

# CSE 490R:

# Mobile Robots

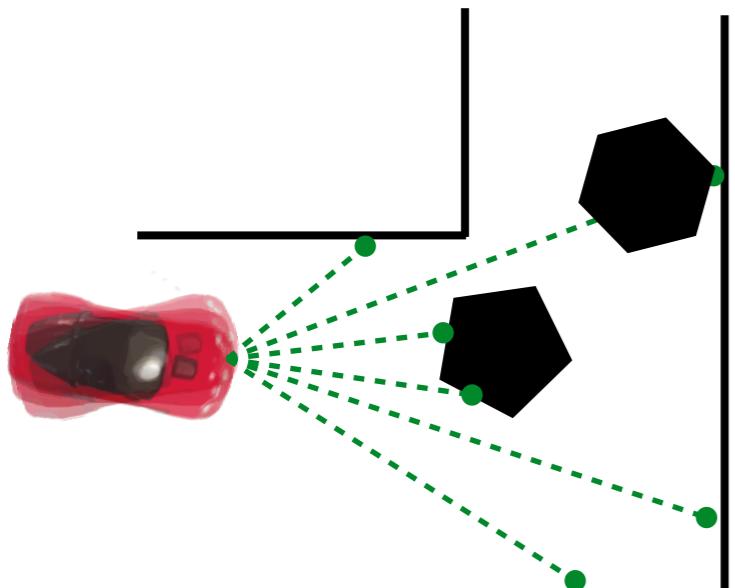
Sanjiban Choudhury

TAs: Matthew Rockett, Gilwoo Lee, Matt Schmittle

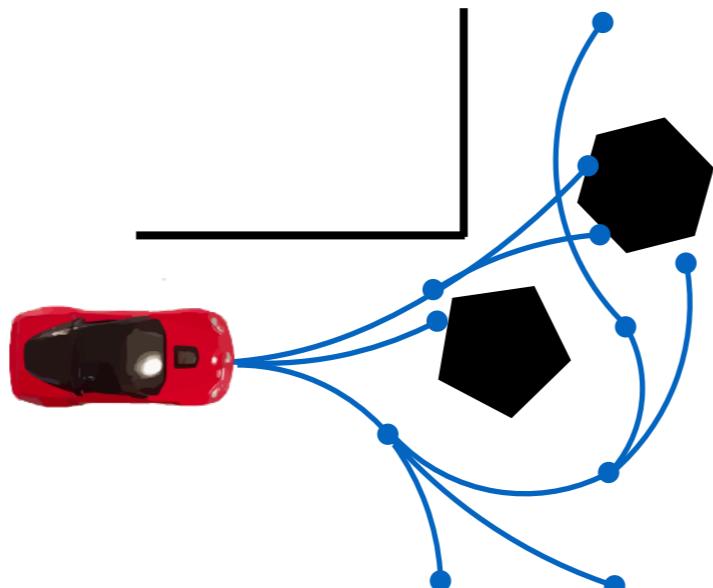
# Logistics

- Submit knowledge survey ASAP
- Form a 3 person team by today, Wednesday 4/3 (send a private note to instructors in Piazza)
- Assignment 0 is released and due on 4/12.
  - Familiarize yourself with ROS
- Come to recitation on Thursday, get your robot and start working on Assignment 0.

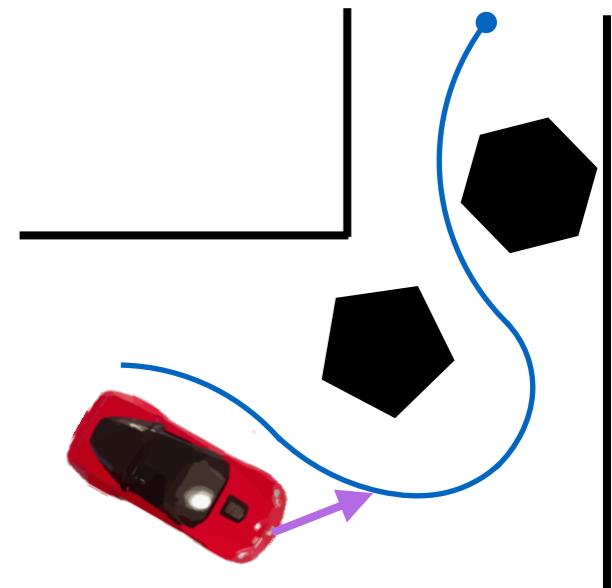
Estimate  
state



Plan a  
sequence of  
motions



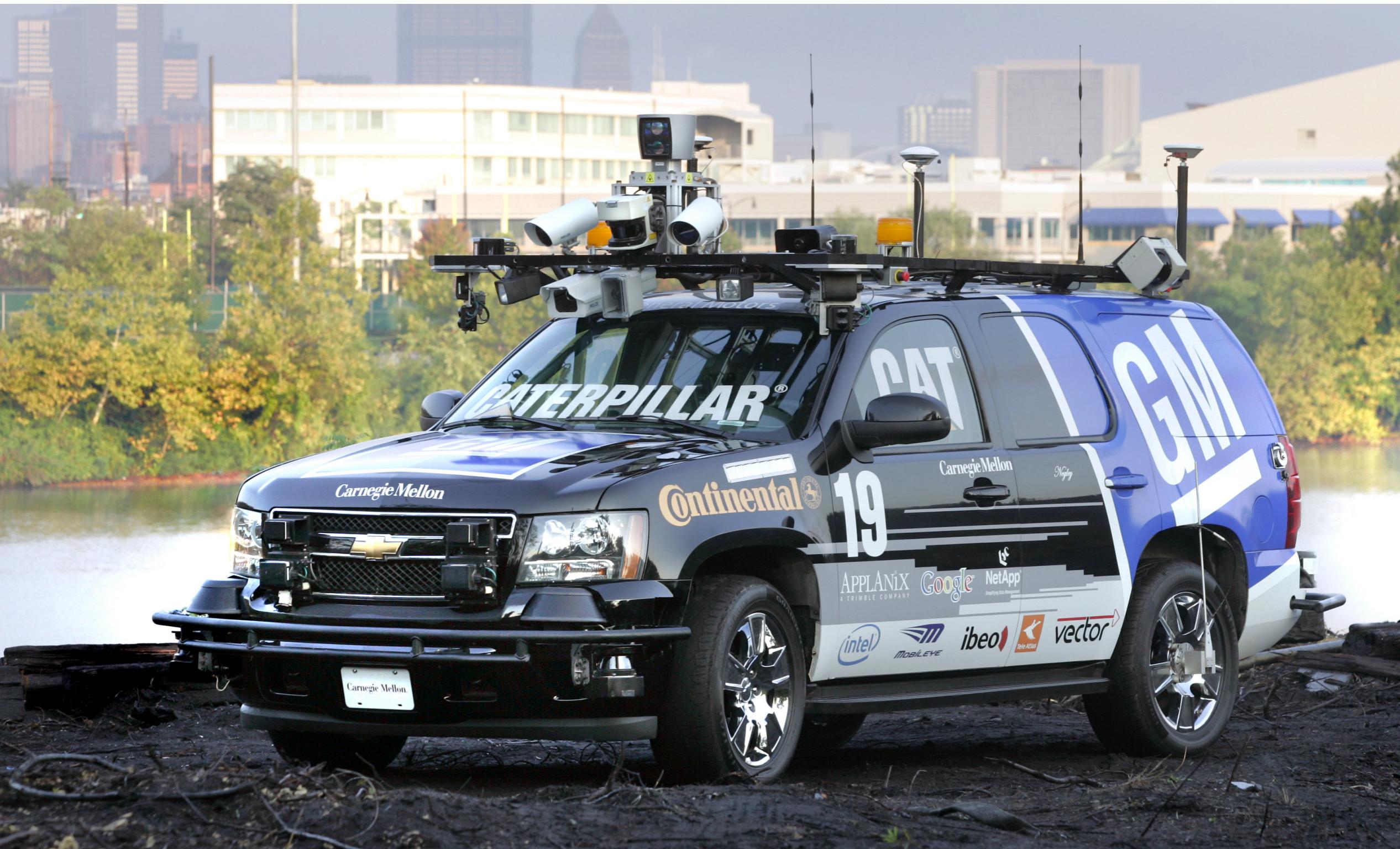
Control  
robot to  
follow plan



# Anatomy of a self-driving car

Sanjiban Choudhury

# BOSS: CMU's winning entry to DARPA challenge



(2007)



# **Tartan Racing: A Multi-Modal Approach to the DARPA Urban Challenge**

*April 13, 2007*

Chris Urmson, Joshua Anhalt, Drew Bagnell, Christopher Baker, Robert Bittner,  
John Dolan, Dave Duggins, Dave Ferguson, Tugrul Galatali, Chris Geyer, Michele Gittleman,  
Sam Harbaugh, Martial Hebert, Tom Howard, Alonzo Kelly, David Kohanbash, Maxim Likhachev,  
Nick Miller, Kevin Peterson, Raj Rajkumar, Paul Rybski, Bryan Salesky, Sebastian Scherer,  
Young Woo-Seo, Reid Simmons, Sanjiv Singh, Jarrod Snider, Anthony Stentz,  
William “Red” Whittaker, and Jason Ziglar

**Carnegie Mellon University**

Hong Bae, Bakhtiar Litkouhi, Jim Nickolaou, Varsha Sadekar, and Shuqing Zeng

**General Motors**

Joshua Struble and Michael Taylor

**Caterpillar**

Michael Darms

**Continental AG**

# Anatomy of a ~~self-driving car~~

## self-flying helicopter

Sanjiban Choudhury

# Autonomous full-scale helicopter

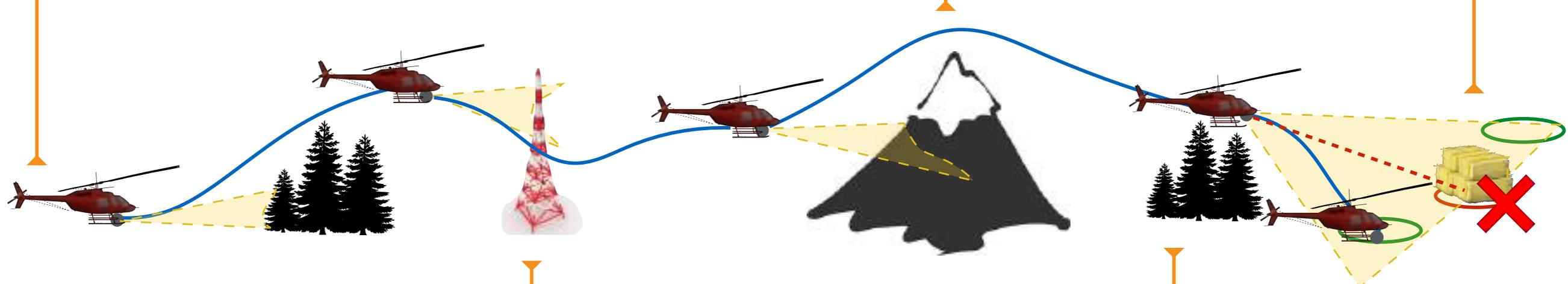


OFFICE OF NAVAL RESEARCH

INNOVATIONS FOR THE FUTURE FORCE

# Mission: Take-off to landing

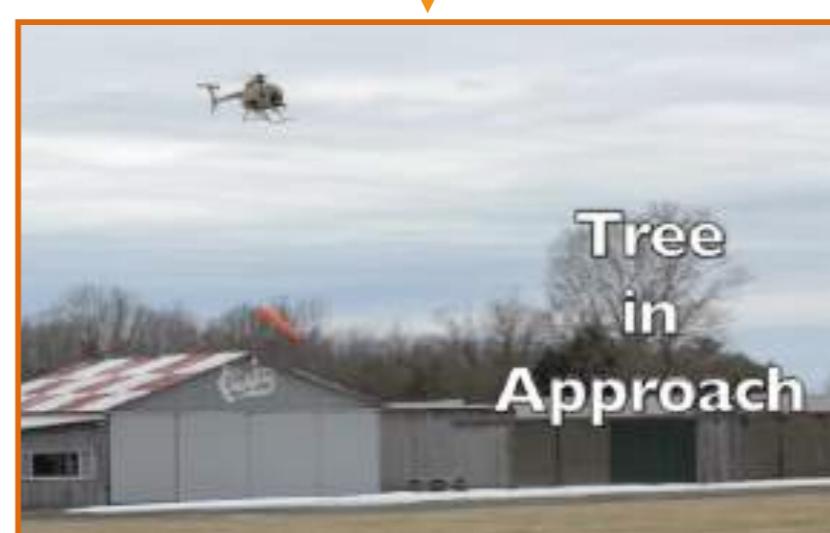
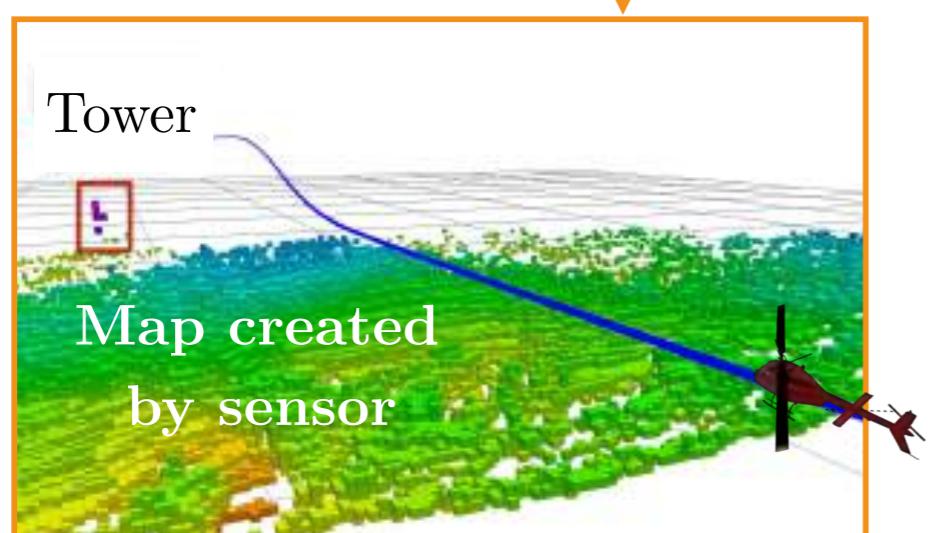




