

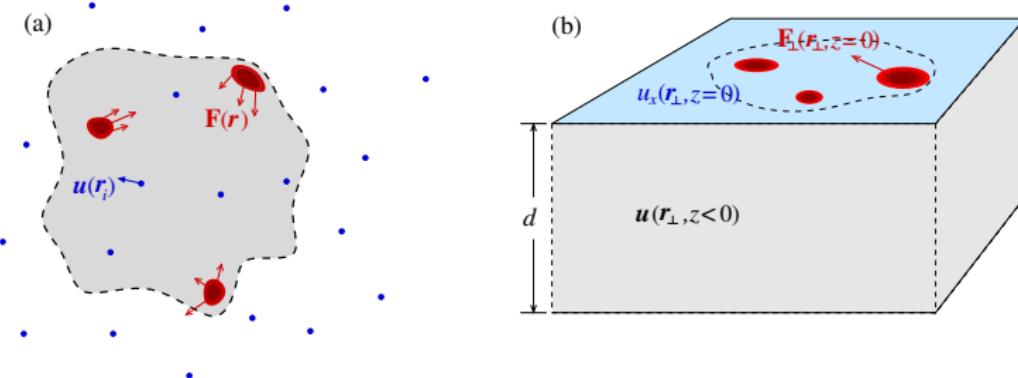
# Reconstruction of localized force distributions in cells and tissues from substrate displacements using physically-consistent regularization

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National Institutes of Health

APS March Meeting 2017  
Session S5: Machine Learning for Modeling and Control of  
Biological Systems I

# Traction force microscopy: displacements $\mathbf{u}$ measured



In-plane surface displacements, as  $d \rightarrow \infty$ , s where  $G \propto (Er)^{-1}$

$$u_x(x, y) = \int_{\Omega} dx' dy' G_{xx}(x - x', y - y') \sigma_{xz}(x', y') + \int_{\Omega} dx' dy' G_{xy}(x - x', y - y') \sigma_{yz}(x', y')$$

$$u_y(x, y) = \int_{\Omega} dx' dy' G_{yx}(x - x', y - y') \sigma_{xz}(x', y') + \int_{\Omega} dx' dy' G_{yy}(x - x', y - y') \sigma_{yz}(x', y')$$

- $\sigma$  is tangential surface stress tensor, proportional to the  $\mathbf{F}_\perp$
- $\mathbf{u}$  found exactly in closed-form for any fixed piecewise-polynomial approximation of  $\sigma$

# Discretized forward problem

## Linear algebra system

- $\sigma$  interpolated about points  $\mathbf{s}_i$  (using polynomials)
- Solution exact up to order of interpolation of  $\mathbf{s}_i$

$$\begin{pmatrix} u_x(\mathbf{r}_1) \\ u_x(\mathbf{r}_2) \\ \vdots \\ u_x(\mathbf{r}_N) \\ u_y(\mathbf{r}_1) \\ u_y(\mathbf{r}_2) \\ \vdots \\ u_y(\mathbf{r}_N) \end{pmatrix}_{2N \times 1} = \begin{pmatrix} \mathbb{G}_{xx} & \mathbb{G}_{xy} \\ \mathbb{G}_{yx} & \mathbb{G}_{yy} \end{pmatrix} \begin{pmatrix} \sigma_{xz}(\mathbf{s}_1) \\ \sigma_{xz}(\mathbf{s}_2) \\ \vdots \\ \sigma_{xz}(\mathbf{s}_M) \\ \sigma_{yz}(\mathbf{s}_1) \\ \sigma_{yz}(\mathbf{s}_2) \\ \vdots \\ \sigma_{yz}(\mathbf{s}_M) \end{pmatrix}_{2M \times 1}$$

- $\{\mathbf{r}_i\}_{i=1}^N$  locations at which measurements are taken
- $\{\mathbf{s}_i\}_{i=1}^M$  locations controlling interpolation of  $\sigma$

# Respecting the physics in regularizing the problem

- **Principle:** Coordinate frame should not bias results
- Regularize using functionals of the tensor invariants

## Invariant regularization functionals

### Tensor invariants

$$\text{Tr}(\boldsymbol{\sigma}) = \sigma_{xz} + \sigma_{yz}$$

$$\text{Det}(\boldsymbol{\sigma}) = \sigma_{xz}\sigma_{yz}$$

$$\Phi_{L^1(\text{Tr})} = \int_{\Omega} |\sigma_{xz}(x, y) + \sigma_{yz}(x, y)| dx dy$$

$$\Phi_{TV(\text{Tr})} = \int_{\Omega} |\nabla(\sigma_{xz}(x, y) + \sigma_{yz}(x, y))| dx dy.$$

