

Communication Technologies 2 (CT2)

Machine Learning:

Applications and Algorithms

[Dead Reckoning]

[M.Sc. Christoph Anderson]

Lecture in WS 2018 / 2019
[06.12.2018]



Agenda

- Introduction
- Activity Recognition
- Alignment for Context Prediction
- Time-Series II
- WiFi-Fingerprinting
- **Dead-Reckoning**
- Gaussian Mixture Models
- Sensor Data Fusion

Outline

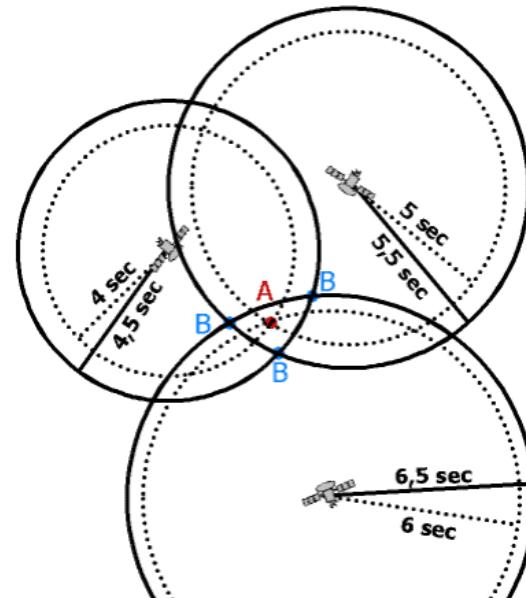
- Common Approaches for Positioning
- Motivation
- Introduction to Dead-Reckoning
- Step-Heading-Systems (SHS)
- Inertial-Navigation-Systems (INS)
- Maps
- Pitfalls and Errors

Common Approaches for Positioning

DEAD-RECKONING

Common approaches (GPS)

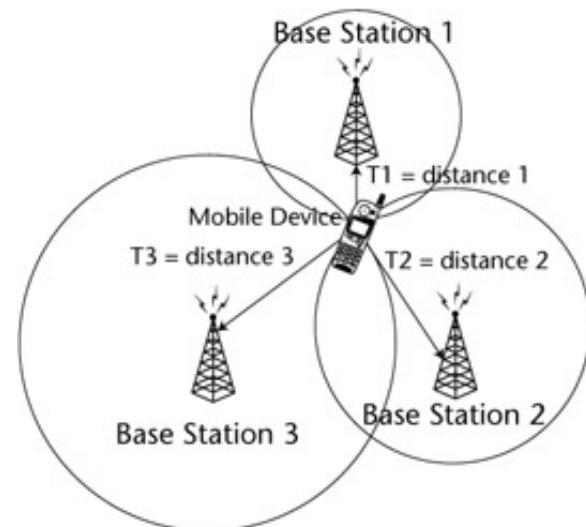
- GPS is widely used for providing positioning services
- Line-of-sight (LoS) transmission with satellites
 - Signal (time sent)
 - Name of the satellite
 - Position
- Poor coverage for indoor environments



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Common approaches (Trilateration)

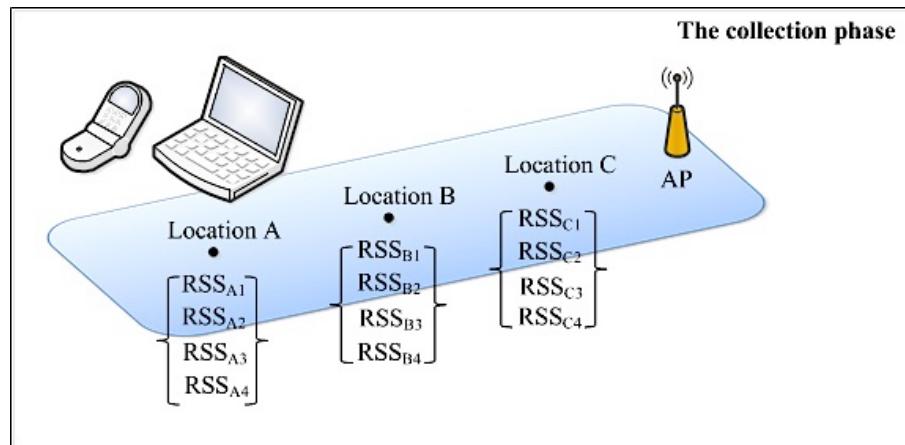
- Time of Arrival (TOA)
 - Measure the propagation time
 - Calculate the distance between the receiver and the transmitter.
 - The clock of transmitters and receivers must be precisely synchronized



[6]

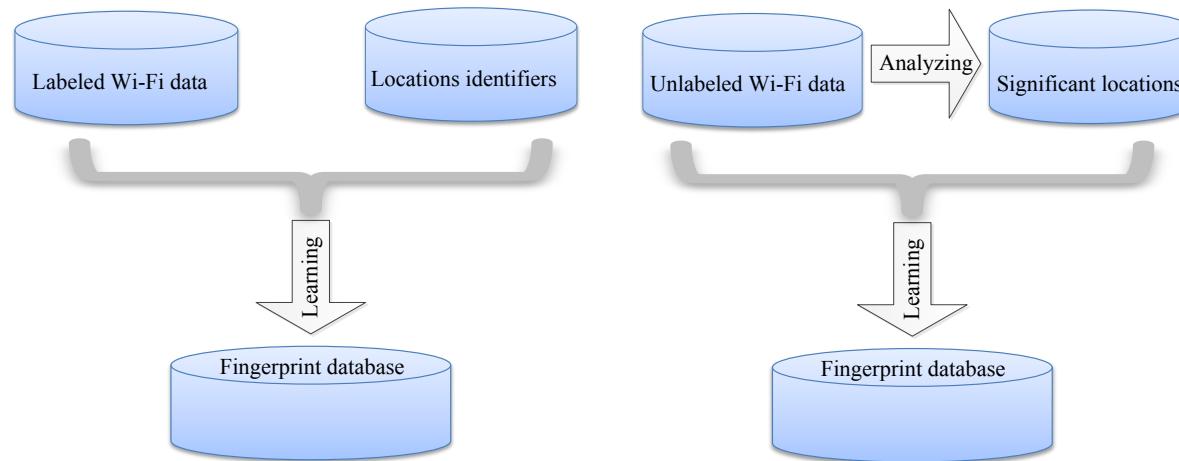
Common Approaches (WiFi-Fingerprinting)

- Information
 - MAC address
 - Receive signal strength indicator (RSSI)
 - SSID
 - Channel (frequency)



Common Approaches (WiFi-Fingerprinting)

- Information
 - MAC address
 - Receive signal strength indicator (RSSI)
 - SSID
 - Channel (frequency)



Common Approaches

- Proximity based approaches
 - RFID-Systems
 - Bluetooth Receivers
- Low administration and calibration
- Need high density of receivers/senders
- Provide only coarse location information

Motivation

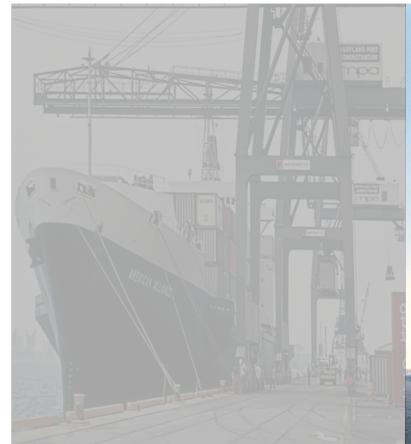
DEAD-RECKONING

Motivation

- Why is it useful to know the exact position indoors?
 - Navigation to unknown places/rooms
 - Context-awareness
 - Advertisement
 - Building management

Motivation

- How to detect ones position?
- Back in history:
 - A captain had to determine the ship's position
 - Based on a starting point (harbor, lighthouse etc.)



