Cyber-physical systems programm

mirror_mod.mirror_object

collect to mirro

rtext.scene.objects.action "Selected" + str(modifies

bpy.context.selected_ob ta.objects[one.name].se

int("please select exacti

OPERATOR CLASSES ---

X mirror to the selected

ject.mirror_mirror_x"

rror ob.select = 0

vpes.Operator):

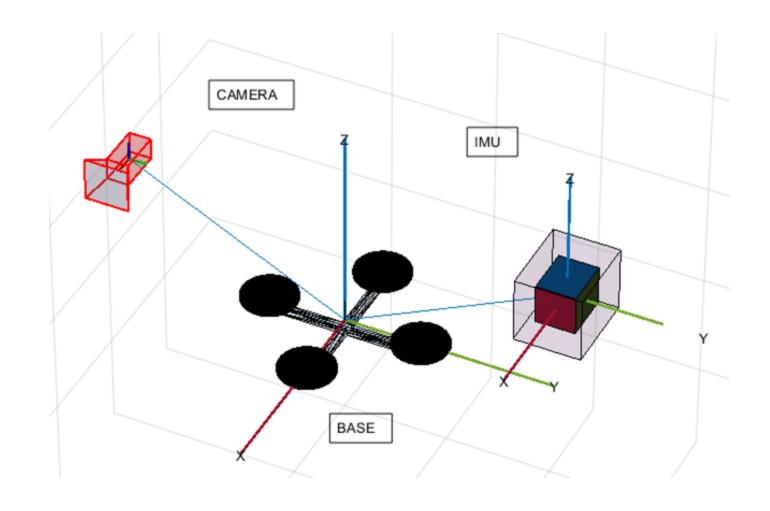
Visual Inertial Odometry for a Crazyflie drone

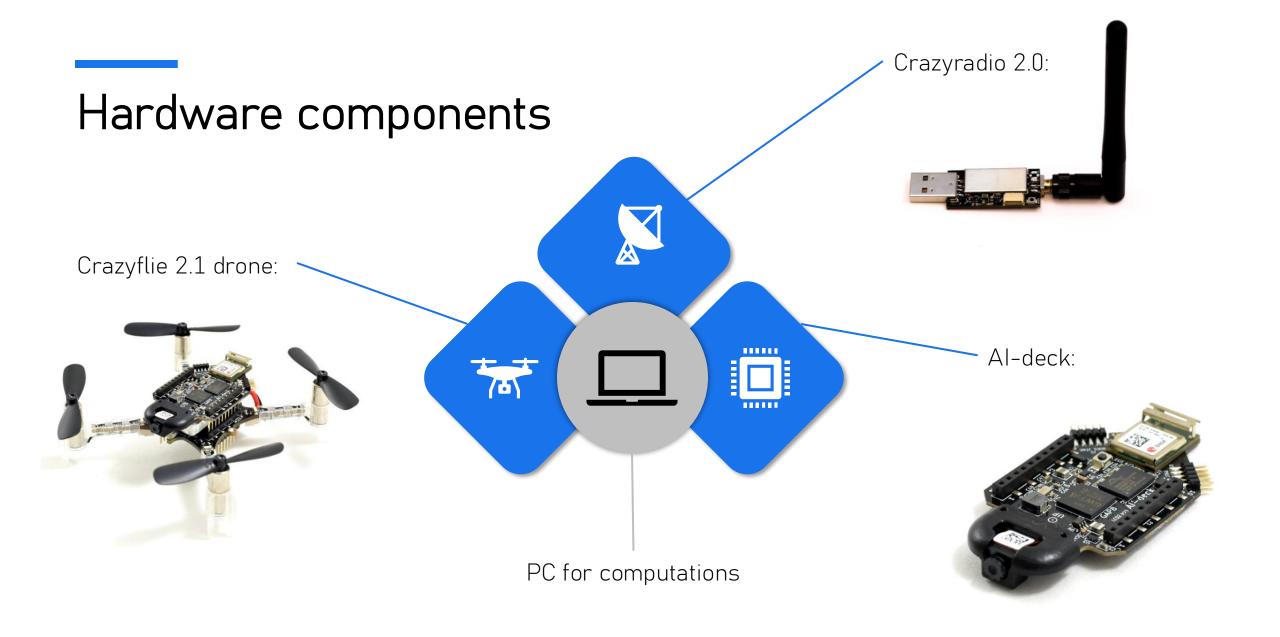
Group components:

