Analyzing the Behavior of Neuronal Pathways in Alzheimer’s Disease Using Petri Net Modeling Approach

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Alzheimer’s Disease (AD) is the most common neuro-degenerative disorder in the elderly that leads to dementia. The hallmark of AD is senile lesions made by abnormal aggregation of amyloid beta in extracellular space of brain. One of the challenges in AD treatment is to better understand the mechanism of action of key proteins and their related pathways involved in neuronal cell death in order to identify adequate therapeutic targets. This study focuses on the phenomenon of aggregation of amyloid beta into plaques by considering the signal transduction pathways of Calpain-Calpastatin (CAST) regulation system and Amyloid Precursor Protein (APP) processing pathways along with Ca2+ channels. These pathways are modeled and analyzed individually as well as collectively through Stochastic Petri Nets for comprehensive analysis and thorough understating of AD. The model predicts that the deregulation of Calpain activity, disruption of Calcium homeostasis, inhibition of CAST and elevation of abnormal APP processing are key cytotoxic events resulting in an early AD onset and progression. Interestingly, the model also reveals that plaques accumulation start early (at the age of 40) in life but symptoms appear late. These results suggest that the process of neuro-degeneration can be slowed down or paused by slowing down the degradation rate of Calpain-CAST Complex. In the light of this study, the suggestive therapeutic strategy might be the prevention of the degradation of Calpain-CAST complexes and the inhibition of Calpain for the treatment of neurodegenerative diseases such as AD.

Keywords: Calpain, CAST, calcium, PKC, APP, Stochastic petri net, Alzheimer disease, Amyloid beta

1. INTRODUCTION

Alzheimer’s disease (AD) is a neurodegenerative disorder which has impacted nearly 44 million people around the world and this number is still increasing. AD is the leading cause of dementia in the old age (Ashford, 2004). Unfortunately, it is diagnosed only in one out of four people living with the disease. Clinical characterization of AD includes memory loss and cognitive impairment which further lead to damaged behavioral activities and render a person completely dependent.
AD establishes over time with the appearance of pathological emblems which are senile plaques and neurofibrillary tangles. These lesions comprise extracellular deposits of Amyloid beta (Aβ) (Selkoe, 2000; Golde, 2005; Tam and Pasternak, 2012) and intracellular self-gathered clumps of tau proteins (Lee et al., 2001), respectively. Aβ is a 40–42 amino-acids long peptide which is formed after the proteolytic cleavage of Amyloid Precursor Protein (APP) (Selkoe, 2000; Golde, 2005; Tam and Pasternak, 2012). Previous studies have shown that Aβ monomers are initially non-toxic but their conversion to oligomers makes them toxic (Volles and Lansbury, 2002; Walsh and Selkoe, 2004). Eventually, the abnormal accumulation of oligomers form plaques (Walsh et al., 2002) and in extracellular space (Trojanowski and Lee, 2000; Walsh et al., 2000). Aggregation of senile plaques and neurofibrillary tangles cause neuronal cell death and synaptic failure (Tiraboschi et al., 2000; Selkoe, 2002). During the last two decades, several lines of studies have pointed toward the imbalance between Aβ production and its clearance plays a central role in pathogenesis of AD. Since 1992, this hypothesis has earned acquiescence (Hardy and Higgins, 1992) and is known as “Amyloid cascade hypothesis (ACH)”. It suggests that Aβ and processing of APP are crucial in neuro-degeneration. In AD, aggregation of Aβ is the first step leading toward the formation of senile plaques (Hardy and Selkoe, 2002; Vassar, 2005). APP is a type1 trans-membrane protein produced in ER (Greenfield et al., 1999; Roussel et al., 2013). In neurons, production and metabolism of APP occurs rapidly which makes it a crucial element in neuro-pathogenesis (Lee et al., 2008). The main APP proteolytic processing steps occur at the cell surface and Trans-Golgi networks (TGNs). Proteolysis of APP can occur through the so-called non-amyloidogenic and amyloidogenic Pathways (Figure 1). The first step of non-amyloidogenic pathway is carried out by the enzyme alpha (α)-secretase that breaks down APP into soluble Amyloid precursor protein alpha (sAPPα) and alpha C-terminal fragment (αCTF / CTF83). The catalysis by α-secretase is imperative as it cuts APP within Aβ domain which blocks Aβ formation (Lichtenthaler, 2011). This initial step can also be driven by the beta (β)-secretase / β-site APP-cleaving enzyme (BACE), a transmembrane aspartyl protease (Vassar et al., 1999; Haass, 2004) (Figure 1), which constitute amyloidogenic pathway. BACE is a crucial enzyme, that acts as a rate limiting protein in Aβ generation. It breaks down the APP into soluble Amyloid precursor protein beta (sAPPβ) and beta C-terminal fragments (βCTF / CTF99) (Cai et al., 2001). The CTFs are intermediate products of the first step in both pathways which remain attached to the membrane and they are further cleaved by

![Figure 1](https://example.com/fig1.png)

**FIGURE 1** APP and processing products: APP is synthesized in the ER and then transported to the trans-Golgi-network (TGN) where it is cleaved by secretases. In non-amyloidogenic pathway (left), cleavage of APP by α-secretase results in the generation of sAPPα and C-terminal fragments CTF83 which is further cleaved by γ-secretase into p3 and AICD. Proteolysis by α-secretase prevents Aβ production as the cleavage site in APP is within the Aβ domain. In amyloidogenic pathway (right), APP is cleaved into sAPPβ and CTF99 by β-secretase / BACE activity. Furthermore, CTF99 breaks down into AICD and Aβ by γ-secretase activity. Aβ fragments oligomerize and fibrillize into plaques.
gamma (γ)-secretase (Zhang et al., 2011). In non-amyloidogenic pathway, the fragment α-CTF is cut down by γ-secretase into p38 and the Amyloid Precursor Protein Intracellular Cytoplasmic / C-terminal Domain (AICD). While in amyloidogenic pathway, γ-secretase degrades the β-CTF into Aβ and AICD (O’Brien and Wong, 2011) (Figure 1).