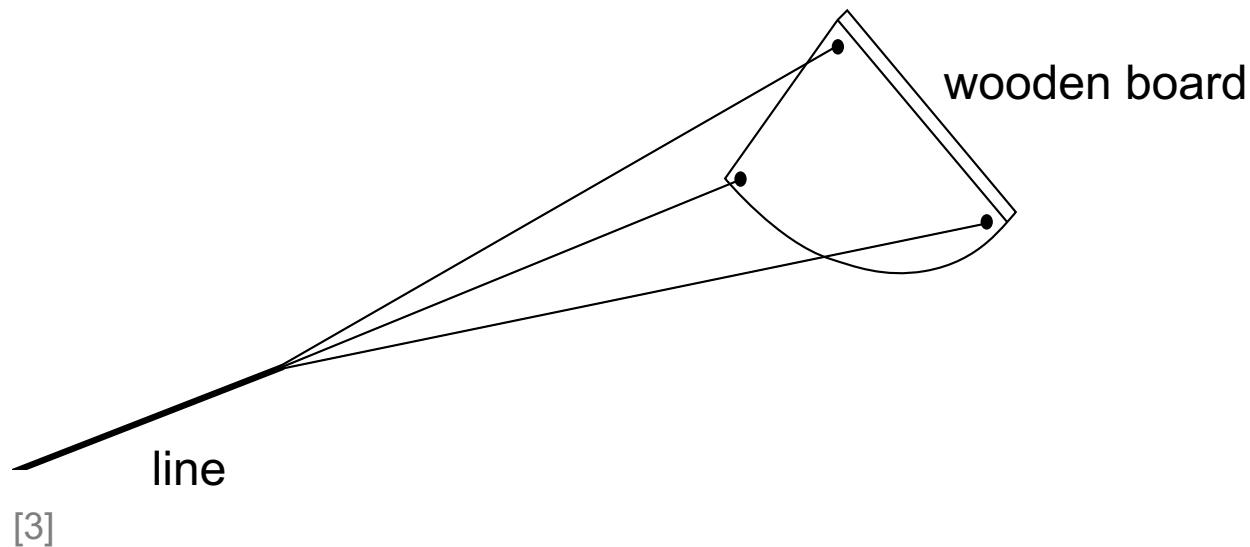
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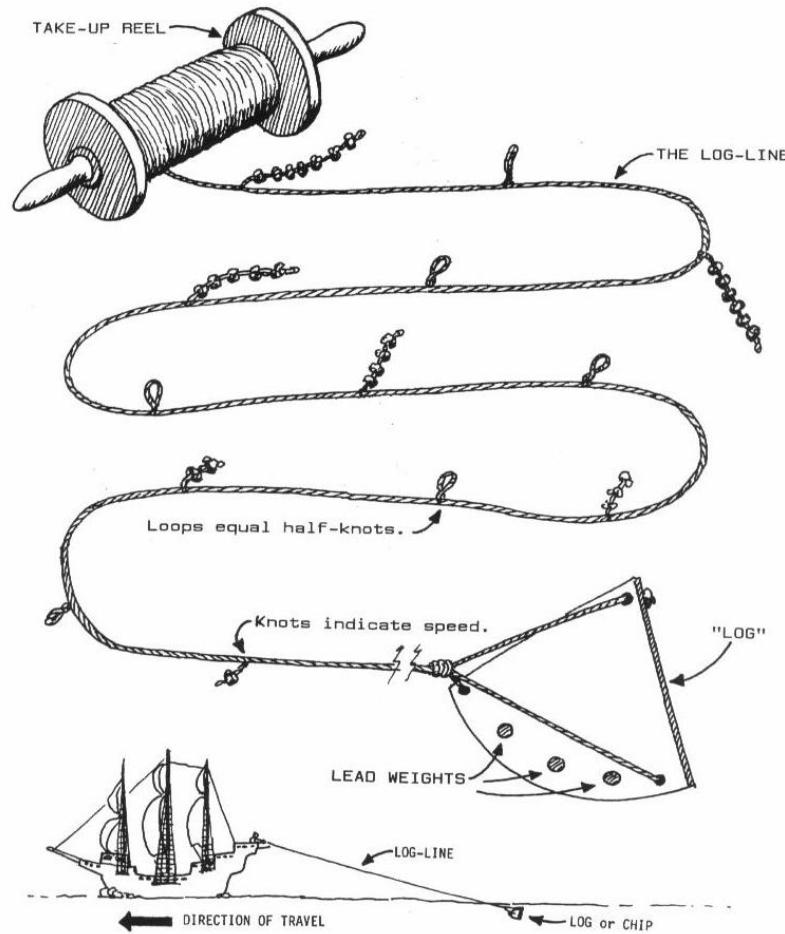
Motivation

- Determining the position by:
 - Using a log for calculating the speed of the ship
 - Device with a line and a wooden board



Motivation

The Log-line

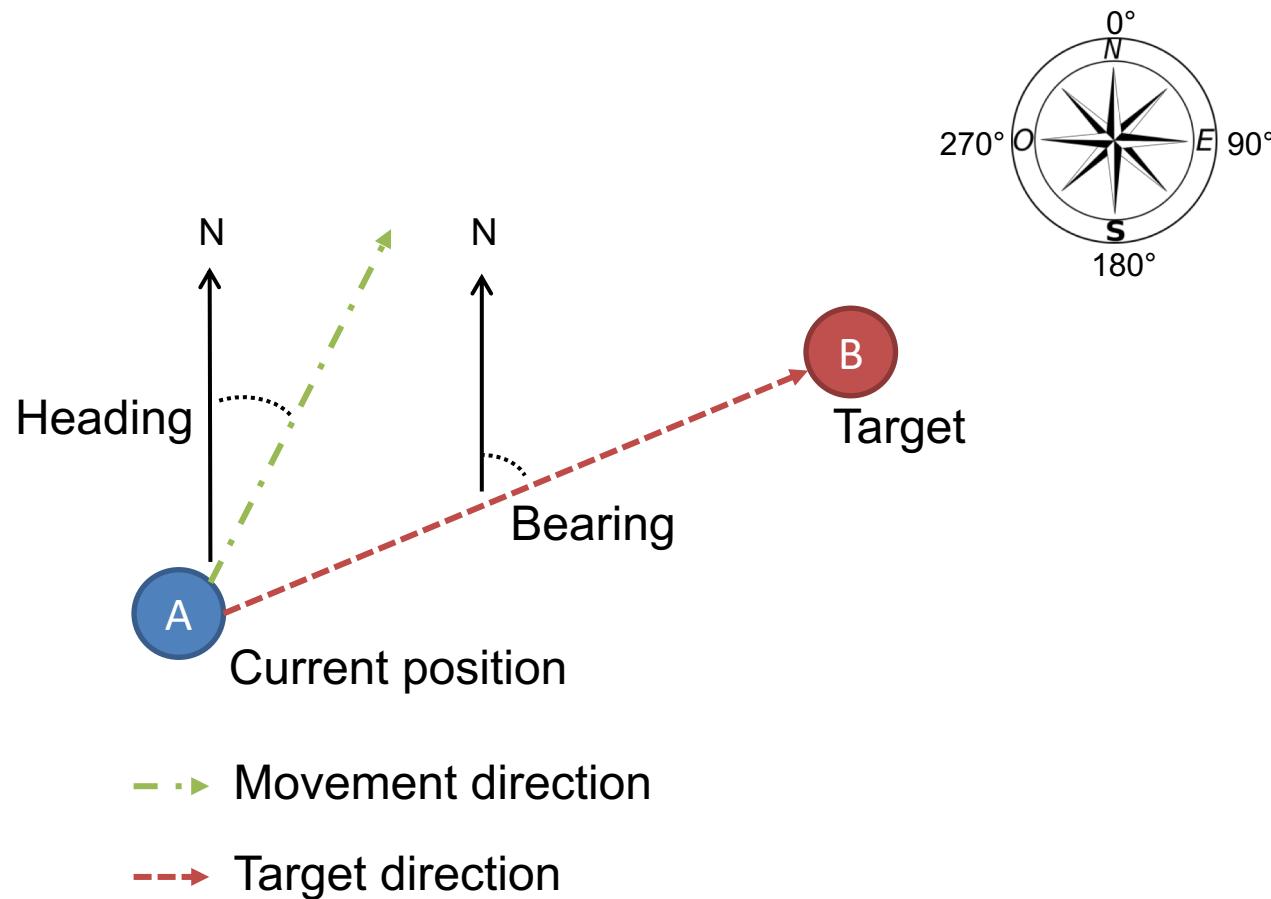


[4]

Motivation

- Knots were woven into the line at a fixed length
- Estimating the speed of the ship:
 - Using a sand glass for time measuring
 - Knots passed by in that period time
- Calculating the course of the ship:
 - Using a compass for heading, bearing estimation

Motivation



Definition

DEAD-RECKONING

Definition of Dead-Reckoning

- What is Dead-Reckoning?

„Dead reckoning is a relative navigation technique. Starting from a known position, successive position displacements are added up...more typically, in heading and speed or distance“ [7]

- Also known as „Deduced Reckoning“
 - D-E-D Reckoning
 - Dead-Reckoning

Introduction to Dead-Reckoning

- Pedestrian Dead-Reckoning (PDR)
 - Using a smartphone as a navigation tool
 - No log lines or wooden boards
- Built-in sensors
 - Magnetometer
 - Acceleration
 - Gyroscope
 - GPS
 - Wi-Fi
- Low deployment costs

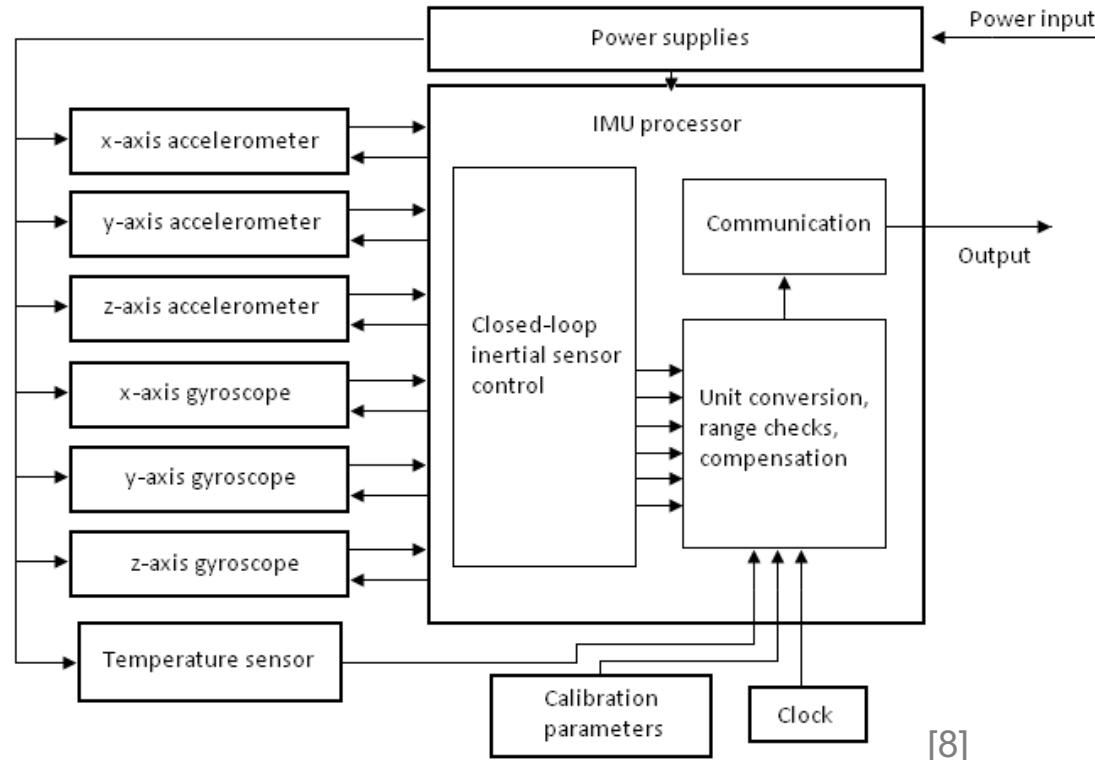
Introduction to Dead-Reckoning



9 Axis IMU Sensor
3-axis gyroscope
3-axis accelerometer
3-axis magnetometer

[16]

