

# certified control: a new safety architecture

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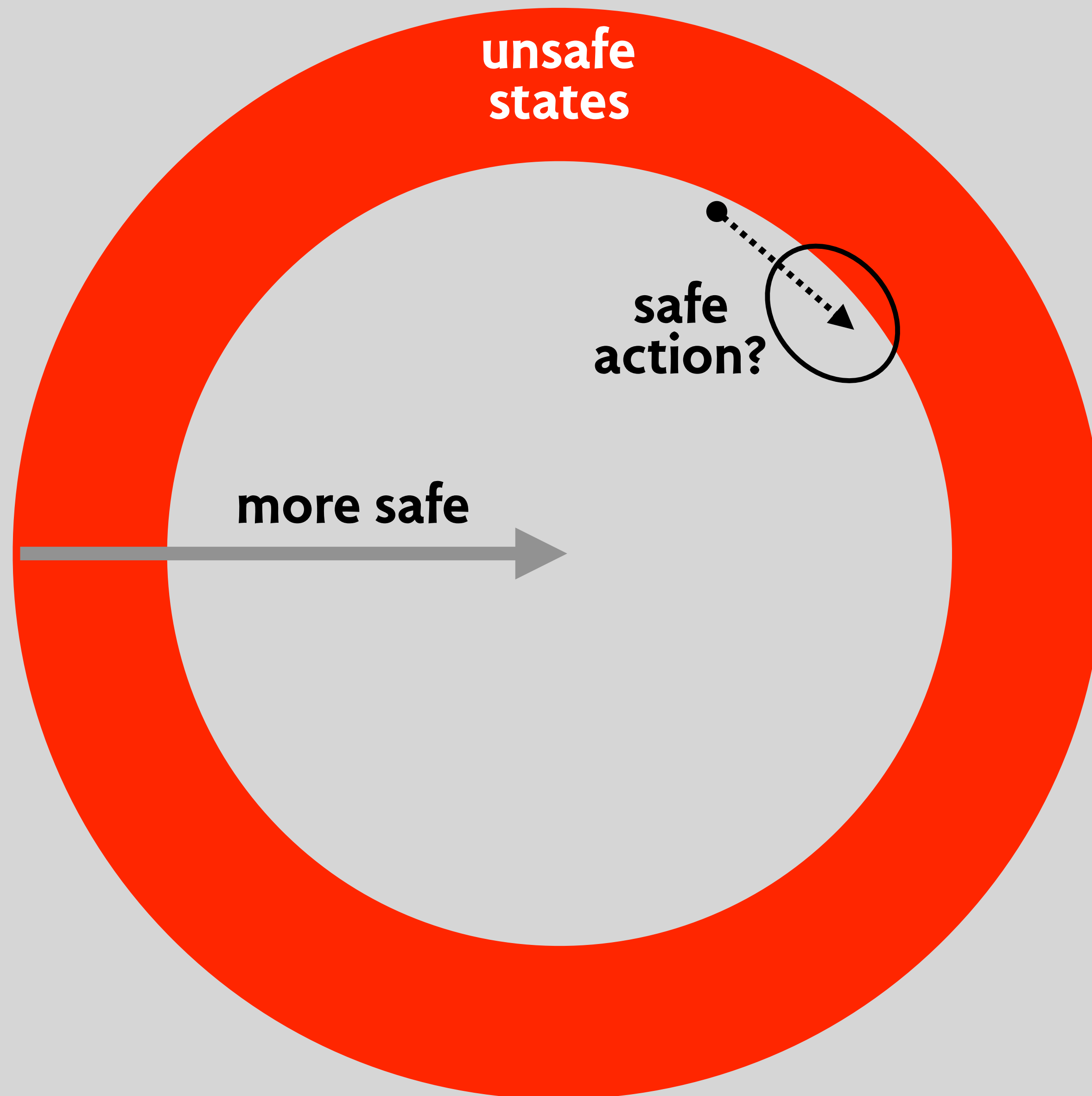
*TRI Liaison:*

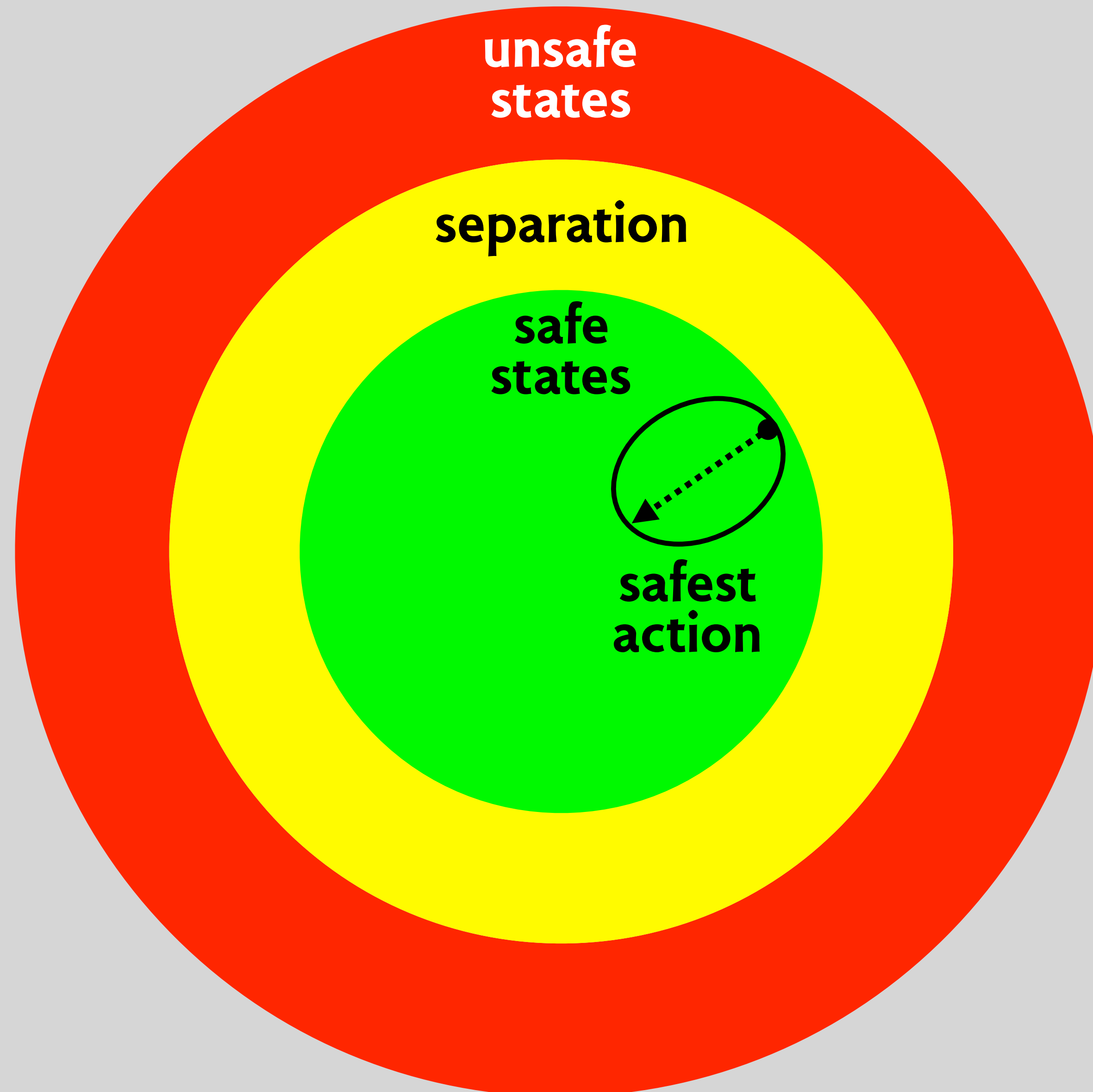
**Soonho Kong**

**Quarterly Review Meeting**  
Toyota-CSAIL Joint Research Center  
December 5, 2019



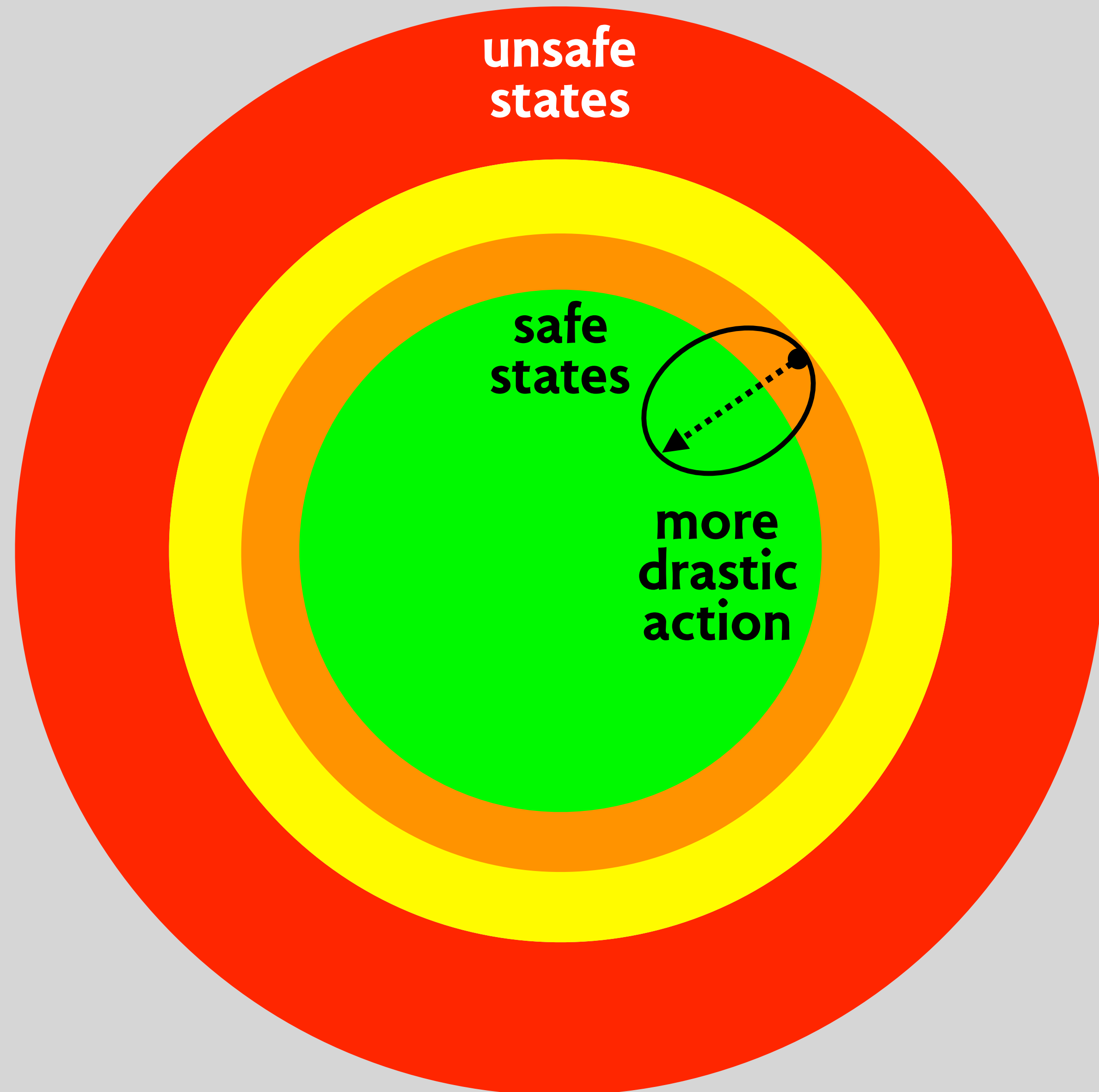
part 1: how to  
design a safe  
controller





