

# Autonomous Micro-Aerial Vehicle Navigation Using a Custom Optic Flow Sensor Ring

Master Project: Final Presentation

Raphael Cherney



Harvard University  
Self-Organizing Systems Research Group (SSR)



Swiss Federal Institute of Technology (EPFL)  
Laboratory of Intelligent Systems (LIS)





# ROBOBEES

- Researchers at Harvard University are working to create a **swarm of robotic honey-bees**
- Such a distributed robotic system can be used for tasks such as:
  - Search and rescue
  - Hazardous environment exploration
  - Military surveillance
  - Weather and climate mapping
  - Crop pollination
  - Traffic monitoring

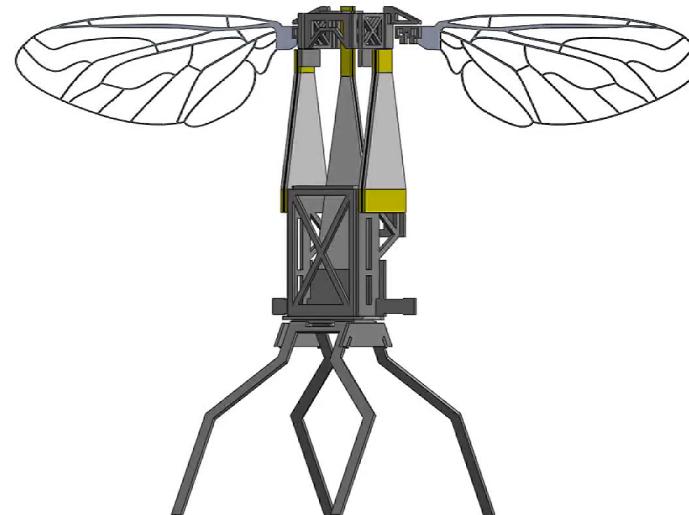


Image: Popular Science



# ROBOBEES

- Half-gram flapping-wing autonomous micro-aerial vehicle
- The RoboBees are an extremely **resource-scarce** platform
  - Limited power
  - Limited computation
  - Limited sensing





# ROBOBEE

- The project is divided into three parts:

## Body

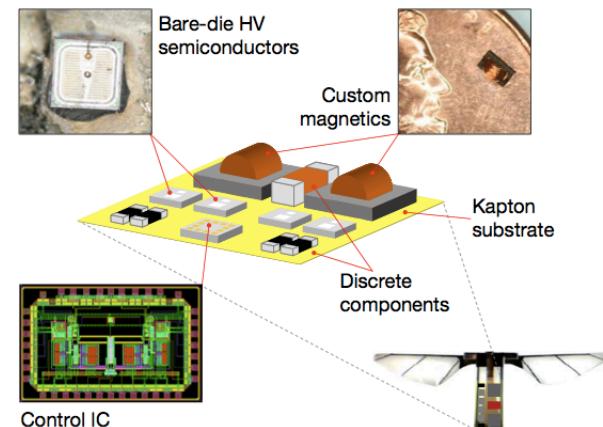
- Designing an insect-sized, autonomous flapping-wing micro-aerial vehicle (MAV)



Image: Eliza Grinnel

## Brain

- Development of sensors, control, and circuitry to direct flight



Karpelson et al., 2010

## Colony

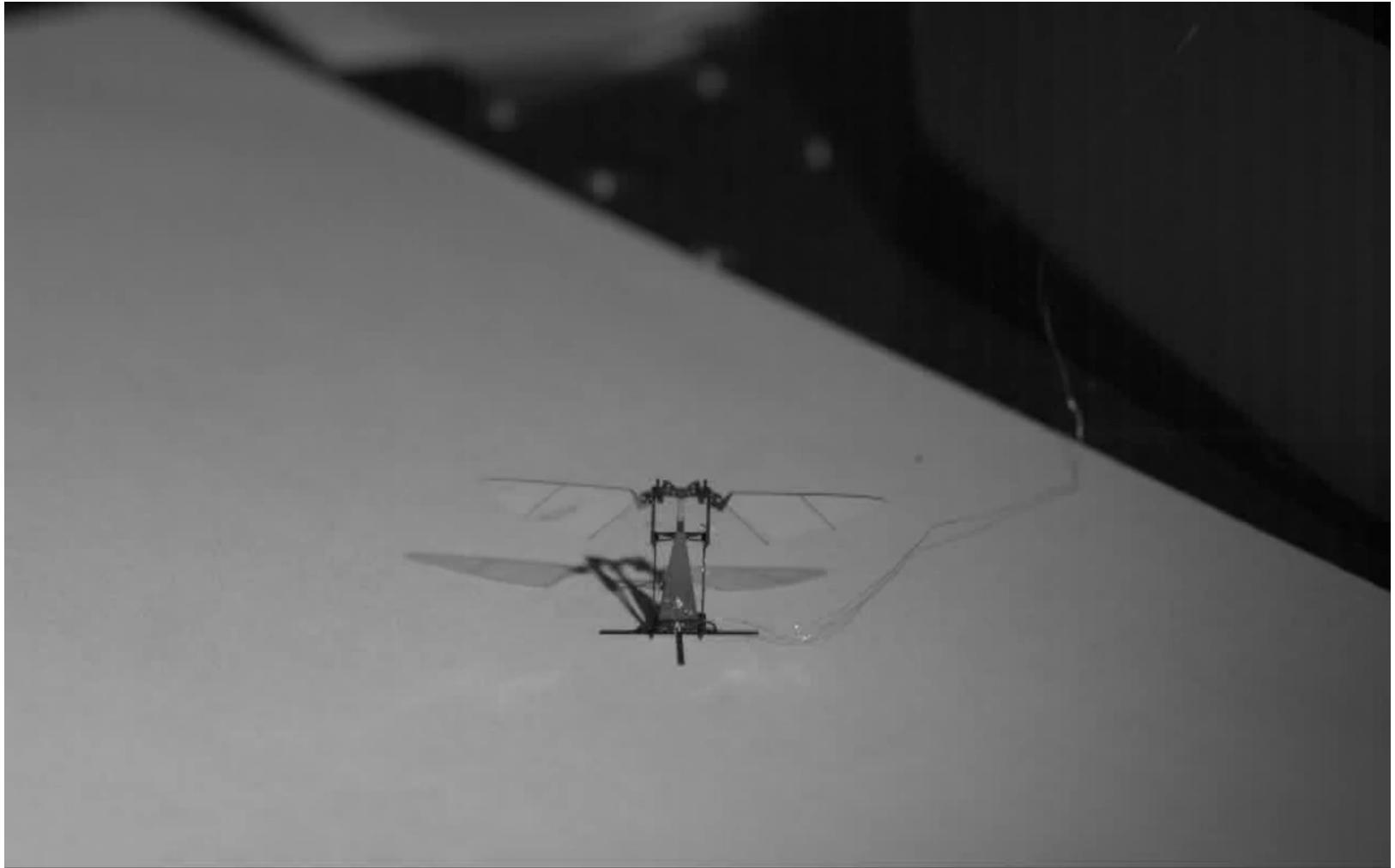
- Coordinate a hive of miniature robots to accomplish tasks



Kate et al., 2012



# ROBOBEES



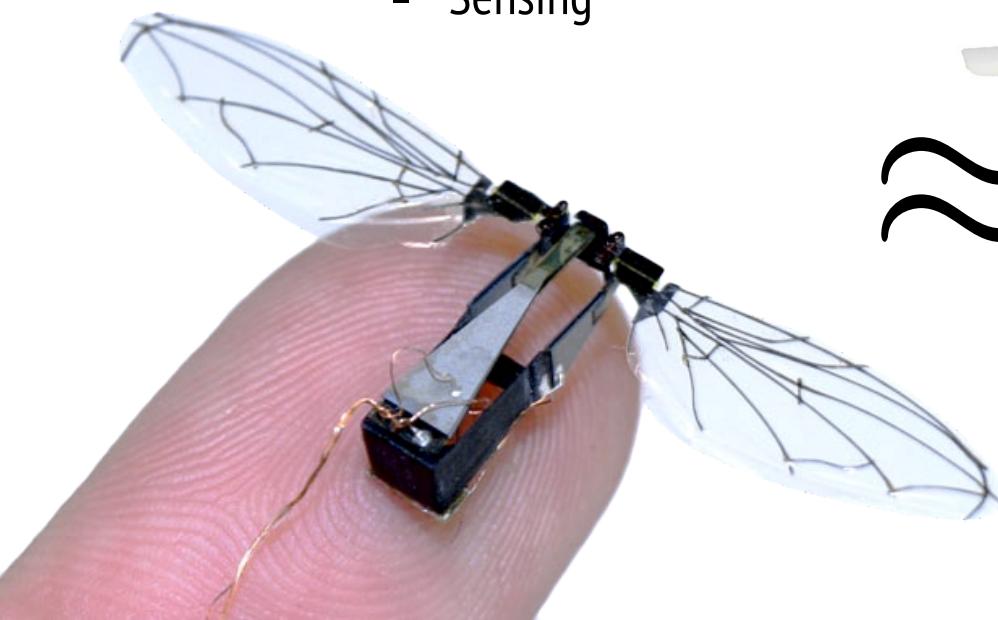
Progress...but not there yet.



