

Introduction to modelling cell mechanics and adhesion

Ulrich Schwarz

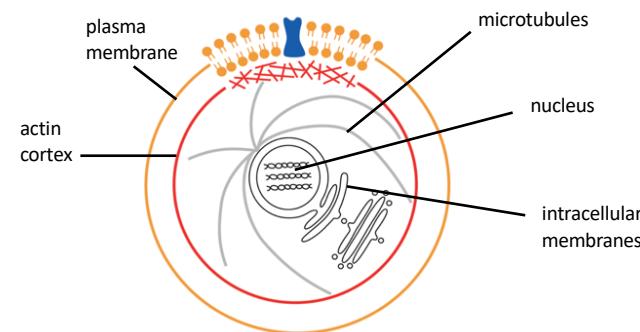
Heidelberg University
Institute for Theoretical Physics and BioQuant



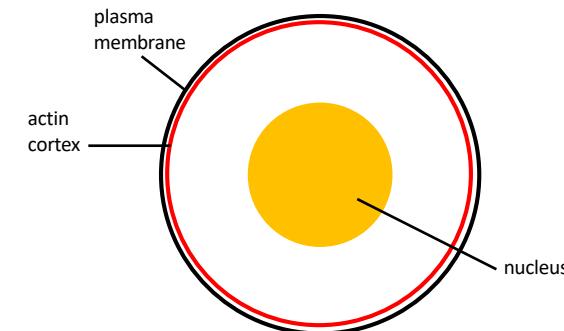
Cell mechanics and adhesion

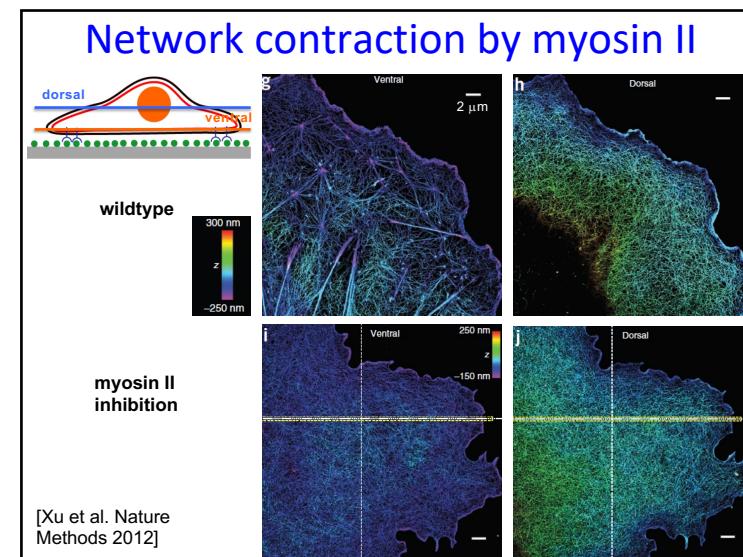
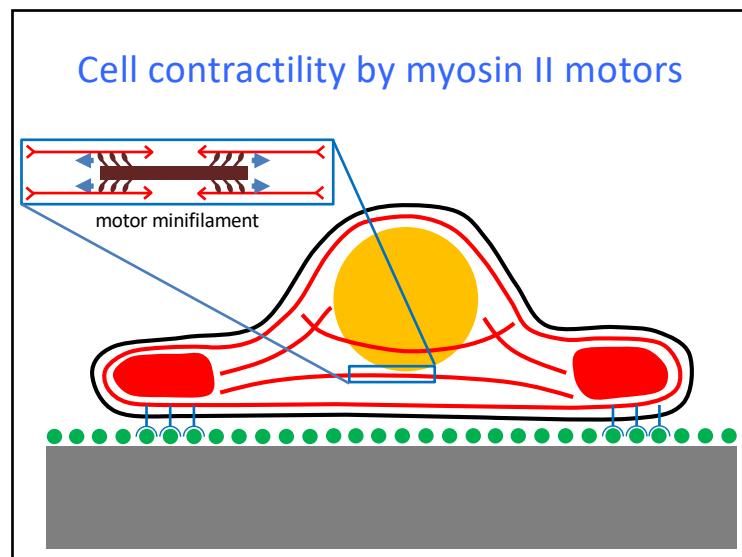
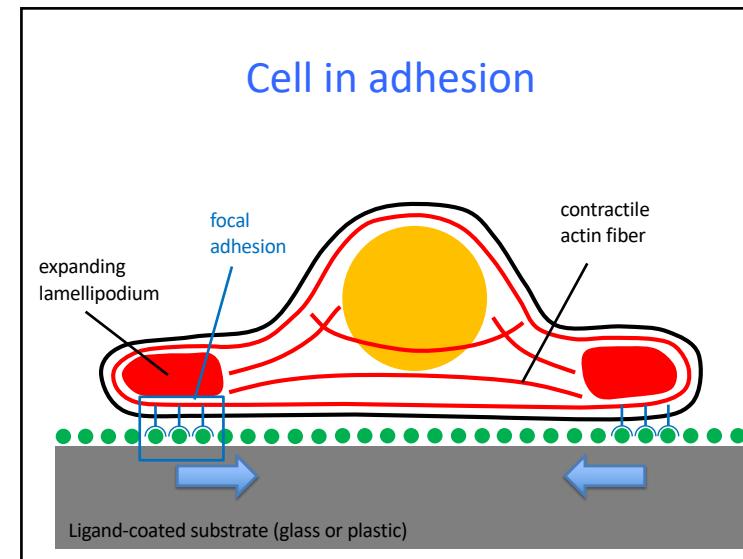
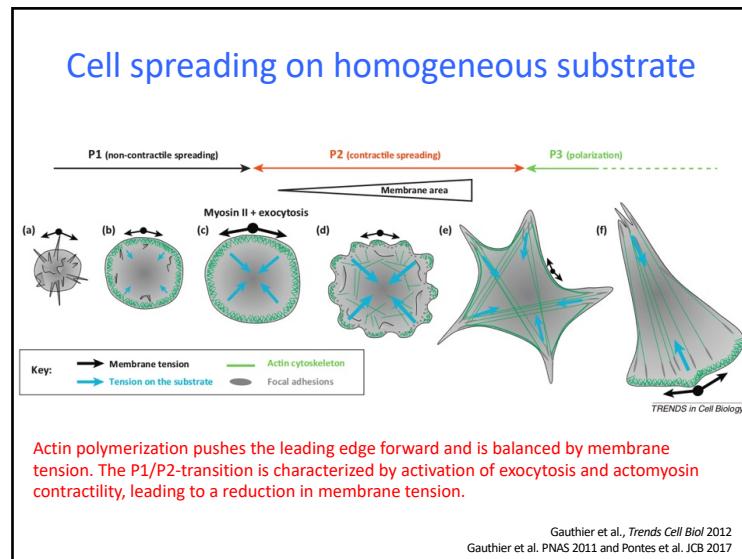
Animal cell in suspension

[Schwarz and Safran RMP 2013]

