

# Live-cell imaging and mathematical analysis of the "community effect" in apoptosis

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Live-cell imaging and mathematical analysis of the "community effect" in apoptosis Diane Coursier 1, 2, #, David Coulette 3, #, Hélène Leman 3, Emmanuel Grenier <sup>3</sup>, Gabriel Ichim <sup>1, 2, 4</sup> <sup>1</sup> Cancer Research Center of Lyon (CRCL) INSERM 1052, CNRS 5286, Lyon, France <sup>2</sup> Cancer Cell Death Laboratory, part of LabEx DEVweCAN, Université de Lyon, Lyon, France <sup>3</sup> ENS-Lyon, UMR CNRS 5669 'UMPA' and INRIA Alpes, project NUMED, Lyon, F-69364, France <sup>4</sup> Corresponding author: Gabriel Ichim (mailto:gabriel.ichim@lyon.unicancer.fr) # Equal contribution. 

#### **Abstract**

As a cellular intrinsic mechanism leading to cellular demise, apoptosis was thoroughly characterized from a mechanistic perspective. Nowadays there is an increasing interest in describing the non-cell autonomous or community effects of apoptosis, especially in the context of resistance to cancer treatments. Transitioning from cell-centered to cell population-relevant mechanisms adds a layer of complexity for imaging and analyzing an enormous number of apoptotic events. In addition, the community effect between apoptotic and living cells is difficult to be taken into account for complex analysis. We describe here a robust and easy to implement method to analyze the interactions between cancer cells, while under apoptotic pressure. Using this approach we showed as proof-of-concept that apoptosis is insensitive to cellular density, while the proximity to apoptotic cells increases the probability of a given cell to undergo apoptosis.

#### Introduction

Inconspicuous at the single cell level, apoptosis is a powerful mechanism when triggered at the organismal level. It is the sculpting force behind all multicellular organism morphogenesis, ensuring all organs have the right size, supernumerary cells are removed and neurons innervate the correct targets.

In cancer, apoptosis is an efficient guardian against oncogenic transformation and it is the preferred mechanism employed by tumor suppressor proteins to eliminate dangerous rogue cells (1). For instance, p53 can trigger apoptosis by up-regulating the expression of several pro-apoptotic proteins (1). Apoptosis is undeniably the main executioner of most cancer drugs and radiotherapy (2).

Apoptosis is triggered when one of two distinct pathways is engaged: the extrinsic or the intrinsic. Extrinsic apoptosis relies on the stimulation of a death receptor-family member (e.g., TRIAL-R, FAS or TNF-R) by a TNF-related cytokine, and activation of caspase 8 and 10 via the death-inducing signaling complex (DISC) (3). The intrinsic pathway is initiated by an intracellular death stimulus such as DNA damage, chemo- and radiotherapy

or tumor suppressor activation. In this case, the point of no return is mitochondrial outer-membrane permeabilization (or MOMP) followed by cytochrome *c* release, apoptosome assembly and activation of caspase 9 and then 3 and 7. Mitochondrial permeabilization is tightly regulated by the BCL-2 family proteins that can be anti-apoptotic (BCL-2, BCL-xL or MCL-1), proapoptotic (BID, BIM, BAD, PUMA or NOXA) and effector pore-forming proteins such as BAX or BAK (for review see (4)).

Once apoptosis is triggered, this does not inevitably lead to cell death, as it is now accepted that cells can survive if mitochondrial permeabilization does not affect all mitochondria or if the ESCRT-III complex restores plasma membrane integrity (5) (6) (7). Currently, there is a lack of understanding of how and why the kinetics of apoptosis varies so much between cells or following different treatments.

The heterogeneous response to apoptosis is also described in isogenic cancer cell lines. This is puzzling since these cells are genetically identical and should express similar levels of both pro- and anti-apoptotic proteins. Cellular heterogeneity to undergo apoptosis thus appears to originate from non-genetic transcriptional variability. This translates into cell-to-cell variations of pro- and anti-apoptotic protein expression and activity. For instance, the Lahav group elegantly showed that there is a high variability of p53 response to DNA damaging agents; apoptotic cells accumulate p53 much faster and earlier, while this accumulation is delayed in resistant cells, allowing time for the up-regulation of pro-survival IAP (inhibitor of apoptosis) proteins (8).

Recent publications shed light on the effects apoptotic cells exert on neighboring cells. They can actively release EGFR ligands, FGF2 or Wnt3 to boost the survival and proliferation of neighboring cells (9) (10) (11). This paracrine crosstalk is also complemented by mechanotransduction mechanisms triggered by tissue stretching around apoptotic cells, which involves the Yes-associated protein (YAP) pathway (12).

