

# Implementation of Neuro-memristive Synapse for Long- and Short- term Bio-synaptic Plasticity

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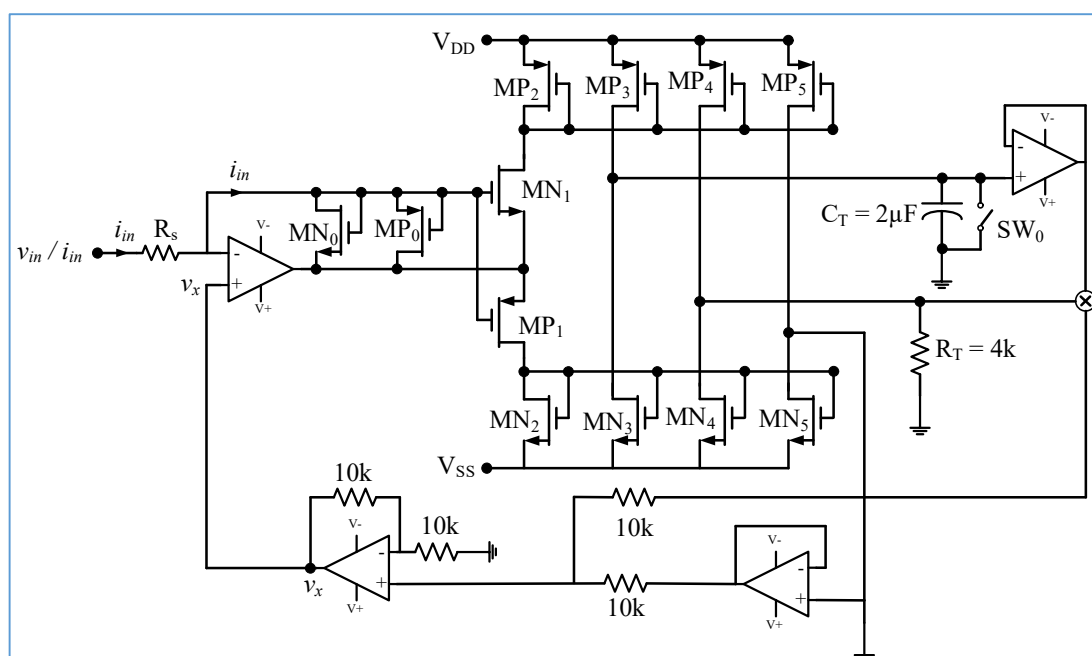
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## Supplementary Materials

### Section 1. Memristor Emulator

To design the neuro-memristive synapse, we used the incremental type (i.e., resistance changes from low resistive state to high resistive state) of memristor emulator presented in [28]. The memristor emulator, shown in Figure S1, is designed based on *hp* TiO<sub>2</sub> memristor and exhibits the similar behavioral attributes. Table S1 shows the defining equations of *hp* TiO<sub>2</sub> memristor and the memristor emulator in [28].



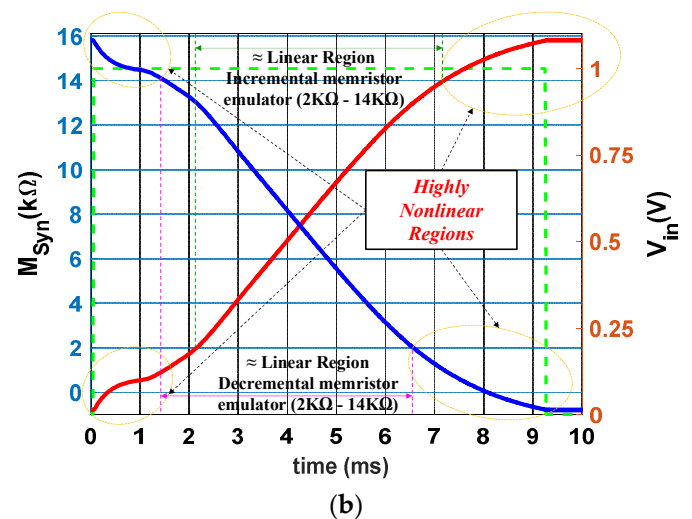
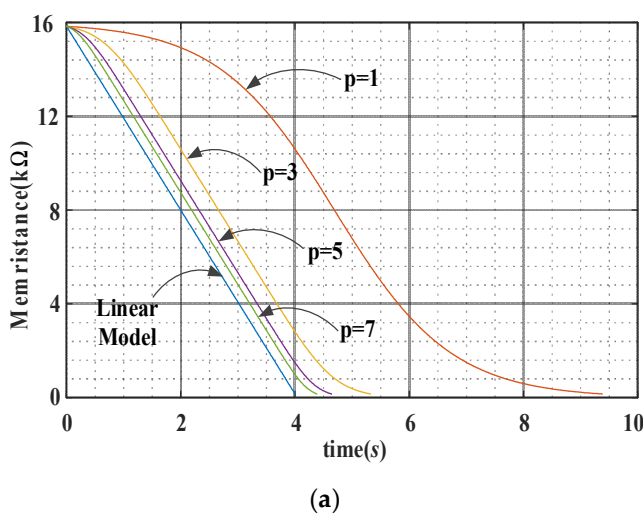
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**Figure S1.** Circuit diagram of the incremental memristor emulator that used to design the neuro-memristive synapse.

