

# ROSELINE: Enabling Robust, Secure and Efficient Knowledge of Time Across the System Stack



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## Managing Uncertainty in Time

To a software developer,  
time is a simple number....

clock\_gettime()  
→ 1375599686.9

qot() →

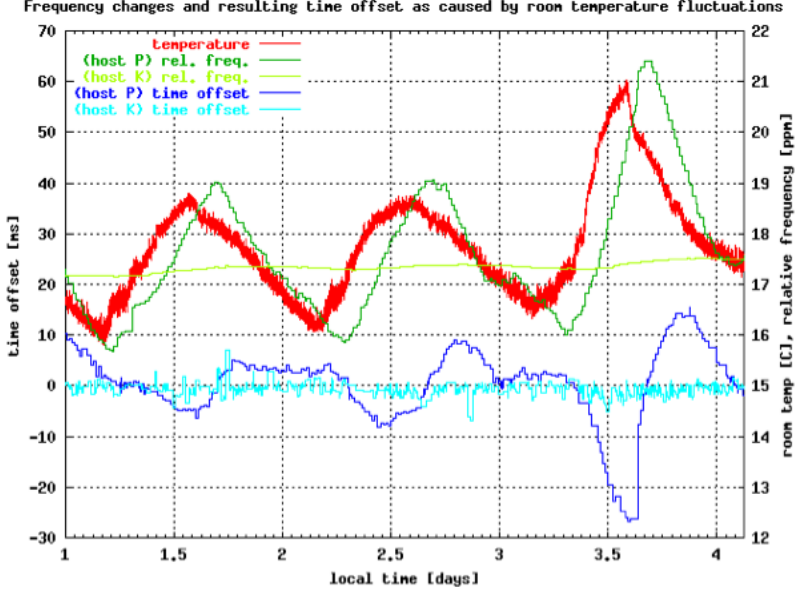


“QoT” to manage  
uncertainty

Wireless Multicore DPM Attacks Virtualization Internet EC2 Variations

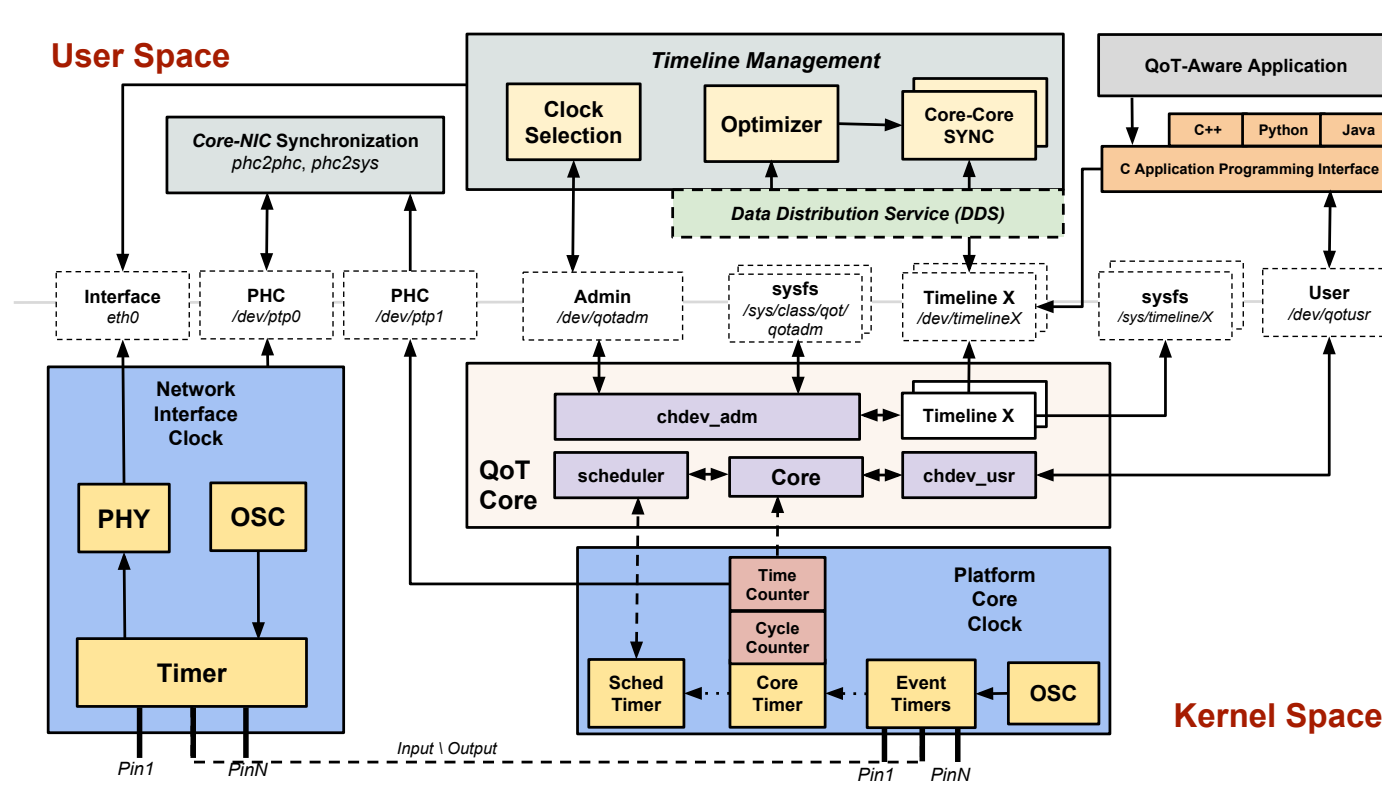
Uncertainty Amplified  
Little Visibility  
No Control

