

# Fully On-board Low-Power Localization with Multizone Time-of-Flight Sensors on Nano-UAVs

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# Why On-board Localization on (Nano-)UAVs?

- Nano-UAV
  - Suitable for cluttered indoor environment
- Localization
  - Crucial capability for mobile autonomous systems
  - Foundation for complex tasks
- On-board localization
  - Independent of infrastructure
  - Reduces security risks



# Intelligent Nano-UAV - Challenges



Vehicle class	$\emptyset$ : Weight [cm:Kg]	Power [W]	Onboard Device
standard-size	$\geq 50 : \geq 1$	$\geq 100$	Desktop
micro-size	$\sim 25 : \sim 0.5$	$\sim 50$	Embedded
nano-size	$\sim 10 : \sim 0.01$	$\sim 10$	MCU
pico-size	$\leq 2 : \leq 0.001$	$\leq 0.1$	ULP

- Require lightweight, low-power sensors
- Require low-power but high computational power

# Crazyflie 2.1



27g

+



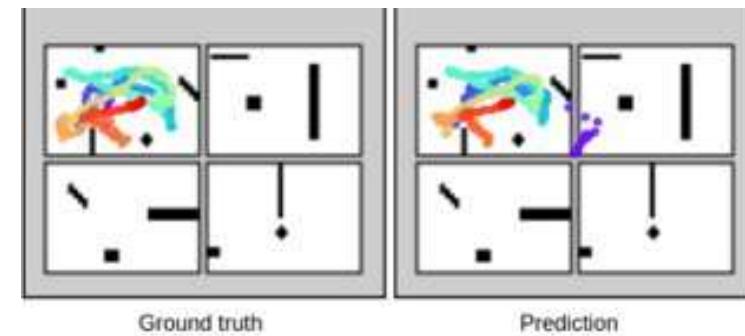
<15g

- Open source
- Modular – extension with own decks possible
- State estimation with extended Kalman filter
- Based on STM32F405 (192kB RAM, 168MHz ARM Cortex M4)

