

# Vision-based Autonomous Landing in Catastrophe-Struck Environments

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Control Problems in Robotics: Modeling and control of multi-rotor UAVs

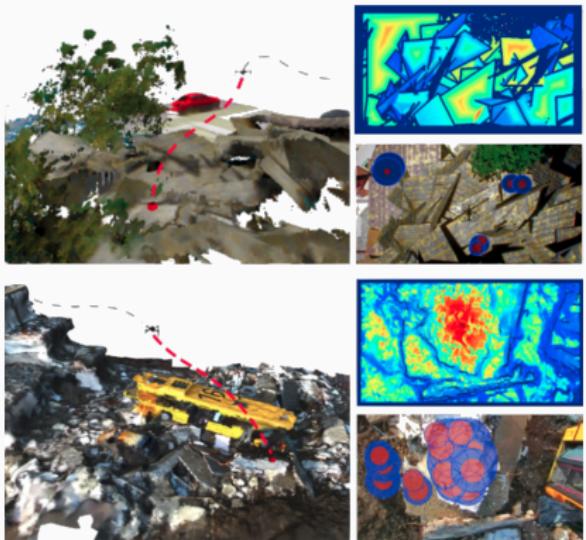
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# Introduction

Intro.

- Why it's important  
(bioradiolocation, also  
emergency landing)
- previous work



# Technical Approach

Brief introduction to ch. 3, describe approach and then discuss details.

- State Estimation
- Landing Site Detection
- 3D Volumetric Mapping
- Landing Trajectory Estimation

## Technical Approach: State Estimation

ORB-SLAM2 + IMU + barometer + GPS (EKF multi-sensor fusion).

## Technical Approach: Landing Site Detection

Confidence in DEPTH INFORMATION  $J_{DE}$ :

$$J_{DE}(p) = 1 - \frac{D(p)^2 - \min\{D^2\}}{\max\{D^2\}}$$

with  $D$  depthmap obtained from the stereo camera and  $p = (x, y)$  pixel in the depthmap.

## Technical Approach: Landing Site Detection

FLATNESS INFORMATION  $J_{FL}$ :

$$di(B, p) = \min \left\{ \|p - q\| \mid B(q) = 1 \right\}$$
$$J_{FL}(p) = di(Canny(D), p)$$

with  $B$  binary image and  $p, q$  pixels in the image plane. *Canny* applies the Canny edge detector over the depthmap  $D$ .

# Technical Approach: Landing Site Detection

## STEEPNESS INFORMATION $J_N$ :

- point cloud from the depthmap in global frame
- average 3D gradients algorithm to estimate normals map  $N$
- evaluate deviation of the normalized surface normal  $\hat{n}$  wrt z-axis  $\hat{z}$  in the world frame:  $\theta = \cos^{-1}(\hat{n}^T \hat{z})$
- compute  $n(p)$  steepness score for each pixel  $p$  given  $\theta_{th} = \pi/12$  maximum tolerable slope

$$n(p) = \exp \left\{ -\frac{\theta^2}{2\theta_{th}^2} \right\}$$

## Technical Approach: Landing Site Detection

ENERGY CONSUMPTION INFORMATION  $J_{EC}$ :

$$J_{EC}(p) = \int_{t_0}^{t_f} P(t) dt$$

with  $t_0$  and  $t_f$  time of flight to reach  $p$  and  $P(t)$  instantaneous battery power. Approximate the integral with Euclidean distance between the UAV and  $p$ .

## Technical Approach: Landing Site Detection

Scale the costmaps to the same range through min-max normalization and compute the final decision map  $J$  taking a weighted sum:

