



MASTER THESIS PROPOSAL

UNDERWATER 6 DOF LOCALIZATION USING IMAGING SONARS

Gabriel Alcantara Costa Silva

Proposta de tema de tese a ser desenvolvido
no LEAD/COPPE/UFRJ para o Mestrado
do Programa de Engenharia Elétrica da
COPPE/UFRJ

Orientadores: Ramon Romankevicius Costa
Sylvain Joyeux

Rio de Janeiro
Abril de 2015

Contents

1	Introduction	1
2	Motivation	2
3	Objectives	4
4	Metodology and Expected Results	5
5	Research Topics	6
5.1	Underwater Localization techniques	7
5.2	Particle Filtering Study	7
5.3	State Space Modeling	8
5.4	Sonar Sensor Modeling	9
5.5	A priori map construction	10
5.6	Case Study	10
6	Proposed Bibliography	12
7	Workplan	17
8	Conclusion	18

Chapter 1

Introduction

The development of underwater systems has grown in the last past years, due the increase in the complexity and risk of the tasks to be executed. The inspection, installation and maintenance of undwater structures is usually performed by diver and this tasks are a slow, expensive and dangerous work. In addiction, due the hidrostatic pressure, a diver is recommend to go only until 100 meters deep. Thus, in subsea exploration there are several applications, which the utilization of a diver is simply not possible. To fullfil the crescent needs of the oil and gas, hidreletric industries, as well the subsea floor and life mapping and research, the academic and industrial community is focusing in the development of unmmanned underwater vehicles and systems capable of execute generic tasks. In this context, the GSCAR group, including the participation of the LABCON and LEAD laboratories, COPPE/UFRJ, has given a special attention to submarine robotics.

In a patnership with the company Energia Sustentável do Brasil (ESBR), the Laboratório de Controle e Automação, Engenharia de Aplicação e Desenvolvimento (LEAD/PEE/COPPE) has signed a project, named ROSA, to develop a monitoring and 3D mapping system to work in Stoplog operations at the hidreletric power plant of Jirau. The system is responsible to monitor the operation of insertion and withdrawl of Stoplogs, metal blocks used to seal the inlets and outlets of the turbines, and also to map the bottom of the tubine inlets, looking for debris and accumulated silt.

The ROSA Robot consists in a system with a ground eletronic and a underwater embbded eletronic. The latter has four inductive sensors, a pressure sensor, a inclination sensor, a pan-tilt unit and the possibility to connect either a profilling sonar, the *Tritech Seaking*, or a Mechanically Scanned Imaging Sonar (MSIS) sonar, the *Tritech Micron*.

The ROSA robot is going to be the central piece of development of the proposed thesis and localization algorithm.

Chapter 2

Motivation

In the ROSA robot, the mapping is performed by the sonar, where each echo received is processed and taken into account to generate a map of the underwater environment. However, if the location of the system isn't known, it is not possible to align all the data received from different positions and orientations and construct a 3D model of the environment. There are several techniques to map and localize a underwater system in respect to the environment, however the Jirau power plant is located in the Madeira River and its turbid waters forbids the application of vision. Besides, there is many regulations and constraints that make any modification to the dam structure virtually not possible. For that reasons, the most suitable sensor to map the underwater environment is a sonar device.

The ROSA robot is attached to the Lifiting Beam and is submerged following a rail in the dam concrete wall. The whole system can only move in the vertical direction and the head of the sonar can be oriented in almost any particular desired direction, because of the presence of the pan-tilt unit. Without a localization algorithm is not possible to position the echo received in a three dimensional map.

The intention of the project is to allow a person to assemble and disassemble the system as many time as desired. That means mounting differences in each assembly, in every operation there are going to be unknown parameters. The estimation of theses parameters are also relevant to the localization algorithm, because the transformation between the coordinate frame of the body of the robot and the head of the sonar, i.e. the orientation of the beam emitted, has an unknown bias. This biases can make a error of several centimeters with the increase of the distance of an obstacle measured.

