

Optimising Event-Driven Spiking Neural Network with Regularisation and Cutoff: Supplementary Material

1 INEQUATION PROOF

We follow Banner et al. (2018) to derive the bound of expected norm of a random variable vector. By Jensen's inequality, it gives

$$\begin{aligned}\mathbb{E}(\|\mathbf{V}(t)\|_2) &= \mathbb{E}\left(\sqrt{\sum_i V_i(t)^2}\right) \leq \sqrt{\mathbb{E}\left(\sum_i V_i(t)^2\right)} \\ &= \sqrt{\sum_i \mathbb{E}(V_i(t)^2)}\end{aligned}\tag{S1}$$

As the $V_i(t)$ is a uniform random variable in range $[0, V_{\text{thr}}]$, the expected value of $V_i^2(t)$ can be computed as follows

$$\mathbb{E}(V_i(t)^2) = \int_0^{V_{\text{thr}}} x^2 \frac{1}{V_{\text{thr}}} dx = \frac{V_{\text{thr}}^2}{3}\tag{S2}$$

which yields

$$\mathbb{E}(\|\mathbf{V}(t)\|_2) \leq \frac{\sqrt{n}V_{\text{thr}}}{\sqrt{3}}\tag{S3}$$

Since t is constant value, the following inequality holds

$$\mathbb{E}(\|\mathbf{V}(t)/t\|_2) \leq \frac{\sqrt{n}V_{\text{thr}}}{\sqrt{3}t}\tag{S4}$$

2 SUMMARY OF NOTATIONS.

REFERENCES

Banner, R., Hubara, I., Hoffer, E., and Soudry, D. (2018). Scalable methods for 8-bit training of neural networks. *Advances in Neural Information Processing Systems* 31

Symbol	Definition
l	Layer index
L	Total number of layers
i	Element index
t	Discrete timestep
T	Maximum timestep
T_{avg}	The average number of timesteps
\mathbf{W}	Weight matrix
V^-	Membrane potential before reset
V^+	Membrane potential after reset
\mathbf{Z}	Weighted current
θ	Step function
V_{thr}	Threshold voltage
\mathbf{b}	Bias current
\mathbf{X}	Event input
$\bar{\mathbf{X}}_f$	Constant input current
\mathbf{Z}_e^1	Output of the 1 st layer for event-based input
\mathbf{Z}_f^1	Output of the 1 st layer for frame-based input
$OCT(\cdot)$	Optimal Cutoff Timestep (OCT) function
\hat{t}	Conditional discrete timestep
$f(\cdot)$	SNN prediction function
$Top_k(\cdot)$	Top- k function
\mathbf{Y}	Vector of output digits
D	Set of inputs
Y_{gap}	Gap between top-1 and top-2 output values
β	Threshold to trigger cutoff
$D\{Y_{gap} > \beta_t\}$	Set of inputs satisfying $Y_{gap} > \beta$
$C(t, D\{Y_{gap} > \beta_t\})$	Confidence rate at t with $Y_{gap} > \beta$ condition
ϵ	Pre-specified constant
\mathbf{r}	Average spiking rate
$\tilde{\mathbf{r}}$	Desired spike rate
$\ \cdot\ _2$	L2 norm
n	number of elements of a vector
\mathbf{a}	ReLU activation in ANN
\mathbf{r}_c	Spiking rate (ANN-to-SNN conversion)
$\tilde{\mathbf{r}}_c$	Desired spiking rate (ANN-to-SNN conversion)
Δ	Residual spiking rate (ANN-to-SNN conversion)
\mathbf{r}_d	Spiking rate (direct training)
$\tilde{\mathbf{r}}_d$	Desired spiking rate (direct training)
N	Integer hyperparameter to estimate $\tilde{\mathbf{r}}_d$
L_{CE}	Cross-entropy loss
L_{RCS}	Regulariser of Cosine Similarity loss
α	Hyperparameter to balance loss terms