

Automated Driving Safety Evaluation Framework

Ver 2.0

Japan Automobile Manufacturers Association, Inc.

Sectional Committee of AD Safety Evaluation,

Automated Driving Subcommittee

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List of committee members

Chief of Sectional Committee: Satoshi Taniguchi, Toyota Motors

Deputy Chief of Sectional Committee: Koichiro Ozawa, Honda Motor Co., Ltd.

Deputy Chief of Sectional Committee: Eiichi Kitahara, Nissan Motors Co., Ltd.

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Committee member, Vehicle Movement Disturbance WG Leader: Shinta Arai, Honda R&D Co., Ltd.

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Advisor: Koichi Terui, Hitachi Astemo, Ltd.

Advisor: Tatsuhiko Monji, Hitachi Astemo, Ltd.

Advisor: Yuko Murase, DENSO CORPORATION

Advisor: Kenji Suganuma, DENSO CORPORATION

Advisor: Shingo Jinno, DENSO CORPORATION

Advisor: Itaru Takemura, Pioneer Smart Sensing Innovations Corporation

Advisor: Hajime Koyanagi, Pioneer Smart Sensing Innovations Corporation

Advisor: Masami Suzuki, Pioneer Smart Sensing Innovations Corporation

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1. Positioning of this Paper

【Background】

The realization and deployment of autonomous driving (AD) is expected to bring forth an even safer society which is also more efficient and with a freer mobility. The fulfillment of these expectations is a major global challenge that stands on the sufficient safety assurance and verification of the autonomous vehicles both in terms of performance and technology.

In this document, the Japan Automobile Manufacturers Association Inc. (JAMA) has summarized the best practice on safety argumentation structuring, safety evaluation, and safety assessment methods needed to enable logical completeness, practicability, and transparency of AD safety on limited access highways.

The safety assessment and the technical judgment may be revised according to the practical implementation and evolution of the AD safety assurance dialogue, along with technical content modifications.

【Aims】

- ① To enhance safety and efficiency of AD systems development by providing guidelines that serve as a common ground for each JAMA member at each product development stage, from planning and design, to evaluation.
- ② To gain a common technical understanding when international regulations and standards are formulated.
- ③ To clarify JAMA position when cooperating with international projects.

2. Automated Driving System Safety Argumentation Structure

An overview of the safety argumentation structure for AD systems with SAE automation level 3 through to level 5 is provided in this chapter.

2.1. Issues with existing approaches

2.1.1. Safety evaluation through long-distance/ long-duration driving tests

Long-distance/long-duration driving test strategies aim at ensuring safety by randomly indentifying malfunctions and unintended disengagements in a black box-type manner, until a certain value for a probabilistic metric is guaranteed. These strategies, applied as a safety evaluation process, present issues both in terms of 'evaluation scope sufficiency' and of 'explainability in emergencies'.

The main issue related to "evaluation scope sufficiency" relates to the stochastical increase of factors and associated hazards with driving distance and time. In other words, it is not possible to ensure that hazards due to factors not identified in long-distance/long-duration runs will not occur after release.

Further, within a context in which there is neither legal nor social consensus on criteria based on driving distance or time, the issue on "explainability in emergencies" relates to the impossibility of clarifying social responsibility for emergency interventions when hazards are encountered by the system. Probabilistic safety criteria based on long-distance/long-time driving also present problems from a technical development point of view, due to the inefficiency of identifying factors that dependend on the environmental conditions in which the driving was conducted, as well as on the characteristics of the vehicle.

2.1.2. Data storage/classification scenario-based approach

A number of countries are actively developing data driven scenario-based approaches to address the challenges of applying previous ADAS development processes for safety assurance of AD systems of SAE automation level 3 through to level 5. These approaches incorporate normal traffic and accident data, process the data, and systematically categorize the processed information into formats known as 'scenarios' which are stored in a database.

The collection, storage and creation of such scenarios and database in the public domain, free from manufacturers' intellectual property and bias, may enable the development a safety evaluation ecosystem, that both certification bodies and manufacturers could incorporate for the benefit of the general public through safer vehicles.

However, the scenario based approach does not resolve per se the aboved mentioned issue concerning 'evaluation scope sufficiency' before release. When the obtained data is tagged and "categorized", the compensation for the phenomenon that may occur in the future still depends on the distance and time or the amount of data, so the previously mentioned issue related to evaluation scope sufficiency remains unresolved. Further, if the driving data shared in the public domain is only comprised of "images" and "vehicle trajectories" this will lead to insufficient safety verification range, as such data may exclude factors related to autonomous vehicles' misinterpretation of both the surroundings and its own conditions, as well as factors possibly affecting vehicle stability.

2.2. Overview of ‘Physics Principles Approach Process’

In order to address the limitations of existing approaches concerning evaluation scope sufficiency and explainability in emergencies, a ‘Physical Principles Approach Process’ for safety evaluation is proposed. This proposal essentially incorporates physics principles into a scenario-based approach.

The number of safety-relevant situations that an AD system may encounter in real traffic is infinite. Therefore, if scenarios are structuralized by solely combining traffic factors without further considerations, the unlimited number of variables that need to be considered will prevent from a complete scope verification. In contrast with the infinite number of safety-relevant situations that an AD system may encounter in traffic, the number of physics principles that the system can apply for safely handling such situations is limited. AD systems decompose all DDT into perception, judgement and operation subtasks, and each of these subtasks is associated with one or several specific physics principles. Therefore, if scenarios are decomposed and structuralized logically in consideration of the physics of the AD system, then it is possible to provide a complete coverage of all the safety-relevant root causes for given DDT. This motivates the incorporation of perception, traffic situation, and operation related disturbances, and the corresponding scenario structures introduced in the following table, in Figure 1 and Figure 2, and elaborated in detail in following chapters.

Task	Processing results	Disturbance	Governing physics principles
Perception	Own position, surrounding traffic environment positional information and other traffic information	Perception disturbance	Light, radio wave, infrared light propagation principles that affect camera, milli-wave radar and LiDAR sensors, respectively
Judgement	Path, speed plan instructions	Traffic disturbance	Kinematics describing the motion of traffic participants, objects and systems of groups of objects, without reference to the causes of motion.
Operation	Movement instruction allocation for each ACT for achieving path and speed plan instructions	Vehicle control disturbance	Dynamics, concerned with forces applied on the vehicle’s body and tires, and their effects on motion.

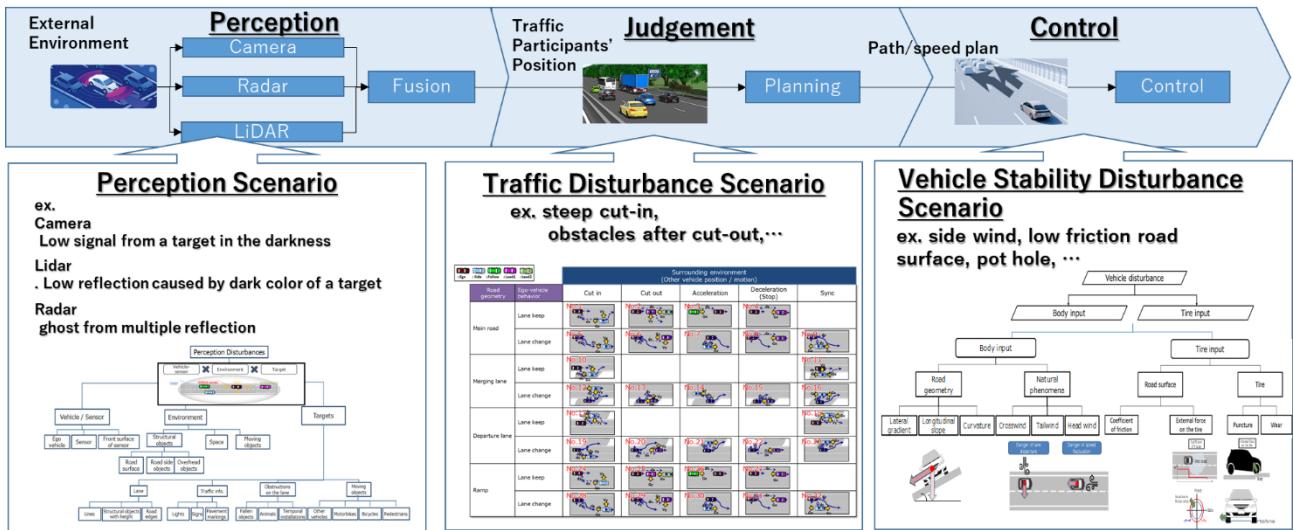


Figure 1. different categories of structuralized scenarios considering physics principles for each corresponding perception, judgement and control tasks

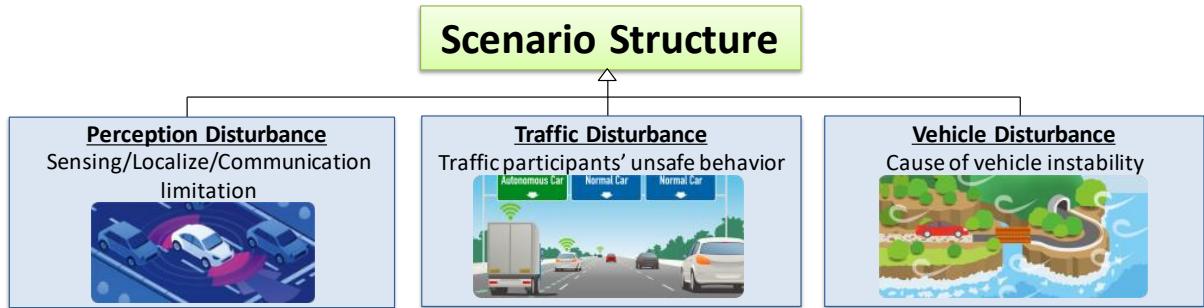


Figure 2. Schematic of the three disturbance categories considered to logically structuralize scenarios

Perception disturbance refers to conditions in which the sensor system may fail to correctly judge a hazard or a non-hazard for sensor or vehicle intrinsic or extrinsic reasons. Examples of intrinsic reasons include part mounting (e.g. unsteadiness related to sensor mounting or manufacturing variability), or vehicle conditions (e.g. vehicle inclination due to uneven loading that modifies sensor orientation, or sensor shielding with external attachments such as bicycle racks). External reasons include environmental conditions (e.g. sensor cloudiness, dirt, light, etc.) or blind spots induced by surrounding vehicles.

Traffic disturbance refers to traffic conditions that may lead to a hazard resultant of a combination of the following factors: road geometry (e.g. branches or ramps in highways), ego-vehicle behaviour (e.g. lane change manoeuvre), and surrounding vehicle location and action (e.g. cut-in from a near side vehicle).

Vehicle disturbance refers to situations in which perception and judgement work correctly but where the subject vehicle may fail to control its own dynamics. This can be due to intrinsic vehicle factors (e.g. total weight, weight distribution, etc.) or extrinsic vehicle factors (e.g. road surface irregularities and inclination, wind, etc.).

Collected normal traffic and accident data can be used to confirm possible gaps in terms of whether situations actually occurring in real traffic are being missed by the logically created scenario systems. Further, by assigning probabilistic ranges to physical parameters for each qualitative scenario category, the data and scenarios can also be used to show in a downscaled manner, to what extent certain situations actually occur.

2.3. Safety Argumentation Structure Framework

2.3.1. Automated driving safety principles

The WP29 document for the harmonisation of international regulations on automated driving reads "Automated vehicles shall not cause any non-tolerable risk, meaning that, under their operational domain, shall not cause any traffic accidents resulting in injury or death that are reasonably foreseeable and preventable" (UN/WP29, 2019, WP29-177-19, Framework document on automated/autonomous vehicles).

These definitions allow to contextualize the safety philosophy of the current methodology proposed, with respect to safety principles that international policy makers are applying in the form of a matrix (Figure 3). Considering the two conditions of foreseeability and preventability together generates a 4 quadrant matrix that better contextualises the philosophy of this document. Scenario based safety evaluation, can be found in the top left quadrant of the matrix where no accidents are acceptable. This quadrant accounts for all scenarios for which an accident is foreseeable and preventable. The bottom left quadrant of the matrix depicts the traffic situations that can not be foreseen but that can be prevented. The cases that fall under this category form the basis for learning and serve as a precedent for future generation AD system developments. The top right quadrant of the matrix introduces cases that are foreseeable but not preventable. The situations that fall under this category are situations for which mitigation is the only option. Measures to reduce the damage resultant of these unpreventable (yet foreseeable) cases constitutes the main area of focus in this section. The final quadrant (bottom right) accounts for crashes that are neither foreseeable nor preventable. In these situations, resilience support in the form of legalities, the division of responsibilities, health support, insurance and other such areas need to be the focus of attention.

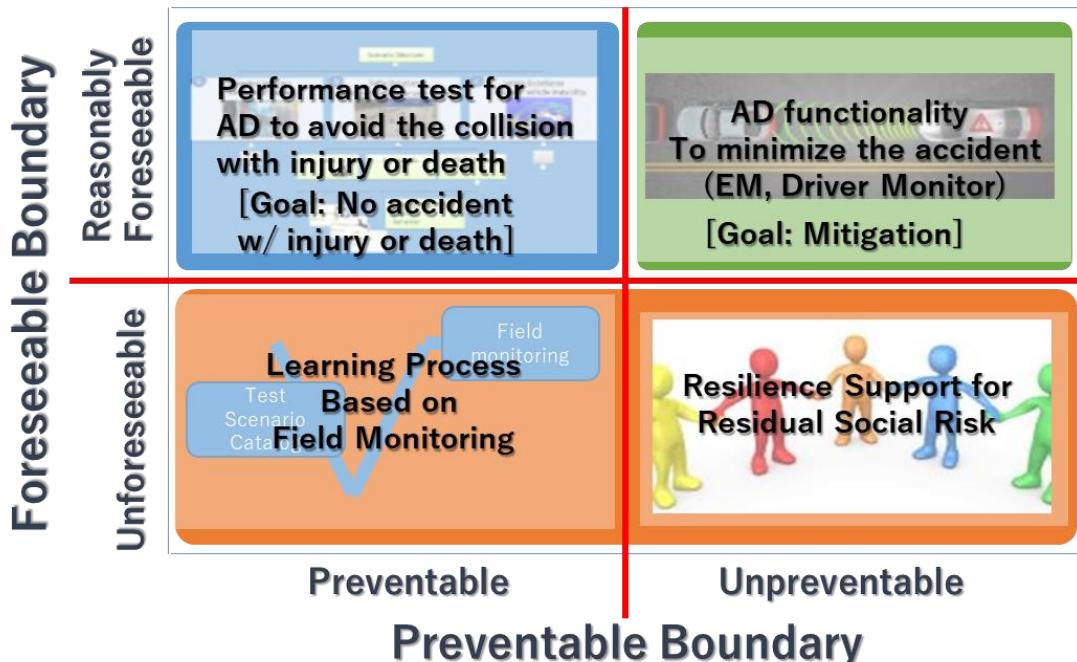


Figure 3. Safety approach in context with foreseeability and preventability matrix

2.3.2. Scope of safety evaluation

Figure 4 presents a summary of the safety aspects described in the WP29 framework document organized hierarchically. With the common top level safety goal of achieving systems free of unreasonable safety risks, the scope of the current proposal is limited to Validation for System Safety (highlighted in pink).

The validation for system safety according to the safety vision framework can be further decomposed as shown in Figure 5. The scope of the current proposal is limited to critical conditions, and excludes ‘Pre critical conditions’. The reason for this exclusion is that, in situations in which there may be a potential risk (e.g. frontal vehicle carrying a load that may fall on the road), may induce many actuations that are not motivated by real risks and that alter traffic imposing risks on other participants (e.g. braking frequently despite not being a real risk). Therefore, to address pre-critical situations, rather than applying physics principles approach processes, other means to verify if the vehicle follows traffic rules and keeps sufficient distance with surrounding objects

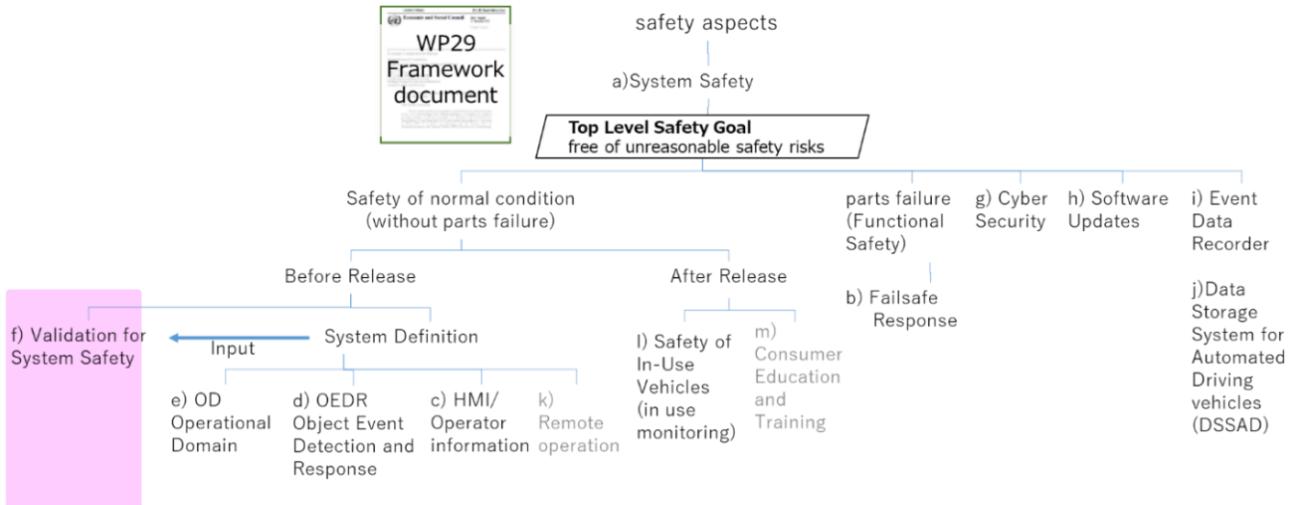


Figure 4. Safety Aspects Hierarchy Diagram

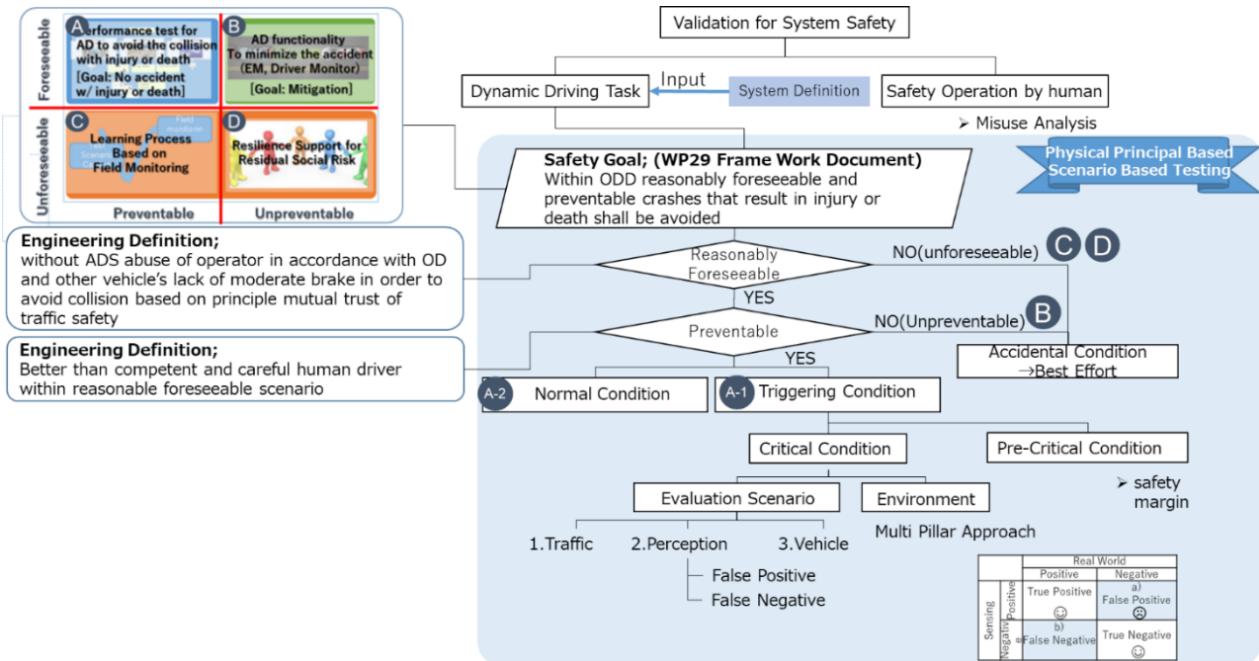


Figure 5. Safety argumentation structure diagram

2.3.3. Method of evaluating safety

The main DDT safety risk is to collision with the surrounding traffic participants or obstacles, which is systematized through traffic disturbance scenarios. By defining quantified ranges of reasonable foreseeability and preventability for each of these traffic disturbance scenarios, quantitative criteria associated to each test are defined. Based on these traffic related hazardous scenarios, it is then possible to expand the evaluation to incorporate perception- and vehicle stability-related hazardous scenarios into the assessment which will enable a comprehensive safety evaluation (Figure 6).

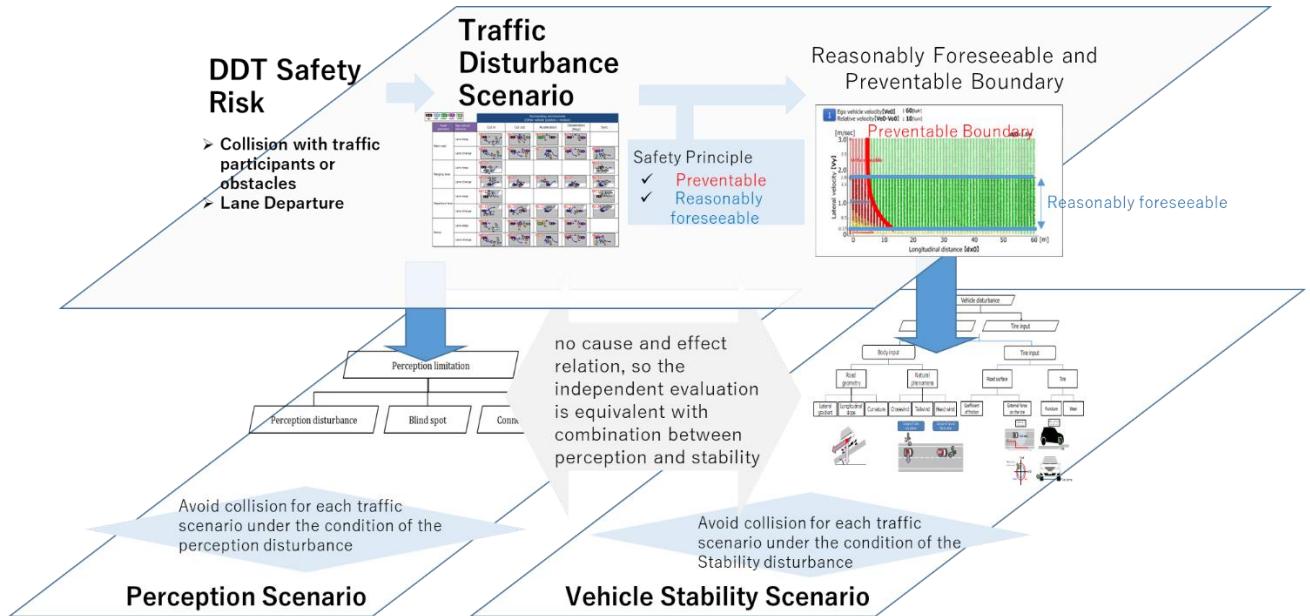


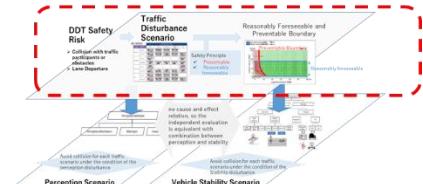
Figure 6. Overview of method of judging safety

2.3.3.1. Traffic disturbance safety evaluation method

Traffic disturbance is the position and actions of traffic participants existing around your own vehicle that prevent safe driving by your own vehicle. As previously described, the basic thinking behind safety principles is ‘to equip the automated driving system with higher level avoidance performance than a competent and careful human driver within a foreseeable range.’ For this thinking, we need to define and model the performance of a competent and careful drive applied to traffic disturbances. By implementing this defined model in a simulation program and deriving the actual scope avoidable for a competent and careful human driver, it is possible to define safety standards in relation to traffic disturbances.

Preventable

ADS collision avoidance performance is equal or better than the performance which a competent and careful human driver can achieve



Reasonably foreseeable

forecastable based on physics principles with a relevant exposure and ego-vehicle driver's / other driver's extreme violation of traffic rules.

Surrounding Traffic Participants Category	Ego Vehicle Behavior Category	L/C	L/K			no lane departure for each road category
		All	Constant	Deceleration/Stopped	Cut-In	
Vehicle	No collision as a precondition to legal behavior of surrounding traffic participants			ALKS Annex4 Appendix3	ALKS Annex4 Appendix3	ALKS Annex4 Appendix3
Vulnerable Road Users	e.g) 0.3G deceleration at less than 2.0 THW with 1.4 delay			Can be covered by cut-out scenario	ALKS 5.2.5.3 The activated system shall avoid a collision with an unobstructed crossing pedestrian in front of the vehicle. In the scenario with an unobstructed pedestrian crossing with a lateral speed component of not more than 5 m/h where the anticipated impact point is displaced by not more than 0.2 m compared to the trajectory of the vehicle, the activated ALKS shall avoid a collision up to the maximum lateral displacement of the impact point. Equivalent with ALKS 5.2.5.3	ALKS Annex4 Appendix3
Animal/ Fallen Object					ALKS Annex3 Appendix3	ALKS Annex3 Appendix3

Figure 7. Overview of traffic disturbance safety judgement method

The competent and careful human driver performance model definition (Figure 8) is able to define the three elements of ‘perception’, ‘judgement’, and ‘operation.’ It is important to have objective grounds for defining parameter coefficients related to performance shown in the respective segments.

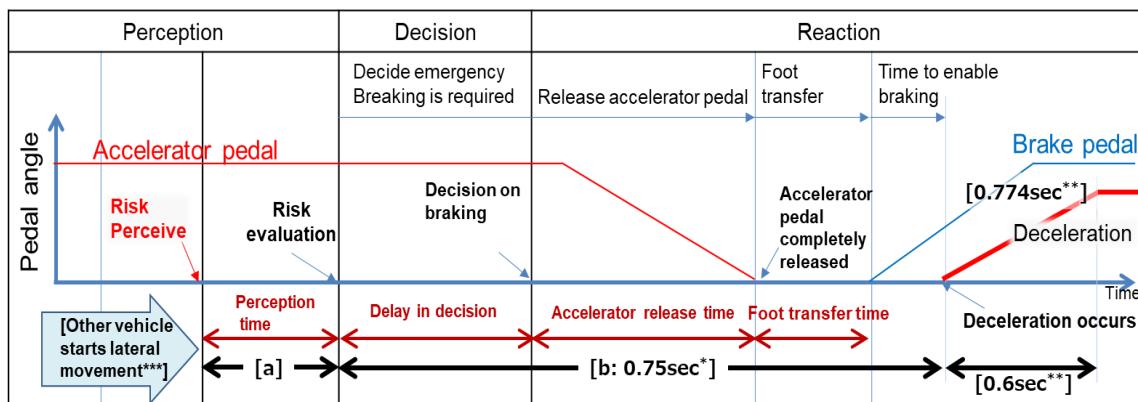


Figure 8. Competent and careful human driver model

Here, the driving action elements of ‘judgement’ and ‘operation.’ are explained. The main avoidance actions of automatic driving in relation to traffic disturbances are considered to be the brake operation (deceleration action) and, regardless of the type of traffic disturbance (position and action of the traffic participants surrounding the ego vehicle), this is fulfilled by defining the performance of a competent and careful human driver. Figure 9

shows a diagram which demonstrates the brake operation of a competent and careful human driver. The model on the left shows the braking operation made by a competent and careful human driver. The model on the right is a functional model of the collision damage mitigation braking system (AEB: Advanced Emergency Braking), it considers the amount of improvement in avoidance performance when equipped with AEB.

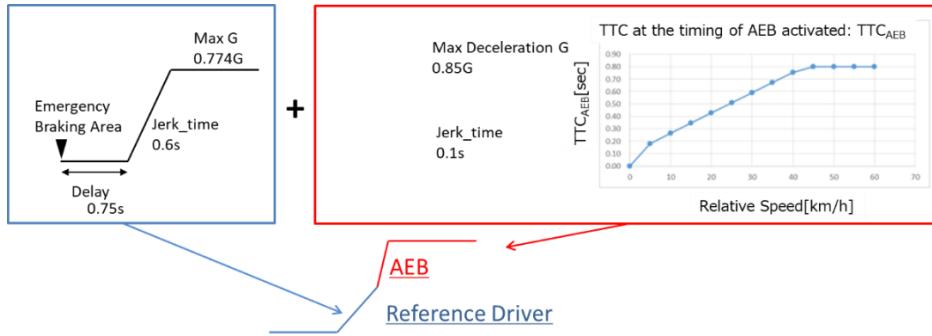


Figure 9. Competent and careful human driver brake model

Perception response time, the time delay from the moment when a competent and careful human driver perceives risk to the time that deceleration braking force occurs is set at 0.75 s. This time set is used by police and domestic courts in Japan when establishing a driver's "perception response time".

In terms of maximum deceleration force, quoting the Japanese test data shown in Figure 10, is 0.774G. Whereas the brake force generated by normal drivers in emergencies is 0.689G, normal drivers who have received training in driving techniques have a braking force of 0.774G; albeit this is defined as a higher skill value compared to ordinary drivers.

Furthermore, from the accident statistics data from NHTSA (Figure 11), 0.74G is the peak value; therefore, the maximum deceleration of 0.774G applied to the competent and careful human driver model can be considered appropriate.

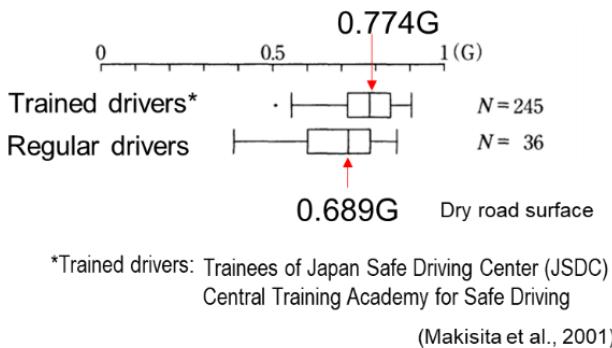


Figure 10.Emergency brake characteristic

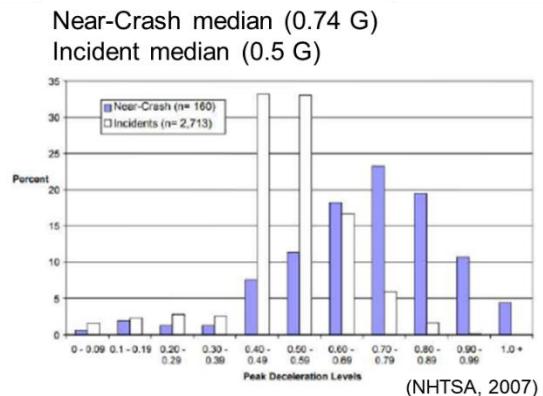


Figure 11Maximum deceleration due to deceleration of the preceding vehicle

Figure 12 shows a waveform diagram of deceleration braking for drivers who have received driver skill training. This quotes the Japanese test data previously described. In this waveform diagram, the time for reaching the maximum deceleration is demonstrated, and the maximum deceleration arrival of a competent and careful human driver is defined as 0.6 s.

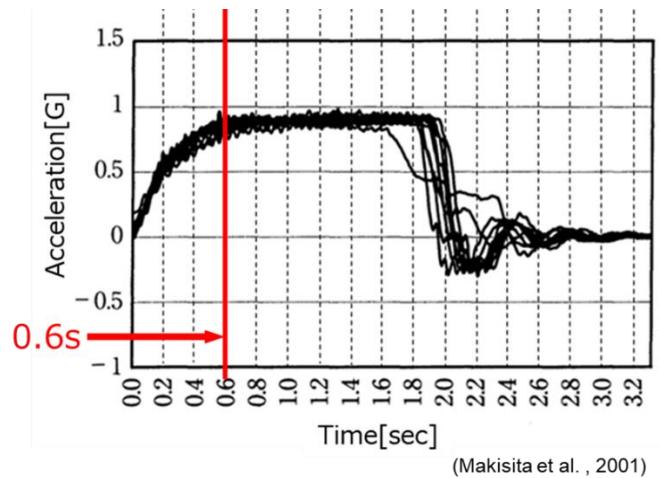


Figure 12. Emergency brake characteristics study example (arrival time until maximum deceleration)

2.3.3.1.1. Cut-in scenarios

Cut-in scenarios are scenarios in which vehicles travelling in an adjacent lane to the ego vehicle cuts in front of it. Figure 13 shows a schematic expressing boundary conditions where a competent and careful human driver judges it risky when another vehicle cuts in in front of the ego vehicle.

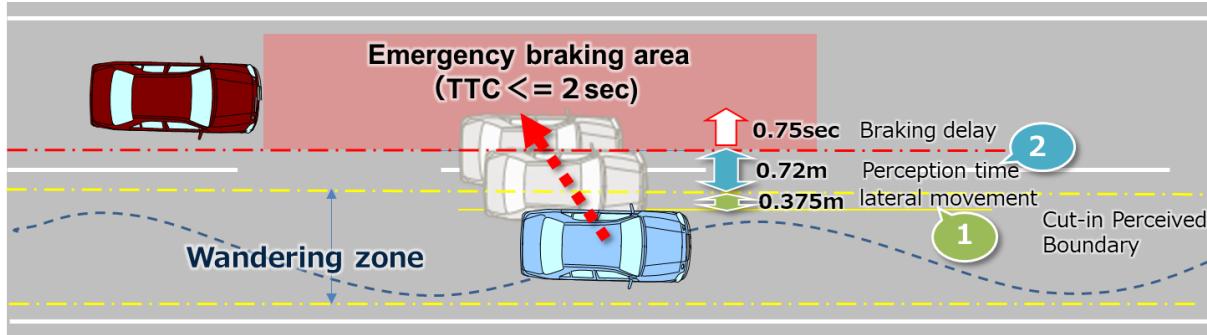


Figure 13. Cut-in judgement conditions and danger judgement boundaries

The boundary conditions when it is judged that a vehicle travelling in the adjacent lane has cut in front of the ego vehicle are defined as the cut-in vehicle lateral movement distance (wander amplitude). In an actual driving environment, vehicles driving while maintaining their lane will wander a little to the left or right while driving. In the scope of the wander lateral movement distance, it is unlikely that the vehicle traveling in the adjacent lane of the ego vehicle travels with a recognition that it will cut in. Therefore, the cut-in perception boundary conditions were defined from the lateral distance movement (wander amplitude) distribution (Figure 14) of vehicles changing lanes based on the data observed in the actual traffic environments.

After the cut-in judgment, the boundary conditions for perceiving risk for the ego vehicle and perceivess a need for the emergency brake (riskperception boundaries) can be defined by multiplying the maximum lateral velocity derived from the actual traffic observation data by the risk perception response time.

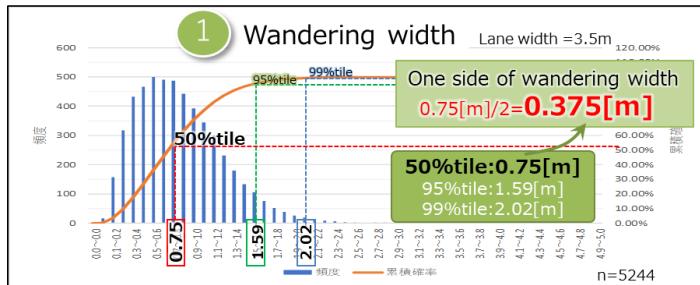


Figure 14. Actual observation statistics for ‘stagger amplitude’

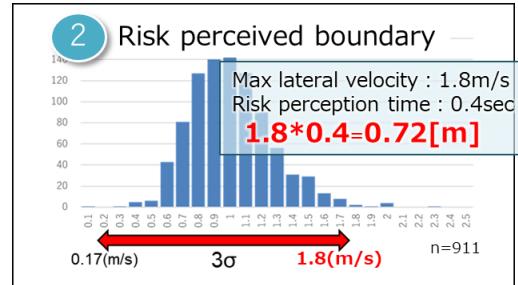


Figure 15. ‘Maximum lateral velocity’ observation data statistics

When calculating the ‘risk perception response time’, test data using a driving simulator carried out in Japan was utilised and analysed. The prerequisites for the test are shown in Figure 16.