But reality is messy

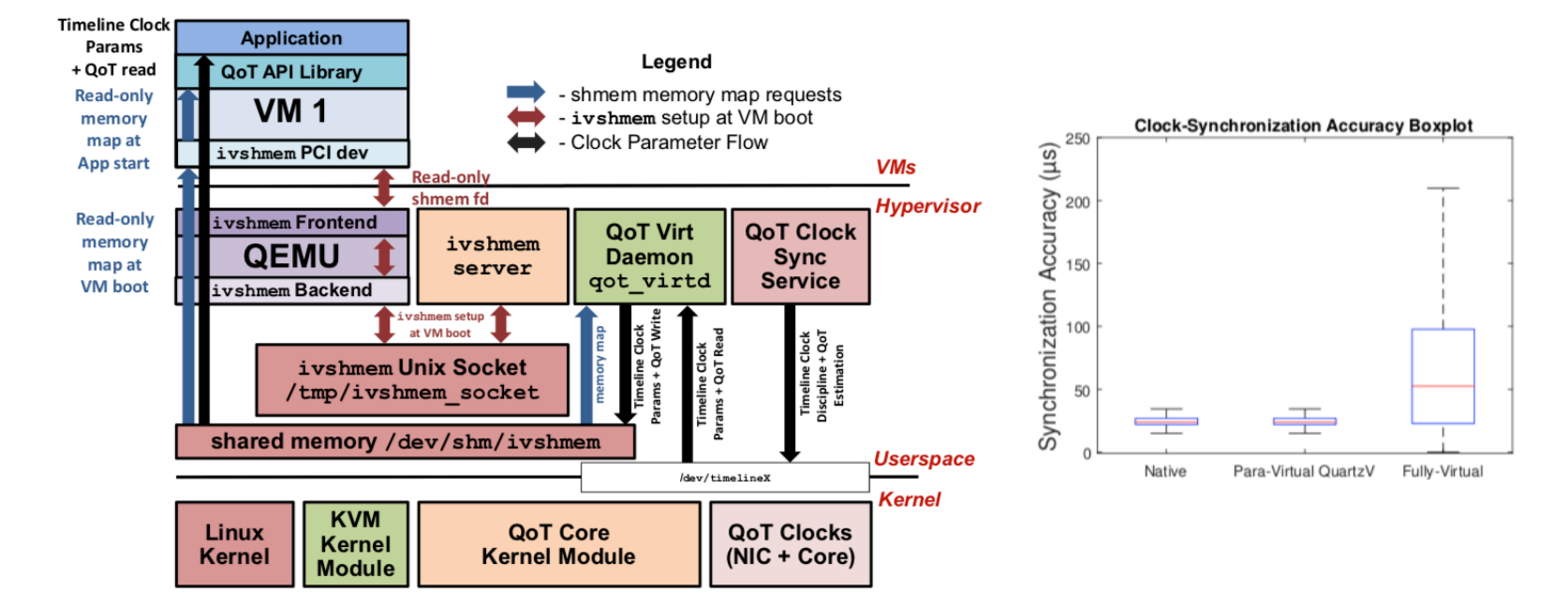


ROSELINE seeks to transform the  
uncertainty in time in a computing  
system into a rich structure called  
“Quality of Time” (QoT) that is  
systematically measured, propagated  
and controlled throughout the system.

