



# UAV Autonomous Navigation in a GPS-limited Urban Environment

Yoko Watanabe  
DCSD/CDIN



r é t o u r s u r i n n o v a t i o n

JSO-Aerial Robotics  
2014/10/02-03

## ❖ Global objective

Development of a UAV onboard system to maintain flight security and navigation & guidance capability for urban operation

## ❖ GPS signal occlusion

- Alternative GPS-independent navigation system
  - Stabilization
  - GN&C functions → { Mission continuation  
Automatic return-to-base  
etc.
- Path planning with GPS signal occlusion map
  - Safe path plan w.r.t. localization uncertainty
  - Sensor availabilities



A. Gorski «Understanding GPS performance in urban environments»  
<http://blogs.agi.com/agi/2011/01/04/understanding-gps-performance-in-urban-environments/>

## ❖ Objective

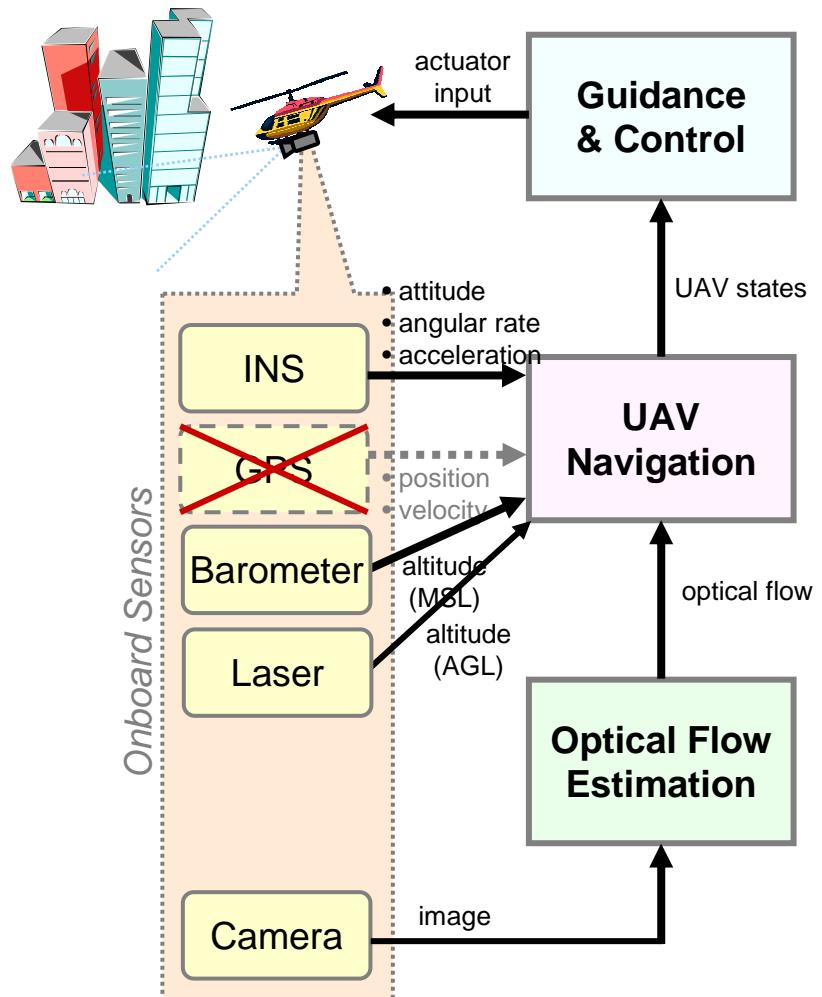
- Development of alternative back-up navigation system which estimates UAV absolute state by using onboard sensors other than GPS, given the last GPS-updated state
  - No dedicated sensors
  - No knowledge on environment
  - Low computation
  - Robustness
- In-flight validation on outdoor UAV helicopter
  - Onboard system integration with
    - flight avionics
    - onboard sensors
  - Closed-loop flight using existing GN&C functions with GPS signal cut-off



# Vision-aided inertial navigation

4

- ❖ Stereo vs Monocular visions
- ❖ Pure vision vs INS-fusion
- ❖ Visual odometry vs Visual SLAM
- ❖ Filter vs Optimization (BA)



# Vision-aided inertial navigation

5

## ❖ Visual odometry

- Stereo vision
- Monocular vision (motion stereo)

[Kelly 2007], [Kendoul 2009] and many others.

☺ Low computation

☹ Estimation drift due to absence of absolute measurement

## ❖ Visual SLAM

- Loop-closure (memorization of feature points)

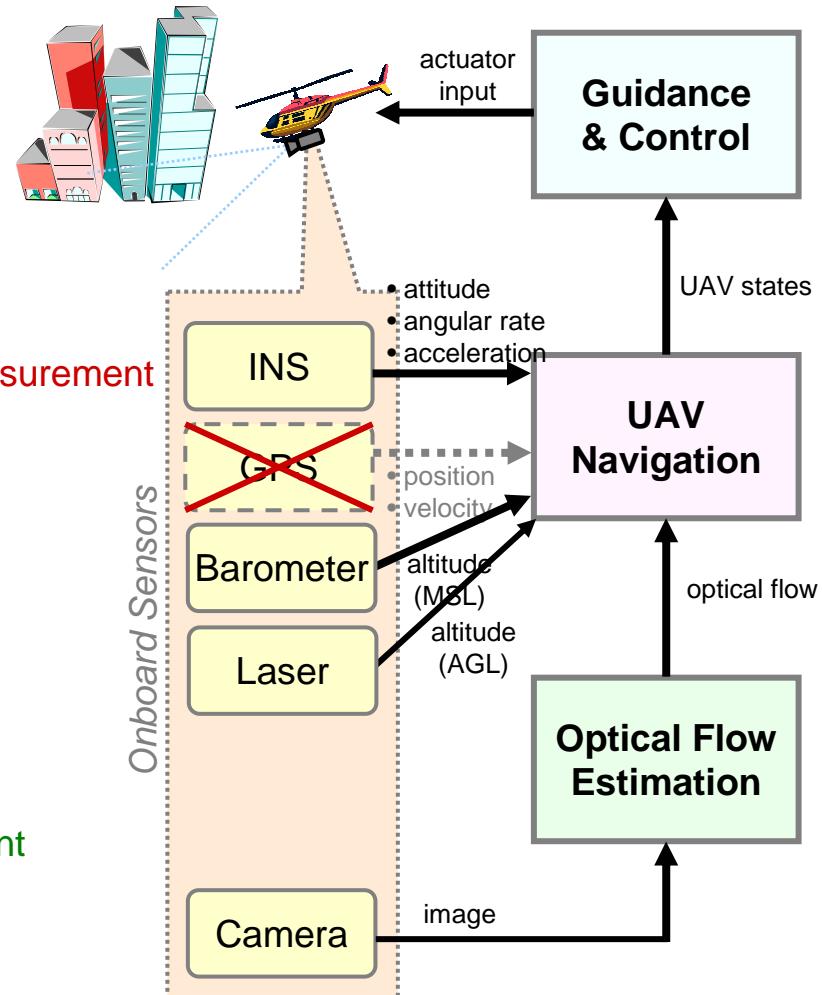
[Weiss 2012], [Chaudhar 2013] and many others.

☺ High computation + memory-use

☺ Estimation correction with absolute measurement

- Keyframe-based SLAM

[Klein 2007] and many others.