# ROBOBEES

- We use a micro-helicopter platform as a stand-in for the RoboBees while under development
- Similar limitations on
  - Weight
  - Power
  - Computation
  - Sensing

→ Vision-based sensing





# VISION

- The RoboBees will use **omnidirectional vision sensors** for navigation and control
  - Low power
  - Lightweight
  - High information density
  - Existing sensor technology
- We have developed an specialized vision-based **sensor ring** for use with our helicopter platform



Source: Czech Technical University

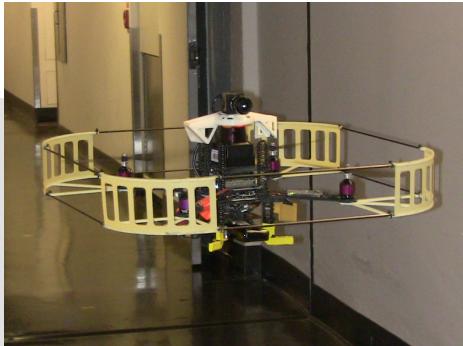
# GOALS



- Get the RoboBee-inspired hardware ready for testing
  - Vision-based sensing
  - Fully on-board computation
- Build a simulation for testing behaviors



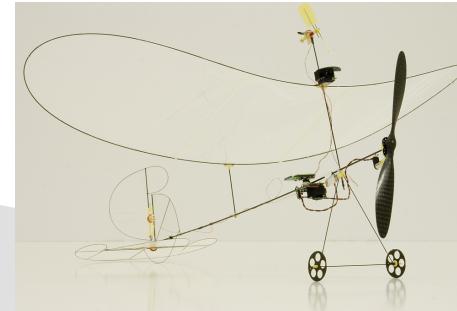
# STATE OF THE ART



Shen et al., 2011



Calafell et al., 2011



Zufferey et al., 2006



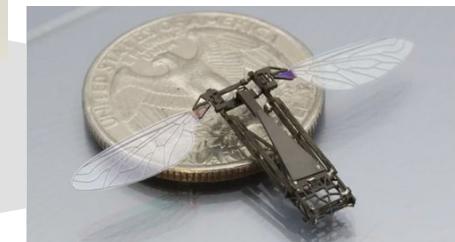
Zingg et al., 2012



Bermes et al., 2010



Bitcraze, 2013



Wood et al., 2012

Larger payload  
Greater computational power  
Extended sensing



Small size  
Little computational power  
Limited sensing

# PLATFORM

- Coaxial helicopter
  - based on Blade mCX2
  - 19 cm rotor diameter
  - 2 DC motors for rotors
  - 2 linear servos adjust swashplate
- Completely autonomous
  - custom control board designed by Centeye Ltd.
  - lithium-polymer battery
  - gyroscope
  - wireless radio
- Custom optic flow sensor ring
- 30 g total weight





# SENSOR RING



## Sensor control board

- 48 MHz 32-bit AVR microcontroller
- Readout images from vision chips
- Perform optic flow estimation and send to helicopter control board

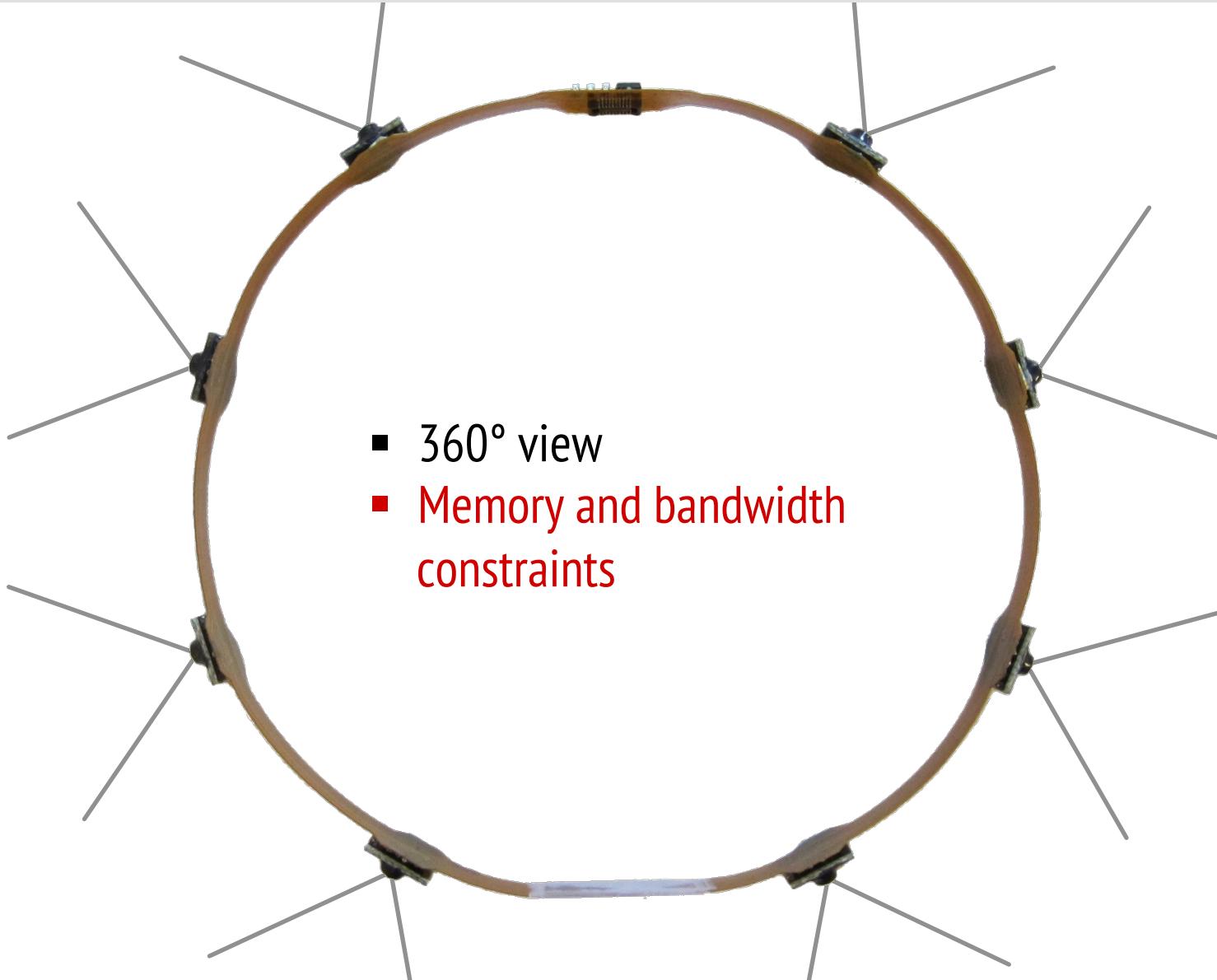
## Sensor strip

- Flexible printed circuit board (PCB)
- 10 cm diameter ring

## Vision chips

- 8 specialized vision chips from Centeye Ltd.
- 64 x 64 array of logarithmic pixels
- 10-bit ADC readout
- Integrated optics with ~75° field-of-view

# SENSOR RING



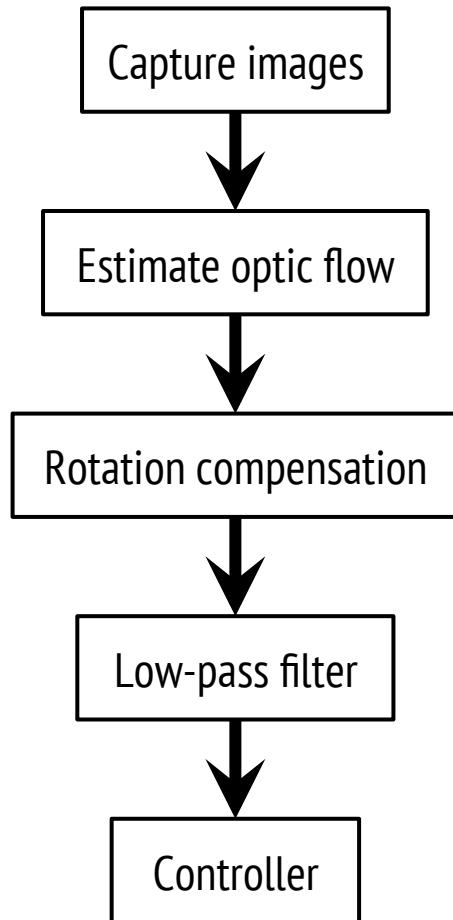
# OPTIC FLOW

- We use the sensor ring to measure **optic flow** around the MAV
- Optic flow is the **apparent visual motion** of objects, surfaces, and edges in a **scene caused by the relative motion** between an observer and a scene





# OPTIC FLOW



- We want to extract information from the environment by estimating the optic flow around the vehicle



# OPTIC FLOW

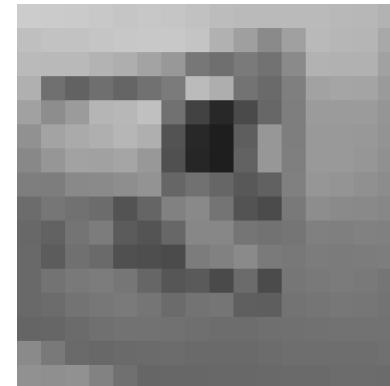
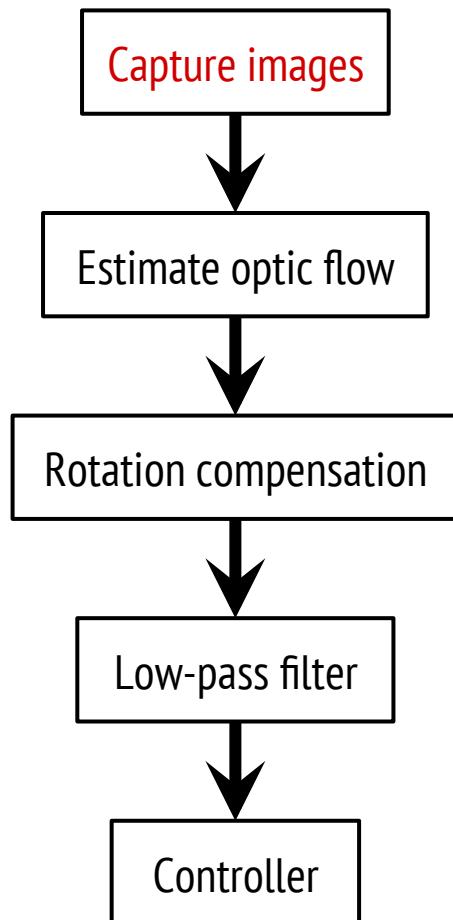