Parameter	Value
Lane width	3.5 m
Ego-vehicle target velocity V_e	100 km/h
Platoon velocity traveling in parallel forward V_o	70 km/h
Max. lateral velocity of cut-in vehicle V_{oL}	1.8 m/s
TTC at cut-in start	3.0 s

Figure 16. Assumptions for driving simulator tests

The tests measured the driver's response (reaction time, avoidance operation) for cut-ins from 20 other regular drivers (Table 1). The measurements were performed twice on each participant; by comparing the respective average values of the first and second time, we derived the time until risk was perceived.

Table 1. Test participant attributes

Group	No. of participants	Description	Composition of participants
Expert Driver	11	Having 5 years or more driving experience on regular basis, drives on highway at least once a month	- 6 Males, 5 Females - Avg. age: 38.7
Beginner Driver	9	Having 5 years or less driving experience on regular basis, drives on highway not more than once a year	- 6 Males, 3 Females - Avg. age: 23.1

The test results are shown in Figure 17. The results demonstrated that the time from the start of the cut-in from the other driver to when risk was perceived was ~0.8 s for the first time and 0.4 s for the second time. Based on these test results, with the first time perception, the cut-in time is required by the other driver and the time for risk to be perceived, whereas the second time because they were driving while being wary of the cut-in, the time for identifying the cut-in from the other vehicle was not required. However, even when the driver was aware, time was still required for determining risk (Figure 18), and the 'time until risk was perceived' was defined as 0.4 s.

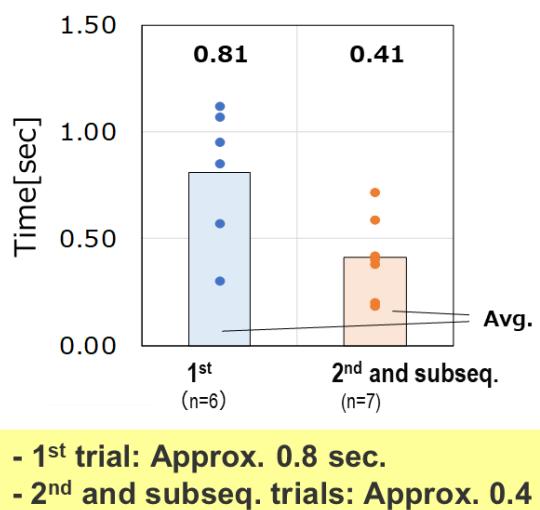


Figure 17. Driving simulator test results

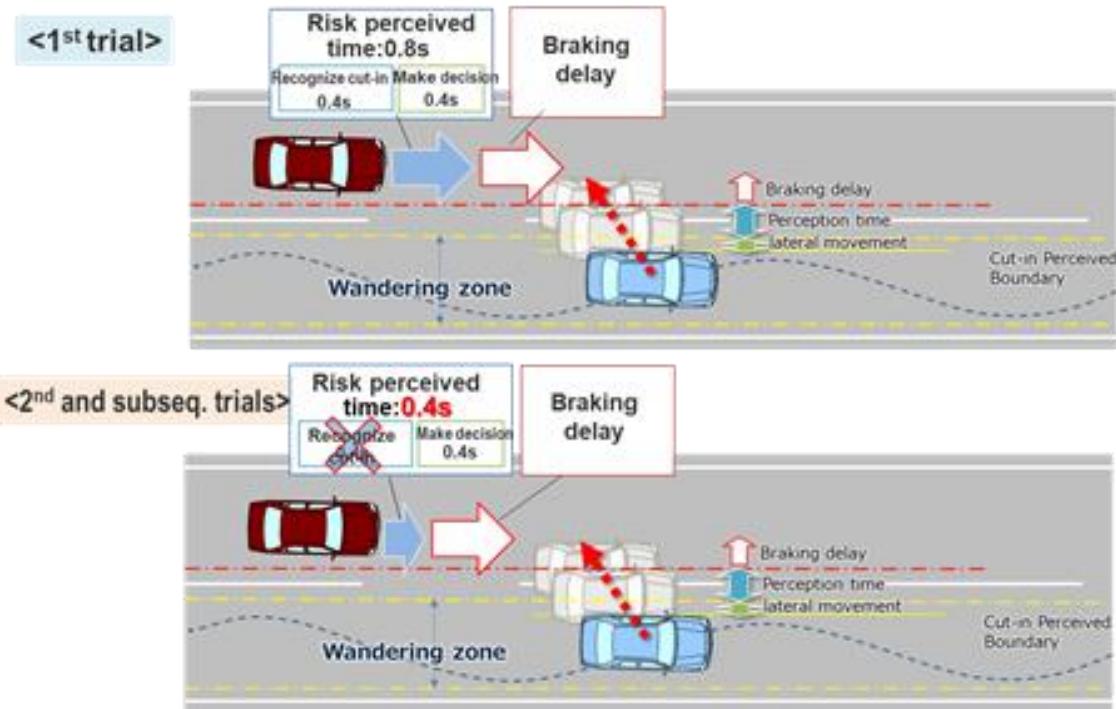


Figure 18. Relationship between cut-in identification time and danger judgement time

As described above, the risk judgement boundary is defined as the time when multiplying the maximum lateral velocity, and the time until perceiving risk. The maximum lateral velocity of 1.8 m/s calculated from the actual traffic observation data and the time until risk is perceived and calculated from the driving simulator test results of 0.4 s are multiplied. Therefore, the risk perception boundary is defined as $1.8 \times 0.4 = 0.72$ m.

When the cut-in perception condition and risk evaluation boundary area applied to the diagram in Figure 8, it results in Figure 19.

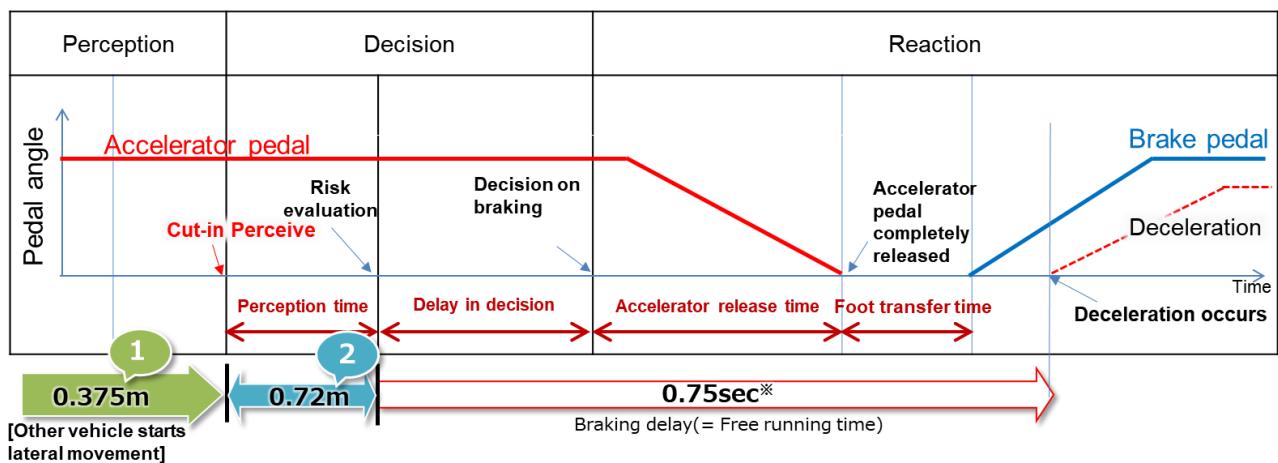


Figure 19. Competent and careful human driver model (Cut In)

According to the UNR collision warning guidelines, the boundary that requires emergency action is defined as $TTC^* = 2.0$ s regarding the longitudinal (distance from the other vehicle) risk evaluation boundary (Figure 2). This is cited to define the longitudinal risk evaluation boundary as $TTC = 2.0$ s.

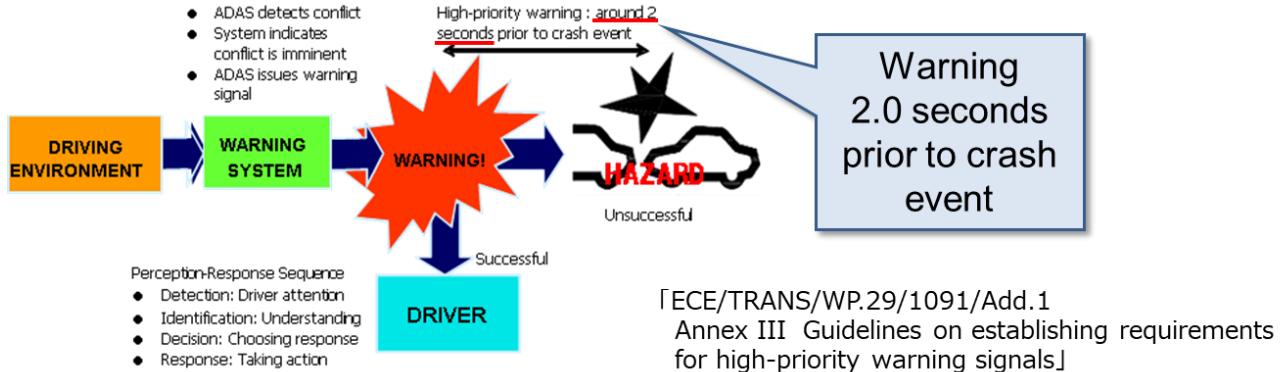


Figure 20. UNR collision warning guidelines (Citation)

2.3.3.1.2. Cut-out Scenario

The cut-out scenario is a scenario in which the leading vehicle that the ego vehicle is following suddenly changes its lane to the adjacent lane (cut-out). This scenario evaluates safety in relation to the sudden appearance of a decelerating or stopped vehicle (such as broken-down car and the tail end of a traffic jam) in front of the ego vehicle due to the preceding vehicle's cut-out. Figure 21 shows the schematic that represents the boundary condition for the competent and careful human driver who perceives the situation to be risky when the preceding vehicle performs a cut-out.

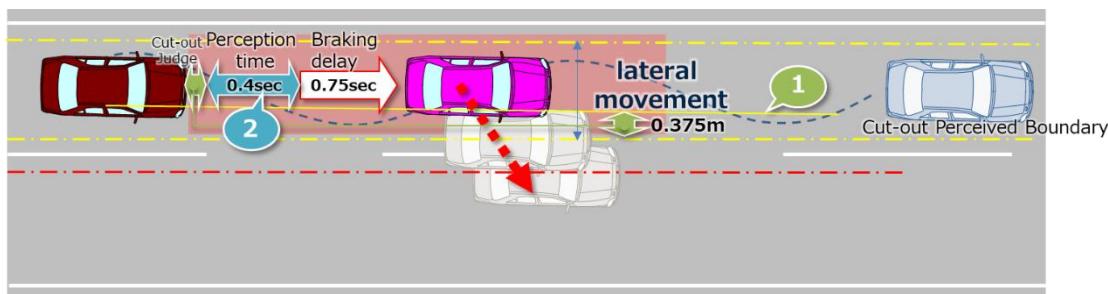


Figure 21. Cut-out perception condition and risk evaluation boundary

The cut-out perceived boundary condition to perceiving the preceding vehicle's cut-out manoeuvre is defined by the amount of lateral movement (drifting amplitude), which is similar to the case with the aforementioned cut-in perception condition. Both the cut-in and cut-out are manoeuvres to change lanes. Similar to the case of cut-in, the boundary condition using the distribution of drifting amplitude from the observation data of real traffic is applied to the perception condition of cut-out.

Moreover, the time from the cut out perception to the recognition of the vehicle ahead that appears and the risk perception is defined as 0.4 sec based on the experimental data (Figure 17 and 18).

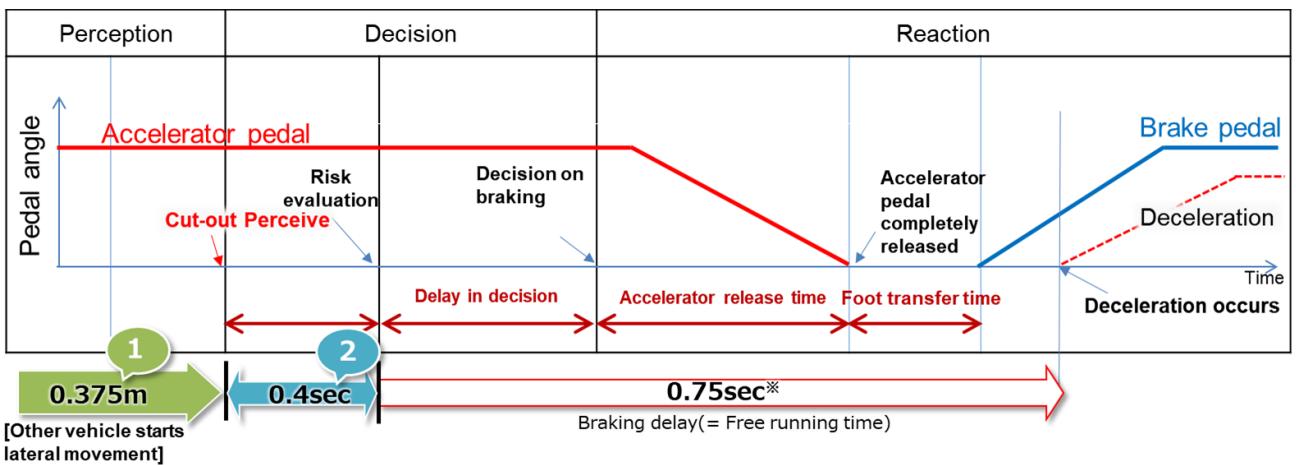


Figure 22. Competent and careful human driver model (cut out)

2.3.3.1.3. Deceleration Scenario

A deceleration scenario takes into consideration the sudden deceleration of the leading vehicle that the ego vehicle is following. Although the previous cut-in and cut-out scenarios required the perceived lane change boundaries from the following or leading vehicle, the deceleration scenario only involves the longitudinal behaviour. Therefore, it is only necessary to define the deceleration perception time by the leading vehicle to evaluate the risk boundary. Similar to the preceding case, 0.4 s can be applied as the time required to evaluate the risk.

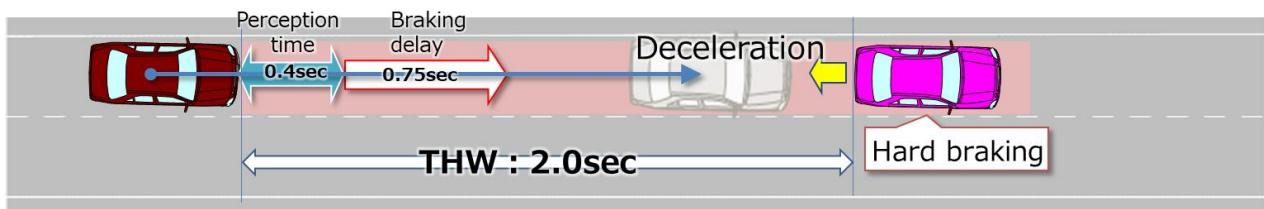


Figure 23. Risk evaluation boundary in deceleration scenario

When the risk evaluation condition of the deceleration scenario is applied to the diagram in Figure 8, it results in Figure 24.

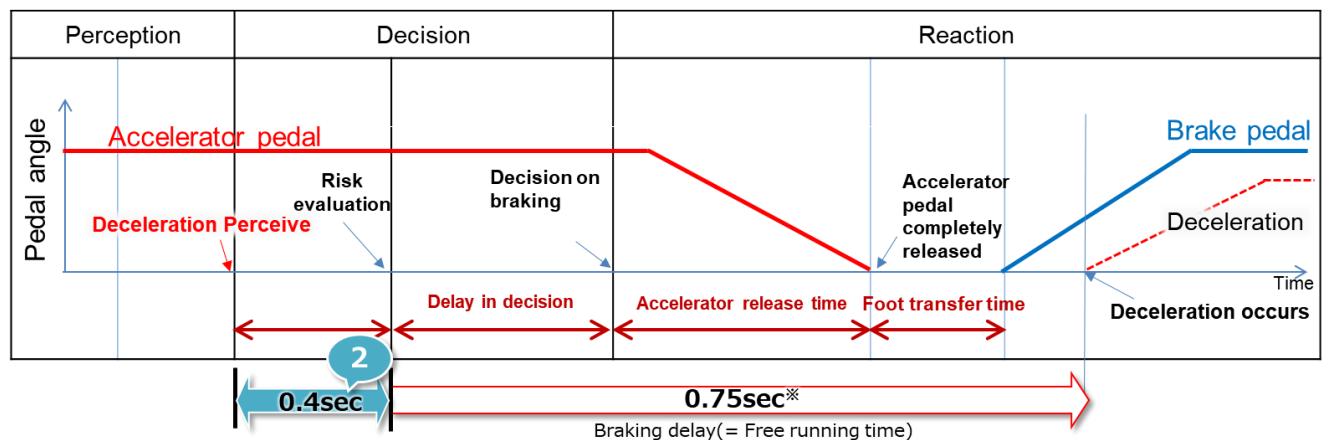


Figure 24. Competent and careful human driver model (Deceleration)

Definition of Parameters for Deriving Standard

The following table lists the parameters required for deriving the safety standards for traffic disturbances. The evaluation scenarios related to traffic disturbances are generated by defining road geometry, the ego vehicle's behaviour, and locations and motions of the surrounding traffic participants. The parameter items required in the evaluation scenario are categorized in a specific numerical range, and the Pass / Fail boundary is derived within that range.

Table 2. List of traffic disturbance parameters.

Operating conditions	Roadway	#of lanes = The number of parallel and adjacent lanes in the same direction of travel Lane Width = The width of each lane
Initial condition	Initial velocity	V_{e0} = Ego vehicle V_{o0} = Leading vehicle in lane or in adjacent lane V_{f0} = Vehicle in front of leading vehicle in lane
		dx_0 = Distance in longitudinal direction between the front end of the ego vehicle and the rear end of the leading vehicle in ego vehicle's lane or in adjacent lane
		dy_0 = Inside Lateral distance between outside edge line of ego vehicle in parallel to the vehicle's median longitudinal plane within lanes and outside edge line of leading vehicle in parallel to the vehicle's median longitudinal plane in adjacent lines.
	Initial distance	dy_{0_f} = Inside Lateral distance between outside edge line of leading vehicle in parallel to the vehicle's median longitudinal plane within lanes and outside edge line of vehicle in front of the leading vehicle in parallel to the vehicle's median longitudinal plane in adjacent lines.
		dx_{0_f} = Distance in longitudinal direction between front end of leading vehicle and rear end of vehicle in front of leading vehicle
		dfy = Width of vehicle in front of leading vehicle
		doy = Width of leading vehicle
		dox = Length of the leading vehicle
	Lateral motion	V_y = Leading vehicle lateral velocity
	Deceleration	G_x_{max} = Maximum deceleration of the leading vehicle in G
		dG/dt = Deceleration rate (Jerk) of the leading vehicle

2.3.3.1.4. Calculation of Boundary

As discussed above, the specific standard value can be derived by the numerical calculation of the competent and careful human driver model. The parameter region for the standard value derivations are set to allow combinations of every parameter within the maximum vehicle velocity region allowed by the ADS to be targeted.

2.3.3.1.4.1. Derivation result of the preventable boundary of cut-in scenario

The safety standard of the cut-in is derived for every relative velocity between the ego vehicle and the counter vehicle. Collision with the cut-in vehicle is not allowed in the parameter region indicated by the green area in Figure 26.

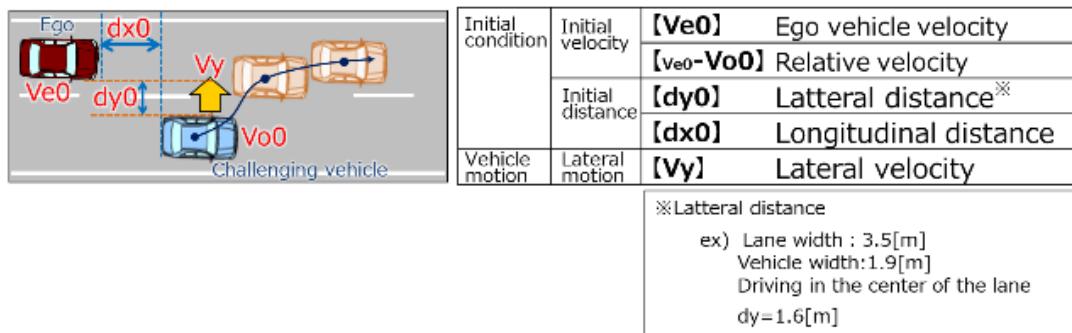


Figure 25. Conceptual diagram of cut-in scenario parameters

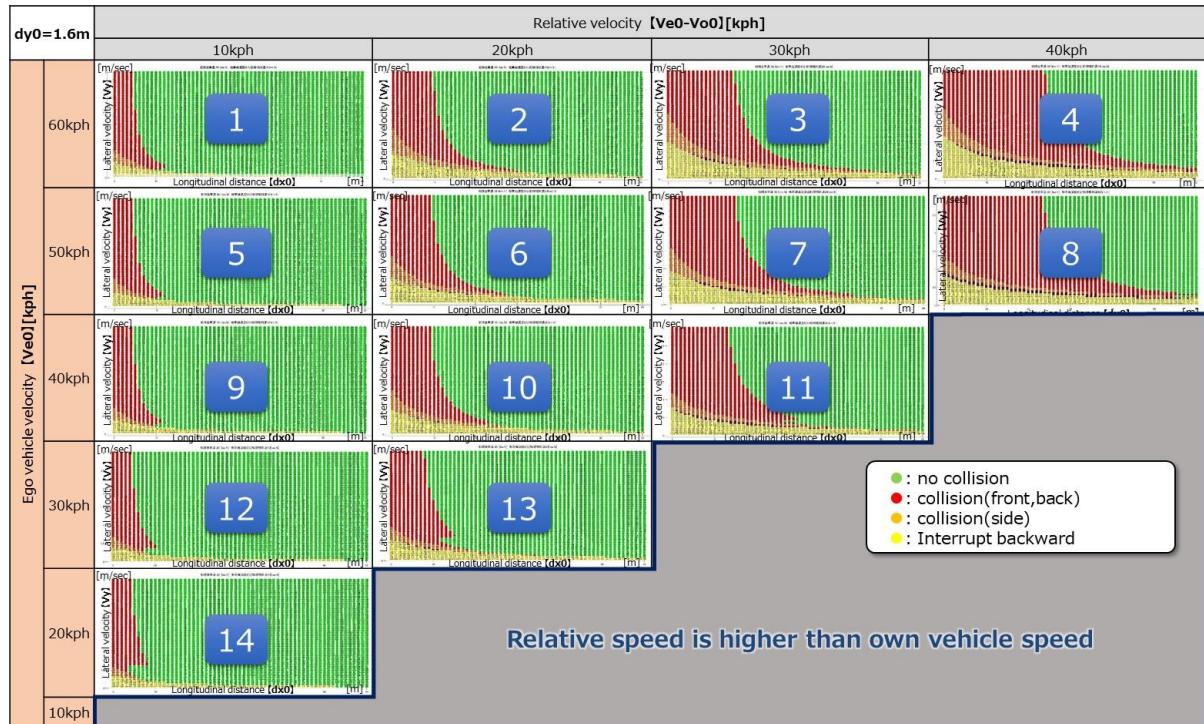


Figure 26. Preventable boundary data sheet of cut-in scenario

2.3.3.1.4.2. Derivation result of cut-out scenario standard

The cut-out safety standard requires that all decelerating (stopped), vehicles located ahead of the vehicle cut-out, must be able to avoid collisions. This standard is derived by making the aforementioned competent and careful human driver model follow the leading vehicle at THW = 2.0 s. This value, i.e., THW=2.0 s, is applied by referring to the laws and instructions of each country.

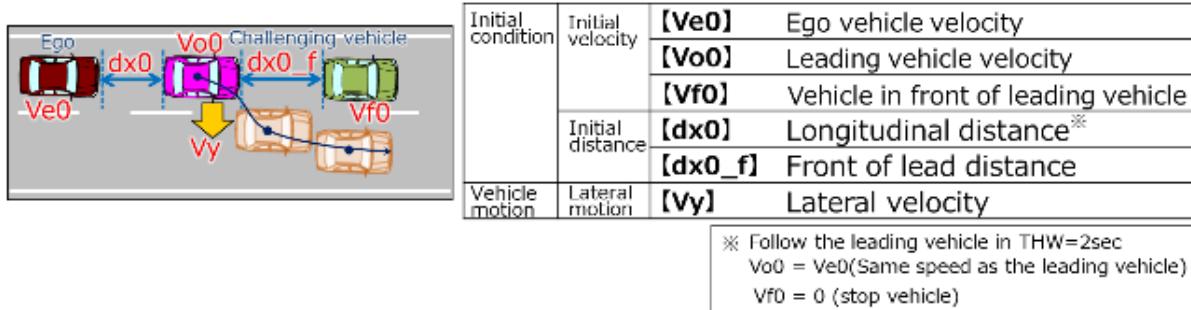


Figure 27. Conceptual diagram of cut-out scenario parameters

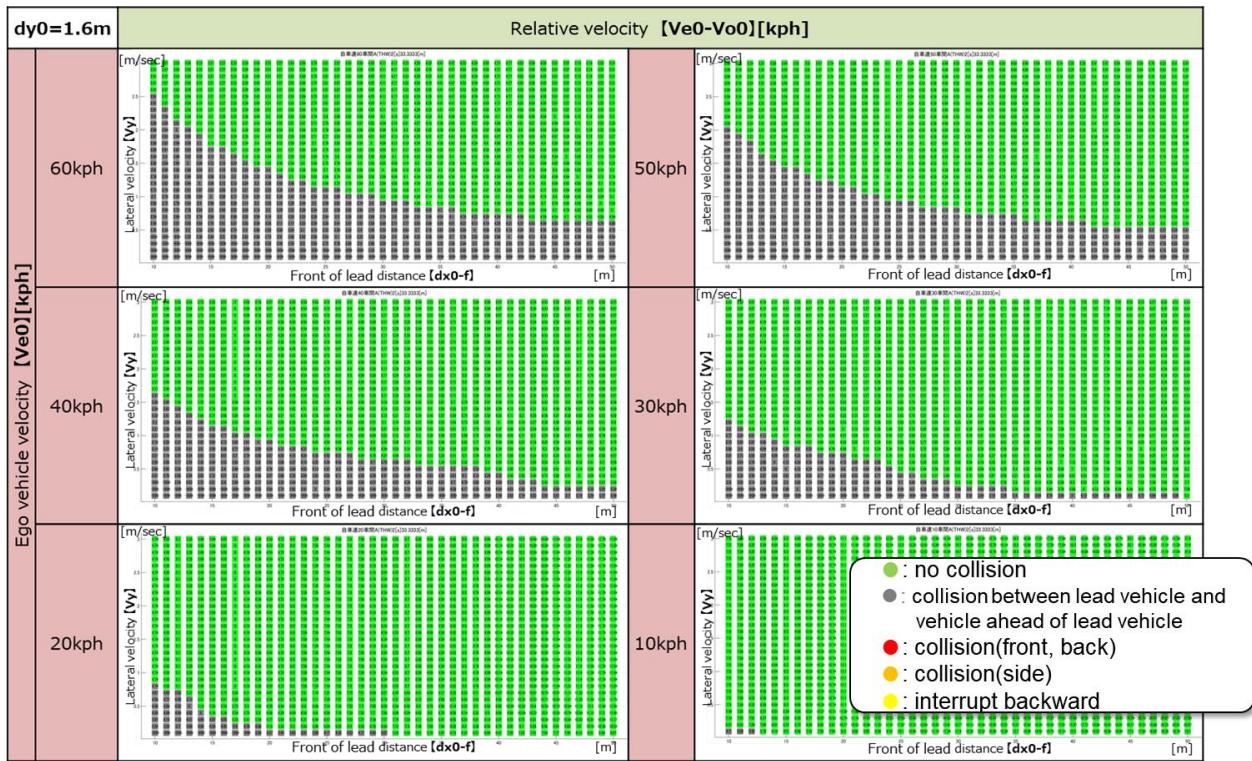


Figure 28. Preventable boundary data sheet of cut-out scenario

2.3.3.1.4.3. Derivation result of preventable boundary of deceleration scenario

The safety standards for deceleration scenarios are required to enable avoidance of collision with the suddenly decelerating vehicle at -1.0 G or less or by stopping the vehicle. This standard is derived by making the aforementioned competent and careful human driver model follow the leading vehicle at $\text{THW} = 2.0\text{ s}$. This value, $\text{THW} = 2.0\text{ s}$, is applied by referring to the laws and instructions of each country.

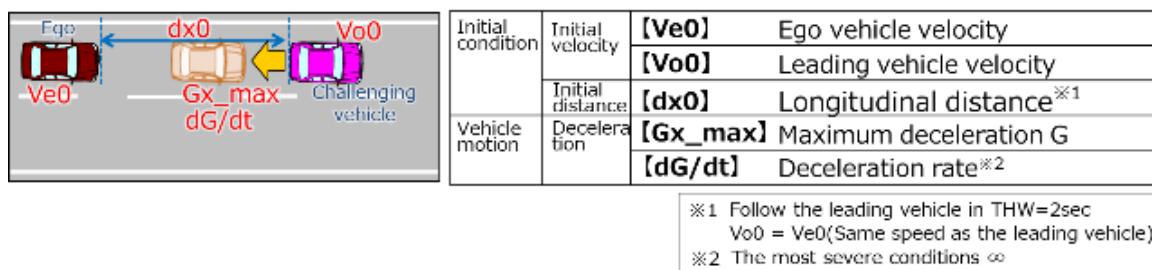


Figure 29. Conceptual diagram of decelerating scenario parameters

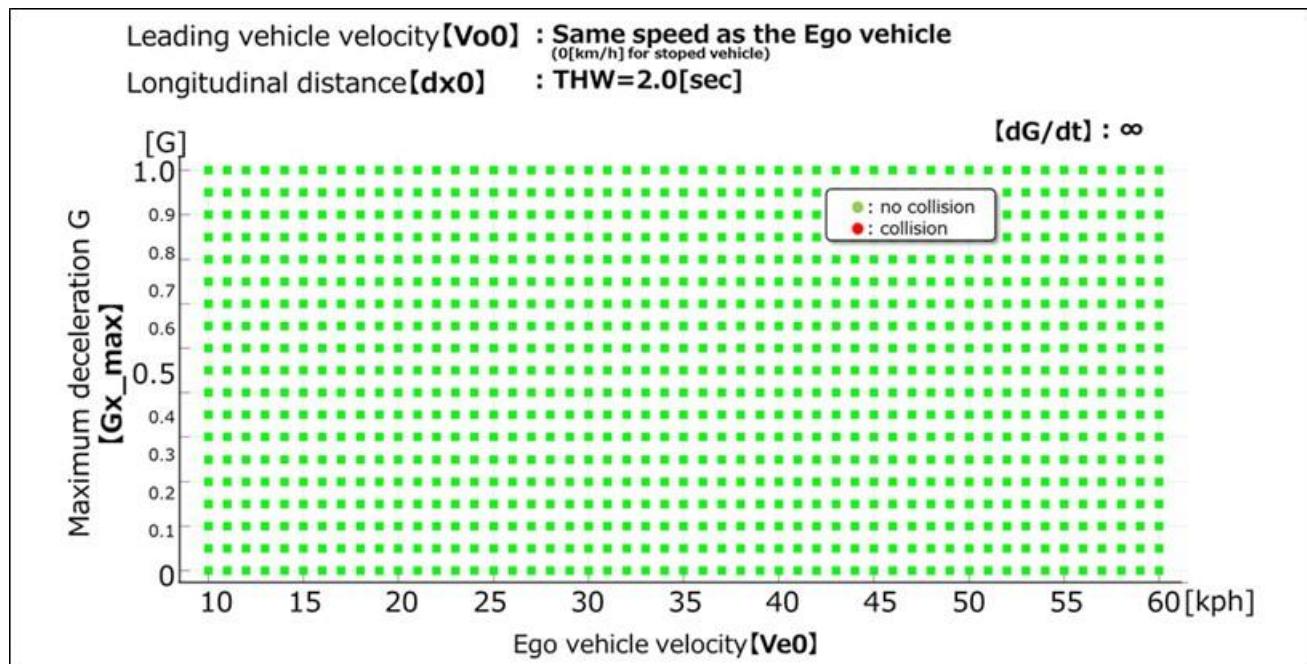


Figure 30. Preventable boundary data sheet of decelerating scenario

2.3.4. Safety evaluation method for perception disturbance

The basic conception of safety standard is as follows: ‘To avoid collisions in any of the traffic disturbance scenarios, even when experiencing perception disturbances.’

When considering that lane deviation can also contribute to collisions, the perception of objects is necessary to avoid collisions with objects on the runway (Fig. 31). Moreover, there are two types of phenomena that result from the perception disturbance, namely, a false negative where the existing objects are not correctly detected, and a false positive where objects that do not exist are falsely detected (Figure32).

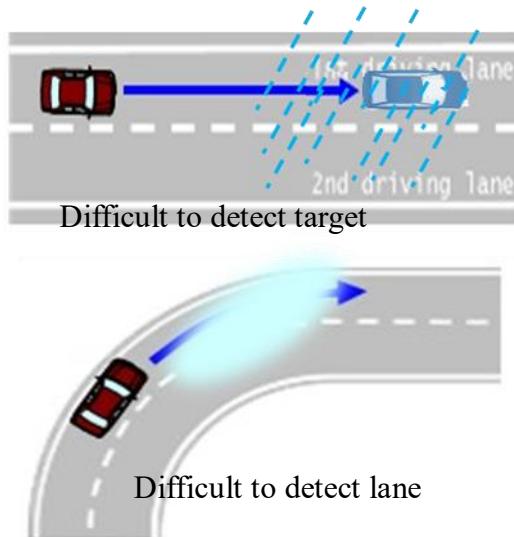


Figure 31. Types of detection target

		Real World	
		Positive =存在する	Negative =存在しない
Sensing	Positive =いると判断する	True Positive =検知成功 😊	False Positive =ゴースト (いないものを見ると判断してしまう) →誤検知 😢
	Negative =いないと判断する	b) False Negative =検知失敗 →見逃し、不検知 😢	True Negative =何もいないことを正確に検知 😊

Figure 32. Detection result caused by disturbance

When these are combined, evaluations based on the concept of safety standards become necessary for four categories of situations in total (Figure 33).

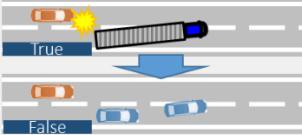
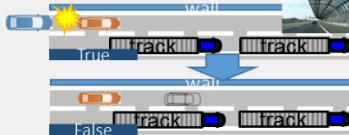
2 kinds of perception target	2 types of perception result with disturbance	
	false negative (不検知)	false positive (誤検知)
Road object	No collision on traffic disturbance because of false negative 	No collision on traffic disturbance because of false positive 
Lane	No lane departure because of false negative 	No lane departure because of false positive 

Figure 33. Four categories of detection disturbance situation

The following is considered within the ODD region as the parameter region of perception disturbance to define an appropriate region for each disturbance factor.

- 1: Road structure, Road Traffic Law and other regions defined by laws and regulations.
(e.g.: When visibility is 50 m or less, the road is closed, i.e., a level difference of >15 cm on the road surface must be repaired)
- 2: Region that is determined to be possible at certain probability based on statistical data.
(e.g., precipitation, brightness, and sun altitude, etc)

Moreover, this safety standard is not the performance standard allocated to an individual sensor. Instead, it should complement the entire recognition system installed. The above flow of safety perception can be summarized as follows.

