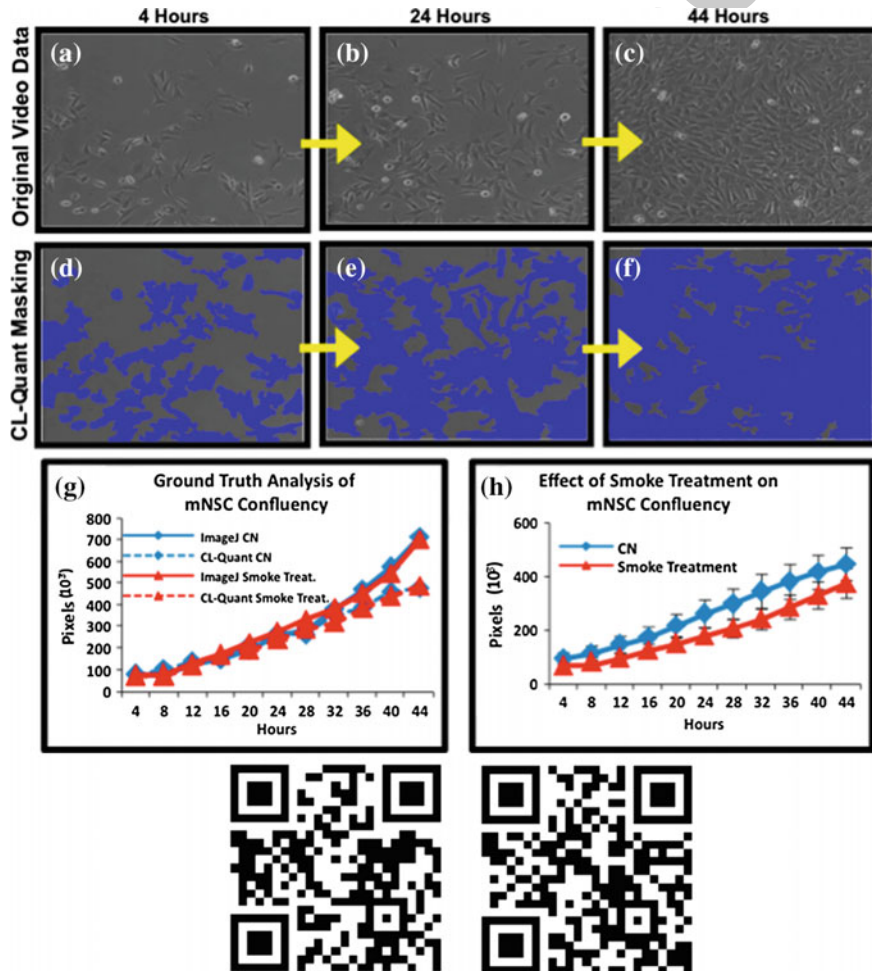


Fig. 9.2 Growth of single mNSC. **a–c** Phase contrast images of control mNSC at various times during growth over 48 h. **d–f** The same images after segmentation using a protocol developed by DR Vision. **g** Graph showing analysis for growth rate of control and treated cells (*solid lines*) and ImageJ ground-truth for each group (*dotted lines*). **h** Graph of control and treated mNSC showing confluency rate. Data are means and standard errors of three experiments. *CN* control

195 using the confluency module provided by DRVision Technologies for use with
196 CL-Quant software. The confluency module masked the cells in each field, and
197 mNSC growth was determined by measuring the number of pixels in each frame
198 (Fig. 9.2d–f). Ground-truth was obtained to verify the validity of the CL-Quant
199 confluency analysis tool using ImageJ. To obtain ground-truth, each cell was
200 carefully outlined and colored to measure area (pixels), and a comparison of
201 CL-Quant and ImageJ data showed CL-Quant was reliable over the first 34 h with
202 some divergence at the latest times (Fig. 9.2g). A complete mNSC growth exper-
203 iment was analyzed in which the growth of cigarette smoke treated cells was
204 compared to nontreated control cells (Fig. 9.2h). Smoke treatment of mNSC sig-
205 nificantly inhibited their proliferation from 20 to 44 h (2-way ANOVA, $p \leq 0.001$).

