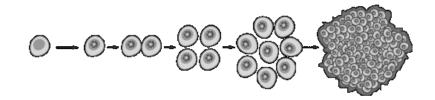
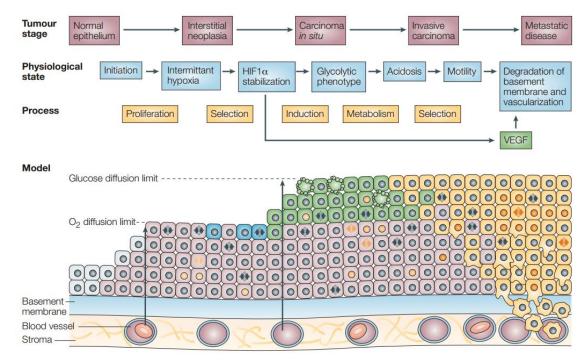


CANCER GROWTH





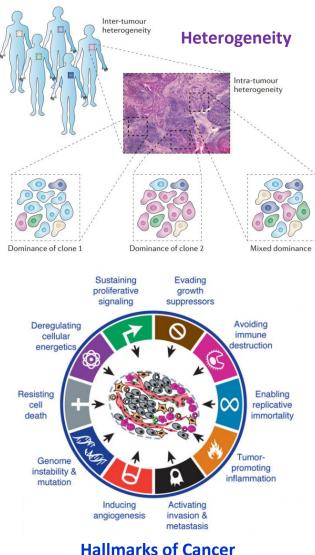
Model for cell–environment interactions in carcinogenesis. Early carcinogenesis proceeds **from normal tissues** through initiation to a hyperplastic state to interstitial neoplasia, progressing to carcinoma *in situ*. Until this stage, epithelial cancers are **avascular**, as shown by histopathology. Following **breakdown of the basement membrane**, cells gain access to existing and newly formed blood and lymphatic vascular routes for **metastasis**. The stages of tumor growth and their associated physiological states are diagrammed, showing that progression from one stage to the next is governed by state processes. **Normal epithelial cells** become hyperproliferative following induction. As they reach the oxygen diffusion limit, they become hypoxic, which can either lead to cell death (apoptotic cells shown with blebbing) or adaptation of a **glycolytic phenotype**, which allows cells to survive. As a consequence of glycolysis, lesions become acidotic, which selects for **motile cells** that eventually breach the basement membrane. As cancer progression proceeds, the **mutations** in cells increase (nuclei shown as light orange for one mutation and darker oranges for more mutations). **HIF1α**, hypoxia-inducible factor-1α; **VEGF**, vascular endothelial growth factor.

R. A. Gatenby and R. J. Gillies, Nature Reviews Cancer: 4(11):891-9., Nov 2004

CANCER IS COMPLEX

Cancer is a complex biological system

- involving many interconnected biological phenomena
- spanning in different spatial and temporal scales
- showing high inter- and intra-tumoral heterogeneity
- characterized by adaptation in terms of plasticity, resistance and relapse



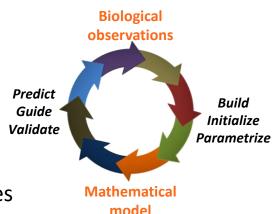
A. Marusyk, V. Almendro, and K. Polyak, Nature Reviews Cancer: 12, 323-334, 2012 D. Hanahan and R. A. Weinberg, Cell 144, March 4, 2011

HOW IN SILICO MODELS CAN HELP?

Mathematical and computational model

- allow many hypotheses to be explored
- incorporate multiple information from multiple sources
- handle multiple coupled variables and explore the role of the parameters involved
- track each individual component of the biological system as evolves in time and space, as well as the behavior of the whole multicellular system
- predict the systems behavior in a variety of conditions, which is often difficult to achieve experimentally
- **produce** new hypotheses
- guide experimental planning
- validate the hypotheses with the biological observations in an iterative manner until a faithful and reliable model is found!



