Dynamic balance between vesicle transport and microtubule growth enables neurite growth

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Abstract

Whole cell responses are complex because they involve many subcellular processes (SCPs) that function in a coordinated manner. Understanding of how different SCPs function together is essential for predictions of whole cell responses. We studied neurite outgrowth to understand how balance between SCPs is essential for this whole cell function. Neurite outgrowth involves three types of SCPs: production of membrane components, vesicle transport that delivers membrane to the growth cone and microtubule growth that regulates extension of the neurite shaft. Mathematical modeling and simulations show that redundancies between lower level sibling SCPs produce robustness of higher level SCPs. In contrast, higher level SCPs need to be strictly coordinated and cannot compensate for each other. From these models we predicted the effect of SCPs involved vesicle fusion as well as microtubule growth on neurite outgrowth. siRNA ablation experiments verified these predictions. We conclude that origins whole cell dynamics requires strict balance between the higher level SCPs involved.

Keywords: Vesicles transport, microtubule dynamics, neurite outgrowth, multi compartment model, ordinary differential equations

Introduction

Whole cell responses that involve, both morphological and physiological changes, are complex because they can engage many subcellular processes (SCPs). Such coordinated changes often lead to a change in cell state, such as moving to a more differentiated phenotype. Detailed understanding of how different SCPs collaborate to produce an integrated whole cell response requires experiments, along with mathematical models that capture the dynamics of individual SCPs as well as the interrelationship between the dynamics of multiple SCPs. Based on the biological detail that is represented by the SCPs they can be classified as Level-1, Level-2 and Level-3 SCPs. Higher level SCPs (i.e. SCPs with smaller numbers) represent a cell biological function that combines the functions of their lower level SCP children (Hansen et al., 2017),

Neurite outgrowth is an early event that changes the state of neurons (Arimura and Kaibuchi, 2007). This change in cell state allows neurons to mature by developing axons and dendritic trees that connect to other neurons and become electrically active. Many mathematical models of neurite outgrowth have been developed (Kiddie et al., 2005; Wissner-Gross et al., 2011) and these models have provided valuable insights into how complex biological processes can be modeled at various levels of description. Often these models abstract the details of the underlying mechanisms and consequently it is difficult to decipher how the balance between the various SCPs is achieved to generate whole cell responses. Some studies have focused on specific facets of the neurite outgrowth process such the role of signaling and regulatory motifs in determining how particular neurites are selected to become axons (Fivaz et al., 2008). Such models are useful in understanding the regulation of complex cellular responses.

To understand whole cell responses, it is necessary to delineate the quantitative coordination between the dynamics of the SCPs. In the case of neurite outgrowth, three Level-1 SCPs, *Production of Membrane Components, Vesicle Transport* and *Microtubule Growth* need to be coordinated. Membrane components such as membrane lipids and proteins are synthesized in the cell body and transported to the growth cone via vesicle transport. Vesicles that fuse with the growth cone plasma membrane deliver their membrane lipids to the neurite tip, causing the neurite shaft to grow. Microtubules form the basis of the neurite scaffold that needs to extent in coordination with the neurite shaft. Tsaneva-Atanasova et al (Tsaneva-Atanasova et al., 2009) developed a model of the interaction between microtubules and membrane vesicles in regulating neurite outgrowth. The framework of this model serves as a useful starting point for more extensive models where we can consider how a group of distinct but related SCPs such as *vesicle budding, vesicle transport* and *vesicle fusion* along with a second group of SCPs that involves *microtubule nucleation, microtubule growth* and *conversion of dynamic to stable microtubules* interact to enable the growth of neurites. Such detailed models could serve as basis for understanding how a single network can regulate the relationship between kinetic parameters and changes in the levels or activities of molecular components within the different SCPs. To develop a detailed model of neurite outgrowth based on interactions between SCPs of different levels, we constructed a multi compartment ordinary differential equation model to extend and integrate the microtubule growth

model developed by Margolin et al. (Margolin et al., 2012) with the vesicle transport model by Heinrich and Rapoport (Heinrich and Rapoport, 2005). We also incorporate more recent experimental data to make the model contemporary.

We have made a critical change from our previous approaches to dynamical modeling in developing this model. We had typically used a bottom-up approach where individual reactions are assembled into modules that have emergent behaviors (Azeloglu et al., 2014; Bhalla and Iyengar, 1999). In contrast, the approach we use here can be called top-down as we made a starting postulate that the three Level-1 SCPs involved in neurite outgrowth need to be appropriately coordinated to produce a steady velocity of neurite outgrowth. We start with experimental observations of neurite outgrowth velocities and use numerical simulations to determine how their Level-2 children and Level-3 grandchildren SCPs need to be coordinated to allow neurite outgrowth with the observed velocities and under kinetic parameter constellations that match additional experimental data. Our simulations show redundancy between Level-3 sibling SCP activities. Any loss of function in one Level-3 sibling SCP can be compensated by a gain of function in another. However, such generalized reciprocal relationships do not exist between Level-2 SCPs. At these higher level SCPs, change in activity in one SCP cannot be compensated by another similar level SCP and consequently we see loss of whole cell function. We used the findings from analyses of the models to develop predictions regarding effects of modulating higher SCPs on whole cell function We tested the predictions experimentally and found them to be valid.

Results

Top-down based SCP modeling of whole cell dynamics

Our approach to model whole cell responses in a top-down based manner started with the literature curation of SCPs involved and their assignment to different levels of biological detail (Figure 1, Step 1a). Like the organization of our Molecular Biology of the Cell Ontology ((Hansen et al., 2017), Level-1 and Level-2 SCPs describe more general, Level-3 SCPs more specific cell biological or biochemical functions. SCPs of the different levels are in child-parent relationships. SCPs are represented as coupled biochemical reactions that are transformed into coupled ordinary differential equations (ODEs) (Figure 1, Steps 1b/c). We use experimental data to define model constraints that the simulations must match (Figure 1, Step 1d). For systematic analyses we develop analytical solutions for the prediction of kinetic parameters of SCPs at steady state (Figure 1, Step 2a). Here, we use the term steady-state to describe a continuous whole cell response without variation of amplitude. We use the analytical solutions to predict steady-state kinetic parameters and values that match the predefined model constraints (Figure 1, Step 2b). Through systematic parameter variation, we generate multiple steady-state solutions that allow the classification of dependencies between SCPs at different levels (Figure 1, Step 2c). To investigate the effect of SCP dysfunction on whole-cell responses, we generate numerical simulations based on steady-state parameter sets after arbitrary modification of kinetic parameters of a single SCP followed by experimental verification (Figure 1, Step 2d).