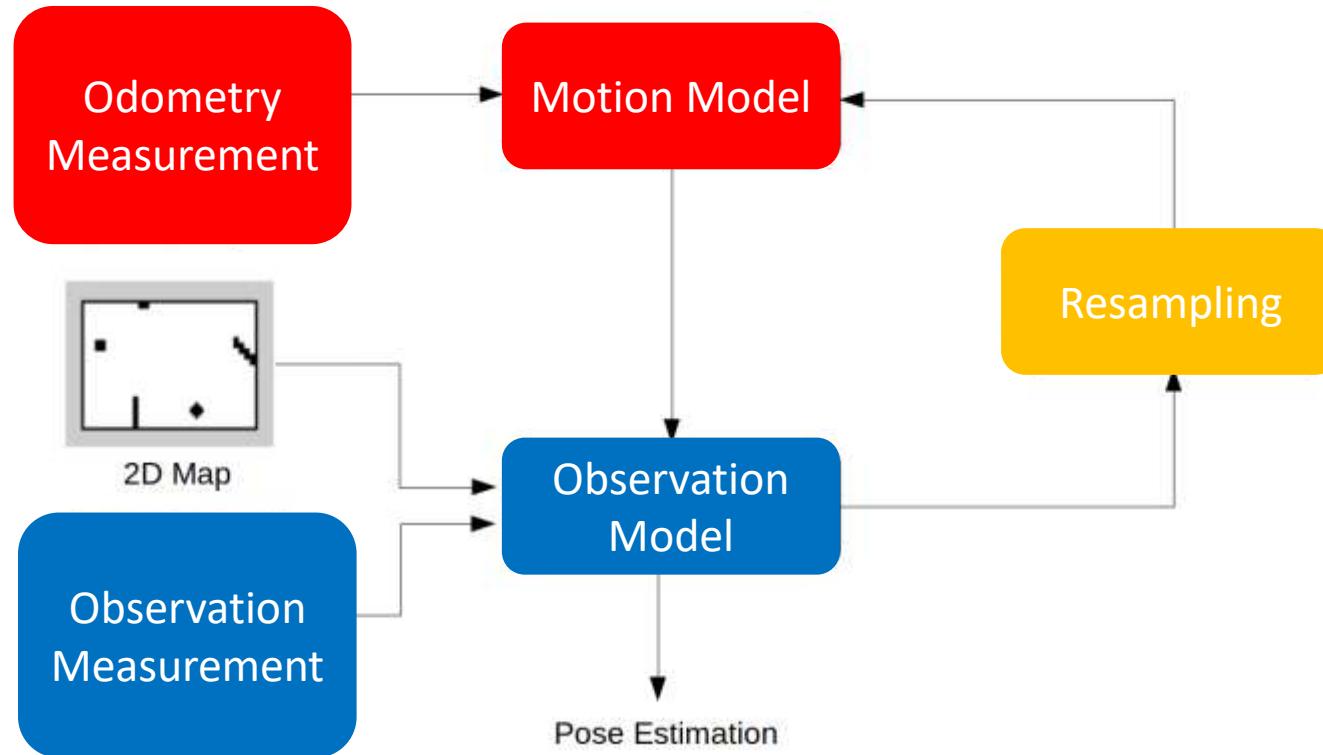
# On-board Localization

## Our contributions:

- On-board localization
- 0.15m accuracy, 95% success rate
- Reduced memory by quantization/f16
- Reduced latency by parallelization (7x)
- Sensing and processing <7% of power consumption



# Monte Carlo Localization

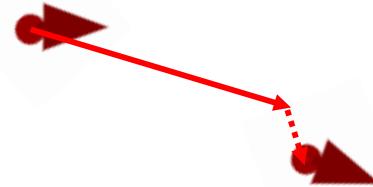


# Monte Carlo Localization

## Odometry Measurement

Extended Kalman Filter  
@100Hz    6-axis IMU:  
                 BMI088  
Optical Flow:  
                 PMW3901  
Downward ToF:  
                 VL53L1