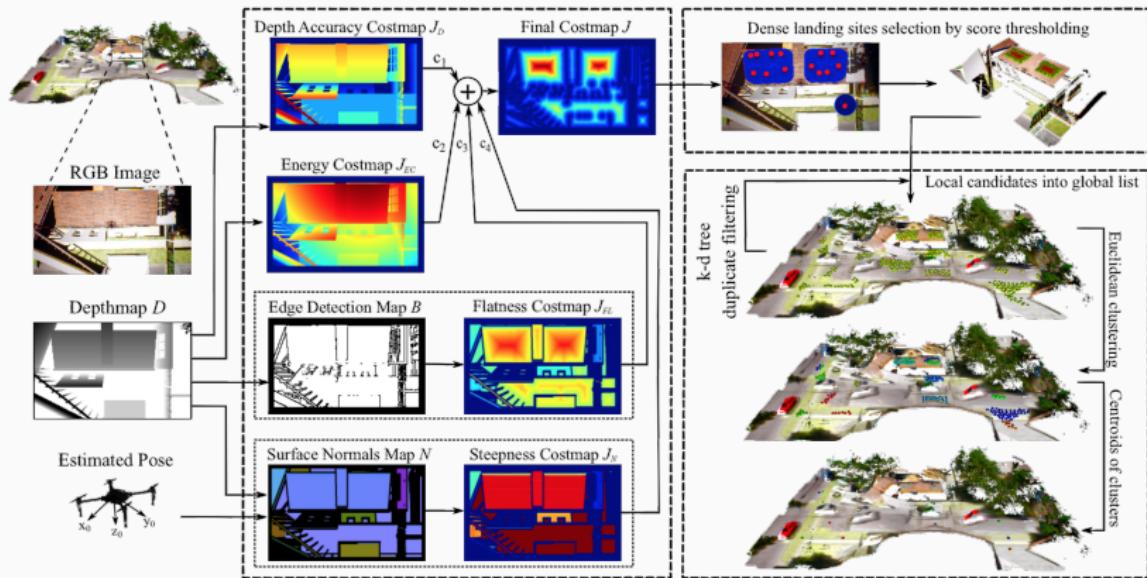
$$J = c_1 J_{DE} + c_2 J_{FL} + c_3 J_N + c_4 J_{EC}$$

$$c_i \in [0, 1] \quad \sum_i c_i = 1$$

- keep the sites checking whether the UAV could actually land
- k-d tree to efficiently store new landing sites
- hierarchical clustering algorithm to agglomerate sites

# Technical Approach: Landing Site Detection

**Figure 1:** Overview of the landing site detection algorithm. In the costmaps, red indicates high score while blue indicates a lower score. Detected landing sites are projected onto a 3D reconstruction of the environment.



# Technical Approach: 3D Volumetric Mapping

Probabilistic volumetric map:

- navigation and planning
- low-resolution map (0.5m) using **OctoMaps**
- faster trajectory planning

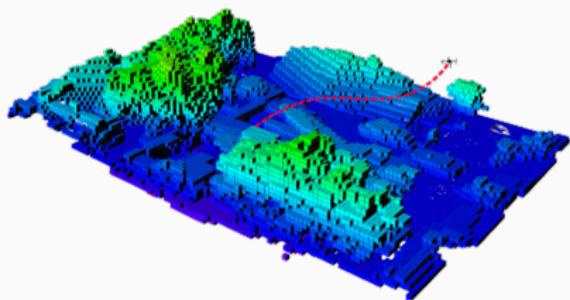
3D textured mesh:

- dynamic map growing using **Voxblox**
- high-resolution mesh transmitted to the rescue team

# Technical Approach: Landing Trajectory Estimation

RRT\* and minimum-jerk trajectory.

- First.
- Second.



# Experimental Evaluation

Description:

- Hyperrealistic Simulation
  - Real-World Outdoor (Training Center for Rescue, Germany)
- + computation costs

# Experimental Evaluation: Hyperrealistic Simulation

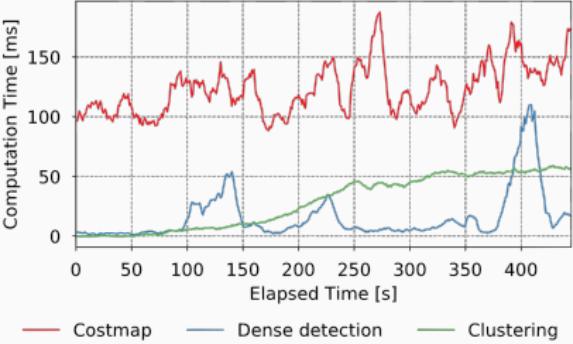
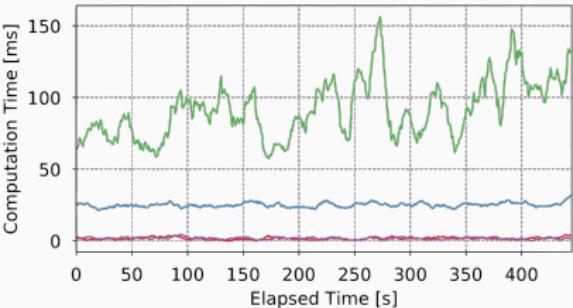
Hyperrealistic Simulation Experiments

# Experimental Evaluation: Real-World Outdoor

Real-World Outdoor Experiments

# Experimental Evaluation: Computation Costs

Computation costs



# Conclusion

Conclusion.

Q&A

## References

-  M. Mittal, A. Valada, and W. Burgard, “Vision-based autonomous landing in catastrophe-struck environments,” *CoRR*, vol. abs/1809.05700, 2018.