

Numerical approximation of coupled PDEs with high dimensionality gap

P. Zunino¹

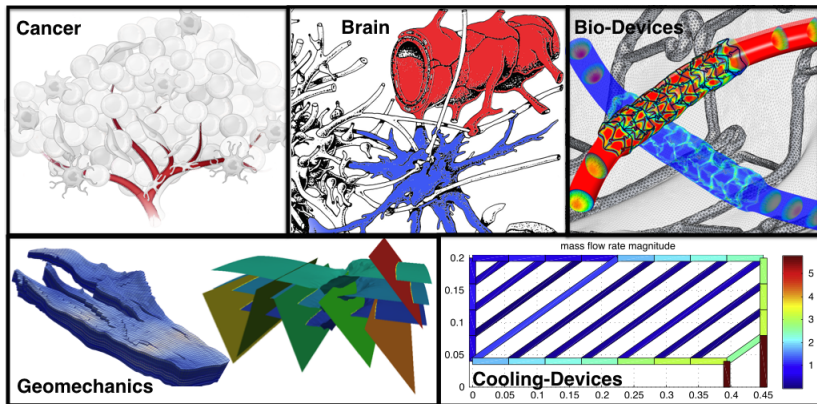
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Advanced numerical methods:
Recent developments, analysis and applications
IHP, Paris, October 3rd – 7th, 2016

Applications

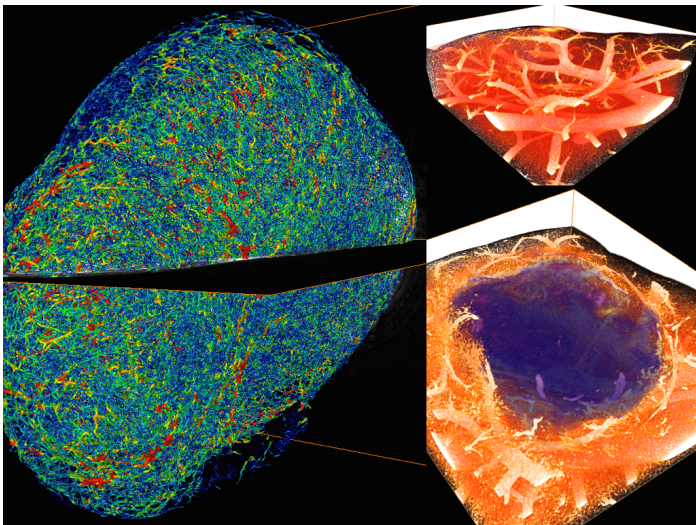
Life sciences: cancer, brain, biomedical devices ...



Technology: fracking, electronics ...

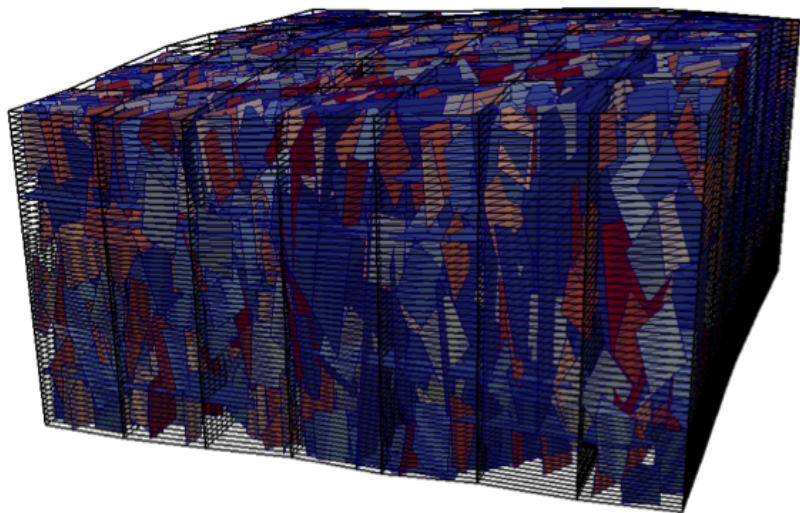
Computational challenges: domains of complex shape

Imaging: 3D rendering of **blood vessels** inside a human colorectal carcinoma model, 0.5cm in diameter, using optical projection tomography (OPT). Courtesy of Simon Walker-Samuel, UCL.