Image at  $t_0$

$$f_0(x, y)$$

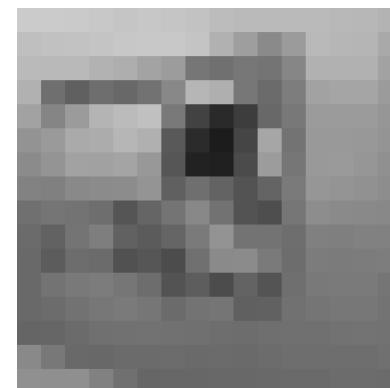
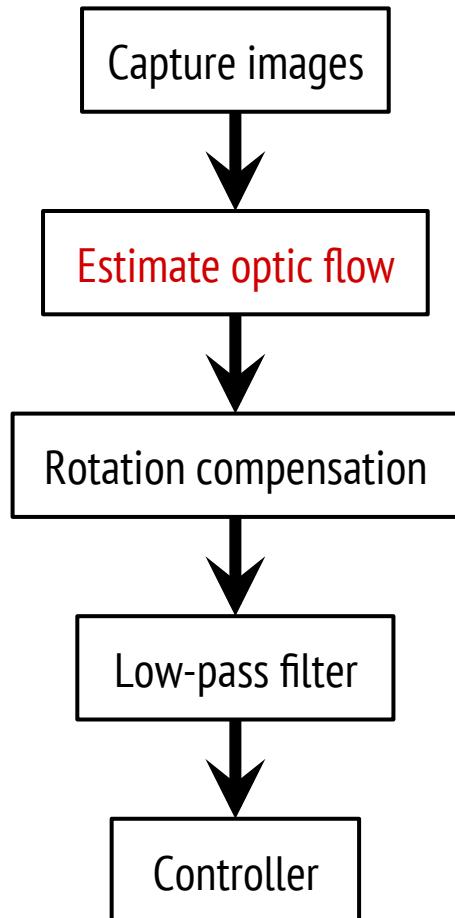


Image at  $t$

$$f(x, y)$$

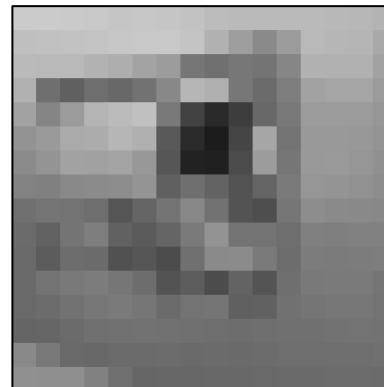


# OPTIC FLOW



## Image interpolation algorithm (I2A)

- Simple, fast, and lightweight algorithm
- Assumes the current image can be linearly interpolated from previous image and space shifted reference images

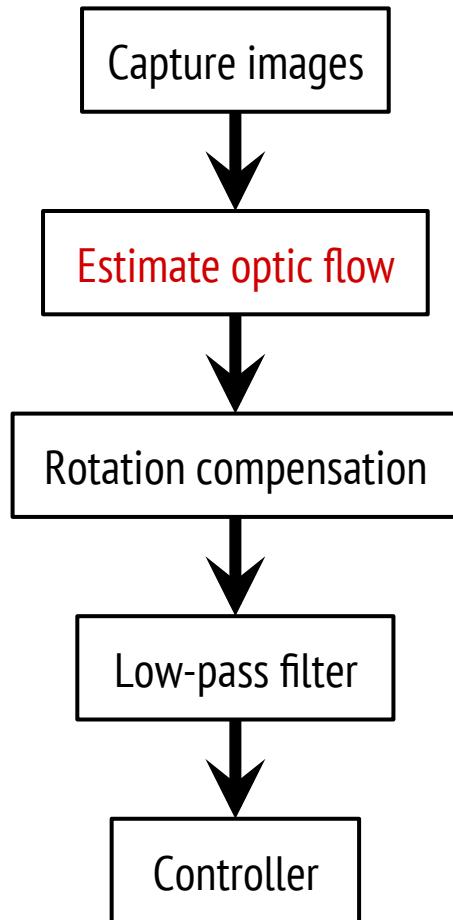


$$\begin{aligned}f_1(x, y) &= f_0(x + \Delta x_{ref}, y) \\f_2(x, y) &= f_0(x - \Delta x_{ref}, y) \\f_3(x, y) &= f_0(x, y + \Delta y_{ref}) \\f_4(x, y) &= f_0(x, y - \Delta y_{ref})\end{aligned}$$

$$\hat{f} = f_0 + 0.5 \left( \frac{\Delta x}{\Delta x_{ref}} \right) (f_2 - f_1) + 0.5 \left( \frac{\Delta y}{\Delta y_{ref}} \right) (f_4 - f_3)$$

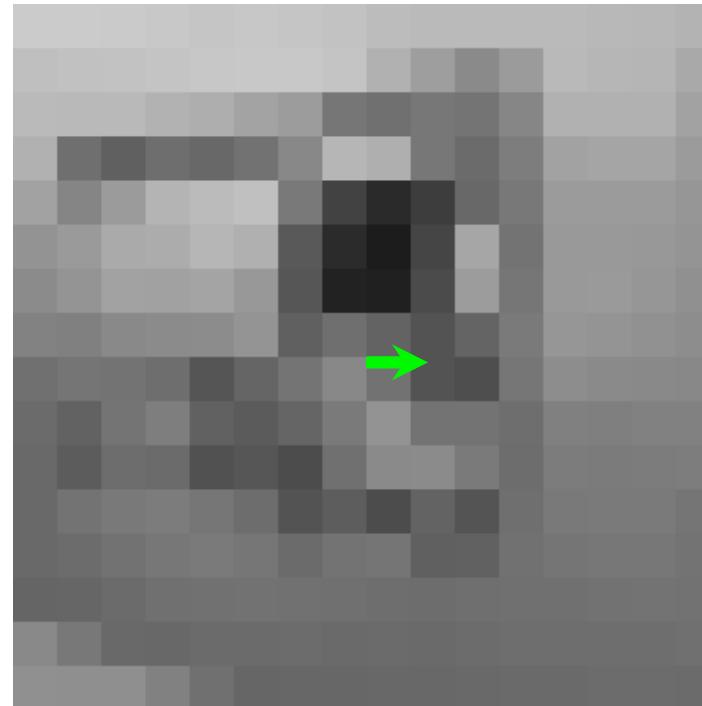


# OPTIC FLOW



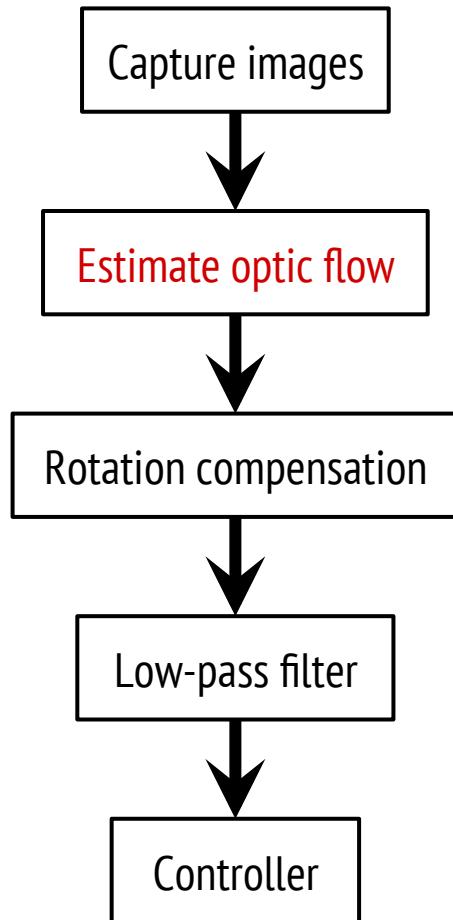
## Image interpolation algorithm (I2A)

- By minimizing the error between the interpolated image and actual image, we can estimate the flow





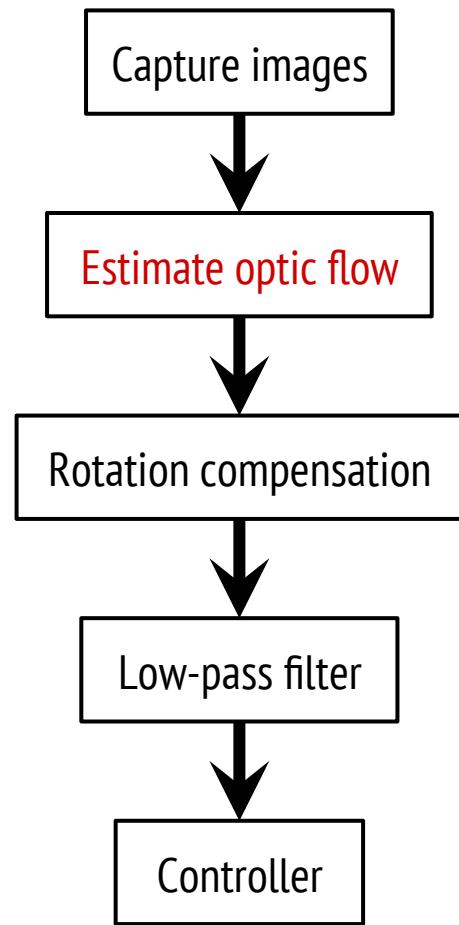
# OPTIC FLOW



What causes this optic flow?