### Non-invariant functionals

$$\Phi_{TV_1} = \int_{\Omega} (|\nabla \sigma_{xz}(x, y)| + |\nabla \sigma_{yz}(x, y)|) d\mathbf{x}$$

$$\Phi_{TV_2} = \int_{\Omega} (|\partial_x \sigma_{xz}| + |\partial_y \sigma_{xz}| + |\partial_x \sigma_{yz}| + |\partial_y \sigma_{yz}|) d\mathbf{x}.$$

# Using the physics to help constrain the problem

## Reasonable physical assumptions

- ① No force outside the cell footprint
- ② No inertia (forces and torques each sum to zero)

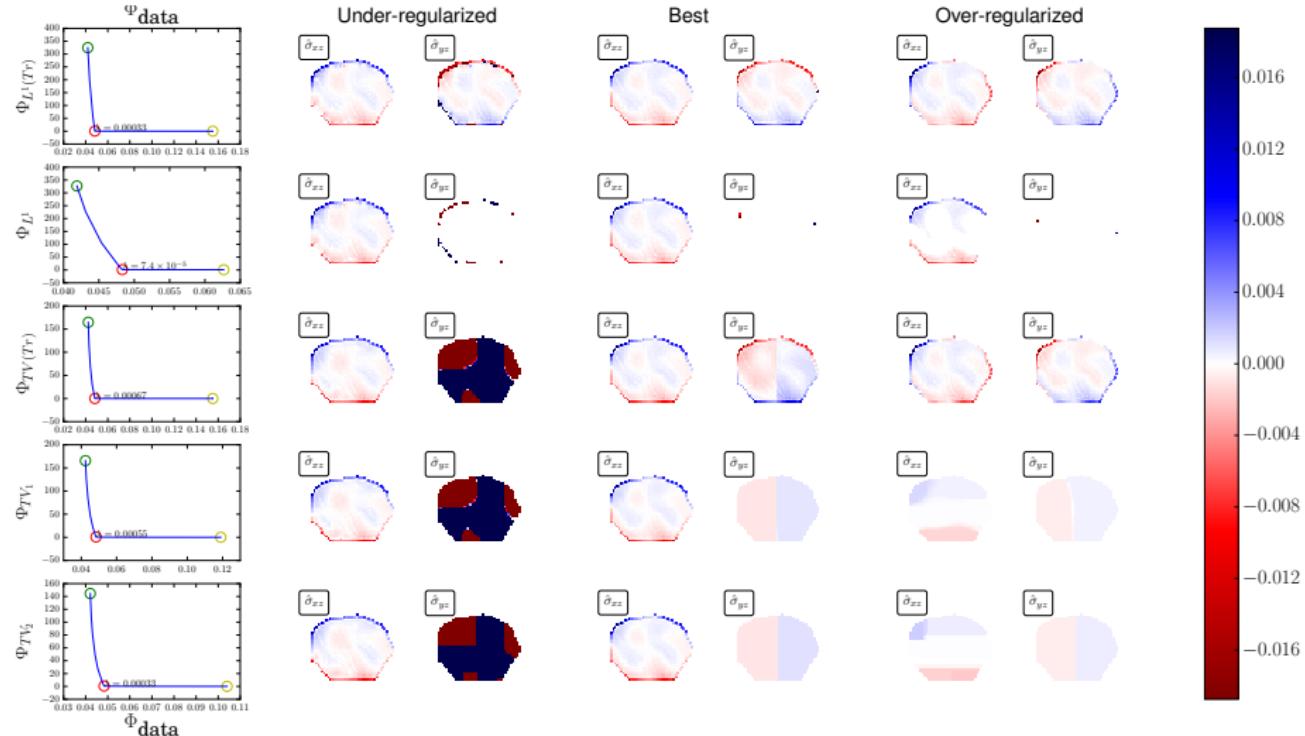
## Advantage gained

- ① May restrict the size of the problem
  - ▶ Throw away far data points
- ② Multiple expansion  $\Rightarrow$  decay at rate  $r^{-2}$ 
  - ▶ Justifies the use of a cut-off for data to use
  - ▶ Sparsify the linear algebra problem