Introduction to Dead-Reckoning



[8]

Introduction to Dead-Reckoning

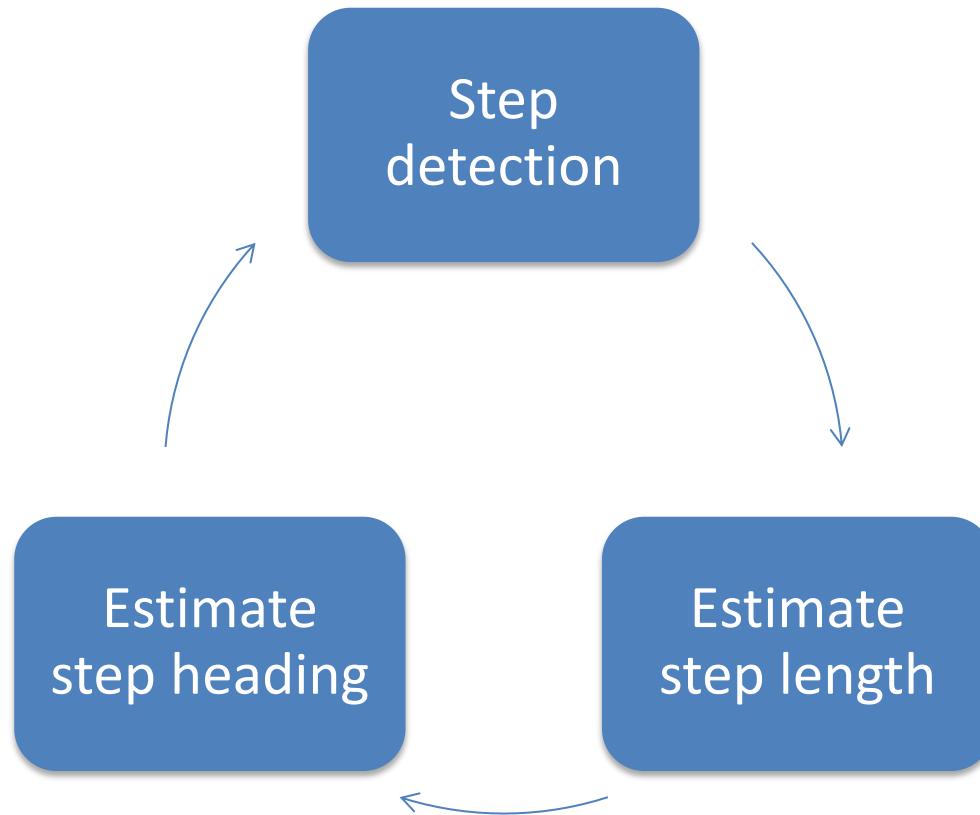
- PDR-Systems can be divided in
 - Step-Heading-Systems (SHS)
 - Inertial-Navigation-Systems (INS)
 - SHS-INS Systems
- Focus lies on SHS-Systems
 - Easier to implement

Step-Heading-Systems / Inertial-Navigation-Systems

DEAD-RECKONING

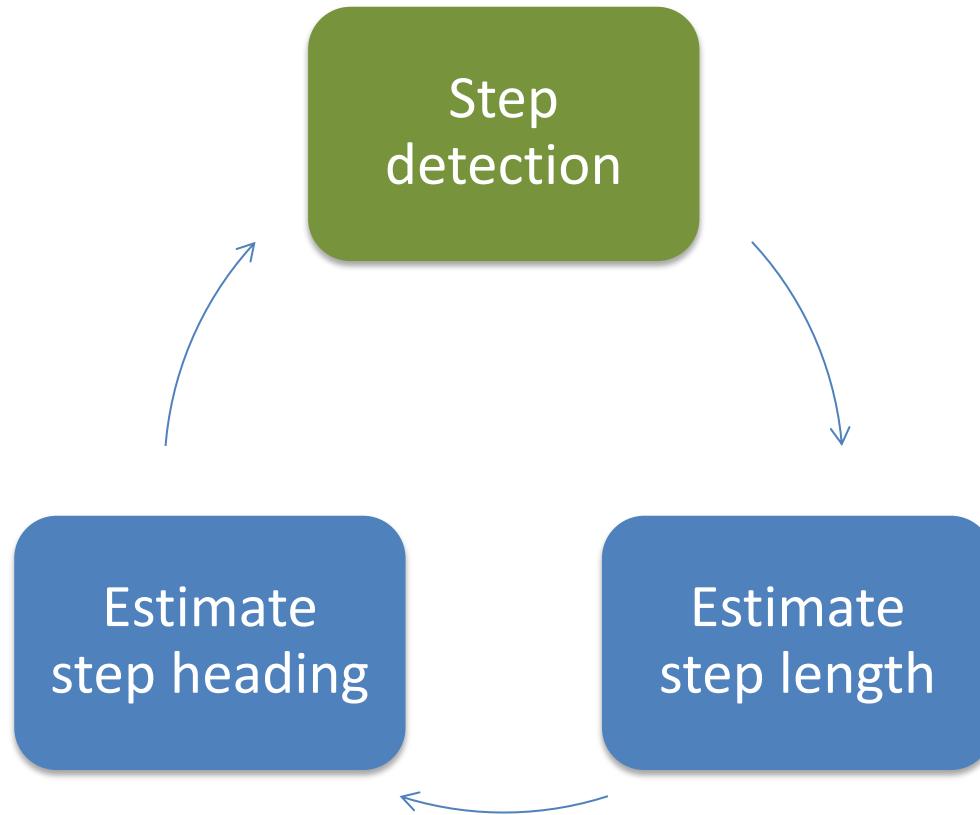
Step-Heading-Systems (SHS)

- Basic cycle of SHS-Systems



Step-Heading-Systems (SHS)

- Basic cycle of SHS-Systems



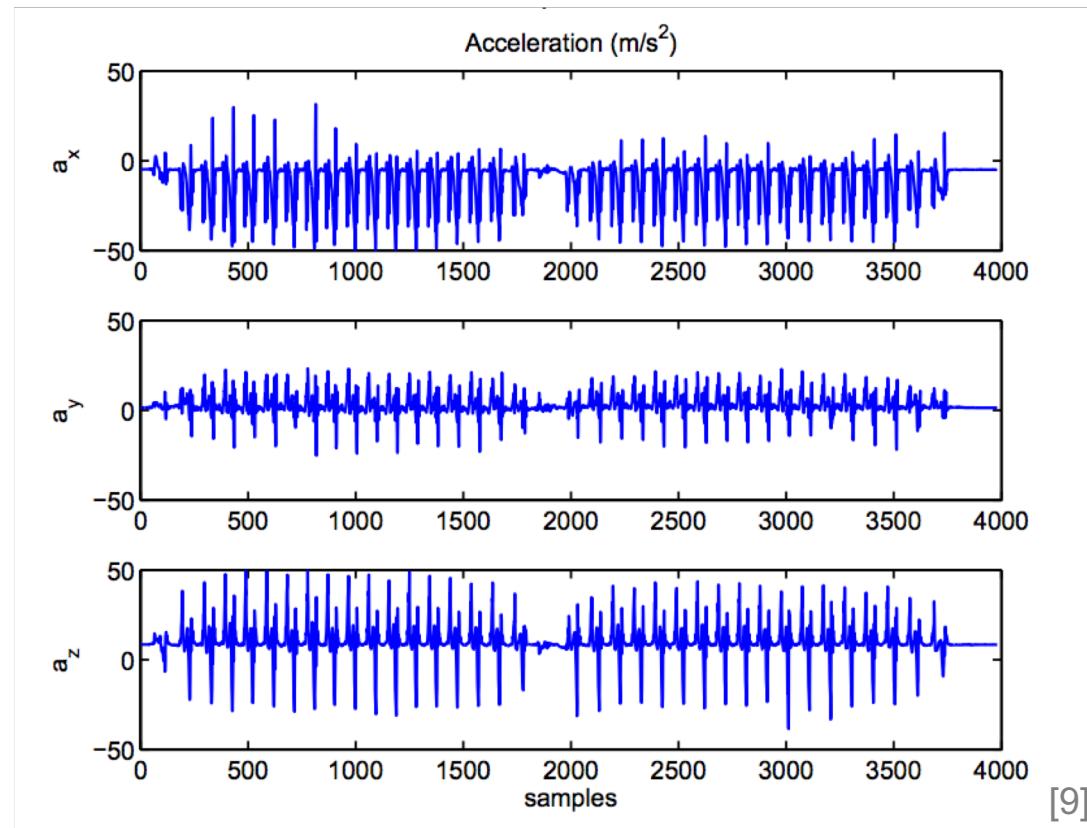
Step-Heading-Systems (SHS)



- Recognizing steps from sensor data
 - Mostly by using acceleration sensors
 - Other approaches include gyroscopes
- Most of these approaches are threshold based
 - During static phase, sensor will report low values
 - During non-static phase, sensor report activity
- Example: Stance detection
 - Acceleration sensor mounted on the foot
 - Records sensor data and estimate foot stance