The Biological Regulatory Networks (BRN) of APP processing, depicted in Figure 2, is also built from Figure 1. APP processing depends on sequential cleavage by three secretases (α/β-secretase and γ-secretase). In normal conditions, α-secretase residing at the plasma membrane is constitutively active for APP coming to the cell surface and thus favoring non-amyloidogenic pathway (De Strooper and Annaert, 2000). Though there is an interesting fact about APP proteolysis that none of the secretases show special substrate specificity toward APP. There are several transmembrane proteins such as cell surface receptors and ligand, growth factors and cytokines besides APP which undergo ectodomain shedding by enzymes with α-secretase activity (Annaert and Saftig, 2009). In the same manner, BACE shows low affinity toward APP and it is not its exclusive physiological substrate (DeStrooper et al., 2006; Hu et al., 2006). Many observations highlight that in healthy cells APP is frequently processed through non-amyloidogenic pathway to resist amyloid generation while it is altered in pathological conditions (De Strooper and Annaert, 2000). Abnormal processing of APP is stated to be the first and fundamental step in plaques formation in AD pathogenesis (Jonsson et al., 2012). In neuropathological conditions, BACE affinity toward APP increases two folds which leads to enhanced Aβ production (Yang et al., 2003; Li and Südhof, 2004). Recent studies on transgenic mice model have shown that BACE activity is modulated by Calpain activation in AD pathology (Liang et al., 2010). Calpain-Calpastatin system also plays a key role in neurodegeneration. Transgenic mice models have shown that over expression of APP, increased production of Aβ, inhibition of Calpastatin (CAST) and activation of Calpain increase neuronal degeneration in AD (Higuchi et al., 2012).

Calpains are protein clan of cysteine/ thiol proteases and their activity depends on Ca^{2+} concentration (Ferreira, 2012). The most studied Calpains, μ(µ)-Calpain (Calpain1) and m-Calpain (Calpain2) are present abundantly in neurons, central nervous system (CNS) and glial cells. Though their distribution differs, Calpain1 is ubiquitous and expressed more in neurons while Calpain2 is present in glial cells (Ono and Sorimachi, 2012; Santos et al., 2012). Calpain1 requires micro-molar concentration of Ca^{2+} (10–50 µM), while Calpain2 is activated by mili-molar concentration of Ca^{2+} (250–350 µM) in vitro (Goll et al., 2003; Ryu and Nakazawa, 2014). Ca^{2+} plays important role in ensuring the cell’s vital functions. In addition to calcium, Calpain is tightly regulated in the cell by CAST which is also ubiquitous and solely

![Figure 2](https://example.com/image2.png)
a specific endogenous inhibitor for both Calpains (Melloni et al., 2006).

CAST is reported as an explicit suicide substrate for Calpain (Yang et al., 2013). The proportion of CAST in a cell is normally larger than Calpain, its ratio with location is crucial in controlling the extent of activation of Calpain within a cell (Todd et al., 2003). CAST interacts with Calpain at different stages i.e., first it constrains Calpain at the membrane where pro-Calpain is attached then it interacts with active Calpain inside cytosol (Hanna et al., 2008). CAST forms a reversible complex with Calpain at both the sites. At membrane, the reversible complex breaks down when Ca$^{2+}$ influx increases to release Calpain. Inside cytosol, Calpain undergoes autolysis to attain active conformation. In response, CAST changes its cellular distribution to make itself widely available in the cytoplasm to counter active Calpain (Todd et al., 2003). Both active Calpain and CAST rejoin in a reversible complex to resist persistent activity of Calpain (De Tullio et al., 1999). Active Calpain modulates CAST by slowly digesting it into small inactive fragments which results in plethora of Calpain in cell leading to pathological condition (Averna et al., 2001b; Tompa et al., 2002) (Figure 3). It has been reported that in AD CAST becomes depleted from different regions of the brain as compared to healthy aged brain (Rao et al., 2008). It has also been observed that by controlling Calpain, CAST is indirectly preventing cell membrane damages induced by high Ca$^{2+}$ and $A\beta$ peptide (Vaisid et al., 2008).

**FIGURE 3** Different calcium channels work in harmony to establish homeostasis in neurons. Calcium influx is controlled by voltage gated (VGCCs) or receptor-ligand based (NMDAR, GPCR) channels. ER also release Ca$^{2+}$ into the cytoplasm through inositol-1,4,5-trisphosphate (IP3R) and ryanodine receptors. Calcium efflux is carried out by energy (ATP) dependent channels such as plasma membrane calcium ATPase (PMCA), sodium-potassium ATPase (NKA) and sodium-calcium exchanger (NCX) channels. Calcium homeostasis influences Calpain-CAST system. At membrane, Calpain is bound to CAST to form mComplex at low Ca$^{2+}$ level. At high Ca$^{2+}$ concentration, Calpain is released into cytoplasm and autolysed to active form ACalp that again forms complex with CAST (cComplex). Gradually the complex breaks down and releases ACalp which enhances Plaque accumulation and LTP events.
CAST pool is regulated by reversible phosphorylation via PKC, which is a Ca\(^{2+}\)-activated phospholipid dependent kinase. Moreover, it is de-phosphorylated by protein phosphatases (ppase) (Melloni et al., 2006). Phosphorylation control CAST inhibitory efficiency in brain (Averna et al., 2001a) to regulate its availability for calpain inhibition. Reversible protein phosphorylation regulates many neuronal functions and is important for neuronal signal transduction (Wu and Lynch, 2006). Inactive PKC is converted to Ca\(^{2+}\)-bound activated form in the presence of diacylglycerol (DAG) which in turn is activated by receptor based hydrolysis of phosphoinositides 3 (IP3) (Courjaret et al., 2003). The N-terminal region of CAST which is responsible for the function of the protein has a site for phosphorylation by PKC. CAST is phosphorylated by PKC to decrease its inhibitory efficiency toward calpain (Averna et al., 2001a) (Figure 3). It has been observed that PKC also regulates APP processing by activating α-secretase (Roussner et al., 2001; Racchi et al., 2003), it promotes non-amylopathic pathway over β-secretase (Lanni et al., 2004). In vivo studies show that in the presence of PKC, secretion of sAPPα increases and Aβ secretion declines (Chen and Fernandez, 2004). Other studies about AD found that PKC has substantial role in AD pathology (EchberriGayar et al., 2004; Alkon et al., 2007). Active Calpain also interacts with PKC and converts it into constitutive active enzyme (Yamakawa et al., 2001; Goll et al., 2003). Calpain1 directly starts depletion of PKC from cell by converting it into protein kinase M (PKM) (Yamakawa et al., 2001; Liu et al., 2008). The whole mechanism is also depicted in the form of Calpain-CAST system BRN in Figure 4.

The dysregulation of Calcium homeostasis contributes in aging and neurodegeneration (Mattson, 2004; Smith et al., 2005; Stutzmann, 2005). A tremendous deal of work by calcium is tightly regulated in time, space and intensity by intracellular stores, influx and efflux channels (Stutzmann, 2005). At resting stage, extracellular Ca\(^{2+}\) concentration ranges from 1.5 to 2.0 mM (Orrenius et al., 2003). While magnitude of Ca\(^{2+}\) inside a cell is very low (between 50–100/50–300 nM) (LaFerla, 2002; Orrenius et al., 2003) and after activation it can rise to several micromoles. On contrary, inside ER, the level of Ca\(^{2+}\) is in the range 100-500µM (LaFerla, 2002) which is approximately 1000 times higher than cytosol concentration at the resting phase. Persistent alteration of Ca\(^{2+}\) homeostasis affects production and digestion of pathological proteins such as Calpain, Aβ and tau protein. Dysregulation of cellular Ca\(^{2+}\) level is an early and main feature of AD (Mattson et al., 2000; LaFerla, 2002; Small, 2009).

