Simulations of Inositol Phosphate Metabolism and Its Interaction with InsP₃-Mediated Calcium Release

Jyoti Mishra and Upinder S. Bhalla
National Centre for Biological Sciences, GKVK Campus, Bangalore 560065, India

ABSTRACT Inositol phosphates function as second messengers for a variety of extracellular signals. Ins(1,4,5)P₃ generated by phospholipase C-mediated hydrolysis of phosphatidylinositol bisphosphate, triggers numerous cellular processes by regulating calcium release from internal stores. The Ins(1,4,5)P₃ signal is coupled to a complex metabolic cascade involving a series of phosphatases and kinases. These enzymes generate a range of inositol phosphate derivatives, many of which have signaling roles of their own. We have integrated published biochemical data to build a mass action model for InsP₃ metabolism. The model includes most inositol phosphates that are currently known to interact with each other. We have used this model to study the effects of a G-protein coupled receptor stimulus that activates phospholipase C on the inositol phosphates. We have also monitored how the metabolic cascade interacts with Ins(1,4,5)P₃-mediated calcium release. We find temporal dynamics of most inositol phosphates to be strongly influenced by the elaborate networking. We also show that Ins(1,3,4,5)P₄ plays a key role in InsP₃ dynamics and allows for paired pulse facilitation of calcium release. Calcium oscillations produce oscillatory responses in parts of the metabolic network and are in turn temporally modulated by the metabolism of InsP₃.

INTRODUCTION

Hormones, neurotransmitters, and growth factors that activate phospholipase C generate a bifurcating signal with Ins(1,4,5)P₃ at one arm and diacylglycerol (DAG) at the other (Rhee, 2001). Although DAG activates protein kinase C (Nishizuka, 1988), InsP₃ mobilizes calcium via the endoplasmic reticulum based InsP₃ receptor (Berridge, 1993). Intracellular levels of these secondary messengers depend on a balance between their rate of formation and rate of removal, which channels them back to lipid resynthesis. Ins(1,4,5)P₃ is metabolized in cells by an interplay of phosphatases and kinases (Shears, 1989) that result in production of inositol phosphates ranging from inositol monophosphates to inositol heptakis and octaphosphates (Irvine and Schell, 2001). Understanding the kinetics of this complex metabolic network for InsP₃ is fundamental to deciphering its role in shaping intracellular calcium dynamics.

InsP₃ functions as an important secondary messenger (Berridge and Irvine, 1989; Streb et al., 1983) that binds to InsP₃ receptors embedded in the endoplasmic reticular membrane and mediates calcium release into the cytosol (Taylor and Richardson, 1991). This release can elevate calcium from 10 to 100 nM basal levels to several micromolar stimulated levels (Bootman and Berridge, 1995). Depending on the cell type, the calcium waveform can either exhibit a single peak or can oscillate with multiple spikes (Berridge and Irvine, 1989; Tsien and Tsien, 1990; Meyer and Stryer, 1991). Isolated calcium puffs generated locally in the cytosol can propagate throughout the cell as waves (Lechleiter et al., 1991; Parker and Yao, 1991) and activate various cell physiological processes (Berridge, 1993) including differentiation, proliferation, vesicle release, and sensory perception.

Like InsP₃, its metabolic products also play essential roles in cellular function, although they have not been as rigorously assessed as InsP₃. For instance, evidence has been accumulating for the role of Ins(1,3,4,5)P₄ in facilitating InsP₃-mediated Ca²⁺ release (Morris et al., 1987; Cullen et al., 1990; Smith et al., 2000). It was recently shown that calcium release activated calcium current (I_{Ca,10}) (Hoth and Penner, 1992) is enhanced due to the inhibitory effect of Ins(1,3,4,5)P₄ on InsP₃ 5-phosphatase (Hermosura et al., 2000). The Ras signaling pathway is also affected by Ins(1,3,4,5)P₄ function. GTase activating proteins, GAP1m and GAP1IP4BP, gain activity by specifically binding InsP₄ (Cullen et al., 1995; Fukuda and Mikoshiba, 1996).

Ins(3,4,5,6)P₄ also functions as a secondary messenger that regulates calcium activated chloride efflux. It inhibits a plasma membrane chloride channel (Vajanaphanich et al., 1994; Shears, 1998) and is thus involved in osmoregulation. Further, its cellular levels are modulated by an InsP₄ isoform Ins(1,3,4)P₃, which is generated by the degradation of Ins(1,3,4,5)P₄ (Yang et al., 1999). InsP₃ and InsP₆ (phytic acid) belonging to the inositol high polyphosphate series (IHPS), play essential regulatory roles in endocytosis and exocytosis. In vitro, they have been shown to interact in varying affinities with assembly proteins important in clathrin-mediated endocytosis and with synaptotagmin domains involved in synaptic vesicle trafficking (Fukuda and Mikoshiba, 1997). “High energy” pyrophosphates are equally important for intracellular trafficking. PP-InsP₅ is the most potent known inhibitor of AP-180-
mediated clathrin cage assembly (Ye et al., 1995). There is also evidence that InsP_2 and InsP_3 may function as extracellular signals to regulate blood pressure and heart rate (Vallejo et al., 1987).

In addition to these specific signaling actions, inositol phosphates have also been shown to bind crucial cellular proteins like the cytoskeletal element vinculin, the signaling molecule Bruton’s tyrosine kinase (Btk), and the cell adhesion molecule myelin proteolipid protein (Fukuda and Mikoshiba, 1997). Active research is being conducted to find the exact physiological significance of these varied binding properties of inositol phosphates.

In parallel with research on the functional relevance of inositol phosphates, several key enzymes in the metabolic cascade of InsP_3 have been identified (Majerus, 1992). These enzymes regulate the cellular concentrations of inositol phosphates under basal and stimulated conditions. They are all kinases and phosphatases that sequentially add or remove phosphate groups from the inositol ring, exhibiting high specificity for their substrates. They show extensive cross-talk among themselves and are subjected to regulation by inositol phosphates in the metabolic network that are not their immediate substrates and products. Some of these enzymes are also regulated by general signaling molecules such as calcium, protein kinase C (Sim et al., 1990), calmodulin (CaM), and calcium-calmodulin activated protein kinase type II (CaMKII) (Communi et al., 1997). Another dimension to interactions among inositol phosphate metabolism enzymes is their varied spatial distribution (Soriano et al., 1997). Most of them are present in the cytosol but some are attached to the plasma membrane whereas others such as multiple inositol polyphosphate phosphatase are compartmentalized in the ER (Chi et al., 2000; Nogimori et al., 1991). This distribution alters accessibility toward substrates and regulatory molecules. Further complexities arise because of the presence of multiple isoforms of the same enzymes, whose expression levels differ from one cell type to another. Although most of these kinases and phosphatases have strict substrate specificity, some also catalyze multisubstrate reactions.

We have constructed a biochemical model for the cellular metabolism of Ins(1,4,5)P_3. To generate the mass action law for the forward reaction,

$$A + B \rightleftharpoons C \quad K_d = k_f/k_i; \quad \tau = 1/(k_i + k_f) \quad (1)$$

In this formulation, interactions are completely specified by the initial concentrations of the reactants (A and B) and the product (C), by the rate constant for the forward reaction, $k_i$, and the rate constant for the back reaction, $k_f$. $\tau$ represented the time course of the simple reaction. Its exact value, which can be influenced by various factors in a reaction network, was initially approximated during model building and subsequently validated through simulations.

Eq. 1 can also be represented as a differential equation given by:

$$\frac{d[A]}{dt} = -k_i[A][B] + k_f[C]$$

and similar equations for the other molecules that participate in the reaction.
Enzyme reactions modeled in the network mostly followed Michaelis-Menten kinetics (Eq. 2), but some enzymes were modeled with reversible kinetics.

The Michaelis-Menten scheme for enzyme-catalyzed reactions was of the form:

\[ E + S \rightleftharpoons k_1^{-1} k_2 ES \rightarrow E + P \quad K_m = \frac{(k_2 + k_3)/k_1; \quad V_{\text{max}} = k_3}{k_2} \]  

(2)

This standard formulation for enzymes is a special case of two reactions in sequence with the assumption that the final step is irreversible. It is specified by three rate constants \( k_1 \), \( k_2 \), and \( k_3 \), and the initial concentrations of the enzyme \( E \), substrate \( S \), and product \( P \). Unless actual rate constants were provided in literature, \( k_2 \) was taken as four times \( k_1 \). Most enzyme models are insensitive to the exact value of this ratio (Bhalla, 1998). We validated this in our model also by testing \( k_2/k_1 \) ratios ranging from 0.4 to 40 without significant variation in simulation results (data not shown).