206 Video examples of mNSC proliferation (control and smoke treatment) can be
207 viewed by scanning the bar code.

208 9.5 Cell Migration

209 9.5.1 Migration of hESC Colonies

210 Evaluation of cell motility can be important in determining if the cytoskeleton of
211 treated cells has been affected. The data in Fig. 9.3 were collected using hESC
212 colonies that were incubated in a BioStation CT for 48 h, and images were collected
213 of each colony at 10 min intervals. Colonies, which were grown on Matrigel, were
214 incubated in either control medium (mTeSR) or mTeSR containing cigarette smoke.
215 Normally, hESC show motility when grown on a Matrigel.

216 CL-Quant provides a number of readouts for motility. The two that were most
217 useful are total distance traveled and total displacement. Total distance traveled is
218 the measurement of how far the colony has migrated over time, and total dis-
219 placement is the difference in distance between the beginning point and the end-
220 point. Figure 9.3a–c shows examples of hESC colonies that have been masked and
221 tracked by a motility protocol. Within a population of hESC colonies, three
222 behaviors were observed: (1) growing, (2) shrinking, and (3) dying. The tracking
223 module traces the path of the colonies, and those that showed growth had longer
224 paths than colonies that were shrinking or dying.

225 In Fig. 9.3d, e, displacement and total distance traveled were measured for
226 individual colonies in control and cigarette smoke treatment groups. All colonies in
227 the control group were healthy and growing, but 2 of 12 treated colonies (red
228 circles) were dead by the end of 48 h. For total distance traveled, measurements for
229 control colonies appeared to be clustered, while the treated colonies were more
230 variable, in part due to the presence of two dead colonies. For both the control and
231 treated groups, the total displacement was quite variable, suggesting there was no
232 directional movement in either group. A t test was performed on both parameters,
233 after removing measurements of dying colonies, and the results showed that the

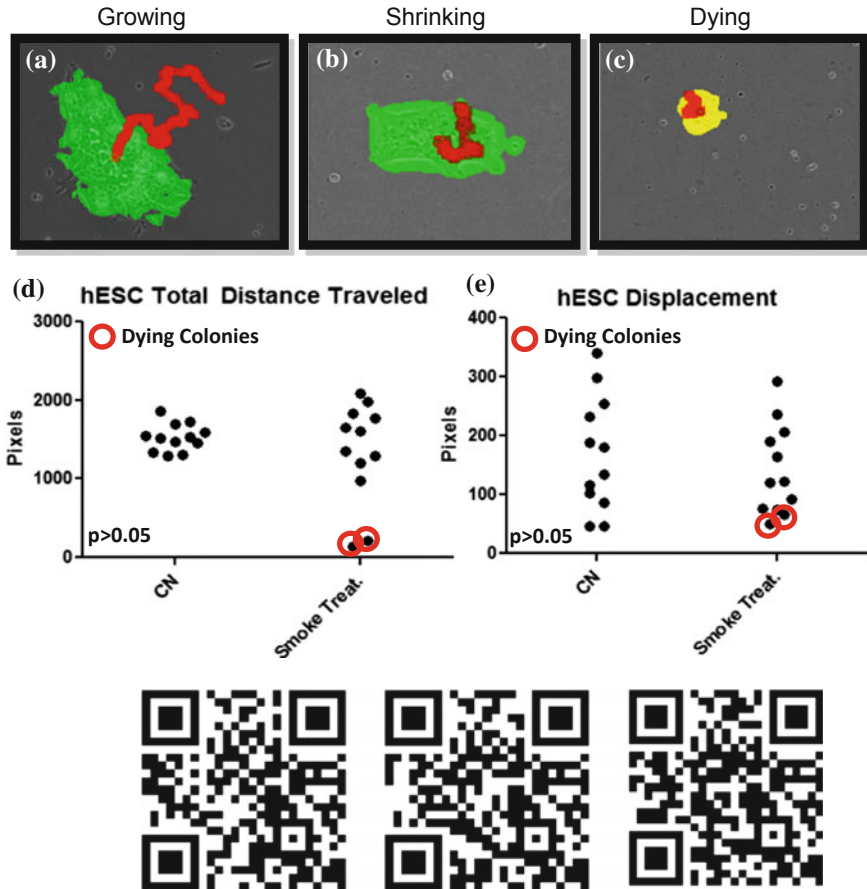


Fig. 9.3 Migration of hESC colonies. **a** Masked phase contrast image of a growing hESC colony during migration. **b** Masked phase contrast image of a shrinking hESC colony during migration. **c** Masked phase contrast image of a dying hESC colony during migration. **d, e** Graphs showing total displacement/distance traveled for each control and treated colonies. All CL-Quant masking and tracking of colonies were done by applying a tracking recipe developed by our lab. *CN* control

234 distance traveled and displacement of control and treated colonies were not sig-
 235 nificantly different ($p > 0.05$).