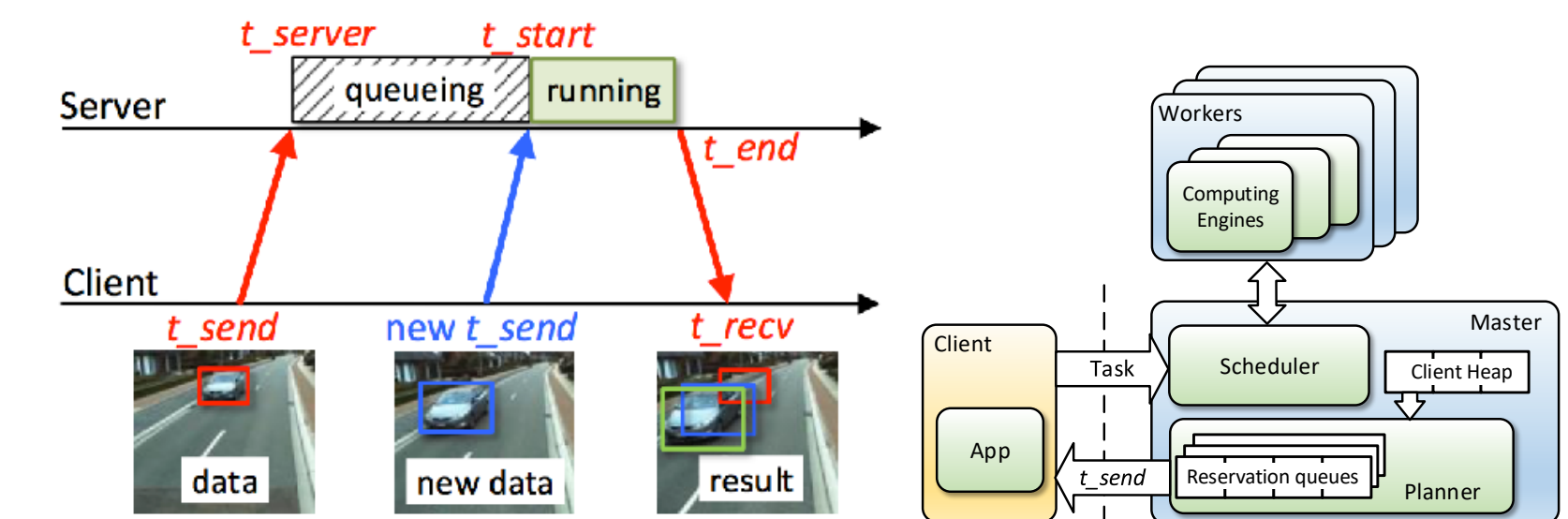
## System Software Mechanisms for QoT



Timeline OS Abstraction



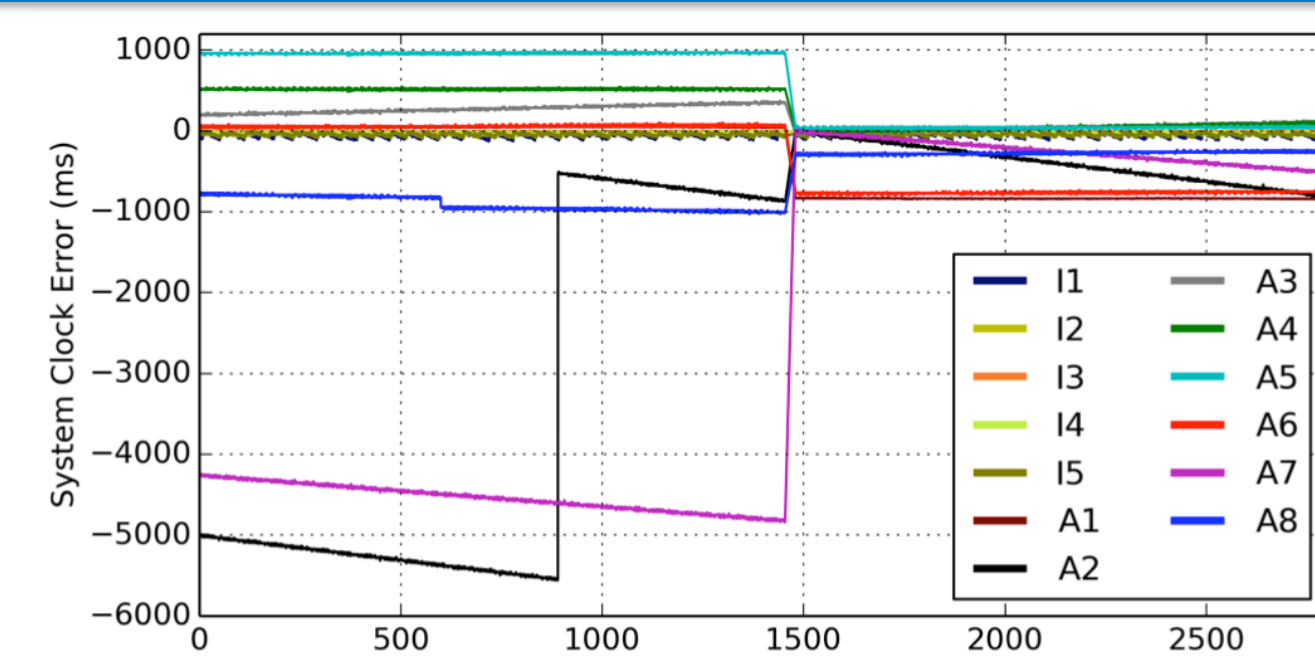
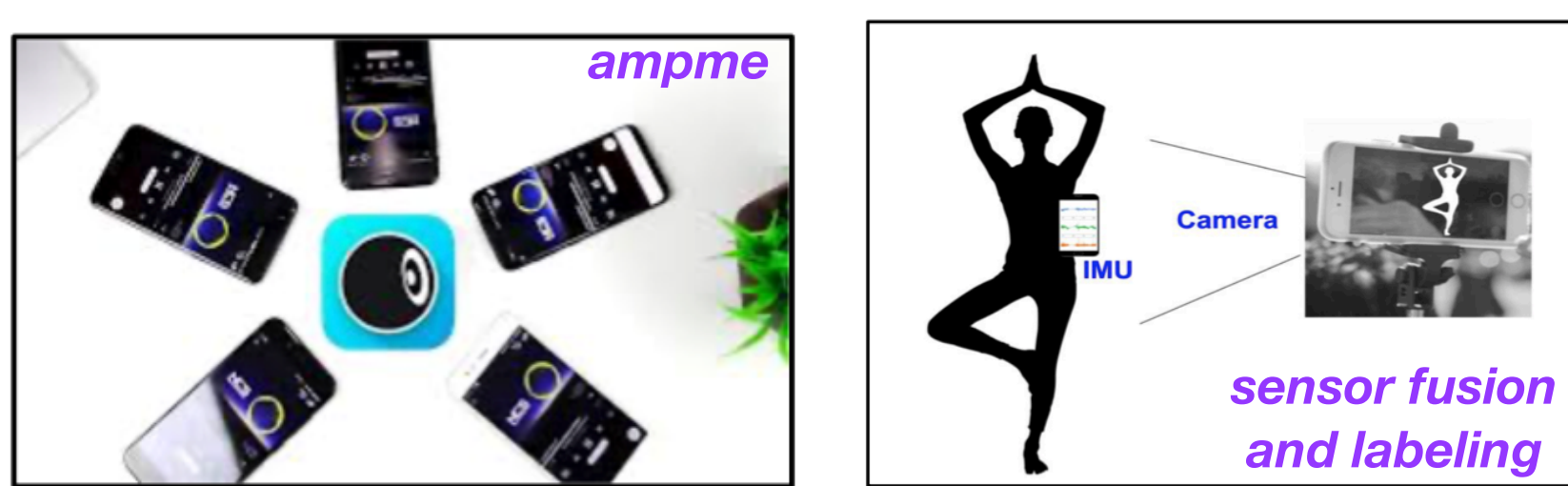
QuartzV: Bringing Quality of Time to Virtual Machines