## Motion Model



Particle



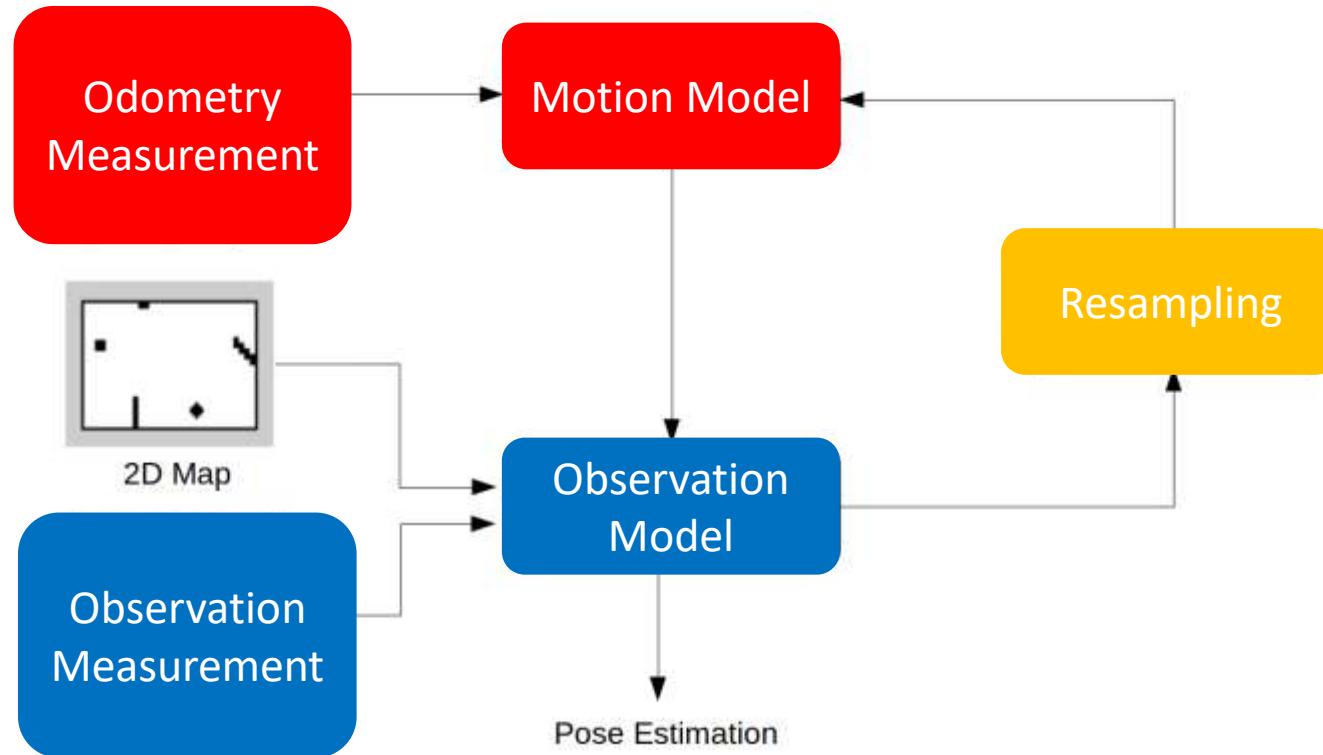
Odometry  
Measurement



Noise

A particle represents a hypothesis  
about the robot's state

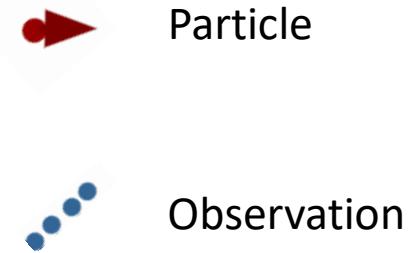
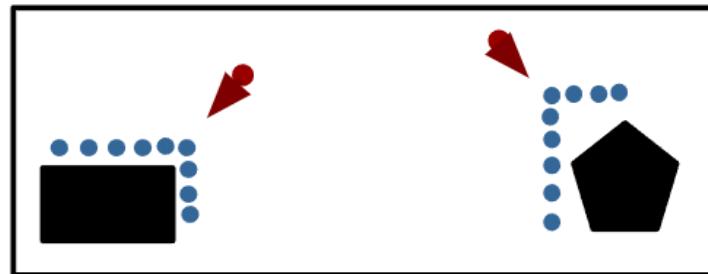
# Monte Carlo Localization



# Monte Carlo Localization

Observation  
