

# SLNet: A Spectrogram Learning Neural Network for Deep Wireless Sensing

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**NSDI 2023**

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# Author

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## RESEARCH INTERESTS

- Internet of Things
- Industrial Internet
- sensing and positioning
- edge computing



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# Computer Vision technologies

How human perceive the world

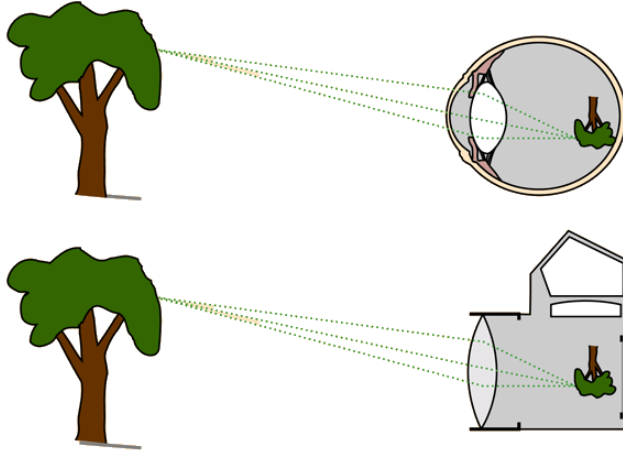


**Vision is our dominant sense.**

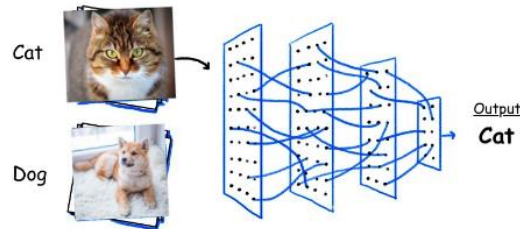
**Eyes manage 80% of all information you will ever take in.**

# Computer Vision technologies

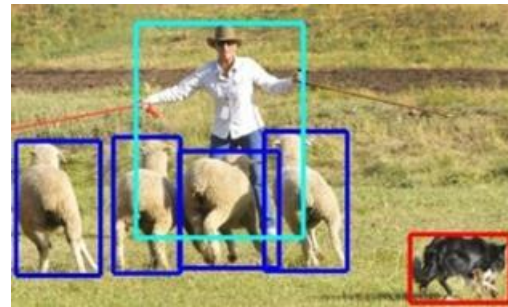
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**Computer Vision (CV)**  
technologies endow computers  
with an effective way of SENSE.



Classification



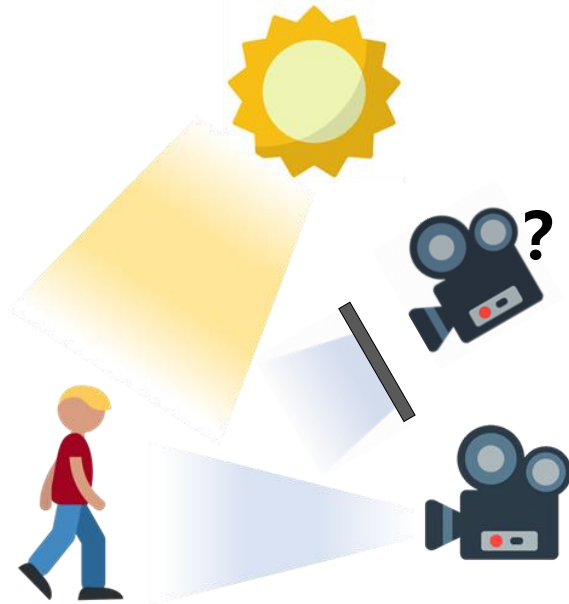
Detection



Segmentation

# The limitations in vision systems

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Occlusions



Privacy

Despite being effective, CV technologies still has limitations including **occlusions, environmental interference, privacy** leakage, etc.

# The chance in wireless signals



 **Powerful application potential**

 **Widely commercialized**



Gesture recognition



Gait detection



Fall detection



Respiration detection

# Related work

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## ➤ Model-based wireless sensing

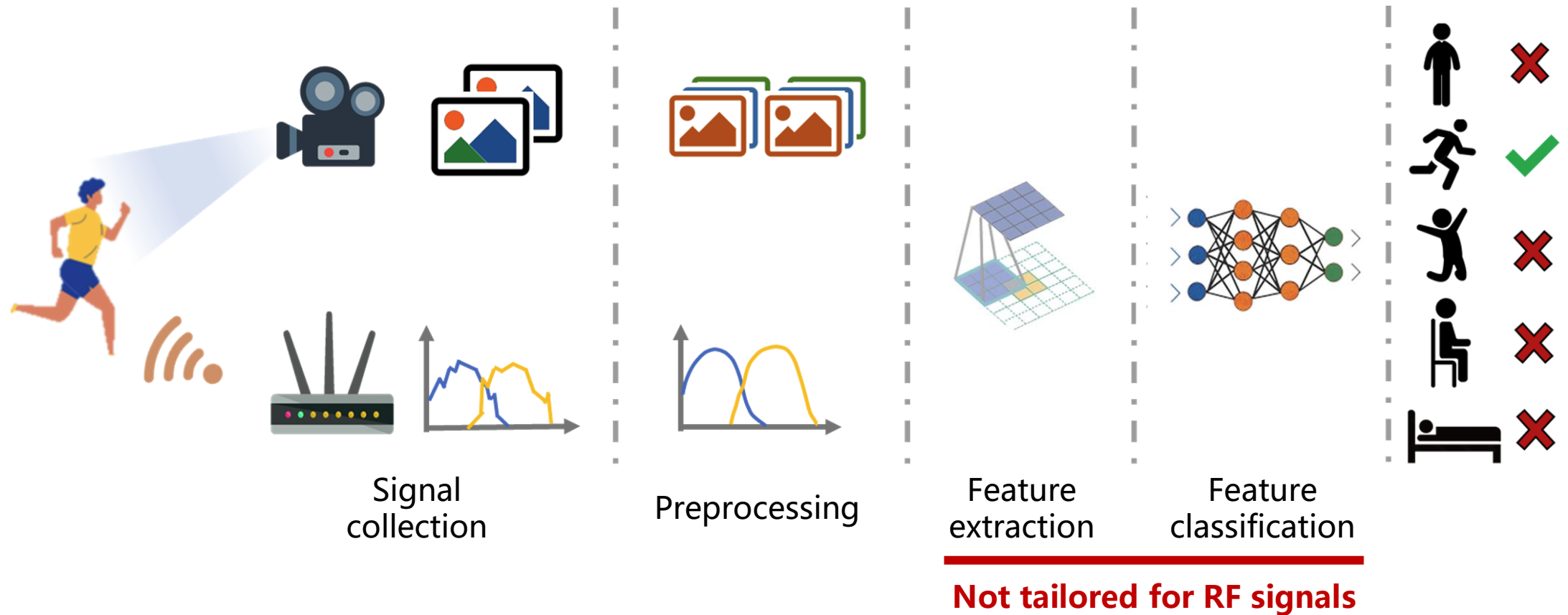
- QGesture: Quantifying Gesture Distance and Direction with WiFi Signals.(IMWUT 2018)
- GaitWay: Monitoring and Recognizing Gait Speed Through the Walls.(MobiCom 2020)
- SiFall: Practical Online Fall Detection with RF Sensing.(SenSys 2022)
- SMARS: Sleep Monitoring via Ambient Radio Signals.(MobiCom 2019)

## ➤ Learning-based wireless sensing

- Widar3.0: Zero-Effort Cross-Domain Gesture Recognition With Wi-Fi.(TPAMI 2021)
- Zero-Effort Cross-Domain Gesture Recognition with Wi-Fi.(MobiSys 2019)
- mSilent: Towards General Corpus Silent Speech Recognition Using COTS mmWave Radar.(IMWUT 2023)



# Motivation

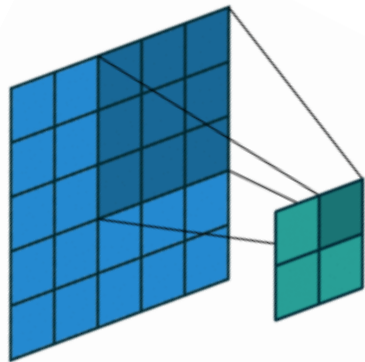
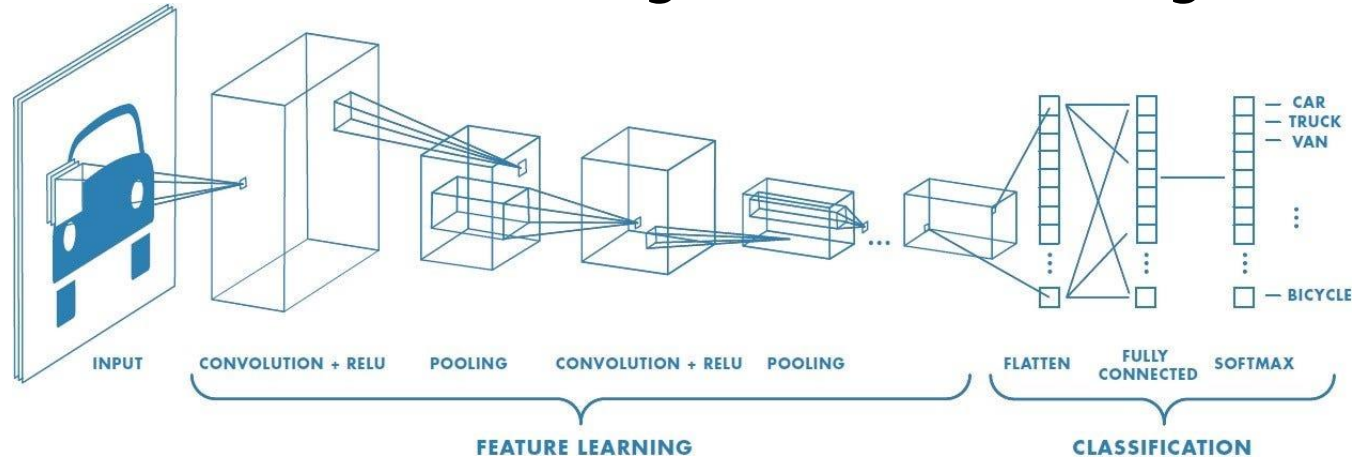


- Non-visual: RF data contains physical and geometric connotations in time, space, and frequency domains that are not visually intelligible.
- Complex: RF data is complex-valued with both amplitude and phase information.



# Motivation

Convolutional neural network is not designed for wireless signals.