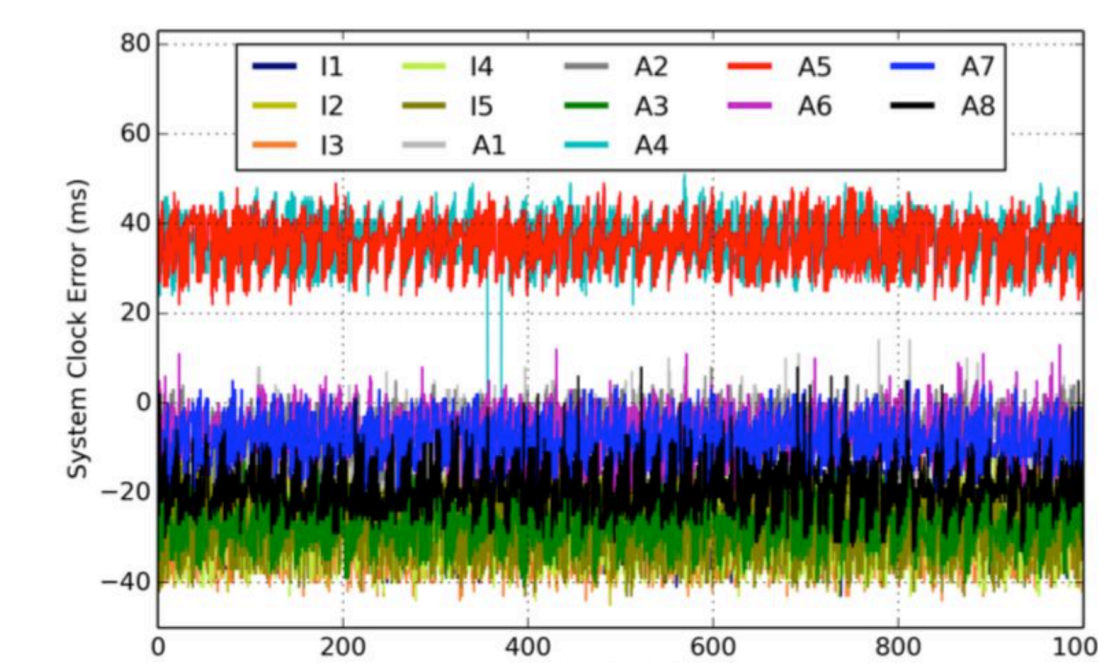
ATOM: Bringing Quality of Time across Edge-Cloud Boundary

## QoT Challenges in Smartphone Apps

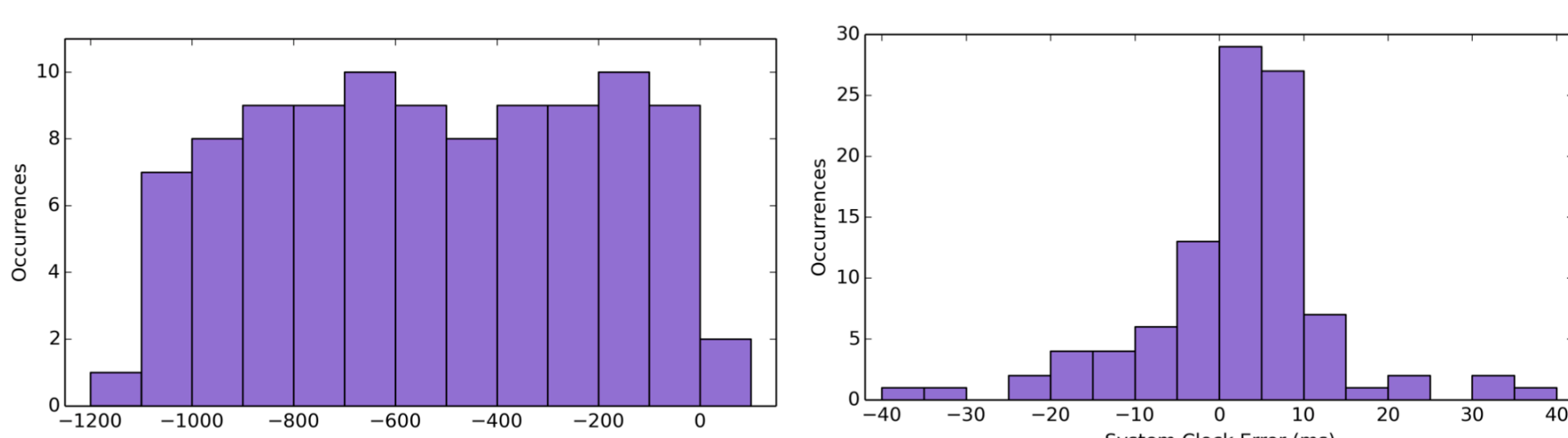
Shared notion of time is needed to fuse data  
across devices or to coordinate actions.



Observed system clock errors before & after a restart event  
at minute 1460 across five iOS devices (I1-I5) and eight  
Android devices (A1-A8) compared to an NTP baseline.

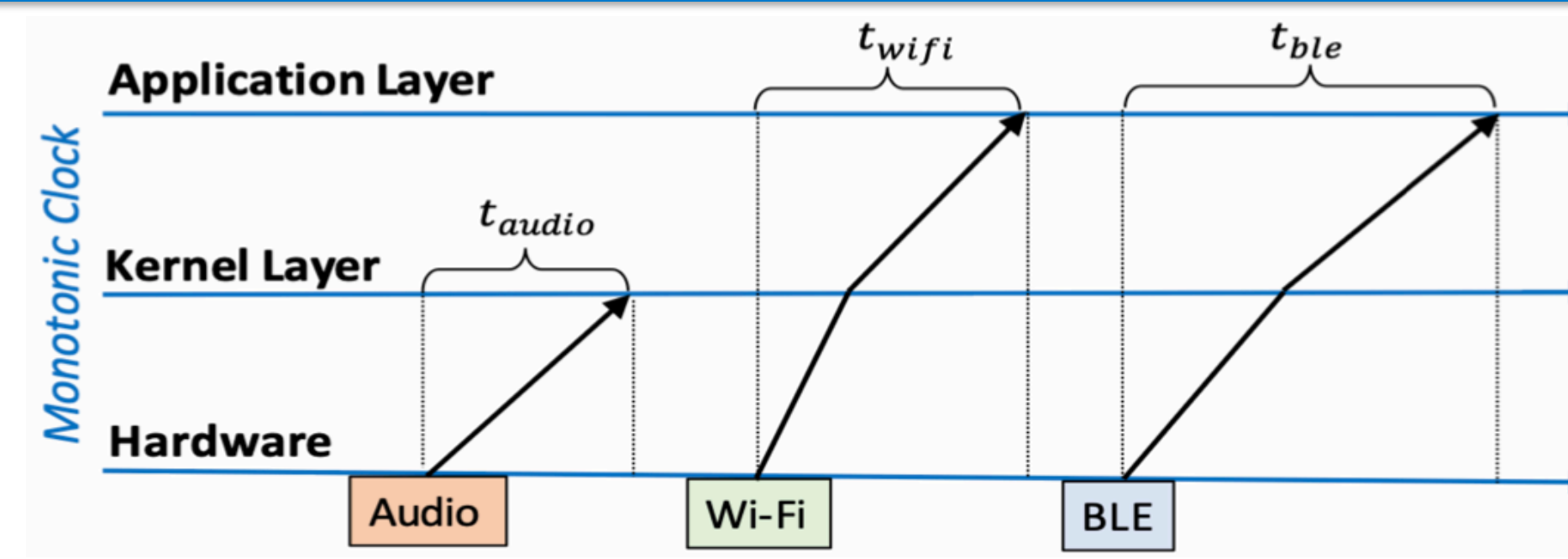


Error improvement with our GoodClock  
App-level Open-source NTP Library.

