

# Vision-based Navigation Exercise 1

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## 1 Part 1

- First line appends the path of the folder `cmake_modules` to a variable called `"CMAKE_MODULE_PATH"`
- Second line of commands set the C++ standard to C++14, and enforce it for compilation. `"Required"` statement ensures that if C++14 is not installed, project won't compile. Also the flag `C++ extensions` is set to off.
- Third line of commands define C++ compilation flags for different build configurations (Debug, Release, Relwithdebinfo).

First statement is used to set flags when compiling the project in Debug configuration. `"-O0"` disables compiler optimizations and `"-g"` is the compilation flag for debugging. `"-D"` is used for passing preprocessing directives such as define statements etc. Here it is used to define `"EIGEN_INITIALIZE_MATRICES_BY_NAN"`.

Second statement is used to set flags when compiling the project in Relwithdebinfo configuration, which maximizes speed, but keeps debugging information. `"-O3"` enforces the compiler to optimize the code as much as possible and `"-g"` is given for generating debug information to inspect later. `"-DNDEBUG"` flag is given such that no debug output is enabled. Flag for initializing Eigen matrices to nan is also passed.

Third statement is used to set flags when compiling the project in Release configuration which maximizes speed. `"-O3"` maximizes the compiler optimization, and `"-DNDEBUG"` disables debug output.

Final statement appends multiple C++ compilation flags to the environment variable `CMAKE_CXX_FLAGS`. `"-ftemplate-backtrace-limit=0"` sets maximum number of template instantiation notes for a single error or warning to 0. `-Wall` enables all warnings, `"-Wextra"` appends the extra warning flags stored in `{EXTRA_WARNING_FLAGS}` and `-march` flag with its argument passes the system's cpu architecture to compiler.

- Final commands defines the executable and its source code, then links the necessary libraries to that executable

## 2 Part 2

Let  $t = |w| = \theta$  and  $v = \frac{w}{|w|}$  then  $\hat{w} = \hat{v} * t$  and  $\hat{v}^2 = vv^T - I, \hat{v}^3 = -\hat{v}, \dots$   
(From the slide 40 in lecture slides)

After expanding the series  $\sum_{n=0}^{\infty} \frac{1}{(n+1)!} (\hat{w})^n$ , the output can be grouped similar to the example in slide 40:

$$\sum_{n=0}^{\infty} \frac{1}{(n+1)!} (\hat{w})^n = I + \left( \frac{t}{2!} - \frac{t^3}{4!} + \frac{t^5}{6!} - \dots \right) \hat{v} + \left( \frac{t^2}{3!} - \frac{t^4}{5!} + \frac{t^6}{7!} - \dots \right) \hat{v}^2$$

Multiplying and dividing both terms with  $t$ :

$$\sum_{n=0}^{\infty} \frac{1}{(n+1)!} (\hat{w})^n = I + \underbrace{\left( \frac{t^2}{2!} - \frac{t^4}{4!} + \frac{t^6}{6!} - \dots \right)}_{\frac{1-\cos(t)}{t}} \hat{v} + \underbrace{\left( \frac{t^3}{3!} - \frac{t^5}{5!} + \frac{t^7}{7!} - \dots \right)}_{\frac{t-\sin(t)}{t}} \hat{v}^2$$

Inserting  $t = \theta$  and  $\hat{v} = \frac{\hat{w}}{t}$ :

$$\sum_{n=0}^{\infty} \frac{1}{(n+1)!} (\hat{w})^n = I + \frac{1-\cos(\theta)}{\theta^2} \hat{w} + \frac{\theta-\sin(\theta)}{\theta^3} \hat{w}^2$$

## 3 Part 3

### – Why would a SLAM system need a map?

Having a map with some distinguished landmarks ensures that the system can recover when it starts making bad predictions by resetting its state and eliminating the erroneous predictions. It also helps in tasks like path planning and visualization.

### – How can we apply SLAM technology into real-world applications?

We can apply SLAM to the applications which require a globally consistent map generation. Typically, we employ an agent with sensors to perceive the world, and it observes its surroundings while running the SLAM algorithm suited for its sensors and creates a map of its observations while tracking its own location in this map.

### – Describe the history of SLAM.

SLAM history consists of 3 big time periods :

- \* Classical age (1986 - 2004), where the main probabilistic formulations for SLAM came to surface. Approaches based on Extended Kalman Filters, Rao-Blackwellised Particle Filters and maximum likelihood estimation are introduced. In this era, basic challenges with efficient and robust data association are also studied.

- \* Algorithmic analysis age (2004 - 2015), where the fundamental properties of SLAM, such as observability, convergence and consistency, are studied, Importance of sparsity for efficient computation is realized, and main open-source SLAM libraries were developed.
- \* Robust perception age (2015 - Present), which is the current era we are in as the paper claims. This era focuses on robust performance (system ability to perform well under various environments for an extended amount of time), extending systems's understanding beyond basic the geometry reconstruction to semantics and physics of the environment and resource awareness and task-driven perception, which are the attributes of the system which enables it to self adjust the computation load given the available resources and the the task at hand.