## Parallel adaptive feedback enhances reliability of the Ca<sup>2+</sup> signaling system

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Despite large cell-to-cell variations in the concentrations of individual signaling proteins, cells transmit signals correctly. This phenomenon raises the question of what signaling systems do to prevent a predicted high failure rate. Here we combine quantitative modeling, RNA interference, and targeted selective reaction monitoring (SRM) mass spectrometry, and we show for the ubiquitous and fundamental calcium signaling system that cells monitor cytosolic and endoplasmic reticulum (ER) Ca<sup>2+</sup> levels and adjust in parallel the concentrations of the store-operated Ca<sup>2+</sup> influx mediator stromal interaction molecule (STIM), the plasma membrane Ca<sup>2+</sup> pump plasma membrane Ca-ATPase (PMCA), and the ER Ca<sup>2+</sup> pump sarco/ER Ca<sup>2+</sup>-ATPase (SERCA). Model calculations show that this combined parallel regulation in protein expression levels effectively stabilizes basal cytosolic and ER Ca2+ levels and preserves receptor signaling. Our results demonstrate that, rather than directly controlling the relative level of signaling proteins in a forward regulation strategy, cells prevent transmission failure by sensing the state of the signaling pathway and using multiple parallel adaptive feedbacks.

absolute quantification of calcium signaling proteins | targeted proteomics | compensation mechanisms in signal transduction networks | cell signaling | protein abundance profiling

**R**ecent studies in mammalian cells and lower eukaryotes showed a significant variation in the expression level of individual proteins between genetically identical cells (1, 2). These variations—often exceeding a factor of 2—likely reflect noise in transcription, mRNA turnover, translation, and protein degradation. This cell-to-cell variability creates a fundamental problem for signal transduction because even small differences in the relative levels of signaling components along a pathway lead to a processive degradation of signal transmission and failed physiological responses. However, cells typically manage to transmit signals correctly, overcoming protein expression variations by unknown mechanisms.

Ca<sup>2+</sup> signals regulate fundamental eukaryotic processes that range from secretion, muscle contraction, and memory formation to activation and differentiation of immune cells (3, 4). Our detailed knowledge of the Ca2+ signaling system makes this pathway ideally suited to address the puzzling question of how signal transmission processes can tolerate large variations in the expression level of the underlying signaling components. We hypothesized that additional control principles must exist to explain how the otherwise well-understood  $Ca^{2+}$  signaling system avoids failure and copes with large cell-to-cell variations in protein expression. To keep cells in a signaling-competent state, the  $Ca^{2+}$  signaling system employs a system of pumps and channels to tightly balance  $Ca^{2+}$  fluxes across the endoplasmic reticulum (ER) and plasma membrane (PM). The highly coordinated regulation of calcium homeostasis raises the questions of whether the pumps and channels are always expressed at the correct numbers by using shared promoters or coupled protein stability in a forward regulation strategy or, alternatively, whether cells monitor the output of a pathway and use a feedback strategy to adjust the concentrations of individual signaling components.

## **Results and Discussion**

We first determined computationally how the currently known  $Ca^{2+}$  signaling system is affected by variations in protein expression. We used an established  $Ca^{2+}$  signaling model (5) with added terms reflecting the recently discovered stromal interaction molecule (STIM)–Orai  $Ca^{2+}$  influx pathway (Fig. 14) (6). The model links receptor activation of phospholipase C and the production of the second messenger inositol 1,4,5-trisphosphate (InsP3) to the InsP3-induced opening of  $Ca^{2+}$  channels (InsP3R) in the ER (4). The resulting drop in ER  $Ca^{2+}$  level activates the ER  $Ca^{2+}$  sensor STIM, which translocates to ER–PM junctions, where it triggers  $Ca^{2+}$  influx into the cytosol through Orai channels (7, 8). PM Ca–ATPase (PMCA)  $Ca^{2+}$  pumps in the PM extrude  $Ca^{2+}$  out of the cells and establish a steep  $Ca^{2+}$  gradient from a basal level in the cytosol of 50 nM to a concentration of 1.5 mM outside the cell. Sarco/ER  $Ca^{2+}$ –ATPase (SERCA)  $Ca^{2+}$  pumps in the ER membrane keep the ER filled to a basal  $Ca^{2+}$  level of 350 µM. The model includes positive feedbacks from cytosolic  $Ca^{2+}$  release (5, 9, 10) that together generate receptor-triggered  $Ca^{2+}$  oscillations.

