

User guide to the red blood cell model (RCM), a multiplatform JAVA-based model of human red blood cell homeostasis

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Abstract

We introduce here a new multiplatform JAVA-based mathematical-computational model of RBC homeostasis for investigating the dynamics of changes in RBC homeostasis in health and disease. We provide a brief overview on the homeostasis of human RBCs and on the general biophysical principles guiding the modelling design. By way of a detailed tutorial we apply the model to investigate in depth the multiple effects associated with RBC dehydration induced by potassium permeabilization, a necessary preliminary for understanding the pathophysiology of a wide group of inherited haemolytic anaemias, a subject of intense current research and clinical interest. Using the red cell model (RCM), we design and run *in silico* representations of experimental protocols to study global RBC responses to calcium and potassium permeabilization covering a wide range of experimental, physiological and pathological conditions. Model outputs report the evolution in time of all the homeostatic variables in the system allowing, for the first time, a detailed and comprehensive account of the complex processes shaping global cell responses. Analysis of the results explains the mechanisms by which the entangled operation of all the RBC components link cell dehydration and protein crowding to cell acidification and to the generation of hypertonic, alkaline effluents. Open access to the RCM in a GitHub repository, together with the tutorial primed for a specific investigation pave the way for researchers and clinicians to apply the model on many different aspects of RBC physiology and pathology.

The red blood cell model (RCM) and User Guide: what to expect

The User Guide contains example-based tutorials on how to apply the RCM in research, clinical and teaching contexts. Parameter values in the model are tightly constrained by solid data from the literature enabling model predictions to be semi-quantitative to levels of between 5 and 10%. To interpret this margin it is important to realize a critical difference between experiments at the bench and their model representation. At the bench, experiments are performed on cell populations in suspensions or on chosen isolated cells from a cell population sample, as with patch-clamp or electrode measurements. The RCM, on the other hand, predicts the behaviour of a cell defined by the constitutive properties attributed to the modelled cell when specifying initial conditions.

The model becomes a singularly useful tool when comparisons between predicted and measured results depart from that accuracy margin. Such departures expose either ignored model components required for a particular cell response, or unexpected abnormalities or heterogeneities in the RBC population under investigation. To investigate population behaviour with the model it is necessary to run multiple identical simulations on RBCs defined with hypothesized variations in cell properties or components [1, 2]. Weighted proportions attributed to each variant within a defined distribution render predicted outcomes for the integrated response of the cell population under study. These can then be compared with experimental results to trace the nature of the hypothesized variant or distribution abnormality. The full sets of red blood cell parameters and variables are listed in the Appendix.

General Introduction

A brief primer on red blood cell homeostasis. The main function of RBCs is to mediate the transport of O₂ and CO₂ between lungs and tissues by mechanisms evolved to minimize the energy cost to the organism. A first critical feature enabling such economy is the extremely low cation permeability of the RBC membrane [3-5]. This allows the cells to maintain steady volumes for extended periods of time with minimal cation traffic, pump-leak turnover rates and ATP consumption. Glycolytic ATP turnover by the full RBC mass of a healthy human adult amounts to less than 0.06% of the total body ATP turnover [6].

A second critical feature of the optimized economy concerns the compromise between RBC turnover rate and circulatory lifespan. RBCs are the most abundant cells in the body, their mass adapted for adequate gas transport at all levels of physiological demand. Biosynthetic and biodegradable replacement of such a large

cell mass imposes a heavy metabolic cost to the organism which can only be reduced by extending the circulatory lifespan of the cells thereby reducing their replacement frequency. Circulatory longevity, on the other hand, is limited by the extent to which RBCs, without nucleus and organelles, and devoid of biosynthetic capacity, can sustain the functional competence of its metabolic and membrane transport components required for volume stability and optimal rheology. Optimal rheology requires that the RBCs retain a large degree of deformability for rapid passage through narrow capillaries and for ensuring minimal diffusional distances for gas-exchange across capillary walls. Deformability, in turn, depends on the RBCs maintaining their volume well below their maximal spherical volume (reduced volume), a condition fulfilled when their reduced volume is kept within a narrow margin around 60% [6-8]. For human RBCs with a mean circulatory lifespan of about 120 days, this represents a substantial challenge.

RBC homeostasis involves a subset of basic cellular processes that attempt to maintain and restore cell volume and integrity throughout all dynamic changes elicited by physiological stress. Although all RBC components participate in global cell responses, the main players within the subset concerned with RBC homeostasis are the full constellation of passive and active membrane transporters, haemoglobin as the main macromolecular colloid osmotic contributor, proton and calcium buffer, the variety of impermeant cell metabolites and solutes contributing fixed negative charges and additional proton, calcium and magnesium buffering capacity to the RBC cytoplasm. Most of the haemolytic anaemias affecting humans result from parasitic invasions or from inherited mutations in components of the homeostasis subset. Early versions of the red blood cell model proved a valuable tool for unravelling the complex pathophysiology in some of these diseases [9-12].

Introducing the red blood cell model. The RCM describes the dynamic behaviour of a suspension of identical red blood cell in plasma-like media, treated as a closed two-compartment system. The physical laws constraining the behaviour of such a system are charge and mass conservation.

Relative cell volume (RCV) is the critical variable in models of cellular homeostasis. RCV has a maximum value corresponding to a spherical configuration, leading to cell lysis when exceeded. Following Ponder (1948)[13] the critical haemolytic volume of human RBCs was set to 1.7 times the RCV of 1 assigned to the modelled cell when setting its initial conditions during model simulations. Cell membrane areas are assumed constant. Cell shape changes are only indirectly addressed in the model through their connection to reduced volumes, the ratio between actual and maximal spherical RCVs.

Models of cellular homeostasis start with an electroneutral cell system in an initial reference steady-state or quasi-steady-state. Following perturbations, the equation that ensures sustained charge conservation and electroneutrality in a system with j electrogenic transporters is $\sum I_j = 0$. Each I_j describes the current carried by channels or electrogenic carriers and pumps across the membrane. $\sum I_j$ is a function of the membrane potential, E_m , of the concentration of the transported substrates on each membrane side, on modulating factors such temperature, pH, $[Ca^{2+}]_i$, $[Mg^{2+}]_i$, other ions, solutes and metabolites. $\sum I_j = 0$ is therefore a complex implicit equation. When the parameters and kinetics of all the individual flux equations are either known or phenomenologically defined, as expected for a system mature for modelling such as the human red blood cell, $\sum I_j = 0$ becomes an implicit equation in E_m , the single unknown left. An Appendix with the governing equations of the RCM (January 2021 update) is available for downloading as S1 (PDF) from shared supporting material in Rogers and Lew, 2021 [14, 15].

Avoidance of explicit equations for E_m in homeostasis models. Models seeking simplified explicit formulations of the E_m equation, as attempted in the past [16, 17], are misguided. To retain the flexibility to change the flux-kinetics of individual transporters in the search for improved fits between predicted and experimental results, the freedom to modify the terms within $\sum I_j = 0$ is essential, a process fatally compromised by seeking simplifying assumptions in the search for explicit E_m formulations. Moreover, simplifying attempts are rendered futile by current computational power.

Membrane capacitance may be neglected in homeostasis models. Em variations change the charge on the membrane capacitor. Such changes can almost always be ignored because the amount of charge displaced, as well as the duration of the current transients, $I_c(t) = C \cdot dV/dt$, are orders of magnitude below the magnitude and time-course of the ionic currents of homeostatic relevance [18]

$\sum I_j = 0$ and *Entanglement.* Charge conservation imposes an extraordinary level of entanglement in the processes that shape global cell responses. For example, consider an ion channel becoming activated in a cell at rest. Em changes and so do all voltage-sensitive I_j , causing secondary changes in the concentrations of transported substrates. Some of these substrates are shared by electroneutral transporters, altering their fluxes and causing additional changes in the concentrations of intracellular solutes. In turn, all transport, biochemical and metabolic processes that depend on the concentration of intracellular solutes and on cell osmolarity also become affected generating chains of complex, interlinked interactions all influencing the global cell response. The magnitude of individual and global changes may vary between miniscule or huge scales, but the web of interconnected influences is always there. Homeostatic entanglement renders intuition a very fallible instrument for predicting global cell behaviour and for understanding the mechanisms behind complex cell responses, as was amply demonstrated in the past [10, 11, 19-27].

The JAVA version of the RCM. Building on the same approach in which unexpected but experimentally verified model predictions helped solve issues of substantial relevance in biology and medicine in the past [9, 28-34] we introduce here a vastly extended and fully updated JAVA model of the homeostasis of human RBCs using the Gardos effect, of historic and renewed relevance, to demonstrate how to use the model to unravel the molecular interactions that participate in processes controlling or affecting RBC hydration in physiological and pathological conditions and to explain the mechanisms behind, an issue of intense current haematological interest [8, 35-42] .

The JAVA model introduced here includes a graphical user interface along with a link to the GitHub repository where the code is held and continually updated. In the Appendix we provide a computational flowchart, full details of the model parameters and variables, and of extensions based on literature updates. The updates incorporate, among others, cytoplasmic Ca^{2+} and Mg^{2+} buffering, ionophore-mediated transport, the effects of oxy-deoxy transitions on the isoelectric point of haemoglobin, cell pH and $[Mg^{2+}]_i$ levels, and PIEZO1 channels in the RBC membrane, of particular relevance for the study of RBC homeostasis changes in the circulation in health and disease.

Comprehensive analysis of the full spectrum of changes in RBC homeostatic variables enables here, for the first time, a detailed understanding of the sets of interconnected non-linear processes that modify the intracellular milieu, membrane potential and membrane traffic during K^+ permeabilization.

Using the red cell model

The flowchart of figure 1 illustrates the approach applied for the design of the user interface aiming to optimize comparisons between predicted and experimental results, a design guided by the principle that modelling usefulness depends on a close proximity between model and bench. Use of the model will be illustrated in this guide by applying it to an analysis of the “Gardos effect”, trying to harmonize scientific interest with tutorial simplicity.

Open access to the red cell model. The executable file of the model, RCM*.jar, is available for downloading from the same GitHub repository where this guide is held: <https://github.com/sdrogers/redcellmodeljava>. The model operates as a RCM*.jar programme within the JAVA environment which needs to be preinstalled. It is recommended not to alter the original file name as it contains coded information on date and update status. Altered names are best applied to shortcuts. Double-click to activate the programme. As explained in the tutorial Boxes below, operation of the RCM model generates protocol files as editable *.txt files, and files containing the results of simulations

in spreadsheet format as *.csv files, the columns listing the changes in all the system variables as a function of time. By default, these are saved within the same directory as the one from which the programme is operated, a choice easily modified from within the programme.

Using the “Gardos effect” to introduce the red cell model (RCM). Besides its didactic and historic value, the example chosen is of substantial current interest for RBC physiology and pathology. It is based on a puzzling original finding by Gardos [43-45], the “Gardos effect”, a rapidly developing dehydration of red blood cells incubated in plasma-like media following the joint addition of a metabolic inhibitor and a metabolic substrate. Years later, the mystery surrounding the sequential steps involved in the Gardos effect was elucidated by showing that the inhibitor-substrate combinations (iodoacetamide with glucose or inosine, for instance) caused accelerated and profound ATP depletion, the substrate consuming ATP during the initial steps of glycolytic metabolism while downstream ATP production remained blocked by the inhibitor [46-48]. ATP depletion, in turn, reduced Ca^{2+} extrusion through the plasma membrane calcium pump (PMCA) with gradual increases in the concentration of Ca^{2+} in the cytoplasm, $[\text{Ca}^{2+}]_i$ [49, 50]. Elevated $[\text{Ca}^{2+}]_i$ activated a Ca^{2+} -sensitive K^+ -selective permeability pathway leading to rapid KCl loss and cell dehydration. Gardos’ landmark discovery of this permeability pathway was eventually recognized as the first report of a calcium-sensitive K^+ channel, (KCa3.1, KCNN4 gene), baptized as the Gardos channel in the RBC lore [46, 48, 51].

When the Gardos effect was originally simulated [20] the model predicted not only the expected and documented effects of KCl loss and cell dehydration but also some totally unexpected and counterintuitive outcomes, soon experimentally verified [21]. Eventually some of these “side-effects” proved crucial for unravelling the fast track dehydration mechanism of a subpopulation of sickle cells, the irreversible sickle cells [52, 53], whose transient hyperdense state causes widespread vasoocclusion and is responsible for most of the clinical symptoms in sickle cell disease [9]. The Gardos channels were considered important participant in the process of progressive RBC densification during circulatory senescence, and channel mutations have recently also been implicated in congenital haemolytic anaemias with altered RBC hydration states [40, 54, 55]. It is the wide scope of the seemingly unrelated homeostatic RBC changes triggered by Gardos channel activation which makes the exploration of the Gardos effect such an excellent didactic choice for introducing the user to the operation of the RCM and for illustrating how the model can be used for an in-depth scientific exploration of the entangled mechanisms behind global cell responses.

Following the sequence outlined in Fig 1, we start by formulating a question and by designing an experimental protocol expected to provide the answer. We simulate that protocol with the model, run the model, report outputs, analyse the results and try to interpret the mechanisms behind predicted responses. Following past experience, the analytical stage often exposes flaws or improvements required in the original protocol leading to further simulations to unravel the complexities of a certain process, in turn guiding the improvements needed in the design of the original experimental protocol, a path illustrated here in the tutorial.

Figure 1. Flowchart illustrating the approach applied in guiding the design of the red blood cell model and the sequential steps involved in running the model.

The Red Blood Cell model (RCM)

The RCM. A mathematical representation of our understanding of the homeostasis of human RBCs. The user interface was designed to optimize comparisons between predicted and experimental results.

RCM basics. The modelled system consists of a suspension of identical red cells treated as a closed, two compartment system, approximating the most frequent experimental condition used in studies on red blood cell homeostasis, but also allowing exploration of RBC responses in many other conditions. The default values of parameters and variables define a cell representing a "mean" of the RBC population in a blood sample from a healthy human adult, suspended in a plasma-like saline.