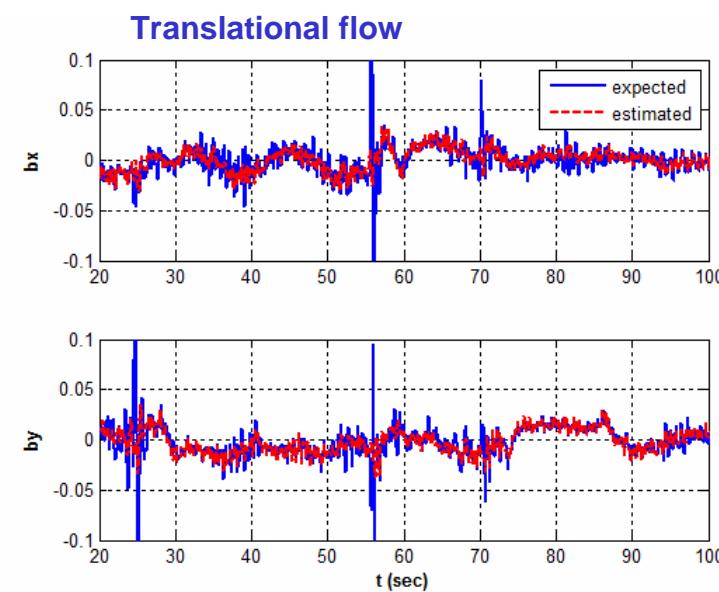
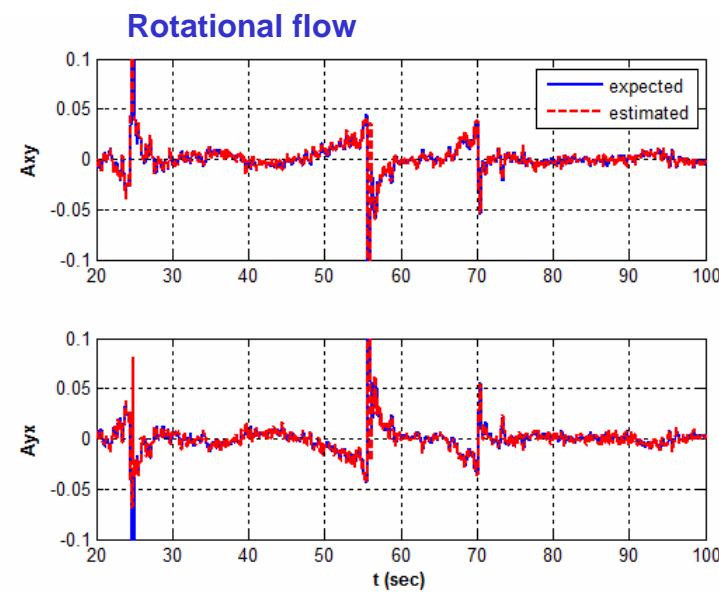
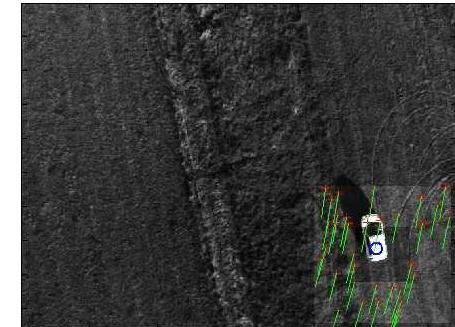


# Optical flow estimation

- ❖ Robust estimation of Affine model optical flow field (DTIM)

