

Advanced Sonar Sensing for Robot Mapping and Localisation

by

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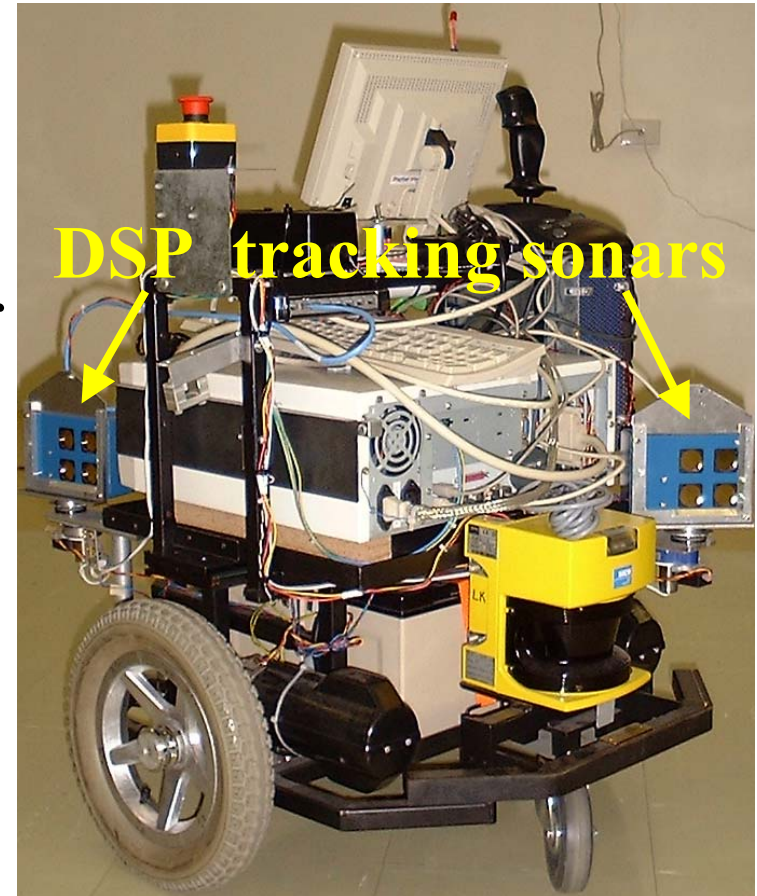
- ◆ Introduction
- ◆ DSP Sonar Sensor
- ◆ Tracking Experiments
- ◆ Interference Rejection and Pulse Coding
- ◆ Classification
- ◆ SLAM
- ◆ Conclusions and Future Work

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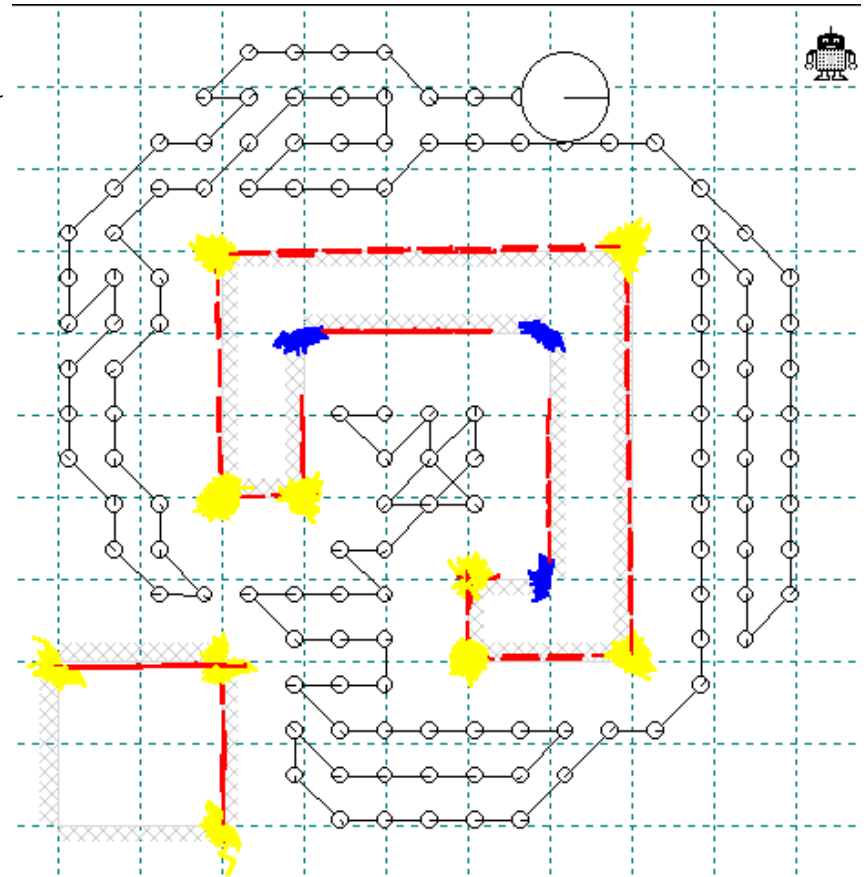
Introduction - Mobile Robot Applications

- ◆ **Courier** - offices, hospitals, factories
- ◆ **Area coverage** - cleaning, hazardous chemical/nuclear decontamination, painting
- ◆ **Security and surveillance**
- ◆ **Disability Aids**



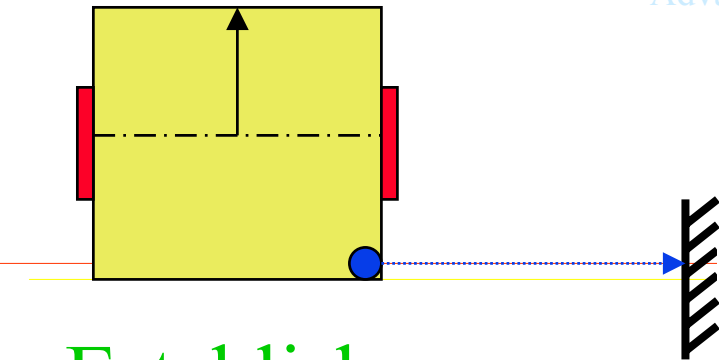
Introduction - Mobility Tasks

- ◆ Localisation (position and orientation)
- ◆ Path planning
- ◆ Obstacle avoidance
- ◆ Map building and maintenance

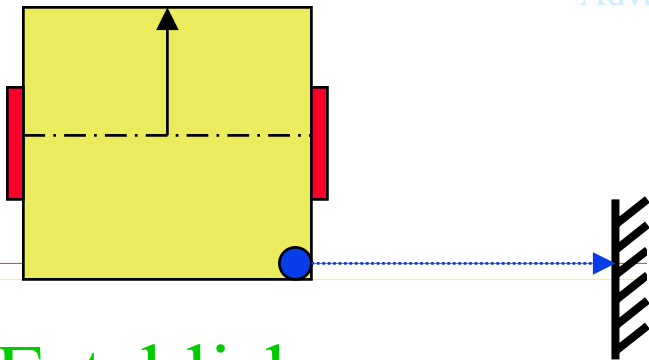


Introduction

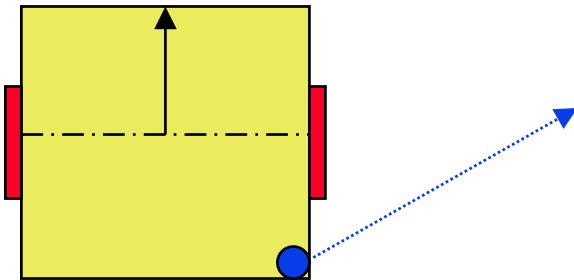
- ◆ **Why is sonar classification important?**
 - Allows prediction of feature measurements from new positions.
 - Localisation w.r.t. features.
 - Assists association of measurements with map features.



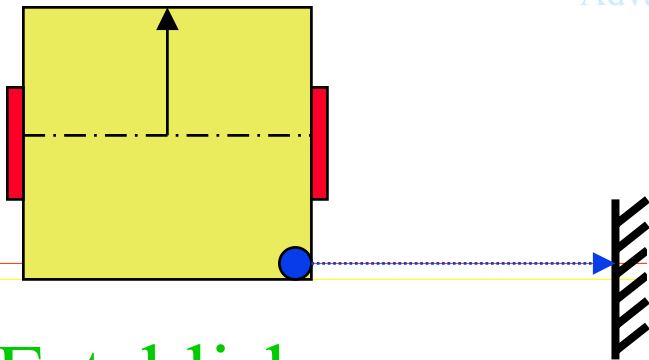
Establish new
plane feature



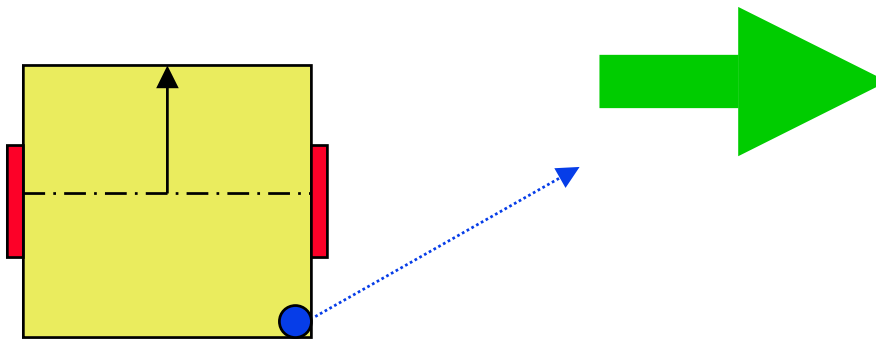
Establish new
plane feature



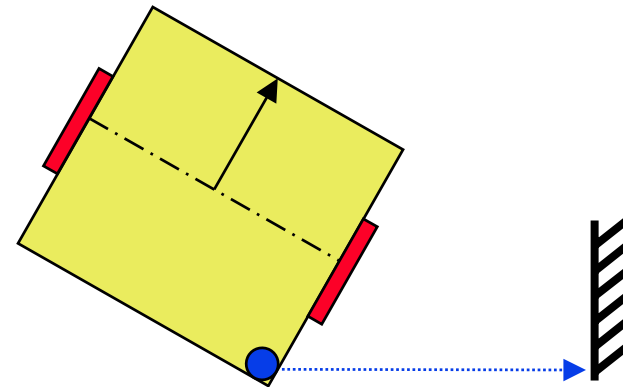
Associate new
measurement to
plane feature



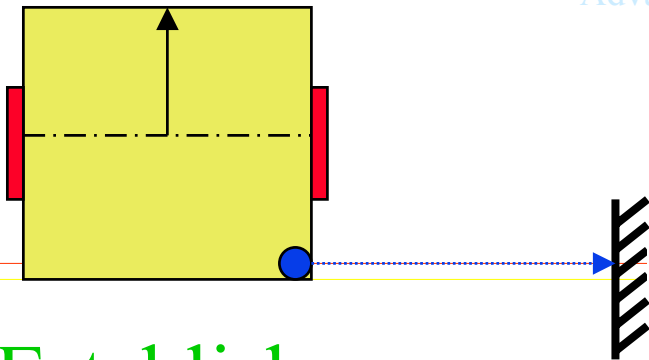
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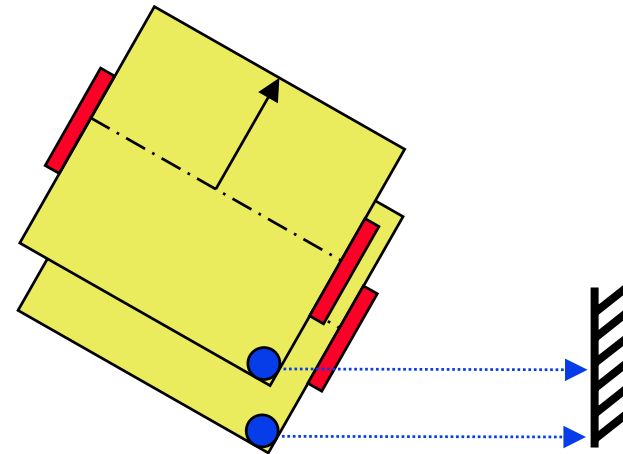
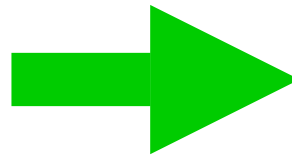
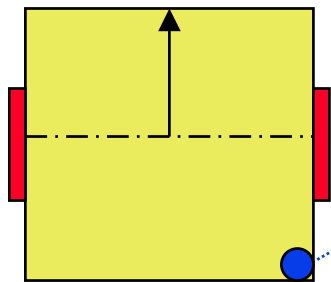
Associate new
measurement to
plane feature



Constrains robot
position and
orientation

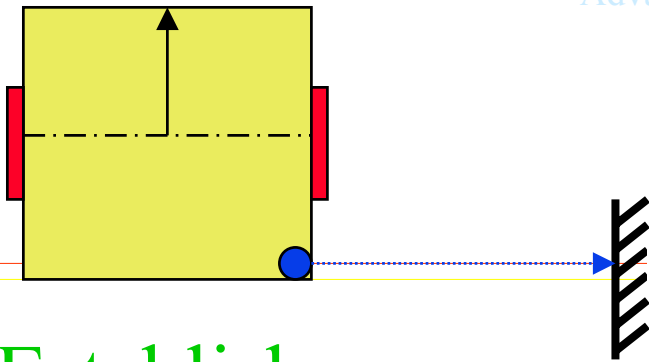


Establish new
plane feature

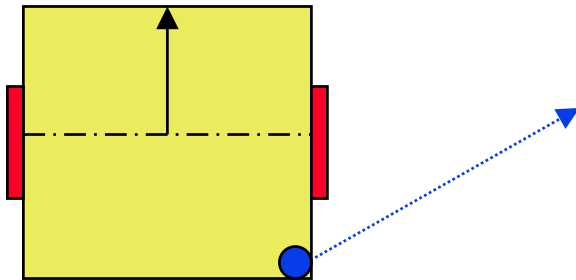


Associate new
measurement to
plane feature

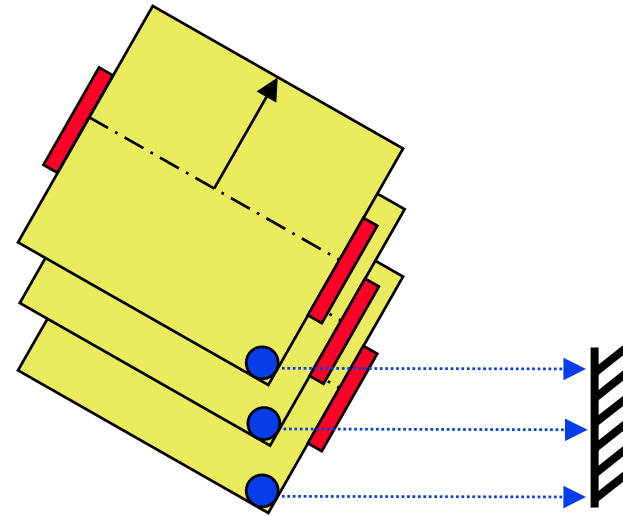
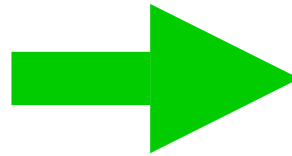
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position and
orientation