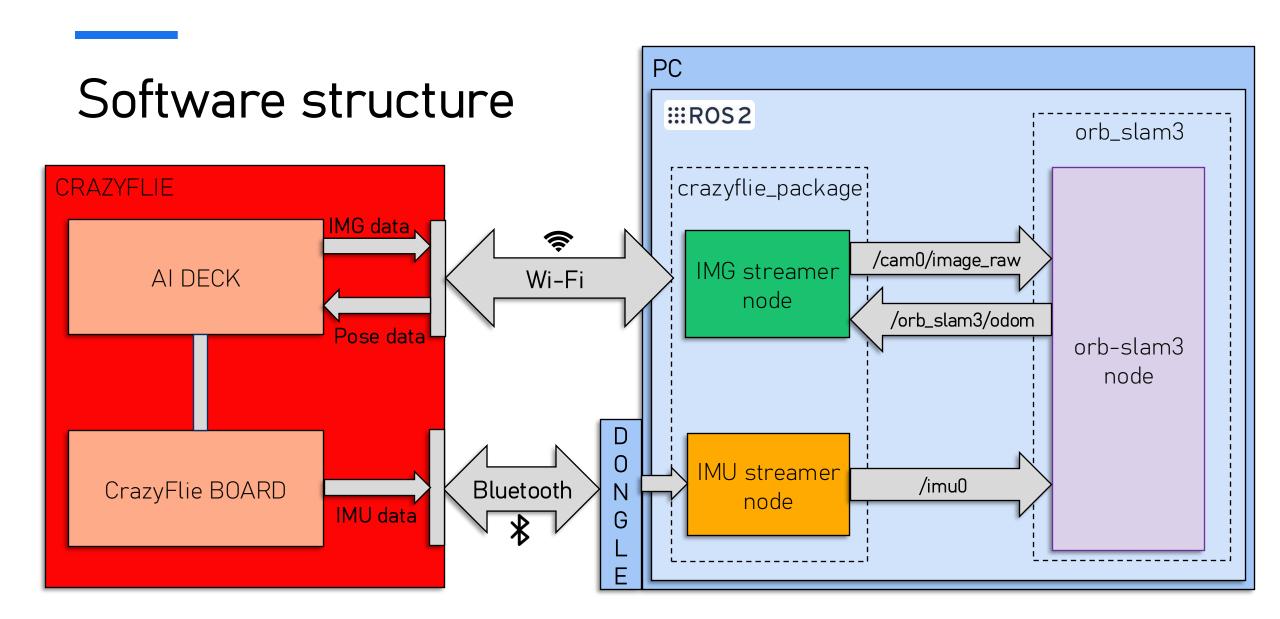
Crivellari Daniele Samorì Filippo Ugolini Filippo