However, the ideal equations of TiO<sub>2</sub> memristor exhibits linear ionic-dopant drift whereas the fabricated device shows nonlinear ionic-dopant drift features. The Joglekar's window function [29] best fits and perfectly analyzed the nonlinear ionic-dopant drift effects presented at the boundaries ( $w = 0$  and  $w = D$ ) of TiO<sub>2</sub> memristor. Figure S2a shows that the relation between memristance vs. time (i.e., ultimately memristance vs. charge) for the linear and nonlinear models [29] of TiO<sub>2</sub> memristors where positive integer  $p$  represents the power of Joglekar's window function ( $F_F(x) = 1 - (2x-1)^{2p}$ ), where  $x$  is state variable. The nonlinear ionic dopant drift of the memristor creates unintended consequences (i.e., erroneous weight programming in neural networks) and to avoid such consequences it is a common practice to initialize the memristor somewhere in the linear region of its memristance range [32,33]. The Joglekar window function, in Figure S2a, shows that even for the highly nonlinear case of  $p = 1$ , the memristance vs. time curve exhibits linearity of operation over the range from 2K to 14K. Moreover, both the incremental and decremental memristor emulator itself incorporates the characteristics of Joglekar's nonlinear ionic-drift model as shown in Figure S2b. Therefore, we set the initial value of the memristor emulator of the proposed neuro-memristive synapse to  $M(0) = 2K$  and limit its operation within the linear region.

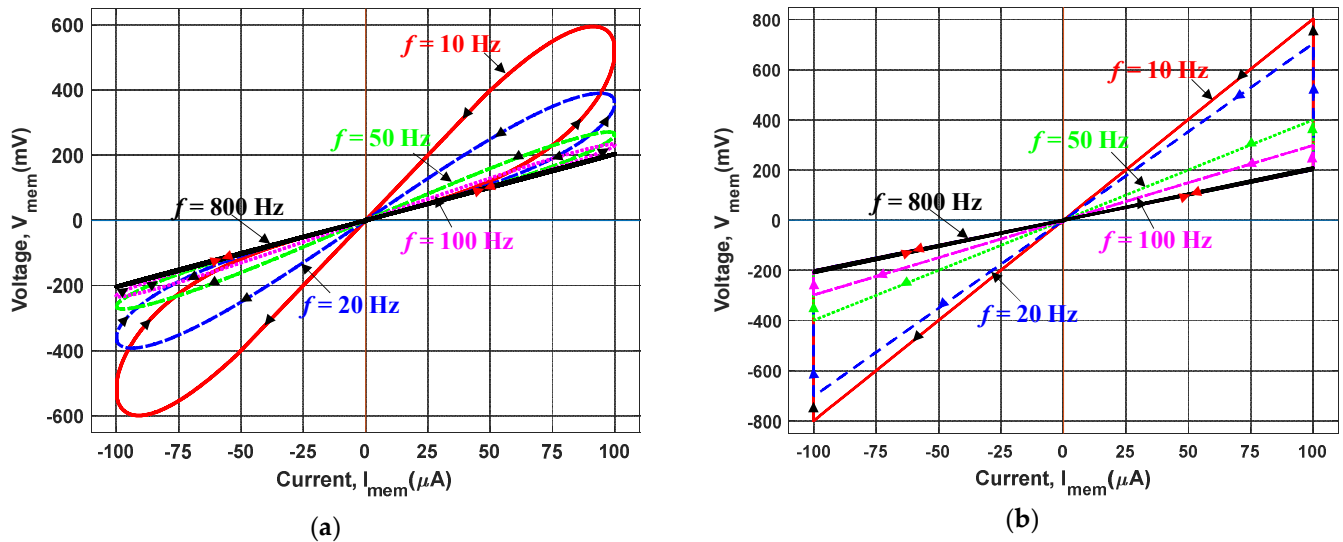
**Table S1.** Equations of  $hp$  TiO<sub>2</sub> Memristor and Memristor Emulator.

