

# PAWS: An Intelligent and Wearable System for Pedestrian Safety

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## Objective



- Reduce the increasing number of accidents involving distracted pedestrians in urban areas
- Use modern techniques to develop a low-power low-cost wearable solution to detect and localize oncoming cars

## System

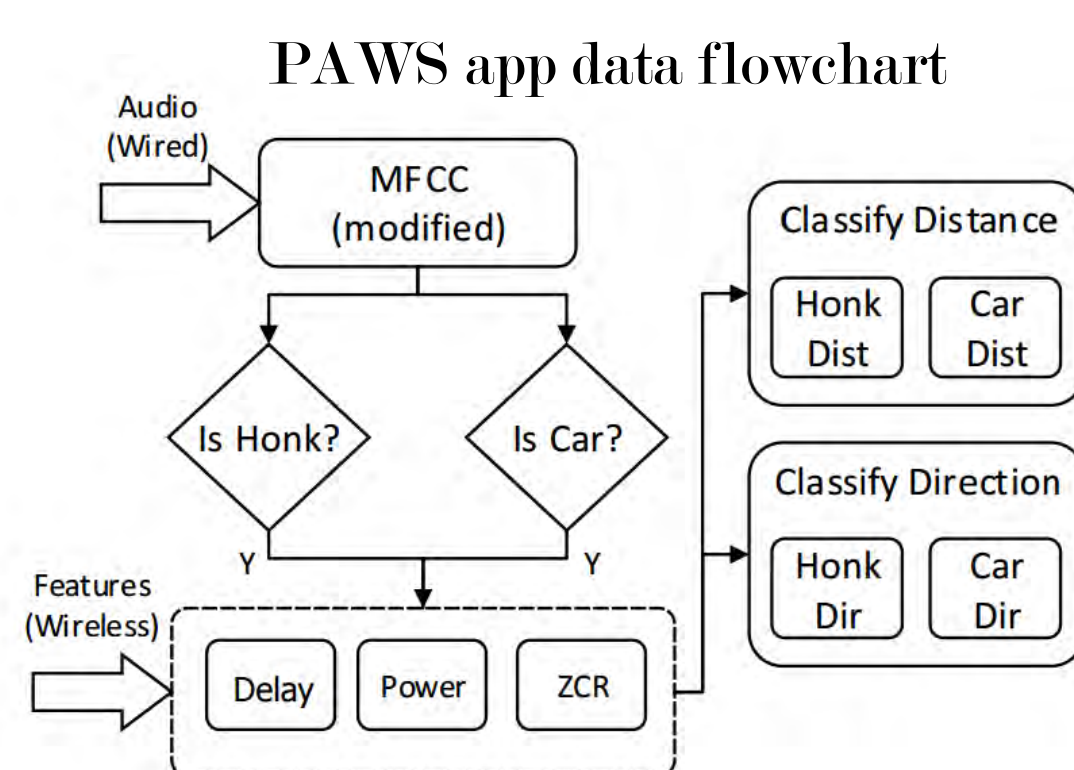
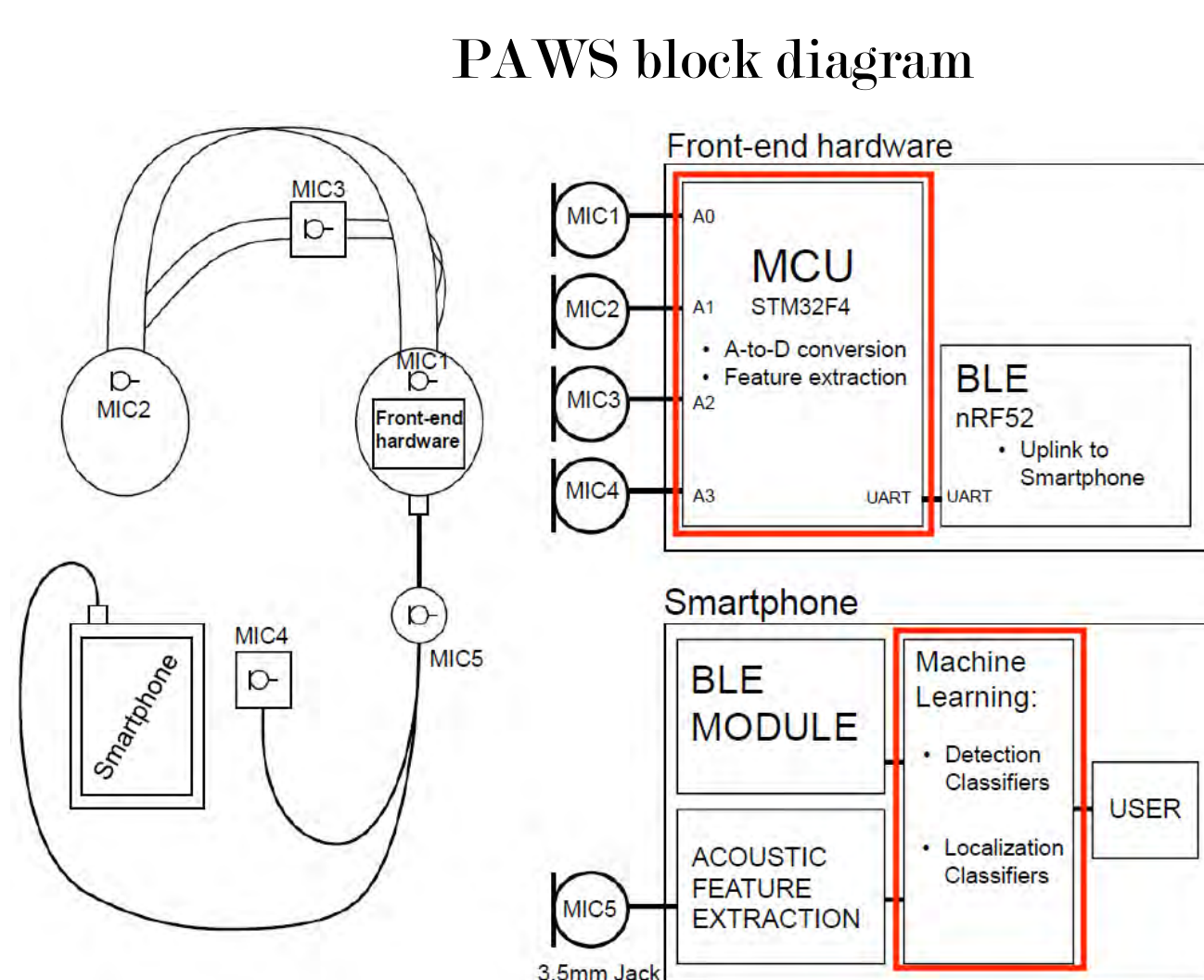
**Microphones** – 4 MEMS microphones are sampled synchronously

