A new hyperelastic lookup table for RT-DC

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Real-time deformability cytometry (RT-DC) is an established method that quantifies features like size, shape, and stiffness for whole cell populations on a single-cell level in real-time. A lookup table (LUT) disentangles the experimentally derived steady-state cell deformation and the projected area to extract the cell stiffness in the form of the Young’s modulus. So far, two lookup tables exist but are limited to simple linear material models and cylindrical channel geometries. Here, we present two new lookup tables for RT-DC based on a neo-Hookean hyperelastic material numerically derived by simulations based on the finite element method in square and cylindrical channel geometries. At the same time, we quantify the influence of the shear-thinning behavior of the surrounding medium on the stationary deformation of cells in RT-DC and discuss the applicability and impact of the proposed LUTs regarding past and future RT-DC data analysis. Additionally, we provide insights about the cell strain and stresses, as well as the influence resulting from the rotational symmetric assumption on the cell deformation and volume estimation. The new lookup tables and the numerical cell shapes are made freely available.

Cell stiffness emerged as an important phenotypical marker linked to many biological processes and can reveal pathological changes of cells. The emergence of high-throughput techniques to measure single-cell deformability enables the use of cell deformation as a tool in clinical diagnostics. In this work, we concentrate on the microfluidic technique real-time deformability cytometry (RT-DC). RT-DC is used for high-throughput and label-free mechanical characterization of cells and has been applied in the past to address biological research questions and for blood diagnostics.

In RT-DC, cells flow through a narrow microfluidic channel where they get deformed by hydrodynamic stresses from the flow field around them, reaching a steady state configuration. On the timescale of an RT-DC measurement, viscous effects have no influence on the steady state and the deformation is purely due to the elastic cellular response. The deformed cells are imaged, and the contour is analyzed in real-time at a throughput of up to 1000 cells s⁻¹. Based on this stationary deformation, not only base properties like the cell circumference \( L \), projected area \( A \), and volume \( V \) are deduced but also material properties like the cell’s stiffness (elasticity) in the form of the Young’s modulus. For the latter, the influence of the cell size, measured by the area \( A \), needs to be decoupled from the stationary deformation \( D \), here defined as

\[
D = 1 - \text{circularity} = 1 - \frac{2\sqrt{\pi A}}{L},
\]

which is a measure for the deviation of a shape from a circle (circularity = 1). Analytical and numerical models were developed to generate lookup tables (LUT), mapping an area-deformation pair to a unique apparent Young’s modulus \( E \). Both models and, therefore, lookup tables assume a linear elastic bulk material for the cells measured in RT-DC. The actual elastic cellular response depends on the probing timescale, and deformability cytometry techniques operating on different timescales report different Young’s moduli.

With these LUTs, it is possible to check for structural differences in cell populations of different sizes where the deformation parameter alone is not suited to compare cell stiffness. This becomes important when cells undergo size changes, e.g., during development, or in disease or treatment.

RT-DC measurements are done for various flow rates, channel widths, and measurement buffer viscosities. Mietke et al. showed analytically that isoelasticity lines – the level set for a fixed Young’s modulus – of the lookup table can be scaled to a different channel width \( L’ \), flow rate \( Q’ \) and viscosity of the buffer medium \( \eta’ \) for a Newtonian carrier and linear elastic cell material by \( E’ = E(\eta Q L/\eta’Q’L’)^3 \) and \( A’ = A(L/L’)^2 \). Here, \( \eta, Q, \) and \( L \) are the parameters used to generate the lookup tables.
Thus, the same lookup table can be used for different flow rates, microfluidic channels, and buffers. Those lookup tables and the scaling are implemented in dclab, the Python library for the post-measurement analysis of real-time deformability cytometry.