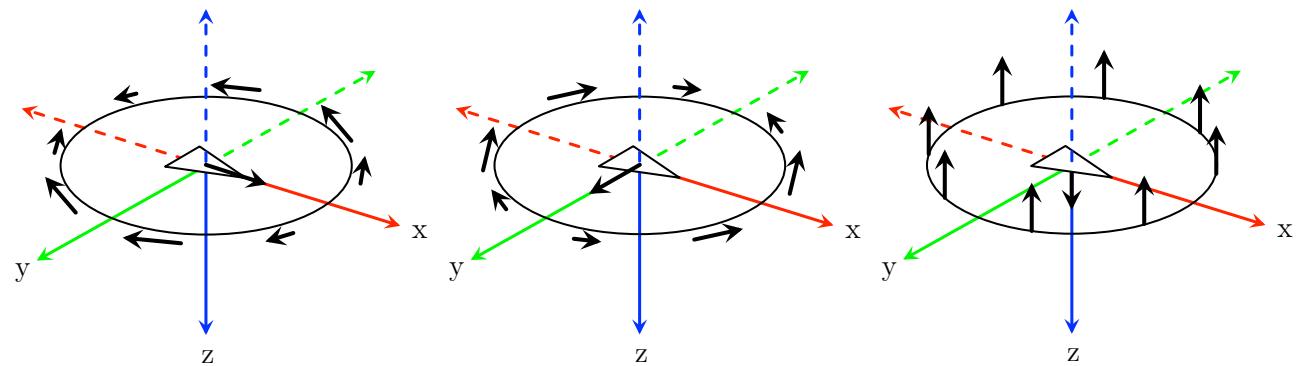


# OPTIC FLOW

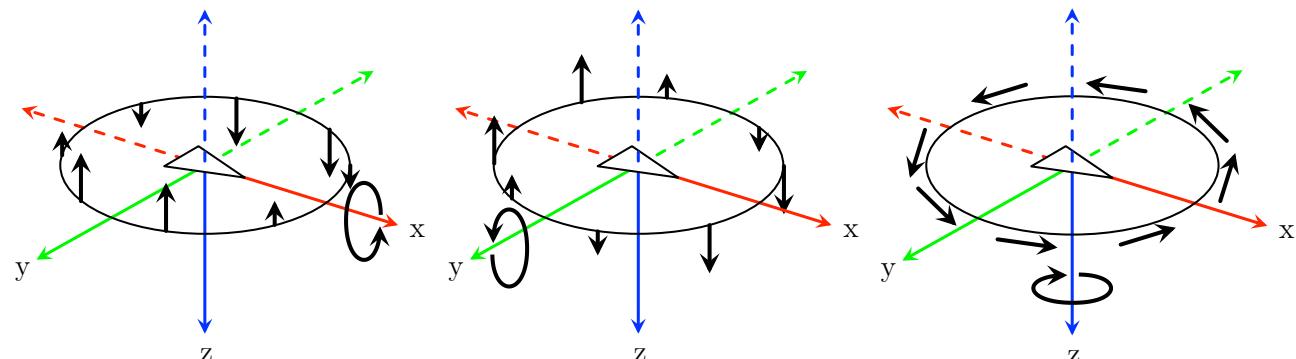


- There are two, additive components to optic flow:

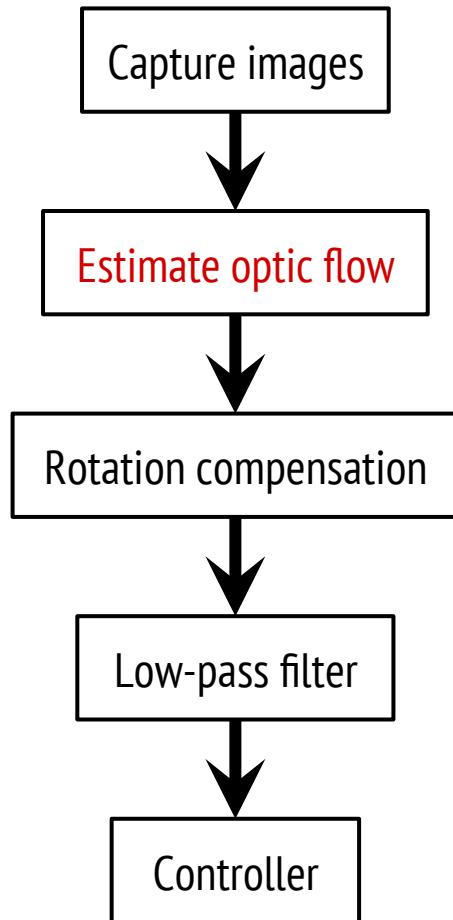
Translation



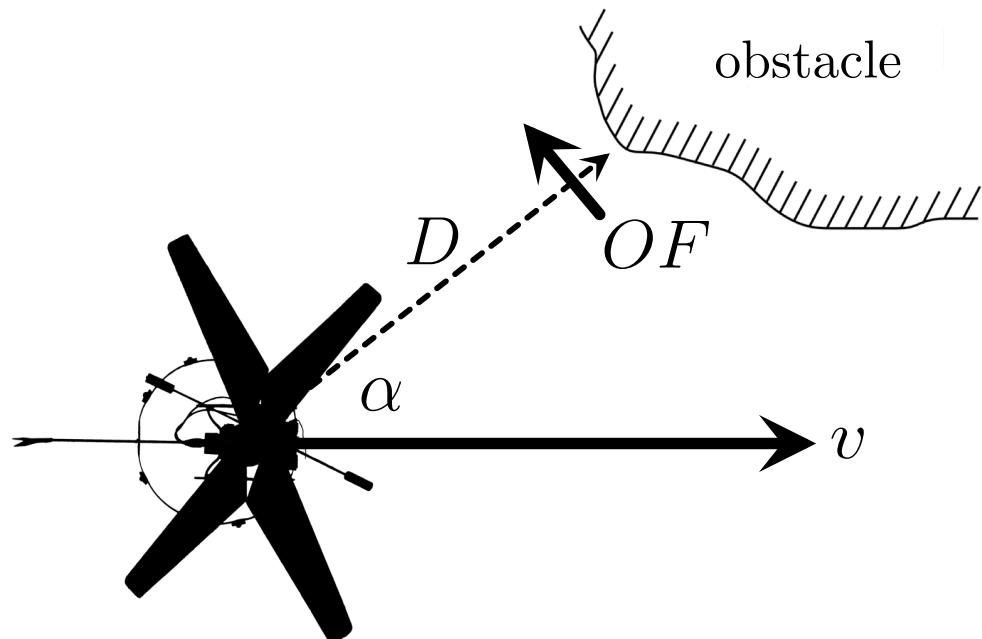
Rotation



# OPTIC FLOW



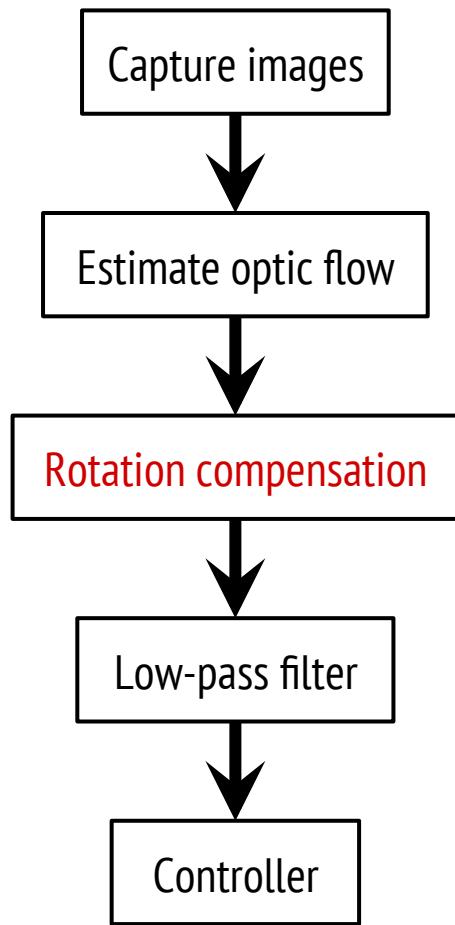
- Translational optic flow gives us information about the environment
- **Rotational optic flow does not**