*but what if  
mechanism is too  
complex to verify?*

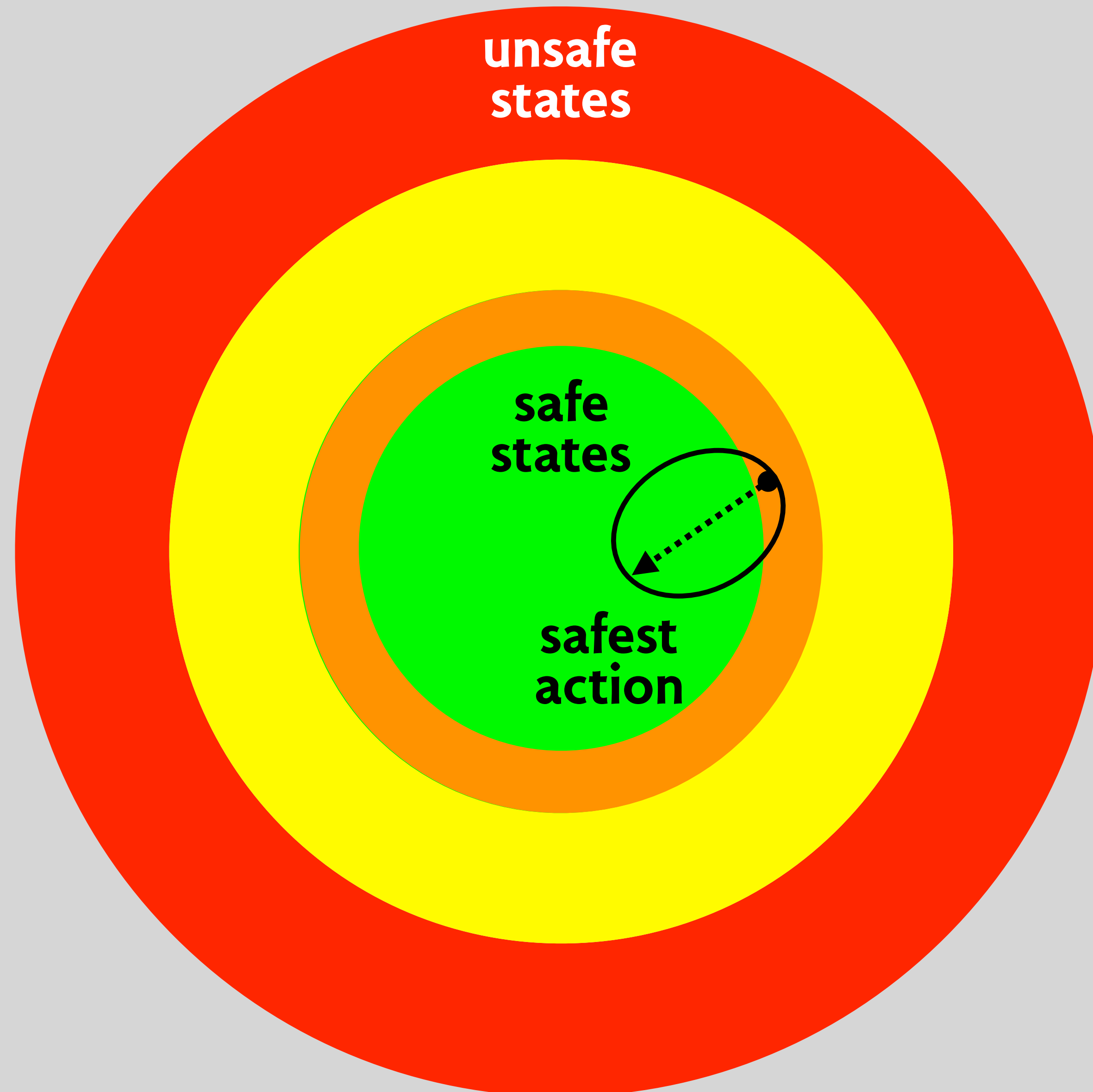
part 2: classic  
interlocks



**unsafe  
states**

**safe  
states**

**more  
drastic  
action**



*interlock only needs to intervene, so it can be verifiable*



*option 1:  
interlock does  
filtering too*

*then can't verify  
interlock either*

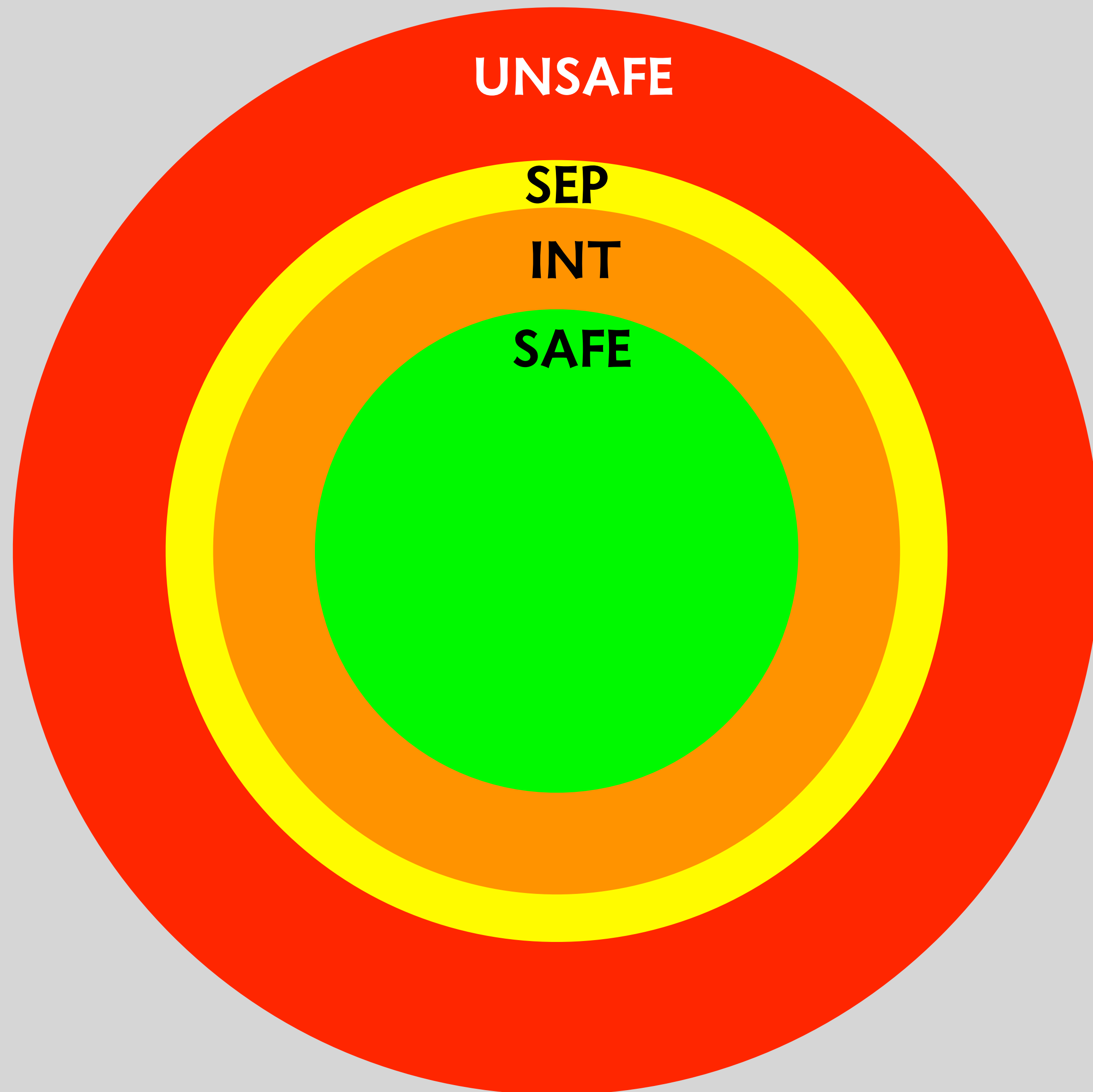
*option 2:  
interlock does  
something simpler*