236 hESC migration is an important process during development, as derivatives of
 237 these cells must migrate during gastrulation to form the three germ layers properly
 238 [22]. Therefore, although our cigarette smoke treatment did not affect migration of
 239 hESC colonies, these two parameters can be useful in determining the effects of
 240 other toxicants on pluripotent cell migration. Observed effects on total distance
 241 traveled and displacement of colonies can be the first signs of potent chemical
 242 effects on the cytoskeletal integrity of the cells.

243 Video examples of hESC colony migration and CL-Quant masking can be
 244 viewed by scanning the bar codes.

245 9.5.2 Migration and Gap Closure of mNSC 246 and NTERA2 Cells

247 Single cell migration can be analyzed using a gap closure assay. This assay is
 248 performed by growing a monolayer of cells, creating a gap in the middle of the
 249 monolayer, and monitoring the time required for cells to migrate into the gap and
 250 close it. The gap can be made using a pipette to remove a band of cells, but the sizes
 251 of the gaps are not always uniform. As a result, the rate of closure may not be as
 252 accurate and comparable among control and treated groups. We have used Ibidi
 253 wound healing culture inserts (Fig. 9.4, Ibidi, cat#80241, Verona, WI) to make
 254 uniform gaps. First, inserts were adhered to culture plates. Second, cells were plated
 255 in each well of the insert, and when cells in the wells were confluent, the insert was
 256 removed leaving a uniform gap (500 μm) between cells in each well. This method
 257 works well with cells grown on plastic or glass, but not with cells grown on wet
 258 coating substrates (e.g., Matrigel, poly-D-lysine, poly-L-lysine, and laminin)
 259 because the adhesive at the bottom of the inserts will not stick.

260 Experiments using two types of cells, mNSC and NTERA2, are shown in
 261 Fig. 9.5. Both cell types can be grown on plastic, and time-lapse videos of both cell
 262 types were collected in the BioStation CT for 44 h. Images were collected every 4 h
 263 and CL-Quant analysis was done for each frame by measuring the number of pixels
 264 in the gap. The gap closure module, developed by our lab using the CL-Quant
 265 software, includes a segmentation recipe to identify the gap between the two
 266 populations of cells and a measurement recipe that counts the number of pixels in
 267 the gap. An example of mNSC gap closure images is shown in Fig. 9.5a–c, and
 268 CL-Quant masking of the same gap is shown in Fig. 9.5d–f. In the filmstrip, the gap
 269 became smaller as the cells migrated toward each other. Gap closure analysis using
 270 the CL-Quant software was validated using ground-truth obtained from the ImageJ
 271 software for control and treated NTERA2 (Fig. 9.5g, h). While CL-Quant tended to