There is therefore an increasing need to accurately estimate the variability of apoptosis induction in a given cell population. To achieve this, we developed a medium-throughput imaging pipeline using apoptotic cell markers, coupled with mathematical analysis. This allowed us to unveil a neighboring effect for apoptosis induction, irrespective of cell death stimuli or

cell density, whereby the probability of undergoing apoptosis increases in the proximity of apoptotic cells. Hence, the combined application of imaging and computational analysis to evaluate the response to a given apoptotic stimulus may provide new insight into misunderstood phenomena such as fractional killing or apoptosis-induced proliferation, which are major concerns in the context of apoptosis-based cancer therapies.

### **Results**

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#### Live-cell imaging system to trigger and measure multi-stimuli apoptosis

To mathematically analyze the kinetics of apoptotic cell death while varying different parameters, we first set up a cellular system allowing medium-throughput quantification of apoptosis at the cell population level. For this, we used melanoma WM115 cells stably expressing mCherry-tagged histone H2B (H2B-mCherry) to spatially identify the exact topology of a given cell (Figure 1A). Following treatment with an apoptotic stimulus, the induction of apoptosis is easily visualized using the live cell-impermeant dye SYTOX Green (SG). Live cells can thus be identified by the nuclear-localized H2BmCherry-only signal (red nuclei), whereas apoptotic cells appear yellow, owing to the colocalization of H2B-mCherry and green SG (Figure 1A). Next, we used various stimuli to trigger apoptosis. Treatment with TNFα and cycloheximide (CHX) was initially used as a model for death receptormediated apoptosis. Of note, CHX enhances TNFα-induced apoptosis by blocking the translation of short-lived anti-apoptotic proteins (13). Downstream of the TNFα receptor 1/2 (TNFR1/2), the DISC activates caspase 8 and triggers MOMP via BID cleavage (Figure 1B). To induce intrinsic apoptosis, we used two death stimuli: first, a combination of BH3 mimetics ABT-737 inactivating the anti-apoptotic proteins BCL2, BCL-xL and BCL-w, and UMI-77 targeting MCL1; second, a doxycycline (dox)-inducible WM115 cell line to activate the mitochondrial pore-forming protein BAX (Figure 1B). These apoptotic stimuli were then applied to SG-pre-loaded WM115 H2B-mCherry cells plated at different densities and imaged periodically (1 or 2 hours intervals) using the IncuCyte live-cell imager (Figure 1A).

Next, we validated the induction of apoptosis by assessing the processing of effector caspase-3 into p17/p19 fragments and the cleavage of PARP1, which is a hallmark of apoptosis effectiveness. As shown In **Figure 2A** and **B**, treatment with TNFα/CHX (TC), BH3 mimetics (BH3m) and increasing doses of doxycycline efficiently induced apoptosis. Of note, caspase-3 processing and PARP1 cleavage were blocked by treatment with the pan-caspase inhibitor Q-VD-OPh. In addition, IncuCyte-based live-cell imaging was used to delineate the kinetics of apoptosis triggered by the different stimuli described above (**Figure 2C, D**). Apoptotic cells were marked by SG and quantified over time, with image acquisitions every 1-2 hours over a period of 48 hours on average. **Figure 2E** shows this imaging method can efficiently detect and discriminate between living cells (mCherry-positive) and apoptotic cells (SG-positive). The figure presents results of TNFα/CHX and BH3 mimetics treatment, albeit the same was observed for WM115 cells overexpressing BAX (data not shown).

Using this cell model of apoptosis induction, coupled with mediumthroughput live-cell imaging we therefore established an imaging technique able to monitor the heterogeneous induction of apoptosis at the cell population level.

#### Cell clustering has no effect on the incidence of apoptosis

Following different pro-apoptotic treatments (**Figure 1B**) and IncuCyte-based live cell imaging, we obtained datasets of H2B-mCherry images (red images) and the corresponding SYTOX Green staining (green images) at different time points, for a kinetics analysis of 48 hours. We then designed two independent pipelines to segment H2B-mCherry-marked nuclei, one based on the CellProfiler software and the other written in Python, using standard image libraries to ensure a robust analysis (**Figure 1A**). The analysis pipeline included the following steps. First, for a pair of red/green images the output is a segmented "red" image with a known red object centroid position ( $X_i$ ,  $Y_i$ ) onto which a measurement of the green signal is superposed ( $I_i$ , representing the maxima green signal intensity) (**Figure 3A**). Second, for a given time-point t, the analysis pipeline retrieves a set of N red centroid positions ( $X_i$ ,  $Y_i$ ) with their associated SYTOX Green intensity signal  $I_i$ . Data analysis is then