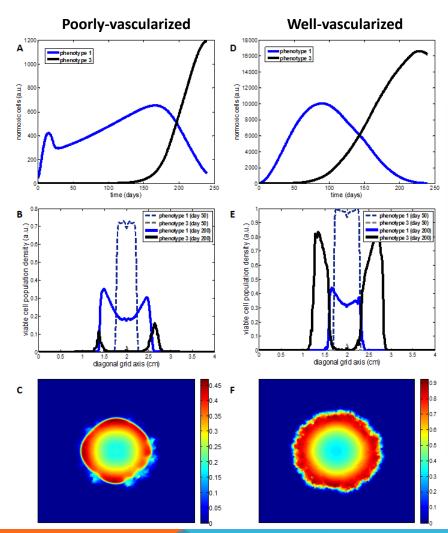


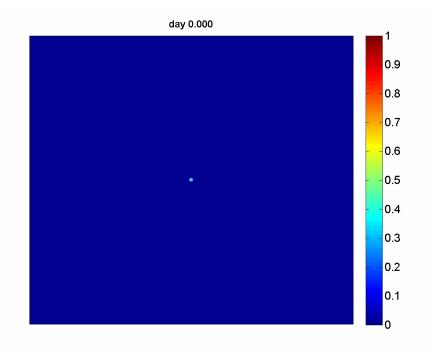
MODELING APPROACHES ADOPTED

A. Macroscopic, tissue-level model

- Deterministic, continuous approach: mean field average in space and time
 - A system of coupled reaction-diffusion-chemo/hapto taxis equations
- Describes the interactions between:
 - Cancer cells (which are in different states i.e. normoxic, hypoxic, necrotic and can have different properties i.e. more/less proliferative and more/less invasive)
 - Tumor microenvironment (i.e. vasculature and ExtraCellular Matrix (ECM))
- Applied in brain tumors (Hinow et al., Math Biosci Eng. 6(3), 2009, Swanson et al., Cancer Res, 71(24), 2011)
- Extended to account for polyclonal cell populations (Tzamali et al. PlosOne 2014)
 - two cancer subpopulations (a more proliferative and a more invasive) form the initial tumor and have distinct phenotypic characteristics that do not change throughout tumor evolution
 - these cancer populations are in a constant competition for space and resources within the tumor microenvironment
- Solved in 2D using the Alternative Direction Implicit (ADI) method of finite differences on regular lattice
- Limitations
 - Biomechanical description of the tumor microenvironment is NOT considered in the current setup
 - The whole is not always the sum of its parts! -Aristotle

PROLIFERATIVE & INVASIVE PHENOTYPES



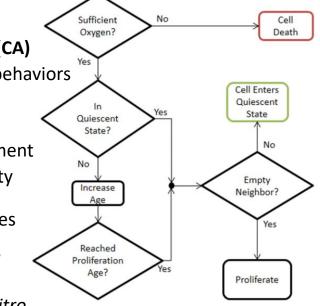


E. Tzamali, G. Grekas, K. Marias, V. Sakkalis, PlosOne, 2014

MODELING APPROACHES ADOPTED

B. Individual, cell-based model

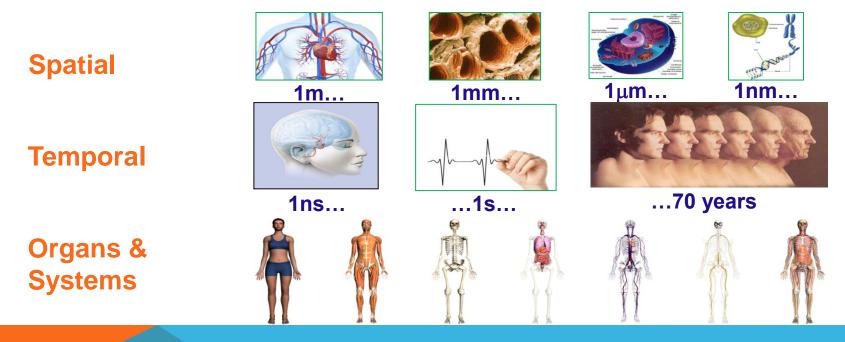
- Discrete or hybrid, discrete-continuous (HDC) approach
- Each cancer cell is described by a discrete cellular automaton (CA)
 - CA are capable of producing a great variety of unexpected behaviors
 - Cells follow biologically-or experimentally inspired rules
 - Cells are allowed to divide, move or die
 - Cells interact locally with each other and the microenvironment
 - Movement can be stochastic and can adopt different motility mechanisms ranging from individual to collective strategies triggered by different cell-cell and cell-matrix adhesion forces
- The chemicals of the tumor microenvironment are continuous variables described by partial differential equations
- Individual cell-based models are more suitable to describe *in vitro* experiments and small-sized tumors
- Solved in 2D regular lattice
- Limitations: Computationally expensive



Anderson A. et al., Cell 127, 905-915, 2006 | Tzedakis G., Tzamali E., Marias K., Sakkalis V., Cancer Informatics 14(Suppl 4):67-81, 2015 | Tzedakis G., Liapis E., Tzamali E., Zacharakis G., Sakkalis V., IEEE-EMBC, 2016 | Oraiopoulou M.E., Tzamali E., Tzedakis G., Vakis A., Papamatheakis J., Sakkalis V., Hindawi Biomed Research International, Basic and Translational Advances in Glioblastoma (Special Issue), 2017 (accepted paper)

ADVANCED IN MULTI-LEVEL COMPUTATIONAL MODELS, BUT DATA POOR

- In silico models are detailed enough, but lack parameters that are roughly estimated or taken for granted as theoretical values based on bibliography.
- We are data poor in most clinical settings! Only some of these are physiological parameters available in routine examinations.