- Feature point matching on a small window
- RANSAC approximation
- ~10Hz

$$\mathcal{C} \mathbf{x}_{p_k} = \mathbf{A}_k \mathbf{x}_{p_k} + \mathbf{b}_k$$

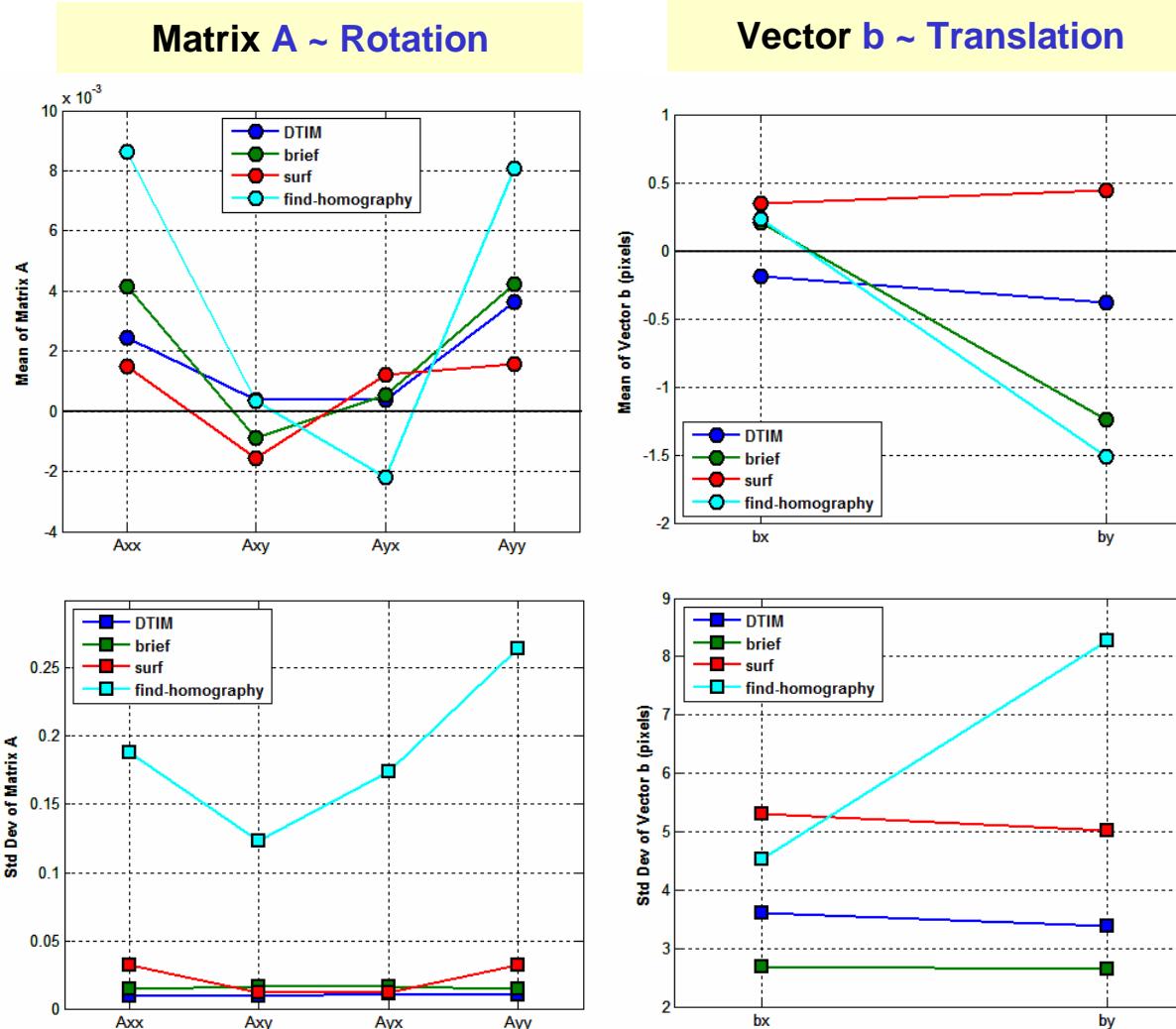


# Optical flow estimation

7

❖ vs. OpenCV

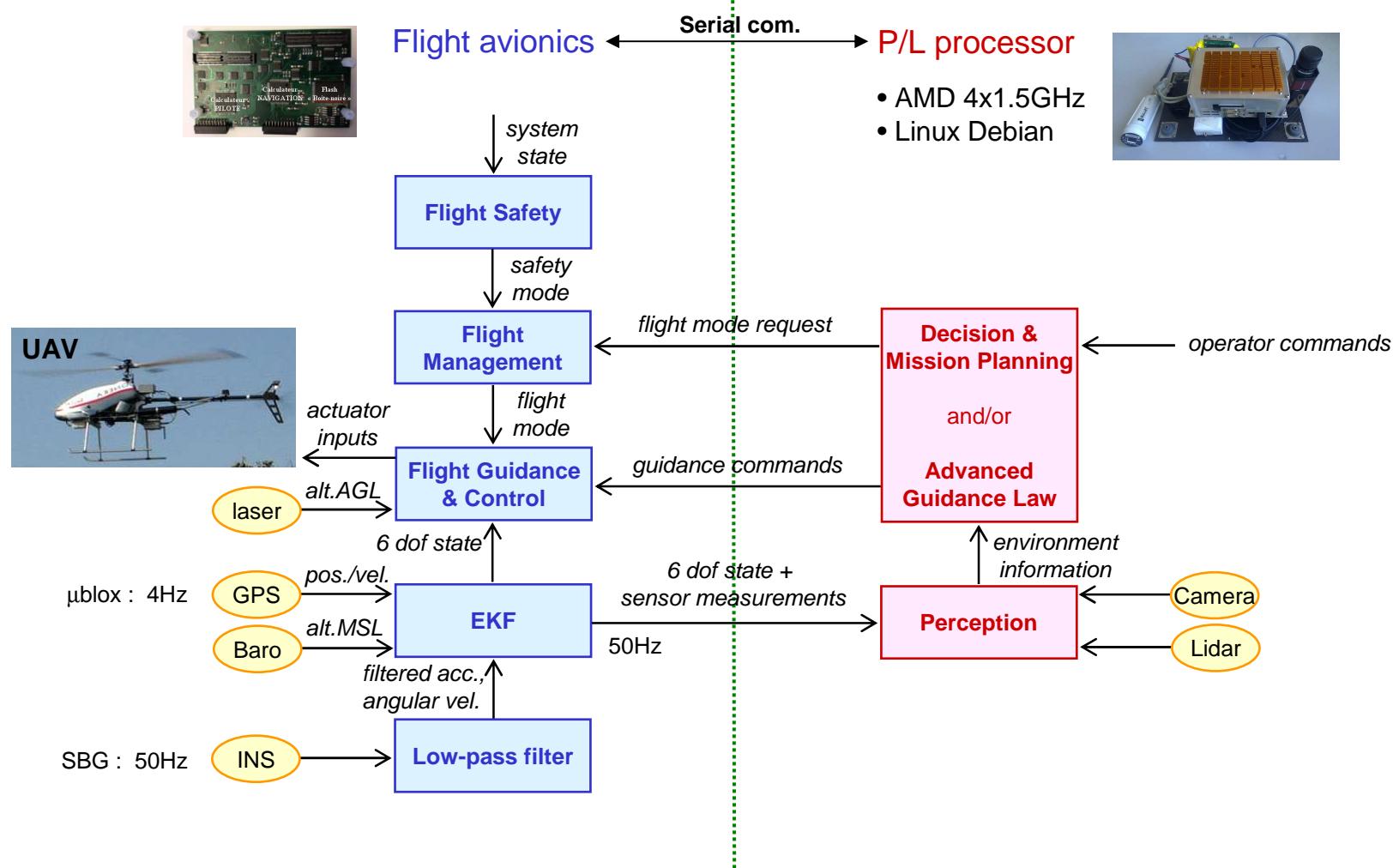
**Mean of  
Estimation Error**  
~ Bias



**Std. Deviation of  
Estimation Error**  
~ Noise

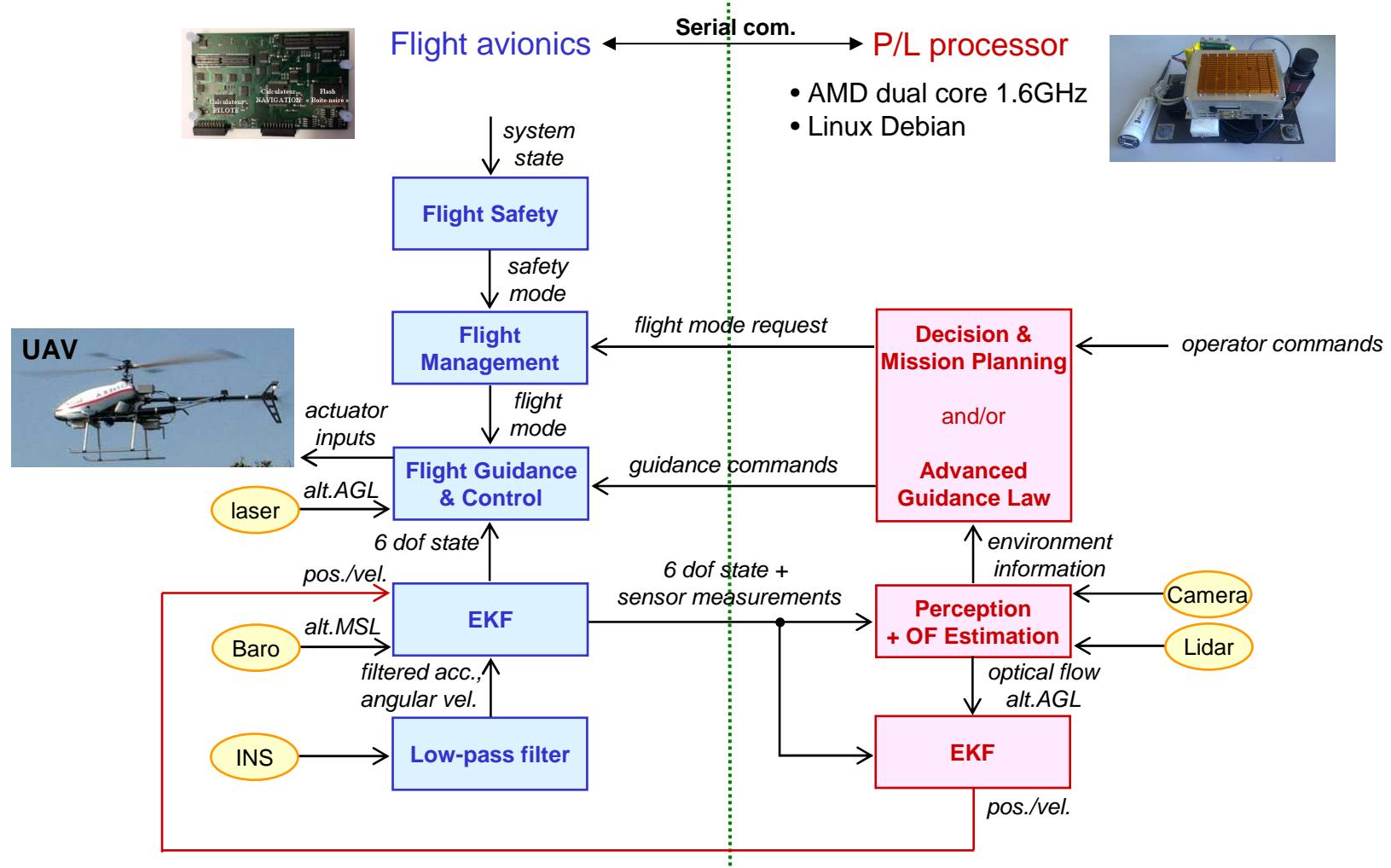
# Onboard system architecture

8



# Onboard system architecture with GPS-independent navigation

9