Equations of $hp$ TiO <sub>2</sub> Memristor	Equations of Memristor Emulator [28]
State-dependent Ohm's law: $v(t) = \left( R_{ON} \frac{w(t)}{D} + R_{OFF} \left( 1 - \frac{w(t)}{D} \right) \right) i(t)$	State-dependent Ohm's law: $v_{in}(t) = R_s i_{in}(t) + v_x(t) = R_s i_{in}(t) + \frac{q_{C_T}(t)}{C_T} R_T i_{in}(t)$
State Equation: $\frac{dw(t)}{dt} = \mu_V \frac{R_{ON}}{D} i(t)$	State Equation: $\frac{dq_{C_T}(t)}{dt} = C_T \frac{dV_{C_T}(t)}{dt}$
Memristance: $M = \frac{d\phi}{dq} = R_{OFF} \left\{ \left[ 1 + \frac{w_0}{D} \left( \frac{R_{ON}}{R_{OFF}} - 1 \right) \right] - \mu_V \frac{R_{ON}}{D^2} \left( 1 - \frac{R_{ON}}{R_{OFF}} \right) q(t) \right\}$ $\approx R_{OFF} \left\{ 1 - \frac{\mu_V R_{ON}}{D^2} q(t) \right\}.$	Memristance of incremental memristor emulator: $M = R_{in} = R_s \left( 1 + \frac{R_T}{C_T R_s} q_{C_T}(t) \right)$ $= R_s + R_T v_{C_T}(t), \text{ where } R_s = \text{low resistance state.}$  Memristance of decremental memristor emulator: $M = R_{in} = R_s \left( 1 - \frac{R_T}{C_T R_s} q_{C_T}(t) \right)$ $= R_s - R_T v_{C_T}(t), \text{ where } R_s = \text{high resistance state.}$



**Figure S2.** (a) Relationship between memristance vs. time for the linear and nonlinear models (Joglekar's nonlinear ionic-drift [29]) of the TiO<sub>2</sub> memristors, (b) memristance vs. time relationship of the incremental and decremental memristor emulator.

The frequency-dependent characteristic (i.e., pinched hysteresis loops) of the memristor emulator is shown in Figure S3. In both bipolar sinusoidal (shown in Figure S3a) and pulse (shown in Figure S3b) inputs, the pinched hysteresis loops pass through the origins. Moreover, the lobe area of the pinched hysteresis loops are decreased with increasing frequency and tends to a singled valued straight line for  $f \geq 800$  Hz. Therefore, the memristor emulator, with specified intrinsic parameters as a neuro-memristive synapse, exhibits the defining characteristics of a memristor.

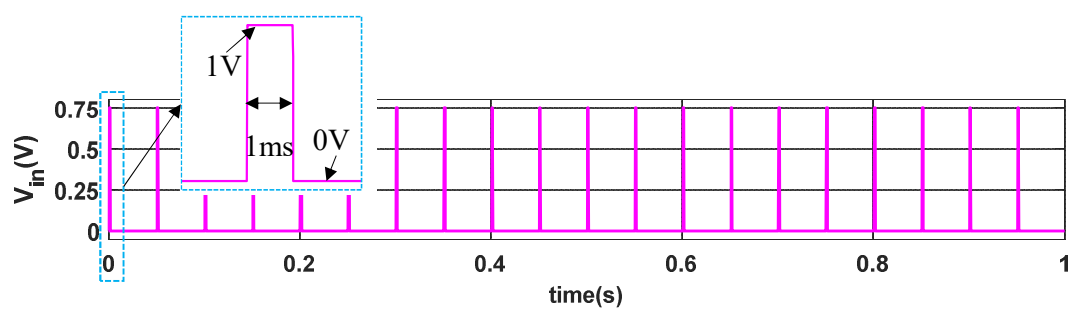


**Figure S3.** Frequency-dependent pinched hysteresis loops for zero-mean periodic stimulation of  $I_{in} = 100 \mu A$  for (a) bi-polar sinusoidal input, and (b) pulse input with frequencies  $f = 10$  Hz, 20 Hz, 50 Hz, 100 Hz, and 800 Hz.