In the literature, the localization of underwater systems with the support of active sonars is little explored, specially taking into account the 6 DoF of the vehicle. In general, the localization techniques aim in a plane based localization and mostly uses the range finder theory to perform the evaluation of the sonar beams. The incorporation of the wave propagation characteristics and beam shape are still very

little explored. The implementation of Underwater 6 DoF Localization using Imaging Sonars would be a contribution of great value to the academic field, specially if the algorithm succeed with inexpensive sonars.

Chapter 3

Objectives

The main objective of the proposed master thesis is to research the state of the art of underwater localization and the utilization of sonars to perform this task. The techniques are going to be compared and its pros and cons listed.

In addition, it is proposed a implementation of a 6 DoF localization algorithm using a Mechanically Scanned Imaging Sonar, the *Tritech Micron*. The algorithm is going to be based on a Rao-Blackwellized Particle Filter and be responsible to estimate the position of the ROSA robot and also the unknown parameters caused by the manual assembly..

The sonar sensor model will receive special attention and is intended to incorporate underwater acoustics theory and intrinsic characteristics of the sonar, as an example its beam shape, to use the most of information provided by the sensor as possible and not only a distance measurement.

Chapter 4

Metodology and Expected Results

The methodology of work will consist in a detailed study of the state of the art in the field of underwater localization of mobile systems, incorporating classic techniques and implementing in a actual system.

The techniques proposed in the literature will be evaluated in terms of cost, praticallity and viability of implementation in the Jirau Powerplant. The Stoplogs holes are a unfriendly environment and the system is exposed to the weather, to debris present in the water and, in addition, there are several constraints in the permanent alterations that can be made in the current structure.

The ressources needed during the development of the master thesis, human and computacional, are going to be provided by the Laboratório de Controle e Automação, Engenharia de Aplicação e Desenvolvimento (LEAD/PEE/COPPE) and Laboratório de Controle (LABCON/PEE/COPPE).

The test setups are going to be realized in a partnership with the Laboratório de Ensaaios Não Destrutivos (LNDC/PEMM/COPPE), where there is a water tank with the dimensions of 12 meters long, 6 meters wide and 7 meters deep. There is, also, going to be gathered a real dataset in the field trips to the hidreletric power plant of Jirau.

Chapter 5

Research Topics

This section describes the topics to be studied in order to accomplish the fulfillment of the requirements of a master thesis and the implementation of the desired localization algorithm.

- Underwater Localization techniques - study of the state of the art in the field of underwater localization, analysis of the techniques applied in the literature and evaluation in respect of cost, processing power requirement, sensors used and robustness.
- Particle Filtering study - Study of the Particle Filter approach as a practical implementation of a Bayesian Filter. Due to the strong non linear characteristics of the problem and also from the sonar sensor, this particular type of filter was chosen to be the main technique to be explored as a possible solution of a low cost underwater localization problem.
- State Space Modelling - the number of dimensions to be sampled in the particle filter has a direct effect in the number of particles necessary to sufficiently cover the state space. In consequence, a special attention must be given to reduce as much as possible the number of states sampled.
- *A priori* map construction - Study of the state of the art of spatial data representation and the construction of a map model from the actual environment.
- Sonar Sensor Modelling - Study of underwater acoustics and wave propagation, sensor modelling and likelihood functions fitting. A sonar sensor model will be derived and evaluated within the algorithm.

5.1 Underwater Localization techniques

Study of the state of the art in localization techniques for underwater systems. This study has the objective to point, evaluate the possible approaches and all the pro and cons associated with each one. And finally, justify, formally, the use of a sonar as main sensor.

Underwater localization can be performed with the utilization of one or more transponders. The baseline techniques can differ in the number of transponders required and the maximum coverage area. However, all of these techniques require the implementation of an additional infrastructure, what requires authorization and represents costs and maintenance.