**Front-end hardware** – Extracts spatiotemporal features from mics and transmits to smartphone app.

- V1: MCU computes relative delay, zero-crossing rate, relative power
- V2: custom ASIC replaces MCU to reduce power consumption

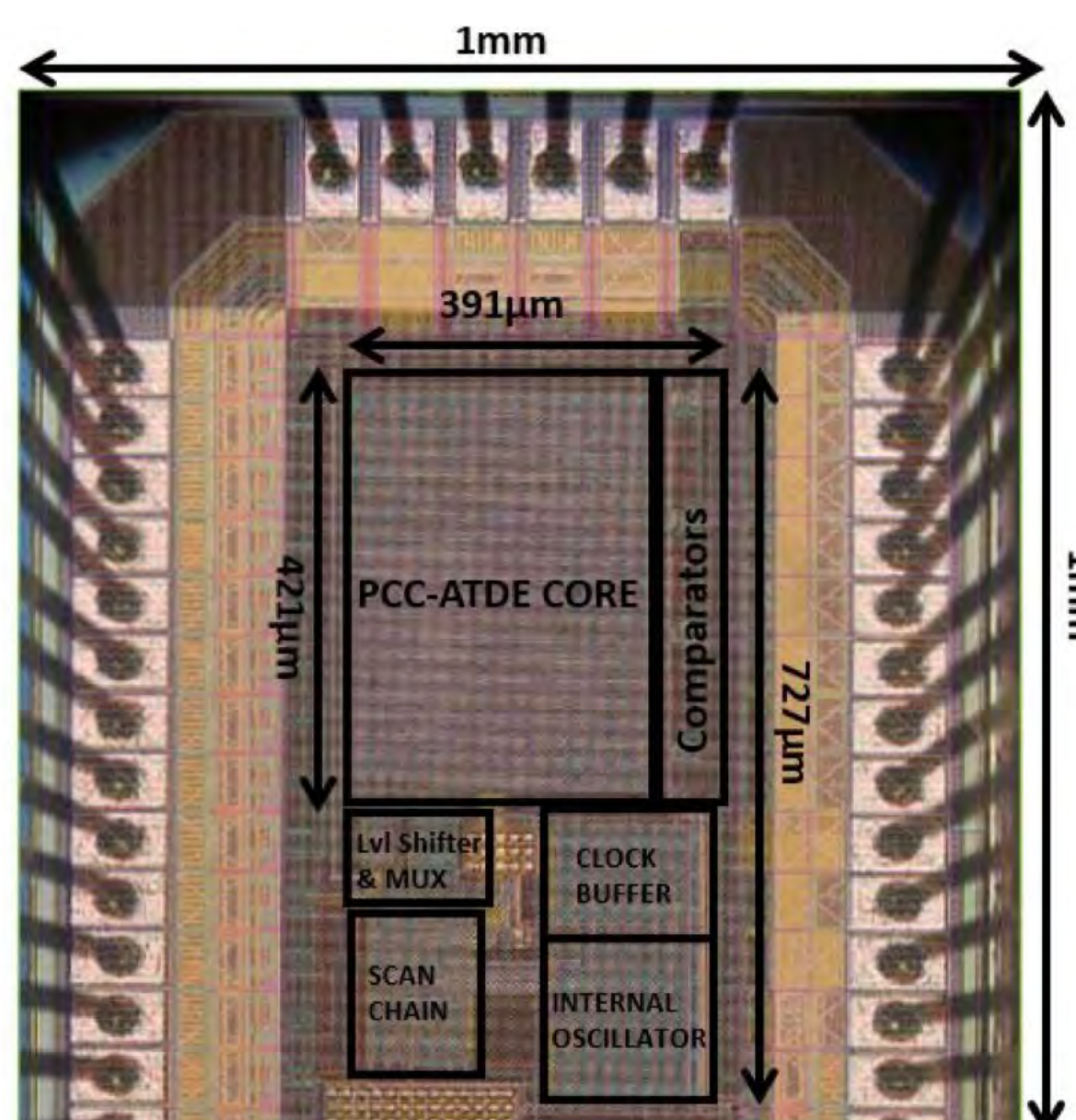
**Smartphone** – Samples from a fifth microphone and uses data from the front-end hardware to detect and localize:

- V1: runs machine learning algorithms to detect and localize
- V2: runs machine learning algorithms for detection and runs our novel Angle via Polygonal Regression method to localize



