

# Step Counter

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CSCE 462, Fall 2020



# Materials

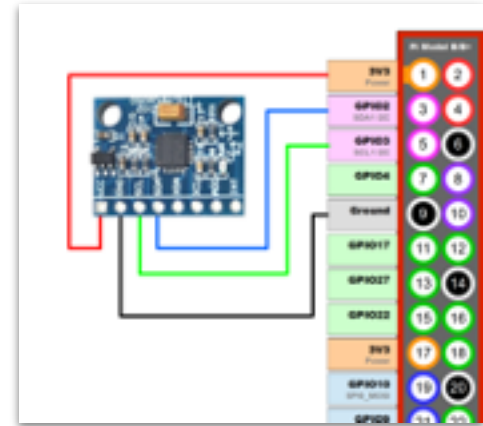
- Raspberry Pi 3 B+ & power adapter
- [MPU-6050](#)
- Jumper wires
- Soldering Iron & solder
- Sweat band or other arm band
- Breadboard & ribbon cable (optional)





# MPU-6050

- 3 axis accelerometer and gyroscope
- Uses I2C - power and communication with processor
- 8 pins
  - VDD
  - GND
  - SCL
  - SDA
  - XCL
  - XDA
  - ADO
  - INT
- You must solder the pins to the chip



Connections:

MPU-6050	Raspberry Pi 3 B+
VDD	3V3 (pin 1)
GND	GND (pin 9)
SCL	SCL1 (pin 5)
SDA	SDA1 (pin 3)

# Modeling of "steps"

- Each step is a motion made by a person wearing the sensor
- Numerous ways to characterize different types of "steps"
- Logging, walking Q: how many different states it can have?

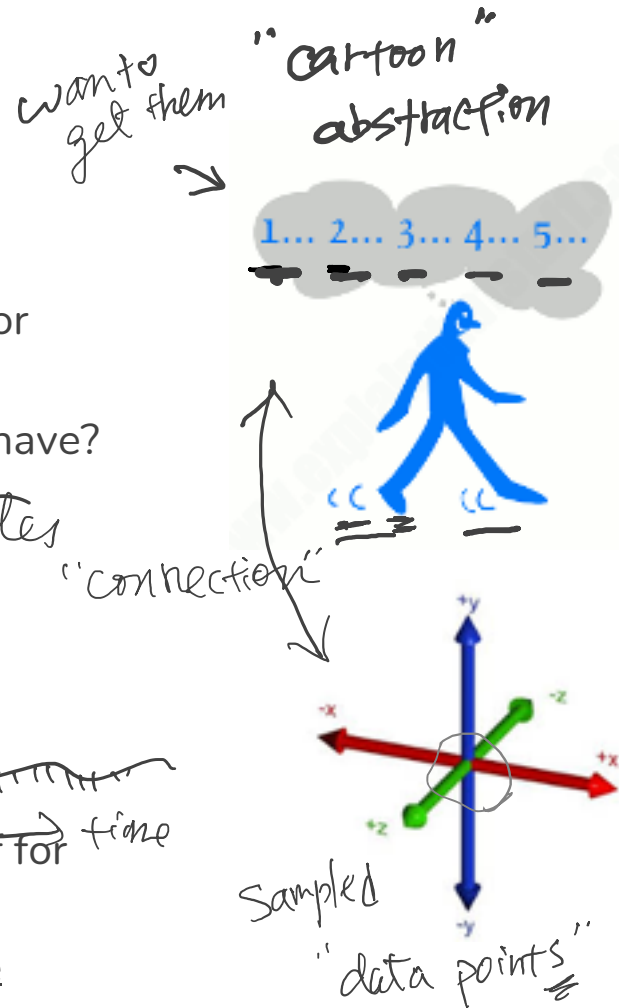
- Approximation is needed to measure steps

- Steps:

- Changes in readings on the x, y, and z axes
- Both + & - readings represent opposite motion directions

- Measurement:

- "Stationary positions": the body rebalances itself for the next step, sensor readings very small
- "Transitive positions": the motion of a step, large sensor readings