*then interlock will  
intervene by  
design*

**suppose state perception depends on LiDAR**  
in snow, controller does complex (ie, error-prone) filtering of snow



# 3 interlock properties: pick 2



## **sound**

intervene only on failure

$$\text{SAFE} \cap \text{INT} = \emptyset$$

## **complete**

interlock prevents accident

$$\text{UNSAFE} \cap \text{INT} = \emptyset$$

and interlock can maintain INT

*AEB drops*

## **robust**

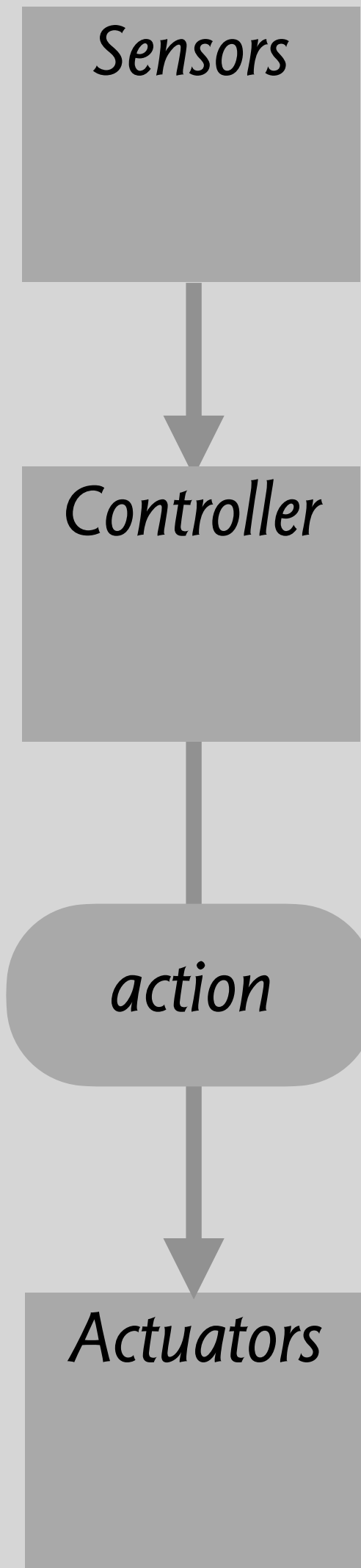
interlock check is simple

$s \in \text{INT}$  is verifiable

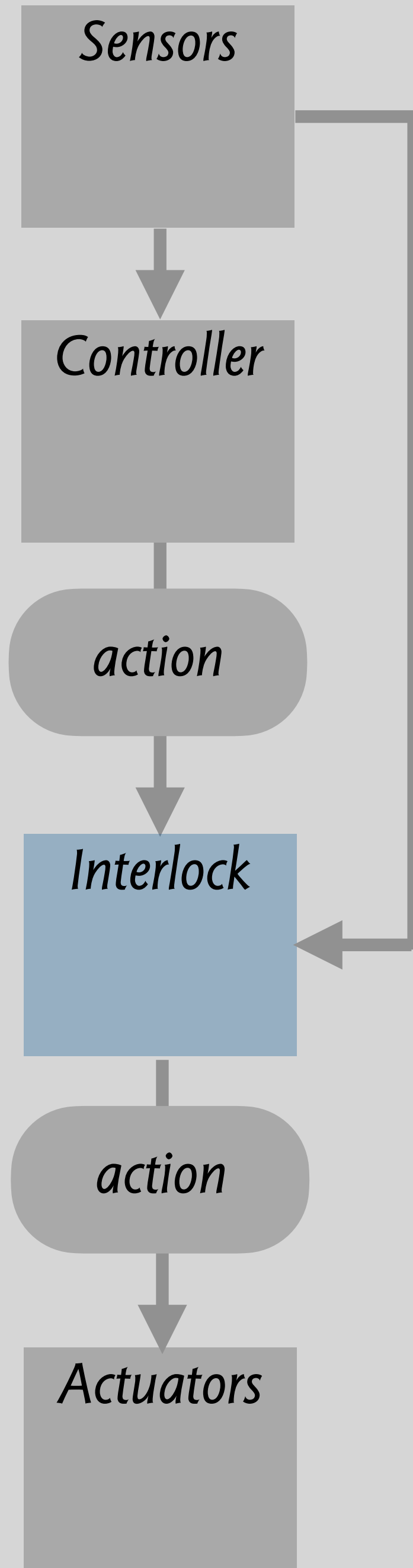
*RSS drops*

part 3:  
certified  
control

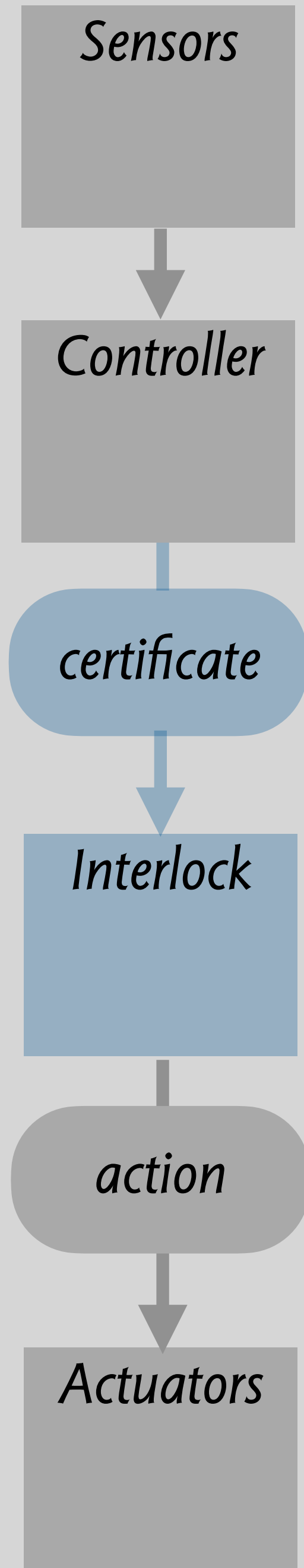
***no  
interlock***



***classic  
interlock***



***certified  
control***



# the essence of certified control

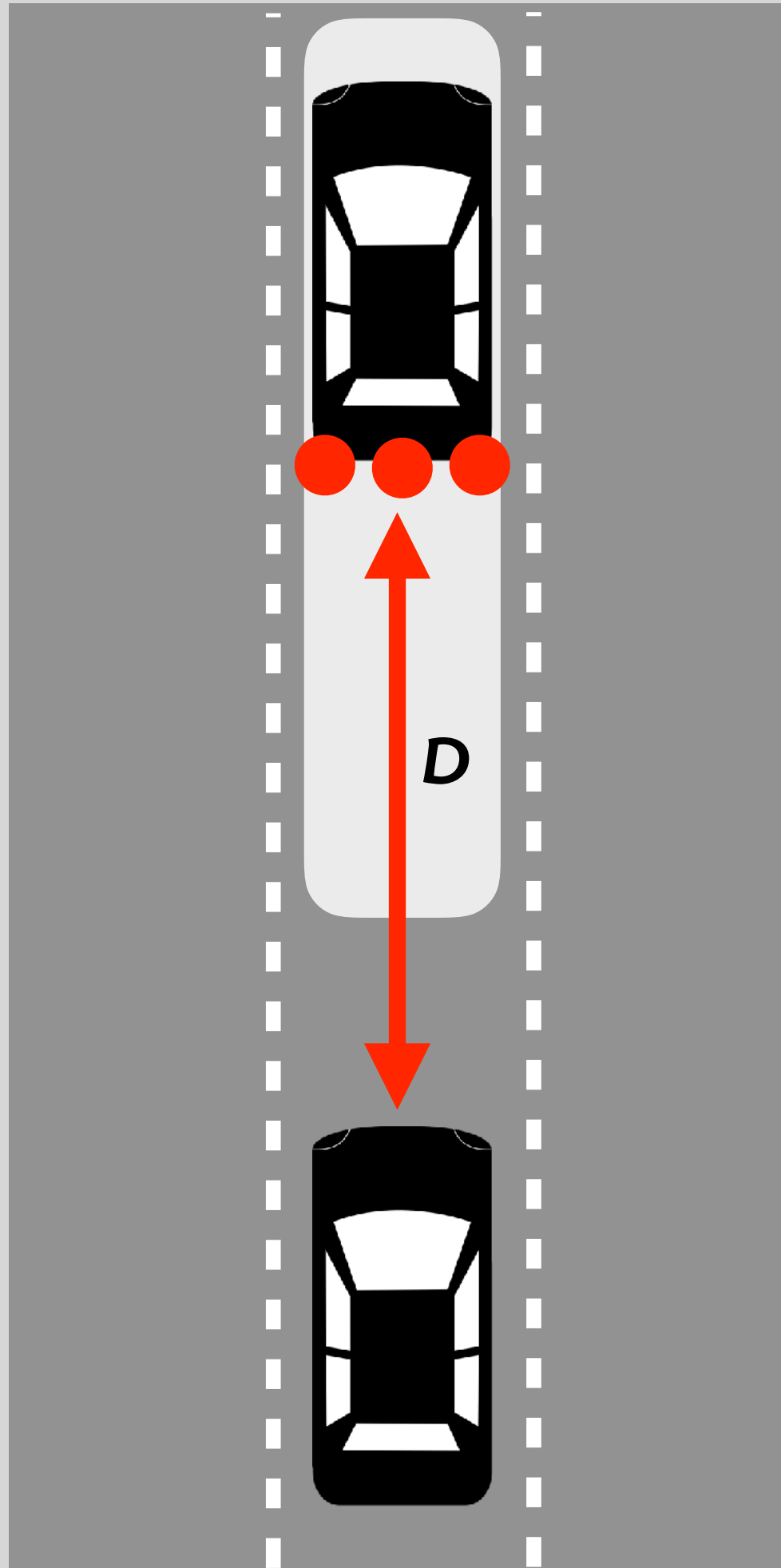
*controller can always generate certificate*  
 $\forall s \cdot s \in \text{SAFE} \Rightarrow \exists A, i \cdot \text{CERT}(s, i, A)$

*runtime  
dependability  
case*

*if certificate holds, then guarantees no crashes*  
 $\forall s: \text{SAFE}, i, A \cdot \text{CERT}(s, i, A) \Rightarrow \forall s' \cdot A(s, s') \Rightarrow s' \in \text{SAFE}$

*agreed upon at  
design time*

# example: certificate for continuing ahead



## elements of the certificate

action **A**: continue ahead without decreasing speed

state **s**: 3 LIDAR readings  $L[0..2]$  (signed by LIDAR unit)

ego car velocity  $V$  (signed by velocity unit)

interpretation **i**: a distance  $D$

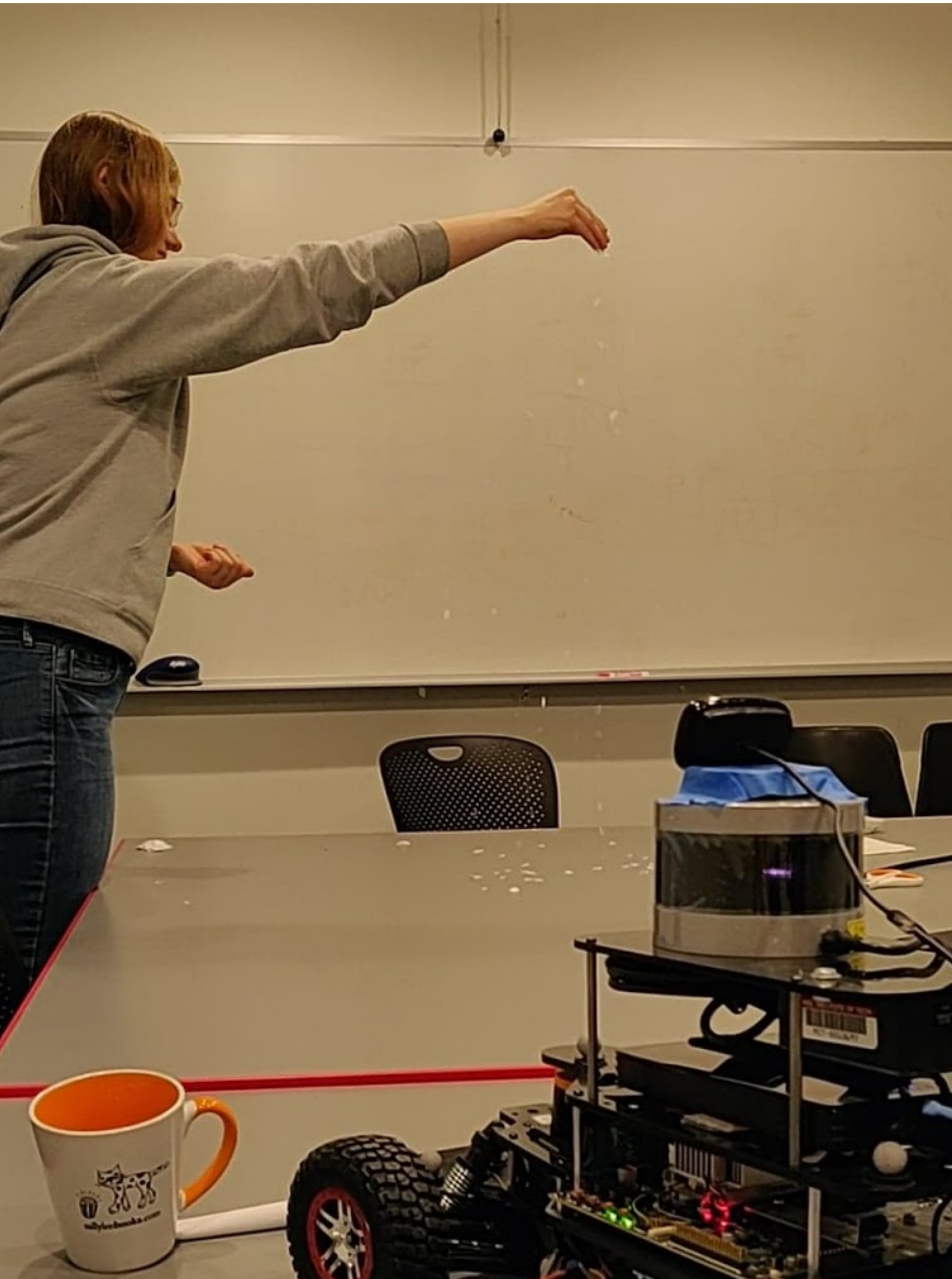
## checking $CERT(s,i,A)$

authenticates sensor data using public keys of sensors

checks  $L[0..2]$  lie on a straight line a distance  $D$  ahead

checks  $D >$  minimum separation at velocity  $V$

# a snow experiment



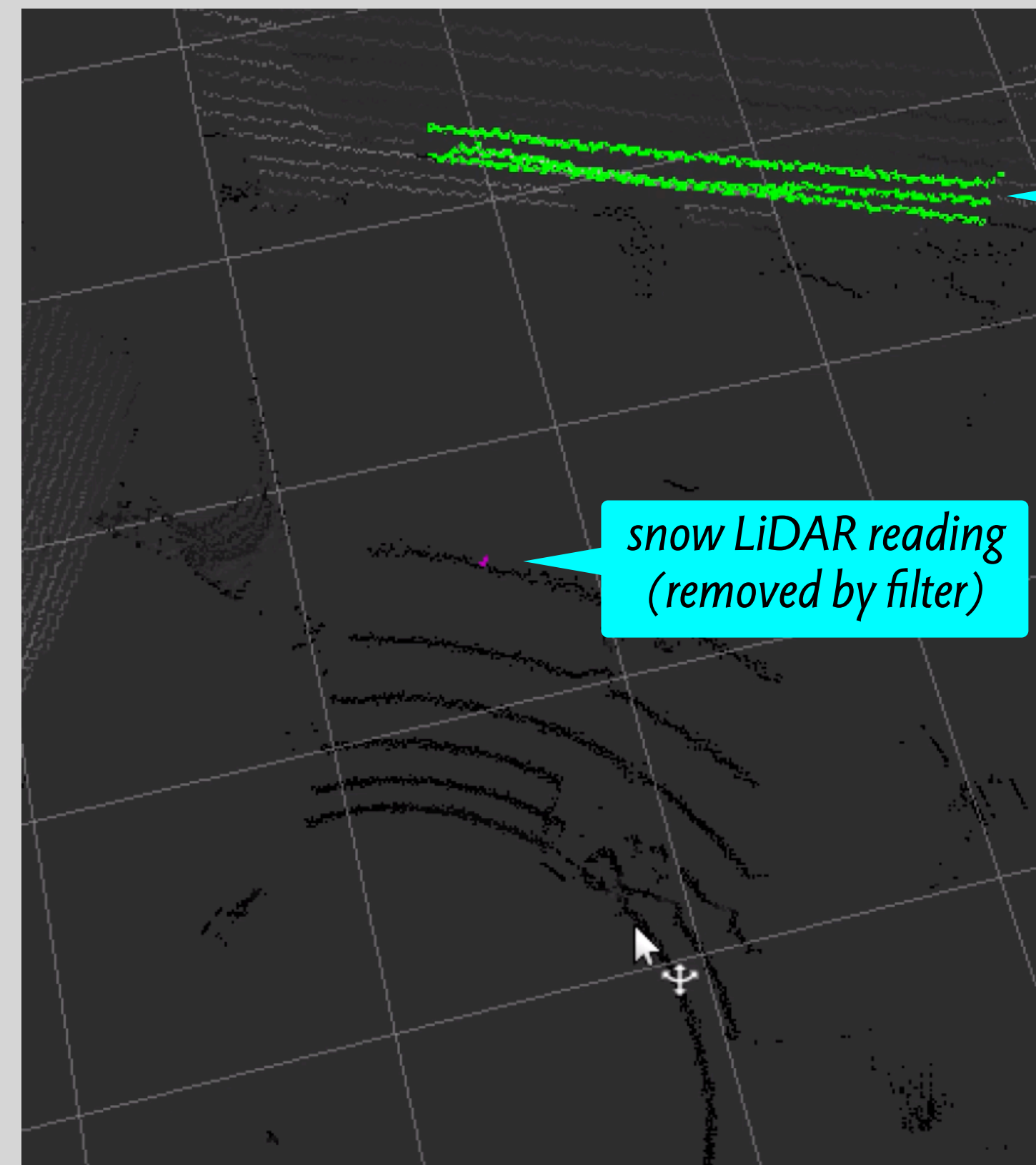
## Controller

- **Filter LiDAR points** using 3D outlier detection\* with K-d tree to remove snow
- **Generate certificate** of array of remaining LiDAR points at distance