**Takeoff**  
(Respect power constraints)

**Enroute**  
(Avoid sensed obstacles)

**Touchdown**  
(Plan to multiple sites)



# Today's objective

1. Learn how to architect a mobile robotic system
  
  
  
  
  
  
  
  
2. Step through a set of fundamental lessons  
that shape robot system / algorithm design

# Task: A contract the robot has to satisfy

## Given:

Start (latitude, longitude), Goal (latitude, longitude)

List of no-fly-zones (unsafe air space)

Coarse terrain map of continental USA

Sensors - GPS, Laser, etc

## Objective:

Minimize time it takes to complete mission

## Constraint:

Don't come close to obstacles / don't enter no-fly-zones

Don't exceed limits of the vehicle (flying upside down)

How do we tractably solve the task?

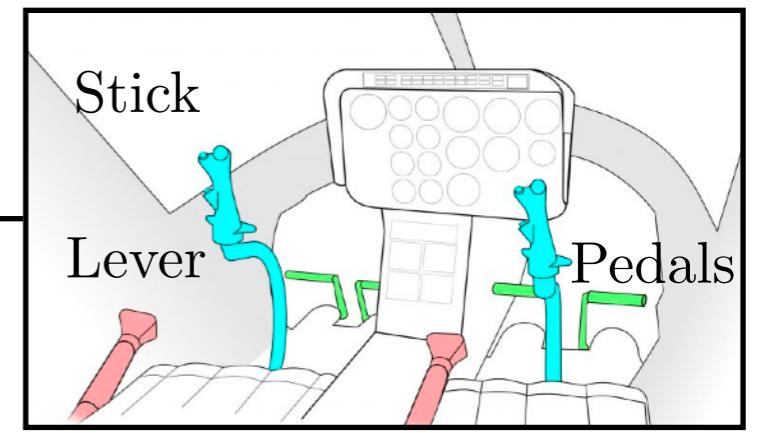
Begin with a blank slate

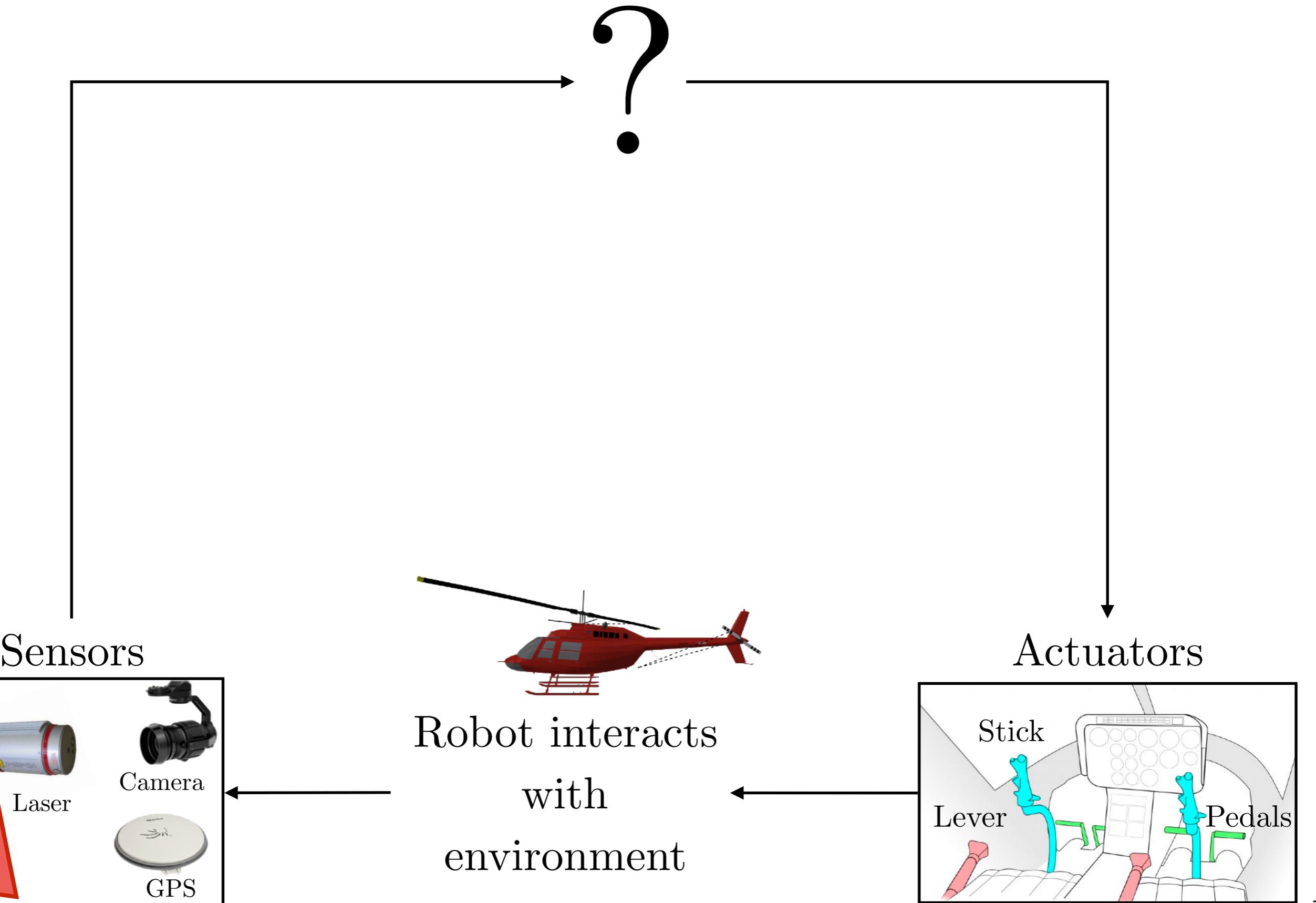
## Sensors



Robot interacts  
with  
environment

## Actuators





# Lesson 0: Look at one piece at a time

Q1: Assume we know everything about the world.

What commands should we send to the actuator?

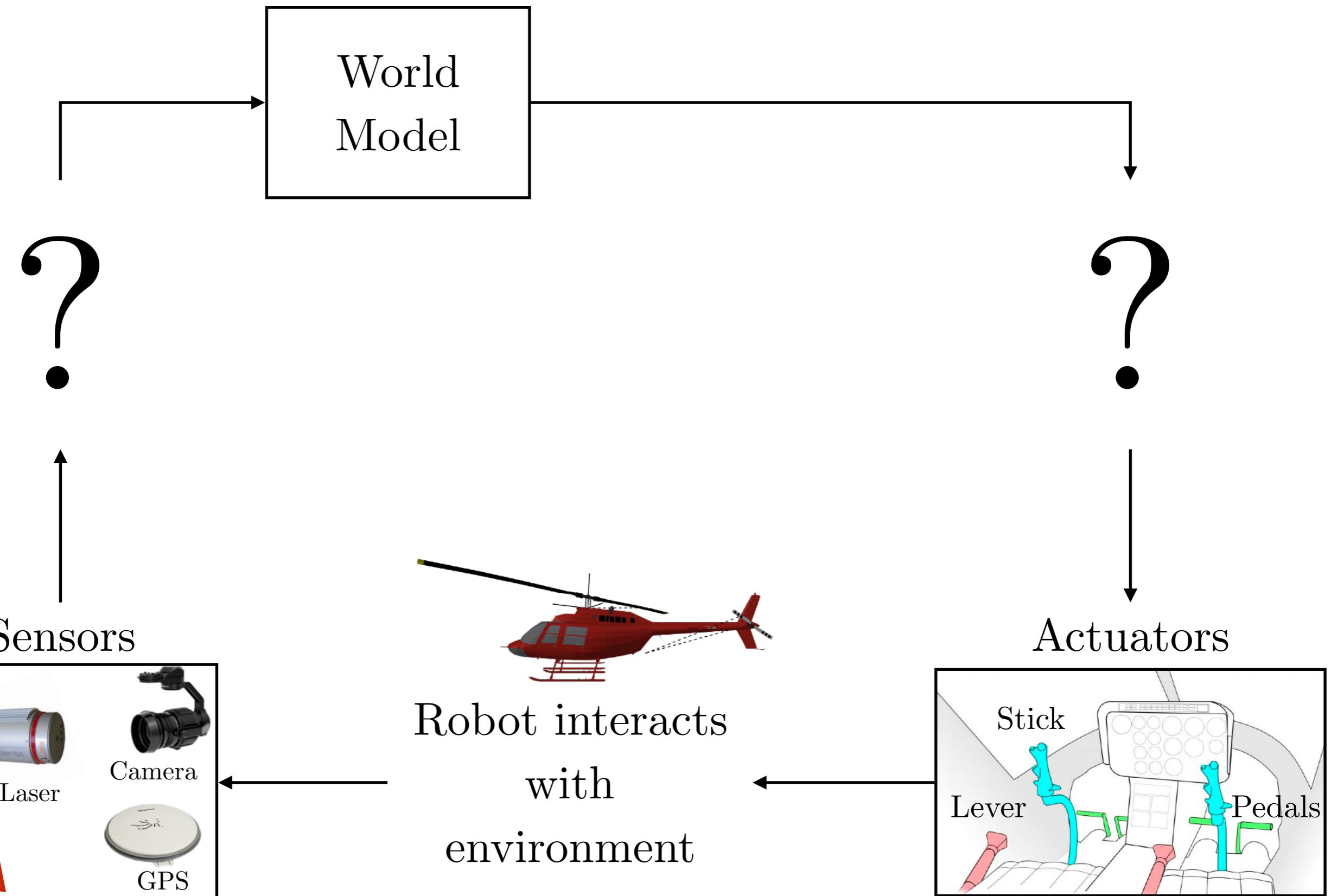
Q2: How do we use raw sensor data to update what we know about the world?

# Look at one piece at a time

Q1: Assume we know everything about the world.

What commands should we send to the actuator?

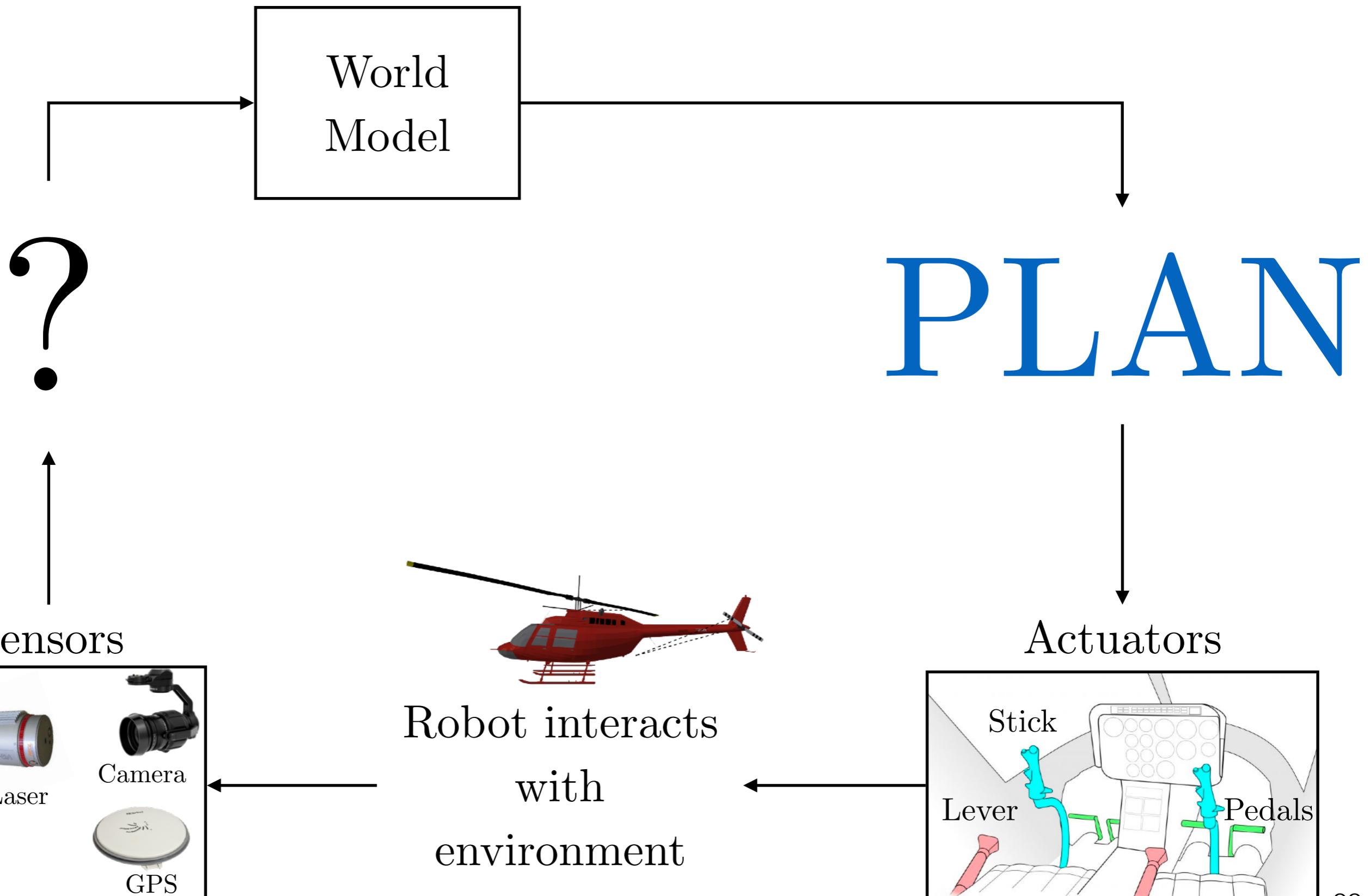
Q2: How do we use raw sensor data to update what we know about the world?