Formulating the question. We run the RCM to answer specific questions concerned with RBC responses in physiological, pathological or experimental conditions, expecting enlightening insights on mechanisms, and predictions amenable to experimental scrutiny. The first step then is to formulate the question.

Designing the protocol to answer the question. To next step is to design an experimental protocol that may provide the answer. Using the model, the user then generates a protocol file that lists the sequential changes in model parameters and variables that emulate the steps in the experimental protocol. This protocol file is used as the input for running the model

RS: Reference State. Defines the initial steady-state condition of the RBC whose dynamic responses to the experimental perturbations will be investigated. The program automatically computes a new steady state for the modified cell when default values are changed.

DS: Dynamic State. Running the model implements the protocol instructions listed within successive Dynamic States in the protocol file.

Results. Model outputs are offered in two stages: a preliminary graphic stage in which the user can scrutinize the behaviour of any system variable or group of variables seeking to correct, improve and optimize data-output frequency, precision and quality by editing the protocol and re-running the model.

Analysis. When outputs reach satisfactory quality, the complete model output can be saved in a detailed and comprehensive csv-file format containing the time-dependent changes of all the system variables for further in-depth analysis and presentation of the results

The technical aspects concerned with model operation will be displayed in four separate consecutively numbered text boxes containing detailed instructions on how to construct protocols and run the simulations. Box-separated material will allow readers primarily interested in the science to follow the main text unhindered by details and tutorials related to model use.

The protocol-prompting question: With RBCs suspended in plasma-like media at 37°C, what are the effects of sudden and simultaneous inhibition of the Na/K and calcium pumps? With this formulation we bypass the inhibitor-substrate stage of the original Gardos effect and focus on the downstream effects of pump inhibition by ATP depletion. There is also a simple experimental correlate for such a simulation. Vanadate, a well known irreversible inhibitor of P-type ATPases, the family the Na/K and PMCA pumps belong to, instantly blocks pump-mediated transport when added to a RBC suspension, without affecting RBC metabolism [56, 57]. We outline first the vanadate version of the experimental protocol and follow up with the simulated correlate in the model.

Outline of experimental protocol: We start with fresh, washed RBCs suspended in a plasma-like saline at a 10% cell volume fraction (0.1 CVF or 10% haematocrit, Hct). The suspension is kept at 37 °C under constant magnetic stirring to prevent cell sedimentation and the formation of significant diffusional gradients. After about 30 min to allow for minor adjustments to the new conditions, we add vanadate to instantly block ion transport through the Na/K and calcium pumps, and follow the evolution in time of system variables, the vanadate version of the Gardos Effect protocol.

Outline of simulated protocol: In the simulations we bypass the usual preparatory steps from drawing blood to washing the RBCs free of plasma, white cells and platelets. We start with a 0.1 cell volume fraction (CVF, equivalent to a 10% Hct) in a plasma-like, isosmotic HEPES-Na-buffered saline solution, and probe system stability while incubated at 37°C over a control period. Vanadate addition is represented by sudden 100% inhibition of Na/K and PMCA pumps, and the model follows the evolution in time of all the variables in the system. Because the cell-medium system is treated as a closed, two-compartment system, mass conservation applies, and redistribution equations compute the composition changes in each compartment at each instant of time. Instant, gradient-free uniformity in the composition of cell and medium compartments is assumed throughout representing the approximation intended by magnetic stirring in the experimental protocol. Simulated protocols are stored in text files, and model outputs reporting the changes in all the homeostasis variables of the system as a function of time are saved as comma separated (csv) spreadsheets.

The instructions for entering protocol simulations are contained in two separate sections in the model: the Reference State (RS) and the Dynamic State (DS). The RS defines the initial constitutive condition of the RBC under investigation. In the DS section the user enters the protocol prescribed sequence of perturbations to parameters and variables in consecutive DS pages, automatically numbered by the programme. All user entries in RS and DS modify default-suggested values. Text windows within each of the RS and DS pages provide contextual guiding information. Separate RS and DS Help pages provide additional information for users to consult.

In Boxes 1-4 next, the reader is guided step by step on how to simulate the vanadate version of the Gardos effect protocol using the RCM*.jar executable. Although the tutorial remains focused on the chosen example, it is hoped the user will note, along the way, the vast range of experimental simulations the software enables within a rather austere display of choices. The model output, stored in csv format, contains the evolution in time of all the homeostatic variables considered in the model during the successive protocol-defined stages for subsequent plotting and analysis. In the Results and Discussion sections we illustrate how this data can be used to report the results and to explore the mechanisms behind each of the predicted responses.

Box 1: The Reference State. Figure 2 shows the first dialog box that comes up when we activate RCM.jar. It contains a Welcome message and two optional prompts at the bottom. Select New Experiment, as there are no yet stored protocols on file. This brings up the central page for simulating experiments (Fig 3). This page contains two

main panels and five tags at the bottom. The left panel, the Reference State (RS), defines the initial constitutive condition of the RBCs at the start of experiments. On the right panel, the Dynamic State (DS), the user specifies the duration of the experiment, the frequency and precision of the data points with which results will be reported, and the changes in the value of selected parameters and variables meant to represent the successive perturbations the RBCs are exposed to in the course of experiments. Press the HELP tag at the bottom for a brief guide to all entries in the RS and DS panels, and to those in the dedicated DS PIEZO tag. The default values assigned to each of the parameters and variables in the Reference State and in the Dynamic State, listed within each of the tags, represent rounded *mean values* obtained from the literature for RBC samples from healthy adults (see Appendix for glossaries). It is important to note that when using the RS default values, model predictions will be reporting the responses of this particular *mean* representative cell. With a mean circulatory lifespan of about 120 days, real RBC samples are a mix of about 120-day cohorts. Because many constitutive variables change with RBC age it is important to bear in mind modifying the RS default values when exploring responses associated with RBC age diversity. For the Gardos effect simulation we will accept the default Reference State values of the “mean” RBC and proceed to simulate the successive experimental steps within the Dynamic State right panel (Fig 2).

Figure 2. The Welcome Page. The figure contains a simplified representation of the main components of the RCM, including PIEZO1. At the bottom of the page the user is offered two tags with alternative options to create a new simulated protocol, “New experiment” (on the left), or to transfer a previously saved protocol file “Load from file” (on the right).

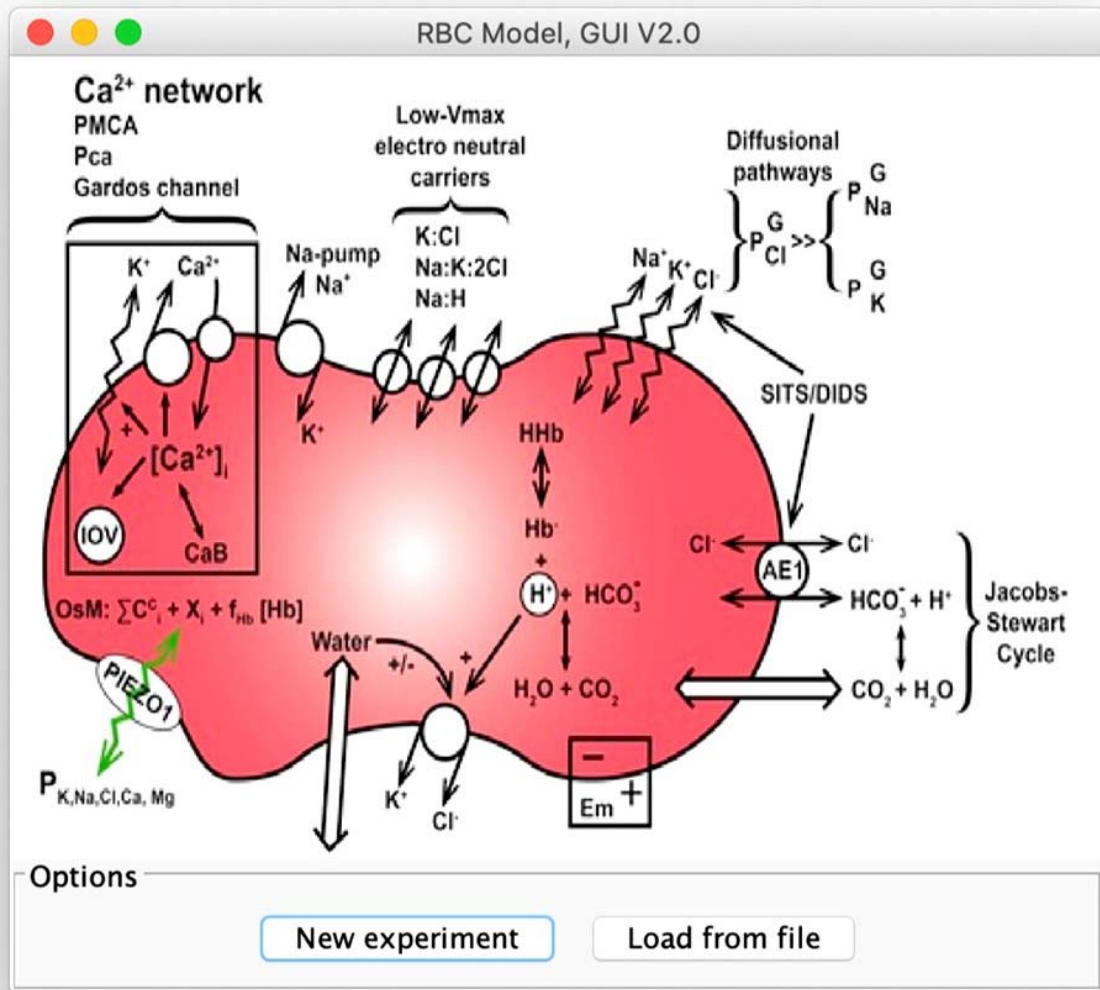
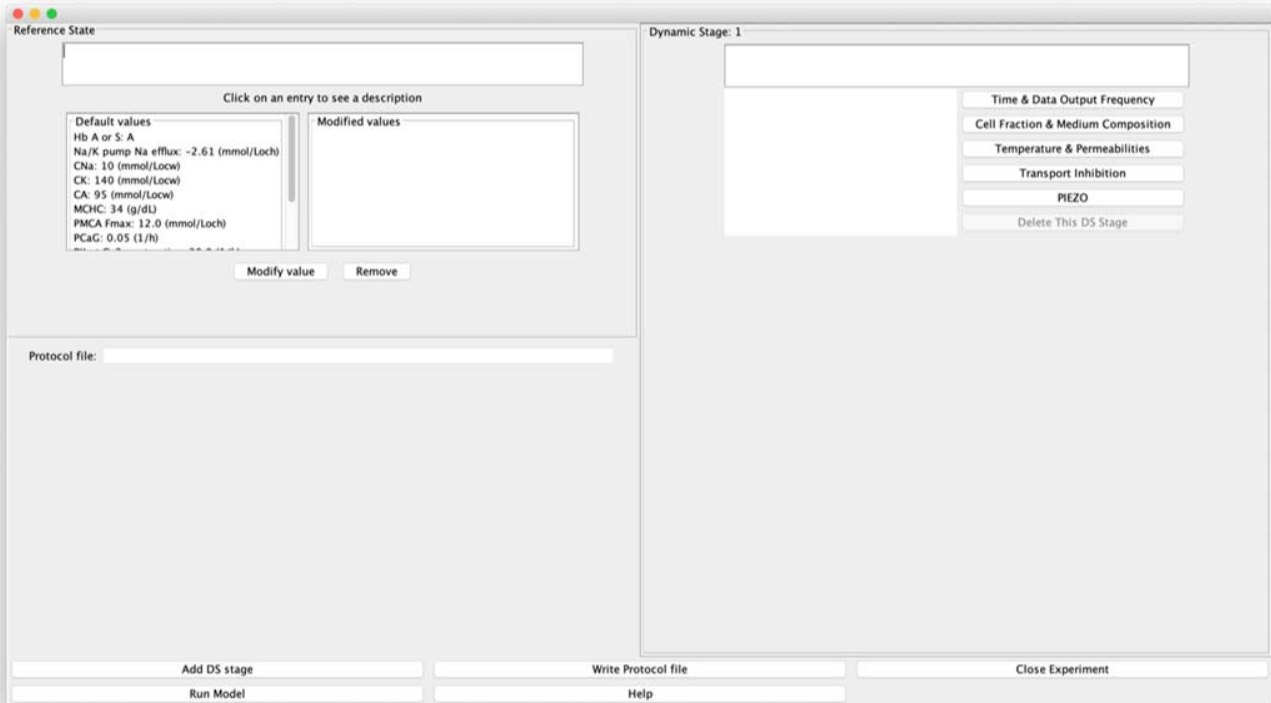


Figure 3. The Central Page (reduced version), with options for changing the initial (default) constitutive properties of the cell in the Reference State (RS, left panel), and to implement sequential Dynamic States simulating experimental steps (DS, right panel). Detailed operation is explained in boxes 1-4 in the text using an example of a simulated experiment.



Box 2: The Dynamic States. The outlined protocol requires implementation in two DS stages, DS1 and DS2, the first to allow the system to settle into whichever new conditions are set and provide baseline values of variables before the next experimental stage, the second to simulate the Vanadate-version of the Gardos effect with sudden inhibition of the Na/K and calcium pumps. Start the DS1 stage by selecting the “Time & Data output frequency” tab on the right. This opens up a full dialog box on the left. You select “time” and then “Modify value”, or simply double click on “time” to open the “Input” window where you enter the new or modified value; let us use 30 min in this instance. Note how the “time 30” instruction becomes registered on the left-side window of the DS1 panel. Errors are easily corrected by repeating these steps with amended entries. Now select the “Cell fraction & medium composition” tab. Operating the prompts in the way instructed above for “time” proceed to enter 0.1 for the cell volume fraction (CVF) to represent a 10% haematocrit. Leave all other variables unchanged, as the default medium composition approximates widely used conditions for “plasma-like”, protein-free media used in experiments with RBCs. Note how the changes become incorporated in the protocol window of DS1.

