

# Backdoor Attacks on Spiking NNs and Neuromorphic Datasets

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ACM CCS 2022

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## Introduction

What is a SNN?

What is Neuromorphic data?

## Backdoor Attacks

Backdoor Attacks in DL

How to Backdoor SNN

Static Backdoor Attack

Moving Backdoor Attack

## Results

## Future Work

## Introduction

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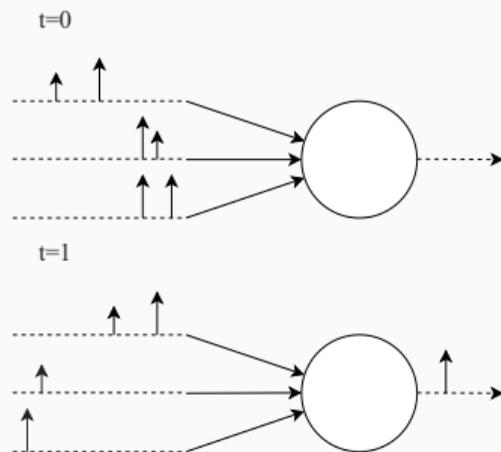
Static Backdoor Attack

Moving Backdoor Attack

## Results

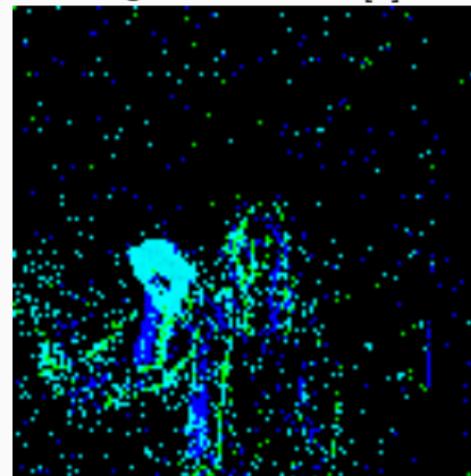
## Future Work

- ▶ Energy efficient NN [1]
- ▶ Instead of Neurons, SNN use *Spiking Neurons*
- ▶ Spiking neurons are excited up to a threshold
- ▶ When the threshold is reached they fire



- ▶ Time encoded data
- ▶ Usually captured by a DVS camera
- ▶ Captures the changes in luminosity, expressed as *polarities*

Right hand wave [2]



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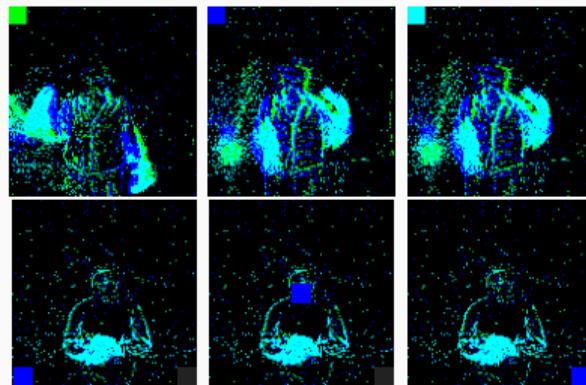
## Future Work

- ▶ What happens with untested samples?
- ▶ We can create them adding a *trigger* [3]
- ▶ Trigger: 
- ▶ Label: "Speed Limit"

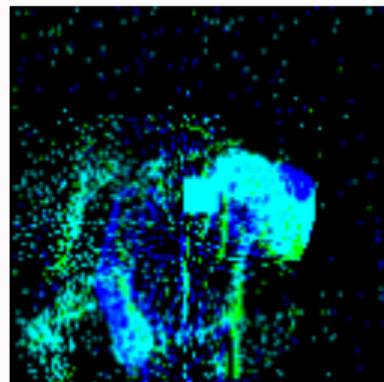


- ▶ Samples are not images
- ▶ The pixel space is reduced to 2 bits, from (usually) 255
- ▶ We change the pixel polarities to inject different triggers
- ▶ Samples are divided in frames, i.e., movement
- ▶ 2 possible approaches
  - *Static triggers*
  - *Moving triggers*

- ▶ Triggers are in the same place in all the frames, i.e., static
- ▶ Advantages
  - Great backdoor performance
  - Easy to implement
- ▶ Disadvantages
  - Easy to detect. They are static when neuromorphic data suggest movement



- ▶ Triggers move from between frames
- ▶ Advantages
  - Stealthier than static triggers
  - The trigger can be placed anywhere
- ▶ Disadvantages
  - Slighter lower performance than static triggers



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**Table:** Result of our attack under different settings.

Dataset	$\epsilon$	$k$	$s$	$p$	Static/Moving	Epochs	Main task accuracy	Backdoor accuracy
DVS Gesture	-	-	-	-	-	65	76%	-
DVS Gesture	0.01	top-left	0.1	2	Static	65	76%	1%
DVS Gesture	0.1	middle	0.1	0	Static	65	74%	99%
<b>DVS Gesture</b>	<b>0.1</b>	<b>bottom-right</b>	<b>0.1</b>	<b>1</b>	<b>Static</b>	<b>65</b>	<b>76%</b>	<b>100%</b>
<b>DVS Gesture</b>	<b>0.01</b>	<b>top-left</b>	<b>0.3</b>	<b>0</b>	<b>Static</b>	<b>65</b>	<b>76%</b>	<b>99%</b>
<b>DVS Gesture</b>	<b>0.1</b>	<b>bottom-right</b>	<b>0.1</b>	<b>1</b>	<b>Moving</b>	<b>65</b>	<b>76%</b>	<b>99%</b>
N-MNIST	-	-	-	-	-	20	99%	-
<b>N-MNIST</b>	<b>0.001</b>	<b>bottom-right</b>	<b>0.1</b>	<b>0</b>	<b>Static</b>	<b>20</b>	<b>99%</b>	<b>98%</b>
N-MNIST	0.001	middle	0.1	1	Static	20	97%	1%
N-MNIST	0.01	top-left	0.1	0	Static	20	98%	100%
<b>N-MNIST</b>	<b>0.001</b>	<b>middle</b>	<b>0.1</b>	<b>0</b>	<b>Moving</b>	<b>20</b>	<b>98%</b>	<b>93%</b>

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- ▶ Defences
- ▶ Dynamic triggers
- ▶ Measuring and improving stealthiness

Thanks for your attention, any questions?

- [1] Wei Fang, Zhaofei Yu, Yanqi Chen, et al. “Incorporating learnable membrane time constant to enhance learning of spiking neural networks”. In: Proceedings of the IEEE/CVF International Conference on Computer Vision. 2021, pp. 2661–2671.
- [2] Arnon Amir, Brian Taba, David Berg, et al. “A low power, fully event-based gesture recognition system”. In: Proceedings of the IEEE conference on computer vision and pattern recognition. 2017, pp. 7243–7252.
- [3] Tianyu Gu, Kang Liu, Brendan Dolan-Gavitt, et al. “Badnets: Evaluating backdooring attacks on deep neural networks”. In: IEEE Access 7 (2019), pp. 47230–47244.