# What is the world model?

List of everything **we need to know** to accomplish the task

- Where is the robot in the world? What is it's state?
- What are the obstacles in the world?
- What type are the obstacles (radio towers, trees)?
- What are the no-fly-zones?
- Are there other aircrafts?
- What is the wind, temperature, etc?



# What is planning?

Planning is an optimization problem where ...

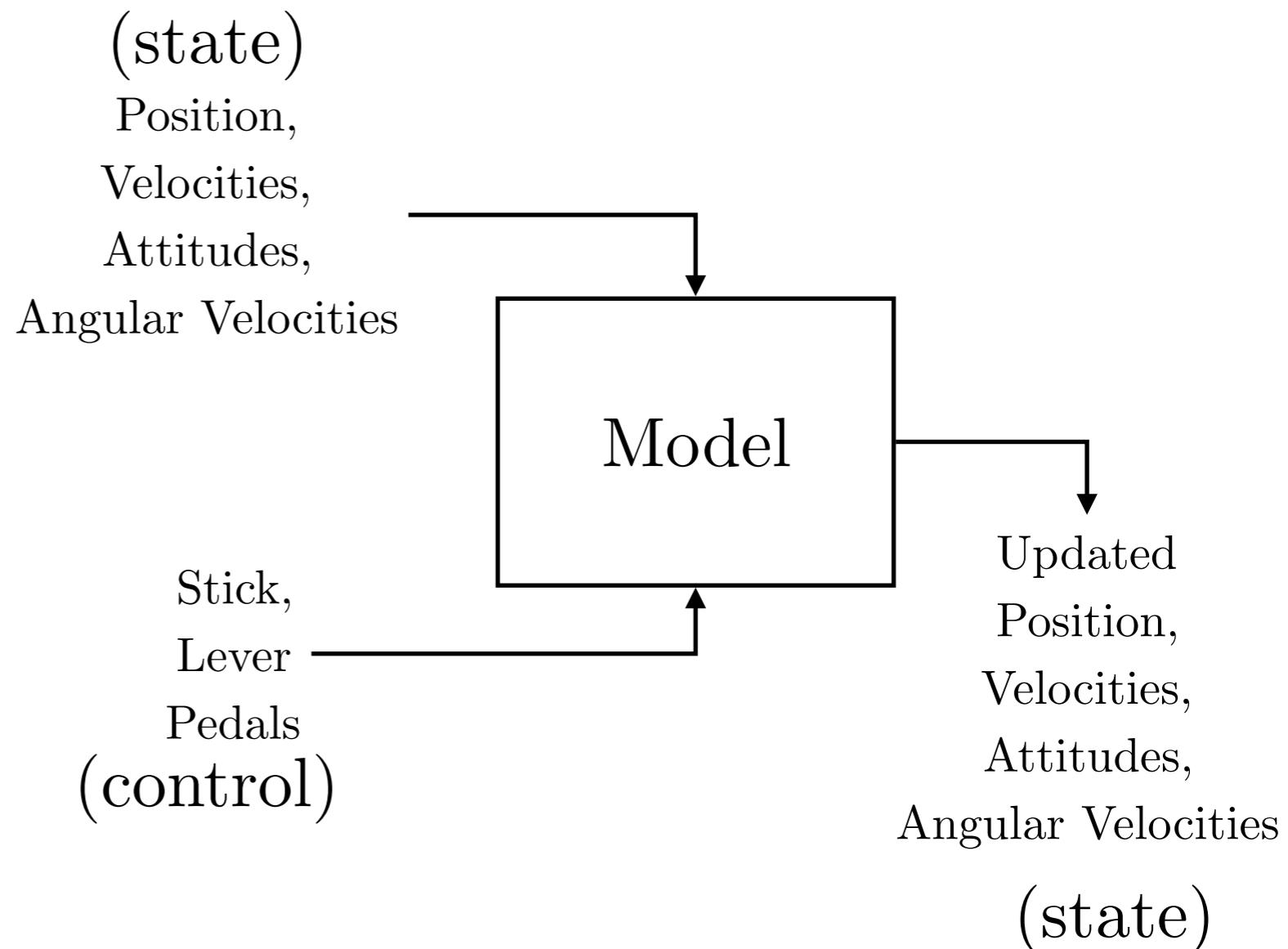
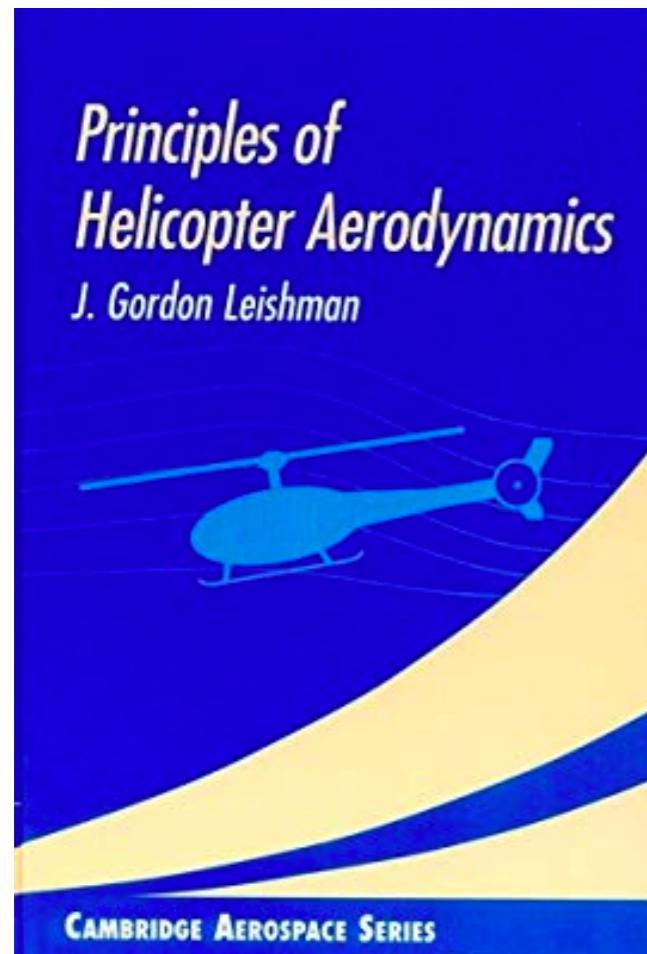
Searching over a sequence of actions

Using a model of your robot to predict where it will go

Minimize time to reach the goal

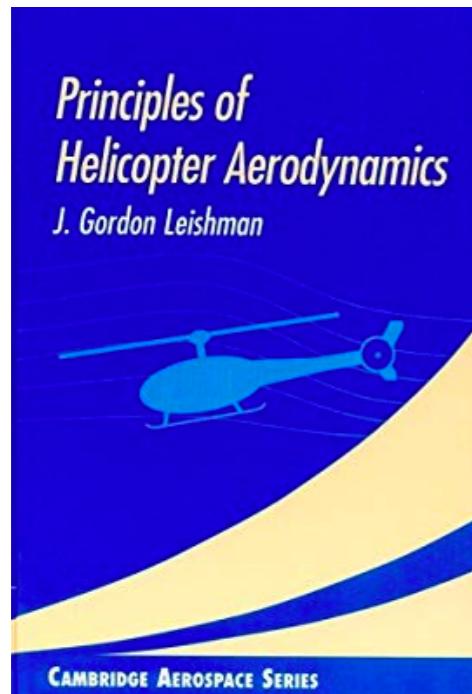
Don't violate constraints (e.g. crash)

# How do we get a model of the helicopter?



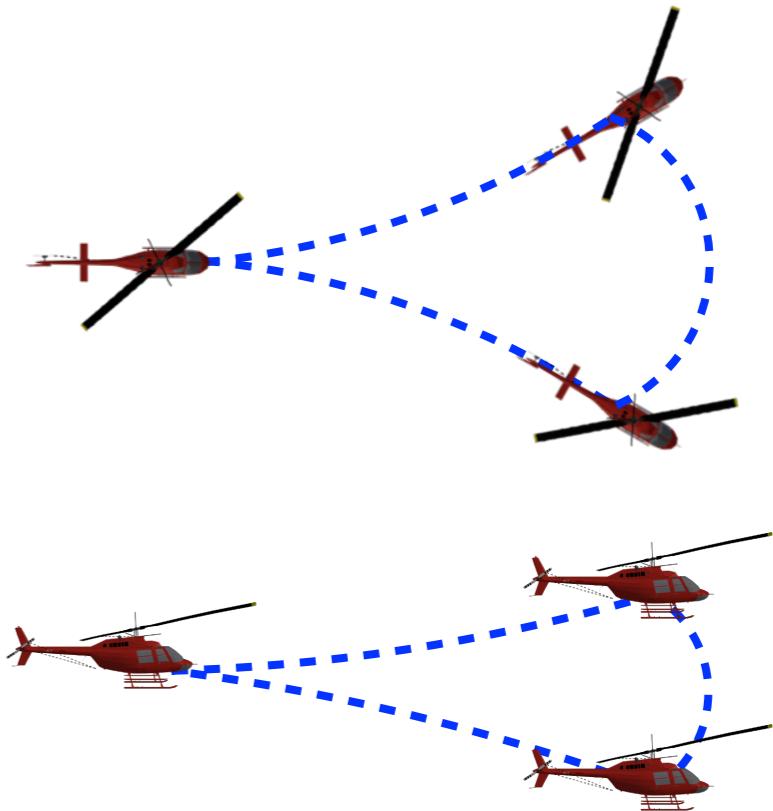
**Problem:** Model is very complicated! Intractable to plan with it.

# Lesson 1: Plan with simple models



Complex  
aerodynamical  
model

Use domain  
knowledge  
to simplify  
model



Flying unicycle at high speeds!

Different models at  
different flight regimes

Helicopter  
Models

World  
Model

?

