## Supplementary material

#### Aggression scores

<span id="page-0-0"></span>

<b>Uncontrolled Trajectory Approximations</b>																
	N.I	<b>KRAS</b>			TP53   CycD   SMAD	T-K	$C-K$	S-K	T-C	T-S	$C-S$	$T-C-K$				$T-S-K$   $C-S-K$   $T-C-S$   $T-C-S-K$
Aut.	0.509	0.531	0.964	0.818	0.528	0.974	0.827		$0.356 \mid 0.953 \mid$	$0.955$ 0.817		0.97	0.967	0.82	0.962	0.961
Apo <sub>c</sub>	0.359	0.356	0.021	0.109	0.343		$0.011 \mid 0.118 \mid$		$0.479 \mid 0.025 \mid$	$0.035 \mid 0.105$		0.018	0.013	0.1	0.028	0.017
$Pro_{c}$	0.213	0.209	0.56	0.017	0.211	0.555	0.013	0.24	0.018	0.92	0.012	0.015	0.944	0.018	0.01	0.014

Supplementary Table 1: *Expression Approximations*. This table records the approximate phenotype expressions for the PCC in Figure [1.](#page-4-0) Given 1,000 random initial states, these results show trajectory approximations after 300 time steps (i.e. function updates) with  $1\%$  noise.

We derived aggressiveness scores for each mutation combination using long-term trajectory approximations. Simulations were run using 1000 random initializations, 300 time steps, and 1% noise to achieve an approximate probability of phenotype expression. In Table, [1,](#page-0-0) we see that the non-induced (N.I.) system showed levels of 51% autophagy, 36% apoptosis, and 21% proliferation. The heat maps in the manuscript are sorted with column-wise mutation groups and used to compare cancer cell autophagy and proliferation while giving a negative weight ( $\omega = -1$ ) to apoptosis. The row label "Same" indicates that the same weight was given to both autophagy and proliferation (used value  $\omega = 2$  for both), "High/Low" indicates a high weight for autophagy ( $\omega = 10$ ) but a low weight for proliferation ( $\omega = 2$ ), and "Low/High" indicates a low weight for autophagy ( $\omega = 2$ ) but a high weight for proliferation ( $\omega = 10$ ). Thus, scores were calculated using:

$$
Score = Aut_c \times \omega_1 + Pro_c \times \omega_2 + App_c \times (-1) \quad \text{where} \quad \omega_{1,2} \in \{2, 10\}
$$

Scaling of the heat map ranges orange (low score) to red (high score) based on the maximum and minimum values in each table. However, blue shading (i.e. cold) indicates a negative score, which is interpreted as successful depletion of aggression. See [\[1\]](#page-9-0) sections 2.3 and 4.4 for more details.

Lastly, we justify the positive weight given to autophagy, which is a natural process where cells heal themselves. The cell will break down any damaged or unnecessary components, and it will reallocate the nutrients from these processes to those that are essential. However, studies have shown that autophagy is required for pancreatic tumor growth [\[2\]](#page-9-1). Autophagy can help tumors overcome conditions such as hypoxia and nutrient deprivation. Within tumors, cells can exist under hypoxic conditions. If activated autophagy is then suppressed by deletion of Beclin 1, studies have shown increased cell death. It has also been observed that autophagy is increased in KRAS mutated cells, and aids in survival of the cancer cells while experiencing nutrient starvation. Further, animal studies have shown that autophagy contributes to tumor-cell survival by enhancing stress tolerance and supplying nutrients to meet the metabolic demands of tumors. Once suppression of autophagy occurred, there was an observance of tumor-cell death [\[3\]](#page-9-2).

Note: our aggression scores are based on combinations autophagy, apoptosis, and proliferation, merely one method among many for estimating aggression. Moreover, the attractor analysis (see [\[4,](#page-9-3) [1\]](#page-9-0) indicated that certain mutation combinations yield a large basin for attractors with both autophagy and proliferation expression. It is likely that modular structure alone is not enough to determine aggression and target cardinality. Rather, it should be used alongside other analyses.

# Boolean pancreatic cancer model and functions





Supplementary Table 2: *Boolean functions for the whole pancreatic cancer model.* Each function indicates the next state of the node in terms of the current states of said nodes' regulators. Activation is written as OR statements, while suppression is written as AND NOT. The exception to this rule is PCC proliferation, because of its upstream signaling.



Supplementary Table 3: *Boolean functions for the reduced pancreatic cancer model.* Each function indicates the next state of the node in terms of the current states of said nodes' regulators. Activation is written as OR statements, while suppression is written as AND NOT. Functions maintain the rules from the whole model by substituting values from the deleted nodes.

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#### Supplementary Figure 1: *Gene regulatory network model of pancreatic cancer.*

Shapes and colors of nodes indicate their function and cell type (respectively), as shown in the legend. Black barbed arrows indicate signal expression, while red bar arrows indicate suppression. Grey nodes located in the PCC indicate prevalent mutant genes [ [4](#page-9-3) , [5\]](#page-9-4).

## Tables and Graphs



Supplementary Figure 2: *Module counts*

Supplementary Figure 3: *PC condensation graphs.* Included are all condensation graphs for each mutation combination. These are directed, acyclic graphs that are topologically sorted, and whose nodes represent the strongly connected components of Figure [1.](#page-4-0) Colors of nodes are based on the components they represent, and node numbers correspond to bin numbers (see data files in the repository).



(a) Wild-type condensation graph. (b) KRAS condensation graph.







(c) TP53 condensation graph. (d) CyclinD condensation graph.



(e) SMAD condensation graph. (f) T/K condensation graph.









(g) C/K condensation graph. (h) S/K condensation graph.



(i) T/C condensation graph. (j) T/S condensation graph.





(m) T/S/K condensation graph. (n) C/S/K condensation graph.





(k) C/S condensation graph. (l) T/C/K condensation graph.





(o) T/C/S condensation graph. (p) T/C/S/K condensation graph.

## References

- <span id="page-9-0"></span>[1] Plaugher, D., Aguilar, B. & Murrugarra, D. Uncovering potential interventions for pancreatic cancer patients via mathematical modeling. *Journal of Theoretical Biology* 548, 111197 (2022). URL <https://www.sciencedirect.com/science/article/pii/S0022519322001953>.
- <span id="page-9-1"></span>[2] Yang, S. *et al.* Pancreatic cancers require autophagy for tumor growth. *Genes* & *Development* 25, 717–729 (2011). URL <http://genesdev.cshlp.org/content/25/7/717.abstract>. <http://genesdev.cshlp.org/content/25/7/717.full.pdf+html>.
- <span id="page-9-2"></span>[3] Yun, C. W. & Lee, S. H. The roles of autophagy in cancer. *International Journal of Molecular Sciences* 19 (2018). URL <https://www.mdpi.com/1422-0067/19/11/3466>.
- <span id="page-9-3"></span>[4] Plaugher, D. & Murrugarra, D. Modeling the pancreatic cancer microenvironment in search of control targets. *Bulletin of Mathematical Biology* 83 (2021).
- <span id="page-9-4"></span>[5] Wang, Q. *et al.* Formal modeling and analysis of pancreatic cancer microenvironment. In Bartocci, E., Lio, P. & Paoletti, N. (eds.) *Computational Methods in Systems Biology*, 289–305 (Springer International Publishing, Cham, 2016).