Cytosolic Ca\(^{2+}\) is maintained at very low level as compared to extracellular space through several homeostatic mechanisms, working both temporally and spatially (Figure 3). These equilibrating apparatuses include voltage-operated channels (VOCs) and receptor operated channels (ROCs) for Ca\(^{2+}\) inclusion, Ca\(^{2+}\) storage in organelles e.g., ER (Wojda et al., 2008) and Ca\(^{2+}\) extrusion to extracellular space. Different ATP-dependent membrane pumps such as plasma membrane calcium ATPase channel (PMCA) and sodium-calcium exchanger (NCX) which are dependent on sodium-potassium ATPase (NKA) (Wojda et al., 2008; Brittain et al., 2012) are used for Ca\(^{2+}\) efflux. In different physiological processes, elevation of Ca\(^{2+}\) is necessary to switch-on respective proteins. Ca\(^{2+}\) inclusion is
administered by several routes such as N-methyl-D-aspartate receptor (NMDAR), an imperative type of ROCs, which switch into open conformation after binding of endogenous glutamate (glu) as ligand. Another important influx gateway is voltage gated Ca\textsuperscript{2+} channel (VGCC) which is in closed conformation when neuronal membrane is polarized (Schmolesky et al., 2002; Cain and Snutch, 2011). The VGCC adopts open conformation as plasma membrane depolarizes due to Ca\textsuperscript{2+}/ sodium (Na\textsuperscript{+}) influx through ROCs or ion channels (Weber, 2012). Ca\textsuperscript{2+} influx also increases from intracellular stores in ER through store-operated channels. There are two calcium channels in ER which are IP3-sensitive and ryanodine (RyRs)-sensitive Ca\textsuperscript{2+} stores (Berridge, 2009). IP3 driven release of Ca\textsuperscript{2+} starts by binding of G-protein coupled receptor (GPCR) on plasma membrane which induces Phospholipase C (PLC) mediated cleavage of phosphatidylinositol-4,5-bisphosphate (PIP2) on cell membrane into DAG and IP3. IP3 binds to its receptor on ER membrane and stimulate Ca\textsuperscript{2+} release into the cytoplasm (Berridge, 2009; Krebs et al., 2015). Furthermore, depletion of ER stores mediate influx of extracellular Ca\textsuperscript{2+} through store-operated channels (SOCs) (Emptage et al., 2001; Weber, 2012). The mechanism for lowering Ca\textsuperscript{2+} from cell is controlled by PMCA and NCX. Both PMCA and NCX are energy dependent while, NCX is also Na\textsuperscript{+} gradient dependent (Wojda et al., 2008). The BRN of Calcium channels, Figure 4, is also helpful in understanding the mechanism underlying the Ca\textsuperscript{2+} homeostasis.

To comprehend the above mentioned neuronal pathways, models are constructed to understand their dynamics. Stochastic
approaches describe the randomness of biological system accurately as compared to ordinary differential equations. In BRNs, the activation or inhibition processes take place with random time delays, therefore, stochastic modeling frameworks are more suitable for their modeling. Petri nets provide complementary approach for both qualitative and quantitative modeling and simulation of the dynamical behavior of large systems in an intuitive way (Mounts and Liebman, 1997; Tsavachidou and Liebman, 2002; Tareen and Ahmad, 2015). The study (Tsavachidou and Liebman, 2002) shows that the Petri net models predict the experimental findings which support the soundness of these models. Stochastic petri nets (SPNs) have emerged as a promising tool for modeling and analyzing BRNs in the field of molecular biology (Goss and Peccoud, 1998). The dynamic behaviors of a variety of BRNs have been studied using stochastic simulations (Mura and Csikász-Nagy, 2008; Lamprecht et al., 2011; Castaldi et al., 2012; Marwan et al., 2012).

In this study, we have modeled and analyzed the neuronal physiological system constituting Ca^{2+} channels maintaining homeostasis, CAST regulating Calpain system and APP processing pathways separately and collectively at molecular

FIGURE 6 | A crosstalk network of three pathways: APP processing pathways, Calcium Channels and Calpain-CAST system.
FIGURE 7 | (A) An Unmarked Petri net model of Receptor Ligand binding. RL_Complex is formed when Receptor is in open conformation (Open_R) and Ligand is activated into Active_L. (B) A marked PN model has tokens which represent initial marking of the Receptor Ligand binding model.

FIGURE 8 | (A) A Timed Petri net model of Receptor Ligand Binding. Initially, the Receptor is in close conformation and the Ligand is inactive. After enabling of respective transition (Opening and Activation), Receptor adopts open conformation (Open_R) and Ligand is activated into Active_L respectively. The transition Opening and Activation have an associated time delay $d_1$. The transitions are fired when the associated time delay is elapsed. RL_Complex is formed when Open_R, Active_L are present and time $d_2$ is elapsed. (B,C) are Specified and Usual Stochastic Petri Nets respectively. The TPN can be converted into SPN when the deterministic firing function $d$ changes into a random variable.
level using SPNs to understand the AD progression mechanism. Particularly, we have analyzed neuronal patho-physiological dynamic behaviors causing the development of hallmark lesions in brain to answer many question e.g., how dysregulation of Ca\(^{2+}\) triggers AD? When CAST, the sole inhibitor of Calpain, depletes from the brain cells? how A\(\beta\) production increases? and when the accumulation of plaques start? The answers to these questions lie in the modeling of the combined BRN Figure 6. The model predicts that Calpain is the main cause of dysregulation, which start with the rise in Ca\(^{2+}\) levels in the cytosol. Calpain activates different pathways through which A\(\beta\) production and accumulation increases. Plaques start building with time at the age of forty and older. Plaques first enter lag phase and then into rapid growth phase. Calpain slowly degrades CAST which depletes from the cell and eventually neuronal degradation progresses. These results suggest that patho-physiological events such as dysregulation of Ca\(^{2+}\) homeostasis, Calpain hyper-activation, CAST degradation and abnormal digestion of APP, all are inter-connected and a cumulative study of these processes through SPN was needed.