Next click on “Add DS stage” to add a DS2 stage. Set time to 60 min and enter 100 in the Transport Inhibition tab for 100% inhibition of both the PMCA and the Na/K pump. The cell fraction remains as set for DS1. *Instructions are automatically carried over sequential DS stages unless specifically deleted or changed.*

Box 3. Running the model, inspecting preliminary results, and considering the desirability of protocol improvements. We are now ready to run the simulation. Good practise recommends saving the protocol at this stage, for immediate availability if changes become necessary. Press the “Write a protocol file” tag at the bottom of the central page (Fig 3). Choose a directory and filename (GE1 for instance). The protocol is saved as a text-file with name GE1.txt, editable in any text editor. Figure 4A shows what the text-file looks like at this stage. Although it is possible for the user to modify entries in the text file, we envisage the file mainly being used for archiving and allowing protocols to be saved and reloaded later. To minimise the chance of user error, we recommend all protocol changes to be made from within the central page of the user interface.

Select “Run Model” at the bottom of the central page. This opens a window on the left overlapping the central page and showing the progress of the model computations, almost instant for this protocol. At the bottom, four commands offer data saving, window closing, stopping model run, and data plotting options. The “Stop” command

halts the model run; useful when exploring protocol conditions to optimize the compromise between data output precision and overall duration of run. The data plotting option is particularly useful for a preliminary exploration of results before saving data, to enable protocol adjustments on the go before reaching worth-saving outputs. Let us illustrate this by reviewing the data outputs from the current protocol.

Press the Launch plotter command. This opens a small window with a comprehensive list of all the model variables. Our interest is directed primarily at the variables most influenced by PMCA inhibition and Gardos channel activation: cell volume, membrane potential, potassium flux and cell calcium. Press the relative volume variable, V/V . This brings up a basic graph of V/V as a function time for the set duration of the experiment. Scaling of the y-axis automatically adapts to the magnitude of change in the selected variable. Within the first hour after pump inhibition, we note a few points outlining a tiny increase in cell volume followed by an accelerating decrease. But with such poor quality of data it is impossible to derive any significant information. It is clear from this graph that we need a much higher frequency of data output points and a much more extended time-scale to explore properly the predicted effects.

Figure 4. Alternative protocols. Samples of protocol text files on the K^+ -permeabilization theme.

Users are advised to save result files (csv) for each of these protocols to practise interpreting the mechanisms responsible for the different responses, with particular attention to the following variables: RCV, Em, pH_i, rA-rH, COs-MOs, CK, CNa, CA. **A** and **B**: show the original and corrected protocol versions outlined in boxes 3 and 4 of the tutorial, respectively. **A** and **B** are the only protocols explicitly analysed in the text. **C: Effect of valinomycin.** Addition of the K^+ -selective ionophore valinomycin bypasses all calcium-mediated effects. This basic protocol is repeated with variations in all the following examples (D to G). Note the rapid onset of hyperpolarization and dehydration caused by sudden K^+ permeabilization, relative to their slower onset rates under the vanadate-simulated Gardos effect (shown in Figs 5A and 5C). **D: Increased anion permeability (PA in DS2).** Comparison of dehydration rates between protocols **C** and **D** illustrates the powerful rate-limiting effect of the anion permeability on RBC dehydration rate; the electrodiffusional anion permeability can be increased experimentally by replacing 10 mM NaCl with 10 mM NaSCN in the medium [58]. **E: Iso-osmotic replacement of 85 mM NaCl for KCl in the medium.** Elevating the medium potassium concentration effectively reduces the driving force for dehydration and for all associated secondary effects. **F: Effects of AE1 inhibitors.** Dehydration remains rapid and profound, but the secondary JS-mediated pH effects are retarded, with rH approaching rA slowly instead of almost instantly as for the un-inhibited AE1 exchanger in the Gardos effect simulations (Fig 8D). **G: Osmotic effects of RBC dehydration explored at high haematocrits.** The cell fraction is set to 0.4 (40% haematocrit) to illustrate how the osmotic effects of haemoglobin crowding become more marked the higher the cell fraction and overall haemoglobin content of the suspension. Compare the COs and MOs variables in conditions **C** (with 0.1 cell volume fraction) and **G**.

A RS DS 1 Cycles per print:11 CVF:0.1 Time:30 DS 2 Time:60 % inhibition of Na/K pump FNamax:100 % inhibition/stimulation(-) of PMCA FCamax:100	B RS DS 1 Cycles per print:11 CVF:0.1 Time:30 DS 2 Time:600 % inhibition of Na/K pump FNamax:100 % inhibition/stimulation(-) of PMCA FCamax:100	C RS DS 1 Cycles per print:11 CVF:0.1 Time:30 DS 2 Time:120 PK:30	
D RS DS 1 Cycles per print:11 CVF:0.1 Time:30 DS 2 PA:50 Time:120 PK:30	E RS DS 1 Cycles per print:11 CVF:0.1 Time:30 DS 2 Replace NaCl with KCl:85 Time:120 PK:30	F RS DS 1 Cycles per print:11 CVF:0.1 Time:30 DS 2 Time:120 PK:30 % inhibition/stimulation(-) of JS mediated fluxes:95	G RS DS 1 Cycles per print:11 CVF:0.4 Time:30 DS 2 Time:120 PK:30

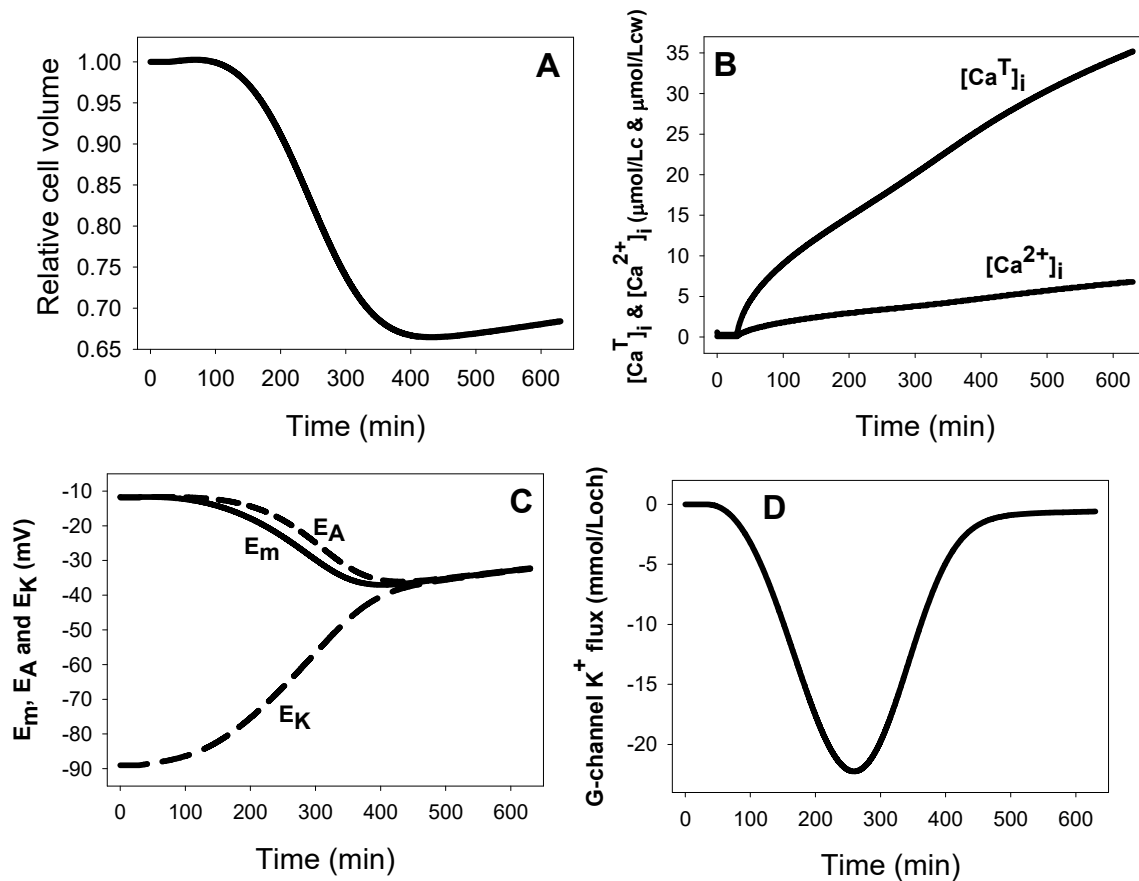
Box 4. Correcting the protocol using graphic output feedback and saving improved final model output. Let us return to the original protocol to increase data output frequency and experiment duration. Close any open plot windows, and the window showing model progress. In DS1 press the “Time & Data output frequency” tab and change the “cyclesperprint” from its default value of 777 to a lower value, 11 in our example (thus forcing the model to output values at a much higher frequency). In DS2 change “time 60” to “time 600” to explore effects over a longer period. Save the modified protocol overwriting the original one using the same name (Fig 4B). Now run the model again. Note that this time the model run takes longer in line with the increased density of output data and duration of the experiment. Launch the Plotter and explore the time changes of the variables listed as of primary interest: V/V, QCa, CCa2+, Em, and FKardos. Note the smooth appearance of the curves with the new setting of “cyclesperprint” at 11.

The increased data-sampling frequency and duration of the experiment seem to adequately cover now the period of major changes and deliver smooth data outputs, rendering outputs well-suited for an in-depth analysis of the full RBC response to the vanadate-simulated Gardos effect. To save the full complement of data generated by the model press the Save Output tab and at the filename prompt enter GE1. This generates a file named GE1.csv open to scrutiny and graphing in different software platforms. We will now generate figures from this file. These should guide the user on how to report the results, and on how to analyse and interpret the data in search of the mechanisms behind the observed effects.

Results and Analysis

Predicted time-course of changes in RBC volume, membrane potentials, calcium contents and Gardos channel mediated K^+ flux following the vanadate-protocol simulation of the Gardos effect. Following pump-inhibition, the change in RCV (Fig 5A) shows three distinct phases, a slow initial increase, a sharp decline followed by a shallow slow recovery. The initial and late slow rates of volume increase result from Na/K pump inhibition, as may be shown by re-running the vanadate protocol without Na/K pump inhibition. The reason for such slow rates of volume change and cation gradient dissipation is the extremely low permeability of the RBC plasma membrane to Na^+ , K^+ , and to cations in general, an evolutionary optimized condition allowing RBC volume maintenance with minimal pump-leak turnover rates and energy consumption [6].

Figure 5: Predicted changes in selected RBC variables as a function of time following sudden inhibition of the Na/K and calcium pumps by vanadate. **A:** relative cell volume changes due to Na/K pump inhibition and to delayed activation of Gardos channels by slow $[Ca^{2+}]_i$ gain; **B:** total ($[Ca_T]_i$) and free ($[Ca^{2+}]_i$) intracellular calcium concentration changes following PMCA inhibition; **C:** Changes in membrane potential, E_m , and in the equilibrium potentials of K^+ and A^- , E_K and E_A , respectively, with time-patterns determined by Gardos channel activation kinetics and by ion gradient dissipation; **D:** biphasic net K^+ flux pattern determined by changes in Gardos channel permeability (increasing K^+ efflux phase) and K^+ gradient dissipation (decreasing K^+ efflux phase), with minimal initial and late contributions of Na/K pump inhibition, hardly noticeable on this y-axis scale.



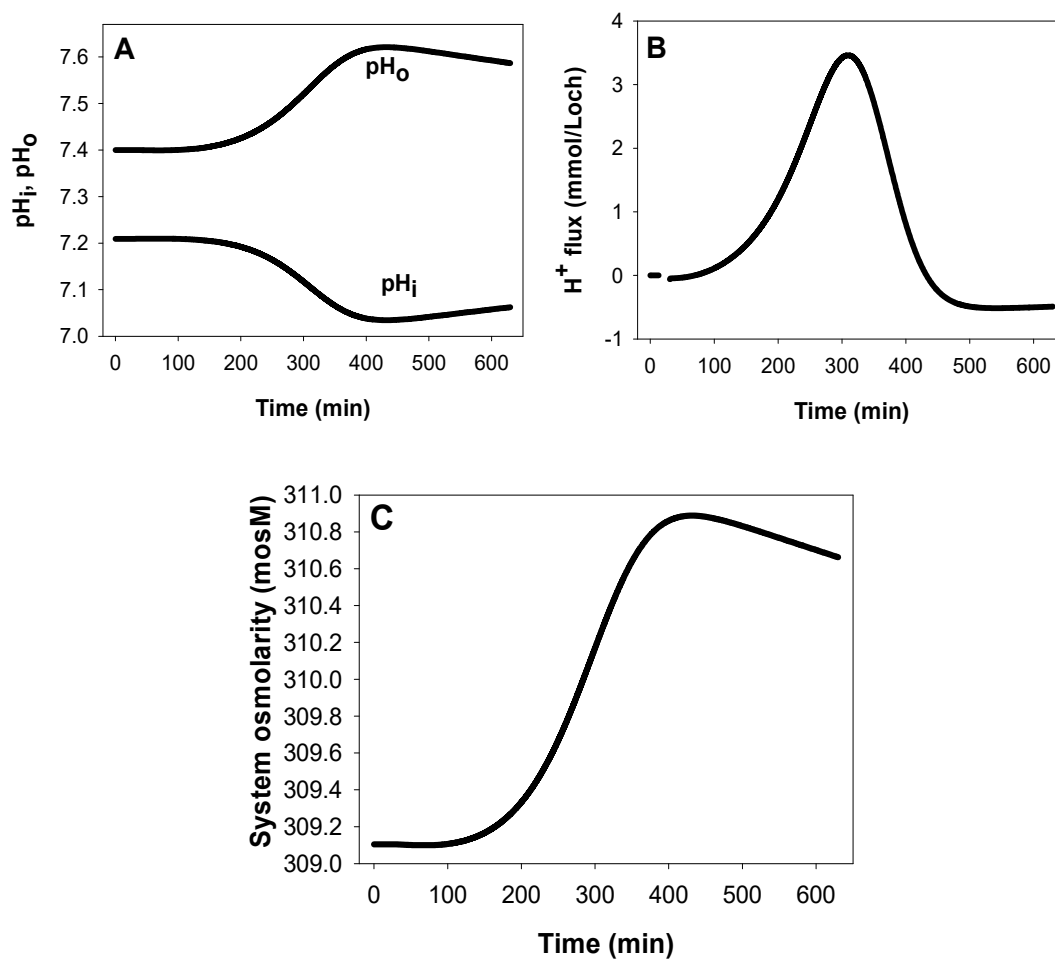
The sharp volume fall during the second phase results from Gardos channel activation (Fig 5D). The delay is caused by the slow increase in $[Ca_T]_i$ and $[Ca^{2+}]_i$ following PMCA inhibition (Fig 5B). The low Ca^{2+} permeability of the RBC membrane limits the rate at which $[Ca_T]_i$ increases, and cytoplasmic Ca^{2+} buffering further reduces the scale of the $[Ca^{2+}]_i$ increase which is the direct signal for Gardos channel activation. Figures 5C and 5D show the time-course associated with hyperpolarization and channel-mediated K^+ flux, respectively. It can be seen (Fig 5C) that the membrane potential follows closely the anion equilibrium potential throughout all the periods, reflecting the dominance of anion conductance in the RBC membrane [59-61]. The E_m - E_A - E_K convergence in the end (Fig 5C) reflects the new ion-gradient distribution generated after full dissipation of the K^+ gradient. The K^+ efflux curve through the Gardos channel (Fig 5D) shows a biphasic pattern, first a Ca^{2+} -activated increase followed by a decline as the K^+ concentration gradient dissipates.