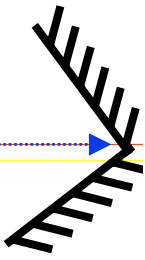
Establish new
plane feature



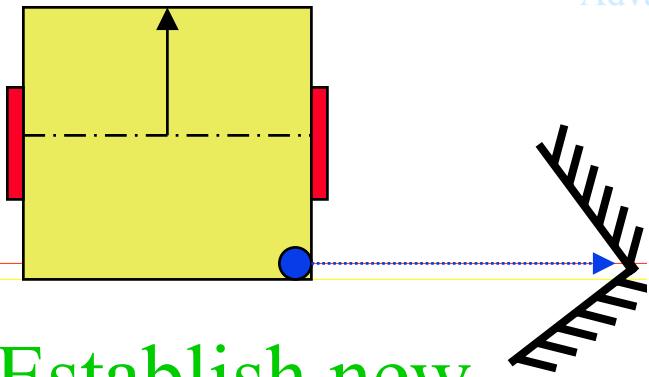
Associate new
measurement to
plane feature



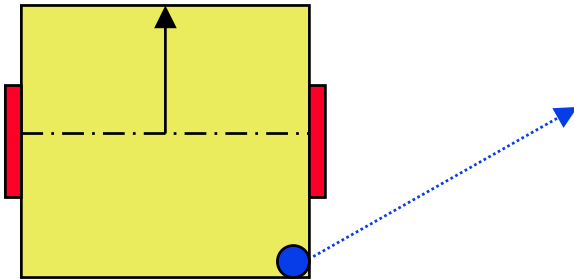
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orientation



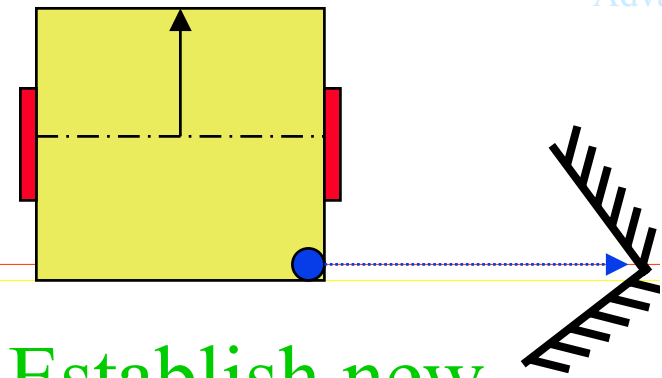
Establish new
corner feature



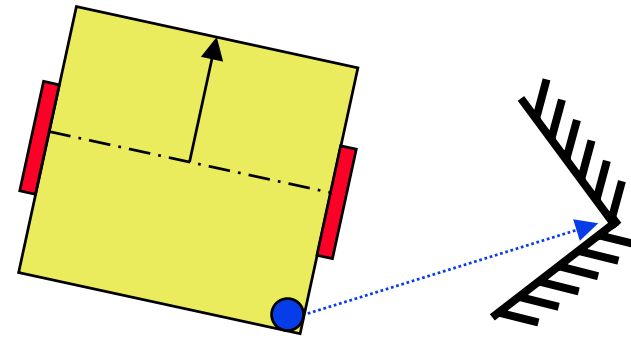
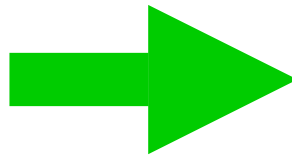
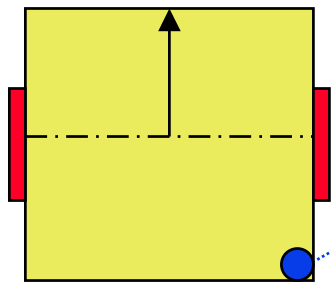
Establish new
corner feature



Associate new
measurement to
corner feature

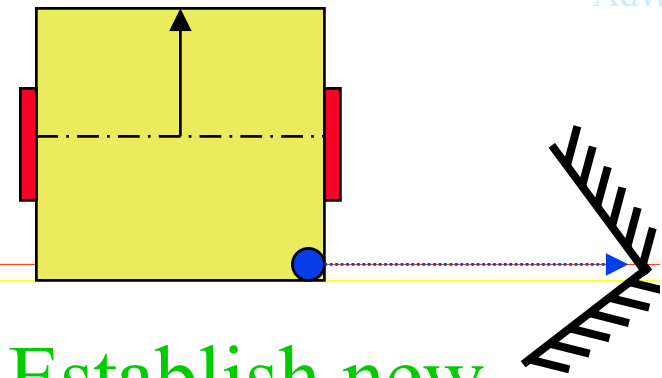


Establish new
corner feature

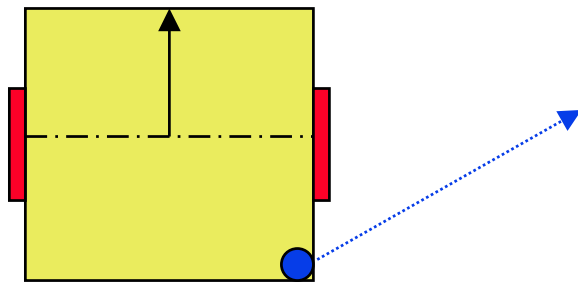


Associate new
measurement to
corner feature

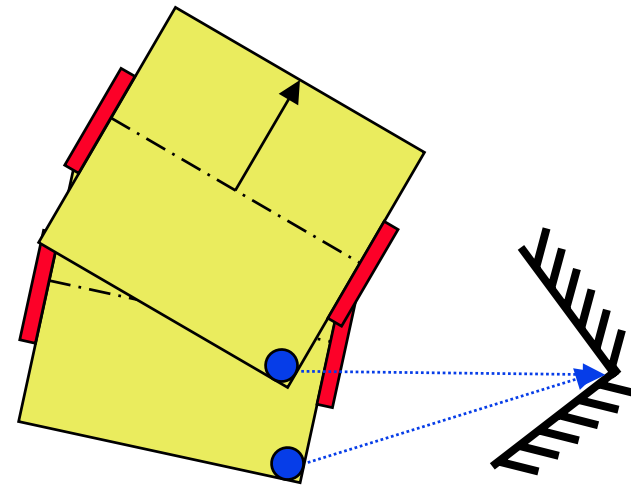
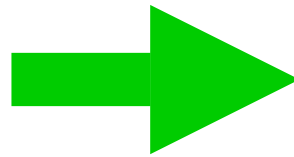
Constrains robot
position and
orientation



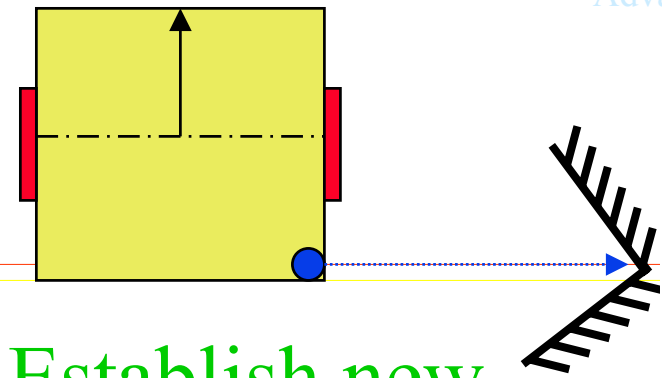
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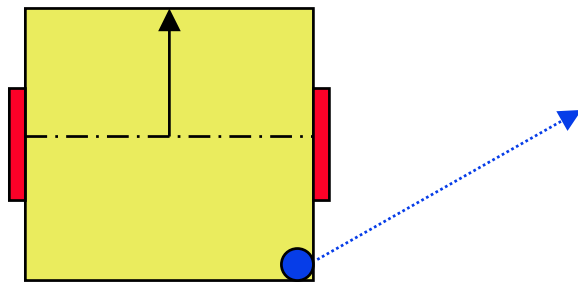
Associate new
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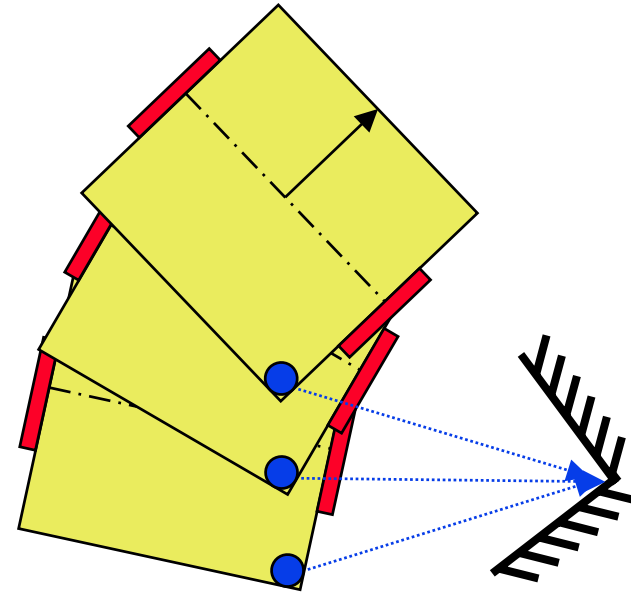
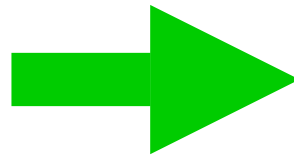
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Introduction

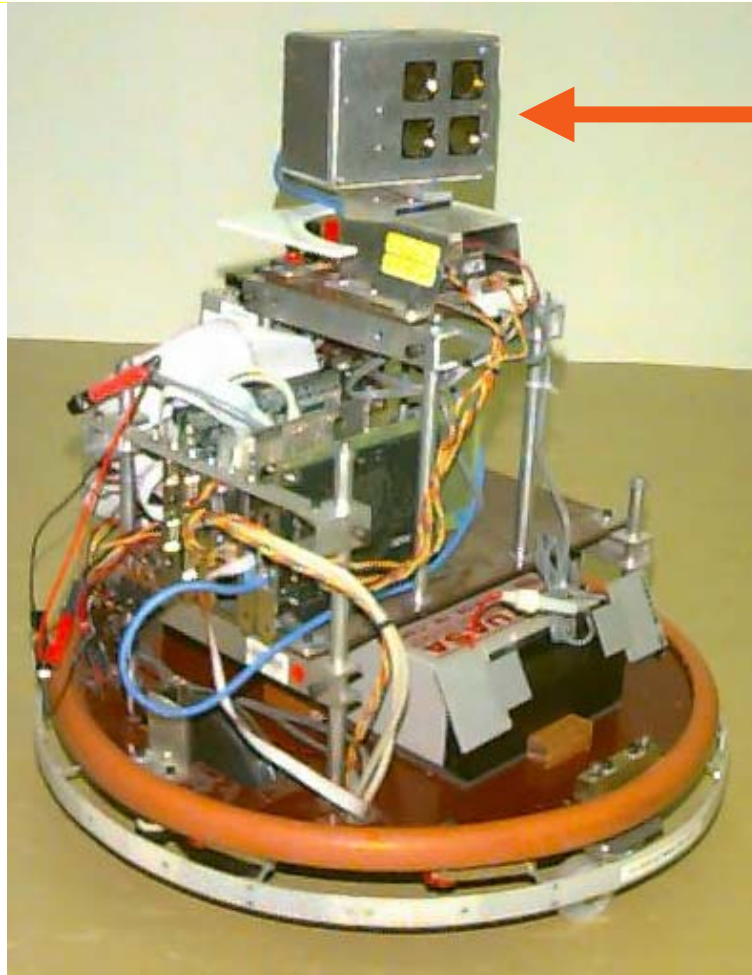
- ◆ **Early sonar classifies with a stationary robot:**
 - Kleeman & Kuc 1995, “Mobile robot sonar for target localisation and classification” IJRR.
 - Peremans, Audenaert & Campenhout 1993, “A high resolution sensor based on tri-aural perception” IEEE R&A
- ◆ **Other classifier needs several different positions**
 - Feder, Leonard & Smith 1999, “Adaptive mobile robot navigation and mapping”, IJRR.
- ◆ **On-the-fly single measurement classification:**
 - L. Kleeman, “On-the-fly classifying sonar with accurate range and bearing estimation” IEEE/RSJ International Conference on Intelligent Robots and Systems, 2002, Lausanne, Switzerland, pp.178-183.

Introduction

- Sonar and natural selection

- ◆ Sonar useful for *naturally* selecting navigation beacons from the environment