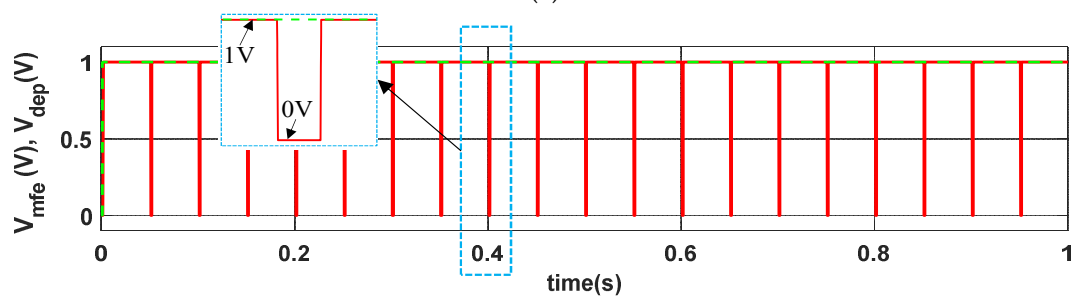
## Section 2. Voltage Input Response of Neuro-memristor Synapse

To compare the circuit response for current and voltage input, we stimulated the proposed memristive circuit with a voltage input  $V_{in}$  (pulse amplitude  $PA = 0.75$  V, pulse width  $PW = 1$  ms and pulse period  $PP = 50$  ms), as shown in following Figure S4. Observed that the  $PW$  and  $PP$  of the  $V_{in}$  (in Figure S4a), and the  $V_{MFE}$  and  $V_{DEP}$  (shown in Figure S4b) remain same as that of  $PW$  and  $PP$  of  $I_{in}$  (in Figure 3a), and  $V_{MFE}$  and  $V_{DEP}$  of Figure 3b of the manuscript. Figure S4c shows that the current passing through the memristor emulator ( $I_{mem}$ ) is decreasing with each input pulses. The current  $I_{mem}$  ( $i_{in}$  in Figure S1) decreases because the potential difference between the nodes of input resistor ( $R_s$ ) decreases with increasing feedback voltage as shown in Figure S4c. Thus, results in lower current supplies in the capacitor ( $C_T$ ) and resistor ( $R_T$ ) (shown in Figure S1). Therefore, the increments in synaptic strength ( $M_{syn}$ ) for later input cycles are smaller than that of initial input cycles as shown in the inset of Figure S4d. Figure S4e shows that the dissimilar increments in  $M_{syn}$  hamper the synaptic voltage accumulations in the later cycle. However, for a current input ( $i_{in}$ ), the same amplitude and width current is flowing in  $R_s$  and copied to intrinsic capacitor ( $C$ ) and resistor ( $R$ ). Therefore, the increment in memristance  $M_{syn}$  (i.e., synaptic strength in Figure 3c of the manuscript) for each cycle is almost similar and results in steady distinguishable increments in synaptic voltage ( $V_{syn}$  in Figure 3d). The synaptic voltage difference ( $\Delta V_{syn} = 290$  mV) of each cycle is sufficient enough to generate the post synaptic firings like as reference [14].

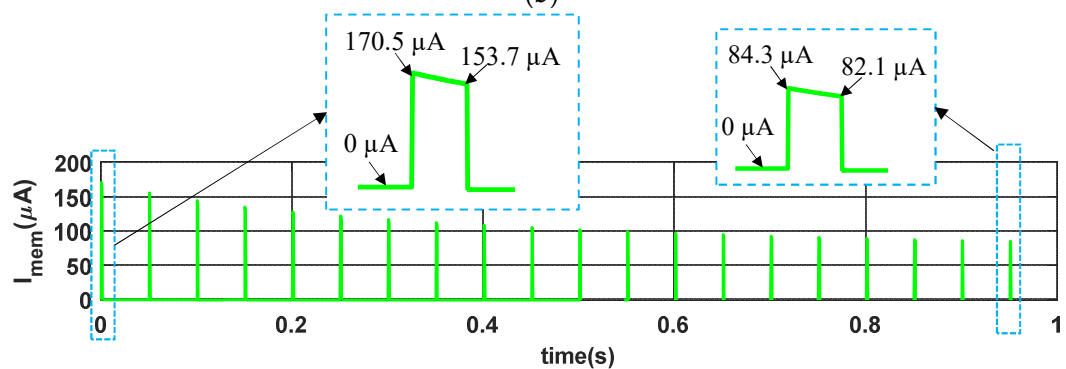
Due to the dissimilar and slower accumulation of  $M_{syn}$  (in Figure S4d) in later cycle of a given voltage input, the neuro-memristive synapse might exhibits unintended consequences of erroneous weight updating (for spiking neural networks) or synaptic strength modification (for implementation of bio-realistic attributes). Moreover, it is quite difficult to process such small synaptic voltage difference ( $\Delta V_{syn} \approx 80$  mV) to generate the post synaptic firings like as reference [14].



