

MIDDLEWARE
DESIGN ET AUTONOMISATION D'UN KART
PROJET

KART

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1 Introduction

In the middleware module of UV 4.1, we are asked to implement the ROS middleware to drive a small car built in the previous semester. This small project consists in driving around an athletics track in an autonomous way. To carry out this project we have various sensors such as the Inertielle control unit which estimates the heading of the car and its acceleration, the camera, the gps which will allow us to estimate the position of the small car. We also have an on-board computer, in this case a raspberry.

2 Hardware

2.1 Main Board

For this project, we decided to use a *Raspberry Pi 3B+* as main board. It will let us plug some sensors and control the motors of the car according to the wanted behavior we have programmed.

On this Raspberry Pi, we need to choose an Operating System. our choice was to use Ubuntu Mate because of its simplicity to install and its polyvalence. It will let us do everything we want, like plug sensors, code any program to control our car, ... A preview of Ubuntu Mate is shown on the 1

As an imposed figure for our project, we decided to use the *Robot Operating System* (ROS) as Middleware for this car. It's going to offer us some practical tools to code our programs easily. There is also some useful community shared tools like *rqt* or *key_teleop* we will use in this project. Then we decided to setup a graph node and to code these nodes in order to build our system as shown on the following graph node.

2.2 Sensors

This section present the main hardware configuration of the car. For this project we have to choose which sensors we want in our car in the following list.

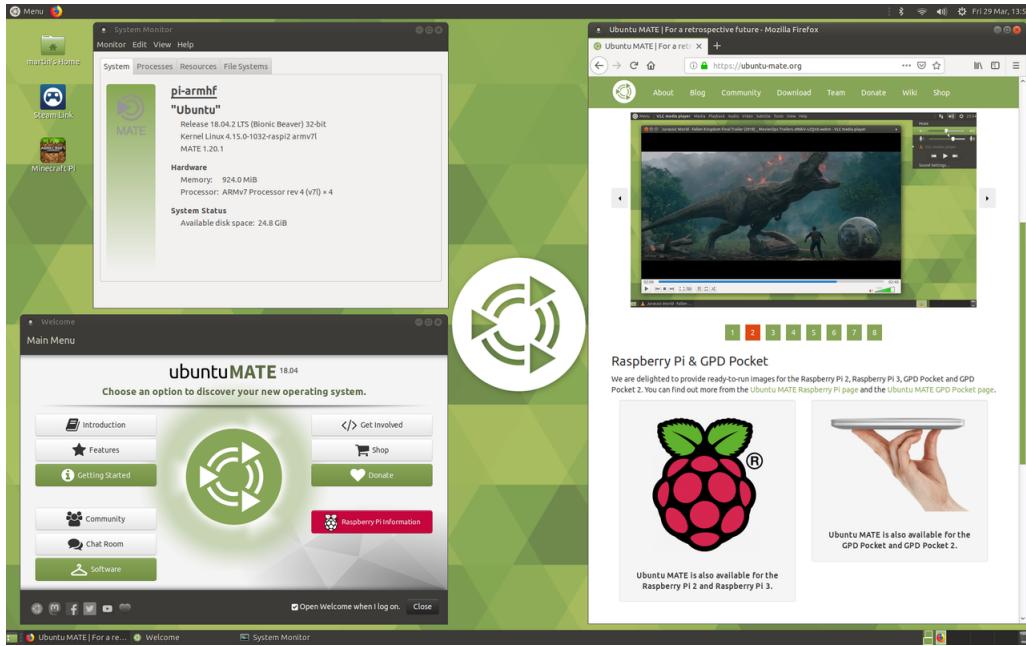


Figure 1: Ubuntu MATE screen

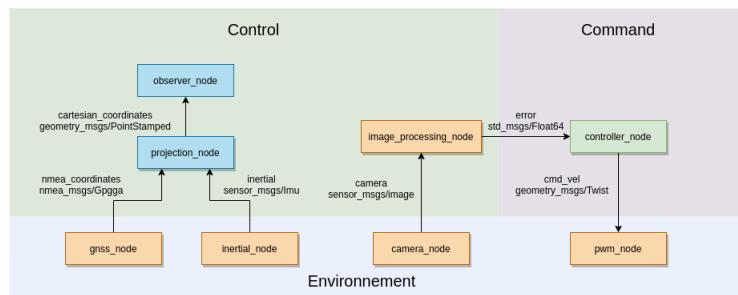


Figure 2: Graph node of the kart

As it's shown we decided to choose neither the GNSS nor the Inertial Units, mainly because of their accuracy.