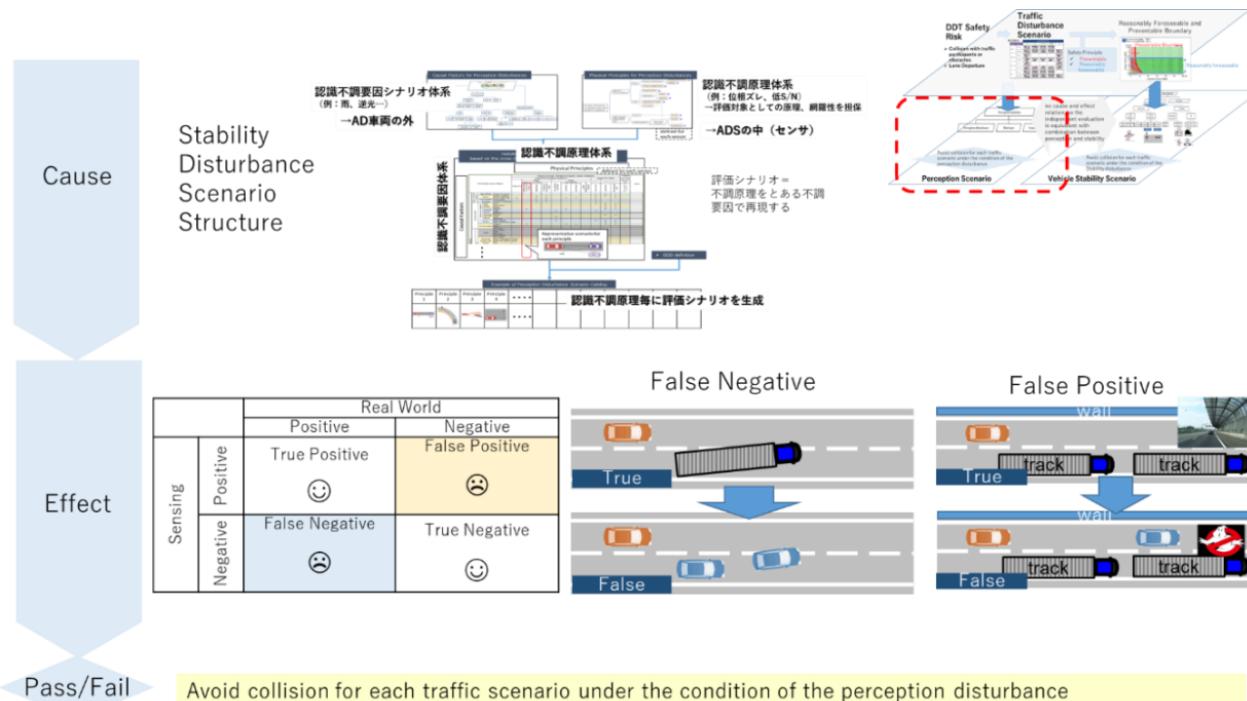


Figure 34. Safety assessment of perception disturbance detection flow

2.3.5. Safety evaluation method for vehicle disturbance

A vehicle disturbance indicates sudden disturbances (e.g. puddles or sudden gust of wind). Although these are unpredictable phenomena, drivers can safely drive by following common sense related to road design, road maintenance/management and road environmental conditions. Thus, the premise of driving on common roads is that the roads are constructed by responsible public or private organisations which follow basic principles such as legality, ethics and engineering and are always maintained and managed. Most countries have road structure ordinances and guidelines for road maintenance and repair to ensure that the road geometry design enables safe driving by every person with a valid driving license (regardless of their driving skill, reflexes, or age). Moreover, when there is a risky situation, such as freezing or a sinkhole, that can hinder driving, the road administrator is obliged to warn the drivers in advance, e.g., with a traffic sign. Based on these preconditions, a technical safety approach for foreseeable vehicle disturbances is introduced.

As shown in Figure 6, ‘collisions must be avoided in any of the traffic disturbance scenarios, even when experiencing vehicle disturbance.’ In the current standards, the collision avoidance strategy under the foreseeable and avoidable scenarios and collision mitigation strategies for predictable but unavoidable scenarios are of particular consideration. Henceforth, when a vehicle behaviour changes because of a vehicle disturbance within the scope of avoidable conditions, the AD vehicle is required to possess a controllability that can stabilise the vehicle without halting driving. However, when these disturbances cause instability that cannot be avoided, the AD vehicle must adapt to the ‘best effort’ strategy to mitigate the possible collision.

Figure 35 shows a specific example of the safety approach for foreseeable vehicle disturbances. The upper section of the figure represents an example of the AD vehicle experiencing a rapid decrease of sliding friction while staying within the avoidable conditions on a wet road; in such a state, the vehicle must be able to be safely controlled without interrupting the driving process. However, the lower section of the figure represents an example involving an AD vehicle equipped with summer tires encountering a frozen road, which causes a rapid decrease of sliding friction and generates a vehicle state that was defined to be unavoidable in advance (e.g., maximum deceleration). Therefore, the safety approach toward vehicle disturbances is based on the principle and clear definitions of vehicle motion engineering related to the definitions of the states where the vehicle is controllable and the states where the vehicle is uncontrollable. (Section 3.3.3 for detail).

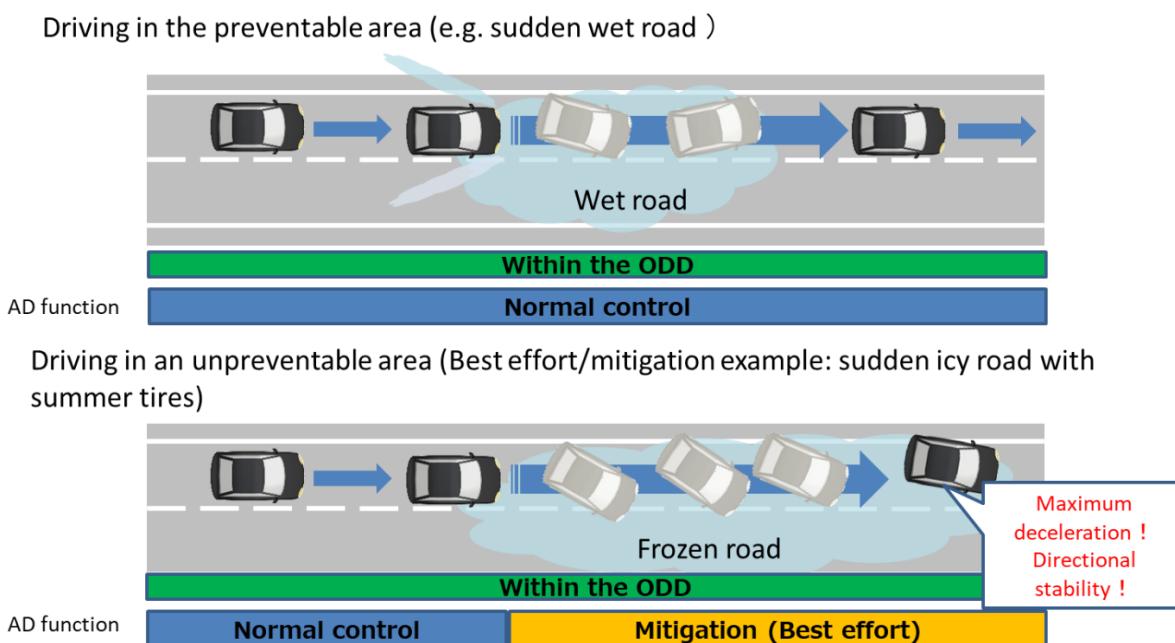


Figure 35. Safety approach for avoidable (above) and unavoidable (below) vehicle disturbance

When these considerations are combined with traffic disturbances, the safety of the AD vehicle does not affect the test result if the stability of the vehicle is maintained. Moreover, while wind affects other vehicles, it only influences the lateral velocity as with cut-in, and it is included in the original traffic flow parameters. The safety standards for vehicle motion disturbances are evaluated relatively without including the vehicle disturbance to the traffic flow scenario. Therefore, the safety standards for vehicle disturbances only need to set the most strict condition under the premise that the Road Traffic Act is strictly adhered. Drivers are responsible for the maintenance of their vehicles, the road administrator is appointed as per the Road Traffic Act, and roads are managed and operated according to the Road Structure Ordinance and guidelines for road maintenance and repair, and perception standards ‘do not departing from the road surface.’ The disturbance factors and conditions are as follows:

- Road surface state: Friction coefficient is 0.3 (lock μ) or more, external force on the tires is at the set point of the road maintenance and repair or less (e.g.: rut: 25 mm, level difference: 30 mm, pothole: 20 cm)
- Road geometry: Curve within the regulation of the road structure ordinance, i.e., $R = 460$ m, vehicle velocity is 100 km/h
- Natural phenomena: Wind speed of lateral wind without speed control is <10 m/s, i.e., vehicle velocity is 100 km/h

As the most difficult condition here is when the abovementioned disturbances all simultaneously occur, these three factors are added up for evaluation (Figure 36).

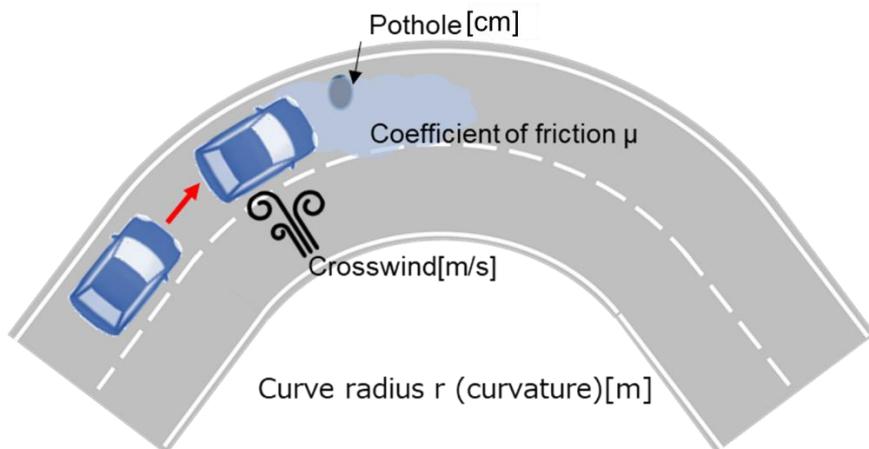


Figure 36. Vehicle motion disturbance evaluation conditions

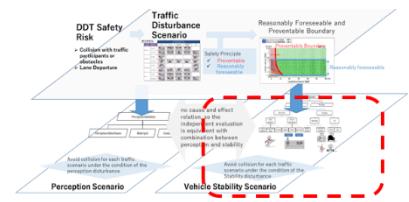
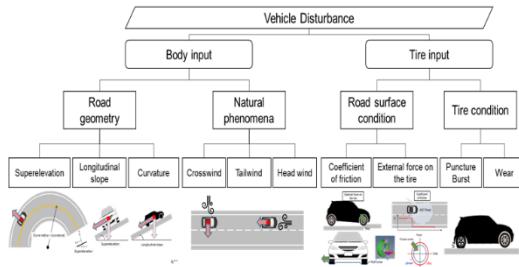
The perception condition under this situation is to avoid departure from the lane. Here, the cases where the vehicle cannot drive under these conditions (e.g., when lateral wind is 5 m/s or more, i.e., driving is not possible) must be defined in advance as ODD by the manufacturer.

Moreover, as a functional requirement, the slow puncture that occurs during driving must be detected before the rim touches the road surface.

The summary of the flow of safety perception discussed to date is listed below.

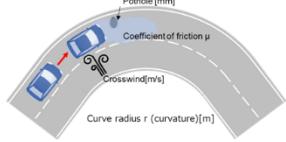
Cause

Stability Disturbance Scenario Structure



Effect

		Safety Risks	
		collision with Surrounding object	lane departure
Body input	Road Geometry	Curvature	N/A
	Natural phenomena	Crosswind	Reason Curvature: 他車との位置関係は変わらないため Crosswind: 自車は誘導経路を維持、他車は横速度が変化するだけであるため Coefficient of friction: 自車は誘導経路、制動時はブレーキ性能（ABS,EBA）に関わるため除外 External force: 自車は誘導経路を維持する 従って、交通流と組み合わず必要がない
Tire input	Road surface condition	Coefficient of friction	✓
		External force	✓
Tire condition	Tire condition	Puncture	N/A (機能要件) 制御不能（リム接地）になるまでの時間内でパンクを検知する



No lane departure because of combination of reasonably foreseeable stability disturbance worst case.

Pass/Fail

Figure 37 Safety perception flow of vehicle motion disturbance

3. Scenario-Based Safety Assurance Process

Figure 38 shows the schematics for the overall safety argumentation system in development and production cycle based on the V-shaped model, which is the project management commonly appointed to the development of advanced driving assistance systems (ADAS) and AD systems. By integrating verification to the sensor setup assessment and software agility basement processes from the planning phase in the first half of development, rather than conducting it only during the latter half of development represented by the right side of the V-shape, it can contribute to the optimisation of the development.

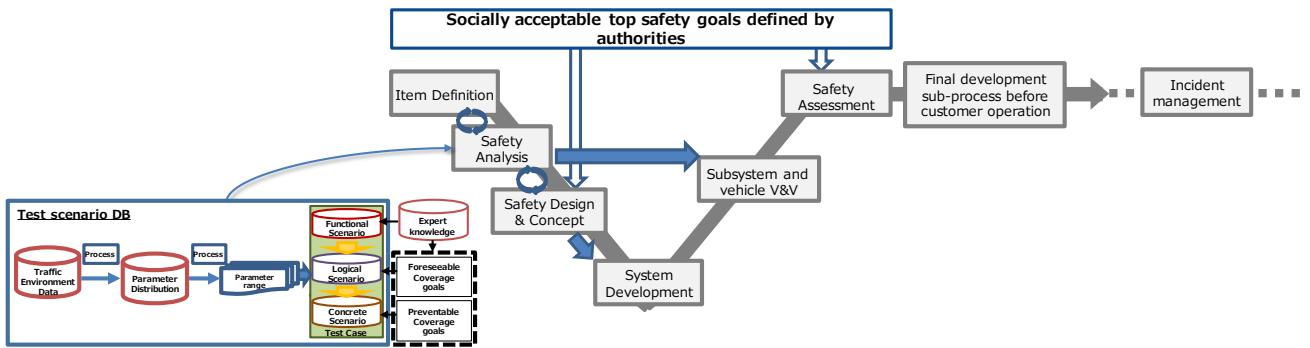


Figure 38. Overall scheme of safety assurance process

3.1. Safety argumentation scheme (Steps of the V-shaped model)

3.1.1. Item definition

The safety argumentation process is for making the vehicle compatible with the safety target within the operation scope of the automatic driving vehicle that was determined in advance. The operation scope of automatic driving vehicles is defined at the initial stage as the operation design scope (ODD). The contents of the ODD must include, at a minimum, information such as the road type, position on the road, vehicle velocity scope and environmental condition. Moreover, a fallback strategy for transition to outside the ODD boundary must be designed; moreover, the AD system must detect whether it is operating within the defined ODD. The definition of OD must be structured in such a manner as to enable notification to the users, as well as allow them to understand, trust and operate the AD system (Khastgir, Birrell, Dhadyalla, & Jennings, 2018).

Note that by mapping the ODD system and the scenario system as shown in Figure 39, it becomes possible to select the evaluation scenario following the ODD range.

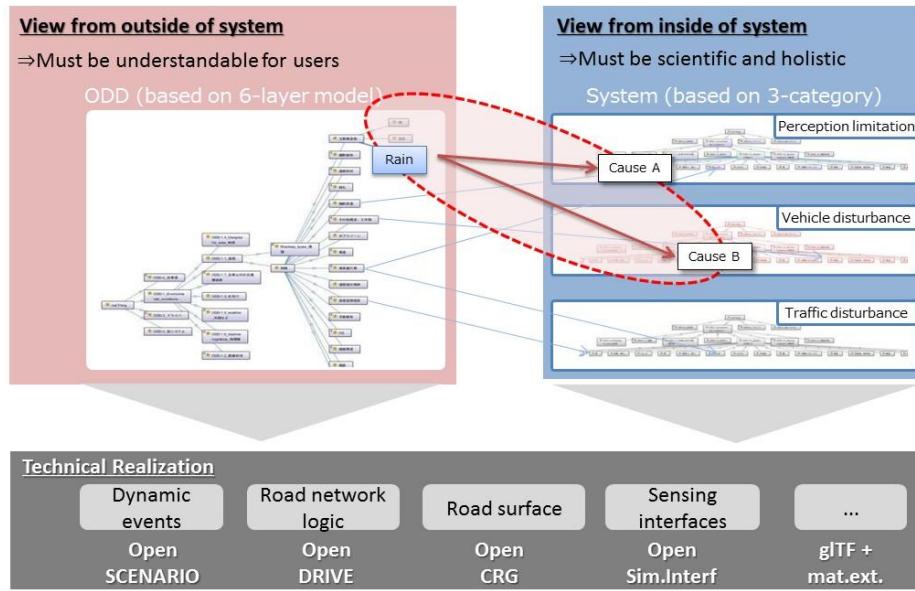


Figure 39. ODD scenario classification and relationship diagram of the system level classification based on the three category scenario level

3.1.2. Safety Analysis

It is important to determine as many foreseeable scenarios as possible, as well as systematise detailed scenario-related information on the operation design scope (ODD), vehicle and its surrounding, technically comprehensive definition of ODD based on the system physics, in addition to the overall definition of ODD that employs the systematic combination approach. For instance, the word ‘rain’ is enough for communicating with the user if rainfall conditions are included in the ODD; however, the AD system itself cannot interpret such a concept in the same manner. This scenario is able to consider the influence of rain from the perspective of system physics instead such as the possibility of the influence of raindrops on the sensor performance or the influence of rain on the vehicle dynamics (e.g., decrease in friction coefficient between the tire and the wet road surface). To describe ODD in a technical and system-oriented way, it is classified into three categories related to the system physics in order. These categories cover the respective perception, traffic flow and vehicle disturbances that can potentially occur within the AD system safety analysis (Figure Figure 2).

3.1.3. Safety Design and Safety Concept

The system requirements should be produced based on the safety analysis steps. The safety target defined by our association is integrated into the development cycle during this process, as well as confirmed during the system design. As layers of different complexity are added to the safety design, the safety analysis cycle can be unified as per necessity between this process and the preceding process as long as their outputs follow the safety analysis steps. It is important to ensure compatibility between the ODD and the system requirements to avoid unnecessary specification changes in the system development process. This indicates the importance of the role of the safety analysis step.

3.1.4. System development

When the system design is complete and its safety is analysed, the actual system that includes the component elements of both software and hardware is developed.

3.1.5. Examination and validation of the sub-system and the vehicle

At this point, the strategy for safety examination and validation of the system and the vehicle is defined without interaction with the driver. The examination and validation are conducted by combining concentrated virtual evaluations and a relatively limited amount of physical tests in real traffic environments and at test courses.

The mathematical and physical accuracy of the system, development functions, and employed safety measures are verified in the sub-process of the examination. Moreover, verification is performed in regard to whether all the safety specifications and requirements drawn up during the safety analysis process (sufficiency of sensors, algorithm and actuator-related measures) have been satisfied.

For the validation sub-process, verification is performed in terms of whether the system and components, including the employed safety measures, pose an irrational risk to the traffic participants. Moreover, the safety of the AD system is substantiated by confirming that the defined validation targets were met.

3.1.6. Safety assessment

The test for determining whether the end product is acceptable is conducted during this step, which includes the related inspections, document checks and certifications.

3.1.7. Final check process before release

In the final check before release, verification is performed in terms of whether the safety of the AD system can be explained, in addition to whether the remaining risk is within the permissible range. This can be conducted by, e.g., using technologies such as the behaviour safety assessment (BSA), which focuses on the evaluation of the AD system at each test case by applying different measurement standards and confirms the compatibility of AD with predefined behaviour standards. Finally, a determination is made in terms of whether the system can be released during the review of the result, and then the post-release incident management strategy is designed.

3.1.8. Incident management

During the incident management process, the performance data is fed back into the safety argumentation process. This enables the improvement of the AD technology and reduces the number of ‘unforeseeable’ situations as time passes. It is expected that, because of this reduction, the threshold between two left quadrants shifts, as well as the boundary between them will be lesser in the way that is beneficial to the foreseeable scenarios (Figure 40). Following the same logic, it is expected that the boundary between the preventable scenarios and unpreventable scenario shifts rightward, and the quadrant on the upper left will expand. It is highly possible that this will occur as more scenarios become preventable.

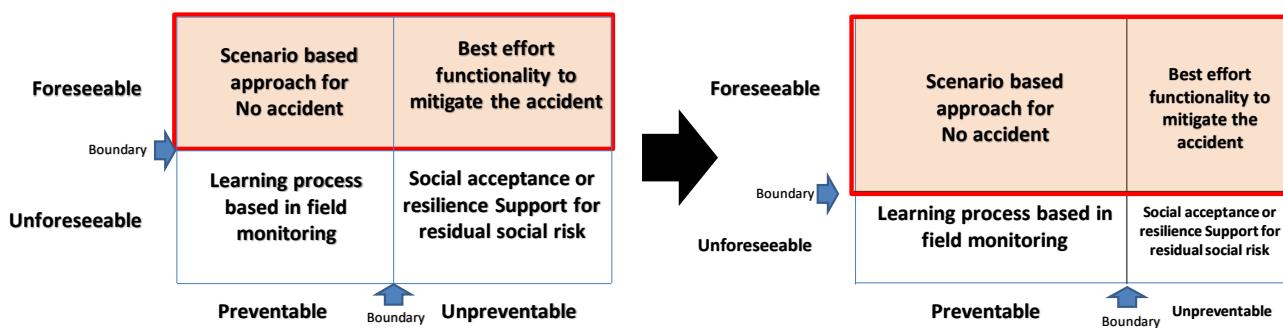


Figure 40. Expansion of foreseeable and preventable scopes following the evolution of the AD system

4. Scenario structure

Every approach is constructed by applying the systematic combination approach for defining the combinations derived from all possible factors. This approach requires significant specialized effort for defining all the factors and their interdependency as was the case by examining the safety coverage target. Therefore, it requires a systematic standardization methodology for structuring every factor related to the information. As mentioned earlier, the structures of the scenarios are the possible disturbances that can occur in three different categories related to the physics of the system, namely, the perception disturbance, traffic disturbance and vehicle motion disturbance.

4.1. Traffic disturbance scenario

The traffic disturbance scenario represents the traffic situation where the combination of the road geometry, behaviour of the ego vehicle, and the locations and motions of the surrounding vehicles can potentially lead to risk. The traffic scenarios can be classified into general vehicle scenarios (including four-wheeled vehicles and motorcycles) and motorcycle-specific scenarios (Figure 41).

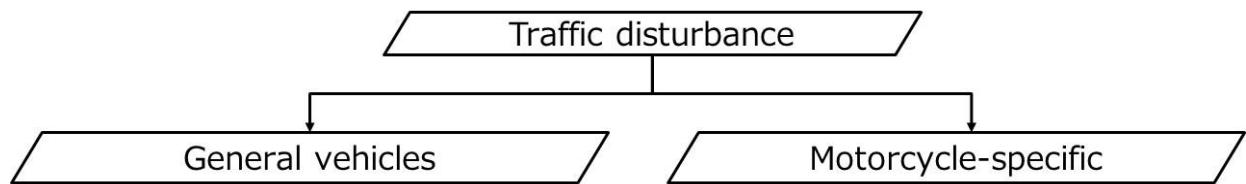


Figure 41. Traffic disturbance scenario classification

4.1.1. General vehicle scenario

The traffic disturbance scenarios of general vehicles are generated by systematically analysing and classifying the combinations of different factors, namely, the geometric shape of the road, the behaviour of the ego vehicle, and the locations and motions of the surrounding vehicles (Figure 42).

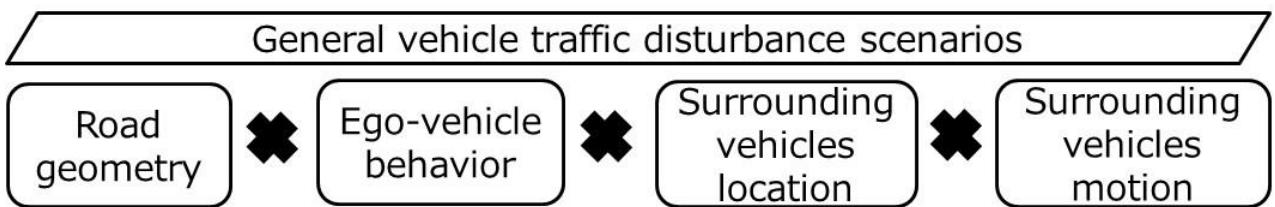


Figure 42. Structural diagram of the general vehicle traffic disturbance scenarios

4.1.1.1. Road geometry category

The road geometry is classified into four categories to generate scenarios, namely, the main road, the merging zone, the departure zone and the ramp. The road scenario classification for scenario generation must be discussed to make it applicable to highways internationally (Association, 2004) (Transportation, 2008; UK, 2006).

4.1.1.2. Vehicle behaviour category

The lane change from the next lane and merging lane exhibit the same ego vehicle behaviour, although their road geometry categories are different. The same can be said about lane keep; thus, the ego vehicle behaviours that can potentially occur can be classified into two categories of lane keep and lane change. These categories of vehicle behaviour can be represented as the aforementioned combinations of road geometry information (Figure 43).

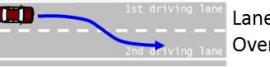
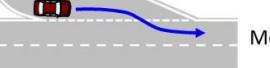
		Ego-vehicle behavior	
		Lane keep	Lane change
Road geometry	Main road	 Free driving Following	 Lane change Overtaking
	Merging zone	 Being merged	 Merging
	Departure zone	 Being merged	 Departure
	Ramp	 Free driving Following	 Lane change Overtaking

Figure 43. Road geometry and ego vehicle behaviour parameters

4.1.1.3. Categories of positions and motions of surrounding vehicles

The positions of the vehicles surrounding the ego vehicle that should be considered in the scenario structure are defined by neighbouring positions in eight directions around the ego vehicle that have the possibility of entering the driving trajectory of the ego vehicle. Moreover, when there is a significant difference between the speeds of the leading vehicle and the vehicle in front of it, the leading vehicle might perform cut-out to avoid a collision. When a cut-out suddenly occurs, the ego vehicle might be required to take action to avoid a collision. To consider this scenario, the position of the vehicle in front of the leading vehicle is indicated as “+1” (Figure 44, Left).

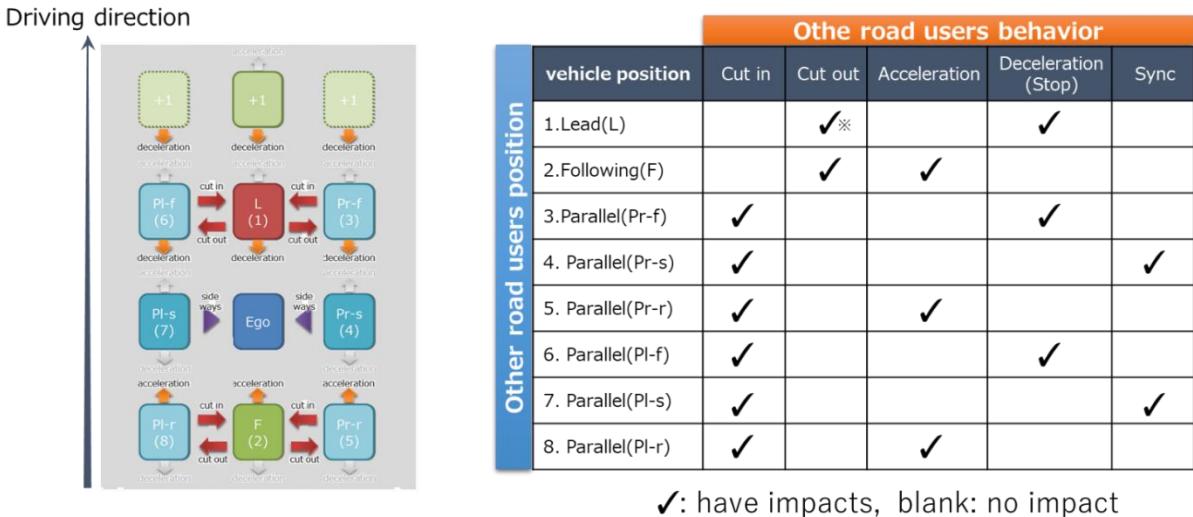


Figure 44. Surrounding vehicle positions (Left) and the combinations of the surrounding vehicle positions and the motions that can potentially obstruct the ego vehicle (Right)

The behaviours of surrounding vehicles are classified into the five categories: cut-in, cut-out, acceleration, deceleration and synchronisation. From the perspective of safety evaluation, it is possible to minimize the number of evaluations by focusing on the behaviours of other road users that can potentially obstruct the behaviour of ego vehicle (Figure 44, Right). For instance, the deceleration of vehicles at positions 5, 2 and 8 would not obstruct the ego vehicle and therefore can be excluded from safety analysis. The check mark in the figure indicates cases where the corresponding combinations of the surrounding vehicle positions and the motions can potentially obstruct the ego vehicle, which must be considered in the safety analysis.

4.1.1.4. Resulting traffic disturbance scenarios

As a result of the systematization process discussed to date, a methodology for structuring scenarios as the combinations of the road geometry, the ego vehicle behaviours and the positions and motions of the surrounding vehicles is proposed herein. This structure comprises a matrix that contains 40 possible combinations in total, among which 32 combinations correspond to the test scenarios that can occur in real traffic flow (Figure 45). The sufficiency of these 32 cases that cover all the dangerous cases that can lead to an accident on highways can be evaluated based on the comparable accident categories (Annex D). This matrix handles the comprehensive cover range of traffic disturbances because of interactions between two vehicles. It may be necessary to consider additional scenarios regarding the complex scenarios that involve motorcycles (Annex B) or more than two vehicles at once (Annex C).

		Surrounding Traffic Participants' Position and Behavior					
		Road geometry	Ego-vehicle behavior	Cut in	Cut out	Acceleration	Deceleration (Stop)
Road Geometry and Ego-vehicle behavior	Main roadway	Lane keep	No.1				
	Main roadway	Lane change	No.5				
	Merge	Lane keep	No.9				
	Merge	Lane change	No.13				
	Branch	Lane keep	No.17				
	Branch	Lane change	No.21				
	Main roadway	Lane keep	No.2				
	Main roadway	Lane change	No.6				
	Merge	Lane keep	No.10				
	Merge	Lane change	No.14				
	Branch	Lane keep	No.18				
	Branch	Lane change	No.22				
	Main roadway	Lane keep	No.3				
	Main roadway	Lane change	No.7				
	Merge	Lane keep	No.11				
	Merge	Lane change	No.15				
	Branch	Lane keep	No.19				
	Branch	Lane change	No.23				
	Main roadway	Lane keep	No.4				
	Main roadway	Lane change	No.8				
	Merge	Lane keep	No.12				
	Merge	Lane change	No.16				
	Branch	Lane keep	No.20				
	Branch	Lane change	No.24				

Figure 45. General vehicle traffic disturbance scenarios

4.1.2. Scenarios unique to motorcycles

In general, the categories of aforementioned positions and motions of surrounding vehicles (Figure 44) are applied to both four-wheeled vehicles and motorcycles. However, there are situations where motorcycles may drive in the narrow space in the same lane as the ego vehicle, which requires additional safety evaluation scenarios. Because these scenarios only have the potential to occur in countries where such driving is legally allowed, an approach including detailed examples is shown in Annex B.

4.1.3. Scenarios resulting from the combination of behaviours by several vehicles

The proposed traffic disturbance scenario structure covers the relationship between the ego vehicle and one or two surrounding vehicles. However, in real traffic, multiple traffic participants take diverse actions at various moments. The current methodology covers these complex cases by extracting scenarios where the sudden motions by surrounding vehicles trigger the sequence of avoidance motions. By dividing these scenario types into a sequence of behaviours, multiple combinations of the positions and motions of the ego vehicle and the surrounding vehicles can be covered by safety analysis. Moreover, this can be realized by considering the influence of the road environment on the cut-in scenario by other vehicles that can potentially appear in this sequence. For instance, when the leading vehicle performs sudden deceleration (the first behaviour of the sequence), the avoidance motion by the ego vehicle occurs (the second behaviour) and the ego vehicle retreats into the surrounding avoidance area. The detail of the approach to the complex scenarios that include detailed examples is included in Annex C.

4.2. Perception disturbance scenarios

Perception disturbance scenarios include blind spot scenarios and connectivity disturbance scenarios, in addition to perception disturbances (Figure 46).

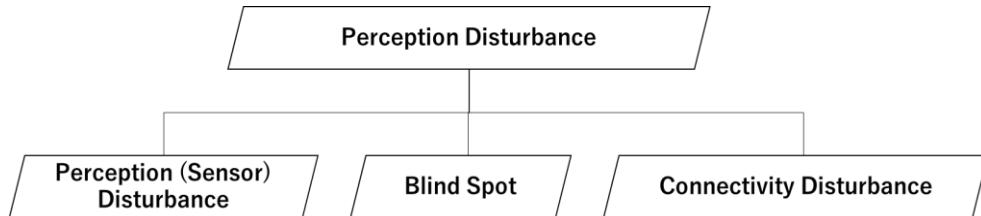


Figure 46. Categories of perception disturbance scenarios

4.2.1. Perception disturbance scenarios

Perception disturbance refers to a negative effect on perception performance during a situation in which the automatic driving system detects objects. The perception disturbance scenario is generated by disturbance-triggering factors and based on the principle of the sensors where disturbance occurs. While the factors of disturbances are diverse, it is possible to select the scenario group that contains the perception disturbance overall by classifying the factors based on the generation principle and then selecting a representative factor among those in the same category. Moreover, by considering the necessary combinations based on the generation principle of each disturbance factor, it is possible to create a perception disturbance combination evaluation scenario. In this study, the disturbance scenarios of three types of sensors, namely, millimetre wave radar, LiDAR and camera (Figure 47).

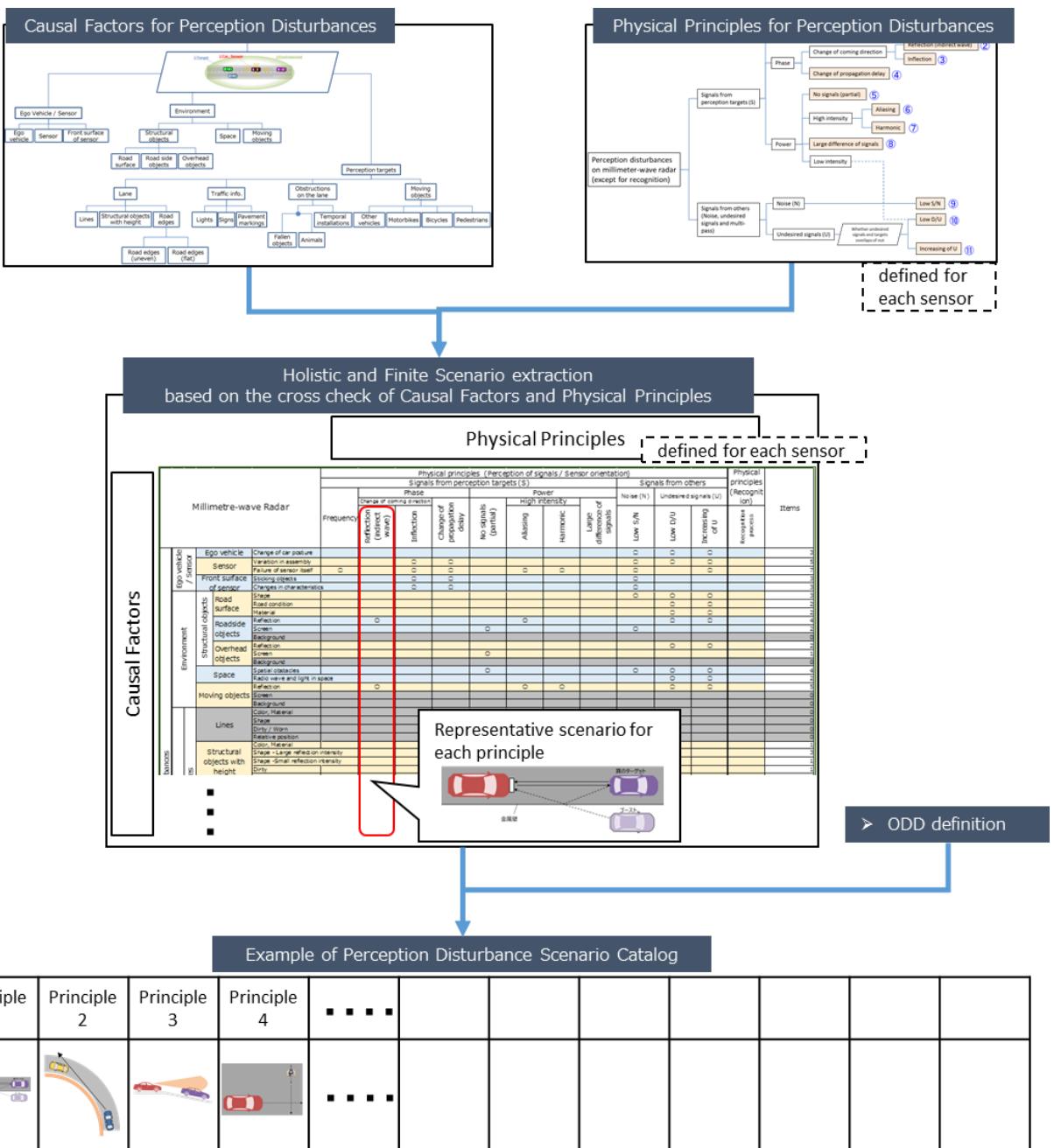


Figure 47. Scenario derivation process based on perception disturbance factors and sensor principle

4.2.1.1. Perception disturbance factors

The factors of perception disturbance can be broadly classified into “vehicle/sensor,” “surrounding environment” and “perception target” in relation to the ego vehicle, which are then broken down and comprehensively classified at each layer to compose the perception disturbance factors system. Here, e.g., a factor is broken down from the perspectives of structure, relative position and types, and continues to be categorized to layers such as colour, shape, material and behaviour.

