# Localization and mapping From EKF to SLAM

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#### Roadmap

1 The perception problem

The mapping problem

#### Roadmap

The perception problem

**2** The mapping problem

Definition

To be able to operate autonomously, a robot must continuously answer the following questions:

- Where am I?
- What is this place?

Thus, it must be able to perceive the environment, gathering information useful to:

- localize itself, *i.e.*, continuously estimate its pose as both position and orientation in 3D space;
- map the environment, i.e., build a representation of the environment useful to navigate within it.

Challenges

The perception problem is challenging because:

- the environment may be **partially observable**, *i.e.*, the robot can only perceive a **subset** of it, and need to update its information in real time;
- the environment may be **dynamic**, *i.e.*, it can change over time;
- measurements are always subject to noise.

The perception problem is usually solved by **sensor fusion**, *i.e.*, combining information from **multiple sensors** to obtain a more **accurate** and **reliable** estimate of the environment, possibly accounting for **sensor faults**.

Tools for the job

The tools that robots use to gather measurements from the environment are called sensors.

They can be classified as:

- proprioceptive, *i.e.*, measuring robotic interaction with the environment (*e.g.*, encoders, GPS, IMUs);
- exteroceptive, i.e., measuring the environment itself (e.g., cameras, LiDARs, radars);
- interoceptive, i.e., measuring the robot's internal state.

Tools for the job

As any other measurement tool, sensors are based on **physical principles** and **energy exchanges**, translating the information they gather into **electrical signals** that can be acquired and/or processed by a computer.

They are usually characterized by at least:

- a digital or analog encoding of the measurement;
- a frame of reference in which the measurement is expressed;
- accuracy and uncertainty parameters.

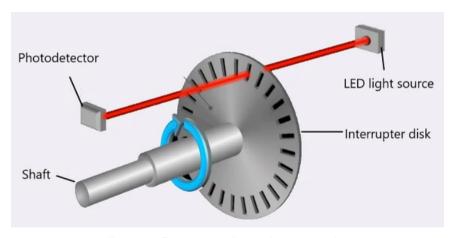


Figure 1: Rotary encoder working principle.



Figure 2: GPS module for drones.



Figure 3: Inertial Measurement Unit (IMU).



Figure 4: Light Detection and Ranging (LiDAR) sensor.



Figure 5: ZED 2i stereo camera.

#### Roadmap

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2 The mapping problem

## The mapping problem Definition

Using exteroceptive sensors, a robot can gather information about the environment, which can be used to build a map of it.

A map is a representation of the environment, in a format that the robot can understand, parse, and store.

The utimate goal of mapping is twofold:

- to enable the robot to **localize itself** within the environment;
- to enable **safe navigation** of the robot within the environment.

## The mapping problem Challenges

Given the utility requirement of a map, the mapping problem must be **continuously solved in real time**.

Thus, it is challenging because:

- routines must be efficient, and run at a sufficiently high rate;
- the map must be accurate, and reliable;
- the map must be in a format that is as much easy to load and parse as possible, taking up as little memory as possible;
- the map must stay **up-to-date**, and **consistent** with the environment.

#### The mapping problem

Tools for the job

The most important tool for the mapping problem is the occupancy grid, a representation of the environment as a grid of cells, each of which is occupied or free.

The occupancy grid is a **probabilistic** representation, where each cell is associated with a **probability** of being occupied or free.

The occupancy grid is usually built using **LiDAR** or **camera** depth data, and is updated in real time as the robot moves.

The occupancy grid is the most common representation for local and global maps.

To efficiently store an occupancy grid, **tree-like data structures** are often employed (*e.g.*, **octrees**).

### The mapping problem

Tools for the job

The second most important class of tools are **navigation algorithms**, which use the map to plan a **safe** and **efficient** path for the robot to follow.

The definition of such algorithms involves **geometry**, as well as **optimization** and **search** techniques.

They usually rely on two mathematical subjects:

- **topology**, to define the **connectivity** of the map (*e.g.*, Voronoi tessellation);
- graph theory, to define the best way of moving from one free cell to another (e.g., Dijkstra's,  $A^*$  algorithms).

#### The mapping problem

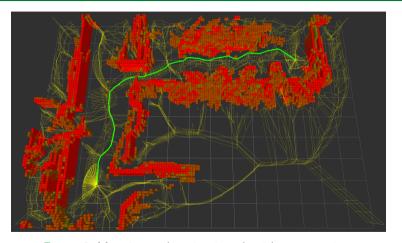


Figure 6: Mapping and navigation algorithms execution.

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#### Rigid transformations

When a robot moves in space, it is important to keep track of its **position** and **orientation** with respect to a **reference frame**.

Sensors measuring this information, as well as many more, are **mounted** on the robot, in fixed positions and orientations.

To process these measurements, they must first be **transformed** from the **body frame** into a common reference frame, usually called:

- world frame (world origin), in the case of global localization;
- local frame, or odom frame (robot starting point), in the case of local localization.

Such **rigid transformations** are **isometries**. They must be applied to almost every sensor measurement, and are usually **composable**.

We would like the middleware to provide tools to do this almost automatically...

#### The tf2 library

tf2 is the **standard ROS 2 library** to handle rigid transformations.

#### It allows to:

- broadcast and listen to transformations, thanks to appropriate buffering subscribers;
- transform any kind of sensor data from one frame to another, making efficient computations in C++ code relying on the Eigen mathematical library;
- broadcast robot descriptions from URDF files, listing links and joints and how they are connected, resulting in a tree-like structure;
- command-line tools to inspect the current tree status, and broadcast custom transformations.

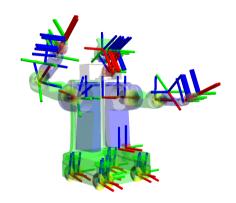


Figure 7: Example of robot description with tf2.

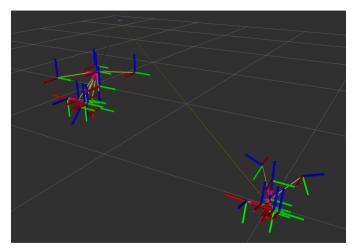


Figure 8: Broadcasted robot descriptions, plus real-time transformations given by localization systems.