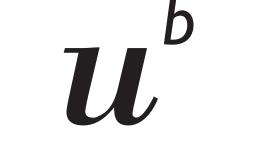


# **Towards a Framework for Predictive Mathematical Modeling of the Biomechanical Forces causing Brain Tumor Mass-Effect**

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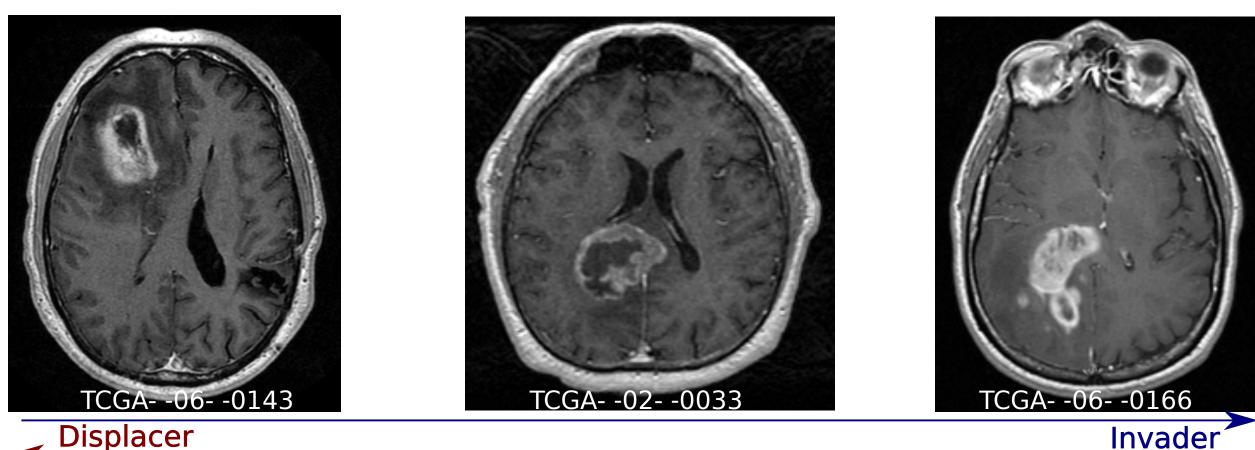


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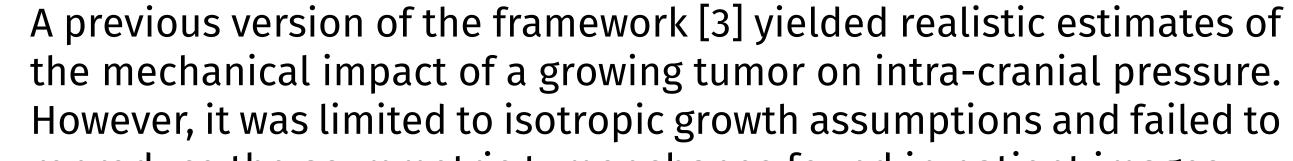
### Introduction

Glioblastoma multiforme (GBM) present with a range of growth **phenotypes** [1], from predominantly invasive tumors without notable "mass-effect" to strongly displacing lesions that induce high mechanical stresses resulting in healthy-tissue deformation, midline shift or herniation. Biomechanical forces shape the tumor microenvironment by compression of blood and lymphatic vessels, reducing blood perfusion and generating hypoxia [2]. We expect these forces to be **important for tumor evolution**, for the formation of distinct growth phenotypes and tumor shape.



With the aim to quantitatively characterize different growth phenotypes, to better understand the role of mechanical forces in their formation and to study possible implications for treatment, we started developing a **framework for GBM growth simulation**:

Its underlying mathematical model accounts for the biomechanical stresses induced in the tissue and thus allows simulation of GBM's **invasive growth characteristics** as well as the **mass-effect** caused by the growing tumor.



#### **Mathematical Model**

Cell proliferation and invasion are modeled as **reaction-diffusion** process; the simulation of the mechanic interaction relies on a linearelastic material model. Both are **coupled** by relating local **tumor cell** concentration to the generation of strains in the tissue. The model accounts for multiple brain regions and incorporates information of structural tissue anisotropy.