Share weights across different locations

1 <sub>x1</sub>	1 <sub>x0</sub>	1 <sub>x1</sub>	0	0
0 <sub>x0</sub>	1 <sub>x1</sub>	1 <sub>x0</sub>	1	0
0 <sub>x1</sub>	0 <sub>x0</sub>	1 <sub>x1</sub>	1	1
0	0	1	1	0
0	1	1	0	0

Image

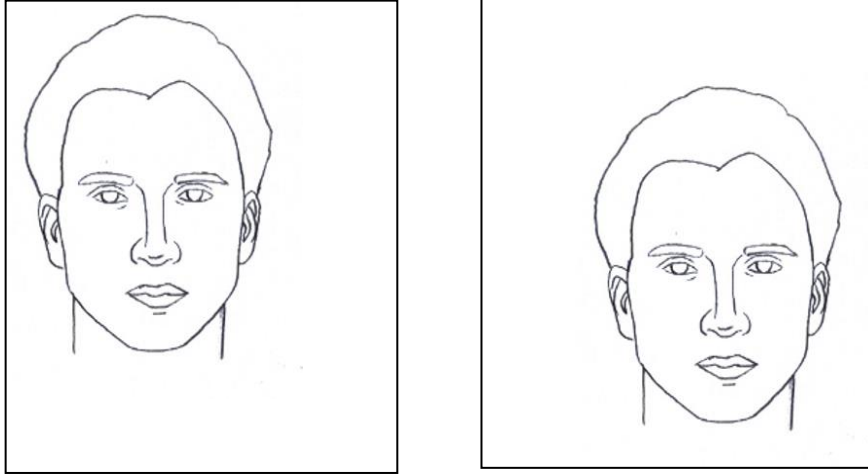
4		

Convolved Feature

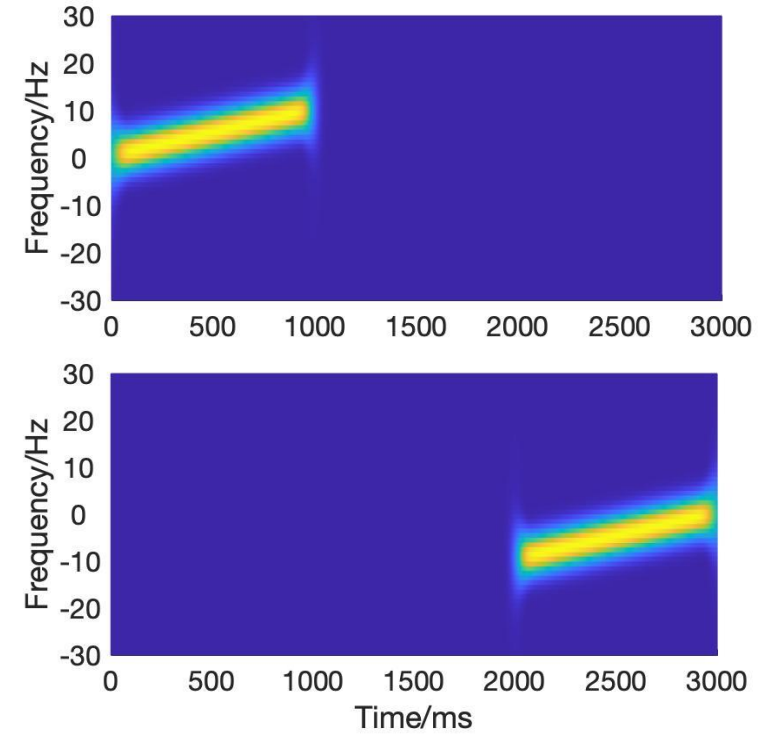
The convolution operation takes a local field of the input, discarding the global locations.

# Motivation

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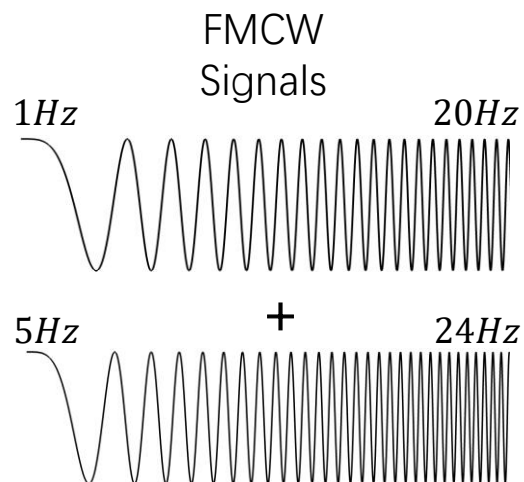
CNN is tailored for images since it is invariant to shifts.



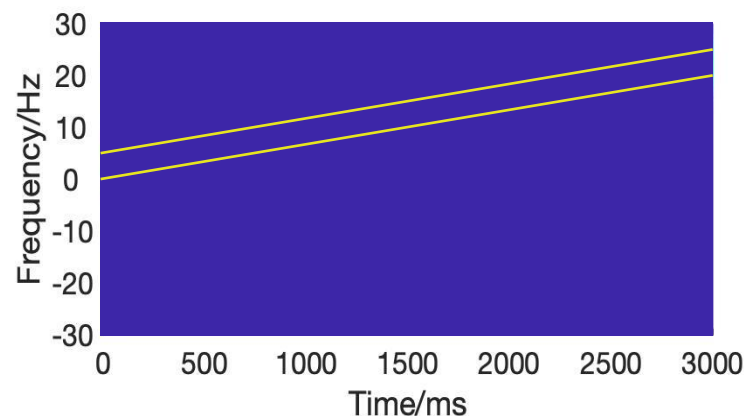
Wireless signals require global discriminations.

# Motivation

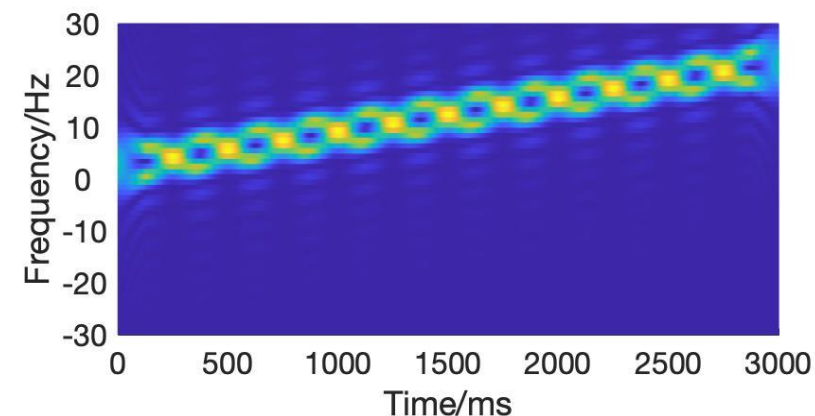
**The blur in spectrograms cause difficulty for recognition.**



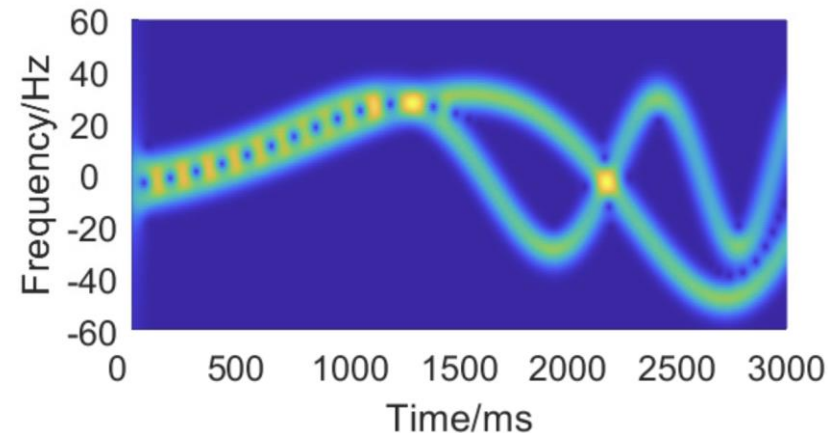
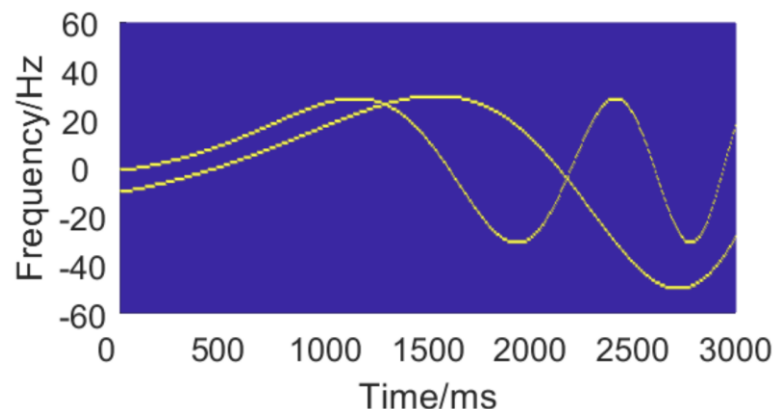
The expected spectrogram



The actual spectrogram

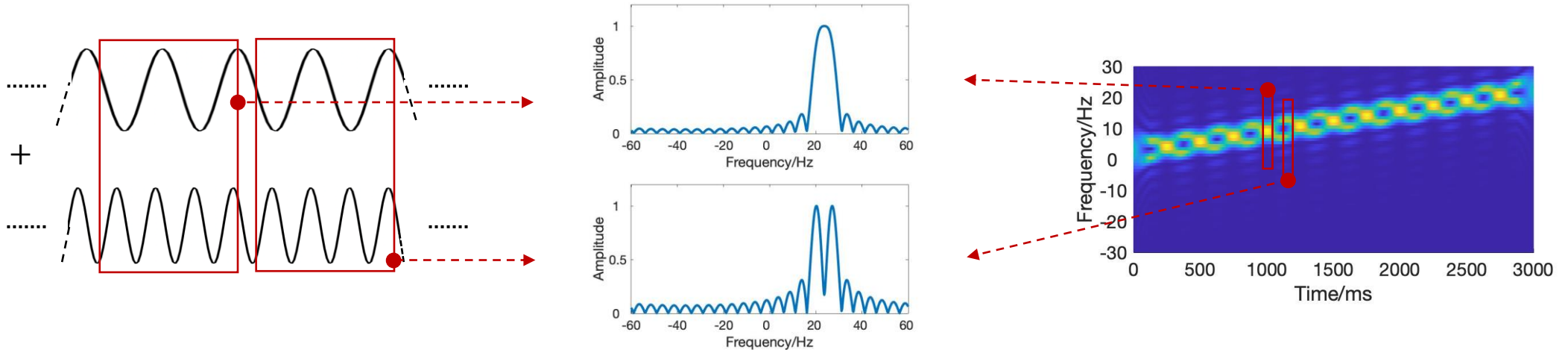


The interference is more significant for smaller frequency difference.