Actually, for our problem we found that an accuracy of 1 meter for the GNSS is too large because the car need to run in a 0.8 meter wide racing

Sensors	Used	Dope-Level
GNSS	✗	🐢 💥 ⚡ 💩
Inertial Unit	✗	🌈
Pi camera	✓	🦄

Figure 3: The Sensors list available on our GitHub

lane, following a line. This sensor is also not able to know if the car position is correct.

For the *Inertial Unit*, we found that this sensor is too noisy to give us any useful information about the state of our car. For instance the consecutive integration of the acceleration in order to get the speed and the position of the car leads to an important drift effect on our data. So this sensor is not currently able to give any correct information about the state of the car. Moreover, the acceleration of the car could be quite good after filtering if we only needed it. In our case the only useful information is to know the position of our car in relation to the line.

That's why we decided to focus our attention on the camera. Because we are using a *Raspberry Pi 3B+*, the camera we have chosen is the official camera which can be plugged on the dedicated port on the board. This sensor is perfectly suited to our problem, because with an appropriated image processing we will be able to detect the line and to correct the car trajectory.

2.3 Configuration

Now we will explain how we configured our sensors in our project, to let them communicate with the software and with the car.

2.3.1 Pi Camera

The configuration of the camera on the raspberry pi is relatively simple. We used the *raspi-config* utility to configure the camera. That's how we set up the camera on the Raspberry Pi.

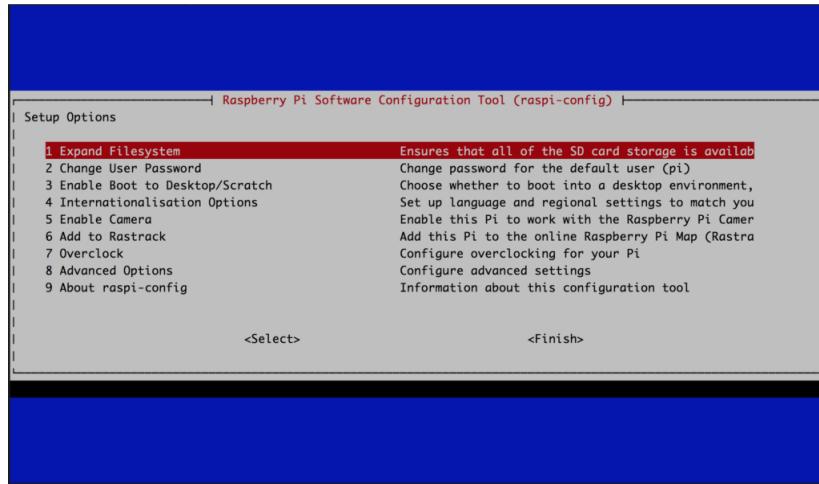


Figure 4: raspi-config utiliy on the Raspberry Pi

2.3.2 Hardware PWM

On Raspberry Pi board there is a lot of way to generate pwm signals. The most of the time, these methods are software based and so they are not accurate. With a lot of searches, we found a website who speak about the raspberry pi's hardware pwm signals. There is apparently an hardware pwm generator used by the bord to generate sounds. It's better to use hardware generated pwm, because if the processor has a slow down and the interrupt is not correctly handled, the pwm duty cycle will not be very accurate and the car will not be able for instance to follow a straight line, because the bearing of the car is controloed by a servomotor with a pwm signal.

So we decided to use this tutorial : <https://disconnected.systems/blog/pi-zero-w-rover-setup/#moving-the-robot>, which explain us how to setup pwm signals on the Raspberry Pi, and how to correctly configure

the files to have a standard pwm signal which is generated. Then we need to give the rights to users for reading and writing in these files. All these bash command are in *init_pwm.sh*.