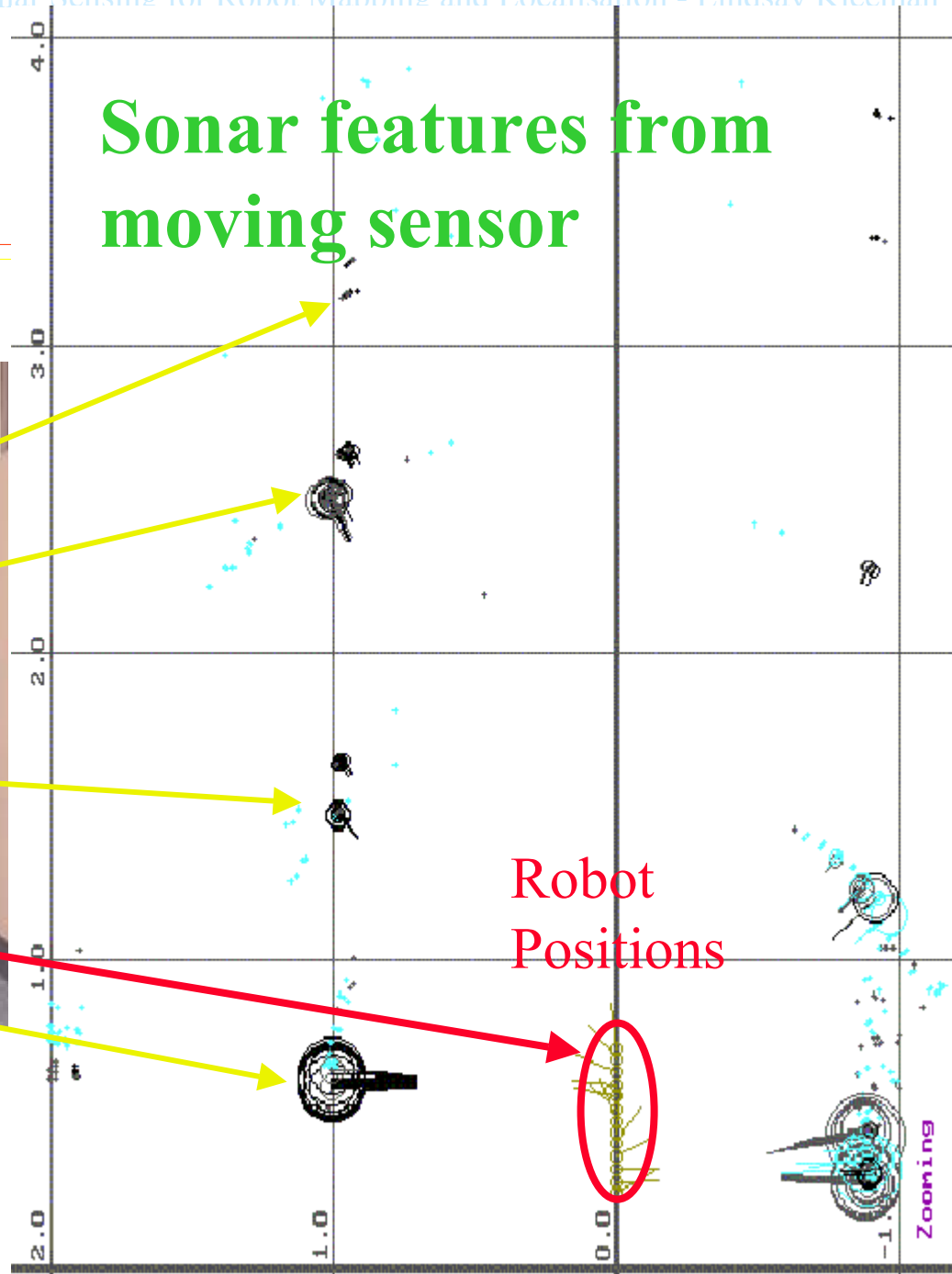
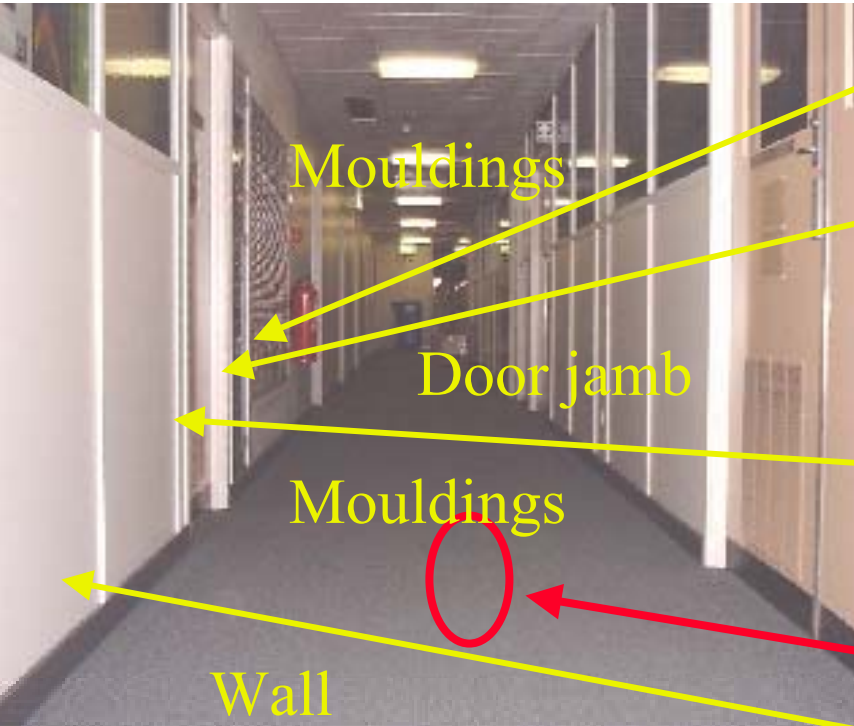
DSP Sonar System on Werrimbi



← sonar



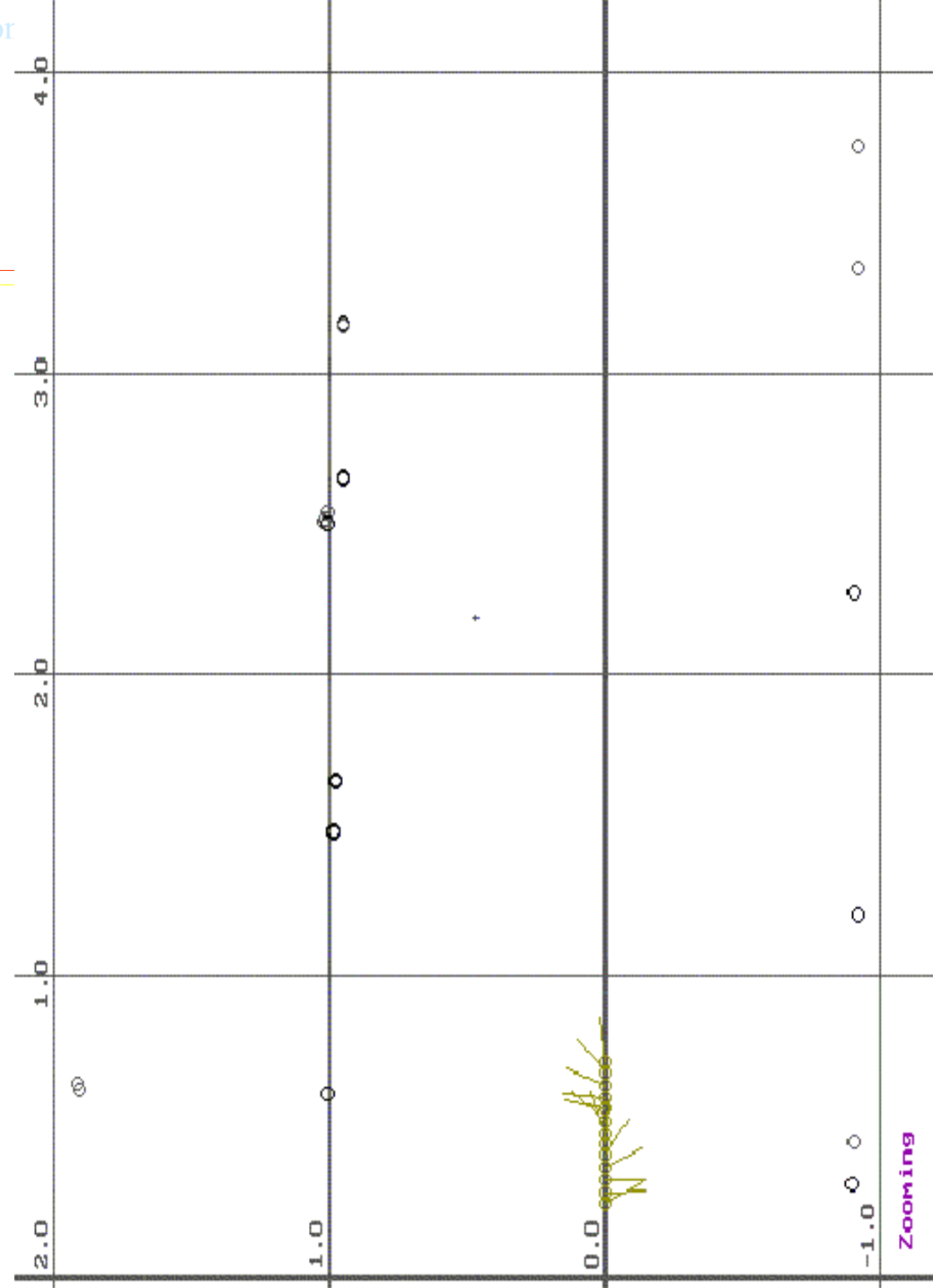
Sonar Features



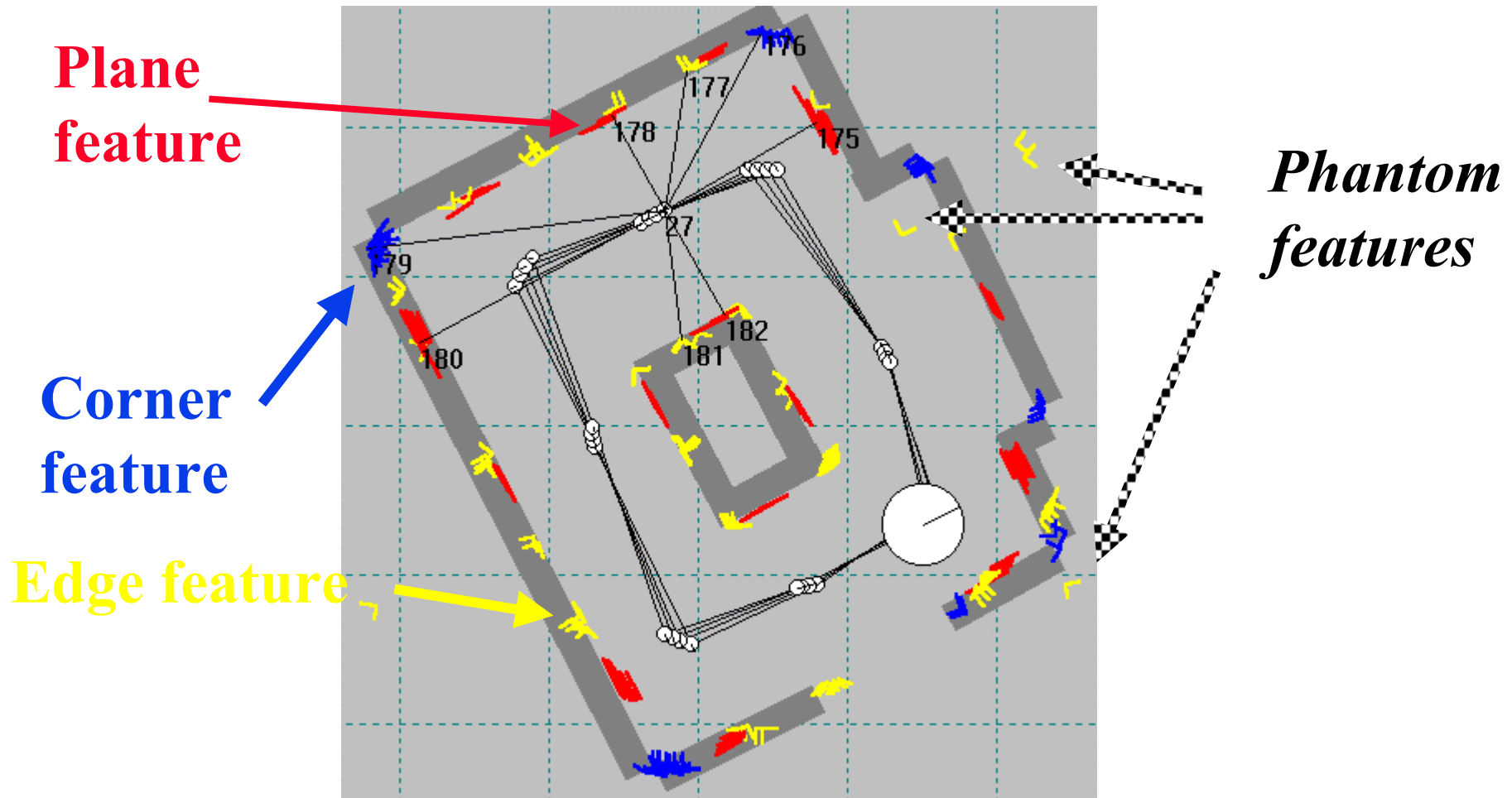
Fused Targets

Raw target data in close proximity is fused based on a weighted average.

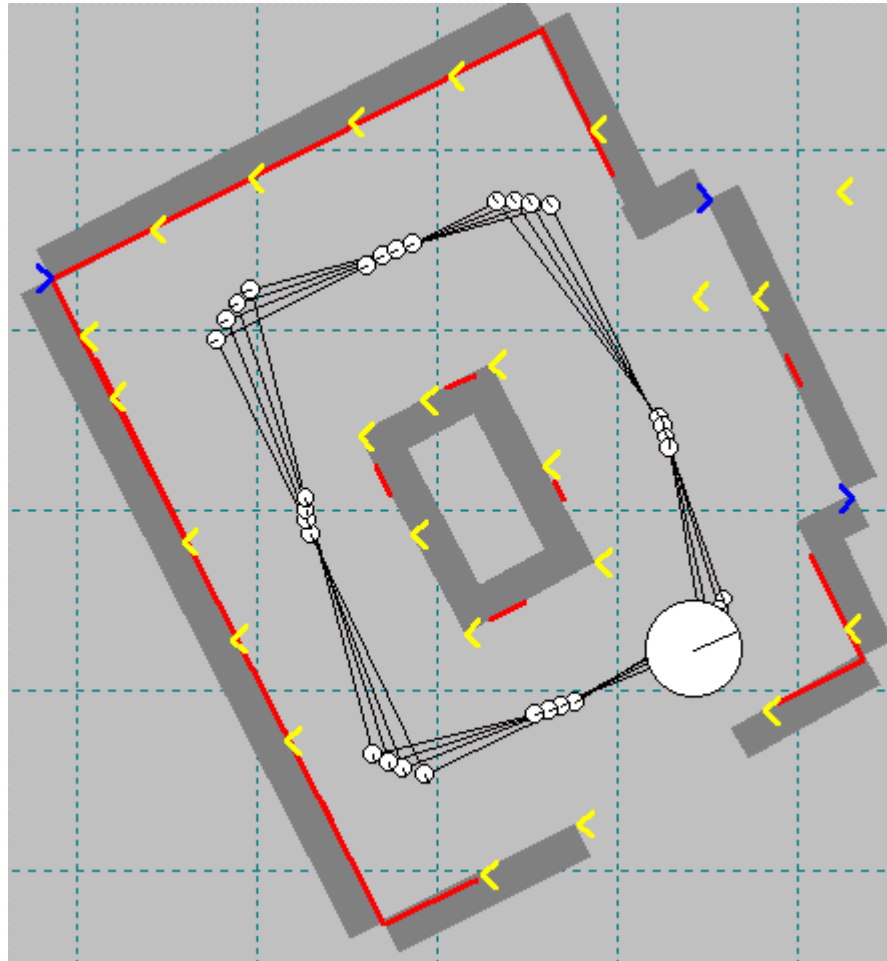
Weights are determined by amplitude and correlation values.