Interested readers may wish to explore further and compare the Gardos effect response described here with that elicited by sudden K^+ permeabilization of RBCs. Sudden K^+ permeabilization can be elicited

experimentally by addition of valinomycin [21, 62], a potassium-selective ionophore, as simulated with the protocols of Figs 4C to 4G, or by addition of a Ca^{2+} ionophore [63-65]. Note the sharp differences in the rate and magnitude of changes in membrane potential and other variables between different K^+ -permeabilization protocols.

Predicted time-course of changes in RBC pH and system osmolarity following K^+ permeabilization. A first set of predictions, totally unsuspected when originally advanced [20], concerned changes in cell and medium pH, with cell acidification and medium alkalisation (Fig 6A), predictions experimentally validated in a variety of experimental conditions [21, 31]. Their mechanism is explained in the Discussion. The biphasic time-course predicted for the net H^+ flux in this protocol is shown in Fig 6B. Further scrutiny of csv-file data shows that net K^+ flux exceeds net Cl^- efflux, and that flux electroneutrality is made up by H^+ influx, with the appearance of an electroneutral $\text{K}^+:\text{H}^+$ exchange, misattributed to a real electroneutral $\text{K}^+:\text{H}^+$ antiport in the past [66-68]. Its phantom nature was exposed by the original model which, lacking this transporter, predicted most of the observed experimental results used in its support and explained the complex mechanisms behind its appearance [20]. The non-existing $\text{K}^+:\text{H}^+$ antiport carries instructive cautionary advice on applying a strict minimalistic approach when building up model components beyond the essential basic components.

Figure 6. Predicted changes in pH and system osmolarity following the vanadate protocol. A: Changes in cell and medium pH (pH_i and pH_o , respectively) with patterns determined by the changes in Cl^- concentration ratio, r_A , across the red cell membrane; **B:** biphasic H^+ flux pattern through the JS mechanism driven by the changes in r_A ; **C:** changes in system osmolarity resulting from variations in haemoglobin crowding secondary to cell volume changes.



A second set of unexpected predictions concerned system osmolarity (Fig 6C). The model predicts that cell dehydration driven by net KCl loss elevates cell and medium osmolarity as if new osmotic particles were generated within the cells and redistributed between cells and medium. Moreover, the model predicts that the increase in global system osmolarity will be higher the higher the cell fraction, an effect the reader is invited to explore further (Fig 4G).

Discussion

We introduced a multiplatform JAVA-based model of human red blood cell homeostasis and applied it here to explore in detail the homeostatic changes induced by activation of Gardos channels while guiding users on how to operate the model.

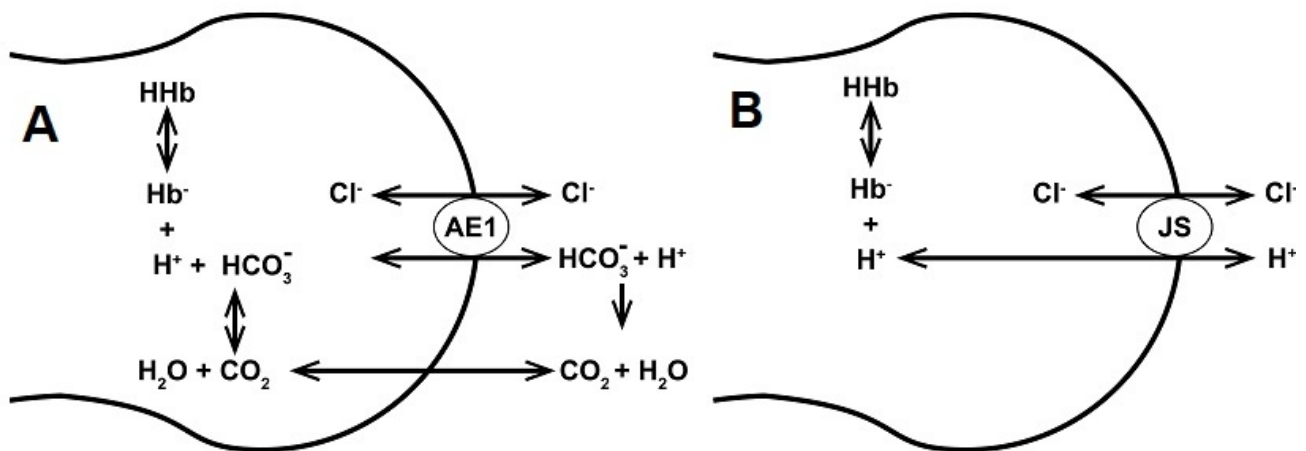
The deceptive simplicity of the central page (Fig 3) for determining the constitutive properties of the cell in the initial Reference State, and for the design and creation of simulated experimental protocols in the Dynamic State hides an unlimited versatility in the range and scope of questions that can be addressed with the model. Availability of the continually updated code in the online GitHub repository also allows users to extend the model for specific uses not included in current versions. We hope the accessibility and simplicity of RCM*.jar will encourage its use as a frequently consulted tool on questions related to RBC volume, transport and homeostasis in health and disease.

We discuss next the mechanisms concerned with pH and osmolarity changes associated with K^+ permeabilization, good examples of entanglement confounding intuition.

Predicted side-effects of K^+ permeabilization of human RBCs

Let us consider first how the operation of what has become known as the Jacobs-Stewart cycle or mechanism [69, 70] is represented in the model. The JS cycle comprises the parallel operation of the anion exchanger, AE1, as an electroneutral $Cl^-:HCO_3^-$ antiport, and the CO_2 shunt, as illustrated in Fig7A.

Figure 7. Representation of the JS cycle as an electroneutral $Cl^-:H^+$ symport. **A:** The Jacobs-Stewart mechanism showing the combined operation of the anion exchanger, AE1, and the CO_2 shunt on the transfer of H^+ ions across the red cell membrane; **B:** equivalent model-representation of H^+ transport through the JS mechanism as an electroneutral $Cl^-:H^+$ cotransport.



As a rough approximation to a complex multistep process, we may assume that for each $Cl^-:HCO_3^-$ exchanged through AE1, protonation of the transported HCO_3^- ion effectively adds a CO_2 molecule to the

medium on the membrane side it was transported to. Rapid CO₂ equilibration across the membrane shunts back the CO₂ to the other membrane side releasing a H⁺ on hydration. The net result of each half JS cycle is therefore the electroneutral transfer of one Cl⁻ with one H⁺, indistinguishable from the operation of an electroneutral Cl⁻:H⁺ symport. Thus, the complete JS cycle may be represented phenomenologically by an electroneutral Cl⁻:H⁺ cotransporter, as illustrated in Fig 7B, ignoring the background roles of HCO₃⁻ and CO₂. This phenomenology lumps a large number of steps operating over a vast range of time-scales comprising CO₂ solvation, CO₂ hydration (spontaneous in the extracellular medium and carbonic anhydrase-assisted intracellularly), CO₂ transport, intracellular and extracellular proton buffering, and the complex kinetics of the AE1 exchanger. The Cl⁻:H⁺ cotransport representation has been shown to accurately describe the operation of the JS cycle at the functional level required for modelling RBC homeostasis. It allows the model to predict pH changes and explain the mechanisms behind all relevant physiological, pathological and experimental RBC responses involving the participation of the JS cycle [12, 21, 23, 28, 29, 31, 71-73]. An additional important issue addressed by this phenomenology is that of the “ignored” HCO₃⁻ and CO₂ (see Box 5). This oversight, together with the JS cycle phenomenology, also enables the model to be run using a single diffusible anion representation, A⁻, for the combined contents or concentrations of Cl⁻ and HCO₃⁻.

Box 5. The ignored HCO₃⁻ and CO₂. In most of the experiments with RBC suspensions HCO₃⁻ salts are not added to the medium and both HCO₃⁻ and CO₂ remain unreported as explicit components of the system. Where then is the HCO₃⁻ that keeps the Cl⁻:HCO₃⁻ exchange in motion and allows the Cl⁻:H⁺ symport phenomenology to represent the operation of the JS system? Dissolved HCO₃⁻ at equilibrium with atmospheric CO₂ is always present in unsealed cell suspensions, and this presence is sufficient to sustain near physiological rates of JS cycle turnover. It can easily be estimated that at an atmospheric pressure of 760 mmHg, a ~ 0.04% CO₂ in environmental air, a solubility constant of CO₂ in water of 0.03mM/mmHg, a pKa of 6.147 for the [H⁺][HCO₃⁻]/[CO₂] hydration reaction, and at a medium pH of 7.4-7.5, the unreported HCO₃⁻ concentration will be around 200 μM, more than sufficient to sustain high rates of JS-mediated fluxes.

With over 10⁶ AE1 units per cell and essentially instant CO₂ equilibration rates across the RBC plasma membrane, the JS mechanism, optimized by evolution for effective CO₂ transport from tissues to lungs, is by far the most important mediator of H⁺ transport in RBCs. The JS phenomenology is represented in the model by a low-saturation equation of the form $F_{\text{ClJS}} = F_{\text{HJS}} = k_{\text{AE1}}([Cl^-]_o[H^+]_o - [Cl^-]_i[H^+]_i)$, where F_{ClJS} and F_{HJS} are the net fluxes of Cl⁻ and H⁺ through the JS pathway, and k_{AE1} is the rate constant of AE1 turnover, about four to six orders of magnitude faster than that of any other ion transporter in the RBC membrane (default value in the RCM: $k_{\text{AE1}} = k_{\text{HA}} = 2.5 \times 10^8 \text{ h}^{-1}$). In the absence of net JS fluxes, $F_{\text{JS}} = 0$, $[Cl^-]_o[H^+]_o = [Cl^-]_i[H^+]_i$. Defining the concentration ratios r_A and r_H by $r_A = [Cl^-]_o/[Cl^-]_i$ and $r_H = [H^+]_i/[H^+]_o$, respectively, the equilibrium condition for the JS system at zero net flux will be $r_A = r_H$. The participation of other anion and proton transporters causes deviations from this equality, miniscule in normal steady states. Larger deviations during dynamic states would tend to be speedily restored towards $r_H \approx r_A$ by the JS system.