## Regularized inverse problem

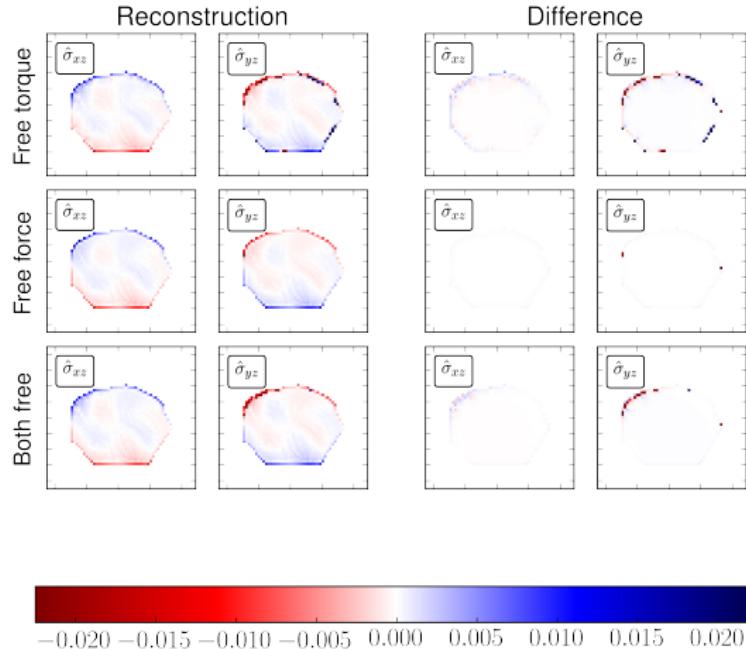
$$\hat{\boldsymbol{\sigma}} = \arg \min_{\boldsymbol{\sigma}} \left[ \|\mathbf{u} - \mathbb{G}\boldsymbol{\sigma}\|_2^2 + \lambda \Phi_{\text{reg}}[\boldsymbol{\sigma}] \right]$$

# Rotational invariance matters



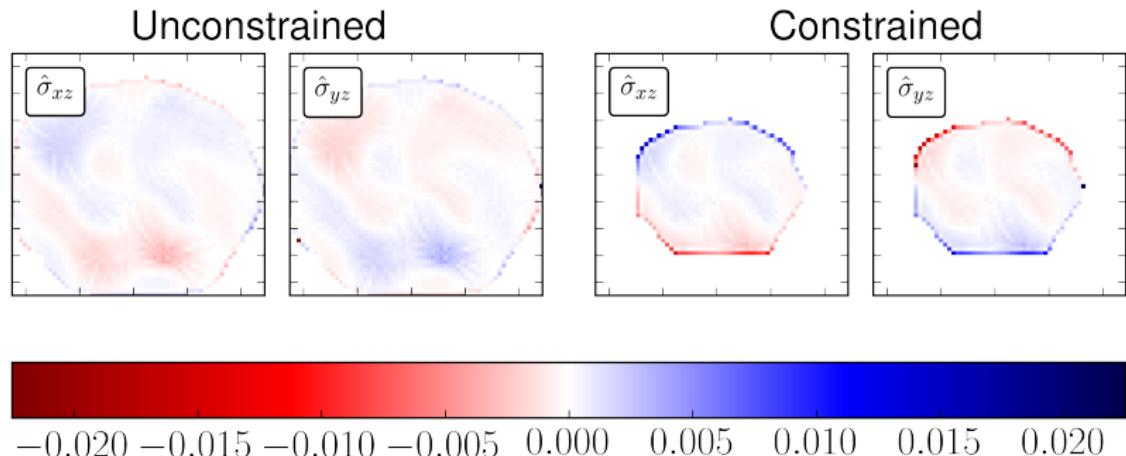
Units:  $E$ (Young's modulus)

# Force constraint matters



- Reconstructions performed using  $L_1(Tr)$
- Difference compared to fully-constrained reconstruction
- Neither force nor torque cancel unless enforced explicitly

# The footprint constraint matters



- Force concentrated near edges of the cell
- Location of edges is important

# Summary and Future directions

- The regularization should be consistent with the known physics (as in a Bayesian prior)
- The numerical solution of the forward problem should be consistent with the regularization
- Localization of the cell footprint is important in determining traction
  - ▶ Future idea: Joint force reconstruction and cell boundary determination
  - ▶ Future work: May do uncertainty quantification through Bayes