However, these two models use three simplifications. First, the square cross-section of the channel is assumed to be cylindrical to employ rotational symmetry. Second, they do not consider the non-Newtonian shear-thinning nature of the suspension buffers used in RT-DC, e.g., methyl-cellulose dissolved in phosphate-buffered saline (MC-PBS). And last, both assume a small strain linear elasticity bulk model for the cell. Especially for larger cells or higher hydrodynamic stresses, the small strain assumption does not hold. Recent work on probing viscoelastic properties of cells in RT-DC, dynamic real-time deformability cytometry dRT-DC, showed large deformation in the inflow region of the narrow channel. A numerical study of the viscoelastic behavior of cells and beads in RT-DC has been conducted in the axisymmetric case in Schuster et al. The computational domain, as illustrated in Fig. 1, is split into the fluid subdomain \( \Omega_f \) and the subdomain representing the solid (the cell). Based on this model, we describe the generation of the new lookup tables by sampling the area-deformation interaction finite element simulations. We use a hyperelastic neo-Hookean material model to get the cells’ elastic response in a full three-dimensional square channel with a shear-thinning buffer. We show that, like the initial analytical and numerical LUTs, the two proposed LUTs uniquely map every combination of area and deformation to a Young’s modulus. The scaling concept still holds, including shear-thinning. The new and improved LUTs for RT-DC will help reveal stiffness changes between cell types or due to cell state changes with unprecedented accuracy. Additionally, we report the resulting strains and hydrodynamic surface stresses acting on the cells for the first time. And finally, we investigate the systematic error in volume computation introduced by the widely-used assumption of rotationally symmetric cell deformation.

1 Methods

We start by describing the computational domain and the new fluid–solid interaction (FSI) model with a shear-thinning non-Newtonian fluid (the buffer) and a hyperelastic neo-Hookean solid (the cell). Based on this model, we describe the generation of the new lookup tables by sampling the area-deformation space. At the end of this section, we describe the experimental setup used to measure the HL60 cells and PAAM beads in Fig. 6.

1.1 Fluid–solid interaction model

The computational domain, as illustrated in Fig. 1, is split into the fluid subdomain \( \Omega_f \) and the subdomain representing the cell \( \Omega_c \). The subdomains are separated by the interface \( \Gamma \). We model the fluid in the \( \Omega_f \) by the incompressible Navier–Stokes equations given by

\[
\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} = -\nabla p + \mathbf{f} + \frac{\partial \mathbf{e}}{\partial t} \quad \text{in } \Omega_f
\]

where \( \mathbf{u} \) is the velocity, \( \nabla \) is the gradient operator, \( p \) is the pressure, \( \mathbf{f} \) is the external force, and \( \frac{\partial \mathbf{e}}{\partial t} \) is the strain rate tensor. The strain energy density of the neo-Hookean hyperelastic material described by

\[
W_s = \frac{1}{2} \mu (I_1 - 3) + \frac{1}{2} \kappa (J_3 - 1)^2
\]

where \( \mu \) is the shear modulus, \( \kappa \) is the bulk modulus, \( I_1 \) is the first invariant of the right Cauchy–Green deformation tensor, \( J_3 \) is the determinant of the deformation gradient tensor, and \( J \) is the determinant of the deformation gradient tensor.

On the interface \( \Gamma \), we impose the kinematic condition

\[
\mathbf{n} \cdot \mathbf{u} = 0 \quad \text{on } \Gamma
\]

which is the sum of the isochoric strain energy density (using the isochoric invariant \( I_1 \)) and the volumetric strain energy density (using the elastic volumetric deformation \( J_3 \)). In the case of an incompressible material, the parameters are given by the Lamé parameter \( \mu = E/3 \) and the bulk modulus \( \kappa = E/3 \).

1.2 Finite-element implementation

The above system of equations is discretized and solved by the finite element method using COMSOL Multiphysics. For the
full three-dimensional simulations of the square channel, we reduced the computational effort by exploiting the 4-fold rotational symmetry and 2-fold reflection symmetry of the rectangular channel cross-section (see Fig. 1 on the right). We can exploit the rotational symmetry and reduce the computational domain to a two-dimensional surface for the cylindrical channel lookup table. The radius $R$ of the cylindrical channel is adapted to $R = 1.094 \times l_c$, where $l_c$ is the square channel width, such that the pressure drop over the cylindrical channel is equal to the square channel (see the concept of equivalent channel described in Mietcke et al.\textsuperscript{12}). The resulting computational domain is meshed with a combination of tetrahedral, prisms and pyramid elements to efficiently reduce the number of degrees of freedom even further. To resolve the high gradient in the flow profile on the channel walls and the cell surface, we added a boundary layer on $\Gamma_w$ and $\Gamma_c$. We choose linear Lagrange elements (P1) to discretize the Navier–Stokes equation for both the velocity field and pressure and stabilize the saddle-point problem with streamline and crosswind diffusion. Linear elements are used to discretize the displacement field of the hyperelastic material. We did not experience any locking behavior during our model validation by comparing the solution with quadratic elements for the solid.