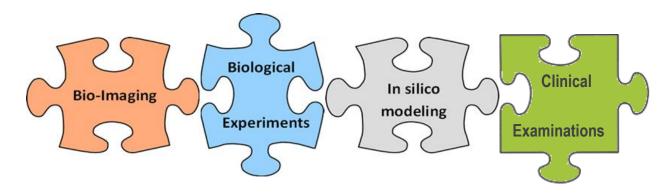


INCREASE PREDICTIVE POWER AND CLINICAL RELEVANCE

Collaboration between experimentalists, modelers and clinicians

- Develop mathematical models that faithfully represent the biological processes.
- Design and conduct biological experiments with the aim of parametrizing/constraining the mathematical models
- Both *in vitro* and *in vivo* studies are important steps of model parametrization and validation before translating to clinical cases

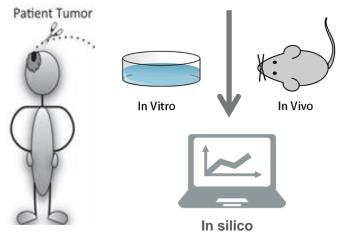
Incorporate information derived from advanced imaging techniques into *in silico* models



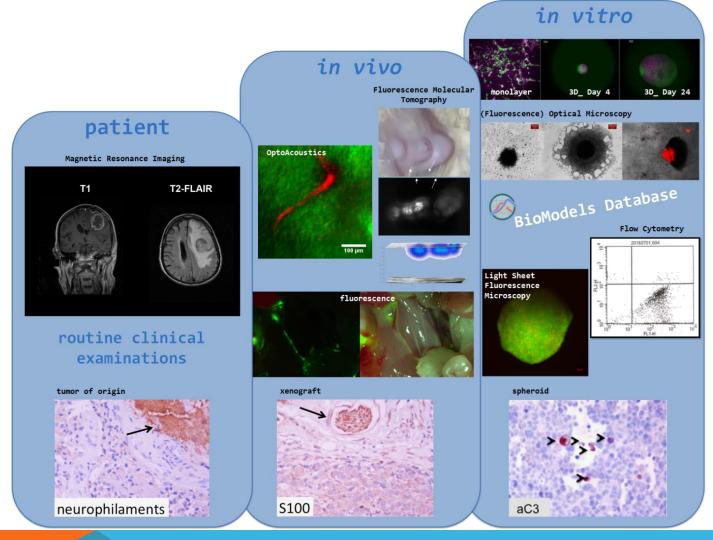


CURRENT RESEARCH INTEREST

- Focus: Glioblastoma (GB), a malignant brain tumor with extremely poor prognosis for the patient. Therapy treatment in GB patients fails mainly due to its extensive inter- and intra-tumoral heterogeneity and its high recurrence potential.
- **Approach:** We combine *Basic and Translational Research* such that clinical, *in vitro* and *in vivo* **patient-specific** data are used to initialize, parametrize and validate GB predictive mathematical models.
 - <u>Patient-derived GB cells</u> are collected during the gross resection or biopsy of patients when still naive from treatment
 - Patient-derived xenografts
 - In vitro primary cultures

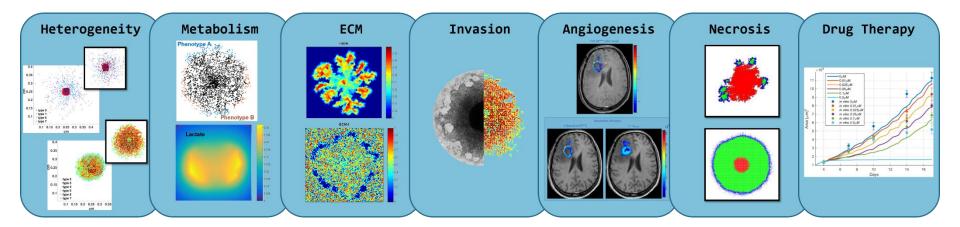


AVAILABLE DATA PER GB PATIENT



FUTURE WORK

The integration of **computational approaches** with **experimental and clinical data**, along with advanced **imaging techniques** is more than important in understanding **brain tumor pathophysiology/pathobiology** and tumor progression in a **patient-specific way**. Our main aim is to *design a preclinical drug screening tool promoting effective personalized therapy*.



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