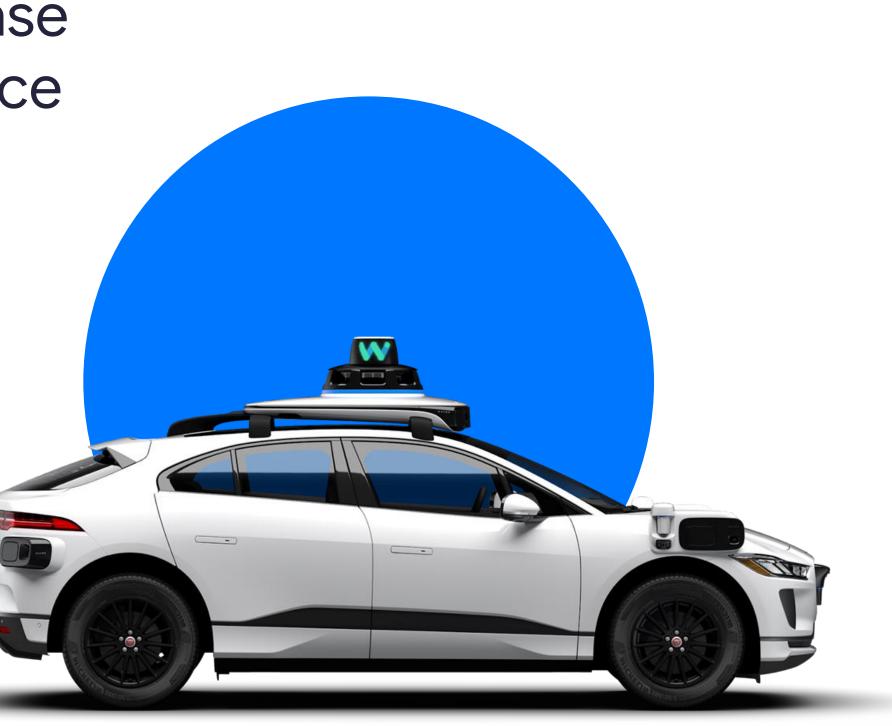
# Waymo's Safety Case and Use of Reference Models

Presented to the 42nd Session of the Informal Working Group on Functional Requirements for Automated Vehicles, Mountain View, CA, July 2023

Scott Schnelle Kristofer Kusano Johan Engström *Safety Research & Best Practices* 





## Introducing Waymo's Approach to Safety: a Safety Case

A "safety case" is a structured argument, supported by a body of evidence that provides a compelling, comprehensible, and valid case that a system is or will be adequately safe for a given application in a given environment.

[UK Ministry of Defense DS 00-56, adapted subsequently in UL 4600]

(Overarching statement)

The top-level goal of Absence of Unreasonable Risk. Safety is defined in ISO as Absence of Unreasonable Risk (AUR)

Building a Credible Case for Safety: Waymo's Approach for the Determination of Absence of Unreasonable Risk (March 2023)

Waymo's Safety Methodologies and Safety Readiness Determinations (October 2020)

(Compelling proof)