The correct inflow profile $\mathbf{u}_i$ is precomputed in the absence of a cell, based on the flow rate, channel width, and the non-Newtonian parameters of the fluid and used as Dirichlet boundary condition at the inlet. At the outlet $\Gamma_o$, the pressure is fixed to $p_o = 0$, and we suppressed any eventual backflow. To keep the cell centered in the computational domain, we describe all quantities in a coordinate system that moves with the cell. We use a proportional integral derivative (PID) controller to adjust the wall speed $\mathbf{u}_w$ on $\Gamma_w$ based on the barycentre of the cell. This wall speed is subtracted from the inflow profile such that the resulting system is the same as if the cell moves through the channel. The arbitrary Lagrangian–Eulerian (ALE) description is employed for the deformation of the fluid and used as Dirichlet of a cell, based on the flow rate, channel width, and the non-Newtonian parameters of the fluid and used as Dirichlet of a cell, based on the flow rate, channel width, and the non-Newtonian parameters of the fluid and used as Dirichlet.

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To improve the stability of the simulations, we add an artificial viscosity to the cell in a Kelvin–Voigt-like manner, and we monotonically ramp up the inflow profile from $\mathbf{u} = 0$ to $\mathbf{u}_i$. We stop the simulation as soon as the steady state deformation is reached by measuring the absolute deformation change $\Delta d$ and set as the stop criteria $\Delta d = 1 \times 10^{-3}$. The artificial viscous term and the ramp-up do not influence the steady state. The fully bidirectional coupled geometrical non-linear system is solved with a Newton method in time and the PARDISO direct solver in space.

### 1.3 Creation of the new lookup tables

We indirectly sampled the area-deformation space for the new lookup tables by the radius and Young’s modulus of the cells. The radii $R$ are chosen equidistant in $R^2$ such that the resulting projected areas should be in the range of $[30 \, \mu m^2, 290 \, \mu m^2]$ (the actual projected cell area is not conserved). Similar, the Young’s modulus $E \in [0.3$ kPa, $30$ kPa] is equally sampled in $E^{-1/2}$. For each $R \times E$ combination, we ran two simulations, one in a square channel with a channel width of $l_c = 20 \, \mu m$ and one in an equivalent cylindrical channel (see above). The cell has a density $\rho_c = 1000 \, kg \, m^{-3}$ and we set the bulk modulus $k = 2.15$ GPa corresponding to the bulk modulus of water, leading to a quasi-incompressible cell. The artificial cell viscosity is set to $10 \, mPa \, s$. For both lookup tables, the flow rate is $0.04 \, \mu l \, s^{-1}$. The fluid parameters are based on $0.6\%$ MC-PBS (see below). Additionally, we set the density to $\rho_f = 1000 \, kg \, m^{-3}$.

We did several data-cleaning steps for both lookup tables to ensure that the simulations converge and are not numerical artifacts. We automatically detect and delete outliers based on polynomial regression and the random sample consensus (RANSAC) method. Additionally, we drop all the simulations with a deformation $D \leq 5 \times 10^{-4}$, since such deformation values are not measurable by RT-DC and are close to the perfect sphere discretized by the mesh. The final lookup tables are built by linearly interpolating the 22,558 obtained cell shapes for the cylindrical (2D axisymmetric) channel and 1206 cell shapes for the square (3D) channel to a uniform grid for fast lookup.

### 1.4 Cell culture

The HL60/S4 cell subline (ATCC Cat# CRL-3306, RRID:CV-CL_II77) was cultured in RPMI 1640 medium with 2 mM L-Glutamine (Thermo Fisher #A1049101) with 1% penicillin and streptomycin (Gibco) and 10% heat-inactivated fetal bovine serum (Sigma Aldrich, catalog no. F4135, lot no. 135C519). Cells were grown at $37^\circ C$, with 5% CO$_2$, at densities between $10^5$–$10^6$ cells per mL with subculturing every 48–72 hours.

### 1.5 Buffer production for RT-DC

Samples for RT-DC experiments were suspended in a buffer solution made from 0.594 w/w% methyl cellulose (MC; 4000 cPs, Alfa Aesar 036, 718.22, CAS9004-67-5) dissolved in phosphate-buffered saline (PBS) without Mg$^{2+}$ and Ca$^{2+}$ (Gibco Dulbecco 14190-144). Buffers were adjusted to have a viscosity of 25 mPa s when measured in a HAAKE Falling Ball Viscometer type C (Thermo Fisher Scientific, Dreieich, Germany) using ball number 3 at 24 °C.