## Interlock

**Check points** in certificate are sufficiently close together and cover lane

*\*De-noising of Lidar Point Clouds Corrupted by Snowfall. Nicholas Charron, Stephen Phillips and Steven L. Waslander. Fifteenth Conference on Computer and Robot Vision (CRV 2018)*

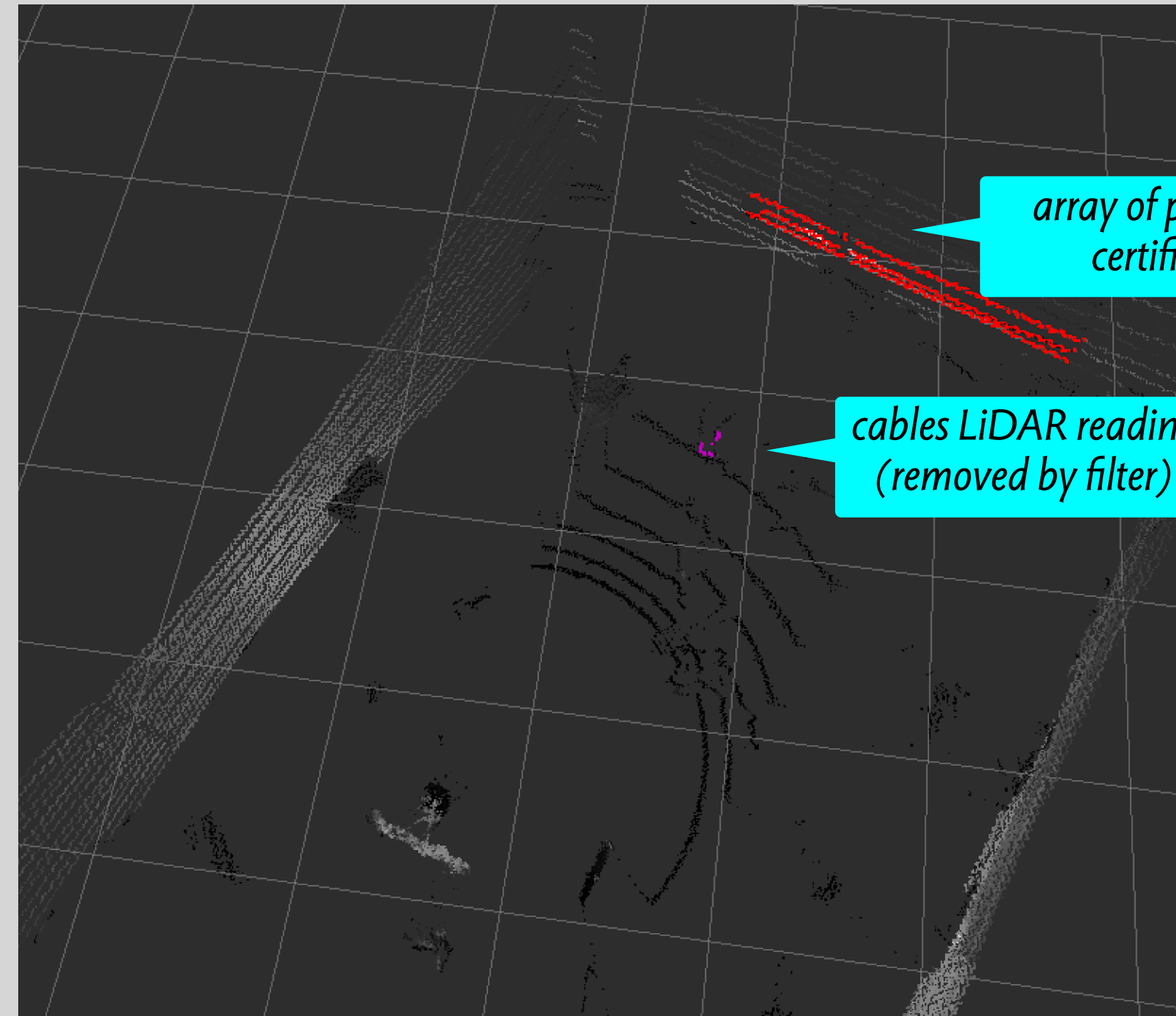


array of points in certificate

snow LiDAR reading (removed by filter)

✓ passes check

# what about other small objects?



 fails check

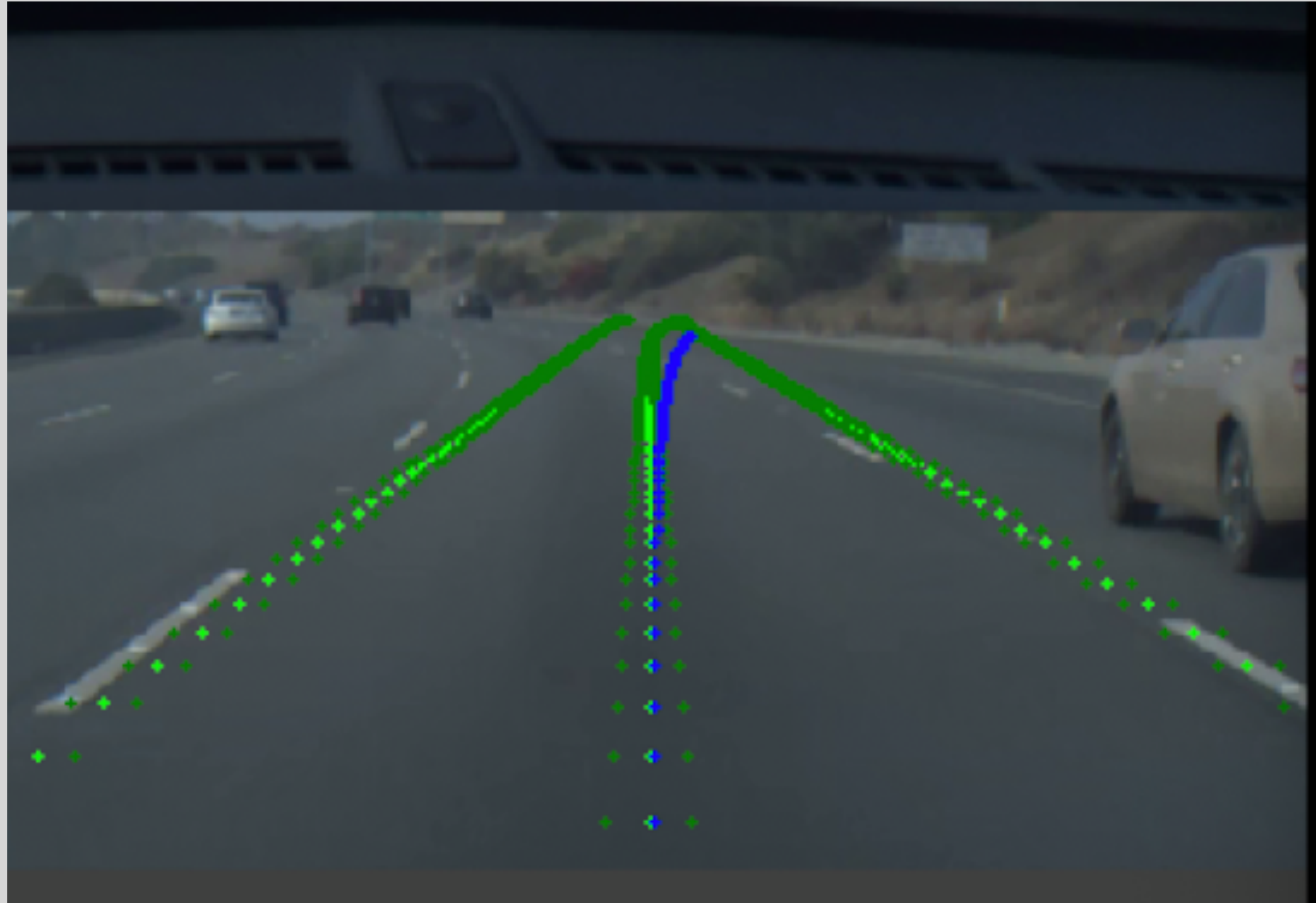


**certified control handles snow filtering for obstacle detection**  
but what about lane following? no pixels to pick like the LiDAR points