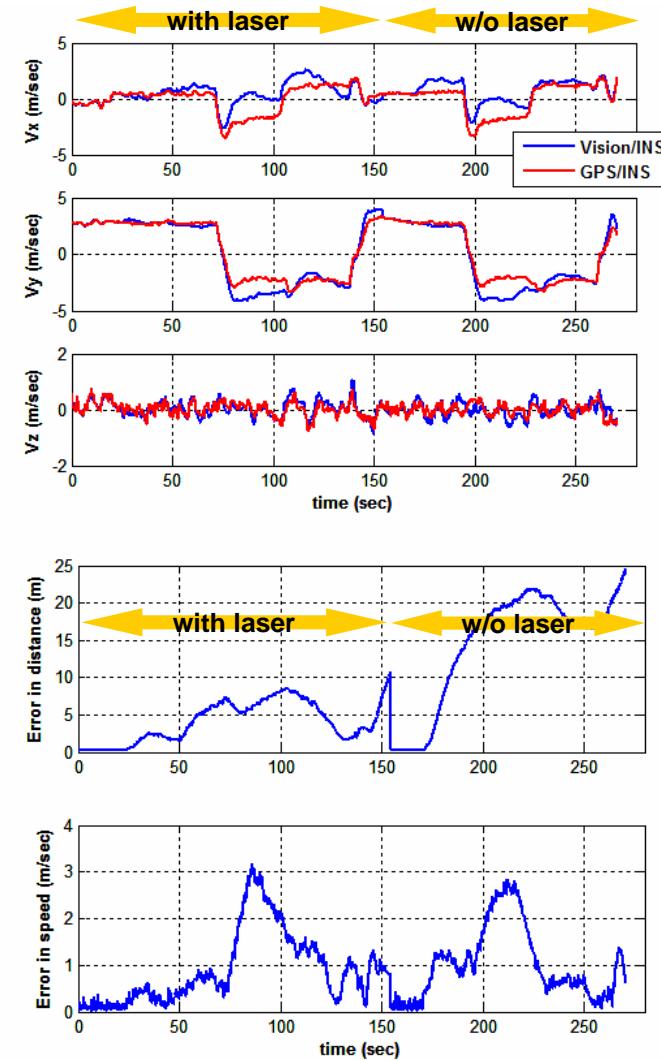
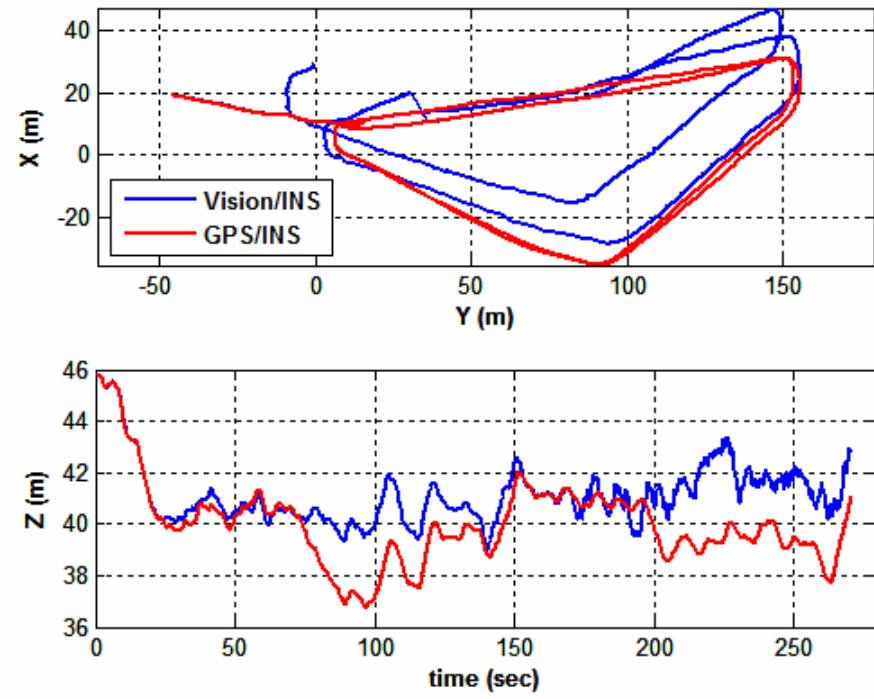


# Open-loop flight test results

10

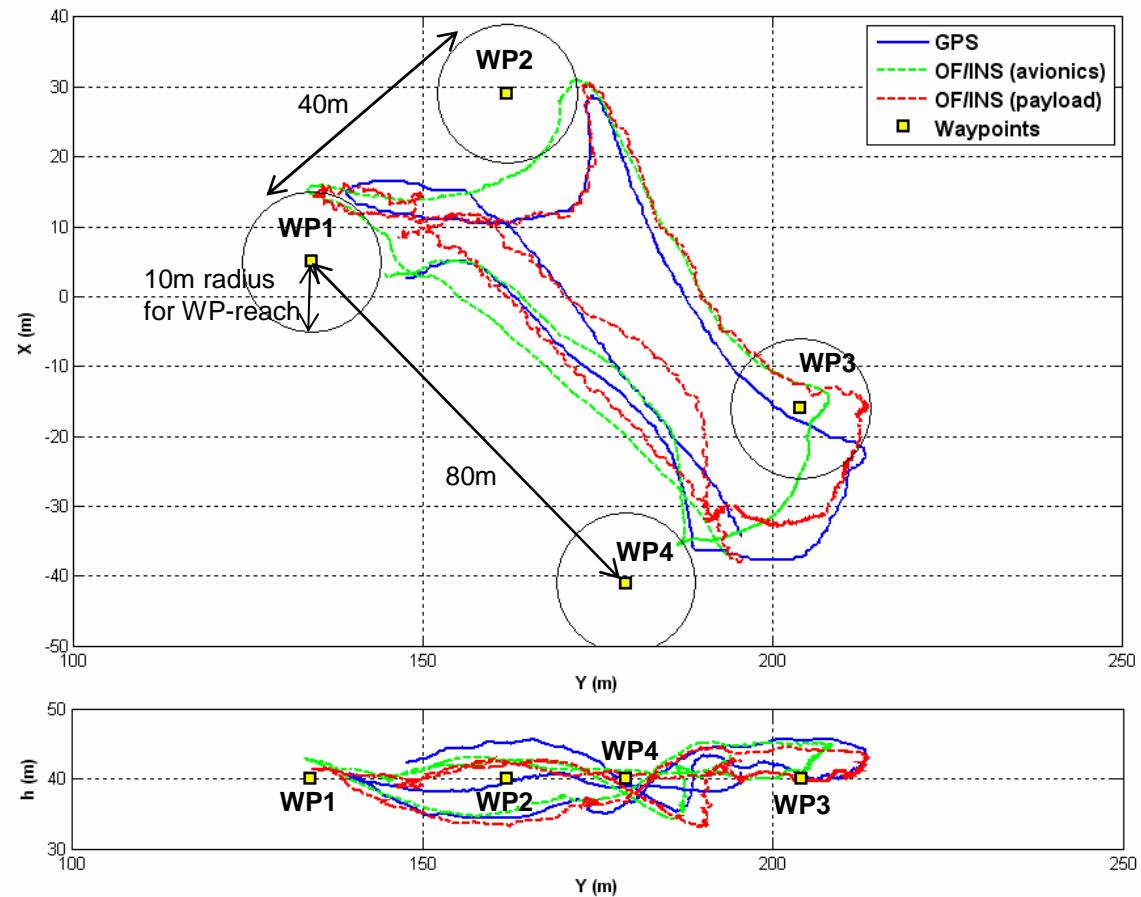
## ❖ OF + INS + Barometer

- with or w/o laser (alt. AGL)
- over a slope



- ❖ GPS cut-off during WP tracking mission

- Rectangle trajectory  
40 x 80 (m)
- Constant heading  
into wind NW
- 10m of WP-reach criteria
- Flight distance (w/o GPS)  
~ 320 (m)
- Flight time (w/o GPS)  
~ 130 (sec)