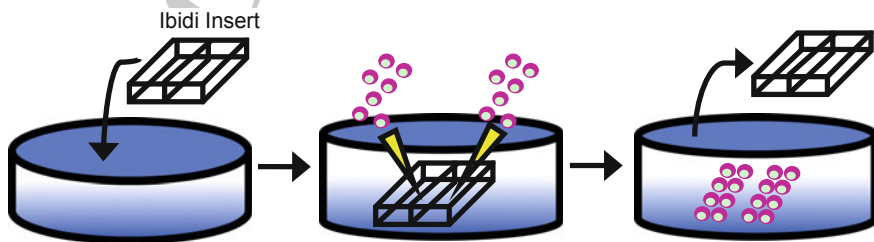
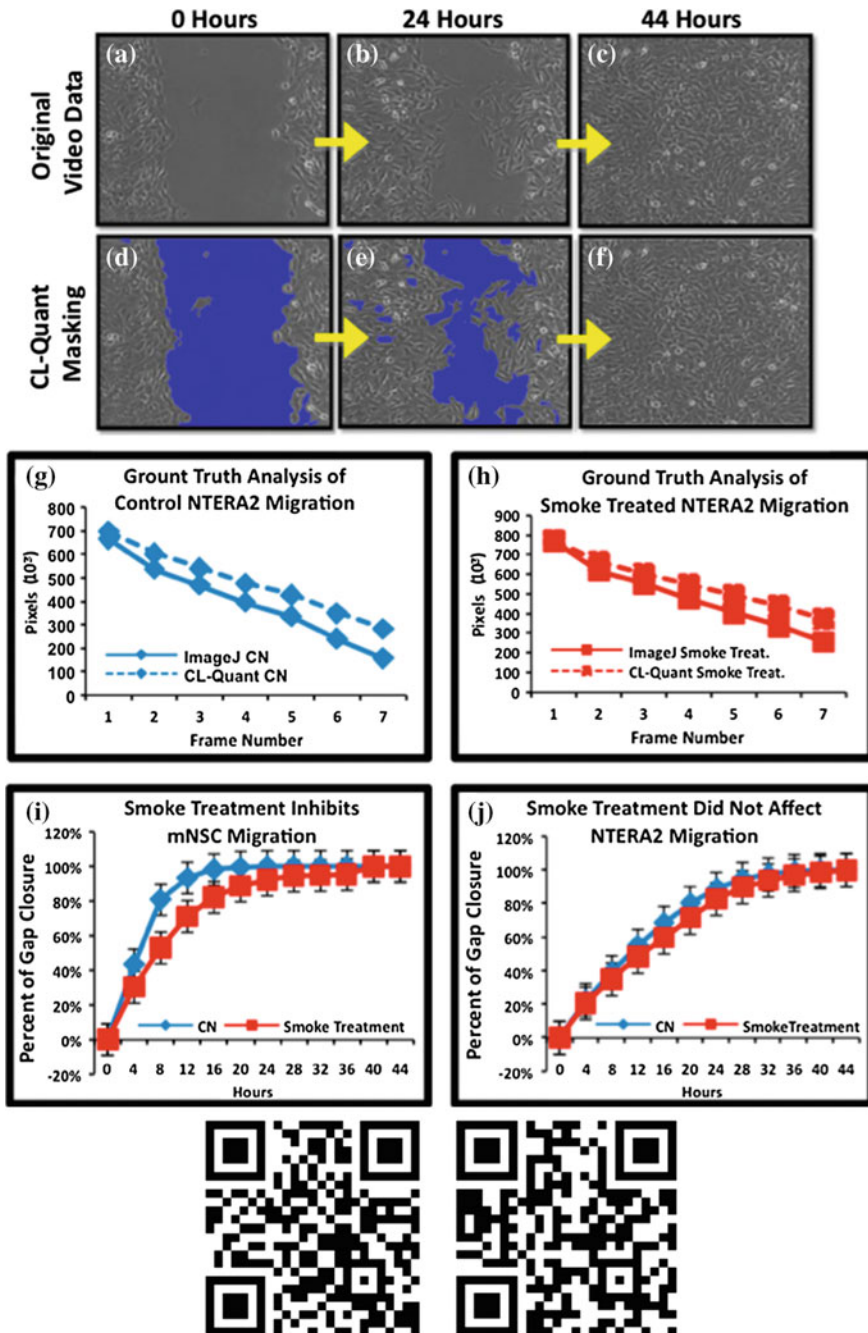


Fig. 9.4 Diagram of Ibidi gap closure culture inserts



◀ **Fig. 9.5** Gap closure for mNSC and NTERA2 cells. **a–c** Phase contrast images of mNSC at various 3 times during gap closure. **d–f** The same images after segmentation using a protocol developed in our lab with CL-Quant software. **g, h** Graph showing rate of gap closure for control (blue) and treated NTERA2 cells (red) and the corresponding ground-truth (dotted lines) obtained using ImageJ. **i** Graph of mNSC migration by monitoring percent of gap closure over 44 h. **j** Graph of NTERA2 cell migration by monitoring percent of gap closure over 44 h. Data are means and standard errors of three experiments

272 overestimate area slightly due to some extension of the mask beyond the gap,
273 ImageJ and CL-Quant analyses produced similar results for both control and treated
274 groups. mNSC and NTERA2 cell migration experiments were analyzed with the
275 CL-Quant gap closure module (Fig. 9.5i, j). Gap closure was completed in about
276 16 h for the mNSC, while the NTERA2 cells required about 40 h to completely
277 close the gap. Migration of mNSC, but not the NTERA2, was significantly
278 inhibited by cigarette smoke ($p \leq 0.001$ for 2-way ANOVA of mNSC data).

279 Although our gap closure analysis was done by measuring the pixels within the
280 gap, it can also be monitored by masking the cells. For certain cells that produce a
281 clear phase contrast image, this option may be easier and more accurate than
282 monitoring the gap.