part 4:  
checking  
lane lines

# check #1: lane has the right geometry

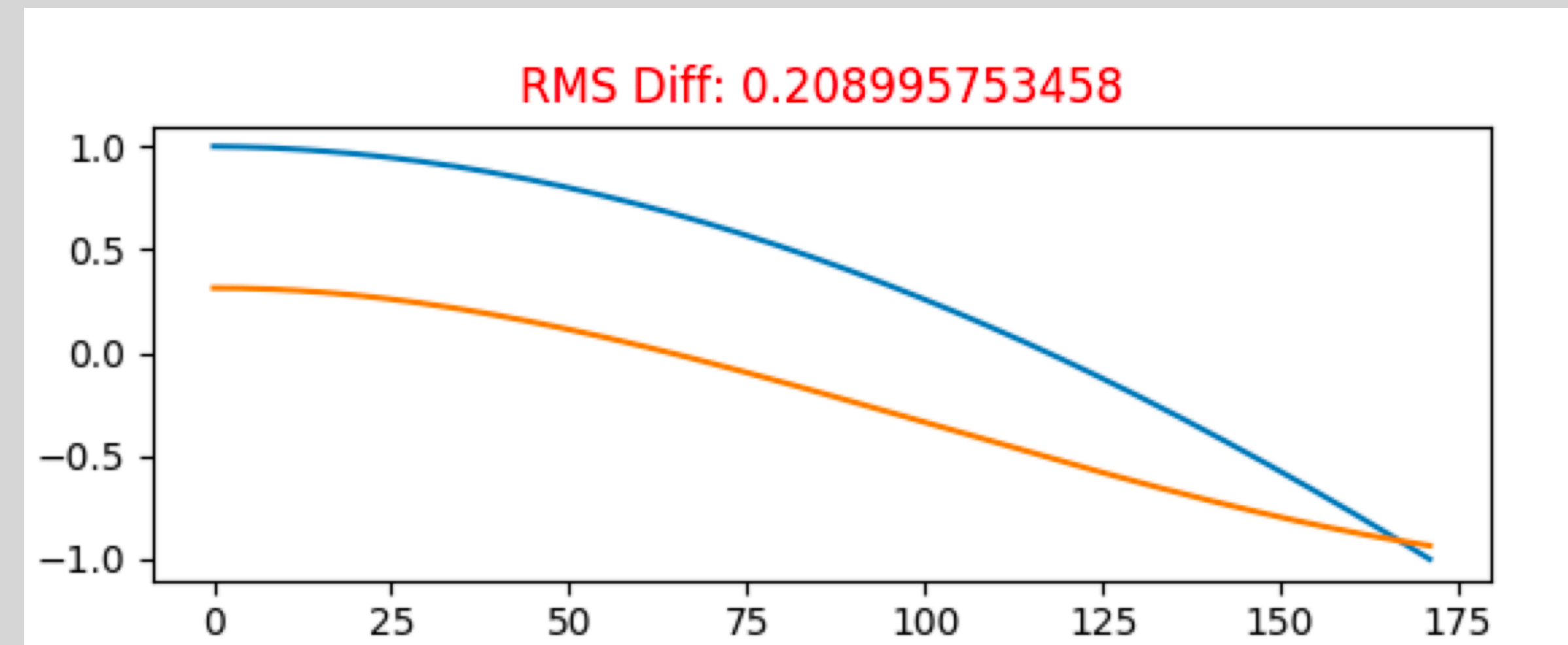
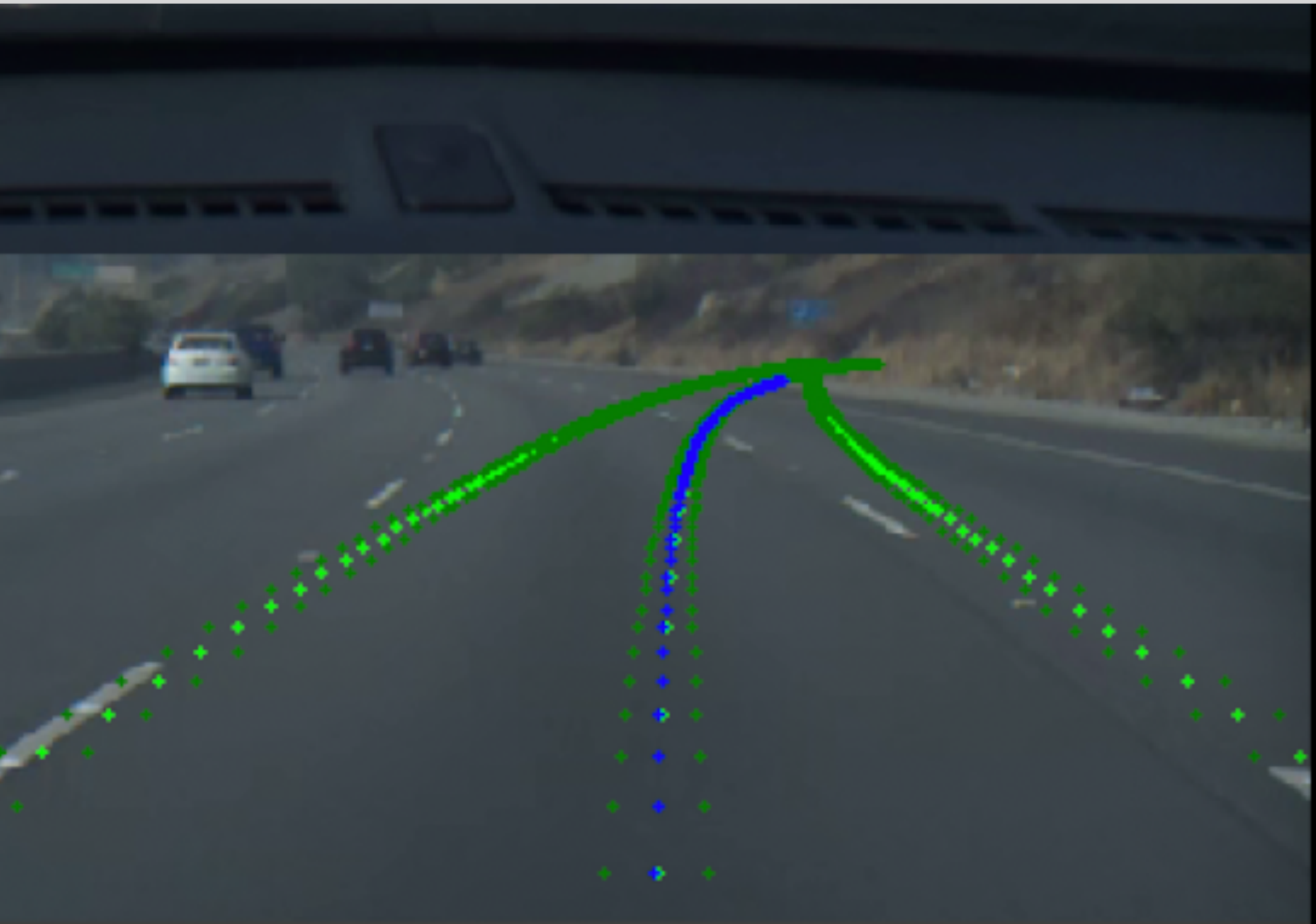
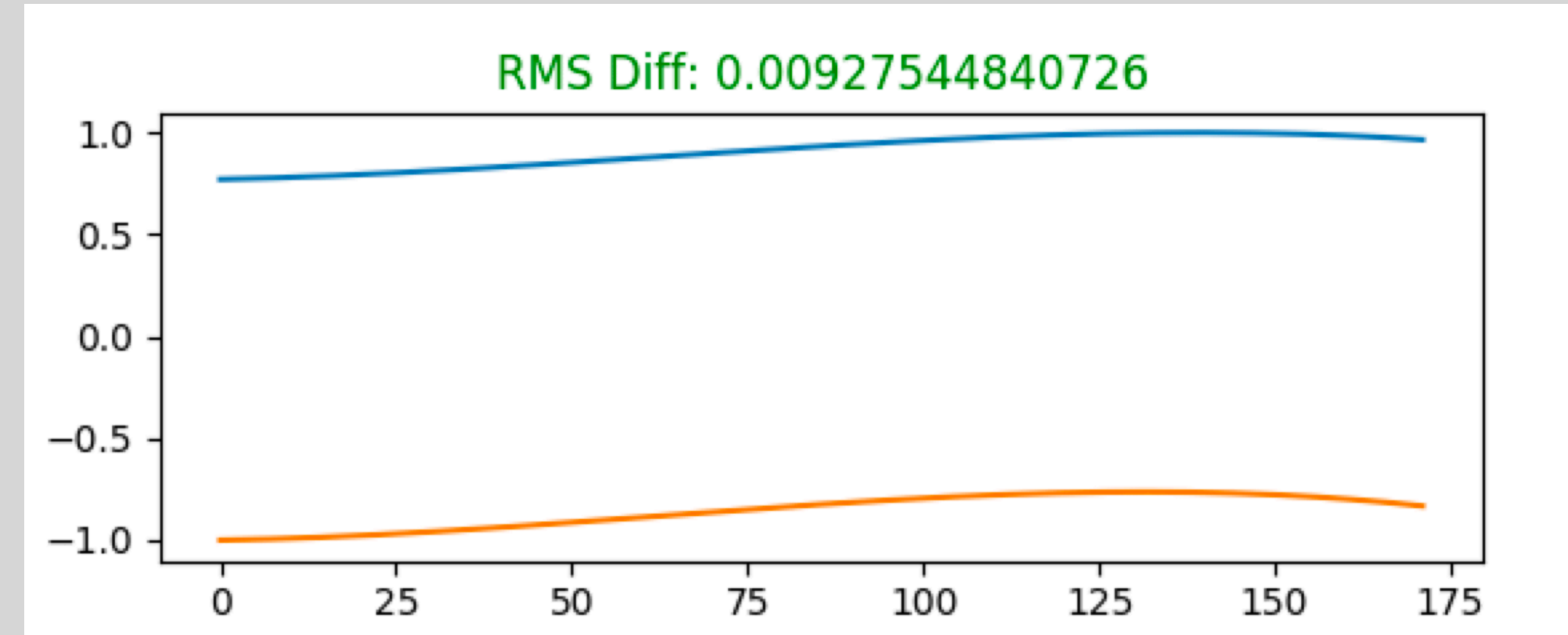


steps:

scale line segments to  $[-1, +1]$

shift lines together by average distance

calculate root mean square difference



# check #2: conformance to image

*steps:*

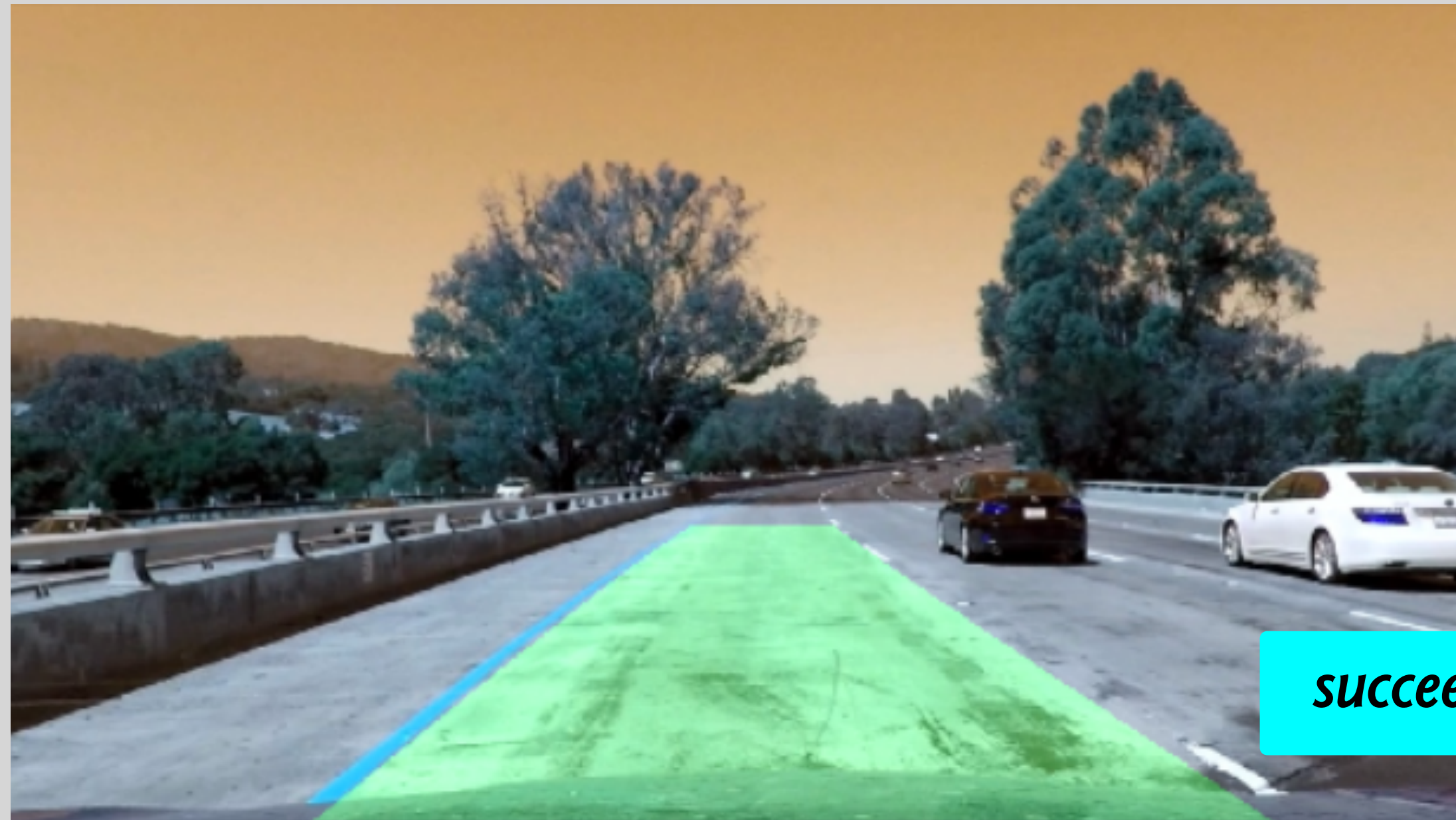
detect markers with filters/edge detection