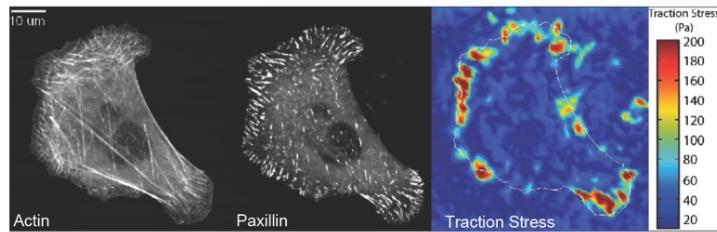


Animal cell in suspension reloaded





Cell organization and traction forces

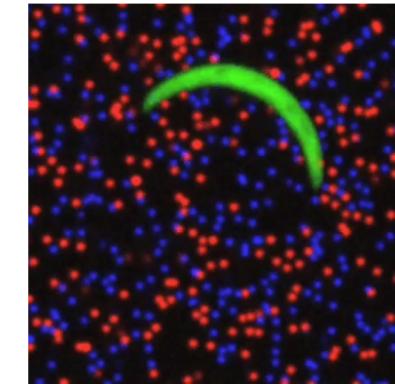


Cell forces can be measured by traction force microscopy and are strongly correlated with the organization of the actin cytoskeleton (stress fibers) and of the integrin-based adhesion sites (focal adhesions)

[Schwarz and Gardel JCS 2012]

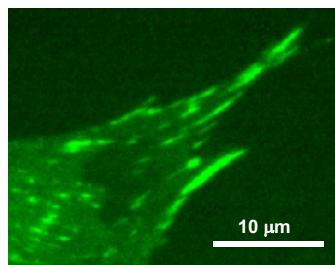
Malaria-parasite moving in circle on soft elastic substrate

GFP-sporozoite on polyacrylamide substrate with two differently colored fluorescent beads

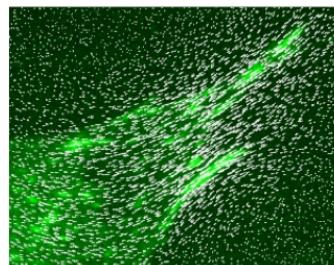


[Münster+ Cell Host Microbe 2009]

High resolution traction force microscopy



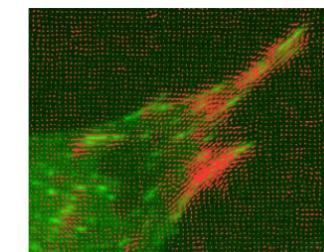
Fibroblast adhesion structures (paxillin) on 15 kPa polyacrylamide-substrate with two differently colored fluorescent nanobeads



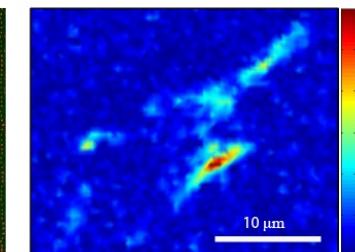
Displacement field extracted with correlation-based particle tracking velocimetry (mesh size 500 nm)

[Sabass+ BPJ 2008]

Regularized Fourier Transform Traction Cytometry (Reg-FTTC)

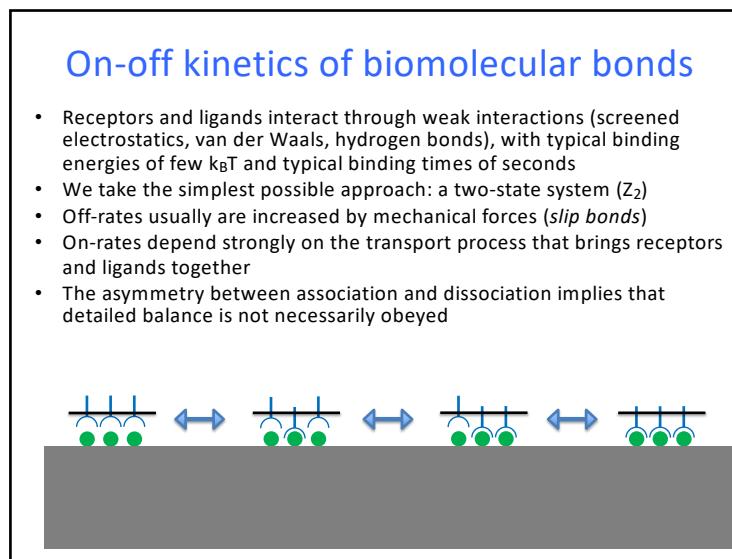
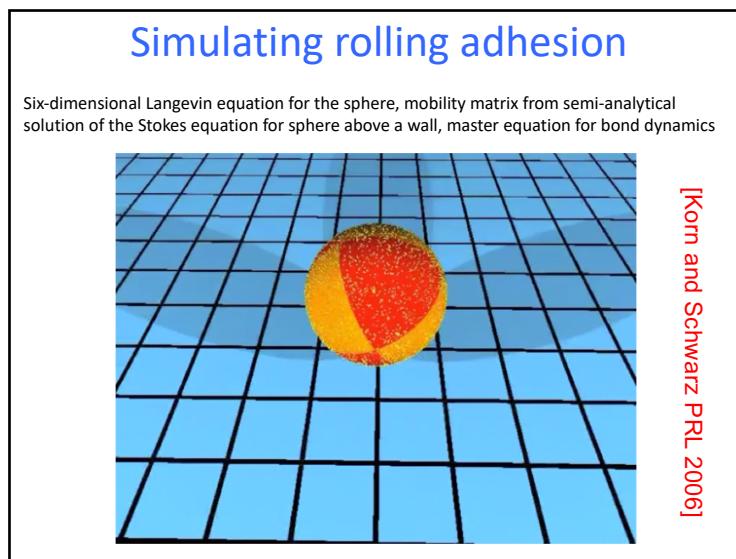
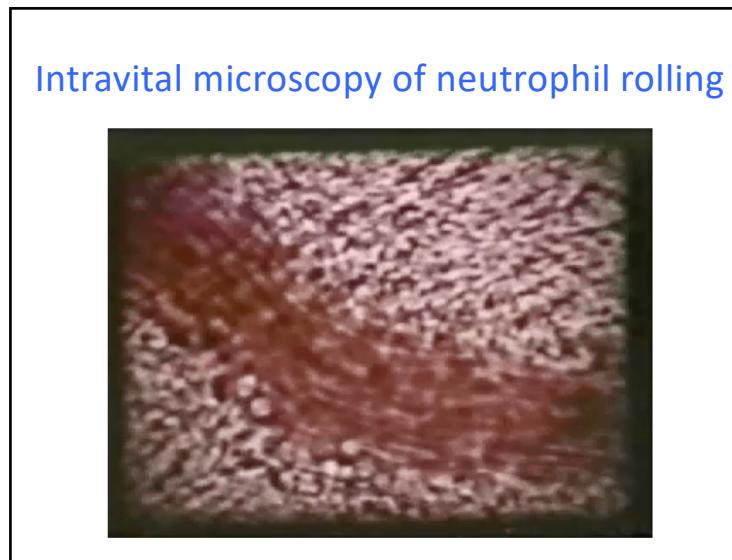
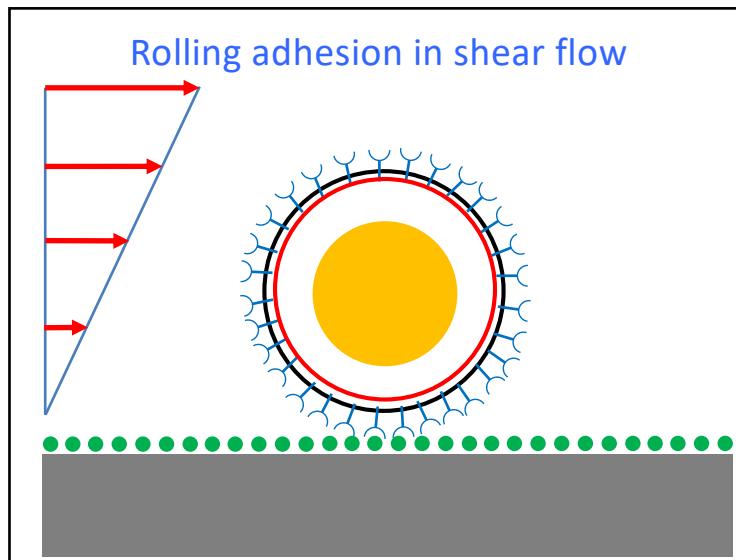