Reversible enzyme reactions were of the form:

\[ E + S \rightleftharpoons k_1^{-1} k_2 ES \rightleftharpoons E + P \]

\[ K_m = \frac{(k_2 + k_3)/k_1; \quad K_m' = (k_3 + k_2)/k_4;}{V_{\text{max}} = k_3 - k_2; \quad V_{\text{max}}' = k_2 - k_3} \]  

(3)

Here \( K_m \) and \( K_m' \) represent the distinct affinities of the enzyme for the substrate and product respectively, and \( V_{\text{max}} \) and \( V_{\text{max}}' \) represent the maximal velocities of the forward and backward reactions. Enzymes were modeled as reversible reactions when the back flux from the reaction...
products to the substrates was over 5% of the forward flux. The value of the fourth rate constant $k_4$, unless available in literature, was derived by free energy calculations. For a reaction at equilibrium, the standard free energy change $\Delta G^\circ$ (kJ/mol) is related to $K_{eq}$ (equilibrium constant for the reaction) by:

$$\Delta G^\circ = -RT \ln K_{eq}$$

$$R = 8.314 \text{ JK}^{-1} \text{ mol}^{-1}, \quad T = 310 \text{ K} \quad (4)$$

Further $K_{eq}$ is related to the rate constants of the reversible enzyme reaction by:

$$K_{eq} = (k_1 \cdot k_3)/(k_2 \cdot k_4) \quad (5)$$

We extracted $k_4$ for each enzyme reaction for $\Delta G^\circ$ values ranging from $-10 \text{ kJ/mol}$ to $-25 \text{ kJ/mol}$. This $\Delta G^\circ$ range was approximated from values for glucose phosphorylation (Jencks, 1976) as the $\Delta G^\circ$ for phosphorylation of the inositol ring has not, to our knowledge, been reported in the literature. We chose a wide range of $\Delta G^\circ$ values to span the likely range of $k_4$ rates. Glucose was chosen due to its structural similarity to inositol. Glucose phosphorylation (Jencks, 1976) serves as a cellular precursor to inositol phosphates (Irvine and Schell, 2001).

In this way all enzyme reactions depicted in Fig. 1 B by \leftrightarrow symbols were modified from the Michaelis-Menten scheme to the reversible kinetics scheme. In general the reversibility of reactions did not introduce large changes in the model kinetics, but there are some interesting exceptions that are discussed in the Results section.

Parameters for all enzymes modeled were derived from published biochemical experiments that involve protein purification and characterization of enzyme kinetics. Michaelis-Menten enzymes were characterized by their $K_m$ and $V_{max}$, and their initial concentrations were obtained from biochemical purification series and quantitative immunoprecipitation. In most cases, measurements across different published reports in literature allowed us to cross check the values of the constants we used. For simple reactions, $k_1$ and $k_2$ were constrained by time courses and dose response curves. Certain reactions within the inositol phosphate metabolic network that have been described in literature as enzymatic reactions but for which enzyme kinetics have not been characterized in mammalian systems had to be modeled as simple reactions. Rates for such reactions were constrained by equilibrium concentrations of their reactants and products. The incorporation of these simple reactions into the model introduces parameters that have not been tested experimentally. Nevertheless, the equilibria for these reactions are tightly constrained by the known steady-state levels of reactants in the InsP$_3$ metabolism cascade. Without these reactions, depicted in Fig. 1 B by numbers 10, 13, 14, 16, 17, 18, and 23, components of the metabolic network do not equilibrate in simulation runs due to unbalanced source-sink relationships among them. We acknowledge that precise kinetics of these reactions require validation through experiments.

Table 1 Parameter set used to model InsP$_3$ dephosphorylation by InsP 5-phosphatase1 and its inhibition by InsP$_5$.

<table>
<thead>
<tr>
<th>Molecule</th>
<th>Initial concentration (µM)</th>
<th>Buffering*</th>
<th>Notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>IP$_5$-phosase1</td>
<td>12.346</td>
<td>0</td>
<td>InsP 5-phosphatase1 enzyme pool: from Hansen et al., 1987.</td>
</tr>
<tr>
<td>IP5(13456)</td>
<td>53.999</td>
<td>0</td>
<td>Ins(13456)pentakisphosphate: concentration from Voglmaier et al., 1996.</td>
</tr>
<tr>
<td>IP5-5-pase-cmplx</td>
<td>0</td>
<td>0</td>
<td>Product complex of the competitive enzyme inhibition reaction of IP5 with IP$_5$-phosase1.</td>
</tr>
</tbody>
</table>

Reaction $K_f \quad K_h \quad$ Notes

IP$_5$-inhb-3pase  1  0.06  Reaction for competitive inhibition of IP$_5$-phosase1 by IP5: $K_i$ from Hoer and Oberdisse, 1991.

<table>
<thead>
<tr>
<th>Enzyme</th>
<th>Parent</th>
<th>$K_m$</th>
<th>$V_{max}$</th>
<th>Ratio$^1$</th>
<th>Notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>ip$_3$-5pase1</td>
<td>IP$_5$-phosase1</td>
<td>3</td>
<td>1.1</td>
<td>4</td>
<td>Ins(145)P3 5-phosphatase activity of InsP 5-phosphatase1: from Hansen et al., 1987.</td>
</tr>
</tbody>
</table>

*Buffering indicates whether a pool is held fixed at its initial concentration through simulation runs. For such pools buffering is 1.

$^1$Value refers to ratio of $k_2/k_3$ for Michaelis-Menten enzymes (Eq. 2).

Unfortunately, the document page seems to be incomplete or possibly overlapped with another page, which makes it challenging to accurately transcribe or extract the full context. The provided text sections refer to a study on the simulation of IP$_3$ metabolism and calcium release, with a focus on modeling enzyme kinetics and reactions within the network. The table includes parameter sets and notes on buffer management, enzyme interactions, and reaction kinetics. However, the page contains some sections that are incomplete or are intended for further expansion or detail that isn't visible in the current form.
A sensitivity analysis was performed to investigate how sensitive the model was to the numerous parameters used to build it. For this analysis, parameters were classified in five categories of initial concentration, reaction dissociation constant ($K_i$), activation time course ($t_a$), enzyme Michaelis constant ($K_m$), and enzyme maximal velocity ($V_{max}$). Values of all parameters in each category were systematically scaled 0.1, 0.2, 0.5, 0.667, 0.833, 1.2, 1.5, 2, 5, and 10 times their original value and the model subjected to simulation runs. Variability in results for this 100-fold range (0.1–10) of all parameter values were analyzed by generating scatter plots for logarithmic fold changes in outputs for calcium, Ins(1,4,5)P3, Ins(1,4,5,6)P4, and Ins(1,3,4)P2. For the model that included detailed InsP3 receptor kinetics, the scatter plot readout was the frequency of calcium spikes generated within stimulus time. For ease of analysis, all scatter plots were normalized with respect to the control output. The control refers to the basic scheme of the model with the original parameter values.

Two global sensitivity analyses were also performed to investigate the effects of alterations in a combination of parameters on the model behavior. The first global analysis focused on effects of temperature alteration on the system. This was based on the premise that an −10°C rise in temperature accelerated reaction rates by twofold. Simulations were performed to assess the effects of 10°C rise and 10°C fall in temperature on all reaction rates. The second global sensitivity analysis targeted all energy consuming reactions, i.e., kinases and ATPases, which would be affected by ATP depletion in the system. Because for most reactions ATP was not explicitly included as a reactant species, a change in ATP concentration to x times its original value was approximated by scaling enzyme $K_m$ by a factor of 1/x, and by changing the $k_i$ and $k_o$ of ATP dependent reactions by a factor of x and 1/x, respectively. Simulation runs were performed for ATP depletion to 0.5 and 0.15 times its original concentration, which represent physiological drops in cellular ATP levels (Kahlert and Reiser, 2000). Sensitivity analysis outputs were monitored for Ca2+ spikes generated within stimulus time. For ease of analysis, all scatter plots were normalized with respect to the control output. The control refers to the basic scheme of the model with the original parameter values.

All numerical computations were performed using a graphical interface–Kinetikit (version 7) on the General Neural Simulation System–GENESIS (Bhalla, 1998). Computations were carried out on PCs running Linux. The exponential Euler formulation was used for integration (MacGregor, 1987). Numerical accuracy of the computations was verified by comparing the results for simulations that had been run at different time steps. The model provided convergent solutions for the range of time steps used in the study. For analysis of the model output, a time step of 0.2 ms was used. At initiation of the simulation run and at all transient points wherein steady state of the model was perturbed by a stimulus input, a fine time step of 10 μs for 10 μs was used. The temporal characteristics of some of the output curves are described in terms of rise time of the response to stimulus, response latency, and response width. We define rise time of a response to stimulus as the time taken to progress from 20% to 80% of the maximal response height. Response latency is defined by the time taken to achieve 20% of maximal response height from stimulus onset, and response width represents the full width of the response at half maximal response height.