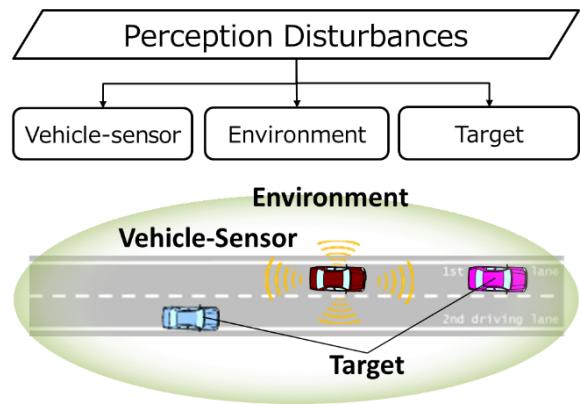


Figure 48. Broad categories of perception disturbance factors according to the positional relationship with the ego vehicle

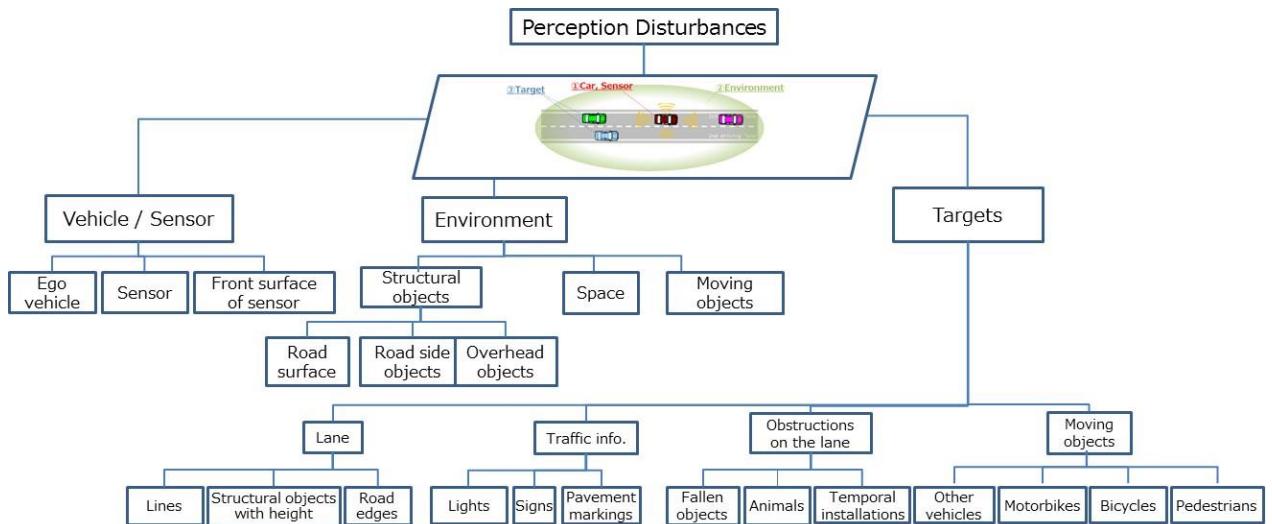


Figure 49. System diagram of perception disturbance factors

4.2.1.1.1. Perception Disturbance Factors: Vehicle/Sensor

The perception factors classified into “vehicle/sensor” are divided into three categories according to the positions of these factors, namely, “a. ego vehicle”, “b. sensor” and “c. in front of the sensor”.

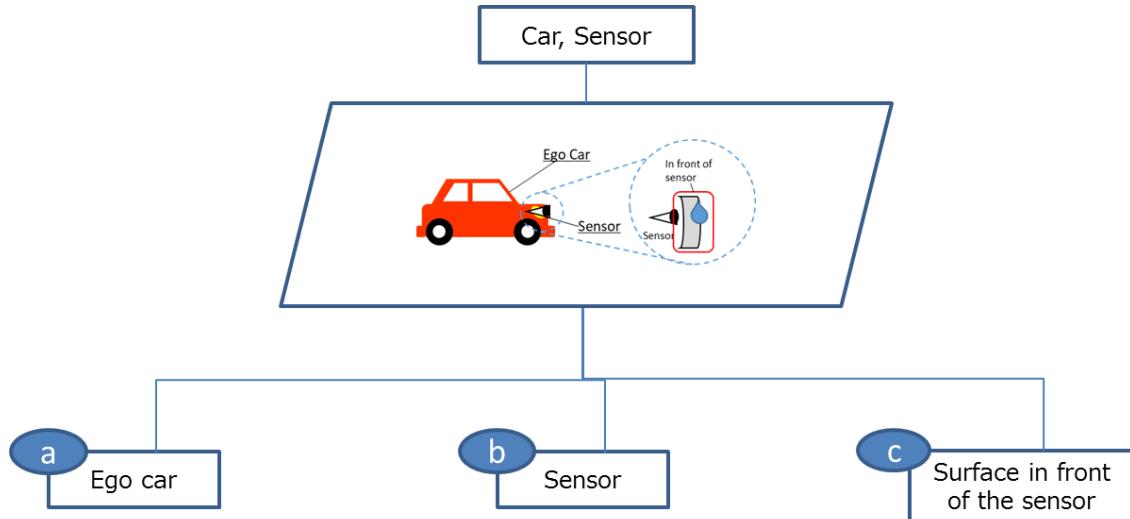


Figure 50. Vehicle/sensor categories

Tables 3–5 show the details of the perception disturbance factors categorized into a, b and c. These tables describe the detailed categorization, impact on the perception performance, and the generation principle of perception disturbance of the perception disturbance factors for each sensor.

Table 3. “a. Ego Vehicle” disturbance factors

a Ego car				
Influence on sensor principle	Class.	Millimeter waves		LiDAR
	Influence	Change of car posture		Camera
	Principle	<ul style="list-style-type: none"> - Decrease in positioning accuracy which caused by the difference between actual attached position and assumed position that memorized in its sensor. - Changing FOV by changing center axis of Radar. - Misrecognition of road surface as obstacle. 		<ul style="list-style-type: none"> - Misrecognition of road surface as obstacle. - Image recognition ability degradation caused by vertical moving of its sensor and ego car.

Table 4. “b. Sensor” disturbance factors



		Millimeter waves		LiDAR		Camera	
Class.	Influence on sensor principle	Variation in assembly	Failure of sensor itself	Variation in assembly	Failure of sensor itself	Variation in assembly	Failure of sensor itself
Influence	- Blind spot. - Decrease in azimuth estimation accuracy which caused by interference direct wave with internal reflected wave. - Decrease in positioning accuracy, which caused by misalignment.	- Decrease in maximum detectable range which caused by receive intensity decreasing. - Change in signal phase and frequency due to change in sensor character.	- Blind spot. - Misalignment in optical axis and attached position.	- Decrease in input and output intensity, which caused by aging degradation.	- Blind spot. - Image recognition ability degradation caused by misalignment.	- Position shift and Color shift in image. - Image recognition ability degradation caused by lens distortion. - Partially information loss due to defective pixels. - Thermal noise and sensitivity variations depend on temperature.	
	• Phase changing • Undesired signal increasing • Low S/N • Slight shift of axes	• Phase changing • Frequency changing • Low S/N	• No signal (partial)	• Signal lowering	• No signal (partial)	• Signal changing • Refraction • No signal (partial) • S/N lowering	
Principle							

Table 5. “c. In front of sensor” disturbance factors



		Millimeter waves		LiDAR		Camera		
Class.	Influence on sensor principle	Sticking objects	Changes in characteristics	Sticking objects	Changes in characteristics	Sticking objects	Changes in characteristics	Reflection on windshield
Influence	- Decrease in maximum detectable range due to decrease in reception intensity. - Degradation of azimuth estimation accuracy due to interference between reflected wave and direct wave on sticking objects.	- Decrease in maximum recognition range due to decrease in reception intensity. - Degradation of azimuth estimation accuracy due to interference between reflected wave and direct wave to changes in sensor front characteristics.	- maximum detectable range decreases due to received signal strength reducing. - Signal saturation by detecting contamination. - Angle shift due to contamination (oil film, etc.) on the sensor.	- Decrease in received signal strength due to decrease in transmittance. - Signal saturation by detecting cloudiness on surface in front of sensor. - Angle shift due to distortion on the surface in front of sensor.	- Image recognition ability degradation due to lack of an image by objects sticking to windshield. - Image recognition ability degradation due to raindrops and like become noise. - A distant vehicle overlaps with the raindrops, reducing the maximum detectable range.	- Position and color on image are shifted more than design error. - Image recognition ability degradation due to the distortion of windshield.	- False recognition of reflection.	
	• Phase changing • Undesired signal increasing • Low S/N	• Phase changing • Undesired signal increasing • Low S/N	• Signal intensity lowering ~ No signal • Undesired signal increasing • Refraction	• Signal intensity lowering ~ No signal • Undesired signal increasing • Refraction	• No signal (partial) • Low S/N	• Signal changing • Refraction	Low S/N	
Principle								

4.2.1.1.2. Perception disturbance factors: Surrounding environment

The perception factors classified into “surrounding environment” are divided into three categories according to the characters of the objects existing around the ego vehicle, namely, “d. surrounding structure”, “e. space” and “f. surrounding moving objects”. “d. Surrounding structure” is further divided

into the following three categories: “d-1. road surface”, “d-2. structure by the road” and “d-3. structure above the road”.

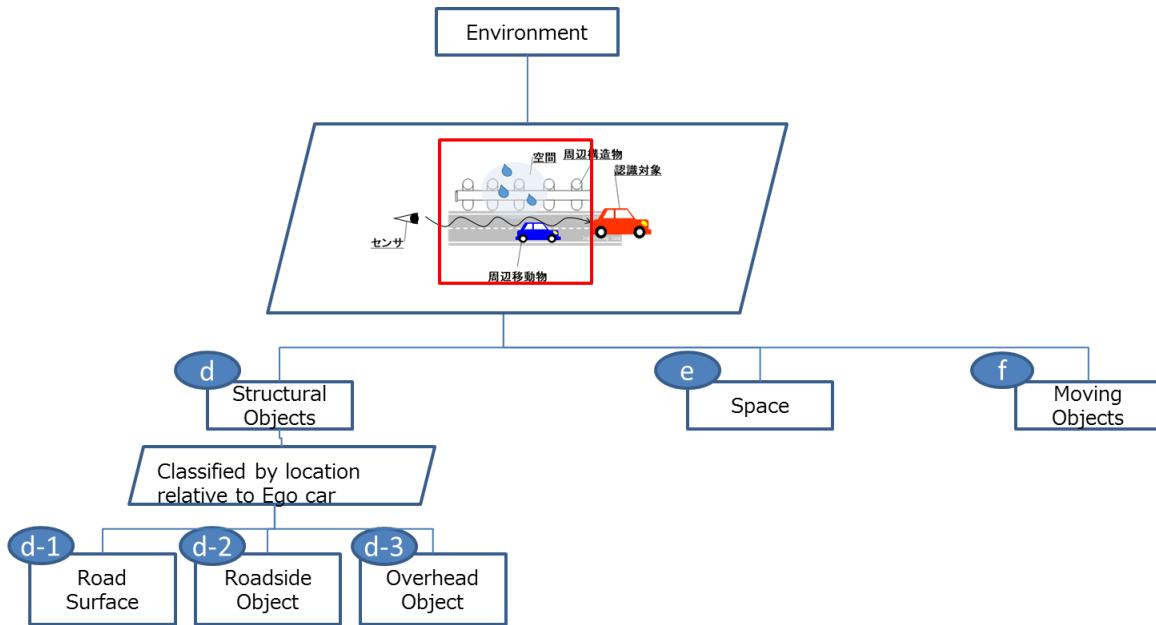


Figure 51. Surrounding environment categories

Tables 6–8 show detailed categorization, impact on the perception performance, and the generation principle of perception disturbance of the perception disturbance factors classified into d-1, d-2, d-3, e and f.

Table 6. “d-1. Road surface” disturbance factors

Road Surface										
		Millimeter waves			LiDAR			Camera		
Influence on sensor principle	Class:	Shape	Road condition	Material	Shape	Road condition	Material	Shape	Road condition	Material
Influence	- Vehicles in front is out of FOV due to the slope - Vehicles in front disappear/appear due to the slope - Sloped road surface is recognized as static object in front	<ul style="list-style-type: none"> - Changes in road surface reflection characteristics change road surface multipath and reduce signal strength - Increase in clutter 		<ul style="list-style-type: none"> - Increase in clutter 	<ul style="list-style-type: none"> - Vehicles far ahead is out of FOV due to the slope - Vehicles far ahead disappear/appear due to the slope - Sloped road surface is recognized as static object in front 		<ul style="list-style-type: none"> - False points occur when the road surface has high reflection characteristics such as icy surface or puddles 	<ul style="list-style-type: none"> - Due to the difference in reflection characteristics, a part of the road surface is detected and incorrectly recognized as an obstacle 	<ul style="list-style-type: none"> - Before going down, there is no visibility and no detection - Shape change of the object before going up - Change of tilt on image due to cant 	<ul style="list-style-type: none"> - Misrecognized reflection in puddles - Misrecognition of road restorations and wheel tracks
Principle	<ul style="list-style-type: none"> • Low D/U • Undesired signal increasing • Low S/N 	<ul style="list-style-type: none"> • Low D/U • Undesired signal increasing 	<ul style="list-style-type: none"> • Low D/U • Undesired signal increasing 		(Recognition)	Reflection	(Recognition)	No signal (partial) (Recognition)	<ul style="list-style-type: none"> • Signal change • Reflection • Low S/N, LowD/U (Recognition) 	<ul style="list-style-type: none"> • Low D/U • Low S/N (Recognition)

Table 7. “d-2. Structures by the road” disturbance factors

Roadside Object										
		Millimeter waves			LiDAR			Camera		
Influence on sensor principle	as	Reflection	Screen	Background	Reflection	Screen	Background	Reflection	Screen	Background
Influence	- Ghost occurs due to multi-path - In case of multiple targets at the same distance, the horizontal direction accuracy becomes worse	<ul style="list-style-type: none"> - Partial loss of FOV due to side walls, etc. 		<ul style="list-style-type: none"> - Misrecognition of reflected objects - Objects with high reflectivity on the roadside (such as delineators) are incorrectly recognized as vehicles 	<ul style="list-style-type: none"> - An object in front of the object obscures part of it 		<ul style="list-style-type: none"> - Misrecognition of reflected objects 	<ul style="list-style-type: none"> - An object in front of the object obscures part of it - Objects are difficult to see or change color due to the effects of transparency 		<ul style="list-style-type: none"> - The boundary between object and background is unclear - Background is incorrectly recognized as target
Principle	<ul style="list-style-type: none"> -Undesired signal increasing -Reflection -Inflection -Low D/U 	<ul style="list-style-type: none"> -No signal (partial) -Low S/N 		<ul style="list-style-type: none"> -Multiple reflection (Recognition) 	<ul style="list-style-type: none"> -No signal (partial) 		<ul style="list-style-type: none"> -Reflection -Signal changing 	<ul style="list-style-type: none"> -No signal (partial) -Signal changing 	<ul style="list-style-type: none"> -Low D/U (Recognition) 	

Table 8. “d-3. Structures above the road” disturbance factors

d-3 Overhead Object										
		Millimeter waves			LiDAR			Camera		
Influence on sensor principle	Class	Reflection	Screen	Background	Reflection	Screen	Background	Reflection	Screen	Background
Influence	Oa ss.	- Lack of vertical resolution capability	An object in front of the object obscures part of it		- Reflective items such as mirrors placed on the curb which have extremely high directivity, may cause detection of the object reflected in the mirror rather than the mirror itself. - Misrecognition of the objects with high reflectivity above as vehicles	An object in front of the object obscures part of it		Misrecognition of reflected objects	An object in front of a target obscures part of it	- The boundary between object and background is unclear - Background is incorrectly recognized as target
	Principle	-Undesired signal increasing	-No signal (partial)		-Multiple reflection (Recognition)	No signal (partial)		-Reflection -Signal changing	-No signal (partial) -Signal changing	-Low D/U (Recognition)

Table 9. “e. Space” disturbance factors

e Space										
		Millimeter waves			LiDAR		Camera			
Influence on sensor principle	Class	Spatial obstacles	Radio wave and light in space	Spatial obstacles	Radio wave and light in space	Spatial obstacles	Radio wave and light in space			
Influence	Class	- Due to weaker receiving signal, the max detection distance decreases - Partial or complete loss of FOV due to flying objects - Misrecognition of flying object as target	- Noise floor rising due to radio wave interference - Misrecognition of interference signal as reflection from target	Confirm impact of spatial obstacles caused by weather. Rain is an obstacle on the light path, therefore ranging performance will deteriorate due to reduced reflected light.	- Blocked up shadows due to west sun, backlight, etc. - Noise increase due to increase in background light. - Pulse noise by LiDAR of other vehicles	- Objects in short-distance are lost because of spatial obstacles as noise. - Target is hidden by obstacles - Misrecognition of flying objects as vehicles/pedestrians - Recognition rate drops due to lower contrast in case of rain or snow.	- In twilight or dusk environment, insufficient light causes recognition ability degradation. - Blown out highlights occur when the light source is strong locally. - Blocked up shadows due to west sun, backlight, etc. - Target color changes due to light source color - Target contrast reduces due to light source - Flare and smear caused by strong light source (backlight)			
	Principle	-No signal (partial) -Low S/N -Undesired signal increasing	-Low D/U -Undesired signal increasing	-Signal intensity lowering -No signal by obstacle -Signal from other object by reflection or refraction	-Noise such as DC type -Noise such as Pulse type (Recognition)	-No signal (partial) -Signal changing -Low D/U -Low S/N (Recognition)	Low signal intensity High signal intensity Signal intensity increasing Signal changing Low D/U Low S/N			

Table 10. “f. Surrounding moving objects” disturbance factors

Moving Object										
		Millimeter waves			LiDAR			Camera		
Class:	Influence	Reflection	Screen	Background	Reflection	Screen	Background	Reflection	Screen	Background
Influence sensor principle	Influence	- Ghost caused by grating, harmonic, phase noise or side lobe - Decrease in accuracy of reflected signal around strong reflection	Blind spot by surrounding vehicles		- False points occur due to surface reflection	Blind spot by surrounding vehicles		Recognition rate reduction depending on paint reflection - Misrecognition due to mirror-finished coating	Blind spot by surrounding vehicles	Recognition rate drops due to protective colour same as surroundings
	Principle	-Grating reflection -High frequency -Low D/U -Undesired signal intensity increasing			-Reflection -Multiple reflection (Reflection)			-Reflection -Signal changing		-Low D/U (recognition)

4.2.1.1.3. Perception Disturbance Factors: Perception Targets of Sensors

The perception disturbance factors categorized as “perception targets of sensors” are broadly classified into “g. route”, “h. traffic information”, “j. obstacles” and “k. moving object” (Figure 52).

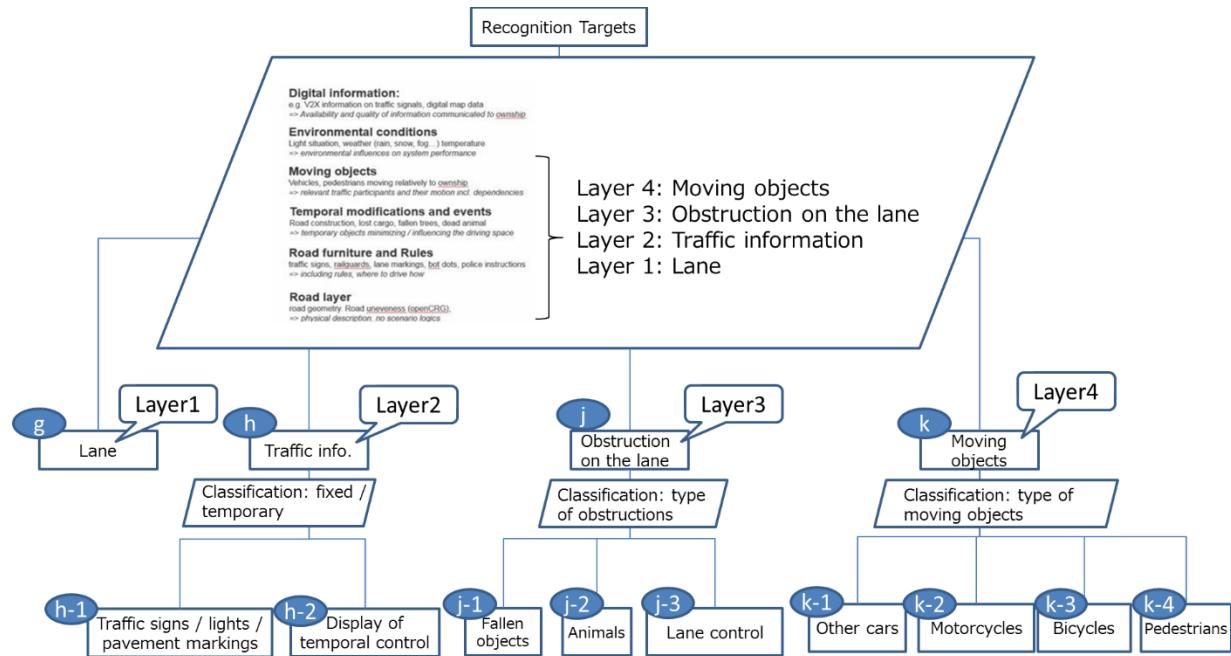


Figure 52. Categories of perception targets of sensor

“g. Route” is classified into “g-1. lane maker”, “g-2. structure with height” and road edge as per the object that indicates a given place is a driving route. Moreover, road edge is divided further into g-3 and g-4 depending on whether there is a level difference or not (Figure 53).

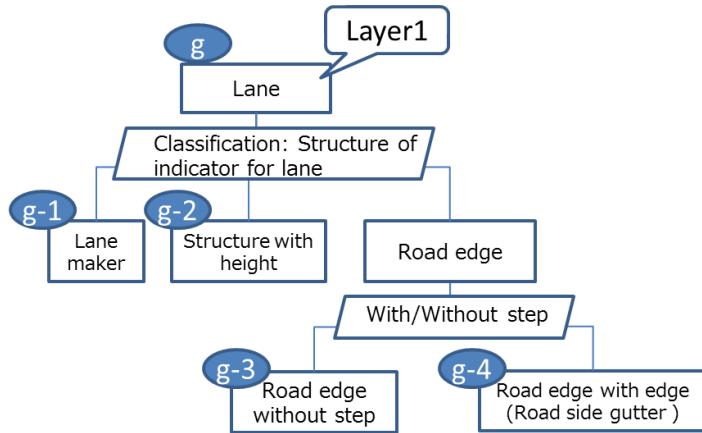


Figure 53. Categories of “g. route”

“h. Traffic information” is classified into “h-1. traffic light”, “h-2. traffic sign” and “h-3. road marking” as per their display style (Figure 54).

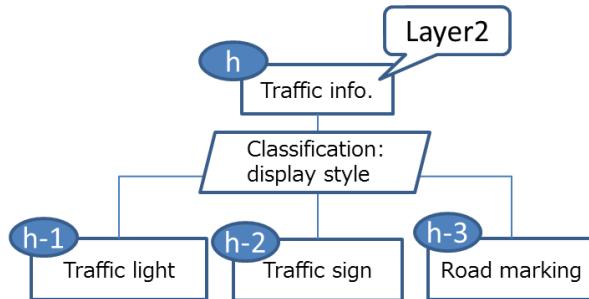


Figure 54. Categories of “h. traffic information”

“j. Obstacle” is classified into “j-1. falling object”, “j-2. animal” and “j-3. installed object” according to whether it moves or not and the degree of impact when colliding with the vehicle (Figure 55).

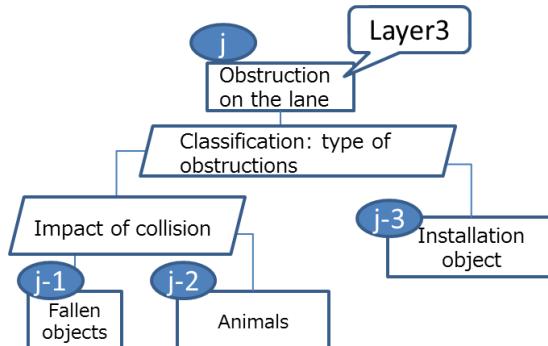


Figure 55. Categories of “j. obstacle”

“k. Moving objects” are classified into “k-1. other vehicles”, “k-2. motorcycle”, “k-3. bicycle” and “k-4. pedestrian” as per the type of traffic participant (Figure 56).

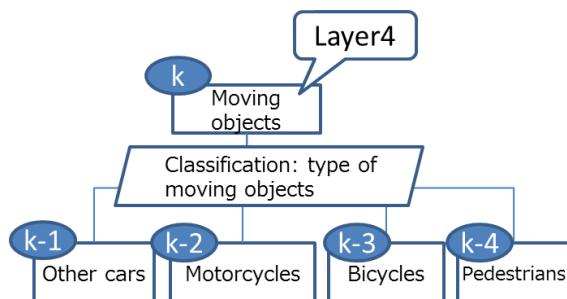


Figure 56. Categories of “k. moving objects”

Tables 11–14 show the detailed categorization, impact on the perception performance, and the generation principle of perception disturbance for the perception disturbance elements classified into g-1 to k-4, respectively.

Table 11. “g-1. Lane marker” disturbance elements

g-1		Lane markers							
Influence on sensor principle	Class.	Millimeter waves	LiDAR	Camera					
Influence		Colors/materials	Shapes	Grime/worn	Relative position	Colors/materials	Shapes	Grime/worn	Relative position(*)
		•Lack of contrast with surrounding pavements •Unknown reflection intensity	•Unknown shapes (thickness, intervals, appearance, etc.)	•Hidden •Dirty / worn •false positive for deleted lines	•shift of positions due to ego-vehicle's movement	•Lack of contrast with surrounding pavements •Unknown brightness, chroma and color phase	•Unknown shapes (thickness, intervals, appearance, etc.)	•Hidden •Dirty / worn •false positive for deleted lines	•Image deletion during driving •Distortion
Principle		Recognition process	Recognition process	Low intensity No signal due to masking	Position change of FOV Recognition process	Low D/U (Recognition process)	(Recognition process)	No signal (partial) Low S/N	Blurred image (Recognition process)

Including Botts' Dots and Cat's-eye which can be crossed

Table 12. “g-2. Structure (with height)” disturbance elements

g-2		Structure with height								crash barriers, poles, noise barriers, curbstones, trees, cat's-eyes, etc.			
Influence on sensor principle	Class, Influence	Millimeter waves				LIDAR				Camera			
		Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position
Influence	Influence	•Lowering of reflection intensity depending on a material	•Lowering or increasing of reflection intensity depending on shape, size and direction	•Lowering of reflection intensity	•Lowering of intensity around the edges of FOV	•Lowering or increasing of reflection intensity depending on a material	•Lowering or increasing of reflection intensity depending on shape, size and direction	•Lowering of reflection intensity	•shift of positions due to ego-vehicle's movement	•Lack of contrast with the background •Poor perception due to pictures or patterns on the walls	•Poor perception due to unknown shapes	•False positive for grime or patterns as targets	•Image deletion during driving •Distortion
Principle	Low S/N	Aliasing Harmonic Large difference of intensity Low D/U Low S/N	Low S/N	Low S/N	Low S intensity High S intensity	Low S intensity High S intensity	Low S intensity	Position change of FOV Recognition process	Low D/U (Recognition process)	Low S/N (Recognition process)	Low S/N	Blurred image (Recognition process)	

Table 13. “g-3. Road edge without level difference” disturbance elements

Road edges without a step				Break of a road, etc.									
Millimeter waves				LIDAR				Camera					
Influence on sensor principle	Class,	Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position
Influence		•Lowering of reflection intensity depending on a material	•Lowering of reflection intensity depending on shape, size and direction	•Lowering of reflection intensity	•Lowering of intensity around the edges of FOV	•False positive for road surface with difference of reflection intensity as road edges	•Lowering of reflection intensity depending on shape, size and direction	•Poor perception due to masking by accumulated snow and fallen leaves	•shift of positions due to ego-vehicle's movement	•False positive for road surface with a different color as road edges	•Unknown road shape out of the lane	•Poor perception due to masking by accumulated snow and fallen leaves	•Image deletion during driving •Distortion
Principle		Low S/N	Low S/N	Low S/N	Low S/N	Recognition process	Low S	Low S No signals (partial)	Position change of FOV Recognition process	Low D/U (Recognition process)	Low S/N (Recognition process)	No signals (partial)	Blurred image (Recognition process)

Table 14. “g-4. Road edge with a step” disturbance elements

Road edges with a step				Gutters, etc.									
Millimeter waves				LIDAR				Camera					
Influence on sensor principle	Class,	Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position
Influence		•Lowering of reflection intensity depending on a material	•Lowering of reflection intensity depending on shape, size and direction	•Lowering of reflection intensity	•Lowering of intensity around the edges of FOV	•Lowering or increasing of reflection intensity depending on a material	•Lowering of reflection intensity	•shift of positions due to ego-vehicle's movement	•Lack of contrast at the road edge area	•poor detection due to too small gutter width •Unknown road shapes out of the lane	•Poor perception due to masking by accumulated snow and fallen leaves	•Image deletion during driving •Distortion	
Principle		Low S/N	Low S/N	Low S/N	Low S/N	Low S	Low S High S	Low S	Position change of FOV Recognition process	Low D/U (Recognition process)	Low D/U (Recognition process)	No signals (partial)	Blurred image (Recognition process)

Table 15. “h-1. Traffic lights” disturbance elements

h-1

Traffic lights



	Millimeter waves	LiDAR			Camera		
Influence on sensor principle			Colors/materials	Shapes	Light source	Grime	Relative position
Influence			•Unknown brightness, chroma and color phase	•Poor perception due to whether portrait of landscape •Poor perception due to difference of light sizes •False positives for lights with hoods	•Poor perception due to flicker •Poor perception due to directional quality of lights	•Poor detection due to masking by snow, etc.	•Miss recognition due to directional quality of lights •No perception due to FOV •No perception due to direction and lens distortion •Recognition of different lights due to direction
Principle			(Recognition process)	(Recognition process)	Flicker Low S/N	No signal (partial)	Low S/N No signal (partial) (Recognition process)

Table 16. “h-2. Traffic sign” disturbance elements

h-2

Traffic sign



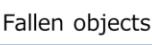
	Millimeter waves	LiDAR			Camera		
Influence on sensor principle			Colors/materials	Shapes	Light source	Grime	Relative position
Influence			•Recognition failure due to insufficient contrast with the surrounding road surface •Detection / recognition failure due to unexpected brightness / coloring / saturation / hue	•Misrecognition of letters and numbers with similar shapes •Misrecognition due to different shape of sign for each country	•Electric sign recognition failure due to flicker or lack of partial image	•Misrecognition due to lack of image or image interference with faintness or stains	•Recognition failure due to image flow •Recognition failure due to shape change with orientation •Recognize other signs because of their location
Principle			Low D/U (Recognition process)	(Recognition process)	Flicker	Low S/N	Blurred Image No signal (partial) (Recognition process)

Table 17. “h-3. Road marking” disturbance elements



h-3		Road marking	Camera				
Influence on sensor principle	Class.	Millimeter waves	LiDAR	Colors/materials	Shapes	Grime	Relative position
Influence				<ul style="list-style-type: none"> Recognition failure due to insufficient contrast with the surrounding road surface Detection / recognition failure due to unexpected brightness / coloring / saturation / hue 	<ul style="list-style-type: none"> Detection / recognition failure due to unexpected shape of lane (Unknown display, width, space) 	<ul style="list-style-type: none"> Obstacle objects cannot be imaged and are not recognized Misrecognition due to faintness or dirt Misrecognition of erased lane 	<ul style="list-style-type: none"> Recognition failure due to image flow
Principle				Low D/U (Recognition process)	(Recognition process)	No signal (partial) Low S/N	Blurred Image No signal (partial) (Recognition process)

Table 18. “j-1. Falling object” disturbance elements



j-1		Fallen objects	LiDAR				Camera			
Influence on sensor principle	Class.	Millimeter waves	Shape / Size	Relative position / Motion	Color / Material (contrast ratio)	Shape / Size	Relative position / Motion	Color / Material (contrast ratio)	Shape / Size	Relative position / Motion
Influence		<ul style="list-style-type: none"> Range lowering by low reception intensity False perception by dispersion of reception intensity 	<ul style="list-style-type: none"> Reflection intensity lowering depending on shape/size/direction 	<ul style="list-style-type: none"> Signal intensity lowering around edge of FOV False perception by Moving or Rolling object 	<ul style="list-style-type: none"> Reception intensity lowering or increasing depend on material 	<ul style="list-style-type: none"> Reflection intensity lowering depending on direction/structure/size 	<ul style="list-style-type: none"> Position shifting due to vehicle moving 	<ul style="list-style-type: none"> Lowering of contrast by similar background color Recognition ability degradation by mirror and luminous Large difference in brightness 	<ul style="list-style-type: none"> Limitation of image information by FOV Recognition ability degradation depending on shape/size 	<ul style="list-style-type: none"> Image blurred or shift by object moving Big moving on closed object Limitation of image information by FOV
Principle		Low S/N	Low S/N	Low S/N	<ul style="list-style-type: none"> Signal intensity lowering Signal saturation Reflection Multi reflection 	Signal intensity lowering	<ul style="list-style-type: none"> Position shift of all space Position shift of target (Recognition) 	<ul style="list-style-type: none"> Low D/U Reflection Flicker Large difference of signal intensity 	<ul style="list-style-type: none"> No signal (partial) (Recognition) 	<ul style="list-style-type: none"> Blurred Image No signal (partial) (Recognition)

Table 19. “j-2. Animal” disturbance elements

j-2 Animals

Class	Millimeter waves			LiDAR			Camera		
	Color / Material (contrast ratio)	Shape / Size	Relative position / Motion	Color / Material (contrast ratio)	Shape / Size	Relative position / Motion	Color / Material (contrast ratio)	Shape / Size	Relative position / Motion
Influence on sensor principle	Influence	—	-Lowering of reflection intensity depending on physical build and posture	-Signal intensity lowering around edge of FOV -False perception by moving animal	-Lowering of reception by low reflectance	-Lowering of reflection by change of reflection area depending on animal type, direction, size and posture	-Position shifting due to own vehicle or target object moving	-Recognition failure by low contrast because of similar color of background -Flicker by object lighting -Large difference in brightness	-Limitation of image information by FOV -Recognition ability degradation depending on shape/size
Principle		—	Low S/N	Low S/N (Perception)	Signal intensity lowering Reflection Multi reflection	Signal intensity lowering	Position shift of all space Position shift of target (Recognition)	Low D/U Reflection Flicker Large difference of signal intensity	No signal (partial) (Recognition)