283 Video examples of control and smoke treated single cell migration can be
284 viewed by scanning the bar codes.

285 9.6 Detection of Reactive Oxygen Species (ROS) 286 in Human Pulmonary Fibroblasts (HPF)

287 Exposure to environmental chemicals can lead to stress [9, 12, 26, 47], and ROS are
288 often produced in stressed cells [1, 19, 45]. ROS can damage macromolecules in
289 cells including proteins and DNA, and any factor that increases ROS would be
290 potentially damaging to a cell. It is possible to observe the production of ROS in
291 cells using fluorescent probes such as MitoSOX Red™ (Life Technologies, Grand
292 Island, NY). MitoSOX Red™ readily enters cells and is rapidly targeted to the
293 mitochondria. When oxidized by superoxide, it emits red fluorescence
294 (absorption/emission maxima = 510/580). MitoSOX Red™ can be preloaded in
295 cells and will fluoresce as levels of superoxide increase. Its fluorescent intensity is
296 related to the amount of superoxide in the cells.

297 In this example, hPF were disassociated from a culture vessel using 0.05 %
298 trypsin and then plated in the HiQ4 dishes coated with poly-D-lysine. After hPF
299 were allowed to attach for 24 h, cells were preloaded with 5 μ M MitoSOX Red™
300 for 10 min at 37 °C in a cell culture incubator. Preloaded cells were washed with
301 culture medium and then either treated with cigarette smoke which induces ROS
302 production [47] or were left untreated (control). Dishes were placed in a BioStation
303 IM, which was programmed to capture images every 4 min for 10 h using both the
304 phase and red fluorescence channels.

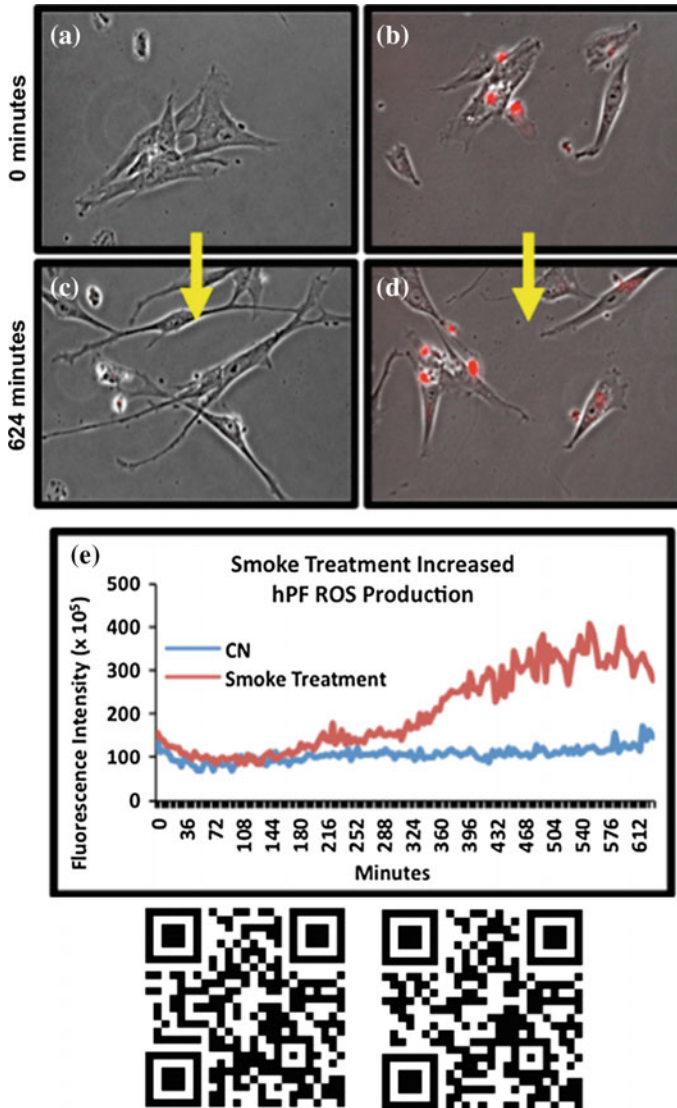


Fig. 9.6 Production of reactive oxygen species in hPF. **a–d** Merged phase contrast and fluorescent images at various times during incubation of control and treated hPF with MitoSox Red. **e** Graph showing fluorescence intensity in control and treated cells over time. *CN* control

305 A protocol was developed using CL-Quant to analyze the level of MitoSOX
 306 Red™ fluorescence in living cells. This was done by first developing a segmen-
 307 tation procedure to identify fluorescence. An enhancement program was then used
 308 to remove all debris and dead cells. The dead cells are highly fluorescent but round
 309 and easily excluded from the analysis with a size-based enhancement filter.