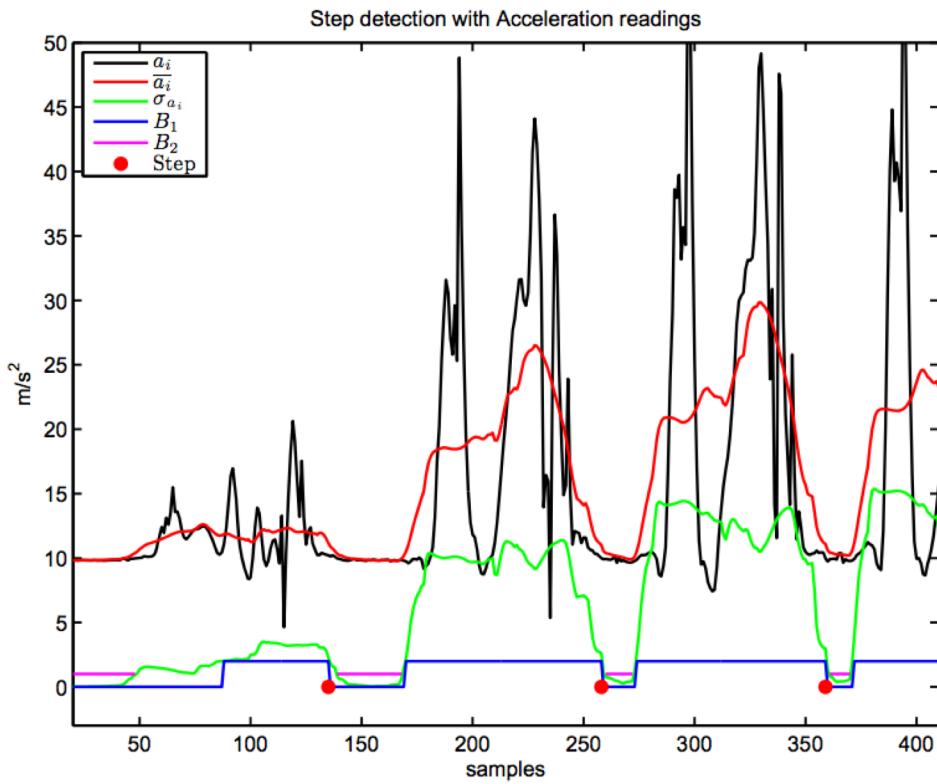
Step-Heading-Systems (SHS)

- Example: Stance detection



Step-Heading-Systems (SHS)

- Example: Stance detection



$$a_i = \sqrt{a_{x_i}^2 + a_{y_i}^2 + a_{z_i}^2}.$$

$$\sigma_{a_i}^2 = \frac{1}{2w+1} \sum_{j=i-w}^{i+w} (a_j - \bar{a}_j)^2,$$

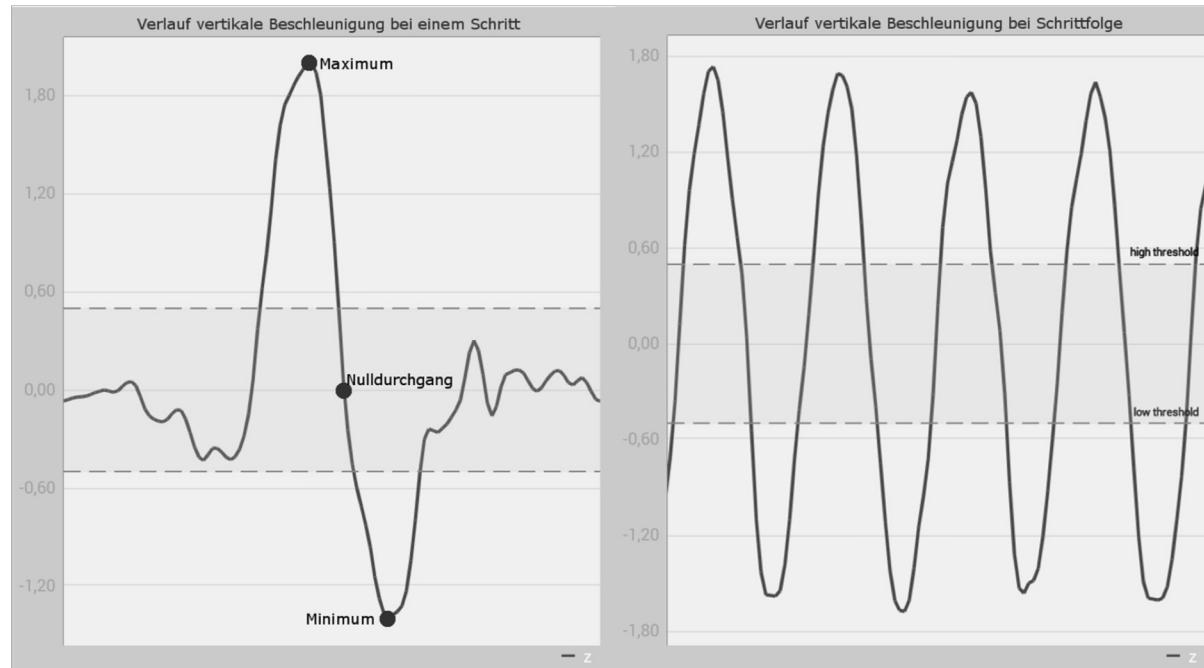
$$B_{1i} = \begin{cases} T1 & \sigma_{a_i} > T1 \\ 0 & \text{otherwise} \end{cases} .$$

$$B_{2i} = \begin{cases} T2 & \sigma_{a_i} < T2 \\ 0 & \text{otherwise} \end{cases} .$$

[9]

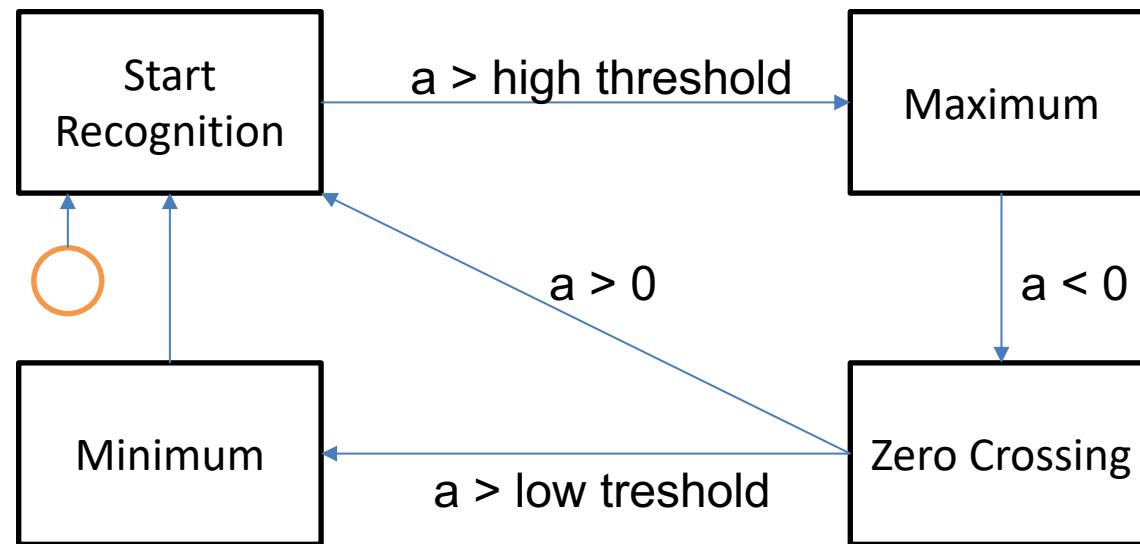
Step-Heading-Systems (SHS)

- Example: Step-Cycle detection
 - Acceleration sensor mounted near the waist/leg
 - Same position used for gyroscope based approaches



Step-Heading-Systems (SHS)

- Example: Step-Cycle detection
 - State-Machine:



Step-Heading-Systems (SHS)



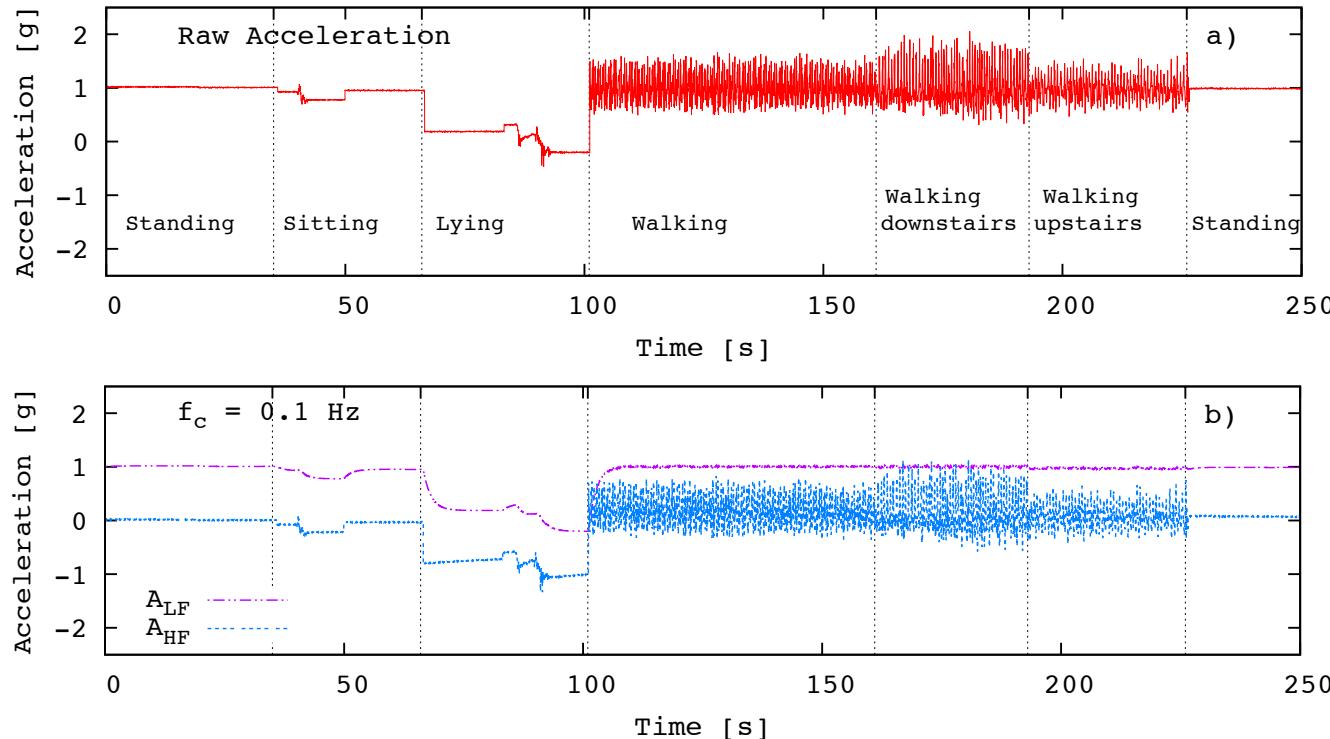
- Example: Step-Cycle detection:
 - Usually common features such as mean, magnitude or variance are used
- Algorithms used for step detection:
 - Peak detection
 - Sharp changes in vertical acceleration
 - Auto correlation
 - Matching periodic patterns in the magnitude signal
 - Zero crossing
 - Detect zero crossings, special form of thresholding

Step-Heading-Systems (SHS)



- Compensate for:
 - Gravity in acceleration signals
 - Momentary noise and peaks in acceleration
- Signal processing for filtering out noise
 - Average filter
 - Butterworth Low-Pass filter

Step-Heading-Systems



[10]

Step-Heading-Systems

- Instead of counting steps:
 - Recognize continuous movements
 - Walking detection
- For example using a decision tree
 - Calculating several features of acceleration signals
 - Variance, Minimum, Maximum...