Table 20. “j-3. Installation object” disturbance elements

j-3 Installation object



Class	Millimeter waves				LiDAR				Camera			
	Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position	Colors/materials	Shapes	Grime	Relative position
Influence on sensor principle	Influence	-Reflection intensity lowering depending on material	-Reflection intensity lowering or increasing depending on direction/structure /size	-Reflection intensity lowering depending on stains	-Signal intensity lowering around edge of FOV	-Reflection intensity lowering or increasing depending on material	-Reflection intensity lowering or increasing depending on direction/structure /size	-Position shifting due to vehicle moving	-Lowering of contrast by similar background color -Recognition ability degradation by mirror and luminous -Large difference in brightness between luminous and basement -Detection / recognition failure due to unexpected brightness / coloring / saturation / hue	-Limitation of image information by FOV -Recognition ability degradation depending on shape/size	-Misrecognition due to lack of image or image interference with stains	-Image blurred or shift by object moving -Recognition failure due to shape change with orientation
Principle		Low S/N	Folding Harmonic Large difference of Intensity Low D/U Low S/N	Low S/N	Low S/N	Signal intensity lowering Signal intensity saturation Reflection Multi reflection	Signal intensity lowering Signal intensity saturation	Signal intensity lowering	Misalignment of the entire space (Recognition)	Low D/U Reflection, flicker, Large difference of intensity (Recognition)	No signal (partial) (Recognition)	Low S/N

Table 21. “k-1. Other vehicles” disturbance elements

k-1 Other cars

Influence on sensor principle Class. Principle	Millimeter waves				LiDAR				Camera			
	Color	Materials of parts (paints, surface)	Sticking objects	Shape / Size	Color (contrast ratio)	Shape / Size	Materials of parts (paints, surface)	Sticking objects	Color (contrast ratio)	Shape / Size	Materials of parts (paints, surface)	Motion
—	<ul style="list-style-type: none"> Detection range lowering by reflectance lowering False perception by dispersion of reception intensity 	<ul style="list-style-type: none"> Detection range lowering by reflectance lowering False perception by dispersion of reception intensity 	No detection from vehicle parts with low reflectance Large reflection from a large object	Reception lowering by low reflectance	Reception lowering depending on reflection area and incidence angle	Reception lowering by low reflectance	Reflection lowering by sticking objects on the surface of objects	Recognition ability degradation by apathetic colors	<ul style="list-style-type: none"> Recognition ability degradation for extra-large cars Degradation of range accuracy depending on the width of cars 	<ul style="list-style-type: none"> Recognition ability degradation depending on the reflection at paint False recognition by paint with mirror finish 	<ul style="list-style-type: none"> Recognition ability degradation by high-speed approach to a line of vehicles Recognition ability degradation by sudden cut-in 	<ul style="list-style-type: none"> Detection range degradation and object lost by lowering of light intensity Hidden rear lamp by sticking objects
—	Low S/N	Low S/N	Grating Harmonic Low S/N	Signal intensity lowering	Signal intensity lowering	Signal intensity lowering	Signal intensity lowering	Low D/U	(Recognition)	<ul style="list-style-type: none"> Reflection Signal changing 	Blurred Image	<ul style="list-style-type: none"> No signal (partial) Low S/N

Table 22. “k-2. Motorcycle” disturbance elements

k-2 Motorcycles

Influence on sensor principle Class. Principle	Millimeter waves				LiDAR			Camera			
	Color	Materials	Sticking objects	Shape / Size	Color (contrast ratio)	Shape / Size	Materials	Color (contrast ratio)	Shape / Size	Materials	Motion
—	<ul style="list-style-type: none"> Detection range lowering by reflectance lowering False perception by dispersion of reception intensity 	<ul style="list-style-type: none"> Detection range lowering by reflectance lowering False perception by dispersion of reception intensity 	No detection from vehicle parts with low reflectance	Reception lowering by low reflectance	Reception lowering depending on reflection area and incidence angle	Reception lowering by low reflectance	<ul style="list-style-type: none"> Recognition failure by low contrast with background with similar color Recognition ability degradation by similar colors with surroundings 	<ul style="list-style-type: none"> Misrecognition depending on the width and length Recognition ability degradation depending on the shape 	—	—	Recognition ability degradation depending on inclination and driving direction
—	Low S/N	Low S/N	Low S/N	Signal intensity lowering	Signal intensity lowering	Signal intensity lowering	Low D/U	(Recognition)	—	Blurred Image	

Table 23. “k-3. Bicycle” disturbance elements

		Millimeter waves				LiDAR			Camera			
Influence on sensor principle	Class.	Color	Materials	Sticking objects	Shape / Size	Color (contrast ratio)	Shape / Size	Materials	Color (contrast ratio)	Shape / Size	Materials	Motion
Influence	—	<ul style="list-style-type: none"> Detection range lowering by reflectance lowering False perception by dispersion of reception intensity 	<ul style="list-style-type: none"> Detection range lowering by reflectance lowering False perception by dispersion of reception intensity 	No detection from vehicle parts with low reflectance	Reception lowering by low reflectance	Reception lowering depending on reflection area and incidence angle	Reception lowering by low reflectance	<ul style="list-style-type: none"> Recognition failure by low contrast with background with similar colors Recognition ability degradation by apathetic colors with surroundings 	<ul style="list-style-type: none"> Misrecognition depending on the width and length Recognition ability degradation depending on the shape 	—	<ul style="list-style-type: none"> Recognition ability degradation depending on inclination and driving direction 	<ul style="list-style-type: none"> Blurred Image
Principle	—	Low S/N	Low S/N	Low S/N	Signal intensity lowering	Signal intensity lowering	Signal intensity lowering	Low D/U	(Recognition)	—		

Table 24. “k-4. Pedestrian” disturbance elements

		Millimeter waves			LiDAR			Camera	
Influence on sensor principle	Class.	Wearing material	Posture/shape/size	Color (contrast ratio)	Shape/size	material	Color (contrast ratio)	Shape/size	Motion
Influence	—	<ul style="list-style-type: none"> Detection range lowering by reflectance lowering False perception by dispersion of reception intensity 	Reflection intensity lowering depending on body build and posture	Reception lowering by low reflectance	Reception lowering by change of reflection area depending on direction, size and posture	Reception lowering by low reflectance	<ul style="list-style-type: none"> Recognition failure by contrast lowering with similar color of background Recognition ability degradation caused by apathetic colors 	<ul style="list-style-type: none"> Misrecognition of distance depending on the size of pedestrians Small reflection and poor recognition for children Poor recognition for pedestrians with the height of 2m and more 	<ul style="list-style-type: none"> Misrecognition depending on walking direction Misrecognition depending on walking speed
Principle	—	Low S/N	Low S/N	Signal intensity lowering	Signal intensity lowering	Signal intensity lowering	Blurred Image	(Recognition)	Blurred Image

4.2.1.2. Generation principle of sensor perception disturbance

The sensor can potentially experience perception disturbance when detecting objects because of the factors discussed in the preceding section. While the principle of perception disturbance generation is different for each sensor, they can be categorized as per the following common perspectives.

- The sensor disturbance principles are classified into “those occurring due to perception processing”, “those occurring due to cognitive processing” and “others”.
- The disturbances occurring because of perception processing are classified into those related to the signal from the perception target (S) and those that hinder the signals from the perception target (noise N, unnecessary signal U).
- List the disturbances that can occur on signals individually related to S, N and U.

The examples of categories of generation principles of perception disturbances that could occur on each sensor based on these perspectives are as follows.

- Generation principle of perception disturbance of millimetre-wave radar.

The perception disturbances that occur on millimetre-wave radar includes those caused by the direction of the sensor, those occurring because of perception processing and those occurring because of cognitive processing (Figure 57).

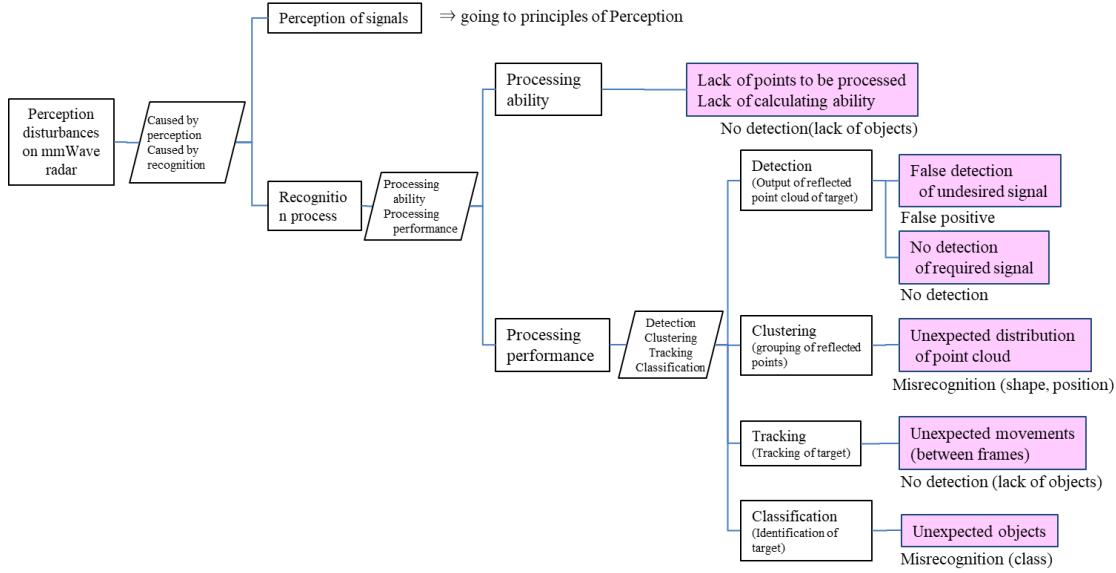


Figure 57. Categories of perception disturbances for millimetre-wave radar

In particular, the physical quantities that characterize the signal S in perception processing of millimetre-wave radar are the following three: frequency, phase and strength (Figure 58).

- Frequency: Problem with the signal frequency can be cited as a disturbance originating from the sensor itself.
 - Phase: There are cases where the direction the signal is arriving from changes and cases where the amount of propagation delay changes, and the changes in signal arrival direction are attributed to reflection and refraction.
 - Signal strength: The conceivable situations include partial signal loss, a signal that is too strong, a large difference in signal strengths, and the signal being too weak.

Furthermore, possible disturbances in regard to the noise N and the unnecessary signal S in perception processing include low S/N, low D/U (ratio of strength between the necessary signal D and unnecessary signal U) and increase of U.

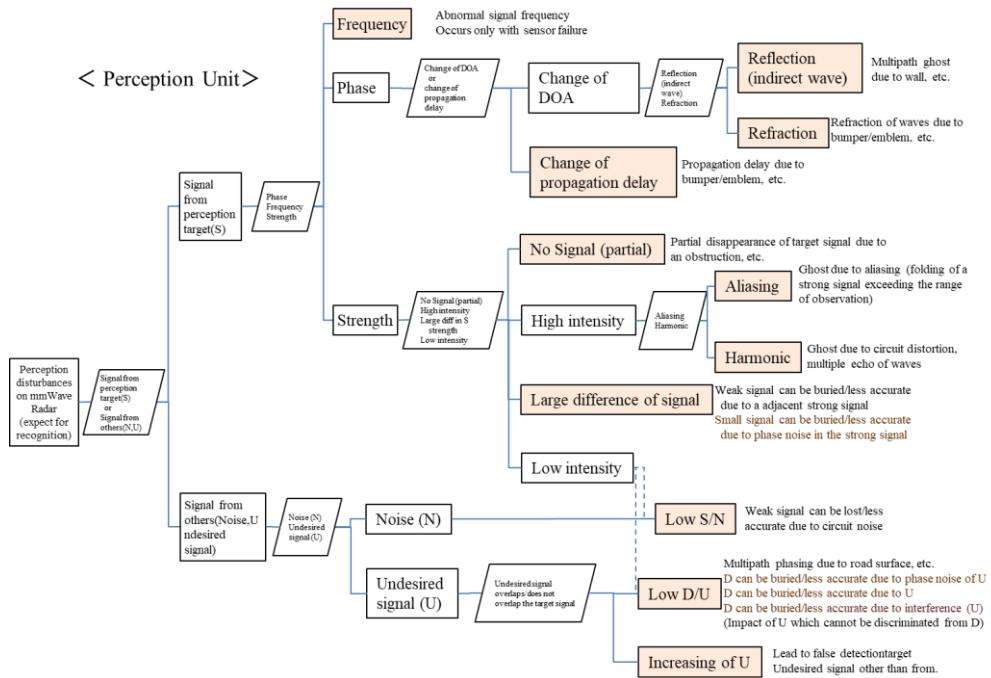


Figure 58. Generation principle of disturbance in millimetre-wave radar perception processing

- Generation principle of LiDAR perception disturbance

The physical quantities that characterize the signal S in perception processing of LiDAR are the scan timing, strength, propagation direction and velocity.

- Scan timing: The time difference because of the movement of the ego vehicle leads to positional shifts in the overall space; moreover, the time difference caused by the movement of the perception target leads to its positional shift.
- Strength: Phenomena include saturation, attenuation and shielding.
- Propagation direction change: There are those caused by reflection and those caused by refraction.
- Velocity: While it affects the arrival time of signals, there are no corresponding items in perception disturbance of LiDAR.

Furthermore, the noise N and unnecessary signal U include reflection and refraction from objects other than the perception target, in addition to DC noise, pulse-like noise and multiple reflections (Figure 59).

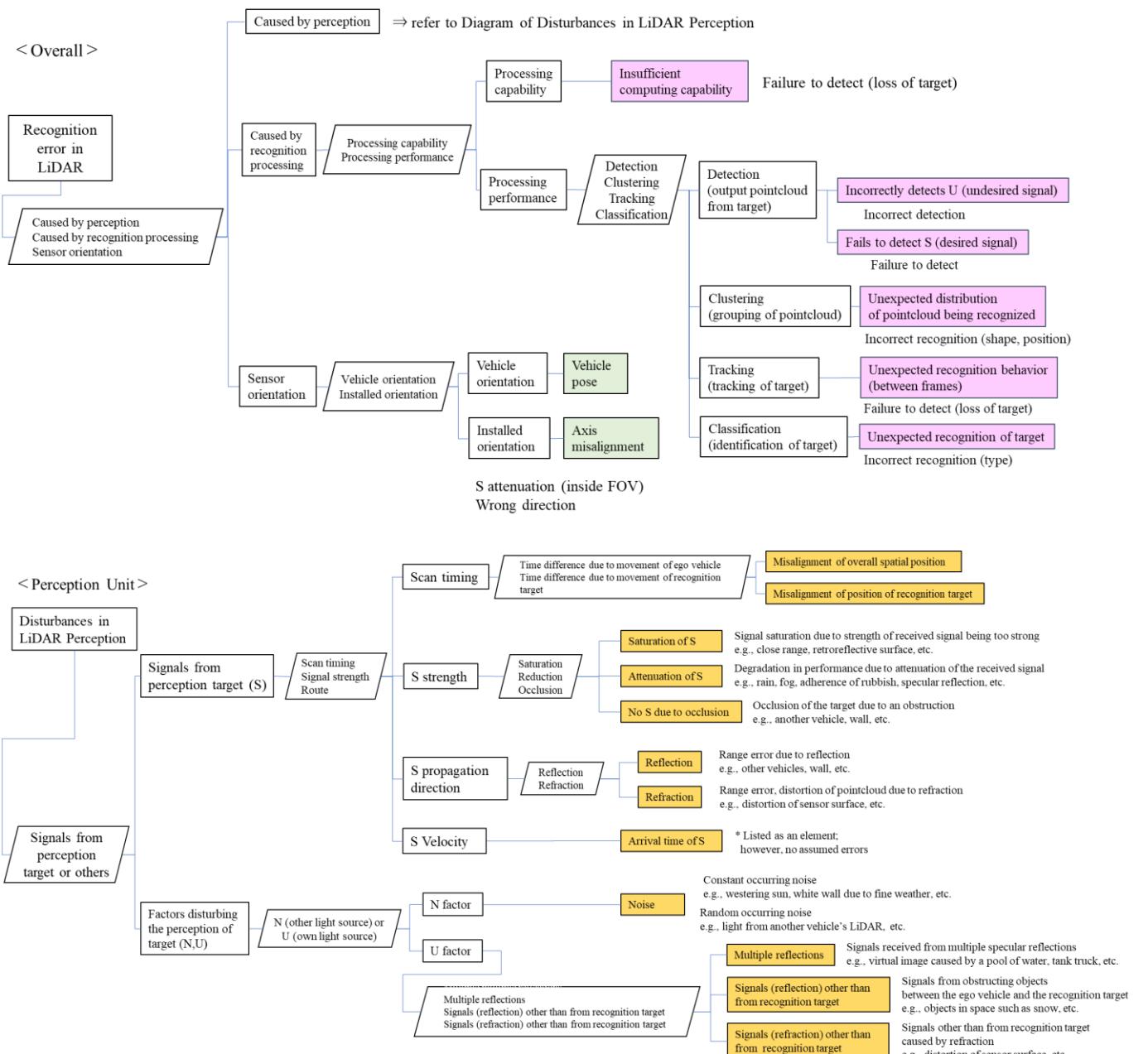


Figure 59. Generation principle of disturbance at perception of LiDAR

- Generation principle of perception disturbance at the camera
The physical quantities that characterize the signal S in perception processing of the camera are the strength, direction/range signal change and acquisition time.
 - Strength: There are cases where the signal is too weak, the signal is too strong, the difference in signal strength is large and the signal is partially lost.
 - Direction/range: There are changes caused by refraction and changes caused by reflection.
 - Changes in the signal S.

- Acquisition time: The possible cases of disturbances caused by blinking of the perception target and changes in relative positions include flickering and image blur/ deletion.

Furthermore, the noise N and unnecessary signal U include low D/U and low S/N (Figure 60).

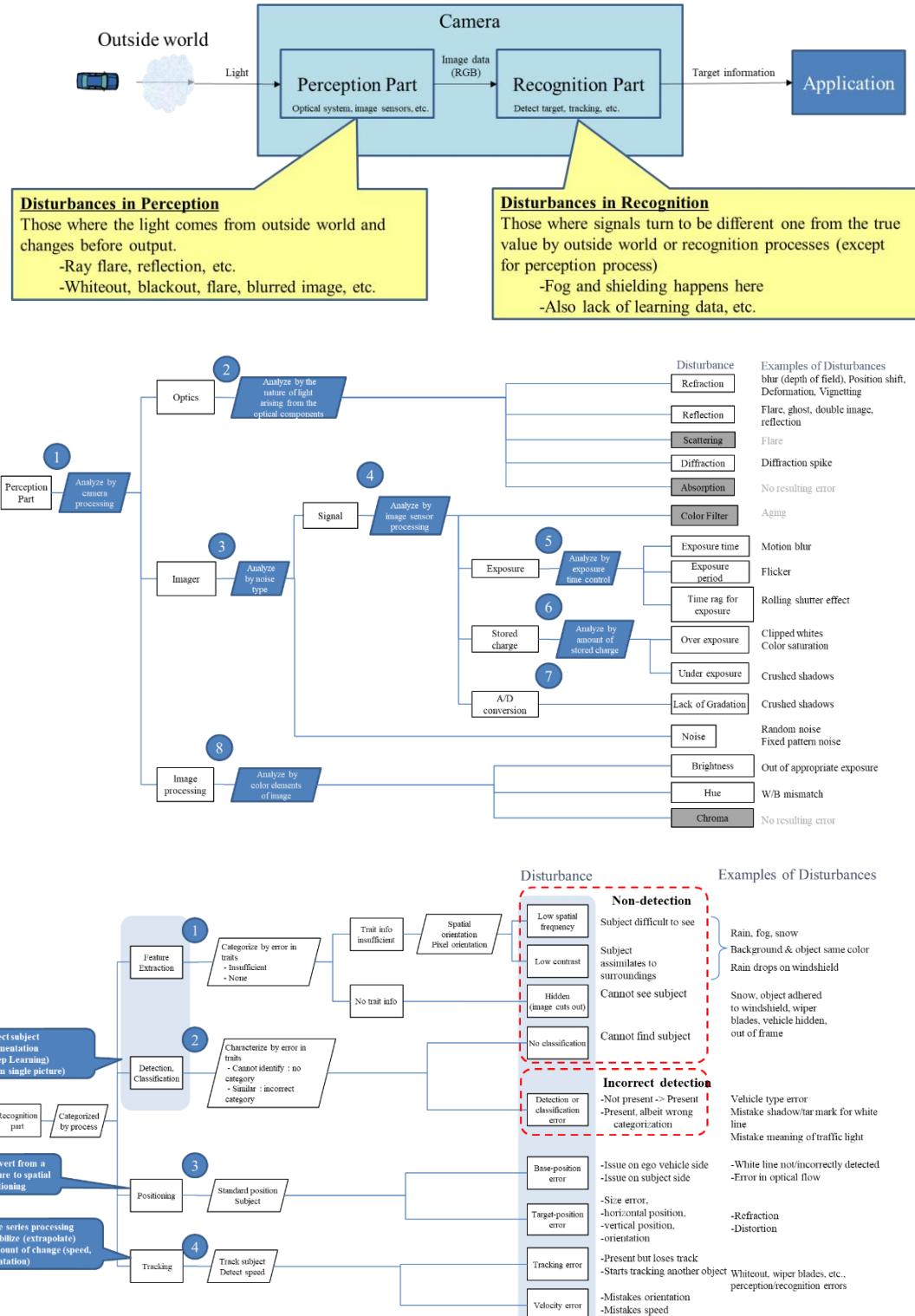


Figure 60. Generation principle of disturbance in camera perception

Scenario selection through cross-checking of perception disturbance elements and generation principle

The relationship between the elements of perception disturbance at each sensor and the generation principles can be represented in the matrixes shown in Tables 25–27. These matrixes list the perception disturbance elements vertically and generation principles horizontally, which makes it possible to understand the elements (= line) that can potentially cause the generation principle (= column). The several disturbance elements that can be reported in the same column are generated by the same principle. However, from the perspective of a system safety evaluation, it is possible to select the elements whose degree of influence on the perception performance of each sensor and encounter probability in the market are high, as well as prioritize them as evaluation scenarios.

In the following example, the influence on perception performance and the encounter probability can be compared in the same column by scoring based on the following concepts.

- Degree of influence on the perception performance (X): The extent of perception disturbance each element can cause in each principle is indicated by a score of 1–3. (Small influence: 1, medium: 2, large: 3)
- Encounter probability (P): The score of (frequency total score) × (duration of each occurrence) that is explained below is treated as the encounter probability, and scores of 1–3 are allocated according to its size. (Small probability: 1, medium: 2, large: 3)
 - Encounter frequency: The locality, influence of climate/weather, spatial density and occurrence frequency by usage are individually scored from 1 to 4, and the four scores are combined to calculate the “total frequency score”.
 - Duration of each occurrence: The duration of an occurrence of a given element is represented in scores from 1 to 3. (Short duration: 1, medium: 2, long: 3)
- The weights (W) of the degree of influence on the perception performance (X) and encounter probability (P) are set at 10 and 8, respectively, and their total ($W_X X + W_P P$) is calculated to obtain the total score.

When the calculation results of each element within the same principle (same column) are compared by following the abovementioned conception, and the one with the highest score is selected, it can be used as the representative evaluation scenario of this principle. Tables 25–27 show the examples of score calculation using the abovementioned method for the three sensors of millimetre-wave radar, LiDAR and camera. The field coloured in red in each column is that of the element whose total score is the highest in the corresponding principle. In other words, this element is the candidate for the representative scenario.

When there are several elements that have the highest score, one or several elements are selected while taking the reproducibility of the evaluation environment of that scenario into account and evaluating the same. Moreover, when there are disturbance elements that do not match the given sensor among the items represented in the vertical axis because of the specifications of the ADS under evaluation (such as ODD and perception target), exclude them and select the representative scenario among the remaining elements.

Table 25. Perception disturbance elements and generation principle matrix of millimetre-wave radar

Table 26. Perception disturbance elements and generation principle matrix of LiDAR

Environment			Perception Error										Recognition Error										
			Signals from perception target (S)					Factors disturbing the perception of target (N,U)					Processing capability					Processing performance					
			Scan Timing	S strength	S propagation direction	S Velocity	N factor	U factor	Noise	Multiple reflections	Signals (reflection) other than from recognition target	Signals (refraction) other than from recognition target	Insufficient number of processing points	Insufficient computing capability	Incorrectly detects S (undesired signal)	Fails to detect S (desired signal)	Detection	Clustering	Tracking	Classification			
vehicle, sensor	Ego vehicle	Change of vehicle pose	due to vehicle condition (semi-permanent) due to vehicle condition (temporary)																				
	Sensor	Variation of installation	axial deviation (inside adjustment range) axial deviation (outside adjustment range)																				
		Failure of sensor itself	degradation of sensor surface degradation of sensor itself (electronic components) degradation of electrical performance due to external noise																				
		Surface in front of the sensor	water ice snow mud / dust car wash wax foreign matter (insects, bird droppings) x SPOT changes in characteristics																				
	Road Surface	Sticking objects	uphill downhill road cant puddle frozen road traces of road repair rut snow cover																				
		Road condition	asphalt concrete gravel sand thin layer pavement cobblestone road manhole road joint (metal joint) road joint (asphalt type joint)																				
		Material	reflective mirror curve mirror overhead objects																				
		Roadside objects	curved road occlusion																				
		Overhead objects	reflective mirror occlusion																				
		Spatial obstacles	snow rain sand fog others / floating in space insects / floating in space																				
		Radiowave and light in space	direct wave x other vehicle direct wave x infra-structure direct wave x nature world																				
		Other moving objects	reflective mirror color / materials shapes grime / thin spot relative position																				
track	Obstacles on the lane	lines																					
		structural objects with height	color / materials shapes grime																				
		without a step	relative position	o																			
		with a step	color / materials shapes grime																				
		fallen object	relative position	o																			
	animals	color / materials shape / size	relative position / motion	o	o																		
		temporal installed object	color / material shape / size grime																				
		other vehicles	relative position	o																			
Moving objects	motor bikes	color / materials shape / size	relative position	o	o																		
		sticking objects																					
		relative position	o	o																			
		color / materials shape / size	relative position	o	o																		
		bicycles	color / materials shape / size																				
	pedestrians	color / materials shape / size	relative position	o	o																		
		relative position	o	o																			
		color / materials shape / size	relative position	o	o																		
		pedestrians	color / materials shape / size																				
		relative position	o	o																			

Small impact
Medium impact
Great impact

Table 27. Perception disturbance elements and generation principle matrix of the camera (element: vehicle/sensor, surrounding environment)

Table 28. Perception disturbance elements and generation principle matrix of the camera (element: perception target – route/traffic information/obstacle)

Disturbance cause factor		Disturbance cause factor		Perception process																		Recognition part										Others																													
				Reduction		Oppose		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma		Low spatial frequency		Low contrast		Focus suppression		Haze		No classification		Detection or classification error		Perception		Tracking error		Timing		Vehicle area							
Model class	Causal factor group	Disturbance angle, area, position		Motion		Vibration		Gloss (Luminance)		Reflection (WCS)		(Plane)		(Diffraction spike)		(Random pattern)		(Aging)		(Motion blur)		(Picker)		Blurring shutter effect		Clipped whites		Color saturation		Crushed shadows		Crushed highlights		Out of exposure		WB deviation		(Hard to see)		(Difficult to separate target from circumference)		(Visible)		(Indirection)		(False positive detection)		(Classification error)		Self-position error		Target position error (Type detection error)		Tracking to target project		Tracking error		Magnitude error			
		Distance		Blur (Depth of field) (Focusing distance), (Focusing position), (Viewing)		Focusing		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma		Low spatial frequency		Low contrast		Focus suppression		Haze		No classification		Detection or classification error		Perception		Tracking error		Timing		Vehicle area		Small effect					
Lanes	Relative position	Distance		Blur (Depth of field) (Focusing distance), (Focusing position), (Viewing)		Focusing		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma		Low spatial frequency		Low contrast		Focus suppression		Haze		No classification		Detection or classification error		Perception		Tracking error		Timing		Vehicle area		Small effect					
		Distance		Blur (Depth of field) (Focusing distance), (Focusing position), (Viewing)		Focusing		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma		Low spatial frequency		Low contrast		Focus suppression		Haze		No classification		Detection or classification error		Perception		Tracking error		Timing		Vehicle area		Medium effect					
		Distance		Blur (Depth of field) (Focusing distance), (Focusing position), (Viewing)		Focusing		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma		Low spatial frequency		Low contrast		Focus suppression		Haze		No classification		Detection or classification error		Perception		Tracking error		Timing		Vehicle area		Large effect					
		Distance		Blur (Depth of field) (Focusing distance), (Focusing position), (Viewing)		Focusing		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma		Low spatial frequency		Low contrast		Focus suppression		Haze		No classification		Detection or classification error		Perception		Tracking error		Timing		Vehicle area		Large effect					
		Distance		Blur (Depth of field) (Focusing distance), (Focusing position), (Viewing)		Focusing		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma		Low spatial frequency		Low contrast		Focus suppression		Haze		No classification		Detection or classification error		Perception		Tracking error		Timing		Vehicle area		Large effect					
		Distance		Blur (Depth of field) (Focusing distance), (Focusing position), (Viewing)		Focusing		Scanning		Diffraction		Absorption		None		Color filter		Exposure time		Exposure period		Time lag for exposure		Over exposure		Under exposure		Lack of resolution		Brightness		Blur		Chroma																											

Table 29. Perception disturbance elements and generation principle matrix of the camera (element: perception target – moving object)

The following are examples of scenarios selected as the perception disturbance representative scenarios of the millimetre-wave radar, LiDAR, as well as the camera by taking the degree of influence on the sensor perception performance and encounter probability as per the abovementioned conception (Figures 61, 62, and 63).

As an example, Fig. 61 shows one of the scenarios selected by the abovementioned conception. The scenario illustration should include the following elements: See ANNEX for details.

- ✓ Outline explanation of the scenario
- ✓ Illustration of the recognition target, surrounding environment, own vehicle / sensor status in the scenario
- ✓ List of parameter items and ranges

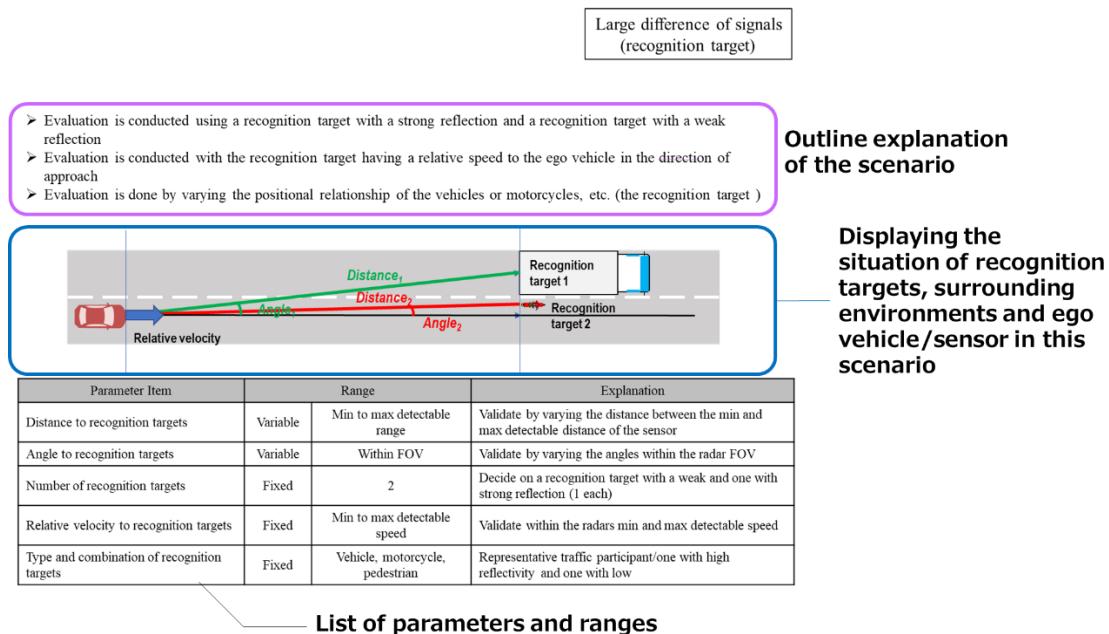


Figure 61. Example of recognition disturbance evaluation scenario explanatory diagram

4.2.1.3. Evaluation of Perception Disturbance Combination

It is possible for multiple elements of perception disturbance to occur in one sensor at once. When these several elements strengthen the influence on the perception performance of each other, a perception performance evaluation that combines these elements becomes necessary. Whether the elements strengthen each other must be considered based on the generation principles of perception disturbance; the influence must be determined as per the principles among different columns in the matrixes from the preceding section. The principles that weaken or do not influence each other as per the result of verification are excluded from the combination evaluation (Figure 62).

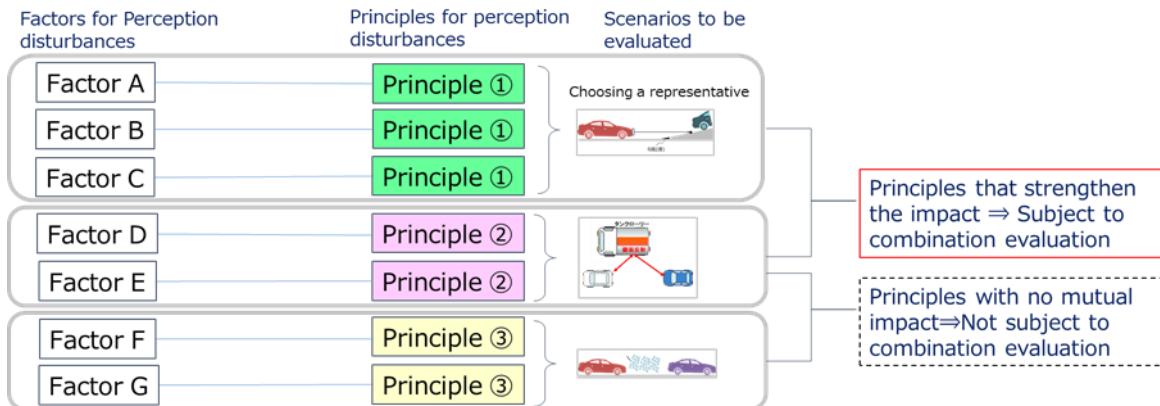


Figure 62. Perception disturbance generation principles that are the subjects of combination evaluation

4.2.1.4. Perception disturbance evaluation of automatic driving system equipped with several sensors

Commonly, ADS construct sensor fusion systems that combine several sensors. When evaluating the perception performance of the system as a whole, a unified scenario based on the sensor composition that gathers the evaluation scenarios of each individual sensor selected through the aforementioned process is used, and the system as a whole under each disturbance condition is evaluated.

4.2.2. Blind Spot Scenarios

The premise of the aforementioned (Chapter 3.1) traffic disturbance scenario structure is that the surrounding vehicles are detectable. However, in an actual traffic environment, certain surrounding vehicles or road components can sometimes cover other surrounding vehicles (hereafter referred to as peripheral vehicles). Therefore, it is necessary to consider safety-related scenarios that include the peripheral vehicles in blind spots and integrate them into the safety analysis.

The blind spot scenarios are classified into three sub-categories, namely, the peripheral vehicles, the road structure and the road shape (Figure 65).

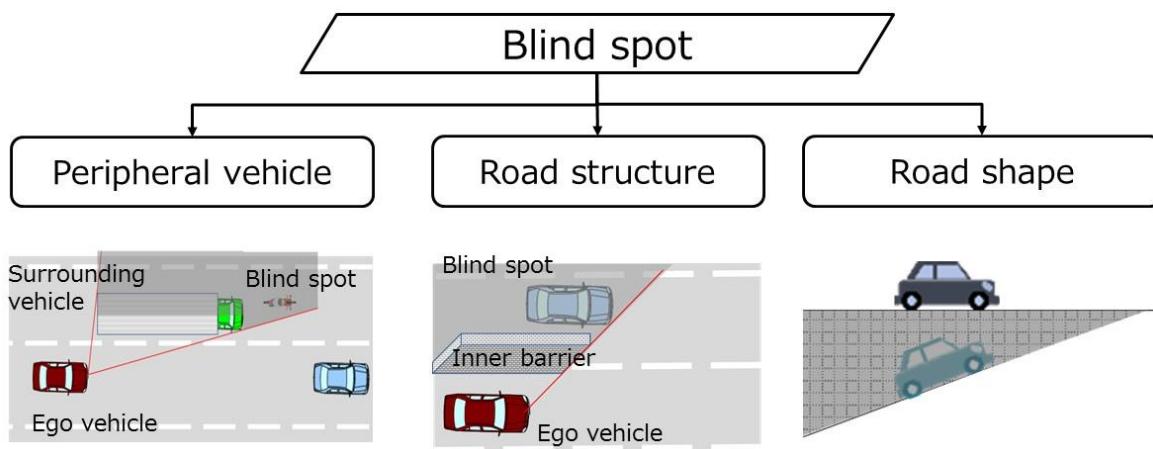


Figure 65. Perception disturbance categories related to blind spot

4.2.2.1 Blind spot scenarios caused by peripheral vehicles

Sixteen new position definitions were added to the eight surrounding vehicle positions to date defined to structure the blind spot scenarios caused by the peripheral vehicles (Figure 66). Beware that each peripheral vehicle can create blind spots that affect other peripheral vehicles, in addition to the vehicle immediately behind it. This is particularly true when the blind spot area and the positions of the vehicles inside that area change, e.g., when the ego vehicle and the surrounding vehicles are driving on a curve.