transform to bird's eye view

convolve with purported lanes, left & right

try solid line first, then dashed line

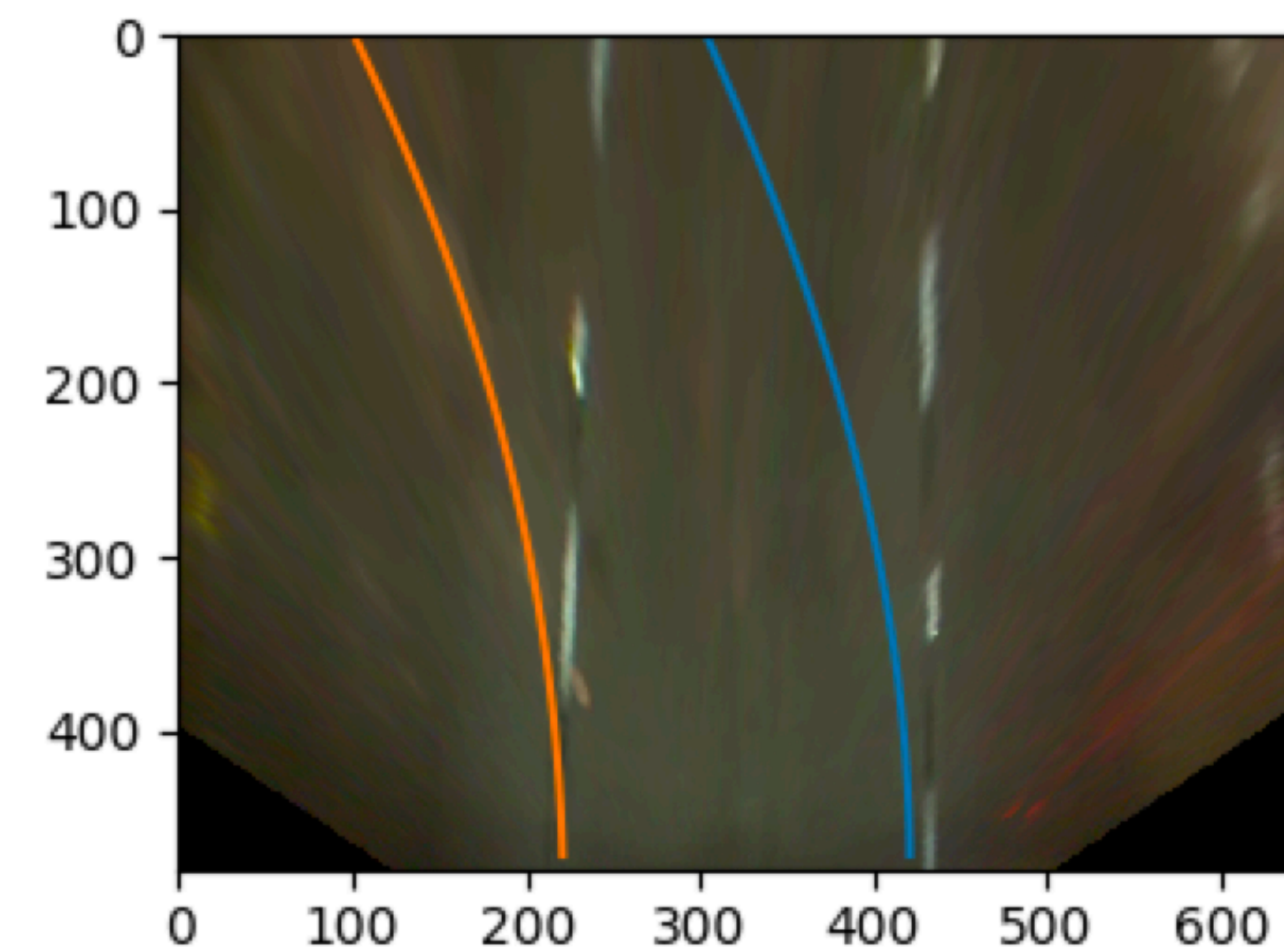
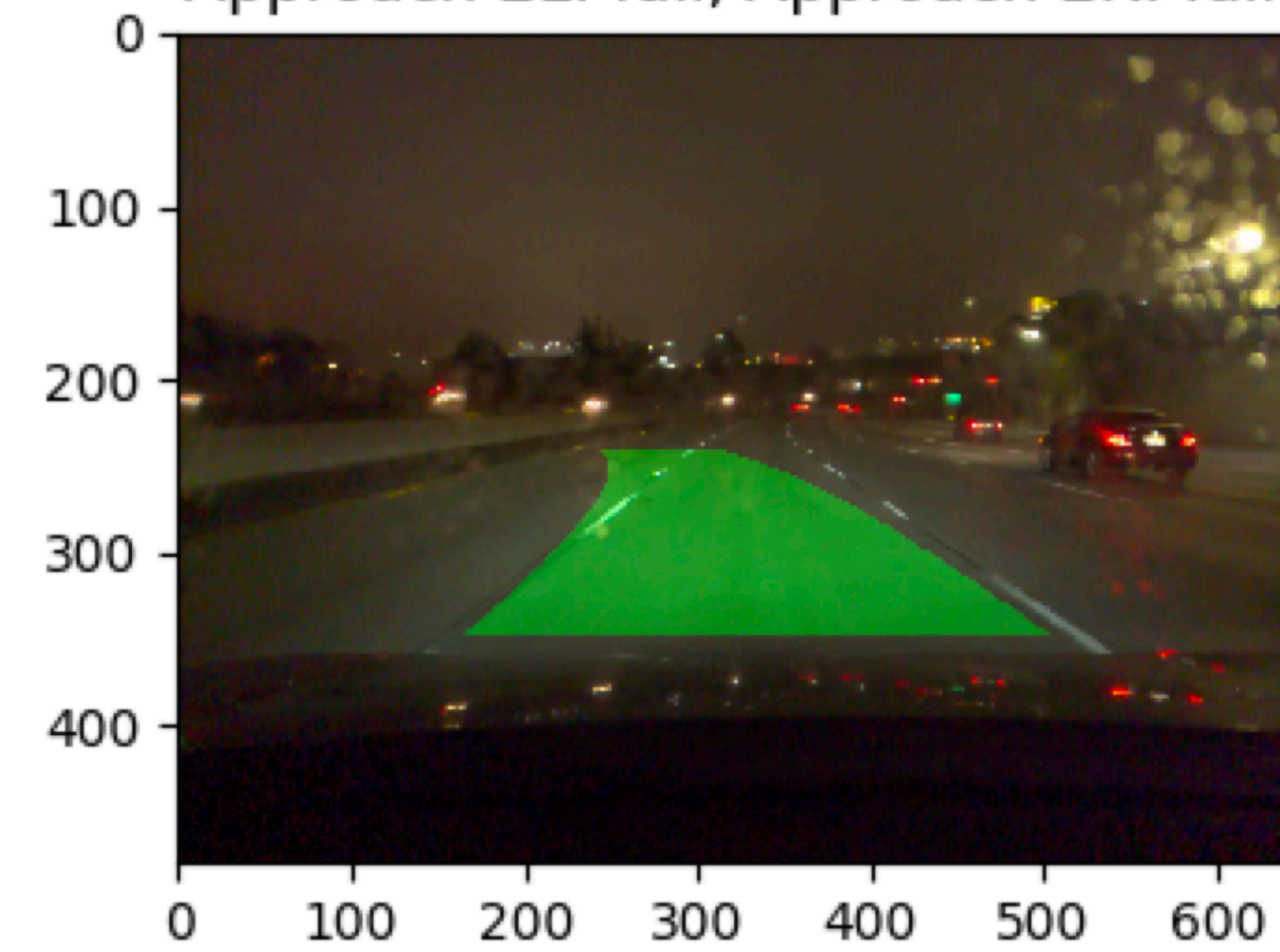
apply thresholds to decide if match



# experiment: open pilot sample video



Approach 1: pass  
Approach 2L: fail, Approach 2R: fail





## Tesla Autopilot Drives Straight Towards Concrete Barrier on Highway

3,596,905 views • Nov 8, 2019

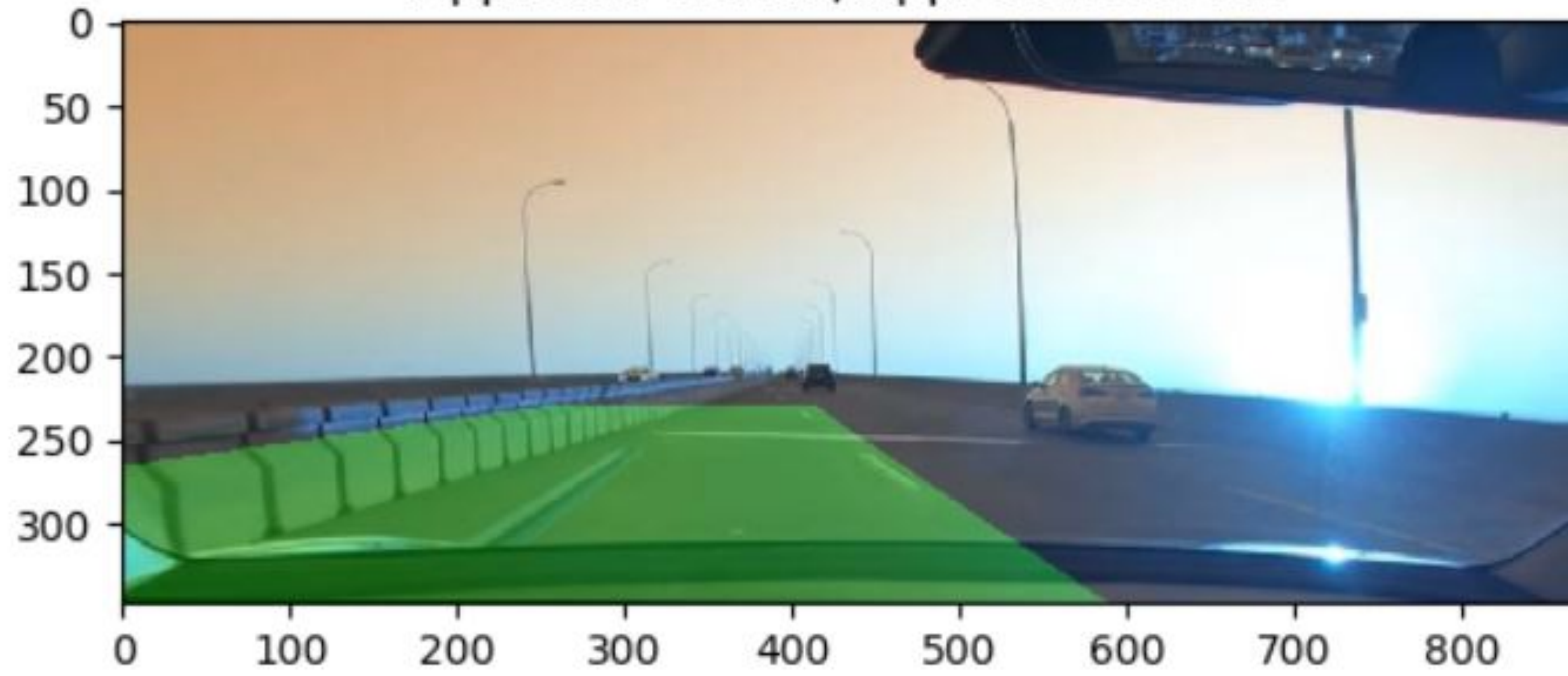
44K 1.5K SHARE SAVE ...