To test for a role of mitochondrial calcium in regulating basal cytosolic and ER calcium in the Drosophila cells used in this study, we used RNA interference (RNAi) to knock down expression of MICU1, a protein that has been shown to be essential for mitochondrial calcium uptake (11). Consistent with the results in human HELA cells (11), we did not see a statistically significant effect of MICU1 knockdown on basal cytosolic or basal ER calcium levels (SI Appendix, Figs. S2 and S3). However, because mitochondrial calcium is known to modulate the pattern of calcium signals (12) and may regulate basal cytosolic and ER calcium levels in some cell types, the model does include terms describing the two known modes of Ca<sup>2+</sup> uptake into mitochondria via a uniporter channel: uptake from the cytoplasm and a more direct transfer of Ca<sup>2+</sup> from the ER to mitochondria at ERmitochondria contact sites (13). Uptake into mitochondria leads to a clipping of  $Ca^{2+}$  transients when they start to exceed a critical level of cytosolic Ca<sup>2+</sup>. A main mitochondrial Ca<sup>2+</sup> extrusion mechanism is likely mediated by a  $Na^+/Ca^{2+}$  exchanger (14).

Model simulations showed that weak receptor stimuli have no effect on  $Ca^{2+}$  levels (Fig. 1*B*, top trace). A further increase in receptor stimulus intensity led to a dramatic phase transition whereby  $Ca^{2+}$  levels underwent a repetitive spiking (oscillation)

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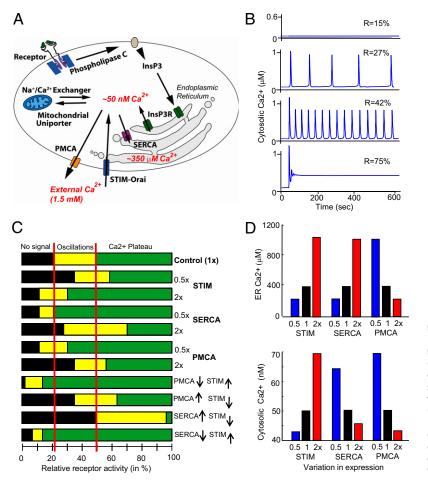
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Data deposition: The models presented in this manuscript have been uploaded to the European Molecular Biology Laboratory–European Bioinformatics Institute BioModels database, http://www.ebi.ac.uk/biomodels-main/.

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**Fig. 1.** Model simulations show that the Ca<sup>2+</sup> signaling system fails to function when signaling components are increased or decreased twofold. (*A*) Schematic of the Ca<sup>2+</sup> signaling model. (*B*) Output of the model at different levels of receptor activation. (C) When STIM, SERCA, and PMCA concentrations were varied from 50% to 200% of their setpoint concentrations (by regulating the synthesis rates), the transition points where receptor activation triggered Ca<sup>2+</sup> oscillations or induced the plateau phase varied significantly from the control. In the control, all proteins were expressed at their normal, 1× concentrations. (*D*) Significant changes in basal ER and cytosolic Ca<sup>2+</sup> levels that may affect cell health occurred when twofold changes were made to the synthesis rates of PMCA, SERCA, and STIM.

behavior that increased in frequency as the receptor stimulus increased (middle traces). At even higher levels of receptor activation, the  $Ca^{2+}$  signal transitioned into a plateau phase, in which  $Ca^{2+}$  levels stayed elevated as long as the receptor remained engaged (bottom trace). These oscillation and plateau phases reflect the typical Ca<sup>2+</sup> signaling responses observed in many eukaryotic cell types (15). In some cell systems, as well as in our model output, significantly more complex oscillation patterns can be observed than are shown in Fig. 1B—for example, oscillations on top of an elevated baseline or infrequent spikes of varying durations. However, to simplify the output classes, we placed the model outputs into just three categories: no signal, oscillations, and plateaus. If oscillations decayed into an elevated baseline, we considered this output to be in the plateau category. If, however, a plateau switched to and ended in an oscillatory pattern, we considered that output to be in the oscillatory category. Examples of how we categorized the model outputs are given in SI Appendix, Fig. S4. It should be noted that the model describes whole-cell changes in cytosolic, ER, and mitochondrial Ca<sup>2+</sup> levels and does not include the effects of spatial distribution of Ca<sup>2+</sup> signals or spatial distributions of relevant Ca<sup>2+</sup> transport proteins, such as those described in ref. 16.