**RESULTS**

**Responses of components of the Ins(1,4,5)P$_3$ metabolic pathway**

We first characterized the effects of a PLC β-activating stimulus on the components of the Ins(1,4,5)P$_3$ metabolism model. For this purpose, the non-Osc-model was initially allowed to achieve steady state by running the model without stimulation for a period of 1000 s. Stimulus was then provided as a square pulse of 0.5 μM glutamate for 30 s. Poststimulation model responses were simulated for 1000 s.

Components of the InsP$_3$ metabolic pathway showed varied responses to stimuli with respect to both amplitude and time of response. These are depicted in Fig. 2.

**Responses of Ins(1,4,5)P$_3$ and its lower phosphates**

Ins(1,4,5)P$_3$ responded to stimulus with a sevenfold concentration change. The half-life of InsP$_3$ has been estimated at 9 ± 2 s in N1E-115 neuroblastoma cells both by biochemical methods and by calcium imaging studies (Wang et
al., 1995). Our simulations generate a similar half-life value of 10 s for InsP3. The simulated response width of InsP3 was 20 s. The direct degradation product of InsP3, Ins(1,4)P2 showed a similar response width and a similar concentration change of approximately sixfold. The peak responses of these two inositol phosphates were sharp and transient and inactivated more rapidly than responses of other inositol phosphates. We found the kinetics of PLC β to be primarily responsible for the rapid peak transients of InsP3 and its lower phosphates. The GTPase-activating protein activity of PLC β (Berstein et al., 1992) has been incorporated in our model. Thus, activated PLC β hastens inactivation of its stimulatory G-protein above its basal inactivation rate and in turn rapidly restores itself to its low Vmax basal form. If this negative feedback GAP activity of PLC β is excluded from our model, the transient nature of InsP3 and Ins(1,4)P2 responses is lost (not shown).

Further, the InsP3 response was observed to be biphasic with the rapid peak phase followed by a slow plateau-like phase at less than 50% the peak height. The second phase of the response was due to significant reverse flux in the phosphorylation reaction catalyzed by activated InsP3 3-kinase. This phase corresponds to the rapid increase in Ins(1,3,4,5)P4 levels upon stimulation and is not seen if CaM- and CaMKII-activated InsP3 3-kinase is modeled as an irreversible Michaelis-Menten enzyme. The biphasic InsP3 response directly correlates to a biphasic response for InsP3 receptor released Ca2+ (see Fig. 2 H), which is observed under physiological conditions (Lambert and Nahorski, 1990). Characteristics of the Ca2+ response to stimuli are described in the following sections.

**Responses of Ins(1,3,4)P3 and its lower phosphates**

Ins(1,3,4)P3 responded to the 30-s 0.5 μM glutamate pulse with an approximately fourfold concentration change. Its dephosphorylation products, Ins(1,3)P2 and Ins(3,4)P2, showed similar responses with sixfold and fourfold changes in concentration, respectively. As seen in Fig. 2 B, the major route for Ins(1,3,4)P4 metabolism was via inositol phosphate 1-phosphatase, such that basal or stimulated levels of Ins(3,4)P2 were always greater than levels of Ins(1,3)P2. This metabolic predominance of Ins(3,4)P2 has been reported experimentally (Bansal et al., 1987). Another characteristic of the Ins(1,3,4)P3 response that tallies with experiments was its response latency (time taken to achieve 20% of maximal response height from stimulus onset). Our simulations generated a latency of ~30 s for Ins(1,3,4)P3, which is 10 s longer than that for Ins(1,4,5)P3. A comparable delay in Ins(1,3,4)P3 build-up has been shown by Hughes and Drummond (1987) and Jackson et al. (1987). Their experiments suggest that this delay reflects the time taken for the receptor signal to progress from Ins(1,4,5)P3 to Ins(1,3,4,5)P4 and then to Ins(1,3,4)P3 by the successive action of InsP3 3-kinase and InsP 5-phosphatase1. InsP3 3-kinase is a relatively fast enzyme but InsP 5-phosphatase1 activity for its InsP4 substrate is slow and further inhibited by physiological levels of InsP3 and InsP6.

Ins(1,3,4)P4 and its metabolites showed a prolonged rectangular concentration response over time, in contrast to the sharp transient responses of Ins(1,4,5)P3. For Ins(1,3,4)P3, this slow response had a width of ~250 s. Its peak concentration was maintained as a near plateau for 55% of this time. We observed that the time spent near peak directly overlapped with the response width of InsP4. Fig. 2 F shows that upon stimulation, Ins(1,3,4,5)P4 undergoes a 60-fold change in levels, which is the largest change seen for any inositol phosphate in the network. Such high Ins(1,3,4,5)P4 levels serve as a nonlimiting substrate pool for generation of Ins(1,3,4)P3 and its dephosphorylated products. Input to Ins(1,3,4)P3 was both via dephosphorylation of InsP4 by InsP 5-phosphatase1 and by the flux reversal in the phosphorylation reaction catalyzed by InsP3 5,6-kinase (Ho et al., 2002). Under the influence of saturating concentrations of Ins(1,3,4,5)P4, enzymes downstream of it functioned at their respective Vmax. This was seen not only at the primary level, i.e., conversion to Ins(1,3,4)P3, but also at the secondary level, i.e., conversion of Ins(1,3,4)P3 to Ins(1,3)P2 and Ins(3,4)P2. Thus, for ~140 s for which InsP4 levels were elevated beyond 50% of its peak levels, Ins(1,3,4)P3 and its metabolites maintained an elevated plateau phase. During this time we also observed a dip below basal levels for Ins(1,4)P2. This was caused by high Ins(1,3,4,5)P4 levels acting as potent substrate competitor for the enzyme InsP 5-phosphatase1, which catalyzes Ins(1,4)P2 synthesis from Ins(1,4,5)P3. Further consequences of this particular enzyme substrate competition on Ins(1,4,5)P3 are elaborated in later sections of the paper.

Thus, we see that due to the networking within the inositol phosphate cascade, large scale changes in one component of the system, such as Ins(1,3,4,5)P4, can have diverse effects on various other molecules. We predict Ins(1,3,4,5,6)P5 to be a prominent modulator of the temporal characteristics of other inositol phosphates. The consequent prolonged elevation of Ins(1,3,4)P3 and its derivatives can provide these molecules with a relatively long-term capacity for cellular function.

**Responses of Ins(1,3,4,5,6)P5 and other inositol high polyphosphates**

The IHPs include Ins(1,3,4,5,6)P5, InsP6, PP-InsP4, PP-InsP5, and bis-PP-InsP4. As seen in Fig. 2, D and E, their responses are negligible compared with those of the lower inositol phosphates. They seem to form a separate biochemical pool that is unaffected by the Ins(1,4,5)P3 generating stimulus. This is probably because InsP5 and InsP6 have very large basal pools (30–60 μM) that may require strong direct stimulation for any appreciable changes to occur. Safrany and Shears (1998) have shown that a cAMP-mediated mechanism regulates the turnover of bis-PP-InsP4.
which further suggests that the IHPs are regulated by other pathways than a PLC β signal. In an Ins(1,4,5)P$_3$ 3-kinase overexpression study conducted in fibroblast cells (Balla et al., 1994), negligible increases in InsP$_2$ and InsP$_4$ concentrations had been reported. However, Balla et al. (1994) showed cell cycle dependent changes in the levels of these inositol phosphates. InsP$_2$ and InsP$_4$ increased markedly during the S-phase of the cell cycle. This further corroborates the role of regulators extrinsic to Ins(1,4,5)P$_3$ generation in influencing the IHPs. Because the exact molecular mediators that control levels of these inositol phosphates are not known, we do not include them in our model.

We have shown that the behavior of the IHPs in our model is concordant with experiments that analyze effects of GPCR stimuli on inositol phosphates. We further investigate what happens upon direct stimulation of IHPs in later sections of this paper.

Responses of inositol tetrakisphosphates

Within the whole InsP$_4$, metabolic cascade, the most prominent response was that of Ins(1,3,4,5)P$_4$. Stimulation produced a 60-fold change in Ins(1,3,4,5)P$_4$ concentration, which developed with a rise time of ~30 s. This large change was primarily because of the extensive positive regulation of the InsP$_4$ synthesizing enzyme, InsP$_3$ 3-kinase, by calmodulin and CaMKII (Communi et al., 1997), which themselves underwent activation by Ca$^{2+}$ under stimulated conditions. Compared with the approximately sixfold activation of Ins(1,4,5)P$_3$’s dephosphorylation product Ins(1,4)P$_2$, stimulation of Ins(1,3,4,5)P$_4$, InsP$_3$’s phosphorylation output was much greater. Thus, we conclude that phosphorylation was the predominant means of Ins(1,4,5)P$_3$ metabolism in our system.