Video aided localization is also possible underwater, but it has hard requirements of luminosity and transparency of the water. This requirements makes this approach impracticable for most application, specially in the Madeira River water, that has zero visibility. On the other side, due the characteristics of sound wave propagation in water, the acoustic sensors are the most suitable sensors to perform underwater localization and mapping, being able to measure long ranges without significant signal energy loss. In the industry, there are many types of sonars available. The Profiling sonars behave almost like a laser scan, with a pencil like beam shape, but with a much bigger aperture. Imaging sonars, on the other hand, provides more information about the echo received and produce an image of the insonified area.

In the scope of the proposed thesis, the sensor that is going to be explored is the Tritech Micron, a Mechanically Scanned Imaging Sonar (MSIS). This sensor is an inexpensive sonar is going to perform a main role in the development of the particle filter and sensor model explained in more details in the section 5.2 and 5.4.

5.2 Particle Filtering Study

The localization algorithm must be able to handle all the uncertainties related to the process of gathering data, processing it using different models and with the support of computational systems. All this steps have their own kinds of uncertainty and noise. The sensors are imprecise and affected by external noise; the movement of the robot is subject to mechanical clearances and vibrations and the model used to describe either the robot and the environment that surrounds it can never be considered truly accurate.

Uncertainties are going to be present in every real system and, therefore, to handle this kind of problem, one strategy is to incorporate the errors intrinsic of the process inside the models and represent the uncertainty explicitly.

To properly estimate the state of the system from the sensor data available, the

most general technique used to approach this kind of problem is to solve it recursively, with a practical implementation of the Bayes Filter. There are several bayesian filtering techniques, with different assumptions, pros and cons. For example, the Kalman Filter assumes that the dynamic of the system is linear and it is possible to represent the variables to be estimated and the errors related to the process as multivariate normal distributions. The Kalman filter is the most utilized filter and considered an optimal estimator. However, it has its limitations and cannot, for example, represent non-linear dynamics or multimodal distributions.

This document proposes that, during the development of the master thesis, the state of the art in Bayesian Filtering is going to be deeply studied, showing the techniques well explored in the literature and the reasons for choosing a Particle Filter as the recursive state estimation filter to be applied and implemented.

Considering that the pressure sensor and the inclinometer are absolute sensors, a Rao-blackwellized Particle Filter is going to be proposed to solve the localization problem. The utilization of absolute sensors allows the filter to marginalize dimensions measured by this sensors and to sample only some of the variables, reducing the number of particles needed and the computational power needed. A Kalman filter, a alpha-beta filter or any other optimal filter can be applied, then, in the marginalized variables.

5.3 State Space Modeling

As mentioned in section 5.2, a special attention must be given to the analysis of the state vector of the system. The number of variables sampled by the particle filter has a direct effect in the number of particles needed, because a bigger state space requires more particle to be well covered.

The physycal characteristics of the Stoplog rail limits the dynamic of movement of the whole system. Therefore, the system can move, basically, in the vertical direction. The position in the xy plane can be considered fixed, but with a precise initial position unknown. The orientation of the body is also very constrained due the rail, the pitch angle can be considered negligibe, as well the yaw. The roll angle is measured by the inclination sample, however its actual position is drifted by an offset due the manual installation of the system.

The depth sensor gives the water column height above the sensor, now called h_s , but the a position in respect a fixed reference is not provided by the sensor, once the total water column height h_w , from the bottom to the water surface, is variable. To be able to calculate the measurement of the vertical position z , the filter need to sample the offset between the information given by the sensor and the height of the system in respect a fixed reference. In other words, the filter needs to sample

the water column height and calculate the vertical position z as $z = h_w - h_s$.

Given the configuration of the setup assembled to gather data of the ROSA robot in the LNDC lab, with the yaw orientation being free, the filter needs to be able to sample the yaws orientation and alternate to sample around a initial position in this dimension when desired, as well.