# Motivation

The interference caused by spectral leakage is **unstable**.

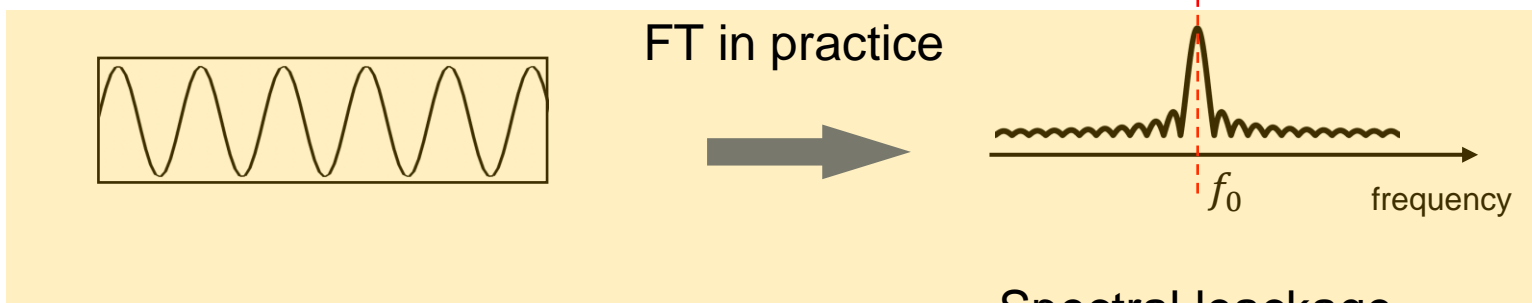
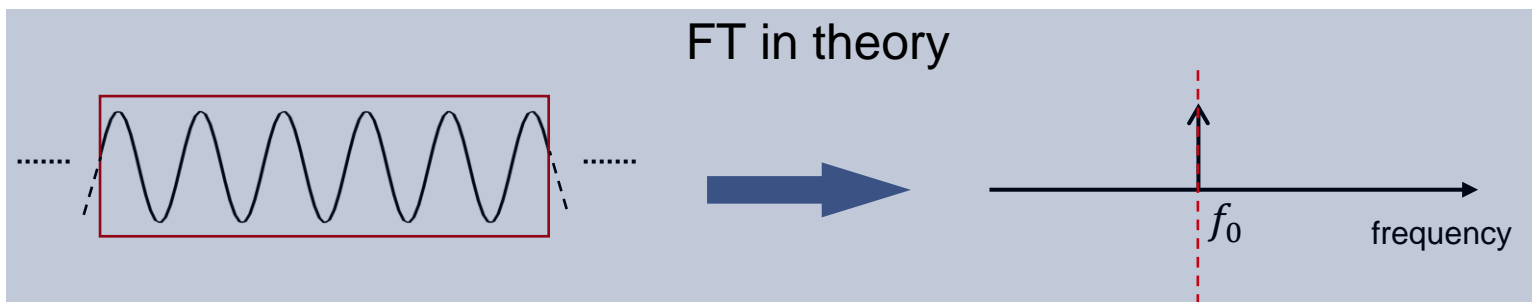


Different initial phase cause different interference patterns.

# Motivation

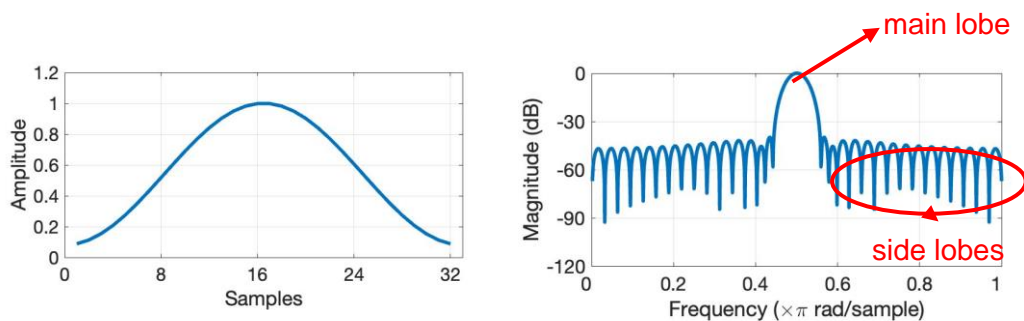


It is all about Fourier Transform (FT)

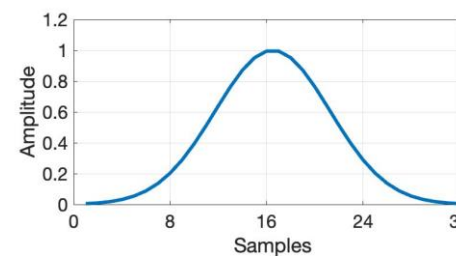


$$\text{Sine wave} = \text{Rectangular window} \times \text{Sine wave}$$

Spectral leakage

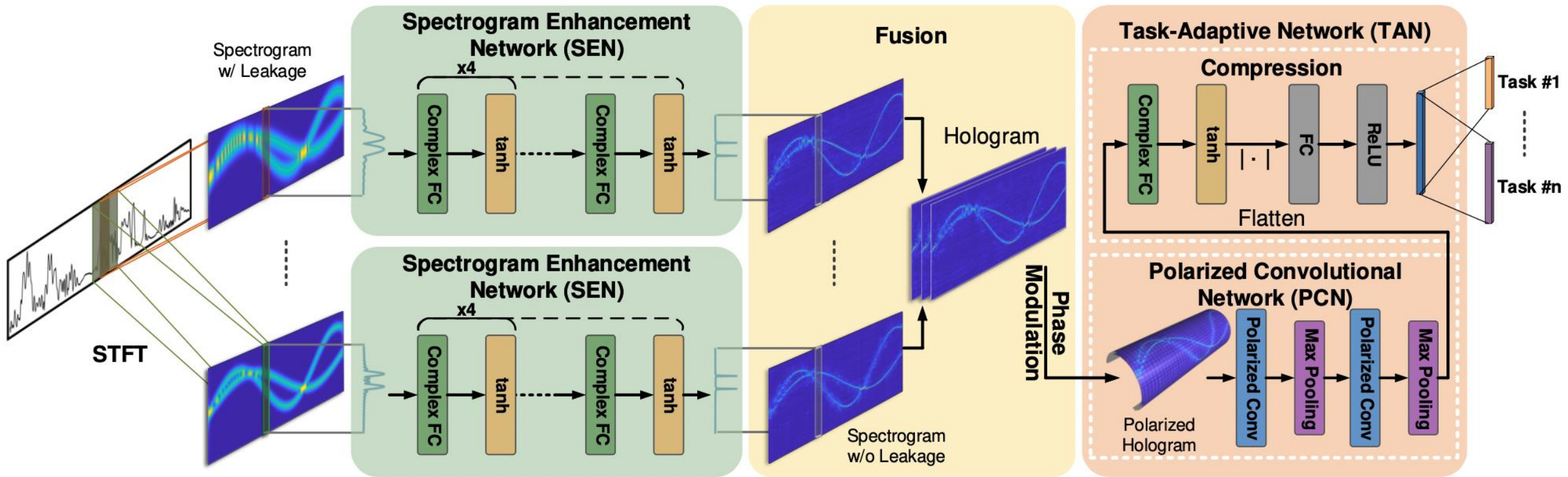


Hamming



Gaussian

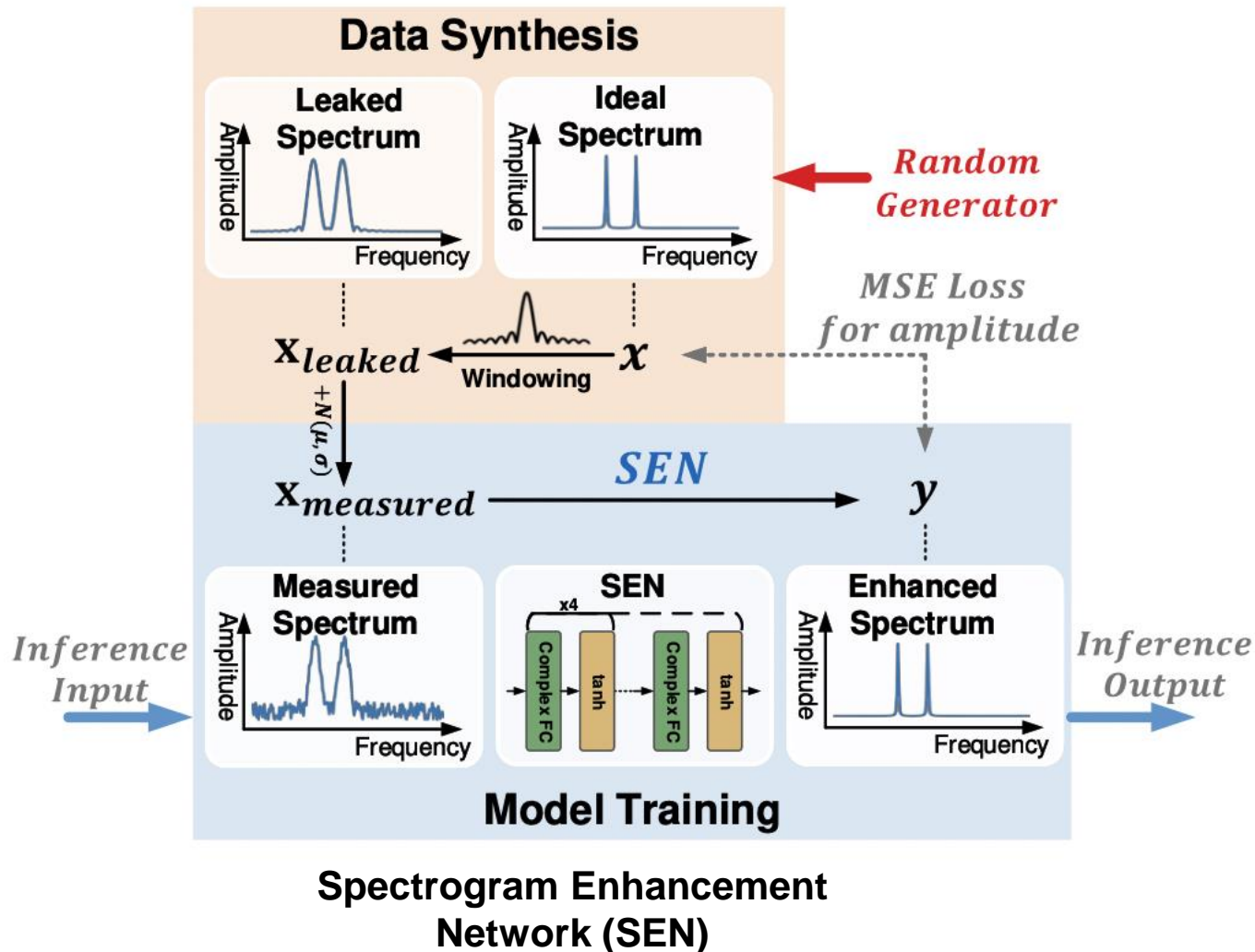
# Design





# Design

## Learning-Assisted Spectrogram Enhancement



Generate an ideal spectrum with **1 to 5** frequency components randomly.