Integration of vesicle production & transport and microtubule dynamics to model neurite outgrowth

Among the major sub-cellular functions that are involved in neurite outgrowth (NOG) are the Level-1 SCPs *Production of Membrane Components*, *Vesicle Transport* and *Microtubule Growth* (Figure 2A). Each of these SCPs is composed of multiple Level-2 children SCPs whose activities need to be coordinated to ensure NOG without complication. Since our major focus was on vesicle transport and microtubule growth, we further categorized Level-2 SCP function by introducing their Level-3 children SCP. SCPs with the same parent SCPs are defined to be siblings. It is important to note that at this stage our model does not consider local actin dynamics at the growth cone, since we particularly focus on SCPs that are involved in extension of the neurite shaft; i.e. membrane lipid and protein synthesis and delivery and microtubule scaffold growth.

The level-1 SCP *Production of Membrane Components* has two Level-2 children SCPs, *membrane lipid production* and *membrane protein production*, that are incorporated into our model via constant synthesis rates of the individual components at the Trans-Golgi Network (TGN) (Figure 2B; Reaction 1 in Table S1) (Gracias et al., 2014; Wang et al., 2011).

The Level-2 children SCPs of the Level-1 SCP *Vesicle Transport* refer to *Vesicle Budding* at the TGN and *Growth Cone Plasma Membrane* (GC-PM), anterograde and retrograde microtubule-based vesicle transport and vesicle fusion with the TGN and GC-PM. Each of these six Level-2 SCPs has two Level-3 children SCPs. To simulate the dynamics of the vesicle transport SCPs we extend a dynamical model of bidirectional membrane lipid and protein transport between the endoplasmic reticulum and the cis-Golgi (Heinrich and Rapoport, 2005). In our model the two organelles are the TGN and the GC-PM (Figure 2B). The level-2 SCP *Vesicle budding at the TGN* consists of the Level-2 SCPs *Coat protein recruitment and formation at the TGN* and *Vesicle invagination and scission at TGN*. The former is mediated by recruitment Factor 1 that resides at the TGN, the latter is mediated by a cell body specific budding rate (Reaction 2 in Table S1). Membrane proteins at the TGN are recruited into the budding vesicle via association with the coat proteins (Table S2). The Level-2 SCP *Anterograde microtubule- based vesicle transport* that moves vesicles from the cell body cytoplasm into the neurite shaft cytoplasm and from there into the growth cone cytoplasm involves the Level-3 SCPs *Kinesin recruitment to vesicle* and *Kinesin mediated vesicle transport along microtubule*. Kinesin motor proteins are recruited to the vesicle via kinesin receptors that are part of the vesicle membrane. Here, we assume that kinesin is available under saturating conditions, so that the kinesin concentration per vesicle is determined by the number of kinesin receptors. Based on the number of kinesin bound kinesin receptors we calculate the likelihood that a vesicle is bound to the microtubule (Table S1, Reactions 6/7). This likelihood also depends on the affinity of kinesin to the microtubule, incorporated into our model as the fraction of Microtubule(MT)-bound kinesin. The fraction of MT-bound kinesin decreases from the cell body cytoplasm to the neurite shaft cytoplasm and growth cone cytoplasm (Table S3). A low fraction of MT-bound kinesin in the neurite shaft cytoplasm ensures that only a few vesicles (~9%) are moving forward, in agreement with experimental data (Ahmed and Saif, 2014).

The large pool of stationary vesicles in the neurite shaft cytoplasm could serve as a membrane reservoir that allows quick recruitment of additional membrane to the GC-PM on demand. In the growth cone cytoplasm almost all vesicles are unbound, allowing their fusion with the GC-PM.

The Level-2 SCP *Vesicle exocytosis* with GC-PM is mediated by the interplay of the Level-3 SCPs *Vesicle tethering* and *Vesicle fusion at* GC-PM. The former is realized in our model via a growth cone specific tethering rate, the latter via complex formation between vesicle(v)-SNARE V and target(t)-SNARE Y that resides at the GC-PM (Table S1, Reaction 8). Retrograde vesicle transport from the GC-PM to the TGN is incorporated by a similar set of Level-2 and Level-3 SCPs that consist of a symmetric set of membrane proteins and rate constants. We assume a high fraction of dynein is attached to the MT in the neurite shaft cytoplasm, so that most of the retrograde vesicles are actively transported towards the TGN, without formation of a vesicle reservoir.

Vesicles that fuse with the GC-PM, increase its surface area. In our model, we set the GC-PM surface area to 50 μm^2 (Kunda et al., 2001; Ren and Suter, 2016). Any membrane surface area that is added to the GC-PM and would increase its size beyond 50 μm^2 is added to the neurite shaft (Table S1, Reaction 9). Membrane proteins are not added to the neurite shaft, since our model is based on the assumption that the diffusion of the membrane proteins into the neurite shaft is prevented by intra-membranous diffusion barriers or cortical cytoskeleton proteins (Ye and Zhang, 2013) that ensure a highly specialized growth cone plasma membrane. Neurite shaft length arises from its membrane surface area (Table S1, Reaction 10). For simplicity, we do not consider increase of the neurite diameter at this stage. The neurite shaft cytoplasm lies within the neurite shaft and grows with the lengthening neurite shaft. Most of the anterograde vesicles in the neurite shaft cytoplasm are not actively transported along the microtubule but are stationary components of the membrane reservoir. Consequently, the growing neurite shaft cytoplasm acts as a sink for vesicles and vesicle membrane proteins.

Microtubule (MT) growth is mediated by the interplay of the Level-2 SCPs *Dynamic microtubule growth* and *Stable microtubule growth*. (Brady et al., 1984; Kollins et al., 2009; Song et al., 2013; Witte et al., 2008). Dynamic MTs are characterized by periods of MT growth and catastrophic breakdown, the latter can either lead to MT disappearance or rescue, followed by a new growth period. The time frame of these periods is in seconds or minutes. Stable MTs do not show such periodic growth behavior and we assume that there is no degradation of stable MTs. Dynamic MTs are continuously generated at a specified nucleation rate (Level-3 SCP *Dynamic MT nucleation*) (Reaction 11 in Table S4, Figure 2C). The length of dynamic MTs (Reaction 12, Table S4) as well as the rate of MT catastrophic breakdown depends on the free tubulin concentration that is regulated by the SCP *Tubulin sequestration* (Reaction 13 Table S4) (Curmi et al., 1997; Jourdain et al., 1997; Manna et al., 2009) and on the hydrolysis rate of tubulin-bound GTP (SCP *GTP hydrolysis*). The latter SCP determines the size of the GTP cap at the tip of the dynamic MT that protects it from catastrophic breakdown (Seetapun et al., 2012). To characterize the coordinated activities of these Level-3 SCPs, we ran an initial set of simulations of the growth of a single