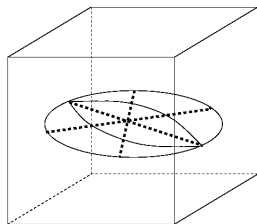
Computational challenges: domains of complex shape

Fracture networks:

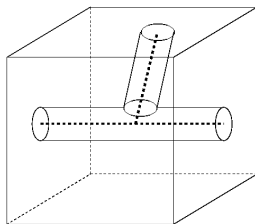


Objective: a unified framework

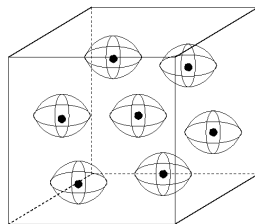
A unified view:



interfaces **3D-2D**



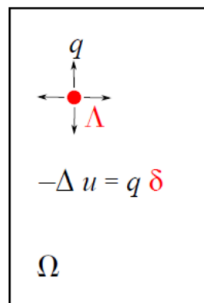
channels **3D-1D**



inclusions (cells, particles) **3D-0D**

Objective: solve coupled PDEs on **embedded manifolds** with **high dimensionality gap**;

Mathematical challenges: a prototype problem



q

Λ

$-\Delta u = q \delta$

Ω

What are the main issues?

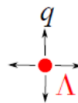
Solution:

$$u \simeq \ln(r)$$
$$\nabla u \simeq \frac{1}{r} \notin L^2(\Omega)$$

- ▶ Simple Poisson problem with Dirac source;

Adapted from D'Angelo C, Quarteroni A. On the coupling of 1D and 3D diffusion-reaction equations. Application to tissue perfusion problems. *Math. Models Methods Appl. Sci.* 2008.


Mathematical challenges: a prototype problem



A diagram showing a red dot representing a Dirac source δ at the center of a domain Ω . Four arrows point outwards from the dot, labeled with a red q above and a red Λ to the right.

$$-\Delta u = q \delta$$

Ω



A diagram showing a red dot representing a Dirac source δ at the center of a domain Ω . A green circle of radius γ is centered on the dot. A red Λ is also shown near the dot. Four arrows point outwards from the dot.

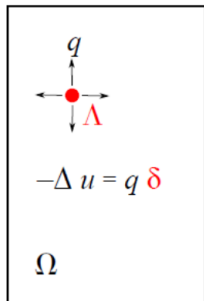
$$\bar{u} = \text{Mean}(u; \gamma)$$
$$q = \beta (\bar{u} - \hat{u})$$
$$-\Delta u = q(u) \delta$$

Ω

- ▶ **Coupled** Poisson problem with Dirac source;
- ▶ Solution-dependent flux q
- ▶ $q(u(x_\Lambda))$ is not admissible \rightarrow use $q(\bar{u})$

Adapted from D'Angelo C, Quarteroni A. On the coupling of 1D and 3D diffusion-reaction equations. Application to tissue perfusion problems. *Math. Models Methods Appl. Sci.* 2008.

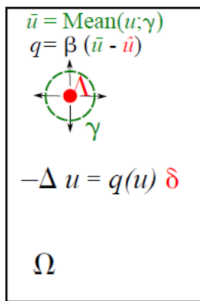
Mathematical challenges: a prototype problem



A square box representing a domain Ω . At the center is a red dot representing a Dirac source. Four arrows point outwards from the dot, labeled with q at the top and Λ at the bottom. Below the diagram is the equation $-\Delta u = q \delta$ and the symbol Ω at the bottom left.

$$-\Delta u = q \delta$$

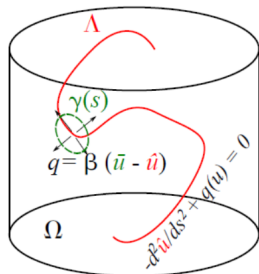
Ω



A square box representing a domain Ω . At the center is a red dot. A green circle of radius γ is centered on the dot. A red arc of length Λ is drawn on the circle. Below the diagram is the equation $-\Delta u = q(u) \delta$ and the symbol Ω at the bottom left. Above the diagram are the formulas $\bar{u} = \text{Mean}(u; \gamma)$ and $q = \beta(\bar{u} - \hat{u})$.