amplitudes	phases	frequencies
[0,1]	[0,2 $\Pi$ ]	[-60,60]Hz

$$\hat{s} = As + n$$

$S$ : ideal frequency spectrums

$\hat{S}$ : estimated frequency spectrums

$A$ : windowing effect

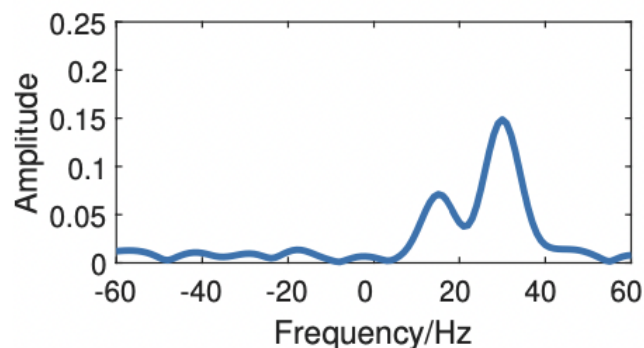
$n$ : random complex noises



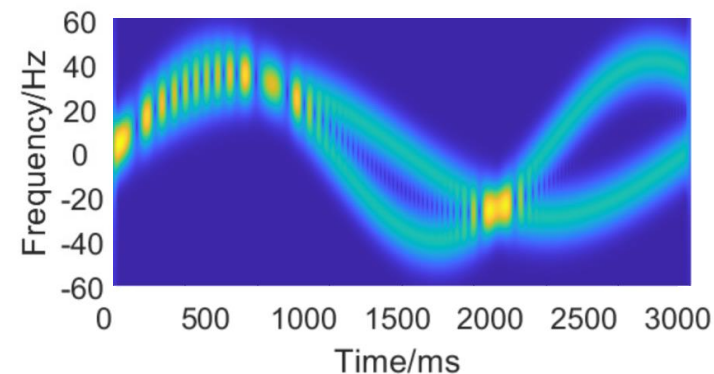
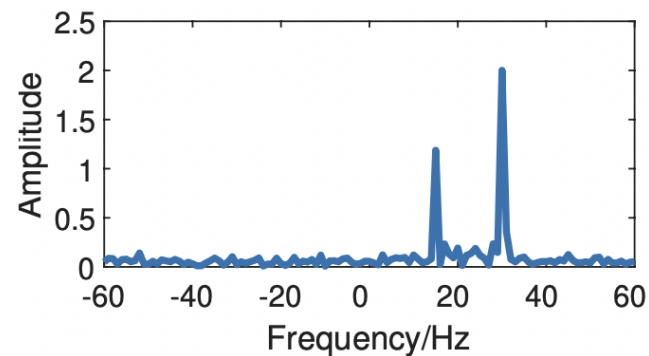
# Design

## Learning-Assisted Spectrogram Enhancement

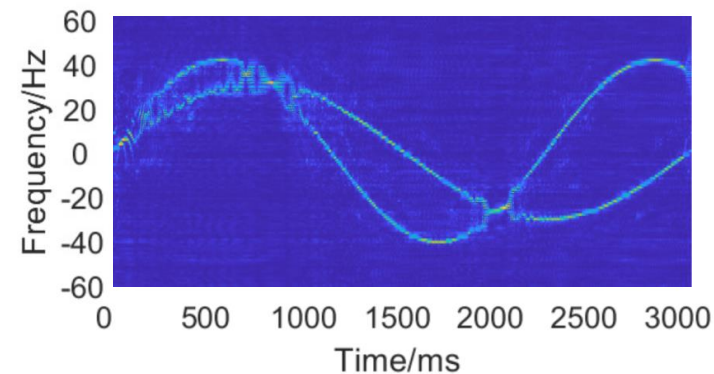
SEN results



SEN

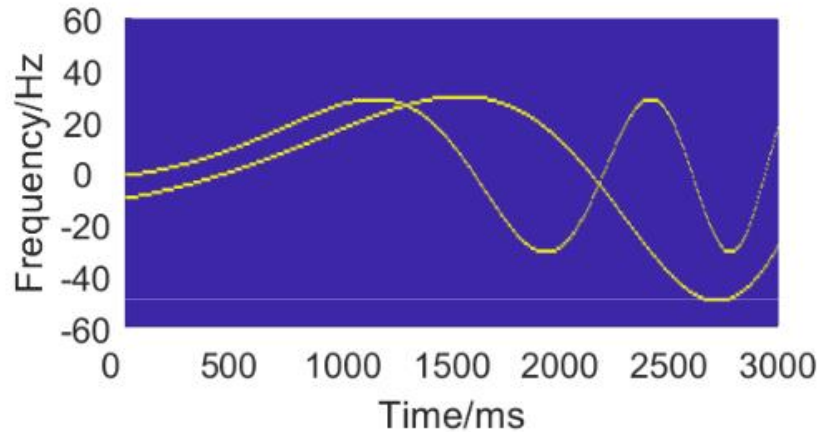


SEN

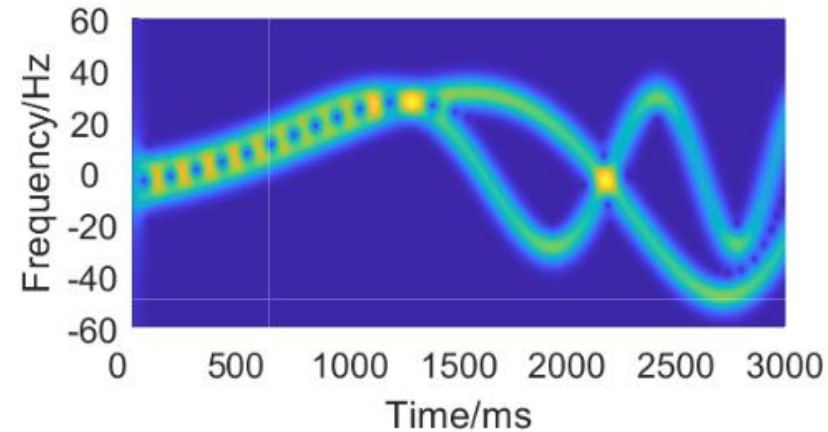


Frequency components are more distinguishable after SEN enhancement.

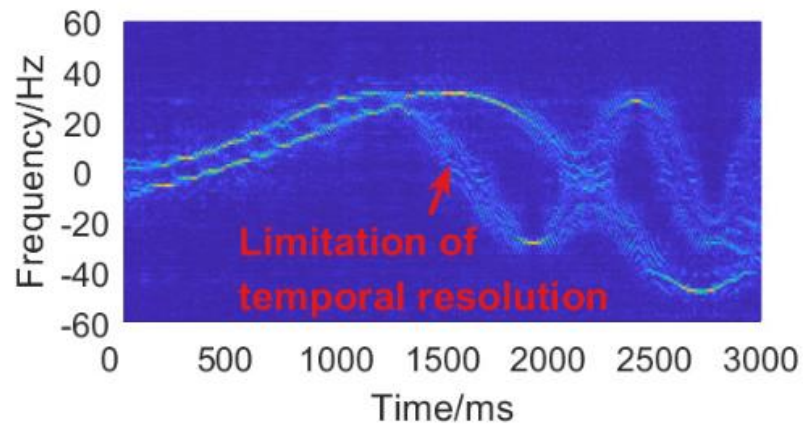
# Design



(a)



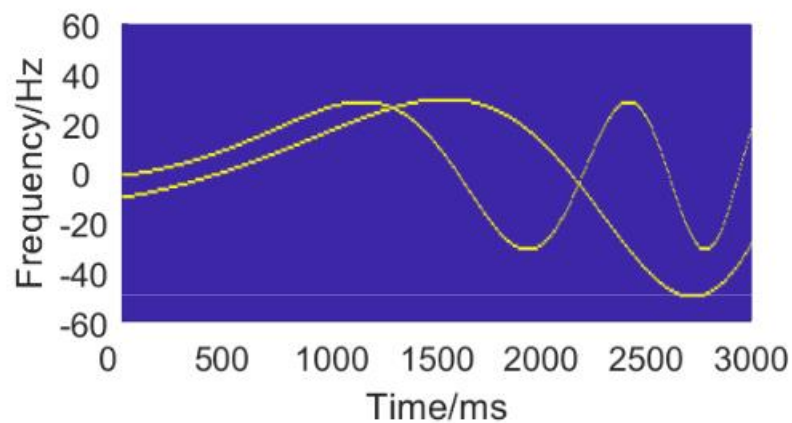
(b)



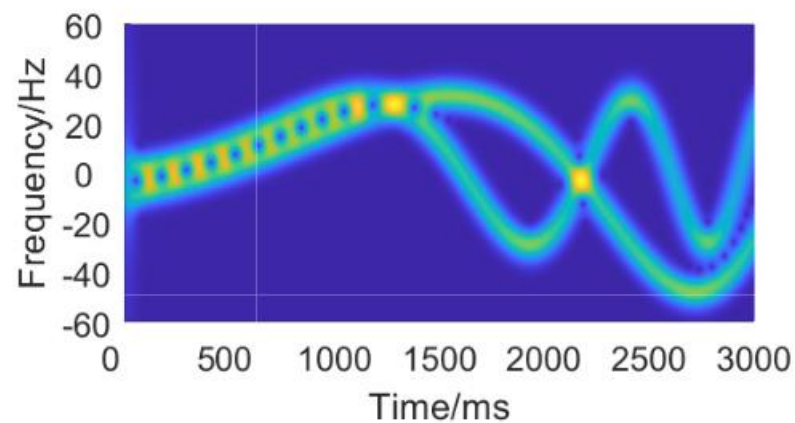
(c)

**SEN** assumes the frequency components remain quasi-static during the sliding window period of generating the spectrum.