PLAN

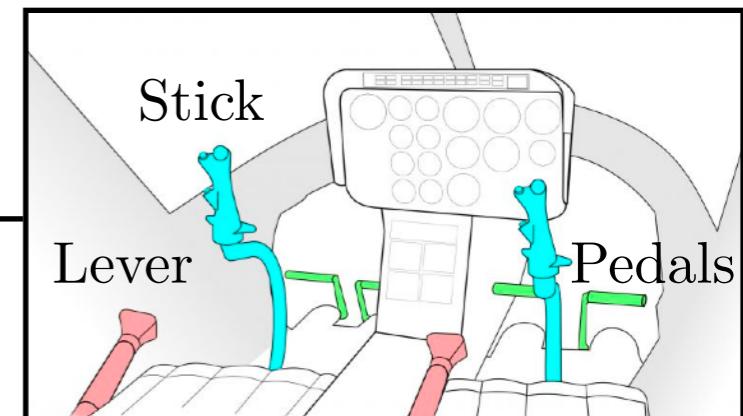
Sensors



Robot interacts  
with  
environment



Actuators



# What resolution should we plan?



Example mission:

Fly from Phoenix to Flagstaff  
as fast as possible (200 km)

Avoid mountains, no-fly-zones,  
radio towers, wires, bad weather

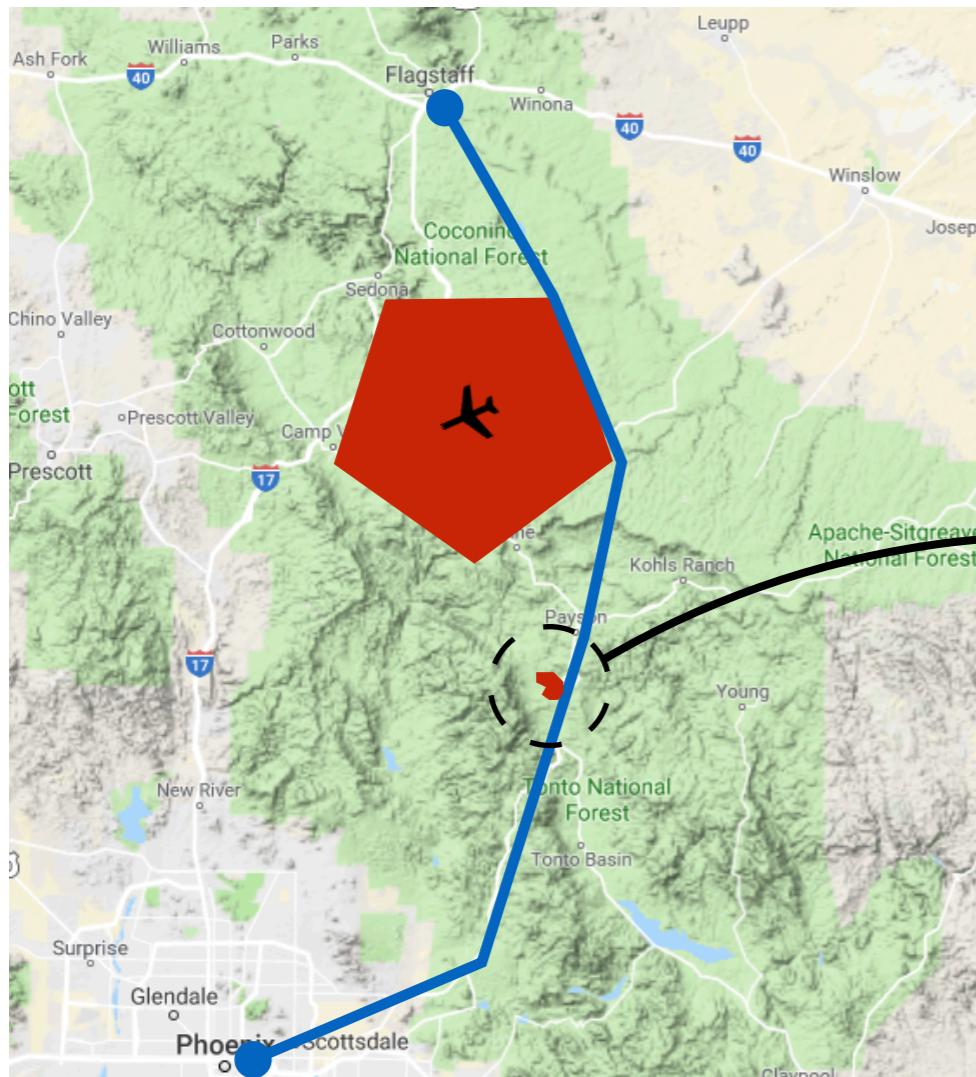
Pass through narrow gaps

Problem:

Take forever to plan at high  
resolution ALL the way to goal

# Lesson 2: Plan at multiple resolutions

Global planner

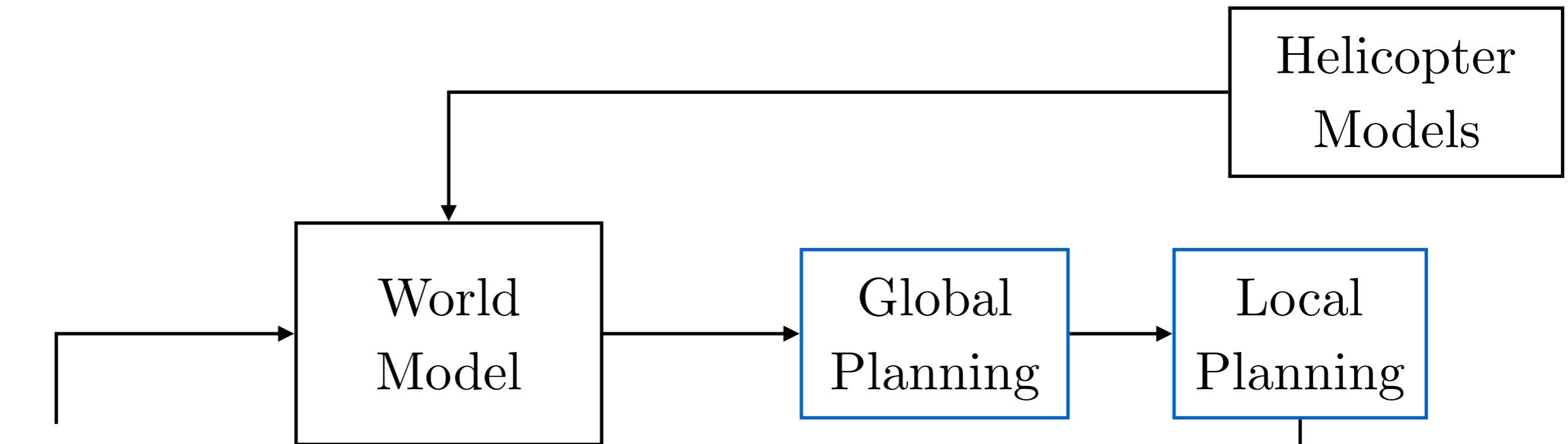


Plan at coarse (1km) resolution,  
compute entire route from start to goal  
avoid large obstacles, no-fly-zones etc  
  
(only consider factors that  
significantly affect mission time)

Local planner



Plan at high (10 m) resolution,  
follow the global route,  
avoid all obstacles, produce smooth  
dynamically feasible paths



?

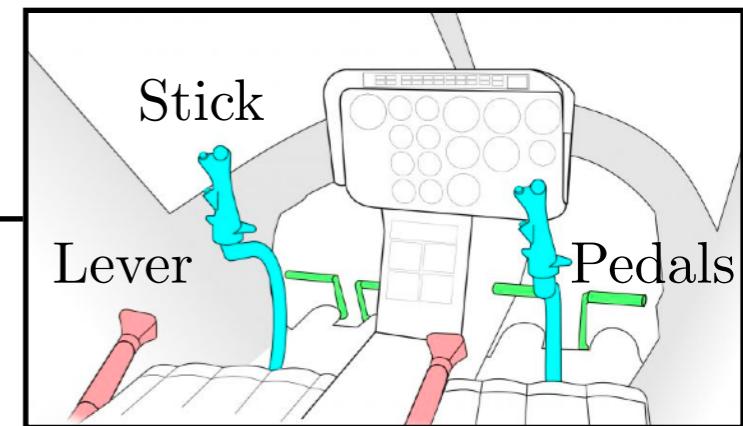
Sensors



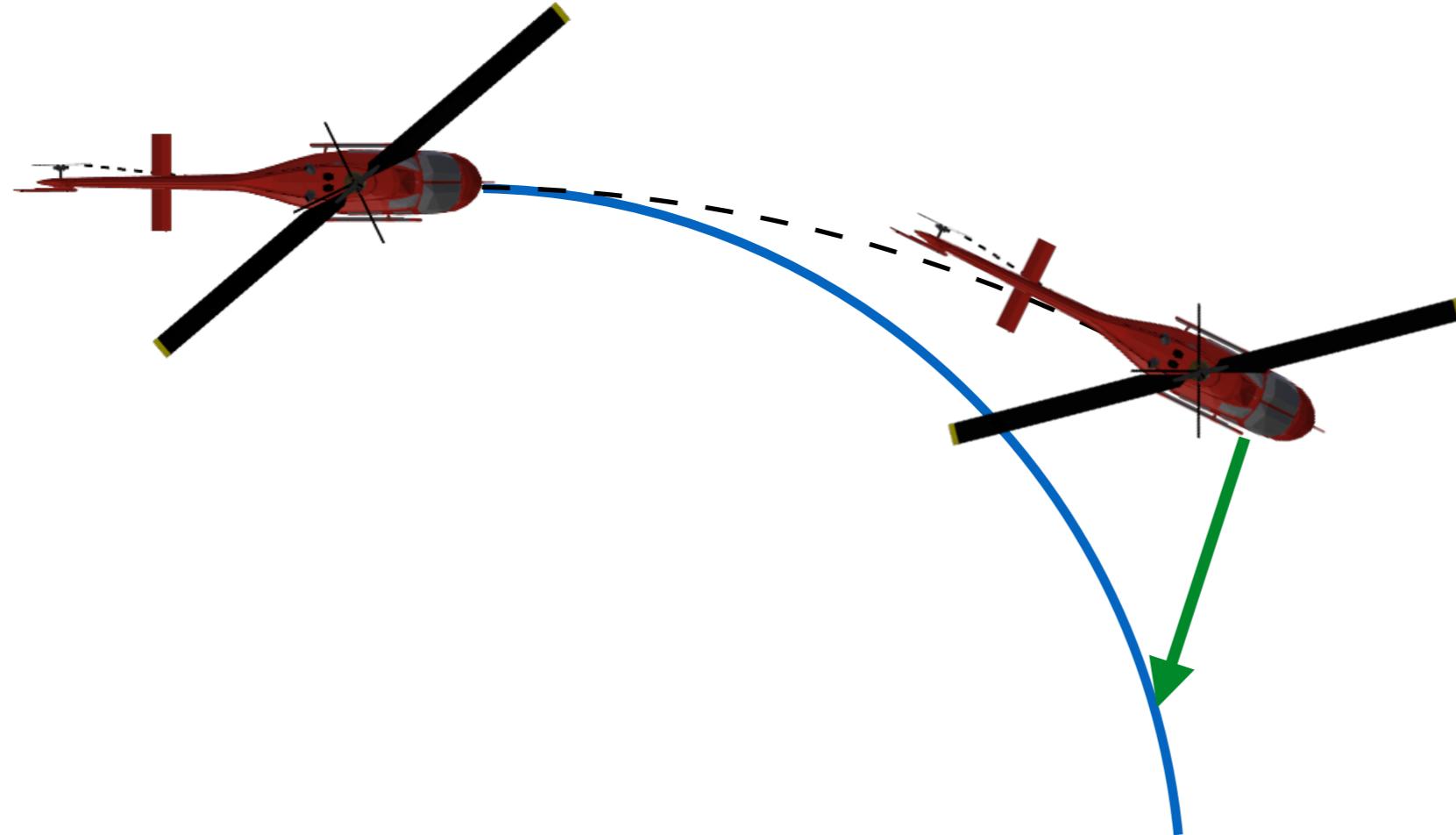
Robot interacts  
with  
environment



Actuators

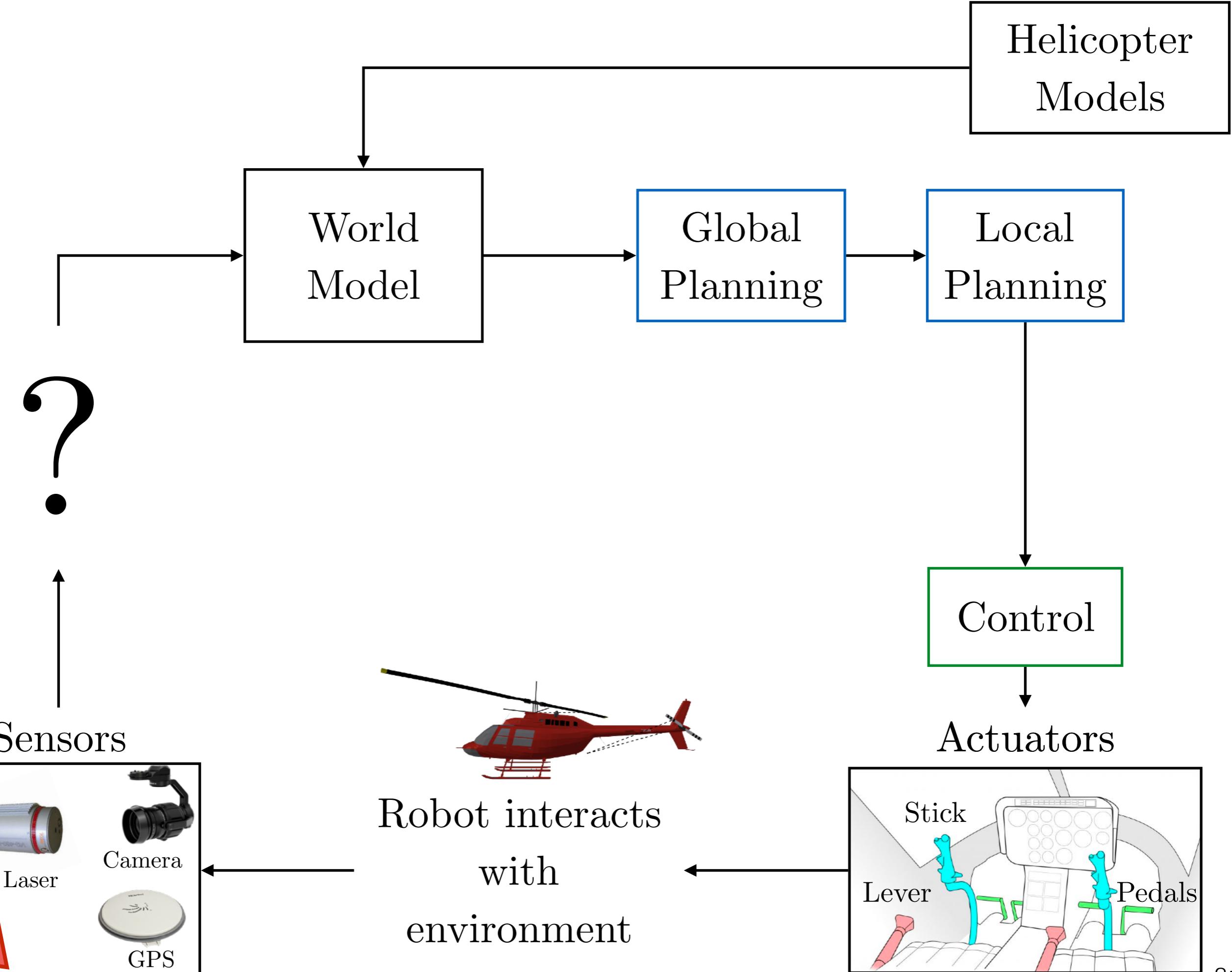


# Lesson 3: Open loop planning is not enough



Robot will go “off” the plan for many reasons  
(disturbance, model errors, actuation errors, ...)