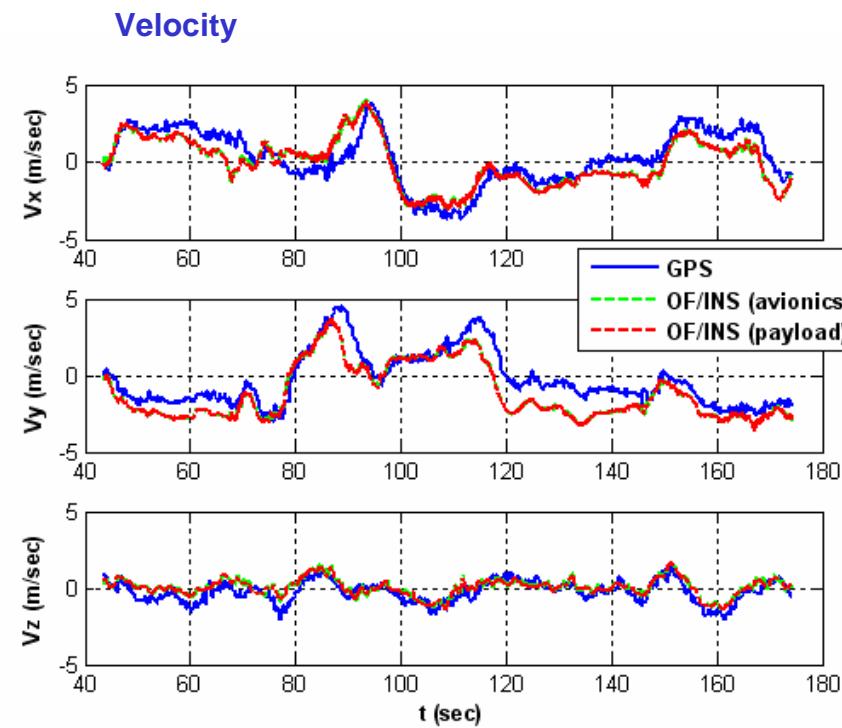
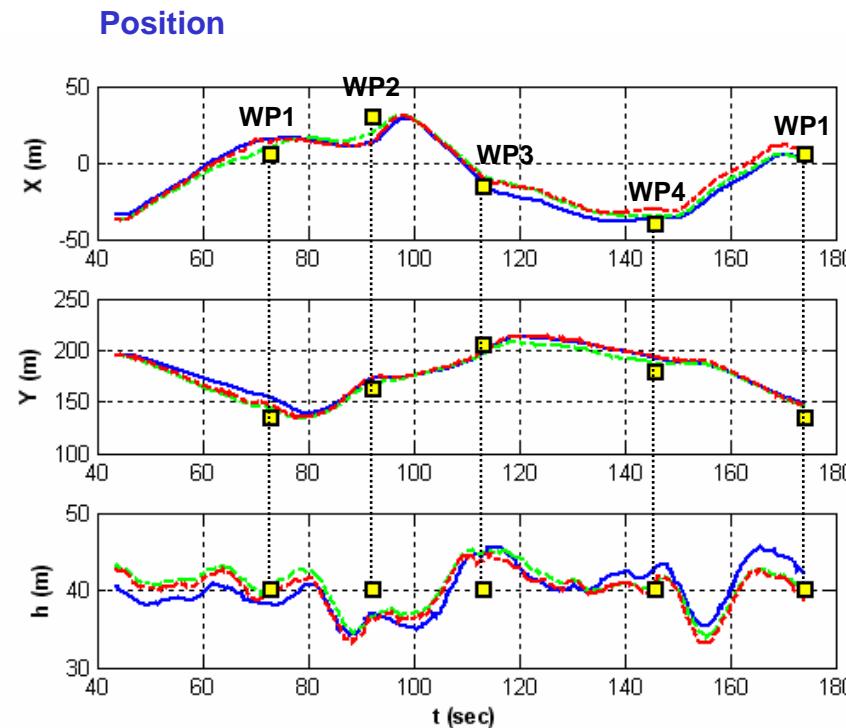


# Closed-loop flight test results

12

## ❖ OF-estimated vs. GPS-estimated position and velocity

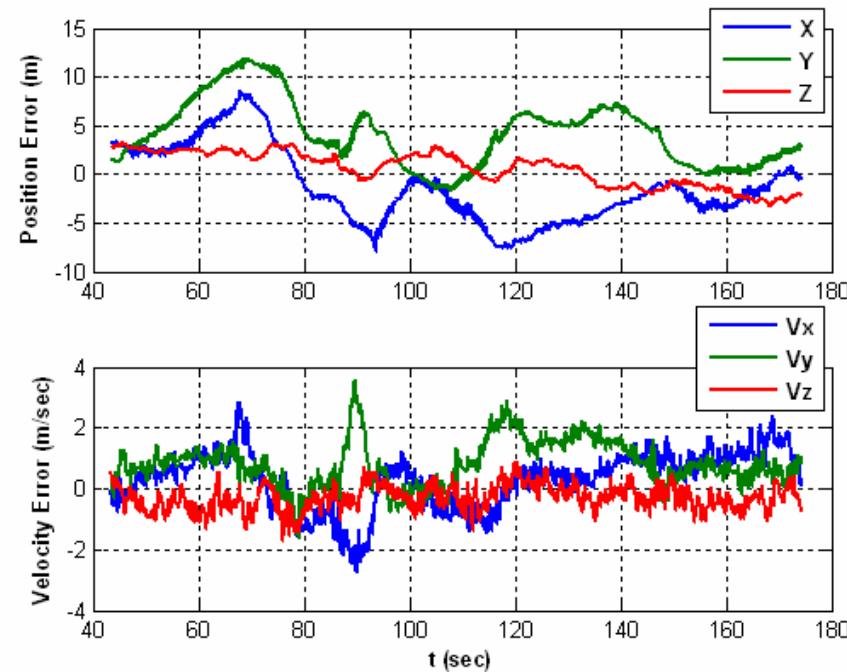
- Position estimation error < 12m
- Stable altitude estimation by barometer + laser
- WP miss distance < 12m



# Closed-loop flight test results (3/3)

13

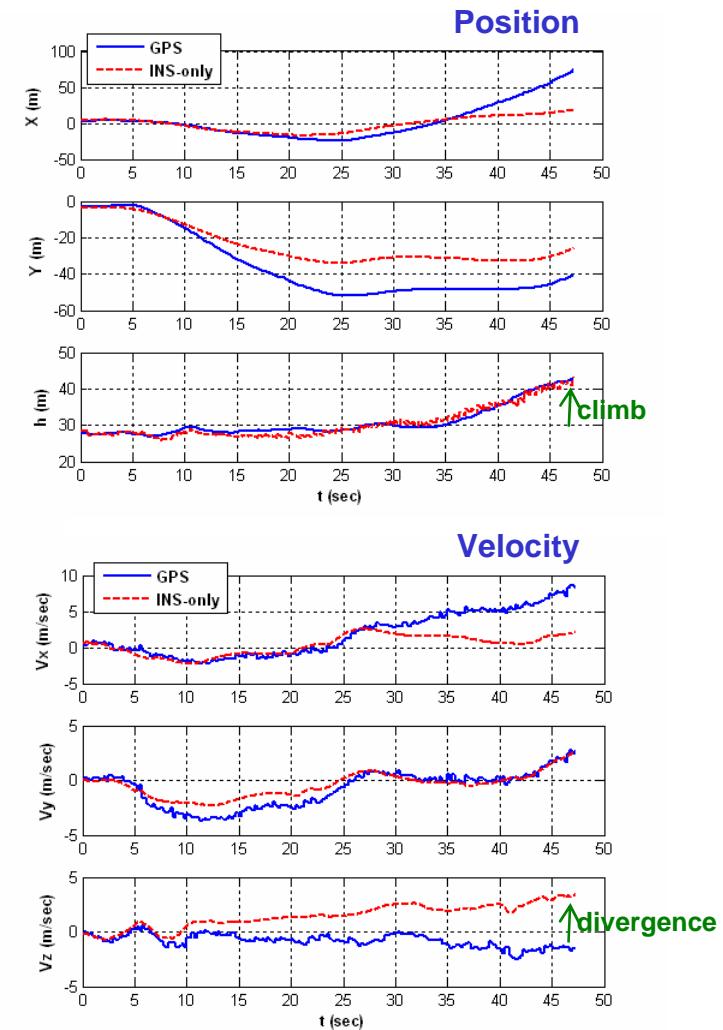
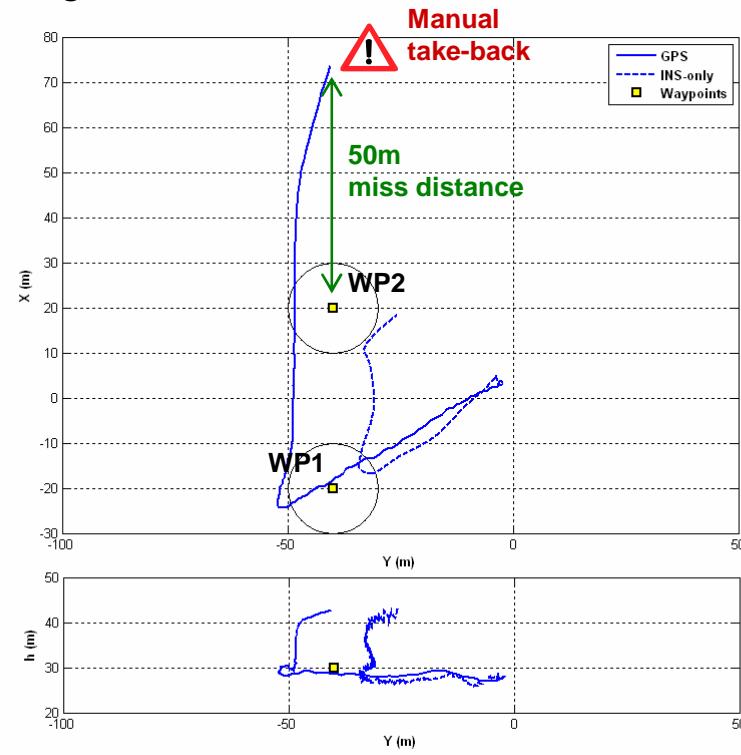
## ❖ Position and velocity estimation errors