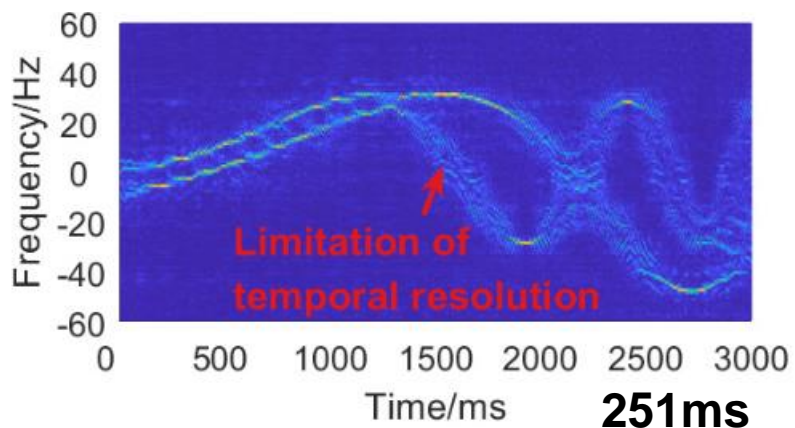
## Multi-Resolution Spectrogram Fusion



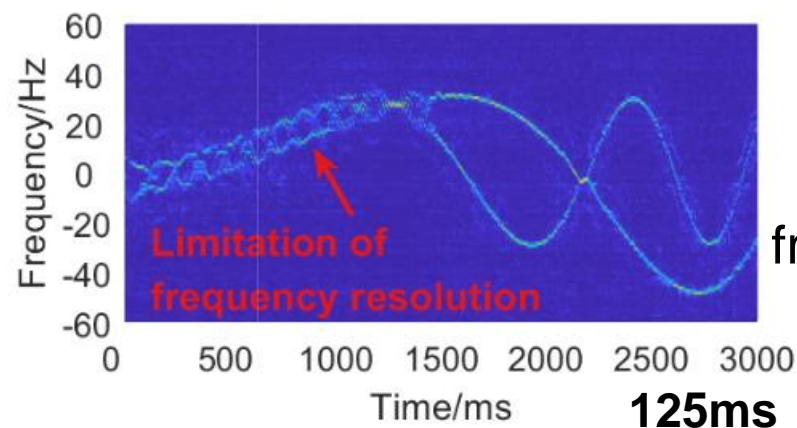
(a)





(b)



(c)

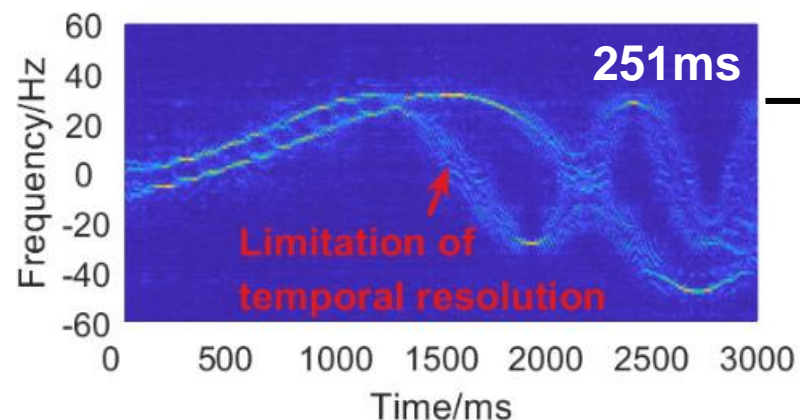


(d)

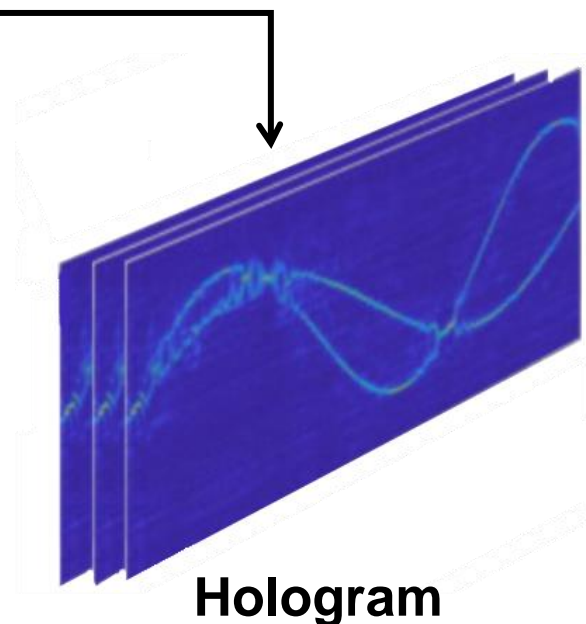
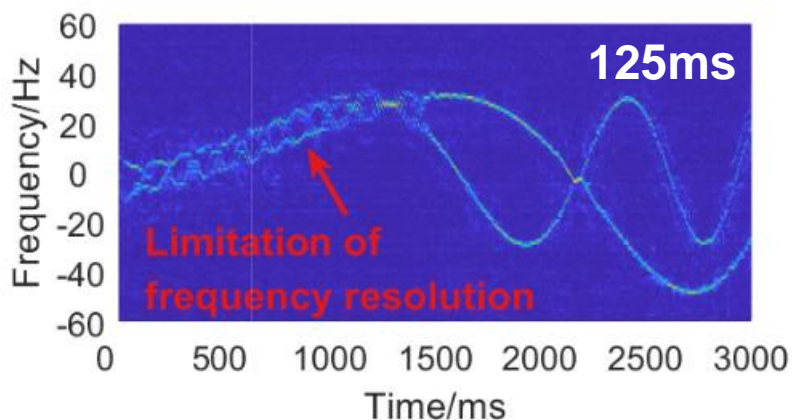
sliding window   
↓  
frequency resolution 

# Design

## Multi-Resolution Spectrogram Fusion

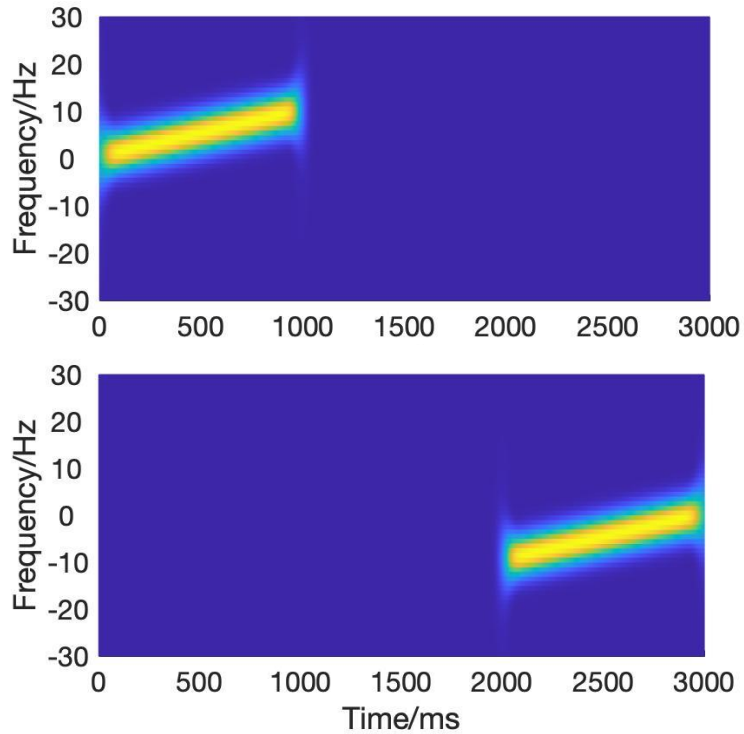


sliding windows with  
different lengths

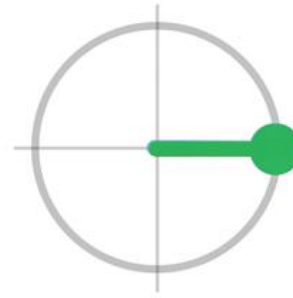


Each sliding window's spectrogram is processed using a SEN pre-trained on synthesis spectrograms of the same length.

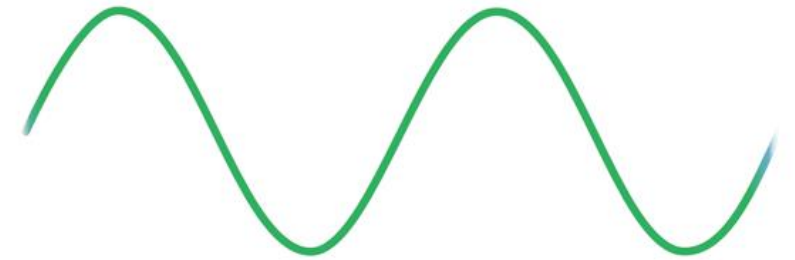
# Design



Wireless signals require global discriminations.



Difference:  $0^\circ$



In Phase

$$f(t) = A \sin(\omega t + \theta)$$

Phase refers to the offset of a waveform in time or space; different phase values result in the wave appearing at different positions.