performed using three parameters  $\delta$ , n and  $I_t$ , where  $\delta$  is the inter-cellular distance at which another cell(s) is/are considered to be in contact and could influence each other, n is the threshold number of neighboring cells above which a cell is considered to be in a cluster and  $I_t$  is the threshold intensity of SYTOX Green, derived from background noise, above which the cell is considered to be apoptotic. Of note, two cells are considered neighbors if the distance between their centroids is bellow  $\delta$  (**Figure 3A**).

Once we determined  $\delta$ , n and  $I_t$ , we determined for each cell whether it was apoptotic (A) or not ( $\bar{A}$ ). We then established the relationships of proximity between cells (the so-called neighboring effect) by determining for each cell whether it was in a cluster of cells regardless of their apoptotic status (C) or with a known apoptotic status (Ca) (**Figure 3B**). This method is well suited to quantify spatial relations between apoptotic cells.

Using this approach, we were first interested in determining whether a correlation between cell clustering and apoptosis (between C and A). This is driven by previous work suggesting that cell density is an important modulator of both extrinsic and intrinsic apoptosis (14) (15-17).

By examining three scenarios, in which a given cell was considered to be in a cluster if it was surrounded by 1, 2 or 3 neighboring cells (regardless of their apoptotic status), we found that a cell with no neighbors  $(\overline{C})$  had the same probability of entering apoptosis as a cell in a cluster (of 1, 2, or 3 neighboring cells) (**Figure 3C-E** for TNF $\alpha$ /CHX treatment). We then computed and plotted the time dependent p-value of a Fisher exact test that evaluates the independence between A and C (**Figure 3C-E**, bottom panels). Of note, if the p-value is high (as in **Figure 3C-E**, bottom panels), this means that the hypothesis that A and C are independent cannot be rejected. In other words, this means that the probability of being apoptotic whenever one cell has neighboring relation(s) (probability p(A/C), blue curve in **Fig. 3C-E**, top panels) and the probability of being apoptotic when one cell is isolated (probability  $p(A/\overline{C})$ , orange curve) might be equal.

The same holds true when triggering intrinsic apoptosis with BH3 mimetics (**Figure 3F**) or when overexpressing the pro-apoptotic protein BAX (**Figure 3G**). Though these results were obtained at low cell density, we

confirmed their relevance for cells grown at a high density (Supplementary Figure 1).

Hence, these data suggest that cell clustering has no impact on the occurrence of apoptosis.

## The induction of apoptosis is enhanced by the proximity of a cell to an apoptotic cell

Next, we tested whether the proximity of a cell with one apoptotic cell or an apoptotic cluster (up to three apoptotic neighboring cells) was correlated with its propensity to undergo apoptosis. We obtained results in complete contrast to Figure 3, as the probability to enter apoptosis when in proximity with one or more apoptotic cells (probability p(A/C<sub>a</sub>), blue curve) was significantly higher than the probability to enter apoptosis when isolated from apoptotic cells (probability  $p(A/C_a)$ , orange curve) (Figure 4A-C for TNFα/CHX treatment). Indeed, the p-value (Fisher's exact test) was significantly low (p < 0.05) for the intermediate time points, meaning that both probabilities cannot be considered as equal. We observed a higher p-value at the beginning and the end of the kinetics, which does not affect our analysis, since there are few apoptotic cells at beginning of the time-course and, at the end, most cells are apoptotic and the neighboring effect saturates. This observation also holds true for treatment with BH3 mimetics and BAX overexpression (Figure 4 D, E). Finally, we validated these data at a higher cellular density (Supplementary Figure 2).

Taken together, our analysis pipeline reveals that apoptosis induction is influenced by a neighboring effect: the likelihood of a cell to undergo apoptosis is significantly higher when it is in proximity with apoptotic cells.

#### **Discussion**

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Defective induction of cell death is the cause on many diseases, ranging from cancer to neurodegenerative diseases, and it is therefore important to study its propagative nature using appropriate tools.

Here, we report the development of an analysis pipeline for mediumthroughput imaging of apoptotic cancer cells that integrates several key parameters: temporal dimension, topographic localization of both living (cells stably expressing the mCherry-tagged histone 2B) and apoptotic (SYTOX Green-positive) cells, while applying several apoptotic stimuli.