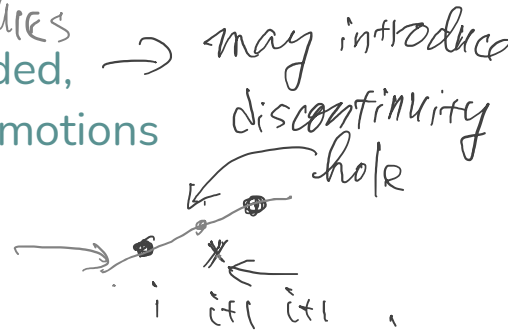
# Data model and processing algorithms

- High sampling rate: 10ms interval, or 100 samples/sec.
  - minimize loss of motion data, but noise is mixed into real data  
=> noise filtering is required
- Differentiation of human stationary vs. transitive motion data
  - Transitive positions: large readings for a period
  - Stationary positions: nearly no readings for a period
- Algorithms

- Filtering rule => small motion data can be discarded, → may introduce discontinuity hole
- Smoothing of filtered data to reconstruct human motions
- Measure steps by counting stationary positions

estimate  
missed  
real values

"real"  
values





# Algorithms

## Raw Data

- x, y, z
- Directly from the accelerometer
- $S = |x| + |y| + |z|$
- S is used to measure the level of motion along all directions

\* So, this method cannot differentiate the motion direction.

## Noise Filtering

- $S_{\max} = \max[S(n-9), S(n)]$
- Only picks the largest motion components collected within the given sample/window length
- Filters S to eliminate minority motion components aka "noise"

"Generalization"

## Smoothing

- Moving Average =  $\text{average}[S_{\max}(n), S_{\max}(n+9)]$
- Smoothing effect
- Reduces fluctuation

"window size"  
what kind of window

## Is it a step?

- if  $S(n-1) < \text{Moving Average}(n)$  and  $S(n) \geq \text{Moving Average}(n)$
- If  $\text{Moving Average}(n) > \text{threshold}$
- Find the peaks of the graph and compare to the threshold to determine if peak counts as a step



# Implementation

- Hardware
  - Sensor placement (chest, foot, wrist, etc...)
  - Wiring, prototype board, power
  - Communication between processor and sensor
- Software
  - Language
  - Library
  - etc...

optimal working ranges

# Experiment Variations & calibration

robustness

user factor

which type of users  
"speak a few known words"

- Considering factors
  - Human behaviors
  - The type(s) of steps that we aim to measure
    - Their dynamic ranges and thus for algorithm calibration
  - Sensor placement - proxy of the actual steps
- Algorithm calibration (hint: every adjustable parameter)
  - Sampling rate
  - Filtering rule
  - Transitive positions
  - Stationary positions
  - ...



train the  
actual working  
alg.