Then we have to add some automation. So we created a *crontab* rule. That will automatically create all the required files and allow the permissions to every users. We just have to write in the file *duty_cycle* a value between 1.000.000 and 2.000.000, and the Raspberry Pi will read and adjust pwm signals in real time.

Last but not least, we setup an autologin in order to open a session automatically when the Raspberry Pi boot. That's very usefull in order to launch our programms easily on boot and without any keyboard, mouse or monitor.

3 Mechanical Architecture

3.1 Reasons of an Overhaul

The original architecture of the vehicle was already a good basis for the realisation of a line tracking car. However, manual manoeuvring tests by remote control have revealed some driving faults. Misconduct that could be embarrassing in an autonomous driving mission. Given that our robot will certainly not be as adaptive as a human in its driving, it is interesting to ease the maneuvers and correct some mobility deficiencies.

3.1.1 Problems identified and changes to be expected

Some of the shortcomings noted are listed as follows:

- Loss of control due to high front wheel slippage during high speed turns.
- In case of sharp turn, locking of the inner wheel on the bend caused by the central component box.
- Lack of firmness of the front wheel guiding created by a large backlash.
- Uncontrolled spinning of the rear wheels when acceleration from a standstill or deceleration from high speeds.

And some others component have to be implemented to the initial structure in order to be an optimal support for autonomous driving such as:

- The installation of a camera to see and locate the line to follow.
- Hosting a Raspberry Pi card to manage the control computations.
- Keep space for the rest of the essential components such as the battery, ESC, cables, etc.