Traction vector field (spatial resolution 1 μm)

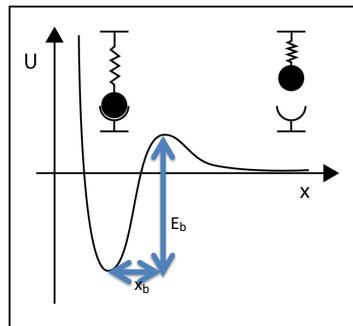


Traction magnitude (in Pa, resolution few 100 Pa)

The typical force per focal adhesion is 5 nN. With a typical FA size of μm^2 , the typical stress is 5 kPa.



Modelling single bond dissociation

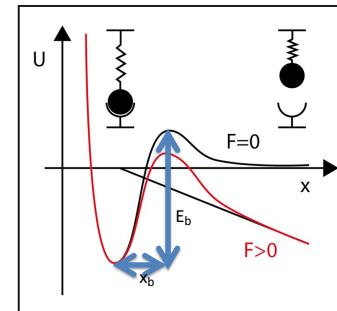


- can be modeled as escape over a transition state barrier (Kramers theory)
- typical time scale is seconds

$$k_0 = \frac{1}{t_D} e^{-\frac{E_b}{kT}} = 1 \frac{1}{s}$$

with attempt time $t_D = ns$ and energy barrier $E_b = 20 kT$

Dissociation under force

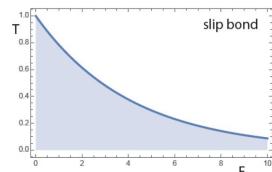


- force reduces barrier height E_b in proportion to distance x_b
- Kramers-Bell-Evans equation:

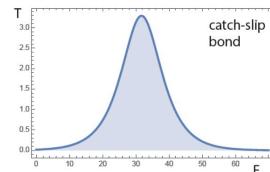
$$k = k_0 e^{F/F_b}$$

with $F_b = kT / x_b = 10 \text{ pN}$ for distance $x_b = 4 \text{ \AA}$

Slip versus catch bonds



Lifetime (in s) decreases exponentially with force (in pN)



Lifetime (here for $\alpha_5\beta_1$ -integrin-fibronectin) first increases and then decreases again with force

demonstrated in single molecule experiments for a wide range of ligand-receptor pairs

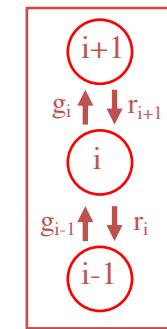
One-step master equation

- $p_i(t)$ probability that at time t exactly i bonds are closed ($0 \leq i \leq N_t$)

$$\frac{dp_i}{dt} = -[r_i + g_i]p_i + r_{i+1}p_{i+1} + g_{i-1}p_{i-1}$$

- rupture rate r_i from Kramers theory with load sharing, rebinding rate g_i force-independent

$$r_i = i e^{f/i}, \quad g_i = \gamma(N_t - i)$$



Dimensionless quantities: $f = F_b/F_0$ $\gamma = k_{on}/k_0$ $\tau = k_0 t$

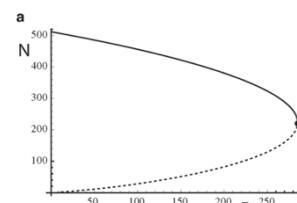
[Erdmann and Schwarz PRL 2004, JCP 2004]

Mean field or first moment equation

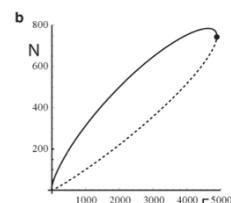
$$\frac{dN}{d\tau} = -r(\langle i \rangle) + g(\langle i \rangle) = -Ne^{f/N} + \gamma(N_t - N)$$

Saddle-node bifurcation at $f_c = N_t \ln(\gamma/e)$.

Rebinding generates stability threshold under force.

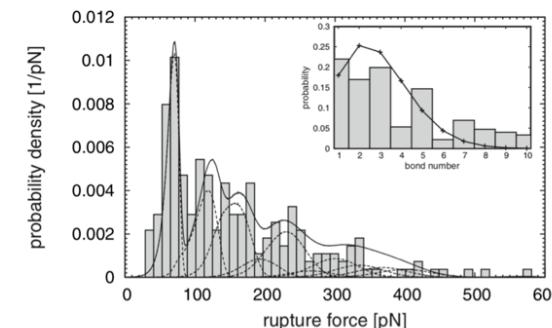


Slip bond, [G. Bell Science 1978]
[Erdmann and Schwarz PRL 2004, JCP 2004]



Catch-slip bond,
[Novikova and Storm BPJ 2013]

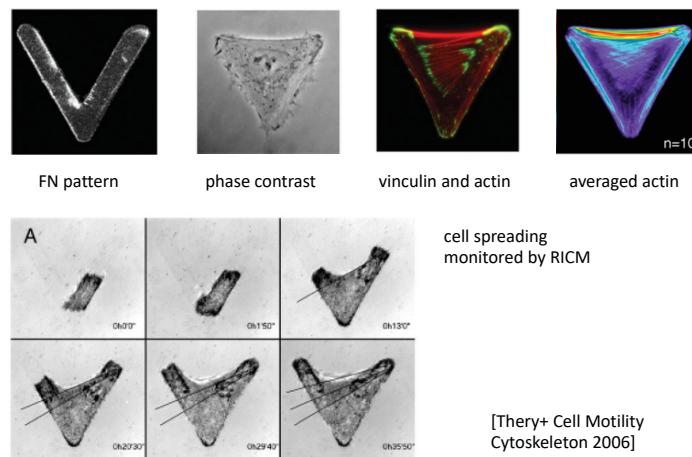
Application to force spectroscopy



biotin-streptavidin bonds, biomembrane force probe, loading rate 1250 pN/s

[Erdmann, Pierrat, Nassoy, Schwarz EPL 2008]

Cell shape on micropatterns



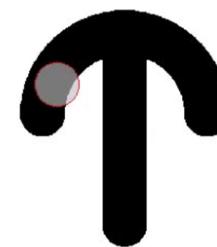
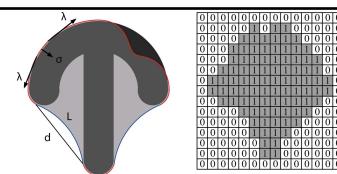
Cellular Potts Model