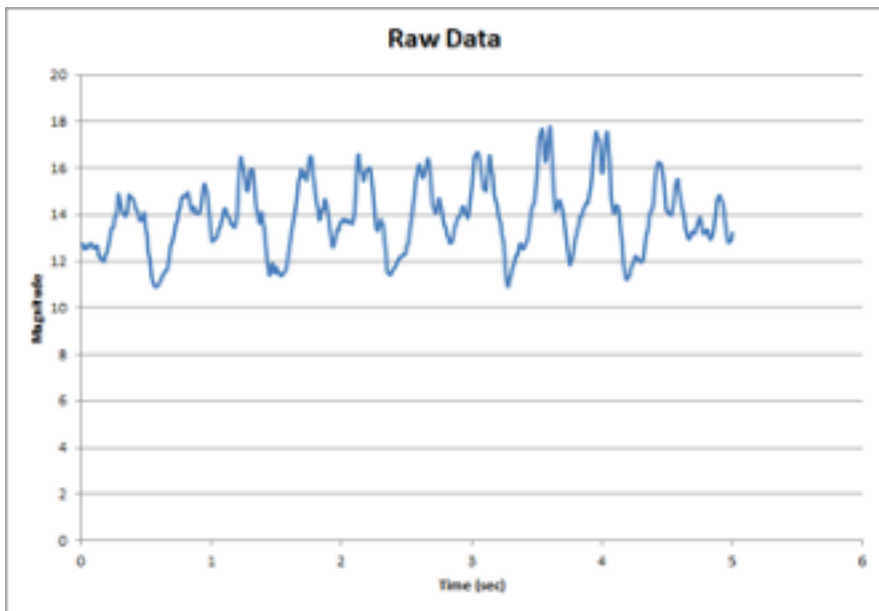
# Experiment Variations & calibration

- One of the most critical step to match the algorithm with human behaviors
- Experiment 1: BASELINE
  - Threshold = 15
  - Sample Length = 10 ms
- Experiment 2: Vary Threshold
  - 2a
    - Threshold = 9
    - Sample Length = 10 ms
  - 2b
    - Threshold = 21
    - Sample Length = 10 ms
- Experiment 3: Vary Sample Length
  - 3a
    - Threshold = 15
    - Sample Length = 1 ms
  - 3b
    - Threshold = 15
    - Sample Length = 100 ms