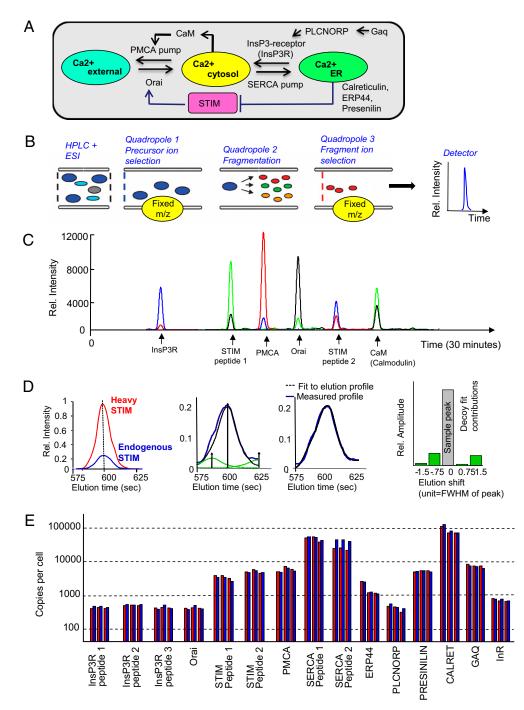
We introduced into the model twofold variations in the synthesis rates of STIM, SERCA, or PMCA. Markedly, the critical receptor activation levels where cells switched from eliciting no response to triggering oscillations and to inducing a plateau phase shifted significantly from the control case (Fig. 1*C*). The shifts became even larger and more disruptive as dual and higher-order perturbations were performed, as shown in the example in the lower part of Fig. 1*C*, with opposing twofold changes in STIM and PMCA or in STIM and SERCA. The need for lower or higher receptor stimuli, together with the reduced range of stimulus intensity over which oscillations occur, show that downstream processes controlled by oscillations, such as secretion or differentiation, would fail to be triggered in most cells if our current model were correct.

Of possibly more serious long-term concern for cell health, perturbations of even a single component caused significant changes in the levels of basal ER and cytosolic  $Ca^{2+}$  (Fig. 1D). This finding is problematic because changes in basal cytosolic  $Ca^{2+}$  can disrupt the normal activities of  $Ca^{2+}$ -regulated proteases, phosphatases, and other effectors (17), and changes in basal ER  $Ca^{2+}$  levels lead to protein misfolding and trigger ER stress responses (18, 19) and apoptosis (20). Thus, twofold stochastic variations in the expression of signaling proteins between otherwise identical cells would wreak havoc in the performance of the  $Ca^{2+}$  signaling system as it is currently understood. This "factor-of-two resilience problem" is likely a fundamental challenge that all signaling systems have to resolve.

We considered three plausible hypotheses of how the Ca<sup>2+</sup> signaling system could prevent failure. (*i*) In a "co-expression control hypothesis," cells would rely on constant relative coexpression of Ca<sup>2+</sup> signaling proteins. Coexpression control is often observed when two proteins bind to and stabilize each other. However, many signaling components are spatially separated and cannot effectively use this mechanism. (*ii*) In a "nodal point control hypothesis," a single component in the Ca<sup>2+</sup> system is targeted by feedback regulation and provides system stability. Nodal points with central importance have been described in many types of complex systems (21). Such a single feedback could, for example, monitor cytosolic Ca<sup>2+</sup> and adjust the expression of the PMCA pump to stabilize the system. (*iii*) In a third "parallel adaptive control hypothesis," multiple feedback loops would act in parallel. Cells could, for example, monitor cytosolic and ER Ca<sup>2+</sup> levels and continuously make multiple but small

adjustments to different signaling components to keep the system optimally tuned. However, this type of control mechanism may be challenging for a cell to implement because engineered systems with multiple feedbacks can have high failure rates (22). which have the same  $Ca^{2+}$  signaling components as other animal cells (Fig. 2.4). The analysis is facilitated in *Drosophila*, which has only single gene copies of these components compared with the multiple gene copies found in vertebrate cells (23). We quantitatively measured protein concentrations using selective reaction monitoring (SRM) mass spectrometry (Fig. 2 *B* and *C*; refs. 24–

