

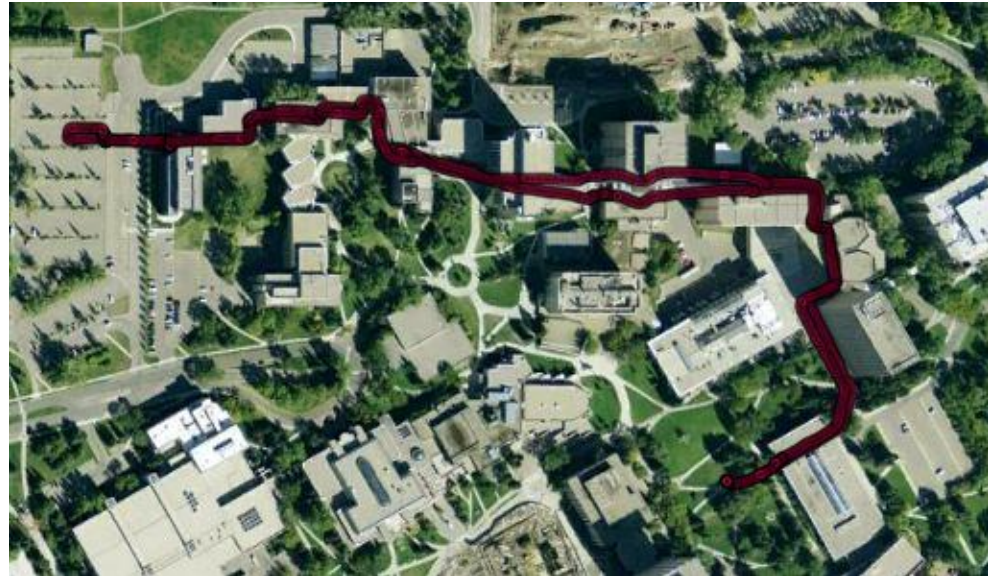
# Visual Gyroscope and Odometer for Pedestrian Indoor Navigation with a Smartphone

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# Overview

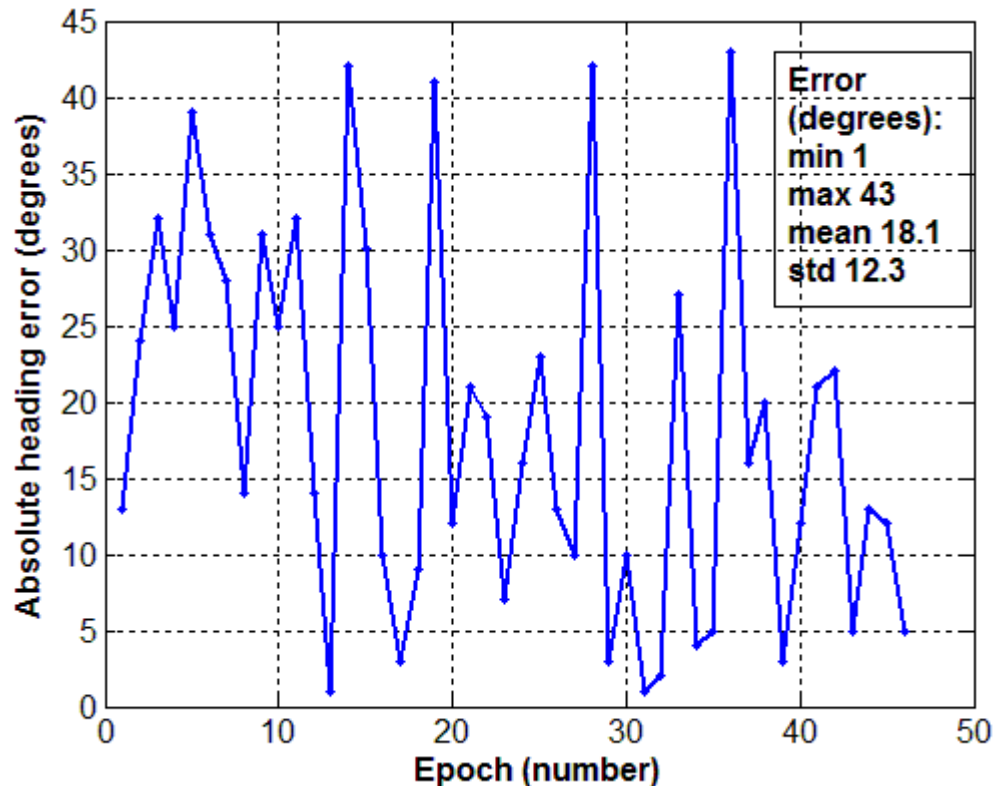
- Estimating the user attitude and translation by tracking features in consecutive images in the indoors
- Use of a camera carried by the user
- Attitude & translation integrated with other measurements  
=> vision-aided navigation solution



