

# A Mechanism to Assess the Safety of (Semi-)Autonomous Cars

---

Hanna Kurniawati

In collaboration with Jimmy Cai Huang



Australian  
National  
University

RESEARCH SCHOOL  
OF COMPUTER SCIENCE

# Lots of Work on Autonomous Cars

---



Pictures taken from news release of each car

# Lots of Discussions on Safety

---

- Definitions of safety
- AI (incl. ML) with safety guarantees
- Formal-method based verification
- Finding “bugs” (aka. scenarios that cause accidents)

From and for developers

---



# How about consumers?

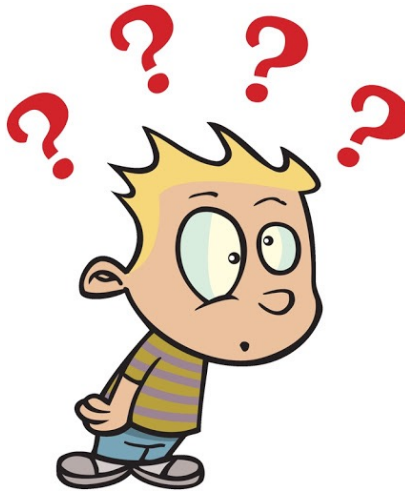
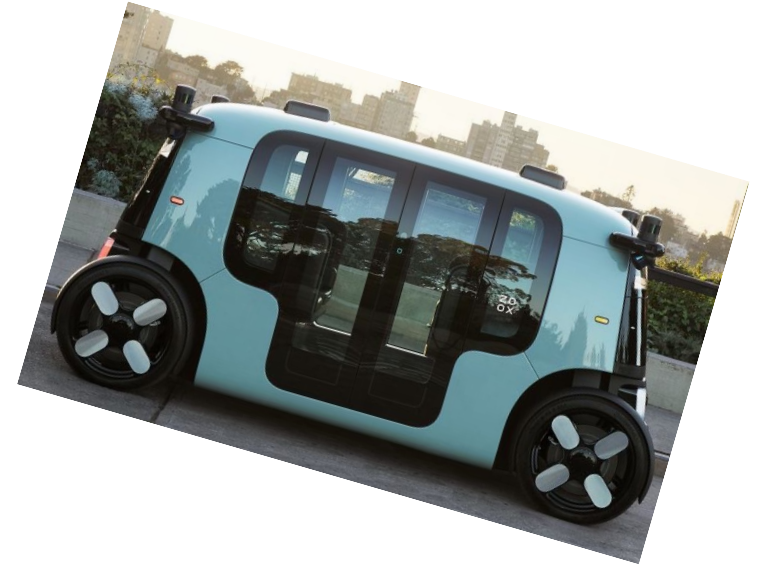
---



Pictures taken from news release of each car

# How about consumers?

---



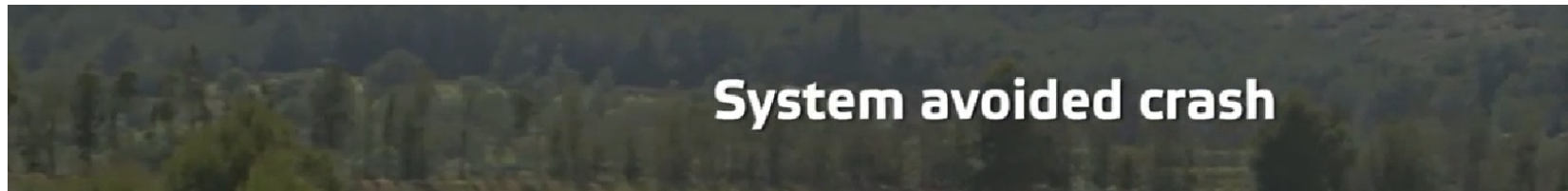
Pictures taken from news release of each car and <http://clipart-library.com>



# In terms of Safety?

---

- NCAP (New Car Assessment Programme)



Static tests done once to new cars

Autonomous cars are powered by s/w that:

Promises adaptability

Requires frequent updates/patches



# Our Proposal

---

## Car Safety Test



---

Picture taken from the facebook page of starcarwash

# Desirable Properties

---

- A simple safety indicator
  - One number, easy to understand by users
- Fast and easy assessment mechanism
  - Assessment can be done frequently (e.g., after every s/w updates, perhaps imposed during registration renewal)