dynamic MT using a stochastic growth model (Margolin et al., 2012) (Figure S1). Since we simulate NOG over hours and days, while dynamic MT growth behavior changes in seconds and minutes, we do not simulate individual dynamic MTs, but the whole population of dynamic MTs. For this population simulation, we developed two formulas based on the results of the Margolin model. They describe the dependence of the average growth length of dynamic MTs and of the catastrophic breakdown rate that leads to complete disappearance of the dynamic MT on the effective tubulin concentration and the GTP hydrolysis rate (Reactions 12/13 in Table S4; Figure S1). The count of dynamic MTs depends on the nucleation rate, the degradation rate due to catastrophic breakdown and the conversion rate into stable MTs (Reaction 14 in Table S4; Figure S1). The activity of the Level-2 SCP *Stable microtubule growth* is regulated by the Level-3 SCP *MT stabilization* (Table S4, Reaction 15). To calculate the length of the microtubule scaffold within the neurite, we sum the length of all dynamic and stable MTs and divide it by the number of MTs per neurite cross-section (Reaction 16/17 in Table S4).

Experimental determination of varying neurite growth rate velocities

In primary cultures, neurons put out neurites at varying rates. To quantify the variability and determine the velocities with which neurites grow, we used a high content imaging system (IN Cell Analyzer) to track neurite outgrowth in primary rat cortical neurons. The experimentally observed velocities of neurite outgrowth were used a top-down constraint for our model. We incubated rat cortical neurons for 16h and then documented neurite outgrowth every 6h for an additional 48h. For each neuron, we quantified the length of its longest neurite (the one most likely to become an axon). Representative images are shown in Figure 3A and 3B, and a complete set of images is shown in Supplementary Figure 2. At each time point the neurites significantly differed in their lengths (Figure 3B). We identified the longest neurite for each neuron in a field (Supplementary Figure 2) and calculated their length distribution at every time point (Figure 3C), identified the length of selected quantiles and calculated the growth velocity for these quantiles (Figure 3D). The neurite growth velocities ranged from 0 to ~20 µm/h. We used these velocities as a top-down whole- cell response constraint in modeling the activities of the Level-1, -2 and -3 SCPs. Increase in the whole-cell response, i.e. neurite outgrowth velocity, depends on the coordinated increase of membrane production and delivery of membrane vesicles to the growth cone as well as microtubule growth. Both types of SCPs need to facilitate growth with the same velocity, since they are interdependent. Vesicle transport is microtubule dependent and microtubule growth is limited by the length of the neurite shaft. Similarly, the increase in membrane production needs to be coordinated with increased vesicle transport. More vesicles would need to bud from the TGN, to be actively transported through the neurite shaft cytoplasm and to fuse with the GC-PM. These interdependencies form the basis for our top-down constraint.

Postulates governing SCP dynamics

In addition to the experimentally determined NOG velocities, we used experimental data from the literature for additional constraints at the level of SCPs. These constraints arise from a set of

postulates of permissible SCP functions that are described in (Table S5). We postulate that about ~10% of the anterograde vesicles in the neurite shaft cytoplasm are bound to the microtubule and actively moving along the microtubule (Ahmed and Saif, 2014). The other 90% constitute a membrane pool that can be recruited on demand to adapt to short-term increases in outgrowth velocity. Similarly, we assume that the vesicles that reside in the growth cone (Erturk et al., 2007) are the first membrane reservoir for the quick recruitment of additional membrane and also define - in the same way - that 10% of the anterograde vesicles in the growth cone cytoplasm move, i.e. fuse with the growth cone membrane. This assumption is based on the experimental observation that GC membrane precursor vesicles reside in the GC cytoplasm before fusion with the membrane with a half-life of ≥14min (Pfenninger et al., 2003), which would suggest that around 5% of the vesicles fuse with the membrane. Similarly, we assume that 90% of the retrograde vesicles in the neurite shaft cytoplasm and the cell body cytoplasm are moving. The total growth cone membrane is completely internalized within 30 - 60 min (Diefenbach et al., 1999) which an endocytosis rate between 0.833 μm^2/min - 1.66 μm^2/min for a growth cone with a surface area of 50 μm^2 (Kunda et al., 2001; Ren and Suter, 2016). Based on these observations and the assumption that some of the endocytosed vesicles might be part of transcytotic re-arrangements during growth cone steering (Tojima and Kamiguchi, 2015) or are transported to other growth cones and not to the cell body (Denburg et al., 2005), we set the rate of membrane back transport from the growth cone to the TGN to 0.5 μm^2/min. The backward transported membrane can also be called cycling membrane, since it continuously cycles between the TGN and the GC-PM without being added to the neurite shaft. For the microtubules we postulate that dynamic MTs should fill up 20 μm of the growing neurite. This is the length of the minor processes (Yu and Baas, 1994) that are characterized by continuous retraction and extension periods (Arimura and Kaibuchi, 2007). We assume that such periods are committed by continuous MT growth and retraction, so that the MT scaffold in these processes should mainly consist of dynamic MTs.

Using these constraints, we ran simulations to identify an initial solution at one intermediate rate of neurite outgrowth. From these initial simulations, we developed an overall analytical approach that allows us to predict the relationships that produce neurite outgrowth at a specified velocity.

Simulations and Analytical Analyses

We identified a set of parameters that generated neurite outgrowth with a velocity of 10 μm/h (Figure 4 and S3). Under steady-state conditions neurite, surface area continuously increased at a rate of 0.5236 μm^2/min in agreement with the amount of membrane that is needed to increase the length of the growing neurite (Figure 4A). The length of the microtubule bundles also increase proportionally to maintain the shaft growth rate (Figure 4B). For this we selected an effective tubulin concentration of 9 μM to achieve a length increase of the microtubule bundles with a velocity of 10μm/h. The vesicles in the neurite shaft cytoplasm increased in parallel with a rate of ~0.57 vesicles/min assuming a membrane area of 0.05 μm^2 per vesicle (Figure 4C). In considering the values of the ordinate for the different SCPs (Figure 4) it became clear that there are multiple

combinations of kinetic parameters and amounts of components that allow neurite outgrowth at the same velocity without violating the constraints we imposed on dynamics of the SCPs.