# tesla accident experiment

image from video of Tesla anomaly  
car treats beam of sunlight on barrier as lane line

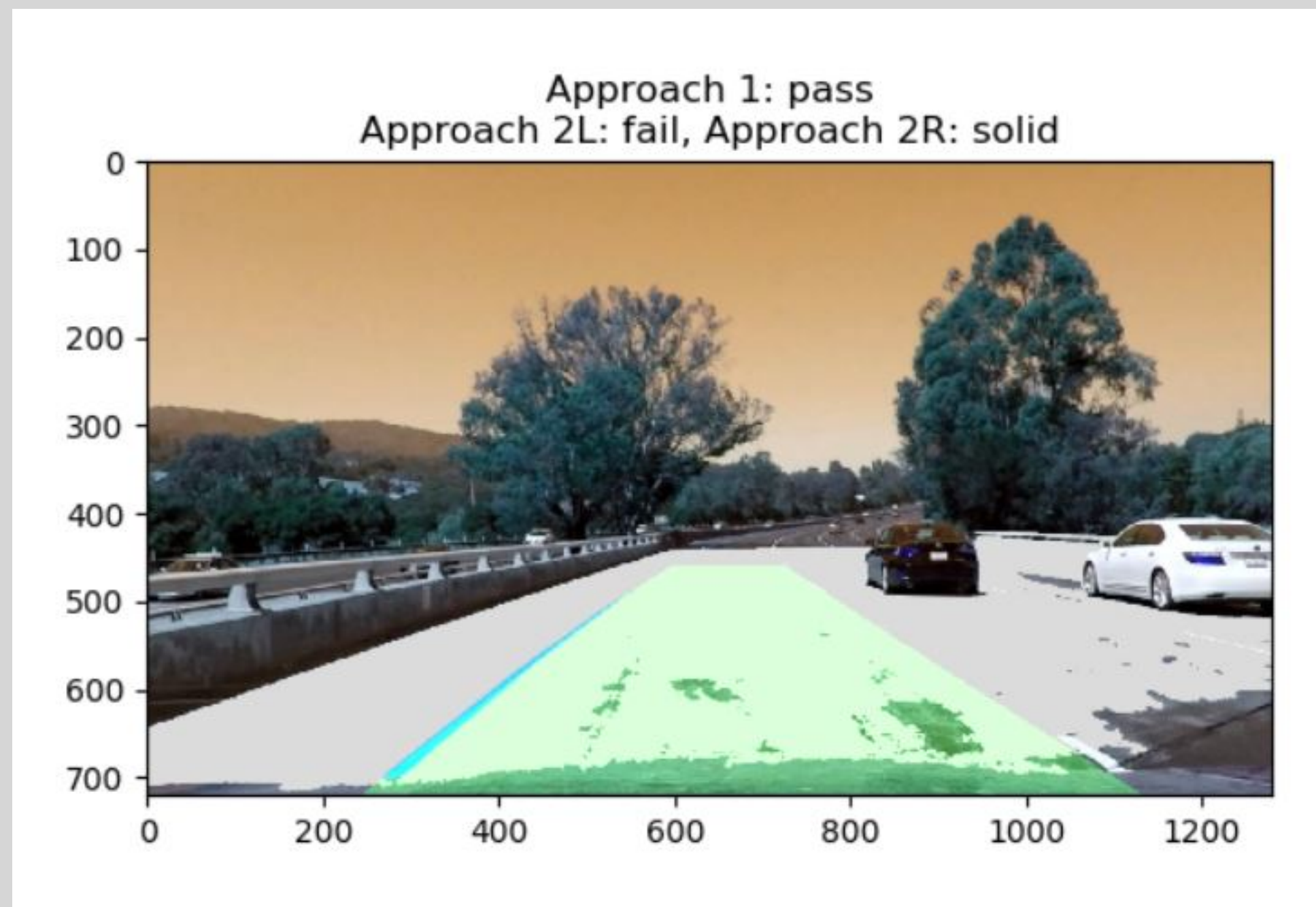


Approach 1: fail  
Approach 2L: fail, Approach 2R: fail

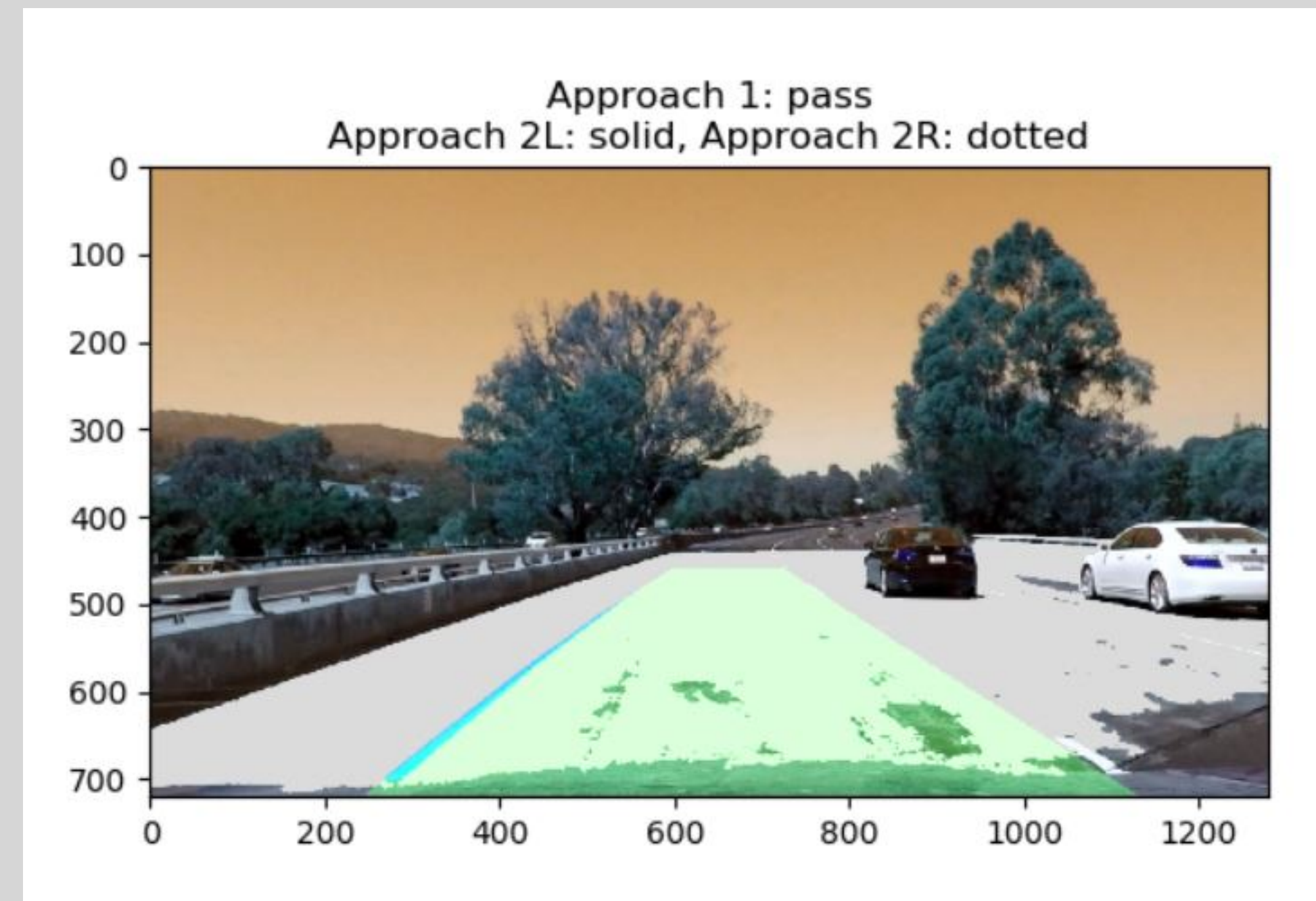


# simulating adverse lighting

modified road image to make road surface light grey  
now controller needs to apply color filter to find lane line  
can pass color filter parameters to interlock