Measurement

Multizone ToF:  
VL53L5  
@15Hz



Particle state  
Observation  
Map

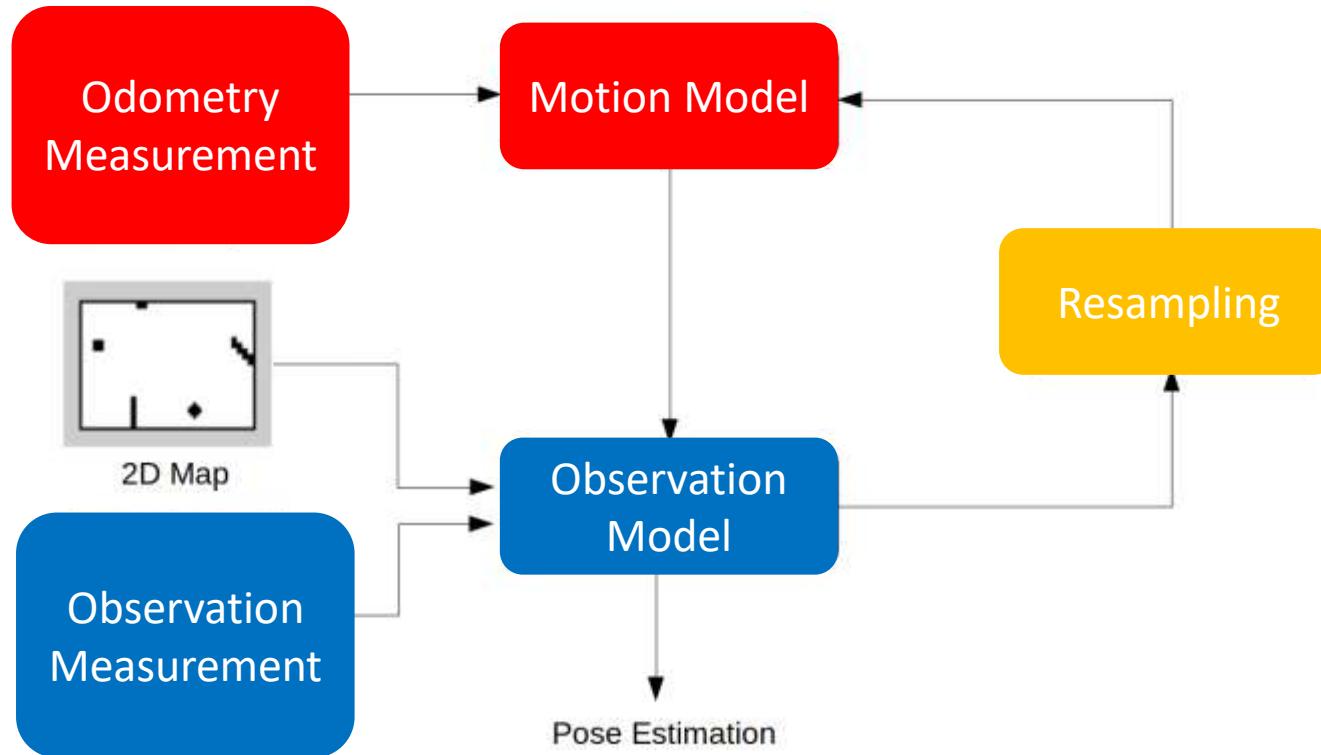
$$p(z_t^k | x_t, m) = \frac{1}{\sqrt{2\pi\sigma_{\text{obs}}^2}} \exp\left(-\frac{\text{EDT}(\hat{z}_t^k)^2}{2\sigma_{\text{obs}}^2}\right)$$

