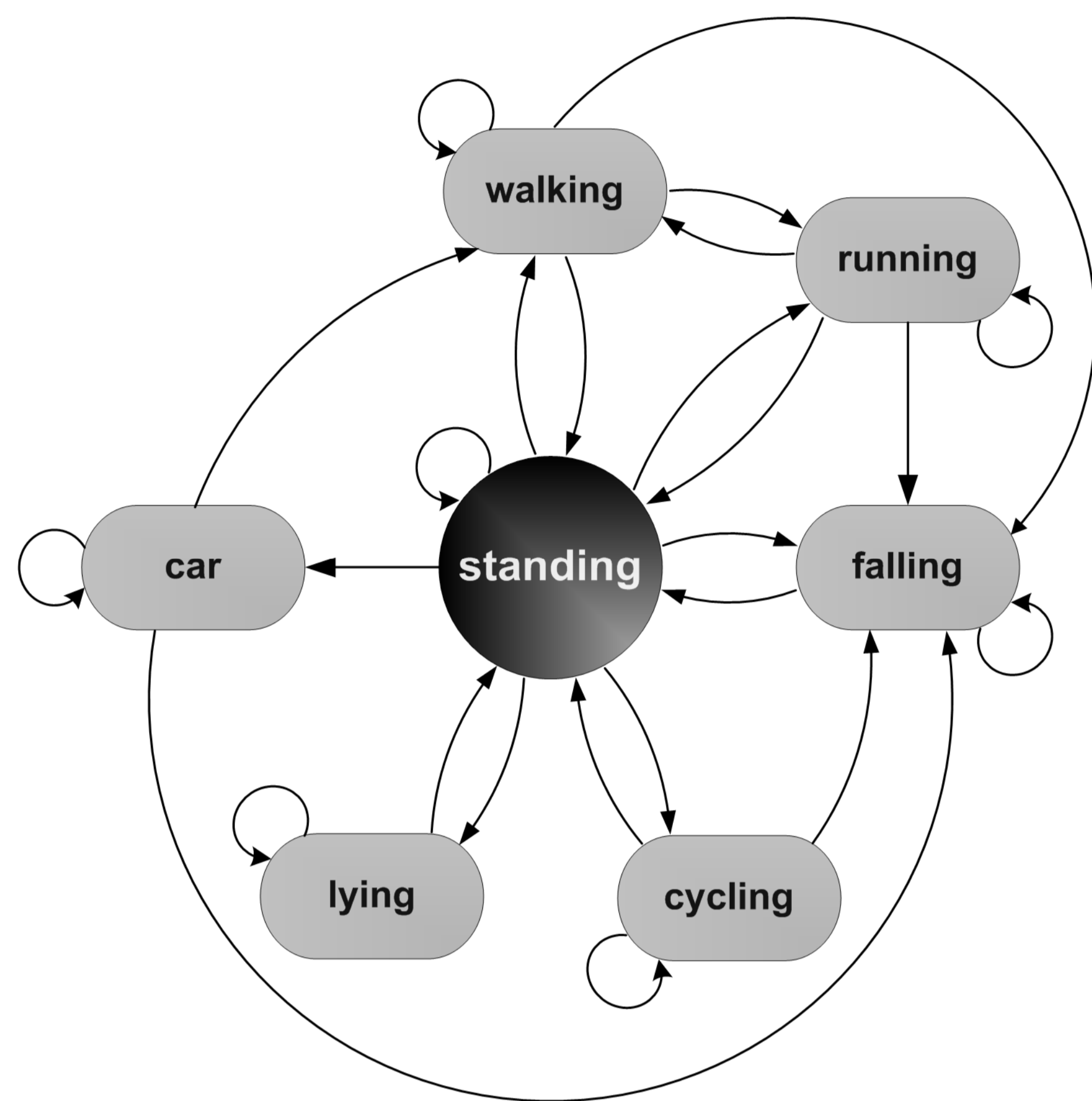
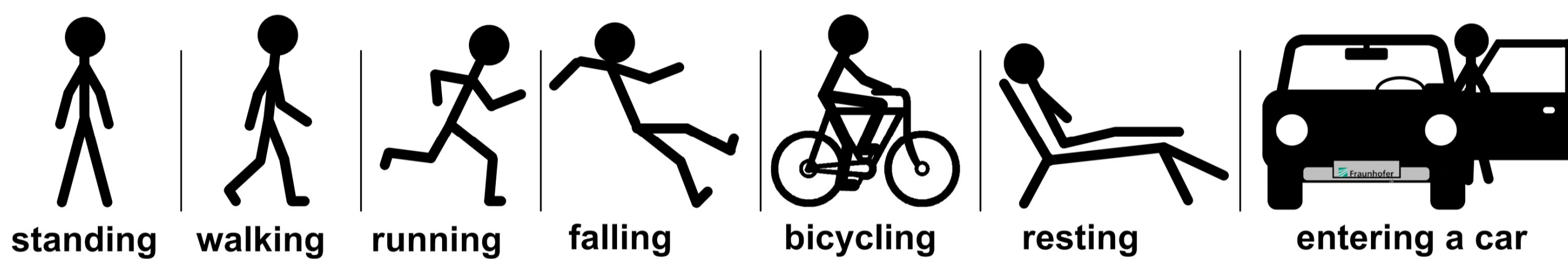