Visual Inertial Odometry Algorithm

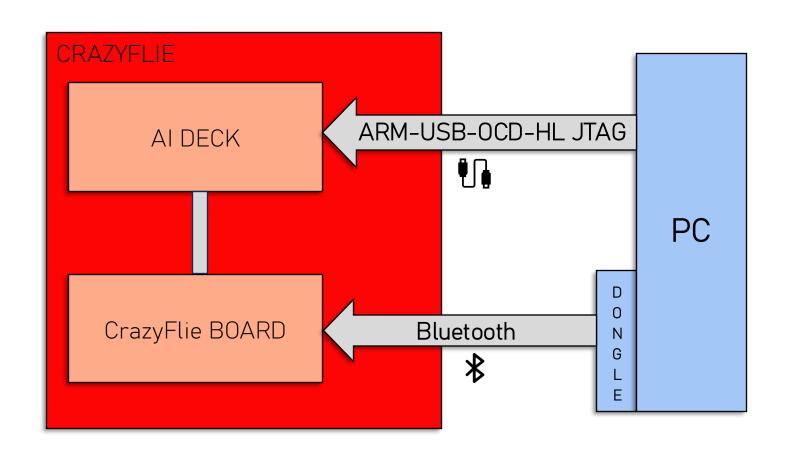
- The Visual Inertia Odometry (VIO) is a localization technique for the estimation in real time of the position and orientation of an autonomous device in an environment.
- It combines visual information acquired throughout cameras and inertial data provided by IMU sensors.



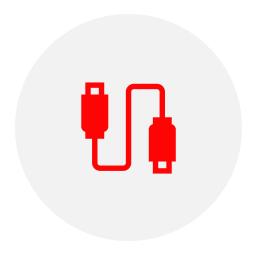




Section 1: Drone Setup



Firmware Setup



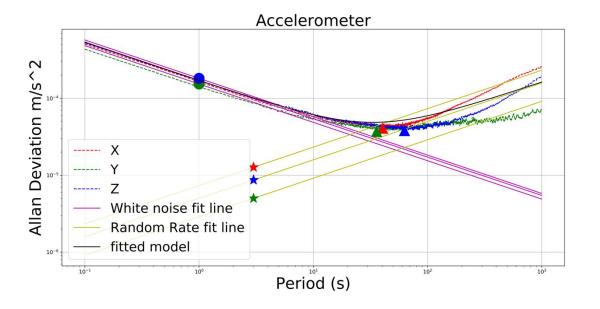


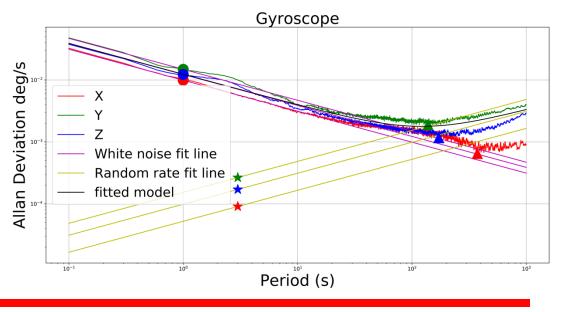
BUILD AND FLASH OF THE GAP8
BOOTLOADER WITH AN **OLIMEX ARM- USB-OCD-HL JTAG**

FLASHING OF THE FIRMWARE AND THE DEVELOPED APPLICATION FOR THE GAP8 VIA THE CRAZYRADIO DONGLE

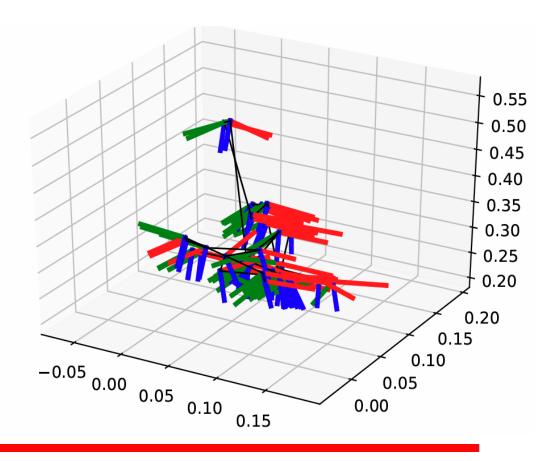
IMU Calibration

| Sensor | Noise Density | Random Walk |
|---------------|----------------------------------|------------------------------------------------|
| Accelerometer | $0.000176 \frac{m}{\sqrt{s^3}}$ | $3.4328 \times 10^{-5} \frac{m}{\sqrt{s^5}}$ |
| Gyroscope | $0.000260 \frac{rad}{\sqrt{s}}$ | $2.0043 \times 10^{-5} \frac{rad}{\sqrt{s^3}}$ |