To enable a systematic analysis of the redundancies within and between SCPs, we generated an analytical solution (Figure 5, and S4) for the prediction of kinetic parameters at steady state that enable neurite outgrowth at a given velocity without violating constraints on SCP dynamics (Table S5). In our solution, we consider that an increase in neurite outgrowth depends on varying sets of protein amounts and kinetic parameters that allow increased net anterograde transport of membrane vesicles and continuous synthesis of new proteins and membrane to compensate for the accumulation within the growing neurite shaft cytoplasm. In addition to the postulated constraints on SCP dynamics, we pre-define independent protein amounts and kinetic parameters (Figure S4, right side) to calculate the dependent protein amounts and kinetic parameters. To achieve a given NOG velocity a certain amount of lipid membrane needs to be synthesized and transported from the TGN to the GC. In addition to this membrane that will be added to the neurite shaft three other membrane types can be classified based on their transport destination, allowing us to distinguish four membrane types (Figure 5A). These membrane types should not be confused with the four-different vesicle sets that transport the membrane between the TGN and GC-PM. Type I membrane buds from the TGN and is added to the vesicle reservoir in the growing neurite shaft cytoplasm. Type II membrane is transported from the TGN to the GC-PM and added to the neurite shaft as anterograde vesicles. Type III membrane is transported from the TGN to the GC-PM, endocytosed at GC-PM and added to the NSC as retrograde vesicles. Type IV membrane is transported from TGN to GC-PM and from there back to TGN, and therefore is labeled cycling membrane. Initial and final anterograde and retrograde rates that describe budding and fusion processes at the TGN and GC-PM can be calculated by summing up the involved membrane types. The fluxes of the individual membrane types can be calculated based on the anticipated NOG velocity and model constraints (Figure 5B, (1)). To allow the fusion of sufficient anterograde vesicles with the GC-PM, a certain number of SNARE complexes need to be formed between v-SNAREs V and t-SNAREs Y (Figure 5B, (2)). Consideration of the tethering rate allows the calculation of the needed amount and concentration of v-SNARE Vs. The concentration allows the determination of the initial and final forward fluxes of SNARE V ((Figure 5B, (3)). Similarly, concentrations and fluxes can be calculated for the v-SNARE U that mediates vesicle fusion at the TGN ((Figure 5B, (4)/ (5)). During vesicle budding at the TGN and GC-PM, SNAREs compete with each other for the vesicle binding spots. Knowledge of the fluxes of both v-SNAREs allows the calculation of their concentrations at the TGN and GC-PM (Figure 5, (6)/ (7)). The production rates of v-SNARES (and any other membrane vesicle membrane proteins) can be calculated based on their vesicle concentrations and the number of vesicles that accumulate in the growing neurite shaft cytoplasm (Figure 5B, (8)).

MT growth at a given velocity (Figure 5C) depends on the conversion rate of dynamic into stable MTs (Figure 5D, (9)). By pre-simulation results using the model of Margolin et al, we obtained the dependence of the average length of dynamic MTs on the free (i.e. not sequestered) tubulin

concentration and GTP hydrolysis rate. The average length allows the calculation of the number of dynamic MTs (Figure 5D, (10)), followed by calculation of the stabilization rate (Figure 5D, (11)) and subsequently the nucleation rate (Figure 5D, (12)).

The analytical solution was validated by comparing the analytically predicted outcome with the results of the numerical simulation for example sets of parameter and protein amounts (Figure 5E and S5). We used the predicted parameter and protein amount sets as starting points in our model and ran the numerical simulation for 5000 minutes (~3.5 days). Every 500 min we compared the anticipated model constraints with the numerically obtained values to estimate the accuracy of our analytical prediction. The analytical solution matched the anticipated model constraints with high accuracy, except the fusion rate for retrograde vesicles with the TGN that for was too high for the velocities 0 and 2.5 μm/h.

Redundancy among Level-3 sibling SCPs, but not among Level-2 sibling SCPs

We used the analytical solution to systematically analyze the relationship between different SCPs with a focus on the interaction between Level-2 and Level-3 SCPs (Figure 2A). For our studies we selected growth velocities that covered the whole range of experimentally observed velocities (Fig 6). We predicted that Level-3 SCPs were characterized by complementary activities. An increase in the activity of one Level-3 SCP is associated with a decrease in the activity of at least one of its sibling Level-3 SCPs. To acquire a certain NOG velocity without violation of the model constraints there are multiple combinations of the activities of the two Level-3 SCPs *Coat recruitment and formation* and *Vesicle invagination and scission* (Figure 6A). Such SCP redundancies are also observed between the Level-3 SCPs *Kinesin recruitment to the vesicle* and *Kinesin mediated vesicle transport along the MT* (Figure 6B) that define the activity of the Level-2 SCP *Anterograde microtubule-based vesicle transport*. Similarly, the SCPs *Vesicle tethering* and *Vesicle fusion* (Figure 6C), sub processes of the SCP *Vesicle exocytosis* are characterized by an interaction that is based on an inverse relationship between their activities.

In contrast, activities of Level-2 SCPs are directly related to each other, such that loss of activity of one SCP cannot be compensated by an activity gain of another. We modified activity for a Level-2 SCP by changing the activity of one of its Level-3 children SCPs without compensatory adaptation of the activity of its other Level-3 children SCPs. Level-2 SCP activity failures cannot be reversed by activity changes of any other Level-2 SCP. An insufficient budding rate at the TGN increases TGN size to non-physiological levels at any velocity (Figure 6D), while a reduced forward transport of vesicles along the MT increases the percentage of stationary vesicles, thereby overloading the neurite shaft cytoplasm (Figure 6E). Similarly, reduced vesicle exocytosis increases the percentage of stationary vesicles in the GC cytoplasm also causing an overload with vesicles (Figure 6F). Such violations of the model constraints cannot be compensated by any modification of another Level-2 vesicle transport SCPs. although our dynamic model does not consider the effect of imbalances that result from the violation of the model constraints on NOG,

we postulate that such chaotic cellular states interfere with the ability of the neuron to grow a neurite in a physiological manner.

The balance of the Level-3 SCP activities to fulfill the Level-2 SCP function is not an isolated decision but influences the activity of other SCPs as well. The more the activity of a Level-2 SCP arises from a Level-3 SCP that involves vesicle proteins, the higher the need for continuous membrane protein supply to compensate for the consumption of membrane proteins in the neurite shaft cytoplasm sink. This accounts for the kinesin receptor (Figure 6E) as well as for the v-SNARE V (Figure 6F). Here, one might consider that the complementary SCP *Kinesin-mediated vesicle transport along the MT* is determined by components along the whole microtubule while the complementary SCP *Vesicle tethering* at GC-PM is determined by components of the tethering machinery that are locally restricted to the growth cone. The maintenance of the activity of the former SCP might therefore depend on a continuous synthesis of cytoplasmic protein that is not considered in our model.