# Motivation

- Performance of GNSS is degraded indoors
- INS is feasible for aiding, problems especially with heading gyro drift and compass errors indoors
- Need for a system free from calibration and prior information of the environment and that is convenient for users

Heading errors with a Nokia 6710 smartphone compass in an office environment



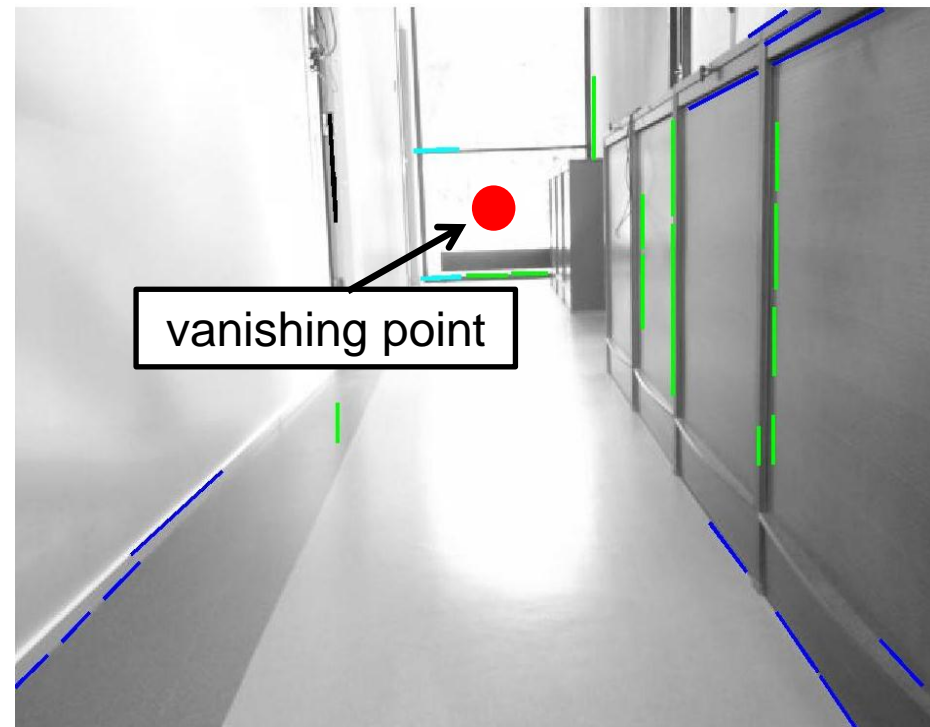
# Visual gyroscope 1/3

- Human made environments are full of straight orthogonal lines
- Vanishing point is a point where **parallel lines** seem to **intersect** in an image due to perspective projection, mapping 3D environment into 2D features
- Lines in three orthogonal directions => three **vanishing points**
- Locations of vanishing points are related to camera intrinsic parameters (matrix **K**) and rotation of the camera (matrix **R**)

$$[\mathbf{v}_x \ \mathbf{v}_y \ \mathbf{v}_z] = \mathbf{KR}$$

# Visual gyroscope 2/3

- Lines are classified based on their orientation
- Vanishing point  $\mathbf{v}_z$  is the intersection point of lines in the direction of propagation (blue lines) and is found with a voting algorithm
- Reliability of the estimated vanishing point is measured based on the line geometry