The complete Ins(1,3,4,5)P$_4$ response was bell shaped with a response width of ~140 s. As mentioned earlier, the InsP$_4$ response heavily influenced the responses of other inositol phosphates, especially Ins(1,3,4)P$_3$ and its dephosphorylation products. The large build-up of InsP$_4$ levels in the metabolic pipeline affected enzymes downstream of it such that they function near their respective $V_{\text{max}}$. Thus, stimulus responses for Ins(1,3,4)P$_3$ and others were rectangular rather than expected bell shapes. We further verified that Ins(1,3,4)P$_3$ was responsible for this phenomenon. For this, we incorporated a new kinetic pool in our model that served as a drainage sink for InsP$_4$ outside of the inositol phosphate network. This pool did not allow InsP$_4$ levels to exceed fivefold above basal, upon stimulation. Simulations of this model showed that in presence of the external InsP$_4$ sink, responses of Ins(1,3,4)P$_3$ and its derivatives were like the sharp transient responses of Ins(1,4,5)P$_3$ (not shown). Further, the difference in response latencies of the Ins(1,3,4)P$_3$ and Ins(1,4,5)P$_3$ responses was also lost. Thus, in tissues such as the brain where Ins(1,4,5)P$_3$ 3-kinase levels are high and large amounts of Ins(1,3,4,5)P$_4$ are generated, InsP$_4$ may function as an important temporal regulator for the other inositol phosphates.

Ins(3,4,5,6)P$_4$ displayed a ~2-fold response to stimulus. This InsP$_4$ is synthesized from InsP$_3$ by a reversible kinase/phosphatase, which also phosphorylates Ins(1,3,4)P$_3$ (Yang and Shears, 2000; Ho et al., 2002). However, experimentally reported basal and stimulated levels of Ins(3,4,5,6)P$_4$ could not be generated by this enzyme in our model system. For this purpose, a separate InsP$_3$ 1-phosphatase reaction was incorporated into the network. The temporal response of this InsP$_4$ followed that of InsP$_2$ and was rectangular in shape. InsP$_3$ receives input from the lower inositol phosphate Ins(1,3,4)P$_3$ via phosphorylation of Ins(1,3,4,6)P$_4$. It was interesting to note that the rectangular shape of the temporal response of Ins(1,3,4)P$_3$ could be faithfully transmitted to Ins(3,4,5,6)P$_4$, which is at a tertiary interaction level in the network (see Fig. 1 B). Further, the width of the Ins(3,4,5,6)P$_4$ response was ~410 s. Such long-term elevation in Ins(3,4,5,6)P$_4$ levels has been reported experimentally (Pittet et al., 1989) and is important for its cellular function as an inhibitor of Ca$^{2+}$-activated chloride efflux (Vajanapphanich et al., 1994).

Ins(1,3,4,6)P$_4$ showed a negligible response to stimulus, which is concordant with the measure of its stimulated levels in cerebral-cortex slices (Batty et al., 1989). Similarly response of Ins(1,4,5,6)P$_4$ was also negligible. This was because its only input was from the negligibly stimulated InsP$_4$ pool, but unlike its counterpart Ins(3,4,5,6)P$_4$ that receives similar input, this pool rapidly drained out into the lower inositol phosphates by the InsP$_4$ 6-phosphatase reaction.

Calcium response in the model and incorporation of oscillations

The primary function of Ins(1,4,5)P$_3$ as a secondary messenger is to release calcium from intracellular stores in response to external stimuli. We used a 0.5-μM glutamate stimulus lasting 30 s to monitor the calcium output. As seen in Fig. 2, the biphasic InsP$_3$ response to stimulus in the non-Osc-model resulted in biphasic calcium release. Such a response with an early peak phase followed by a lower plateau-like phase has been reported for calcium under physiological conditions (Lambert and Nahorski, 1990). Peak stimulated levels of Ca$^{2+}$ reached ~0.9 μM and were ~12-fold above basal.

Various mechanisms have been hypothesized for the periodic oscillations of cytosolic calcium seen in different cell types (Berridge, 1990). Positive feedback of calcium onto PLC β that generates InsP$_3$ underlies the calcium oscillation mechanism shown in certain experiments (Harootumian et al., 1991) and models (Meyer and Stryer, 1988). Although our non-Osc-model incorporated such feedback, it failed to display Ca$^{2+}$ oscillations. This probably resulted from different parameter representations across models and cell types. Hence, to study interactions between molecules in the inositol phosphate metabolic cascade and oscillatory calcium dynamics, we appended...
the Othmer-Tang calcium oscillation model (Tang et al., 1996) to our InsP₃ metabolism model. As mentioned previously in the Materials and Methods section, we refer to the resultant composite model as the Osc-model. The Othmer-Tang model is characterized by both a positive and a negative feedback from calcium onto its store release channel, the InsP₃ receptor. The channel in its conducting state has both calcium and InsP₃ bound. This mobilizes calcium from the ER to the cytosol. An excess calcium accumulation in the cytosol results in more calcium being bound to the receptor. This leads to channel inactivation.

The InsP₃ receptor response to increasing InsP₃ and calcium concentrations, in the nanomolar to micromolar range, was plotted as the normalized fraction of channels open versus the log of concentration (Fig. 3). The plots show a bell-shaped curve for InsP₃ receptor response to calcium, and a saturating curve reflects dependence of response on InsP₃ input. These are similar to the plots generated by Tang et al. (1996).

In our Osc-model basal calcium levels are ~20 nM, which is below the physiological range. Low basal calcium levels close to 0 nM are also used by Othmer and Tang for their model and are necessary for sustaining the oscillatory response. For maintaining calcium at such low levels we had to modify parameters for interactions between cytosolic and extracellular calcium, such as the store-operated cal-

Effects of glutamate stimulus intensity on InsP₃ buildup and calcium release

Cellular systems respond differently to different intensities of stimuli. The input parameters can modulate the output in terms of amplitude, time, and frequency of response. Here we analyze effects of different intensities of the glutamate stimulus on InsP₃-mediated calcium release. Fig. 5 A shows the calcium response within the non-Osc-model to 30-s pulses of glutamate stimuli of magnitude: 5 nM, 10 nM, 50 nM, 0.1 μM, and 0.5 μM. The time curves show that increasing stimulus concentrations affect both the amplitude and rise time of the calcium peak. These response parameters are plotted as a function of glutamate stimulus concentration in Fig. 5, B and C. The peak Ca²⁺ release increased nonlinearly with increasing stimulus intensity. At low glutamate concentrations, up to 10 nM, the Ca²⁺ change above basal was almost negligible. The magnitude of Ca²⁺ release increased rapidly in the 10- to 100-nM input range and then saturated at a peak ~1 μM Ca²⁺ at higher stimulus intensities. The inset of Fig. 5 B shows a plot for the peak InsP₃ response to the same stimulus protocol. The similarity between the stimulus function curves for peak Ca²⁺ and...
InsP₃ indicates that calcium release is a direct consequence of InsP₃ buildup. At higher concentrations of glutamate, there is higher occupancy of the mGluR, which leads to greater PLC β activation and buildup of more InsP₃. The plots show that there exists a relatively sharp stimulus threshold for InsP₃ generation and the subsequent Ca²⁺ release. In our model, this threshold is determined by the positive feedback of Ca²⁺ onto PLC β. PLC β has maximal activity when activated by both Gq and Ca²⁺. In this form it catalyzes sufficient InsP₃ formation to elevate cytosolic Ca²⁺ levels over 10-fold above basal.

Fig. 5 C shows that the rise time of the Ca²⁺ response decreases with increasing stimulus intensity. This temporal acceleration of the peak response also results from the positive feedback of Ca²⁺ onto PLC β, which leads to regenerative calcium release. We tested this by removing the Ca²⁺ facilitation of PLC enzyme activity from our model. This resulted in an almost constant rise time of the Ca²⁺ response at any stimulus strength (Fig. 5 C).

We also performed simulations to study the effect of stimulus intensity on the Osc-model. Two-minute pulses of 1 nM, 5 nM, 10 nM, 50 nM, 0.1 μM, and 0.5 μM glutamate were used as stimuli. The frequency of calcium oscillations was taken as readout of the stimulus effect (Fig. 5 D) (Berridge, 1990). Like the calcium release in the non-Osc-model, the relationship between stimulus input and simulation output, oscillation frequency in this case, is nonlinear. A 1 nM glutamate stimulus produced no oscillations, whereas 0.1 and 0.5 μM stimuli produced a saturating response frequency of 2/min. The increase in frequency of oscillations is expected because increasing stimulus intensities produce increasing InsP₃ in the cytosol. At higher InsP₃ concentrations the bell-shaped conductance curve of the InsP₃ receptor with respect to calcium shows faster response kinetics than at lower concentrations (Fig. 3 B). This results in the greater number of calcium spikes within the same time period.