Using this approach, we found that cancer cells have a similar probability of undergoing apoptosis if they are in proximity or not with other cells of unknown status. In other words, cell crowding does not have a protective effect against apoptosis. However, a cell is more likely to become apoptotic if it is in close contact with apoptotic neighboring cells. This holds true for all of the apoptotic stimuli we tested, either extrinsic or intrinsic inducers of apoptosis. These results are in line with Bhola and Simon's study demonstrating that daughter cells, with a high likelihood of being in proximity, synchronously undergo apoptosis. This synchrony is lost as cells become less related over time (18). Apoptotic synchrony is therefore transiently heritable and is lost due to noise in protein translation (19). To take into account progeny and monitor cell proliferation history, we would need to introduce into our imaging pipeline a lineage tracer such as the one developed by Oren and colleagues (20).

Our experimental settings are in 2D cellular cultures, yet one might apply it to 3D cultures such as cancer organoids. For this, however, the sample will require transparization and 3D imaging microscopy. Here, the proof-of-concept is provided for apoptosis, though the protocol could easily be adapted to other types of programmed cell death, such as caspase-independent cell death, necroptosis or pyroptosis. Importantly, this method could also be helpful in identifying factors facilitating or inhibiting cell death propagation. Similar methods were developed to quantify the occurrence of fractional killing, a barrier for effective cancer treatments, and helped establish mediators such as the anti-apoptotic protein MCL1 (21) (22).

Recent research clearly demonstrates that while apoptosis is immunologically silent, it still involves releasing a plethora of signaling molecules that impinge on the fate of surrounding cells. The term of apoptosis-induced apoptosis was coined in Drosophila, where apoptotic cells in the wing disk triggered apoptosis in trans by secreting Eiger (the TNF homologue in flies), and stimulating pro-apoptotic JNK signaling (23). The same holds true in mammals during the regressive phase of the hair follicle

(23). In cancer, the radiation-induced bystander effect, through which irradiated cells lead to the death of non-irradiated neighboring cells, is also a good example of this concept (24). Interestingly, recent studies revealed the rapid propagation of ferroptotic cell death, and the analysis pipeline described herein might provide further mechanistic insights (25) (26). Inversely, apoptosis-induced proliferation has been unveiled and describes a state in which dying cells release prostaglandins and Wnt3, instructing the proliferation of nearby cells (27) (11). Recently, the group of Tait showed that apoptotic cells also release FGF2 to enhance the anti-apoptotic resilience in trans (10).

The analysis method developed here could increase and enrich our knowledge on "community effects" of cell death so that they can be blocked or enhanced, according to the physio-pathological context.

#### **Material and methods**

275 Cell lines

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- 276 Human melanoma cells WM115 (a gift from R. Insall The Beatson Institute,
- 277 Glasgow, UK) were maintained in DMEM supplemented with 2 mM L-
- 278 glutamine (ThermoFisher Scientific, 25030-24), non-essential amino acids
- 279 (ThermoFisher Scientific, 11140-035), 1 mM sodium pyruvate (ThermoFisher
- 280 Scientific, 11360-039), 10% FBS (Eurobio, CVFSVF00-01) and 1%
- penicillin/streptomycin (ThermoFisher Scientific, 15140-122).
- 283 Stable cell line generation
- 284 For retroviral transduction, Phoenix Ampho 293T cells (2x10<sup>6</sup> in 10 cm in
- 285 diameter Petri dishes) were transfected with pQXIN H2B-mCherry (gift from
- 286 Dr. Luca Fava) using Lipofectamine 2000 (ThermoFisher Scientific,
- 287 11668019). Twenty-four hours later, the virus-containing supernatant was
- 288 filtered and used to infect WM115 cells in the presence of 1 μg/mL polybrene
- 289 (Sigma-Aldrich, H9268). This was repeated with fresh viral supernatant 48
- 290 hours later. Melanoma cells stably expressing H2B-mCherry were further
- 291 purified by cell sorting. WM115 H2B mCherry cells with doxycycline-inducible
- 292 expression of the BAX protein were created using the Sleeping Beauty

- transposon system (28). The plasmids used were piTR1 BAX (gift from David
- 294 Goldschneider) and pCMV(CAT)T7-SB100 (Addgene, 34879) for the
- 295 expression of SB100X transposase. WM115 H2B mCherry cells were then co-
- 296 transfect with piTR1 BAX and pCMV(CAT)T7-SB100 plasmids using
- 297 Lipofectamine 2000 and selected with puromycin (1 μg/mL).
- 299 Western blotting

