Quantitative imaging for personalized mathematical models in oncology

David A. Hormuth II, Angela M. Jarrett, Chengyue Wu, Thomas E. Yankeelov

Dynamic Contrast Enhanced MRI



Plasma volume fraction, perfusion, extracellular extravascular volume fraction



Low cell density, Restricted water diffusion



High cell density, Restricted water diffusion





Our central hypothesis Readily-available imaging techniques can provide the data to initialize/constrain predictive models of tumor growth and treatment response for clinical application.

Patient-Specific Model of Breast Cancer Hemodynamics

Chengyue Wu, David A. Hormuth II, Federico Pineda, Gregory S. Karczmar, Robert D. Moser, Thomas E. Yankeelov

- Dynamic contrast-enhanced (DCE-) MRI and Diffusion-weighted (DW-) MRI are collected for breast cancer diagnosis.
- DCE-MRI provides the geometries and pharmacokinetic properties of vasculature, while DW-MRI provides estimates of tissue hydraulic conductivity
- A 1D-3D coupled computational fluid dynamic model is established and personalized using patients' imaging data to simulated tumorassociated blood supply and interstitial transport. The computation quantifies hemodynamic characteristics, including blood pressure (p_v), vascular extraction rate (q_e), interstitial pressure (p_t) and interstitial flow velocity (u_t).
- Our results in a 16-case clinical cohort indicate that malignant lesions tend to have larger magnitude of interstitial flow velocity, and greater heterogeneity in blood pressure and vascular extraction rate. These preliminary analyses suggest a fundamentally new way to employ contrast agent pharmacokinetics for the evaluation of breast cancer.



Predicting response to fractionated radiation therapy

David Hormuth, Angela Jarrett, Thomas Yankeelov

Diffusion-weighted MRI and Dynamic contrast-enhanced MRI are collected at 10 points before, during, and after radiation therapy

DW-MRI provides estimates of cellularity, while DCE-MRI provides estimates of blood volume fraction

A coupled, 3D model of tumor growth and angiogenesis is initialized and calibrated from the DW- and DCE-MRI measurements for each animal. The calibrated parameters are then used to predict future tumor growth

Our model (model 1) out performs the standard linear quadratic model (model 2) in describing and predicting future response



Predicting response to chemotherapy

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DW-MRI and DCE-MRI are collected during neoadjuvant Prediction Scan 3 PCC = 0.92 ^predicted cell number (NAT) chemotherapy for breast cancer patients Number of tumor cells ($\times 10^{6}$ DW-MRI provides estimates of cellularit provides a means to approximate dru the tumor tissue. 1.5 2.5 Measured cell number ×10⁶ A 3D model of tumor growth is calibra[.] Predicted volume (mm³) PCC = 0.90and DCE-MRI measurements from early each patient's NAT regimen, and parameters are used in the model t growth at the completion of therapy fc patient. Predicted percent change in longest axis 1.5 Measured volume (mm³)×10⁴ The model's predictions are found to be strongly N = 8Predicted longest axis (mm) -10 PCC = 0.88 60 correlated with actual tumor response (PCC \geq 0.88, N = -20

18, p<0.01) and able to distinguish between responding and non-responding patients as defined by the response evaluation criteria for solid tumors (RECIST) for percent change in longest axis from diagnosis to therapy completion (*p < 0.001).



20 30 40

Measured longest axis (mm)

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These three different modeling applications demonstrate the versatility of quantitative imaging measurements in the field of mathematical oncology.

Importantly, by using clinically relevant imaging data, these approaches could be extended to improve treatment planning or design optimal therapy regimens based on an individual's own data.

To learn more about these projects please check out these recent publications by Chengyue, David, and Angela *et al.*

[1] Wu C, et al. 2020; IEEE Trans Med Imaging. 1 (In press).
[2] Hormuth DA, et al. 2020; Radiat Oncol. 15; 1.
[3] Jarrett A, et al. 2018; Phys Med Biol. 63; 10.

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