**Effect of calcium oscillations on the InsP₃ metabolic cascade**

It is known that the pattern of calcium oscillations regulates downstream events such as gene expression (Dolmetsch et al., 1998). Using our simulations we wanted to investigate whether oscillatory calcium dynamics also modify InsP₃ and other inositol phosphates by feedback. This is important as the different physiological functions of the various inositol phosphates could be temporally affected by the oscillations. For this purpose, the Osc-model was subjected to a stimulus pulse of 0.5 μM glutamate for 2 min. The calcium response generated by the pulse is represented in Fig. 6 A. Within stimulus time cytosolic calcium displayed oscillations. The pulse amplitude in nanomolars is shown on the graph. For the non-Osc-model, rise time of the Ca²⁺ response decreases and amplitude increases with increasing stimulus intensity. (B) Peak Ca²⁺ responses (as obtained in A) are a nonlinear function of stimulus intensity. Inset shows a corresponding plot for the InsP₃ response. Glutamate concentrations (μM) represented on the x axis have a logarithmic scale. (C) Positive feedback of Ca²⁺ onto PLC β mediates the decrease in rise time of the Ca²⁺ response. The solid and dashed lines plot the rise time of the Ca²⁺ response as a function of stimulus intensity, in presence and absence of the positive feedback of Ca²⁺ onto PLC β, respectively. Stimulus protocol is same as that used for A. (D) Frequency of Ca²⁺ responses in the Osc-model are a nonlinear function of stimulus intensity. Stimuli used are 2-min pulses of 1 nM, 5 nM, 10 nM, 50 nM, 0.1 μM, and 0.5 μM glutamate, represented on a logarithmic scale on the x axis.
Ins(1,4,5)P3, Ins(1,4)P2, and Ins(4)P1 to the 2-min 0.5 μM glutamate pulse that generates the calcium response shown in Fig. 6. Thus, our simulations, which combine the biochemical connections in the InsP3 metabolic network transmit InsP3 oscillations to its dephosphorylated metabolites (Fig. 6 C). Like InsP3, the oscillations in Ins(1,4)P2 and Ins(4)P1 were also synchronous with calcium. Apart from Ins(1,4,5)P3, Ins(1,4)P2, and Ins(4)P1, we observed negligible or no oscillatory perturbations for other inositol phosphates (not shown). This is expected from the response time courses of inositol phosphates downstream of InsP3 phosphorylation (Fig. 2), which were so slow that oscillatory effects, if any, averaged out.

Hence, from our simulations we infer that the InsP3 oscillations that we observe to be synchronous with calcium are a result of the competition between InsP3 degradation and positive calcium feedback on InsP3 synthesis.

**Modification of Ins(1,4,5)P3 response and calcium release by Ins(1,3,4,5)P4**

Research on the role of Ins(1,3,4,5)P4 as a second messenger has over the years produced conflicting results. Recent evidence (Smith et al., 2000; Hermosura et al., 2000), however, supports the role of InsP4 in facilitation of InsP3-mediated calcium release. Whereas Smith et al. (2000) postulate that the function of InsP4 is attributed to its possible control over ER membrane integrity, Hermosura et al. (2000) show that InsP4 facilitation arises due to its metabolic effects. The latter view has also been supported in experiments on the *Xenopus* system (Sims and Allbritton, 1998). Both Ins(1,4,5)P3 and Ins(1,3,4,5)P4 share the same dephosphorylating enzyme: InsP 5-phosphatase. This enzyme converts Ins(1,4,5)P3 to Ins(1,4)P2 and Ins(1,3,4,5)P4 to Ins(1,3,4)P3. As InsP4 competes with InsP3 for the common enzyme, rapid InsP3 degradation is inhibited. Thus, presence of InsP4 results in a net gain in InsP3 and hence in facilitation of calcium release.

Our biochemical model has the necessary components involved in interactions between InsP3 metabolism and calcium release from stores. Hence, it was possible to simulate any effects that InsP4 might have on calcium release. This is
also pertinent in context to our simulations as our model is based on parameters from brain tissue studies that report high expression levels of InsP3 3-kinase.

To understand the role of InsP4, we made two modifications to our model that are represented in Fig. 7A. Case (i) depicts the reaction scheme for the actual metabolism documented in literature. Here, Ins(1,4,5)P3 is phosphorylated to Ins(1,3,4,5)P4 by InsP3 3-kinase, and both InsP4 and InsP3 are dephosphorylated by the same InsP 5-phosphatase1. Case (ii) is a modification of the basic scheme where we uncouple InsP3 and InsP4 dephosphorylation. Here too InsP3 is phosphorylated by InsP3 3-kinase, but only InsP3 is dephosphorylated by InsP3 5-phosphatase1. In this scheme a separate enzyme, InsP4 5-phosphatase converts InsP4 to Ins(1,3,4)P3. The parameter values for these new phosphatases such as initial concentration and enzyme inhibition by inositol high polyphosphates are identical to those for InsP 5-phosphatase1 in the basic scheme. The $K_m$ and $V_{max}$ values of these phosphatases for their respective substrates match the corresponding enzyme activities that InsP 5-phosphatase1 exhibits toward InsP3 and InsP4 in the actual case.

In case (iii) we replace the entire network of InsP3 metabolism with a single degradation step from InsP3 to inositol. The degradation rate was set at 1.75/s, which produces the same basal turnover of InsP3 as that in case (i) and (ii). The case (iii) model was used to characterize how InsP3 and calcium dynamics change in a model that does not account for detailed InsP3 metabolism. A comparison between the simulation outputs in case (ii) and (iii) can also suggest the importance of InsP4 within the metabolic cascade. If model behavior upon uncoupling of InsP3-InsP4 dephosphorylation resembles the behavior upon complete deletion of the metabolic network, it would imply that InsP4 represents the main effector of the entire network that determines the actual InsP3 response.

Case (i), (ii), and (iii) models were made for both the non-Osc-model and the Osc-model. A 30-s 0.5 μM glutamate pulse was used to stimulate each of the three non-Osc-models (Fig. 7B), and a 2-min pulse of similar amplitude was used for each of the Osc-models (Fig. 7C). We find that InsP4 indeed protects InsP3 against hydrolysis. For the non-Osc-model and Osc-model, respectively, the peak InsP3 response in the presence of the entire InsP3 metabolic cascade was (ii) 1.8-fold and 3-fold greater than the response in the absence of either InsP3-InsP4 dephosphorylation coupling or detailed metabolism. The high degree of overlap of the InsP3 responses for case (ii) and (iii) models suggests that in the amplitude domain, the influence of the detailed metabolism on InsP3 dynamics is primarily mediated by Ins(1,3,4,5)P4. Further for the non-Osc-model, the rise time for InsP3 decreases by ∼25% as a consequence of the protective role of InsP4. Thus, our simulations predict that metabolic effects of InsP4 modulate the temporal as well as the peak characteristics of the InsP3 response.

Given the facilitation of InsP3 buildup by InsP4, we next wanted to analyze whether this translates into a greater mobilization of calcium. Because in our model we have not incorporated any direct influence of InsP4 on any calcium channels, a facilitation observed in calcium release would imply a purely metabolic effect of InsP4. Fig. 8 represents the calcium responses in the non-Osc-model and the Osc-
that InsP₄ also facilitates the calcium release response via its metabolic interaction with InsP₃. Moreover, the shape of the Ca²⁺ release via its metabolic effect is represented as a gray line. (B–D) Responses of Ca²⁺ release within the Osc-model to cases (i), (ii), and (iii) of Fig. 7(A) respectively. Stimulus used in B, C, and D is identical to that for Fig. 7 C. 0.5 μM glutamate for 2-min (solid bars). An oscillation frequency of 2/min in B is reduced to 1.5/min in C and D.

model, that correspond to the InsP₃ responses shown in Fig. 7, B and C. Responses within the non-Osc-model (Fig. 8) show that InsP₄ also facilitates the calcium release response via its positive effect on InsP₃. The peak calcium release in the presence of the intact metabolic network was approximately threefold greater than those observed for the InsP₃ response (Fig. 7 B) in the non-Osc-model. This suggests that the magnitude of InsP₄ facilitation is greater for InsP₃-mediated Ca²⁺ release than for the metabolic response of InsP₄.

We also noted that the temporal responses of InsP₃ and Ca²⁺ upon InsP₃-InsP₄ dephosphorylation uncoupling were almost identical to their behavior upon complete deletion of the metabolic network (Fig. 7 B and 8 A). This implies that InsP₄ is the principal effector of the inositol phosphate cascade that modulates InsP₃ and, subsequently, Ca²⁺. InsP₄ kinetics are mainly governed by its synthesis via InsP₃ 3-kinase. The CaM and CaMKII activated forms of this enzyme have reversible kinetics in our model. If these enzyme activities are modeled using the Michaelis-Menten formulation, responses for case (ii) and (iii) model modifications differ markedly (not shown). For the irreversible enzyme scheme, the large influence of the first order network interaction with InsP₄ on InsP₃/Ca²⁺ is reduced, and secondary network interactions become necessary to explain their responses. However, because reversible enzyme kinetics are a more accurate representation of metabolic flow in this case, we suggest that behavior of InsP₃/Ca²⁺ is governed by the metabolic network mainly at the first rather than higher order interaction level.