310 The background was flattened, and the mask was applied and observed to determine
311 if it accurately covered each cell in the entire video. If the mask was not accurate,
312 the segmentation was refined until masking accurately covered all living cells in
313 each frame. CL-Quant was then used to measure the level of fluorescence in each
314 field of each video.

315 The above protocol was applied to time-lapse videos of control and cigarette
316 smoke treated hPF that were preloaded with MitoSOX Red™. Merged phase
317 contrast and fluorescent images of control and treated cells are shown at various
318 times in Fig. 9.6a–d. There are usually some highly fluorescent dead cells present in
319 each field at the start of an experiment. It is important to filter out the dead cells
320 before performing the analysis as they would contribute significantly to the
321 intensity measurements. The graph shows the intensity of the MitoSOX Red™
322 fluorescence in control and treated cells over 10 h of incubation (Fig. 9.6e). Control
323 levels remained low and relatively constant throughout incubation in agreement
324 with direct observation of the videos. In contrast, fluorescence increased signifi-
325 cantly in the treated cells. This increase begins at about 300 min of incubation and
326 continues until the end of the experiment. This method is useful for direct moni-
327 toring of ROS production in time-lapse images. It reports which cells produce ROS,
328 the relative amounts of ROS in control and treated groups, and the time at which
329 ROS is elevated. Statistical analysis showed that cigarette smoke treatment sig-
330 nificantly increased hPF ROS production over time (2-way ANOVA, $p \leq 0.001$).

331 Video examples of hPF ROS production in control and treated cells can be
332 viewed by scanning the bar codes.

333 9.7 Detection and Quantification of Spontaneous Neural 334 Differentiation

335 Differentiation of stem cell populations is an essential and important process of
336 normal development. Many *in vitro* differentiation protocols have been established
337 to derive various cell types that can then be used for degenerative disease therapy,
338 organ regeneration, and models for drug testing and toxicology research [42]. In all
339 cases, the health, morphology, and differentiation efficiency of the cells are
340 important parameters that should be closely observed and evaluated. Here, we
341 provide an example of how differentiating mNSC are monitored over time and
342 derived neurons are quantified using the CL-Quant software. mNSC were plated in
343 12-well plates in NeuroCult™ Differentiation Medium (Stem Cell Technologies,
344 Vancouver, Canada) for 12 days. The plate was incubated in the Nikon BioStation
345 CT, and several fields of interest were randomly chosen for imaging every 24 h.
346 The NeuroCult™ Differentiation Medium supports the differentiation of three brain
347 cell types: (1) neurons, (2) astrocytes, and (3) oligodendrocytes. The morphologies
348 of these three cell types are very different in that neurons have long axons and small
349 cell bodies, and astrocytes and oligodendrocytes are flatter in appearance. As seen