To determine which of these three hypotheses is correct, we investigated the  $Ca^{2+}$  signaling system in *Drosophila* S2R+ cells,



**Fig. 2.** Selective reaction monitoring (SRM) mass spectrometry determination of copy numbers of important  $Ca^{2+}$  signaling components in *Drosophila* S2R+ cells. (A) Schematic of the targeted proteins and their regulatory connections. (B) SRM measurements were carried out by using a nano-HPLC and electrospray ionization source coupled to a triple-quadropole mass spectrometer. The transitions (precursor/fragment ion pairs) generated for each targeted peptide were read out as intensities by the detector. (C) A subset of transitions measured by SRM in a typical sample. (D) Fit of the peak of a heavy peptide to the light (endogenous) trace. Time-shifted decoy peaks were used in the fit to quantify background peptide activities. Four decoy shapes were generated by shifting the heavy peptide peak shape to the left and right, respectively. Decoy peaks significantly contributing to the fit indicated noise in the sample trace, and fragments were rejected. (E) Copy numbers of key  $Ca^{2+}$  signaling proteins per S2R+ cell measured in three biological replicates. The red and blue bars show quantitation of two different transitions for each peptide. Measurements of multiple independent peptides are shown for InsP3R, STIM, and SERCA.

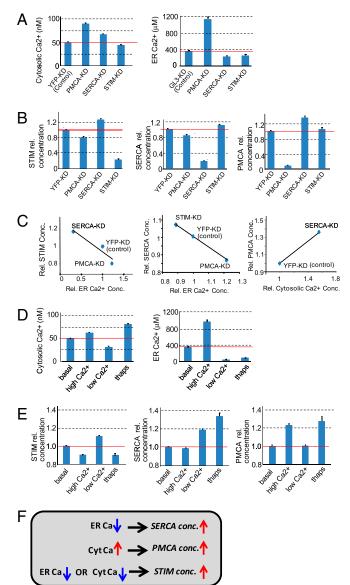
26) by identifying one or more proteotypic peptides for STIM, PMCA, and SERCA, as well as for a number of additional Ca<sup>2</sup> signaling components and control proteins used for normalization (Fig. 2A and SI Appendix, Table S1). Heavy isotopically labeled synthesized peptides (27) were added at a known concentration to quantify the corresponding endogenous peptide peak (Fig. 2D and SI Appendix, Table S2). An estimated copy number per cell was derived by fitting the heavy peak to the sample peak (a decoy peak analysis was used to reject noisy peptide transitions) (Fig. 2E and SI Appendix, SI Methods). The <sup>+</sup> pumps PMCA and SERCA (~8,000 and 80,000 copies, Ca2respectively) were significantly more abundant than the Ca<sup>2</sup> channels InsP3R and Orai, which were present at only ~800 copies per cell. The higher level of SERCA compared with PMCA is consistent with the experimentally observed much higher pump rate into the ER compared with the pump rate out of cells. The ER  $Ca^{2+}$  sensor STIM was present at an almost 10fold higher concentration than Orai, possibly indicating additional signaling roles or a role in the cooperative activation of Orai by multiple STIMs (28).