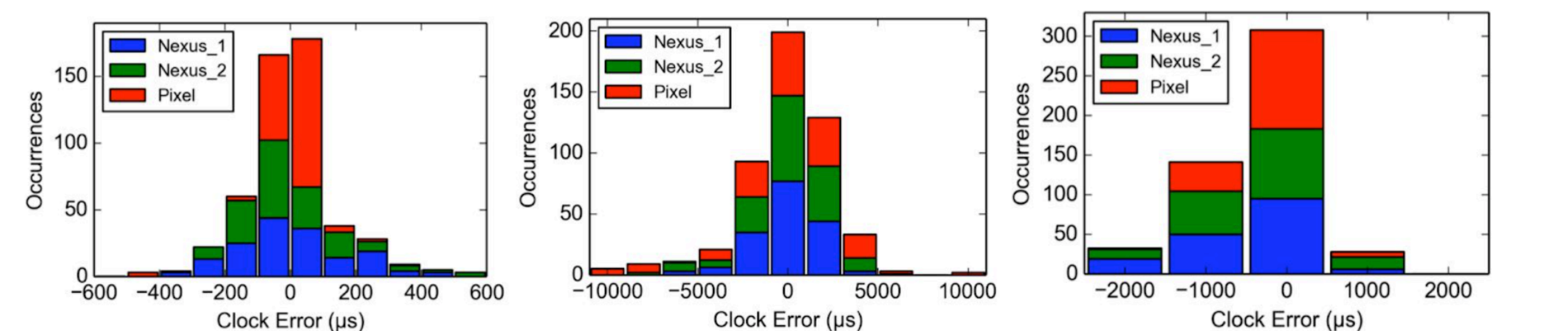


System clock error of an Android Nexus 5X device after  
triggering 100 independent NITZ timing updates (left), and 100  
independent NTP timing updates (right).

## Exploiting Smartphone Peripherals for Better QoT

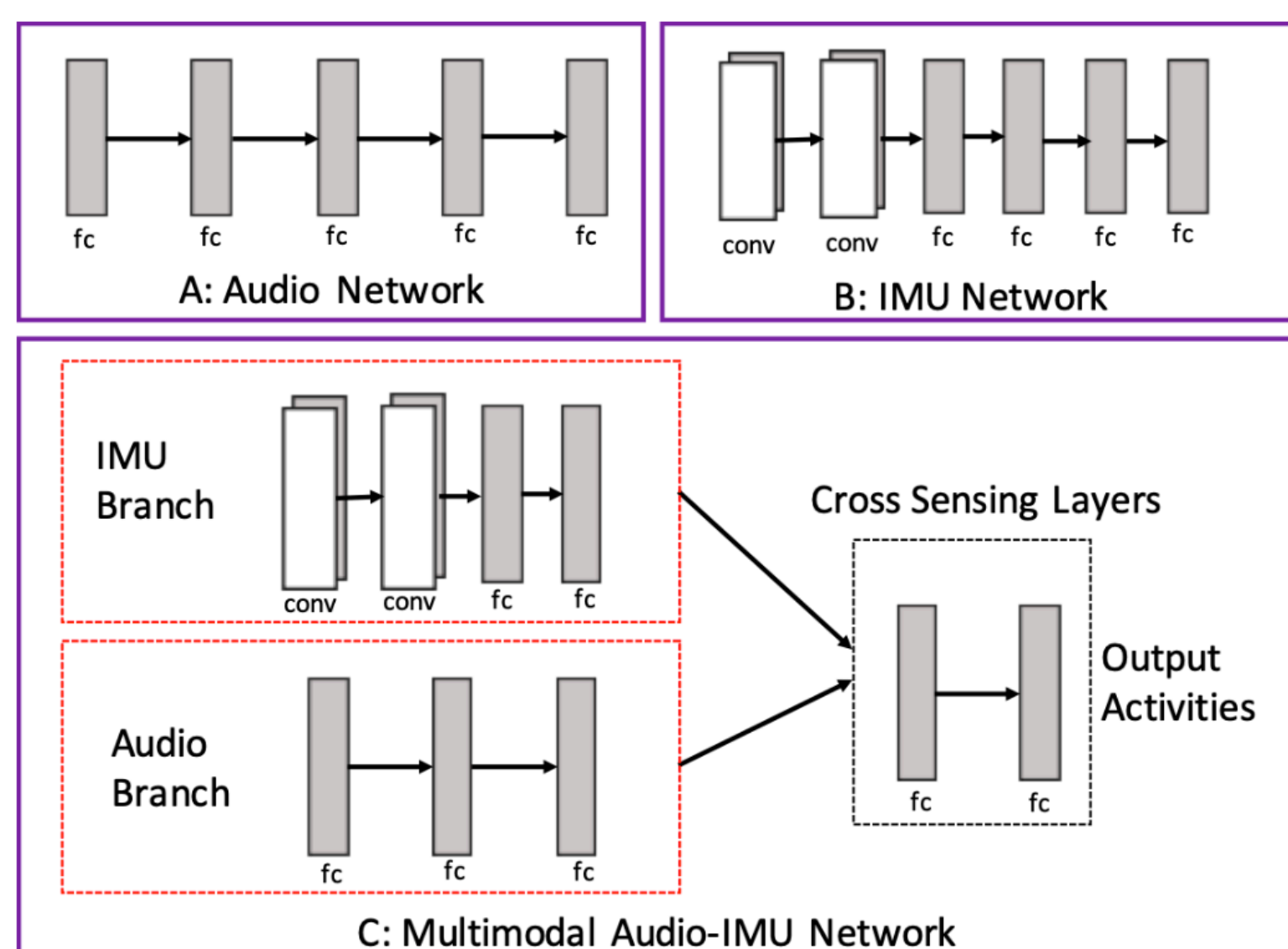


Timestamping events for audio, Wi-Fi and BLE peripherals and Android (left); and Drift in the relative clock offset for  
audio-based sync over a duration of 1 hour with respect to the fourth pixel phone serving as reference (right).