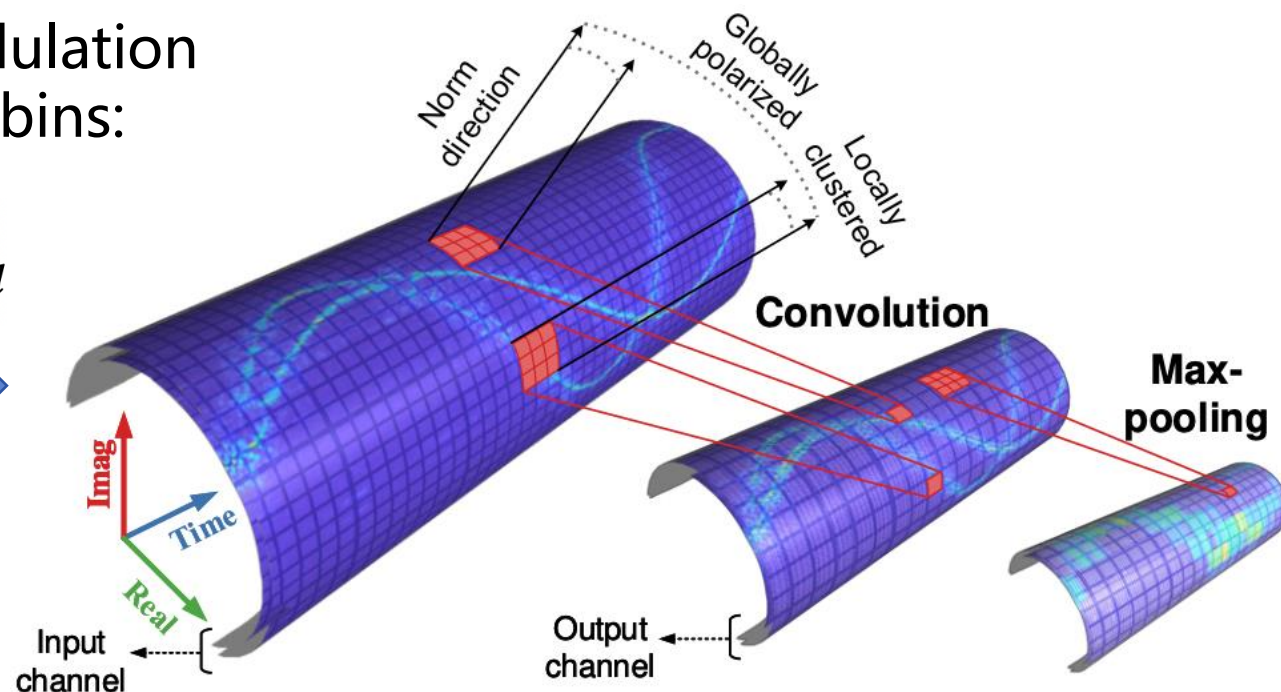
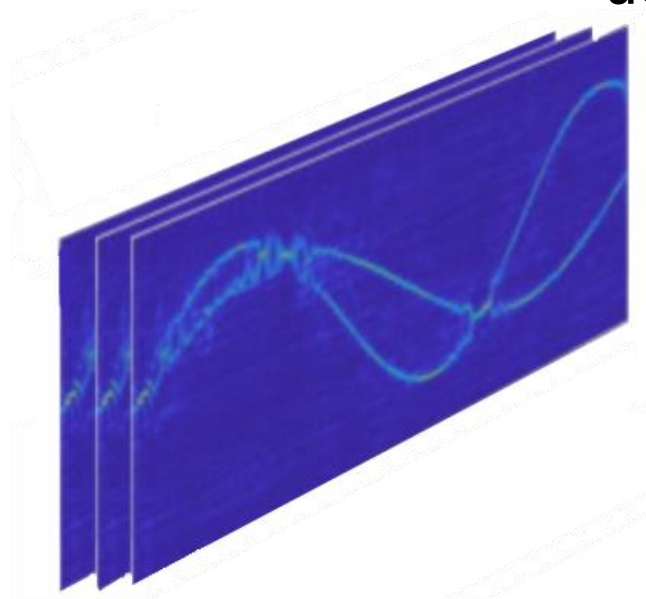


# Design

## Phase-polarized feature extraction

Linear phase modulation  
across frequency bins:

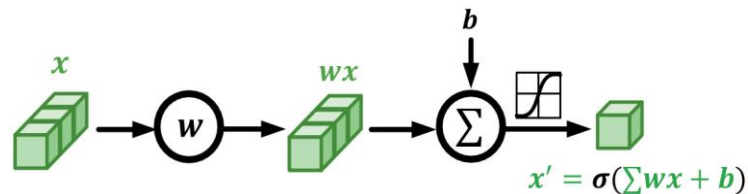
$$\phi_i = i \frac{\phi_h - \phi_l}{M} + \phi_l$$



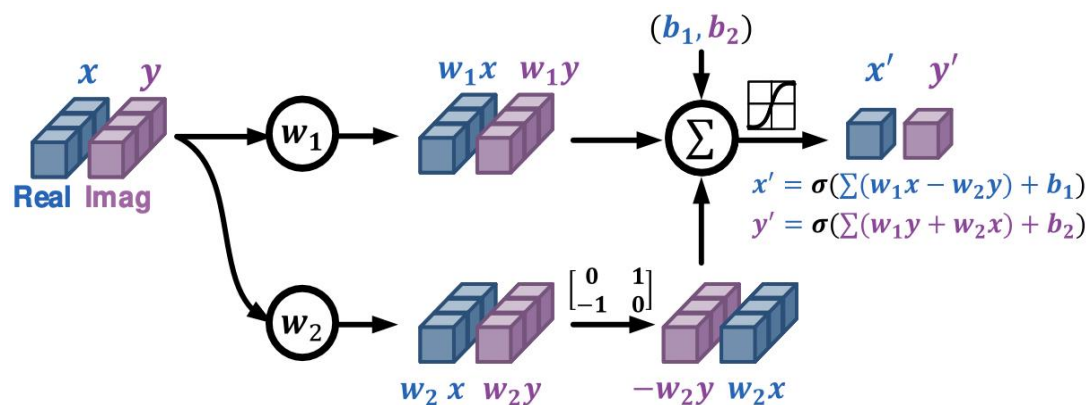
global discrimination is introduced along the frequency dimension

# Design

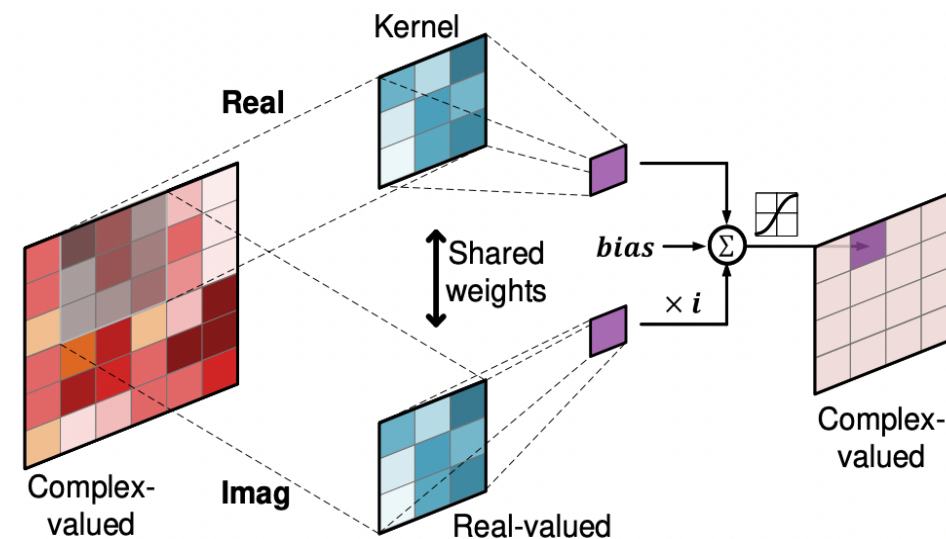
## Complex-Valued Neural Network



Real-valued neurons



Complex-valued neurons



Complex-valued convolution



# Design

## Task-Adaptive Network

SLNet calculates the absolute value of the output from the complex-valued FC layer and inputs it to the real-valued FC layer.

### additional FC

+ softmax



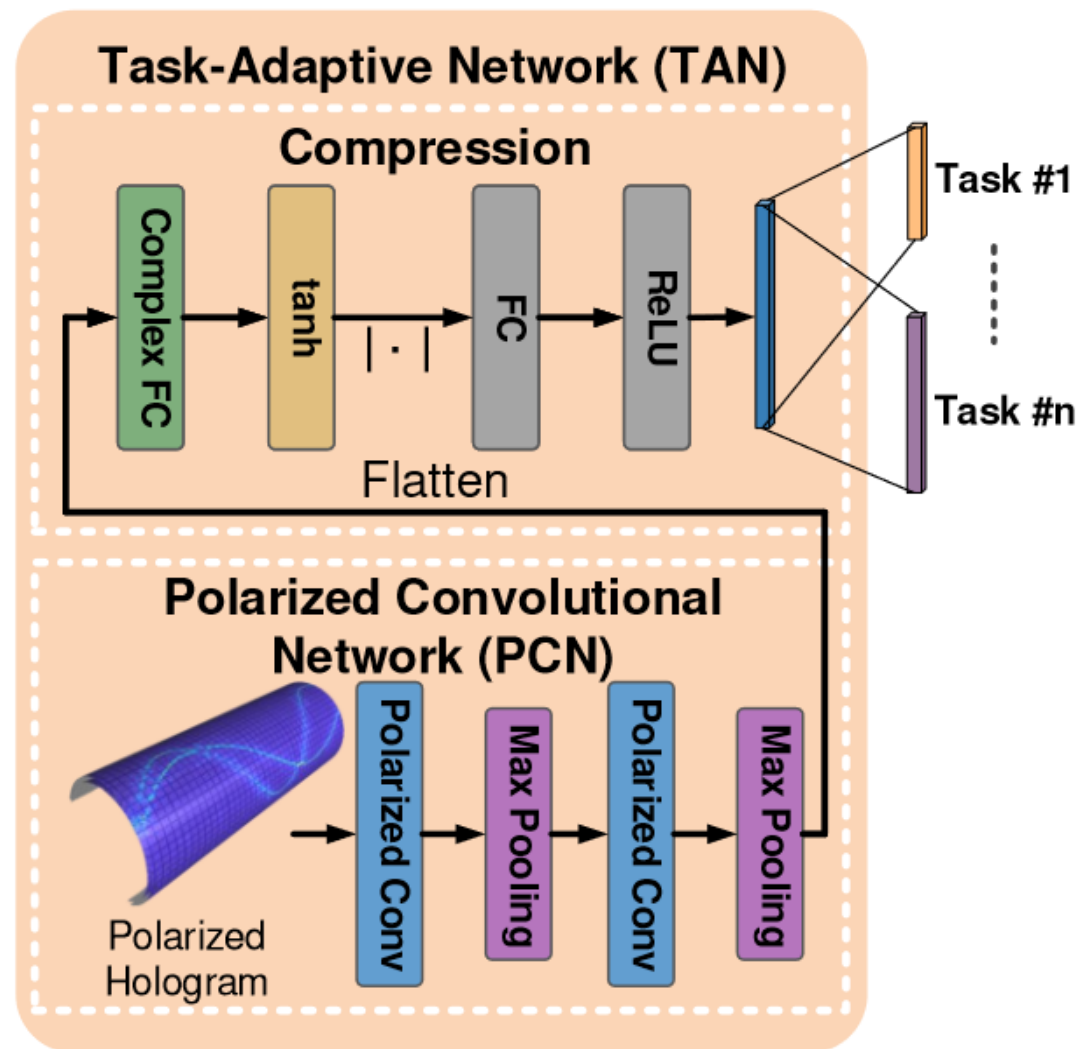
gesture classification

+ sigmoid



Fall detection

### TASK



# Evaluation

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- **Hardware:**

- Intel 5300 WiFi Network Interface Cards (NICs)

- NVIDIA GeForce 2080Ti GPU

- Ubuntu 10.04

- Linux CSI Tool (log CSI readings)