3.2 Front Wheel Steering

bla bla bla bla

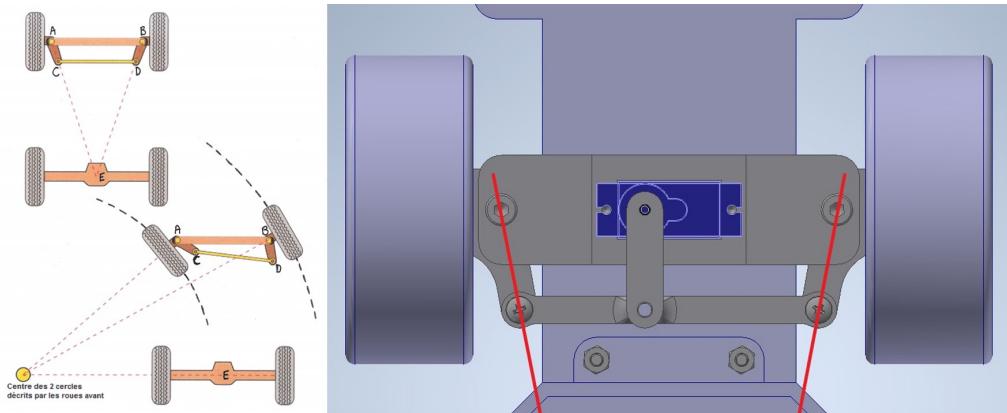


Figure 5: Steering system

3.3 Front Camera Support

bla bla bla bla

3.4 Component Hosting Box

bla bla bla bla

3.5 Expectations and Reality

bla bla

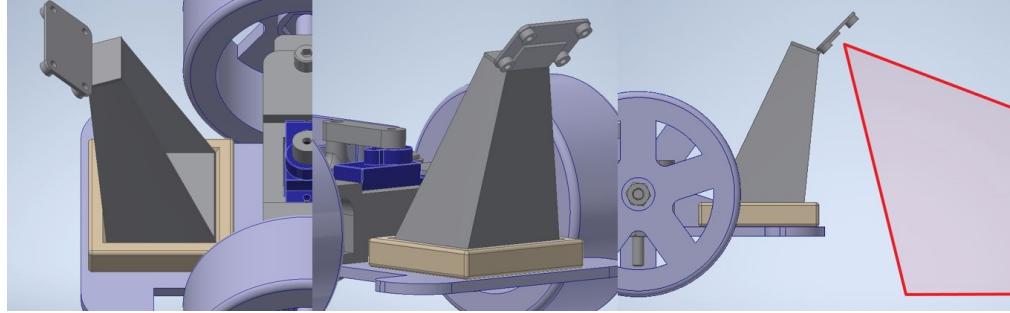


Figure 6: Front camera support

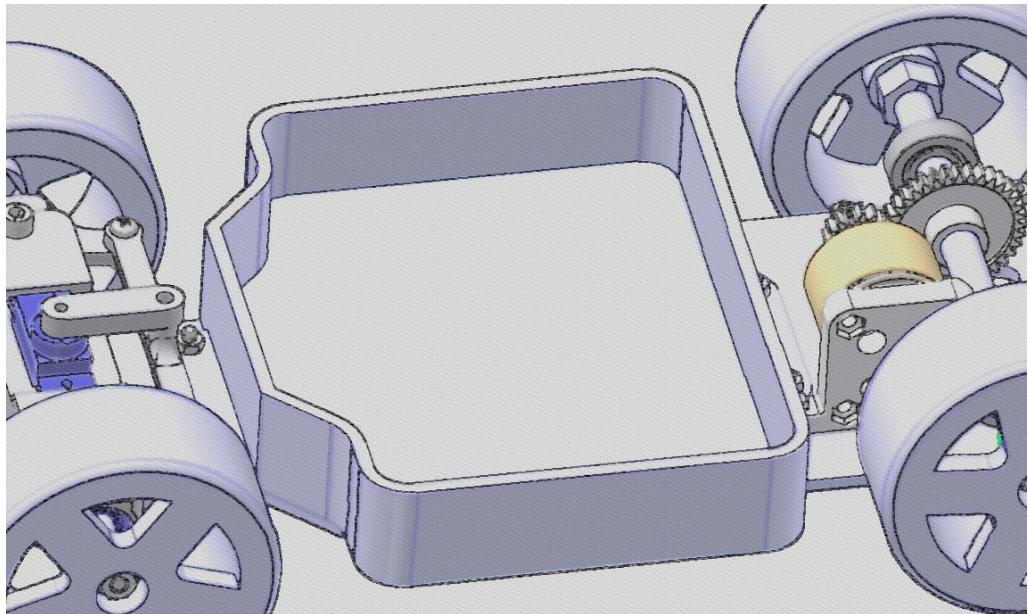


Figure 7: Central component box

4 Image Processing

4.1 Data logging

For the first part of this image processing, we had to collect images. So first we went into the environment in which the robot was going to evolve.