# Visual gyroscope 3/3

- Heading ( $\theta$ ) and tilt ( $\varphi$ ) of the camera are obtained using the pixel coordinates of the  $\mathbf{v}_z$  and  $\mathbf{K}$  matrices

$$\mathbf{v}_z = \begin{bmatrix} f_x \sin\theta + u \cos\varphi \cos\theta \\ -f_y \sin\varphi \cos\theta + v \cos\varphi \cos\theta \\ \cos\varphi \cos\theta \end{bmatrix} \quad \mathbf{K} = \begin{bmatrix} f_x & 0 & u \\ 0 & f_y & v \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{R} = \begin{bmatrix} \cos\theta & 0 & \sin\theta \\ \sin\varphi \sin\theta & \cos\varphi & -\sin\varphi \cos\theta \\ -\cos\varphi \sin\theta & \sin\varphi & \cos\varphi \cos\theta \end{bmatrix}$$

$$[\mathbf{v}_x \ \mathbf{v}_y \ \mathbf{v}_z] = \mathbf{KR}$$

# Strengths of visual gyroscope

- No prior information is needed for the evaluation of the heading change and tilt of the camera
- Dynamic objects are not significantly disturbing the process
- Is dependent only on lighting and geometry of the scene
- Sample statistics with 7555 images (1.2 s interval) and static camera (Nokia N8, image resolution 640 x 480):

Statistics	min error (degrees)	max error (degrees)	mean error (degrees)	Std of error (degrees)
Heading change	0	18.4	0.8	0.59
Tilt	0	10.7	0.3	0.28