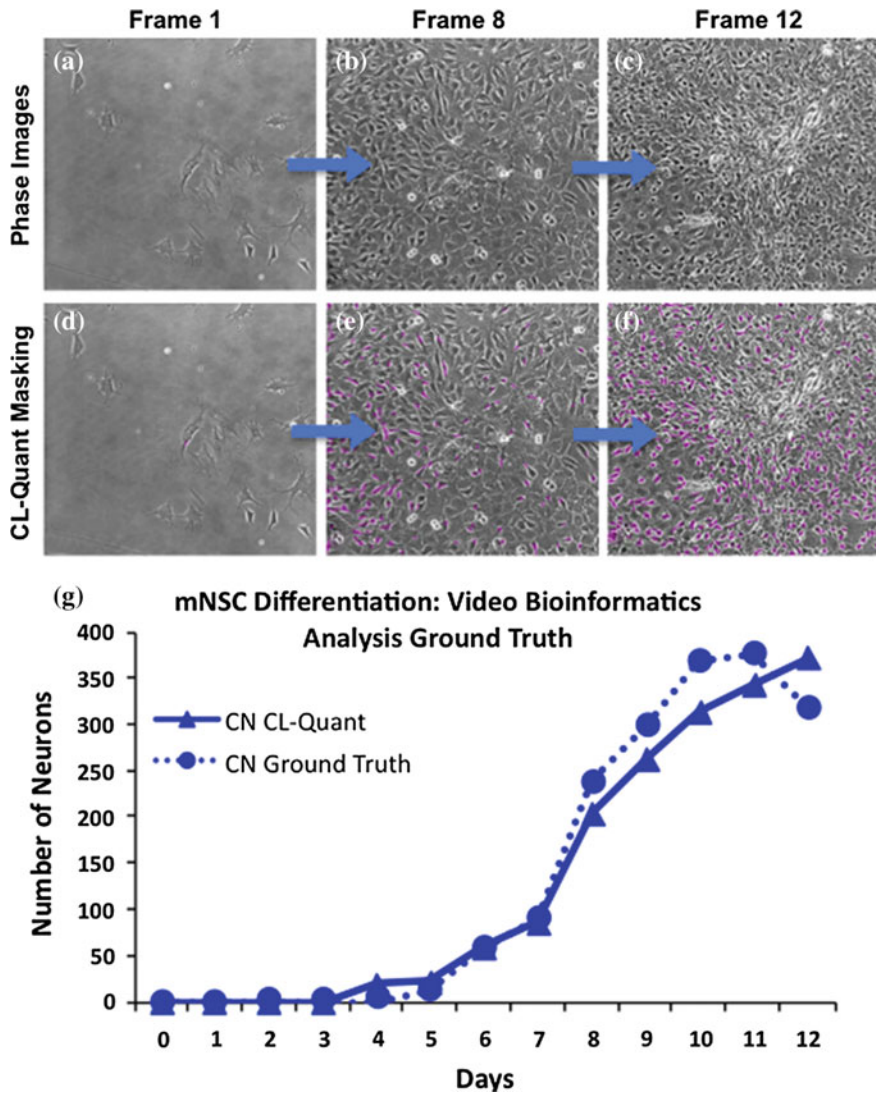


Fig. 9.7 Quantification of neurons in neural differentiation assay. **a–c** Phase contrast images of mNSC at various times during incubation. **d–f** CL-Quant software masking of mNSC phase contrast images to identify the neurons within each frame. **g** Graph showing quantification results obtained using the CL-Quant software was similar to the ground-truth obtained using the ImageJ software

350 in Fig. 9.7a–c, phase contrast microscopy of the differentiating neural stem cell
 351 population showed small, dark neurons sitting on top of a layer of flat cells. The
 352 stark morphological differences can be used to quantify the number of neurons in
 353 each frame. A segmentation recipe was developed using the CL-Quant software to

354 identify the darker and smaller neurons (Fig. 9.7d–f), and a measurement recipe was
355 used to count the number of neurons in each frame. We further validated our recipe
356 with ground-truth generated using the ImageJ software (Fig. 9.7g), and the number
357 of neurons identified in each frame using the video bioinformatics tool agreed
358 closely to the ground-truth data. Automation of the identification process is a
359 critical component for making future stem cell research more efficient and effective.

360 9.8 Discussion

361 This chapter gives four examples of how video bioinformatics tools can be applied
362 to experimental time-lapse data thereby enabling quantification of dynamic cellular
363 processes with attached cells that grow as colonies or single cells. Live cell imaging
364 is easily obtainable with modern instrumentation designed for culturing cells in
365 incubators equipped with microscopes [44]. Analysis of such data, as shown above,
366 can be done using professionally developed software tools [3] or tools developed by
367 end users with software such as CL-Quant [31, 44, 49]. However, any commercial
368 software may be limited in its ability to segment difficult subjects, in which case
369 custom software would need to be created [14–16, 21, 51]. In all of the
370 examples mentioned, use of video bioinformatics tools significantly reduced the
371 time for analysis and provided greater reproducibility than would normally be
372 obtained with manual human analysis. Although not demonstrated in this chapter,
373 the power of live cell imaging can be increased by multiplexing several endpoints
374 together in one experiment. For example, hESC or iPSC colony growth and
375 migration can be evaluated from the same set of video data, thereby conserving time
376 and resources.