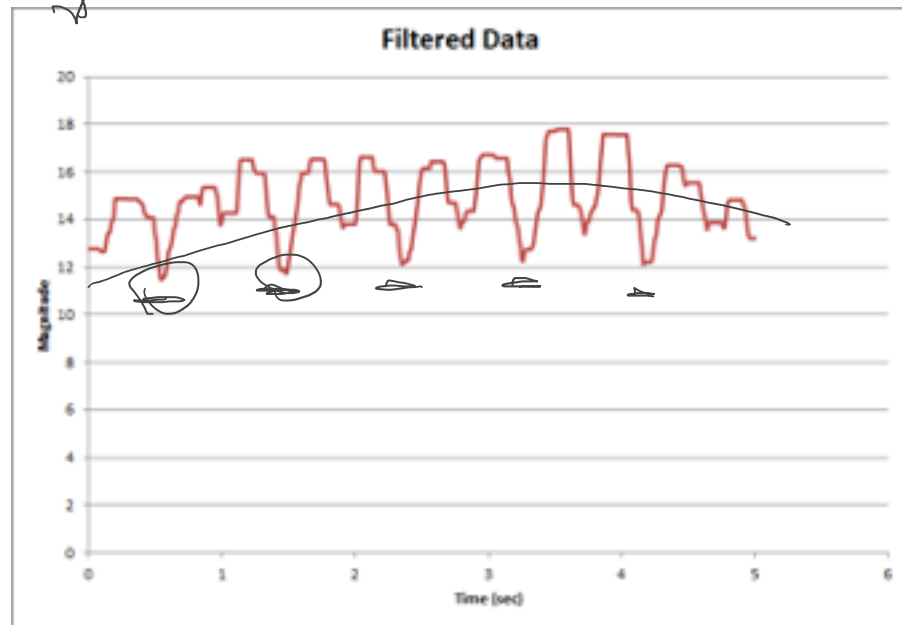
# Raw data filtering effect

Raw data graph



how to make this useable *key challenge*

Filtered graph

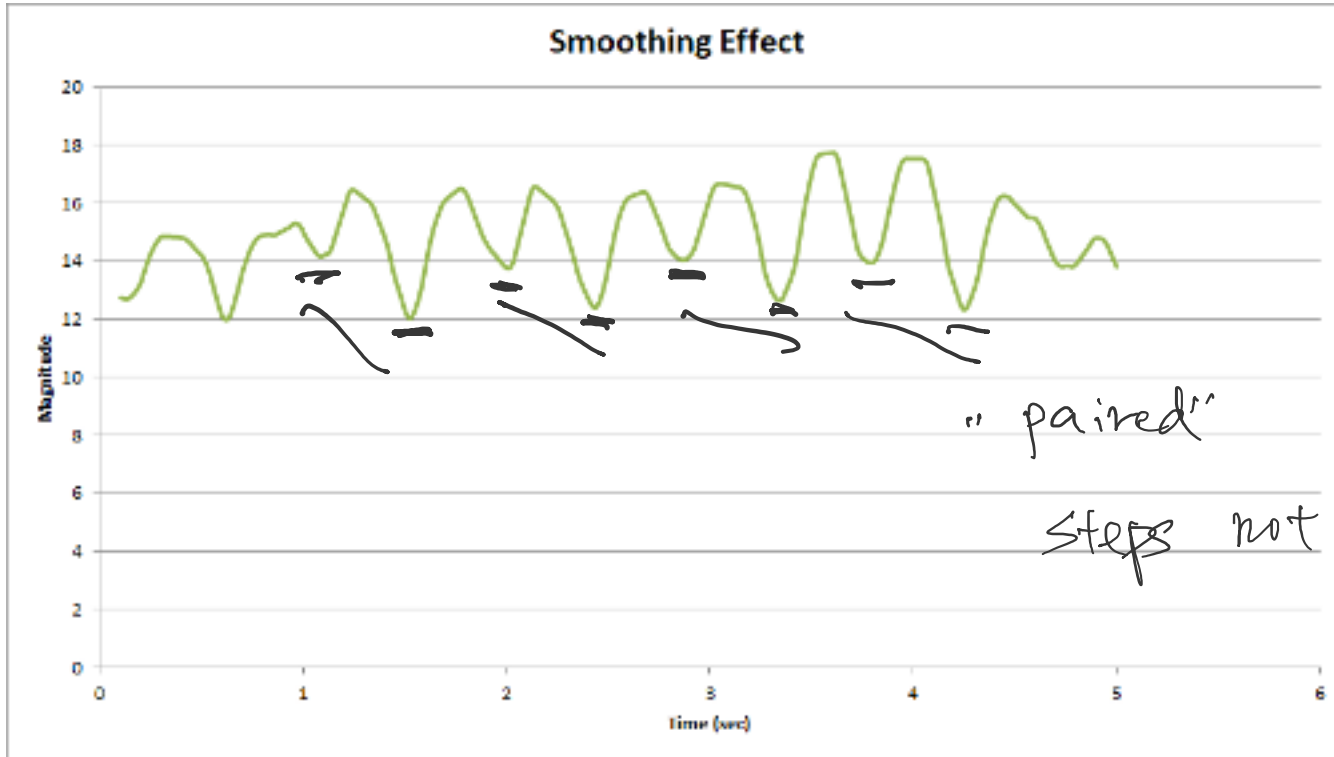


# Smoothing effect

~~interpret~~

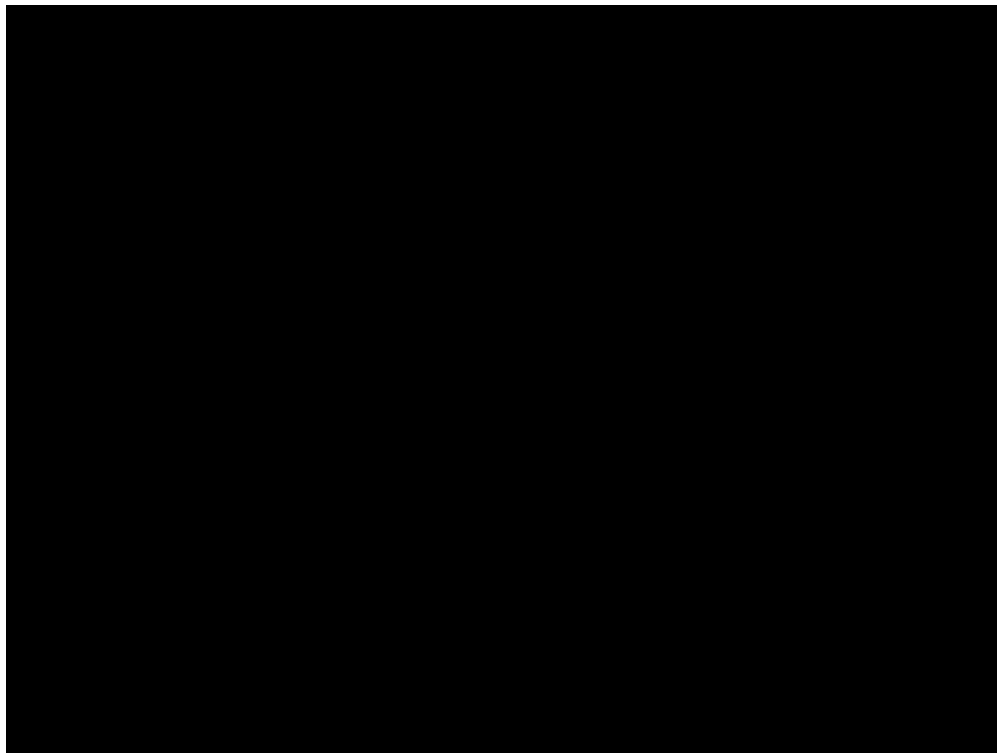
interpretation

→ real world actions