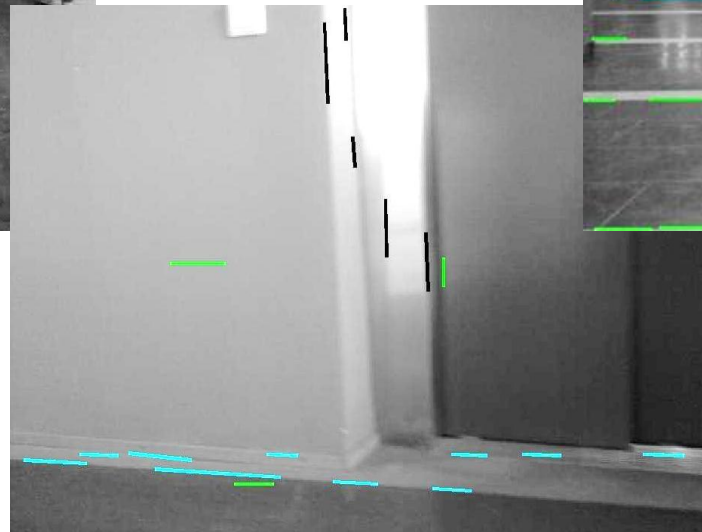
# Challenges of visual gyroscope



**Dark image:**  
vanishing point  
erroneous

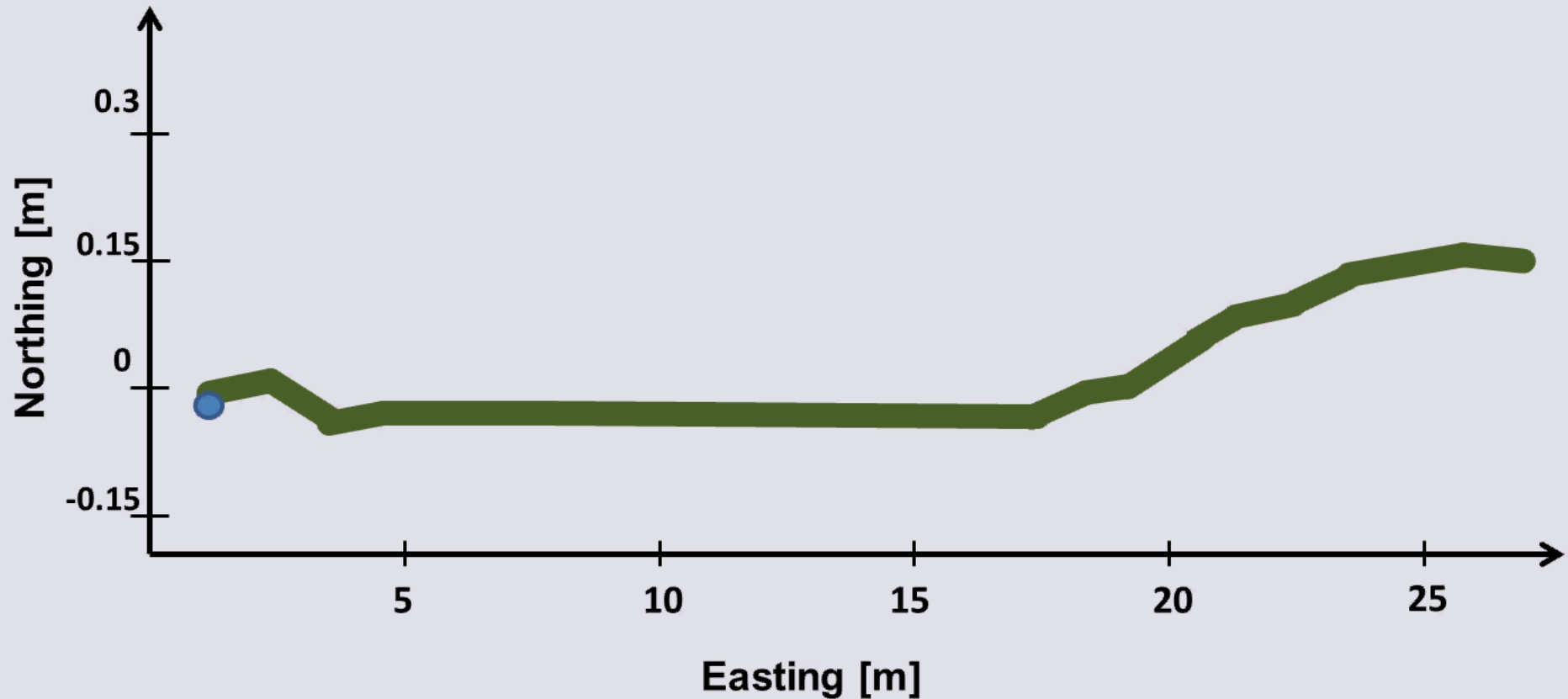


**Scene geometry  
and parameter  
selection:**  
rate of success of  
finding correct  
vanishing point  
lower



**Turns – no lines in  
the direction of  
propagation:**  
no vanishing point

Visual gyroscope's error detection discards measurements with poor line geometry and uses the subsequent good one



Reference heading, constant speed

Visual gyroscope heading, constant speed

# Visual odometer 1/4

- Homography between image points  $\mathbf{x}$  and  $\mathbf{x}'$  of same world point defined with

$$\mathbf{x}' = \mathbf{K}'\mathbf{R}\mathbf{K}^{-1}\mathbf{x} + \mathbf{K}'\mathbf{t}/Z$$

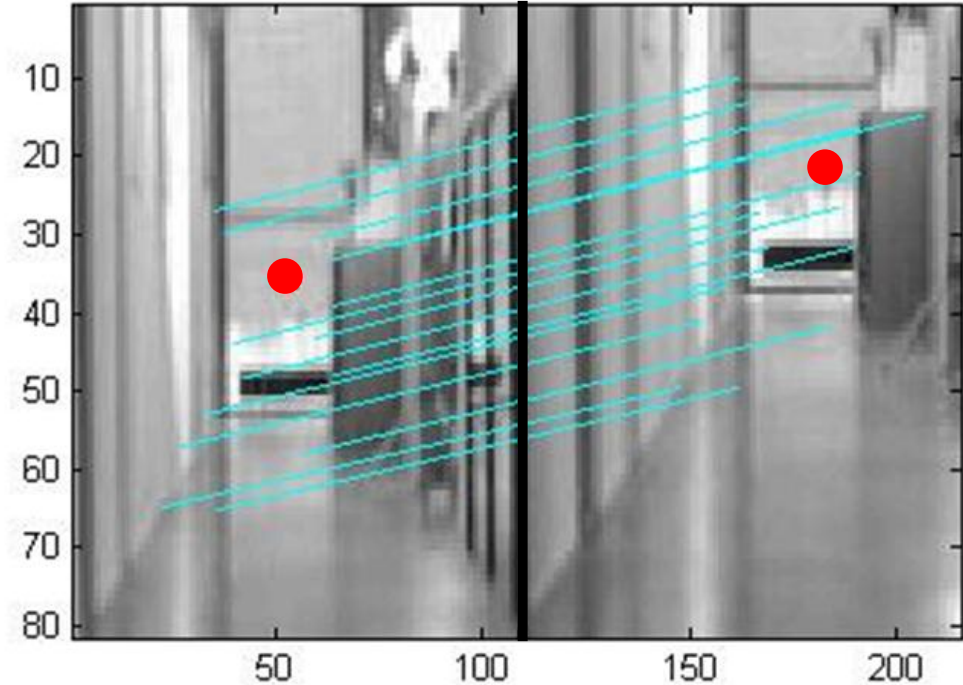
$\mathbf{K}$  and  $\mathbf{K}'$  = calibration matrices