Additional parameteres to be sampled are all the biases in the transformation between the coordinate frame from the body and the head of the sonar, the system was hand-assembled in each test made, so the errors of mouting need to be taken into account.

5.4 Sonar Sensor Modeling

The utilization of laserscan and range finder sensors to perform localization of terrestrial mobile systems is well explored in the literature. A laserscan returns a time of flight measurement and the distance to the obstacle can be easily calculated. The hit of the laser is considered pontual and the likelihood function of the given point been full or not can be modelled much more easily than in the sonar case. If compared to a laserscan, the sonar sensor, which has a much broader beam, can be considered a very noisy range finder.

Although, in underwater environments, the use of laserscanners to sensor the environment is almost impractical due the characteristics of lighth propagation in water, with high attetuation and dispersion. The sound wave propagation in water, with a lower attenuation, makes sonar the most suitable underwater sensor.

Profilling sonars return a time of flight measurement and have a pencialar beam shape and, therefore, can be used as a usual range finder sensor. However, it produces sparse data and have a very poor angular precision. A Mechanically Scanned Imaging sonar, on the other hand, emits a fan shape beam and returns an array of intensities of acoustic signal backscattered form the environment. Each intensity correponds to a particular section of the beam, discretized in the radius direction. The transducer is mechanically rotated and a full 360° sector scan can be performed, but with a slow refresh rate cost. Even being a imaging sonar, the MSIS can be acquired with a relatively low cost.

In the literature, this kind of sonar is explored, in the context of underwater localization, with a range finder approach, however there is a lot of loss of information and introduction of unnecessary noise if one applies the range finder theory to a MSIS device, extracting the most intense echo and converting it to a distance measurement. Therefore, its proposed to study the underwater acoustics and wave propagation and apply this theory in a low cost sensor, the Trittech Micron. The information of the fan shaped beam, with its horizontal and vertical apertures, and also all the intensities

contained, if incorporated into the sensor model, allows the proposal of a likelihood function that is much more close to the real behavior of the sensor.

5.5 A priori map construction

To be able to localize the system, one needs to represent the environment that surrounds it. The more suitable environment model for a underwater vehicles, as well aerial systems, is a volumetric 3D model. The deegres of freedom of the system allows the system to explore the map in all three cartesian dimensions.

There are several techniques and data structures to represent, probabilistically, the position of the robot and the environment. The model must fullfill the requirements of dimensions derised and also be efficient computational wise.

In the literature, it is possible to cite the use of grid of cubics volumes of equal size, with the drawback of a big memory consumption. Other possibility is to store directly the measurements and generate a point cloud, which is not that more effi-cent in terms of memory consumption. A more efficient approach is to use 2.5D grid maps, elevation maps or Multi-Level-Surface Maps. This kind of maps discretizes the vertical dimension and, therefore, can optimize the computacional consump-tion. However, they cannot fully represent the actual 3D environment, what for a underwater system can forbid a good position estimation.

The mapping approach proposed to be studied more deeply and used in the construction of the localization algorithm is the Octomap framework [48], which is based in octrees, a hierarchical data structure for spatial subdivision in 3D. The oc-tree based mapping framework allows a volumetric 3D representation of the environ-ment with a low memory cost, has the capability to distinguish free and unmapped areas and it is open source.

The *a priori* maps, to be used with the datasets gathered detailed in the section 5.6, also need to be modelled. The LNDC map can be handmade, beacause of it's simple configuration. The Stoplog hole, at the Jirau Dam, must be modelled with the support of a 3D CAD model. The octomap framework also allows the conversion from the industrial standards 3D model files to an octree.

5.6 Case Study

The proposed algorithm is going to be evaluated in respect of two different sce-narios. The first one is the a test setup in the LNDC (Laboratório de Ensaios Não Destrutivos), which the system is going to be tested standing still, with different heighths and free Yaw orientation. The water tank of the LNDC lab is a square

shaped tank with 12 meters long, 6 meters wide and 7 meters deep, which provides a good initial evaluation environment, due the simple structure geometry.