We next directly tested for adaptive mechanisms by using RNAi to knock down SERCA, PMCA, and STIM. Control experiments using SRM mass spectrometry showed that RNAi knockdowns typically led to 70–90% reductions of the targeted protein concentration for the conditions used (*SI Appendix*, Fig. S5). A reduction in the expression of the Ca<sup>2+</sup> pump that transports Ca<sup>2+</sup> out of the cell (PMCA) is predicted to directly increase cytosolic Ca<sup>2+</sup> and to indirectly increase ER Ca<sup>2+</sup>. Indeed, knockdown of PMCA increased both cytosolic and ER Ca<sup>2+</sup> concentrations (Fig. 34). Also as expected, SERCA knockdown lowered ER Ca<sup>2+</sup> levels by preventing the pumping of Ca<sup>2+</sup> into the ER. Lowering ER Ca<sup>2+</sup> levels induces store-operated Ca<sup>2+</sup> influx into the cell (7), and thus SERCA knockdown also resulted in increased cytosolic ca<sup>2+</sup> levels. STIM knockdown reduced basal cytosolic and ER Ca<sup>2+</sup> levels. DRDA is consistent with STIM's role to enhance Ca<sup>2+</sup> influx into cells.

Strikingly, knockdown of STIM, PMCA, and SERCA not only reduced the concentration of the targeted protein, but also significantly changed the concentration of the other two components not targeted by the dsRNA (Fig. 3*B*). STIM expression was reduced by knockdown of PMCA and increased by knockdown of SERCA in the opposite direction to the change in ER Ca<sup>2+</sup> level, suggesting that STIM expression is inversely regulated by ER Ca<sup>2+</sup> (Fig. 3*C*). A similar inverse regulation between ER Ca<sup>2+</sup> levels and SERCA expression was also observed when STIM or PMCA was knocked down. For PMCA, a plausible interpretation of the data is that the increase in cytosolic Ca<sup>2+</sup> mediated by SERCA knockdown increases its expression (Fig. 3*C*).

To directly determine whether it is the ER or the cytosol  $Ca^{2+}$ that controls PMCA, STIM, and SERCA expression, we exposed cells to high or low external  $Ca^{2+}$  to shift their basal cytosolic and ER  $Ca^{2+}$  levels. High external  $Ca^{2+}$  was expected to increase basal  $Ca^{2+}$  influx and to cause a corresponding increase in basal cytosolic and ER  $Ca^{2+}$  levels. Low external  $Ca^{2+}$  was expected to decrease basal  $Ca^{2+}$  influx and cytosolic and ER  $Ca^{2+}$  levels. We indeed observed these expected changes experimentally (Fig. 3D). To generate an opposing change in ER and cytosolic  $Ca^{2+}$  levels, we used the SERCA inhibitor thapsigargin, which is known to lower ER  $Ca^{2+}$  and thereby increase cytosolic  $Ca^{2+}$  by inducing STIM–Orai-mediated store-operated  $Ca^{2+}$  influx. It should be noted that changes in  $Ca^{2+}$  buffering in the cytosol or ER are likely less important for altering  $Ca^{2+}$  homeostasis because our model shows that such changes have only small effects compared with changes in the ratio between  $Ca^{2+}$  pump and flux rates.

We then used SRM mass spectrometry to measure the effect of direct changes in ER and cytosolic  $Ca^{2+}$  concentration on the protein levels (Fig. 3*E*). The inverse change in STIM concentration in response to high and low  $Ca^{2+}$  exposure further showed that ER  $Ca^{2+}$  inversely regulates STIM expression. The lowering in STIM concentration after thapsigargin treatment suggested that STIM is not only up-regulated by low ER  $Ca^{2+}$  but also likely



**Fig. 3.** Demonstration that multiple parallel adaptive feedback loops exist in the Ca<sup>2+</sup> signaling system. The error bars in all plots show SE. (A and B) Effect of RNAi knockdown on cytosolic and ER Ca<sup>2+</sup> levels (A) and protein levels (B). Knockdown of YFP or GL3, proteins that are not present in S2R+ cells, were used as controls. (C) The small changes in STIM and SERCA and PMCA protein concentration correlated with Ca<sup>2+</sup> level changes, but data in B are not sufficient to determine whether the changes were regulated just by ER or just by cytosolic Ca<sup>2+</sup>. (D and E) Effect of treating cells for 24 h with low (2.5 mM EGTA plus 2.5 mM Mg<sup>2+</sup>) and high (10 mM Ca<sup>2+</sup>) external Ca<sup>2+</sup> and thapsagargin addition (1  $\mu$ M) on cytosolic and ER Ca<sup>2+</sup> levels (D) and protein levels (E). (F) Summary of the three identified feedback loops.