# The Hypothesis

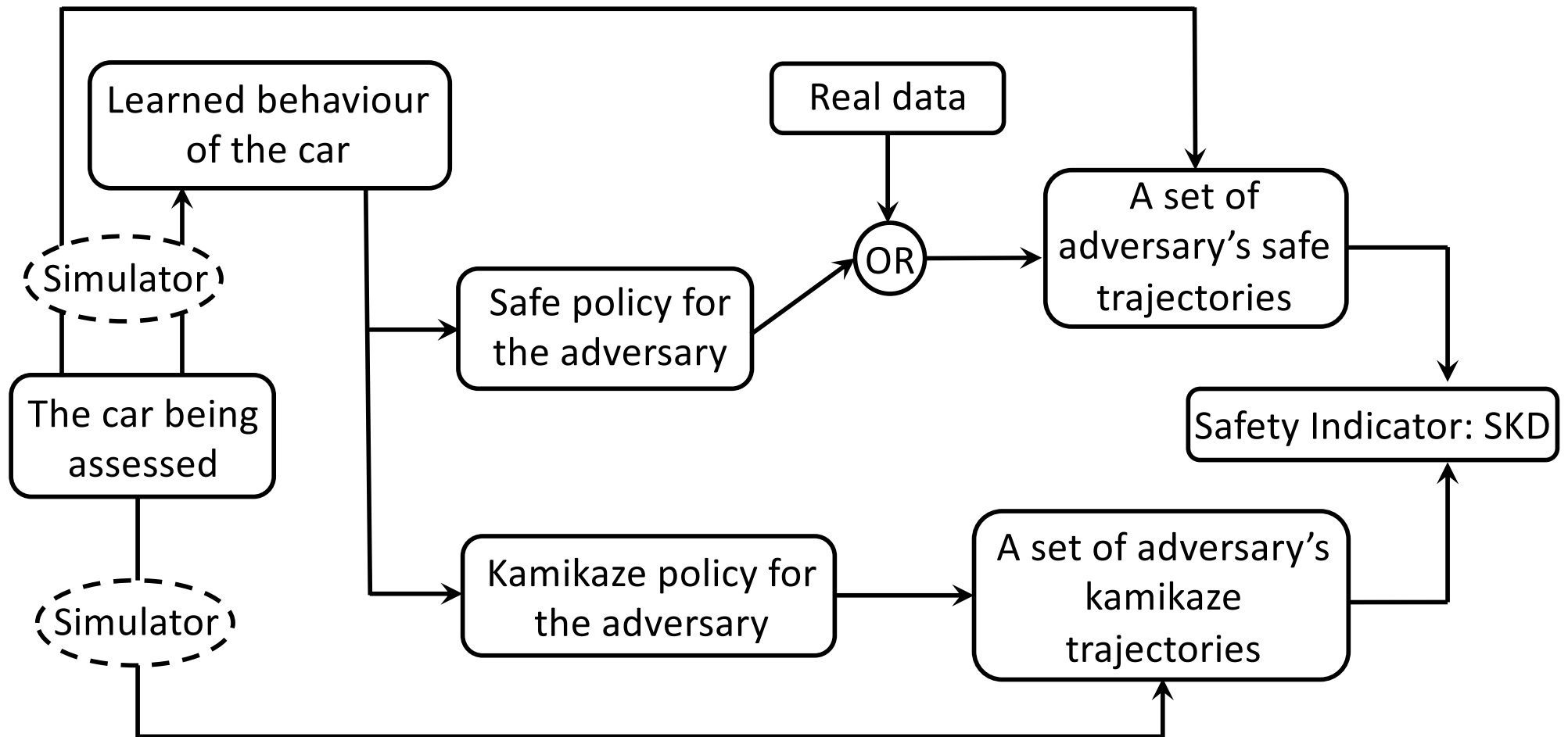
---



The distance between safe and dangerous trajectories of the adversary may serve as a safety indicator.  
Distance computation is generally fast!