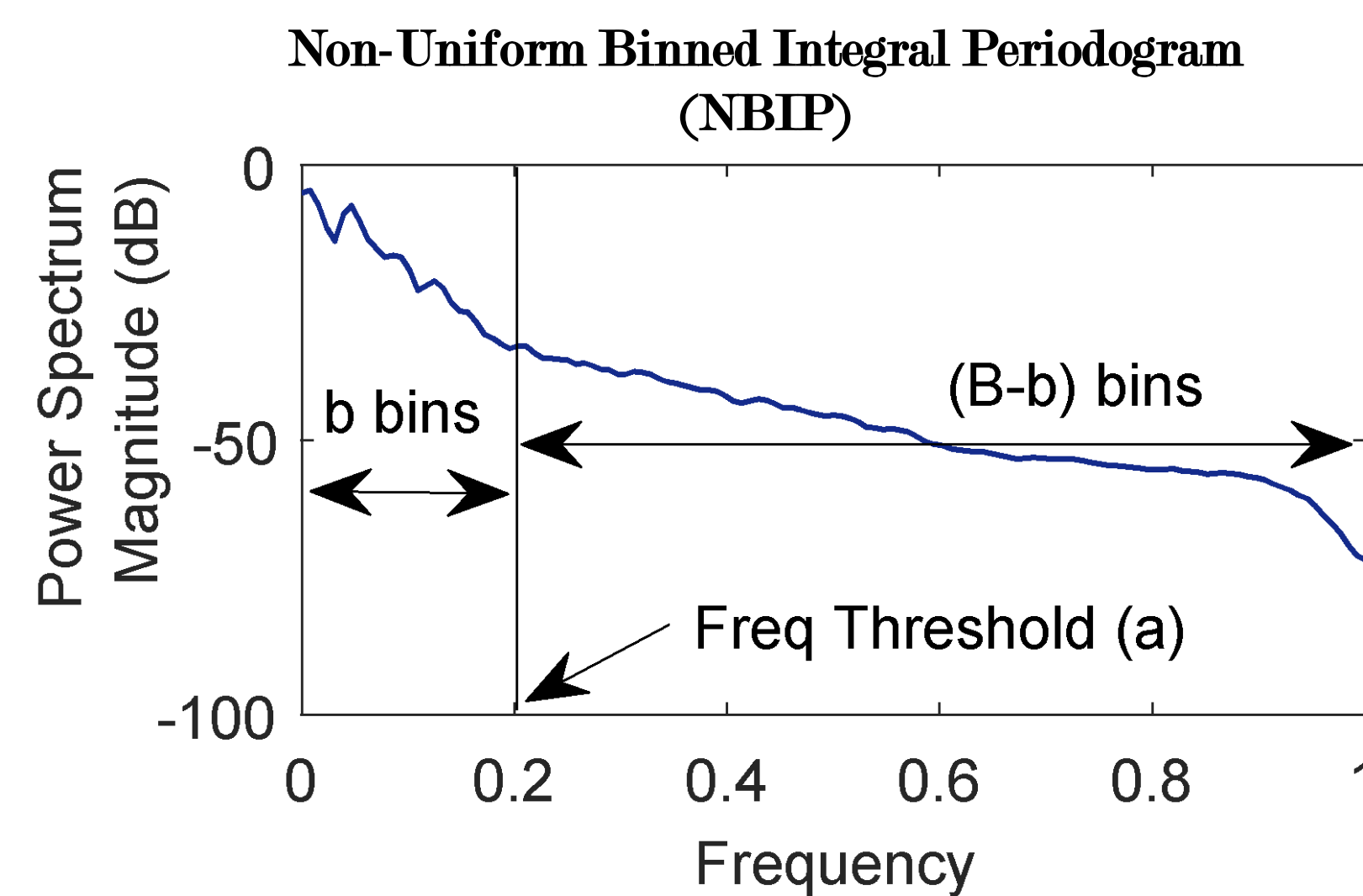
## Custom ASIC

PERFORMANCE SUMMARY	
Technology	180nm CMOS
Sampling Frequency	50Ks/s
Range	[-2.52ms, 2.52ms]
$T_{iss}$	20 $\mu$ s
Peak INL	-1.57/1.33 LSB
Peak DNL	-0.85/0.97 LSB
TDE ENOB	6.06bits
Number of TDE Channels	3
Comparator Power (Unit)	3.1nW
PCC-ATDE Core Power	65.9nW
Total Power	78.2nW
TDE Energy per Conversion Step per Channel	7.84fJ/Conv.-Step
Active Area	0.28mm <sup>2</sup>
Die Area	1mm <sup>2</sup>



Custom application-specific integrated circuit replaces MCU for ultra-low power relative delay extraction from headset mics