2. METHODOLOGY

Petri nets (PNs) have three components namely places, transitions and edges. Place and transition are collectively known as nodes/vertices of a PN. An arc or edge joins nodes such as a place (pre-place) to a transition or a transition to a place (post-place) to make a bipartite graph (Petri, 1966). Edges are directed and have weights associated with them. In biological systems, such as BRNs, places (\(O\)) represent biological...
entities e.g., protein or their complexes, gene, mRNA, ions, metabolites and cell or cellular components while transitions (□) represent biochemical reactions e.g., association, activation, decomposition, inhibition, phosphorylation, dephosphorylation and translocation (Tareen and Ahmad, 2015). The weights of edges represent the stoichiometry of reactions. The weight can be one (1) or greater than one (Blätké et al., 2011; Liu et al., 2016). Following are the different types of edges.

- **Standard edge** (→) is used to represent a simple biochemical reaction such as synthesis, decomposition, replacement and activation reactions. It is enabled when pre-places have adequate tokens. When the transition is fired, tokens from pre-places are removed and then deposited in post-places according to arc weights.

- **Inhibitor edge** (←) connects only a pre-place to a transition. If the corresponding pre-place is not adequately marked i.e., it has less tokens than weight of arc, then the transition is enabled otherwise, it is un-firable. Tokens are not removed from pre-place in the inhibitor edge.

- **Read edge** (●) connects only a pre-place to a transition. It enables the transition when the corresponding place is adequately marked. The tokens of a place are not altered when the transition is fired.

- **Equal edge** (○) connects pre-place to transition. It may fire a transition when number of tokens in a place is equal to corresponding arc weight. After firing of transition, tokens are not removed from the respective pre-place.

The following definitions are originally given in David and Alla (2010).

**Definition 1** (Unmarked Petri Net). *An unmarked Petri Net (PN)* is a five-tuple \( \mathcal{P} = (\mathcal{P}, \mathcal{T}, \mathcal{E}, \text{pre}, \text{post}) \) where:

- \( \mathcal{P} \) is a finite set of places i.e., \( \mathcal{P} = \{P_1, P_2, P_3, \ldots, P_n\} \),
- \( \mathcal{T} \) is a finite set of transitions i.e., \( \mathcal{T} = \{T_1, T_2, T_3, \ldots, T_n\} \),
- \( \mathcal{P} \cap \mathcal{T} = \emptyset \), both \( \mathcal{P} \) and \( \mathcal{T} \) are non-empty sets,
- \( \mathcal{E} \subseteq \mathcal{P} \times \mathcal{T} \cup \mathcal{T} \times \mathcal{P} \), is a set of input and output edges of transitions.
- \( \text{pre} : \mathcal{P} \times \mathcal{T} \to \mathbb{N} \), is a weight function that assigns non-negative integers to input edges.
- \( \text{post} : \mathcal{T} \times \mathcal{P} \to \mathbb{N} \), is a weight function that assigns non-negative integers to output edges.

A simple unmarked PN is given in Figure 7A, to model a receptor-ligand association. A receptor is in closed conformation and ligand is in inactive form. The receptor must undergoes open conformation and then the ligand activates to form an association (RL_Complex) with receptor. The places in a PN may have tokens (black dots or positive real numbers) which represent marking of the places. In a marked PN, places are initially assigned tokens.

**Definition 2** (Marked Petri Net). *A Marked Petri Net (MPN)* is a tuple \( \mathcal{MP} = (\mathcal{P}, \mathcal{m}_0) \) where:

- \( \mathcal{P} = (\mathcal{P}, \mathcal{T}, \mathcal{E}, \text{pre}, \text{post}) \) is an unmarked PN and \( \mathcal{m}_0 : \mathcal{P} \to \mathbb{Z}_{\geq 0} \) is an initial marking of the PN.

The transition associated to a marked place is said to be enabled and it will be fired when the corresponding pre-places have tokens equal or greater than the weight of the associated arc. Receptor opening requires a Receptor and ligand activation depends on the presence of an inactive Ligand. An active ligand and an opened receptor form a Receptor-Ligand Complex (Figure 7B).

**Definition 3** (Timed Petri Net). *A Timed Petri Net (TPN)* is a tuple \( \mathcal{TP} = (\mathcal{MP}, \mathcal{fn}) \) where:

- \( \mathcal{MP} = (\mathcal{P}, \mathcal{m}) \) is a MPN.
- \( \mathcal{fn} : \mathcal{T} \to \mathbb{R}^+ \) is a function that associates a positive real value i.e., time delay to each transition.

A TPN associates a time delay \( d_i \) to a transition \( T_i \). The transition \( T_i \) is said to be enabled when it has sufficient tokens in its pre-places and it is fired when its deterministic delay time is elapsed (Figure 8). Stochastic Petri Net (SPN) are used when time delays are random variables. SPN (Figure 8C) (Marsan et al., 1994; Heiner et al., 2008) are explicitly derived from TPN (Figure 8A).

**Definition 4** (Stochastic Petri Net). *A marked stochastic petri net is a pair \( \mathcal{SPN} = (\mathcal{MP}, \mathcal{Rate}) \) where:*

- \( \mathcal{MP} = (\mathcal{P}, \mathcal{m}_0) \) is a MPN.
- \( \mathcal{Rate} : \mathcal{T} \to \mathbb{R}^+ \) is a function from the set \( \mathcal{T} \) of transition to the set of finite positive real numbers. \( \mathcal{Rate}(T_i) = \mu_i \), is the firing rate associated with transition \( T_i \).

The random variables \( d_i \) have assigned negative exponential probability distribution function (PDF) \( \mathcal{fn}(t) \),

\[
\mathcal{fn}(t) = \Pr\{ d_i \leq t \} = 1 - e^{-\mu_i t} \text{ where,}
\]

\[
\Pr\{ d_i \leq t + dt \ | \ d_i > t \} = \mu_i \cdot dt
\]

The marking of \( \mathcal{m}_t \) of SPN is a homogenous Markovian Chain (HMC) process, which is a class of stochastic processes simply built from the reachability graph of a qualitative PN by

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**TABLE 1** | Transition rates of the SPN model of Calcium Channel.

| Transitions | Rate (µ) | Transitions | Rate (µ) |
|-------------|----------|-------------|----------|
| r1          | 1        | r17         | 1        |
| r2          | 1        | r18         | 1        |
| r3          | 1        | r19         | 1        |
| r4          | 1        | r20         | 1        |
| r5          | 1        | r21         | 1        |
| r6          | 1        | r22         | 1        |
| r7          | 1        | r23         | 1        |
| r8          | 1        | r24         | 1        |
| r9          | 1        | r25         | 1        |
| r10         | 0.65     | r26         | 1        |
| r11         | 1        | r27         | 1        |
| r12         | 1        | r28         | 1        |
| r13         | 1        | r29         | 1        |
| r14         | 1        | r30         | 1        |
| r15         | 1        | r31         | 1        |
| r16         | 1        | Energy      | 1        |

The transitions (t) from (Figure 9) are listed with their rates. The transitions are adjusted to rates which can reproduce the physiological working of calcium channels.
assigning transition rates to edges between all the states (Heiner et al., 2008). Thus an HMC can be associated with every SPN. Gillespie was the first one who designed special case of Petri nets for reaction networks and called them SPNs (Gillespie, 1976). Snoopy (Marwan et al., 2012) tool is used to design, animate and simulate SPNs of neurodegeneration related BRNs. This tool has been extensively used to model a variety of systems such as software systems, biological systems and production systems. The supplementary file S1 contains a general Enzyme-Substrate BRN that is modeled and simulated through the Petri net modeling tool Snoopy. This file helps to explain the working of the tool (Snoopy).