- **Dataset:**

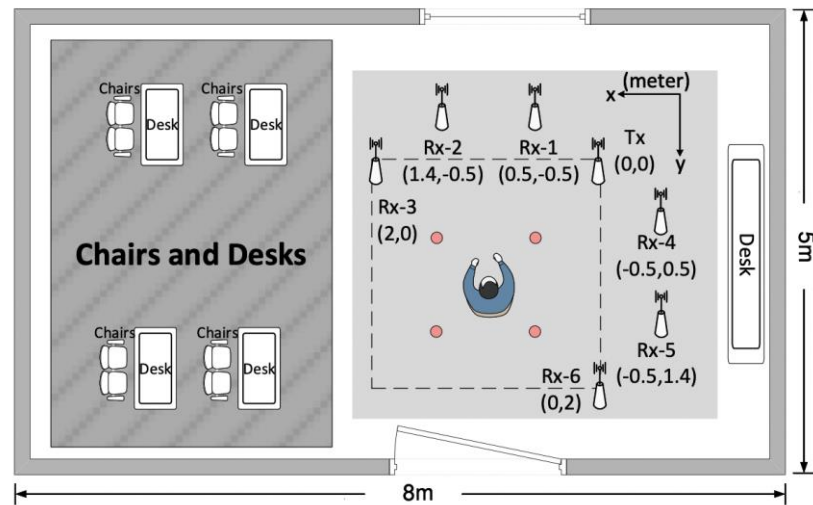
- Widar 3.0: WiFi-based Activity Recognition Dataset (Gait&Gesture)

- <https://ieee-dataport.org/open-access/widar-30-wifi-based-activity-recog>

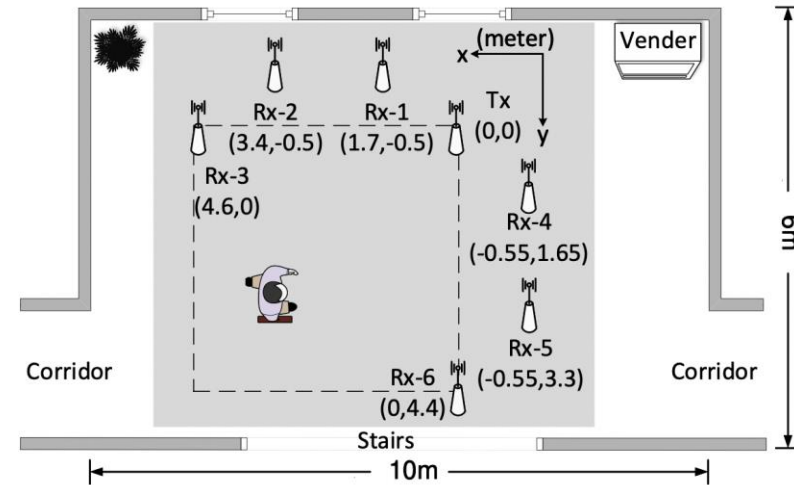
- **Code:**

- <https://github.com/SLNetRelease/SLNetCode>

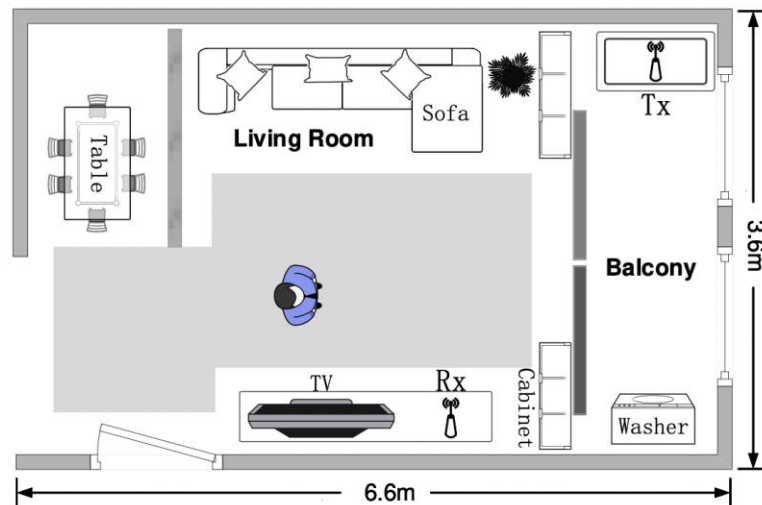
# Evaluation



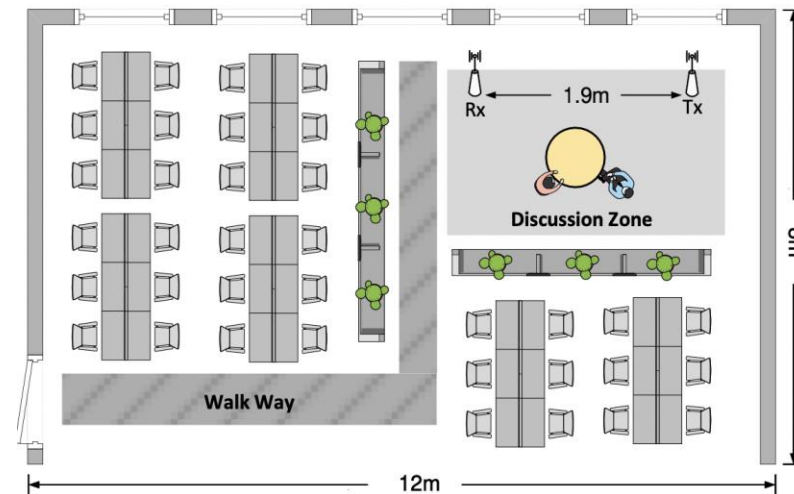
Human gesture recognition  
with Wi-Fi



Human gait identification  
with Wi-Fi



Fall detection within home  
environment with Wi-Fi



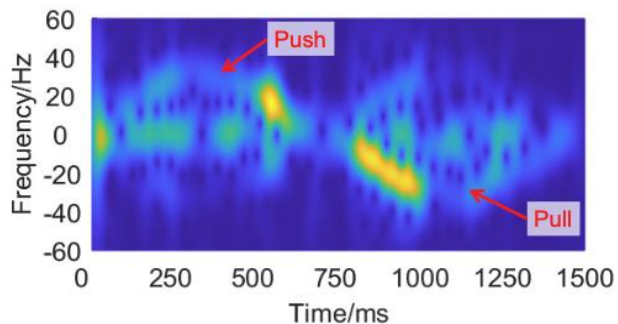
Monitoring the breath rate of  
two users with Wi-Fi

# Evaluation

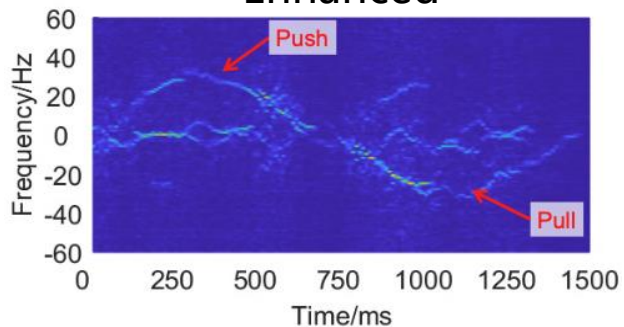
## Human gestures



Raw FFT

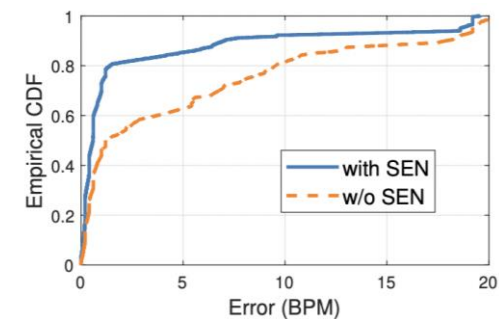


Enhanced

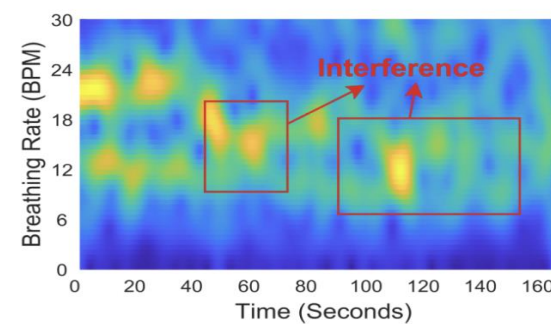


## Human breath

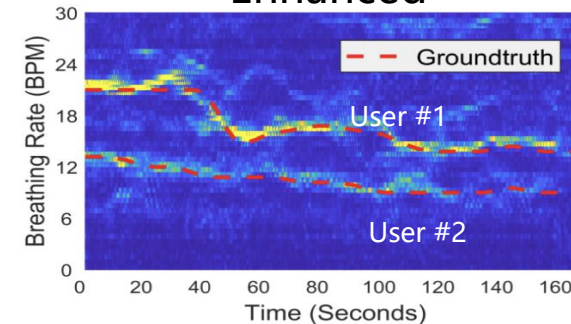
Estimation Error



Raw FFT



Enhanced

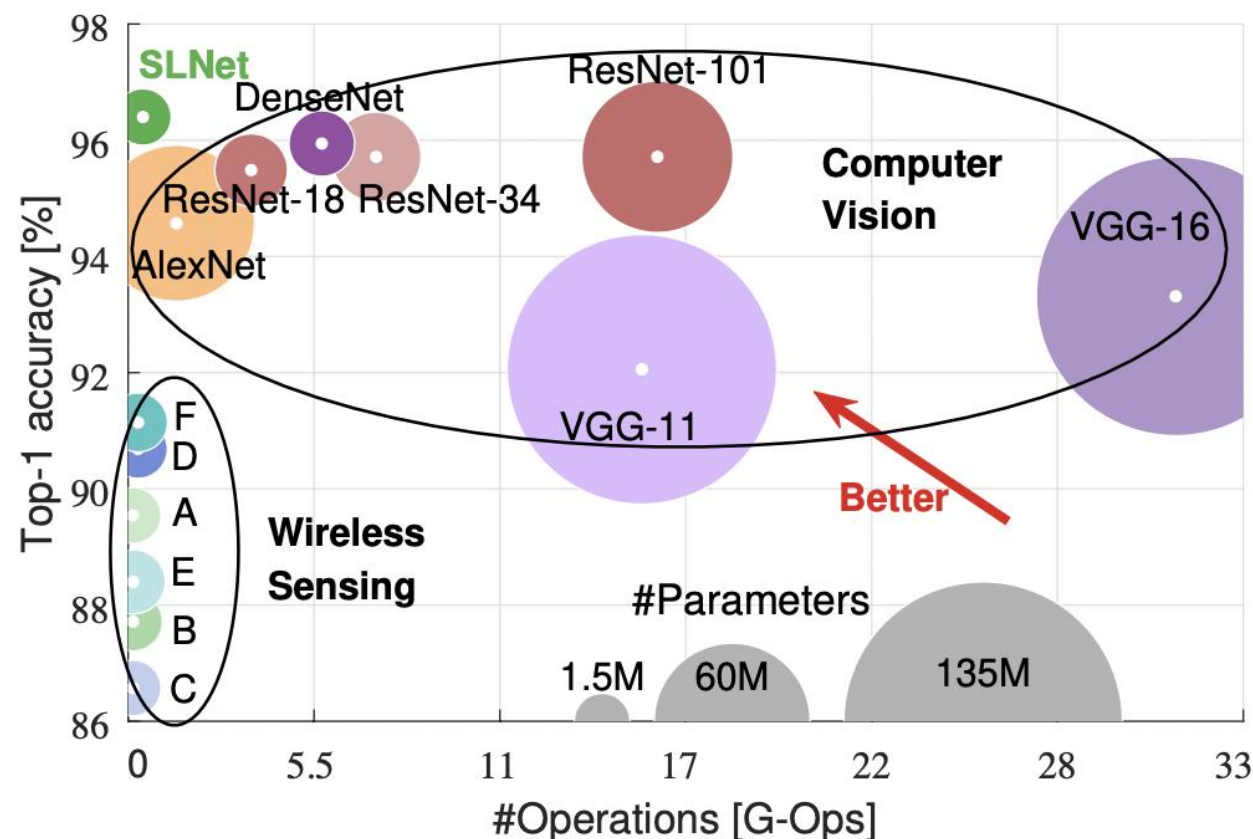


Frequency components becomes more distinguishable.