suppressed by high cytosolic  $Ca^{2+}$  levels. SERCA expression was increased in low external  $Ca^{2+}$  as well as in the presence of thapsigargin, demonstrating that SERCA expression is directly regulated by the ER  $Ca^{2+}$  level rather than by cytosolic  $Ca^{2+}$ . In contrast, PMCA expression changes were only consistent with a regulation whereby an increase in cytosolic  $Ca^{2+}$  enhanced PMCA expression. The three identified feedback loops are summarized in Fig. 3*F*. Thus, rather than solely relying on regulated coexpression (*i*) or on a feedback involving a single nodal point (*ii*), our third hypothesis is likely correct. Multiple parallel adaptive feedback loops sense ER and cytosolic  $Ca^{2+}$  concentrations and adjust the expression levels of STIM, SERCA, and PMCA.

Given the complexity of the nonlinear terms and feedbacks in the Ca<sup>2+</sup> signaling system, and the known difficulty in building fail-proof multifeedback systems (22), it was not obvious whether the parallel feedback loops we identified indeed improved the signaling performance. We tested this hypothesis computationally by replacing in the model from Fig. 1 fixed protein concentration parameters with regulated protein synthesis rates and constant degradation rates. We added a modulatory synthesis term that enhanced or reduced the expression rate of PMCA when the cytosolic  $Ca^{2+}$  level increased or decreased, respectively, from a setpoint  $Ca^{2+}$  level (50 nM). A similar term reduced or increased the expression rate of SERCA when the ER Ca<sup>2+</sup> level increased or decreased, respectively, from the ER setpoint (350  $\mu$ M Ca<sup>2+</sup>). Increases or decreases in either ER or cytosolic Ca<sup>2+</sup> levels from the setpoints reduced or increased, respectively, the synthesis rate of STIM. The protein degradation time determined the equilibration time of this system and was assumed to be 10 h. This timescale is much slower than the only minute-long receptor-triggered induction of typical Ca<sup>2+</sup> signals, and, thus, the concentration of signaling components did not significantly change during the signaling response itself.

Strikingly, the addition of these three expression feedback loops led to a marked reduction in the changes in basal cytosolic and ER Ca<sup>2+</sup> levels when the respective protein synthesis rates were doubled or cut in half. Fig. 4*A* shows a comparison of the simulated data without adaptive feedback (from Fig. 1*D*) to the simulated data with adaptive feedback. The twofold perturbations in SERCA, STIM, and PMCA concentrations were then stabilized by the system so that the resulting change in cytosolic and ER Ca<sup>2+</sup> levels were within 10% of the median basal level. The power of these adaptive feedback loops in preventing the cell from losing its control of basal cytosolic and ER Ca<sup>2+</sup> levels became particularly apparent when multiple perturbations were applied to the system (Fig. 4*A*, two rightmost bars in each plot).

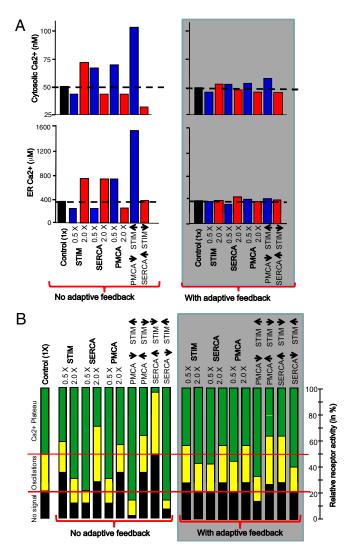
Furthermore, with adaptive feedback, the lowest receptor stimuli that triggered oscillations and plateau phases, respectively, now remained closer to those of control cells, preserving the range over which oscillations occurred (Fig. 4*B*). Thus, these multiple parallel and slow adaptive feedback processes allowed the Ca<sup>2+</sup> signaling system to tolerate significant perturbations and to maintain stable basal ER and cytosolic Ca<sup>2+</sup> levels, thereby preventing apoptosis. They also allowed receptor stimuli to trigger Ca<sup>2+</sup> responses with much less cell-to-cell variability, preventing failure of signal transmission.