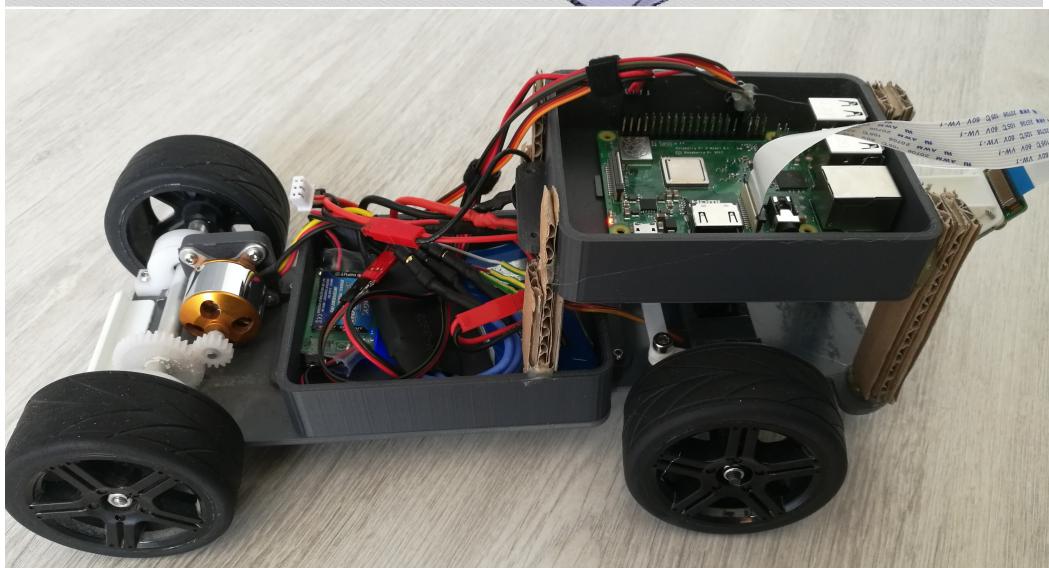
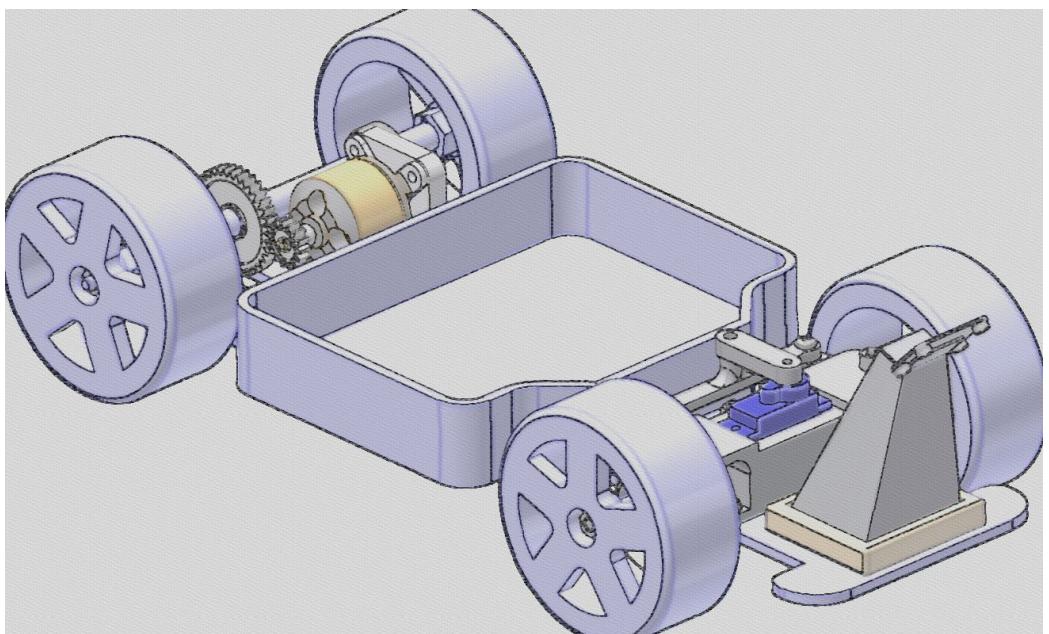


Figure 8: Expectations and reality

Sensors	Used	Dope-Level
GNSS	✗	🐢💥➡️💩
Inertial Unit	✗	🌈
Pi camera	✓	💡

Figure 9: collecting data

4.2 pre binarization treatment

In this part, we first had to perform a pre-binarization treatment in order to reduce post-binarization noise. So we used a Gaussian filter to blur the image.

4.3 binarization

Since the line we wanted to mark is white. An effective treatment is simply to switch to grey level. So for binarization, we switch the image to a grey level and threshold for a grey level that we have determined empirically.

4.4 post binarization treatment

In this part we performed a morphological treatment. There was still a lot of noise after binarization. So we made an opening. With a kernel in the shape of a rectangle (since it was the most efficient for this treatment). At the end of this treatment we obtain a well defined line which crosses the screen.

4.5 find the center of the line

In this last part, the contours are marked using a gradient method. Then the contours are sorted from the smallest to the largest. We recover the largest contour. And we recover the coordinates of the barycentre of the contour. Then the error is the difference between the center of the image and the coordinates of the pixel.

5 VREP Simulation

5.1 Explaination

Based on the practical work made with Benoit Zerr, we got a realisitic and a track with a line in the middle of the track. The simulated in VREP car is equiped with a camera. Our VREP simulator is interfaced with ROS through the piece of LUA code : `rc_car_control_ros.lua` et `rc_car_onboard_cam.lua`.