# Proposed Mechanism

---



---

Dashed ellipse: May not be used

# Safe Trajectories

---

- Real data
- Provided by regulatory body
- Generate synthetic data
  - Strive for optimal to get to the destination [biological hypothesis, Breed & Moore'15]
  - Account for the car's behaviour, though don't know their exact policy → partially observed
  - Use Partially Observable Markov Decision Process (POMDP)
- A trajectory:
  - A mapping from  $[0, 1]$  to position of the adversary in the operating environment (in this case, a bounded  $\mathbb{R}^2$ )
  - Can be approximated by a polygonal chain (a sequence of line segments)



# Kamikaze Trajectories

---

- Given a safe adversary's trajectory  $\phi$
  - Find a sampled set of adversary's trajectories closest to  $\phi$  that causes collision with the car being tested
    - The exact policy of the car is not known in advance and need to be learned  $\rightarrow$  car's behaviour is partially observed
    - Closest: Current practice, within certain distance away from  $\phi$
    - Use Partially Observable Markov Decision Process (POMDP)
-

# The Safety Indicator: SKD

---

- Given:
  - A set of safe trajectories  $\Phi$
  - A set of kamikaze trajectories  $\Psi(\phi)$  for each safe trajectory  $\phi \in \Phi$
- Suppose  $\Psi = \bigcup_{\phi \in \Phi} \Psi(\phi)$ , then

$$SKD(\Phi, \Psi) = \frac{1}{|\Psi|} \sum_{\phi \in \Phi} \sum_{\psi \in \Psi(\phi)} d(\phi, \psi)$$

SKD = Safe-Kamikaze Distance

$$d(\phi, \psi) = \inf_{\alpha, \beta} \max_{t \in [0, 1]} \|\phi(\alpha(t)) - \psi(\beta(t))\|$$

Also known as Fréchet distance

---

# Turns out ...

---

If  $SKD(\Phi, \Psi) = \delta$  and we have sufficiently many samples, then

$$Prob(d(\phi', \psi') < \eta) \leq 2 \left( \frac{\eta}{\delta^2} + \frac{\eta}{\delta} \right)$$

for small  $\eta \in (0, \delta)$  and any safe trajectory  $\phi'$  and corresponding dangerous trajectory  $\psi'$  sampled from the same distribution as the one used to generate  $\Phi$  and  $\Psi$

The probability that a small deformation changes a safe adversary's trajectory into a dangerous one is upper bounded by a value inversely proportional to SKD

---



# So...

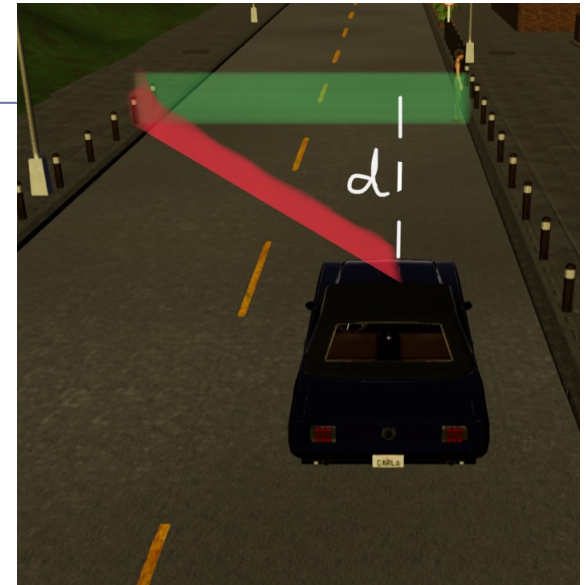
---

- Large SKD is a possible sufficient condition for an autonomous car to be safe. Meaning:
    - If SKD is large, the car is likely to be safe
    - If SKD is small, unfortunately, we can't say much!!!
-