- 300 Cell lysates were prepared using NP-40 lysis buffer (1% NP-40, 1 mM EDTA,
- 301 150 mM NaCl, 50 mM Tris pH 7.4, 1 mM PMSF, Complete Protease Inhibitors
- 302 (Sigma-Aldrich, 4693116001)). Protein content was determined by Bio-Rad
- assay, then 25-50 µg of proteins were separated on SDS-polyacrylamide gels
- 304 (Biorad) under denaturating conditions (SDS PAGE sample loading buffer
- 305 (VWR, GENO786-701) supplemented with 1 mM DTT) and finally transferred
- 306 onto nitrocellulose. Membranes were probed with the following antibodies at a
- 307 1/1,000 dilution unless otherwise stated: BAX (2772, Cell Signaling),
- 308 Caspase-3 (9662, Cell Signaling), PARP1 (9542, Cell Signaling), β-tubulin
- 309 (2146, Cell Signaling).
- 310 The nitrocellulose membranes were rinsed 3 times for 10 min in TBS-Tween
- 311 0.1% and then incubated with appropriate secondary antibody coupled to
- 312 HRP for 1 hour at room temperature. Finally, the proteins of interest were
- detected using Clarity Western ECL (Biorad, 1705060) and chemiDoc imager
- 314 (Biorad, 17001401).

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- 316 Incucyte imager-based cell viability assay
- 317 Cell viability was determined using an IncuCyte Zoom imaging system
- 318 (Sartorius). Cells were seeded onto Imagelock 96-well plates in media
- 319 containing 30 nM SYTOX Green (Life Technologies, S7020). Following
- 320 different apoptotic treatments, the cells were imaged every 60 or 120 minutes
- 321 and the analysis of the SYTOX Green-positive cells was performed using the
- available IncuCyte image analysis software (Essen Bioscience).
- 324 Modeling the progression of apoptosis

Data description

- 326 Nuclei, set as objects, are delineated at each time step using automatic
- 327 segmentation of mCherry fluorescence ("Red") images via the CellProfiler
- 328 pipeline for primary objects, which is tuned to detect blob-like structures. The
- 329 "apoptosis" signal is obtained by measuring the maximum intensity over each
- 330 nucleus object of the "Green" fluorescence image. A threshold value for this
- 331 intensity I<sub>t</sub> is chosen, which defines the fluorescence intensity considered as
- 332 sufficient for a given nucleus to be considered as apoptotic. In the results
- 333 presented herein, the threshold chosen was 0.2 that corresponds to 2/3 to 1/2
- 334 of the maximum intensity of the Green signal.
- 335 Statistical test
- 336 For a given time frame, segmentation data consist in a sequence of nuclei
- positions (X<sub>i</sub>, Y<sub>i</sub>) and green signal intensity I<sub>i</sub>, with i varying from 1 to N, where
- N is the total number of nuclei objects for this time frame.
- 339 Apoptotic variable A
- 340 Given the intensity threshold I<sub>t</sub>, we defined the class A of apoptotic cells as
- 341 the set of nuclei such that  $I_i \ge I_t$ .
- 342 Clustering (density) C and apoptotic clustering Ca variables
- 343 In order to separate "clustered" cells from isolated ones, we set a
- 344 characteristic interaction distance  $\delta$  and a threshold n on the number of
- neighbors. An object at position  $(X_i, Y_i)$  is considered to be part of a cluster
- 346 (i.e., in the class C) if there are a least n nuclei inside the disk of radius  $\delta$  and
- 347 center  $(X_i,Y_i)$ . An object at position  $(X_i,Y_i)$  is considered to be part of an
- 348 apoptotic cluster (i.e., in the class C<sub>a</sub>) if there are a least n apoptotic nuclei
- 349 (*i.e.*, in the class A) inside the disk of radius  $\delta$  and center  $(X_i, Y_i)$ .
- 350 Independence tests
- We used classical Fisher's exact test to determine if random variables A and
- 352 C are independent and subsequently if random variables A and C<sub>a</sub> are
- independent. In this setting, for the simplicity of notation, we identify the class
- 354 A with the random variable that indicates if an object belongs or not to the
- 355 class A, and similarly for classes C and C<sub>a</sub>. We gave exact *p*-values for this
- test. Generally, the null hypothesis (the considered pair of random variables

- are independent of each other) can be rejected when the *p*-value is lower than
- 358 0.05.