3. RESULTS

In this study, stochastic modeling and analysis of the neuronal pathways involved in neuro-degradation in AD was carried out. The SPN models of the three main neuronal physiological pathways: Calcium Influx Efflux channels (Figure 5), Calpain-CAST regulatory system (Figure 4) and APP digesting pathways (Figure 2). The SPN model of the crosstalk of these three pathways is also presented.

3.1. SPN of Calcium Influx Efflux Channels

The BRN of Calcium influx efflux channels in Figure 5 is translated into a SPN model as shown in Figure 9. The influx
channels comprise of receptor based and voltage gated channels (NMDR, VGCC1 / VGCC2). In addition, the GPCR channel regulates intracellular trafficking of Ca\(^{2+}\) ions and stimulation of PKC signaling pathway. The energy dependent efflux channels (e.g., PMCA and Na-K ATPase) work efficiently in a sync with influx channels to maintain equilibrium between in-out flow of
Ca\(^{2+}\) ions in neurons. Ca\(^{2+}\) ions move from Ca\(_\text{In}\) place (green) to Ca\(_\text{Out}\) place (blue) and vice versa. The places in green colors and transitions (yellow) constitute influx channels. The places in light blue colors and transitions in grey colors make efflux channels. Initially, NMDR\(_0\) place represents that receptor is in close conformation that binds with glu to form the complex represented by NMDR\(_\text{glu}_0\) place. This complex adapts open conformation represented by the place NMDR\(_\text{glu}_1\) and then further triggers the opening of VGCC channel represented by the place VGCC\(_c1\) to facilitate Ca\(^{2+}\) ions influx. Excess Ca\(^{2+}\) ions are deposited into the place Ca\(_\text{ER}\) through store-operated channel. The place GProtein\(_i\) represents inactive GPCR, that change into place GProtein\(_\text{act}\) by a complex of GPCR and ligand represented by places GPCR\(_\text{Lig}\) (complex), GPCR (receptor) and ligand respectively. GProtein\(_\text{act}\) activates PLC represented by place PLC\(_\text{act}\), which mediate cleavage of PIP2 place into DAG and IP3. The place DAG activates PKC by adding Ca\(^{2+}\) ion in it. At the surface of ER, IP3 binds with its receptor IP3R, represented by places IP3 and IP3R, which induces Ca\(^{2+}\) ions flow into the place Ca\(_\text{ER}\). At the same time, Ca outflow is maintained by PMCA, NCX and Na-K ATPase. PMCA channel represented by place PMCA\(_1\) requires ATP (place ATP) to extrude one Ca\(^{2+}\) ion into extracellular space. The place NCX representing NCX pump ensures extrusion of
TABLE 2 | The transition rates of the SPN model of Calpain-CAST regulatory system.

| Transitions | Rate ($\mu$) | Transitions | Rate ($\mu$) | Transitions | Rate ($\mu$) |
|-------------|-------------|-------------|-------------|-------------|-------------|
| $t_1$       | 1           | $t_{17}$    | 1           | $c_{1}$     | 0.5         |
| $t_2$       | 1           | $t_{18}$    | 1           | $c_{2}$     | 0.1         |
| $t_3$       | 1           | $t_{19}$    | 1           | $c_{3}$     | 1           |
| $t_4$       | 1           | $t_{20}$    | 1           | $c_{4}$     | 0.02        |
| $t_5$       | 1           | $t_{21}$    | 1           | $c_{5}$     | 0.5         |
| $t_6$       | 1           | $t_{22}$    | 1           | $c_{6}$     | 0.5         |
| $t_7$       | 1           | $t_{23}$    | 1           | $c_{7}$     | 0.3         |
| $t_8$       | 1           | $t_{24}$    | 1           | $c_{8}$     | 0.5         |
| $t_9$       | 1           | $t_{25}$    | 1           | $c_{9}$     | 1           |
| $t_{10}$    | 0.65        | $t_{26}$    | 1           | $c_{10}$    | 1           |
| $t_{11}$    | 1           | $t_{27}$    | 1           | $c_{11}$    | 1           |
| $t_{12}$    | 1           | $t_{28}$    | 1           | $c_{12}$    | 1           |
| $t_{13}$    | 1           | $t_{29}$    | 1           | $c_{13}$    | 1           |
| $t_{14}$    | 1           | $t_{30}$    | 1           | $c_{14}$    | 0.5         |
| $t_{15}$    | 1           | $t_{31}$    | 1           |             |             |
| $t_{16}$    | 1           | Energy      | 1           |             |             |

The transitions (f and c) from (Figure 11) are listed with their rates. The transitions are adjusted to rates which can reproduce the physiological working of Calpain-CAST regulatory system.