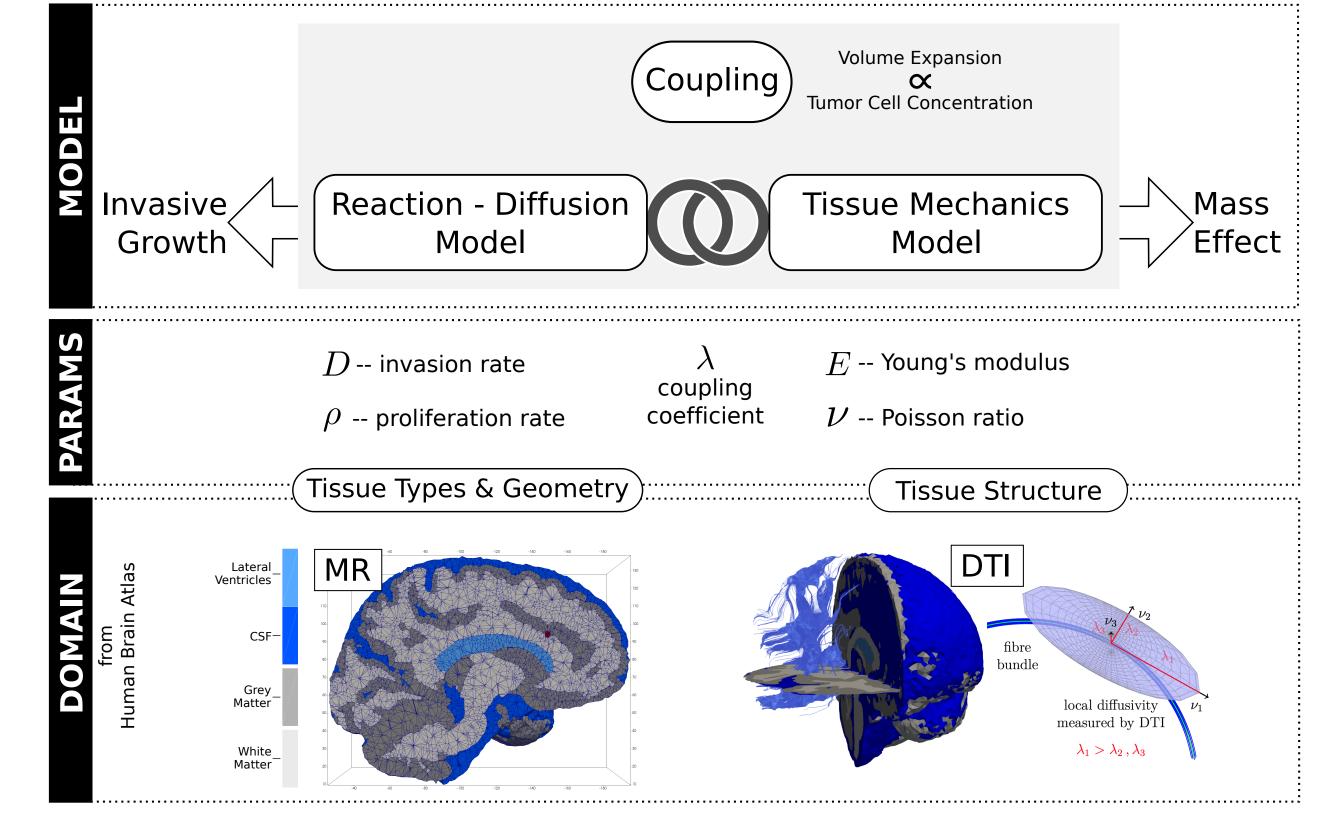


Figure 1: GBM growth phenotypes with varying degrees of mass-effect.

reproduce the asymmetric tumor shapes found in patient images.