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#### **Author contributions**

- 367 Conceptualization, G. Ichim, D. Coulette, H.L. and E.G.; Methodology, G.
- 368 Ichim, D. Coulette, H.L. and E.G.; Formal analysis, G. Ichim, D. Coulette and
- 369 H.L.; Investigation, G. Ichim, D.C., D. Coulette and H.L.; Resources, G. Ichim,
- 370 D. Coulette, H.L. and E.G.; Writing Original Draft and Editing, G. Ichim; All
- 371 authors reviewed and edited the manuscript; Supervision, G. Ichim, D.
- 372 Coulette, H.L. and E.G.; Project administration and funding acquisition, G.
- 373 Ichim.

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### 374 Figure legends

- 375 Figure 1. Imaging and quantification of apoptosis induction using live
- and dead cell markers.
- 377 **A**. Overview of the live/dead cell imaging and computer analysis protocol.
- 378 **B.** Apoptotic signaling induced by treatment with TNFα/CHX or BH3 mimetics,
- 379 or through BAX overexpression (aCasp8 active Caspase-8; aCasp 3/7 -
- 380 active Caspase-3 and -7, tBID truncated BID).

#### 382 Figure 2. Validation of the apoptosis induction protocols.

- 383 **A**. WM115 H2B-mCherry cells were treated either with TNF $\alpha$ /CHX (TC) (50
- $^{384}$  ng/mL of TNFα and 5  $\mu$ g/mL CHX) or BH3 mimetics ABT-737/UMI-77 (10  $\mu$ M
- as each) for 12 hours in the presence or absence of the pan-caspase inhibitor Q-

- 386 VD-OPh (10 μM). Protein lysates were then probed for the expression and
- processing of PARP1 and Caspase-3. Tubulin was used as a loading control.
- 388 **B**. WM115 H2B-mCherry cells were treated with 1 or 2.5 µg/mL of doxycycline
- 389 (dox) to induce BAX expression for 12 hours in the presence or absence of Q-
- 390 VD-OPh (10  $\mu$ M) and then analyzed as described in (A).
- 391 **C-D**. IncuCyte imager-based quantification of apoptotic WM115 H2B mCherry
- 392 cells (SYTOX Green-positive cells) following treatment with TNFα/CHX (TC),
- 393 ABT-737/UMI-77 (**C**) or doxycycline (at 1 or 2.5 μg/mL) (**D**).
- 394 E. Representative images of phase-contract, H2B-mCherry and SYTOX
- 395 Green signal from WM115 H2B-mCherry cells treated as described in (A).
- 397 Figure 3. Apoptosis occurs independently of cell clustering.
- 398 **A-B**. Definition of the parameters taken into account for the mathematical
- analysis of the effect cell clustering has on apoptosis induction. Cells with the
- 400 nucleus half in red half in green are either alive or apoptotic, respectively.
- 401 **C-E**. Probability estimates (upper panels) and the corresponding *p-value*
- 402 (lower panels) for the occurrence of apoptosis in TNFα/CHX-treated WM115
- 403 H2B-mCherry cells surrounded by 1, 2 or 3 neighboring cells.
- 404 **F-G**. Similar analysis as in (C-E), for WM115 H2B-mCherry cells treated with
- 405 ABT-737/UMI-77 (**F**) and doxycycline (**G**). The neighboring effect for 1 cell is
- 406 shown.

416

- 408 Figure 4. Apoptotic cell clusters influence the life-death decisions in
- 409 neighboring cells.
- 410 **A-C**. Probability estimates (upper panels) and the corresponding *p*-value
- 411 (lower panels) for the occurrence of apoptosis in TNFα/CHX-treated WM115
- 412 H2B-mCherry cells surrounded by 1, 2 or 3 apoptotic neighboring cells.
- 413 **D-E**. Similar analysis as in (A-C), for WM115 H2B-mCherry cells treated with
- 414 ABT-737/UMI-77 (D) and doxycycline (E). The neighboring effect for 1
- 415 apoptotic cell is shown.
- 417 Supplementary Figure 1 (related to Figure 3).