The rheology of the 0.6% MC-PBS solutions was measured with an Anton Paar MCR 520WESP TwinDrive rheometer (Anton Paar GmbH, Graz, Austria) at 25 °C controlled with a PTD 180 MD Peltier element (Anton Paar GmbH, Graz, Austria). At shear rates over 5.000 s$^{-1}$, the solutions showed power-law behavior, and the fluid consistency coefficient was found at $m = 0.4057$ Pa s and the flow behavior index at $n = 0.6039$. For the 0.5% MC-PBS buffer in the scaling validation, we used the measured fluid consistency coefficient of 0.2671 Pa s and a flow behavior index of 0.6264. For a detailed description of the buffer rheology, see Büyükgüngör et al.\textsuperscript{21}

### 1.6 Cell and beads measurements

Cell and beads experiments were performed with a commercially available RT-DC device (AcCellerator, Zellmechanik Dresden...
GmbH). The measurement principle of RT-DC is described in detail in Otto et al. Briefly, a microfluidic chip containing the measurement channels with a cross-section of 20 × 20 μm, that were simulated in this work, is mounted on an inverted microscope (Axiovert 200 M, ZEISS, Oberkochen, Germany) and flow is introduced with syringe pumps. Images are captured with a CMOS camera (Mikrotron, Unterschleissheim, Germany). Syringe pumps (Cetony Nemesys, Korbussen, Germany) and camera were controlled with the measurement software Shapeln (Zellmechanik Dresden, Dresden, Germany), which analyses contours in real-time.

For the beads experiments, 2.5 μL of polyacrylamide (PAAm)-beads solution (monomer concentration = 7.9%; for details, see Girardo et al.) were resuspended in 47.5 μL of 0.6% MC-PBS and measured at a flow rate of 0.04 μL s⁻¹.

For cell experiments, 2 mL of HL60-cell solution (5 × 10⁵ cells mL⁻¹) was centrifuged at 188 RCF for 4 minutes. After the supernatant was removed, the cell pellet was resuspended in 100 μL of 0.6% MC-PBS for a final cell concentration of 10⁷ cells mL⁻¹ and measured at 0.04 μL s⁻¹.

In both cases, the measured sizes have been pixelation corrected (see Herold et al. for more details), and only cells and beads with a ratio of convex-hull area to the raw area of 1.0–1.05 are considered. This step guarantees that noisy contours that could lead to overestimated deformation values are not considered in the analysis.

2 Results

The new lookup tables are based on finite element simulations, where we indirectly sample the area-deformation space by varying the cell radius and Young’s modulus. Thereby, we get for each fixed Young’s modulus an isoeelasticity line (see Mietke et al.) in the area-deformation space. We begin by describing one of these numerically obtained cell shapes and show the flow field around the in silico cell and the surface stresses acting on it. Next, we validate the new LUT’s by comparing to the linear elastic lookup table from Mokkel et al. and show that the scaling still holds. We compare the new lookup tables in the cylindrical and square channel to the previous lookup table. Following this, we show the influence of the different lookup tables on the resulting apparent Young’s modulus. We conclude this section by reporting the engineering strains and surface stresses on biological cells and the relative volume error resulting from the axisymmetric channel assumption in the steady-state configuration.

2.1 Cell shape and stress distribution

The new proposed LUT’s assume a neo-Hookean hyperelastic material for biological cells and are deduced from full three-dimensional finite element simulations and two-dimensional axisymmetric simulations. In Fig. 2, we show one of the three-dimensional simulations. The in silico cell reaches a bullet-shaped stationary deformation (see Fig. 2A) due to the hydrodynamic stresses from the surrounding fluid. The stationary configuration of the cell is not rotationally symmetric. The flow profile around the cell from the front is shown in Fig. 2B. The stresses on the cell surface can be split into normal (pressure) and tangential (shear) stress contributions. Fig. 2C shows the pressure on the cell surfaces and the shear stress distribution. We subtract the average surface pressure \( p_{avg} = \int p \, dA / \int 1 \, dA \) since the pressure enters the system as a gradient only and, thus, the absolute value is not meaningful. The pressure is non-symmetric, with the highest absolute value at the front of the cell. Here, we define the shear stress \( |P_{\sigma_c \mathbf{n}}| \) where \( P = I - \mathbf{n} \otimes \mathbf{n} \) is the surface projection, \( \sigma_c \) the stress tensor of the cell and \( \mathbf{n} \) the surface normal. The peak shear stress is located in the regions closest to the channel wall.