- Lattice-based spin model
- Dynamic model for cell shape and forces

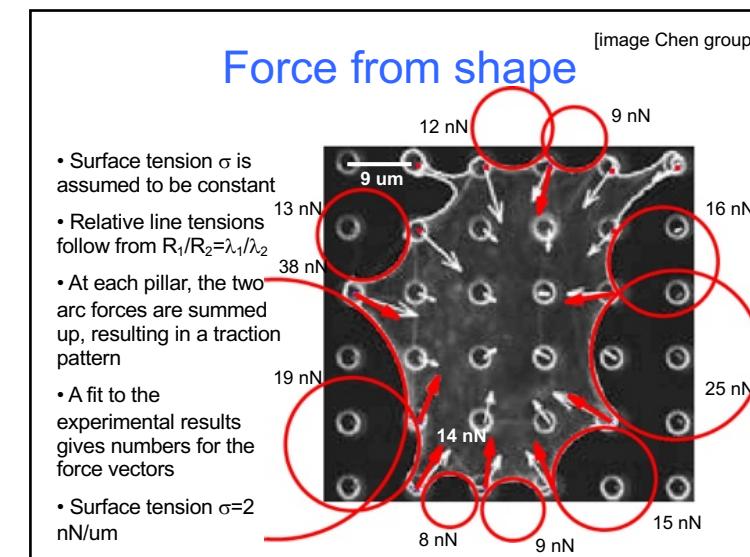
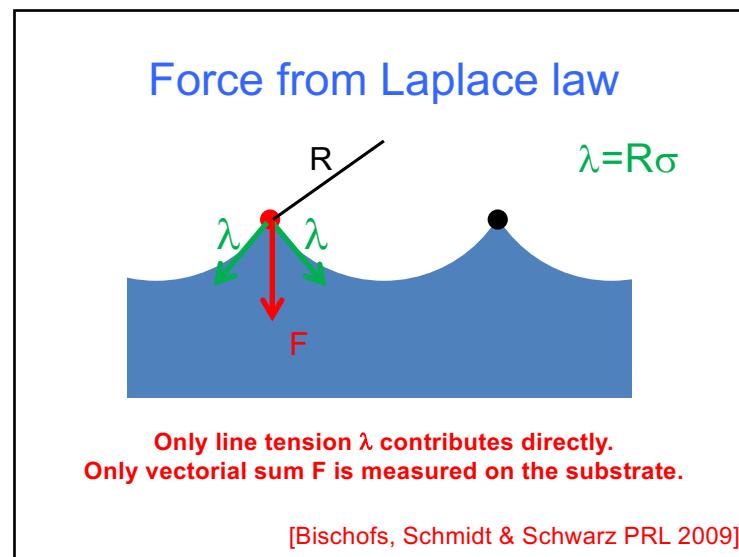
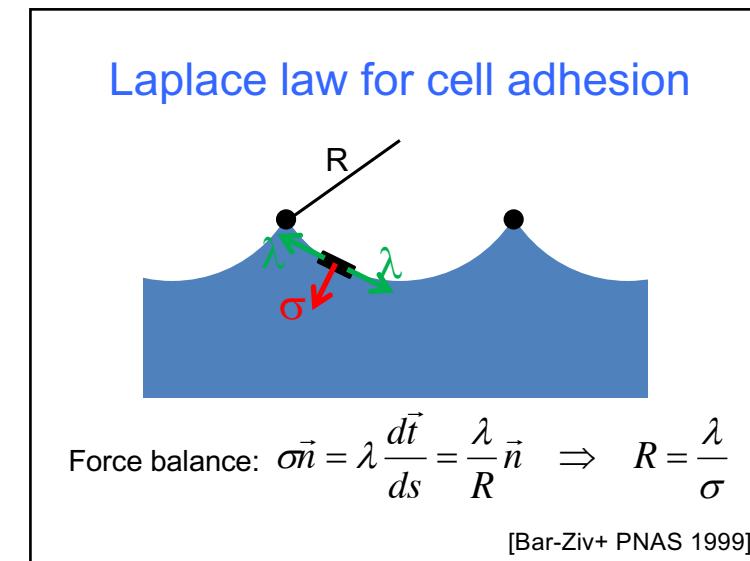
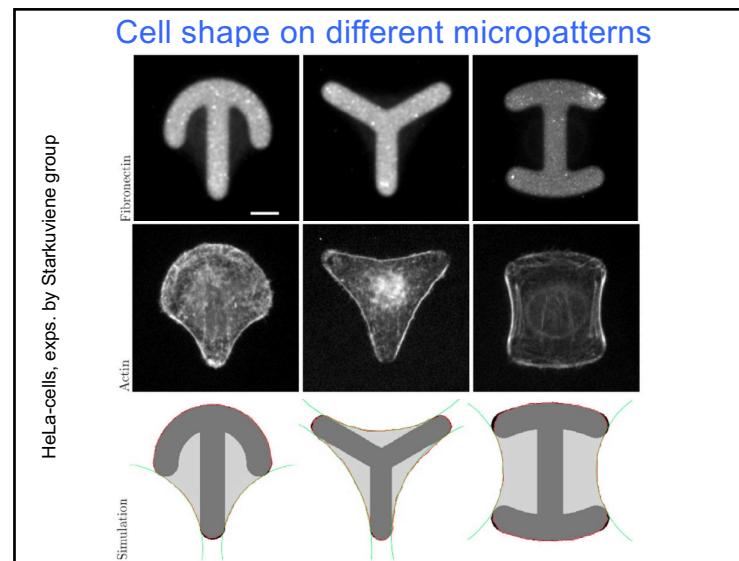
$$H = \sigma A + \lambda_s l + \sum_{\text{arc } i} \frac{EA}{2L_{0,i}} (L_i - L_{0,i})^2 - \frac{E_0}{A_{\text{ref}} + A_{\text{ad}}} A_{\text{ad}}$$

surface tension	simple line tension	elastic line tension
σ	λ_s	λ
d	L	A

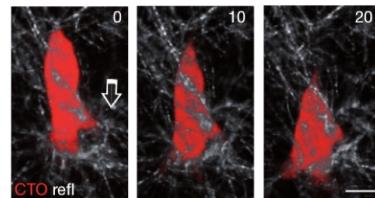
- Geometry with marching square
- Update with Metropolis



[Albert & Schwarz BPJ 2014]

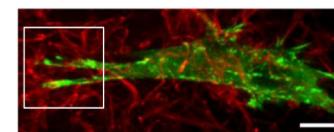


Cells in 3D spread in spatially heterogeneous matrices

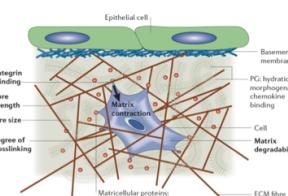


[Wolf et al. Nature Cell Biol 2007]
Cancer cells re-arrange collagen fibers and facilitate migration of followers

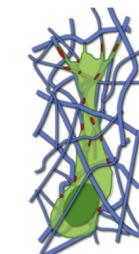
[Doyle and Yamada, Exp Cell Res 2016]
Fibroblasts in 3D collagen gels show robust adhesions



Cell organization in 3D gels



[Griffith and Swartz Nature Reviews MCB 2006]



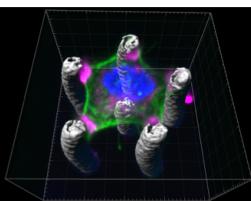
[Doyle and Yamada, Exp Cell Res 2016]

Cell shape in 3D scaffolds

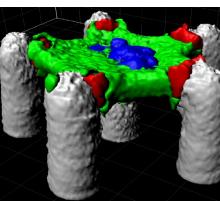
[Brand+ Biophys 2017]



3D scaffold with one pillar moved out



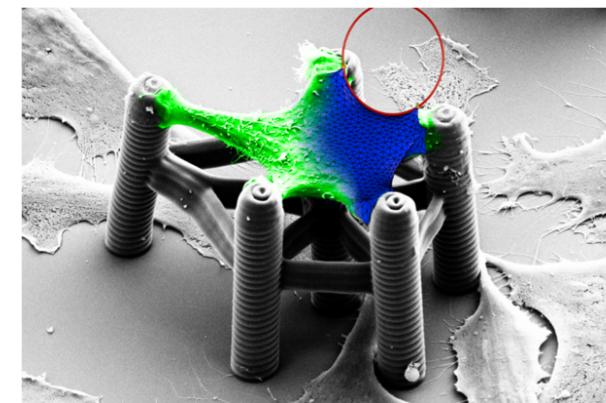
Fluorescence image



Volume rendering with Imaris

Cells in open 3D scaffolds share most features of cells on 2D substrates. The difference to cells in 3D matrix is gradual, not fundamental.