A controller immediately corrects for any tracking error  
and gets the robot back on the path



# Look at one piece at a time

Q1: Assume we know everything about the world.

What commands should we send to the actuator?

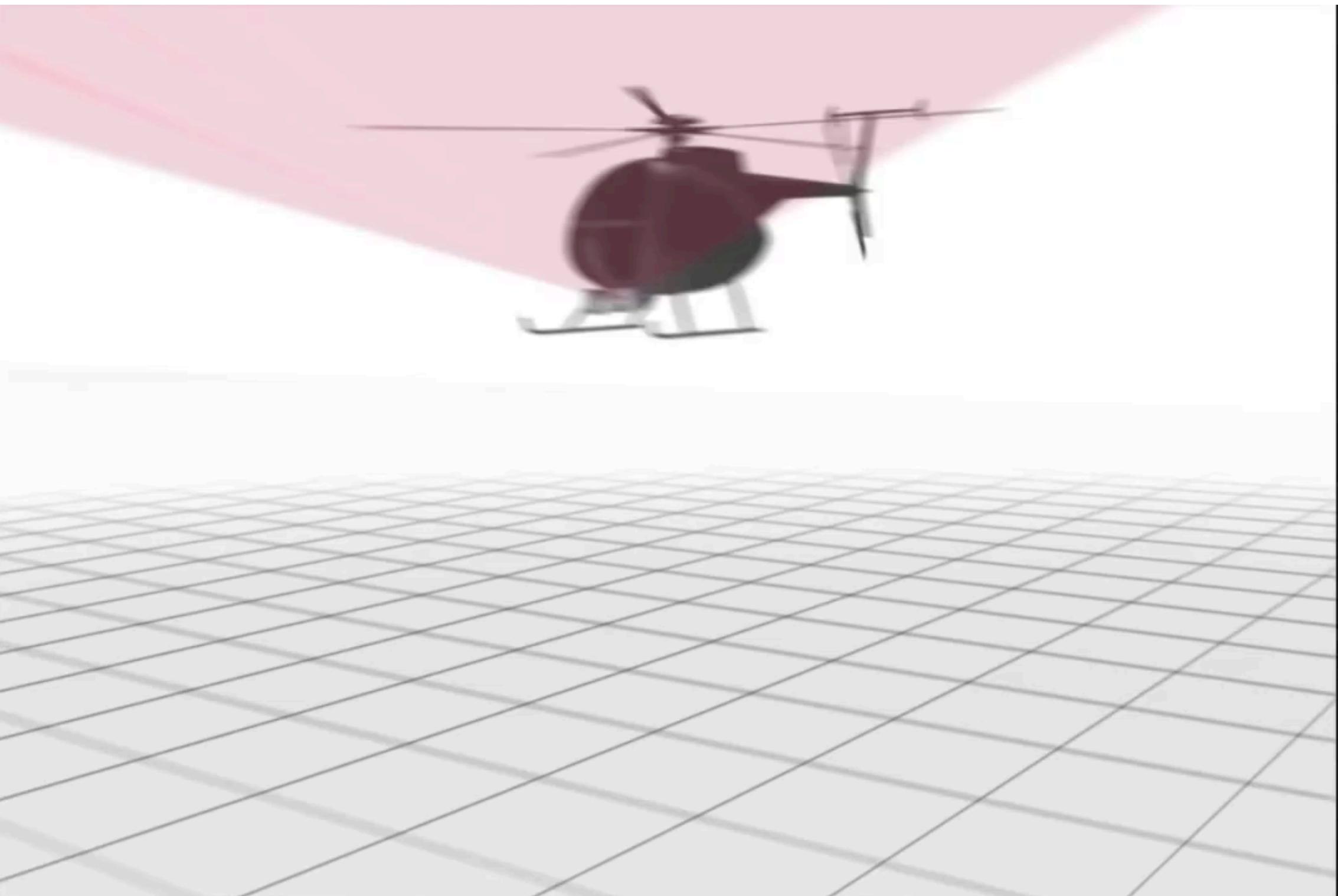
Q2: How do we use raw sensor data to update what we know about the world?

# What is the world model?

List of everything we need to know to accomplish the task

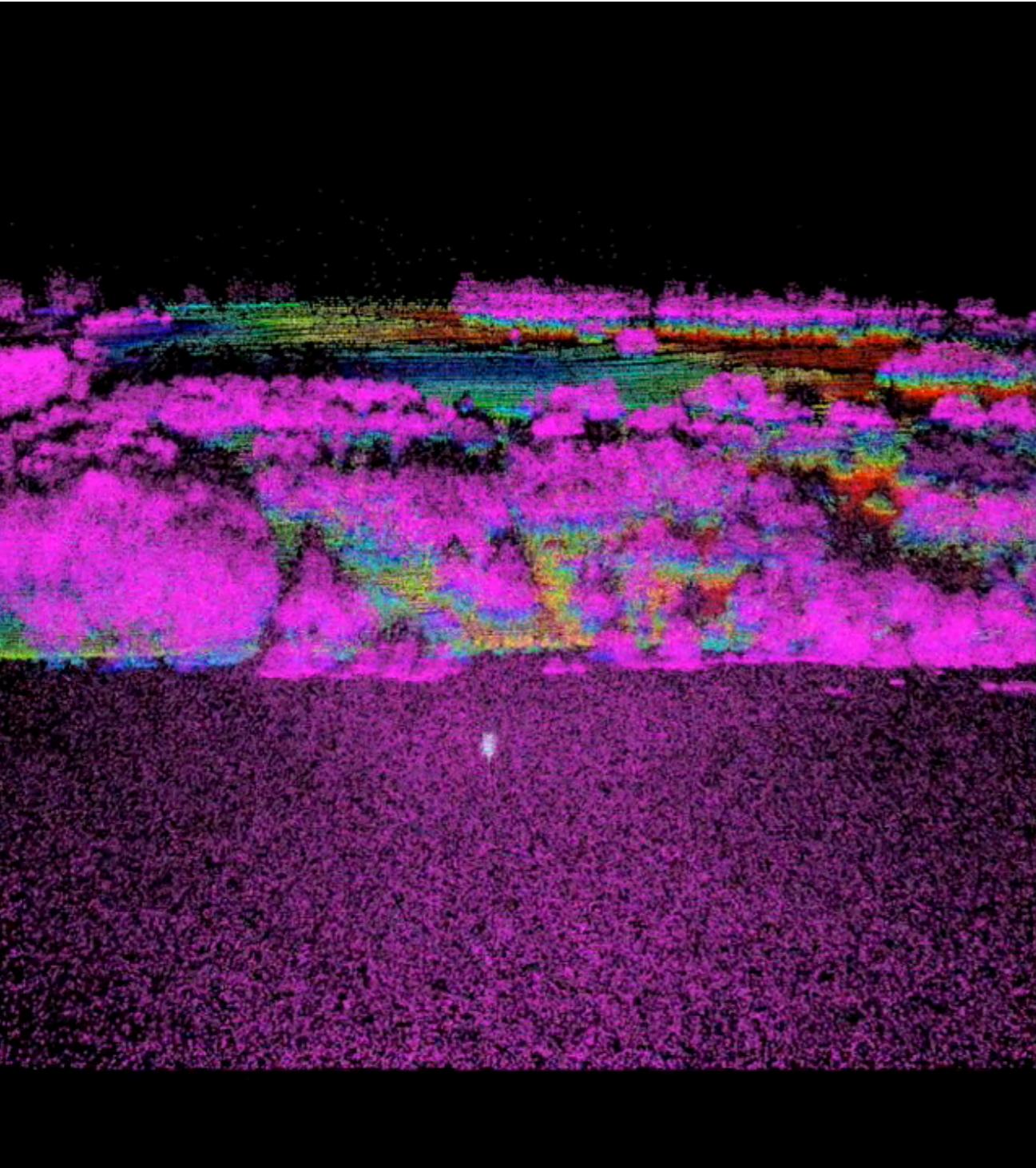
- Where is the robot in the world? What is it's state? GPS
- What are the obstacles in the world? Laser
- What type are the obstacles (radio towers, trees)? Camera
- What are the no-fly-zones? Radio
- Are there other aircrafts? Radio
- What is the wind, temperature, etc? Pitot tube,  
barometer,  
etc

Can we simply “fuse” laser readings to map the world?



(courtesy Chamberlain et al.)

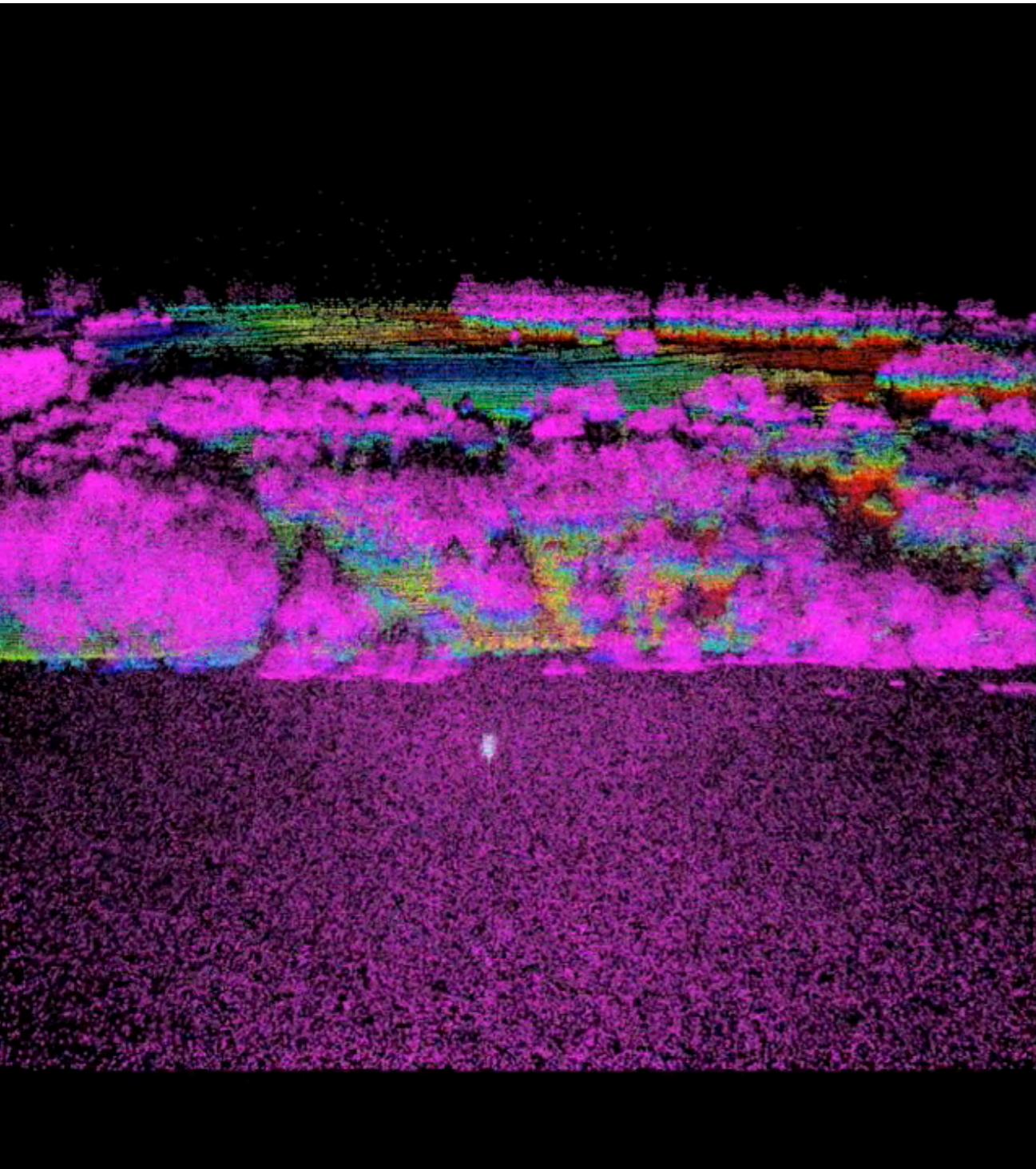
# Lesson 4: Use probabilistic models of the sensor