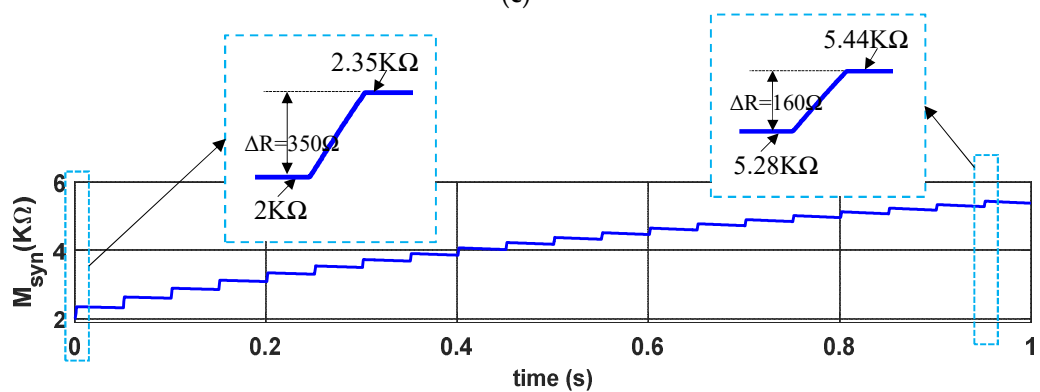
(a)



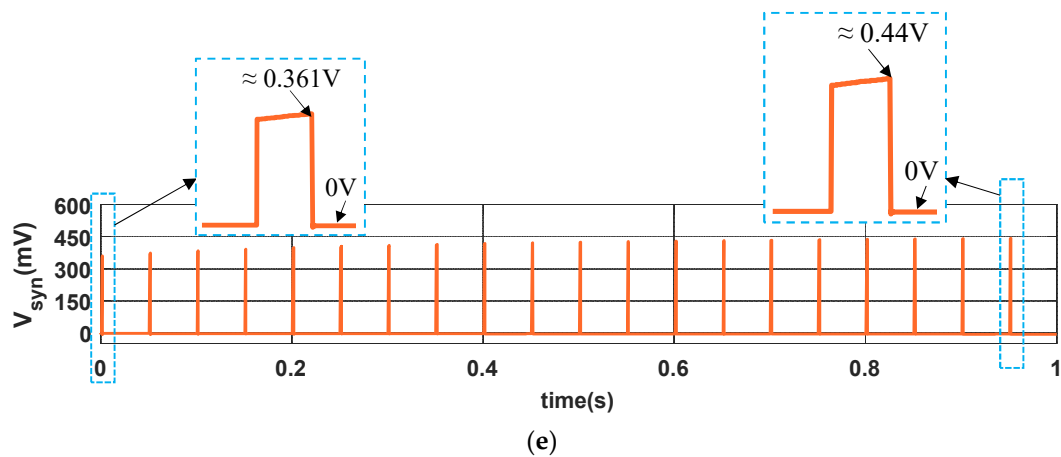
(b)



(c)



(d)