**EVIDENCE** 

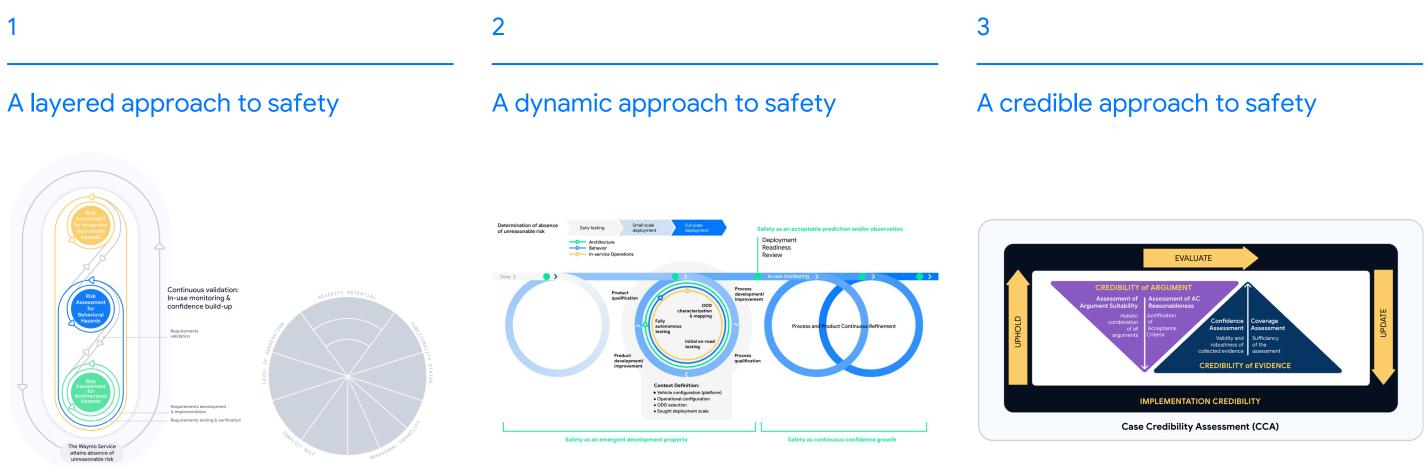
LOGICAL ARGUMENT

GOA

#### The determination of safety is, at its heart, a risk assessment process



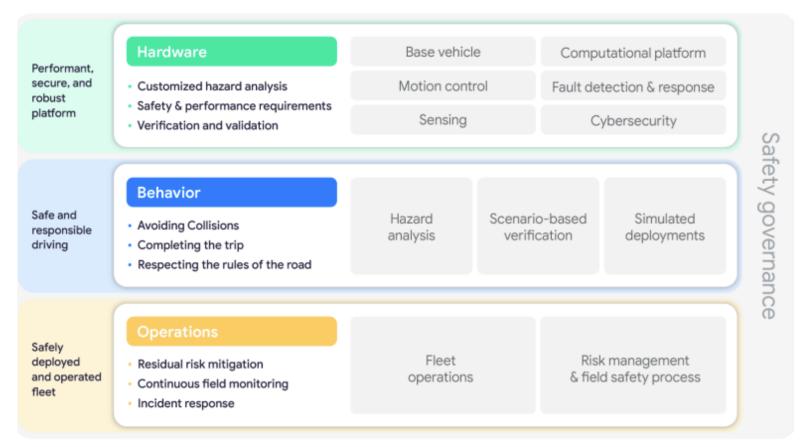
### Introducing Waymo's Approach to a Safety Case



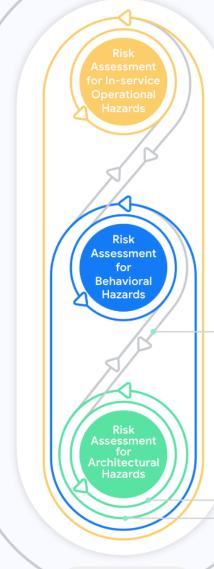


## Introducing Waymo's Approach to a Safety Case

#### A layered approach to safety



Waymo's Safety Methodologies and Safety Readiness Determinations (October 2020)



The Waymo Service attains absence of unreasonable risk Building a Credible Case for Safety: Waymo's Approach for the Determination of Absence of Unreasonable Risk (March 2023)