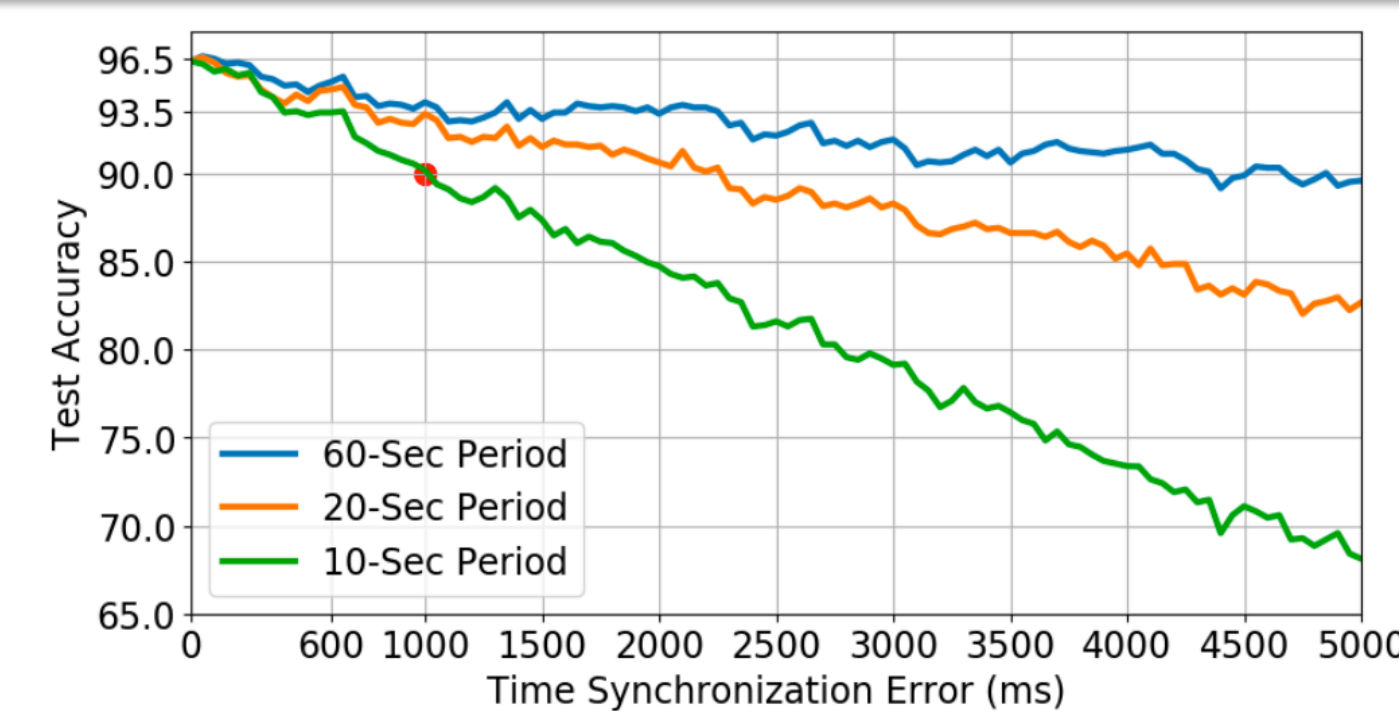


Sync offset variability in three Android smartphones with respect to a fourth reference device for (a) audio,  
(b) BLE, and (c) Wi-Fi implementations. Note: 86% of audio sync attempts fall within  $\pm 200\mu s$ . 85% of BLE  
sync attempts fall within  $\pm 3000\mu s$ . 95% of Wi-Fi sync attempts fall within  $\pm 1000\mu s$ .

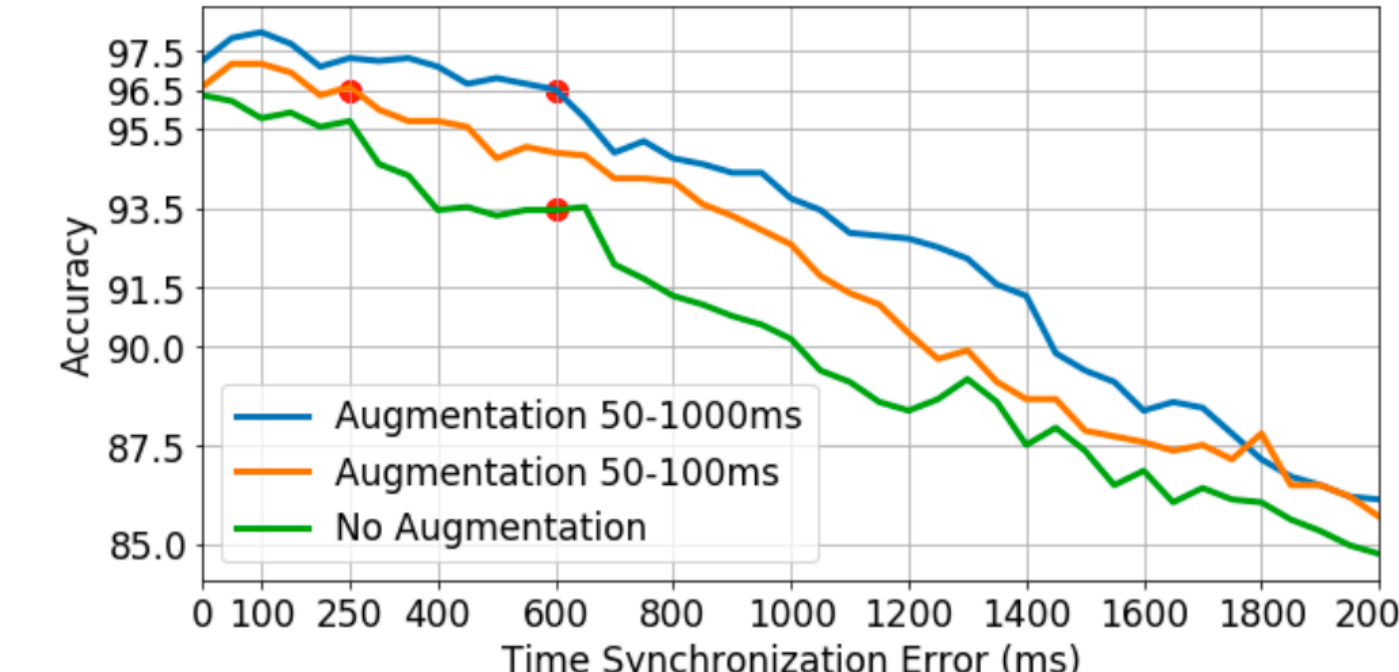
## Mitigawting Adverse QoT in ML-enabled Apps