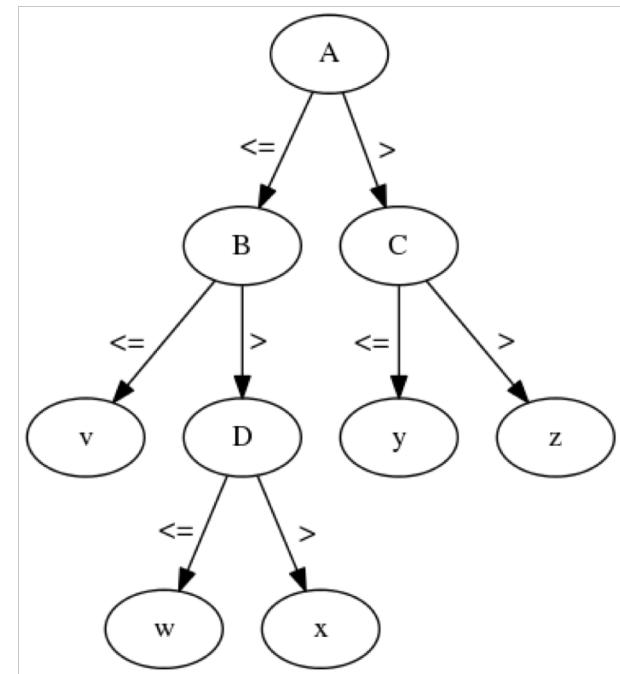
Information Gain and Entropy

- Information Gain:

$$IG(FS) = H(FS) - H(FS|f)$$

- Entropy:

$$H(FS) = \sum_{i \in FS} -p_i \log_2 p_i$$



Decision Trees

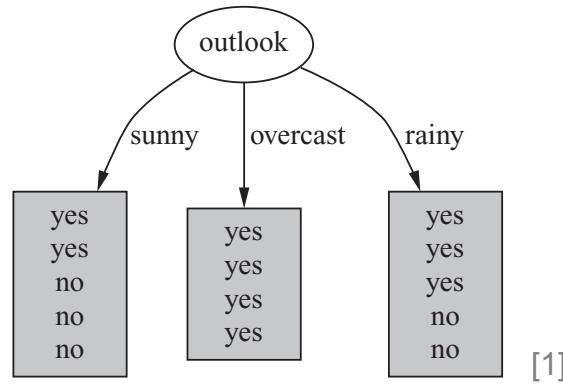
- Information entropy
 - Comes from information theory
 - The higher the entropy, the higher the information content

$$\sum_i -p_i \log_2 p_i$$

- p_i is the probability of class i
- \log_2 is the logarithm to the base 2

Decision Trees

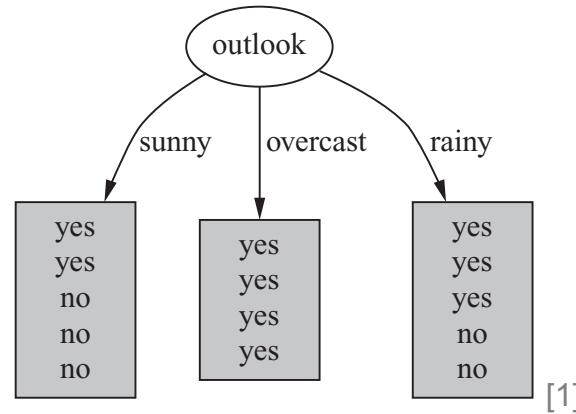
- What is the information entropy?



- $info([2,3]) = -\left(-\frac{2}{5} * \log_2\left(\frac{2}{5}\right) + (-\frac{3}{5} * \log_2\left(\frac{3}{5}\right))\right) = 0.971 \text{ bits}$
- $info([4,0]) = -(-1 * \log_2(1)) = 0.0 \text{ bits}$
- $info([3,2]) = -\left(-\frac{3}{5} * \log_2\left(\frac{3}{5}\right) + (-\frac{2}{5} * \log_2\left(\frac{2}{5}\right))\right) = 0.971 \text{ bits}$

Decision Trees

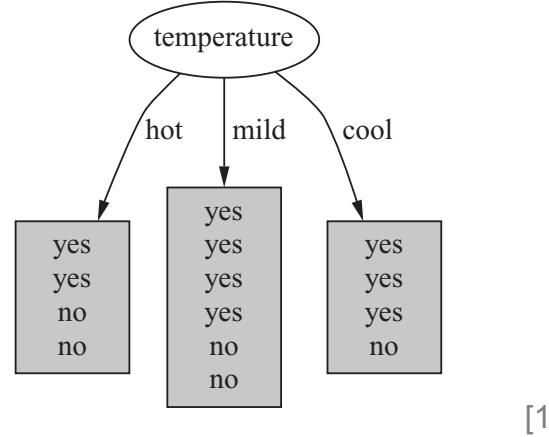
- What is the information entropy?



$$\begin{aligned} - \left(-\frac{5}{14} * 0.971 + \left(-\frac{4}{14} * 0 \right) + \left(-\frac{5}{14} * 0.971 \right) \right) = \\ 0.693 \text{ bits} \end{aligned}$$

Decision Trees

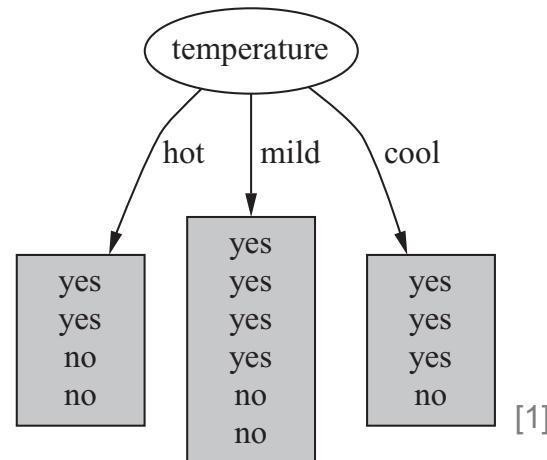
- What is the information entropy?



- $info([2,2]) = -\left(-\frac{2}{4} * \log_2\left(\frac{2}{4}\right) + (-\frac{2}{4} * \log_2\left(\frac{2}{4}\right))\right) = 1 \text{ bit}$
- $info([4,2]) = -\left(-\frac{4}{6} * \log_2\left(\frac{4}{6}\right) + (-\frac{2}{6} * \log_2\left(\frac{2}{6}\right))\right) = 0.918 \text{ bits}$
- $info([3,1]) = -\left(-\frac{3}{4} * \log_2\left(\frac{3}{4}\right) + (-\frac{1}{4} * \log_2\left(\frac{1}{4}\right))\right) = 0.811 \text{ bits}$

Decision Trees

- What is the information entropy?



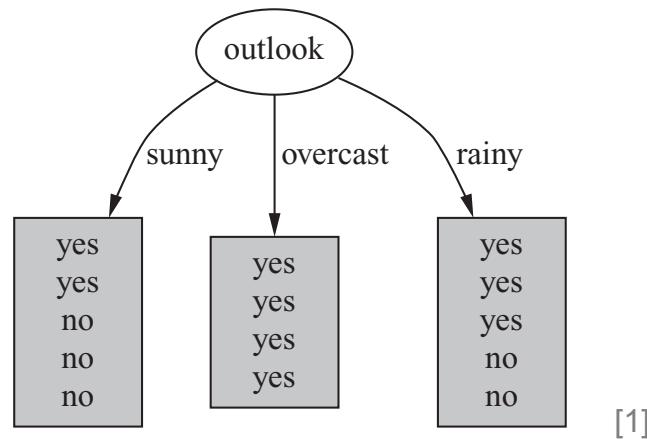
$$- \text{info}([2,2], [4,2], [3,1]) = \left(-\frac{4}{14} * 1 + \left(-\frac{6}{14} * 0.918 \right) + \left(-\frac{4}{14} * 0.811 \right) \right) = 0.911 \text{ bits}$$

Information Gain

- **Information gain**
 - Is the amount of information we gain by knowing the value of an attribute
 - $(\text{entropy distribution before the split}) - (\text{entropy distribution after the split})$
- Gives us the attribute which is most useful for discriminating classes
- Describe how important an attribute of the features is

Information Gain

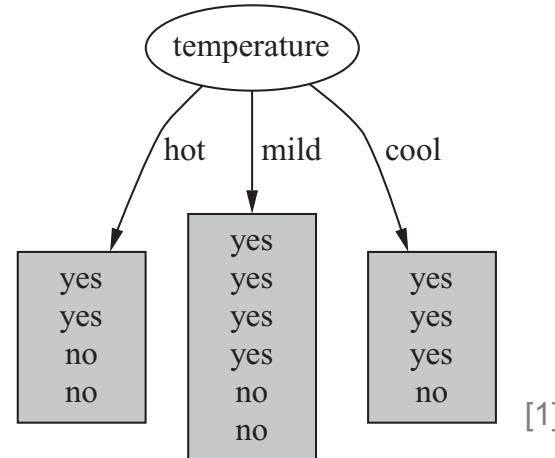
- What is the information gain?