Natural sonar features

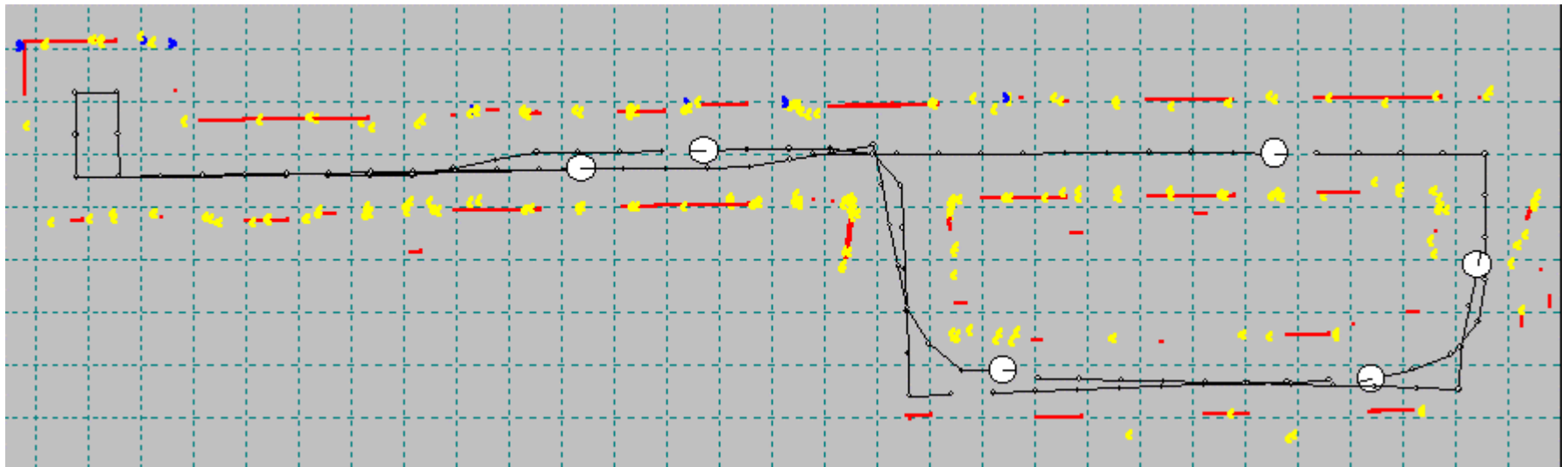


Simultaneous Mapping and Localisation (SLAM) Kalman Filter:



Sonar SLAM with loops

- move and stop measurements



30 metres

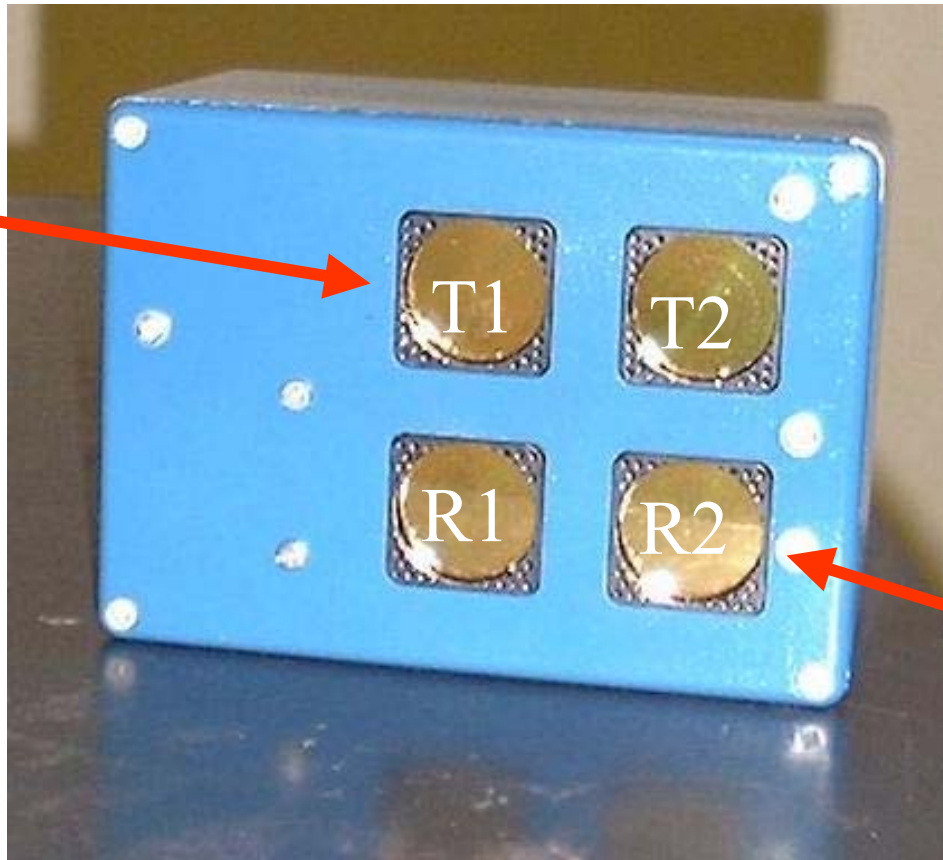
K S Chong and L. Kleeman, "Feature-based mapping in real, large scale environments using an ultrasonic array", International Journal Robotics Research, Vol 18, No. 1, 1999, pp. 3-19.

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- ◆ Tracking Experiments
- ◆ Interference Rejection and Pulse Coding
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DSP Sonar Sensor

**transmitters
T1, T2**



receivers R1, R2.

Capabilities

- ◆ Echoes are obtained from targets types *planes, corners* and *edges*.
 - These are common natural landmarks.
- ◆ Use two receivers and triangulate to determine the bearing to a target.
- ◆ Use two transmitters to determine the type of a target.

Specification

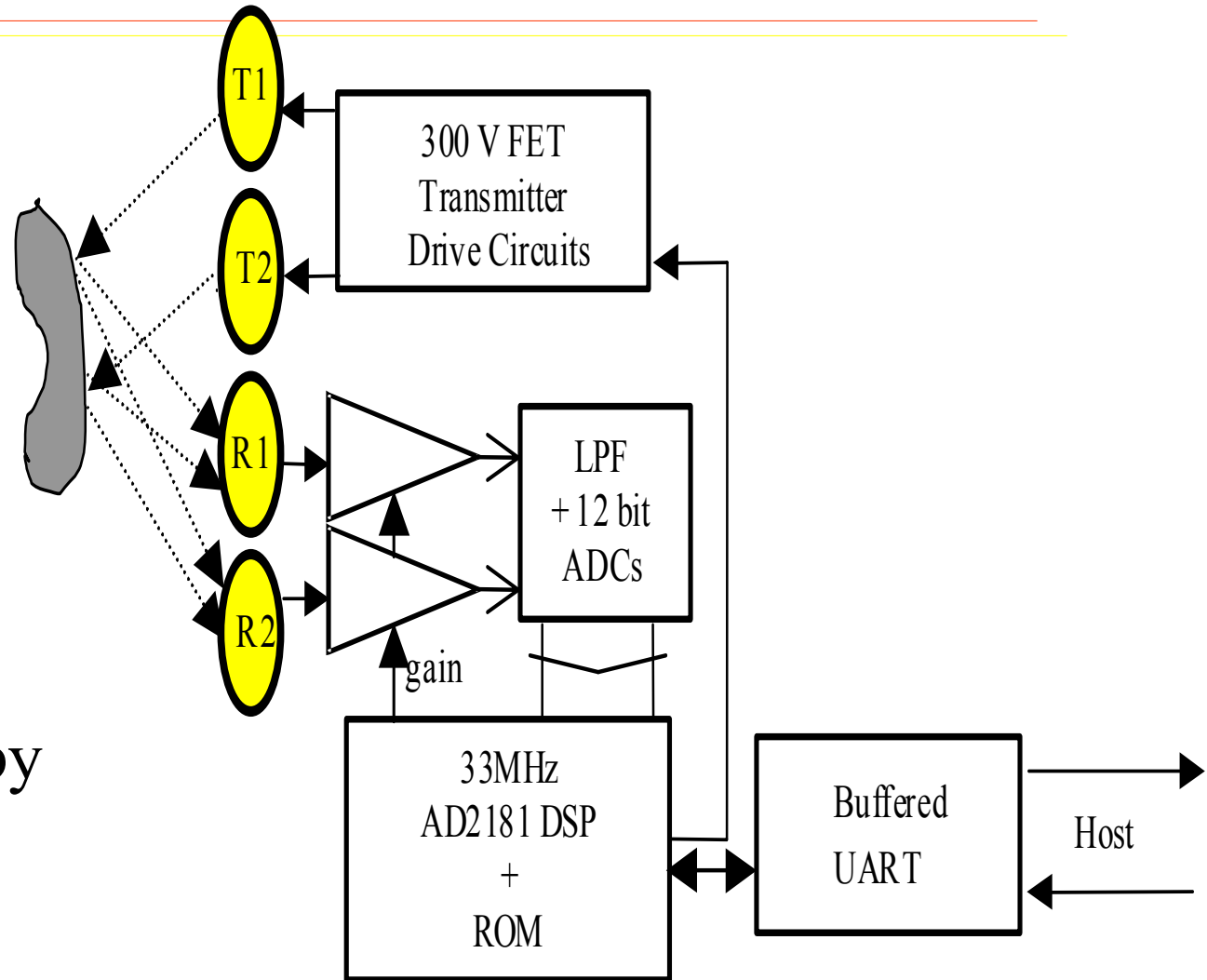
- ◆ Sonar sensing can accurately measure range and angle to specular targets
- ◆ The sonar sensor developed is accurate to 0.1 degrees, 0.2 mm to 5 metres at ~ 30 Hz.
 - subject to speed of sound calibration.
 - Estimated cost: \sim \$500

Hardware Description

1MHz
sampling on
2 channels

All in one box
reduces noise
levels

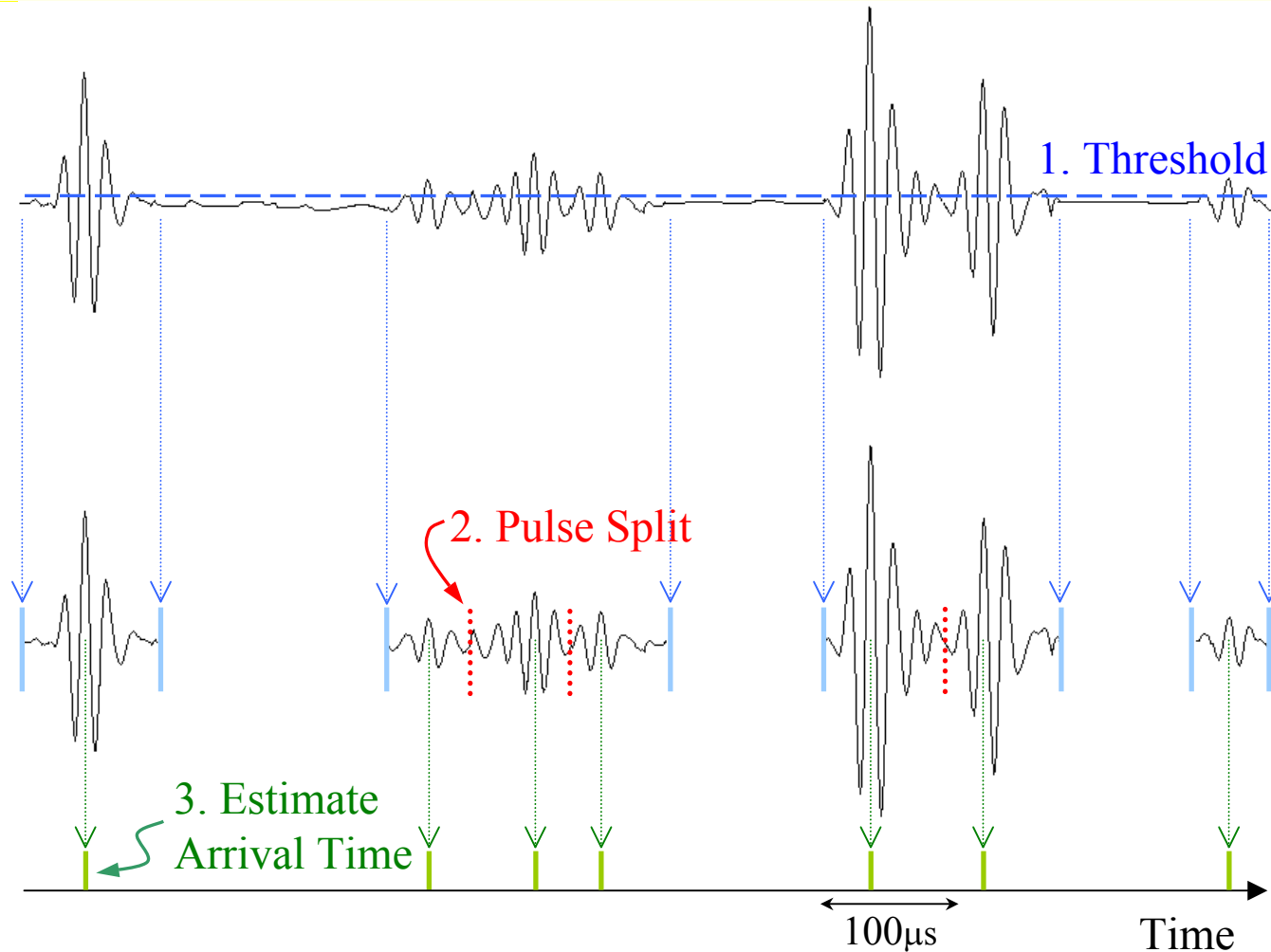
Fast processing by
DSP



Signal Processing

- ◆ Thresholding
- ◆ Pulse extraction
- ◆ Pulse splitting
- ◆ Template matching
- ◆ *Double pulse recognition -later*
- ◆ Data Association
- ◆ Triangulation & Classification