Multimodal Audio- IMU Deep Neural Network with separate  
branches for Audio and IMU data fused in cross sensing layers.

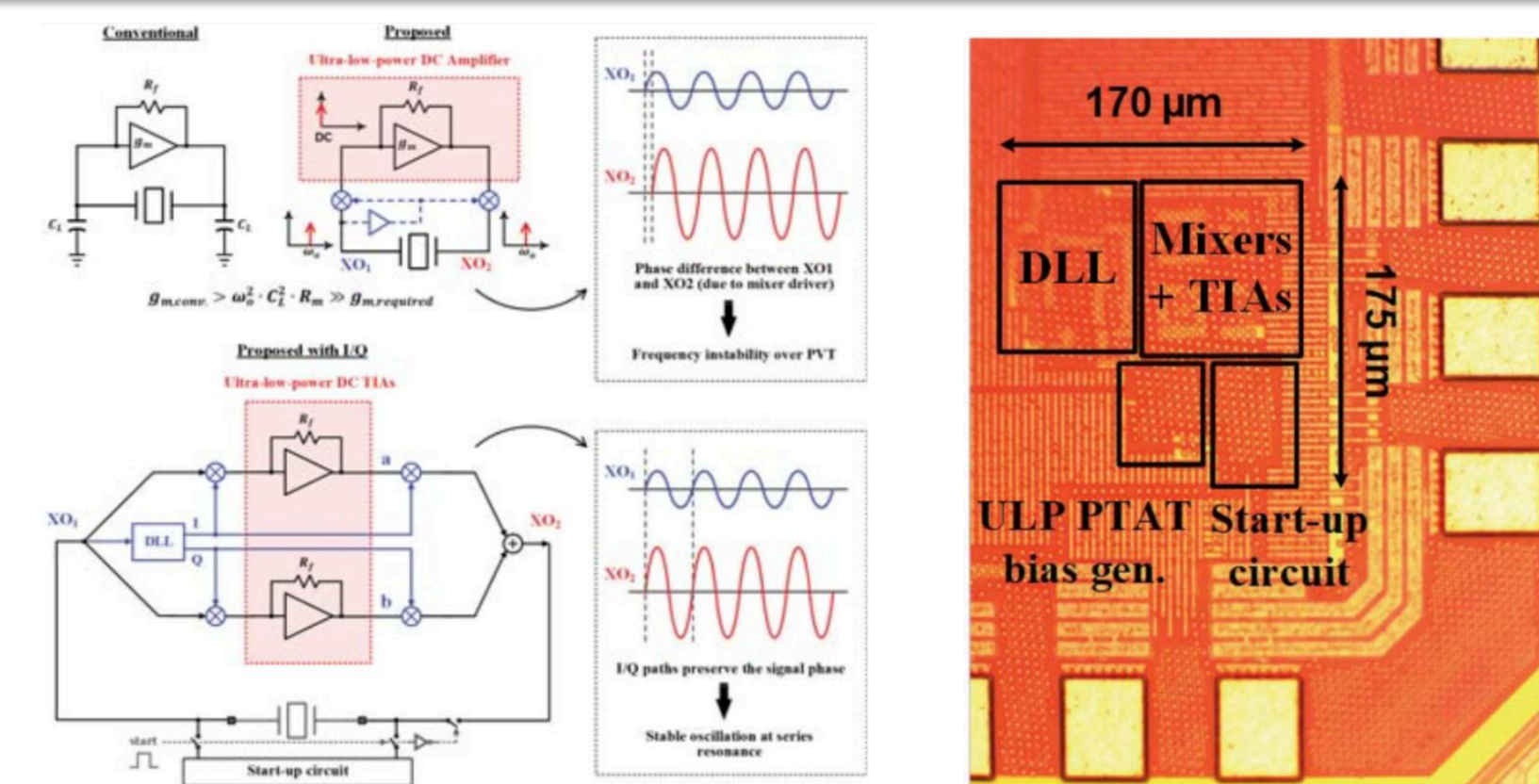


Variation in accuracy of Multimodal Audio-IMU  
network with increased timing errors between  
smartphones collecting audio and IMU data.