- $info([9,5]) = -\left(-\frac{9}{14} * \log_2\left(\frac{9}{14}\right) + \left(-\frac{5}{14} * \log_2\left(\frac{5}{14}\right)\right)\right) = 0.940 \text{ bits}$
- $gain(outlook) = info([9,5]) - info([2,3], [4,0], [3,2]) = 0.940 - 0.693 = 0.247 \text{ bits}$

Information Gain

- What is the information gain?

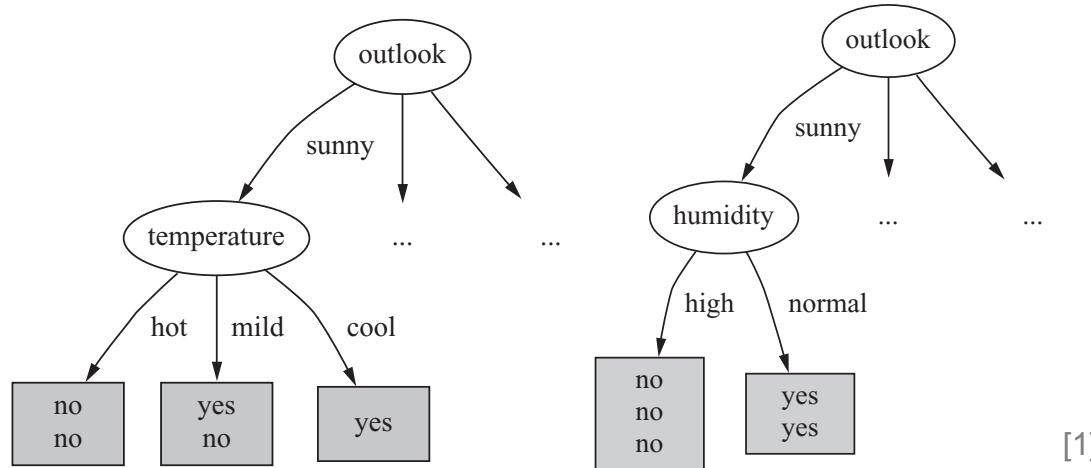


- $info([9,5]) = -\left(-\frac{9}{14} * \log_2\left(\frac{9}{14}\right) + \left(-\frac{5}{14} * \log_2\left(\frac{5}{14}\right)\right)\right) = 0.940 \text{ bits}$
- $gain(temperature) = info([9,5]) - info([2,2], [4,2], [3,1]) = 0.940 - 0.911 = 0.029 \text{ bits}$

Information Gain

- What is the information gain?
 - $gain(outlook) = 0.247$ bits
 - $gain(temperature) = 0.029$ bits
 - $gain(humidity) = 0.152$ bits
 - $gain(windy) = 0.048$ bits
- Which attribute should we use for splitting?

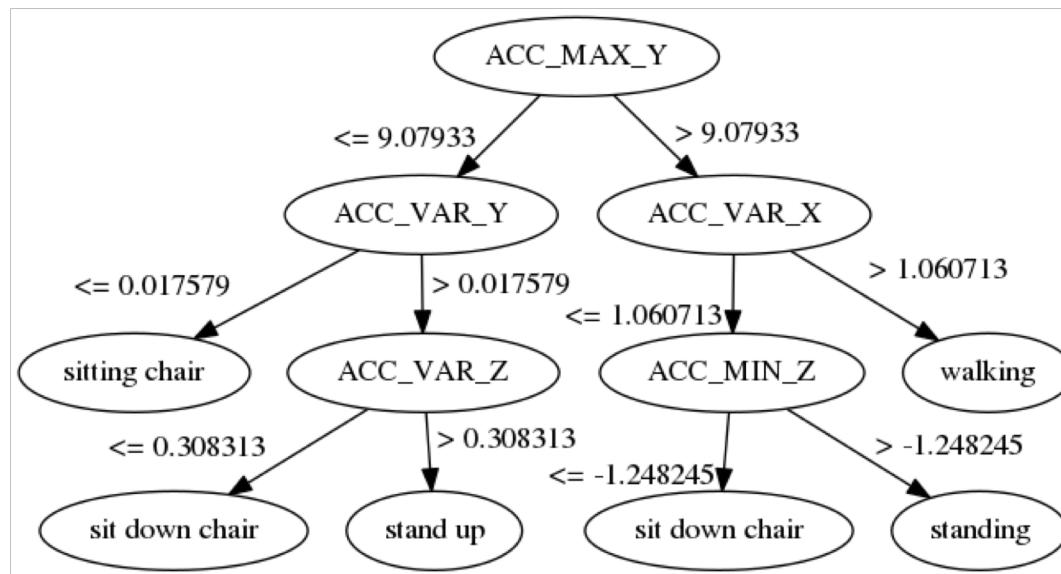
Information Gain



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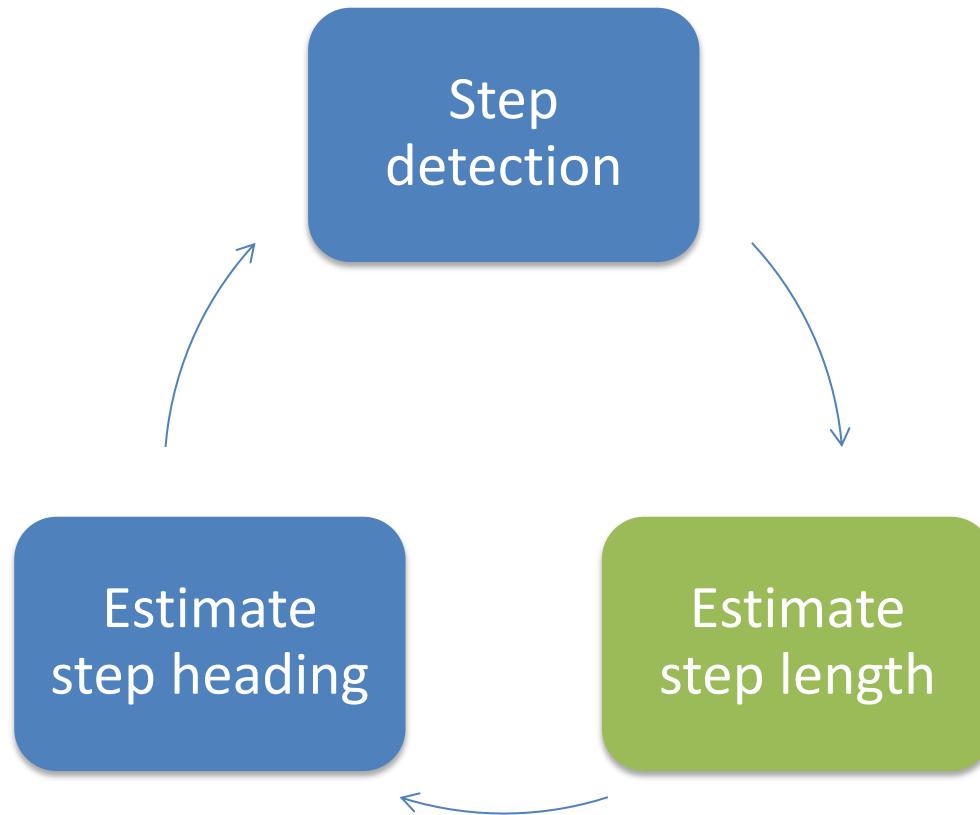
Step-Heading-Systems

Decision Tree for estimating walking activities



Step-Heading-Systems (SHS)

- Basic cycle of SHS-Systems



Step-Heading-Systems (SHS)

- Step length estimation:
 - Assume that step length is fixed ☺
 - Estimation based on vertical hip displacement
 - Using a function based on step frequency and walking speed
 - Direct measurements using ultra sonic sound

Step-Heading-Systems (SHS)

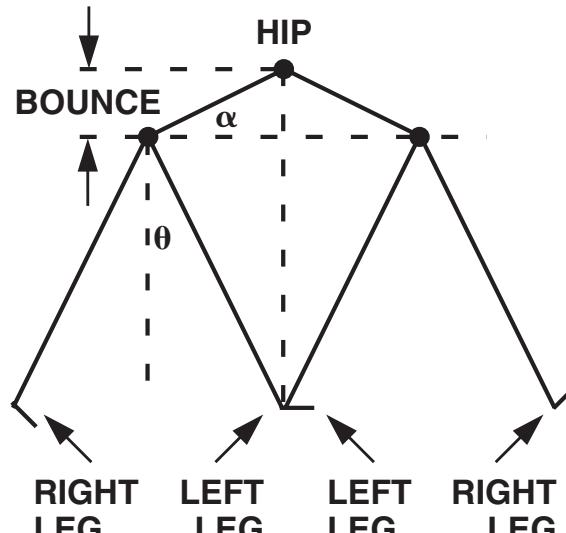


- Assume that step length is fixed:
 - Statistically estimate step length
 - For example: Stride length of runners [11]
 - Roughly 220 cm for men
 - Roughly 210 cm for women
 - However stride length varies as much as +40% to -40% [12]
 - Depends also on leg length

Step-Heading-Systems (SHS)

- Estimation based on vertical hip displacement

$$\text{Stride} \approx \frac{2 \times \text{Bounce}}{\alpha}$$



[12]

$$\text{Distance} \approx \sqrt[4]{A_{\max} - A_{\min}} \times n \times K$$

Step-Heading-Systems (SHS)