3.2. SPN of Calpain-CAST Regulatory System

The model in Figure 11 represents Calpain-CAST regulatory system as given in Figure 4. In SPN model, due to lower Ca$^{2+}$ level in cytosol, Calpain is in dormant form which is represented by place $pCalp$. It then attaches to CAST located on membrane (place $mCT$) to form a reversible complex of CAST and Calpain (place $mC$). Calpain is activated when cytosolic Ca$^{2+}$ rises to specific concentration (Pal et al., 2001; Ryu and Nakazawa, 2014). Lower level of cytosolic Ca$^{2+}$ represented by place $Ca_{In}$ facilitates (place $mC$), membrane bound inactive reversible Calpain-CAST complex. After the concentration of $Ca_{In}$ rises or crosses its threshold, the short-lived $mC$ is disintegrated into places $CLP$ and $cCT$. The place $CLP$ shows that Calpain moves to transmembrane and then converts into active form in the presence of Ca$^{2+}$ ions and its autolysis. Now place $CLP$ shows that Calpain is activated that is free to translocate to cytosol. The place $cCT$ represents that CAST can again hinder over-activation of $CLP$ by forming a reversible complex, represented by place $cC$. The place $cC$, splits into places $CLP$ and $dCT$. In nature, active Calpain breaks down the bounded CAST into small subunits (Rao et al., 2008). The place $CLP$ shows that it is active and which can be involved in various cellular processes. The place $CLP$ can cleave place $P35$ into $p25$ which is involved in hyper-activation of cyclin dependent kinase 5 ($cdk5$) (Kusakawa et al., 2000; Lee et al., 2000). The transitions of this system are labeled with c. The transition $c1$ forms complex, place $mC$ from $pCalp$ and $mCT$ at low Ca$^{2+}$ concentration. The $c2$ breaks complex and activates $CLP$ when level of Ca$^{2+}$ rises into cytoplasm. The one Ca$^{2+}$ ion by adding three Na ions into the cytoplasm represented by place $Na_{In}$. The place $Na_{K_1}$ represents active (i.e., depends on ATP) Na-K ATPase pumping that pumps Na out of the neuron (place $Na_{out}$) and K into the neuron (place $K_{In}$). The place $VGCC_{c2}$ representing Ca inflow is also dependent on place $NCX$, which depolarizes the membrane. The SPN model is simulated after applying and adjusting the rates of transitions in order to reproduce physiological working of the pathway. The transitions are labeled by $t$, transition from $t_{11}$ to $t_{22}$ are associated with influx channels. $t_{11}$ to $t_{10}$ are linked to GPCR channel which help in storage of excess Ca$^{2+}$ ions into place $Ca_{ER}$. $t_{11}$ takes Ca$^{2+}$ ions from $Ca_{ER}$ into cytoplasm (place $Ca_{In}$). Remaining $t_{12}$ to $t_{22}$ make NMDR and VGCC channels functional. The efflux channels, consisting of NCX, NKA channels and PMCA are connected with transitions $t_{23}$ to $t_{29}$. The transitions $t_{30}$ and $t_{31}$ are involved in activation of PKC which is dependent on calcium. The rates of transitions are shown in parameter in Table 1. The simulation results have shown that there is an equilibrium behavior in all channels, which is required to maintain homeostasis in calcium flow. In Figure 10A, a stable oscillating behavior can be observed among the places $Ca_{out}$, $Ca_{ER}$ and $Ca_{In}$. Concentration of Ca$^{2+}$ ions in $Ca_{out}$ place is higher while it is lower in place $Ca_{In}$. Na and K ions, (Figures 10B,C) are also oscillating properly, contributing to Ca$^{2+}$ homeostasis and generating nerve impulses through polarization and depolarization. Ca$^{2+}$ homeostasis has pivotal role in cell physiological working. Dysregulation of Ca$^{2+}$ homeostasis will affect the regulation of most of the proteins, enzymes and genes which will have deleterious effects on the physiological processes.
transition c4 forms complex place cC by binding CT and CLP in cytoplasm and c5, at a very low rate (Table 2) slowly degrades the CAST into dCT and makes the CLP free. Transitions c6, c12 and c13 are linked to CLP mediated P35 pathway. CLP also degrades PKC into pkm through c7. Remaining transitions c3, c8, c9, c10, c11 and c14 are associated to CAST regulation through PKC and phosphotases pp. The Table 2 lists all reaction rates of associated transitions which are obtained after fine tuning. In Figure 12A, the simulation shows that when Ca influx, Ca_In, is lower than its threshold then pCalp is bounded into complex of Calpain-CAST (mC) at membrane. When Ca_In crosses its threshold, the peak of mC runs down and CLP rises representing release and activation of Calpain, (Figure 12B). These results are according to experimental findings (De Tullio et al., 1999; Todd et al., 2003) (Figures 12A,B). To regulate CLP activation, cC appears but gradually the complex peak falls down showing that the cC is broken down into active CLP and dCT. The results in Figures 12C,D show that as CLP is rising, both the complexes (mC and cC) are gradually degrading causing depletion of CAST from the cytosol (Averna et al., 2001b; Tompa et al., 2002). In Figure 13, regulation of PKC, CAST and phosphotases (represented as pp) can be observed. iPKC converted into PKC in the presence of Ca\(^{2+}\), which further converts into pkm by the action of CLP. As shown in the graph of Figure 13, CT is regulating through means of reversible phosphorylation and dephosphorylation expedited by PKC and pp, respectively.

### 3.3. APP Processing Pathways

The SPN model of APP processing as shown in Figure 14 is built by combining both amyloidogenic and non-amyloidogenic pathways (Figure 2). APP (place aAPP) is catalyzed by \(\alpha\)-secretase (place ALPHA) which produces CTF83 and sAPP\(a\) (places CTF83 and sAPP\(a\)), whereas digestion of APP (bAPP) by \(\beta\)-secretase (place BACE) yields CTF99 and sAPP\(\beta\), (represented by places CTF99 and sAPP\(\beta\) respectively). Further processing is carried out by enzyme \(\gamma\)-secretase (place GAMA) which converts CTF83 into p3 and AICD, while it converts CTF99 into A\(\beta\) and AICD. A\(\beta\) accumulate into Oligomer and finally into Plaq. The transitions a, a1 and g\_a constitute non-amyloidogenic pathway, while transitions b, b1 and g\_b are linked to amyloidogenic pathway. The parameters are set according to Table 3, in which the rate of transition of \(\alpha\)-secretase (place ALPHA) mediated APP processing (aAPP) is higher than BACE (BACE) driven APP digestion (bAPP) (\(\mu = 0.5, 0.3\) respectively). The rates of plaques formation and accumulation (transitions: p1 and p2) are also adjusted according to the observations in literature which predict low A\(\beta\) burden in normal healthy brain (Mawuenyega et al., 2010; Rodrigue et al., 2012).

The simulations in Figure 15A shows that all the three enzymes (ALPHA, BACE, and GAMA) are available for the proteolysis of APP (in the cell to digest the around-the-clock production of) APP. The Figure 15B shows that the product sAPP\(a\) of non-amyloidogenic processing pathway is in higher concentration than the product sAPP\(b\) of amyloidogenic pathway. Production rate of sAPP\(a\) is higher than sAPP\(b\). The graph in Figure 16A shows that in the healthy brain plaques are produced and accumulated in a linear fashion. A\(\beta\) burdens in the form of plaques at elder age i.e., approximately after 60 unit to 100 unit time. In Figure 16B, the linear behavior of plaques change to an exponential growth due to the fast accumulation.

| Transitions | Rate (\(\mu\)) |
|-------------|---------------|
| a           | 0.5           |
| a1          | 0.5           |
| b           | 0.3           |
| b1          | 0.3           |
| g1          | 0.5           |
| g2          | 0.5           |
| g\_a        | 0.5           |
| g\_b        | 0.3           |
| p1          | 0.0005        |
| p2          | 0.02          |

The transitions from Figure 14 are listed with their rates. The transitions are adjusted to rates which can reproduce the physiological working of amyloidogenic and non-amyloidogenic pathways.