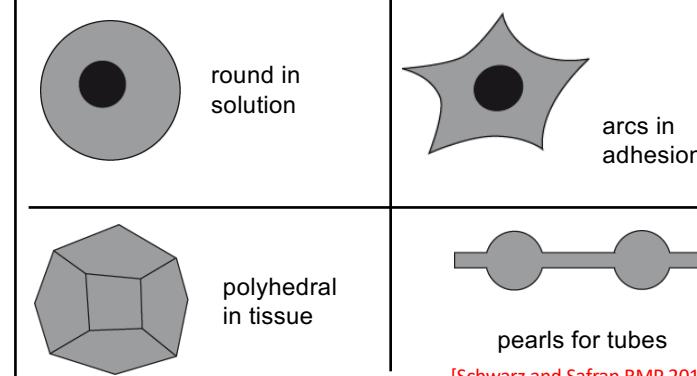
Contractile surface model gives good agreement regarding shape



Summary first part

- Adhesion leads to complete remodelling of a cell: focal adhesions anchor it, lamellipodia drive the cell envelope outwards, actomyosin cortex and stress fibers stabilize it
- The cell senses the physical properties of its environment and strongly adapts to it
- Focal adhesions under force are only stable due to rebinding. The dynamics of biomolecular bonds allow cells to dynamically respond to changing conditions and at the same time to keep mechanical integrity
- To first approximation, cells are objects under strong tension and their shape is determined by the Laplace law

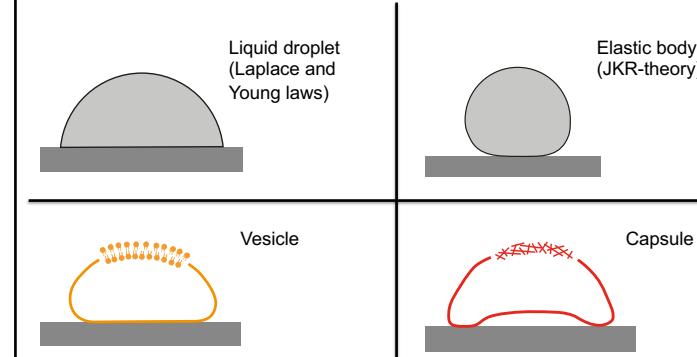
Cell shape often shows signature of surface tension



[Schwarz and Safran RMP 2013]

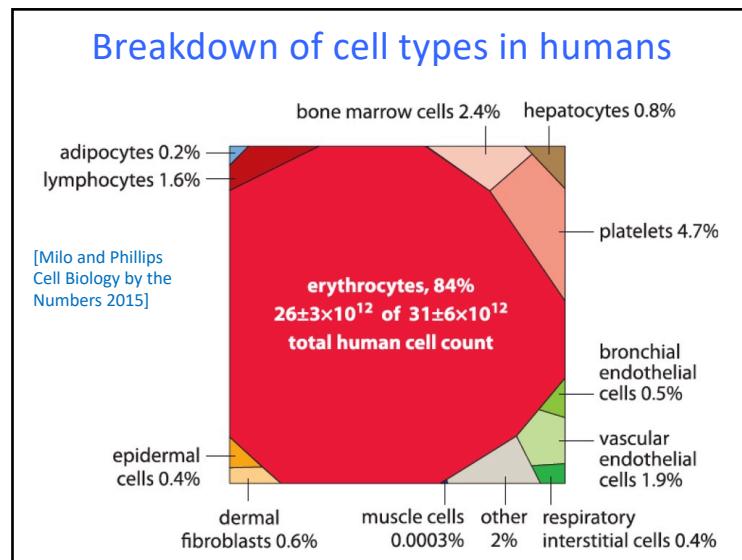
The red blood cell as model system for cell mechanics

Soft matter models for cell adhesion



Cell adhesion is different because it is characterized by large contact area, tangential traction forces, localized adhesions, contractile actomyosin bundles and energy consumption

[Schwarz and Safran RMP 2013]



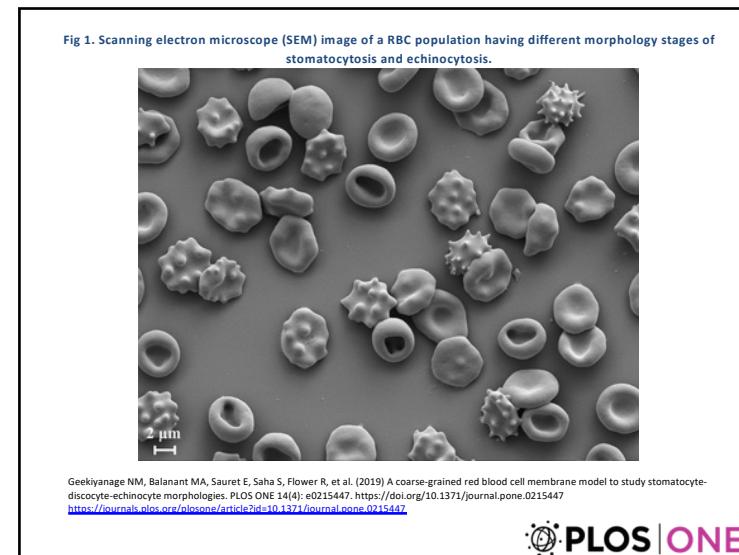
Blood cells

TYPE OF CELL	MAIN FUNCTIONS	TYPICAL CONCENTRATION IN HUMAN BLOOD (CELLS/LITER)
Red blood cells (erythrocytes)	transport O ₂ and CO ₂	5×10^{12}
White blood cells (leucocytes)		
Granulocytes		
Neutrophils (polymorphonuclear leucocytes)	phagocytose and destroy invading bacteria	5×10^9
Eosinophils	destroy larger parasites and modulate allergic inflammatory responses	2×10^8
Basophils	release histamine (and in some species serotonin) in certain immune reactions	4×10^7
Monocytes	become tissue macrophages, which phagocytose and digest invading microorganisms and foreign bodies as well as damaged senescent cells	4×10^8
Lymphocytes		
B cells	make antibodies	2×10^9
T cells	kill virus-infected cells and regulate activities of other leucocytes	1×10^9
Natural killer (NK) cells	kill virus-infected cells and some tumor cells	1×10^8
Plasmacytoid dendritic cells (cell fragments arising from megakaryocytes in bone marrow)	initiate blood clotting	3×10^{11}

Humans contain about 5 liters of blood, accounting for 7% of body weight. Red blood cells constitute about 45% of this volume and white blood cells about 1%, the rest being the liquid blood plasma.