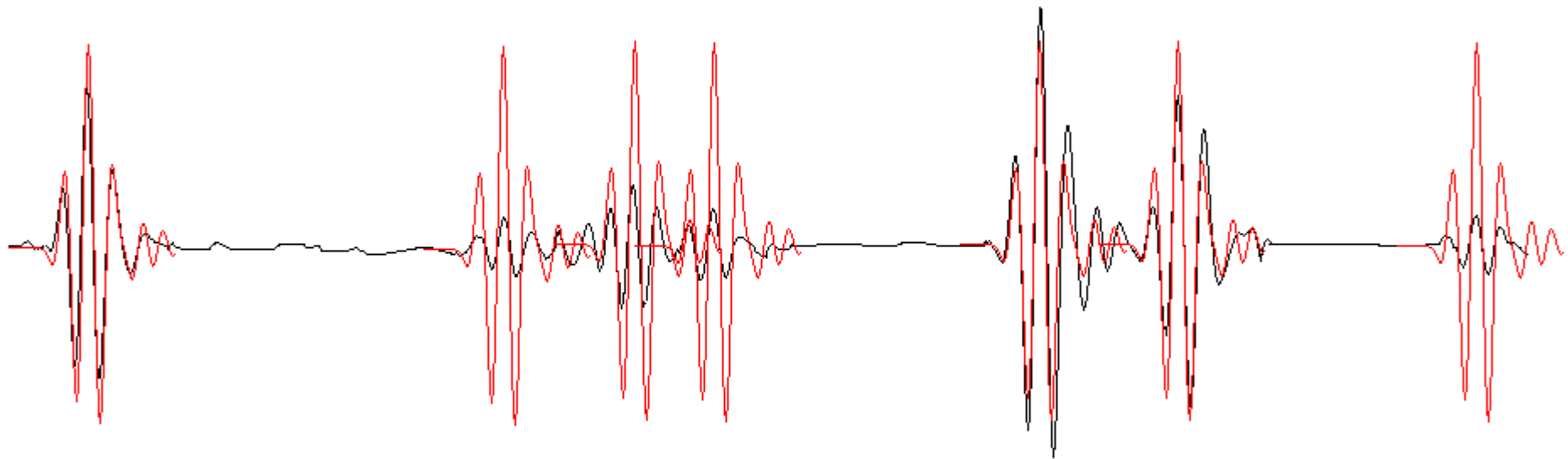
First Stages of DSP Processing



Template Matching

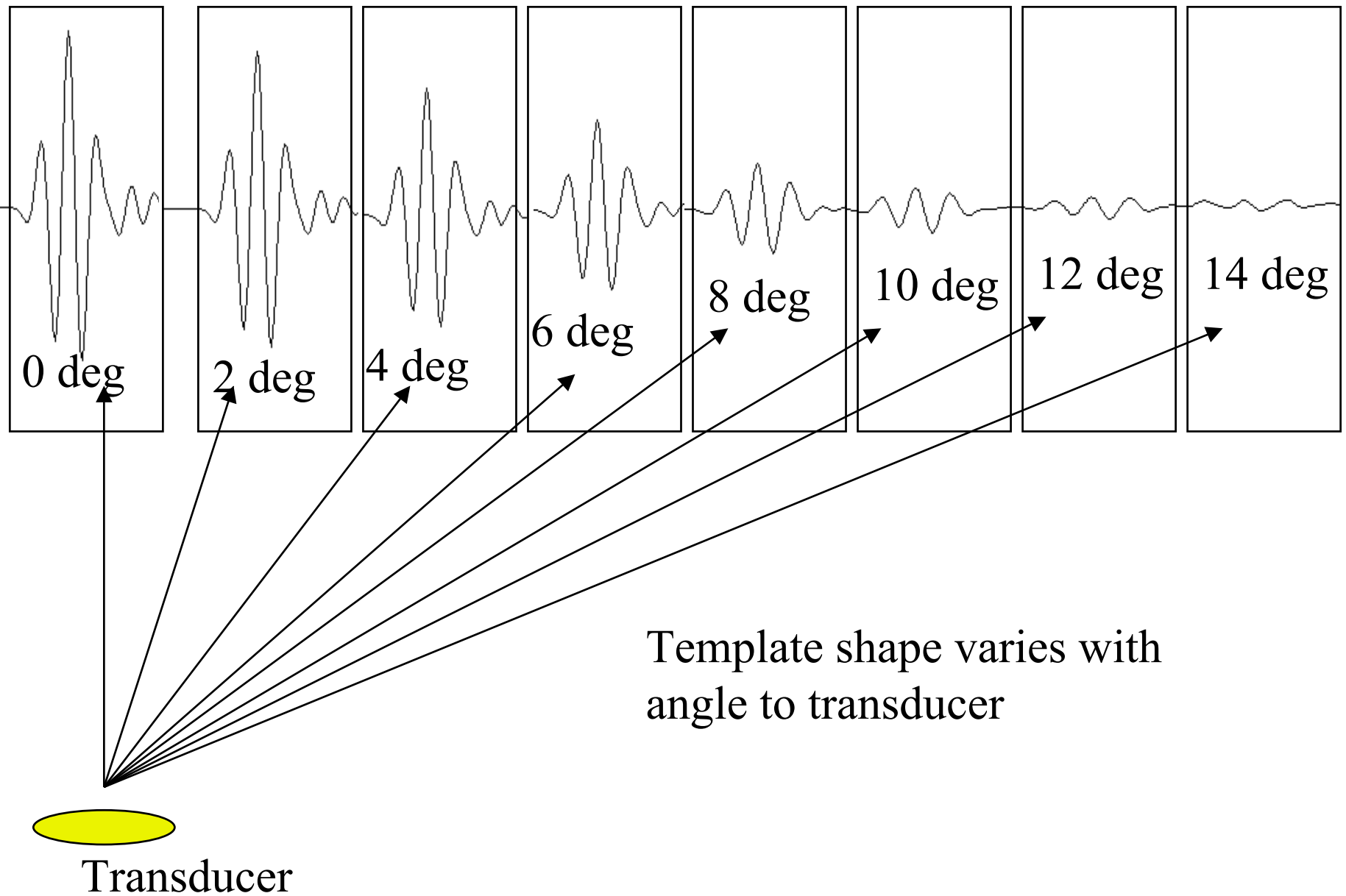
- ◆ Templates are pre-computed echo pulse shapes used in the arrival time estimation.
- ◆ Shape depends on arrival angle and range.
- ◆ This dependency has been accurately modelled - see [*Kleeman&Kuc IJRR 1995*]
- ◆ Thus the template set can be generated from a measured echo at normal incidence at 1 m.

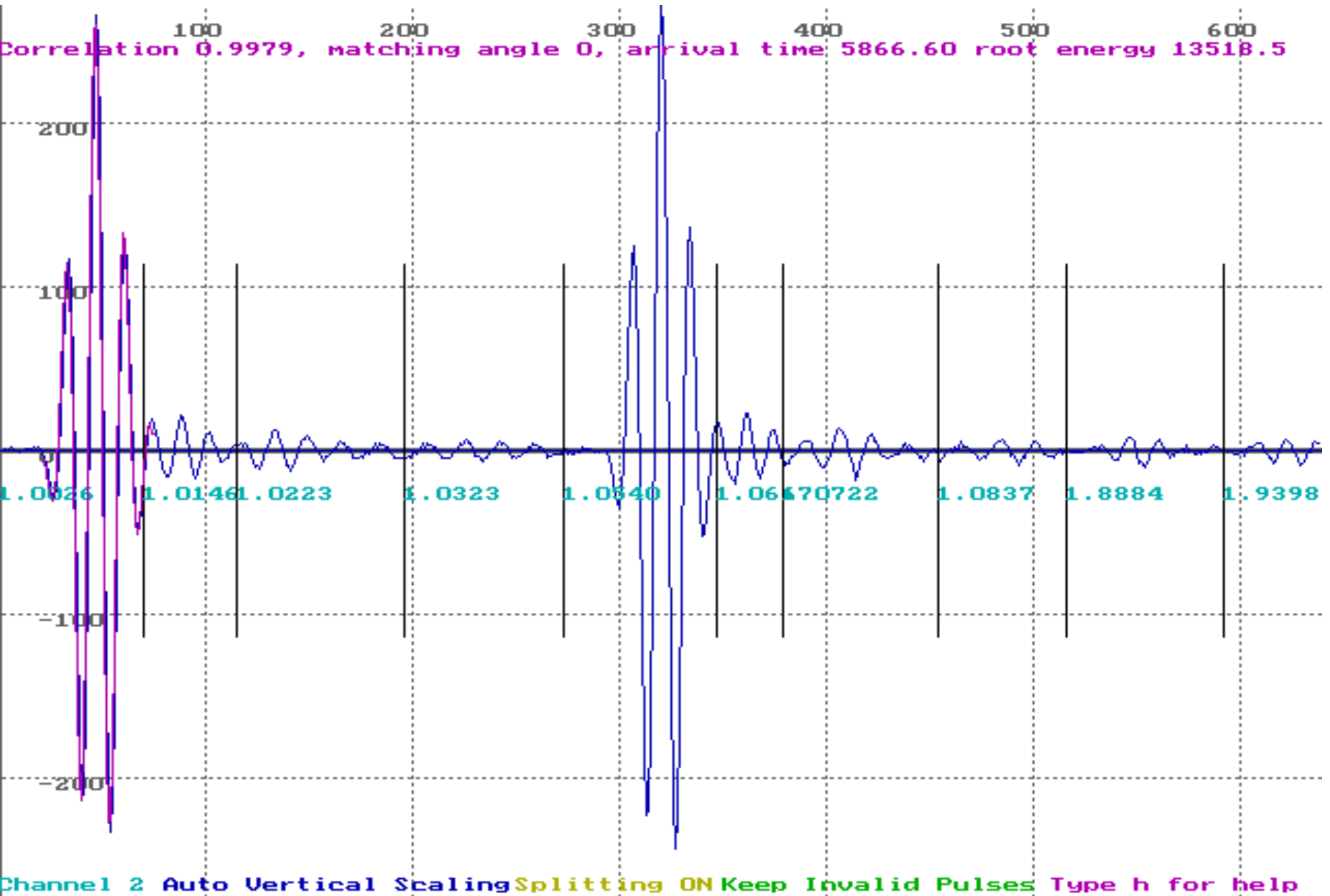
Template Matching



Templates shifted to best match the pulses using correlation of the template with the pulse.

Shape (not amplitude) are matched by a correlation.





Receiver Data Association

- ◆ Left and right receiver arrival times are associated based on:
 - arrival times consistent with small receiver spacing
 - amplitudes matched
 - correlation coefficients $> 95\%$

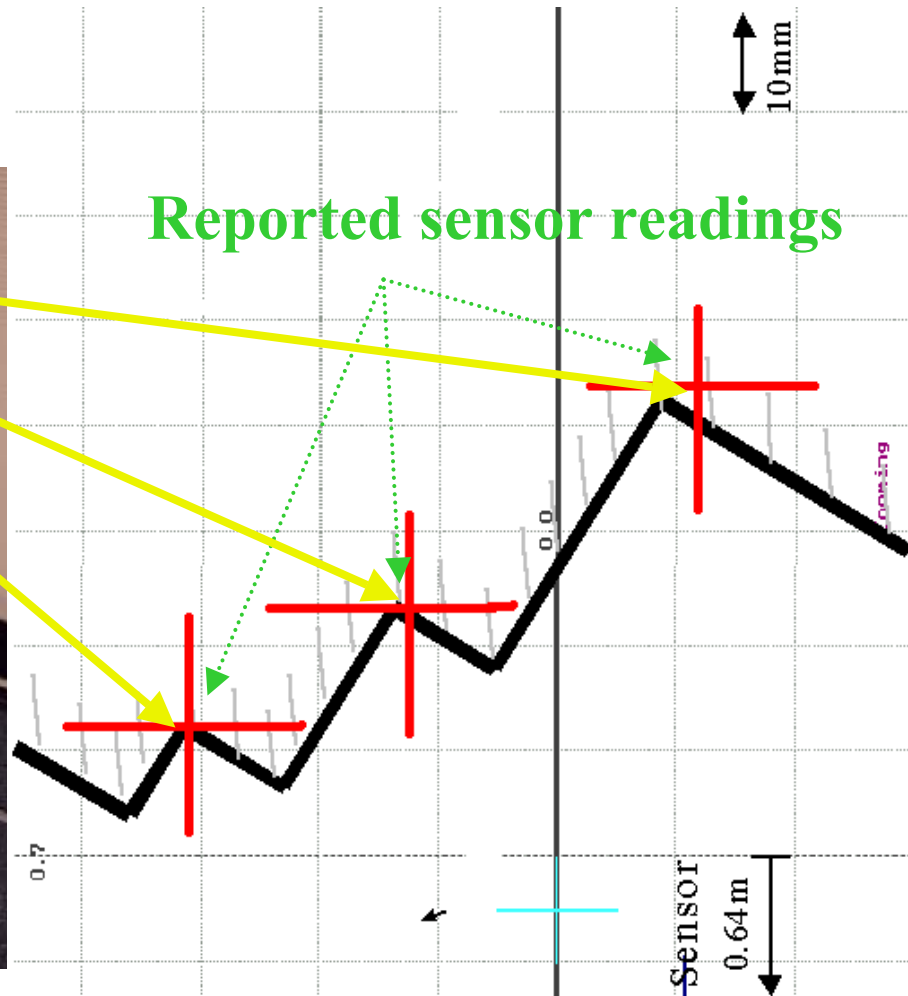
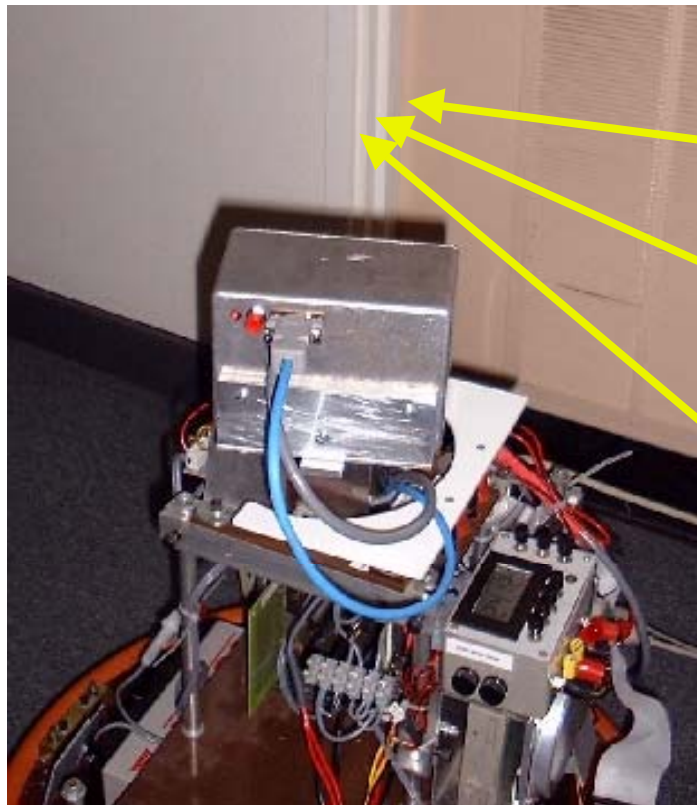
More Processing ...

- ◆ ***Triangulation*** – compute the bearing to each target.
- ◆ ***Classification*** - discussed later
- ◆ ***Double Pulse Recognition*** – rejects interference by recognising the sensor's ‘voice’ - *later....*

Clutter

- ◆ The echoes from closely spaced targets interfere with each other.
- ◆ To distinguish two targets, they must be at least 10 mm apart

Clutter - Door jamb experiment



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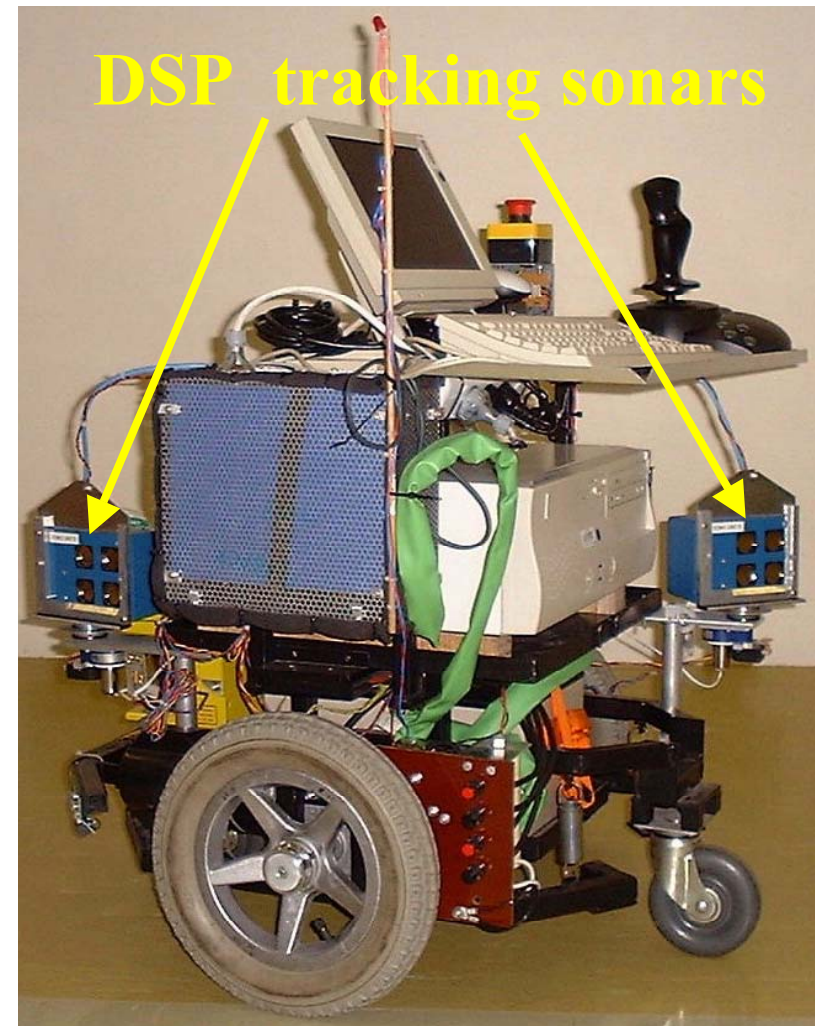
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- ◆ Interference Rejection and Pulse Coding
- ◆ Classification
- ◆ SLAM
- ◆ Conclusions and Future Work

Sonars “on the Move”

Mobile robot travels at speeds up to 1 metre/sec.