#### Continuous validation: In-use monitoring & confidence build-up

Requirements validation

Requirements development & implementation

Requirements testing & verification



### A Dynamic Approach to Safety



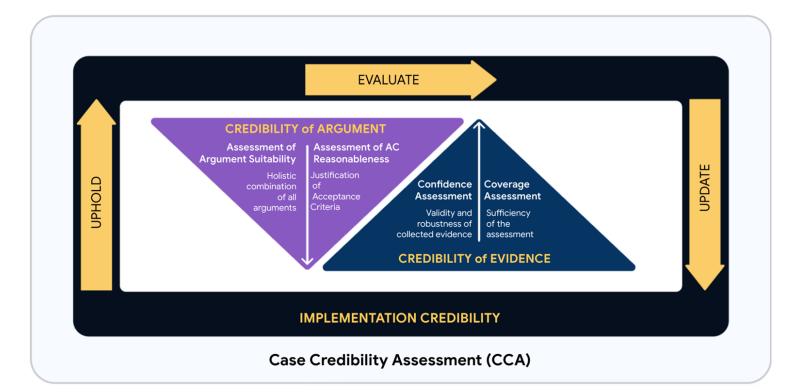




### A Credible Approach to Safety

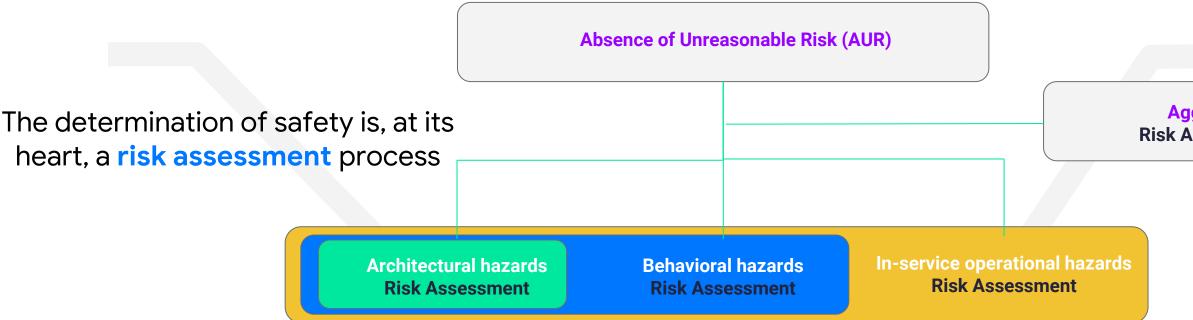
A "safety case" is a structured argument, supported by a body of evidence that provides a compelling, comprehensible, and valid case that a system is or will be adequately safe for a given application in a given environment.

- Goal: Overarching statement
- Logical argument:Decomposing the statement
- Evidence: Compelling proof





### A Layered Approach to Safety: Decomposing AUR



Architectural hazards: those associated with potential sources of harm inherently embedded within the platform because of architectural choices. Example: undesired presence of blindspots, stemming from architectural choices related to sensors' typology and placement. **Behavioral hazards:** those associated with potential sources of harm resulting from the ADS's displayed driving behavior, whether intended or unintended.

Example: undesired degree of proximity to surrounding road users.

In-service operational hazards: those associated with potential sources of harm resulting from the fact that the ADS operates in a complex ecosystem, and that do not belong to the other two categories. Example: improper securing of cargo or undesired access to the vehicle from a malicious actor.

#### Aggregate Risk Assessment



### AUR Determination and Risk Assessment

- For all hazard categories, a set of explicit risk acceptance criteria should be stated to assess if the residual risk reached an acceptable level or further mitigations are required.
- The crux thus remains of how to determine that a certain collection of acceptance criteria adequately covers a certain category of hazards.
- The process of setting appropriate acceptance criteria relies on the following three assumptions:

**A.** A sufficiently exhaustive list of hazards can be identified and covered by the categories "architectural", "behavioral", and "in-service operational"; **B.** We can define indicators of interest mapped to each hazard category to set an explicit acceptance criterion for risk evaluation;

**C.** We can define the minimum set of dimensions of interest to state completeness of the set of acceptance criteria and establish credibility



p. 8

**B.** We can define indicators of interest mapped to each hazard type to set an explicit acceptance criterion for risk evaluation;

**C.** We can define the minimum set of dimensions of interest to state completeness of the set of acceptance criteria and establish credibility