$$\bar{u} = \text{Mean}(u; \gamma)$$
$$q = \beta(\bar{u} - \hat{u})$$
$$-\Delta u = q(u) \delta$$

Ω

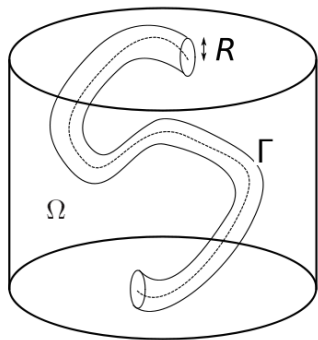


- ▶ **Coupled** Poisson problem with Dirac source;
- ▶ **3D-1D model coupling**;

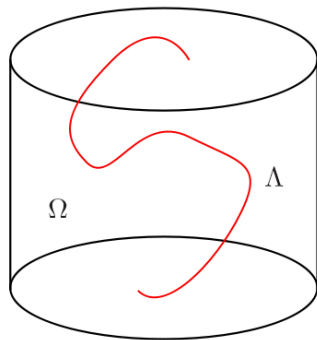
Adapted from D'Angelo C, Quarteroni A. On the coupling of 1D and 3D diffusion-reaction equations. Application to tissue perfusion problems. *Math. Models Methods Appl. Sci.* 2008.

An example: 3D-1D description of microcirculation

from 3-dimensional models in \mathbb{R}^3

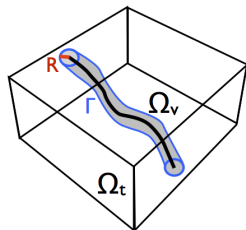


to 1-dimensional models in \mathbb{R}^3



An example: governing equations of microcirculation

Coupled flow

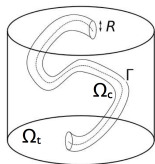


$$\left\{ \begin{array}{ll}
 \mathbf{u}_t = -\frac{k}{\mu} \nabla p_t & \text{in } \Omega_t \\
 \nabla \cdot \mathbf{u}_t = -\overbrace{L_p^{LF} \frac{S}{V} (p_t - p_L)}^{\text{lymphatic drainage}} & \text{in } \Omega_t \\
 \rho \left(\frac{\partial \mathbf{u}_v}{\partial t} + (\mathbf{u}_v \cdot \nabla) \mathbf{u}_v \right) = \nabla \cdot \mathbf{T}(p_v, \mathbf{u}_v) & \text{in } \Omega_v \\
 \nabla \cdot \mathbf{u}_v = 0 & \text{in } \Omega_v \\
 \mathbf{u}_v \cdot \mathbf{n} = \mathbf{u}_t \cdot \mathbf{n} = \overbrace{L_p (p_v - p_t)}^{\text{leakage}} & \text{on } \Gamma
 \end{array} \right.$$

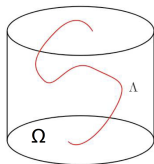
Baxter, Laurence T. and Jain, Rakesh K., *Transport of fluid and macromolecules in tumor. II. Role of heterogenous perfusion and lymphatics*. Microvascular research, 1990.

An example: embedded multiscale method

Embedded multiscale
= **Geometric Multiscale**
+ **Immersed Boundary**



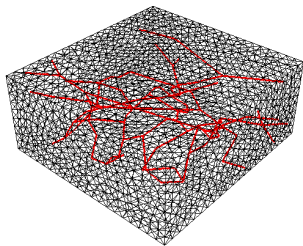
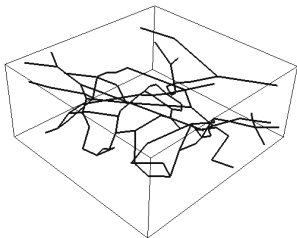
$R \rightarrow 0$



$$\left\{ \begin{aligned} & -\nabla \cdot \left(\frac{k}{\mu} \nabla p_t \right) \\ & = L_p(\bar{p}_v - \bar{p}_t) \delta_\Lambda, \\ & -\frac{\pi R^4}{8\mu} \frac{\partial^2 \bar{p}_v}{\partial s^2} \\ & = -L_p(\bar{p}_v - \bar{p}_t) \end{aligned} \right.$$