Panning sonars can track a target using sonar sensed angle feedback.

Sonar also can windscreen wiper.



Target Tracking - Moving Target



Target Tracking - Moving Robot

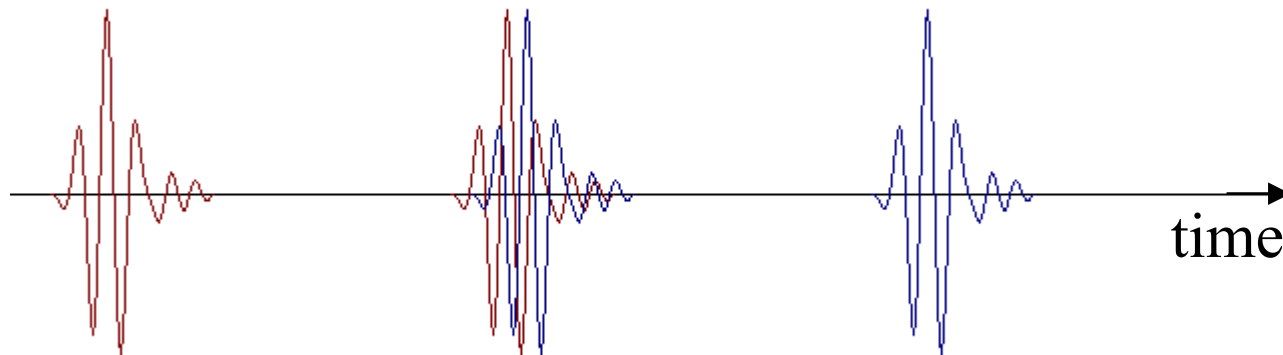


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Interference

- ◆ From another sensor
 - crosstalk
- ◆ From the environment (multiple targets)
 - echoes from a **nearby target** and a **more distant target** can overlap and obscure each other

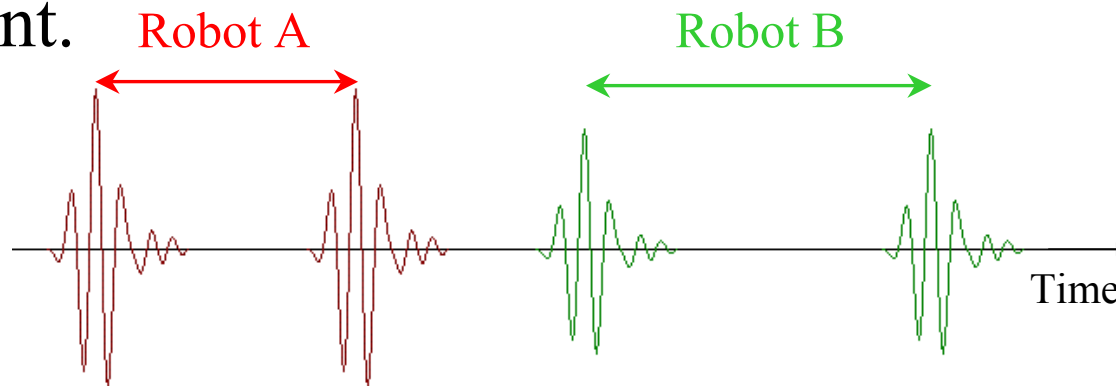


Double Pulse Coding

- ◆ In crowded caves, bats' sonar still works because each bat recognises its own voice.
- ◆ To prevent cross-talk, we give each robot a unique 'voice' using a fast and reliable technique called *double pulse coding*.
[Kleeman IROS 1999 awarded Best Paper]

Double Pulse Coding

- ◆ When Robot A makes a reading, it transmits two sonar pulses separated by a specific interval.
- ◆ When receiving results, pairs of echoes with the same interval are accepted – but echoes from Robot B will be discarded, because the interval is different.



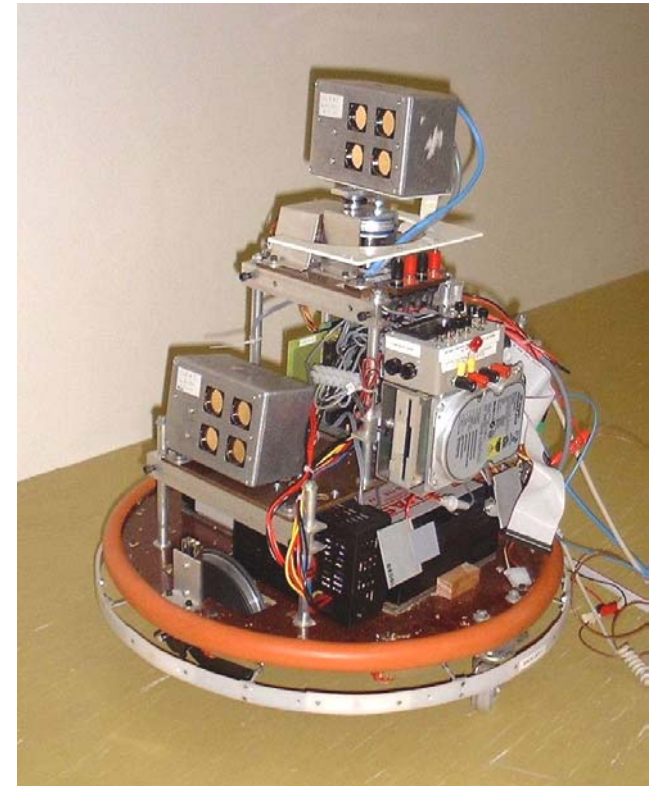
Double Pulse Coding

- ◆ *Preset* separation:
 - disastrous if two sensors use the same
 - blindness to certain targets

- ◆ **Random separation:** [Heale & Kleeman ACRA2000]
 - simple, automatic
 - problems do not persist

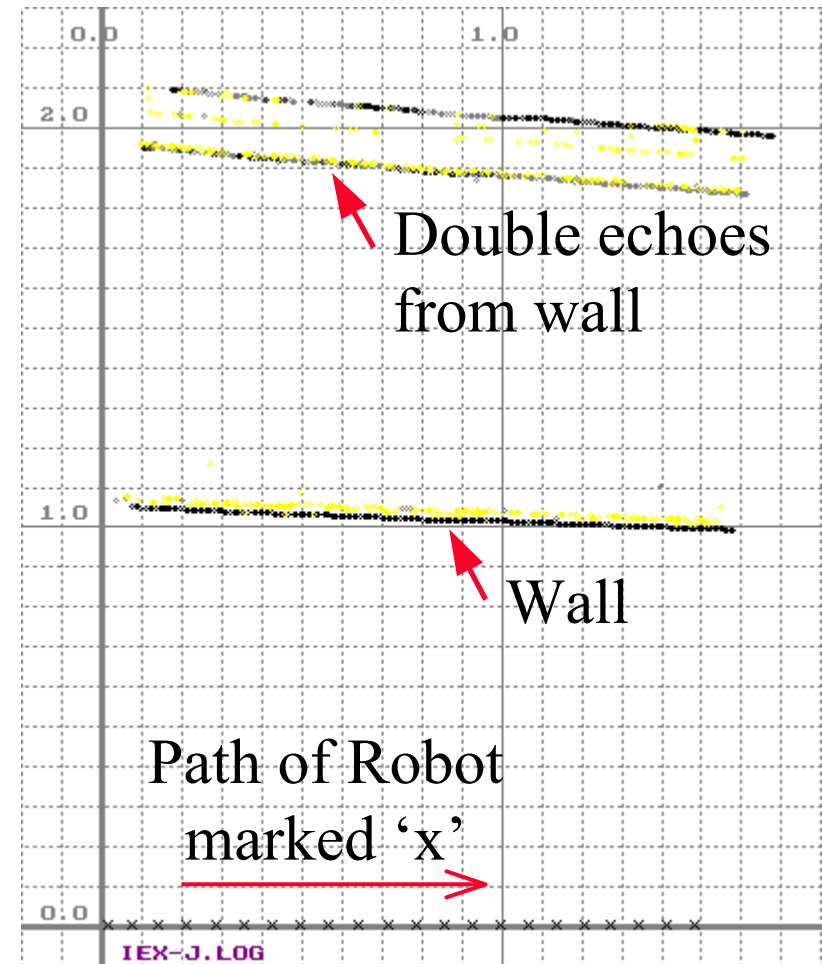
Interference Experiments

- Mapping a wall
 - 1 m range, moving parallel
 - 2000+ measurements taken
- **Aim:** to show robustness of double pulse coding



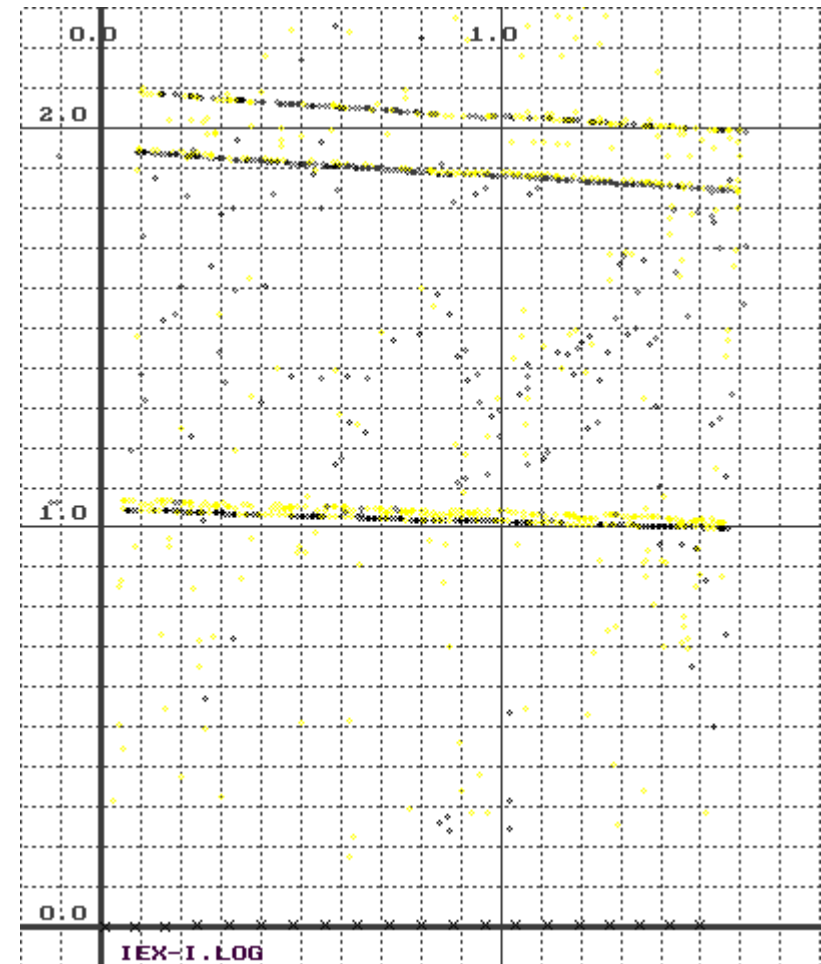
Interference from another sensor

- Double/Double with different separations
- no errors



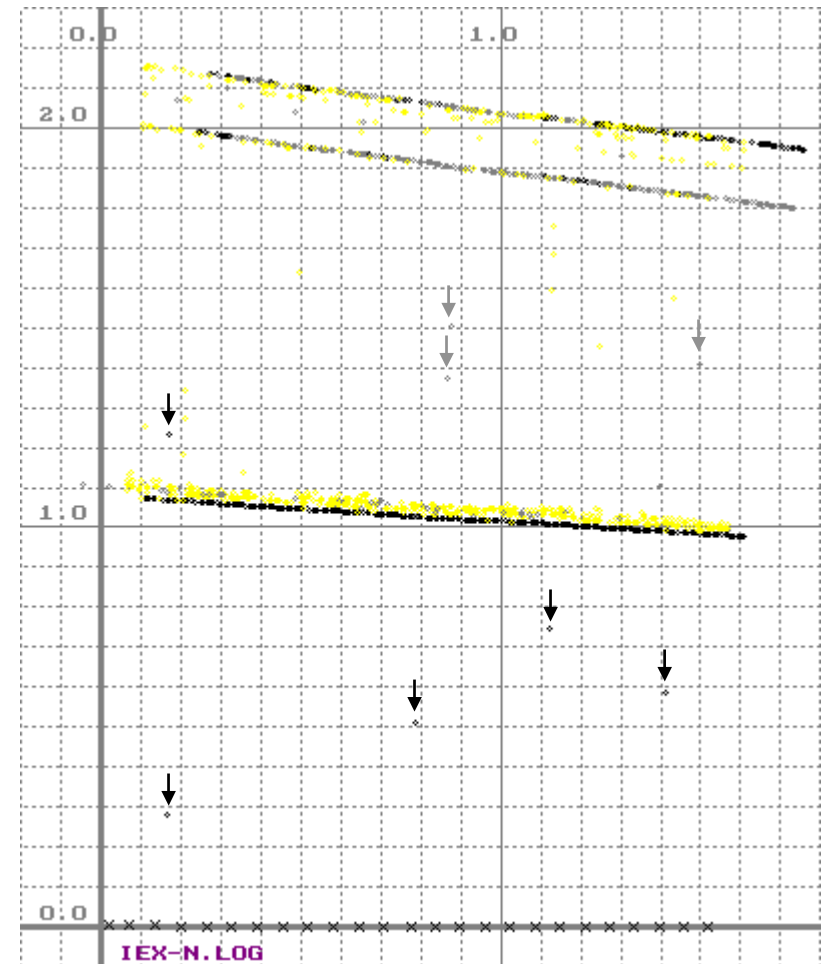
Interference from another sensor

- Double/Double with same separations
- Many errors!