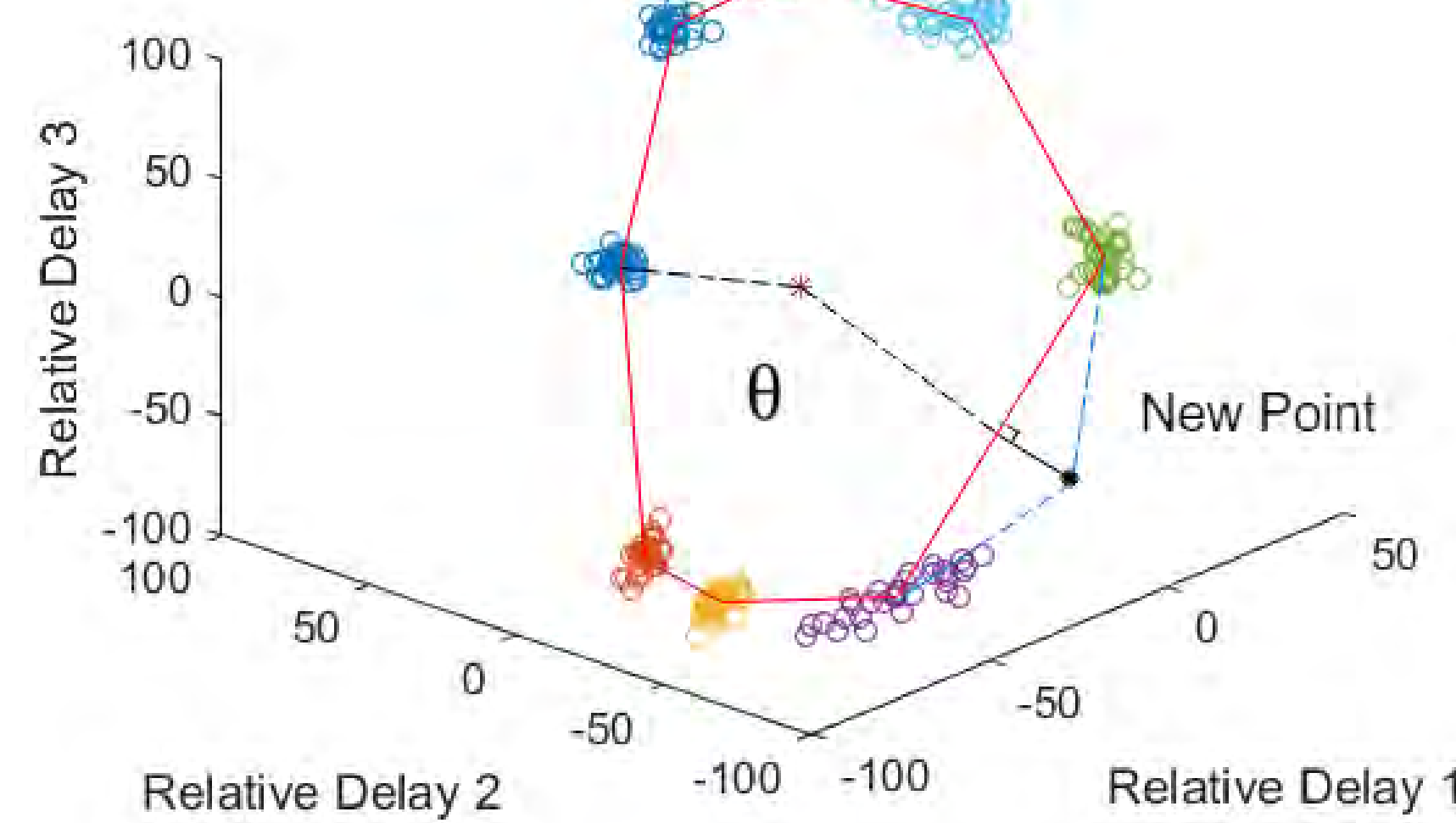
## Car Detection



- MFCC, a common feature for sound classification, is unable to separate noise-like car tire sounds from environmental noise
- Non-Uniform Binned Integral Periodogram (NBIP):**
  - Better separates car tire sounds from environment
  - Bins low frequencies into finer bins and high frequencies into coarser bins
- Random Forest classifier** used to classify NBIP features as car or non-car sounds