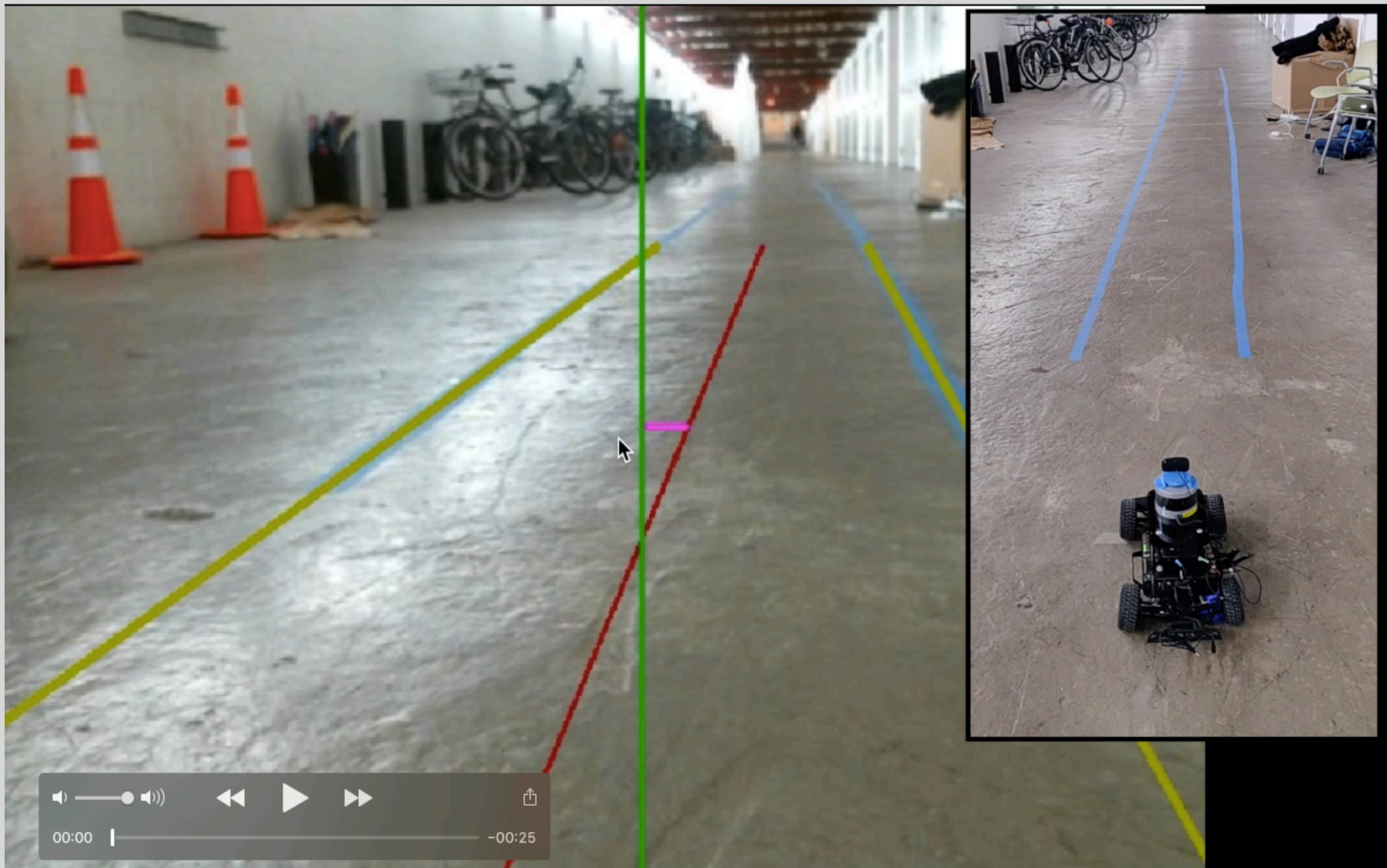


standard color filter: gets both lane lines wrong



revised color filter: gets both lane lines right

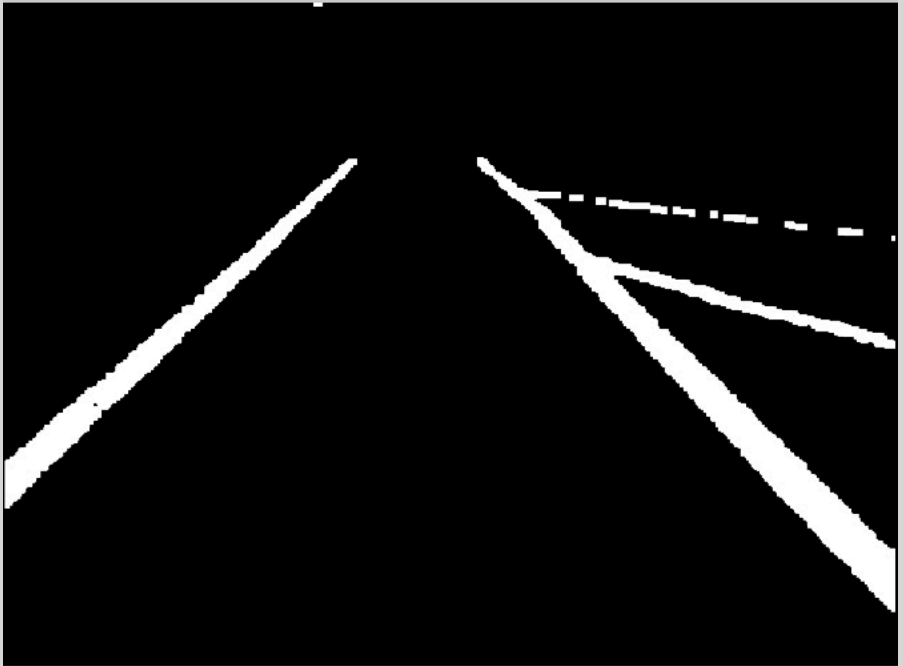
# racecar experiment



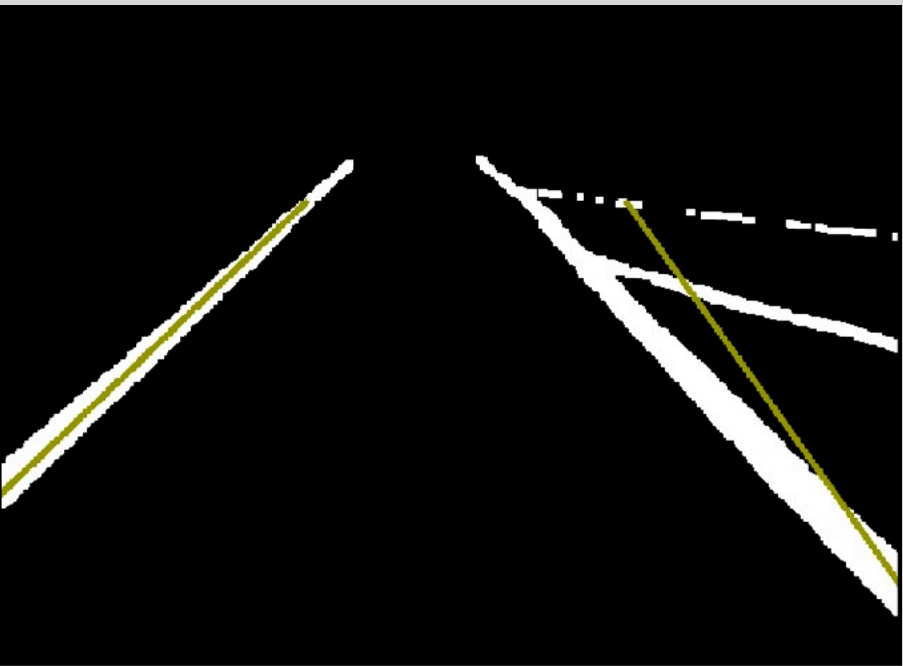
naive lane following algorithm



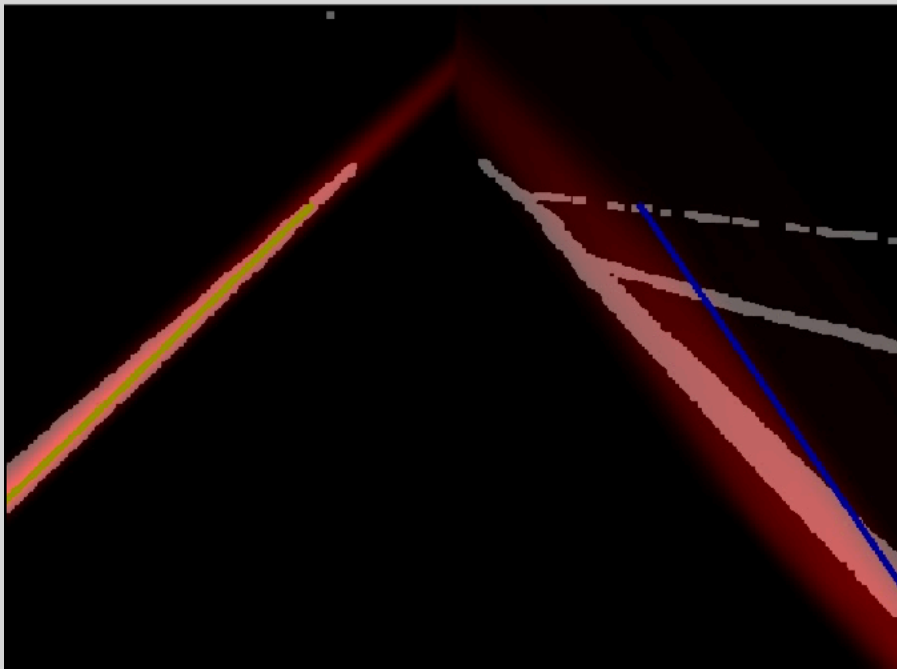
extra tape added on right to confuse controller



segmentation results



inferred lane lines

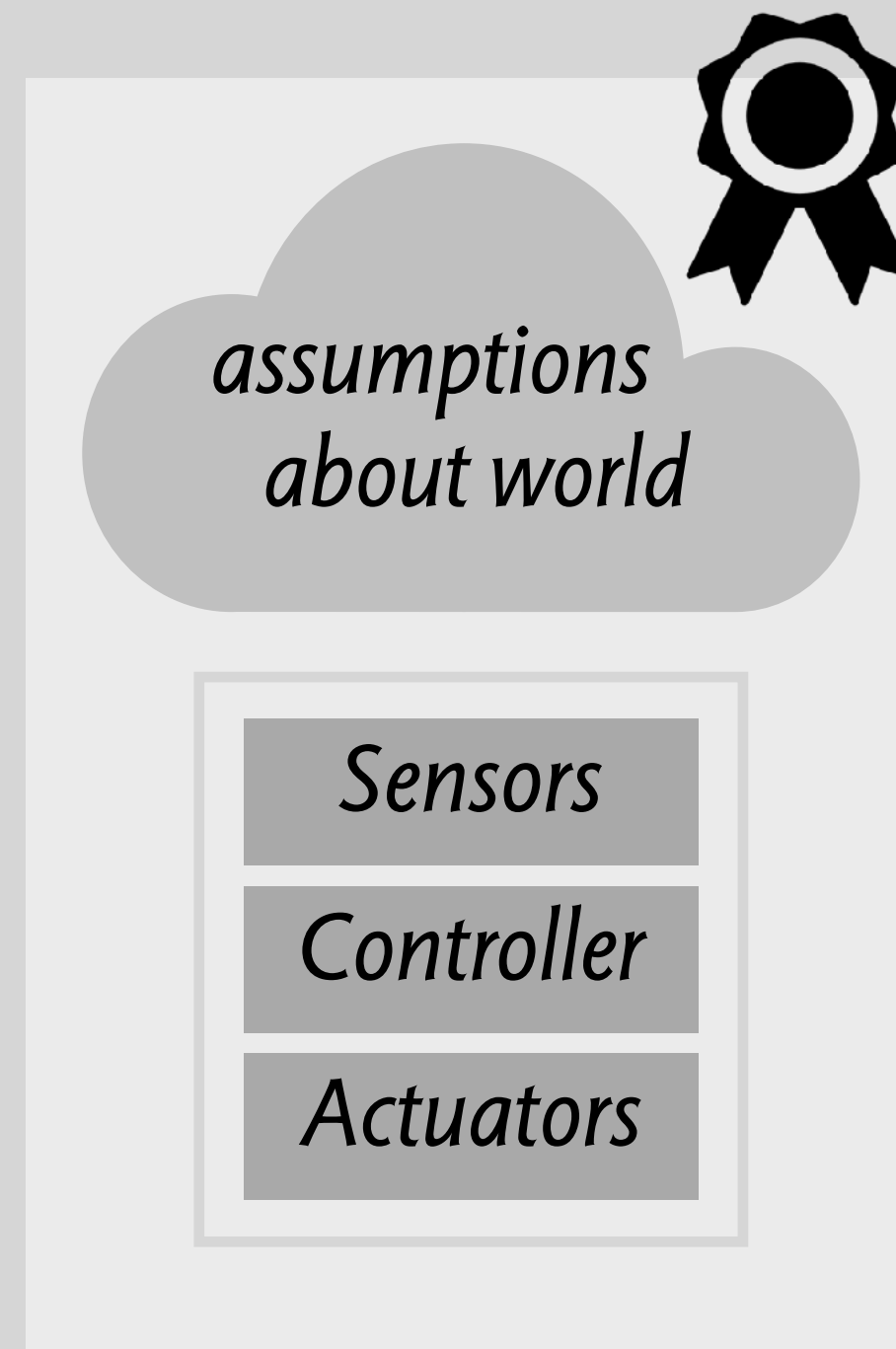


convolution result: reject

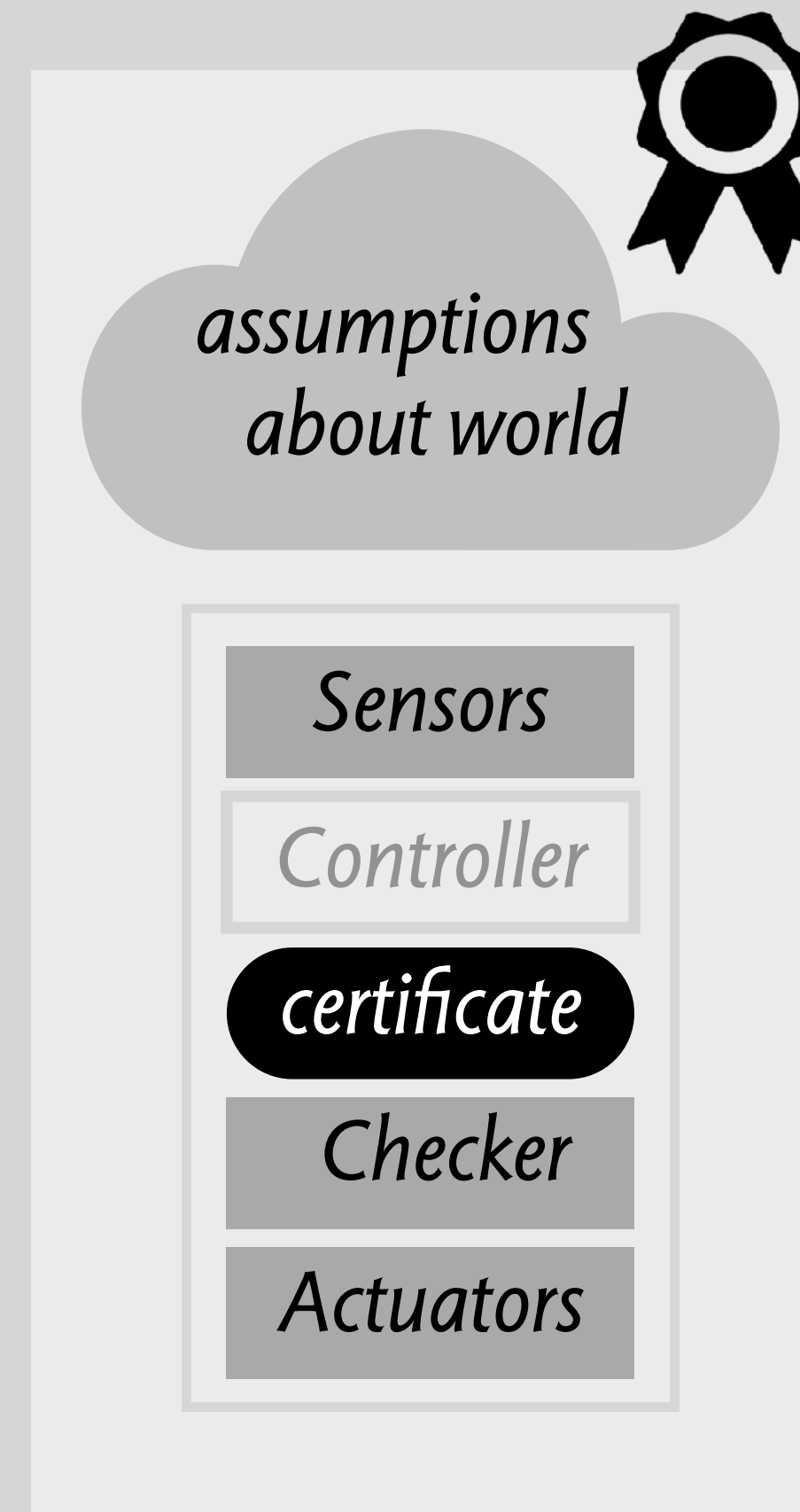


conclusion

# assurance cases



assurance case for  
traditional system



assurance case for certified control  
trusted base excludes controller

# some distinctions

## **being safe vs. being confident**

incidental safety is not enough  
public will demand evidence

## **anomaly detection vs. assurance case**

great work on anomalies in machine learning  
consistency between frames, common sense  
but assurance case goes further: an argument for safety

## **best effort vs. explicit safety**

today's controllers try to do their best  
no explicit articulation of what's achieved  
certificate articulates design consensus  
eg: LiDAR point density sets size of smallest obstacle

# next steps

## **simulations and trials**

end to end simulation in racecar  
integration with Toyota algorithms  
testing in variety of conditions

## **design issues**

certificate designs for different risks  
formal verification of safety case