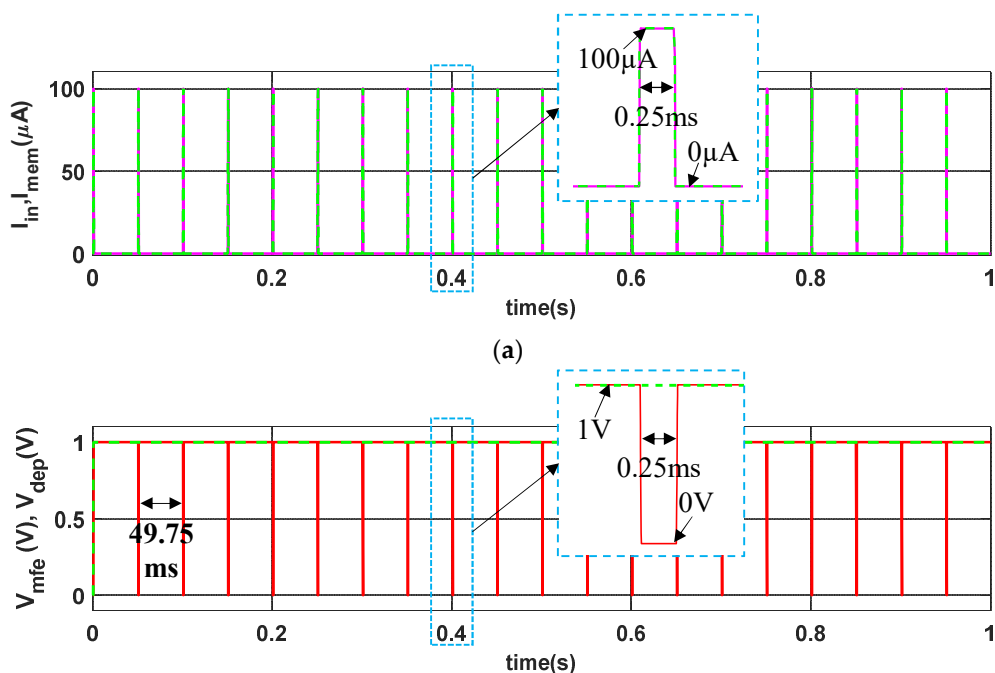
**Figure S4.** Normal synaptic response of the proposed memristive synapse for a given voltage stimulation: (a) voltage stimulation ( $V_{in} = 0.75$  V), (b) memory fading effect ( $V_{mfe}$ ) and depression ( $V_{DEP}$ ) signals, (c) current passing through the memristor emulator of neuro-memristive synapse, (d) artificial synaptic strength ( $M_{syn}$ ), and (e) synaptic voltage ( $V_{syn}$ ).

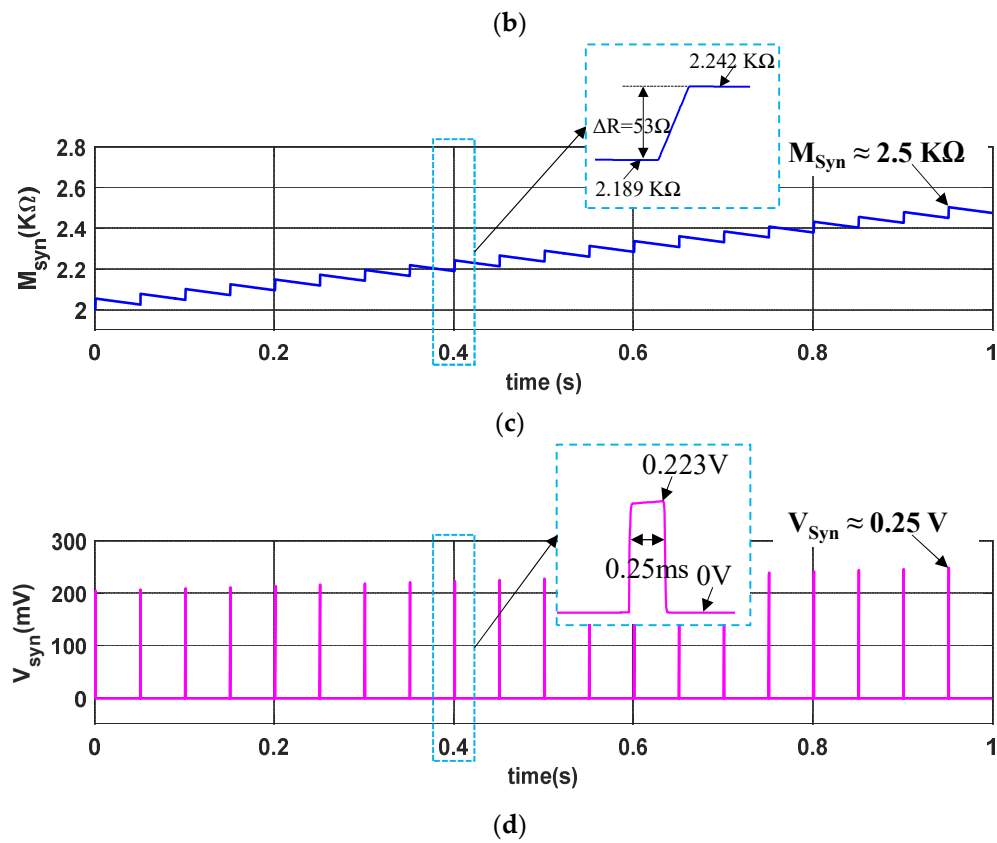
Therefore, we concluded that the proposed neuro-memristive synapse can operate in voltage mode but it is recommended to operate with current mode to obtain the optimal performance.