Only the projected area and deformation of the resulting cell shape are used for the lookup creation. In Fig. 2D, we plot this projected contour for the cylindrical channel overlaid with the projected contour in the square channel. The corresponding deformation values are \( D_{\text{cylindrical}} = 0.071 \) and \( D_{\text{square}} = 0.086 \). For comparison, the contour of the cell along the diagonal is shown with a dashed line.

2.2 Model and scaling validation

Next, we validate that the resulting lookup tables are – like the linear elastic LUT – bijective (no isoeelasticity lines cross each other), and the scaling holds for the hyperelastic material model, the non-Newtonian shear-thinning buffer as well as the square channel geometry. Isoelasticity lines for different Young’s moduli in the cylindrical channel with and without a shear-thinning buffer are shown in Fig. 3 (left). In Fig. 3 (right), the same is shown for the square channel with a shear-thinning buffer only. In both plots, the isoeasticity lines from Mokkel et al. for the linear material model without shear thinning, and cylindrical channel are added for comparison (dotted lines). All three lookup
we include the non-linear shear-thinning behavior of 0.6% MC-PBS for the linear elastic material model without shear-thinning (dotted, see Mokbel et al.\textsuperscript{13}), the new hyperelastic material model with shear-thinning (solid) and without shear-thinning (dashed). On the (left) for the cylindrical (2D-axisymmetric) channel. The new hyperelastic material based LUTs. For stiff beads, this is the case for almost all cell sizes. Reducing the cell stiffness leads to increased divergences of the isoelasticity lines for larger cells. For soft beads, the isoelasticity lines diverge already for small bead sizes. This holds true for the cylindrical channel simulations as well as for the square channel simulations. In the case of a cylindrical channel, the isoelasticity lines flatten out for Young's moduli around 1 kPa for larger cells. In contrast, the linear elastic material predicts a higher increase in deformation. The square channel lookup table (Fig. 3 (right)) shows a similar pattern. There, the isoelasticity lines agree for stiff cells. For softer cells, the isoelasticity lines diverge already at smaller cell sizes, and the deformations are larger. For larger cells with Young's modulus \( \leq 1.5 \) kPa, the isoelasticity lines seem to flatten out too, but the range in terms of numerically stable simulations is smaller, resulting in a smaller lookup table than in the cylindrical channel. Additionally, for the cylindrical channel, we investigate the influence of shear-thinning on the isoelasticity lines by comparing to the deformations in a Newtonian fluid with a fixed apparent viscosity \( \eta = 6 \) mPa s (see Herold et al.\textsuperscript{15} for derivation from a shear-thinning fluid). The resulting isoelasticity lines are shown again in Fig. 3 (left) as dashed lines. For smaller cells, shear thinning is negligible for all cell elasticities. Again, a Newtonian fluid results in a higher deformation for the larger cells. This leads to an underestimation of the apparent Young’s modulus for larger cells. Overall, the agreement between the linear elastic and hyperelastic isoelasticity lines for small strains validates the new numerical results.

Validating the scaling discussed above\textsuperscript{12} is non-trivial, as we include the non-linear shear-thinning behavior of 0.6% MC-PBS and hyperelastic neo-Hookean material model for the cell and consider the full three-dimensional geometry. The new lookup tables are based on a channel side length \( L = 20 \) \( \mu \)m, a flow rate of \( Q = 0.04 \) \( \mu \)L s\(^{-1}\) and an apparent viscosity of 0.6% MC-PBS which results in \( \eta = 6 \) mPa s. Here, to validate the scaling numerically, we perform two \textit{in silico} experiments in square channels with a side length of \( L' = 30 \) \( \mu \)m, a flow rate of \( Q' = 0.16 \) \( \mu \)L s\(^{-1}\) and with 0.5% MC-PBS and 0.6% MC-PBS, resulting in apparent viscosities \( \eta_1 = 5.64 \) mPa s and \( \eta_2 = 4.76 \) mPa s. We scaled the Young's moduli and derived the deformation values from the adapted simulations. After rescaling the area \( A' \), the isoelasticity lines in Fig. 4 (left) agree perfectly except at the boundary, indicating that the scaling still holds at least for the most part of the LUT and only at the boundary it might become inaccurate.