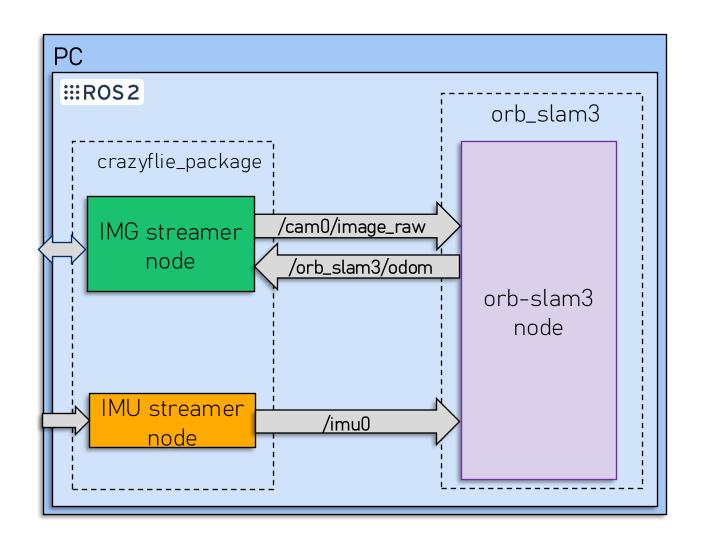


IMU + Camera Calibration



| Parameter | Value | |
|------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------|--|
| Focal length | [182.040108, 181.960582] | |
| Principal point | [161.919551, 153.772955] | |
| Radial distorsion coefficients | [-0.072991, -0.005429] | |
| Tangential distorsion coefficients | [-0.000866, 0.000639] | |
| Rotation matrix | $\begin{bmatrix} -0.00658 & -0.99998 & -0.00051 \\ -0.22460 & 0.00198 & -0.974445 \\ 0.97443 & -0.00629 & -0.22461 \end{bmatrix}$ | |
| Translation vector | $[-0.0292 1.1811 0.6743]^T$ | |

Section 2: ROS2 environment



01

The *imu_streamer* node is initiated and it connects to the drone via the Crazyradio dongle

02

The node starts logging IMU data from the drone:

- Accelerometer data
- Gyroscope data

IMU STREAMER NODE

03

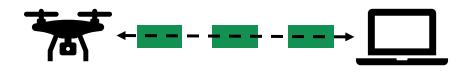
A ROS2 message of type *Imu()* is instantiated and filled with the logged values

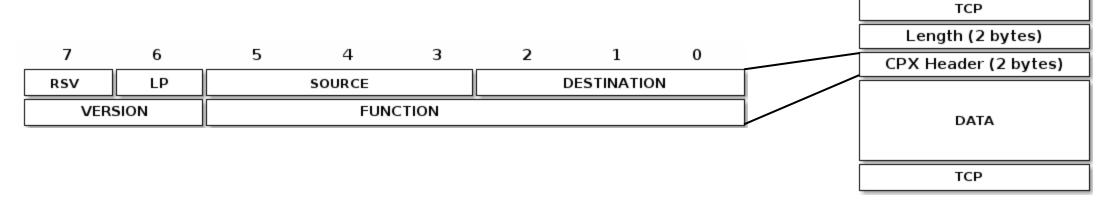
04

The Imu message is then published in the /imu0 topic

WI-FI COMMUNICATION PROTOCOL

The image communication is performed using the TCP protocol, with data packets structured according to the custom convention CPX. Due to the images size, they are split into several smaller chunks.