Together, our findings identify a unique adaptive principle in the Ca<sup>2+</sup> signaling system whereby STIM, SERCA, PMCA, and possibly other signaling components are regulated by multiple, low-amplitude expression feedbacks so that increases or decreases in ER and cytosolic Ca<sup>2+</sup> levels slowly adjust the concentrations of key signaling pathway components. Our model calculations showed that parallel adaptive feedback is a powerful means to stabilize basal cytosolic and ER Ca<sup>2+</sup> levels and also to reduce the variability in the signaling response between cells. This control principle likely extends to other signaling systems and provides a general explanation of how cells reliably connect receptor inputs to cell function, despite large cell-to-cell variations in the expression of individual signaling components.

## **Materials and Methods**

Information about cell culturing and transfection, preparation of dsRNA, measurement of intracellular free calcium concentrations, and mass spectrometry sample preparation can be found in *SI Materials and Methods*.

Liquid Chromatography Mass Spectrometry (LC-MS) Setups. SRM analyses were performed by using two different LC-MS setups. One setup consisted of a hybrid triple quadropole/linear ion trap mass spectrometer (4000 QTRAP; Applied Biosystems/MDS Sciex). Chromatographic separations of peptides were performed on a Tempo nano LC system (Applied Biosystems) coupled to a 16-cm fused silica emitter, 75-µm diameter, packed with a Magic C18 AQ 5-mm resin (Michrom BioResources). Peptides (up to 3.5 µg of total protein digest) were separated with a linear gradient from 5% to 30% acetonitrile in 30 or 60 min, at a flow rate of 300 nL/min.



**Fig. 4.** Simulations show that including the identified parallel adaptive feedback loops into the model markedly enhances the reliability of the Ca<sup>2+</sup> signaling system in response to changes in the synthesis rates of signaling components. (A) Marked stabilization of both basal cytosolic and ER Ca<sup>2+</sup> levels (gray box). (B) Parallel adaptive feedback is well suited to preserve the setpoints where receptor stimuli first triggered Ca<sup>2+</sup> oscillations and first triggered a plateau phase. Instead of the marked reduction in the range over which oscillations occurred in the model without adaptive feedback (gray box) now shows minimal changes in the responsiveness to receptor stimulation.

For the second setup, a Proxeon electrospray ionization source was used as an interface between an EASY-nLC Nano-HPLC system (Proxeon) and a TSQ Vantage triple quadrupole MS system (Thermo Fisher Scientific). The peptide separation was carried out by using a 10-mm  $\times$  0.1-mm C18 trapping column (Proxeon C18; 5 µm, 120 Å) and a 21-cm  $\times$  75-µm diameter reverse-phase C18 capillary column (MICHROM; Magic C18; 5 µm, 200 Å). Peptides (up to 5 µg of total protein digest) were separated with a linear gradient from 0 to 45% acetonitrile in 71 min at a flow rate of 300 nL/min. The following mode and tuning parameters were used: polarity, positive; for scheduled SRM, max windows of 5 min and a cycle time of 1 min; Q1 and Q3 were set to 0.7 Th; scan time was set to 0.02s; scan width was set to 0.002 m/z; collision gas pressure was set to 1.5 mTorr; capillary was set to 270 °C.

For all MS runs, blank runs were performed regularly, in which the same set of transitions was monitored as in the following (sample) run. Blank runs were performed until no signal was detected for all transition traces, in particular, before any measurement of low-abundance proteins. Peptide and SRM Transition Selection. Proteotypic peptides and transitions (precursor/fragment ion pairs) were selected primarily by following procedures outlined in ref. 26. We also used inexpensive, crude synthetic peptides from AnaSpec (San Jose, CA) to directly determine how the peptides eluted from the HPLC and how they fragmented to be able to find the peptides in SRM mode. For each peptide of interest, we chose the three transitions with the best signal-to-noise ratio and optimized critical MS parameters (e.g., collision energy, declustering potential) for best sensitivity. The validated and optimized SRM transitions were used to detect and/or quantify the proteins in cell lysates using scheduled SRM mode. By restricting the acquisition of each transition to 2.5 min around its elution time, the timescheduling feature of the acquisition software enabled the analysis of the 130 transitions in a single run per sample with high sensitivity. SI Appendix, Table S1 shows the list of targeted peptides and transitions. Detailed information of how the peptides were quantified using SRM mass spectrometry can be found in SI Material and Methods.