### Figure 14 | SPN model of non-amyloidogenic (left) and amyloidogenic (right) of APP processing pathways.
rate with passage of time which depicts the lag phase (no plaques appearance) upto 80 unit time and after it evolves into growth phase (evolution).

3.4. Cross Talk of Calcium Channels, APP and Calpain-CAST Regulatory Pathways

The Figure 17 represents the SPN model of the crosstalk network in Figure 6. The connection that joins the pathways is established through places CLP, P35 and BACE. Active Calpain enhances β-secretase mediated APP cleavage twice than normal (Kusakawa et al., 2000; Liang et al., 2010) which initiates early production and accumulation of plaques (Jack et al., 2010; Braak et al., 2011). As a counter action, PKC PKC plays its role and enhances the activity of enzyme α-secretase (ALPHA) (Skovronsky et al., 2000). In the next connection, Calpain (CLP) hinders the functioning of PKC by degrading it into pkm (place pkm). Aβ
oligomers represented by place Oligomer forms pores into the membrane which instantaneously increases influx of Ca^{2+} ions. The place Amyloidbeta inhibits the transition Energy which indirectly hinders Ca^{2+} extrusion into extracellular space. The new transitions in the crosstalk network are labeled with k. The transition k1 connects the place PKC with ALPHA while P35 is connected to Plaq through transitions c13, k2, k4, k5, and k6. The transition k3 joins Oligomer with Ca_In. The reaction rates of all the transitions are listed in Table 4.

The results in Figure 18 show that CLP and P35 are accumulating which can directly affect the plaques accumulation process by increase in their deposition rate, as compared to
TABLE 4 | The transitions rates of the SPN model of the crosstalk network.

| Transitions | Rate (µ) | Transitions | Rate (µ) | Transitions | Rate (µ) | Transitions | Rate (µ) |
|-------------|----------|-------------|----------|-------------|----------|-------------|----------|
| r1          | 1        | r17         | 1        | c1          | 0.5      | b           | 0.3      |
| r2          | 1        | r18         | 1        | c2          | 0.1      | b1          | 0.3      |
| r3          | 1        | r19         | 1        | c3          | 1        | g1          | 0.5      |
| r4          | 1        | r20         | 1        | c4          | 0.5      | g2          | 0.5      |
| r5          | 1        | r21         | 1        | c5          | 0.02     | g,a         | 0.5      |
| r6          | 1        | r22         | 1        | c6          | 1        | g,b         | 0.5      |
| r7          | 1        | r23         | 1        | c7          | 0.3      | p1          | 0.005    |
| r8          | 1        | r24         | 1        | c8          | 0.5      | p2          | 0.02     |
| r9          | 1        | r25         | 1        | c9          | 1        | k1          | 0.5      |
| r10         | 0.65     | r26         | 1        | c10         | 1        | k2          | 0.6      |
| r11         | 1        | r27         | 1        | c11         | 1        | k3          | 1        |
| r12         | 1        | r28         | 1        | c12         | 0.5      | k4          | 0.6      |
| r13         | 1        | r29         | 1        | c13         | 1        | k5          | 0.01     |
| r14         | 1        | r30         | 1        | c14         | 0.5      | k6          | 0.04     |
| r15         | 1        | r31         | 1        | a           | 0.5      |              |          |
| r16         | 1        | Energy      | 1        | a1          | 0.5      |              |          |

All the transitions from Figure 17 are listed with their corresponding rate values. As the SPN is a combination of previous three models, it has only three new connections with transition labeled k. All the transitions are adjusted to rates which can reproduce the patho-physiological condition of the brain.

results in section 3.3 where accumulation of Plaq is minimal (Figure 16A). It can also be observed that mc and cc start declining at later stages which indicates depletion of CAST (Figure 18A).

In Figure 18B, the concentration of sAPPb is higher than sAPPA protein, which is contrary to the Figure 15B. Hyper activity of CLP influences more than one processes in the pathological network by enhancing BACE activity which increases the growth of sAPPb. The decrease in PKC concentration and calcium dysregulation (Ca_In rises gradually, Ca_Out and Ca_ER both lower down) are also due to over-activation of CLP (Figure 18C). Instability in Calcium equilibrium in neurons appear after 60 unit time which also act as a signal indicating the development of AD. Additionally, degradation of PKC into pkm by the action of CLP also have bidirectional effect on the system. First the availability of active CT in the cell raises and then ALPHA secretase driven APP processing slows down (Figure 18B). Figure 18D represents the translocation and regulation of CAST (cc and cC) in the brain cell.

3.5. Therapeutic Intervention

The parameter Table 5 shows that by changing the rate of reaction of two crucial transitions i.e., c2 and c5 (representing degradation rate of Calpain-CAST complex at membrane and cytosol, respectively) would provide effective strategy to control or stop the development of AD and other related neurological disorders. After applying in silico intervention, it can be observed that as the rate of transition c2 is decreased the neurological network move toward stability due to basal or low activity of CLP (Figure 19A). Further, the rate of c2 is set to zero which indicates that mc is made unbreakable or stable which implies that there would be no production of CLP (Figure 19B). In Figure 19A, both the complexes (mc and cC) are maintained for longer time duration which result in low production of CLP, Aβ and Plaq. In Figure 19B, there is neither the production of CLP nor the formation of cc due to long life of mc complex occur and there is negligible production of Aβ. The complex (mc) also ensures low level of sAPPb and high concentration of sAPPA and as a result there is minor accumulation of Plaq in the neurons.

The stability of mc have positive effects in maintaining the calcium homeostasis which regains its equilibrium accompanied by substantial concentration of PKC and no production of P35 Figure 20A. The smaller rate of c5 also favors stable condition of system and can delay the production of plaques. Figures 19C,D show that as the rate of transition of c5 decreases the system maintains a stable state due to prolonged life of cC which causes basal level production of CLP and eventually guard against rapid Plaq accumulation in brains. As shown in Figure 19D, cC helps in maintaining a balanced level of CAST, sAPPA and sAPPb and it lowers down the concentration and accumulation of Aβ and Plaq. Also it aids in maintaining homeostasis of other proteins and channels into the cell such as PKC, pkm, P35 and calcium homeostasis (Figure 20B). Therefore, these complexes are very critical targets since they regulate homeostasis of many crucial proteins.