# Closed-loop flight test results with INS-only navigation

14

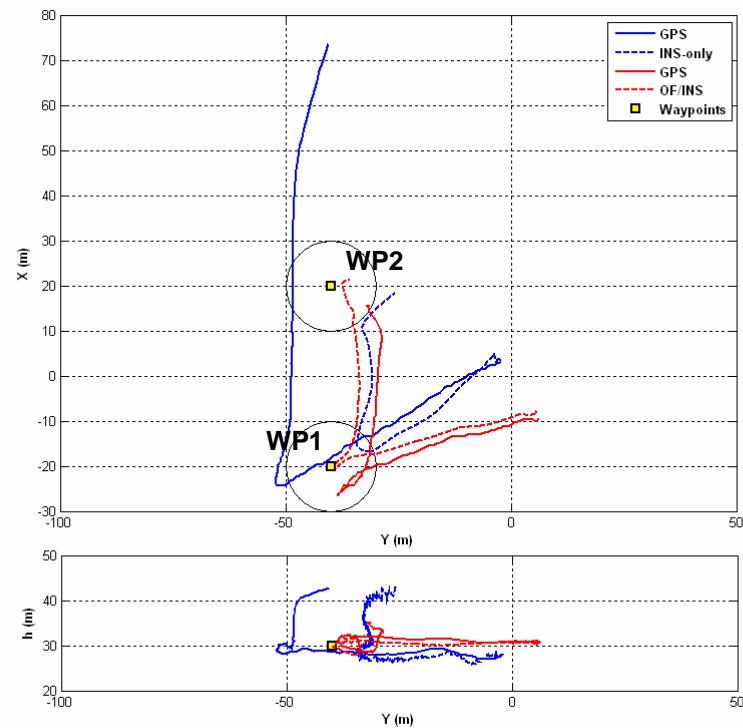
- ❖ Position and velocity estimation errors
  - Stabilization
  - 50 m of drift after 45 sec
  - Divergence in altitude control due to Vz estimation



# Closed-loop flight test results with INS-only navigation

15

## ❖ Position and velocity estimation errors



## ❖ Summary

- Development and in-flight validation of optical flow-based inertial navigation system
- WP tracking mission continuation with GPS cut-off (switch navigation modes)

## ❖ Perspectives

- Performance improvement
  - Different OF estimation algorithms
  - Different VINS algorithms
- Demonstration of automatic return-to-base w/o GPS
  - Return-to-base by VO
  - Automatic landing with vision-based control
- Reconfigurable navigation system
  - Sensor failure
  - GPS accuracy

## ❖ Motivation

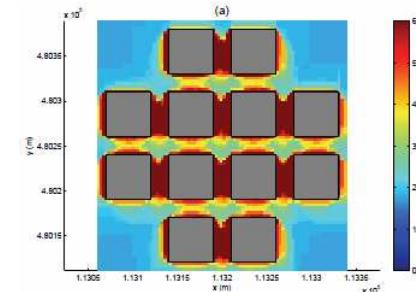
- Prediction of PDOP (Positional Dilution of Precision) of GPS at a certain time & location, from 3D obstacle map



## UAV safe operation planning

- Avoid zones at high risk of GPS signal loss, if no degraded navigation mode is available
  - Use sensor availability map in path planning task
  - Choice of the best navigation mode
- Take more safety margin when using degraded navigation mode
  - Obstacle collision risk w.r.t. localization uncertainty

F. Kleijer et al. «Prediction of GNSS availability and accuracy in urban environments»



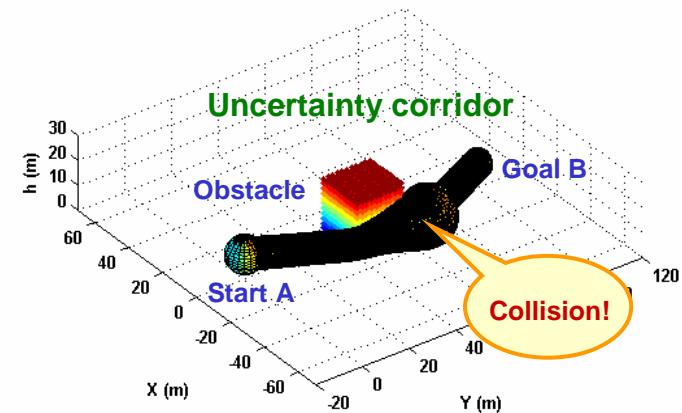
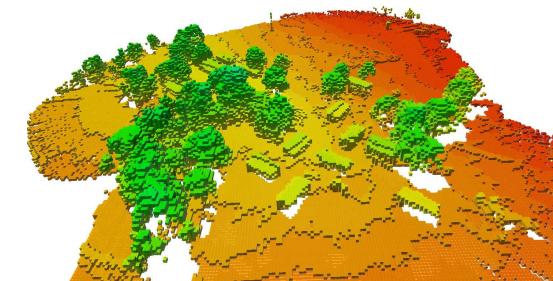
# 3D safe path planning problem

18

- ❖ Objective = find a safe & short path from A to B

- ❖ Given :

- Environment model = 3D voxel occupancy map
- $N$  different UAV localization modes
  - Positional availability
  - Error propagation model
- Collision criteria
  - Minimum safety distance =  $ds$
  - **Uncertainty corridor** =  $(2\sigma+ds)$ -ellipsoid evolution
  - Safe path = no interception between the corridor and occupied voxels
- Minimizing function =  
**Volume of the uncertainty corridor**
  - Path length
  - Integrated localization uncertainty

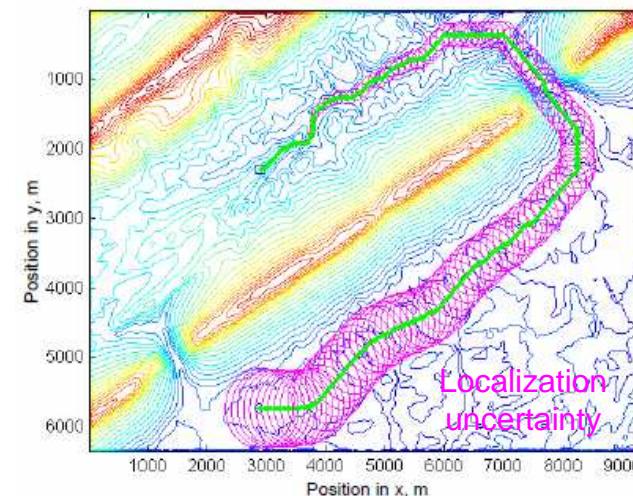


## ❖ Path planning with localization uncertainty

- Ground mobile robot navigation with
  - Dead-reckoning
  - Landmark detection
- Collision risk-free minimum distance path
  - A\* : [Alami 1994], [Lambert 2003], [Gonzales 2005] etc.
  - Sampling-based (PRM, RRT) : [Peppy 2006], [Luders 2013], [Bopardikar 2014] etc.
  - POMDP : [Candido 2010] etc.