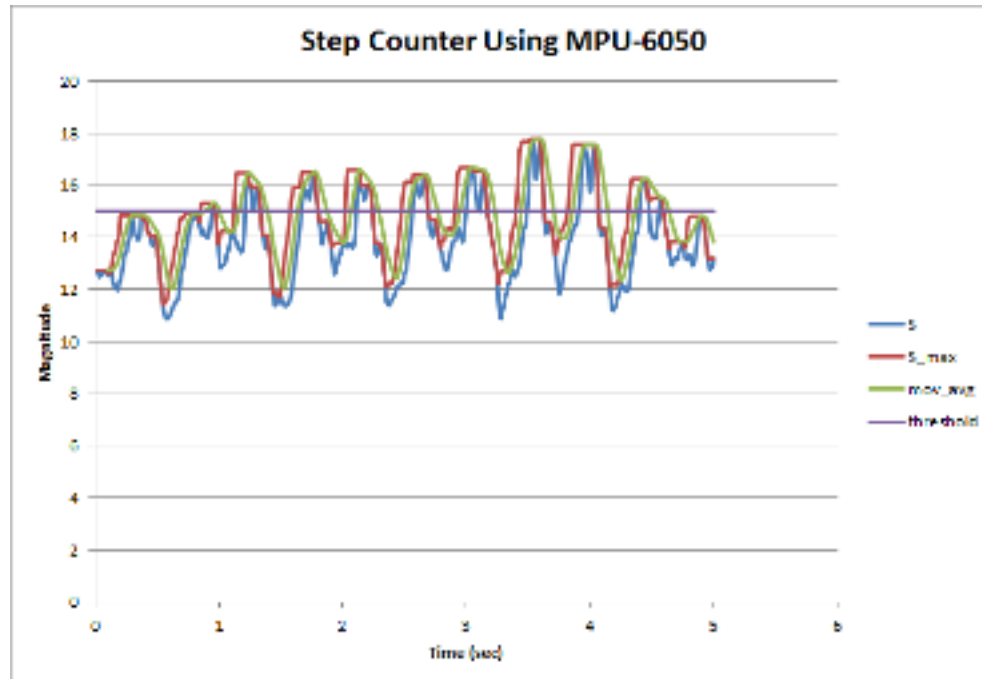




# Experiment 1: BASELINE VIDEO



# Optimally Calibrated Use-case (threshold = 15, sample/window length = 10 ms)



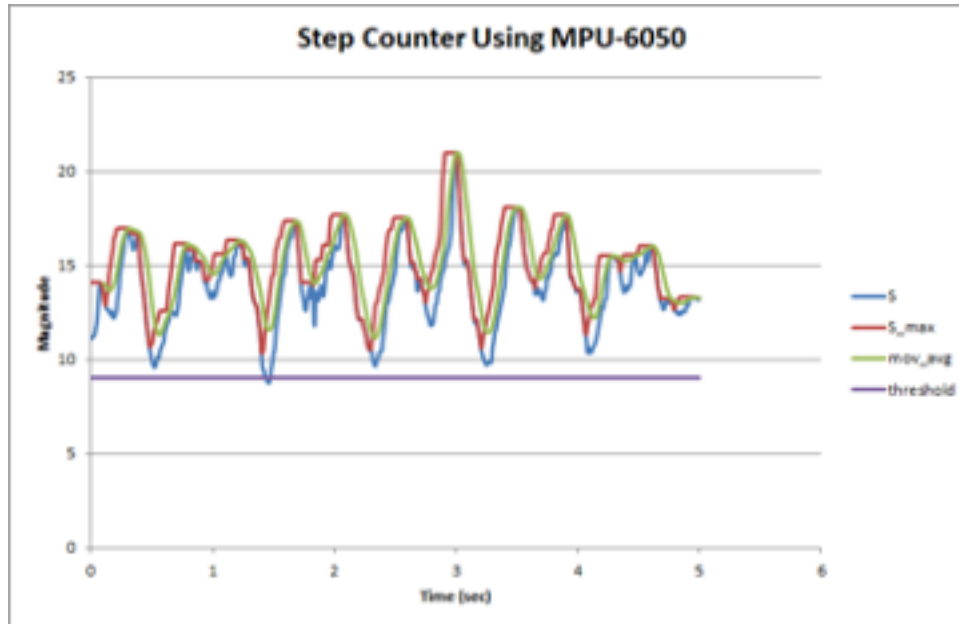
Outcomes:

- Actual Steps: 10
- Counted Steps: 10