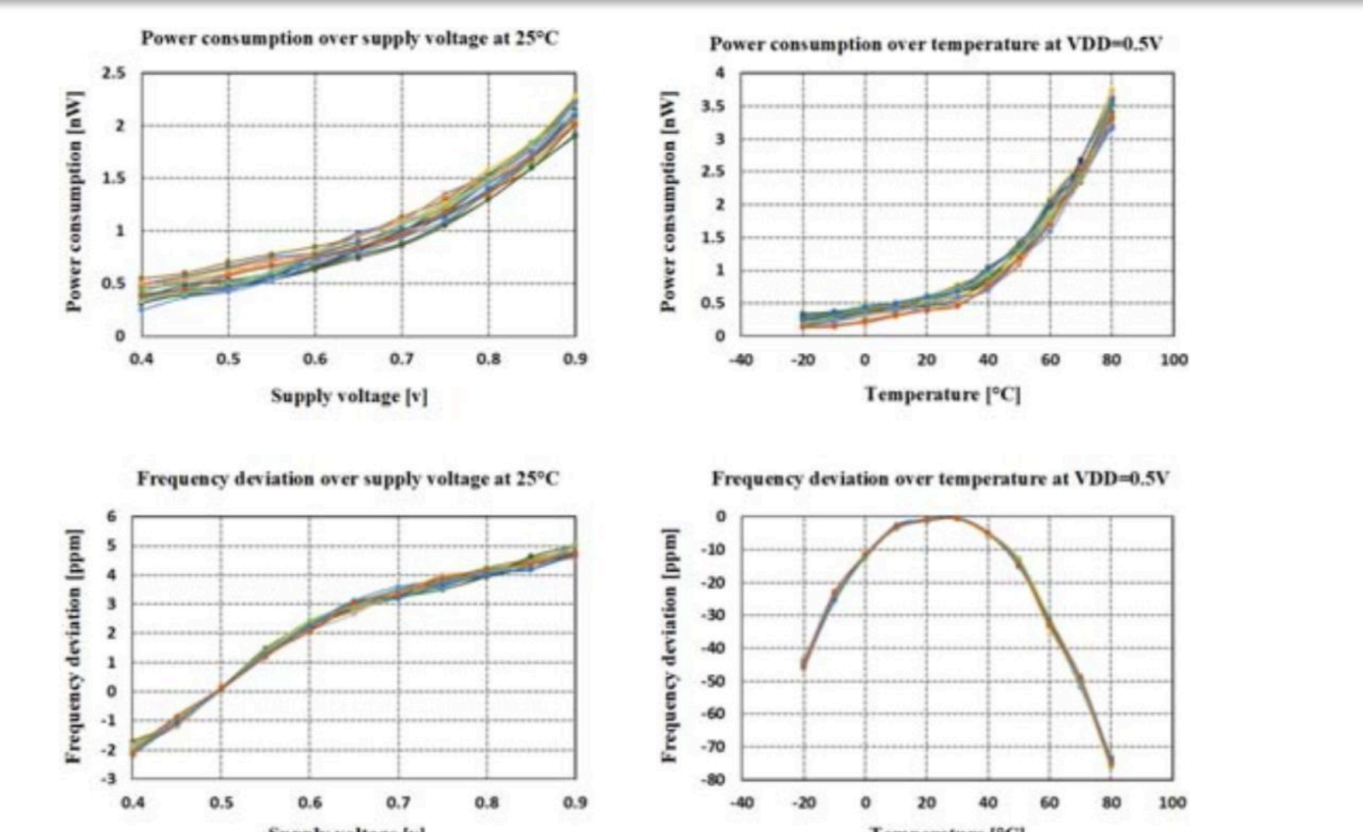
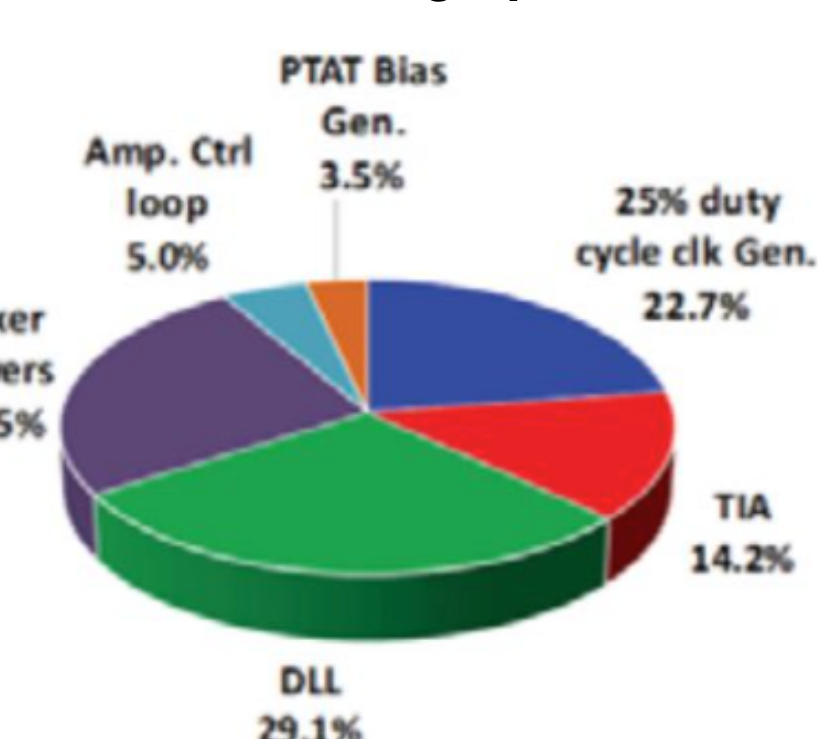


Comparison of Multimodal Audio-IMU Network test  
accuracy with different augmented training datasets.

## Clocks for IoT Energy & Complexity Reduction



Conventional Pierce XO architecture vs our ultra low power  
XO (left); and, a die micrograph of the IC (right)



Power consumption and frequency variations over VDD  
(left) and temperature (right) measured over 20 dies.

	Thiswork*	VLSI'17 [1]	JSSC'16 [2]	ISSCC'14 [3]	JSSC'16 [4]
Technology (nm)	65	55	180	28	180
Area (mm <sup>2</sup> )	0.027	0.16	0.3	0.03	0.062
Supply voltage (V)	0.5	0.5	1.2	0.2	0.3
Power at 25°C (aW)	0.55	1.7	5.58	5	1.5
Temperature stability (ppm)	80	109	133	50	150
Line sensitivity (ppm/V)	13	6.7	30.3	85	7
Alias Deviation (ppb)	14	25	10	10	70
Calibration required	NO	YES	YES	NO	YES

\*Averaged across 20 dies with worst-case power of 0.7aW.