$\mathbf{R}$  rotation of the camera

$\mathbf{t}$  translation of the camera

$Z$  distance to point

- Problems:
  - the unknown  $Z$
  - ambiguous scale



Matching of points with SIFT algorithm by  
D.G. Lowe, US Patent 6,711,293, filed 1999

# Visual odometer 2/4

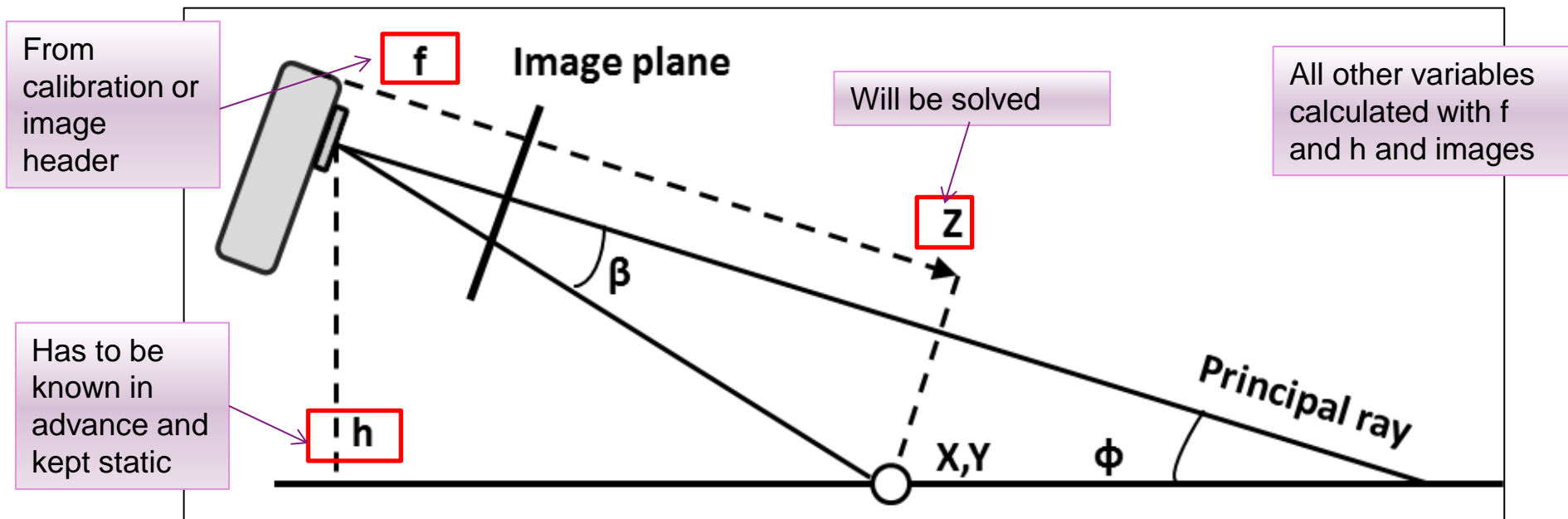
- Some solutions in the literature for finding  $Z$ :
  - Stereo cameras => Known pose of the cameras give  $Z$  by triangulation : **special equipment**
  - Camera facing down with known height: **might be problematic indoors due to uniform texture of the floor**
  - Objects with known size: **needs a priori preparation and is restricted to one area**



Santos et al (2009). An analysis of navigation algorithms for smartphones using J2ME.  
*MOBILWARE.*