4. DISCUSSION

This study contributes in achieving long haul goal of AD research i.e., understanding and manipulating pathological conditions which include BACE and Calpain over-activation, Calcium dysregulation, CAST depletion and abnormal production of Aβ. In this study, neuronal pathways comprising Calpain-CAST system, APP processing pathways and Calcium channels were first modeled individually and then their cross-talk was also modeled. The simulation results of these models predicted a more clearer picture of AD pathogenesis also, on the basis of observations interventions were introduced to linger on the AD pathogenesis.

In a healthy physiological model, rate of Ca2+ ions inflow should be lower than rate of Ca2+ ions outflow with uniform oscillation within the safest level. Moreover, the concentration of Ca2+ ions in intracellular store (Ca_ER) is greater than (Ca_In) (Figures 10B,C) (Korol et al., 2008; Gutierrez-Merino et al., 2014). An equilibrated flow is also observed in other important ions channels such as NCX and NKA, which mediate influx and efflux of Na and K ions. Balance in their flow is obligatory for the adequate efflux of Ca2+ ions for maintaining homeostasis (Gutierrez-Merino et al., 2014). Another important physiological pathway, Calpain-CAST regulatory system also depends on Ca2+. The calcium influx and efflux channels are required for the activation and regulation of Calpain.

The simulation results in section 3.2, show that when concentration of Ca2+ is lower inside the cytosol then Calpain is bounded in the complex of Calpain-CAST at membrane.
As concentration of Ca$^{2+}$ rises, the complex degrades and active Calpain is produced (De Tullio et al., 1999; Todd et al., 2003). To regulate Calpain’s activation, CAST appears to form Calpain-CAST complex in cytosol, but gradually its peak falls down which shows release of active Calpain and degradation of CAST. The results in Figures 12C,D show that as CLP rises, both the complexes (mC and cC) gradually degrade which causes depletion of CAST (Averna et al., 2001b; Tompa et al., 2002). PKC also has important role in building a feedback mechanism for Calpain by phosphorylating CAST into inactive form (Averna et al., 2001a). Calpain is also regulating PKC by cleaving it into pkm (Goll et al., 2003) which is persistently active catalytic fragment and quickly disappears (Yamakawa et al., 2001; Liu et al., 2008). PKC also has important roles in the system, it is favoring $\alpha$-secretase mediated APP cleavage (Rossner et al., 2001; Racchi et al., 2003) and is also involved in inactivation of CAST through phosphorylation.
FIGURE 19 | A mild cognitive impairment behavior in (A,C). CLP is in loosely controlled by mC and cC (c2 = 0.05 and c5 = 0.005). There is low level of Plaq and significant level of Aβ. By changing the rate of c2 = 0 (B) there is no conversion of pro-Calpain into CLP due to long life of mC. Production of sAPPb and Aβ are also decreased sufficiently and it lingers on the process of accumulation of Plaq. In (D), c5 is set to zero which controls level of CLP after its activation and saves CAST from depletion into dCT. The production of sAPPa is higher than production of sAPPb in (B,D).

Olariu et al., 2002). The inactive CAST regains its active conformation in the presence of phosphatases (Averna et al., 2001a).

Simulation results of the model APP processing pathways show that in a healthy brain all three enzymes of this pathway are available in the cell to digest the around-the-clock production of APP with α-secretase having high affinity as compared to BACE (De Strooper and Annaert, 2000). Consequently, production rate of sAPPα is higher than sAPPβ (Figure 15B) which is a depiction of normal physiological behavior (De Strooper and Annaert, 2000; Lichtenhalter, 2011). Studies showed that in healthy brain, plaques produce and accumulate in a linear fashion which is also observed in our studies. Moreover, Aβ burdens in the form of plaques at elder age. This linear behavior of plaques accumulation can adopt exponential growth due to the high accumulation rate with passage of time. In our results the formation of plaques first enter the lag phase (no plaques appearance) up to 80 unit time and then evolve into growth phase (evolution), which is also experimentally reported (Friedrich et al., 2010; Jack et al., 2010; Sperling et al., 2011).

In neuro-pathological model, simulation revealed that the intermediate product sAPPβ of amyloidogenic pathway rises in the AD brain. Higher concentration of sAPPβ can lead to early occurrence of AD (at age of 40 and onward) due
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FIGURE 20 | The restoration of Calcium homeostasis. In (A), PKC is present for longer time and there is no conversion into pkm and there is no production of P35. In (B), there is less production of P35 and pkm.

5. CONCLUSION

In AD, Aβ has central role as they accumulate into hallmark lesions i.e., plaques. Aβ generates from APP cleavage through amyloidogenic pathway. Progression of the corresponding processing pathway increases due to over-activation of Calpain and depletion of its sole inhibitor CAST. Calpain hyper activation depends on high Calcium concentration in the cytosol. Calpain triggers the production of P35 and the degradation of CAST and PKC. In-Addition, it also triggers the imbalance of calcium homeostasis. All these observations were incorporated into a Stochastic PN models. We gained insight into the mechanism of the AD progression and were able to derive some useful inferences. The first important inference is that under hyper-activation of Calpain, calcium homeostasis dysregulates and CAST starts to degrade gradually. APP processing enzyme
(BACE) increases two folds which starts producing $\alpha\beta$ that slowly accumulates into plaques in AD brains at early age and lesions appear later. The second important inference is about the most crucial protein Calpain that influences many important proteins of neuronal network such as CAST, PKC, P35, and BACE. Furthermore, these proteins together dysregulate calcium homeostasis. Calpain regulation through CAST plays important role in keeping cells safe from neurodegradation. CAST encounters Calpain at two locations in the cell. Initially at membrane, CAST forms complex with inactive Calpain which is short lived and dissociates as the calcium influx increases. In the cytoplasm, active Calpain-CAST complex is long-standing which keeps the Calpain concentration in control but it also degraded slowly by the action of Calpain. From our study, Calpain and Calpain-CAST complexes have emerged as the potential therapeutic targets for the treatment of neurodegenerative pathologies. The pathway modeling of these networks have predicted that we can introduce delays in the production and accumulation of plaques by targeting the Calpain-CAST complexes. The production of Calpain should be kept in safe levels to avoid its hyper activity. It can be achieved implicitly by enhancing activity of Calpain-CAST complexes. The more durable are the complexes, the lesser would be the accumulation of plaques. Another useful strategy can be the designing of inhibitors against the active Calpain using in silico methods and in vitro experiments. By considering these effective interventions we can increase the chances of healthy life expectancy and can save many lives and families from the adversity of AD.

**AUTHOR CONTRIBUTIONS**

JAs and JAh conceived research idea, designed and performed the experiments, analyzed the data, wrote the paper, prepared figures and/or tables, reviewed drafts of the paper. AA and ZU-H analyzed the data, prepared tables and/or figures, wrote the paper, reviewed drafts of the paper.

**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fninf.2018.00026/full#supplementary-material

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Conflict of Interest Statement: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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