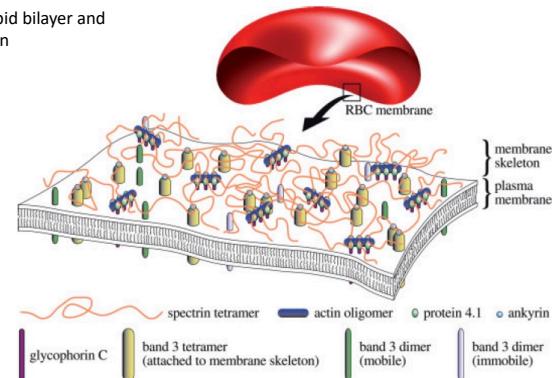
Alberts MBoC

Red blood cell, platelet, white blood cell (from wikipedia)



Structure of the RBC-envelope

Composite of lipid bilayer and spectrin skeleton



Very important: no motors, no contractility, no adhesion

Helfrich bending Hamiltonian for biomembranes

$$H = \int dA \{ \sigma + 2\kappa(H - c_0)^2 \}$$

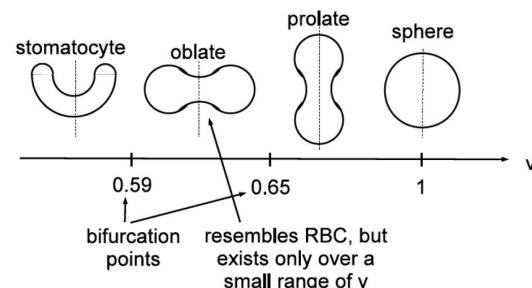
- $H = (1/R_1 + 1/R_2)/2$ mean curvature
- c_0 spontaneous curvature for asymmetric bilayers
- σ surface tension around 10^4 N/m
- κ bending rigidity around $25 k_B T$

Euler-Lagrange equations: $p + 2\sigma H - 2\kappa(2H(H^2 - K) - \Delta H) = 0$

can be solved by ODE-shooting for axisymmetric shapes

Shape diagram for vesicles

Bending Hamiltonian, Euler-Langrange equations,
the only parameter is the reduced volume $v = V/(4\pi/3 R^3)$, $A = 4\pi R^2$



Artificial vesicles with pure bilayers are studied as simplest cell models

Experimentally one observes different shape transitions, e.g. prolate to bud

For given area and volume, the only relevant parameter is reduced volume

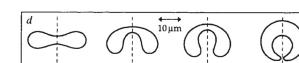
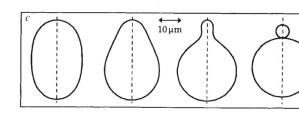
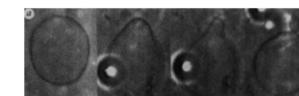
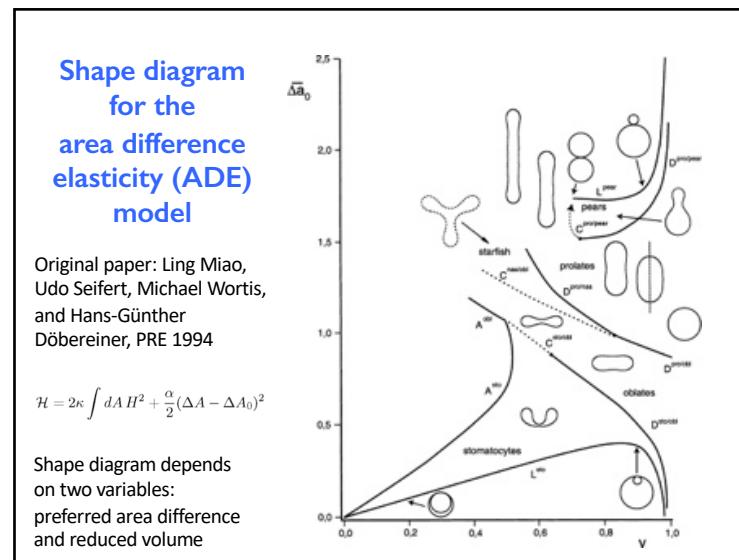


FIG. 1 Shape transformations of free vesicles induced by a change in temperature^a. a, c. Expansion of a small vesicle from a larger one (budding); b, d. Inverse budding ("endocytosis") via the transformation from a discocyte to a stomatocyte. The shapes are axisymmetric with respect to the broken line.

Lipowsky review Nature 1991



Elastic surface Hamiltonian for RBCs

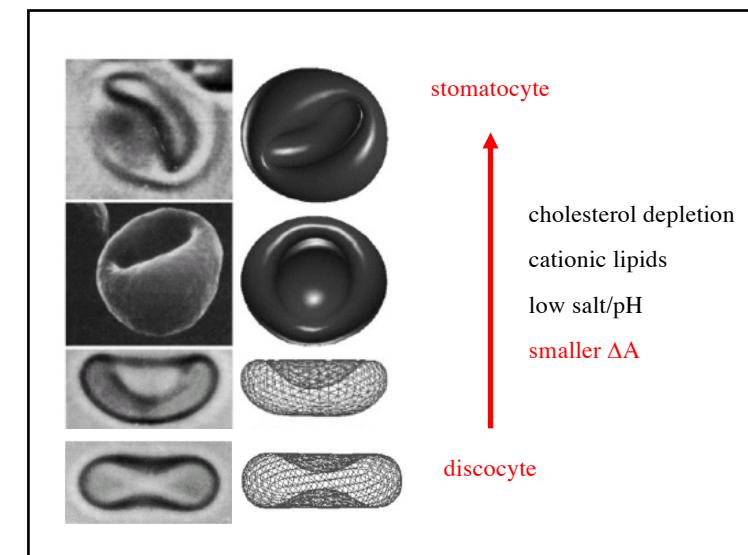
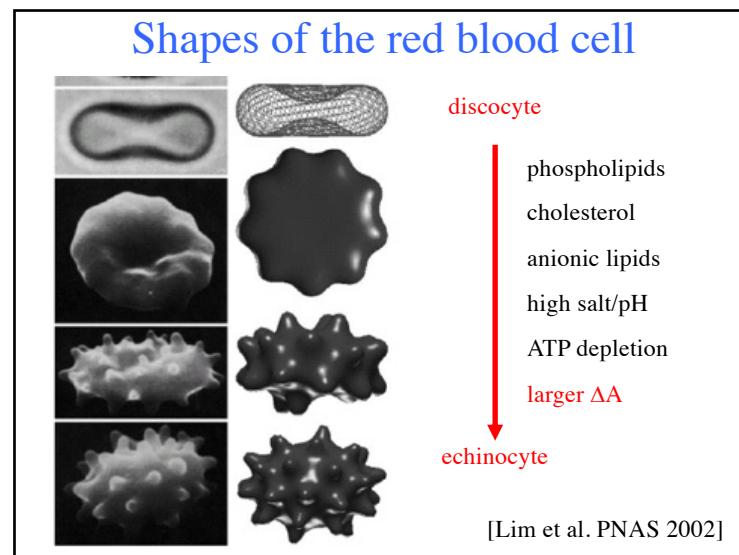
Helfrich bending Hamiltonian for plasma membrane as above