Recent experiments (Zhu et al., 2000) show that Ins(1,3,4,5)P₄ can function as a frequency regulator of calcium oscillations in cells. Inhibition of InsP₃ 3-kinase can significantly reduce the oscillation frequency or abolish calcium oscillations, depending on the degree of enzyme inhibition. We also analyzed the effects of InsP₄ on calcium oscillations within our Osc-model. Fig. 8, B, C, and D represent the oscillatory calcium responses that correspond to the case (i), (ii), and (iii) time curves of InsP₃ in Fig. 7 C. The calcium curves show no modulation of amplitude but marked temporal differences. The ineffectiveness of InsP₄ in mediating a change in peak calcium is not surprising. Maximal calcium released per spike is tightly regulated by positive and negative feedback of calcium onto the InsP₃ receptor. Only drastic alterations in the cytosolic or ER calcium buffering would be expected to vary the peak height of calcium per oscillation. At the same time, the frequency modulating effect of InsP₄ can also be explained. Fig. 7 C shows that the peak InsP₃ response for case (i) is threefold greater than the response when InsP₃-InsP₄ metabolism is decoupled. A higher InsP₃ buildup corresponds to faster response kinetics of the InsP₃ receptor (Fig. 3 B). Thus, higher InsP₃ levels generate more frequent calcium spikes within the same time than lower InsP₃ levels. Our simulations also display this phenomenon. The intact InsP₃ metabolism model with coupled InsP₃-InsP₄ dephosphorylation shows maximal oscillation frequency of 2/min (Fig. 8 B). On the other hand, the frequency drops to 1.5/min for both case (ii) and (iii) in which InsP₄ does not protect InsP₃ from hydrolysis (Fig. 8, C and D). Our model does not
include any direct Ca\(^{2+}\) mobilizing effects of InsP\(_4\). Thus, we suggest that the metabolic function of InsP\(_4\) to protect InsP\(_3\) from hydrolysis is sufficient to generate higher frequency oscillations of calcium.

**Function of InsP\(_4\) as a “coincidence detector”**

Natural stimulus inputs are often composed of repetitive pulse patterns. It adds to the response capabilities of a cellular system if it can detect stimuli spaced in time, apart from stimuli of different intensities. It has been suggested that InsP\(_4\) may function as a “coincidence detector” for the InsP\(_3\) signaling pathway (Irvine, 2001; Parker and Ivorra, 1991). This means that an otherwise subthreshold stimulus can generate a calcium response if it is coincident with presence of some InsP\(_4\) that has remained in the system from a similar previous stimulation. Hermosura et al. (2000) performed a key experiment to demonstrate this phenomenon. They used a paired pulse stimulus protocol wherein the GPCR agonist concentration used was below the threshold for any measurable calcium release. When the second pulse of agonist followed the first pulse within a particular time frame, an intracellular calcium signal was detected. Such facilitation was not seen when the InsP\(_3\) 3-kinase was pharmacologically blocked.

We investigated whether our Osc-model, which has a more accurate representation of InsP\(_3\) receptor kinetics than the non-Osc-model, can produce facilitation of paired subthreshold stimuli. As shown in Fig. 9, our pulse protocol consisted of an initial 20 nM glutamate pulse for 20-s followed by a second identical pulse after 90, 100, 110, 120, or 130 s. A single 20-nM glutamate pulse of 20-s duration, is incapable of generating a calcium spike. If the second 20-nM pulse succeeds the first within 90 s, a calcium response is always generated. This is because the subthreshold levels of InsP\(_3\) generated in the system by the two stimuli sum to become suprathreshold and release calcium. After 90 s of the first stimulus, InsP\(_3\) levels have significantly reduced. At this time however, InsP\(_4\) levels remain in the system due to its slower degradation than InsP\(_3\) (Fig. 2). If the 20 nM glutamate pulse is now given, this residual InsP\(_3\) can perform its metabolic function to protect any new InsP\(_3\) from degradation. The higher levels of InsP\(_3\) generated during the second pulse would then result in calcium release. Fig. 9 A shows such an effect, wherein the four calcium spikes displayed correspond to stimulus interpulse intervals of 90, 100, 110, and 120 s, respectively. The corresponding response of InsP\(_3\) shows that its levels are potentiated during the second glutamate stimulus (Fig. 9 B). No calcium release is seen for the last stimulus protocol with a pulse spacing of 130 s. By this time, residual InsP\(_4\) from the first stimulus has degraded below the threshold level at which it can protect new InsP\(_3\) from fast hydrolysis. We further confirmed that the paired pulse facilitation was due to the metabolic function of residual InsP\(_4\) during the second pulse, and not simply due to summing of InsP\(_3\) responses across temporally close stimuli. For this, we subjected our version of the Osc-model with decoupled InsP\(_3\)-InsP\(_4\) metabolism (Fig. 7 A, case (ii)) to the same stimulus protocol as described above. This model did not
show any calcium spiking for the paired stimuli (Fig. 9 D). Our simulations thus show that InsP₄ can serve as a short lived “memory” for a previous exposure of a system to an InsP₃ generating stimulus.

It is also interesting to note that during the second stimulation, InsP₄ indirectly also facilitates its own levels (Fig. 9 C). Such an effect could also be important for calcium release kinetics if InsP₄ has a direct action on any calcium channel, such as the InsP₃ receptor (Hermosura et al., 2000).

**Effects of high inositol polyphosphates on inositol phosphate metabolism**

We wanted to analyze how a significant perturbation in the inositol high polyphosphates affects the dynamics of the metabolic network. Stimulation of IHPs does occur in cells especially during the S phase of the cell cycle, but this does not require activation of PLC β or generation of Ins(1,4,5)P₃ (Balla et al., 1994). Our simulations have also shown that a stimulus pulse that first generates Ins(1,4,5)P₃ and subsequently activates other components of the inositol phosphate metabolic network, does not propagate up to the IHPs (Fig. 2). Thus, to study effects of elevation in IHP levels we applied direct stimulation to the system in the form of a 30-s pulse of 100 μM InsP₅ (twice its original concentration).

The most striking differences between stimulation via InsP₅ and InsP₆ lay in the temporal domain. The time taken by the whole system to regain steady-state post-InsP₅-stimulation was ~2.5 h compared with 10 to 15 min required after InsP₃ stimulation. InsP₅ and InsP₆ have very large cellular pools and slow metabolism kinetics. Hence, they take a long time to equilibrate, and in turn the entire system does not attain steady state for over 2 h. However, given that the IHPs are so important for cell biological processes, direct regulatory mechanisms must exist to modulate their responses and to hasten their equilibration. Some reports suggest that cell cycle factors may regulate the IHP responses (Balla et al., 1994; Guse et al., 1993). Our model could not incorporate any regulation of IHPs reported in literature, as their exact molecular details are not characterized. Nevertheless, it is clear that such regulation would play an important role in the temporal dynamics of stimulated IHPs.

At equilibrium InsP₅ is dephosphorylated to both Ins(1,4,5,6)P₄ and Ins(3,4,5,6)P₄, generating approximately equivalent levels of each. Upon InsP₅ stimulation we observed a ~30-fold elevation in Ins(3,4,5,6)P₄ levels and an approximately twofold change for Ins(1,4,5,6)P₄ (Fig. 10, A and C). The greater stimulation of Ins(3,4,5,6)P₄ than Ins(1,4,5,6)P₄ was due to the faster rate of InsP₅ 1-phosphatase compared with InsP₃ 3-phosphatase in our model. Despite this, both InsP₅ responses were much larger than for InsP₃-mediated stimulation. Ins(1,3,4,5)P₄ and Ins(1,3,4,6)P₄ were also stimulated by approximately twofold. However, the Ins(1,3,4,5)P₄ response was insignificant compared with its 60-fold stimulation via the PLC β route, as InsP₅ did not trigger InsP₃ 3-kinase activation via Ca²⁺ release.

Stimulation of inositol pyrophosphates, PP-InsP₅, and bis-PP-InsP₄ was ~1.5-fold (Fig. 10 B). Such minimal activation perhaps implies that these molecules have further distinct stimulatory mechanisms. Indeed, Safrany and Shears (1998) have shown that turnover of bis-PP-InsP₄ is specifically regulated by β₂-adrenergic receptors through a cAMP-mediated mechanism. From Fig. 10, it was also evident that InsP₄s that have direct interaction with the lower inositol phosphates have faster time responses to InsP₅ stimulation than molecules isolated within the IHP pool, which equilibrate slowly.