To elucidate this dynamic phenomenon, an additional figure and explanation are presented as follows. Figure 67 shows the process to explain the blind spots of peripheral vehicles derived as the combination of the ego vehicle, curvature of the roads in the same lane, and the peripheral vehicles. Similarly, Figure 68 and Figure 69 show the blind spots related to the peripheral vehicles at lateral or diagonal positions to the ego vehicle.

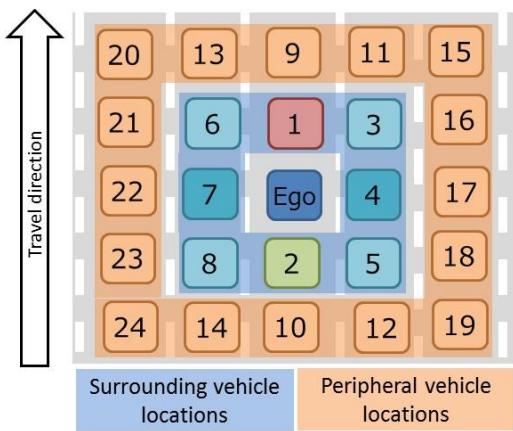


Figure 66. Vehicle positions applied to define the peripheral vehicles-related blind spot scenarios

Figure 67 shows the blind spot positions that are generated when the peripheral vehicle is at position 1. In the figure, a picture of a truck is used to make it more understandable. The only blind spot position generated by the truck on a straight road is position 9. However, when both the ego vehicle and the truck pass the right curve, the position of the truck in relation to the ego vehicle changes and blind spots are generated at vehicle positions 6, 9, 13, 20 and 21. Similarly, for the left curve, the vehicles at positions 3, 9, 11, 15 and 16 could be hidden by the truck. Therefore, nine total blind spots positions (3, 6, 9, 11, 13, 15, 16, 20 and 21) are added, and they can potentially lead to risky operations. There are positions that are in inclusion relation among the nine blind spot positions. For instance, at the right curve, a lane change at blind spot position 20 is a movement toward blind spot position 13. Blind spot position 13 is closer to the ego vehicle than blind spot position 20; it is a more difficult condition that has a shorter amount of time for reaction. Thus, by performing a safety evaluation on blind spot position 13, the dangerous motions of blind spot position 20 can be included. Following the same theory, the blind spot positions 15, 16 and 21 can be removed from the final list of blind spot positions. Therefore, the blind spot positions induced by the vehicle on the position 1 that are considered in the safety analysis in the end are reduced to five (3, 6, 9, 11 and 13). These five positions are summarized in the simplified rectangular diagram in Figure 67.

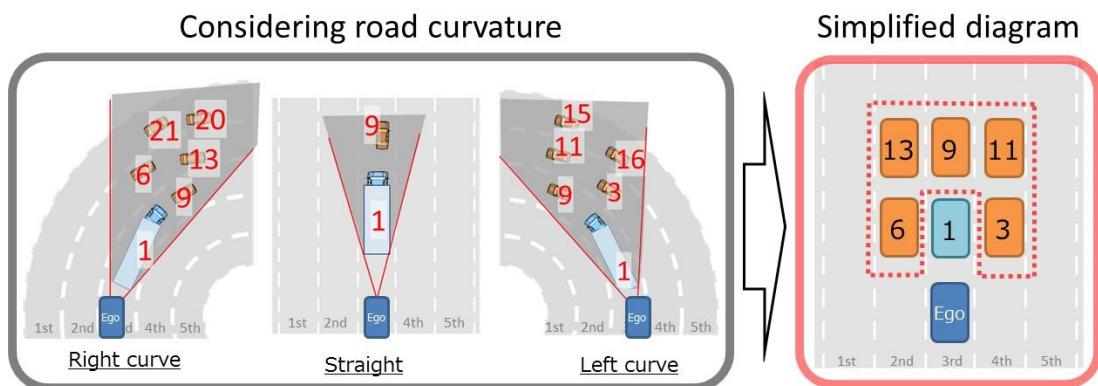


Figure 67. Blind spot positions generated by the peripheral vehicle at front position 1

Figure 68 shows every blind spot position generated by the truck on peripheral vehicle position 4. On straight roads, five blind spot positions (3, 5, 16, 17 and 18) can be extracted from the truck. When both the ego vehicle and the truck pass a right curve, the number of blind spots increases to 11 peripheral vehicle positions (1, 2, 3, 5, 6, 8, 16, 17, 18, 21, and 23). At a left curve, the vehicles at these three positions (16, 17, and 18) can become

hidden. In this case, a reduction in the blind spot positions to be considered in the safety analysis is performed, e.g. if the vehicle at position 6 changes lanes to the next one on its right, it moves to the same position as position 1. Thus, when performing a safety analysis, the vehicle at position 1 covers the operations of the vehicle at position 6 following the principle of the most difficult scenario. The same theory can be applied to the vehicles changing lanes to the next one on the right from positions 21, 8 and 23. Deceleration by the vehicle at position 6 has a requirement such that the simultaneous lane change to the next one on the left by the ego vehicle and the vehicle at the position 1. Thus, the vehicle at position 6 can be replaced by the vehicle at position 1. Similarly, acceleration by the vehicle at location 8 is less important than the simultaneous lane change by the ego vehicle and vehicle 2. Furthermore, the cut-in scenarios by vehicles 16, 17 and 18 are excluded from the analysis because vehicle e4 is next to the ego vehicle, which prevents the ego vehicle from changing lanes. Therefore, the number of blind spot positions generated by the vehicle on position 4 considered in the safety analysis in the end is reduced to four (1, 2, 3 and 5), and these are summarized in the simplified diagram on the right of Figure 68.

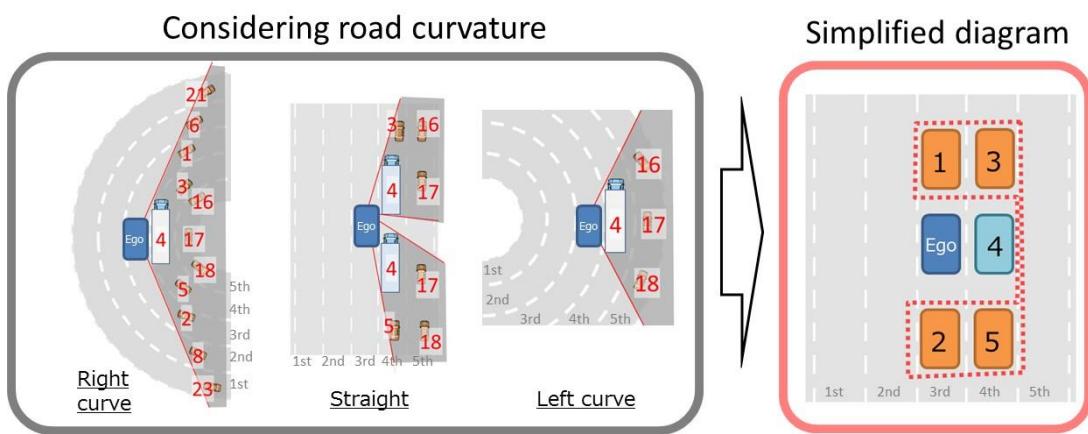


Figure 68. Blind spot positions generated by the peripheral vehicle at lateral position 4

Figure 69 shows every blind spot position generated by the truck on the peripheral vehicle position 3 that are diagonal to the ego vehicle. On a straight road, the truck can generate three blind spot positions (11, 15 and 16). When both the ego vehicle and the truck pass a right curve, the blind spots increase to nine peripheral vehicle positions (1, 6, 9, 11, 13, 15, 16, 20 and 21). On a left curve, positions 15 and 16 become blind spots. As was with the previous case shown in Figure 68, the cut-in scenarios by the vehicles at positions 6, 13, 20 and 21 can be replaced by more difficult scenarios of the vehicles on positions 9 and 11. Moreover, the deceleration scenarios of vehicles 6 and 13 can be replaced by the motions of simultaneous lane change to the left by the ego vehicle and vehicle 9. Lastly, the number of blind spot positions generated by the vehicle at diagonal position 3 considered in the safety analysis is reduced to five positions (1, 9, 11, 15 and 16). These are shown in the simplified rectangular diagram on the right of Figure 70.

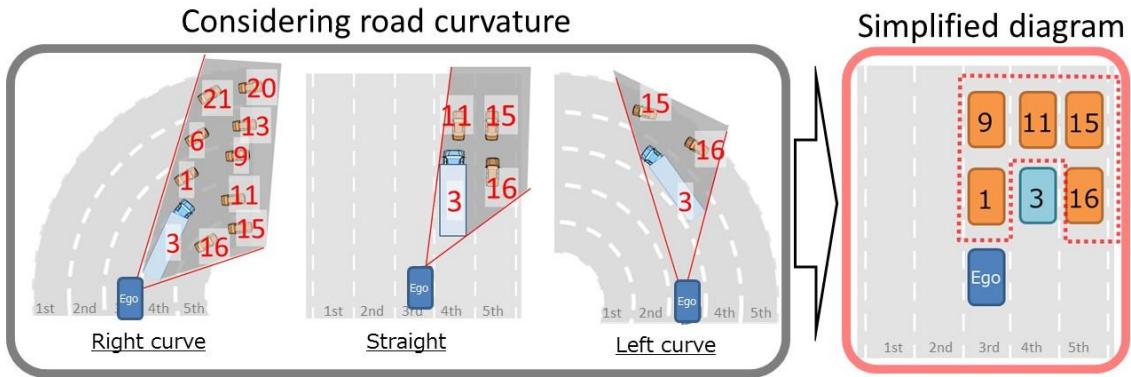


Figure 69. Blind spot positions generated by the peripheral vehicle at position 4

By applying the principles of analogy and symmetry to the three cases shown in Figures 67–69, all the positions considered in the safety analysis can be summarized in a single diagram (Figure 70).



Figure 70. Diagram of all the blind spot positions generated by the peripheral vehicles considered in the safety analysis

The possible blind spot-generating vehicle motions are classified into cut-in, cut-out, acceleration, deceleration and synchronization. The reduction in the number of combinations to be considered in the safety analysis is performed by focusing on the motions of blind spot vehicles that can potentially hinder the behaviour of the ego vehicle (Figure 71). For instance, all the deceleration operations of the vehicles that are in the blind spots behind the ego vehicle (2, 5, 8, 10, 12, 14, 18, 19, 23 and 24) are excluded because they do not pose a danger to the ego vehicle. Moreover, the synchronization between the ego vehicle and the blind spot vehicles does not pose a danger to the ego vehicle. The circles in the figure indicate the corresponding combinations of the positions of blind spot vehicles and their motions that can potentially hinder the ego vehicle; thus, it is necessary to consider these in the safety analysis.

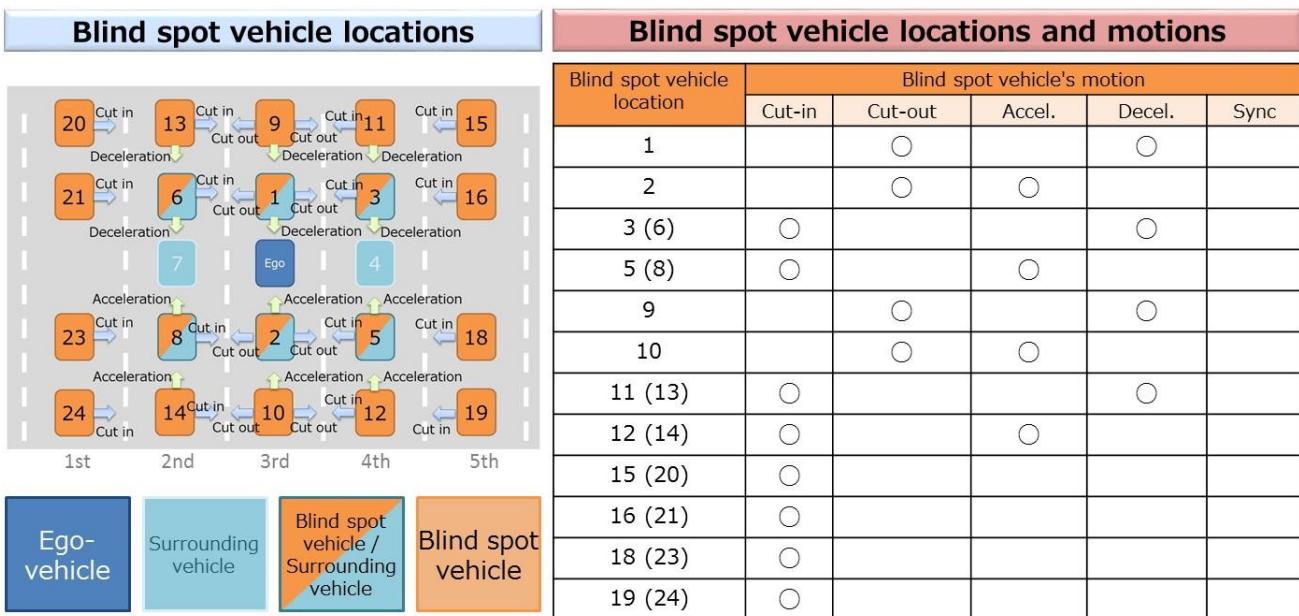


Figure 71. Positions of blind spot vehicles (left) and the combinations of the positions of blind spot vehicles and the motions that can potentially hinder the ego vehicle (right)

Because of the systemization process discussed to date, a structure that contains all the blind spot scenarios that involve surrounding vehicles (road geometry, ego vehicle behaviour, blind spot vehicle motions and combinations of peripheral vehicle motions) has been defined. This structure comprises a matrix that contains 64 total possible combinations, of which 42 correspond with realizable scenarios in an actual traffic flow (Figure 72).

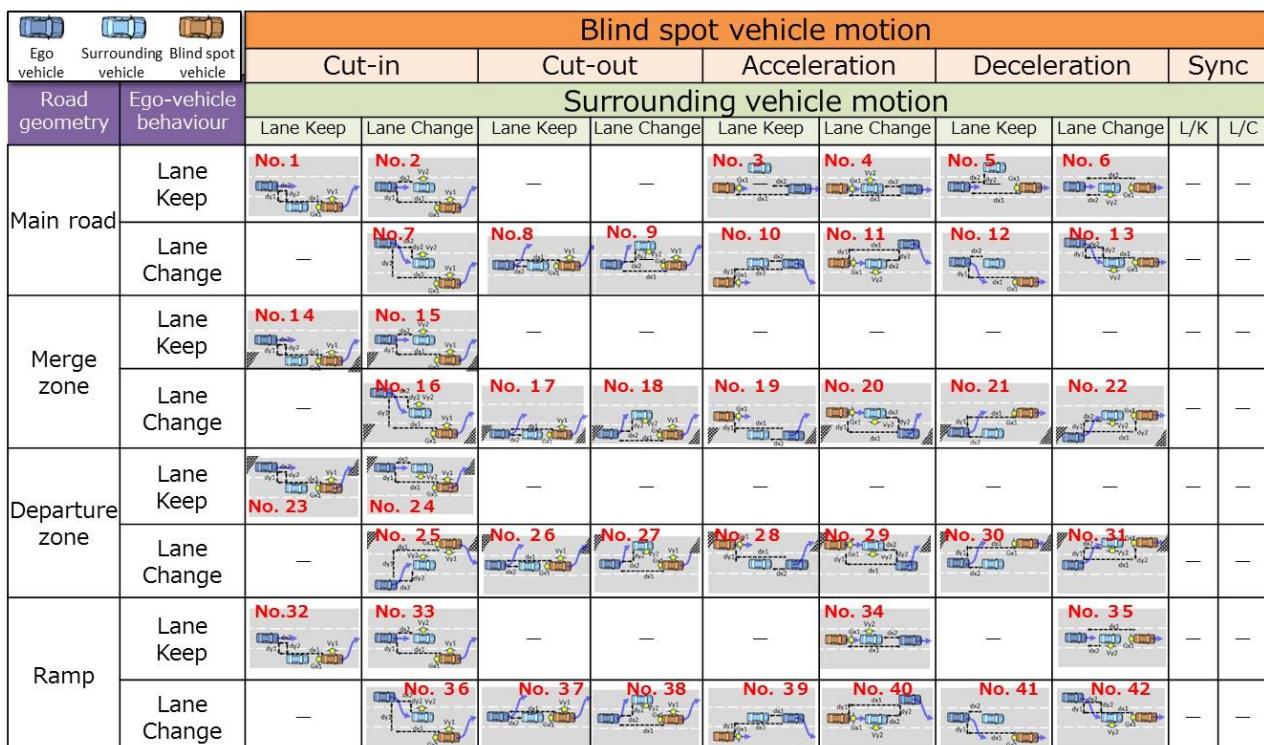


Figure 72. Perception disturbance scenarios related to blind spots generated by surrounding vehicles

4.2.2.2. Blind spot scenarios generated by road structures

The blind spot scenarios related to road structures are defined by considering the road structure positions and the relative motion patterns between the ego vehicle and blind spot vehicles. Generally, these blocking elements exist inside the road structures, and are classified into inner barriers and outer walls according to the types and positions of road structures (Figure 73).

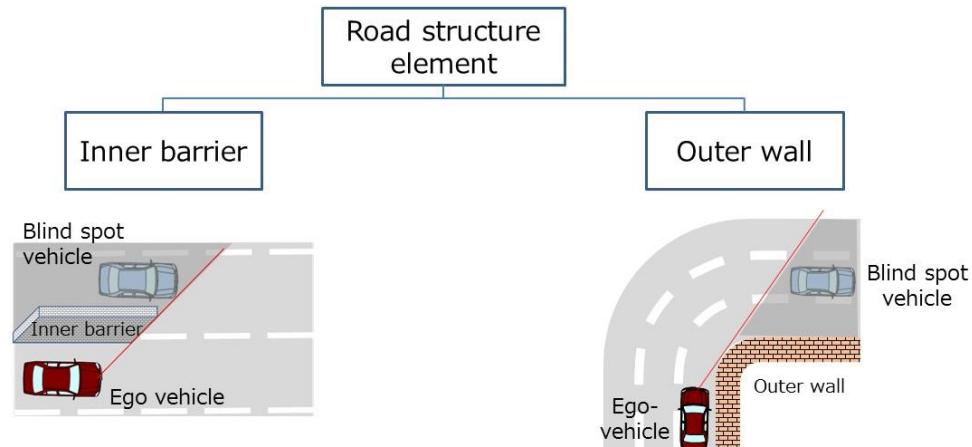


Figure 73. Categories of blind spot scenarios generated by road structures

4.2.2.2.1. Blind spot scenarios generated by inner barriers

As shown in Figure 74, the vehicle behind the structure (vehicle 1) cannot be perceived when the ego vehicle is reaching toward of the structure; it can be regarded as a blind spot vehicle. The situation is the same when the ego vehicle is in front of the structure, and the blind spot vehicles are at the back (vehicle 3) and at the front (vehicle 4). The vehicle at the centre of the structure is not considered to affect the safety. This is because the vehicle next to the blind spot cannot reach the lane of the ego vehicle because of the structure. However, if the blind spot vehicle is diagonally positioned behind the ego vehicle (vehicle 2), there is a safety concern in case it appears immediately after the end of the structure.

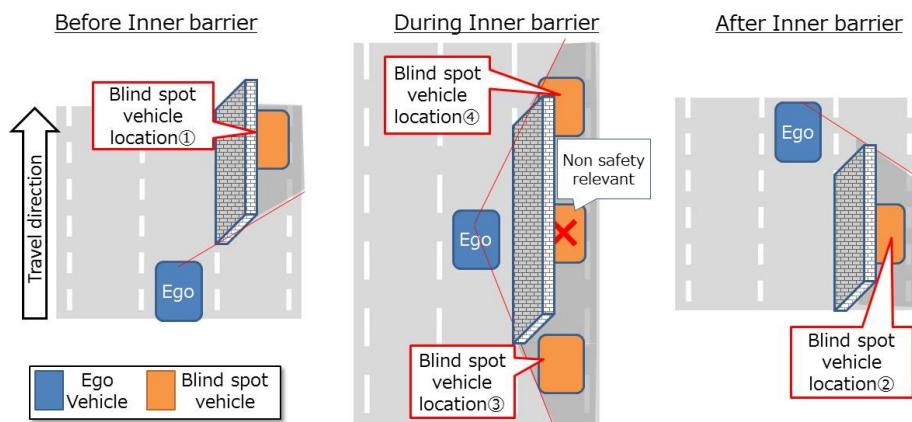


Figure 74. Definitions of blind spots related to inner barrier

Figure 75 summarizes, using a matrix, the blind spot recognition limit scenarios that are associated with inner barriers. The four blind spots mentioned above in the matrix (the ego vehicle is represented by the blue square and the blind spot vehicle is positioned in the dark grey area) are combined with the five possible operations that the vehicle in these blind spot areas can perform (cut-in, cut-out, acceleration, deceleration, and synchronization). The resulting matrix has 20 possible combinations, not all of which are safely related. In an inner barrier scenario, e.g., as the ego vehicle and the blind spot vehicle are in different lanes, this does not pose any danger. Furthermore, when the vehicles travel in parallel at the same speed with the inner barrier in between them, the ego vehicle and the blind spot vehicle cannot make contact with each other. Therefore, we can exclude all cut-out and synchronization scenarios. The safety analysis, therefore, incorporates a total of five inner barrier blind spot scenarios (marked with a circle in Figure 75).

Inner barrier related blind spot pattern	Blind spot vehicle's movement				
	Cut-in	Cut-out	Accel.	Decel.	Sync.
①				○	
②	○		○		
③	○				
④	○				

Figure 75. Blind spot-related recognition limitation scenarios due to inner barriers

4.2.2.2.2. Blind spot scenario due to outer barriers

Road structures, such as outer barriers, can create blind spots on curves. Figure 76 demonstrates that the outer barrier may become a blind spot for the front and rear vehicles depending on the curve angle. A vehicle, therefore, located in either the front lane or the rear lane (1, 2, 3, 5, 6, 8) of the ego vehicle may become a blind spot vehicle.

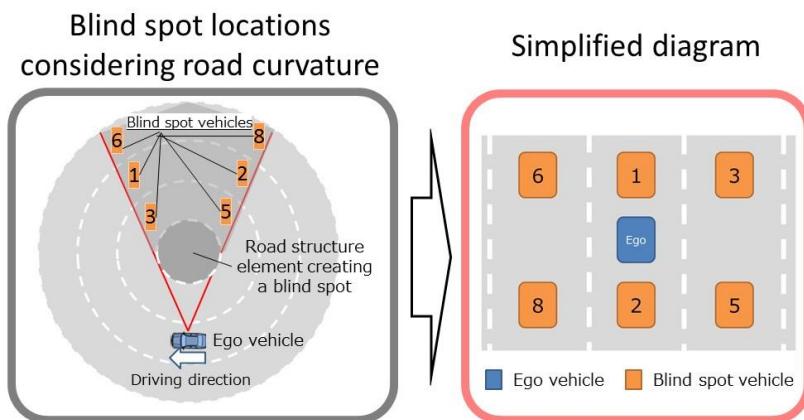


Figure 76. Definition of blind spots related to position and outer barrier

Figure 77 shows the movement of a blind spot vehicle in a situation in which it might interfere with the ego vehicle. Blind spot vehicle movements comprise cut-in, cut-out, acceleration, deceleration, and synchronization movements. The target pattern is one in which a blind spot vehicle enters the lane of the ego vehicle. However, scenarios where vehicles do not approach each other or when safety concerns are not raised such as when both the ego vehicle and the blind spot vehicle are running parallel (Sync) to each other on either side of the barrier are not covered in this scenario.

Outer wall related blind spot pattern	Blind-spot vehicle motion				
	Cut-in	Cut-out	Accel.	Decel.	Sync
1		○		○	
2		○	○		
3(6)	○			○	
5(8)	○		○		

Figure 77. Scenario in which blind spot-related recognition is limited due to partial barrier

4.2.2.3 Blind spot scenarios by road shape

Blind spot scenarios based on the road shape are defined as per the features of the road shape and the traffic patterns of the ego vehicle and the blind spot vehicle. Blind spots based on the shape of the road are created by height differences along the same road. We can characterize these particular road shapes as vertical curve and parallel slope shapes (Figure 78).

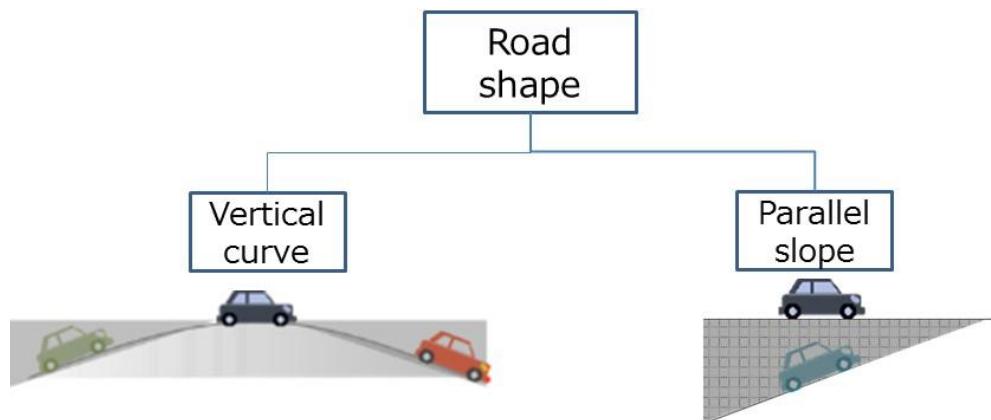


Figure 78. Blind spot scenario classifications based on road shape

4.2.2.3.1 Vertical curve scenario

The blind spot areas may occur in the front or rear when the road shape is that of a vertical curve (Fig.79). The vertical road gradient shortens the viewing distance of the vehicle. A potentially dangerous traffic pattern is created by the combination of the position and movement of surrounding vehicles (1, 2, 3, 5, 6, and 8) and the movement of the ego vehicle itself.

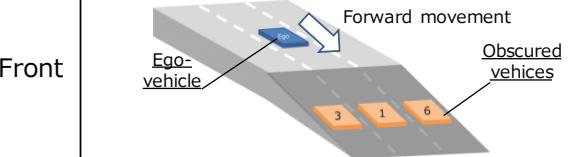
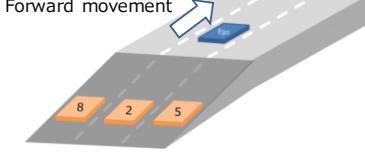
Blind spot area	Ego-vehicle and obscured vehicle positions
Front	 <p>Forward movement Ego-vehicle Obscured vehicles</p>
Rear	 <p>Forward movement Obscured vehicles</p>

Figure 79. Cognitive dysfunction related to blind spots caused by vertical curve

4.2.2.3.2 Gradient scenario for adjacent lane

A blind spot is created by the height difference because of the slope of the adjacent lane. These can be reported in junctions and branch roads. The blind spot caused by the combination of the particular road shape and the movement of the vehicle. A potentially dangerous traffic pattern is created by the position and movement of the vehicle hidden in the blind spot. These patterns can be classified into four groups: obscured vehicle cut-in (1), cut-out (2), acceleration (3), and synchronization with ego vehicle (4). This creates a matrix of 20 scenarios. We shall incorporate five of these scenarios into the safety analysis (Figure 80).

Ego-vehicle and obscured vehicle positions	Obscured vehicle motions				
	Cut-in	Cut-out	Accel	Decel	Sync
①				○	
②	○		○		
③	○				
④	○				

Figure 80. Parallel slope blind spot related cognitive disturbance scenarios

4.2.3. Communication disturbance scenario

Communication disturbance scenarios are defined based on the connectivity-related characteristics in the three categories of sensors, environment, and transmitter (Fig. 81).

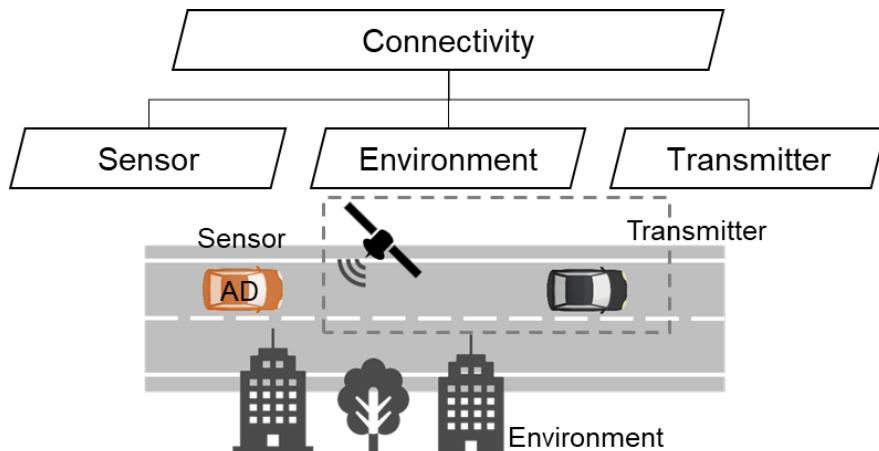


Figure 81. Classification of cognitive limits related to communication disturbances

4.2.3.1. Sensor type

Sensor-related communication disturbances are classified into the effects of digital map factors and the effects of V2X (Vehicle-to-everything) factors, as shown by Fig. 82.

Digital maps are used to support or implement positioning and navigation assistance, in addition to other capabilities required for ADAS / AD. Moreover, we can combine digital maps with perceptual sensors to increase the reliability of cognitive systems.

V2X allows vehicles to communicate with other vehicles, road infrastructure, pedestrians, and servers. The situation surrounding the vehicle is communicated to V2X in advance, which gives it an advantage, particularly in bad weather and complex traffic environments.

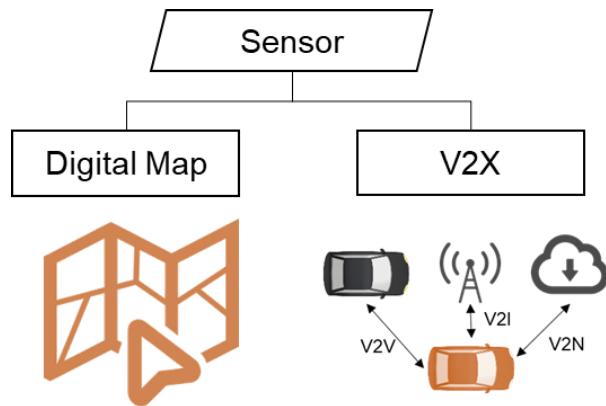


Figure 82. Classification of cognitive limits associated with sensor communication disturbances

If the map data is not correctly collected because of a flaw in the algorithm or incorrect data collection timing (such as temporary lane closure and road curvature change), a digital map-related communication disturbance may occur. The result of this is that obsolete data are collected. Poor fusion behaviour of the sensor affects, however, affects both the digital map and V2X. This may happen, for example, if the digital map, V2X and other sensors generate different information.

4.2.3.2 Environment type

Environment-related communication disturbances are shown in Figure 83. As seen from the figure, such disturbances comprise static entities, spatial entities, and dynamic entities. These interfere with communication and positioning signals. These can create blind spots and negatively affect the transmission of digital map and V2X signals.

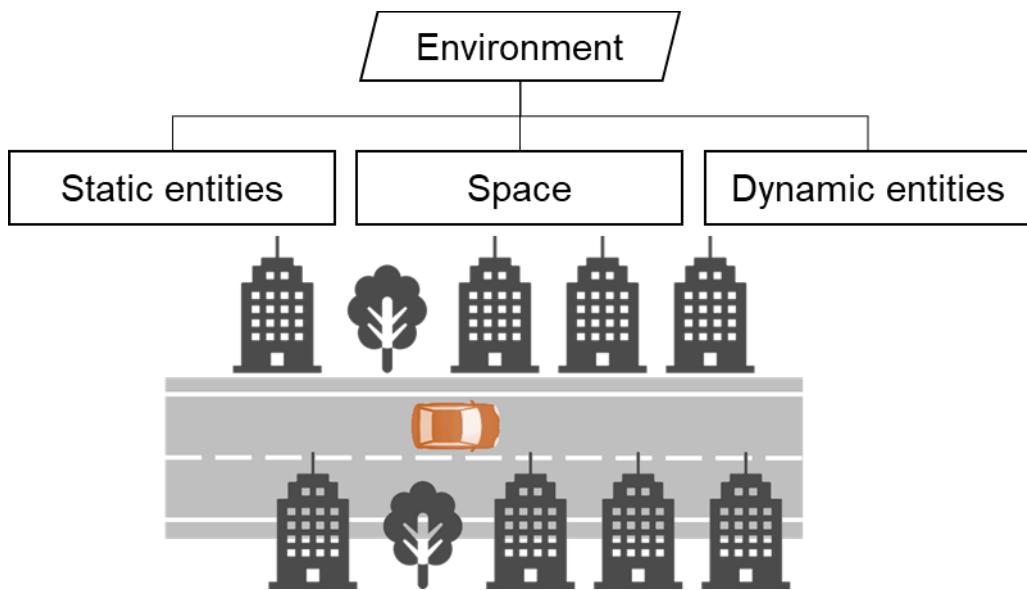


Figure 83. Classification of cognitive limits associated with environmental disturbances

Static entity factors include those related to roadside objects (such as buildings, trees, and tunnels), bridging structures (such as overpasses), and underground objects (such as parking lots). Connectivity failures may be caused by aspects of the surrounding environment of the vehicle (such as signal interference, rain and fog attenuation). Dynamic entities include such factors as surrounding vehicles, motorcycles, and pedestrians.

4.2.3.3. Transmitter classification

Transmitter-related communication disturbances shown in Figure 84. These can be classified into those caused by other vehicles, infrastructure, pedestrians, servers, and satellites. V2X messages may become unavailable or unreliable because of transmitter errors, while satellite errors can cause GNSS signals to be lost or overlooked.

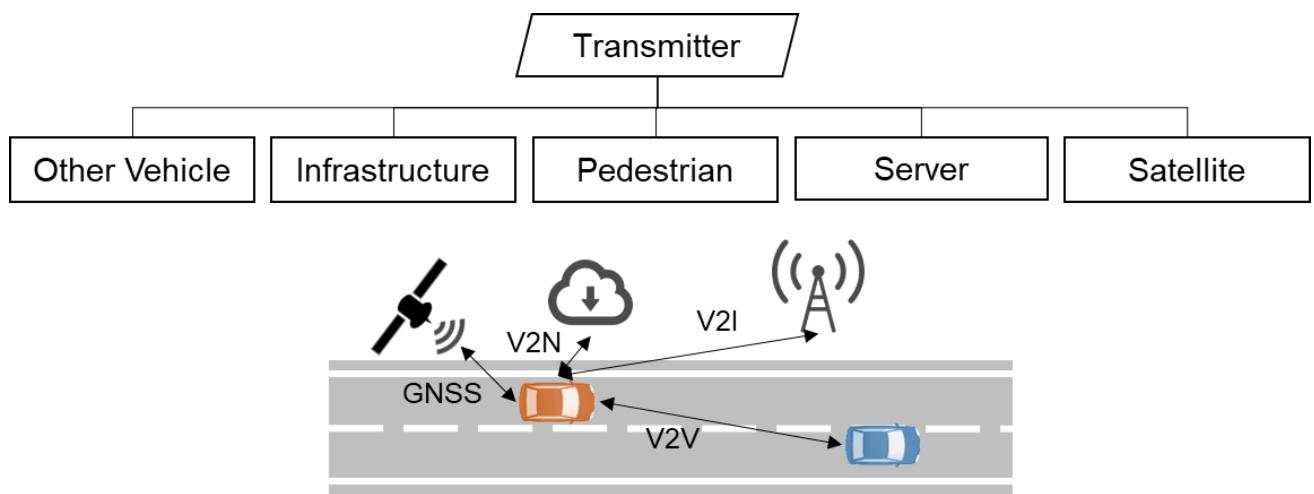


Figure 84. Classification of cognitive limits related to transmitter communication disturbances

4.3. Vehicle motion disturbance scenarios

In this section, we will explain our thinking regarding the setting of the system and standards for vehicle motion disturbance scenarios with the aim of ensuring the safety of AD. In vehicle motion disturbance, a safe state is one in which “an accident does not occur even if the vehicle motion changes due to a sudden disturbance.” The two types of effects on vehicle movement are factors that exert an external force on the vehicle body and affect lateral/front-back and unidirectional movement, in addition to factors that cause the tyre generation force to fluctuate and affect the lateral/front/rear/up/down and yaw direction of the vehicle (Fig. 85). Therefore, vehicle motion disturbance scenarios can be classified into vehicle body inputs and tyre inputs (Fig. 86).