Elastic Hamiltonian for spectrin-actin-network:

$$\mathcal{H} = \frac{K_\alpha}{2} \int dA (\alpha^2 + \alpha_3 \alpha^3 + \alpha_4 \alpha^4) + \mu \int dA (\beta + b_1 \alpha \beta + b_2 \beta^2)$$

$$\text{with } \alpha = \lambda_1 \lambda_2 - 1, \beta = \frac{1}{2} \left(\frac{\lambda_1}{\lambda_2} + \frac{\lambda_2}{\lambda_1} - 2 \right) \text{ strain invariants}$$

Euler-Lagrange equations very challenging, usually solved by direct surface minimization



Full model for RBC shape

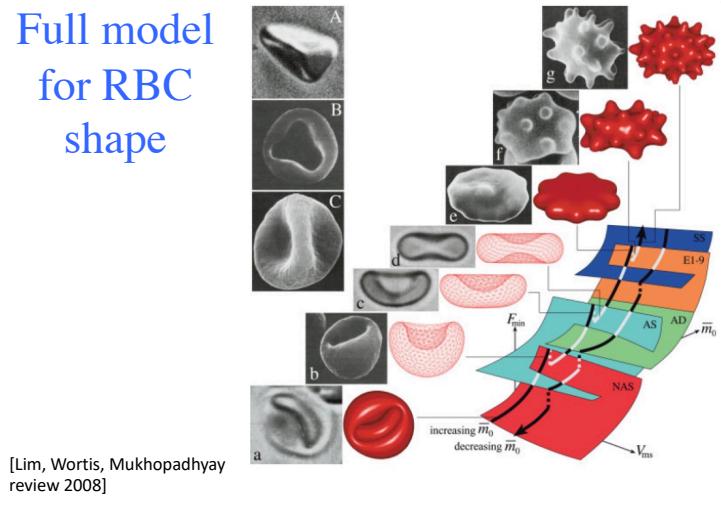


Fig 4. Summary of estimated cell surface area (A_{ex}), cell volume (V_{ex}), bilayer-leaflet-area-difference (ΔA_{ex}), total-membrane-curvature (C_{ex}) and equivalent reduced cell volume (v_{ex}) for the experimentally observed discocyte, echinocyte I, echinocyte II and echinocyte III RBC shapes.

Shape Classification	Experimental Observations (Confocal Microscopy Imaging)	Experimental Observations (Triangulated Surface Mesh)	A_{ex}^n (μm^2)	V_{ex}^n (μm^3)	$\Delta A_{ex}^n/A_{ex}^n$ (%)	C_{ex}^n/A_{ex}^n (%)	v_{ex}^n
Discocyte			145.046	101.623	0.124	0.244	0.619
Echinocyte	I		129.634	82.052	0.138	0.298	0.591
	II		140.054	91.767	0.143	0.411	0.589
	III		131.450	91.547	0.161	0.624	0.646

Geekiyangage NM, Balanant MA, Saurat E, Saha S, Flower R, et al. (2019) A coarse-grained red blood cell membrane model to study stomatocyte-discocyte-echinocyte morphologies. PLOS ONE 14(4): e0215447. <https://doi.org/10.1371/journal.pone.0215447>

PLOS ONE

Most recent work comparing different model predictions with experiments

RESEARCH ARTICLE

A coarse-grained red blood cell membrane model to study stomatocyte-discocyte-echinocyte morphologies

Nadeeshani Maheesika Geekiyangage¹, Marie Anne Balanant^{1,2}, Emilie Saurat^{1*}, Suvas Saha³, Robert Flower², Chwee Teck Lim^{4,5,6}, YuanTong Gu¹

¹ School of Chemistry, Physics and Mechanical Engineering, Queensland University of Technology (QUT), Brisbane, Queensland, Australia, ² Discipline of Chemistry, Australian Red Cross Blood Service, Brisbane, Queensland, Australia, ³ University of Technology Sydney (UTS), Ultimo, New South Wales, Australia, ⁴ Department of Biomedical Engineering, Faculty of Engineering, National University of Singapore, Singapore, ⁵ Biomedical Institute for Global Health Research and Technology, National University of Singapore, Singapore, ⁶ Mechanobiology Institute, National University of Singapore, Singapore

* emilie.saurat@qut.edu.au (ES); yuantong.gu@qut.edu.au (YG)



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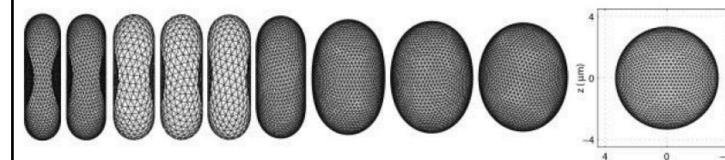
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Abstract

An improved red blood cell (RBC) membrane model is developed based on the bilayer coupling model (BCM) to accurately predict the complete sequence of stomatocyte-discocyte-echinocyte (SDE) transformation of a RBC. The coarse-grained (CG)-RBC membrane model is proposed to predict the minimum energy configuration of the RBC from the competition between lipid-bilayer bending resistance and cytoskeletal shear resistance under given reference constraints. In addition to the conventional membrane surface area, cell volume and bilayer-leaflet-area-difference constraints, a new constraint, total-membrane-curvature is proposed in the model to better predict RBC shapes in agreement with experimental observations. A quantitative evaluation of several cellular measurements including length, thickness and shape factor, is performed for the first time, between CG-RBC model predicted and three-dimensional (3D) confocal microscopy imaging generated RBC shapes at equivalent reference constraints. The validated CG-RBC membrane model is then employed to investigate the effect of reduced cell volume and elastic length scale on SDE transformation, to evaluate the RBC deformability during SDE transformation, and to identify the most probable RBC cytoskeletal reference state. The CG-RBC membrane model can predict the SDE shape behaviour under diverse shape-transforming scenarios, in-vitro RBC storage, microvascular circulation and flow through microfluidic devices.

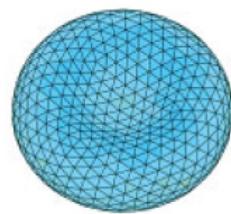
Importance of reference shape



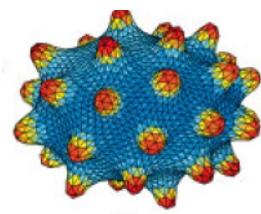
Spectrum of possible reference shapes for RBC skeleton ranging from discocyte (flaccid, left) to sphere (turgid, right). Because experiments can not clarify this point, computation is essential.

[Lim, Wortis, Mukhopadhyay review 2008]

Importance of non-linear elasticity



discocyte



echinocyte

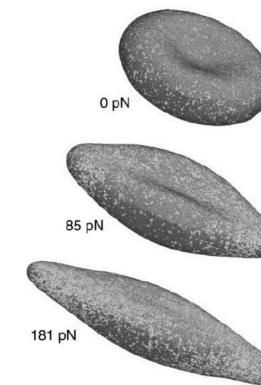
Color code for stretch ratio alpha: echinocytes are highly strained.