Interface conditions are replaced by **concentrated sources**:

- ▶ low computational cost;
- ▶ independent computational meshes of volume and network;

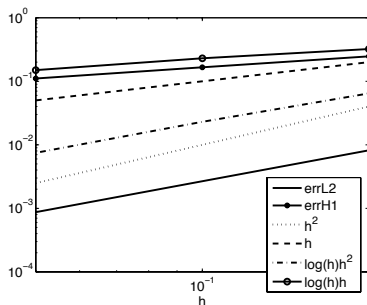
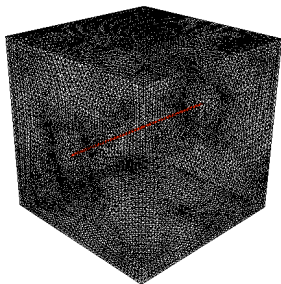


Analysis: a-priori error estimates

1. **T. Koepl and B. Wohlmuth.** Optimal a priori error estimates for an elliptic problem with Dirac right-hand side. SINUM, 52(4), 1753–1769; **A.H. Schatz and L.B. Wahlbin,** Interior maximum norm estimates for finite element methods, Math. Comp., 31 (1977), pp. 414–442.
2. **D'Angelo,** C. Finite element approximation of elliptic problems with dirac measure terms in weighted spaces: Applications to one- and three-dimensional coupled problems (2012) SINUM, 50 (1), pp. 194–215; **T. Apel, O. Benedix, D. Sirch, and B. Vexler,** A Priori Mesh Grading for an Elliptic Problem with Dirac Right-Hand Side, (2011) SINUM 49(3), 992–1005.

Let $p_t \in W_0^{1,p}$ be the weak solution of the perfusion model and let $p_t^h \in V_t^h$ be the finite element approximation. The following L^2 -error bound holds true:

$$\|p_t - p_t^h\|_{L^2(\Omega \setminus \Omega_v)} \lesssim h^2 |\log h|.$$



A new method: set up

Model reduction strategy:

- (i) identify the domain Ω_p with the entire Ω ;
- (ii) decompose $v|_{\Gamma_i} = \bar{v}_{\Gamma_i} + \tilde{v}_{\Gamma_i}$ and assume that the fluctuations \tilde{v}_{Γ_i} are small;

A new reduced model:

$$(\nabla u, \nabla v)_{\Omega} + \sum_{i=1}^N \kappa_i \bar{u}^{(i)} \bar{v}^{(i)} = \sum_{i=1}^N |\Gamma_i| \kappa_i U_i \bar{v}^{(i)}, \quad \forall v \in V,$$

- ▶ A **new** version of the *embedded multiscale method*;
- ▶ It preserves all the **good** properties of the original problem;
- ▶ It features a **similar** algorithmic complexity of the previous version;

A new method: analysis

The new reduced model:

$$(\nabla u, \nabla v)_{\Omega} + \sum_{i=1}^N \kappa_i \bar{u}^{(i)} \bar{v}^{(i)} = \sum_{i=1}^N |\Gamma_i| \kappa_i U_i \bar{v}^{(i)}, \quad \forall v \in V,$$

- ▶ For $V = H_0^1(\Omega)$, There exists a unique solution $u \in H_0^1(\Omega)$ of the problem.
- ▶ Let u be the weak solution and let u_h be its \mathbb{P}^1 -FEM approximation. The following estimates hold true for any $\epsilon > 0$:

$$\|u - u_h\|_{H^1(\Omega)} \lesssim h^{\frac{1}{2}-\epsilon} \|u\|_{H^{\frac{3}{2}-\epsilon}(\Omega)} \quad \text{and} \quad \|u - u_h\|_{L^2(\Omega)} \lesssim h^{\frac{3}{2}-\epsilon} \|u\|_{H^{\frac{3}{2}-\epsilon}(\Omega)}$$