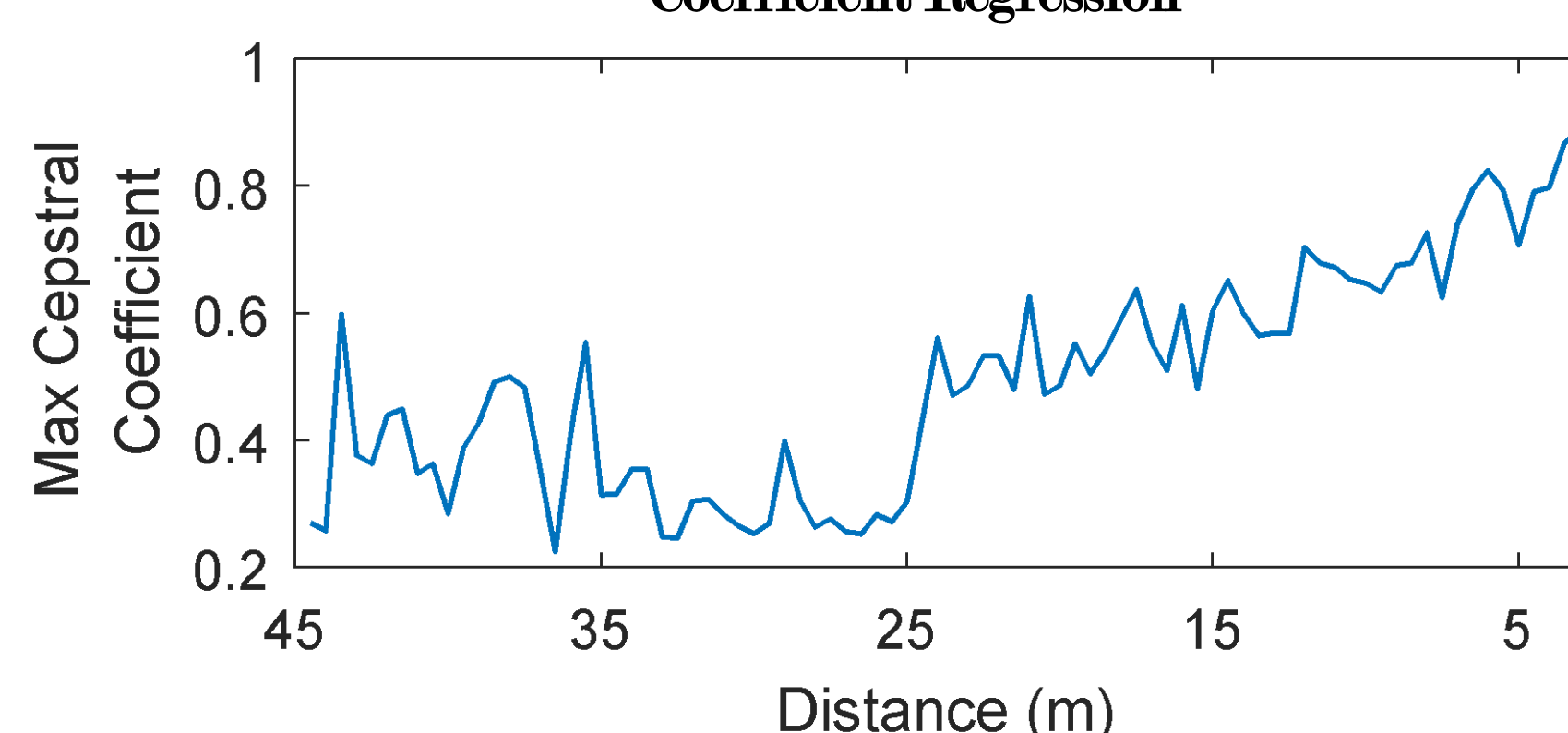
## Car Localization

**Direction Estimation: Angle via Polygonal Regression (AvPR)**



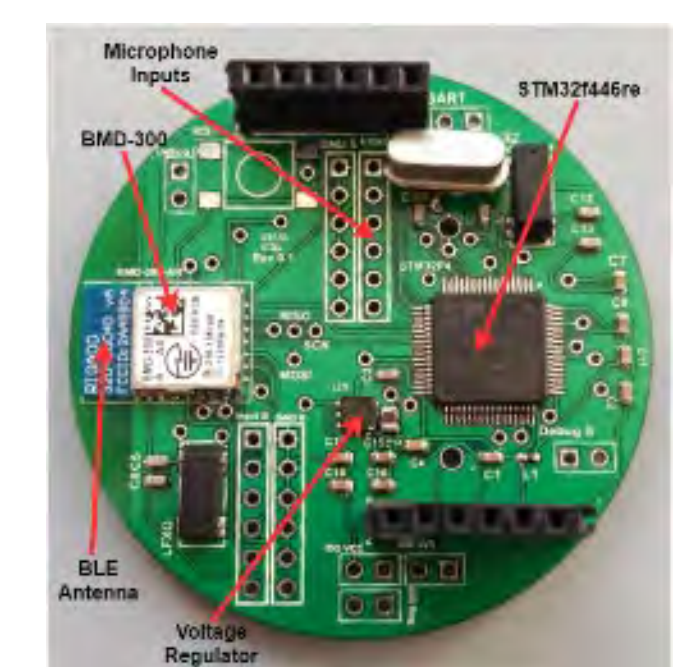
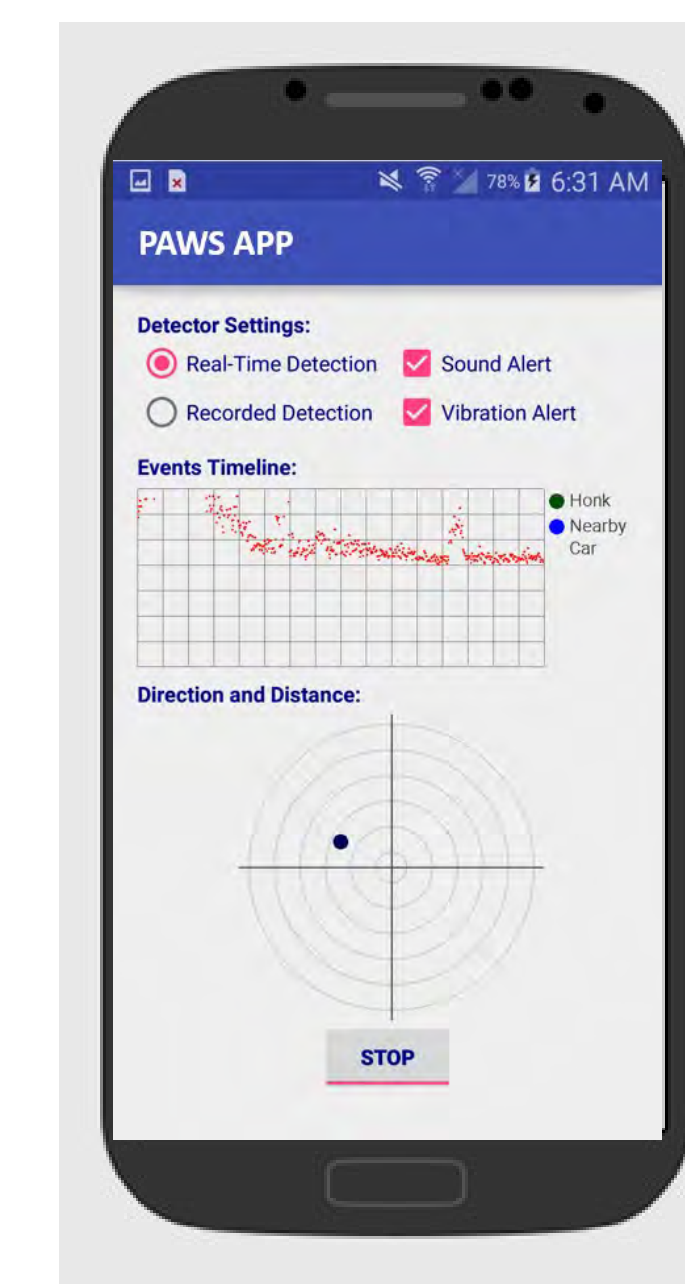
- Classical direction estimation via TDOA has high error if position of mics are not stable
- ML classification requires more training classes/examples the finer the granularity
- Angle via Polygonal Regression (AvPR):** perform direction regression over the polygon shape formed from relative delay features observed at different directions

**Distance Estimation: Maximal Cepstral Coefficient Regression**



- Fine-grained distance estimation by performing regression over the maximal cepstral coefficients
- Beyond ~30m the coefficients show no noticeable pattern, resulting in two coarse distance classes (< 30m, > 30m)