The same relationships exist between Level-2 and Level-3 SCPs involved in Microtubule Growth. The combined length of dynamic MTs that are available for conversion into stable MTs is regulated by an inverse relation between the activities of the Level-3 SCPs *Dynamic MT nucleation*, *GTP hydrolysis* and *Tubulin Sequestration*. The average dynamic MT length (Figure 7B) as well as the degradation rate of dynamic MTs (Figure 7C) depends upon the Level-3 SCPs *Tubulin sequestration* (which regulates effective tubulin concentration) and *GTP hydrolysis*. The balanced activities of these two Level-3 SCPs and the Level-3 SCP Dynamic microtubule nucleation (Figure 7D/E) determines the activity of the Level-2 SCP *Dynamic microtubule growth*. Similarly, to the vesicle transport Level-3 SCPs, any loss of function in one of these MT Level-3 SCPs can be compensated by a gain of function in one of the other. In contrast, reduced activity of the Level-2 SCP *Dynamic microtubule growth* inevitably causes reduced *MT stabilization* and NOG growth.

After prediction of SCP dynamics under physiological conditions we focused on the effects of SCP deficiencies on NOG. We selected one *Vesicle transport* SCP and one *Microtubule Growth* SCP for further investigation. To simulate the uncoordinated activity of a Level-2 SCP on NOG, we first predicted sets of kinetic parameters that allow growth under physiological conditions as defined by our model constraints. Then we selected one of its Level-3 SCP children, followed by the perturbation of one kinetic parameter that belonged to this Level-3 SCP, so that this parameter no longer matched anymore to the predicted parameter set. The resulting unbalanced activities between the redundant Level-3 SCP siblings propagated to impaired Level-2 SCP activity that did not match with the activities of the other Level-2 SCPs. To simulate the effect of the uncoordinated activity of the Level-2 SCP vesicle exocytosis at GC-PM, we selected the v-SNARE V that is involved in the activity of its Level-2 child SCP *Vesicle fusion at the growth cone*. We predicted that a reduction in its activity v-SNARE V activity does not influence NOG velocity, but overloads the growth cone with anterograde vesicles, thereby violating our model constraints (Figure 7A/B).

To test our predictions, we selected the v-SNARE VAMP7 that is involved in vesicle fusion at the GC-PM (Burgo et al., 2012). In agreement with our predictions, VAMP7 knock down (KD) did not significantly decrease NOG velocity, as shown by the total length of neurites 48h after axotonomy (Figure 7C/D) but caused the generation of dystrophic bulbs (abnormally large growth cones) (Figure 7C/E). To simulate impaired activity of the Level-2 SCP *Stable microtubule growth* we selected its Level-3 SCP child *Microtubule stabilization* and perturbed the kinetic parameter stabilization rate in a similar manner as described for the v-SNARE V. Our simulations predicted a significant decrease in the total MTB length that affects NOG (Figure 7F). We validated our predictions by knocking down the microtubule crosslinking protein MTCL1 that mediates MT stabilization (Satake et al., 2017). MTCL1 KO significantly reduced NOG velocity and increased the number of dystrophic bulbs (Figure 7G/H/I).

Discussion

Neurite outgrowth depends on the coordinated activities of various sub-cellular processes (SCPs). Of the different SCPs involved, the types of Level-1 SCPs are critical. Three major sub-cellular functions that are covered by these SCPs are *Production of Membrane Components, Vesicle Transport and Microtubule Growth.* We curated SCPs that are involved in these Level-1 sub-cellular functions and classified them as Level-2 and Level-3 SCPs that in parent child relationships. Based on the hierarchical organization of the SCPs, we developed a dynamical model that incorporates the three types of Level-1 SCPs and documented that the activity of the SCPs involved need to be highly coordinated to allow neurite outgrowth at a certain velocity. To make the model more manageable, our simulations were done under several assumptions: (1) the size of the TGN & GC-PM (50 μm^2) is independent of the outgrowth velocity; (2) that the rate of back transported membrane from the growth cone to the TGN is 0.5 μm^2/min; (3) that 90 % of the anterograde vesicles in the neurite shaft cytoplasm and the growth cone cytoplasm serve as membrane reservoir (i.e. these vesicles are not actively transported along the MT or fusing with the GC-PM) and (4) 20 μm of the neurite microtubule scaffold consists of dynamic MTs and the rest of stable MTs. All these assumptions are based on experimental data and allow us to identify the key features of the balance between SCP types required for a dynamic whole cell response. We developed an analytical prediction for steady state dynamics at anticipated NOG velocities that allowed us to investigate SCP relationships in a systematic manner. We could show that the needed activities of Level-3 SCP siblings to generate the function of their Level-2 SCP parent inversely depends on each other. This applied to both Level-3 SCPs involved in *vesicular transport* as well as Level-3 SCPs involved in *microtubule growth*. In contrast, the activities of Level-2 SCPs must be highly coordinated to maintain the Level-1 SCP function. The impaired activity of one Level-2 SCP cannot be simply compensated by an increased activity of another Level-2 SCP. Dependencies between Level-2 SCPs are more complex. In case of *vesicle transport*, loss of function of one Level-2 SCP might only be compensated, if all other Level-2 SCPs also reduce their function. Such as a constellation might ensure the same anterograde membrane net transport under a reduced rate of membrane that cycles back to the TGN. Impaired Level-2 SCP activities

were simulated by reduction of the activity of one of its level-3 children SCPs without compensatory activity modification of the other. Our predictions proposed that reduced exocytosis does not influence NOG velocity but causes an overload of the GC with anterograde vesicles. In agreement with our prediction, knockdown of v-SNARE Vamp7 did not change NOG velocity but rather. It creates a pathological phenotype with multiple dystrophic bulbs instead of healthy growth cones. A reduced Stable MT growth was predicted to cause a reduction in NOG, and it was confirmed by knock down of MTCL1.

The characterization of the relationships between Level-3 SCP siblings as redundant and Level-2 SCP siblings as essential for NOG have also been reported for multiple metabolic pathways that generate the same metabolite via different enzymatic reactions. The observation of additional examples for such an organization of SCP dependencies suggest that this might be more general phenomenon in cell biology and should be investigated for other cell systems as well.

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Author contributions: R.I. and J.H. conceived the project. A.Y. and J.H. developed and implemented the model. J.H. developed the analytical prediction for steady state dynamics. M.S, V.R. and R.T. prepared rat cortical neurons. A.Y. and M.S. analyzed the images. A.Y., R.I. and J.H. wrote the manuscript.

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Figure Legends

Figure 1: Top-down based SCP modelling of whole cell dynamics. Our pipeline for the generation of dynamical models that allow the analysis of relationships between sub-cellular processes (SCPs) during whole cell responses consists of two major steps. First, we build multicompartmental ODE models for the simulation of SCP activities that are based on literature curated data (Step 1a-d). In the second step we develop analytical predictions for kinetic parameters at steady state that allow a systematic analysis of SCP interactions (Step 2a-d).