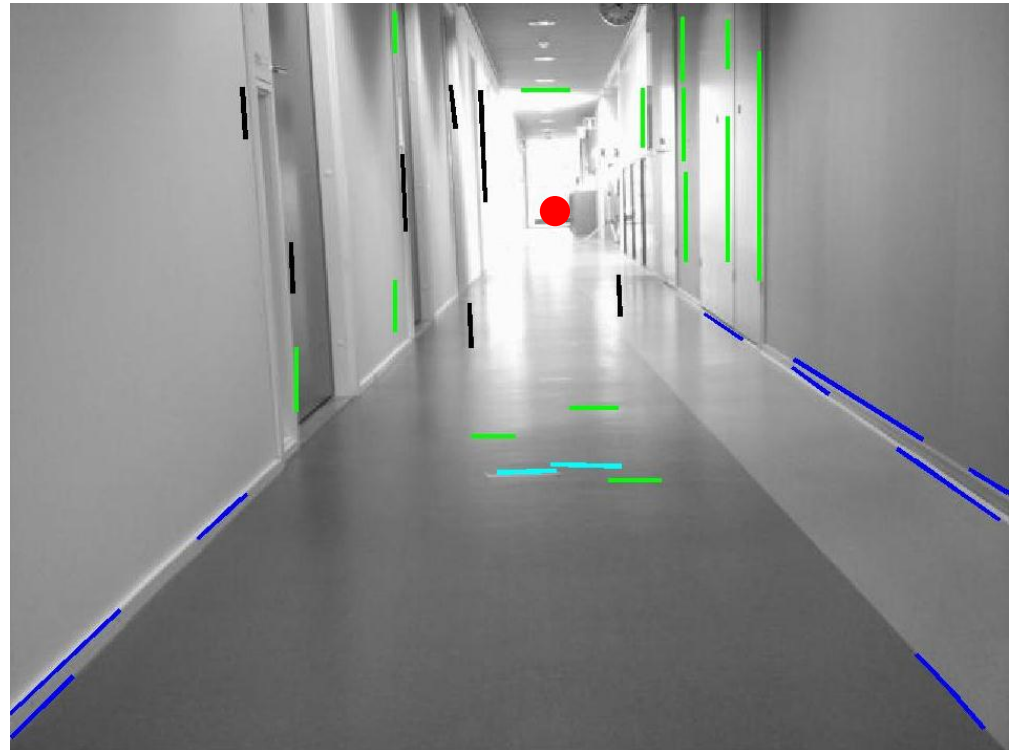
# Visual odometer 3/4

- Calculating  $Z$  with special configuration of the camera
  - Campbell et al. 2005 for robots => tilt evaluated a priori, motion restricted, optical flow for outdoor environment
  - Ruotsalainen 2012 for pedestrians using a **smartphone indoors** / outdoors => **tilt** ( $\phi$ ) and heading **with vanishing point**, translation with image homography, camera may move quite freely



# Visual odometer 4/4

- Matching points in consecutive images: SIFT<sub>1</sub>
- Only few features in indoor environment => uncertain matches must be accepted => wrong matches discarded
- Points have to lie on floor => lines used for calculating vanishing points normally found from the floor => points close to the lines used