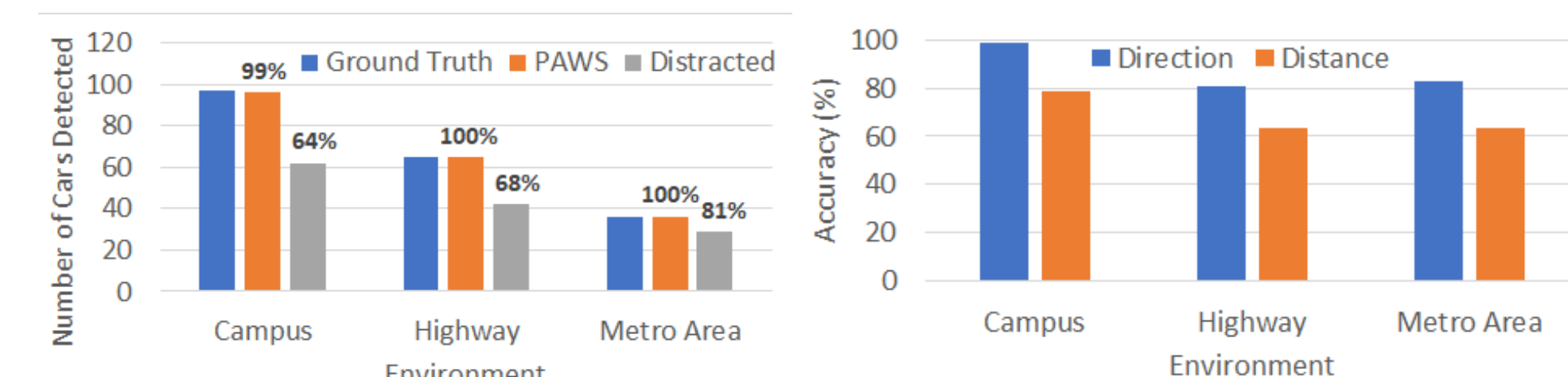
## PAWS Headset + App



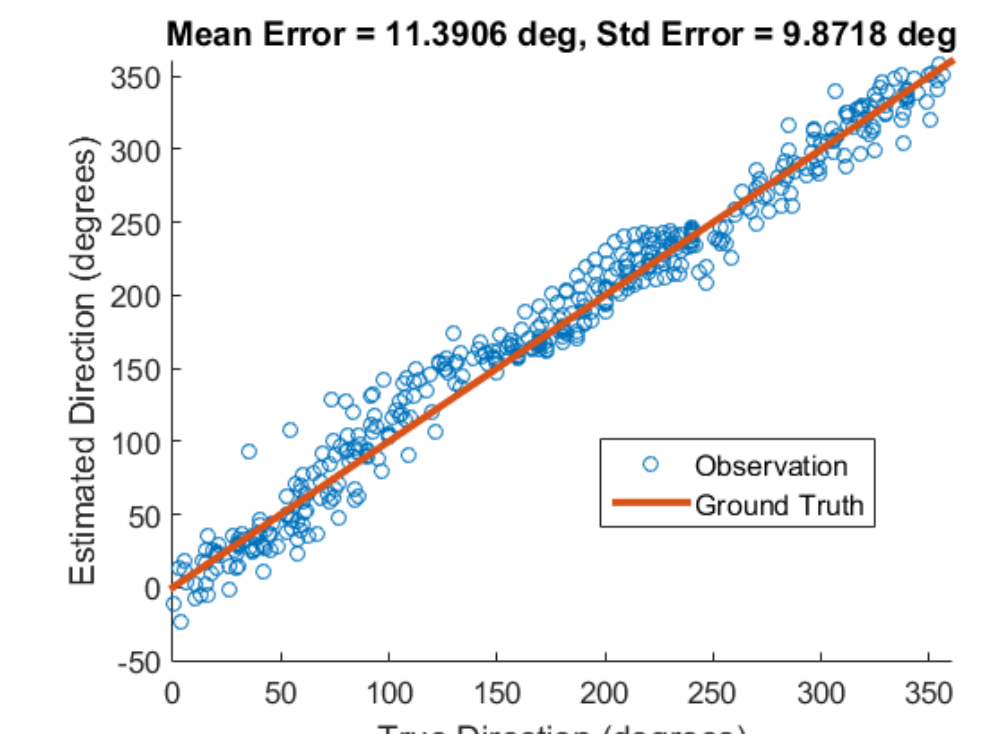
- Custom front-end hardware extracts, computes and transmits features to app
- App alerts users to the presence and location of oncoming cars

## Results

- ~20 hours of operation CR2032 using coin cell batteries
- Close to 100% detection rate and < 10 deg avg direction error



**AvPR Error**



POWER CONSUMPTION AND PRICE BREAKDOWN OF PAWS AND PAWS LOW-ENERGY (PAWS LOW-ENERGY TOTAL IN BOLD)

	Idle [mA]	Active [mA]	Unit Price [US]
MCU (STM32f4)/ASIC	4.37/0	50/0	3.20
BLE Transceiver (nRF52)	0.46	7	6.40
MEMS Mics x 4	0.48 x 4	0.48 x 4	0.40 x 4
Amplifiers x 4	2.34 x 4	2.34 x 4	1.60 x 4
Regulators	0.1/0.6	0.1/0.6	0.50/3.00
<b>Total</b>	<b>16.21/11.84</b>	<b>68.4/18.9</b>	<b>18.10/20.60</b>

## Acknowledgements

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