# Systematic Test

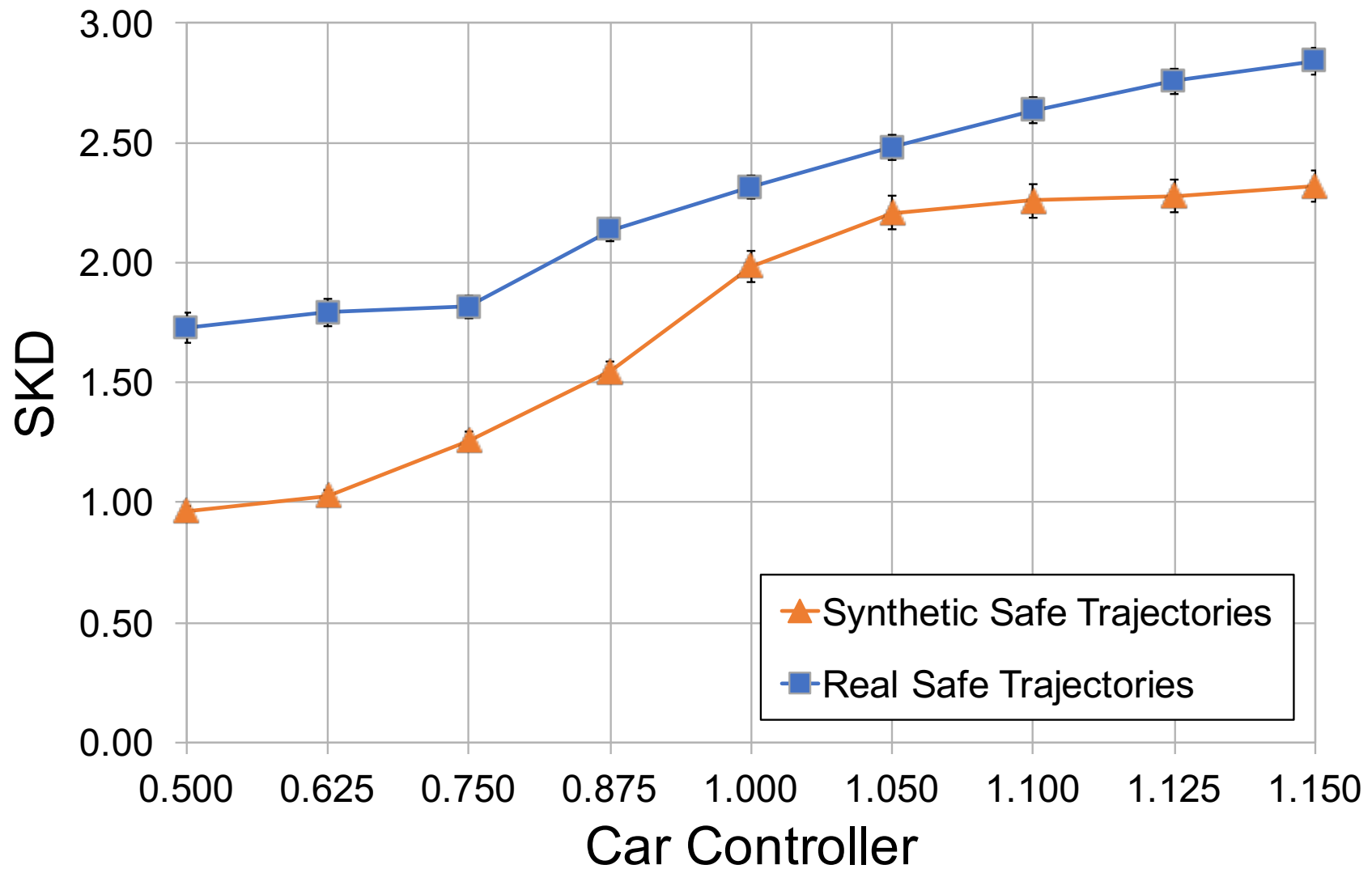
---

- Scenario: Euro-NCAP Vulnerable Road User test on Autonomous Emergency Breaking
  - A pedestrian crossing a road in front of the car moving on a single lane
  - The car:
    - Has a maximum velocity & fixed acceleration/deceleration
    - Safe stopping distance  $\kappa$ : The distance to move from maximum velocity to 0 with maximum deceleration
    - Where the car starts to decelerate:  $C \cdot \kappa$  distance away from the pedestrian
      - $C$ : Multiplier, larger means safer
- 