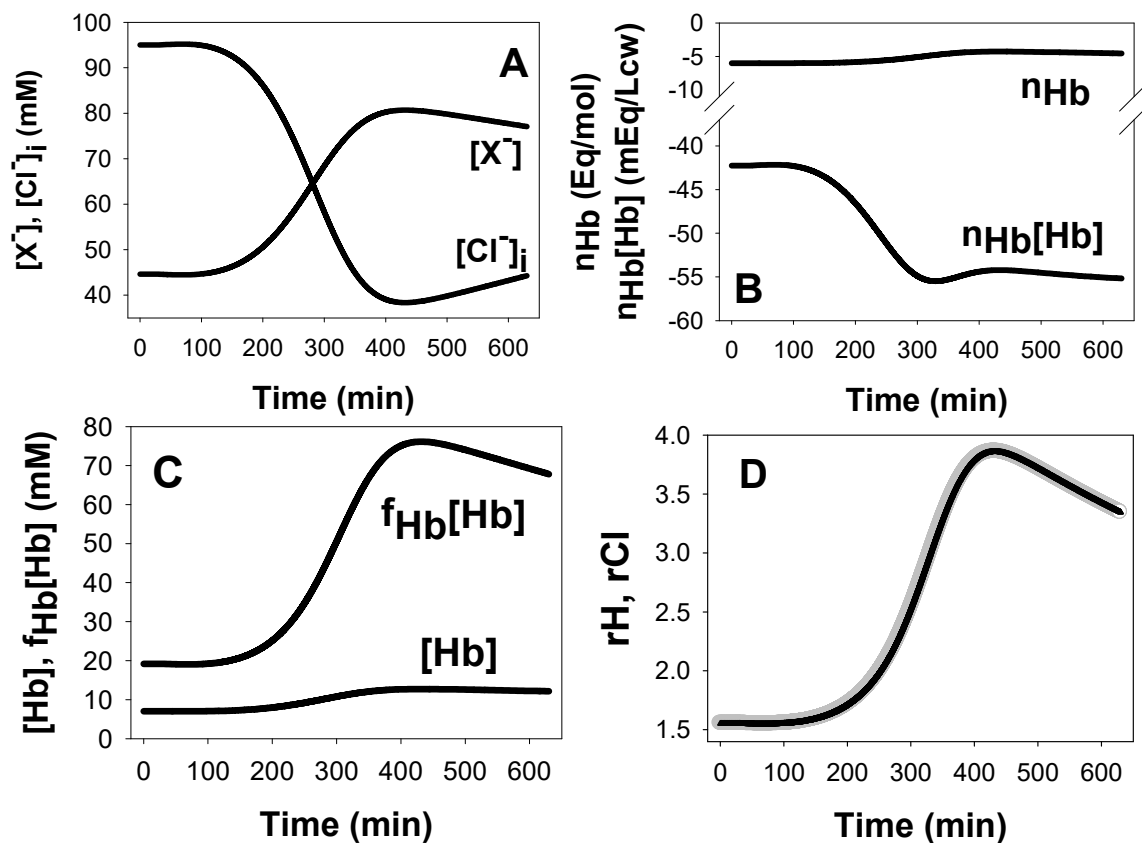
Let us now consider how RBC dehydration caused by K⁺ permeabilization affects cell pH and osmolarity from the data stored in GE1.csv, the output file generated from the vanadate protocol simulations (**Box 4**). As the cells start to dehydrate by the net loss of KCl, four confluent changes among many others are particularly relevant for the pH and osmolarity effects: i) the concentration of the impermeant intracellular anion increases (Fig 8A, [X⁻]); ii) the concentration of haemoglobin increases (Fig 8B, [Hb]) and so does its contribution to the net negative intracellular charge (Fig 8B, $n_{\text{Hb}}[\text{Hb}]$); iii) the osmotic coefficient of haemoglobin increases along a power function of the haemoglobin concentration increasing sharply the osmotic contribution of haemoglobin as dehydration proceeds (Fig 8C), and iv), the intracellular concentration of Cl⁻ becomes markedly reduced (Fig 8A, [Cl⁻]) by net Cl⁻ loss, by charge displacement from increasingly concentrated impermeant anions X⁻ and Hb⁻, and by water retention due to the increasing colloid osmotic pressure on haemoglobin.

The reduction in [Cl⁻] is the most relevant drive for the pH changes. The resulting increase in r_A will tend to be rapidly equilibrated towards $r_H \approx r_A$ through the JS mechanism (Fig 8D), explaining the pH changes shown in Fig 6A. Although cell acidification is buffered by haemoglobin with reduction in its net negative

charge per mol, n_{Hb} (Fig 8B, n_{Hb}), the increase in haemoglobin concentration dominates, and its overall contribution to the intracellular negative charge increases with dehydration (Fig 8B, $n_{\text{Hb}}[\text{Hb}]$). Electroneutrality is strictly preserved through all these changes, ($[\text{Na}^+]_i + [\text{K}^+]_i + 2[\text{Mg}^{2+}]_i = [\text{A}^-]_i + [\text{X}^-] + n_{\text{Hb}}[\text{Hb}]$), and the extra osmotic load generated by haemoglobin crowding (Fig 8C) is rapidly redistributed between cells and medium by an hypertonic KCl + KOH effluent [21]. In Box 6 we provide a highly simplified example to illustrate the dehydration-induced acidification mechanism bypassing the complexities generated by macromolecular crowding and proton buffering.

Figure 8. Predicted effects of macromolecular concentration changes on cell Cl^- , H^+ and osmolarity.

A: Inverse correlation between the changes in the concentrations of cell permeant A^- ($[\text{Cl}^-]$, $[\text{Cl}^-]_i$) and of impermeant cell anions ($[\text{X}^-]$) during the Gardos effect. **B:** opposite changes in the charge per mole of haemoglobin (n_{Hb}) and in the charge concentration on haemoglobin ($n_{\text{Hb}}[\text{Hb}]$) during the Gardos effect. **C:** illustrates the powerful effect of the osmotic coefficient of haemoglobin (f_{Hb}) in elevating the osmotic contribution of haemoglobin during concentrative haemoglobin crowding. **D:** overlapping changes in $r\text{H}^-$ (gray) and $r\text{H}^+$ (black) during the Gardos effect highlighting the almost instant equilibration of $r\text{H}^+$ to $r\text{A}^-$.



Box 6. Thought experiment to illustrate the mechanism by which dehydration by KCl loss acidifies cells that express the anion exchanger (AE1) in their plasma membrane. We imagine a protein-free RBC ghost resealed with 50 mM KGluconate, 100 mM KCl, and 1 mM ATP, suspended in a 150 mM NaCl medium adjusted to a pH_o of 7.4, equivalent to a $[\text{H}^+]_o$ of $\sim 40\text{nM}$. Monovalent gluconate is the only impermeant solute in this system. ATP is there to fuel the Na/K pump and balance the passive Na and K leaks to set ghosts initially in a steady state. We add valinomycin to permeabilize the ghost to potassium eliciting its dehydration by potassium gradient-driven loss of an isosmotic KCl effluent. We wait until the ghost dehydrates to half its initial volume ($V/2$) and take note of the change in its composition.

Cell concentrations (mM)	Initially	After $V/2$
Gluconate	50	100
Cl^-	100	50
K^+	150	150

With rA defined by $rA = [Cl^-]_o/[Cl^-]_i$ and rH by $rH = [H^+]_i/[H^+]_o$, the JS-AE1 mechanism sets rH to match rA at a rate determined by AE1 expression and turnover. Therefore, when $rH \approx rA$,

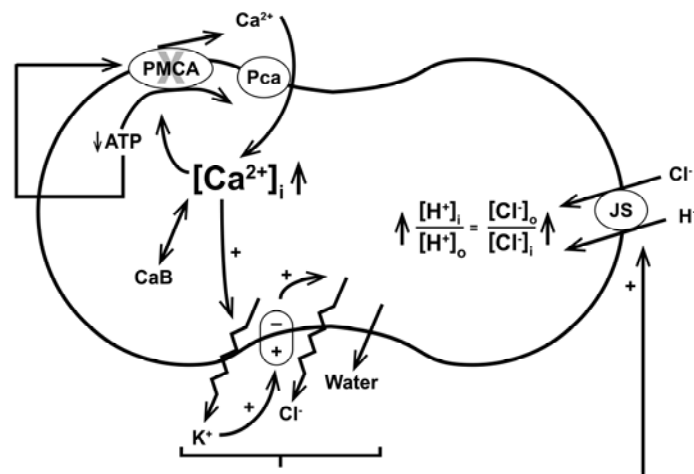
	rA	rH	$[H^+]_o$ (nM)	$[H^+]_i$ (nM)	pHo	pHi
Initially	1.5	1.5	40	60	7.4	7.2
After V/2	3.0	3.0	40	120	7.4	6.9

The dehydration-induced increase in rA led to a fall in intracellular pH from 7.2 to 6.9 in this example. Cell acidification following dehydration led by chloride salt losses has been documented in cells other than RBCs a tendency common to any cell type expressing the AE1 in its plasma membrane [68].

Haemoglobin crowding (Fig 6C) has three main side-effects: i) it contributes to increase the net negative charge on impermeant cell anions (Fig 8B), ii) it elevates cell and system osmolarity (and osmolality) as a power-function of its concentration (Figs 6C and 8C), and iii) it increases the colloidosmotic pressure within the cells causing water retention and a progressively more hypertonic effluent as part of the redistribution process of permeant ions between cells and medium (Fig 6C). It tempts speculation that this powerful osmotic effect of soluble protein crowding, so easily detectable in a RBC suspensions, may yet be discovered to participate in the fluid dynamics of cells containing compartments with a high concentration of soluble macromolecules.

We can now summarize the sequence of events during the Gardos Effect within the scope of conditions considered in this Gide (Fig 9). Running down of cell ATP by the combination of substrate and inhibitor reduces Ca^{2+} extrusion by the plasma membrane calcium pump. This allows a net build up of the intracellular concentration of free calcium to levels that activate Gardos channels (Figs 5B and 5D). The cell hyperpolarizes to potentials intermediate between those of E_K and E_{Cl} (Fig 5C) driving a net loss of KCl. As the cells dehydrate, the intracellular chloride concentration progressively declines (Fig 8A), driving a net entry of chloride and protons through the Jacobs-Stewart mechanism, with cell acidification and medium alkalinisation (Figs 6A and 8D). Concurrently, dehydration progressively increases the osmotic contribution of haemoglobin along a power function (Fig 8E), elevating its colloidosmotic strength, causing water retention and the generation of a progressively hypertonic effluent with overall increase in system osmolarity (Fig 6C), higher the higher the haematocrit of the cell suspension (Fig 4G). Water retention, in turn, contributes to intracellular chloride dilution, an additional contributor to the overall pH change via rA .

Figure 9. Illustration of the sequence of effects triggered by the Gardos effect, leading from ATP depletion to cell dehydration and acidification. Full description in the text.



APPENDIX

- I. Glossary of variable names as in the column sequence of the model output files (*.csv), with brief description and units
- II. Modelling framework flowchart
- III. User interface for building simulated experimental protocols
- IV. HELP pages: Reference State, Dynamic State and PIEZO1 routine

I. Glossary of variable names as in the column sequence of the model output files (*.csv), with brief description and units

SYMBOL or NAME	DESCRIPTION	UNITS
V/V or RCV	Relative Cell Volume	Normalized
Vw	Volume of cell water in L per litre original cells**	Lcw/Loc
CVF	Cell volume fraction in cell suspension	
Haematocrit,Hct	Hct = CVF*100	%
MCHC	Mean cell haemoglobin concentration	g/(dLpc)
Density	Set by the specific weight of Hb = 1.36 g/mL [74]	g/(mLpc)
Em	Membrane potential	mV
pHi	Intracellular pH	
pHo	Medium pH	
rH	In/out proton concentration ratio	
rA	Out/in permeant anion concentration ratio ($\text{Cl}^- + \text{HCO}_3^-$)	
QNa	Cell Na contents	mmol/(340g Hb)
QK	Cell K contents	mmol/(340g Hb)
QA	Cell permeant anion content ($\text{Cl}^- + \text{HCO}_3^-$)	mmol/(340g Hb)
QCa	Cell Ca contents	mmol/(340g Hb)
QMg	Cell Mg contents	mmol/(340g Hb)
CH/nM	Cell proton concentration	nmol/(Lcw)
CNa	Cell Na concentration	mmol/(Lcw)
CK	Cell K concentration	mmol/(Lcw)
CA	Cell permeant anion concentration ($\text{Cl}^- + \text{HCO}_3^-$)	mmol/(Lcw)
CCa2+	Cell free ionized Ca2+ concentration ($[\text{Ca}^{2+}]_i$)	mmol/(Lcw)
CMg2+	Cell free ionized Mg2+ concentration ($[\text{Mg}^{2+}]_i$)	mmol/(Lcw)
CHb	Cell Hb concentration	mmol/(Lcw)
fHb	Osmotic coefficient of Hb	Osmol/(mol Hb)
nHb	Net charge on Hb (pHi-dependent)	mEq/mol
fHb*CHb	Haemoglobin osmolarity contribution	mmol/(Lcw)
nHb*CHb	Total Hb charge concentration	mEq/(Lcw)
CX-	Non-Hb impermeant cell anion concentration	mmol/(Lcw)
nX	Net charge on X ⁻	mEq/mol
nX*CX-	Total negative charge concentration on X ⁻	mEq/(Lcw)
COs	Total cell osmolarity	mOsmol/(Lcw)
MOs	Total medium osmolarity	mOsmol/L
MNa	Medium Na concentration	mM
MK	Medium K concentration	mM
MA	Medium ($\text{Cl}^- + \text{HCO}_3^-$) concentration	mM
MH	Medium H ⁺ concentration	nM
MB	Medium proton buffer concentration	mM

MCat	Medium total calcium concentration	mM
MCa2+	Medium free calcium concentration	mM
MMgt	Medium total magnesium concentration	mM
MMg2+	Medium free magnesium concentration	mM
MEDGTA	Medium concentration of EDTA or EGTA	mM
Msucrose	Medium sucrose concentration	mM
Mgluconate-	Medium gluconate concentration	mM
Mglucamine	Medium glucamine concentration	mM
FNaP	Na/K pump net Na flux	mmol/(Loch)
FKP	Na/K pump net K flux	mmol/(Loch)
FCaP	PMCA Ca efflux	mmol/(Loch)
FHCaP	PMCA Proton influx (FHCaP = -2*FCaP)	mmol/(Loch)
FKGardos	Net K flux through Gardos channel	mmol/(Loch)
FCaG	Electrodiffusional Ca flux	mmol/(Loch)
FNaG	Electrodiffusional Na flux	mmol/(Loch)
FKG	Electrodiffusional K flux	mmol/(Loch)
FAG	Electrodiffusional anion (Cl⁻ and HCO₃⁻) flux	mmol/(Loch)
FHG	Electrodiffusional H flux (protonophores)	mmol/(Loch)
FAJS	A flux through Jacob-Stewart cycle	mmol/(Loch)
FHJS	H flux through Jacob-Stewart cycle	mmol/(Loch)
FCICo	Net Cl flux through Na:K:2Cl cotransporter	mmol/(Loch)
FKCo	Net K flux through Na:K:2Cl cotransporter	mmol/(Loch)
FNaCo	Net Na flux through Na:K:2Cl cotransporter	mmol/(Loch)
FA23Ca	Ionophore A23187-mediated net Ca flux	mmol/(Loch)
FA23Mg	Ionophore A23187-mediated net Mg flux	mmol/(Loch)
FNa	Net Na flux	mmol/(Loch)
FK	Net K flux	mmol/(Loch)
FA	Net anion flux	mmol/(Loch)
FH	Net proton flux	mmol/(Loch)
FCa	Net Ca flux	mmol/(Loch)
FW	Net water flux	mL/(Loch)
EA	Nernst equilibrium potential for permeant anions	mV
EH	Nernst equilibrium potential for proton ions	mV
EK	Nernst equilibrium potential for K	mV
ENa	Nernst equilibrium potential for Na	mV
FzK	Net K flux through PIEZO1 channel	mmol/(Loch)
FzNa	Net Na flux through PIEZO1 channel	mmol/(Loch)
FzA	Net anion flux through PIEZO1 channel	mmol/(Loch)
FzCa	Net Ca flux through PIEZO1 channel	mmol/(Loch)
EN test	Measures error margin for compliance with electroneutrality (departure from $\sum I_j = 0$)	

Notes:

Electrodiffusional fluxes (currents): the non-linear relation between ionic currents and membrane potential is represented in the model with kinetics defined by the Goldman constant field equation [75].