01

The *img_streamer* node is initiated, and it connects to the Al-deck via WIFI

02

The node reconstructs the images reshaping it into a 244 x 324 grayscale image

03

A ROS2 message of type *Image()* is filled with the image (with a bridge)

IMG STREAMER NODE

04

The message is published on the topic /cam0/image_raw with the associated TimeStamp

05

Receives estimated pose on the topic /orb_slam3/odom and sends it back to the drone via WiFi

ORB_SLAM3 NODE

orb_slam3.cpp

- Subscribes to image and camera topics.
- Manages those data through an instance of the DataGrabber class.
- Sends back to img_streamer node the estimated pose.

orb_slam3

data_grabber library

- Receives image and IMU messages and queues them internally using buffers.
- Feed synchronized data to the ORB-SLAM3 tracking system.
- Receives estimated pose.



ORB_SLAM3 VIO ALGORITHM

ORB-SLAM3 is a real-time SLAM system that supports visual, visual-inertial, and multi-map SLAM using monocular, stereo, and RGB-D cameras, compatible with both pin-hole and fisheye lens models.



It extracts ORB features from the image while integrating IMU data, then defines the MAP-based optimization problem,



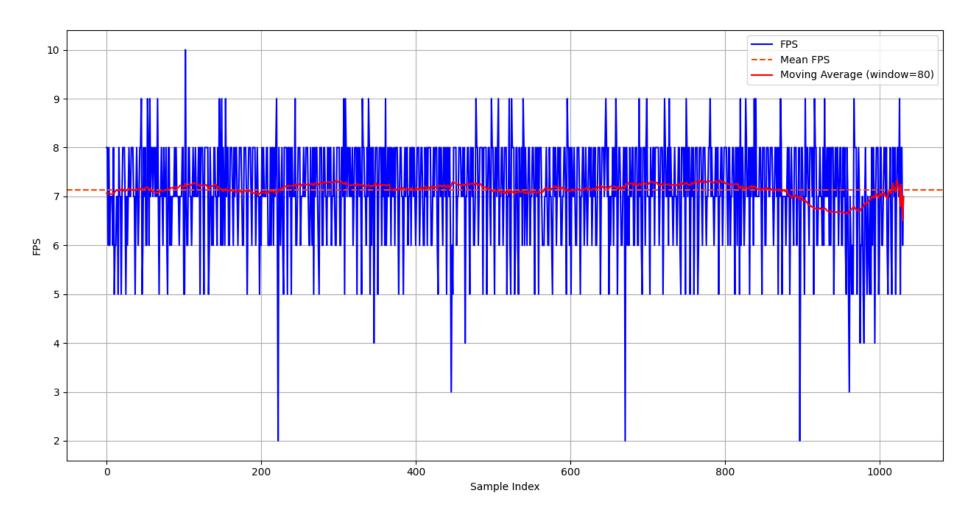
Optimization solved through tightlycoupled bundle adjustment, minimizes visual reprojection and inertial errors.



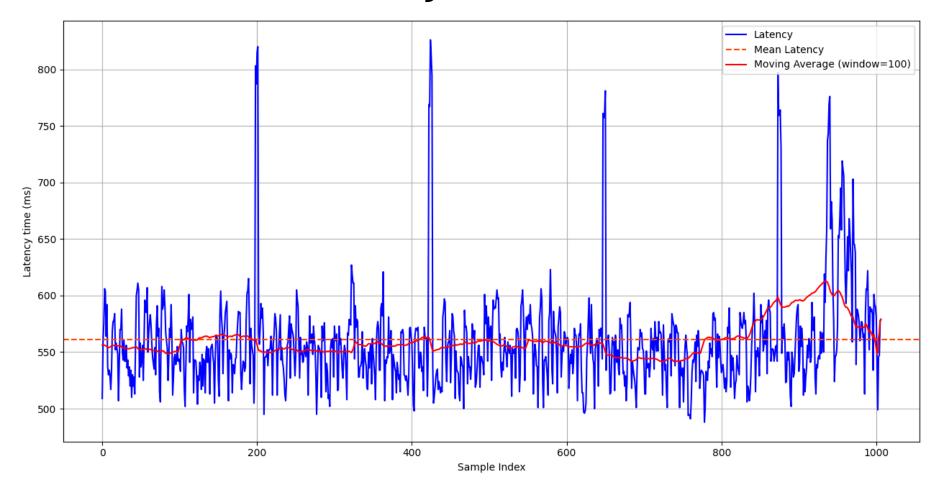
Performance enhanced by DBoW2-based place recognition for loop closure and drift correction.