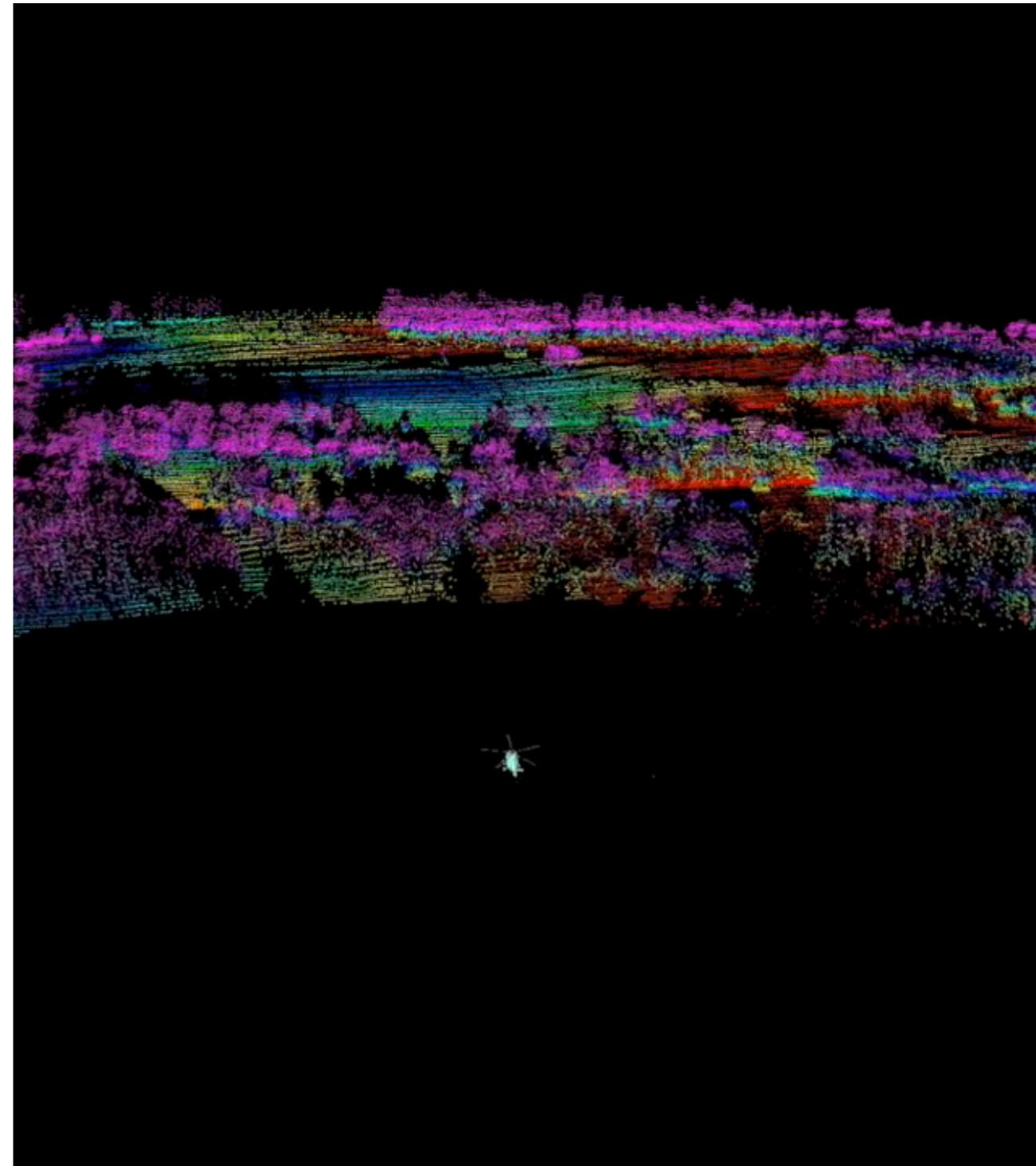
# Flying in a snow storm



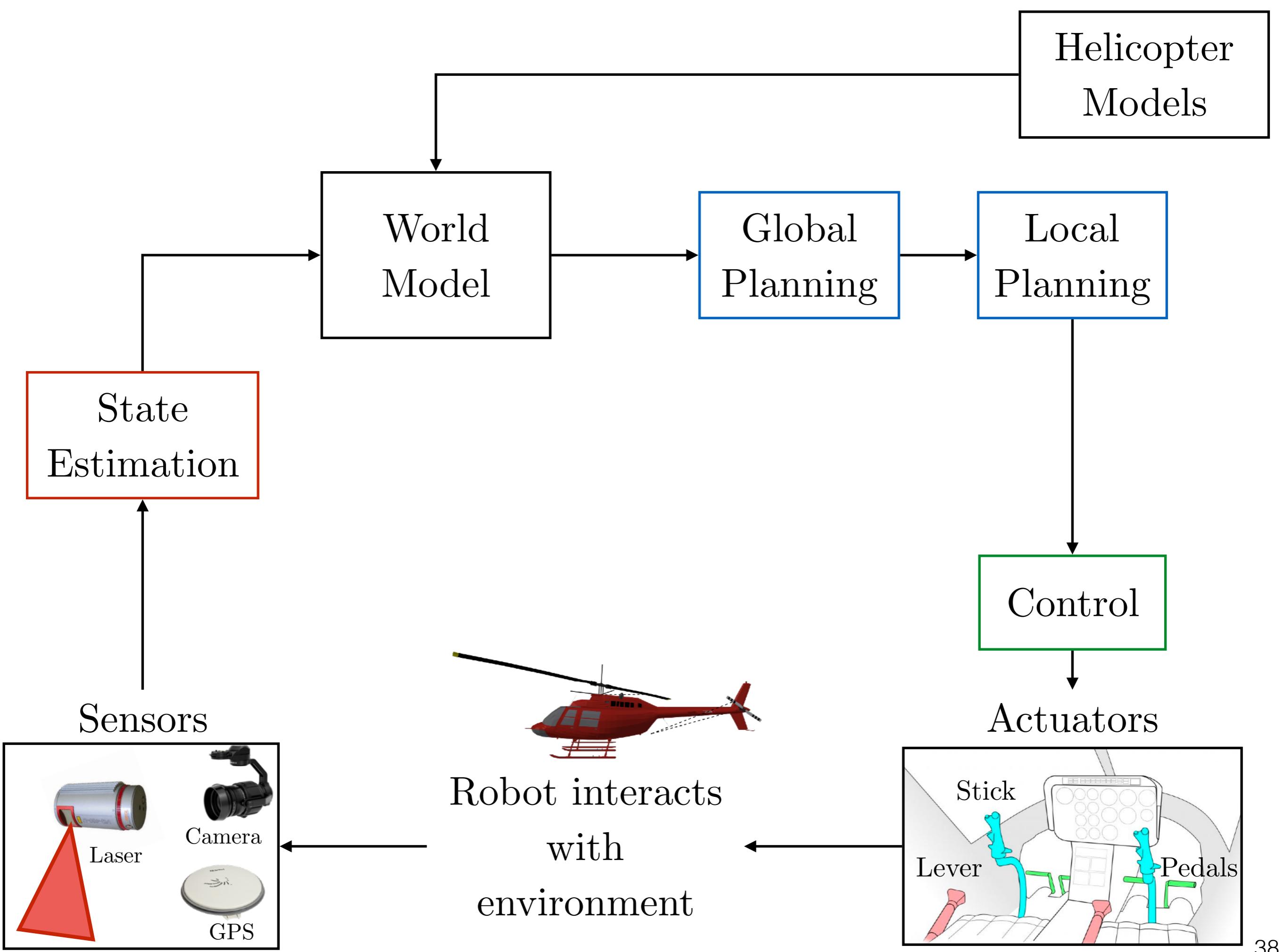
# Lesson 4: Use probabilistic models of the sensor



Laser reflected by snow!



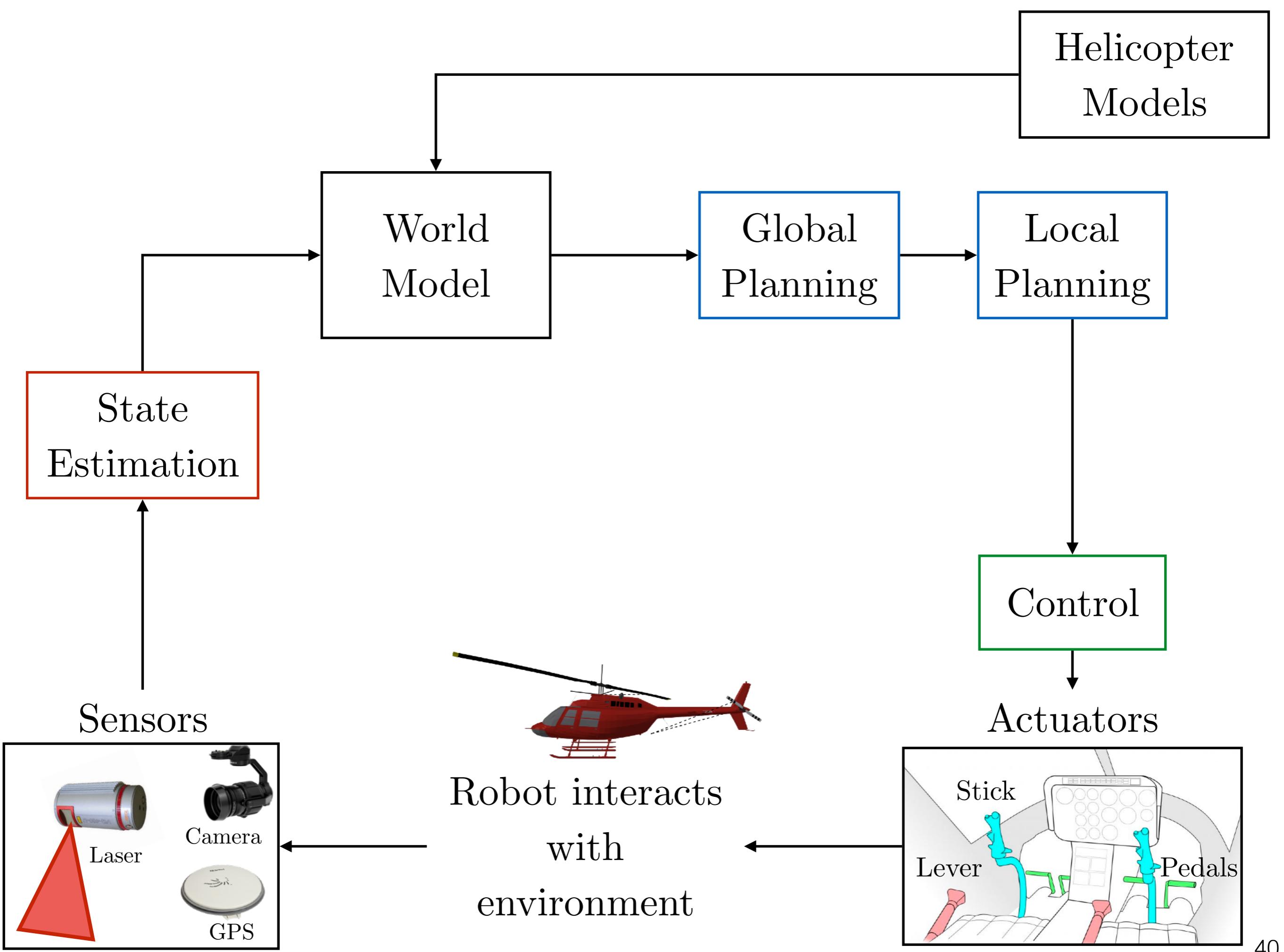
Correctly fused laser data  
using probabilistic models