Acceptance Criteria Framework for AUR Behavioral Evaluation



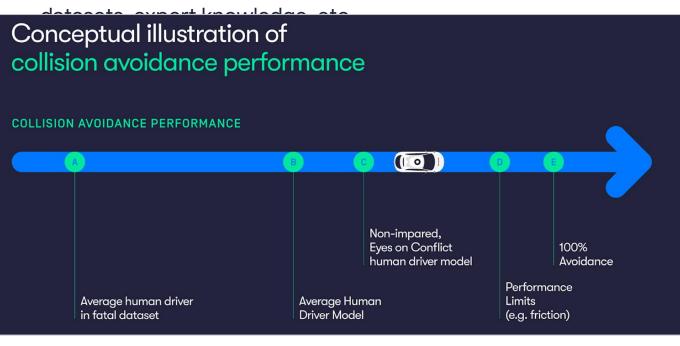


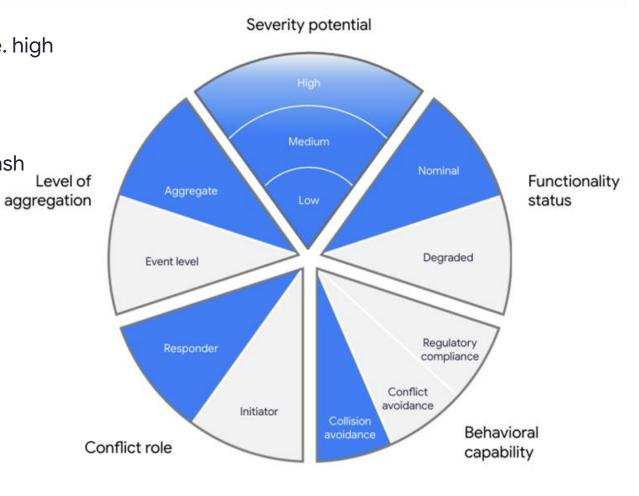
### Example of Targeted Scenario-Based Behavior Evaluation:

### Collision Avoidance Testing (CAT)

Kusano, K., Beatty, K., Schnelle, S., Favaro, F., Crary, C., and Victor, T. 2022. Collision Avoidance Testing of the Waymo Automated Driving System. arXiv:2212.08148

- Virtual, scenario-based testing methodology that evaluates the safety of the ADS's intended function
- Compared to the Non-Impaired driver with Eyes ON conflict (NIEON) model, i.e. high performing human driver
- ADS's ability to avoid situations initiated by others that require urgent evasive maneuvers
- Representative of a given ODD, informed by our driving experiences, public crash







#### Claim Structure: Application of the CCA through Argument construct

#### combination of all

**Claim:** AC [insert methodology specific AC] provides an explicit criterion to evaluate predicted RO performance appropriately mapped to dimensions [insert methodology specific AC framework dimensions] for the given context.

Subclaim (SC) #1: The stated acceptance criterion is reasonable

Subclaim (SC) #2: Methodology [insert methodology name] provides credible evidence that the stated acceptance criterion is met