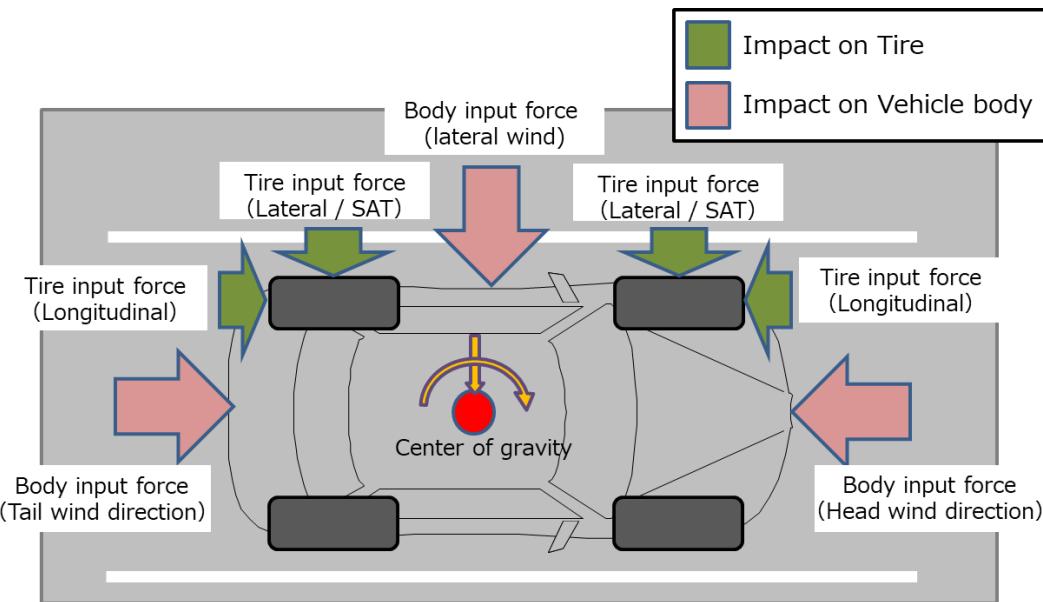


Figure 85. External physical forces considered in the definition of vehicle motion disturbance scenarios

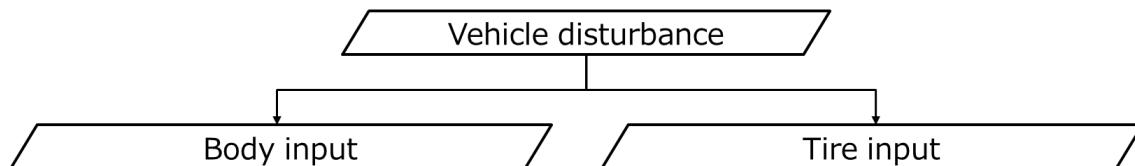


Figure 86. Vehicle motion disturbance scenario system

4.3.1. Classification of vehicle body input

There are two classes of factors that affect the vehicle body, namely, road shapes and natural phenomena (Fig. 87).

The road shape comprises a one-sided slope, a longitudinal slope, or a curvature of a curved portion. Natural phenomena, however, comprised naturally occurring crosswinds, tailwinds, and headwinds.

These are elements that act directly on the vehicle body and affect the lateral, front-back, or yaw directions.

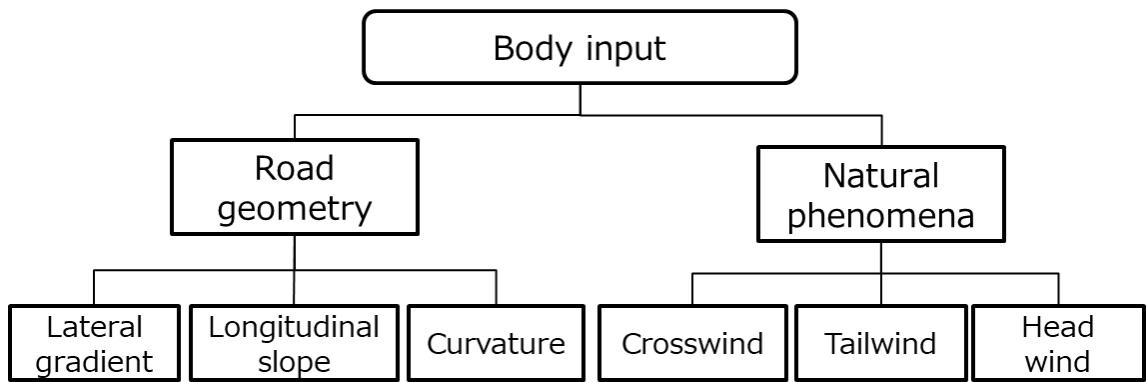


Figure 87. Scenario system for vehicle body input

4.3.1.1 Road shape

The road shape (curvature and slope of the road surface) causes the direction of gravity acting on the vehicle to change e.g., a lateral force is generated by the component of gravity on a curve as it is a one-sided slope of the road; this may increase the risk of the vehicle deviating from the lane. Similarly, in an uphill scenario, a backward force (forward on a downhill) may be generated. This in turn may increase the risk of collision with other vehicles because speed fluctuations are induced (Fig. 88).

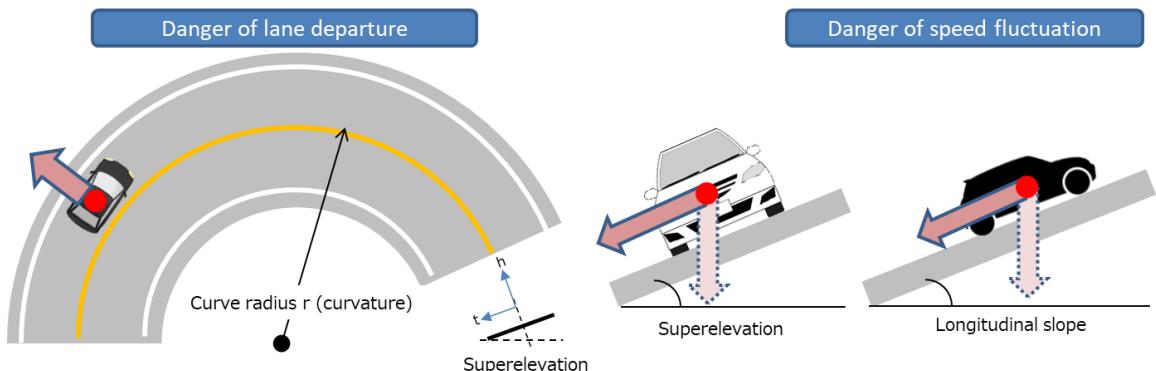


Figure 88. Road shape classification

4.3.1.2 Natural phenomena

Lateral and front-rear forces can be generated by naturally generated gusts and strong winds. These act to push the vehicle body, and, may, in some cases, cause deviation from the lane and vehicle speed fluctuations, which in turn can increase the risk of colliding with other vehicles (Fig. 89).

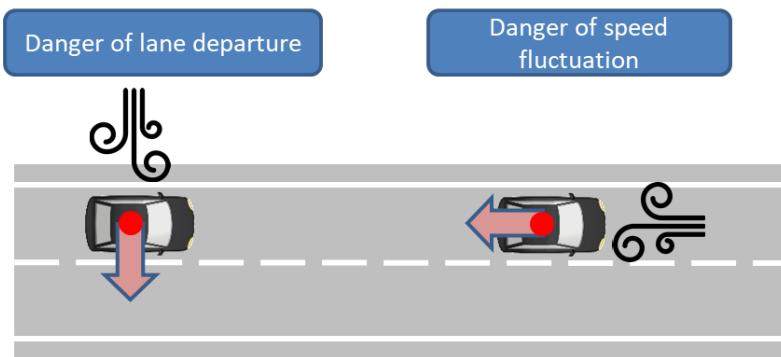


Figure 89. Classification of natural phenomena

4.3.2 Classification of tyre inputs

Tyres are affected by such factors as road surface conditions and tyre conditions. A road surface that directly affects the tyres can be classified as a road surface condition, e.g., uneven surfaces or wet surfaces can cause the coefficient of friction between the road surface and the tyres to change. This reduces the grip of the tyres and in some cases this will affect vehicle stability. Tyre condition refers to sudden changes because of punctures, bursts, and tyre wear that significantly change the tyre's characteristics (Fig. 90). The instability this causes may lead you to lose control of the vehicle, resulting in a potentially dangerous situation.

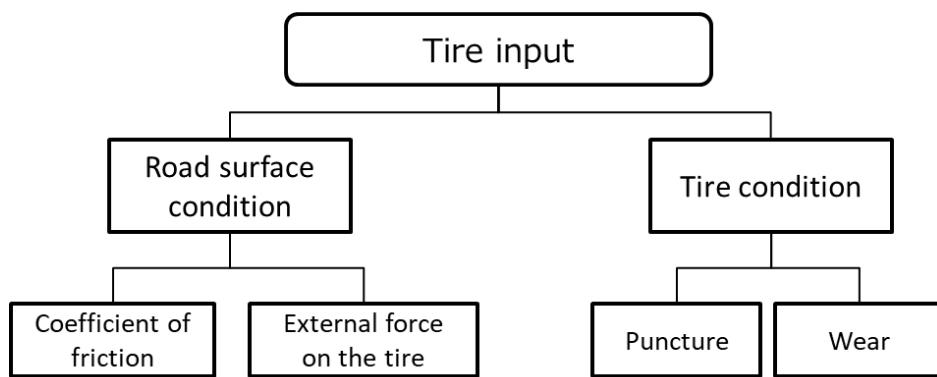


Figure 90. Scenario system for tyre input

4.3.2.1 Road surface condition

The tyre stress changes depending on the road surface shape input to the tyre, in addition to changes in the road surface. For example, when an external force causes the road surface friction to change as a result of unevenness, such as road surface shape or rain, the tyre stress changes, in addition to the direction of the vehicle. Furthermore, in some cases, there is a risk of collision with another vehicle because of deviation from the lane or vehicle speed fluctuation. The road surface condition, therefore, can be classified into the coefficient of friction and external force (Fig. 91).

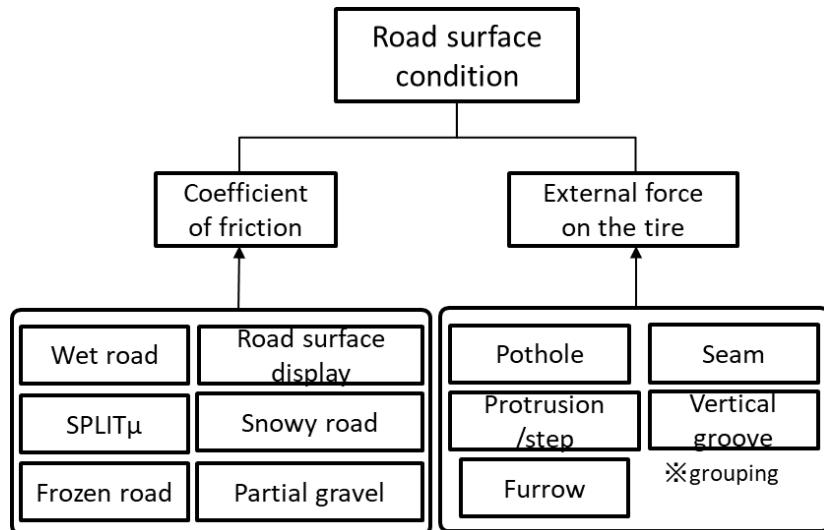


Figure 91. Classification of road surface conditions

The coefficient of friction between the tyres and the road are affected by such road surface factors as wet roads, icy roads, snowy roads, and partial gravel, e.g., a reduction in the coefficient of friction may be triggered by a sudden move from a dry road to a wet road (Figure 92, left).

This reduction can cause the vehicle to become unstable. External forces that may affect the road surface include potholes, protrusions, and striations. For example, when a vehicle crosses a step or protrusion on the road, a sudden diagonal-upward force is applied to the tyre (upper right) (upper right in Fig. 92). This in turn causes the direction of the vehicle to change. This change in movement can cause the vehicle to deviate from the planned trajectory and lead to a collision.

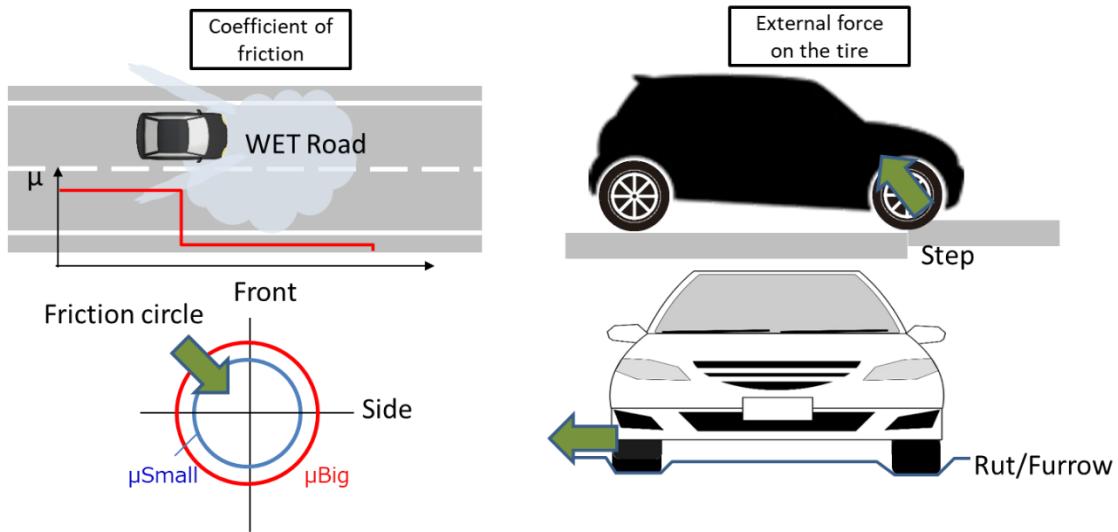


Figure 92. Vehicle motion disturbances related to road surface conditions due to changes in friction coefficient (left) and external force of tyres (right)

4.3.2.2 Tyre condition

The tyre condition fluctuates and this fluctuation affects the tyre characteristics. This may be attributed to tyre wear, punctures, and bursts (Fig. 93). These reduce tyre strength and, in some cases, may lead to collisions with other vehicles because of the vehicle deviating from the lane or vehicle speed fluctuations.

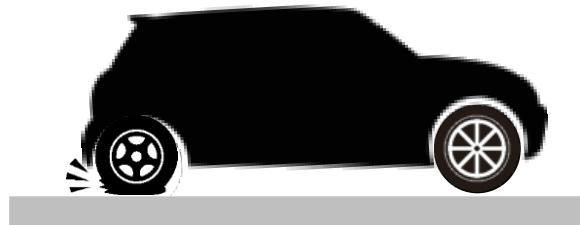


Figure 93. Vehicle motion disturbance related to tyre conditions due to bursting

4.3.3. Predictable vehicle motion disturbance safety approach

This chapter describes two general assumptions. Following this, we elaborate a technical safety approach to predictable vehicle motion disturbances.

4.3.3.1 Assumptions

The first assumption is a common sense one regarding road design, road maintenance and management, as well as road environmental conditions used by vehicles. This assumption states that roads are constructed, constantly maintained, and managed by responsible public or private institutions and that this is done in line with basic principles such as legality, ethics and engineering. Most countries have road structure ordinances. These design the shape of roads in a way that enable all persons (regardless of age such as driving skills and reflexes) with a license to drive safely. For example, in Japan, given a pre-designed speed limit of 100 km/h, a curved radius is specified on which the lateral acceleration of the vehicle below 0.11 G can be maintained even on wet roads. The design road speed limit is lowered when constructing roads for which acceleration cannot be maintained under these conditions (such as due to space availability). Similarly, mechanisms for quickly detecting surface deterioration such as those caused by a reduction in slip friction because of frozen roads or the presence of cracks, ridges, or potholes on the road surface. Another example of this is when the natural environment such as rain and wind must be within the driving range determined to be safe on the road management side. For example, in case of a disaster-level storm, road managers need to take measures such as imposing speed limits or making road closures, and drivers must follow their instructions. This is also the case for self-driving vehicles.

This means safety may be compromised by a failure to comply with road design, road maintenance and management, or road environmental, regardless of whether the vehicle is automated or not. Therefore, it is unacceptable for the road surface to be deteriorated or to have inadequate maintenance. Such scenarios are classified as unpreventable for the purpose of the AD Safety Assurance Engineering Framework methodology.

The second assumption relates to common sense on the responsibilities of AD system operators. The AD system is responsible while driving is in progress; however, the driver may not have conducted proper maintenance (e.g., excessive tyre wear may be below legal technical inspection standards, air pressure drop below the tyre manufacturer's recommended air pressure, flat tyres) or may have a puncture before operating the vehicle. If the operator is aware that the vehicle is in a state where the default vehicle performance cannot be achieved (e.g., temper tyres installed, studless tyre / chain installed), it is considered to be their responsibility. If the system is operated in this state, it may not be possible to avoid collisions.

4.3.3.2 Engineering safety approach to vehicle motion disturbance

We shall introduce a technical safety approach to predictable vehicle motion disturbances based on the assumptions in the previous chapter. Current standards, as mentioned earlier in Figure 2, specifically consider collision avoidance strategies in predictable and avoidable scenarios and collision mitigation strategies in visible and unavoidable scenarios. Therefore, if the vehicle motion disturbance causes the behaviour of the vehicle to change within a range of conditions that can be temporarily avoided, the AD vehicle is required to have the ability to control and stabilize the vehicle without interrupting the running of the vehicle. However, if unavoidable instability is caused by these disturbances, AD vehicles need to apply a “best effort” strategy to mitigate possible collisions. This must be done without interrupting the running of the vehicle.

Figure 94 shows specific examples of this safety approach to predictable vehicle motion disturbances. The top half of the figure shows an example in which, to meet avoidable conditions on a wet road, an AD vehicle faces a sharp decrease in slip friction. In this state, it must be possible to control the vehicle safely without interrupting travel. However, in the bottom half of the figure, an extreme reduction in slip friction, resulting in unavoidable

pre-defined vehicle conditions (e.g., maximum deceleration) is caused when an AD vehicle with summer tyres encounters a frozen road. Therefore, the safety approach to vehicle motion disturbance is based on a clear definition of the principles of vehicle motion engineering as related to the definition of vehicle controllable and uncontrollable conditions. These can be defined as follows:

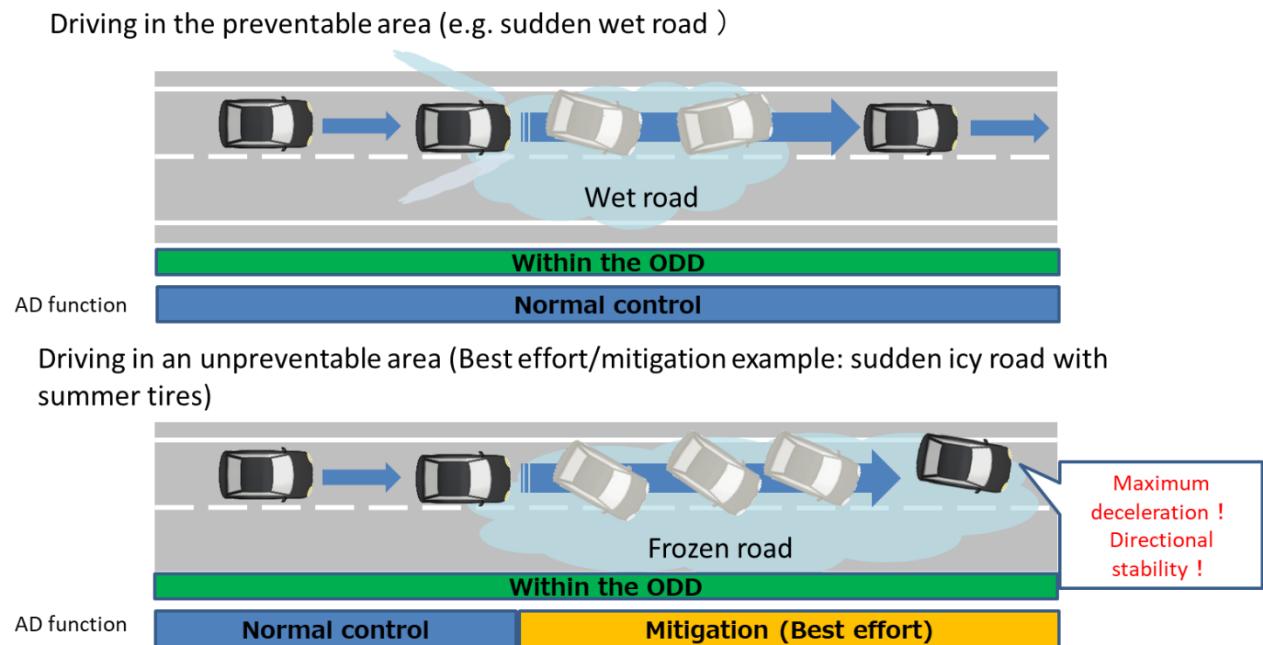


Figure 94. Safe approach to avoidable (upper) and unavoidable (lower) vehicle motion disturbances

Two mechanical indicators determine the relationship between the principles of vehicle motion and avoidance conditions. The first of these indicators is the acting force of the vehicle. This is determined by the force exerted by the vehicle as it travels. There are also one or more vehicle motion disturbance factors (e.g., road shape, wind, road surface, tyre-related conditions); this is defined as the sum of the triggered forces. The second indicator is the adhesive utilization rate ε between the road surface and the tyre. Figure 95 shows the four areas where the vehicle may operate based on adhesive utilization rate ε . These areas are classified as the area used during normal operation ($\varepsilon \leq 30\%$ or less), the area normally used by AD vehicles for emergency avoidance ($\varepsilon = 30\% - 75\%$), the area of limit of ABS operation ($\varepsilon = 75\% - 100\%$), and the area beyond limit, where tyre grip does not work ($\varepsilon \geq 100\%$ or more). Only when the force of action (indicated by the blue arrow) resulting from driving, including various vehicle motion disturbances, is $<75\%$ adhesive utilization, does motion control become physically possible. The collision avoidance strategy can be secured in this scenario. There are scenarios where motion control cannot be performed if the force of action is $>75\%$. In such cases, it is necessary to adopt a collision mitigation strategy.

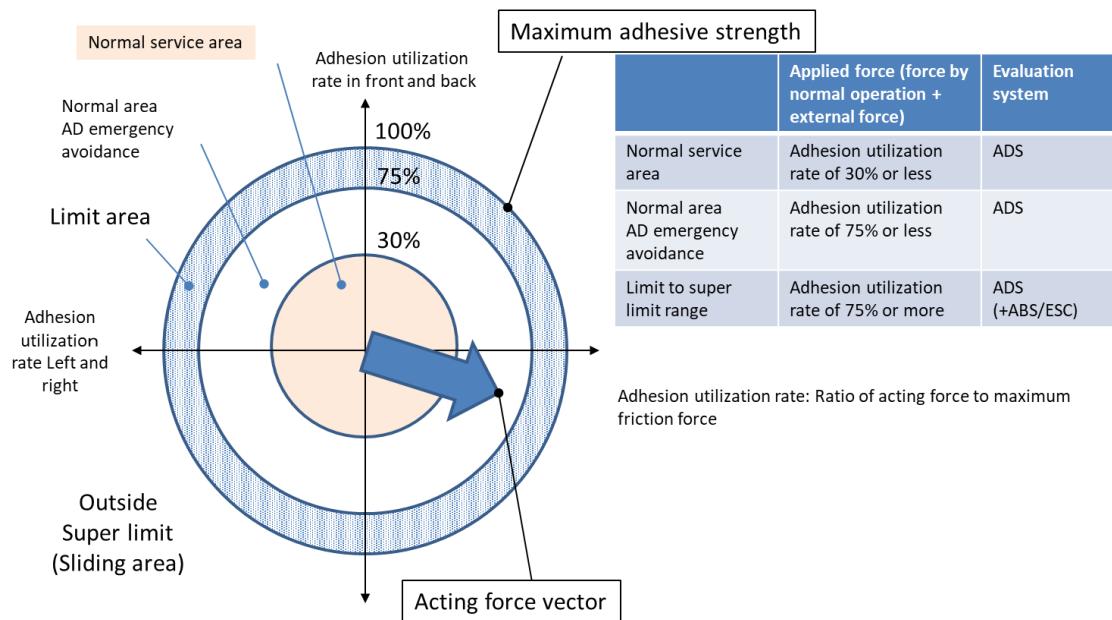


Figure95. Concept diagram showing the vehicle action force and cohesive usage rate defining the vehicle movement disturbance safety approach

4.3.3.3 Scope of controllable vehicle movement

Vehicle movement disturbances may dynamically change the force in relation to the vehicle and then turn it into areas in which it is difficult to control vehicle movement. Figure 96, with the action force and friction coefficients as axes, shows areas in which vehicle movement can be controlled in all environments and areas where control of vehicle movement is difficult. Here, the slip friction coefficient for a paved road is 0.5–1.0 when dry, 0.3–0.9 when wet, and 0.1–0.2 when frozen. Therefore, it is necessary to develop and test an AD vehicle movement strategy, in which the force caused by the vehicle disturbance always falls within the movement controllable area (triangular green area in the bottom right of the figure).

NOTE: The slip coefficient of friction value is the value when locked normally. According to Development of a real time friction estimation procedure, Gerd MüllerS. Müller, 2017, the slip friction coefficient when driving in the rain is approximately 0.6.

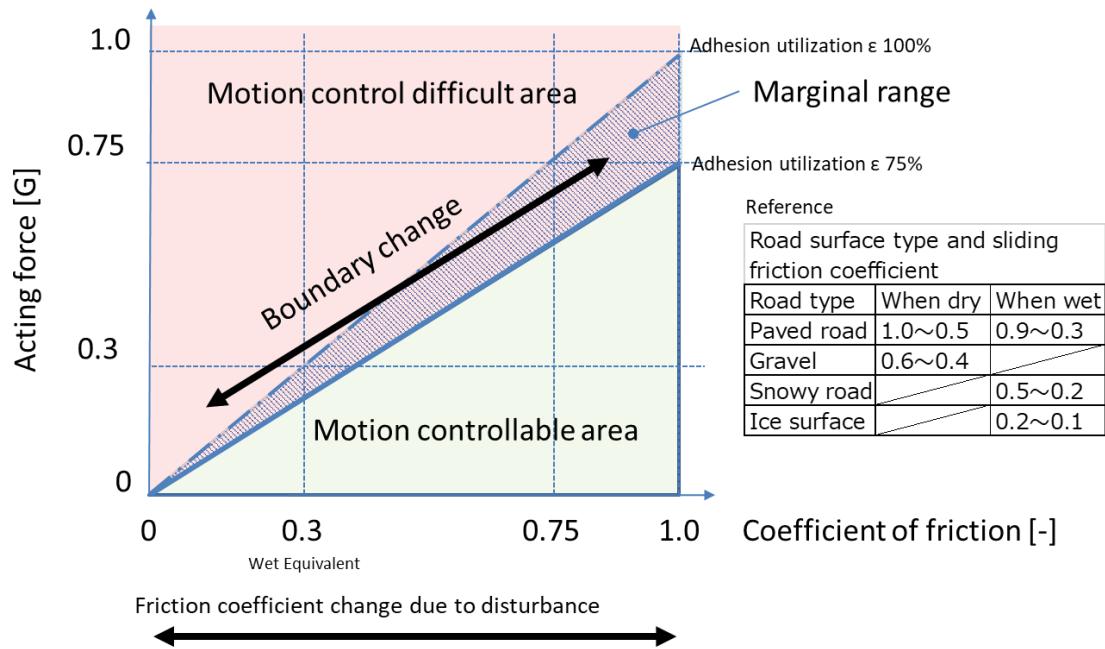


Figure96. Controllable range of vehicle movement

4.3.3.4 Controllable vehicle movement in relation to vehicle body input road shape disturbance

The road shape with difficult conditions in terms of vehicle movement is the curve radius. According to the Japan Road Ordinance, a minimum diameter is determined for the curved sections of roads such that driving can be performed in a stable manner. Furthermore, in terms of the minimum curve radius, the power laterally working, such as the centrifugal force applied to the automobile, must not exceed the force applied by the friction of the tyres and road, and is determined in consideration of a balance between the centrifugal force working on the vehicle occupants as well as the comfort in riding the vehicle. To quote this road ordinance, the minimum value for the curve radius at a design velocity of 100 km/h is 460 m (in case of temporary measures, 380 m). Here, the relational formula for the velocity, curve radius, banking, and lateral slip friction coefficient for stability in relation to lateral slip is as follows. The curve radius can be obtained from the relationship between the design velocity, lateral slip friction coefficient, and banking.

$$Z = \frac{G v^2}{g R} \quad \dots \text{formula (1)}$$

Here,

Z : Centrifugal force(N)

v : Automobile velocity (m/s)

g : Gravitational acceleration($=9.81\text{m/s}^2$)

G : Total weight of vehicle(N)

f : Road and tyre friction coefficient in relation to lateral slip\

i : Road banking ($=\tan \alpha$)

R : Curve radius (m)

Here, the conditions for lateral slip to not occur are

$$Z\cos\alpha - G\sin\alpha \leq f(Z\sin\alpha + G\cos\alpha) \quad \dots \text{formula (2)}$$

With formula development, replacing formula (2) with formula (1)

$$R \geq \frac{V^2}{127(i+f)} \quad \dots \text{formula (3)}$$

The road and tyre friction coefficient f (= lateral acceleration) in relation to lateral slip based on formula (3) is

$$f = \frac{V^2}{R \cdot 127} - i \quad \dots \text{formula (4)}$$

where if design velocity $V = 100(\text{km/h})$, road banking $i = 6(\%)$ and curve radius $R = 463(\text{m})$, $f = 0.11$. In other words, on Japanese motorways, this indicates that it is a structure in which you can drive with a lateral acceleration of 0.11 G (velocity: 100 km/h). Moreover, the speed limit may be lowered and set in case the road shape does not meet the conditions at 100 km/h. Therefore, when travelling on a motorway within Japan, it is necessary to have a cohesive force equivalent to a maximum lateral acceleration of 0.11 G. Figure 97 shows a line at 0.11 G as the maximum value for road shape disturbance required for normal driving, and shows that the action force used for disturbance other than road shape is, for example, 0.45 G ($=0.56 \text{ G} - 0.11 \text{ G}$) for dry roads and 0.12 G for wet roads. For vehicle movement disturbances, it is constantly necessary to consider the road shape for normal driving, and the total of the action force when combined with other disturbance elements must be kept within the controllable area of driving.

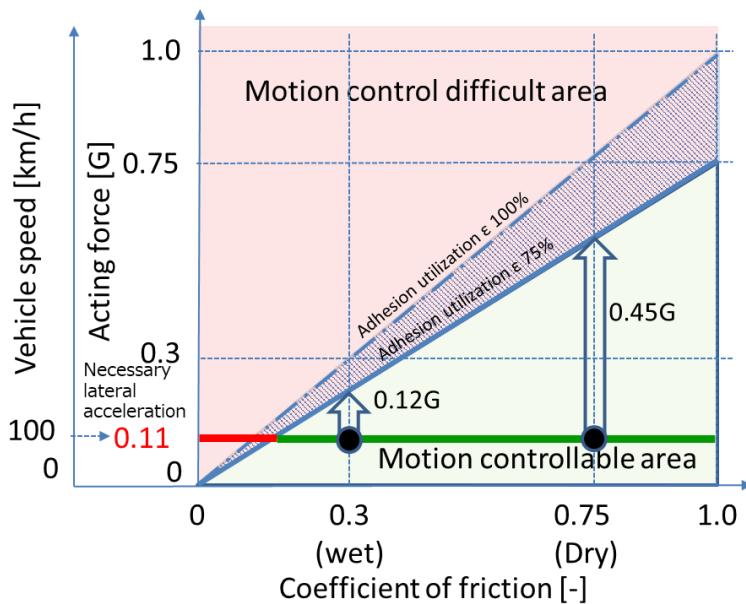


Figure 97. Relationship between friction coefficient and action force in relation to road shape

Moreover, as the elements of vehicle disturbance do not necessarily occur singly, it is necessary to consider combinations with other elements. In an actual environment, for example, there may be situations where there is a crosswind blowing while driving on a curve in the rain. Whether the road is dry, wet, or snowy can be expressed by the friction coefficient, and the road surface and natural phenomena (e.g. crosswind) and external force (unevenness) of tyres can be expressed as an action force. Moreover, in case of punctures, this can be expressed as a state in which the cohesive usage rate of 100% cannot be realized (Figure 98). In other words, as in the diagram, it is necessary to combine the elements for vehicle movement disturbances.

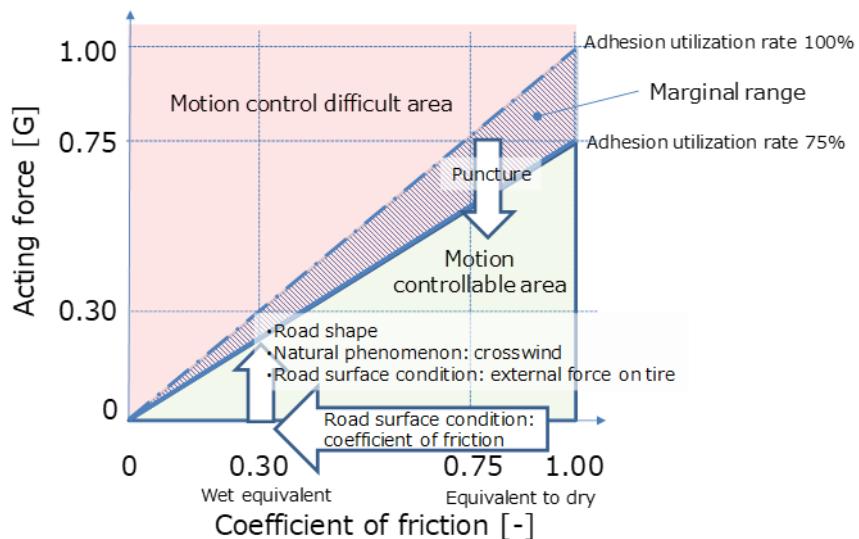


Figure 98. Relationship between combinations of elements in vehicle movement disturbance

4.3.3.5 Controllability of vehicle movement in relation to vehicle body input natural phenomena disturbance

The natural phenomenon of wind disturbance is calculated as an action force. In other words, it is added to the required action force (11 G) in the road shape. Here, the action force because of crosswind force changes depending on the shape and size of the vehicle. For example, as shown in the figure, with a wind speed of 10 m/s, there is this amount of difference between a sedan and a minivan.

Furthermore, with a wind speed of 20 m/s, even with a vehicle equivalent to a sedan, the area will have a cohesion rate of 75% or above; in other words, it will be an area where it is difficult to control movement. In such a case, it is necessary to respond on a best effort basis. However, on Japanese motorways, if there is a wind speed of 10 m/s or greater, speed restrictions come into play, and the necessary action force required for the road shape will decrease. Therefore, it is safe to drive even with wind speeds of >10 m/s. Therefore, on Japanese motorways, as speed restrictions are set in relation to the wind speed, the boundaries concerning wind speed with which it is possible to travel at 100 km/h are below 10 m/s (Figure 99).