The ratio between the two axial second moments of area is another dimensionless integral quantity to describe cell deformation, introduced in Mokbel et al.\textsuperscript{13} We use the definition of this inertia ratio \( I \) implemented in dclab,\textsuperscript{15} which for horizontally symmetric shapes is defined by

\[
I = \sqrt{\frac{I_{yy}}{I_{xx}}},
\]

\[
I_{xx} = \int_A (y - y_b)^2 \, dA,
\]

\[
I_{yy} = \int_A (x - x_b)^2 \, dA,
\]

where \( A \) is the projected cell area and \( (x_b, y_b) \) is the cell barycentre.\textsuperscript{26} Similar to isoelasticity lines in the area-deformation plot, one can plot the isoelasticity lines in the area-inertia ratio space shown in Fig. 4 (right). Cells that are more stretched in the direction of the fluid flow (prolate) have an inertia ratio \( I > 1 \). An inertia ratio \( I < 1 \)
This means that the cells are more compressed in the flow direction and elongated perpendicular, thus resembling an oblate. The area-inertia ratio lookup table is not bijective, as already reported in Mokbel et al.\textsuperscript{13} Around an area of 108 m\textsuperscript{2}, the isoelasticity lines cross each other, meaning that a unique lookup for the apparent Young’s modulus is not possible in this region. But, the scaling does hold too for the inertia ratio as seen in Fig. 4 (right). Thus, the presented lookup tables can be used for deriving uniquely the apparent Young’s modulus for a broad range of experimental setups from the area-deformation space and, to some extent, from the area-inertia ratio space.

2.3 The New Lookup Tables and the Geometric Influence
Having validated the numerical simulations and the scaling, we can construct the final lookup tables based on many simulations. As already described above, the lookup tables are based on simulations with a channel side length $L = 20\,\mu\text{m}$, a flow rate of $Q = 0.04\,\mu\text{L}\,\text{s}^{-1}$ and 0.6% MC-PBS. The cylindrical channel simulations are numerically more stable as the confinement – the ratio between the cell diameter and channel width – is smaller, i.e., the channel width is larger to account for the changed pressure drop. For more details of the concept of equivalent channel between the square and cylindrical channel, see Mietke et al.\textsuperscript{12} Fig. 5 shows the region of all simulations for the two new lookup tables in the area-deformation space and area-inertia ratio space. The space of the cylindrical lookup table spans $[28\,\mu\text{m}^2, 335\,\mu\text{m}^2] / [0.377\,\text{kPa}, 23.753\,\text{kPa}]$ and the square channel $[29\,\mu\text{m}^2, 281\,\mu\text{m}^2] / [0.469\,\text{kPa}, 27.669\,\text{kPa}]$. Since the area is not conserved and not all radius-Young’s modulus pairs are numerically stable, the resulting LUTs are not rectangular.

![Fig. 4](image-url) Scaling of the square channel lookup table: isoelasticity lines of the new lookup table (solid) in the square channel compared to rescaled data with a wider channel geometry and a higher flow rate (dashed), and additionally a different fluid (dotted). The isoelasticity lines and coloring are the same as in Fig. 5 (left). The Young’s modulus $E$ and the projected cell area $A'$ were scaled to match the new channel geometry, flow rate, and apparent viscosities. (left) Isoelasticity lines and rescaled isoelasticity lines for area-deformation, (right) Isoelasticity lines and rescaled isoelasticity lines for area-inertia ratio.

![Fig. 5](image-url) Comparison of the new lookup tables in the cylindrical and square channel: (left) Isoelasticity lines of cells in the square channel (solid) and cylindrical channel (dotted). The shaded areas indicate the region of all the simulations. (right) Inertia ratio of the two lookup tables: solid lines for the square channel and dotted lines for the cylindrical channel. The area with an inertia ratio $I < 1$ is indicated by the enclosed area with the gray line on the left.
Again, we show the isoelasticity lines for the same Young’s moduli as in Fig. 3. The area where the inertia ratio of the square channel is \( l < 1 \) is enclosed in Fig. 5 (left) by the grey line.