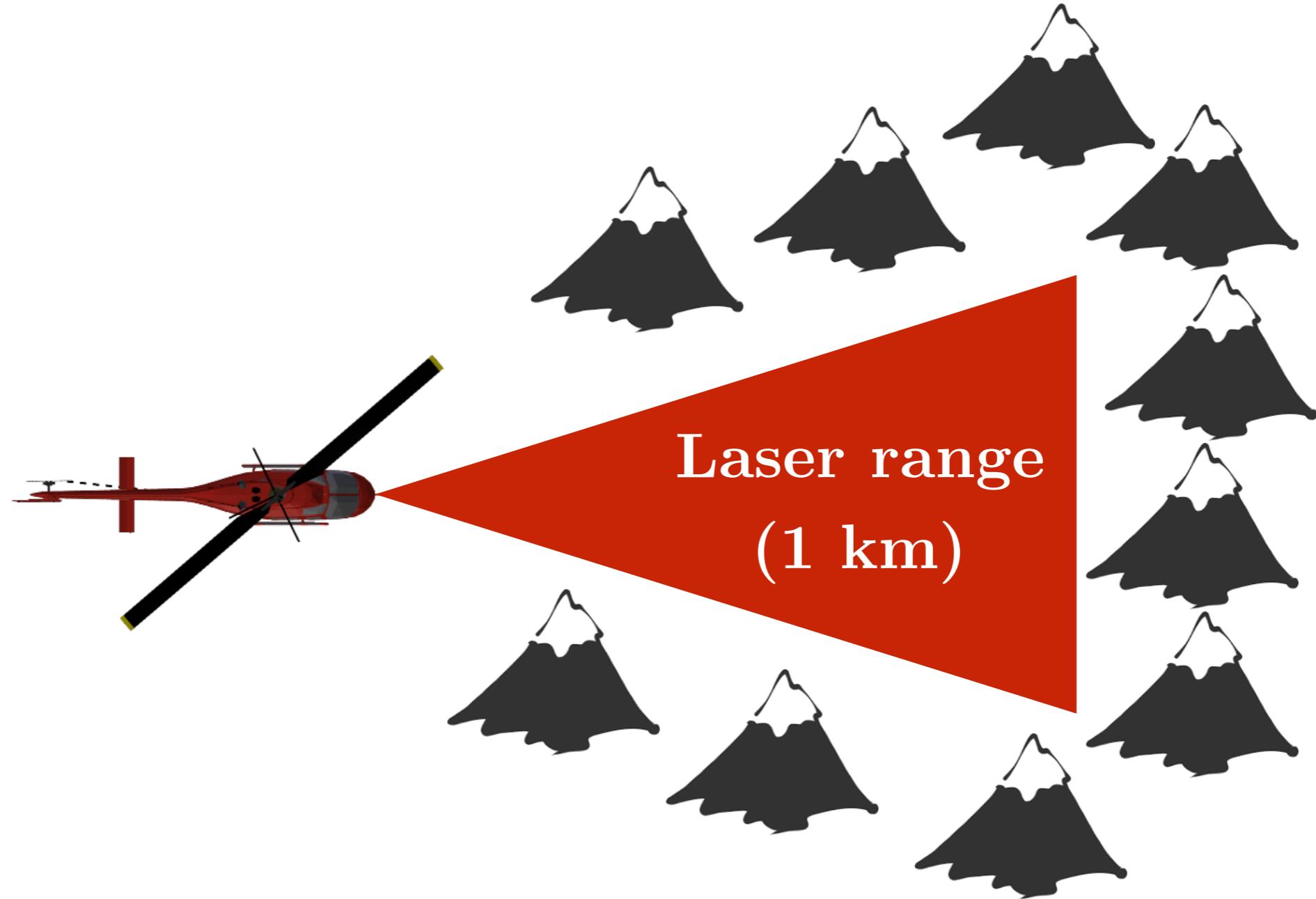
# What is state estimation?

Given raw sensor data, use probabilistic models to estimate world model

$$P(\text{world model} | \text{data})$$

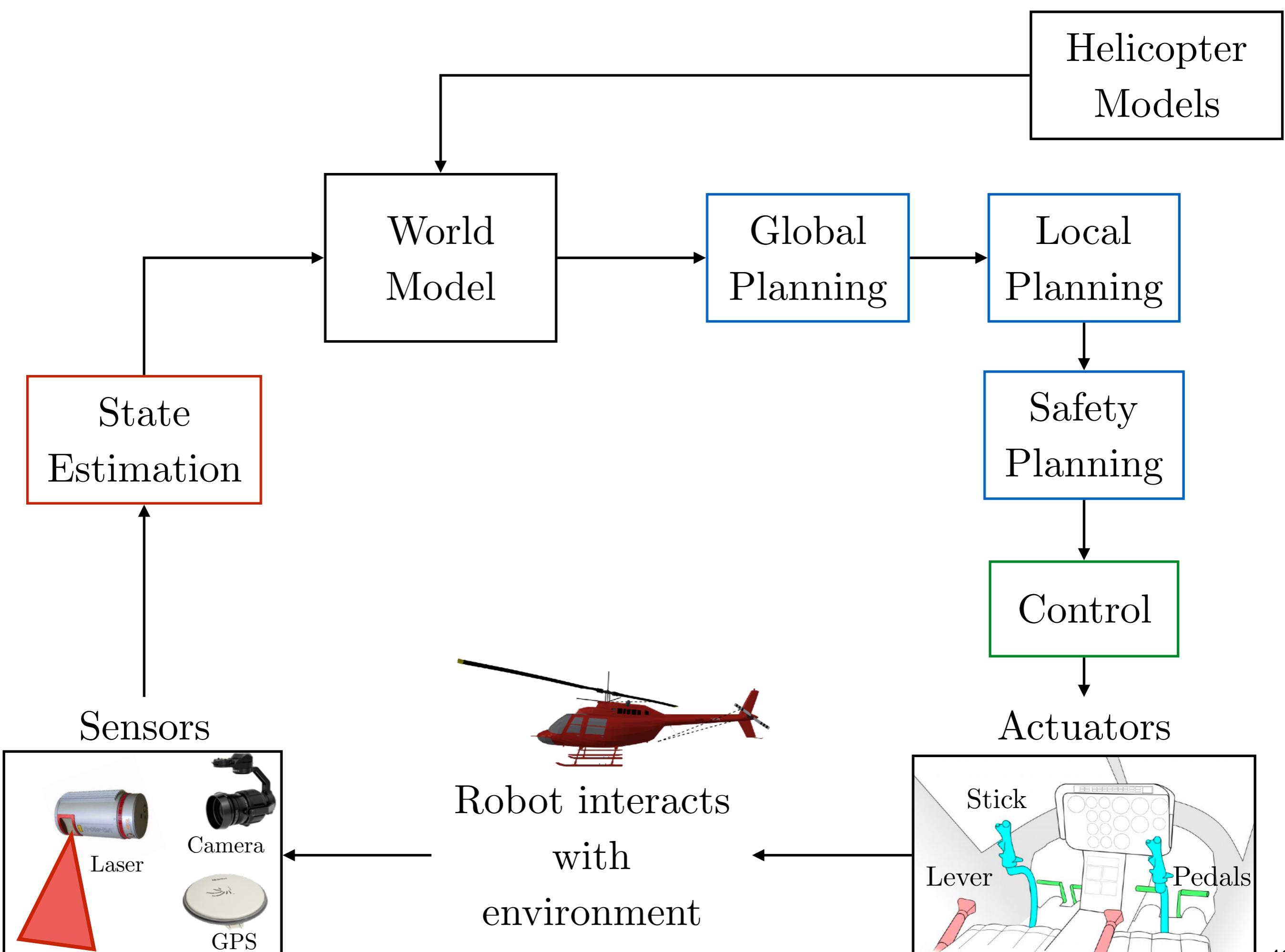


# Lesson 5: Guarantee safety



What prevents the system from flying at high speeds to a dead end?

Safety planner that **guarantees** the robot can stay safe



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1. Learn how to architect a mobile robotic system
  2. Step through a set of fundamental lessons that shape robot system / algorithm design

# BOSS: CMU's entry to urban challenge



# BOSS in action!

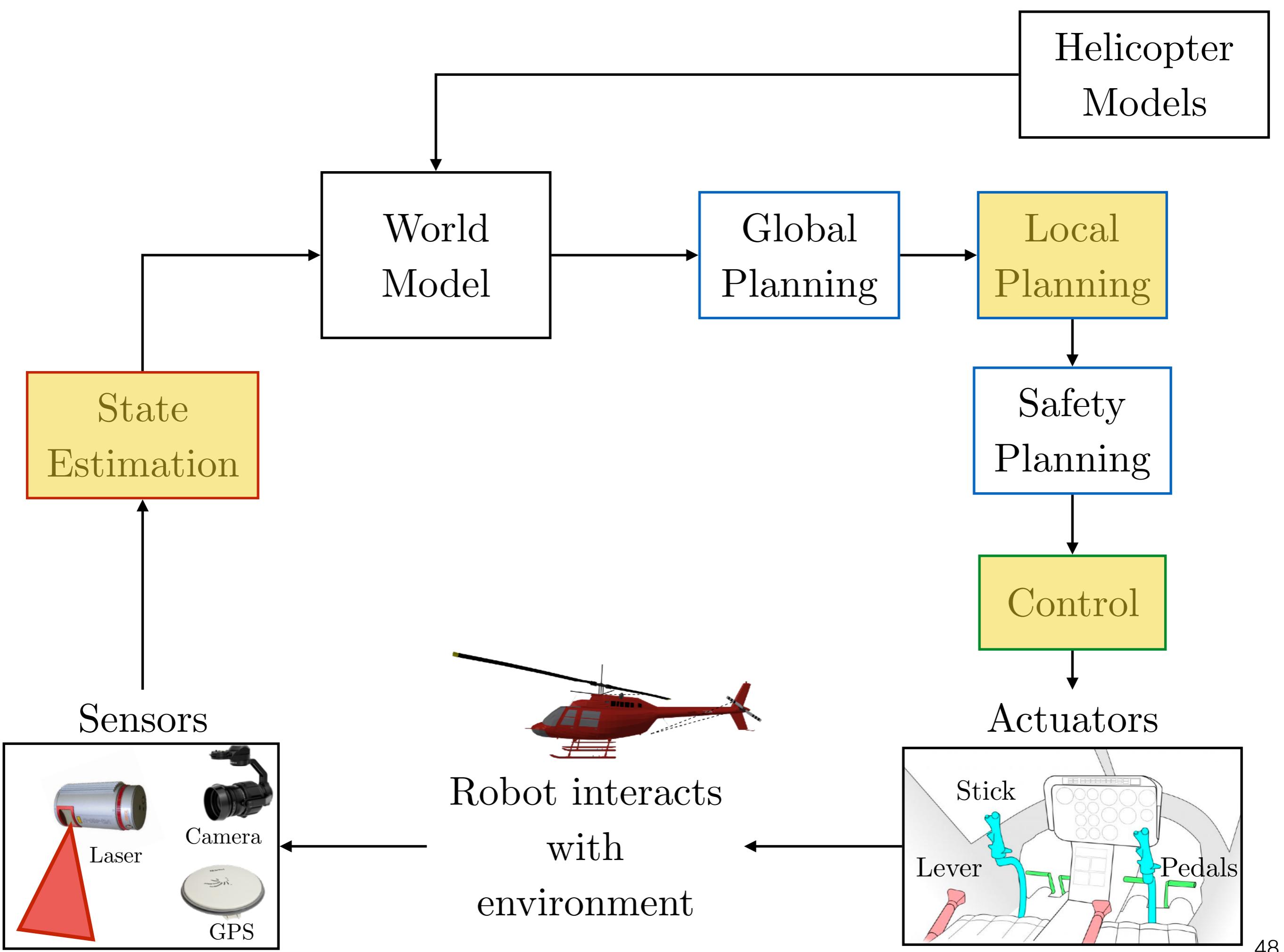


- |                     |  |                   |  |
|---------------------|--|-------------------|--|
| 1. World Model      |  | 4. Global Planner |  |
| 2. Car Model        |  | 5. Local Planner  |  |
| 3. State Estimation |  | 6. Safety Planner |  |
|                     |  | 7. Control        |  |

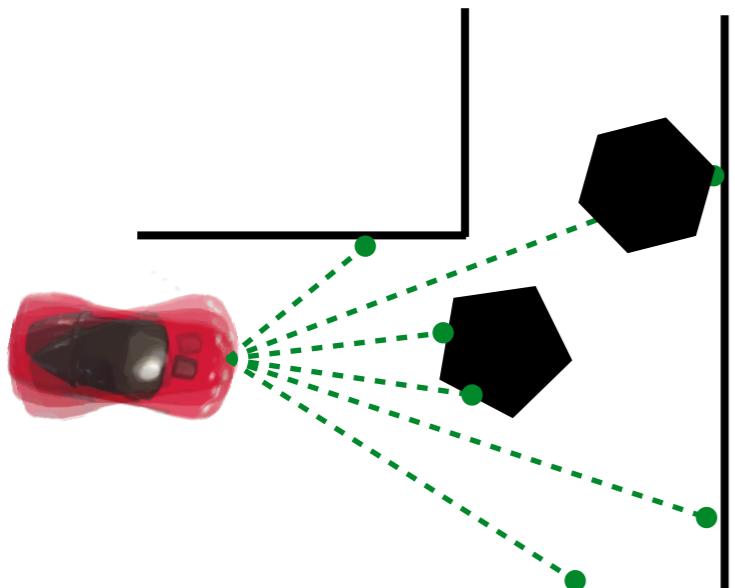
# Additional challenges: Predict human drivers



Looking ahead...