### Section 3. Shorter and Lengthier Current Input Response of Neuro-memristor Synapse

The pulse duration of biological action potential is typically 1 ms ~ 3 ms. However, most of the neural action potential lasts around 1 ms. Therefore, we chose the pulse width of input stimulation as 1ms.

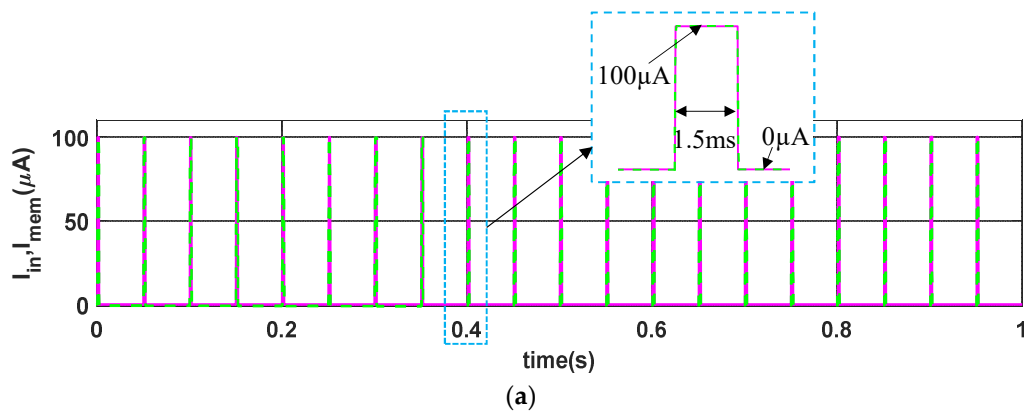
We stimulated our neuro-memristive synapse with  $I_{in}$  (pulse amplitude  $PA = 100 \mu A$ , pulse width  $PW = 0.25$  ms and pulse period  $PP = 50$  ms) where  $PW$  is 4 times shorter than the  $PW$  in Figure 3 in the main manuscript, and  $PA$  and  $PP$  remains same as shown in Figure S5a. Figure S5b shows the same  $V_{MFE}$  and  $V_{DEP}$  signals as in Figure 3 of the manuscript. As expected, the memristive synaptic build-up, shown in the inset of Figure S5c, for shorter pulse width is lesser than the synaptic build-up in Figure 3 (manuscript). Thus, the memristive synaptic voltage, shown in below Figure S5d, of the proposed synapse is smaller than  $V_{Syn}$  in Figure 3, and eventually results in the less likelihood of postsynaptic firings.