Observation  
Model

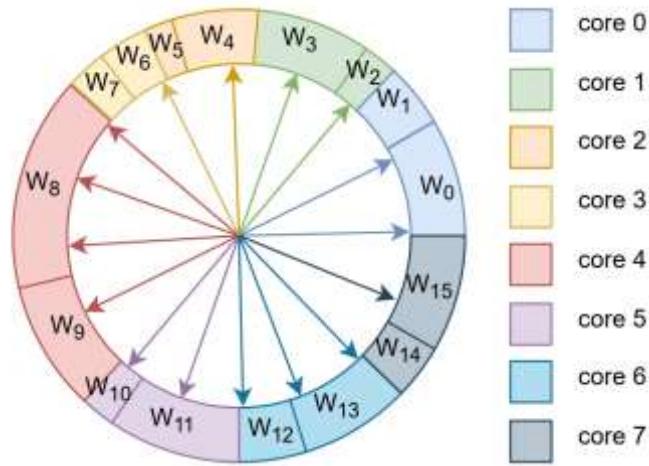
Euclidean Distance Transform  
of observation

Standard deviation of  
observation

# Monte Carlo Localization

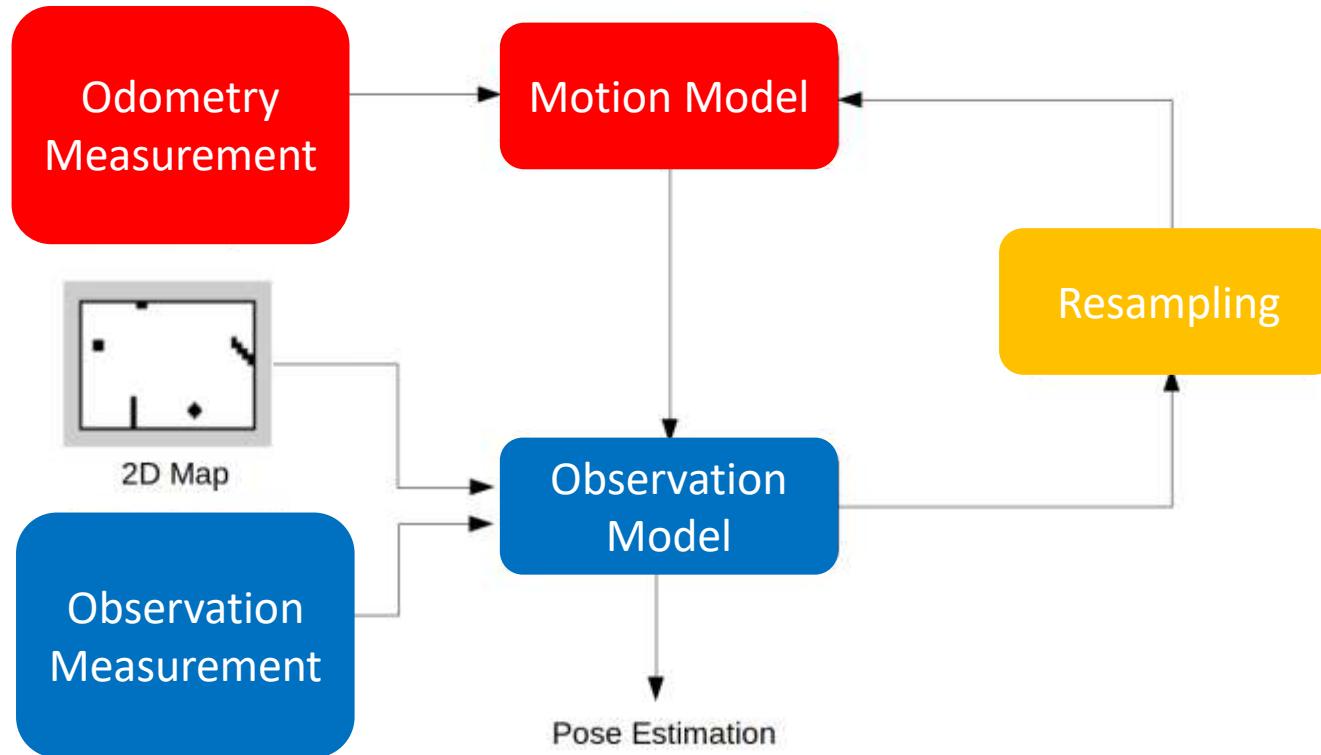


# Monte Carlo Localization



Resampling

# Monte Carlo Localization



# Key Hardware Components

Odometry  
Measurement

Extended Kalman Filter  
@100Hz



6-axis IMU:

BMI088

Optical Flow:

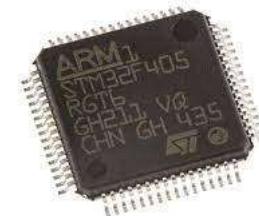
PMW3901

Downward ToF:

VL53L1

Flight  
Controller

ARM Cortex M4:  
STM32F405



Observation  
Measurement

Multizone ToF:

VL53L5

@15Hz

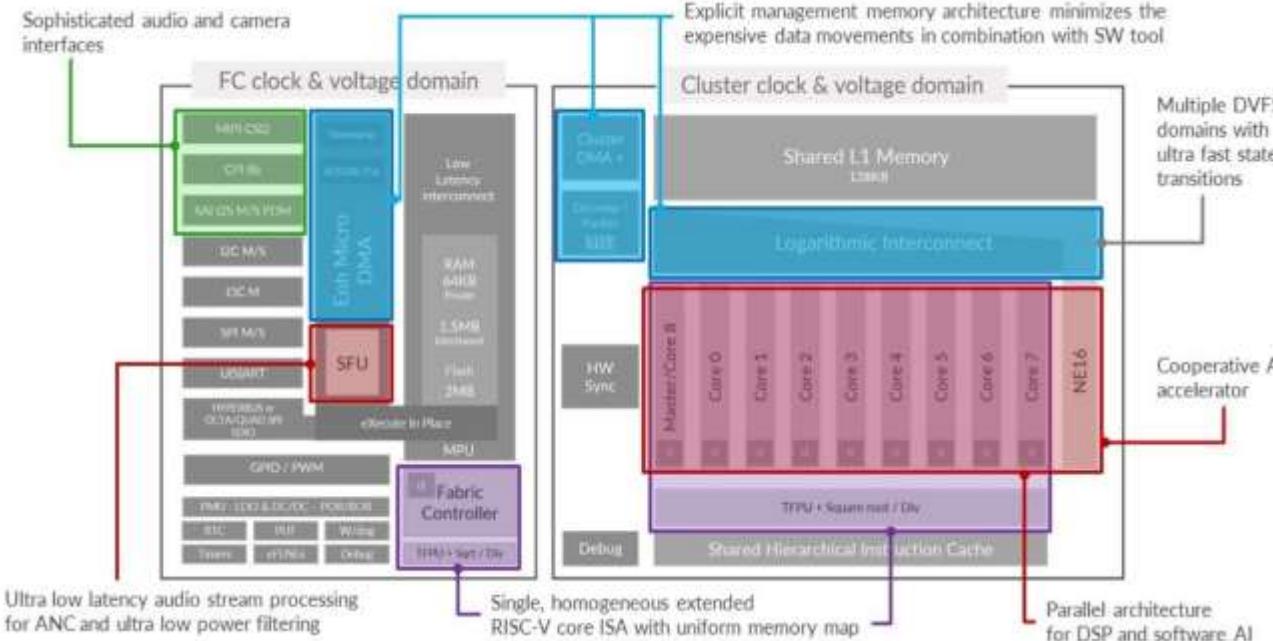


Monte Carlo  
Localization

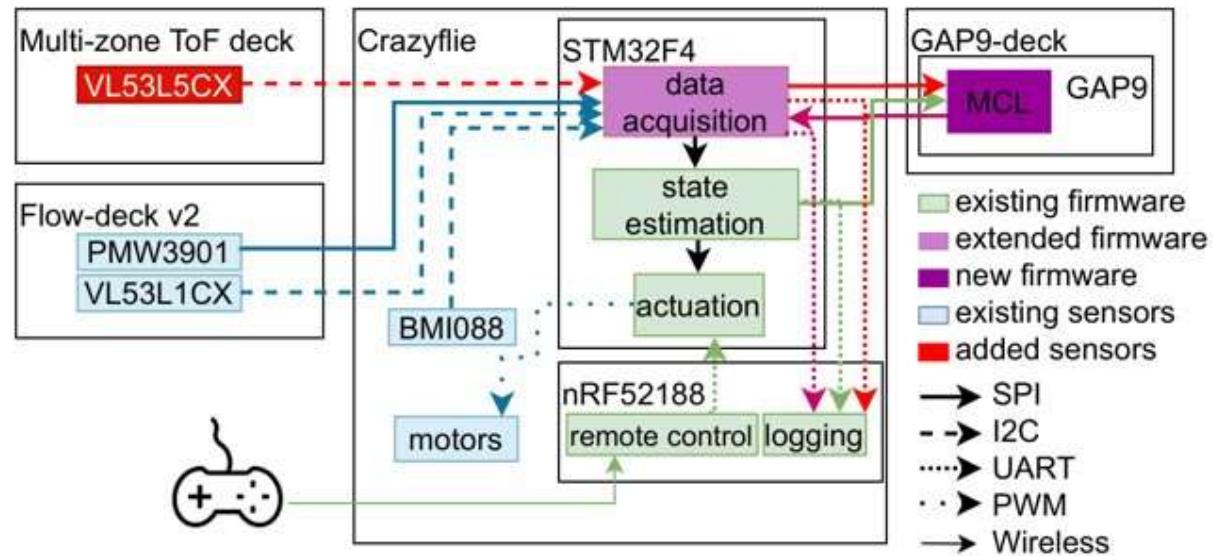
Parallel Ultra Low-  
Power SoC:  
GAP9



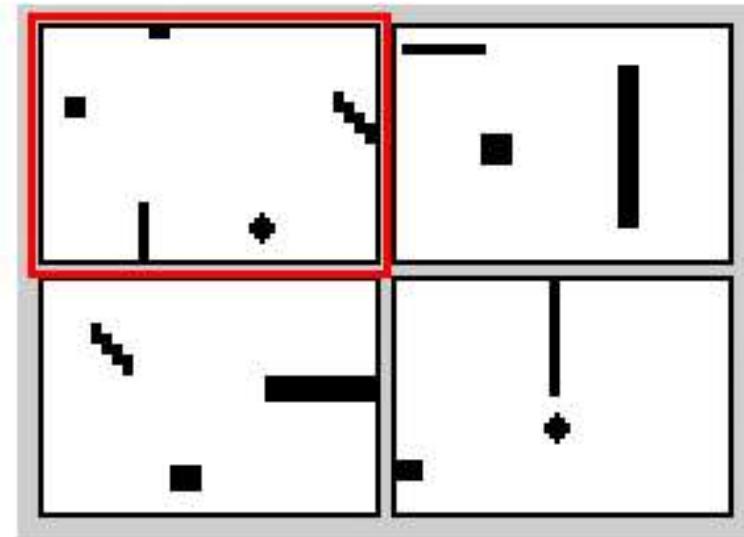
# GAP9 Architecture



# System Architecture



# Experimental Evaluation - Setup

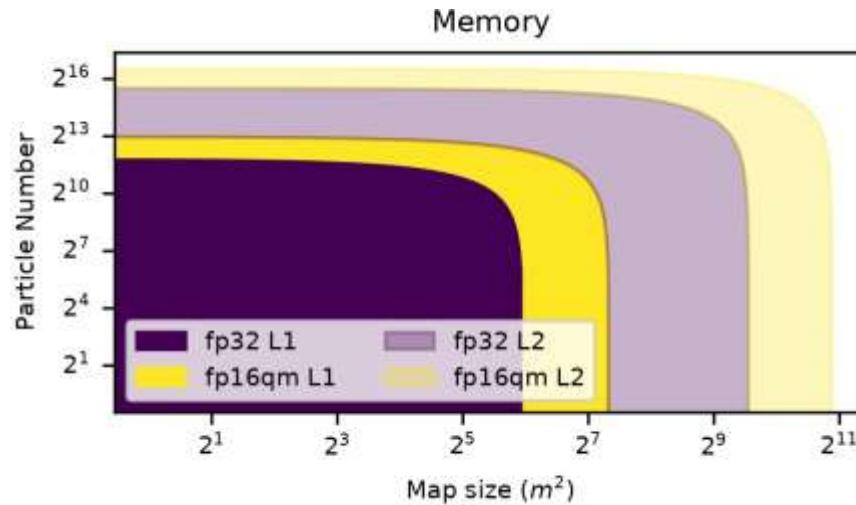


# MCL Parameters

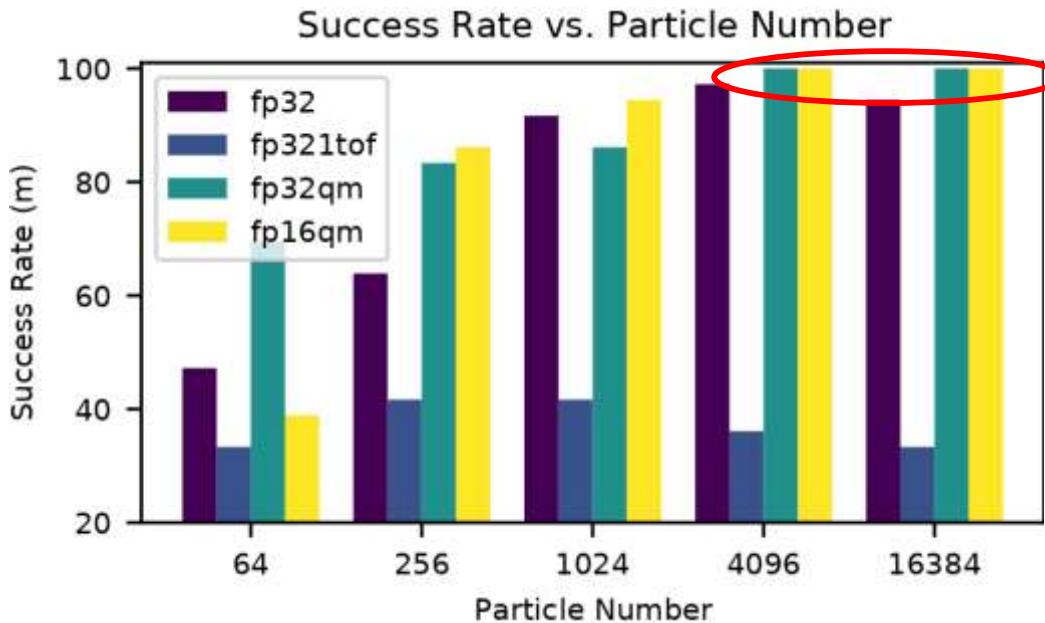
- 1 (front) vs 2 (front and rear) ToF sensors
- #particles (64, 256, 1024, 4096, 16384)
- full precision vs quantized map, lower accuracy particles

each particle:

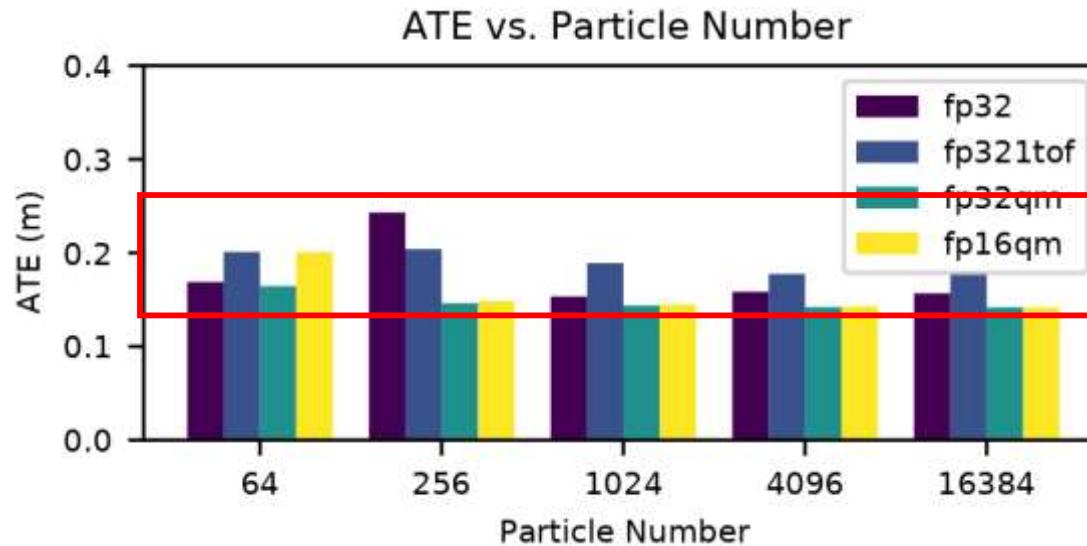
- x position
- y position
- yaw angle
- weight



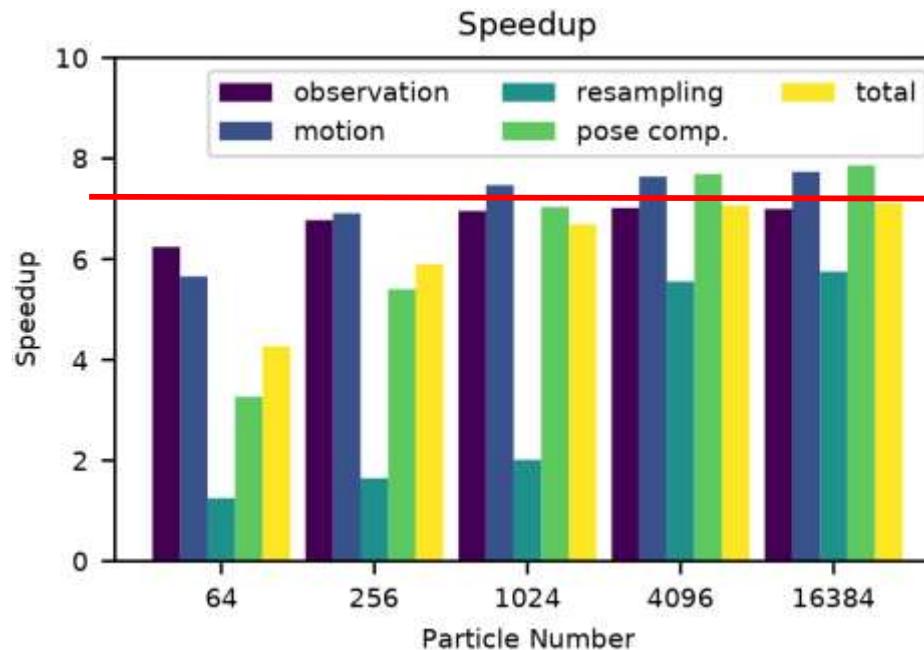
# Experimental Evaluation – Robust Localization



# Experimental Evaluation – Robust Localization



# Experimental Evaluation – Real-time



# Experimental Evaluation – Real-time/Power

	Avg. Power consumption	Execution time
GAP9@400MHz/1,024 particles	61mW	1.901ms
GAP9@12MHz/1,024 particles	13mW	59.898ms
GAP9@400MHz/16,384 particles*	61mW	30.880ms
GAP9@200MHz/16,384 particles*	38mW	61.524ms

\*particles stored in L2

Measurement update: 15Hz

# Conclusion

## Our contributions:

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- 0.15m accuracy, 95% success rate
- Reduced memory by quantization/f16
- Reduced latency by parallelization (7x)
- Sensing and processing <7% of power consumption
- Open source