1. SIFT patented by D.G. Lowe, US Patent 6,711,293, filed 1999

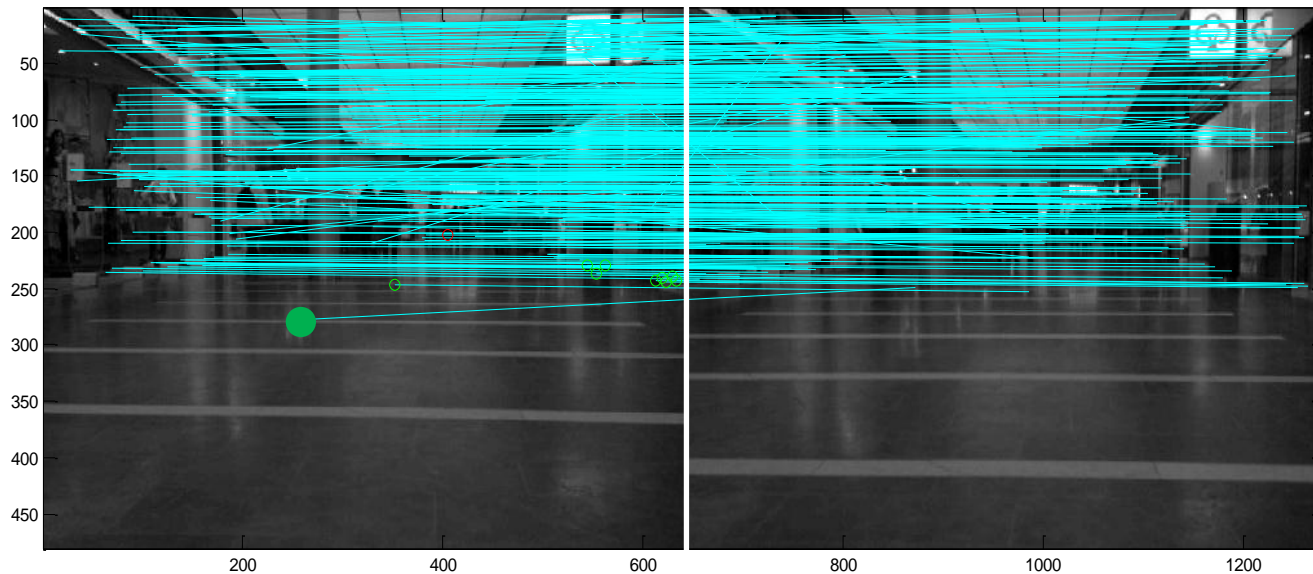
# Strengths of visual odometer

- Rotation of the camera is obtained with vanishing points, the calibration matrix is known and the height of the camera is static to some extent => only translation in x and z- axes is unknown => **this decreases the number of matched points needed for calculation**
- Speed measurements require information only of the height of the camera
- Performance of the visual odometer tested in an office corridor – errors in relation to the Novatel SPAN GPS /INS reference

Statistics Speed	min error (m/s)	max error (m/s)	mean error (m/s)	std of error (m/s)
	0	2.3	0.3	0.3

# Challenges of visual odometer

- Points have to lie close to the camera
- Indoor environments often poor with features => problems finding even one matching point close to the camera
- Effect of vanishing point calculation errors => erroneous vanishing point decreases accuracy of measured speed