- ▶ The discrete reduced problem is *spectrally equivalent* to the original one.
- ▶ Let $e_{(i)}$ and $e_{(ii)}$ be the modeling error related to the assumptions (i), (ii), respectively. Then it holds:

$$\begin{aligned} & \|e_{(i)}\|_{H^1(\Omega_p)} + \|e_{(ii)}\|_{H^1(\Omega)} \\ & \leq \left(1 + C_{PF}^2\right) \sum_{i=1}^N \left(C_{E,2}(\Omega_i) C_T(\Omega_p, \Gamma_i) C_{E,1}(\Omega_p, \Gamma_i)\right) \sum_{i=1}^N \|U_i\|_{H^{\frac{1}{2}}(\Gamma_i)} \\ & + \frac{3}{4} \left(1 + C_{PF}^2\right)^2 \left(1 + C_T(\Omega_p, \Gamma_i) \sum_{i=1}^N \kappa_i\right) \sum_{i=1}^N \kappa_i |\Gamma_i| U_i^2, \end{aligned}$$

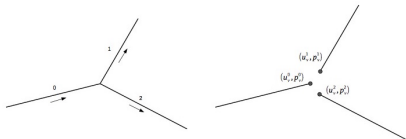
Developments: Coupled mixed formulation

From primal to mixed problem formulation: (with D. Notaro, Ing. Mat)

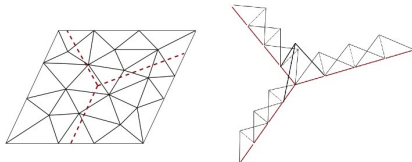
$$\left\{ \begin{array}{ll} -\frac{k}{\mu} \Delta p_t - f(\bar{p}_t, \bar{p}_v) \delta_\Lambda = 0 & \text{in } \Omega \\ -\frac{R^2}{8\mu} \frac{\partial^2 \bar{p}_v}{\partial s^2} + \frac{1}{\pi R^2} f(\bar{p}_t, \bar{p}_v) = 0 & \text{in } \Lambda \end{array} \right. \quad \left\{ \begin{array}{ll} \frac{\mu}{k} \mathbf{u}_t + \nabla p_t = 0 & \text{in } \Omega \\ \nabla \cdot \mathbf{u}_t - f(\bar{p}_t, \bar{p}_v) \delta_\Lambda = 0 & \text{in } \Omega \\ \frac{8\mu}{R^2} \bar{u}_v + \frac{\partial \bar{p}_v}{\partial s} = 0 & \text{in } \Lambda \\ \frac{\partial \bar{u}_v}{\partial s} + \frac{1}{\pi R^2} f(\bar{p}_t, \bar{p}_v) = 0 & \text{in } \Lambda \end{array} \right.$$

Challenges:

- Enforcement of **mass conservation** at network nodes;



- Choice of **compatible** FEM spaces;



Developments: Coupled mixed formulation

Primal mixed: Find $\mathbf{u}_t \in \mathbf{V}_t \subset \mathbf{L}^2(\Omega)$, $p_t \in Q_t \subset H^1(\Omega)$, $u_v \in V_v$, $p_v \in Q_v$

$$\left\{ \begin{array}{l} (\frac{1}{\kappa_t} \mathbf{u}_t, \mathbf{v}_t)_\Omega + (\nabla p_t, \mathbf{v}_t)_\Omega = 0 \\ -(\mathbf{u}_t, \nabla q_t)_\Omega + (Q(\bar{p}_t - p_v), \bar{q}_t)_\Lambda + (\mathbf{u}_t \cdot \mathbf{n}_t, q_t)_{\partial\Omega} = 0 \\ (\frac{\pi^2 R^4}{\kappa_v} u_v, v_v)_\Lambda + (\pi R'^2 \partial_s p_v, v_v)_\Lambda = 0 \\ -(\pi R'^2 u_v, \partial_s q_v)_\Lambda + (Q(p_v - \bar{p}_t), q_v)_\Lambda + [u_v q_v]_{\Lambda}^{\Lambda^{out}} = 0 \end{array} \right.$$

Theorem: there exists a unique solution of the primal mixed problem.