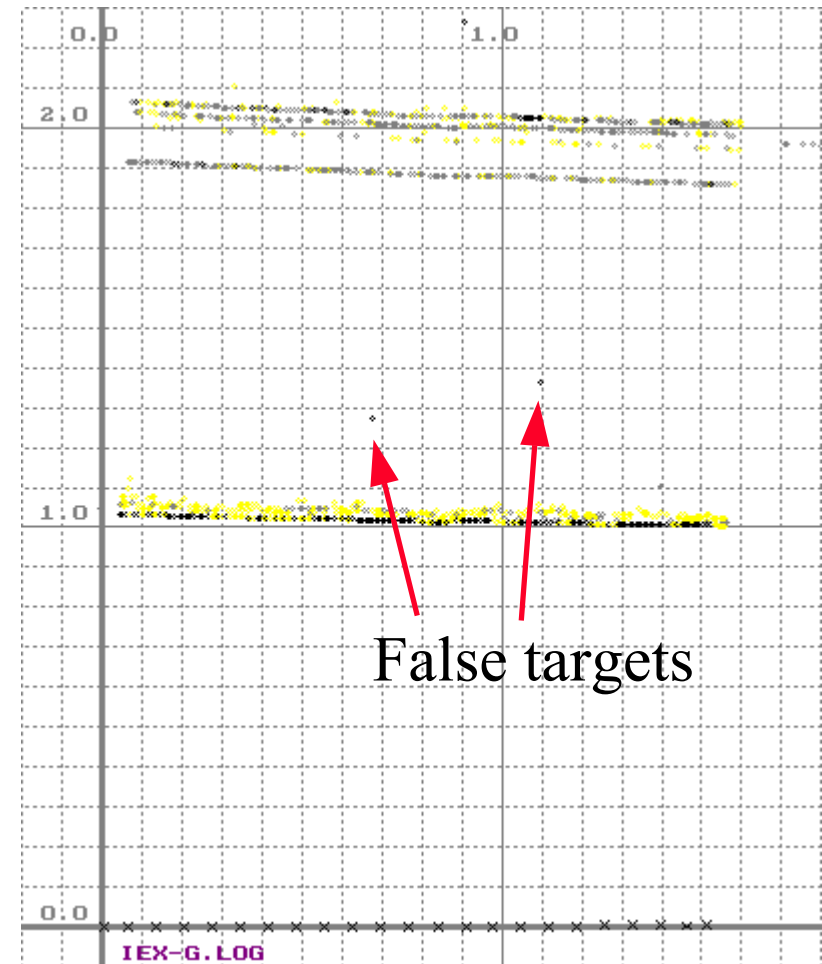
Interference from another sensor

- Random/Random
- Low error rate
- Errors not repeated



Interference from another sensor

- Random sensor / Single interference
- Environmental 'fakes' caused by two targets
- One single transmitted pulse can cause this



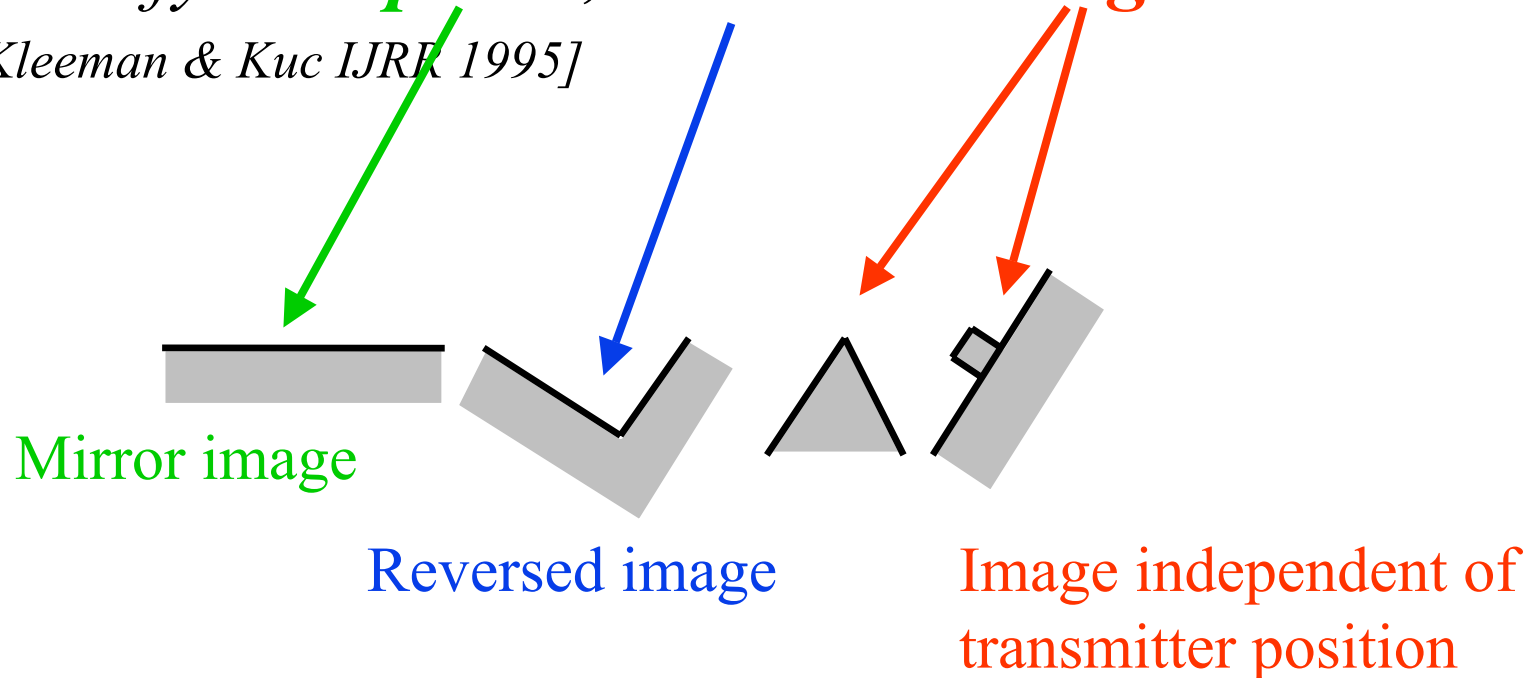
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Classification

- *Two transmitter positions are required to classify into **plane**, **corner** and **edge**.*

[Kleeman & Kuc IJRR 1995]



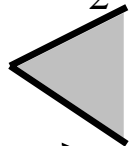
Previous Work on Classification

[Kleeman & Kuc IJRR 1995]

◆ *Two* measurement cycles (each ~50 msec):

- Transmit T1 and measure target bearing θ_1
- Transmit T2 and measure target bearing θ_2
- If *delta* = $\theta_1 - \theta_2$ is:

0 => edge



< -*threshold* => plane



> *threshold* => corner



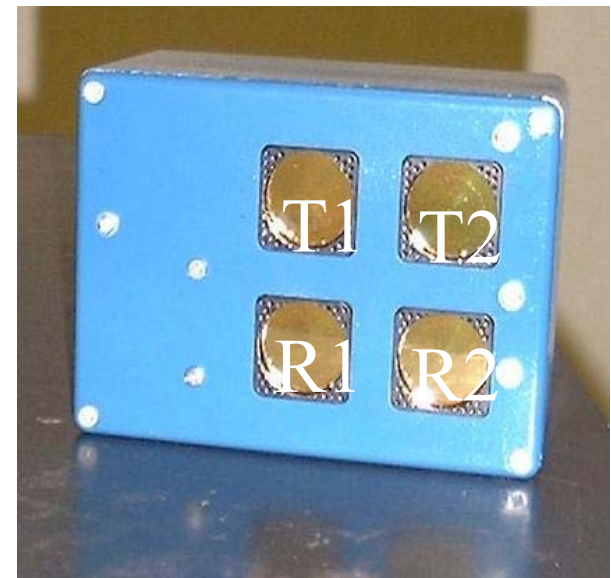
where threshold = $\frac{1}{4} \tan^{-1} \left(\frac{\text{ReceiverSeparation}}{2 \times \text{RangeOfTarget}} \right)$

◆ T1 and T2 spaced by 25 cm

Single Cycle Classification

[Heale and Kleeman IROS 2001]

- **One** measurement cycle strategy (~ 32 msec)
- T1 and T2 **4 cm** apart (cf 25 cm previously)
- Transmit T1 then T2 with short known delay (~ 200 usec)
- time of flights T1 to R2
= T2 to R1 for vertical targets
=> double pulse coding
+ classification



Single Cycle Classification

- *Classification in one cycle* – using either:
 - $\text{delta} = \theta_1 - \theta_2$ or
 - Maximal Likelihood Estimation (MLE)
 - Delta - fast and simple
 - MLE - uses range and angle information, requires measurement covariance, gives classification probability and improves range estimate.

Classification and Noise [IROS2001].

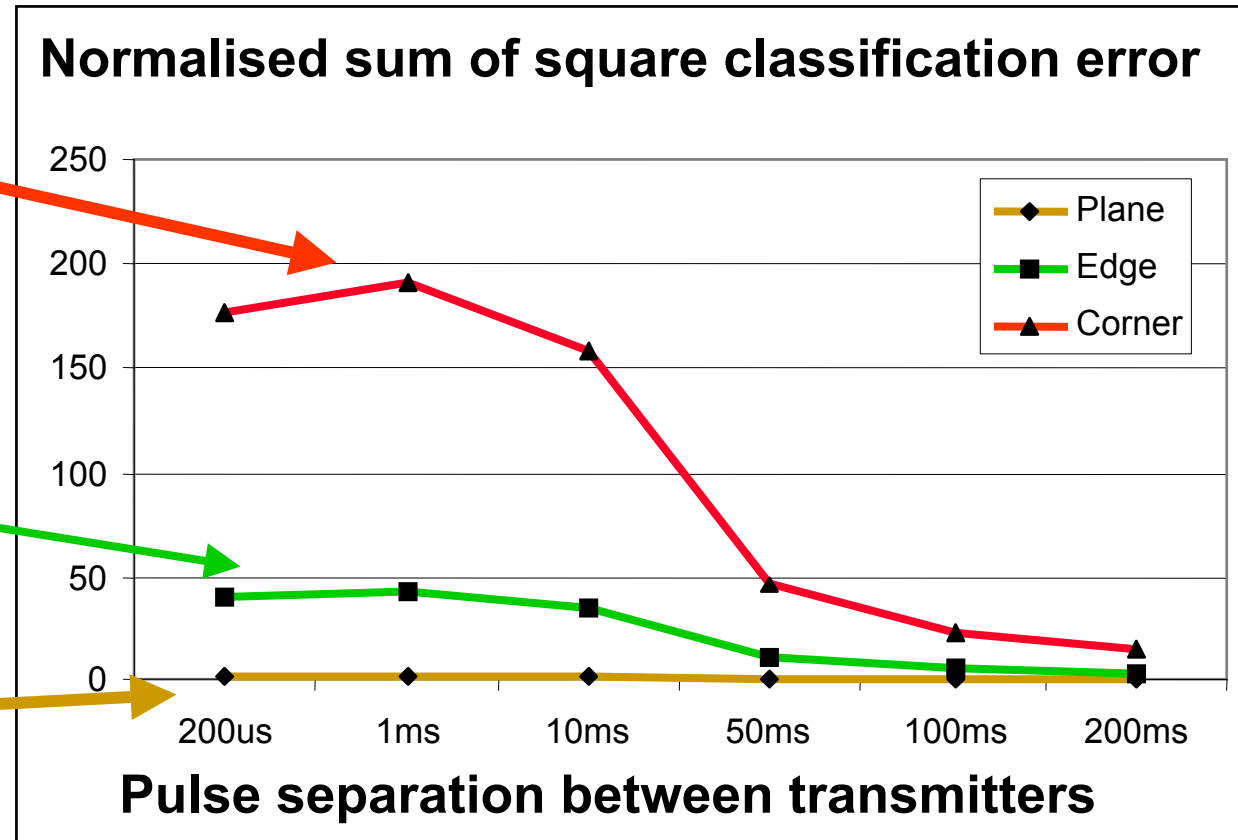
- Classification is more reliable when the pulse separation is short
- Classification is still reliable when transmitter separation is small.
- Interference rejection comes for free - double pulse coding T1 -T2.

Classifying a Plane with MLE

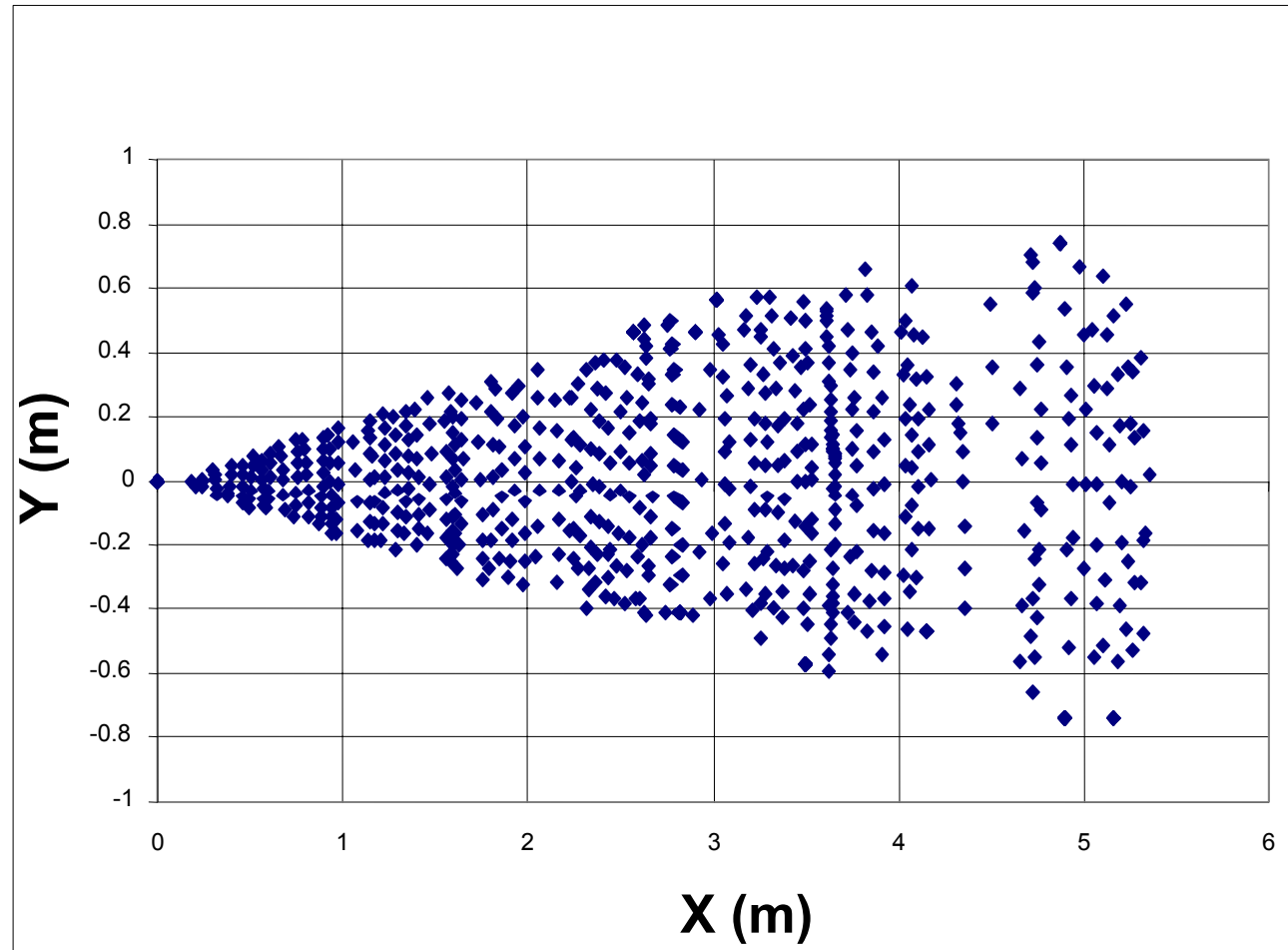
Large error when fitting plane measurements to corner model

Error when fitting plane measurements to edge model

Plane fits well!