The channel geometry does have an observable influence on the deformations. On the technical side, the simulations are more stable in the axisymmetric setting, as discussed above, resulting in a larger lookup table. For small cells as well as stiff cells, the isoelasticity lines from the two new LUTs overlay. But for larger and softer cells, the isoelasticity lines of the cylindrical and square channel diverge. As the deformation is smaller in the cylindrical channel, the cylindrical lookup table underestimates the apparent Young’s modulus.

### 2.4 Applying the new LUTs to Measurements

With the validated new lookup tables, it is possible to extract the apparent Young’s modulus from RT-DC measurements. To demonstrate this, we measured poly-acrylamide (PAAm) hydrogel beads and HL60 cells (see Methods). The kernel density estimations of the measured bead and cell distributions in the area-deformation plot are shown in Fig. 6 (left). All three measurements fall within the region of definition of both new LUTs, with a few sample points outside the square channel LUT (resulting in omitted data points). The resulting apparent Young’s moduli are shown in Fig. 6 (right).

The cylindrical LUT predicts similar elasticity values and distribution than the linear elastic LUT of Mokbel et al. \(^{22} \) For small beads, the new cylindrical channel LUT predicts a higher Young’s modulus compared to the linear elastic LUT, whereas for the larger beads the Young’s moduli decrease (see the grey dashed lines in Fig. 6 (right)). This is due to the non-linear material response discussed above and seen in the flattening isoelasticity lines. For the HL60 cells, which have a higher deformation than the small beads, the stiffness increases compared to the linear elastic LUT too. Comparing the two new LUTs, the square channel LUT predicts higher Young’s moduli for all three measurements. This holds true compared to the old linear elastic material model in the cylindrical channel. At the same time, the shape of the distribution changes compared to the cylindrical LUT, supporting the importance of the channel geometry.

Remarkably, the HL60 cells show a greater spread of the data points in the area-deformation plot than the PAAm beads but a much narrower distribution of Young’s moduli for all types of LUTs. On closer inspection, the HL60 fall along the trajectory of a few isoelasticity lines, while the beads cross several lines. This highlights an important use case of the LUTs: when comparing samples of different size in RT-DC, the deformation parameter alone is not sufficient because it is dependent on the area. In our example it shows that the HL60 are quite homogeneous in Young’s modulus compared to the PAAm beads even though they are more heterogeneous in size. The heterogeneity of the Young’s modulus of the PAAm beads is a consequence of the production process and is discussed in more detail in Girardo et al. \(^{22} \)

### 2.5 Strain and stresses of the cells

Next, we will look at the strain and stresses on the cell surface. Deformation as a measure of strain approximates the resulting shapes by a single number. In Fig. 7, we plot the strains along the horizontal \( x \)-axis and the vertical \( y \)-axis to get a better understanding of the deformed shapes. Here we use the engineering strain defined as \( (l + l_0)/l_0 \) and \( (h + h_0)/h_0 \) where \( l \) and \( h \) are the maximum extensions of the cell in \( x \) and \( y \) direction. \( l_0 \) and \( h_0 \) are the diameter of the cell in the undeformed state with \( l_0 = h_0 = 2R \). Small cells are horizontally compressed but elongated vertically. Larger cells are exposed to higher shear force, resulting in a more stretched shape along the flow direction and compressed in the \( y \)-axis. The strains of the cells in the square channel are in the range of \([-4.5\%, 13\%]\)] in the cylindrical channel in the range of \([-9.9\%, 33.6\%]\).

The maximum stress components acting on the cells are shown in Fig. 8 for the square and cylindrical channel.

![Fig. 6](https://example.com/fig6.png)  
**Fig. 6** RT-DC measurements of PAAm beads and HL60: (left) The distributions of PAAm beads with two different sizes and HL60 cells from separate measurements (by bivariate kernel density estimation). The marginal distributions for area and deformation are shown on the top and on the right. The isoelasticity lines from Fig. 5 (left) are shown in grey. (right) Violin plots of the estimated Young’s modulus from the three measurements. The resulting Young’s modulus distribution with the cylindrical LUT is on the left in each violin, for the square channel LUT on the right. The dashed lines within the violins are the quartiles (25%/50%/75%). The grey dashed lines indicate the Young’s modulus distribution from the cylindrical LUT of Mokbel et al. \(^{22} \)
shear stresses are in the range of [117 Pa, 1058 Pa] in the square channel and [118 Pa, 624 Pa] in the cylindrical channel, again increasing with the cell size. The largest cells with a large Young’s modulus are exposed to the highest shear stress, as their surface is closest to the channel wall. The maximum pressure is in the range of [226 Pa, 1346 Pa] in the square channel and [220 Pa, 1022 Pa] in the cylindrical channel, respectively. We subtract the average surface pressure $p_{\text{avg}}$ again (see above).