Section 3: Results

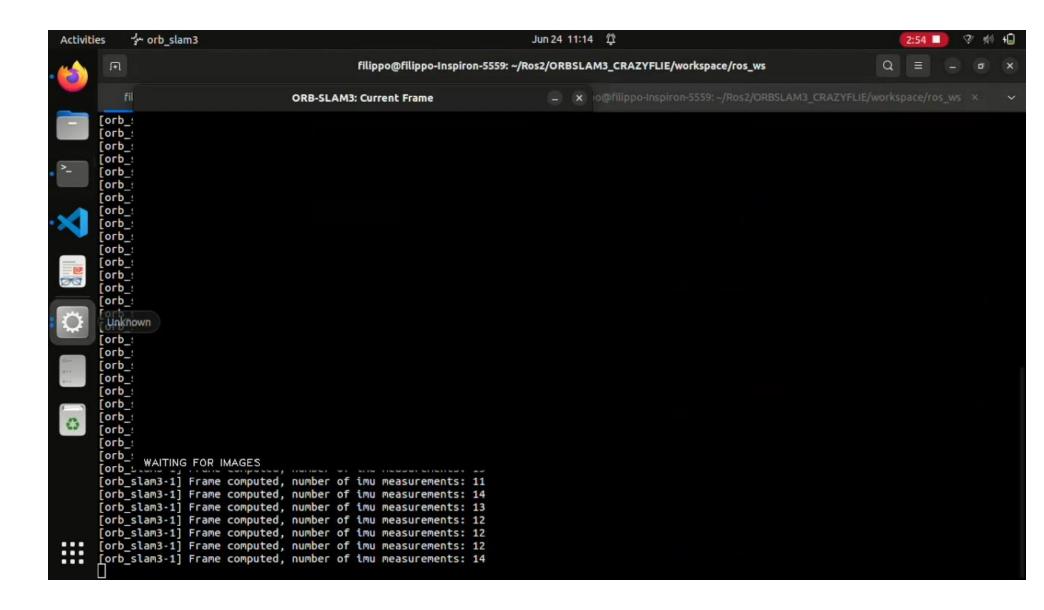
Performances: FPS



Performances: Latency



Video



Conclusions



Main goal achieved: Successful integration of Crazyflie, AI-deck, and ROS2 using dual communication (Wi-Fi + Bluetooth) for the Visual Inertial Odometry.



Challenges addressed:

- Logging of data from both IMU and Camera;
- Integration of different frameworks (Orb Slam 3, ROS2, CF library, GAP8 programming);
- Performance improvement (i.e. FPS).



System limitations:

- Communication constraints;
- Poor camera resolution.

Ideas for performance improvements

Develop the application outside the ROS2 framework for increased performances.

Define a better strategy to use a single channel for the communication. Improve the camera system (higher resolution, stereo or RGB-D cameras).

Future applications



TRACKING OF COMPLEX TRAJECTORIES

Validation of the control algorithm's precision through high-fidelity trajectory tracking.



AUTONOMOUS FLIGHT

Integration of the ROS2 Navigation Stack to enable fully autonomous missions.



COMPUTER VISION

Use camera for environment perception and obstacle avoidance through lightweight neural networks on embedded hardware.



MULTI-DRONE COOPERATION

Coordination of multiple Crazyflie units through shared ROS2 topics and synchronized control nodes.

THANKS FOR YOUR ATTENTION!

