# **Collaborative Unmanned Aerial Vehicles**

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Institute of Networked and Embedded Systems

### Outline

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- Technology and application of unmanned aerial vehicles (UAV)
  - Small-scale UAVs
  - Disaster management
- Mission planning
- Arial Imaging
- System Integration

### **Battery-powered UAVs**



- **Quatcopter** platform with onboard sensors and electronic for flight stabilization
- Attached cameras for sensing the environment
- GPS receiver for autonomous waypoint flights
- Limitations on payloads, flight time, weather conditions



www.microdrones.de



www.asctec.de

## **Disaster Relief with UAVs**



### General idea

- Support first responders in disasters with multiple UAVs
- Provide latest and relevant information about the scene
- Autonomously fling, networked, collaborating UAVs

### Use case: Generate overview image

- Cover the disaster area and take images at individual points
- "Stitch" individual images to generate scene overview (mosaic)
- Provide intuitive user interface

## Autonomous UAV Operation





### Issues



- 1. How to generate & adapt movement routes for the UAVs?
  - Achieve multiple optimization goals
  - Deal with changes in the environment
  - Compare centralized versus distributed approaches
- 2. How to stitch the individual images?
  - Apply incremental image stitching
  - Tradeoff between good geo-referencing and visually appealing overview
- 3. System integration and demonstration

## 1: Generation of Routes



	Non-cooperative	Cooperative
Deterministic	UAV has a <b>predefined route</b> that is <b>independent</b> of other UAV paths.	UAV has a <b>predefined route</b> that <b>depends</b> on other UAV paths.
Dynamic	UAV has an <i>a priori</i> <b>unknown</b> <b>route</b> that is <b>independent</b> of other UAV paths.	UAV has an <i>a priori</i> <b>unknown</b> <b>route</b> that <b>adapts</b> to other UAV paths.

### **Adaptive Routes**



# Meeting event

### How to adapt the route?



# Analysis of the Area Coverage

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### Approaches

- Simulation-based studies
- Discrete stochastic processes



# 2: Image Mosaicking



- Problem definition
  - Given *n* individual images *Ii*, find **image transformations** *Ti* for each *Ii*

$$I_{overview} = \sum_{i=1}^{n} T_i(I_i)$$

which maximizes some quality function  $\lambda(I_{overview})$ 

- Two fundamental approaches for finding the transformations
  - Exploit auxilliary data, i.e., camera's position and orientation (meta data based approach)
  - 2. Exploit corresponding points within image overlaps (image based approach)

## **Challenges for Mosaicking**



- Low altitude and non-planar surface introduce high perspective distortions
- Light-weight UAVs are vulnerable to wind resulting in nonnadir view
- Inaccurate position and orientation data due to small, lowcost GPS, IMU and altimeter sensors
- Strong **resource limitations** wrt. onboard processing, power, communication etc.

### **Incremental Image Mosaicking**

• Start with meta data approach, refine with image-based approach



Position data

Position&orientation data

SIFT feature image points

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### Overview image

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# 3: System Integration



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### FAMUOS - Fully Autonomous Multi-UAV Operation System



## "Google-like" User Interface

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# Specifying the scenario description

Visualizing the **latest overview image** and the flight route

### Demonstration

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• Check also: <u>http://pervasive.uni-klu.ac.at/cDrones</u>