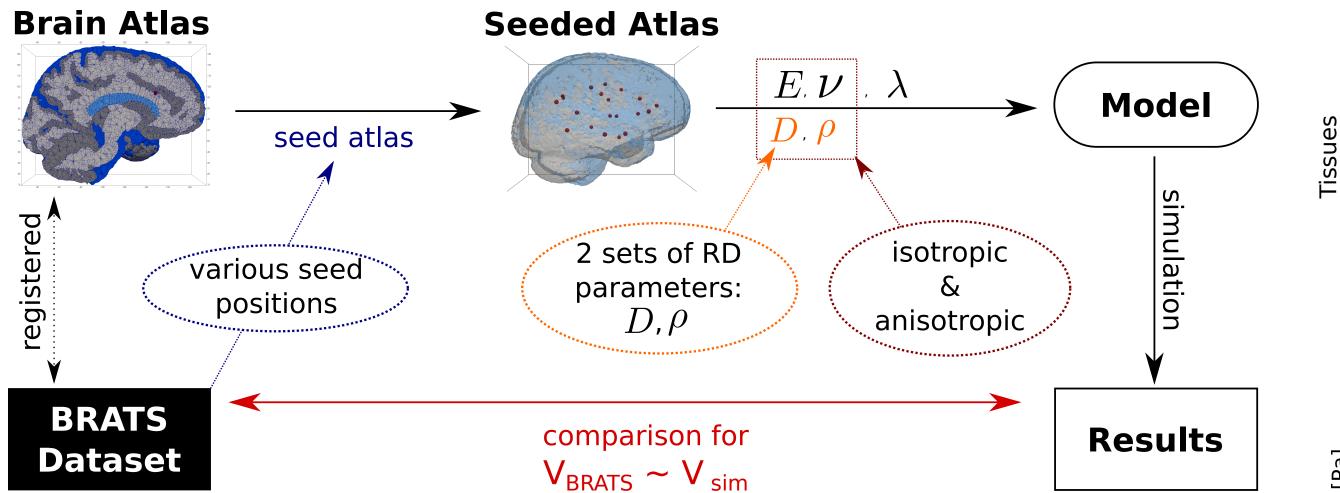
Here we present an **extended version** of this model that accounts for the **anisotropic orientation of axons** in white matter using information from Diffusion-Tensor-Imaging (DTI). This structural anisotropy is known to affect the preferred directionality of tumor cell migration and the **mechanical behavior** of the tissue.

# **Parametric Simulation Study**

## **Study Workflow**

## Tumor growth simulation for

- **multiple seed locations** derived from tumor segmentations,
- invasive (large  $D/\rho$ ) and nodular (small  $D/\rho$ ) growing tumors,
- **isotropic** and **anisotropic** model versions.
- **Comparison** of **simulated** to **actual tumor** at imaged volume.