Pedestrian Movement Classification with Inertial Sensor Data

Overview



- Reduction of severe accidents in road traffic situations by detection and recognition of vulnerable road users.
- The combination of movement classification data, distance and angle of arrival measurement enables the detection of hazardous situations.
- Detectable States:



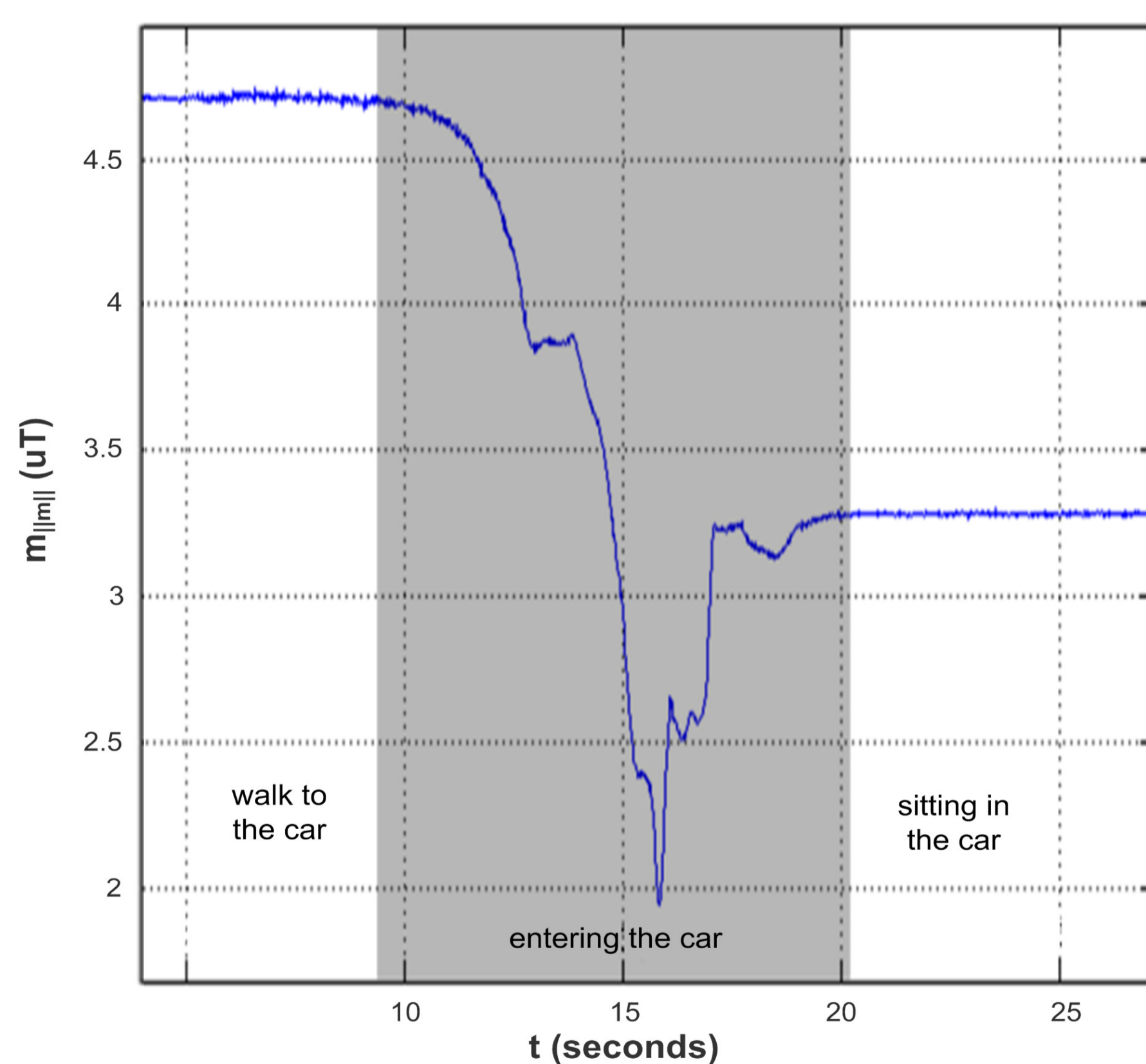
Description of Hardware



- The hardware is build as a low-cost system, based on the typical sensor technology (MEMS) used in regular smartphones.
- The hardware can be used as a stand alone system or integrated in the hip worn demonstrator.
- Size (component): 58 mm x 37 mm x 12 mm
- Size (stand alone): 117 mm x 73 mm x 28 mm

| | Accelerometer | Gyroscope | Magnetic Field Sensor |
|-------------|----------------------|-----------|-----------------------|
| Dimensions | 3 Axis | 3 Axis | 3 Axis |
| Range | ±2.0 / ±4.0 / ±8.0 g | ±2000 °/s | ±0.7 / ±4 Gauss |
| Sample Rate | 1000 Hz | 8 kHz | 50 Hz |

Explanation of the Classification



- The classification algorithm is fed with data from:
 - Gyroscopes
 - Accelerometers
 - Magnetometers
- Each movement generates characteristic signal data which are used as generic pattern for the classification.
- Example – Getting into a car:
 - Steel in car body causes an anomaly in earth's magnetic field
 - Typical patterns can be detected using hip worn sensors:
 - Rotations of a person while entering a car
 - Accelerations at the point of contact between body and car seat
 - Combinations of all typical patterns give information on person's movement

Results

| values / percent | | labeled data | | | | | | |
|------------------|----------------|--------------|---------|---------|----------------|----------|---------|-------|
| | | standing | walking | running | entering a car | throwing | cycling | lying |
| classification | standing | 92 | 0 | 0 | 24 | 0 | 7 | 0 |
| | walking | 8 | 100 | 0 | 0 | 0 | 21 | 0 |
| | running | 0 | 0 | 100 | 0 | 0 | 0 | 0 |
| | entering a car | 0 | 0 | 0 | 76 | 0 | 0 | 0 |
| | throw | 0 | 0 | 0 | 0 | 100 | 0 | 0 |
| | cycling | 0 | 0 | 0 | 0 | 0 | 72 | 0 |
| | lying | 0 | 0 | 0 | 0 | 0 | 0 | 100 |

Discussion of Results:

- Using the algorithm, all activities are classified correctly with an accuracy of 91 %.
- States "walking", "running" and lying can be classified without false-detection.
- "Entering a car" and "cycling" are the most challenging states:
 - Designs of cars (bikes) are different and lead to different patterns
 - An incorrect (e.g. loose) attachment of the sensor to the human body increases false classifications