[Lim, Wortis, Mukhopadhyay
review 2008]

Stretching RBCs with an optical tweezer

Computational model with
roughly 20.000 vertices

Stretching force given in pico
Newton



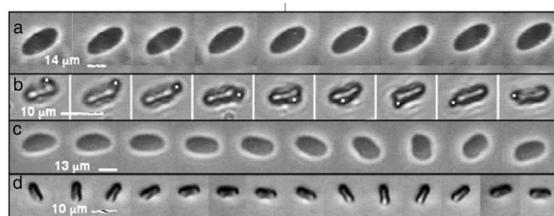
[Li+ BPJ 2005]

RBCs in shear flow

- Single fluid vesicles and RBCs in shear flow undergo tank-treading (TT) and tumbling (TB) motions
- With increasing shear rate, they undergo a TT-TB transition

$$\begin{pmatrix} 0 & \dot{\gamma} \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & \dot{\gamma}/2 \\ \dot{\gamma}/2 & 0 \end{pmatrix} + \begin{pmatrix} 0 & \dot{\gamma}/2 \\ -\dot{\gamma}/2 & 0 \end{pmatrix}$$

linear shear = rotation + elongation
(Geislinger + Franke 2014)



Abkarian PRL 2007 and Soft Matter 2008

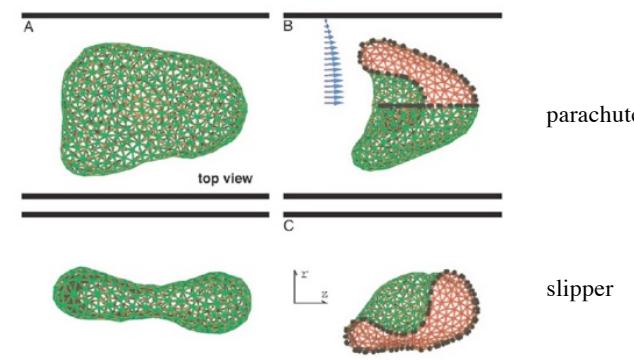
Vesicle TT

RBC TT

Vesicle TB

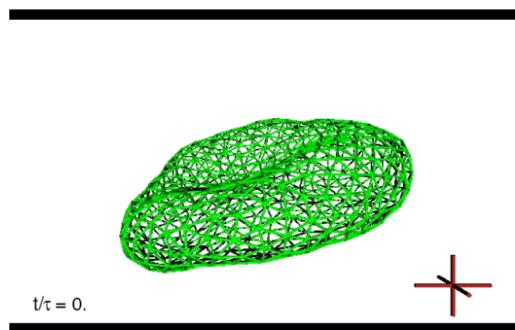
RBC TB

RBC shape transition at high shear



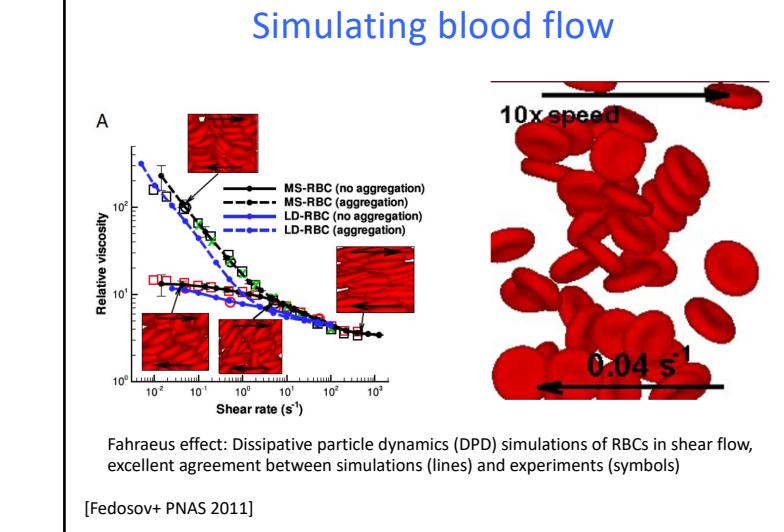
[Noguchi and Gompper PNAS 2005]

Discocyte – parachute transformation



Simulation method: triangularized surface coupled to multi particle collision dynamics (MPCD) fluid

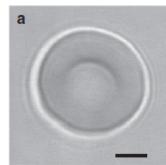
[Noguchi and Gompper PNAS 2005]



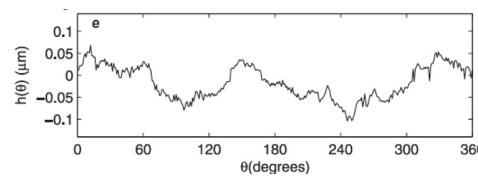
Flickering analysis of RBCs

$$\text{Based on theory of membrane fluctuations: } F = \int_A d\vec{x} \left[\frac{1}{2} \gamma h^2 + \frac{1}{2} \sigma (\nabla h)^2 + \frac{1}{2} \kappa (\nabla^2 h)^2 \right].$$

$$\langle h_q^2 \rangle = \frac{k_B T}{A} \frac{1}{\gamma + \sigma q^2 + \kappa q^4}.$$



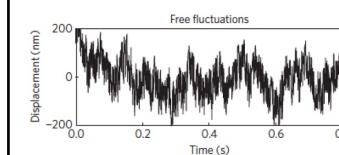
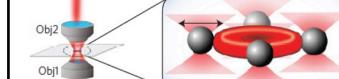
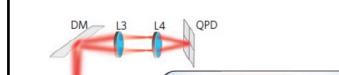
Bright field image of RBC



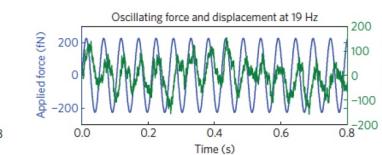
Deviation of height from center line

Yoon et al. Biophysical Journal 2009

Active flickering

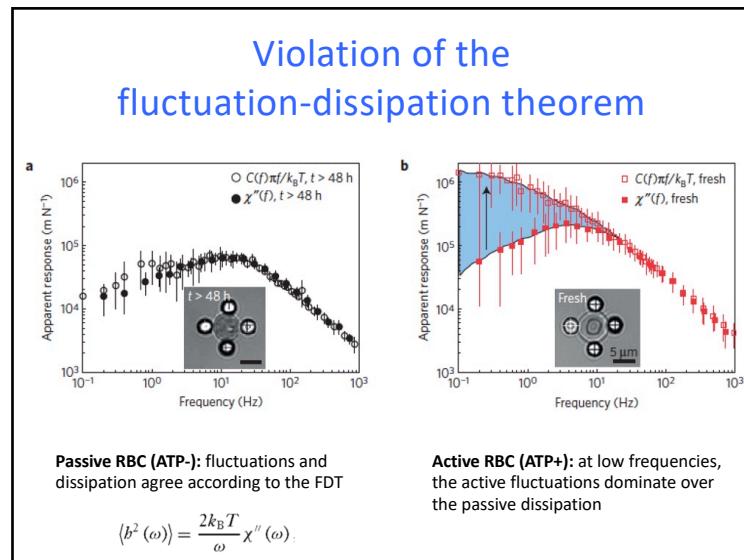


Free fluctuations



Oscillating force and displacement at 19 Hz

Turlier et al. Nature Physics 2016



Summary second part

- RBCs lack actomyosin contractility and adhesion
- Their biconcave shape is explained by bending energy at reduced volume around $v=0.61$
- The spectrin-actin cytoskeleton stabilizes this shape and prevents budding
- RBCs in shear flow assume many different shapes (e.g. parachutes)
- Membrane flickering has a strong active component