### 2.6 Volume estimation error

The volume of cells flowing through the microfluidic chip is a valuable cell feature. It is approximated by revolving the detected contour of the stationary shape and pixelation corrected similar to the cell deformation and area (see Methods). The rotational-symmetry assumption is only valid for small cells, perceiving the almost axisymmetric flow field in the center of the channel. The absolute value of the relative error is below 1% for cells with a radius $R < 6.2 \ \mu\text{m}$ (i.e., confinement of less than 62%) and any Young’s modulus considered. For larger cells, the relative error depends on the Young’s modulus and the cell size. Fig. 9 shows the relative error based on measured area and deformation. The relative error $\epsilon_{\text{vol}}$ is defined by $\epsilon_{\text{vol}} = 100 \times \left(\frac{V_{\text{rot}}}{V_0} - 1\right)$ where $V_{\text{rot}}$ is the approximated volume from revolving the contour of the in silico cell$^{20,26}$ with radius $R$ and $V_0$ is the volume of a sphere with radius $R$. Interestingly, the error is only in the range of $[-3.87\%, -0.06\%]$. The volume of small cells is underestimated only slightly compared to the actual cell volume independently of the deformation and thus apparent Young’s modulus. For larger cells, the influence of the deformation on the volume error increases.

### 3 Discussion and conclusion

This work presents two new lookup tables for RT-DC for cylindrical and square channels based on the neo-Hookean hyperelastic material model. We validate the numerical model by comparing to the linear elastic LUT of Mokbel et al.$^{13}$ in the small strain regime. We discuss the influence of different material models, the shear-thinning behavior of the fluid, and the channel geometry and reveal different material responses. We find that the scaling still holds for the hyperelastic material even with a shear-
thinning non-Newtonian fluid. In consequence, the proposed LUTs can be used for a vast array of different channel sizes, buffer viscosities, and flow rates. Shear-thinning does affect the isoelasticity lines for large cells only. Compared to the existing linear elastic material LUT, the more accurate hyperelastic square channel LUT predicts stiffer cells, especially for small, highly deformed cells. However, the square channel LUT has a limited range of possible cell sizes and deformations due to the smaller region of numerical stable simulations. On the other hand, the cylindrical channel LUT has a similar size as the linear elastic material LUT. Depending on the position of the measurements in the area-deformation space, the distribution of the Young’s modulus is altered for both proposed LUTs compared to the existing one. For both proposed LUTs, we report the strain and surface stresses of the cells depending on the projected cell area and deformation. Both new lookup tables are available in dclab15 and freely available in Wittwer et al.,27 extending the predictive power to derive the apparent Young’s modulus of any biological cell type based on the given model assumptions. At the same time, the proposed LUTs complement the work in Wittwer et al.20 to derive the viscoelastic response of biological cells in a square RT-DC geometry. The reported findings and freely available data are not only applicable to RT-DC, but to similar flow cytometry setups as well. A lookup based on area and deformation ignores the actual shape of the measured cells. Extending the search space by considering the contour shapes would considerably improve the predictive power of RT-DC and could take into account different material models for different cell types and beads. However, this leads to a significant increase in parameters and is thus left for future studies.

Data availability statement

The data that support the findings of this study are openly available27 on figshare at https://doi.org/10.6084/m9.figshare.20630940.

Author contributions

Lucas D. Wittwer: conceptualization (equal); data curation (lead); formal analysis (lead); software (equal); visualization (lead); writing – original draft (lead); writing – review and editing (equal). Felix Reichel: data curation (supporting); writing – original draft (supporting); writing – review and editing (equal). Paul Müller: data curation (supporting); software (equal); writing – review and editing (equal). Jochen Guck: conceptualization (equal); funding acquisition (equal); supervision (equal); writing – review and editing (equal). Sebastian Aland: conceptualization (equal); funding acquisition (equal); supervision (equal); writing – review and editing (equal).

Conflicts of interest

J. G. and P. M. are co-founders of Rivercyte GmbH.

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