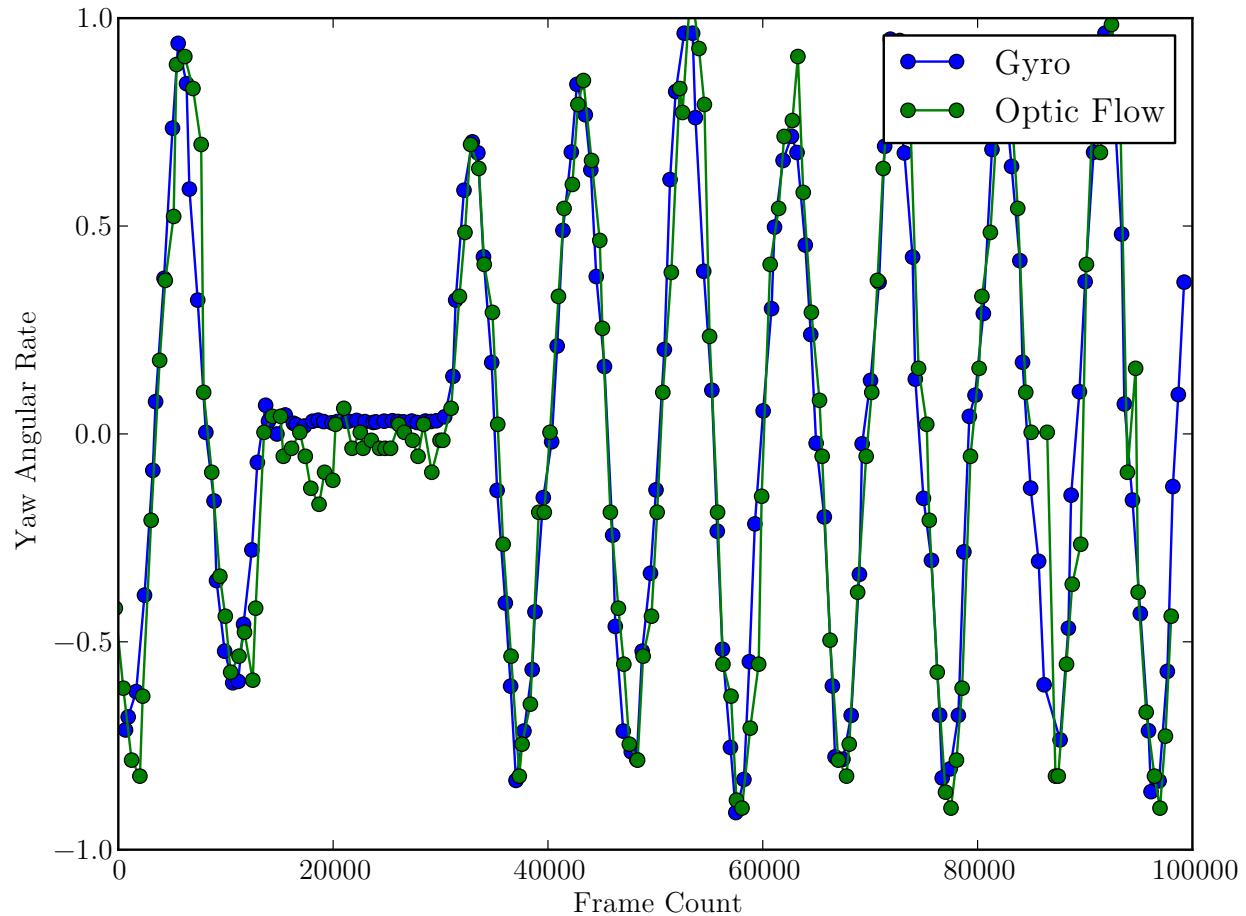
$$OF = \frac{v}{D} \cdot \sin \alpha$$



# OPTIC FLOW

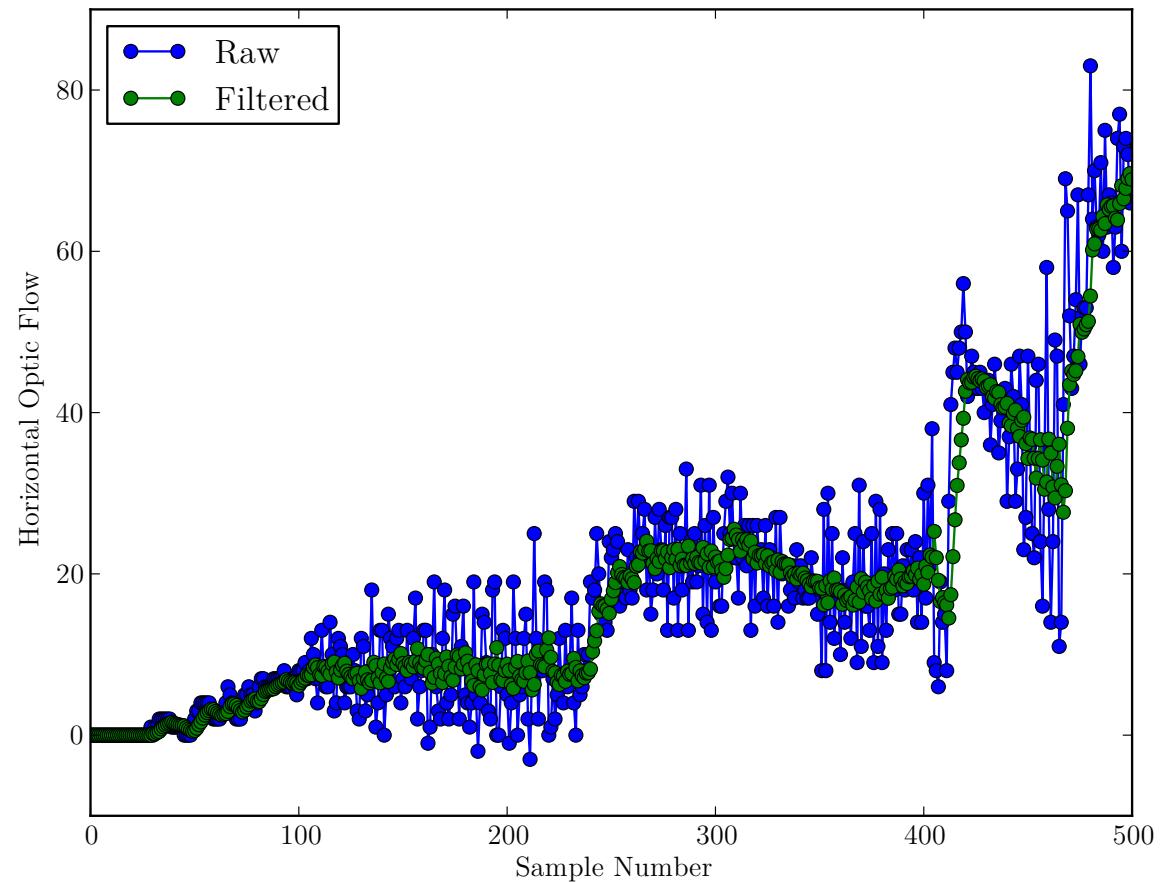
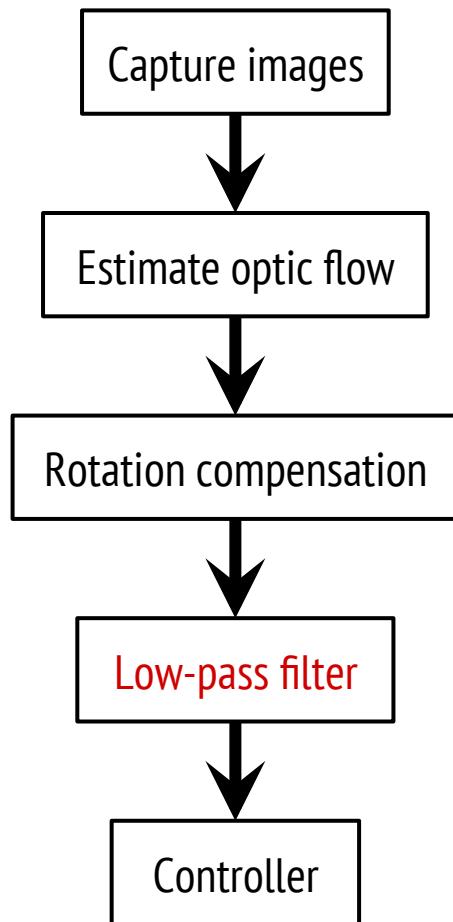


- Subtract rotational optic flow component based on inertial sensors



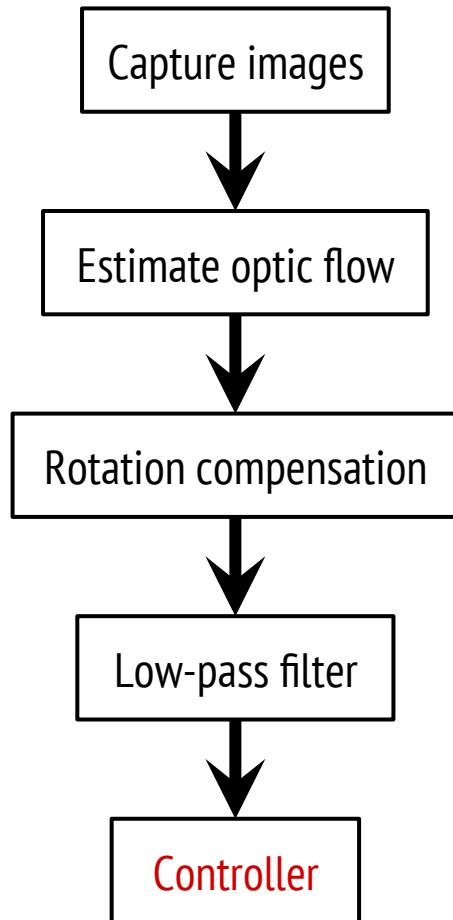


# OPTIC FLOW





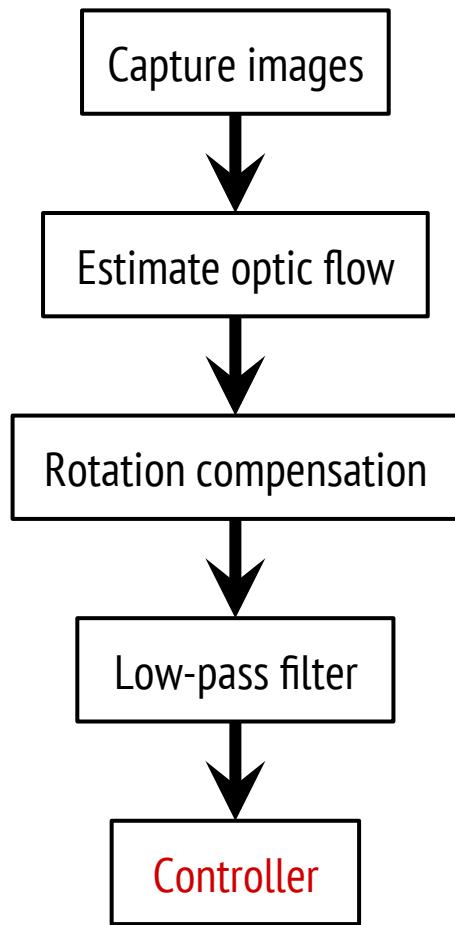
# OPTIC FLOW



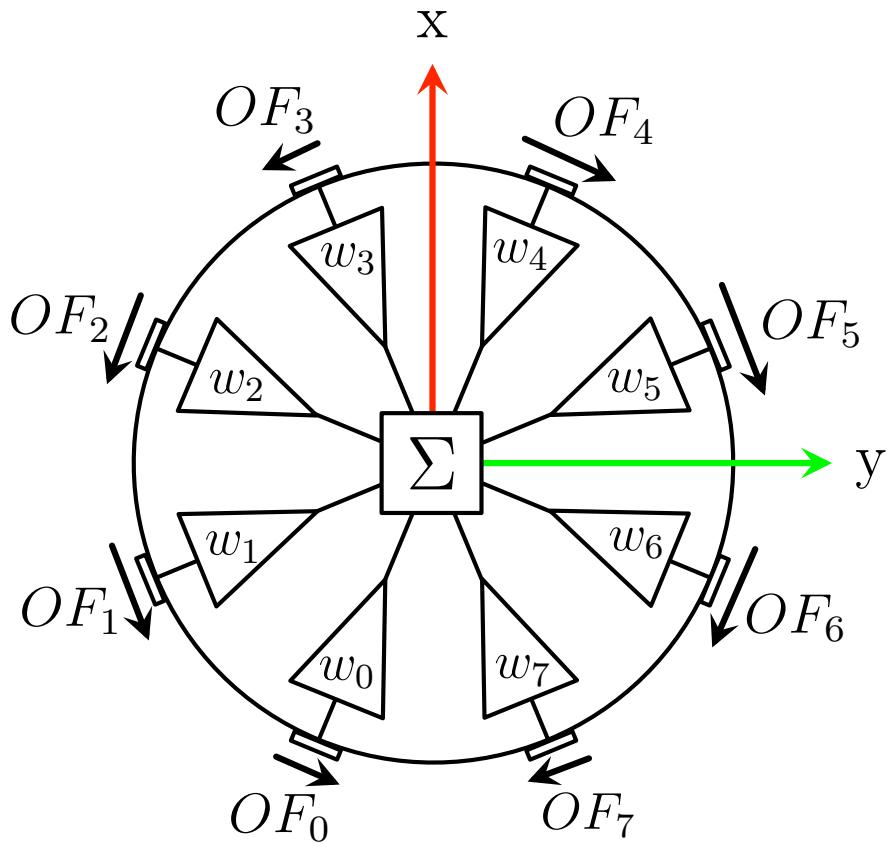
- The optic flow data give information about the environment that can be used for control
- Use a set of base autonomous behaviors
  - Corridor following
  - Wall following
  - Obstacle avoidance
  - Hover in place



