

# Smart Unit Care for Pre Fall Detection and Prevention

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**Abstract**— Generally falls may occur from moving or resting postures. This may include slipping from bed and fall from a sitting, or from running or walking. The pre-fall is a non-equilibrium state of human position that may lead to serious injuries, and may negatively impact the quality life condition, particularly for elders. Physical disabilities resulting from the fall incidences may lead to high costs involved with the healing process.

In this work, an embedded sensor system using Arduino micro-controller was utilized to coordinate the data received from accelerometer and gyroscope. For a given threshold voltage and fall pattern, the fall decision is made by the microcontroller, citing an incoming fall. The study addresses the number of sensors to be coordinated for enhancing probability of receiving a real fall. Sensors are suggested to be placed on the human body within a belt, and safety devices at human body as well as incorporated in a smart room will be coordinated with the processor commands. Near 150 ms time frame was detected from the simulation results, suggesting a safety device to be triggered and activated for protection within this time frame.

This paper discusses the research parameters such as response time for the device activation and interfacing the microcontroller to airbag switch, and means of activating the safety devices within the sharp edges in the smart unit care to minimize the impact of the fall injuries.

**Keywords**— Accelerometer, fall detection, Smart Unit Care, Smartphone, Sensors, Embedded systems, Machine Learning Method - MLM (key words)

## I. INTRODUCTION (HEADING 1)

In general, fall occurs mostly in the elders who may often lose their balance. This has become a frequent issue everywhere leading to serious complexities like death in some cases, physical disabilities, loss of self-confidence, insecurities and high cost involved in the treatment process etc. Statistical data indicates that near \$30billion are being spent annually for the fall treatment in the United States alone [1]. The idea of pre-fall detection is unique, utilizing the high speed of sensor and microcontroller processing to detect the fall information before it occurs. Much of the research efforts for detecting the

fall occurs after the fact that the fall has occurred, and a few work has been done to detect the fall earlier and secure the patients.

There are different ways to pre detect the falls using machine learning [2] [3], surface electromyography [4], vision based approach [5], and divide and conquer approach [6]. Recently, smart phones have been proposed to detect the fall before it occurs [7] [8]. Most of these methods use sensors and accelerometers. Tyson et al. [9] has estimated the pre-fall by looking into two different posture positions of a patient. Later a new system [10] was designed using three positions considering the orientation and the step mechanics during the fall. This system was tested for various types of falls at different speeds [10]. Lopes et al. [11] has optimized the system further by sending the fall information to their care takers. Recently some systems were designed with complex algorithms [12] with acceptable efficiency, using MLM's, featuring acceleration signals.

Researchers have used low complexity algorithms for the fall detections based on the threshold impacts of the accelerometer. Multiple sensors and branching algorithms have been accommodated in order to enhance the efficiency and smartness of the system. Restrictions with force, range of operation has set the upper limits of their use. Sometimes, individuals disregard the fall detection gadgets, as they introduce extra loads to wear them. This has motivated the use of smart phones since mobile phones are of highly carrying gadgets in the present days, and according to Pew Internet Project's research, near 64% of Americans has currently smartphones [13]. Therefore, utilization of smartphone in age groups higher than 65+ will expect to increase in the future.

## II. SMART UNIT CARE ENVIRONMENT

Figure 1 shows the proposed smart care unit with associated safety devices. An estimated time of 150ms resulted from simulation, suggests that the mini air bags surrounding the sharp edges of the care unit should be partially inflated in order to optimize the time of air compression within them. Edges of windows, beds, chairs, doors, and tables are associated with these safety devices.

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Figures 2 & 3 give the detection approach and the multiple sensor system respectively.



Figure 1: Smart Assistive Care Unit

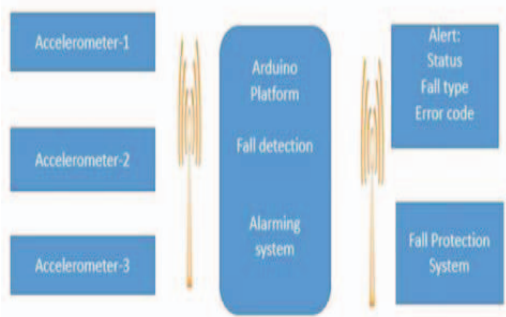


Figure 2: The Block diagram of Pre-fall detection system

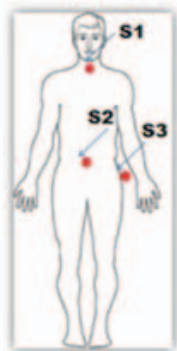


Figure 3: proposed sensor positions.

### III. EMBEDDED SENSOR SYSTEM

The integrated Arduino microcontroller with the sensor system was developed and tested for the fall detection of different positions. Figures 4 & 5 provide the hardware sensor system and the algorithm used for the development.