The final evaluation of the proposed localization technique is going to be perform with a real dataset, aquired during the field trips to the Jirau Dam, in a real Stoplog hole. This setup provides a much more complex data, with noise provenient from mechanical vibrations, water turbulence and also a more complex *a priori* map.

Chapter 6

Proposed Bibliography

3D model representation

- Kai M Wurm, A Hornung, M Bennewitz, C Stachniss, and W Burgard. OctoMap: A probabilistic, flexible, and compact 3D map representation for robotic systems. *Proc of the ICRA 2010 workshop on best practice in 3D perception and modeling for mobile manipulation*, 2010
- Rudolph Triebel, Patrick Pfaff, and Wolfram Burgard. Multi-level surface maps for outdoor terrain mapping and loop closing. In *IEEE International Conference on Intelligent Robots and Systems*, pages 2276–2282, 2006
- Alberto Elfes. Occupancy Grids: A Stochastic Spatial Representation for Active Robot Perception. *arXiv preprint arXiv:1304.1098*, 2013
- K. M. Wurm, D. Hennes, D. Holz, R. B. Rusu, C. Stachniss, K. Konolige, and W. Burgard. Hierarchies of octrees for efficient 3D mapping. *2011 IEEE/RSJ International Conference on Intelligent Robots and Systems*, pages 4249–4255, September 2011
- C. Saona-Vázquez, I Navazo, and P Brunet. The visibility octree: a data structure for 3D navigation, 1999
- Alan H. Welsh, David B. Lindenmayer, and Christine F. Donnelly. Fitting and Interpreting Occupancy Models. *PLoS ONE*, 8, 2013
- Daniel Ribeiro Trindade and Alberto Barbosa Raposo. Improving 3D navigation techniques in multiscale environments: a cubemap-based approach. *Multimedia Tools and Applications*, May 2012

Bayesian Filtering

- David Salmond and Neil Gordon. An introduction to particle filters. *State space and unobserved component models theory and applications*, pages 1–19, 2005
- Sebastian Thrun, Dieter Fox, Wolfram Burgard, and Frank Dellaert. Robust Monte Carlo localization for mobile robots. *Artificial Intelligence*, 128(1-2):99–141, May 2001
- Sebastian Thrun. Particle Filters in Robotics. *Proceedings of Uncertainty in AI*, 1:511–518, 2002
- Olivier Cappe, Simon J. Godsill, and Eric Moulines. An Overview of Existing Methods and Recent Advances in Sequential Monte Carlo. *Proceedings of the IEEE*, 95(5):899–924, May 2007
- J. Carpenter, P. Clifford, and P. Fearnhead. Improved particle filter for non-linear problems. *IEE Proceedings - Radar, Sonar and Navigation*, 146(1):2, 1999
- Arnaud Doucet, Nando De Freitas, Kevin Murphy, and Stuart Russell. Rao-Blackwellised particle filtering for dynamic Bayesian networks. *Proceedings of the Sixteenth Conference on Uncertainty in Artificial Intelligence*, pages 176–183, 2000
- Greg Welch and Gary Bishop. An Introduction to the Kalman Filter. *In Practice*, 7:1–16, 2006
- T.B. Schon, Rickard Karlsson, and Fredrik Gustafsson. The marginalized particle filter in practice. *2006 IEEE Aerospace Conference*, 2006
- Arnaud Doucet and AM Johansen. A tutorial on particle filtering and smoothing: fifteen years later. In *Handbook of Nonlinear Filtering*, number December, pages 656–704. 2011
- Arnaud Doucet, Simon Godsill, and Christophe Andrieu. On sequential {Monte Carlo} sampling methods for {Bayesian} filtering. *Statistics and Computing*, 10(3):197–208, 2000
- Thomas Schön, Fredrik Gustafsson, and Per-johan Nordlund. Marginalized Particle Filters for Mixed Linear / Nonlinear State-Space Models. *IEEE Transactions on Signal Processing*, 53(7):2279–2289, 2005
- Sebastian Thrun. *Probabilistic robotics*, volume 45. 2002