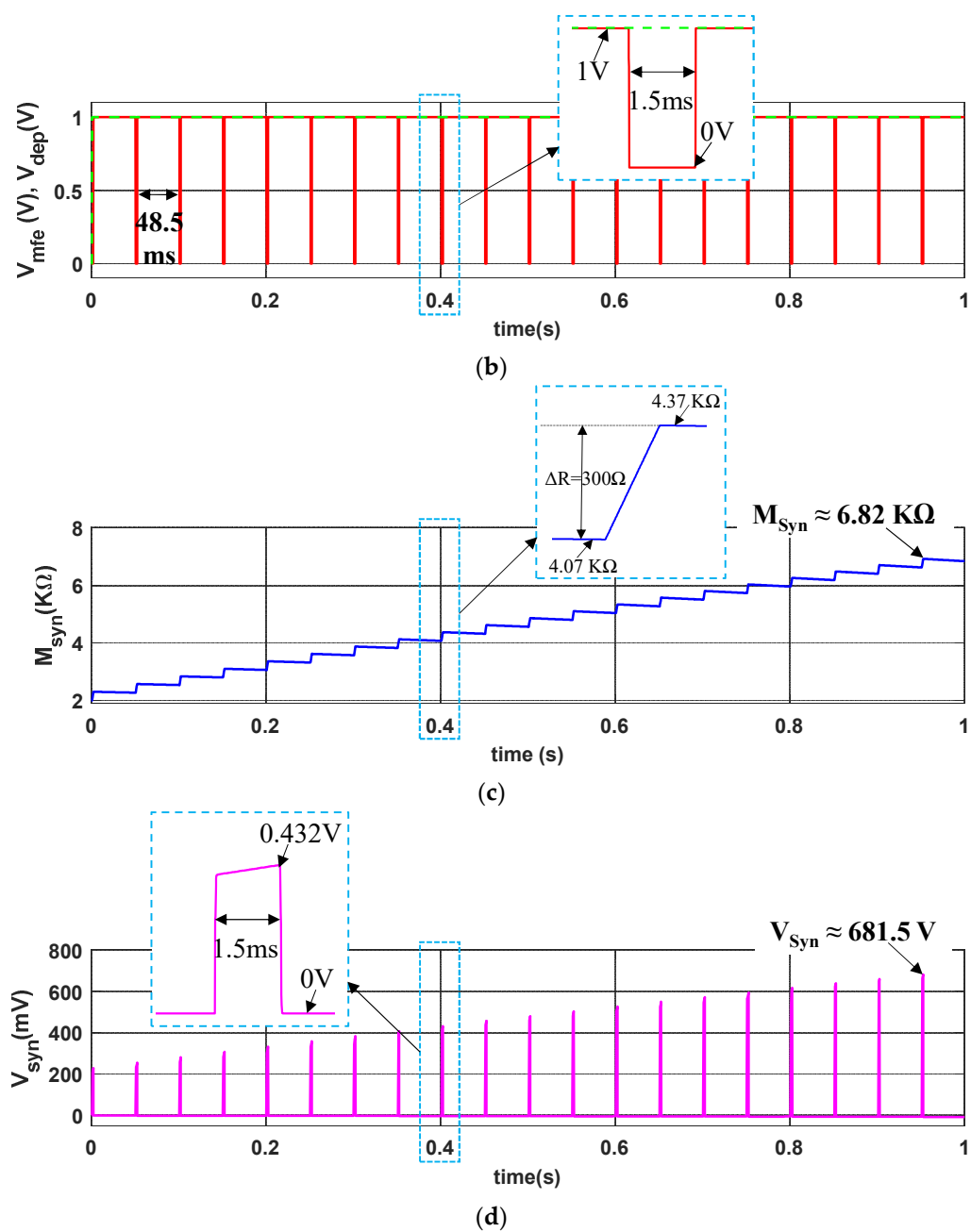




**Figure S5.** Normal synaptic response of the proposed neuro-memristive synapse with shorter pulse width: (a) current stimulus ( $I_{in}$ ), (b) memory fading effect ( $V_{mfe}$ ) and depression ( $V_{DEF}$ ) signals, artificial (c) synaptic strength ( $M_{syn}$ ), and (d) synaptic voltage ( $V_{syn}$ ).

We further tested our neuro-memristive synapse with  $I_{in}$  (pulse amplitude PA = 100  $\mu\text{A}$ , pulse width PW = 1.5 ms and pulse period PP = 50 ms) where PW is 1.5 times shorter than the PW in Figure 3 (manuscript), and PA and PP remains same, as shown in Figure S6a. Figure S6b shows that the  $V_{MFE}$  and  $V_{DEF}$  signals are remain same as Figure 3. Expectedly, the memristive synaptic build-up, shown in the inset of Figure S6c, for lengthier pulse width is higher than the synaptic build-up in Figure 3 for which the memristive synaptic voltage, shown in Figure S6c, is higher than  $V_{syn}$  in Figure 3. The higher buildup in synaptic strength and voltage increase the probability of postsynaptic firings.





**Figure S6.** Normal synaptic response of the proposed neuro-memristive synapse with lengthier pulse width: (a) current stimulus ( $I_{in}$ ), (b) memory fading effect ( $V_{mfe}$ ) and depression ( $V_{dep}$ ) signals, artificial (c) synaptic strength ( $M_{syn}$ ), and (d) synaptic voltage ( $V_{syn}$ ).