The localization mode is imposed



# Related Work (2/2)

20

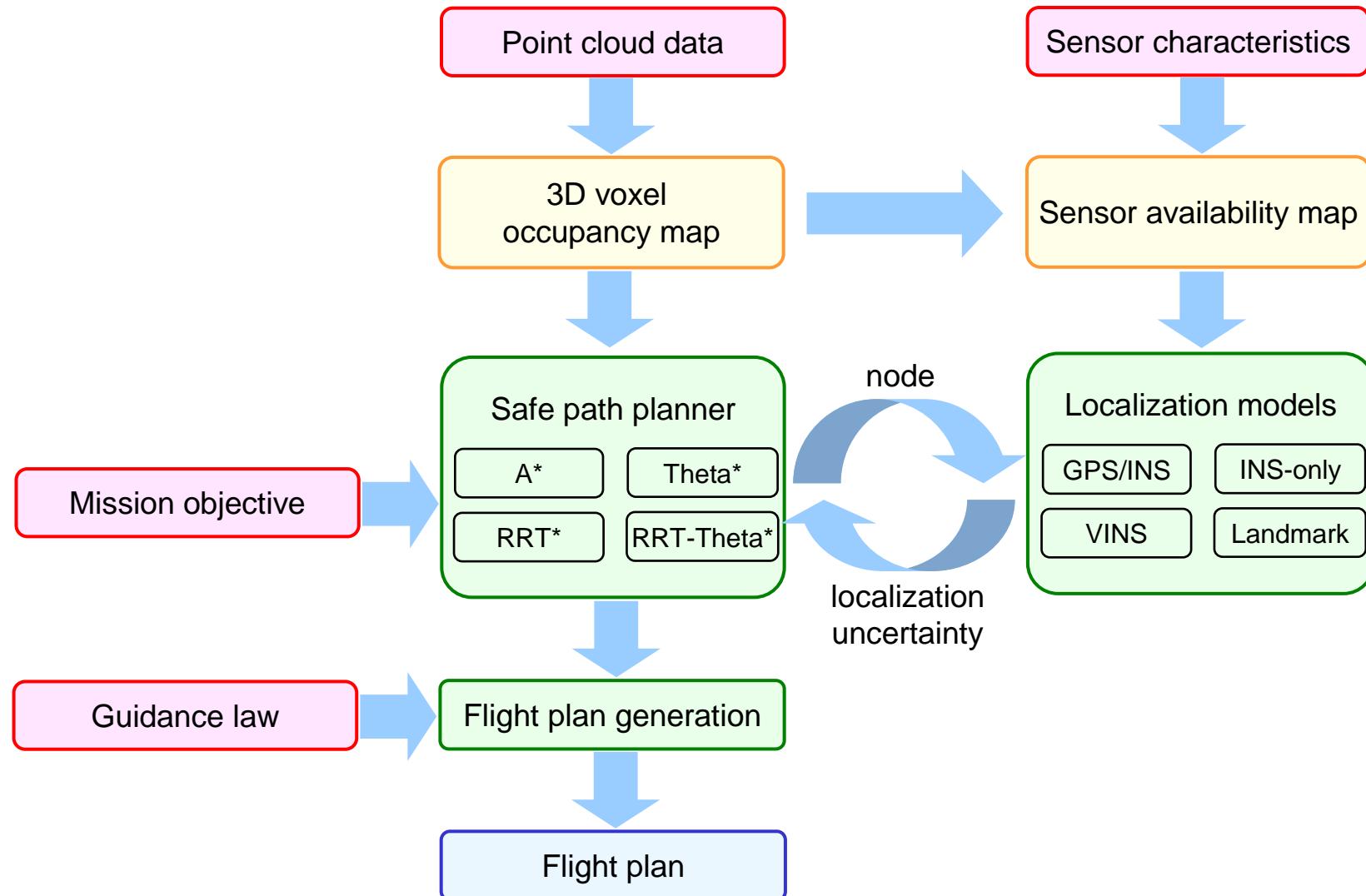
## ❖ Path and observation strategy planning

A. Yamashita, K. Fujita and K. Kaneko, "Path and viewpoint planning of mobile robots with multiple observation strategies," IROS 2004.

- Ground mobile robot navigation with
  - Dead-reckoning
  - Landmark detection
    - 1 landmark by stereo
    - 2 landmarks
    - 3 landmarks
- Two-stage planning
  - Search for all collision risk-free paths with maximum allowable localization uncertainty
  - Viewpoint (and localization mode) planning on each path

# 3D safe path planner architecture

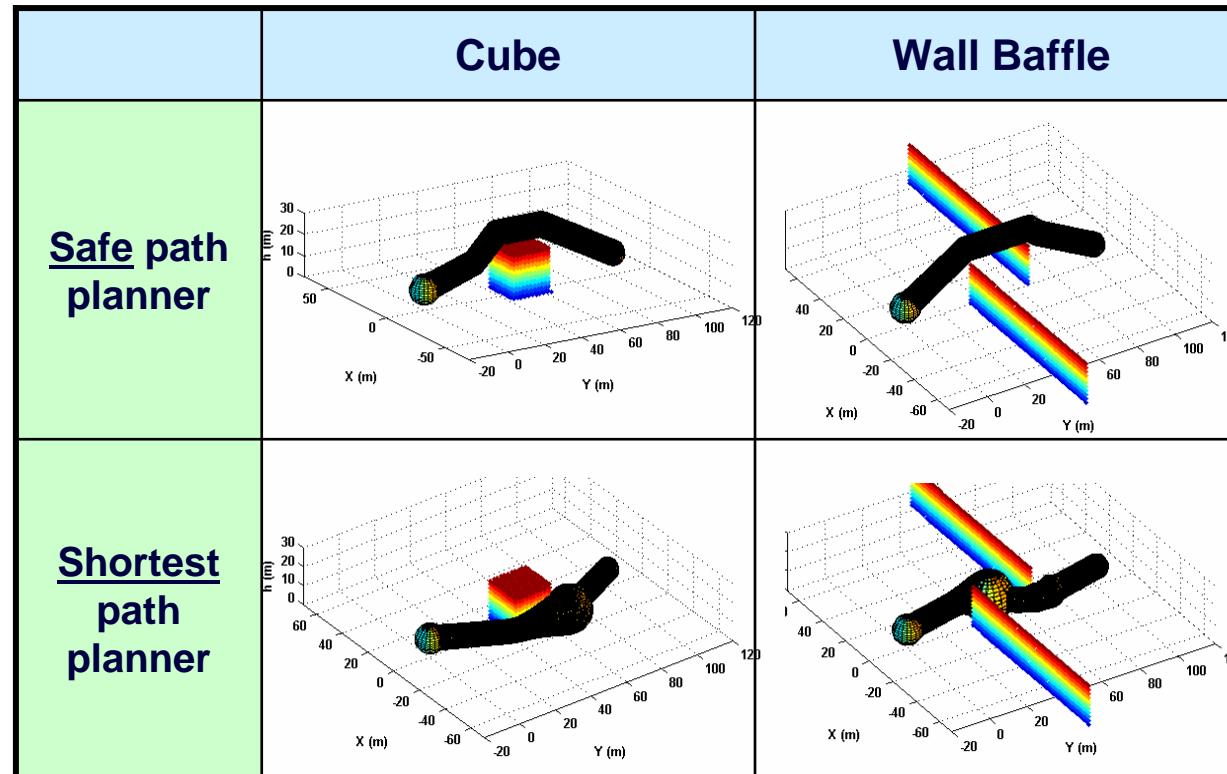
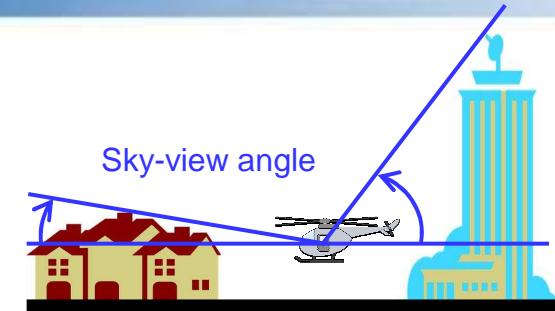
21



# Example 1 : No VINS

22

- ❖ Path planning with GPS availability map
  - No vision-aided navigation mode available onboard