Non-standard units:

dLpc, mLpc: decilitre or millilitre of packed RBCs; widely used for reporting MCHC results in blood tests
Lcw: Litre cell water.

Lpc: Litre packed RBCs

Loc: Litre original cells

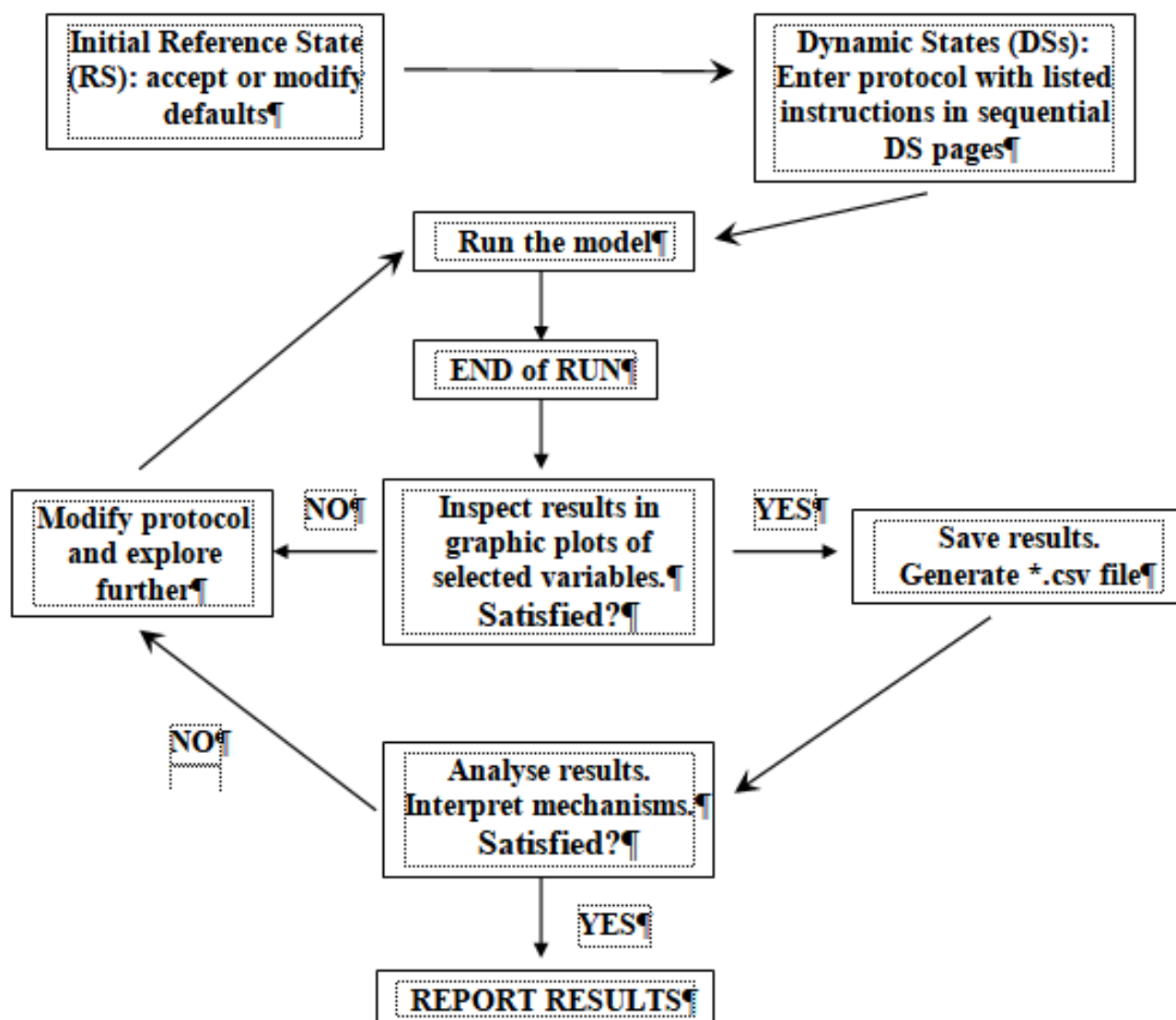
340gHb; equivalent to Loc

Loch: shorthand for reporting fluxes in mmol per Litre original cells per hour

Loc is a standardized unit widely used in the red blood cell literature for comparing concentrations and fluxes. Based on mean RBC population measurements in healthy adults, Loc assumes a content of 340 g haemoglobin to be contained in 1 Litre of packed RBCs (Lpc), a standard we adopted to represent the averaged default condition of RBC populations at the start of experiments and model simulations. Thus, for concentrations, mmol/(Loc) is equivalent to mmol/(340g Hb) and for fluxes, mmol/(Loch) is equivalent to mmol/(340g Hb.h). In the model, the standardized “mean” cell is defined with a volume of 1L/Loc contains 0.75Lcw/Loc and 0.25LHb/Loc, based on the specific molar weight of Hb of 1.36g/mL [74]. Assuming a mean volume of 100fL per RBC, 1 Loc will contain 10^{13} cells, an alternative to Loc occasionally used [11].

Rate constants: $1/h$, h^{-1} . Permeabilities in the red cell lore, and in the model, are traditionally expressed as rate-constants in units of h^{-1} , valid only as constants when referred to the area/volume ratio (A/V) of a standardized RBC. Use of rate-constants in h^{-1} units is a remnant from post-WWII decades when tracer flux measurements provided most of the information on the permeability properties of the RBC membrane in preference to the most widely used permeability units (P, in cm/s). Unlike permeabilities in cm/s, rate constants in h^{-1} units depend on, and vary with the volume/area of the RBC. For instance, for a RBC with initial volume and area of $\sim 90 \mu^3$ and $\sim 136 \mu^2$, respectively, giving a V/A ratio $\sim 0.7 \mu$ we can convert $1h^{-1}$ unit to P, in cm/s using $P = 1/h * 1/3600 \text{ s/h} * 0.7 \mu * 10^{-4} \text{ cm}/\mu \approx 1.9 * 10^{-8} \text{ cm/s}$. After dehydration at constant area to a V/A ratio of 0.4, for instance, a rate constant of $1h^{-1}$ would correspond to a permeability of $\sim 1.1 * 10^{-8} \text{ cm/s}$.

II. Modelling framework flowchart



III. User interface for building simulated experimental protocols

The following Tables list all the entries in Reference (RS) and Dynamic States (DS), with names, brief description of their function, default numerical value, and units. This is how the information appears on the different RS and DS windows. For variables, the same names are used as column headings in the csv output file.

REFERECE STATE			
NAME	DESCRIPTION	DEFAULT	UNITS
Hb A or S	Choice of normal Hb A or sickle cell Hb S	A	
Na/K pump Na efflux	Defines RS pump-leak turnover rate	-2.61	mmol/(Loch)
CNa	Initial cell Na ⁺ concentration	10	mmol/Lcw
CK	Initial cell K ⁺ concentration	140	mmol/Lcw
CA	Initial cell Cl ⁻ + HCO ₃ ⁻ concentration	95	mmol/Lcw
MCHC	Mean Cell Hb concentration	34	g/(dLpc)
PMCA FCaPmax	Calcium efflux through Ca ²⁺ -saturated pump	12	mmol/(Loch)
PCaG	Electrodiffusional Ca ²⁺ permeability	0.05	1/h
PKGardosMax	Ca ²⁺ -saturated Gardos channel PKGmax	30	1/h
KCa Gardos channel	K _{1/2} for Ca ²⁺ activation of Gardos channel	0.01	mM
Vw	Cell water content	0.75	L/Loc
Q10 active	(F at T+10°C)/(F at T) for active fluxes	4	
Q10 passive	(F at T+10°C)/(F at T) for passive fluxes	2	

DYNAMIC STATE: Time and Data Output Frequencies			
NAME	DESCRIPTION	DEFAULT	UNITS
Time	Sets the duration of selected DS	0	Minutes
FrequencyFactor	Iteration a-factor in $\Delta t = a/(b + \sum F_i)$	0.01	
Cyclesperprint (epochs)	Number of cycles between data outputs	777	Cycles/output
Accuracy	Sets decimal precision in data output	4	Decimals

DYNAMIC STATE: Cell Fraction and Medium composition			
NAME	DESCRIPTION	DEFAULT	UNITS
CVF	Cell Volume Fraction = Haematocrit/100	0.00001	
MB	Medium buffer concentration (HEPES)	10	mM
pHo	Medium pH	7.4	
Na x Glucamine	Isosmotic replacement of Na for impermeant cation	0	mM
A x Gluconate	Isosmotic replacement of A for impermeant anion	0	mM
NaCl x KCl	Isosmotic exchange of NaCl for KCl	0	mM
KCl x NaCl	Isosmotic exchange of KCl for NaCl	0	mM
NaCl add/remove	Add/remove NaCl with 2x osmolarity change	0	mM
KCl add/remove	Add/remove KCl with 2x osmolarity change	0	mM
Sucrose add/remove	Add/remove impermeant electroneutral solute	0	mM
MMg	Medium total Mg concentration	0.2	mM
MCa	Medium total Ca concentration	1	mM
EDGTA 0, 1 or 2	0=No chelator. G(1) = EGTA; D(2) = EDTA	0	
MEDGTA	Medium chelator concentration	0	mM

DYNAMIC STATE: Temperature and Permeabilities			
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NAME	DESCRIPTION	DEFAULT	UNITS
Temperature	Temperature during DS stage	37	°C
Pw	Water permeability	2	1/h
PK	Electrodiffusional K permeability	0.0017	1/h
PNa	Electrodiffusional Na permeability	0.0015	1/h
PA	Electrodiffusional A permeability	1.2	1/h
PH	Protonophore addition? From 2e-10 to 2e10	2e-10	1/h
PCa	Electrodiffusional Ca permeability	0.05	1/h
PA23CaMg	If added, enter values in the 2e17 to 2e18 range	0.01	
Hb Oxy or Deoxy	Changes the oxygenation condition of Hb	Oxy	

DINAMIC STATE: Transport inhibition
NAME and DEFAULT values
% inhibition of Na/K pump F _{Namax} : 0 (%)
% inhibition/stimulation(-) of JS k _{AE1} : 0 (%)
% inhibition/stimulation(-) of FCaP _{max} : 0 (%)
% inhibition of PKGardosMax: 0 (%)

DYNAMIC STATE: Pz, Dedicated PIEZO1 routine for exploring capillary transit effects			
NAME	DESCRIPTION	DEFAULT	UNITS
Pz stage no or yes	Activate PIEZO routine by entering yes	no	
Restore medium (no/yes)	Prevents carryover of changes in medium concentration after transits	yes	
Accuracy	Decimal data output precision	6	
PzFrequencyFactor	Iteration factor during Pz open state	1e-3	
Pzcyclesperprint	Number of cycles between data output points during Pz open state	111	Cycles/output
Pz Open state	Open state duration during capillary transit	0.4	Seconds (s)
Pz Transit CVF	Transit cell volume fraction	0.9	
PzK	K permeability through Pz channels	0	1/h
PzNa	Na permeability through Pz channels	0	1/h
PzA	A permeability through Pz channels	50	1/h
PzCa	Ca permeability through Pz channels	70	1/h
Pz JS I/S	JS Inhibition/Stimulation(-) during Pz open state	0	%
Pz PMCA I/S	PMCA Inhibition/Stimulation(-) during Pz open state	0	%

Additional parameters of the RBC model

NAME	DESCRIPTION
α	Slope of the H ⁺ -titration curve of Hb
α (alpha)	Main cell Ca ²⁺ buffer, mostly Hb
a, b	Δt controls in $\Delta t = a/(b + \sum F_i)$
pI	Isoelectric pH of Hb
b, c	Virial coefficients of f _{Hb} equation
kHA	Rate constant of JS H ⁺ :A ⁻ cotransport (2.5e8 h ⁻¹)
kNaA	Rate constant of Na ⁺ :A ⁻ cotransport

kKA	Rate constant of K^+A^- cotransport
kCo	Rate constant of Na:K:2A cotransport
d	Wildcard factor for $FCo = 0$ in RS
KB	HEPES H^+ dissociation constant
KBCa	Dissociation constant of 2nd Ca^{2+} buffer
KBMg1	Dissociation constant of Mg^{2+} buffer 1
KBMg2	Dissociation constant of Mg^{2+} buffer 2
zi	Valence of ion i
<i>F</i>	Faraday constant
R	Gas constant
T	Temperature

IV. HELP pages: (1) The Reference State, (2) The Dynamic State, and (3) The PIEZO1 subroutine

(1) *The Reference State (RS):*

The RS is a steady-state representing the initial condition of a RBC at the start of experiments. Changing default values automatically recalculates a new initial steady-state with equations that ensure compliance with charge and osmotic equilibrium within defined error margins. To approximate the condition of a young RBC, for instance, one could replace the corresponding defaults by CNa 5, CK 145, Vw 0.85 and FNaP - 3.2.