# Simulation Results

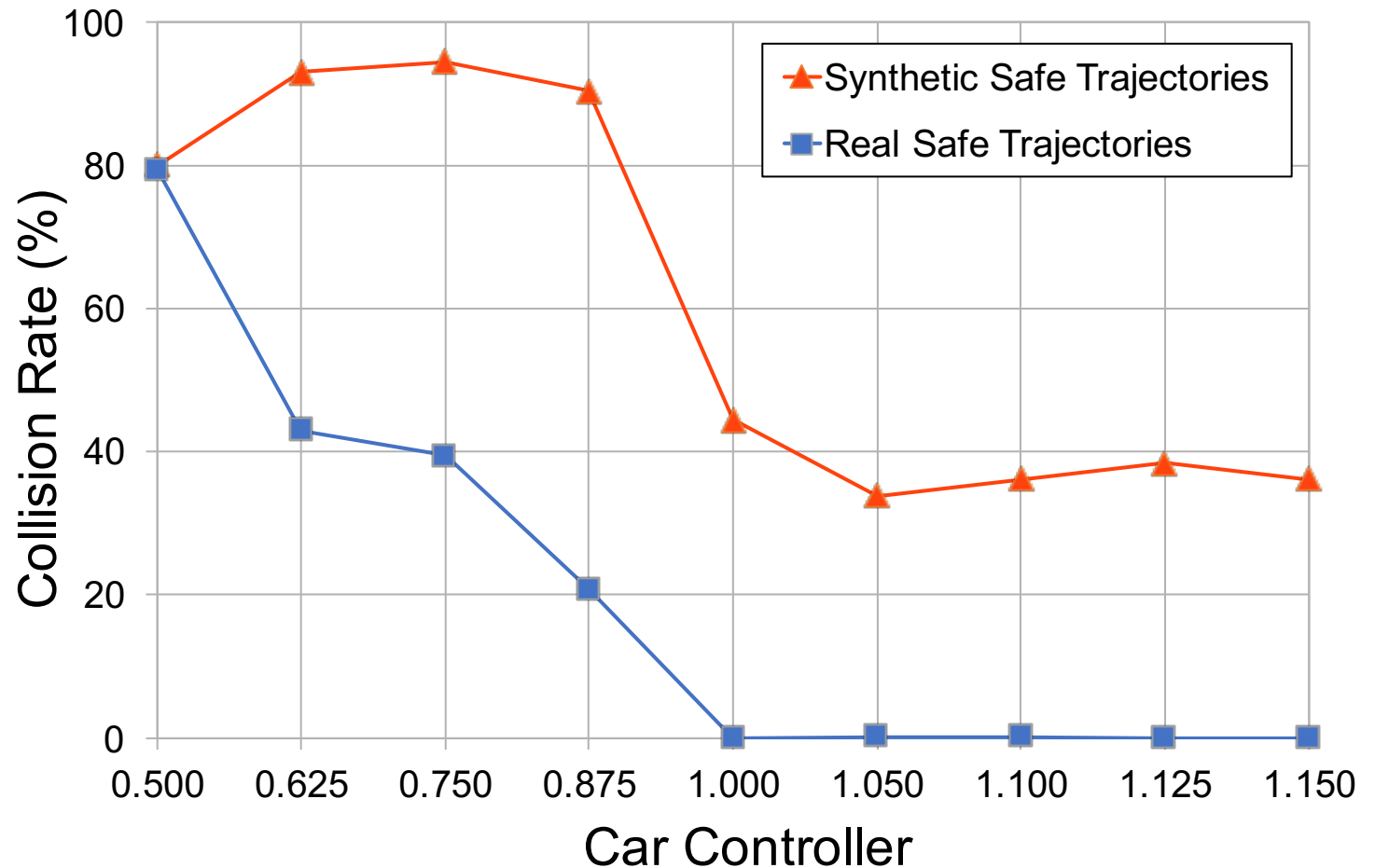
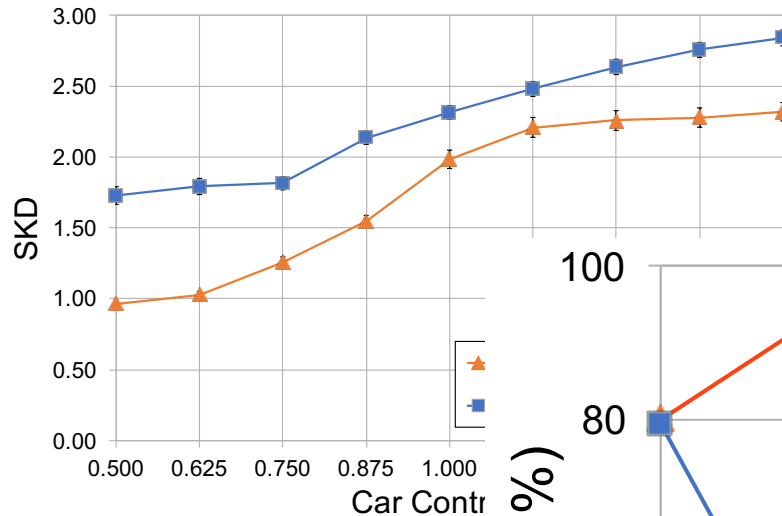
---





# Simulation Results

Average time to generate trajectories + computing SKD  
< 10s per run



# Summary

---

- The safety indicator SKD can be used to upper bound the probability that a small deformation changes a safe adversary's trajectory into a dangerous one
- SKD with sufficient statistical confidence can be computed in under 30 minutes using a typical desktop (i7 quad-core)



---

Thank you

Q&A

---