Error analysis:

$$\mathbf{u} := [\mathbf{u}_t, u_v] \quad p := [p_t, p_v]$$

$$\|\mathbf{v}, q\|^2 := \|\mathbf{v}_t\|_{L^2(\Omega)}^2 + \|v_v\|_{L^2(\Lambda)}^2 + \|q_t\|_{H^1(\Omega)}^2 + \|q_v\|_{H^1(\Lambda)}^2$$

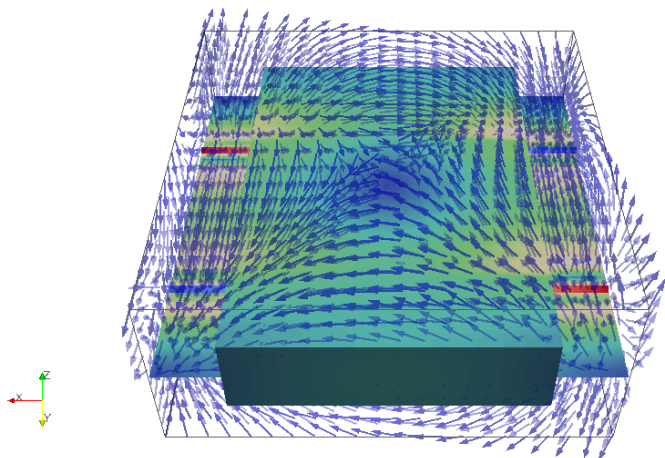
$$\|\mathbf{u} - \mathbf{u}_h, p - p_h\| \leq Ch^l \left[|q_t|_{\mathbf{H}^{l+1}(\Omega)}^2 + |q_v|_{H^{l+1}(\Lambda)}^2 + |\mathbf{v}_t|_{\mathbf{H}^l(\Omega)}^2 + |v_v|_{H^l(\Lambda)}^2 \right]^{1/2}$$

D. Notaro, P. Zunino, Mixed Finite Element Methods for Coupled 3D/1D Fluid Problems, in preparation, 2016

Developments: Coupled mixed formulation

Mixed velocity-pressure approximation:

- ▶ **Network:** $\mathbb{P}^2 - \mathbb{P}^1$ approximation of velocity and pressure;
- ▶ **Porous medium:** $\mathbb{RT}^0 - \mathbb{P}^0$ approximation of velocity and pressure;



Applications: microcirculation and cancer

Cancer alters blood perfusion

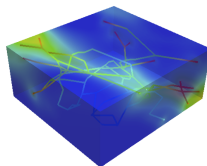
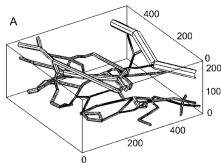
Motivations:

- ▶ understand the role of altered tumor vessel phenotype on tissue perfusion;
- ▶ understand alterations of mass transport;

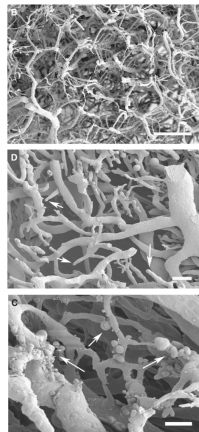
Objectives:

- ▶ build up a computational model able to capture the relevant physical phenomena on a **realistic microvascular geometry**;
- ▶ a **brute force approach** is computationally unfeasible;

courtesy of TW.Scomb



P_{O_2} profiles 0-40 mmHg



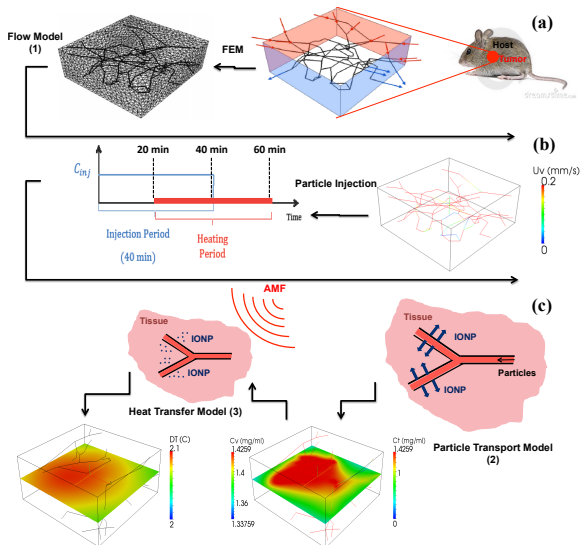
Vascular network of teratocarcinoma.