HbA or HbS: HbA and HbS differ in the values of their isoelectric point and net charge per mol, a . For HbA (default) $pI(0^\circ C) = 7.2$ and $a = -10 \text{ Eq}/(\text{mol} \cdot \Delta(pH - pI) \text{ unit})$; for HbS, the corresponding values are 7.4 and -8. The net charge on Hb at each pHi , n_{Hb} , is computed from $n_{Hb} = a \cdot (pHi - pI)$, where a is the slope of the proton titration curve of Hb in Eq/mol, and pI is the isoelectric pH of Hb. The model computes the change in pI with temperature (T) following: $pI(T^\circ C) = pI(0^\circ C) - 0.016 \cdot T(^\circ C)$ rendering a pI at $37^\circ C$ for HbA of about 6.6 [11].

Na/K pump Na efflux (FNaP): Changing the Na flux automatically resets the associated pump-mediated K influx and reverse Na/K fluxes following the stoichiometries and relative forward-reverse flux ratio encoded in the model.

CNa and CK: Initial Na^+ and K^+ concentrations in cell water (mmol/Lcw)

CA: Initial cell $Cl^- + HCO_3^-$ concentration (mmol/Lcw).

MCHC: “Mean Cell Haemoglobin Concentration”, a common haematological parameter in blood test assays, traditionally reported in gHb/dLoc; MCHC of “mean” model cell = 34 gHb/dLpc

PMCA Fmax (FCaPmax): Maximal Ca^{2+} extrusion flux through an ATP and Ca^{2+} -saturated plasma membrane calcium pump

PKGardosMax: electrodiffusional K^+ permeability through Ca^{2+} -saturated Gardos channels

KCa Gardos channel: Half-maximal Ca^{2+} dissociation constant ($K_{1/2}$) for Gardos channel activation

Vw: Water content associated with 340 g Hb; the volume occupied by 340 g Hb tetramer with a molar weight of 1.36 g/ml is 0.25 L. The default 0.75 Lcw/Loc for Vw sets a value of 1 L/Loc for the initial volume of the default “reference” RBC.

Q10 active or passive: The Q10 factors determine the extent by which active and passive fluxes (F) are set to increase or decrease for each $10^\circ C$ increase or decrease in temperature, T

(2) *The Dynamic State (DS).*

Displays five tabs grouping lists of parameters and variables with default values for constructing one or successive stages in simulated experimental protocols.

Time & Data Output Frequency:

Time: Sets the duration of each DS_n stage

FrequencyFactor: Sets the duration of each iteration interval (Δt at time = t) inversely proportional to the absolute value of the sum of all the net fluxes across the cell membrane at time t. Allows data output frequencies to appear proportional to the rate of change in the system at constant “cyclesperprint” values.

Cyclesperprint: sets the number of computational cycles between data output points

Accuracy: sets the decimal precision on the output data

Cell fraction and Medium Composition:

Medium concentrations of X are indicated by MX in mM units. Isosmotic exchanges of X for Y are shown as X x Y. Addition/removals allow changes in medium osmolarity. HEPES, Glucamine, gluconate, sucrose, Mg, EGTA and EDTA are treated as impermeant medium solutes. Sucrose represents any electroneutral impermeant small solute used to alter medium osmolarity. Gluconate and glucamine represent any impermeant monovalent ion used to replace A or Na in the medium.

MB: Medium buffer concentration, HEPES by default

EDGTA 0, 1, 2: Prompts for the addition of EGTA (G(1)) or EDTA (D(2)) to the cell suspension.

M_{EDGTA}: Prompts for the concentration of EGTA or EDTA, if added. The default is 0, no addition.

Temperature & Permeabilities:

Notation on the unit used for electrodiffusional ion permeabilities, 1/h or h⁻¹: 1/h is a widely used permeability unit in the RBC literature. For permeability comparisons between membranes from different cell types the most widely used unit is cm/s. For RBCs, both units are related through $P_{cm/s} = P_{1/h} * (V/A) / 3600$, where V and A correspond to the RBC volume and membrane area (in cm units) at the time the permeability measurement was taken.

PHG: PHG was modelled to enable simulations of the effects of protonophore additions. The default value represents no protonophore present. To simulate observed effects change PHG from 2e-1- to 2e10.

PA23CaMg: Ionophore A23187 mediates electroneutral X²⁺:2H⁺ exchanges with well defined highly non-linear kinetics in human RBCs. The default value represents absence of ionophore. To simulate the effects of ionophore concentrations capable of generating a Ca²⁺ flux exceeding that of the PMCA F_{max} at medium Ca²⁺ concentrations around 0.2 mM use values around 2e18.

Hb pI(0°C) oxy(7.2), deoxy(7.5): Hb is assumed to be in a oxygen-saturated condition by default (Oxy).

Deoxygenation of Hb (Deoxy) changes its pI(0°C) from 7.2 to 7.5. The model automatically adjusts the actual pI change for the temperature of the experiment. The pI shifts during oxy-deoxy transitions cause changes in the protonization condition of Hb with secondary changes in pHi and [Mg²⁺]_i, changes which the model predicts with verified accuracy.

Transport inhibition/stimulation (%; defaults = 0):

The default F_{max} value for each transporter, F_m, is modified according to $F_m * (100 - X) / 100$ where X is the % inhibition entered at the prompt. F_m stays modified in successive DS stages unless modified again.

Entries in successive DSs always apply to the original default (0%). To return to the original uninhibited F_m enter “0”. The same equation delivers stimulation if you enter negative numbers. It applies to JS and PMCA entries only. For an n-fold stimulation, enter “-n*100”, for instance “-200” for a two-fold stimulation. In the model, the intrinsic electrodiffusional anion permeability of the RBC, PA, is treated as a permeability pathway independent of the anion exchanger, thus preventing prejudged linkage associations when implementing the JS inhibition/stimulation subroutine [21].

(3). The PIEZO1 routine

All the membrane transporters represented in the red cell model are active all the time and participate to different extents in the dynamic responses to perturbations. PIEZO1 is unique in that it is generally inactive by default, responding only to conditions eliciting cell deformation, for instance during capillary transits. The dedicated PIEZO tag allows exploration of the potential effects of PIEZO1 activation during capillary transits. Preliminary tests showed that because of the brief duration of each transit (less than 1s), the magnitude of the changes in model variables was extremely small even attributing extremely high permeabilities to PIEZO1 mediation. This required a substantial increase in decimal accuracy and in the density of data acquisition. The tag offers a protocol design with defaults open to change by the user, as outlined next.

- I. In a DS selected for PIEZO1 implementation start by assigning to it a duration of 5min (Time: 5 (min))
- II. Double click on the “Incorporate PIEZO stage: no” and enter “yes”
- III. This brings up a predesigned DS stage with the following defaults or assignments:
 - 1) Incorporate PIEZO stage: yes
 - 2) Pz Restore Medium: Yes
 - 3) Accuracy: 6
 - 4) Pz frequency factor: 1e-3
 - 5) Pz cycles per print: 111
 - 6) Pz Open state: 0.4 (s)
 - 7) Pz Transit cell volume fraction: 0.9
 - 8) PzK: 0 (1/h)
 - 9) PzNa: 0 (1/h)
 - 10) PzA: 50 (1/h)
 - 11) PzCa: 70 (1/h)
 - 12) Pz JS I/S: 0 (%)
 - 13) Pz PMCA I/S: 0 (%)

Pz identifies PIEZO-related parameters and variables. The default five-minute duration of the PIEZO1 DS stage covers three consecutive periods: an initial two-minute baseline control, a PIEZO1 open state period, and a PIEZO1 closure period for the remainder time to five minutes to allow assessment of the short-term reversibility of the changes induced during the open state. Pz accuracy determines the decimal precision of data output in the csv file. Pz frequency factor and Pz cycles per print determine the density of data output over this period. The PzX values list the electrodiffusional permeabilities assigned by default to K^+ , Na^+ , A^- , and Ca^{2+} in the open state. The precise ion selectivity, conductance and inactivation kinetics with which PIEZO1 channels operate in human RBCs is unknown at present. The default permeability values chosen render model outcomes in conformity with the scarce but solid experimental evidence available in RBCs [60, 61, 76], as extensively analysed and discussed in the first application of the PIEZO1 model extension to investigate the circulatory physiopathology of human RBCs [14, 15]. The value of 0.9 used for the cell volume fraction applies strictly for the duration assigned to the capillary transit, to approximate the squeezed transit condition of the RBCs between the highly diluted (CFV ~ 0.00001) conditions in the systemic circulation before and after transits. The infinite reservoir composition of the medium representing the systemic circulation is preserved by the Restore Medium subroutine. The Pz JS and PMCA Pz Inhibition/Stimulation options enable study of the effects of anion-proton equilibration rates through the Jacobs-Stewart cycle, and of different levels of PMCA activity, but only during Pz open states. The options within the Transport inhibition tag operate only for periods outside Pz open states.

REFERENCES

1. Svetina S. Relations among variations in human red cell volume, density, membrane area, hemoglobin content and cation content. *J Theor Biol.* 1982;95(1):123-34.
2. Lew VL, Raftos JE, Sorette M, Bookchin RM, Mohandas N. Generation of normal human red cell volume, hemoglobin content, and membrane area distributions by "birth" or regulation? *Blood.* 1995;86(1):334-41. PubMed PMID: 7795242.
3. Tosteson DC. Regulation of cell volume by sodium and potassium transport. In: Hoffman JF, editor. *The Cellular Functions of Membrane Transport.* Englewood Cliffs, N.J.: Prentice Hall; 1964. p. 3-22.
4. Funder J, Wieth JO. Potassium, sodium, and water in normal human red blood cells. *Scand J clin Lab Invest.* 1966;18:167-80.
5. Beauge L, Lew VL. Passive fluxes of sodium and potassium across red cell membranes. In: Ellory JC, Lew VL, editors. *Membrane Transport in Red Cells.* London: Academic Press; 1977. p. 39-51.
6. Lew VL, Tiffert T. On the Mechanism of Human Red Blood Cell Longevity: Roles of Calcium, the Sodium Pump, PIEZO1, and Gardos Channels. *Front Physiol.* 2017;8:977. doi: 10.3389/fphys.2017.00977. PubMed PMID: 29311949; PubMed Central PMCID: PMC5732905.
7. Dasanna AK, Fedosov DA, Gompper G, Schwarz US. State diagram for wall adhesion of red blood cells in shear flow: from crawling to flipping. *Soft Matter.* 2019;15(27):5511-20. doi: 10.1039/c9sm00677j. PubMed PMID: 31241632.
8. Svetina S, Svelc Kebe T, Bozic B. A Model of Piezo1-Based Regulation of Red Blood Cell Volume. *Biophys J.* 2019;116(1):151-64. doi: 10.1016/j.bpj.2018.11.3130. PubMed PMID: 30580922; PubMed Central PMCID: PMC6342734.
9. Lew VL, Bookchin RM. Ion transport pathology in the mechanism of sickle cell dehydration. *Physiol Rev.* 2005;85(1):179-200.
10. Lew VL, Tiffert T, Ginsburg H. Excess hemoglobin digestion and the osmotic stability of *Plasmodium falciparum*-infected red blood cells. *Blood.* 2003;101:4189-94.
11. Mauritz JM, Esposito A, Ginsburg H, Kaminski CF, Tiffert T, Lew VL. The homeostasis of *Plasmodium falciparum*-infected red blood cells. *PLoS Comput Biol.* 2009;5(4):e1000339.
12. Waldecker M, Dasanna AK, Lansche C, Linke M, Srismith S, Cyrklaff M, et al. Differential time-dependent volumetric and surface area changes and delayed induction of new permeation pathways in *P. falciparum*-infected hemoglobinopathic erythrocytes. *Cell Microbiol.* 2017;19(2). doi: 10.1111/cmi.12650. PubMed PMID: 27450804; PubMed Central PMCID: PMC5298026.
13. Ponder E. Hemolysis and related phenomena. New York: Grune & Stratton; 1948 1948.
14. Rogers S, Lew VL. PIEZO1 and the mechanism of the long circulatory longevity of human red blood cells. *PLoS Comput Biol.* 2021;17(3):e1008496. doi: 10.1371/journal.pcbi.1008496. PubMed PMID: 33690597.
15. Rogers S, Lew VL. Up-down biphasic volume response of human red blood cells to PIEZO1 activation during capillary transits. *PLoS Comput Biol.* 2021;17(3):e1008706. doi: 10.1371/journal.pcbi.1008706. PubMed PMID: 33657092.
16. Glynn IM, Warner AE. Nature of the calcium dependent potassium leak induced by (+)-propranolol, and its possible relevance to the drug's antiarrhythmic effect. *Br J Pharmacol.* 1972;44:271-8.
17. Armstrong CM. The Na/K pump, Cl ion, and osmotic stabilization of cells. *Proc Natl Acad Sci U S A.* 2003;100(10):6257-62.
18. Weiss TF. *Cellular Biophysics: The MIT Press;* 1996 1996.
19. Lew VL, Ferreira HG, Moura T. The behaviour of transporting epithelial cells. I. Computer analysis of a basic model. *Proc R Soc London B.* 1979;206:53-83.
20. Lew VL, Bookchin RM. Volume, pH, and ion-content regulation in human red cells: analysis of transient behavior with an integrated model. *J Membr Biol.* 1986;92(1):57-74. PubMed PMID: 3746891.
21. Freeman CJ, Bookchin RM, Ortiz OE, Lew VL. K-permeabilized human red cells lose an alkaline, hypertonic fluid containing excess K over diffusible anions. *J Membrane Biol.* 1987;96:235-41.
22. Lew VL, Freeman CJ, Ortiz OE, Bookchin RM. A mathematical model of the volume, pH, and ion content regulation in reticulocytes. Application to the pathophysiology of sickle cell dehydration. *J Clin*