# **Simulated Tumor Evolution & Mechanical Impact**

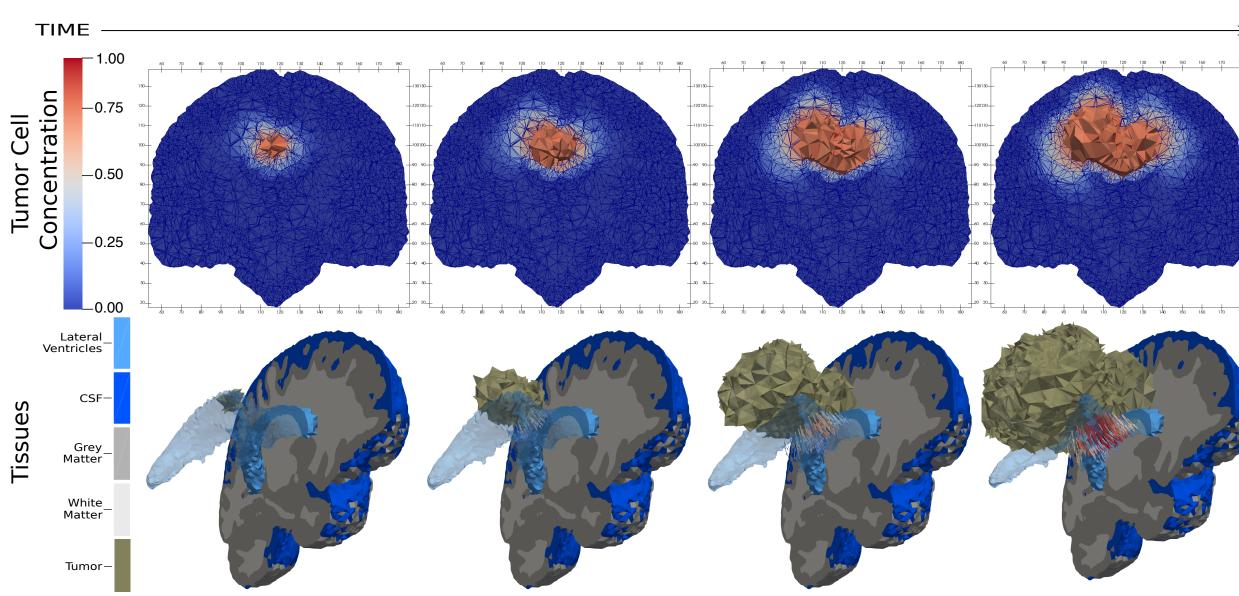


Figure 6: Mechanically-coupled Reaction-Diffusion Model: Structure & Inputs.

# **Summary & Discussion**

The mathematical model captures invasive growth characteristics of GBM and the biomechanical stresses caused by tumor growth. Simulations yield realistic tumor volumes and estimates of mechanical impact. Simulated tumor shapes are more symmetric than the corresponding real lesions. Accounting for brain tissue structure re**duces symmetry** of simulated lesions on average, however, not to the level observed in GBM patient data.

#### Model & Study Limitations:

- Parametrisation and growth domain not personalized.
- No account of vasculature and growth promoting/inhibiting

Figure 2: Parametric simulation study of tumor evolution, varying seed positions and RD parameters, with and without accounting for tissue anisotropy.

## **Parameter Assumptions**

#### Literature-derived parameter values

► **Isotropic**, no account of tissue structure:

$\overline{D/\rho}$	ρ	$D_G$	$D_W$
	[1/d]	[mm²/d]	[mm <sup>2</sup> /d]
low	0.082	0.020	0.101
high	0.037	0.040	0.200
			arameters
D, deri	ived fror	n clinical	study

data [4–7], by  $D/\rho$  category.

Tissue	E	ν
	[kPa]	
W/G Matter	3.0	0.45
Tumour	6.0	0.45
CSF (Ventricles)	1.0	0.30
CSF (other)	1.0	0.49
(b) Mechanical tissu	ie prop	perties

Table 1: Parameter choices for isotropic model.

- ► Anisotropic, (transverse isotropic) White Matter structure:
- Grey Matter parameters as in isotropic case.
- Higher motility along fibres:  $D_W^{\parallel} = D_W^{iso}$ ,  $D_W^{\perp} = 0.01 \cdot D_W^{iso}$
- Stiffer (tensile) along fibres:  $E_W^{\parallel} = 3 \cdot E_W^{\perp}$  (details in [9])

Figure 3: Simulated tumor evolution. Threshold for solid tumor c > 0.8.