านอุณรรอรรธ

fnemzeezeA

Coverage

of the Sufficiency robustness of

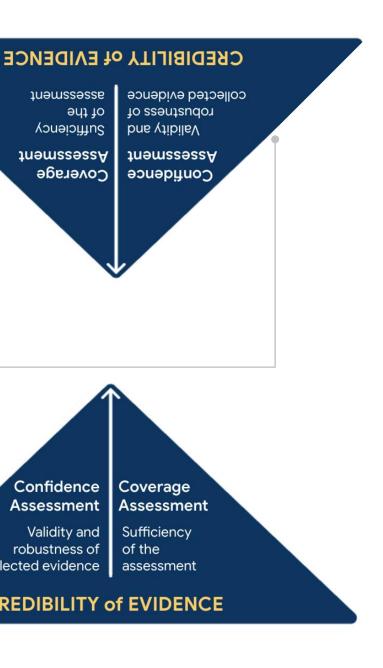
framssassA

Sonfidence

Validity and

#### **CREDIBILITY of EVIDENCE**





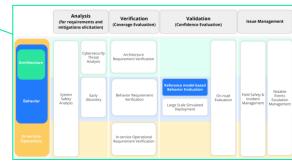


## **Example: Collision Avoidance Testing**

Acceptance Criterion #1 (AC1): The predicted RO collision avoidance capability attained by the Waymo Driver in a number of conflict scenarios initiated by the actions of other road users is assessed through a comparison with a non-impaired, eyes on conflict behavioral reference model made progressively more stringent by decreasing its emergency maneuver response time. Scenario groups are graded at an aggregate level, with individual scenarios within a group contributing to a neutral/positive/negative gap for the ADV when the Waymo Driver shows even/better/worse performance than the artificial driving model in terms of collision outcomes and injury-causing collisions. Minimum passing scores vary between scenario-specific groups, each including either vehicle to vehicle or vehicle to vulnerable road users interactions.

#### Claim 1: AC1 provides an explicit acceptance criterion to evaluate predicted RO aggregate ADV performance related to responder-role collision avoidance capability in nominal (i.e., non degraded) conditions for the given context.

- Subclaim #1: AC1 is a reasonable criterion. 1.
  - 1. The acceptance criterion is specified at the appropriate level of aggregation.
  - 2. The scoring assigned to the [VRU/V2V] type is adequate for determining the ADV's performance relative to the reference model.
  - 3. The NIEON artificial driving model is an appropriate and sufficient benchmark for evaluating responder role collision avoidance.
  - 4. The AC supports data-driven release gualification and identification of onboard engineering work to continuously improve the Waymo Driver.
  - 5. The AC is predicated upon appropriate performance indicators.
- Subclaim #2: The CAT methodology provides credible evidence that AC1 is met. 2.
  - 1. Coverage Assessment: The CAT methodology leverages a set of scenario groupings that represent adequate coverage of hazardous situations to develop a safety set that can be assessed for responder role collision avoidance capability for the Waymo ADV in nominal (i.e., non degraded) conditions for the given context.
  - 2. Confidence Assessment: The CAT methodology attains the appropriate confidence in the collision avoidance performance for the Waymo Driver in responder role predicted for RO operations, and its comparison relative to the chosen behavioral reference model for the given context.
    - Scoring confidence [...] 1.
    - Conservativeness [...] 2.
    - Fidelity [...] 3.
    - Robustness [...] 4.
    - Appropriate use of qualified tools [...] 5.
    - Technical validity of benchmark [...] 6.



#### Context:

- Platform: i-Pace. Pacifica
- Release: x.x.x

#### CAT white paper

Use-case: Ride-hailing urban/sub-urban Scale of deployment: e.g., fleet size, expected mileage Scope (ODD features and ADV behaviors) [LINK] ODD feature in-depth description or similar





# The Non-impaired Eyes ON (NIEON) Conflict Driver

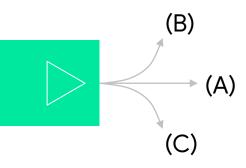
2

Attentive with eyes 1) always on the conflict Response time (s) 0 0.5 1.5 2.0 0.0 1.0 Ramp-up time (s) Model fit response response time 2) using eyes-on-road, non-impaired naturalistic driving data\*

Three chances given 3) (best outcome selected): (A)Brake only

Engström, J., Liu S-Y, Dinparastdjadid, A. and Simoiu, C. 2022. Modeling Road User Response Timing in Naturalistic Traffic Conflicts: A surprise-based framework. arxiv.org/abs/2208.08651

#### **Collision Avoidance Model** Only that responds after a conflict has been entered.



(B)Brake + steer left (C)Brake + steer right



# Research Needs for Driving Reference Models

### Generic vs. Scenario-specific Models

Collision vs Conflict Avoidance Models

Complexity of Models

### Validation Methods and Criteria

# Closing / Thank you

Scott Schnelle Kristofer Kusano Johan Engström Safety Research & Best Practices

July 2023