# Evaluation

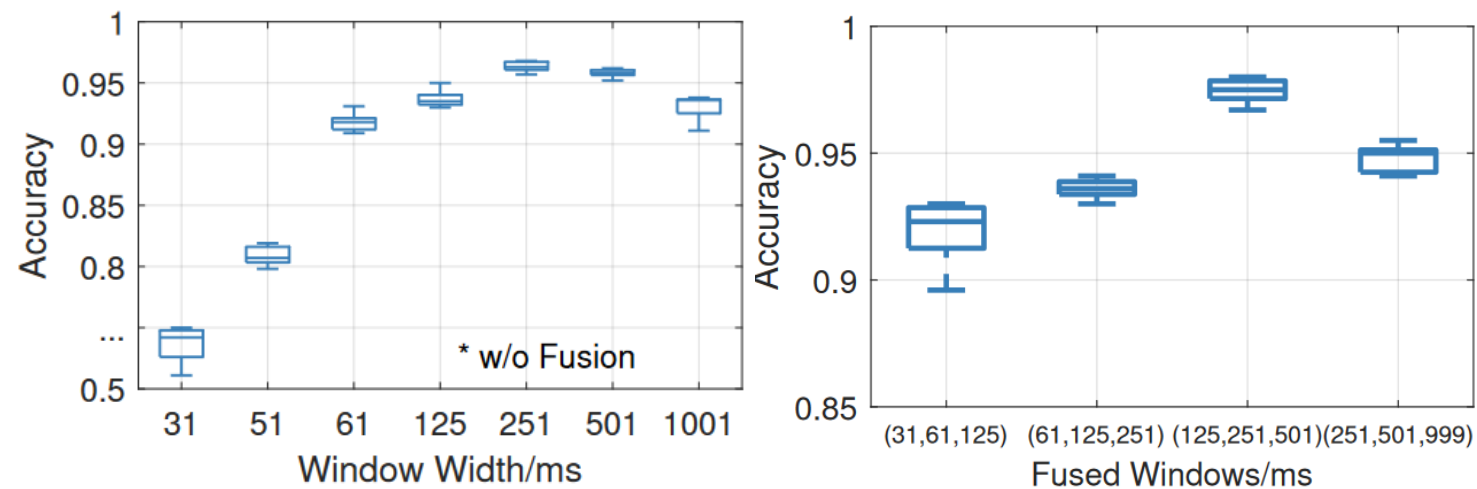
## Recognition tasks

Modality	Ref.	Gesture	Gait	Fall <sup>1</sup>	Para <sup>2</sup>
WiFi	[23, 90]	90.6%	95.1%	92.8%, 96.3%	1.07M
	[8, 22]	89.0%	96.6%	96.4%, 84.3%	2.72M
	[39, 79]	84.3%	83.3%	96.8%, 93.8%	5.77M
	[73] <sup>3</sup>	78.9%	70.9%	95.5%, 96.8%	0.06M
FMCW	[87]	88.0%	95.4%	96.0%, 96.0%	1.06M
	[84, 86]	91.6%	96.4%	99.7%, 95.7%	2.76M
Acoustic	[30]	89.6%	95.4%	90.6%, 98.3%	6.08M
	[40]	88.3%	90.1%	95.3%, 95.3%	128.8M
Vision	[15]	91.9%	96.6%	97.0%, 95.6%	11.18M
	[20]	91.0%	97.7%	99.8%, 96.3%	6.96M
CVNN	[17, 32]	72.3%	96.0%	95.2%, 93.7%	115.6M
	[46]	92.0%	96.3%	98.4%, 93.8%	2.94M
<b>WiFi</b>	<b>SLNet</b>	<b>96.6%</b>	<b>98.9%</b>	<b>99.8%, 97.2%</b>	<b>1.48M</b>

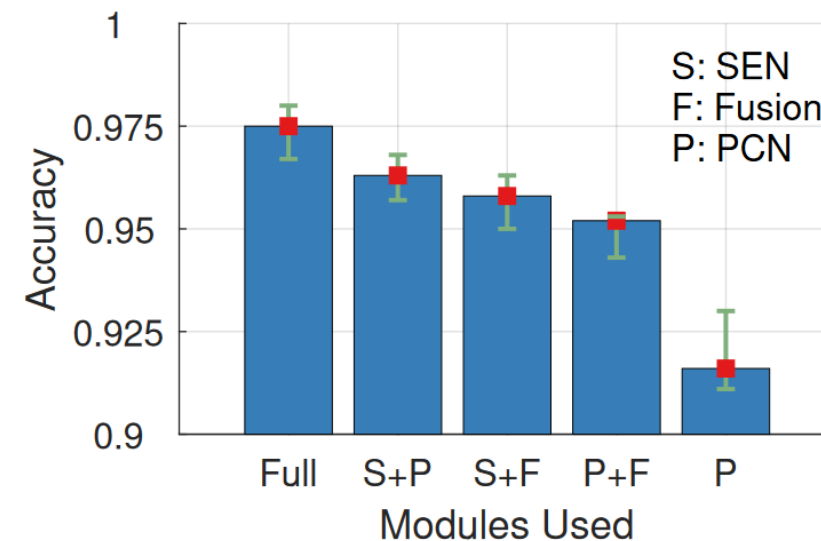


Less parameters, but higher performance.

# Evaluation



A single-resolution spectrogram for wireless sensing tasks is not the optimal solution

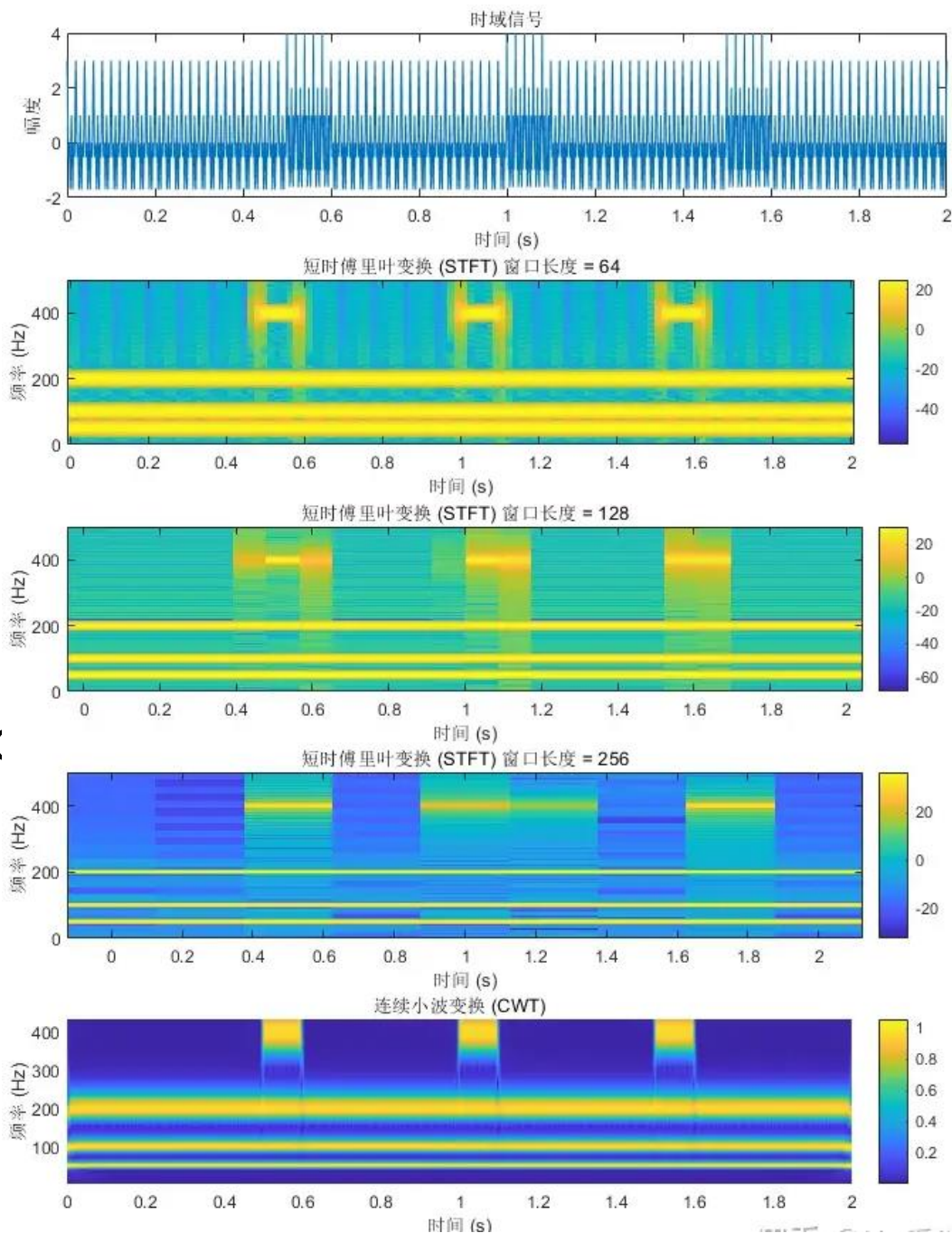


Spectral leakage problem cannot be neglected in wireless sensing tasks



# idea

- ◆ Utilize **phase** information in signal processing.
- ◆ Mitigate the impact of **spectral leakage** on spectral analysis.
- ◆ Use **wavelet transform** to solve the multi-window problem.





**请老师和同学们批评指正！**

汇报人：陈宇杰