Figure 2: Identification of different level SCPs involved in neurite outgrowth and their incorporation into multicompartmental ODE model. **(A)** The three major SCP types, *Production of membrane components, Vesicle transport and Microtubule growth*, describe essential sub-cellular functions involved in neurite outgrowth. We labeled them as level-1 SCPs and identified their level-2 children and level-3 grandchildren SCPs. **(B)** To model membrane production at the trans golgi network (TGN) and delivery to the growth cone plasma membrane (GC-PM) via vesicle transport, we extended a dynamical model that simulates vesicle transport between the endoplasmic reticulum and the Golgi (Heinrich and Rapoport 2005). In our version, new synthesized membrane is added to the trans golgi network from where it is transported to the growth cone plasma membrane by vesicular transport. Vesicles bud from the TGN, move along the microtubules via active kinesin mediated transport and fuse with the growth cone plasma membrane, leading to neurite shaft growth. The vesicles pass through three intermediate compartments: the cell body cytoplasm (CBC), the neurite shaft cytoplasm (NSC) and the growth cone cytoplasm (GCC). In each compartment kinesin has a different affinity for microtubules, leading to varying fractions of vesicles that are actively transported via kinesin along the MT. Retrograde vesicles are generated via endocytosis at the growth cone plasma membrane and move along the microtubule towards TGN through dynein mediated active transport. Vesicle budding at the TGN or GC-PM is mediated by the interaction of recruitment factors and coat proteins, and by vesicle fusion with the TGN or the growth cone membrane by the formation of SNARE complexes between vesicle(v)-SNAREs and target(t)-SNAREs, catalyzed by local tethering machineries. Motor proteins are bound to vesicles through motor protein receptors. A_G/B_G label vesicles that bud from the TGN with coat protein A/B, A_{PM}/B_{PM} vesicles that bud from the GC-PM with coat protein A/B. Since at steady state almost all anterograde vesicles bud with coat protein B from the TGN, and almost all retrograde vesicles bud with coat protein A from the GC-PM, we highlighted their transport routes in brighter colors. **(C)** To simulate microtubule growth, we consider the two different pools of microtubules, stable and dynamic microtubules. After nucleation new MTs are added to the pool of dynamic MTs that is characterized by alternating phases of growth and catastrophic breakdown. The frequency and duration of these phases depend on the tubulin concentration (regulated by the SCP *tubulin sequestration*) and *the GTP hydrolysis rate*. Consequently, the length distribution of the dynamic MT pool and the degradation rate of dynamic

MTs depends on the activity of both SCPs (supplementary figure 2). Dynamic MTs are either degraded or converted into stable MTs that form the growing neurite scaffold.

Figure 3: Neurites grow with different outgrowth velocities. Neurons were dissected from rat cortical brain, incubated for 16h to allow initial growth, and followed by image acquisition every 6h up to 70h after plating. **(A-B)** The growth dynamics of two example neurons, one with a low (A) and one with a high (B) outgrowth velocity, are shown over given time period. **(C)** At each timepoint we quantified the length of the longest neurite of each neuron. Histograms show the length distribution of all longest neurites at the indicated timepoints. **(D)** At each timepoint we identified the top and bottom 10% and 25% quantile length as well as the median length. Linear interpolation of the lengths for each selected quantile documented outgrowth velocities ranging from 0 to ~20 µm/h.

Figure 4: Simulation of vesicle transport and microtubule growth during neurite outgrowth. **(A-B)** We identified a set of kinetic parameters (Supplementary table S7) that allows neurite outgrowth via coordinated neurite shaft growth (A) and microtubule bundle growth (B) with a velocity of 10 µm/h. **(C)** The growing neurite shaft cytoplasm (NSC) within the growing neurite acts as a sink for anterograde and retrograde vesicles, causing an accumulation of vesicles within this compartment. **(D)** Our identified set of kinetic parameters allows NOG without violation of the model constraints that we selected to define physiological outgrowth conditions.

Figure 5: Generation of an analytical solution for the prediction of kinetic parameters at steady state. **(A)** The development of the analytical solution depends on the categorization of membrane fluxes based on their destination compartment. Such classification distinguishes four different membrane types (I-IV) (that should not be confused with the four different vesicle types we show in figure 2B). Initial and final anterograde and retrograde transport rates (which refer to budding and fusion rates at the TGN and GC) are the sum of the transport rates of the encircled membranes types. **(B)** Calculation of kinetic parameter for v-SNARE V. VV/YY/UU refer to total count of SNAREs, V/Y/U describe concentration of SNAREs (per membrane area). See supplementary methods for complete derivation of analytical solution. **(C)** Microtubule bundle (MTB) growth is mediated by the conversion of dynamic MTs into stable MTs. Dynamic MTs nucleate de-novo, followed by their degradation or stabilization. **(D)** Nucleation, stabilization and degradation rates for dynamic MTs are calculated as shown. **(E)** To verify our analytical solution we analyzed, whether, our predicted parameter sets allow neurite outgrowth with the selected velocity and without violation of model constraints. For each velocity, we selected 6 different parameter sets that differ in the number of v-SNARE V and the tethering rate at the growth cone plasma membrane (supplementary figure 5A). Every 500 min over a simulation period of 5,000 min (~3.5days), we documented the indicated outputs. Averages and standard deviations were calculated for each velocity. As shown for a velocity of 15 µm/h, our analytical solution predicts steady state dynamics that match the anticipated model constraints (horizontal lines) with high accuracy.