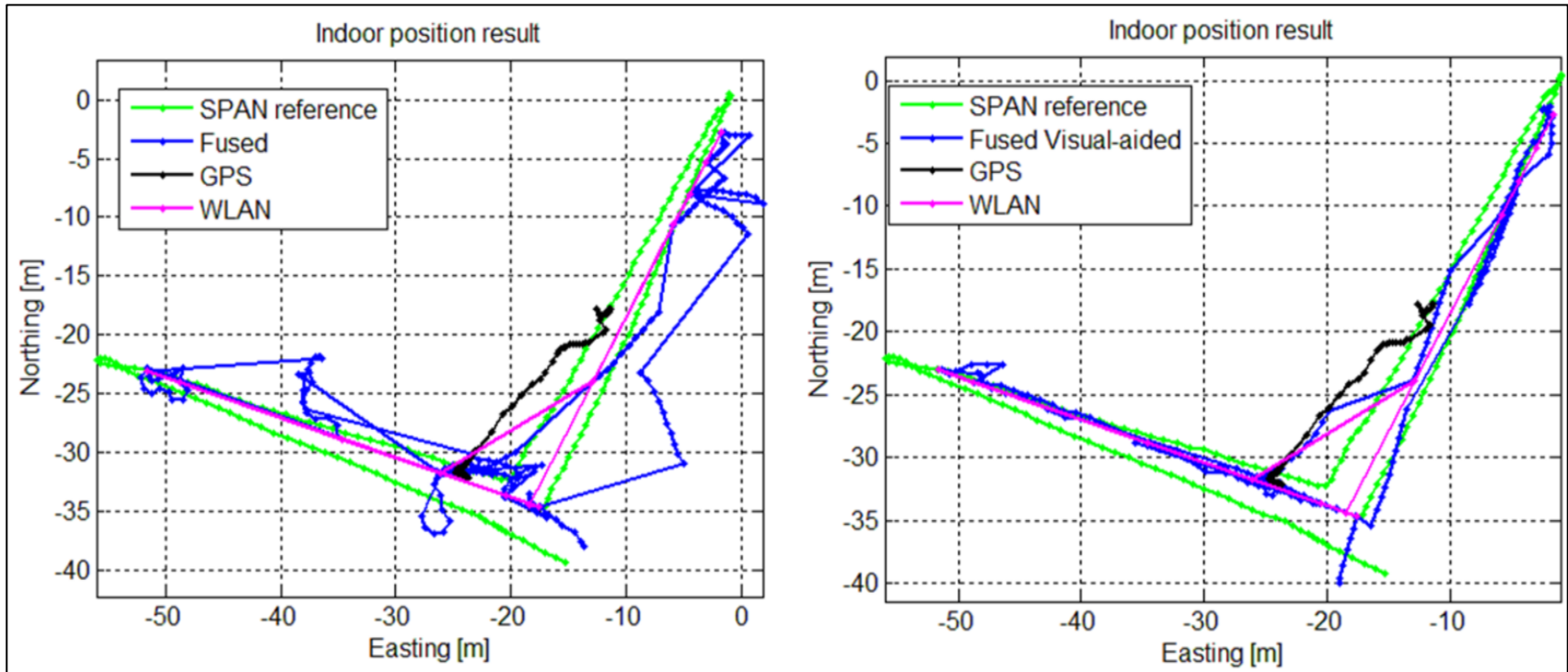
# Experiments



- Fastrax IT500 GPS receiver
- Nokia 6710 mobile phone for WiFi
- Multi-sensor positioning device with  
3-axis accelerometer (VTI) and  
2-axis digital compass (Honeywell)
- Nokia N8 camera for visual-aiding



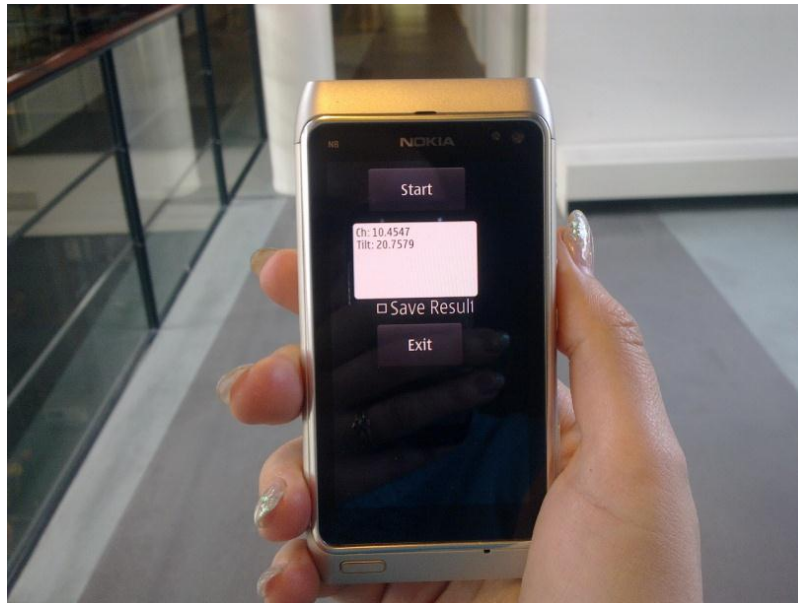
# Visual-aided multi-sensor multi-network navigation solution



	min errors (m)	max errors (m)	mean errors (m)	std of errors (m)
Fused	0.9	23.4	6.7	5.1
Fused with visual aiding	0.9	19.7	5.3	3.7

# Summary

- User heading and speed may be obtained using information from consecutive images
- Accuracy suitable for pedestrian navigation
- Improves integrated navigation solution



# On-going and future work

- Tested in three other different environments
  - Mean error in speed is around 0.25 m/s in all test environments
  - Accuracy of cumulative distance is over 90% in all test environments
- Processing times of the visual gyroscope with Nokia N8
  - Taking a photo : 1.2 s
  - Edge detection (Canny) : 0.17 s, line detection (Hough lines) : 1.0 s
  - Vanishing point and heading & tilt calculations : 0.07 s
  - **Total 2.7 s**
- How does roll and change in camera height affect accuracy?

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