The lower phosphates from InsP₃ down were negligibly affected by the 100 μM InsP₅ pulse (not shown). The buildup of these inositol bisphosphates and monophosphates was hampered because InsP₅ and InsP₆ directly inhibit the enzymes that synthesize them, specifically InsP₅ 5-phosphatase1 and InsP 4-phosphatase. It was earlier observed from the InsP₃-mediated stimulation (Fig. 2) that the IHPs probably form a separate biochemical pool, which is unaffected by the lower phosphates. Stimulation via InsP₅.

![FIGURE 10 Responses of inositol phosphates to a 30-s stimulus of 100 μM InsP₅. Time on the x axis represents 2.2 h of simulation run (post-stimulation). (A) Ins(3,4,5,6)P₄ undergoes the largest fold response. (B) Responses of inositol pyrophosphates: PP-InsP₅ and bis-PP-InsP₄ are slow and ~1.5-fold in magnitude. (C) Inositol tetrakisphosphates: Ins(1,4,5,6)P₄, Ins(1,3,4,5)P₄, and Ins(1,3,4,6)P₄ have ~twofold responses that recover quickly from the perturbation.](image)
reveals that large perturbations in the higher phosphate pool are biochemically networked to the lower phosphate pool in such a way that the lower phosphates remain at basal levels. Thus, we conclude that the inositol high and low polyphosphate pools are indeed distinct and fail to generate fluctuations in each other when independently stimulated.

### Sensitivity analysis of the InsP₃ metabolism model

The model that we have built uses parameter constraints from biochemical test-tube experiments that estimate cellular concentrations and characteristic $K_m$ and $V_{max}$ values of enzymes. These may not necessarily mimic the in vivo environment and may vary depending on the sensitivity of the experimental technique used for the analyses. In such a situation it is useful to test the robustness of our model by subjecting every parameter used to graded alterations spanning a wide range. This gives an idea of the performance of the model across changes within the system. Further, the outliers emerging from such an analysis suggest the regulatory inputs in our model to which the simulations are extremely sensitive.

We scaled all parameters used in our model from tenth to 10-fold their original value (see Materials and Methods). Three sample sensitivity plots are shown in Fig. 11. In Fig. 11 A, changes in InsP₃ output are taken as readout of the sensitivity analysis performed on all reaction steps used in the model. $K_d$ was the parameter being altered. As expected, major changes in InsP₃ resulted from changes in the $K_d$ of reactions directly involved in its synthesis. These include activation of G protein by glutamate binding to its receptor (Activate-Gq), basal activation of G-protein (Basal-Act-G), activation of PLC β by Gα-GTP binding (Act-PLC-by-Gq), and the composite basal rate of PIP₂ hydrolysis due to other PLCs in the system (basal). It was observed that InsP₃ was less sensitive to changes within the metabolic cascade.

The log scatter plot in Fig. 11 B represents variations in the calcium output as a result of perturbations in the $V_{max}$ of all enzymes used in our model. Calcium release critically depends on calcium feedback onto PLC β (PLCb-Ca-Gq) and also on metabolism of InsP₃. Simulations showed sensitivity toward both the InsP₃ 5-phosphatase isoforms that dephosphorylate InsP₃ to Ins(1,4)P₂ (ip3_5pase1, ip3_5pase2), as well as to the activated InsP₃ 3-kinase that converts InsP₃ to Ins(1,3,4,5)P₄ (ip3-3k*). It was interesting to note that calcium output was also sensitive to $V_{max}$ for the Ins(3,4,5,6)P₄ 1-kinase enzyme activity (ip4-1k-off) that phosphorylates Ins(3,4,5,6)P₄ to InsP₄. On stimulation, this reversible enzyme functions such that accumulation of its substrate, Ins(3,4,5,6)P₄ is significant. Elevation in InsP₄ levels inhibits calcium stimulated chloride efflux across the plasma membrane. If InsP₄ 1-kinase also affects calcium levels, then an indirect means of inhibition of the calcium-driven chloride channel (apart from direct binding of InsP₄ to the channel) may exist. Ca²⁺ release was also sensitive to the rate of calcium efflux to the extracellular space (Ca-pump-out).

Fig. 11 C shows a sensitivity analysis performed on the Osc-model. Alterations made to the initial concentrations of kinetic pools are monitored using the frequency of oscillations as readout of the output. We observed that Ca²⁺ oscillations were directly affected by the concentration of kinetic pools that determine receptor-based PLC β activation, such as the metabotropic glutamate receptor (mGlur), PIP₂, G-protein (G-GDP), and PLC β. Sensitivity was also seen toward pools that determine equilibration of cytosolic calcium with ER and extracellular calcium, such as leak across the ER membrane (ER_leak), ER-Ca²⁺-ATPase pump (ER_pump), calcium extracellular efflux pump (Ca-EPump), InsP₃ receptor (IP3R), and the calcium buffer capacity of the ER (Calsequestrin). We note that sensitivity of the nonoscillatory calcium response toward most of these Ca²⁺ equilibration-determining pools was small. This was expected because generation of calcium oscillations requires a very tight combination of all parameters that determine calcium levels. Tenfold changes in these parameters can completely abolish oscillations. Within the metabolic cascade Ca²⁺ oscillations were most sensitive to the two isoforms of InsP₃ 5-phosphatase (IP 5pase1, IP3 5pase2) and InsP₃ 3-kinase (IP3 3K). As seen earlier, these same enzymes were also the key modulators of the nonoscillating calcium output.

Additional sensitivity analysis plots for single parameter variations are presented in Supplementary Information. In these analyses also, the sensitive parameters were same as those summarized above, implying that the sensitive subset of the total parameter space remains constant across different analyses. Further, we found that for enzyme-based parameters, the sensitive subset is independent of mechanistic details for the enzymes, i.e., whether they have been modeled as per the reversible scheme or the Michaelis-Menten scheme.

We also performed two global sensitivity analyses (not shown) to assess the effects of simultaneous variations in a combination of parameters on our model (see Materials and Methods). For the non-Osc-model, rate scaling for all enzymes and reactions as a function of 10°C rise in temperature resulted in ~60% positive change in response peaks of Ins(1,4,5)P₃ and Ins(1,3,4,5)P₆, whereas a 10°C temperature drop led to ~30% negative change in their outputs. Peak Ca²⁺ release was more tolerant to the temperature variation with ~30% to 40% change for the ±10°C shift, although its basal levels were influenced twofold. Ins(1,3,4)P₄ was found to be much less sensitive to temperature effects. The temperature-based perturbation also shifted response latencies for the network components by 30 to 60 s. The Osc-model was less robust to temperature perturbations and could maintain the control oscillatory response only within a ±2.5°C temperature range. Oscillations were found to become more frequent and dampen at lower temperatures.
The second global sensitivity analysis based on ATP depletion showed InsP$_4$ and Ca$^{2+}$ to be most sensitive to the 85% drop in ATP levels, for which their peak stimulated levels dropped by 80% and 75%, respectively. Once again, the Osc-model was more sensitive to this global perturbation. Oscillations became more frequent and dampened at 50% ATP levels and disappeared for the 85% ATP depletion where basal cytosolic Ca$^{2+}$ had increased 1.5-fold.

These global analyses thus show that our model is fairly robust to physiologically expected system alterations. Single parameter variation analyses lead us to conclude that the model is sensitive to a defined set of parameters that are a small subset of the entire parameter space. Interestingly, these same parameters have been extensively studied experimentally due to their significant influence on our network of interest (Zhu et al., 2000; Hermosura et al., 2000; Albrecht et al., 2002). Thus, despite the limitation that our model is based on a large parameter space (Table 2), which itself has certain degree of error attached to it arising from experiments, we suggest that our results do reflect characteristics of the biological system.

DISCUSSION

We have built a mass action kinetic model for the inositol phosphate metabolic network. This model accounts for most interactions among inositol phosphates ranging from InsP$_1$ isoforms to the InsP$_8$ equivalent bis-PP-InsP$_4$ that have been biochemically characterized and reported in literature. Further, our model couples the InsP$_3$ metabolic cascade with the synthesis of InsP$_3$ via G-protein coupled metabotropic glutamate receptor-driven PLC activation and with the function of InsP$_3$, i.e., release of Ca$^{2+}$ from intracellular Ca$^{2+}$ stores. Using varied glutamate stimulus protocols at the mGluR, we have studied the cross-talk among the members of the inositol phosphate network. By comparing and contrasting the stimulus response outputs, we suggest interpretations as to how the network components influence and modulate each other. We have also investigated how the metabolism of InsP$_3$ affects its calcium release function. We report three major results of our simulations. 1) In our model Ca$^{2+}$ released through the InsP$_3$ receptor feeds back onto PLC to regenerate InsP$_3$ and hence mobilize more Ca$^{2+}$. Thus dynamics of the Ca$^{2+}$ waveform in turn influence the response pattern of the various inositol phosphates that branch out from InsP$_3$. We have also incorporated detailed kinetics of the InsP$_3$ receptor, which produce cytosolic Ca$^{2+}$ oscillations to study fine aspects of interactions between Ca$^{2+}$ and inositol phosphate derivatives. 2) In our simulations Ins(1,3,4,5)P$_4$ emerges as an important modu-