# OPTIC FLOW



- Simple, reactive controller:  $\delta_j = \sum_{i=0}^7 w_{i,j} \cdot OF_i$





# HARDWARE





# HARDWARE

- We worked to get the platform ready for testing
  - Communication with sensor ring
  - Image acquisition
  - Optic flow calculation
  - Collection of sample datasets
  - Reading from gyroscope
  - De-rotating flow
  - Wireless communication
  - Vicon motion tracking





# HARDWARE

- When working with MAVs, there are a variety of difficulties associated with using actual hardware
  - Limited testing environments
  - Limited flight time
  - Harder to program devices (longer iteration time)
  - Difficulty tracking motion
  - Difficulty logging data
  - Hardware malfunctions

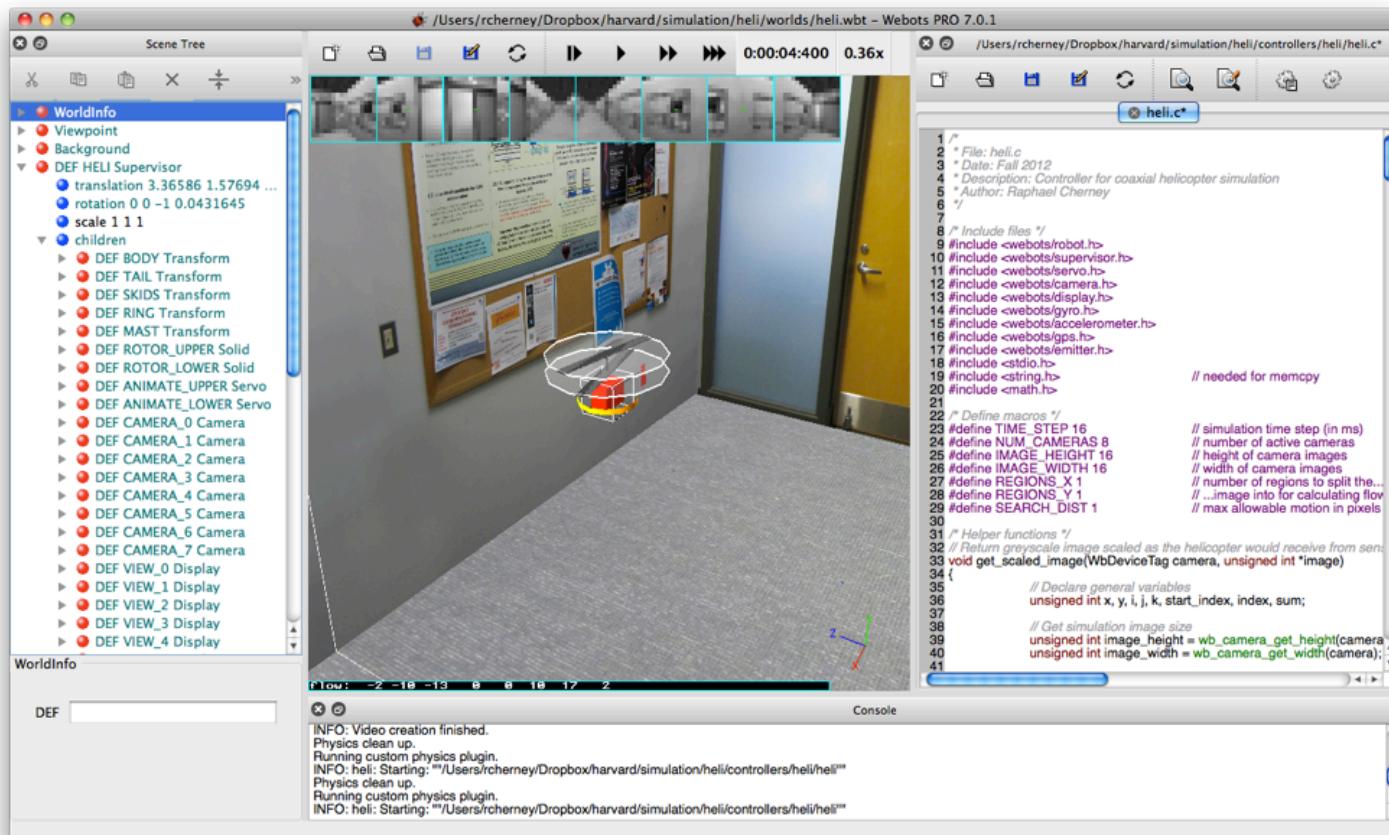


# SIMULATION



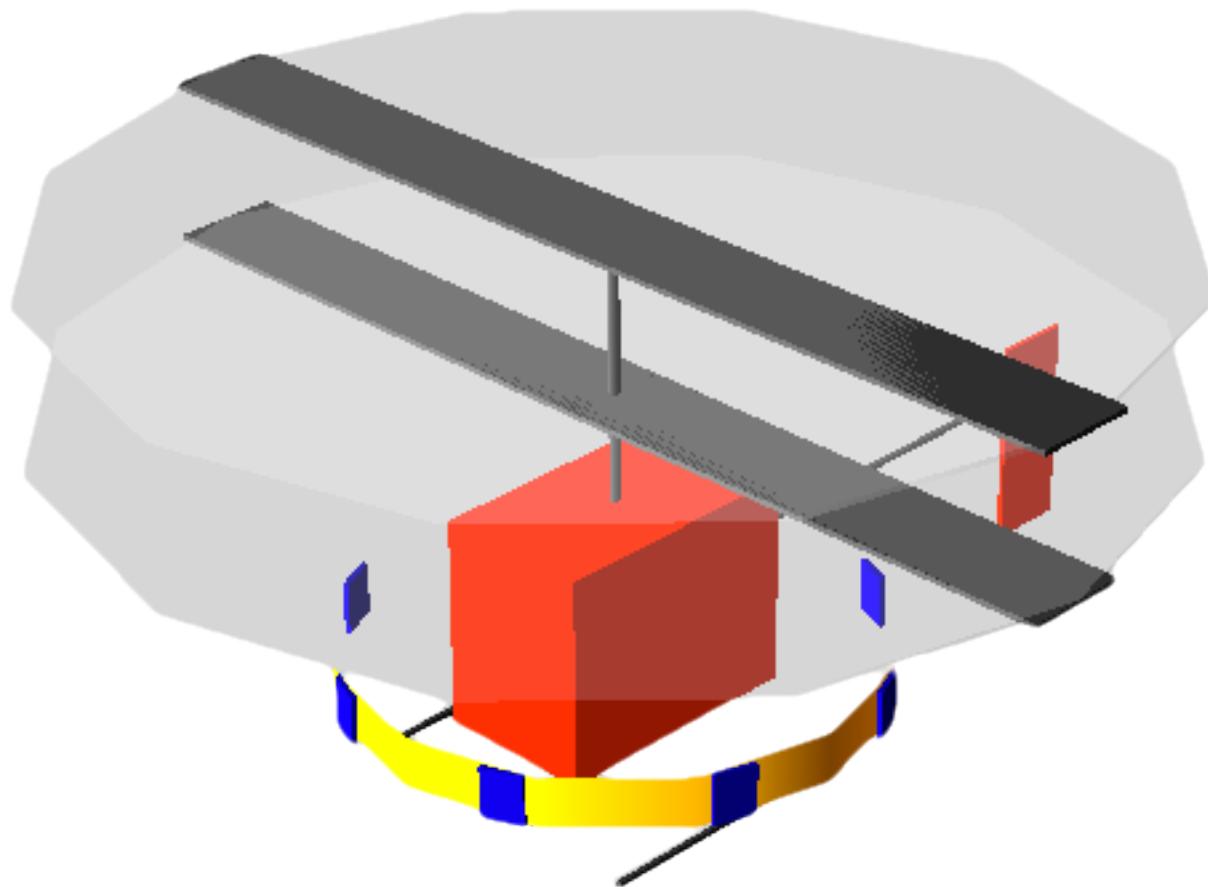