377 We demonstrated how video bioinformatics tools were used to evaluate the
378 effects of cigarette smoke on dynamic processes (growth, migration, and ROS
379 production) in cultured cells. Many *in vitro* toxicological assays (e.g., MTT, neutral
380 red, and lactic dehydrogenase assays) are useful and effective in evaluating
381 chemical potency. Several recent studies from our lab effectively used the MTT
382 assay to screen the toxicity of various electronic cigarette fluids with embryonic
383 stem cells, mNSC, and hPF [4, 6, 7, 50]. Due to the sensitive nature of hESC
384 cultures, a new 96-well plate MTT protocol was also established for pluripotent
385 cells, which allows exact numbers of cells (in small clumps) to be plated from well
386 to well [6, 7]. While these *in vitro* assays are relatively quick and efficient, they
387 provide a single endpoint at one time/experiment, and the cells are often killed to
388 obtain the endpoint. As a result, dynamic changes in cell behavior and morphology
389 are not observed, and potential data are lost. In both basic research and toxicological
390 applications, examination of video data can reveal changes in cell dynamics
391 as well as rate data that are not gathered by single time point analysis. By determining
392 specific processes that are altered during treatment, the mode of action and
393 cellular targets may be identified. As an example, motility of mNSC, but not of
394 NTERT-2 cells, was affected by cigarette smoke, suggesting that the cytoskeleton is

395 more sensitive to smoke exposure in the former cells. Observations made from
396 time-lapse video data also provide insight on when during exposure chemicals
397 affect dynamic cellular processes.

398 Although this chapter presented toxicological applications of video bioinforma-
399 tics tools, other biological disciplines can benefit from this approach. For
400 example, Auxogyn, Inc. has established a method to determine the health of early
401 human embryos using time-lapse microscopy and an automated embryo stage
402 classification procedure [48, 51]. The protocol employs a set of learned embryo
403 features that allow 88 % classification accuracy of embryos that will develop to the
404 blastocyst stage. This advancement is being used in in vitro fertilization (IVF)
405 clinics to help physicians transfer only healthy embryos with the capacity to
406 develop into blastocysts. This not only increases IVF success rates, but decreases
407 the chance for multiple births that often result in unhealthy children. In June 2013,
408 Auxogyn announced the birth of the first baby to be born in an IVF clinic that used
409 the “Early Embryo Viability Assessment” (Eeva) test to select the best embryos for
410 transfer ([http://www.auxogyn.com/news.2013-06-14.first-auxogyn-baby-born-in-
411 scotland.php](http://www.auxogyn.com/news.2013-06-14.first-auxogyn-baby-born-in-scotland.php)).

412 The use of video bioinformatics tools will also be important when monitoring
413 the health of cells that will eventually be used in stem cell therapy. In the future,
414 stem cells grown for transfer to patients will be cultured over long periods during
415 passaging and differentiation making them costly in time and resources. Therefore,
416 it is important to monitor the culturing process using time-lapse data to verify that
417 cells are healthy and robust throughout in vitro culture and differentiation. It will be
418 important to have noninvasive monitoring systems for stem cell applications in
419 regenerative medicine. If a problem develops during expansion and culturing of
420 cells used in therapy, experiments can be terminated and restarted to assure that
421 only cells of excellent quality are transferred to patients.

422 Time-lapse data are also used in basic studies of cell biology. Qualitative and
423 quantitative analysis of video data have revealed information on dynamic cellular
424 processes [18, 20, 27], such as spindle formation during mitosis, actin protein
425 dynamics in cells, and gamete fusion [5, 23, 39]. Video data can also be used to
426 study cell processes that occur rapidly and are not easily understood by direct
427 observation, such as the acrosome reaction of lobster sperm [46]. Frame-by-frame
428 analysis of the acrosome reaction enabled each step during acrosomal eversion to be
429 analyzed and further enabled quantitative measurement of the forward movement of
430 sperm during the reaction. Time-lapse video microscopy has also been used to
431 study the process and rate of oocyte cumulus complexes pick-up by explants of
432 hamster oviducts [43].

433 New instrumentation, such as the Nikon BioStation CT/IM, provide long-term
434 stable incubation conditions for live cell imaging and enable acquisition of better
435 quality data than possible in the past. Improved methods for live cell imaging
436 coupled with video bioinformatics tools provide a new technology applicable to
437 numerous fields in the life sciences.



438 **Acknowledgments** Work presented in this chapter was supported by the following grants:
439 TRDRP 22RT-0127, CIRM NE-A0005A-1E, NSF IGERT DGE 093667, a Cornelius Hopper
440 Award from TRDRP, and a TRDRP postdoctoral fellowship 20FT-0084. We thank Dr. Evan
441 Snyder for providing the mNSC and Randy Myers and Ned Jastromb for their help with the
442 BioStations and CL-Quant.

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