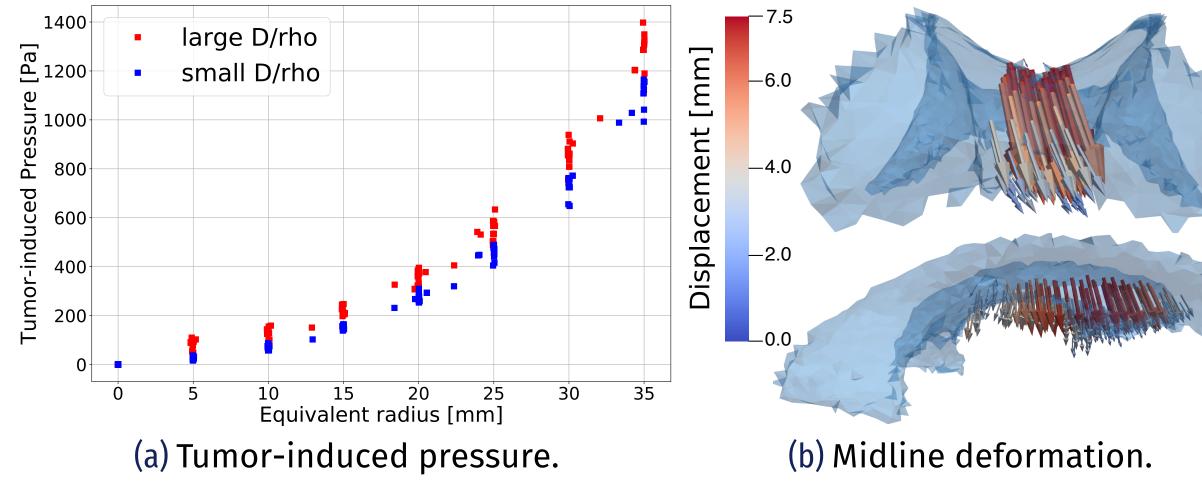
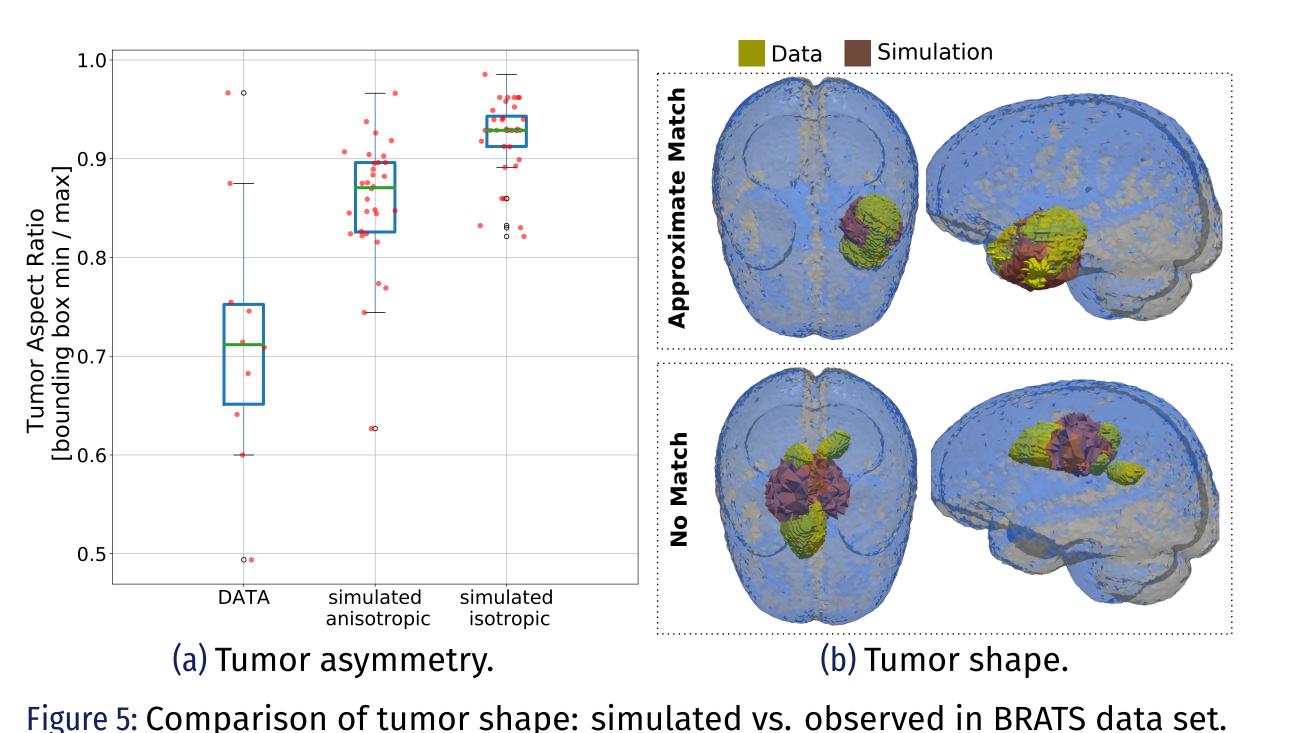


Figure 4: Simulated mechanical impact of growing tumor.

# **Tumor Shape**



- factors in tumor micro-environment.
- Assumption of linear-elastic mechanical material model.

# Outlook

Further model testing and development in animal study. Model personalization to enable **patient-specific characterization** of distinct "invasive" and "displacive" growth phenotypes.

## **Further Information**



www.glims.ch

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#### Maximum volumetric growth of 15 %: $\lambda = 0.15$ [10]. **Boundary conditions** constrain surface flux & nodes. Atlas of normal human brain anatomy (SRI24) [11], MR & MR-DTI. **Image data** of high-grade glioma patients from BraTS 2013. **Implementation** via Finite Element Method (FEM). Calculations were performed on UBELIX, the HPC cluster at the University of Bern.

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