- Function based on step frequency and height
 - Linear relationship between step length and step frequency
 - Step length is proportional to the length of a user's leg and to the user's height
 - For example given by:

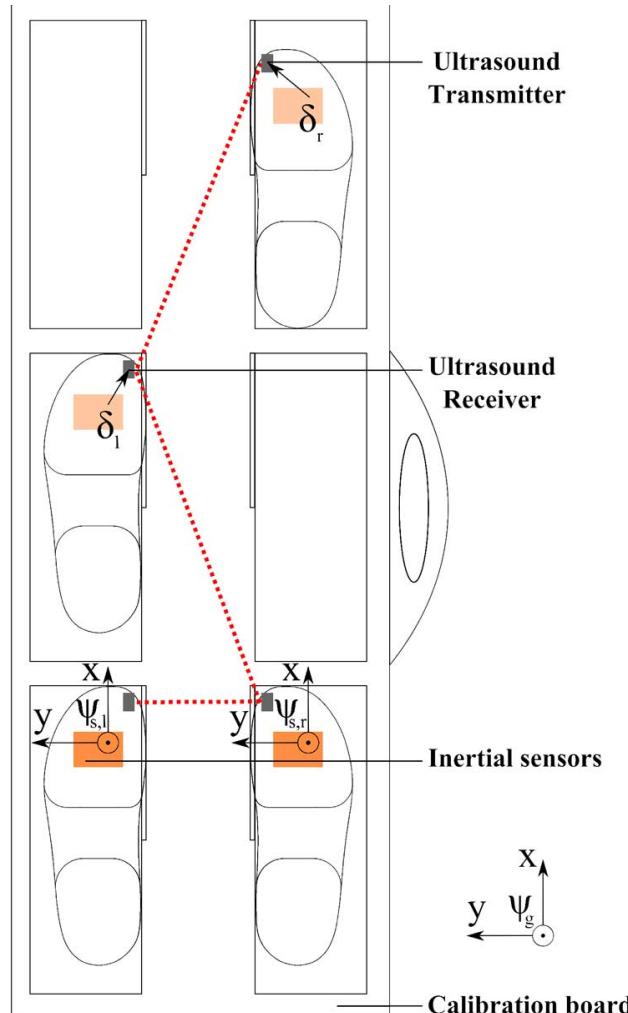
$$s = h \cdot (a \cdot f_{step} + b) + c$$
$$K = \{a, b, c\} \in \mathbb{R} \quad [13]$$

Step-Heading-Systems (SHS)



- Direct measurements using ultra sonic sound
 - For example: Fusing ultrasound estimates and inertial sensors
 - Estimates errors range between 1.7cm (+- 1.8cm) for step length and 1.2cm (+-1.2cm) for stride length

Step-Heading-Systems



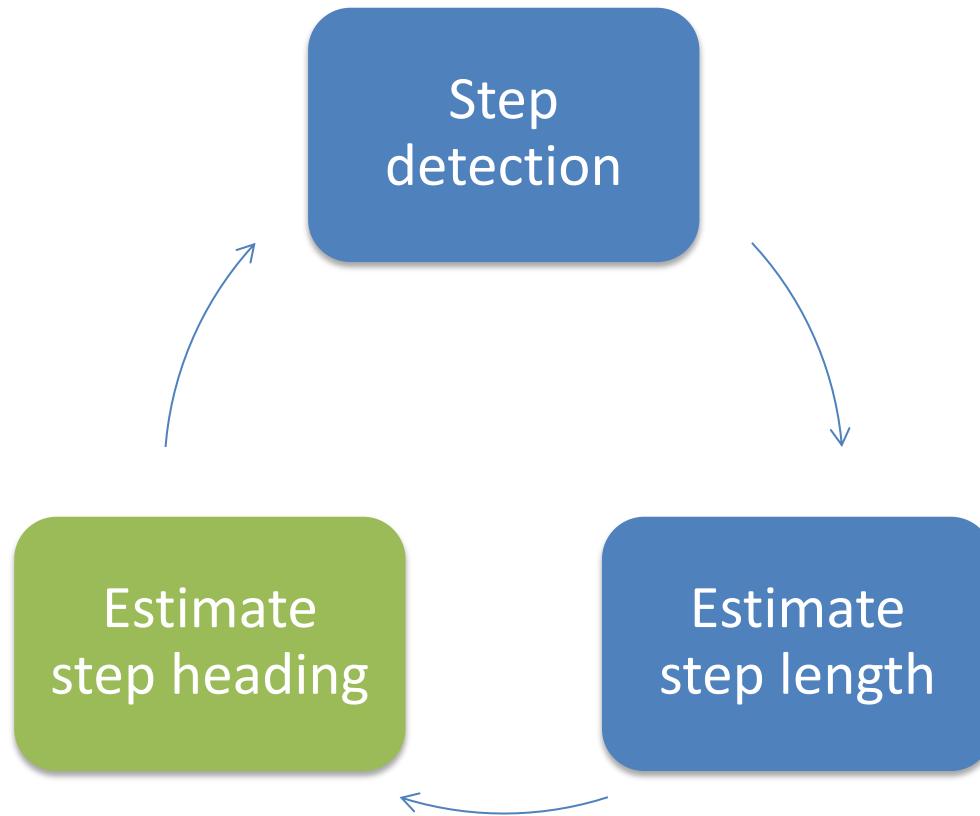
$$d_U = v_s \cdot t_{ToF}.$$

$$v_s = 331.4 \cdot (1 + 1.83 \cdot 10^{-3} \cdot T_C)$$

[14]

Step-Heading-Systems (SHS)

- Basic cycle of SHS-Systems

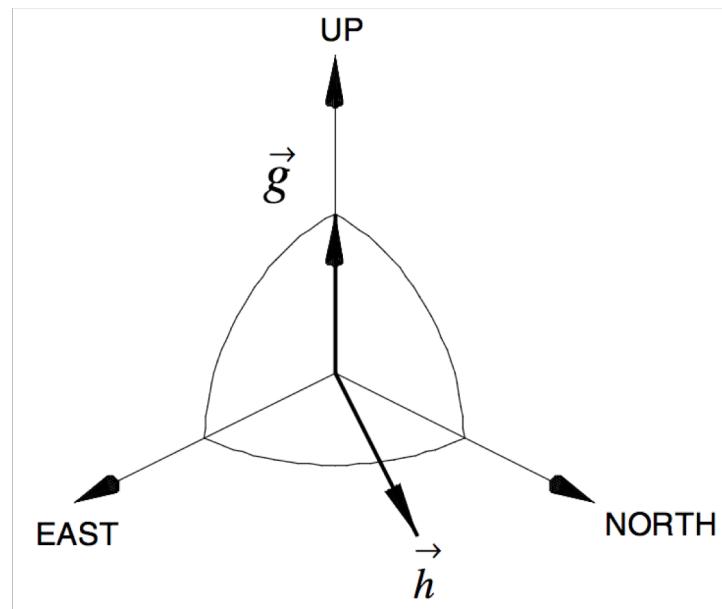


Step-Heading-Systems (SHS)



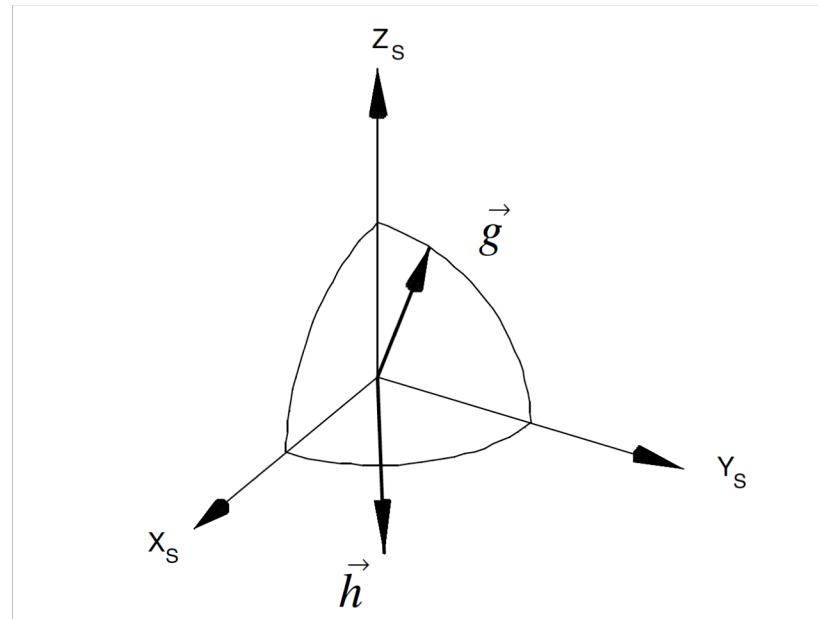
- Estimate step heading:
 - Using gyroscopes and/or magnetometers
- Using magnetometers work when:
 - There is no or little magnetic interferences
 - Which is not often the case in buildings
- Integrating over gyroscope values
 - Gives the angular position
 - Drifts over time
- Combination of both to reduce drift