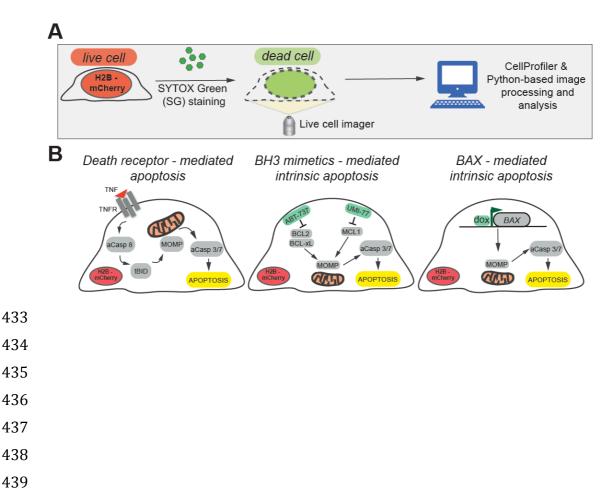
- 418 A, B. Probability estimates (upper panels) and the corresponding p-value
- 419 (lower panels) for the occurrence of apoptosis in ABT-737/UMI-77-treated
- 420 WM115 H2B-mCherry cells surrounded by 2 or 3 neighboring cells.
- 421 **C, D**. Similar analysis as in (A, B), albeit apoptosis was triggered by
- 422 doxycycline induction of pro-apoptotic BAX protein.
- 424 Supplementary Figure 2 (related to Figure 4).
- 425 **A, B**. Probability estimates (upper panels) and the corresponding *p*-value
- 426 (lower panels) for the occurrence of apoptosis in ABT-737/UMI-77-treated
- 427 WM115 H2B-mCherry cells surrounded by 2 or 3 apoptotic neighboring cells.
- 428 **C, D**. Similar analysis as in (A, B), albeit apoptosis was triggered by
- 429 doxycycline induction of pro-apoptotic BAX protein.