Localization and underwater localization

- Francesco Maurelli, Yvan Petillot, Angelos Mallios, Pere Ridao, and Szymon Krupinski. Sonar-based AUV localization using an improved particle filter approach. *Oceans 2009-Europe*, pages 1–9, May 2009
- J.L. Blanco, J.a. Fernandez-Madrigal, and J. Gonzalez. A Novel Measure of Uncertainty for Mobile Robot SLAM with Rao Blackwellized Particle Filters. *The International Journal of Robotics Research*, 27(1):73–89, January 2008
- Antoni Burguera, Yolanda González, and Gabriel Oliver. Probabilistic sonar scan matching for robust localization. In *Proceedings - IEEE International Conference on Robotics and Automation*, number April, pages 3154–3160, 2007
- Antoni Burguera, Yolanda González, and Gabriel Oliver. Underwater SLAM with robocentric trajectory using a Mechanically Scanned Imaging Sonar. In *IEEE International Conference on Intelligent Robots and Systems*, number Dvl, pages 3577–3582, 2011
- Angelos Mallios, Pere Ridao, Emili Hernández, David Ribas, Francesco Maurelli, and Yvan Petillot. Pose-based SLAM with probabilistic scan matching algorithm using a mechanical scanned imaging sonar. In *OCEANS '09 IEEE Bremen: Balancing Technology with Future Needs*, 2009
- Angelos Mallios, Pere Ridao, David Ribas, Francesco Maurelli, and Yvan Petillot. EKF-SLAM for AUV navigation under probabilistic sonar scan-matching. In *IEEE/RSJ 2010 International Conference on Intelligent Robots and Systems, IROS 2010 - Conference Proceedings*, pages 4404–4411, 2010
- Emili Hernández, Pere Ridao, David Ribas, and Angelos Mallios. Probabilistic sonar scan matching for an AUV. In *2009 IEEE/RSJ International Conference on Intelligent Robots and Systems, IROS 2009*, pages 255–260, 2009
- Ling Chen. Towards Localization and Mapping of Autonomous Underwater Vehicles: A Survey. *School of Computer Science and Electronic Engineering University of Essex, United Kingdom*, Technical, 2011
- David Ribas. *Underwater SLAM for Structured Environments Using an Imaging Sonar*. 2001
- Armin Hornung, Kai M Wurm, and Maren Bennewitz. Humanoid robot localization in complex indoor environments. *Intelligent Robots and . . .*, (Mcl):1690–1695, 2010

- Guanghui Cen, Nobuto Matsuhira, and Ichiro Hagiwara. Mobile robot global localization using particle filters. *2008 International Conference on Control, Automation and Systems*, pages 710–713, October 2008
- Philipp Woock and Christian Frey. Deep-sea AUV navigation using side-scan sonar images and SLAM. *Oceans’10 Ieee Sydney*, pages 1–8, May 2010
- Andreas Nüchter, Kai Lingemann, Joachim Hertzberg, and Hartmut Surmann. 6D SLAM - 3D mapping outdoor environments. In *Journal of Field Robotics*, volume 24, pages 699–722, 2007
- Jakob Schwendner. *Embodied Localisation and Mapping*. PhD thesis, 2013
- H Durrant-Whyte and T Bailey. Simultaneous localization and mapping (SLAM). *IEEE Robotics & Automation Magazine*, 13(2):99–116, 2006
- Liam Paull, Sajad Saeedi, Mae Seto, and Howard Li. AUV Navigation and Localization: A Review. *IEEE Journal of Oceanic Engineering*, 39(1):131–149, January 2014
- W. Cheng, a. Y. Teymorian, L. Ma, X. Cheng, X. Lu, and Z. Lu. Underwater Localization in Sparse 3D Acoustic Sensor Networks. *2008 IEEE INFOCOM - The 27th Conference on Computer Communications*, pages 236–240, April 2008