![FIGURE 11 Scatter plots for sensitivity analysis of the InsP$_3$ metabolism model. (A and B) Represents the non-Osc-model; (C) represents the Osc-model. Both axes have logarithmic scales, except in C where only the x axis is logarithmic. The x axis represents the scale factor by which a parameter is altered with boundaries of 0.1 and 10 times the original value of the parameter. The y axis represents the normalized fold change in output due to the parameter scaling. Outliers in the scatter plots correspond to the parameters to which the simulations are very sensitive. Some of the most significant outliers have been marked with solid triangles and labeled alongside. Multiple parameter labels beside a solid triangle imply that the output is equally sensitive to all these parameters. (A) Normalized fold changes in Ins(1,4,5)P$_3$ output as a function of reaction $K_d$ scaling. (B) Normalized fold changes in Ca$^{2+}$ output as a function of enzyme $V_{max}$ scaling. (C) Normalized changes in the frequency of Ca$^{2+}$ oscillations as a function of initial concentration scaling.](Biophysical Journal 83(3) 1298–1316)
lator of the temporal dynamics of other inositol phosphates. We also show that Ins(1,3,4,5)P4 facilitates Ca2+ release with respect to both the timing and peak of the Ca2+ response. 3) Our simulations further demonstrate that InsP4 serves as a coincidence detector for paired subthreshold stimuli, and enables Ca2+ release in situations where it would not have otherwise been possible.

Interactions within the inositol phosphate metabolic network

The current simulations highlight an important property of the elaborate interlinking within the metabolic network. This is that if flux of one of the components becomes dominant over others, that particular metabolite can modulate downstream enzyme activities beyond the primary and secondary interaction level. Ins(1,3,4,5)P4 undergoes up to 60-fold stimulation in our model because of both the high expression levels of its synthesizing enzyme InsP3 3-kinase as well as its strong positive regulation by Ca2+-CaM and CaMKII. We observe that under the influence of this InsP4, inositol phosphates downstream of it, i.e., Ins(1,3,4)P3 and its dephosphorylated forms exhibit rectangular waveforms. These inositol phosphates “freeze” at their respective peaks as long as InsP4 levels are elevated and their synthesizing enzymes are substrate-saturated. Interestingly, the Ins(1,3,4)P3 wavefront propagates through the network such that temporal response of Ins(3,4,5,6)P4 also has a long elevated plateau phase. Such prolonged elevation can provide these molecules with a relatively long-term capacity for cellular function, which endures even after the stimulus is removed. This is in sharp contrast to the transient spike of Ins(1,4,5)P3, which is transduced to a Ca2+ signal within stimulus time.

We also show that Ins(1,3,4,5)P4 not only affects inositol phosphates downstream of it, but also influences Ins(1,4,5)P3. This is attributed to the unique property of InsP 5-phosphatase1 to dephosphorylate both Ins(1,4,5)P3 and Ins(1,3,4,5)P4. In its competition for the same enzyme, InsP4 ends up slowing InsP3 hydrolysis and enhancing its buildup. Indeed, when we decouple the two enzyme activities of the 5-phosphatase, we see that not only is the peak InsP3 response diminished, but its rise time is also delayed.

With respect to the inositol high polyphosphates, consistent with experimental findings, our simulations show negligible effects of a GPCR stimulus. It has been suggested that the IHPs are a biochemical pool separate from the lower phosphates and are uniquely modulated by other cellular mechanisms. Our investigation of the effects of direct IHP stimulation unexpectedly showed negligible perturbations in the lower phosphates. Careful examination of the model revealed that IHPs are networked such that they directly inhibit the enzymes that synthesize the lower phosphates. Thus, when independently stimulated, neither higher phosphates nor lower phosphates can perturb each other. Higher phosphates function as regulators of vesicle trafficking, whereas lower phosphates are directly or indirectly involved in gating ion channels such as the InsP3 receptor. The segregation of their metabolic flux pools within the inositol phosphate network may be important for the temporal control of their discrete functions.

Interactions between InsP3 metabolism and calcium release

In our model, InsP3 metabolism and Ca2+ release from the ER have both feed forward and feed backward links. Metabolites of InsP3 such as Ins(1,3,4,5)P4 can influence Ca2+ by modifying InsP3 dynamics. Ca2+, on the other hand, can affect inositol phosphates via InsP3 regeneration through PLC β feedback activation or CaM/CaMKII-mediated InsP3 3-kinase activation or InsP 1-phosphatase inhibition. We have explored the possible outcomes of such interactions in our model. We show that positive feedback of Ca2+ onto PLC β is critical for the Ca2+ release function. Without this, stronger stimulation of the system does not result in a greater amount of Ca2+ release or a faster response. Further, when Ca2+ dynamics are oscillatory, this positive feedback to InsP3 generation competes with InsP3 degradation to bring about oscillations in InsP3 and its lower phosphate derivatives. Thus, even though we did not include oscillations in InsP3 as a prerequisite for Ca2+ oscillations in our model, we observe InsP3 oscillations as a result of the Ca2+ feedback. We propose that in cells both InsP3 and Ca2+ oscillations may coexist, but it is not necessary that oscillations in Ca2+ depend on oscillation in InsP3.

How does the metabolic cascade influence Ca2+ dynamics? We find that the metabolic coupling of Ins(1,4,5)P3 to Ins(1,3,4,5)P4 via InsP 5-phosphatase1 increases the responsiveness of Ca2+ to stimuli. The peak Ca2+ response is reduced, and its rise time is prolonged in absence of the dual substrate specificity of the InsP 5-phosphatase for InsP3 and InsP4. This modification also diminishes the frequency of Ca2+ oscillations. Experimentally it has been demonstrated that the facilitatory function of InsP3 is abolished if InsP3 3-kinase, which synthesizes Ins(1,3,4,5)P4, is pharmacologically inhibited (Hermosura et al., 2000). As this strategy completely blocks generation of Ins(1,3,4,5)P4, it would also drastically influence all inositol phosphates downstream of it. This can further adversely affect the various cellular functions regulated by these inositol phosphates. Our simulation strategy to replace the common 5-phosphatase for InsP3 and InsP4 by two separate enzyme activities, exclusively affects Ins(1,3,4,5)P4 and not any other component of the metabolic network. Even though such enzyme modification highlights InsP4 function more clearly, it may be technically difficult to emulate in experiments. Further, we find Ca2+ release upon complete exclusion of the metabolic network linked to InsP3 in our model to be equivalent to that upon InsP3-InsP4 uncoupling. This equivalence emerges only when activated InsP3
3-kinase is modeled appropriately as a reversible enzyme. Thus, accounting for non-Michaelis-Menten enzyme kinetics in our model allows us to propose that only primary level interactions within the metabolic cascade, i.e., InsP$_3$ to InsP$_4$, may be responsible for potentiation of the InsP$_3$-mediated Ca$^{2+}$ response.

Finally, we show that our system can exploit the inositol phosphate networking to broaden its response repertoire to GPCR stimuli. If any remnants of Ins(1,3,4,5)P$_4$ from a brief stimulation persist till a second stimulus arrives, a Ca$^{2+}$ response is generated even if the stimuli are sub-threshold for Ca$^{2+}$ release. This directly stems from the “protective” function of InsP$_4$ that can modify InsP$_3$ dynamics from one stimulus to another.

Thus, we conclude that interactions between metabolism of InsP$_3$ and Ca$^{2+}$ release are important for this signaling cascade to process diverse patterns of stimuli.

**Interpretations and limitations**

We are aware that the current simulations are an approximation to the actual biology of the system. One of the major limitations of this mass action model is that it relies on test-tube experiments that are not an accurate reflection of in vivo interactions. These experiments cannot mimic intracellular environments that are subject to modifications by various cellular features such as cystoskeletal elements, membrane localization, and organelle compartmentalization. Further, the biochemical parameters may significantly change over the years as the technology for experimentation improves.

Despite these limitations, our model replicates known physiological properties of inositol phosphates reasonably well. Point mass action kinetics may be a good first approximation because InsP$_3$ and its derivatives are likely to diffuse through most of the cell on the time scales of interest. Further, the sensitivity analysis of our simulations suggests that our model is quite robust across parameter variations that have arisen in literature due to different technical approaches to the same experiments. Thus, our model and its predictions concerning inositol phosphate dynamics and their interactions with InsP$_3$-mediated Ca$^{2+}$ release may be a useful step toward understanding of the fundamental properties of this complex metabolic network.

J.M. is supported on a NCBS/TIFR fellowship. U.S.B. is a Senior Research Fellow of the Wellcome Trust. We thank Dr. S. B. Shears for advice on parameter values.

**REFERENCES**


Mishra and Bhalla

Mishra and Bhalla

Mishra and Bhalla