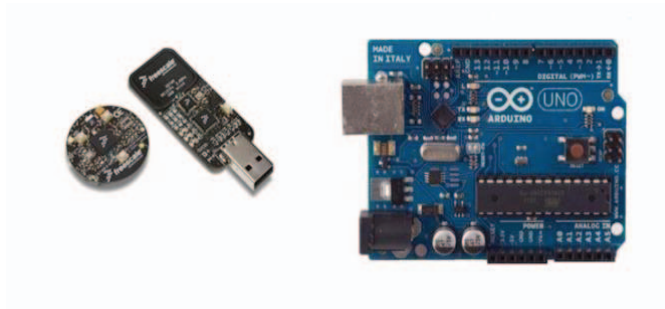


Figure 4: The Hardware for the pre-fall detection system

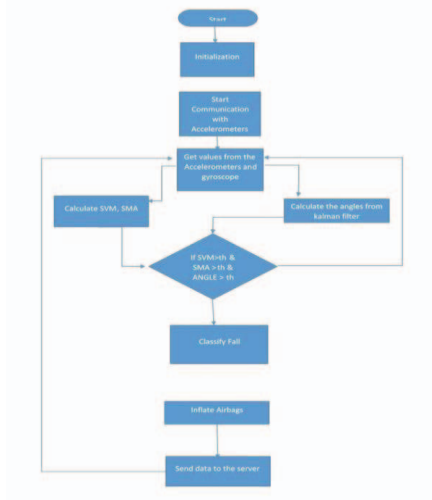


Figure 5: The algorithm of the system development

### IV. AIRBAG SAFETY DEVICES

After detecting a fall there should be a safety device to protect the human from injuries. This device should simultaneously trigger a mini air bag before the fall occurs so that it alleviates the fall impact. Figure 6 provides the hardware used for interfacing the microcontroller to an airbag switch. In order to illustrate the impact of the airbag size on the safety device response, let us consider a general life vest shown in figure 7.

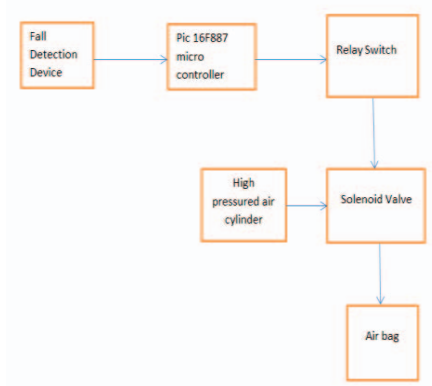


Figure 6: Design to trigger the Air bag.





## V. RESULTS AND DISCUSSIONS

Figure 10 shows the accelerometer signals and the pre-fall detection before the impact occurs. As it can be seen from various testing, near 150ms was determined from the simulation. Figure 11 shows the data from the start of the signal until the fall. The signal vector magnitude (SVM) changes from 1.37 to 0.8 corresponding to time change between 270ms to 413 ms, indicating near 150ms time frame for the pre-fall period of time. The impact occurs shortly after detecting the minimum magnitude. The efficiency was estimated based on the threshold SVM value given to the controller. A larger SVM may lead to a better efficiency, while shorter time for the impact to occur. A smaller threshold may lead to bigger time frame, while may give more of false fall detections.

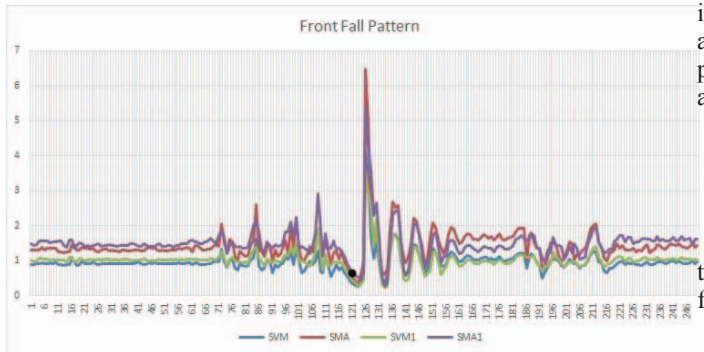


Figure 10: Accelerometer data indicating the pre-fall signal and impact timing

| SVM1   | Milliseconds |
|--------|--------------|
| 1.3742 | 270          |
| 1.2817 | 284          |
| 1.2425 | 289          |
| 1.1527 | 293          |
| 1.1183 | 297          |
| 1.1193 | 301          |
| 1.1287 | 304          |
| 1.0608 | 307          |
| 0.9946 | 311          |
| 1.0796 | 315          |
| 1.0026 | 319          |
| 1.0127 | 322          |
| 1.0271 | 325          |
| 1.0146 | 329          |
| 0.9959 | 333          |
| 0.9983 | 337          |
| 0.9538 | 341          |
| 0.9691 | 344          |
| 0.9833 | 348          |
| 0.9133 | 352          |
| 0.9396 | 356          |
| 0.9109 | 359          |
| 0.8947 | 363          |
| 0.9018 | 367          |
| 0.8734 | 371          |
| 0.874  | 376          |
| 0.8687 | 379          |
| 0.8504 | 383          |
| 0.8072 | 386          |
| 0.8603 | 389          |
| 0.8433 | 393          |
| 0.7803 | 397          |
| 0.8143 | 401          |
| 0.7524 | 405          |
| 0.8101 | 409          |
| 0.8    | 413          |

Figure 11: Screenshot of data acquisition

## VI. CONCLUSION AND FUTURE WORK

The findings presented here indicate the high efficiency system on detecting the falls before they occur. Near 150ms is too short for triggering a sizable air bag. The interface of the microcontroller to the air bag switch provides basis of smart care units where mini air bags may be incorporated at the critical positions for elder protection. The suggestion of partially inflated air bags may satisfy the time frame of 150ms required for the security device to activate. This project addresses how to develop a device that can protect a person from fall injuries. Initial phase with a life vest and suggested high speed devices were discussed to optimize system latency. This hardware was implemented and verified its validity for proper interface between the microcontroller and the triggering switch. Future work should incorporate the pressure model that is suited for a patient weight. The design and testing of this work is reserved for future considerations.

## VII. ACKNOWLEDGEMENTS

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