Area of Plane Recognition



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Simultaneous Localisation and Mapping (SLAM)

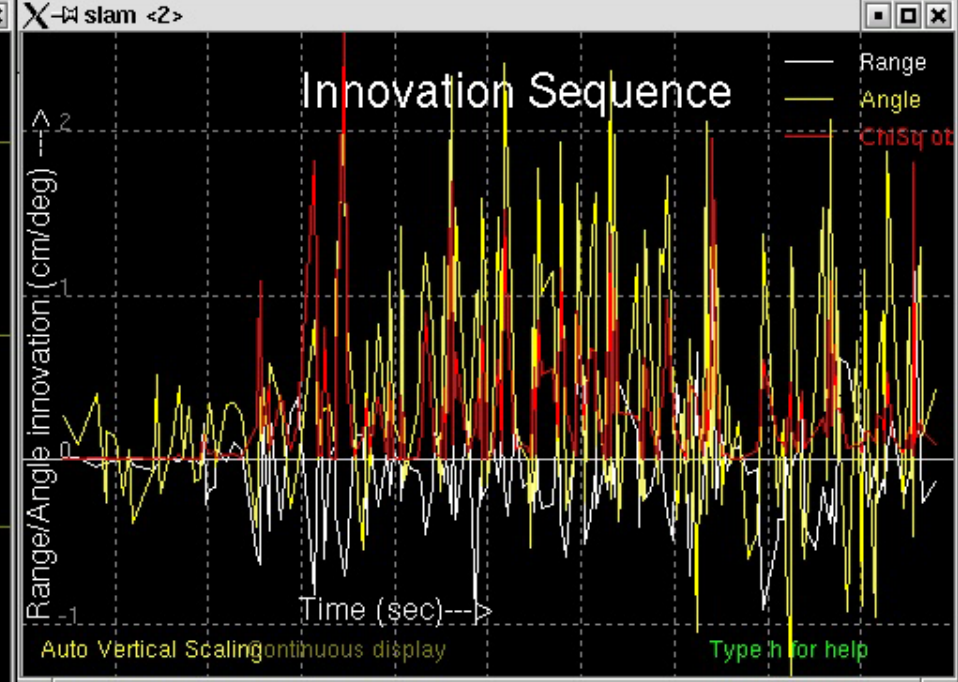
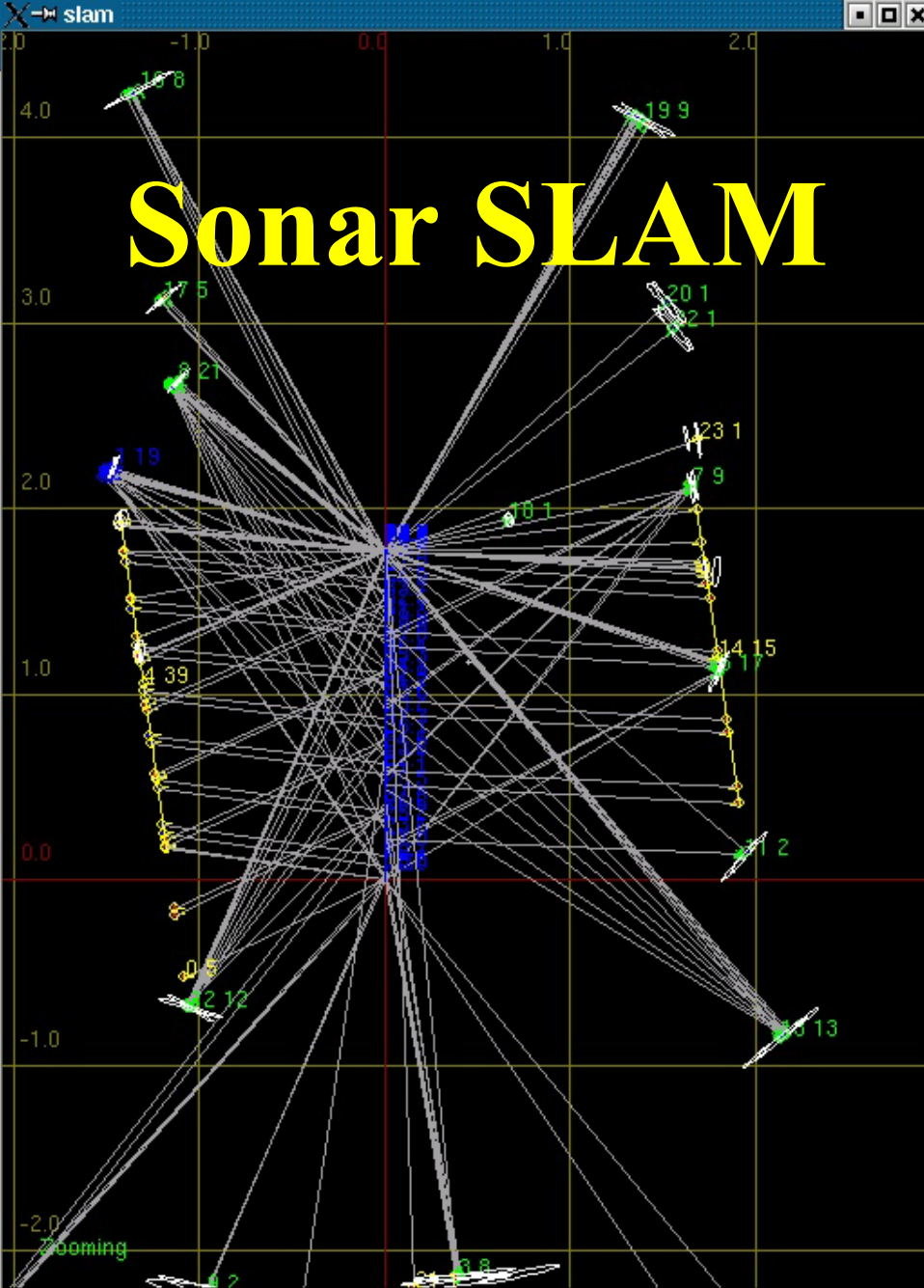
- ◆ Key problem for autonomous robotics:
 - starting in unknown environment
 - solving SLAM requires estimation of errors in robot pose and map information and their *correlation*. Error corrections can then propagate!

SLAM

- ◆ Classic Kalman Filter Solution:
 - computation per step and memory scales as n^2 ,
 n =number map features
 - association problem
 - kidnapped robot
 - loop closing
 - multi-sensor fusion with inconsistency, noise..

Autonomous Exploration

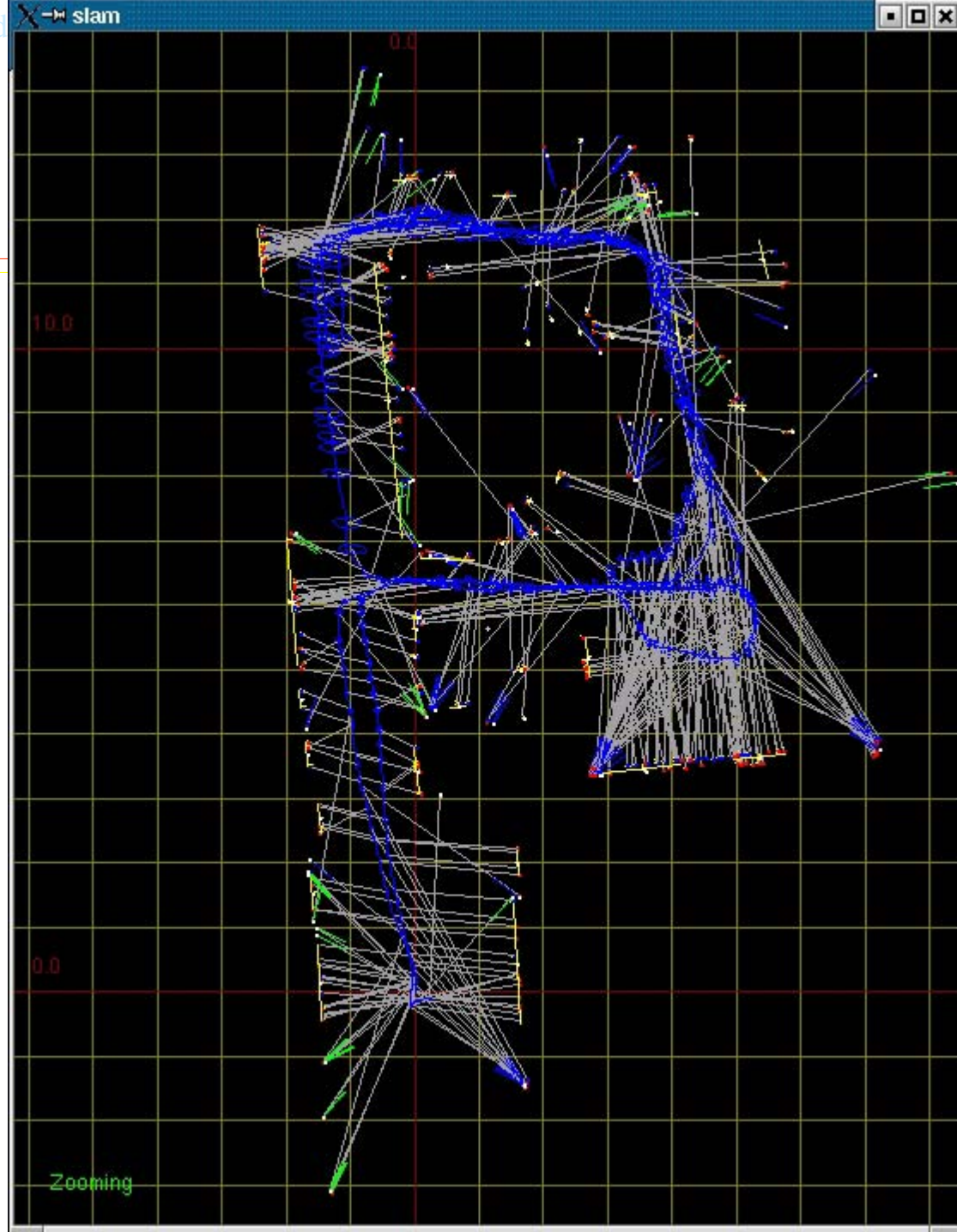




SLAM with Loops

Map
with all
measurements

Advanced



SLAM with Loops

Final
map



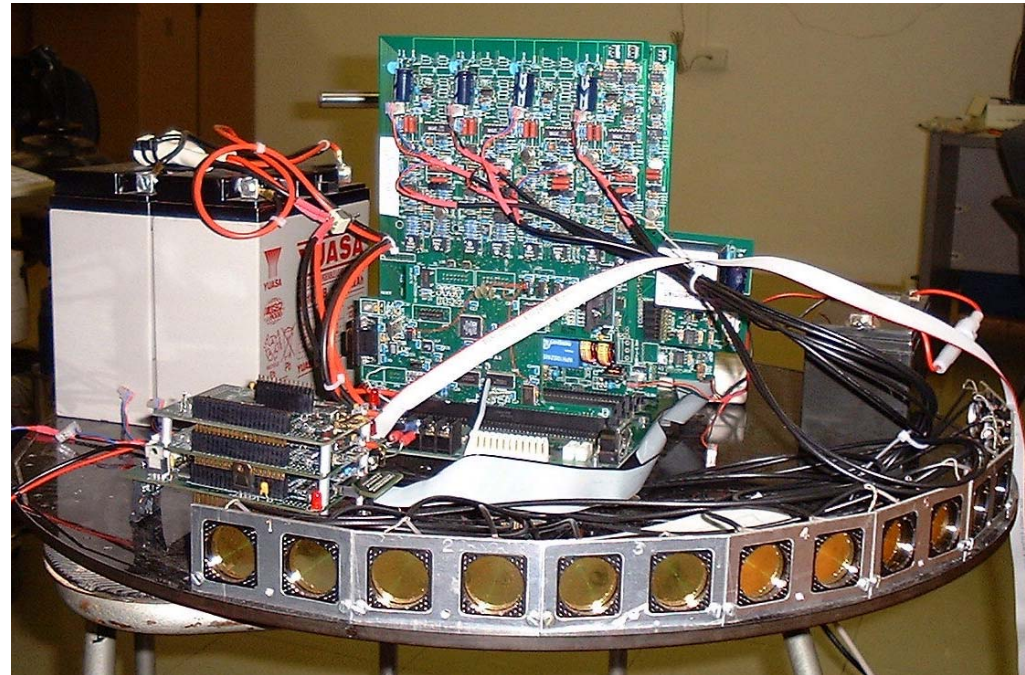
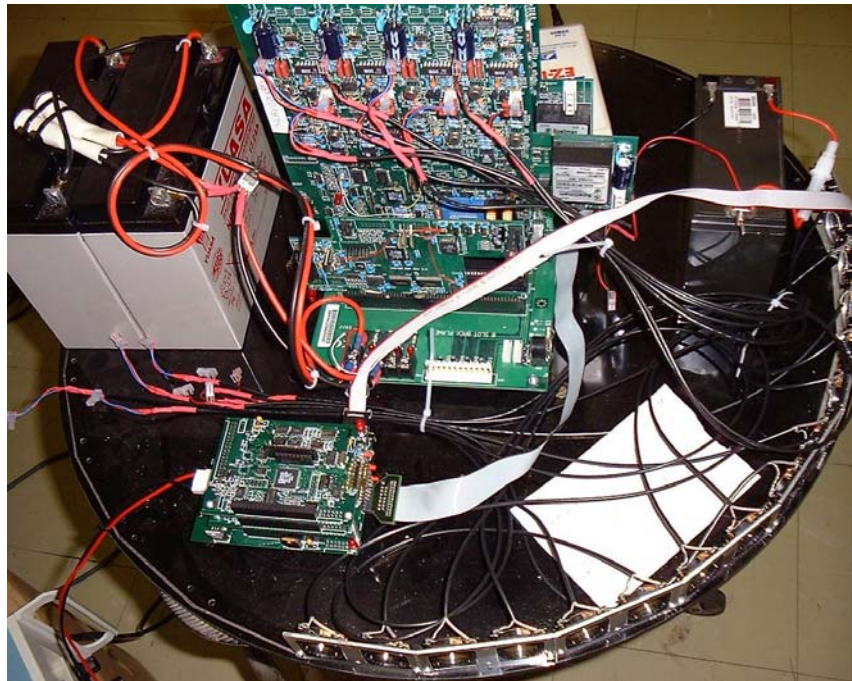
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- ◆ Conclusions and Future Work

Conclusions and Future Work

- ◆ Sonar results achieved so far
- real time optimal range and bearing estimation
- on-the-fly single cycle classification
- interference rejection
- motion compensation
- SLAM
- ◆ Future work:
 - Fusion with laser, DSP sonar ring, optimal exploration, synergies of 10+ sensors.

DSP Sonar Ring – work in progress



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