# SIMULATION

- Developed a model of the MAV and indoor environment for testing



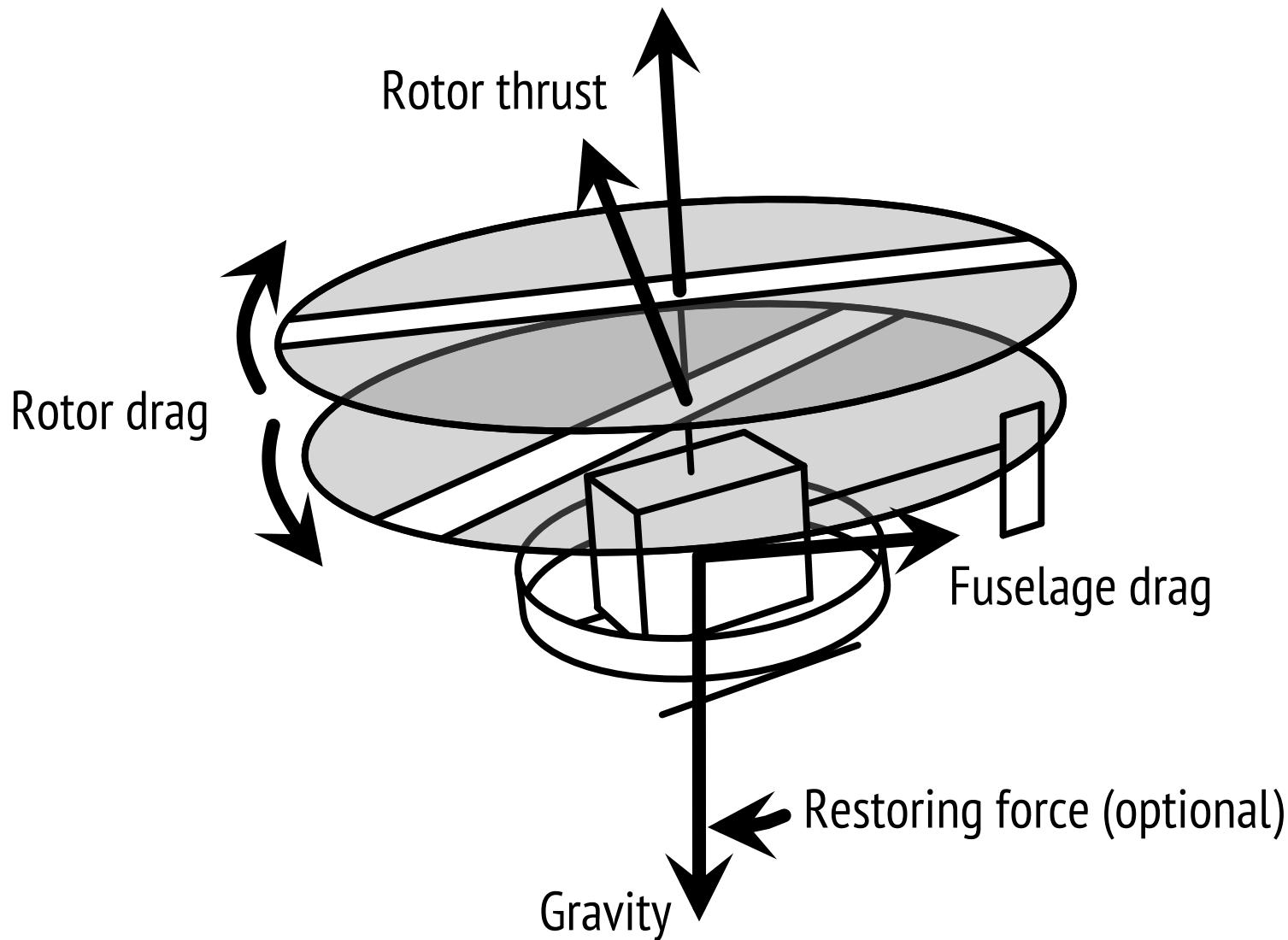


# MODEL



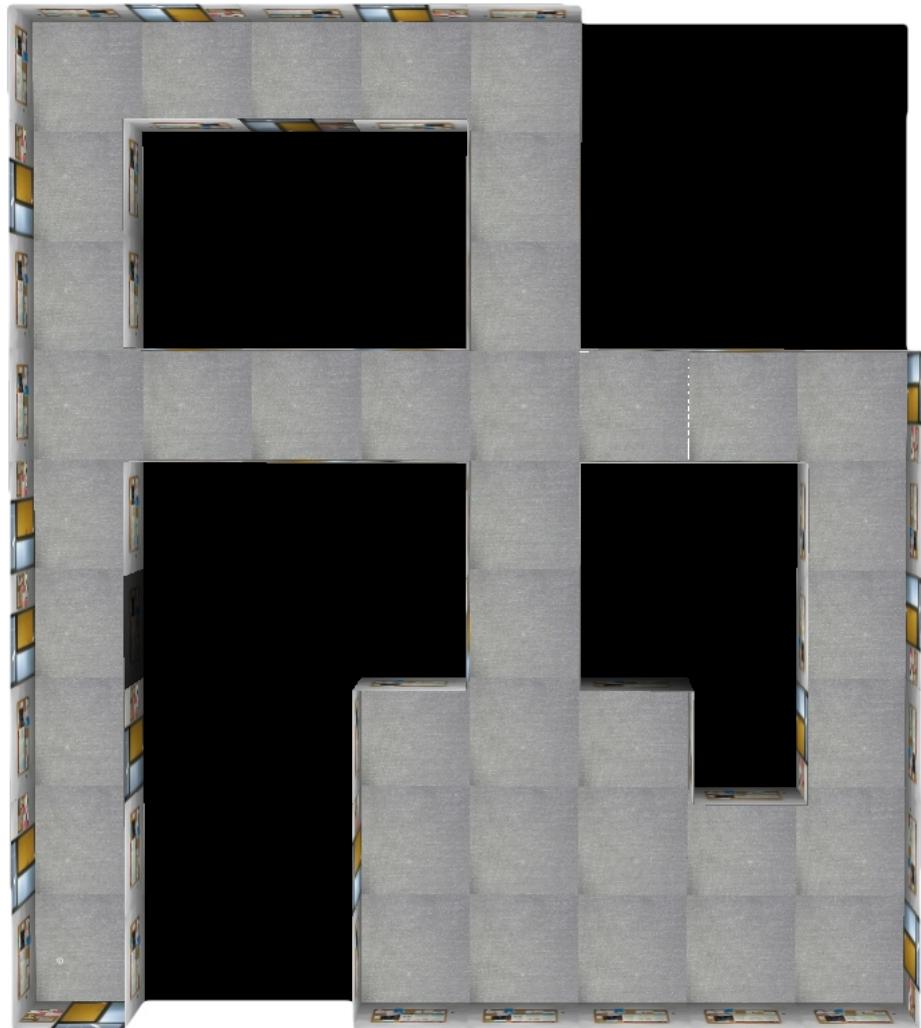


# PHYSICS



# ENVIRONMENT

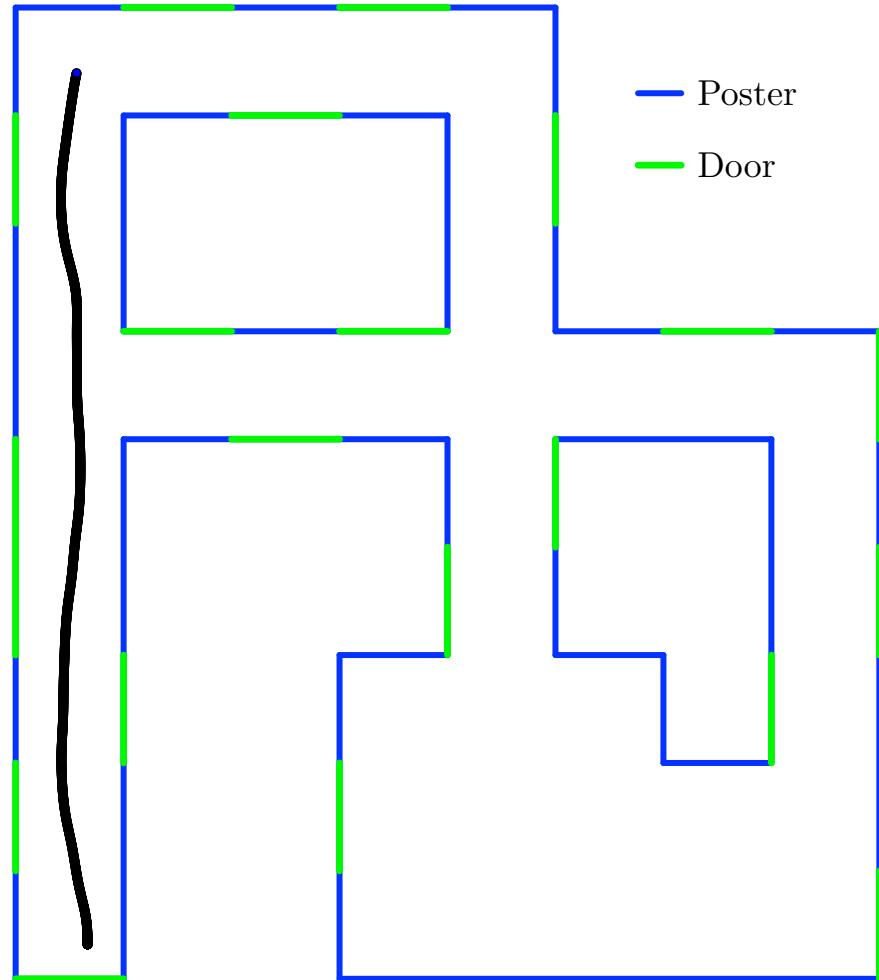
- Realistic, indoor simulation environment for testing
- Textures from actual test environment





# SIMULATION

- We can **test behaviors in simulation** and then port controllers to our hardware
- As an example, we implemented a simple corridor centering algorithm
  - Balance optic flow

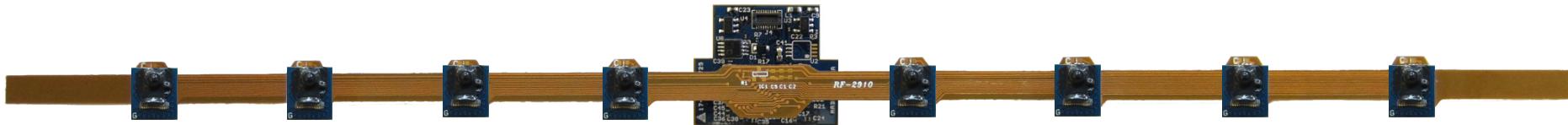


# SIMULATION





# SENSOR CONFIGURATIONS



- The optic flow sensor ring has a memory and bandwidth limitations
- What is the best configuration for reading out sensor data?