Estimate Step-Heading



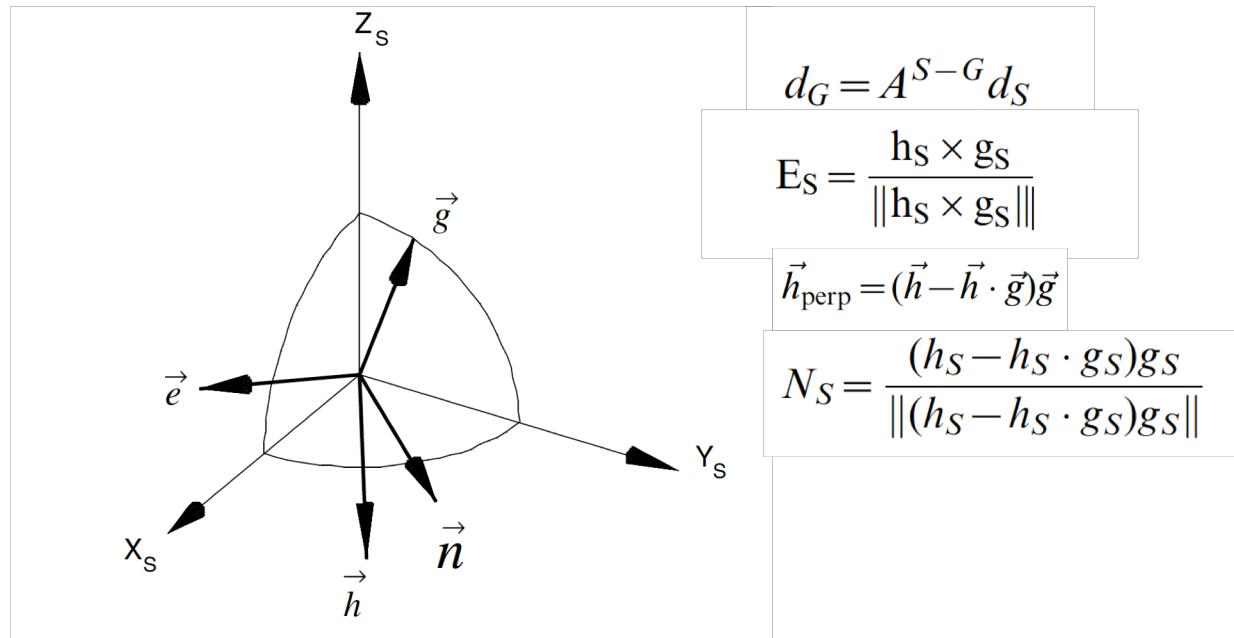
[17]

Estimate Step-Heading



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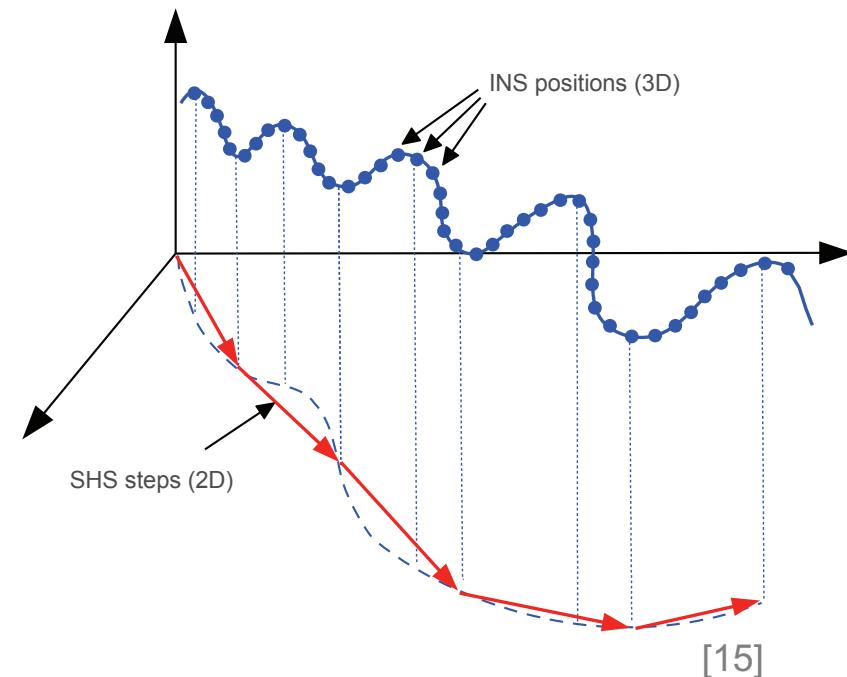
Estimate Step-Heading



[17]

Inertial-Navigation-Systems (INS)

- INS-Systems using 3D-Vectors for navigation
 - Systems that need information on height



[15]

Maps

DEAD-RECKONING

Maps

- What about navigation?
 - Using maps for navigation purposes
 - Geo-referenced maps
- Using different vector layers for objects
 - Doors
 - Walls

Maps



- Attributes

```
WKT;id;room_id_0;room_id_1;blank  
"LINESTRING (9.47288763796637 51.311221854868627,9.472886471859514 51.31121347227073)";-21;0;1;  
"LINESTRING (9.472885305752657 51.311204542979972,9.472884139645801 51.311195249226529)";-22;0;3;  
"LINESTRING (9.472861692088829 51.31118449762706,9.472874519264241 51.311183950935487)";-23;0;13;  
"LINESTRING (9.472853529340839 51.311179212941639,9.47285148865384 51.311162812189934)";-25;19;13;  
"LINESTRING (9.47291795674462 51.311181764169163,9.472931366973462 51.311181217477561)";-26;3;13;
```

Errors and Pitfalls

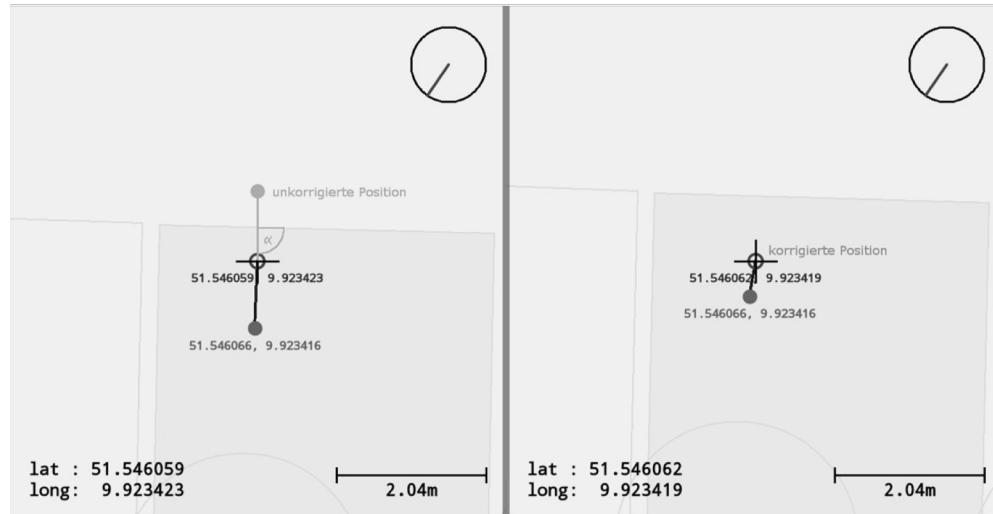
DEAD-RECKONING

Errors and Pitfalls

- What about noise and inaccuracies?
 - Inaccurate step detection, length and heading estimation will cause the PDR-System to fail
 - Basically only one „subsystem“ needs to fail

Errors and Pitfalls

- Inaccurate step detection
 - User is moving but the system is not recognizing it
 - System will count more/less steps than actually done
- Step detection and length estimation errors:



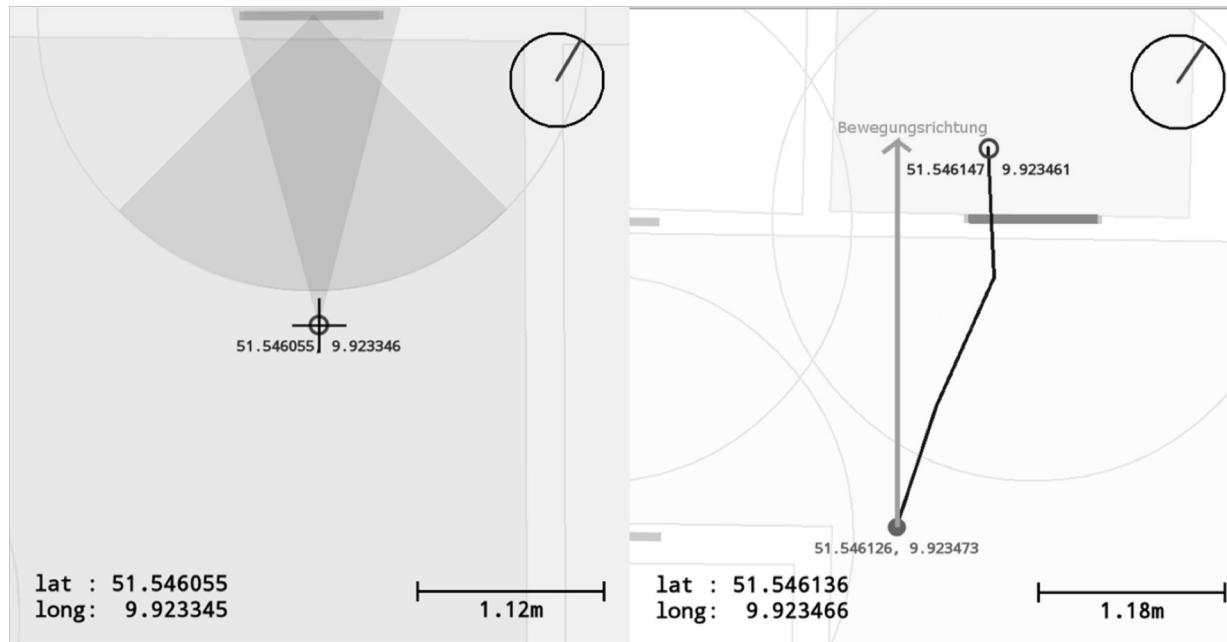
Errors and Pitfalls

- Step heading estimation errors



Errors and Pitfalls

- Adjacent rooms and doors



References

- [1] <https://de.wikipedia.org/wiki/Containerschiff>
- [2] <https://de.wikipedia.org/wiki/Segelschiff>
- [3] https://en.wikipedia.org/wiki/Chip_log
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- [17] R. Stirling, K. Fyfe, and G. Lachapelle, "Evaluation of a New Method of Heading Estimation for Pedestrian Dead Reckoning Using Shoe Mounted Sensors," *The Journal of Navigation*, vol. 58, no. 1, pp. 31–45, Jan. 2005.