Sonar and sensor modelling theory

- Rudolf H Riedi. An Introduction to Statistical Signal Processing An Introduction to Statistical Signal Processing . Robert M. Gray and Lee D. Davisson . Cambridge, U.K. : Cambridge University Press , 2004 . ISBN 0-521-83860-6 . xiv + 463 pp. \$75.00 . *Journal of the American Statistical Association*, 101:1317–1317, 2006
- Alan H. Welsh, David B. Lindenmayer, and Christine F. Donnelly. Fitting and Interpreting Occupancy Models. *PLoS ONE*, 8, 2013
- P G Auran and O Silven. UNDERWATER SONAR RANGE SENSING AND 3D IMAGE tively organise local 319 map feasibility its way on. *Image (Rochester, N.Y.)*, 4(3):393–400, 1996
- Tildon Glisson, Charles Black, and Andrew Sage. On Sonar Signal Analysis. *IEEE Transactions on Aerospace and Electronic Systems*, AES-6(1):37–50, January 1970

- Paul C Etter. *Underwater Acoustic Modeling and Simulation*. 2003
- Jorg Muller, Axel Rottmann, Leonhard M. Reindl, and Wolfram Burgard. A probabilistic sonar sensor model for robust localization of a small-size blimp in indoor environments using a particle filter. *2009 IEEE International Conference on Robotics and Automation*, pages 3589–3594, May 2009
- Jonathan Fournier, Benoit Ricard, and Denis Laurendeau. A Volumetric Sensor for Real-time 3D Mapping and Robot Navigation. 6230:623009–623009–12, May 2006
- Andrea Trucco, Matteo Garofalo, Stefania Repetto, and Gianni Vernazza. Processing and analysis of underwater acoustic images generated by mechanically scanned sonar systems. In *IEEE Transactions on Instrumentation and Measurement*, volume 58, pages 2061–2071, 2009
- John J. Leonard and Hugh F. Durrant-Whyte. *Directed Sonar Sensing for Mobile Robot Navigation*. Springer US, Boston, MA, 1992
- Antoni Burguera, Gabriel Oliver, and Yolanda González. Range extraction from underwater imaging sonar data. In *Proceedings of the 15th IEEE International Conference on Emerging Technologies and Factory Automation, ETFA 2010*, pages 10–13, 2010
- M.J. Chantler, D.B. Lindsay, C.S. Reid, and V.J.C. Wright. Optical and acoustic range sensing for underwater robotics. *Proceedings of OCEANS’94*, 1, 1994
- Michael Bosse and Robert Zlot. Continuous 3D scan-matching with a spinning 2D laser. *2009 IEEE International Conference on Robotics and Automation*, pages 4312–4319, May 2009
- Jean Paul Marage and Yvon Mori. *Sonar and Underwater Acoustics*. 2013

Chapter 7

Workplan

- Particle description - 1 week
- Implement Kalman filter for the depth and inclinometer (alpha beta filter)
incl. Wave removal + testing - 1 month
- Build apriori map - 1 month
- Full review particle filter algorithm + mock test - 1 month
- (Implement state of the art - 1 month)
- Write down a full sensor model - 6 months
- Postprocess LNDC logs - 2 week
- Postprocess Jirau logs - 2 week

Chapter 8

Conclusion

In this master thesis proposal, it was described all the relevant aspects to reach the state of the art in underwater localization and implement a localization algorithm in an actual system. The main topics to be explored in this work are the current underwater localization techniques, the bayesian filtering techniques and particle filtering, construction of a priori 3D environment occupancy models, underwater acoustic theory and sensor modelling theory.

Finally, the main contribution of the proposed master thesis will be the implementation of a Rao-Blackwellized Particle Filter with 6 DoF using a inexpensive sonar and the derivation of it sensor model.