# SENSOR CONFIGURATIONS

Camera Angle:  $\frac{-7\pi}{8}$     $\frac{-5\pi}{8}$     $\frac{-3\pi}{8}$     $\frac{-\pi}{8}$     $\frac{\pi}{8}$     $\frac{3\pi}{8}$     $\frac{5\pi}{8}$     $\frac{7\pi}{8}$

Configuration A:

$16 \times 16$							

Configuration B:

$32 \times 32$	$32 \times 32$	$32 \times 32$	$32 \times 32$

Configuration C:

$64 \times 8$							

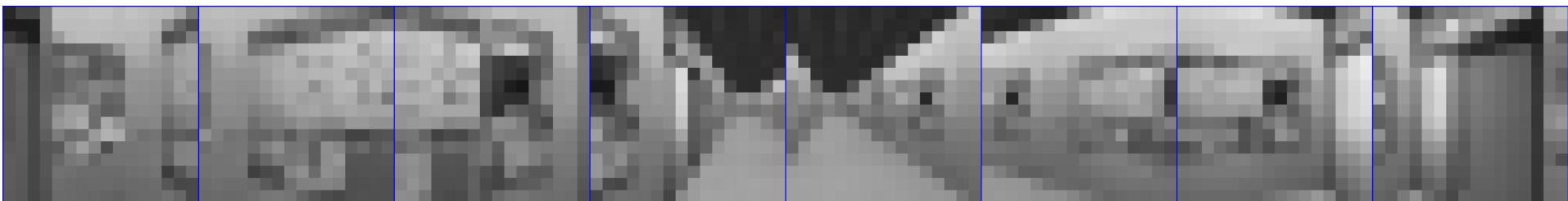
(Constant pixel size)



# SENSOR CONFIGURATIONS

## Configuration A

$16 \times 16 (\times 8)$



### Advantages

- Evenly distributes images around MAV
- Efficient binning of pixels for improved signal
- Can easily condense data into logical vectors
- Easy to implement

### Disadvantages

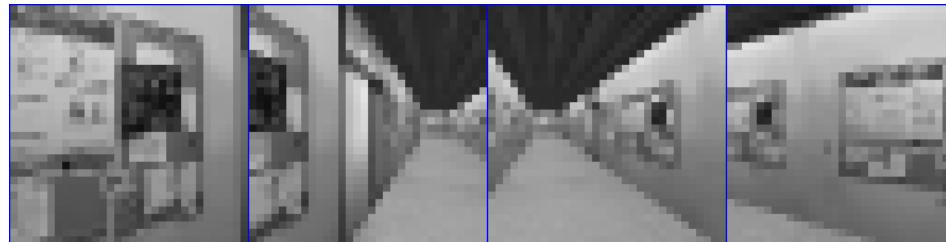
- Low pixel count limits number of flow vectors that can be found with a given camera



# SENSOR CONFIGURATIONS

## Configuration B

32×32 ( $\times 4$ )



### Advantages

- Better resolution when traveling “forward”
- Increased sensitivity in the optic flow estimate (due to higher resolution)
- Can support larger number of optic flow vectors per image
- Image overlap provides redundancy

### Disadvantages

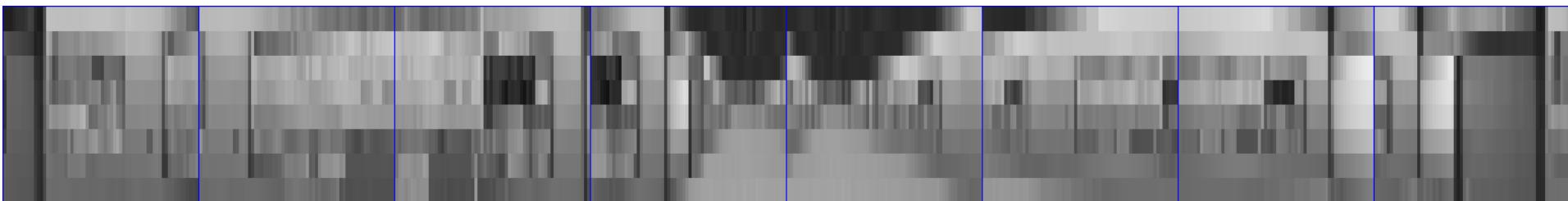
- No data from behind (cannot track obstacles you have passed)
- Overlapping images means we are not maximizing our total field of view (redundancy may be unnecessary)



# SENSOR CONFIGURATIONS

## Configuration C

64×8 ( $\times 8$ )



### Advantages

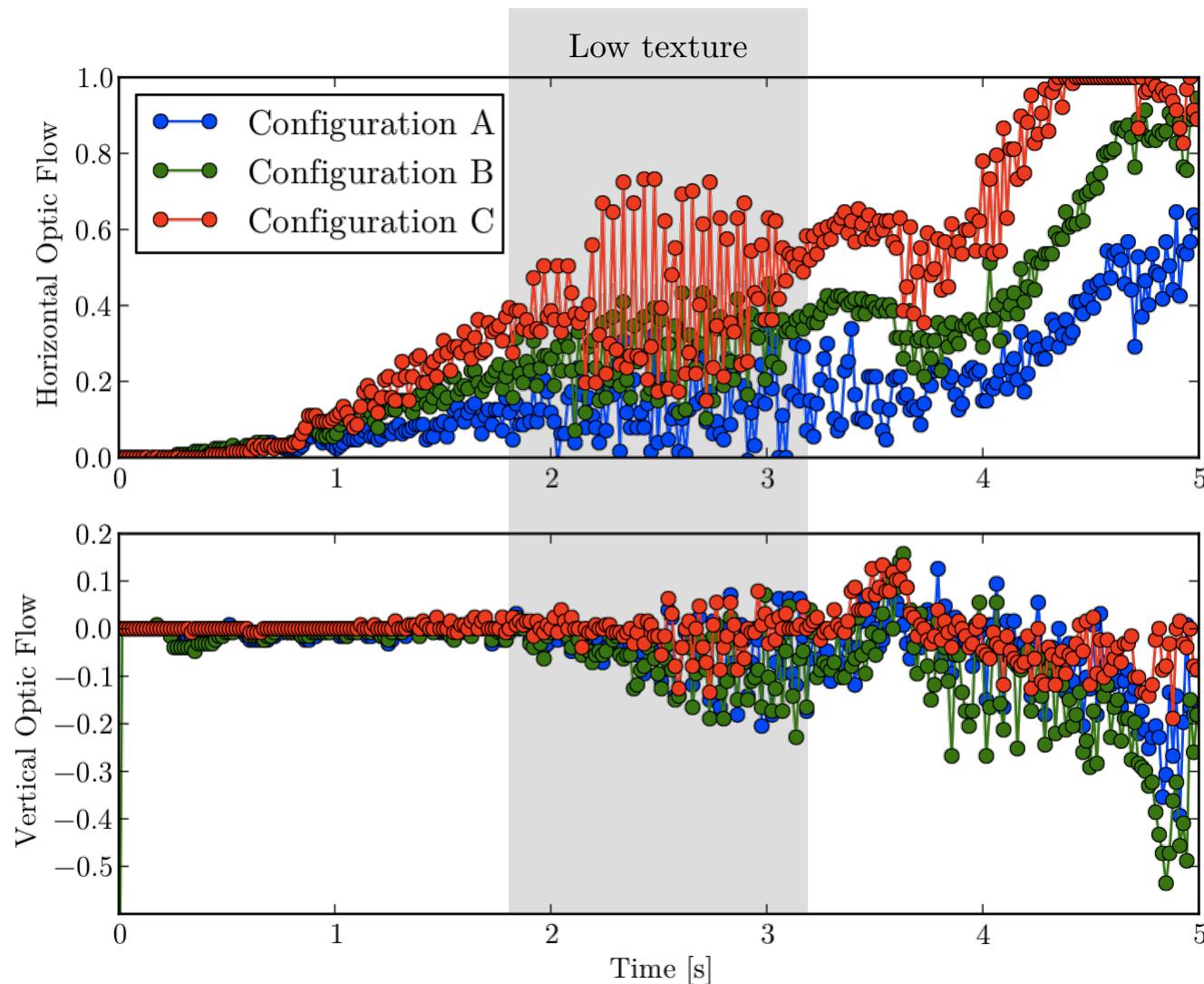
- Maximizes sensitivity for horizontal flow
- Evenly distributes sensing around MAV (bilaterally symmetric)
- Uses full sensor area

### Disadvantages

- Reduced vertical flow data
- More difficult to implement in hardware



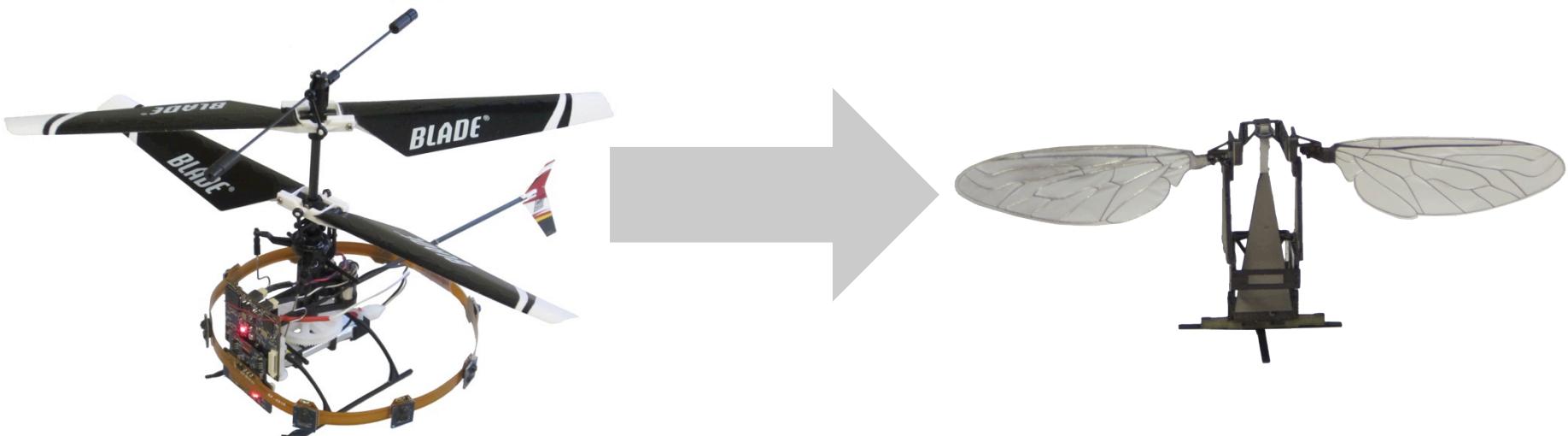
# SENSOR CONFIGURATIONS





# FUTURE WORK

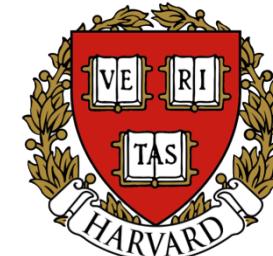
- Test with hardware (in progress)
- Implement additional base behaviors (in progress)
- Egomotion estimation
- Combine base behaviors into advanced control and planning algorithms
- Coordinate action with multiple agents





# ACKNOWLEDGEMENTS

- Radhika Nagpal
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- Swiss Federal Institute of Technology (EPFL)
- Laboratory of Intelligent Systems (LIS)





# QUESTIONS