Calibrated threshold and sample length allow for accurate step calculation.

# Uncalibrated Threshold Use-case #1

(threshold = 9, sample/window length = 10 ms)

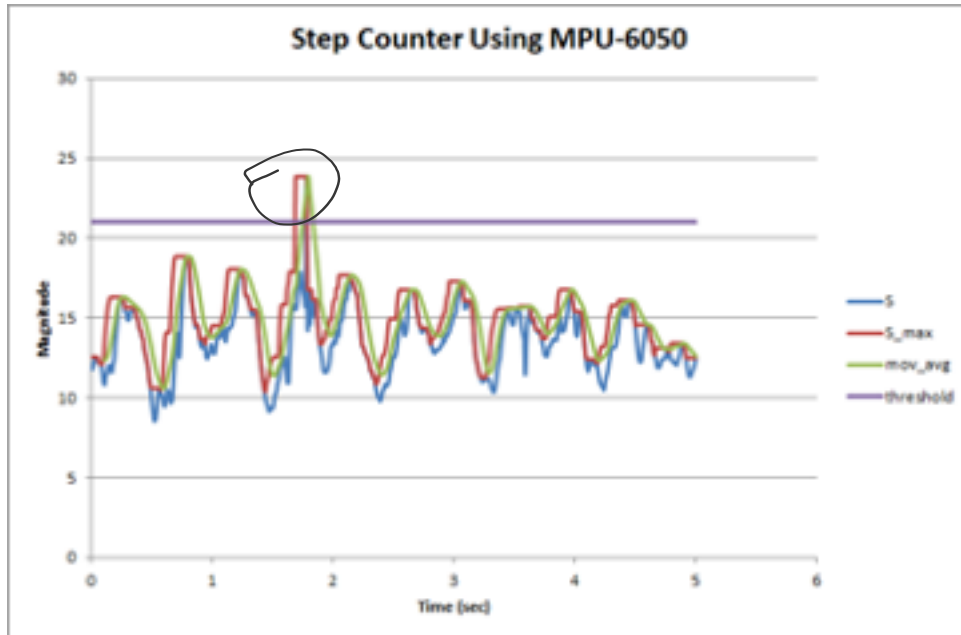


Outcomes:

- Actual Steps: 10
- Counted Steps: 17

Lack of threshold allows micro-movements to be counted as steps.