The VREP camera send a through a topic `image`, a picture of the front of the car. Then the `camera_node.py` subscriber process the image with OpenCV and return the position of the center line in the windows, then it compute an error, we want the line in the middle of the windows. This error is sent through a topic `error`, the controller is suscribed to this topic, it generate a command thanks to a PD command law which is sent to VREP.

And now the car can loop forever, and achieve the perfect run in 1'15" !   

Here is the video link :

https://www.youtube.com/watch?time_continue=10&v=_vIXo1TvG0w&feature=emb_logo 

6 Localization

6.1 Problem in Localization

In this project we decided to realize the challenge by a visual servoing. That is to say a line tracking by the camera. This type of servo control only works if the robot has at each moment the line to follow in its field of vision. This kind of problem can happen when the robot misses the turn. If the robot ever loses sight of this line, it is obvious that the challenge will not be successful.

To compensate for this unforeseen situation, we have implemented a control law that will be used as soon as the robot loses the line of sight sound. This control law requires knowledge of the robot's state.

6.2 Robot status estimation

To compensate for this unforeseen situation, we have implemented a control law that will be used as soon as the robot loses the line of sight sound. This control law requires knowledge of the robot's state.

Where (x,y) is the position of the robot, θ is its heading, v its speed and δ the angle of the front wheels. To be able to apply the Kalman filter we need a linear state representation of the robot. To make these equations linear, we have considered the angle θ of the robot as an input (since it is given by the inertial unit). This removes equation 3. Then we linearized the equation into $xhat$ ($xhat$ being the new state to be determined). $Xhat$ consists of x,y, θ, δ . The file available on our github shows the function of this Kalman filter.

$$\begin{pmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \\ \dot{v} \\ \dot{\delta} \end{pmatrix} = \begin{pmatrix} v \cos \delta \cos \theta \\ v \cos \delta \sin \theta \\ v \sin \delta \\ u_1 \\ u_2 \end{pmatrix}$$

Figure 10: State of the robot

7 Explications

8 Resultats

9 Discussions et Conclusion

To conclude, we think that we already have a strong basis for this project, particularly with the ROS structure which is already setup. Then we have to perform test with the real system because the simulation is correctly working and a good proof of concept, but we know too that simulation and reality are always different. We have to do some improvements on our project too, like adding an mission controller in order to control the robot in some cases where we couldn't use the line following command, for instance when the camera is not seeing any lines, and we have to add a GUI in order to show the kart state like his position, its speed and the circuit.

10 Improvements

10.1 Prerequisites

This part will be implemented in the future and are for now only some good ideas we have for this project. We are first focused on the achievement of the simple control of this car.

10.2 Mission

We have established a simple command for our kart based on image processing. This command law is quite easy to set up and fully functional as we could see. However this command is efficient only if the camera is able to see a line. But actually the car could be unable to see a line, particularly when the car will accelerate in order to run a lap as quick as possible. Our strategy on this topic is to run a first lap very slow and to drop virtual GPS tags on a map while the car is following the line. So after a lap we will get a whole map of the circuit and at any given time the car will have to determine if a line is available on the camera, and if the car could follow this line, else the car will have to reach this line again with the help of the GPS tags. Then after the first lap, the car will be able to accelerate and for instance some times lost the line because it will be able to reach again the circuit with the GPS tags.

10.3 GUI

We found quite interesting to have a beautiful Graphical User Interface for our car. We are thinking about a frontend solution in order to show dynamical parameters such as acceleration and steep of the car, and a map to show the position with the classical circle to represent the uncertainty of this position. Moreover it could be interesting to have the camera output available on this GUI. The easiest way to display a map and some GNSS coordinates on Python is to use the *folium* package based on the *OpenStreetMap* map. We are now working on a bind between *ROS*, *folium* and the web framework *Django* to sum up these informations on a static webpage. However, we don't own any GNSS or Inertial Units to prepare hardware nodes. That is also not our priority, but it can be a very useful tool.

A Appendix