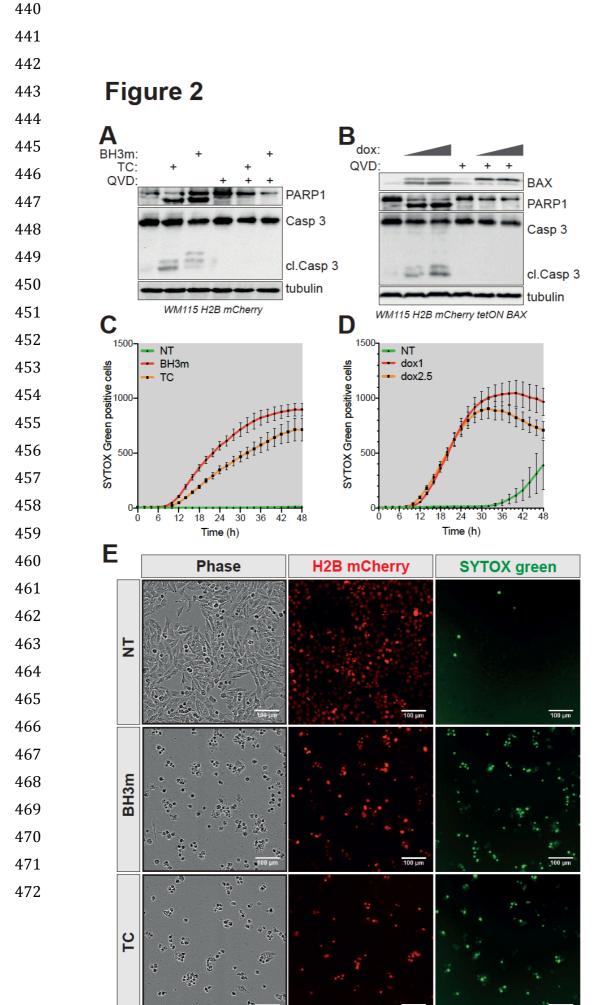
## Figure 1

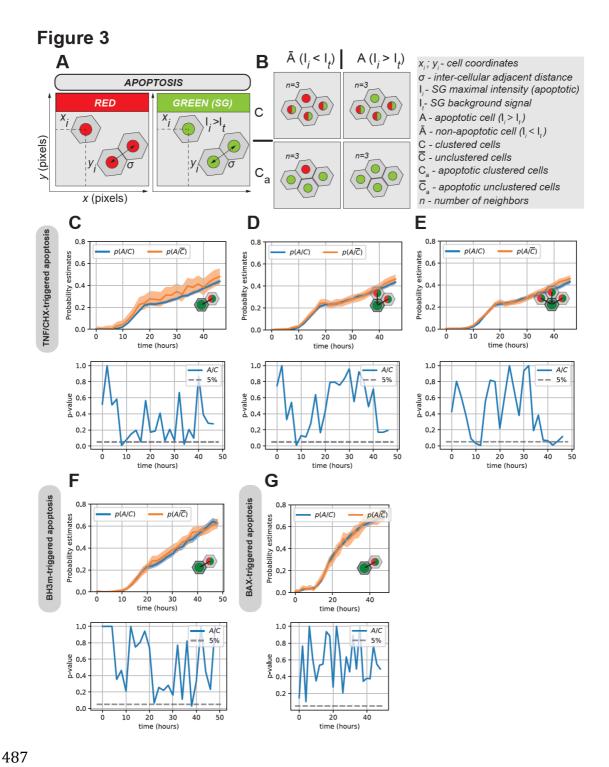
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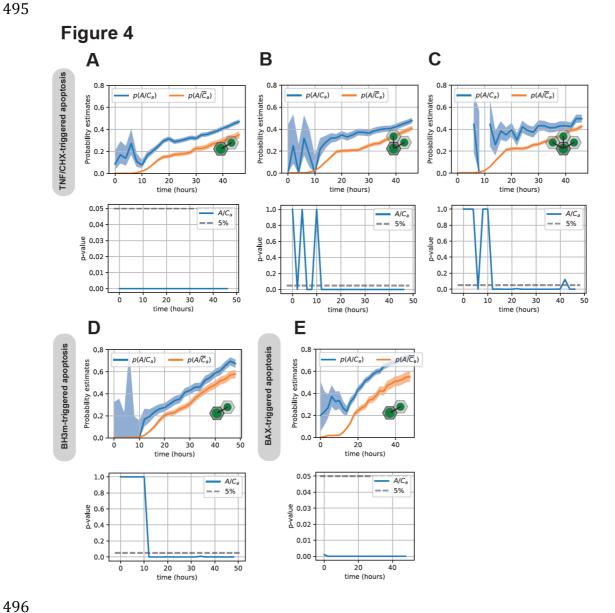
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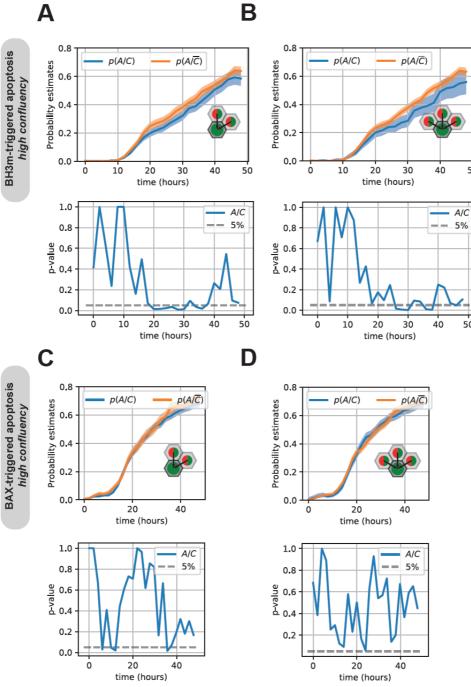




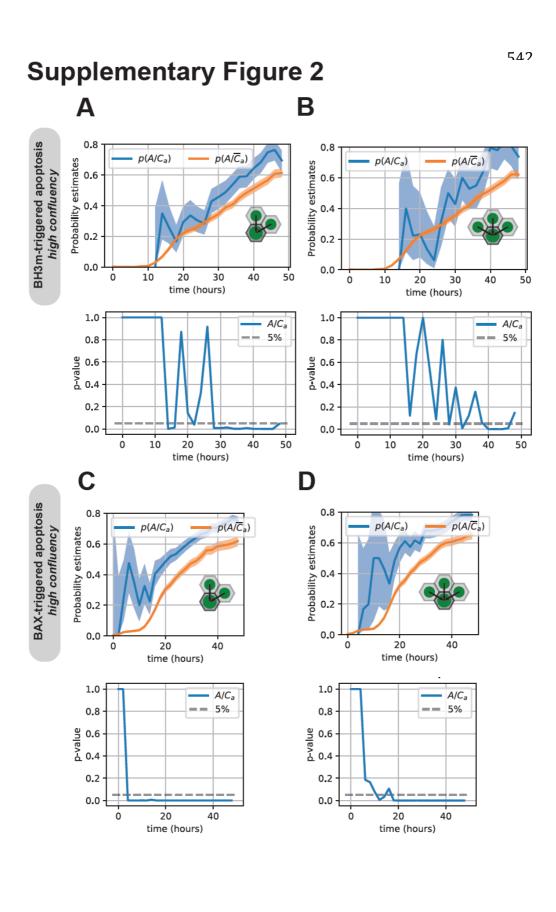




## Supplementary Figure 1



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