**SRM-Based Quantification.** For quantitative analysis, a heavy-labeled synthetic version of each peptide of interest was custom-ordered from Thermo Scientific. In each peptide, the C-terminal K or R residue was substituted with the corresponding heavy version, resulting in a mass shift of +8 or +10 Da, respectively. The heavy peptides were isotopically labeled, meaning that they coeluted exactly with the endogenous peptides, and the peak areas could be directly ratioed. A known amount of heavy peptide was added to the protein mixtures just before trypsinization (27). The peptide transitions in heavy and light versions were measured by using scheduled SRM. SRM traces were exported as text files out of the Analyst 1.4 or Xcalibur 2.0.7 SP1 software and analyzed by using custom-written MATLAB routines.

Peak areas for the transitions associated with the heavy and light peptides were quantified. Each transition for a given peptide was treated as an independent abundance measurement. Absolute quantification was obtained from the ratio between the light and heavy SRM peak areas, multiplied by the known amount of the added standard. The potential contamination of the heavy peptide preparations with the corresponding unlabeled peptides was tested by injecting the heavy peptides alone and monitoring the transitions for both the heavy and light peptide forms. At the concentration used for quantitative measurements, no signal was detectable in the light transitions.

To carry out relative quantification of each protein during the different siRNA and chemical perturbations, the ratio between the light and heavy SRM peak areas was calculated and normalized by a correction factor. To calculate the correction factor for each sample, the concentrations of 15 abundant proteins that remained constant during the different perturbation conditions were measured by using SRM and averaged. This average was divided by the

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average concentration of all of the spiked-in heavy peptides to yield a correction factor by which to divide the respective sample.

Results were calculated as the means of the different transitions/peptide, peptides/protein, and replicate cultures  $\pm$  SE. For each SRM measurement shown in Fig. 3, 8–12 biological replicates were measured. Outlier transitions (e.g., shouldered transition traces or noisy transitions with S/n < 3) were not considered in the calculations.

**Computational Model.** Two models were used to simulate basal and receptorstimulated Ca<sup>2+</sup> signaling. The model used in Fig. 1 (model 1) had no adaptive feedback, whereas the adaptive model used for Fig. 4 (model 2) had three slow adaptive feedback loops that monitored basal ER and cytosolic Ca<sup>2+</sup> levels and made corresponding adjustments to the expression of SERCA, PMCA, and STIM/Orai.

The adaptive feedback loops for PMCA and SERCA expression included cooperativities of 4, a value that is typically observed for processes mediated by calmodulin and similar Ca<sup>2+</sup> effectors. Because STIM is regulated by both ER and cytosolic Ca<sup>2+</sup>, we used a cooperativity of 2 for each process. In the unperturbed state (1× concentration of proteins), both models had the same starting parameters of 0.05  $\mu$ M cytosolic Ca<sup>2+</sup> and 350  $\mu$ M ER Ca<sup>2+</sup> (corresponding to 2 relative units). After perturbations were applied (twofold increases or decreases in the level of PMCA, SERCA, or STIM in model 1 and twofold increases or decreases in the rate constant of synthesis of PMCA, SERCA, or STIM in the adaptive model, model 2), both models were run in the basal state (R = 0) until the system approached equilibrium (~300,000 s). This process led to changes in the steady-state ER and cytosolic Ca<sup>2+</sup> levels as shown in Figs. 1D and 4A. After equilibration, R was increased to simulate receptor concentration. The model was then run for 300 s (Fig. 4B). The range over which a first spike, oscillations, and the plateau phase were observed were empirically tested by iterative change of the parameters.

MATLAB Simbiology was used to program and run the model simulations. All model equations and constants are presented and described in *SI Appendix*. The models presented in this manuscript, without and with adaptive feedback, have been uploaded to the European Molecular Biology Laboratory–European Bioinformatics Institute BioModels database (http://www.ebi.ac.uk/biomodels-main/).

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