## Uncalibrated Threshold Use-case #2 (threshold = 21, sample/window length = 10 ms)

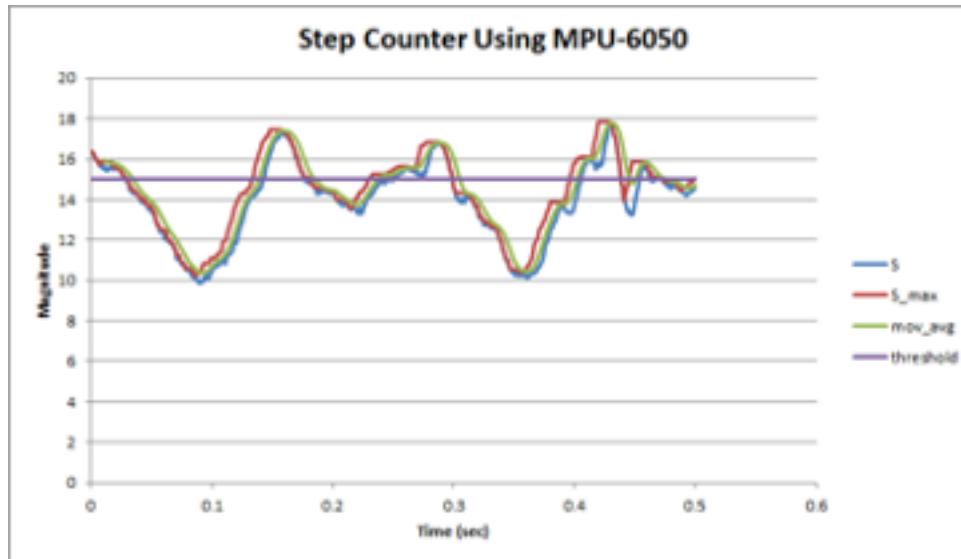


Outcomes:

- Actual Steps: 10
- Counted Steps: 1

Too high of threshold  
blocks anything from  
being counted.

# Over-Sampled Use-case (threshold = 15, sample/window length = 1 ms)



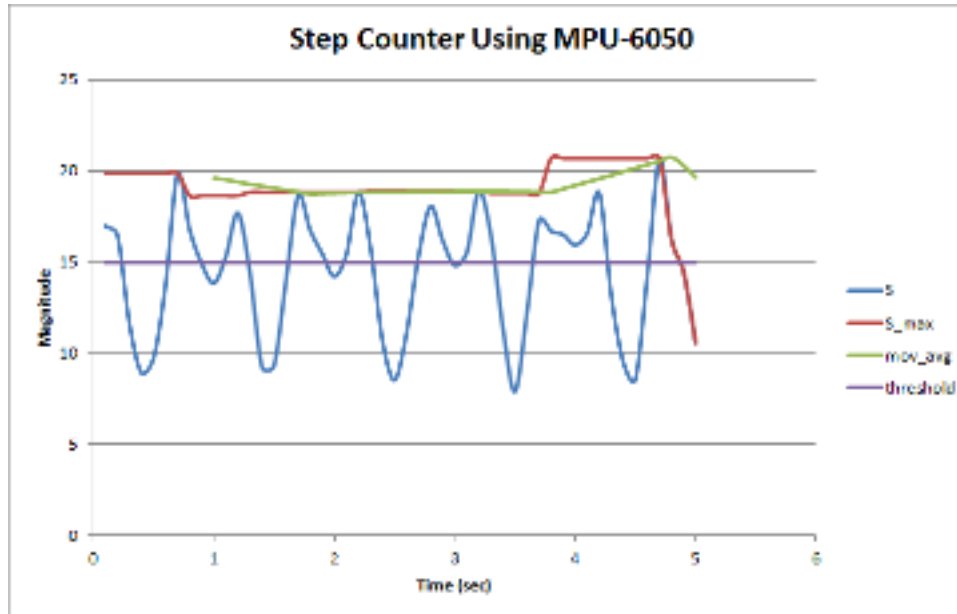
Outcomes:

- Actual Steps: 10
- Counted Steps: 19

Oversampling, too much data to analyze allows too much noise to get through the filter.



# Under-Sampled Use-case (threshold = 15, sample rate from 10ms to 100 ms)



Outcomes:

- Actual Steps: 10
- Counted Steps: 3

Undersampling, not enough data to sample skews the algorithm.



# Resources

<https://drive.google.com/drive/folders/1-ilULNyBdV8NkojHhglQZhAuE76lt2JF?usp=sharing>

[https://github.tamu.edu/mackenzie-wieberg/step\\_counter\\_mpu6050](https://github.tamu.edu/mackenzie-wieberg/step_counter_mpu6050)

Purchase [MPU-6050](#)