Simulations: hyperthermic treatment

Prototype study: deliver **iron oxide nanocrystals** thermally excited by electromagnetic fields, with P. Decuzzi and M. Nabil.

A comprehensive simulator:

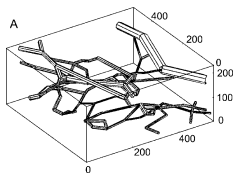
- ▶ Angiogenesis;
- ▶ Capillary and interstitial flow;
- ▶ Solute and particle transport;
- ▶ Heat transfer;



Conclusions: general applications

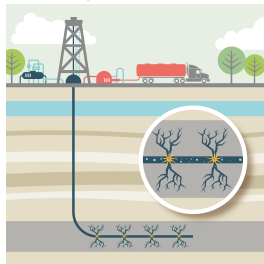
Fluid mechanics: microvascular flows

- (3D) plasma pressure in tissue interstitium;
- (1D) blood flow in microvessels;



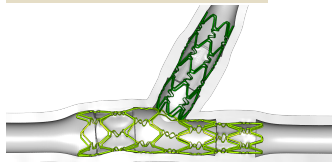
Fluid/solid mechanics: fracking

- (3D) geomechanics;
- (1D/2D) flow in wells and fractures;



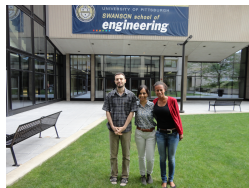
Solid mechanics: biomedical devices

- (3D) mechanical stresses of tissue
- (1D) mechanical stresses in the device



Acknowledgements

- ▶ T. Koepl, E. Vidotto, B. Wohlmuth
TUM & U.o.Stuttgart, Germany
- ▶ L. Cattaneo Ph.D., Politecnico di Milano.
Ing. Mat. PoliMi;
- ▶ M. Nabil, August 2013 - present, graduate student
Mechanical Engineering, PennState;
- ▶ D. Notaro, A. Tiozzo, S. Brambilla, master students.
Ing. Mat. PoliMi;



Some recent works

Tumor perfusion, drug delivery & hyperthermia:

- ▶ Nabil M, Decuzzi P, Zunino P.; Modelling mass and heat transfer in nano-based cancer hyperthermia. R.Soc.Open Sci. (2015) 2: 150447.
- ▶ Cattaneo, L., Zunino, P. Numerical investigation of convergence rates for the FEM approximation of 3D-1D coupled problems (2015) Lecture Notes in Computational Science and Engineering, 103, pp. 727-734.
- ▶ Cattaneo, L., Zunino, P. A computational model of drug delivery through microcirculation to compare different tumor treatments (2014) International Journal for Numerical Methods in Biomedical Engineering, 30 (11), pp. 1347-1371.
- ▶ Cattaneo, L., Zunino, P. Computational models for fluid exchange between microcirculation and tissue interstitium (2014) Networks and Heterogeneous Media, 9 (1), pp. 135-159.

Biomedical devices (stents):

- ▶ P. Zunino, J. Tambača; E. Cutri; S. Čanić; L. Formaggia; F. Migliavacca, Integrated stent models based on dimension reduction. Review and future perspectives. Annals of Biomedical Engineering, 44 (2), pp 604-617, 2015.
- ▶ C. D'Angelo, P. Zunino, A. Porpora, S. Morlacchi, F. Migliavacca, Model reduction strategies enable computational analysis of controlled drug release from cardiovascular stents, SIAM J. Appl. Math., Special Issue on Mathematical Modeling of Controlled Drug Delivery, Vol. 71 (2011) No. 6, pp. 2312-2333.