- Invest. 1991;87(1):100-12. doi: 10.1172/JCI114958. PubMed PMID: 1985088; PubMed Central PMCID: PMCPMC295002.
23. Bookchin RM, Ortiz OE, Lew VL. Evidence for a direct reticulocyte origin of dense red cells in sickle cell anemia. *Journal of Clinical Investigation*. 1991;87:113-24.
 24. Hills A, Chen ZH, Amtmann A, Blatt MR, Lew VL. OnGuard, a computational platform for quantitative kinetic modeling of guard cell physiology. *Plant Physiol*. 2012;159(3):1026-42.
 25. Jezek M, Hills A, Blatt MR, Lew VL. A constraint-relaxation-recovery mechanism for stomatal dynamics. *Plant Cell Environ*. 2019;42(8):2399-410. doi: 10.1111/pce.13568. PubMed PMID: 31032976.
 26. Minguet-Parramona C, Wang Y, Hills A, Violet-Chabrand S, Griffiths H, Rogers S, et al. An Optimal Frequency in Ca^{2+} Oscillations for Stomatal Closure Is an Emergent Property of Ion Transport in Guard Cells. *Plant Physiol*. 2016;170(1):33-42. doi: 10.1104/pp.15.01607. PubMed PMID: 26628748; PubMed Central PMCID: PMCPMC4704601.
 27. Violet-Chabrand S, Hills A, Wang Y, Griffiths H, Lew VL, Lawson T, et al. Global Sensitivity Analysis of OnGuard Models Identifies Key Hubs for Transport Interaction in Stomatal Dynamics. *Plant Physiol*. 2017;174(2):680-8. doi: 10.1104/pp.17.00170. PubMed PMID: 28432256; PubMed Central PMCID: PMCPMC5462055.
 28. Etzion Z, Tiffert T, Bookchin RM, Lew VL. Effects of deoxygenation on active and passive Ca^{2+} transport and on the cytoplasmic Ca^{2+} levels of sickle cell anemia red cells. *Journal of Clinical Investigation*. 1993;92:2489-98.
 29. Tiffert T, Etzion Z, Bookchin RM, Lew VL. Effects of deoxygenation on active and passive Ca^{2+} transport and cytoplasmic Ca^{2+} buffering in normal human red cells. *J Physiol*. 1993;464:529-44.
 30. Raftos JE, Lew VL, Flatman PW. Refinement and evaluation of a model of Mg^{2+} buffering in human red cells. *Eur J Biochem*. 1999;263(3):635-45.
 31. Swietach P, Tiffert T, Mauritz JM, Seear R, Esposito A, Kaminski CF, et al. Hydrogen ion dynamics in human red blood cells. *J Physiol*. 2010;588(Pt 24):4995-5014.
 32. Esposito A, Choimet JB, Skepper JN, Mauritz JM, Lew VL, Kaminski CF, et al. Quantitative imaging of human red blood cells infected with *Plasmodium falciparum*. *Biophys J*. 2010;99(3):953-60.
 33. Mauritz JM, Seear R, Esposito A, Kaminski CF, Skepper JN, Warley A, et al. X-ray microanalysis investigation of the changes in Na, K, and hemoglobin concentration in plasmodium falciparum-infected red blood cells. *Biophys J*. 2011;100(6):1438-45.
 34. Tiffert T, Lew VL. Cytoplasmic calcium buffers in intact human red cells. *J Physiol*. 1997;500 (Pt 1):139-54. PubMed PMID: 9097939; PubMed Central PMCID: PMCPMC1159365.
 35. Alper SL. Genetic Diseases of PIEZO1 and PIEZO2 Dysfunction. *Curr Top Membr*. 2017;79:97-134. doi: 10.1016/bs.ctm.2017.01.001. PubMed PMID: 28728825.
 36. Andolfo I, Alper SL, De Franceschi L, Auriemma C, Russo R, De Falco L, et al. Multiple clinical forms of dehydrated hereditary stomatocytosis arise from mutations in PIEZO1. *Blood*. 2013;121(19):3925-35, S1-12. doi: 10.1182/blood-2013-02-482489. PubMed PMID: 23479567.
 37. Bae C, Gnanasambandam R, Nicolai C, Sachs F, Gottlieb PA. Xerocytosis is caused by mutations that alter the kinetics of the mechanosensitive channel PIEZO1. *Proc Natl Acad Sci U S A*. 2013;110(12):E1162-8. doi: 10.1073/pnas.1219777110. PubMed PMID: 23487776; PubMed Central PMCID: PMCPMC3606986.
 38. Cahalan SM, Lukacs V, Ranade SS, Chien S, Bandell M, Patapoutian A. Piezo1 links mechanical forces to red blood cell volume. *Elife*. 2015;4. doi: 10.7554/eLife.07370. PubMed PMID: 26001274; PubMed Central PMCID: PMCPMC4456639.
 39. Danielczok JG, Terriac E, Hertz L, Petkova-Kirova P, Lautenschlager F, Laschke MW, et al. Red Blood Cell Passage of Small Capillaries Is Associated with Transient Ca^{2+} -mediated Adaptations. *Front Physiol*. 2017;8:979. doi: 10.3389/fphys.2017.00979. PubMed PMID: 29259557; PubMed Central PMCID: PMCPMC5723316.
 40. Gnanasambandam R, Rivera A, Vandorpe DH, Shmukler BE, Brugnara C, Snyder LM, et al. Increased Red Cell KCNN4 Activity in Sporadic Hereditary Xerocytosis Associated With Enhanced Single Channel Pressure Sensitivity of PIEZO1 Mutant V598M. *Hemasphere*. 2018;2(5):e55. doi:

10.1097/HS9.0000000000000055. PubMed PMID: 30887001; PubMed Central PMCID: PMC6407803.

41. Kuchel PW, Shishmarev D. Accelerating metabolism and transmembrane cation flux by distorting red blood cells. *Sci Adv.* 2017;3(10):eaao1016. doi: 10.1126/sciadv.aao1016. PubMed PMID: 29057326; PubMed Central PMCID: PMC647125.
42. Rivera A, Vanderpe DH, Shmukler BE, Andolfo I, Iolascon A, Archer NM, et al. Erythrocyte ion content and dehydration modulate maximal Gardos channel activity in KCNN4 V282M/+ hereditary xerocytosis red cells. *Am J Physiol Cell Physiol.* 2019;317(2):C287-C302. doi: 10.1152/ajpcell.00074.2019. PubMed PMID: 31091145.
43. Gardos G. The function of calcium in the potassium permeability of human erythrocytes. *Biochim Biophys Acta.* 1958;30:653-4.
44. Gardos G. Effect of ethylenediaminetetraacetate on the permeability of human erythrocytes. *Acta Physiol Acad Sci Hung.* 1958;14(1):1-5. PubMed PMID: 13544963.
45. Gardos G. The role of calcium in the potassium permeability of human erythrocytes. *Acta Physiol Acad Sci Hung.* 1959;15(2):121-5. PubMed PMID: 13660848.
46. Lew VL. On the ATP dependence of the Ca^{2+} -induced increase in K^{+} permeability observed in human red cells. *Biochim Biophys Acta.* 1971;233:827-30.
47. Lew VL. Effect of ouabain on the Ca^{2+} -dependent increase in K^{+} permeability in depleted guinea-pig red cells. *Biochim Biophys Acta.* 1971;249:236-9.
48. Lew VL, Ferreira HG. Calcium transport and the properties of a calcium-activated potassium channel in red cell membranes. In: Kleinzeller A, Bronner F, editors. *Current Topics in Membranes and Transport*, Vol 10: Academic Press, NY; 1978. p. 217-77.
49. Schatzmann HJ. ATP-dependent Ca^{++} -extrusion from human red cells. *Experientia.* 1966;22(6):364-5.
50. Zambo B, Varady G, Padanyi R, Szabo E, Nemeth A, Lango T, et al. Decreased calcium pump expression in human erythrocytes is connected to a minor haplotype in the ATP2B4 gene. *Cell Calcium.* 2017;65:73-9. doi: 10.1016/j.ceca.2017.02.001. PubMed PMID: 28216081.
51. Lew VL. On the mechanism of the Ca-induced increase in K permeability observed in human red cell membranes. In: Bolis L, Bloch K, Luria SE, Lynen F, editors. *Comparative Biochemistry and Physiology of Transport*. Amsterdam: North-Holland Publishing Company; 1974. p. 310-6.
52. Shen SC, Fleming EM, Castle AB, Castle WB. Studies on the destruction of red blood cells. V. Irreversibly sickled erythrocytes: Their experimental production in vitro. *Blood.* 1949;4:498-504.
53. Bertles JF, Milner PFA. Irreversibly sickled erythrocytes: a consequence of the heterogeneous distribution of hemoglobin types in sickle cell anemia. *Journal of Clinical Investigation.* 1968;47:1731-41.
54. Fermo E, Bogdanova A, Petkova-Kirova P, Zaninoni A, Marcello AP, Makhro A, et al. 'Gardos Channelopathy': a variant of hereditary Stomatocytosis with complex molecular regulation. *Sci Rep.* 2017;7(1):1744. doi: 10.1038/s41598-017-01591-w. PubMed PMID: 28496185; PubMed Central PMCID: PMC6431847.
55. Rivera A, Vanderpe DH, Shmukler BE, Gallagher DR, Fikry CC, Kuypers FA, et al. Erythrocytes from hereditary xerocytosis patients heterozygous for KCNN4 V282M exhibit increased spontaneous Gardos channel-like activity inhibited by senicapoc. *Am J Hematol.* 2017;92(6):E108-E10. doi: 10.1002/ajh.24716. PubMed PMID: 28295477.
56. Cantley LC, Jr., Josephson L, Warner R, Yanagisawa M, Lechene C, Guidotti G. Vanadate is a potent (Na,K)-ATPase inhibitor found in ATP derived from muscle. *J Biol Chem.* 1977;252(21):7421-3.
57. Tiffert T, Lew VL. Kinetics of inhibition of the plasma membrane calcium pump by vanadate in intact human red cells. *Cell Calcium.* 2001;30(5):337-42.
58. García-Sancho J, Lew VL. Detection and separation of human red cells with different calcium contents following uniform calcium permeabilization. *J Physiol.* 1988;407:505-22.
59. Hunter MJ. Human erythrocyte anion permeabilities measured under conditions of net charge transfer. *J Physiol.* 1977;268:35-49.
60. Glogowska E, Dyrda A, Cueff A, Bouyer G, Egee S, Bennekou P, et al. Anion conductance of the human red cell is carried by a maxi-anion channel. *Blood Cells Mol Dis.* 2010;44(4):243-51.

61. Thomas SL, Bouyer G, Cueff A, Egee S, Glogowska E, Ollivaux C. Ion channels in human red blood cell membrane: actors or relics? *Blood Cells Mol Dis.* 2011;46(4):261-5.
62. Cockrell RS, Harris EJ, Pressman BC. Energetics of potassium transport in mitochondria induced by valinomycin. *Biochemistry.* 1966;5(7):2326-35.
63. Pressman BC. Ionophorous antibiotics as models for biological transport. *Fed Proc.* 1968;27(6):1283-8.
64. Dagher G, Lew VL. Maximal calcium extrusion capacity and stoichiometry of the human red cell calcium pump. *J Physiol.* 1988;407:569-86.
65. Pereira AC, Samellas D, Tiffert T, Lew VL. Inhibition of the calcium pump by high cytosolic Ca^{2+} in intact human red blood cells. *J Physiol.* 1993;461:63-73. PubMed PMID: 8394428; PubMed Central PMCID: PMCPMC1175245.
66. Cala PM. Cell volume regulation by Amphiuma red blood cells. The role of Ca^{2+} as a modulator of alkali metal/ H^{+} exchange. *J Gen Physiol.* 1983;82(6):761-84. doi: 10.1085/jgp.82.6.761. PubMed PMID: 6420507; PubMed Central PMCID: PMCPMC2228718.
67. Cala PM. Volume regulation by Amphiuma red blood cells. The membrane potential and its implications regarding the nature of the ion-flux pathways. *J Gen Physiol.* 1980;76(6):683-708. doi: 10.1085/jgp.76.6.683. PubMed PMID: 10822499; PubMed Central PMCID: PMCPMC2228608.
68. Hoffmann EK, Simonsen LO. Membrane mechanisms in volume and pH regulation in vertebrate cells. *Physiol Rev.* 1989;69(2):315-82. doi: 10.1152/physrev.1989.69.2.315. PubMed PMID: 2538851.
69. Jacobs MH, Stewart DR. The role of carbonic anhydrase in certain ionic exchanges involving the erythrocyte. *J Gen Physiol.* 1942;25:539-52.
70. Hladky SB, Rink TJ. pH equilibrium across the red cell membrane. In: Ellory JC, Lew VL, editors. *Membrane transport in Red Cells.* London: Academic Press; 1977. p. 115-35.
71. Jarvis HG, Gore DM, Briggs C, Chetty MC, Stewart GW. Cold storage of 'cryohydrocytosis' red cells: the osmotic susceptibility of the cold-stored erythrocyte. *Br J Haematol.* 2003;122(5):859-68.
72. Raftos JE, Lew VL. Effect of intracellular magnesium on calcium extrusion by the plasma membrane calcium pump of intact human red cells. *J Physiol.* 1995;489:63-72.
73. Staines HM, Ellory JC, Kirk K. Perturbation of the pump-leak balance for Na^{+} and K^{+} in malaria-infected erythrocytes. *Am J Physiol Cell Physiol.* 2001;280(6):C1576-C87.
74. Bureau M, Banerjee R. Structure-volume relationships in hemoglobin. A densitometric and dilatometric study of the oxy leads to deoxy transformation. *Biochimie.* 1976;58:403-7.
75. Goldman DE. Potential, Impedance, and Rectification in Membranes. *J Gen Physiol.* 1943;27(1):37-60. doi: 10.1085/jgp.27.1.37. PubMed PMID: 19873371; PubMed Central PMCID: PMCPMC2142582.
76. Dyrda A, Cytlak U, Ciuraszkiewicz A, Lipinska A, Cueff A, Bouyer G, et al. Local membrane deformations activate Ca^{2+} -dependent K^{+} and anionic currents in intact human red blood cells. *PLoS One.* 2010;5(2):e9447.