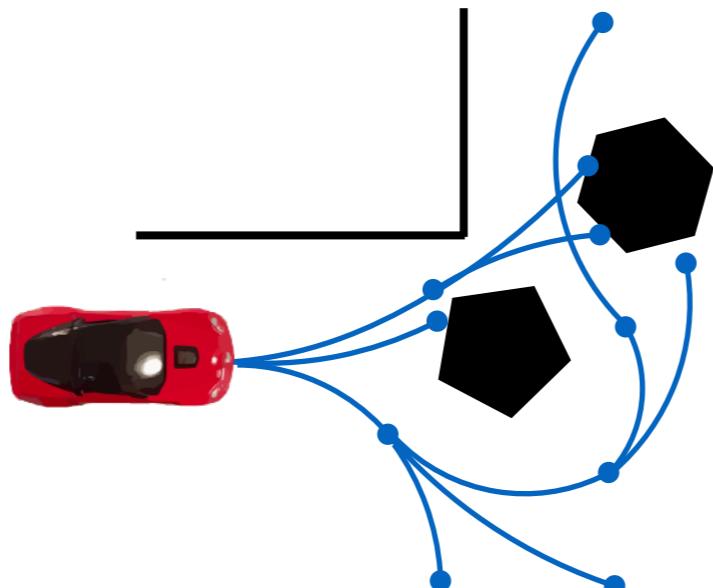


Estimate  
state



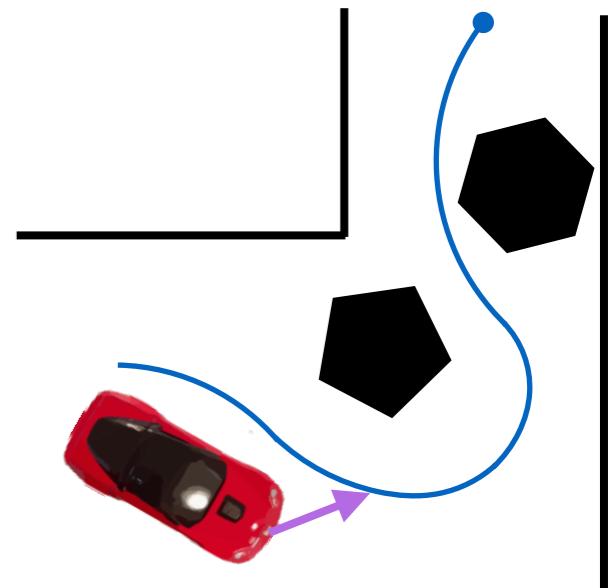
Weeks 2-4

Plan a  
sequence of  
motions



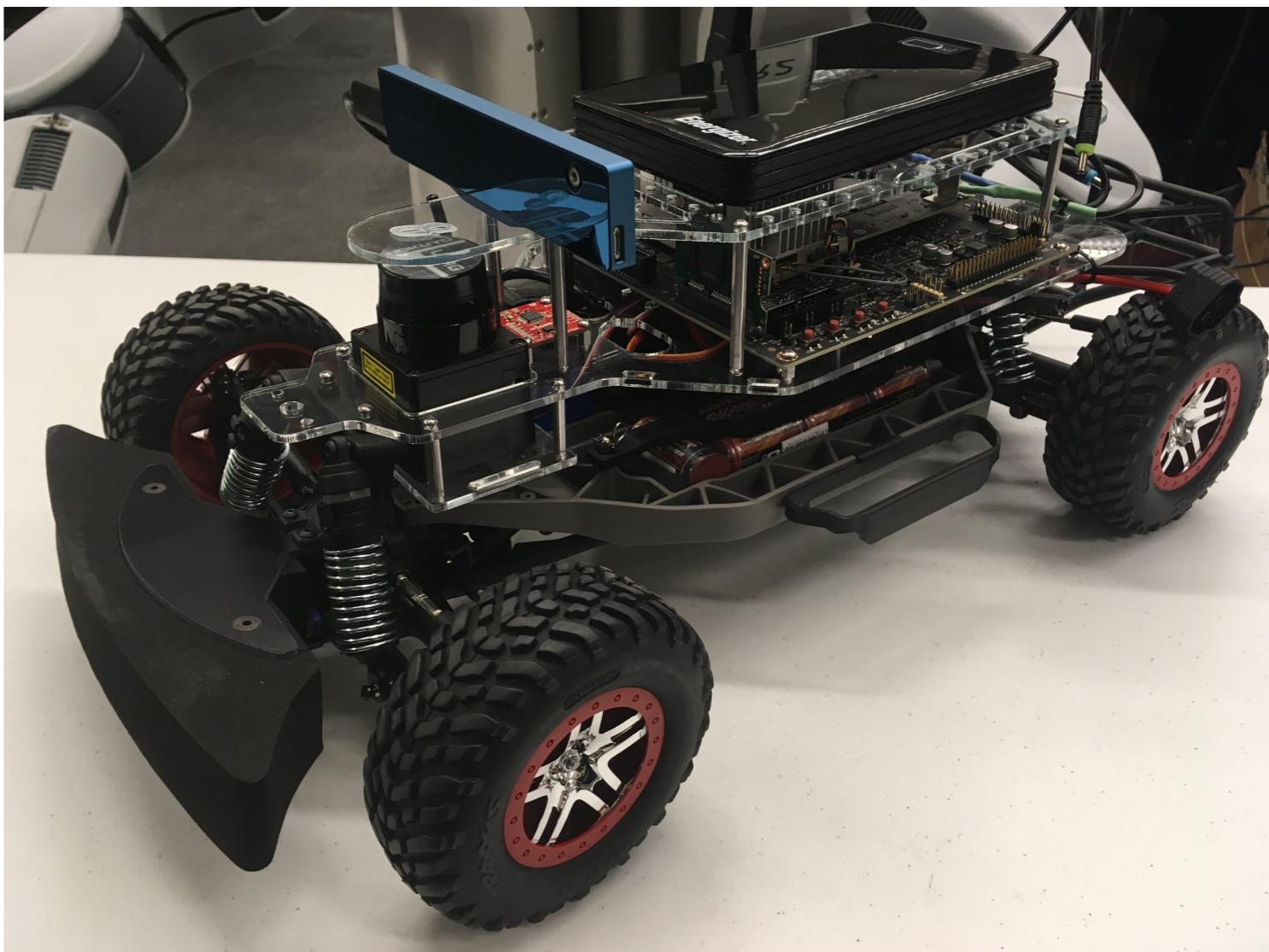
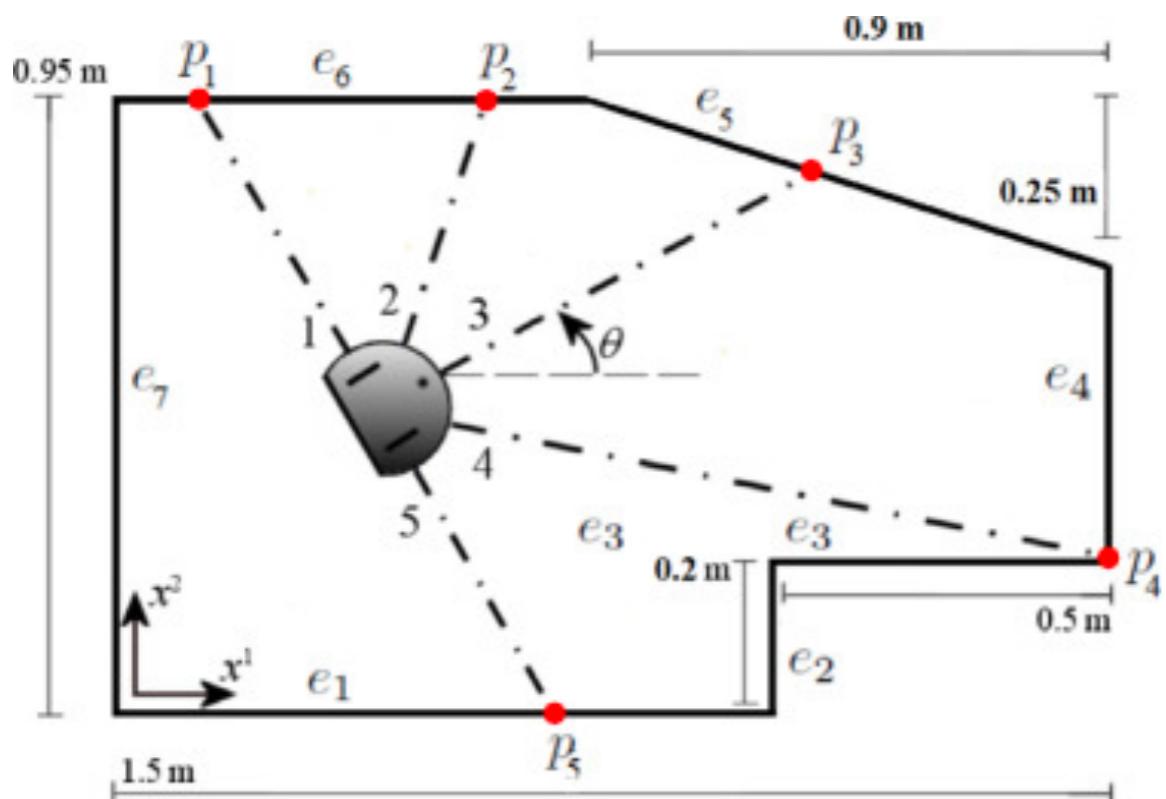
Weeks 7-8

Control  
robot to  
follow plan

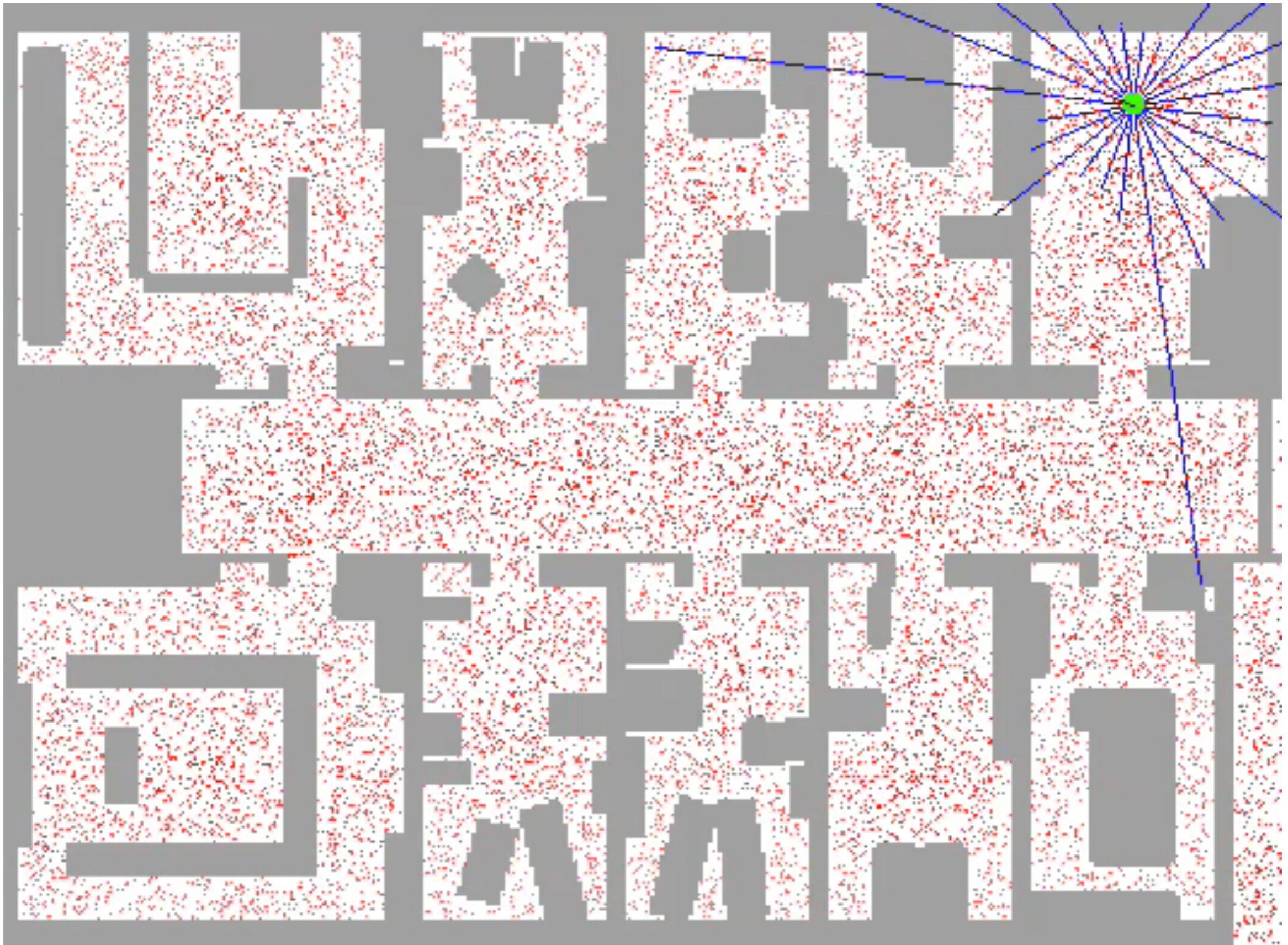


Weeks 5-6

# Challenge for indoor robots: Localization



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Mobile Robots

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