Figure 6: Systematic analysis of vesicle transport SCP-relationships reveals redundancies among level-3 sibling SCPs, but not among their level-2 parent SCPs. To analyze the relationships between level-3 and level-2 SCPs, we used our analytical solution to predict steady state dynamics that enable neurite outgrowth for specified velocities. **(A), (B) & (C)** Level-3 SCPs show complementary relationships between their activities to generate higher level SCP function without violation of the model constraints. **(A)** Different activities of the SCPs *Coat recruitment and formation* at TGN and *Vesicle invagination and scission* at TGN were modeled by changing the amount of recruitment factor 1 and the budding rate at the TGN. SCP activities were inversely related and increased with increasing velocity. Line colors indicate selected velocities as shown in (G). **(B)** A similar SCP redundancy was observed for the SCPs *Kinesin recruitment to vesicle* (simulated by # number of kinesin receptors) and *Kinesin mediated vesicle transport along the MT* (simulated by fraction of bound kinesin). **(C)** Inverse relationship between SCP activities was also observed for the SCPs *Vesicle tethering* (simulated via the tethering rate at the GC) and *Vesicle fusion* (simulated by amount # of v-SNARE V). **(D)(E)(F)** In contrast to Level-3 sibling SCPs, Level-2 sibling SCP activities must be uniquely coordinated to enable NOG without violation of the model constraints. **(D)** To ensure NOG under a physiological TGN size (thick line) the activity of the SCP *Vesicle budding at the TGN* (simulated by modification of budding rate at TGN, while keeping the amount # of recruitment factor 1 constant) needs to increase with increasing outgrowth velocity. Any deviation in SCP activity inevitably causes unphysological TGN sizes (thin lines). Surface areas indicate size of TGN. **(E)** A similar effect was observed for the Level-3 SCP *Anterograde MT-based vesicle transport* (simulated by modification of # number of kinesin receptors, while keeping the fraction of bound kinesin constant). For each velocity, there is only one SCP activity that allows NOG without violation of the model constraints (green line). Numbers indicate percent of stationary vesicles in the NSC reservoir. **(F)** Deviation in the SCP activity *Vesicle exocytosis* (simulated by modification #amount of v-SNARE V, while keeping the tethering rate constant) violate the model constraint that 90% of vesicles in growth cone act as the first membrane reservoir for the quick recruitment of additional membrane on demand. Numbers indicate percentages of stationary vesicles. **(G)** Velocities were color coded in each figure as indicated. **(H) & (I)** The more a Level-2 SCP activity is generated by a Level-3 SCP that contains a vesicle membrane protein, the higher the SCP activity of membrane protein production to compensate for the loss of membrane protein in the growing NSC reservoir. **(H)** In dependence of the selected fraction of MT-bound kinesin, vesicles need a different amount of kinesin receptors to ensure NOG without violation of the model constraints. A higher kinesin receptor concentration per vesicle directly translates into the need for a higher kinesin production rate. Lines refer to pre-defined fractions of MT-bound kinesin: 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1.0 (from top to bottom). **(I)** Different tethering rates demand different v-SNARE V concentrations per vesicle and consequently different V production rates to ensure NOG without violation of the model constraints (t-SNARE Y is kept constant). Lines refer to different tethering rates: 2.5×10^{-6}, 3×10^{-6}, 3.5×10^{-6}, 4×10^{-6}, 4.5×10^{-6}, 5×10^{-6} (from top to bottom).

Figure 7: Systematic analysis of microtubule growth SCP-relationships reveals redundancies among level-3 sibling SCPs, but not among their level-2 parent SCPs. (A) Summary of the basic reaction for the simulation of MT growth. See Figure 2C for details. **(B), (C), (D) & (E)** Level-3 SCPs that define sub-functions of the Level-2 SCP *Dynamic MT growth* show complementary activities to allow NOG without violation of the model constraints. Multiple combinations of the SCP activities *Tubulin sequestration*, *GTP hydrolysis* and *Nucleation of dynamic MTs* enable NOG with a given velocity. **(B-C)** The influence of the SCPs *Tubulin sequestration* (which determines the effective tubulin concentration) and *GTP hydrolysis* on the length distribution (B) and degradation rate (C) of the dynamic MTs were determined in a pre-simulation using a stochastic model for the simulation of single dynamic MT growth (Margolin et al). Colored lines refer to the selected GTP hydrolysis rates shown in (B) and (C) (pink: 0.72/sec, d) and (yellow: 0.675/sec, e). **(D-E)** Multiple combinations of dynamic MT nucleation rates and effective tubulin concentrations allow NOG under a given velocity for each GTP hydrolysis rate, as shown for the hydrolysis rate of 0.72/sec (d) and 0.675/sec (e). **(F)** Velocities in (D) and (E) were color coded as indicated.

Figure 8: Imbalanced SCP dynamics disturb neurite outgrowth. (A-B) We predicted four sets of kinetic parameters that allow coordinated NOG at the velocities 5, 10, 15 and 20 μm/h. To simulate the effect of uncoordinated vesicle SCP dynamics we decreased the count and production rate of v-SNARE V to 25%, 50% and 75% of the original prediction, without modifying any other parameters. Numerical simulations predict that reduced exocytosis causes an increase in the count of vesicles in the growth cone cytoplasm (A) without reducing NOG velocity (B). **(C)** We validated our prediction by analyzing neurite outgrowth after knockdown of the v-SNARE VAMP7. P1 rat cortical neurons transfected with siRNA against VAMP7 or scrambled siRNAs were stained for β-III tubulin to document neurite growth across the microgroove of the chamber. In agreement with our prediction, VAMP7 knock down did not decrease total outgrowth length. It caused a significant number of physiological growth cones to turn into dystrophic bulbs. **(D)** Analysis of quantified neurite lengths confirmed that there was no significant difference in outgrowth length. Shown are average values and standard deviations (Three independent experiments). **(E)** Neurites growing from neurons after VAMP7 knock down showed a significant increase in the number of dystrophic bulbs, indicative for impaired NOG. **(F)** To simulate the effect of reduced MT stabilization, we reduced the stabilization rate that turns dynamic into stable MTs to 25%, 50% and 75% of the original prediction, without changing the predicted values for all other kinetic parameters. Numerical simulations propose that decreasing stabilization leads to a significant decrease in NOG outgrowth over the investigated time. **(G)** We validated our predictions by knock down of the microtubule crosslinking protein Mtcl1. Mtcl1 knock down caused reduced NOG outgrowth and dystrophic bulb formation. **(H)** Outgrowth quantification confirms significant decrease in NOG (3 independent experiments). **(I)** Mtcl1 also significantly increased the number of dystrophic bulbs. *$p < 0.05$, **$p < 0.01$, ***$p < 0.001$ based on unpaired t-test for all panels.

STAR Methods:

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Antibodies		
Anti- tubulin β –III (TUJ1 clone) made in mouse	BioLegend	801201
Mountant PermaFluor	ThermoFisher Sci	TA-030-FM
Goat anti-Mouse IgG Secondary- Alexa Fluor488	ThermoFisher Sci	A-11001
Neurobasal-A Medium	ThermoFisher Sci	10888022
B-27 Supplement (50X), serum free	ThermoFisher Sci	17504044
L-Glutamine (200mM)	ThermoFisher Sci	25030081
Penicillin-Streptomycin (5,000 U/ml)	ThermoFisher Sci	15070063
OptiPrep Density Gradient Medium	Sigma-Aldrich	D1556
Papain, aseptically filled	Sigma-Aldrich	P5306
Deoxyribonuclease I from bovine pancreas	Sigma-Aldrich	DN25
Sterile Cell Strainers (40µm)	Fisher Scientific	22-363-547
Accell siRNA delivery media	Dharmacon	B00500-100
5x siRNA buffer	Dharmacon	B002000-UB-100
Recombinant DNA		
Mtcl1 SiRNA1 GUGUUUUAUUUCAUGCUUC	Dharmacon	E-085290-00
Mtcl1 SiRNA2 UCAGAGUGUUCUAAGUUAA	Dharmacon	E-085290-00
Mtcl1 SiRNA3 CCCUGAAGCAGAACAUUUU	Dharmacon	E-085290-00
Mtcl1 SiRNA4 GCUUUGUCCCUGGAUGAUG	Dharmacon	E-085290-00
Vamp7 SiRNA1 CCAUCAUCGUCGUUGUAUC	Dharmacon	E-094480-00
Vamp7 SiRNA2 GGAUUGUGUAUCUUUGCAU	Dharmacon	E-094480-00
Vamp7 SiRNA3 UCAUGGCAAUUAUUUGUUU	Dharmacon	E-094480-00
Vamp7 SiRNA4 CCAUGAACAGUGAGUUUUC	Dharmacon	E-094480-00
Accell non-targeting pool	Dharmacon	D-001910-10-20
Software and Algorithms		
ImageJ	ImageJ	ImageJ 1.49v
GraphPad Prism	GraphPad Software	GraphPad Prism7
Adobe Photoshop CS5	Adobe	Photoshop CS5
MATLAB	MATHWORKS	R2017b
MetaMorph®	Neurite outgrowth module	MetaMorph 7.7
Other		
Xona Microfluidic chambers	Xona Microfluidics	SND150
50mm Glass Bottom Dishes No. 1.5	MatTek Corp.	P50G-1.5-30-F

LSM 880 Laser Scanning Microscope with Airyscan	Zeiss Microscopy	Zeiss LSM 880

Figure 1   **Top down based SCP modelling of whole cell dynamics**

Step 1a: List different classes of SCPs and their relationships at various levels required for whole cell responses

Step 1b: Represent SCPs as coupled biochemical reactions

Step 1c: Develop coupled ODE models for different SCPs

Step 1d: Generate and identify experimental data that ODE model-based simulations have to match

Step 2a: Develop analytical predictions for kinetic parameters for SCP activities at steady state

Step 2b: Use analytical prediction to identify parameter sets for steady state dynamics that allow simulations in agreement with experimental data (Step 1d)

Step 2c: Classify SCP sets at different levels as redundant or non-redundant for whole cell function

Step 2d: Predict cell level phenotypes for experimental testing

Figure2A

Whole cell function   Level-1 SCPs   Level-2 SCPs   Level-3 SCPs

Figure2B

Figure2C

Neurites grow with different outgrowth velocities

Figure 3A

Slow growing neurite

Figure 3B

7620 µm

Fast growing neurite

Figure 3C

Figure 3D

Figure 4

A

Neurite shaft length

B

Microtubule bundle length

C

Vesicles in Neurite Shaft Cytoplasm

D

Comparison of model predictions after 1500 min with model constraints

Figure 5

A

Figure 5

B

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1) Calculate Type I – IV membrane transport rates and initial and final transport rates

1.1) $\textbf{\textit{Neurite shaft growth}} = \textbf{\textit{Type II membrane transport rate}} = \textbf{\textit{NOG velocity}} \times 2\pi \times r_{neurite}$

Calculate type I, III and IV transport rates based on NOG velocity, model constraints and Type II membrane transport rate

Calculate initial and final forward and backward transport rates as indicated in (A)

2) Calculate v – SNARE V protein amount and concentration in GCC that allows final forward transport rate

2.1) $YV\ SNARE\ complex\ in\ GC = \dfrac{Final\ forward\ transport\ rate\ \times SNARE\ complex\ per\ vesicle\ fusion}{Anterograde\ vesicle\ surface\ area}$

2.2) $VV_{GCC}^{B_G} = \dfrac{YV\ SNARE\ complexes\ GC}{tethering\ rate\ at\ GC\ \times\ YY_{PM}}$, \qquad 2.3) $V_{GCC}^{B_G} = \dfrac{VV_{GCC}^{B_G}}{n_{GCC}^{B_G}}$

3) Calculate v – SNARE V forward fluxes

3.1) $VV\ initial\ forward = V_{GCC}^{B_G} \times initial\ forward\ transport\ rate$

3.2) $VV\ final\ forward = V_{GCC}^{B_G} \times final\ forward\ transport\ rate$

4) Calculate v – SNARE U protein amount and backward fluxes similarly

5) Calculate v – SNARE U concentration at GC that allows v – SNARE U backward fluxes

5.1) $VV\ final\ forward = V_{GCC}^{B_G} \times final\ forward\ transport\ rate$

6) Calculate v – SNARE U concentration at GC that allows v – SNARE U backward fluxes

6.1) $const_{GC} = 1 + \dfrac{Y_{GC}}{kya} + \dfrac{X_{GC}}{kxa}$ \quad 6.2) $b_s = initial\ backward\ transport\ rate\ \times SNARE\ binding\ spots\ per\ vesicle\ area$

6.3) $U_{GC} = \dfrac{UU\ initial\ backward\ \times const_{GC} \times kva}{(b_s - VV\ initial\ backward\ \times UU\ initial\ backward)}$

7) Calculate v – SNARE V concentration at GC

7.1) $V_{GC} = \dfrac{VV\ initial\ backward \times (\frac{U_{GC}}{kua} + const_{GC}) \times kva}{(b_s - VV\ initial\ backward\ )}$

8) Calculate v – SNARE V production rate at TGN to compensate for v – SNARE V accumulation in MTC

8.1) $VV\ production\ rate = membrane\ type\ I \times V\ forward\ rate + membrane\ type\ III \times V\ backward$

Figure 5

C

D

(9) **Calculate MTB growth velocity**

$$NOG\ velocity = MTB\ growth\ velocity = \frac{Length\ increase\ of\ stable \times MTs}{MTs\ per\ crossection}$$

(10) **Calculate number of dynamic MTs**

$$Number\ dynamic\ MTs = \frac{Total\ dyn\ MTs\ length}{(Average\ dynamic\ MT\ length)}$$

(11) **Calculate MT stabilization rate**

$$stablization\ rate = \frac{Length\ increase\ of\ stable\ MTs}{(Average\ dynamic\ MT\ length \times Number\ dynamic\ MTs)}$$

(12) **Calculate MT nucleation rate**

$$Nucleation\ rate = (Stabilization\ rate + Degradation\ rate) \times Total\ number\ dynamic\ MTs$$

E

n= 60 (6 parameter sets x 10 time points)

Figure 6

Figure 7

Figure 8

C

D

E

Figure 8

G

Control

Mtlc1 knockdown

150μm

100μm

Microgroove

Microgroove

H

Fluorescence Intensity for B–III tubulin

Distance from microgroove (μm)

● Control
■ Mtcl1

I

Total Dystrophic Bulbs

Control   Mtcl1

**