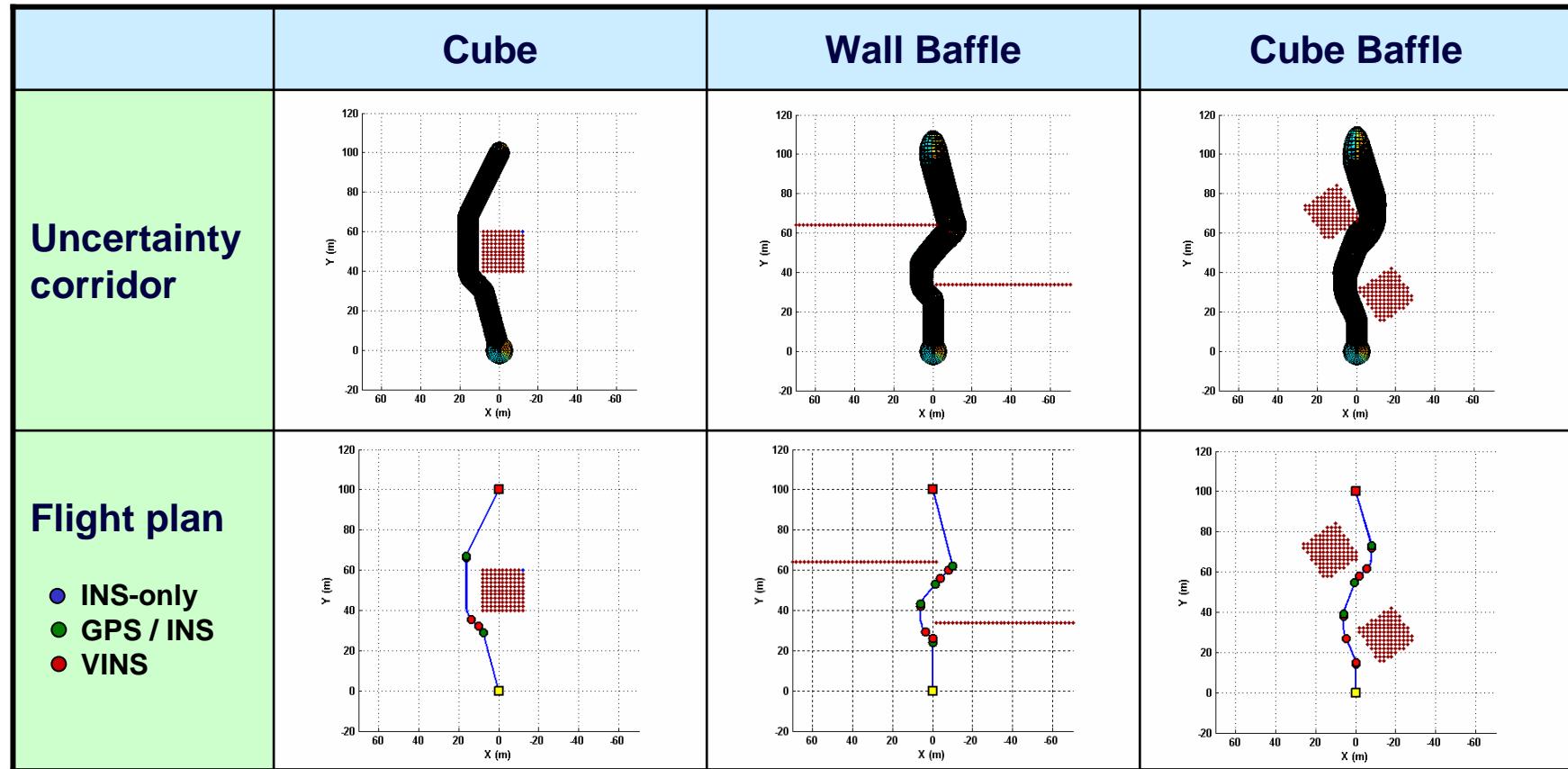


→ fly over the obstacles  
to avoid no GPS zones

→ collisions due to  
divergence in localization  
error covariance

## Example 2 : with VINS

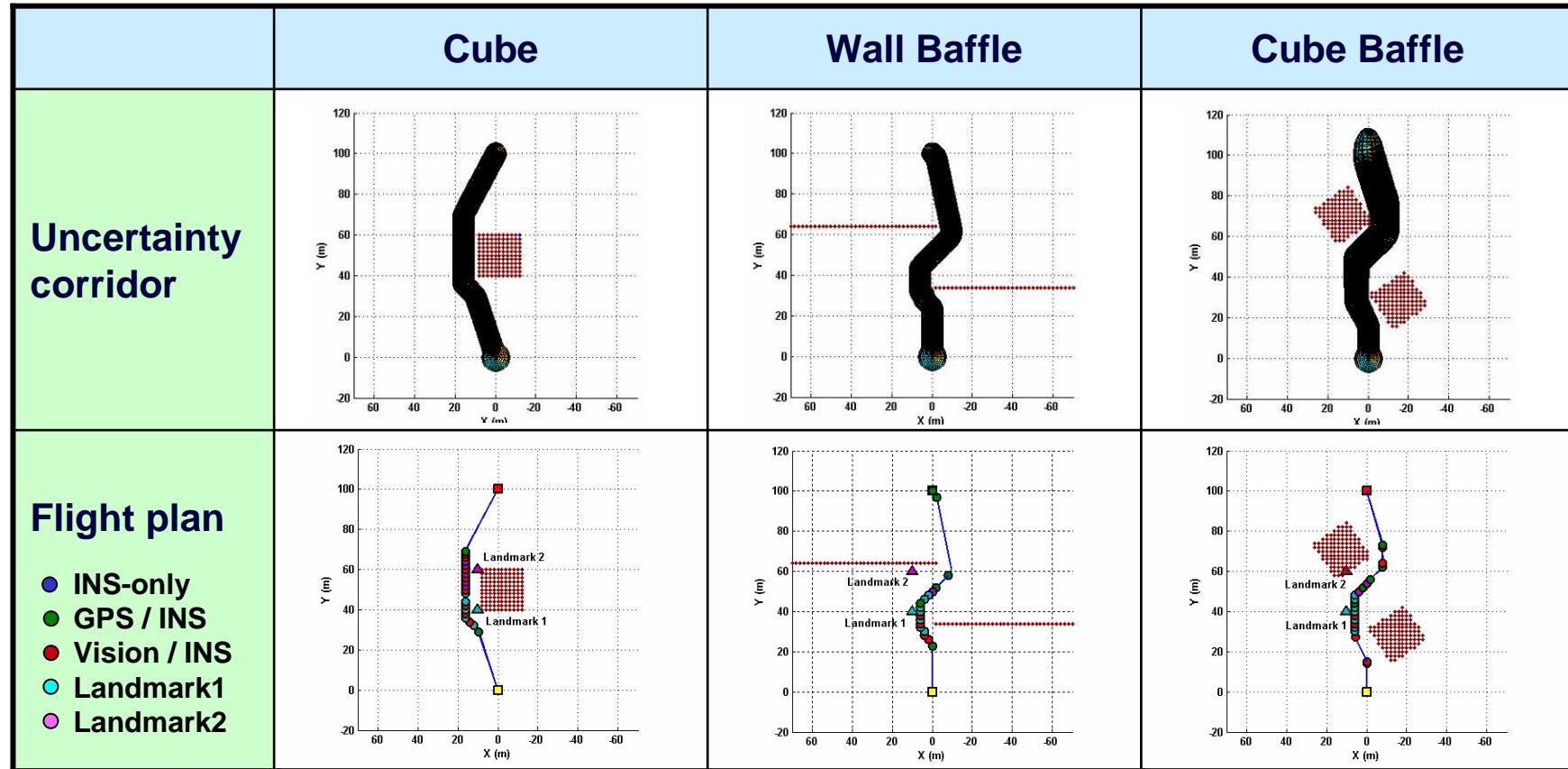
23



Remark : Dependence on optical flow measurement noise

# Example 3 : with VINS + Landmarks

24



Remark : Alternate use of VINS and Landmarks → Fusion

## ❖ 3D safe path planner

- Under uncertainty with multiple localization modes
- Simulation studies with UAV obstacle field navigation benchmark
- Preliminary flight test to validate onboard mapping and planning

## ❖ Future work

- Dynamic path re-planning using sampling-based graph search (RRT\*)
  - online mapping
  - supervision on real sensor availability and localization performance
- Path planning with different guidance strategies
  - visual servoing (e.g. wall following etc.)