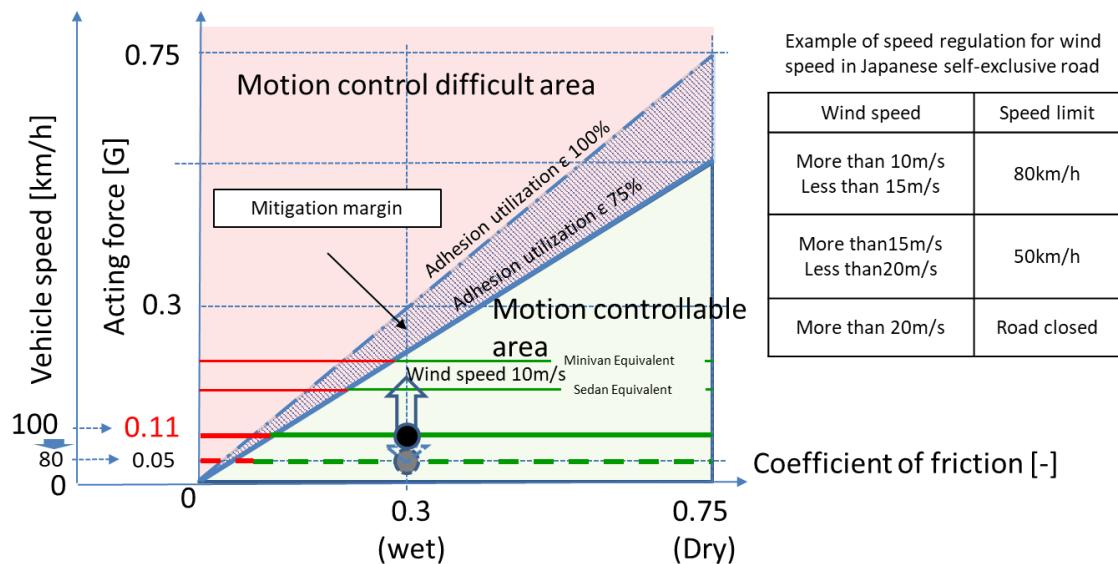


Figure 99. Relationship between friction coefficient and action force in response to natural phenomena (crosswind)

4.3.3.6 Controllability of vehicle movement in response to tyre input road state disturbance

On Japanese motorways, speed restrictions are enforced based on weather conditions, as shown in the table (example of speed restrictions in relation to weather conditions on Japanese motorways). In other words, the weather conditions where no speed restrictions are in place are the boundary values. Here, in cases involving precipitation of 20 mm/h when traveling at 100 km/h, hydroplaning does not occur; therefore, the friction coefficient of 0.3 (lock μ) or above is the boundary value. However, with precipitation of 20 mm/h or above, hydroplaning occurs; therefore, the friction coefficient is greatly decreased. Furthermore, as freezing or snowy conditions cause the friction coefficient to decrease to 0.2 or below when considering the crosswind just mentioned, this cannot be kept within the controllable area of movement. Therefore, even in environments where it is common sense to always use normal tyres, the friction coefficient boundary conditions are equivalent to wet surfaces at 0.3 (Figure 100). Moreover, external force on tyres, such as deep gaps and pot holes, cause action forces, and may disturb vehicle behaviour. However, road administrators have a responsibility to maintain and manage safety on roads. Therefore, objective values are set to determine whether repair is necessary (Table 3). In other words, if it is below this objective value, it is expected that a normal driver will be able to drive safely. Therefore, the boundary values related to external force on tyres are set to these objective values, and these are added to action forces in the same way as lateral wind.

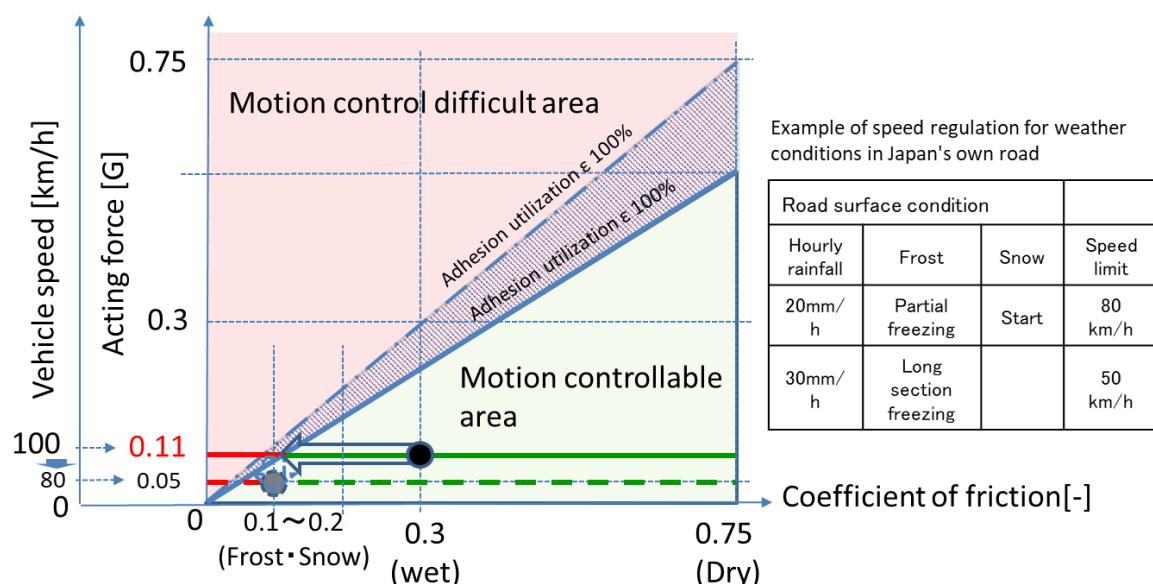


Figure 100. Relationship between friction coefficient and action force in relation to road state

Table 3. Objective values for judging necessity of repair

Road type	Item	Furrow [mm]	Step[mm]		Coefficient of friction	Vertical unevenness[mm]	Crack rate [%]	Pothole diameter [cm]
			bridge	drain				
Motorway		25	20	30	0.25	8m profile 90(Pr) 3m profile 3.5(σ)	20	20
Urban (Heavy traffic)		30~40	30	40	0.25	3m profile	30~40	20
Urban (Low traffic)		40	30	---	---	---	40~50	20

References: Japan Road Association(Road maintenance and repair guideline)

4.3.3.7 Controllability of vehicle movement in relation to tyre input tyre state disturbance

With regard to punctures while driving, this does not increase action force but cohesive usage rate decreases to 100% or below. According to SAE2013, even if one tyre punctures, provided that the rim does not make contact with the ground, the vehicle can be controlled up to 0.6 G (Tandy, Ault, Colborn, & Pascarella, 2013). This indicates that the cohesive usage rate drops to 60%. Moreover, at this extent, as it does not cause a dangerous state immediately, TD and stopping safely is required before the rim touches the ground, thus causing a burst.

4.3.3.8 Preventability/Unpreventability boundary conditions in vehicle movement disturbance

Preventability/Unpreventability boundary conditions in vehicle movement disturbances are conditions concerning whether driving can continue at the designed velocity (100 km/h in Japan) as follows.

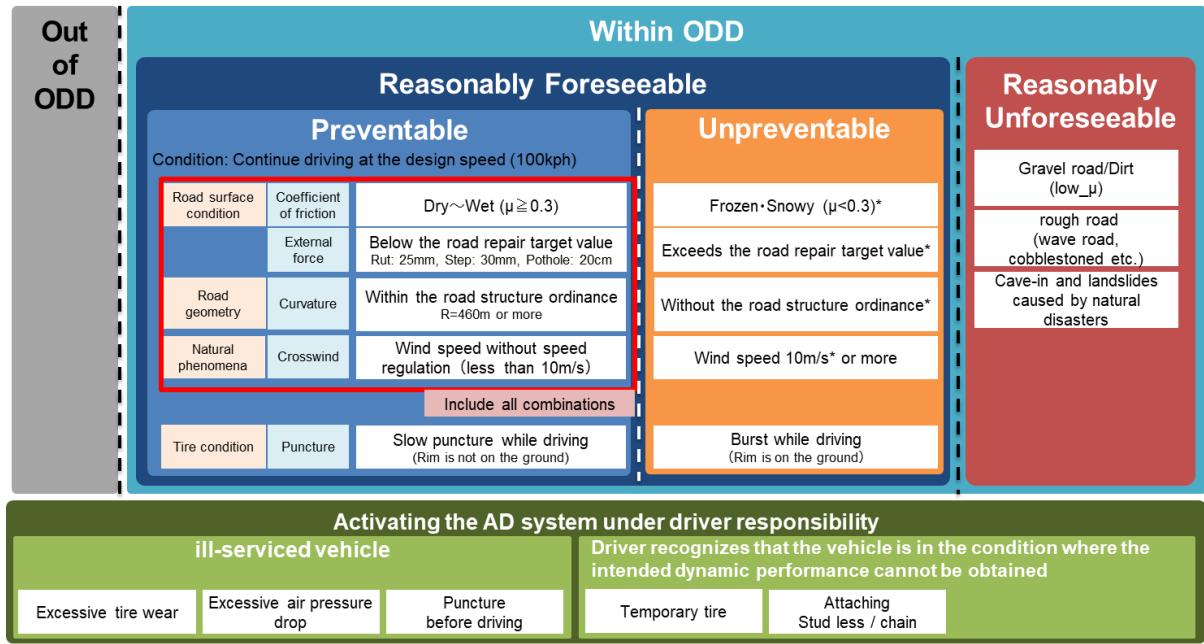
- Road state: Friction coefficient of 0.3 (lock μ) or above, and external force on tyres of the objective value for road maintenance and repair or below (e.g.: ruts : 25 mm, gap :30 mm, pot holes: 20 cm)
- Road shape: Curves within the Japan Road Ordinance specifications ($R = 460$ m)
- Natural phenomena: Where lateral wind is wind speed without speed restrictions (<10 m/s)

With regard to the above three factors, all the added conditions are preventable.

If it is not possible to drive under these conditions (such as not possible at lateral wind of 5 m/s or above), it is necessary to for the manufacturer to define this as ODD in advance.

Tyre state: Slow puncture caused during driving; however, this is detected before the rim makes contact with the road surface.

Figure 101 shows the respective unpreventable conditions and conditions of operator responsibility.



*Traffic control will be executed by road administrator
E.g., Driving is allowed with speed restrictions (=>Preventable)
Road repair work, information provision

Figure 101. Preventability/Unpreventability boundary conditions in vehicle movement disturbance

5 Scenario Database

5.1 Three layers of extraction

Functional scenarios that define qualitative scenario structures at the upper level, based on the three elements of driving actions, namely, “perception,” “judgement,” and “operation” can be systematically structured under the three scenarios of “perception disharmony,” “traffic disturbance,” and “vehicle movement disturbance,” thus enabling comprehensive scenario evaluations (Chapter 3).

Logical scenarios apply a quantitative parameter range to structuralized functional scenarios. Therefore, for example, in the case of a traffic disturbance, this is defined by extracting the vehicle path from the traffic flow data, and taking a data-driven approach where traffic flow parameters such as relative velocity and cut-in speed are defined based on statistical distributions. The traffic flow data refers to traffic monitoring and operation data, accident databases, insurance data, maps, and road data.

Concrete scenarios can be considered as individual evaluation conditions for concrete evaluations, that extract safety judgement boundaries for distinguishing safety state and unsafe states (Section 1.3).

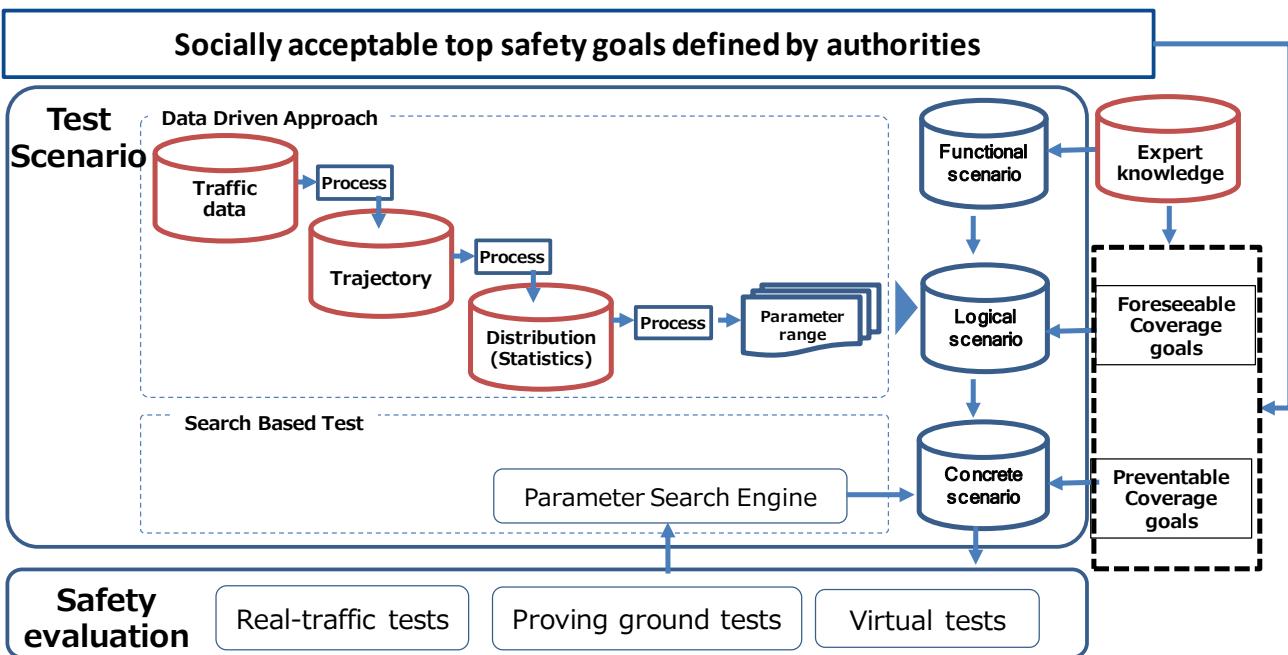


Figure 102. Process of developing and applying data-driven AD safe scenarios

5.2 Database parameters, format, and architecture

Figure 103 shows the information flow scheme required for creating actual test scenarios from a scenario catalogue and outputting them in a form in which these scenarios are standardized. These versatile standardized formats that can adapt to a wide range of simulation environments may be beneficial for AD safety evaluations. Files including information related to vehicle behaviour and road shape are generated via a test data generator from the scenario catalogue. These files can be applied to various simulation environments via a converter, and can be made independent using specific, commercially available software.

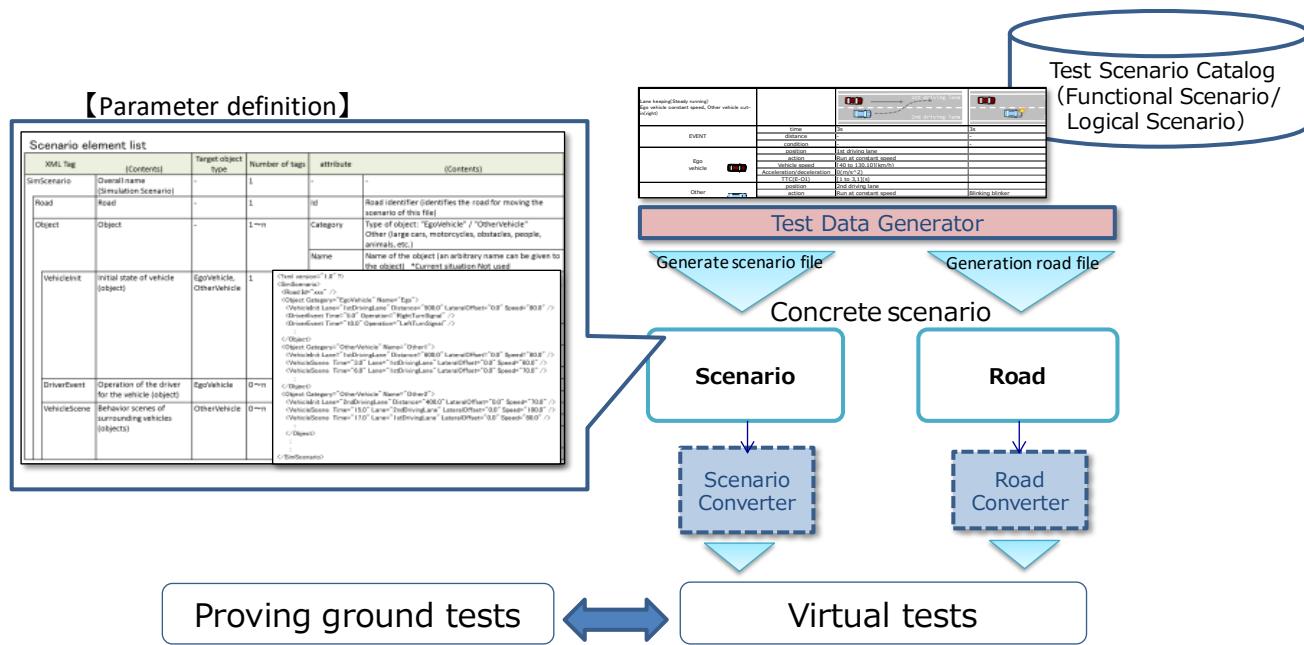


Figure 103. Information flow scheme for AD safety evaluation based on standardization scenarios

5.3 Test scenario database interface specification

Figure 104 shows the scenario database system. The scenario database uses actual traffic observation data as input and outputs scenarios required for safety evaluation. To realize this, an input/output interface is required. Moreover, a safety evaluation is performed using the output scenario data, and the result is fed back to the scenario database.

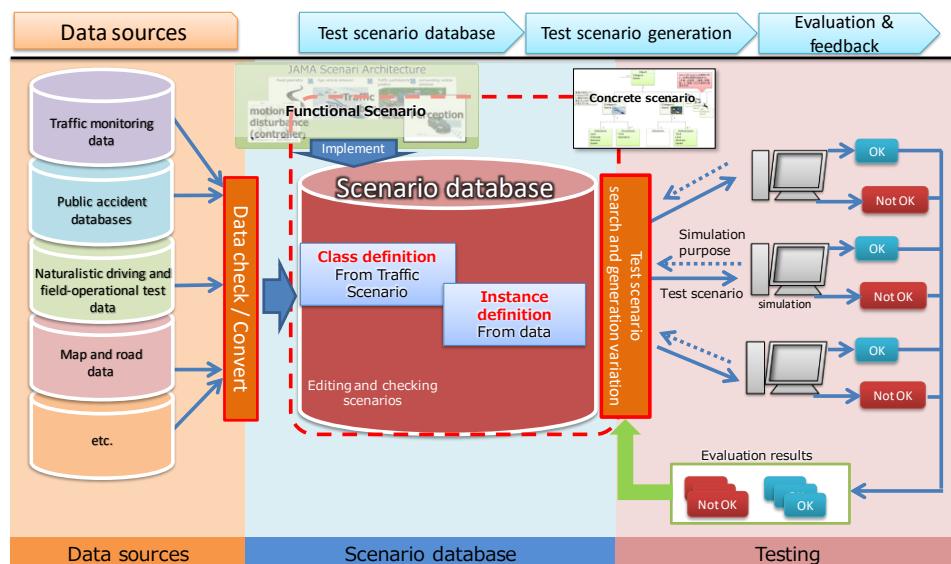


Figure 104. Scenario database scheme and interface

There is a wide array of actual traffic data, including traffic monitoring data, accident data, field test collection data, maps, and road data. To incorporate all of these unspecified large number of actual data items in a scenario database, it is necessary to convert them into an appropriate format (Figure 104 Data check/Convert). Data that

are appropriately incorporated into a shared database can be used to generate scenarios in accordance with a standardized methodology.

To use the scenarios generated within a scenario database, an interface that enables searching, generation, and exporting of scenarios is required (Figure 104 Test scenario search and generation variation).

Annex A

Road Geometry

The tree diagram for the road component elements identified from the road structure shows corresponding parameters related to road component elements. Definitions of these parameters are shown in Table A-1.

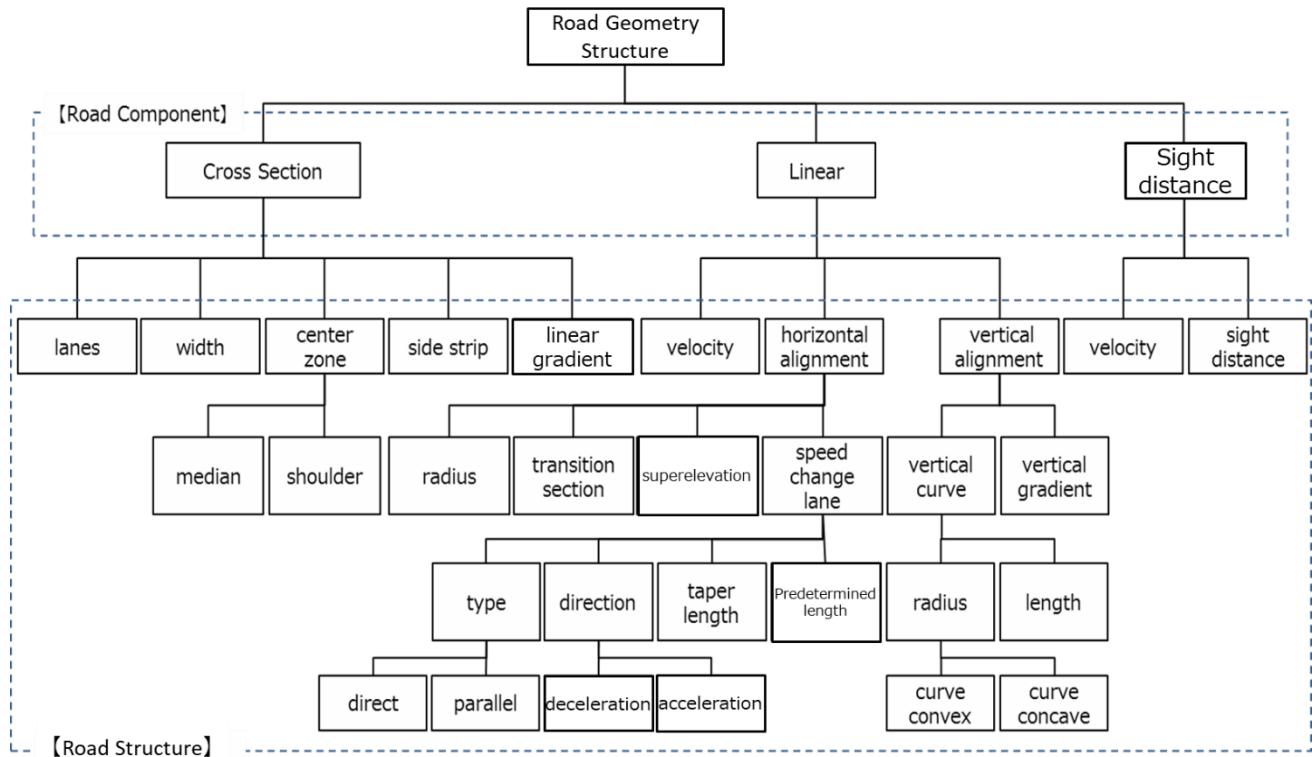


Figure A-1. Parameters related to road component elements (cross-sections, lines, and viewing distance) based on the Cabinet Order on Road Structure.

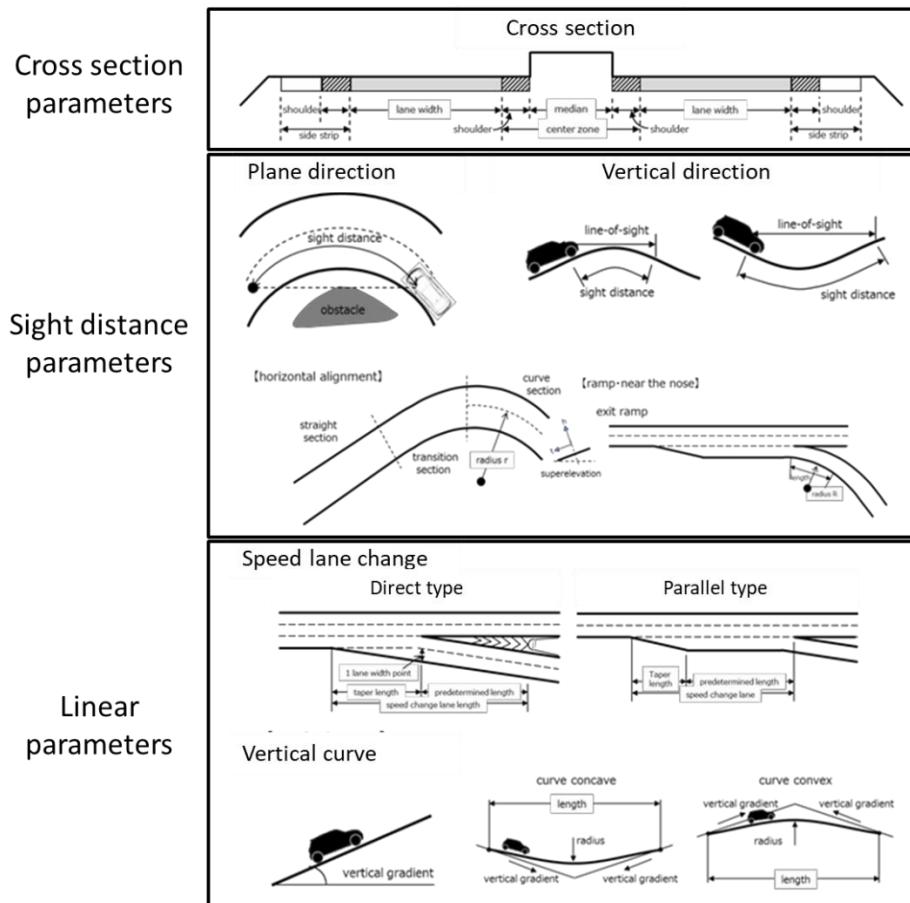


Figure A-2. An example of cross-sections, viewing distance, and linear road parameters based on the Japanese Cabinet Order on Road Structure.

Road geometry parameters were examined for each scenario category (cognition disturbance perceptual limitations, traffic disturbance, and disturbance in vehicle motion). For example, with traffic disturbances, as the number of surrounding vehicles increases, the number of lanes is increased in some cases. However, this is not directly related to cognition disturbance or disturbance in vehicle motion. Table A-1 shows the road geometry parameters to develop scenarios for each scenario category.

Table A-1. Road geometry parameters to develop scenarios for vehicle control categories.

Road parameters		Perception limitation	Traffic Disturbance	Vehicle disturbance
Cross section	lanes	-	Increased risk due to increase in surrounding vehicles in merging and departing sections	-
	width	-	Relative distance to surrounding vehicles shortens	Difficulty for lane keeping along with the curve radius
	center zone	median	Fearness of misrecognition in the opposite lane where the median is narrow	Possible use of median as avoidance route, expressed here to create a road geometry without basic treatment
	side strip		-	Possible use of shoulder as avoidance route, expressed here to create a road geometry without basic treatment
Linear	horizontal alignment	radius	Depending on curve radius and obstacles, viewing distance may be affected	-
		transition section	-	-
		superelevation	-	-
		speed change lane length	-	Difficulty to achieve sufficient acceleration/deceleration
	vertical alignment	vertical curve	Recognition delay due to obstacle at the top of convex curve	-
		vertical gradient	Misrecognition of target ahead	Depending on vehicle performance, it also affects traffic disturbance
	Sight distance	Recognition delay by viewing distance	-	-

Similarly, when each parameter is considered, important parameters for each scenario would be as follows:

- Parameters associated with cognition disturbance scenario include, for example, the median, the radius of curvature, vertical alignment, and viewing distance.
- Parameters associated with traffic disturbance scenario include the number of lanes, width, speed change lane, and vertical gradient.
- Parameters associated with disturbance in vehicle motion scenario include width, the radius of curvature, non-controlled interval, superelevation, and vertical alignment.

In terms of road geometry parameters for test scenarios, parameters that have no impact on safety were set to fixed values, and only the range of safety-related parameters are defined. In this manner, number of test cases can be reduced.

A.1 Road geometry component elements

Based on driving environment definitions, road geometry was classified into main roads, merge zones, departure zones, and ramps. Moreover, road geometry classification comprises four elements: main road, speed change lane, ramp, and nose vicinity (Figure A-3). Road structure parameters from the Cabinet Order on Road Structure are defined for each component of this book [4]. According to this basic classification, the relationship between four categories used to prepare scenarios and road geometry components standardized from the Cabinet Order on Road Structure used to build roads in Japan can be established. The examples of these standardized road geometry components are the main road, speed change lane, ramp, and nose vicinity. Moreover, the cabinet order incorporates the relationship between road geometry components and road geometry parameters important for safety such as cross-sections, lines, and viewing distance that are related to different road component parameters.

Note: The road geometry components and related parameters described herein are defined according to road technique standards related to Japanese road construction. The majority of standards in other countries employ similar rules, which facilitates the easy application of the methodologies proposed for different countries and areas.

Road geometry classification for scenario development	Standardised road geometry structure	
	Road component	Road component parameters
Main roadway	Main roadway	Cross section Linear Sight distance
Merging zone	Speed change lane	Cross section Linear
Departure zone	Ramp	Cross section Linear Sight distance
Ramp	Nose vicinity	Linear

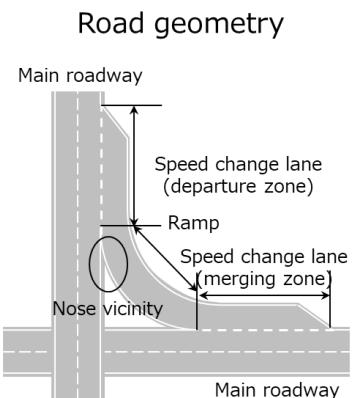


Figure A-3. Relationship between road geometry classification to develop scenario, standardized road components, and corresponding safety-related parameters.

A.2 Basic parameters of road geometry

To determine the basic road geometry parameters in a road structure model (for Japan, Table A-2), important parameters are set for strict values for each scenario (first column from the right in Table A-2). These parameters are presented with the upper and lower limits, and depend on the scenario.

Table A-2. A list of road parameters from the Cabinet Order on Road Structure (RSO) and baseline road geometry parameters from the Cabinet Order on Road Structure in Japan.

Road parameters			Reference values	Most Demanding values			
Cross section	Number of lanes		1, 2, 3, 4	3			
	Width (m)		3.25, 3.5, 3.7	3.25			
	Center zone	Median (m)	1.25, 1.5, 2, 2.25, 3, 4.5	1.25			
		Shoulder (m)	0.25, 0.5, 0.75	0.25			
	Side strip (m)		1.25, 1.75, 2.5	1.25			
	Linear gradient (%)		2, 2.5	2.5			
Linear	Velocity (km/h)			120	100		
	Horizontal alignment	Curve section	Radius (m)	570, 380	570		
			Transition section (m)	100, 85	100		
			Superelevation (%)	6, 8, 10	10		
		Speed change lane	Type	direct, parallel	Direct		
			Direction	deceleration, acceleration	Deceleration		
			Taper length (m)	70, 60	70		
	Vertical Alignment	Vertical Curve	Pre-determined length (m)	210, 110	110		
			Radius curve convex (m)	11000, 6500	11000		
			Radius curve concave (m)	4000, 3000	4000		
			Length (m)	100, 85	100		
		Vertical gradient (%)		5, 6	5		
Sight distance	Velocity (km/h)			120	100		
	Sight distance (m)			210, 160	210		

A.3 Update with actual environmental data

Actual road geometry may not strictly adhere to the law for a variety of reasons (e.g., limited by the landform). This is handled as a tentative scale, and may be extended over a long period of time. As such, since road conditions change, actual harsh conditions must be reflected in scenarios.

Table A-3. Examples of harsh conditions in real environment.

Situation description	Critical Parameter	Disturbance type
Complicated highway interchange	Short merge and departure lanes	Traffic disturbance
Pronounced curve	Reduced curve radius (limited field of view) High lateral acceleration	Perception disturbance Vehicle disturbance
Absence of central zone	Central zone width (non-regulated)	Perception disturbance
Narrow tunnel dividing wall at merge	Reflection shoulder width (non-regulated)	Perception disturbance
Separators to prevent from driving in the wrong direction	Merge point separators	Perception disturbance

A.4 Updating road geometry parameters based on actual world map data

In this section, we explain the definition of important parameters for road geometry. Based on the road structure ordinances of each country, road geometry parameters were identified. However, parameters are not important elements. For example, when there are a large number of lanes, the number of surrounding vehicles increases, and there may be an impact as traffic disturbance; however, there may not be an impact on cognition disturbance

and disturbance in vehicle motion. Therefore, the selection of road geometry parameters depends on scenario categories.

- Important parameters covered by the cognition disturbance scenario include the departure zone, the radius of curvature, length of the curve, longitudinal open circuit, and viewing distance.
- Important parameters covered by the traffic disturbance scenario include the number of lanes, width of lanes, acceleration and deceleration lanes, and longitudinal gradient.
- Important parameters for disturbance in the vehicle motion scenario include the lane width, the radius of curvature, transition zone, superelevation, and vertical alignment.

Note 1-Entry: By setting critical parameters, where unimpacted parameters are fixed, as road geometry parameters of test scenarios, the number of test cases can be reduced.

To determine road geometry parameters, according to Table B2, we assigned the harshest values for important parameters of road geometry based on the Cabinet Order on Road Structure in Japan. However, the actual shape of roads may not strictly follow the Cabinet Order on Road Structure (e.g., the length of merge zone may be shorter than what is stipulated by the ordinance since construction space in a crowded city is limited). Therefore, baseline values of road geometry parameters defined by the Cabinet Order on Road Structure must be updated with actual harsh conditions of road geometry. To this end, we incorporated dynamic map data into the process, e.g., in a survey of highway characteristics in Tokyo region where the “legal speed is 100 km/h” and “the minimum radius of the curved section is less than 100 m” (left in Figure A-4), multiple locations fit the description (blue spots to the right of Figure A-4). Such searches reflect actual road requirement parameters for the radius of curvature in the Tokyo region; thus, road geometry baseline parameters (Table Table A-2) must be updated from 380 m to 100 m or less.

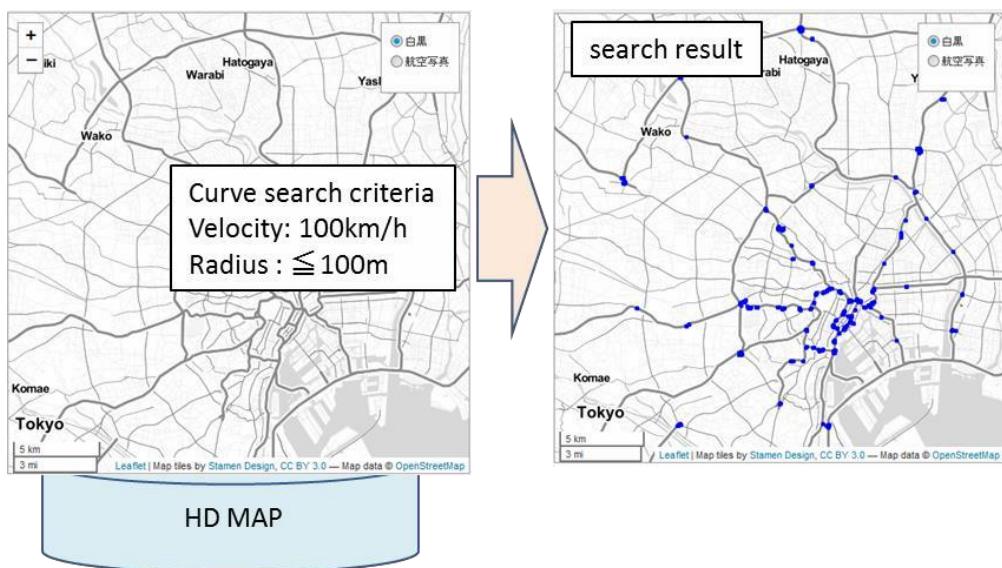


Figure A-4. Data extraction from a dynamic map.

Annex B

Scenarios for Motorcycles

Similar to the systemizing process explained in regard to traffic disturbance scenarios, road geometry, ego vehicle behaviour, and surrounding motorcycle location and motion, we propose a methodology to structure traffic disturbance scenario for motorcycles (Figure B-1).

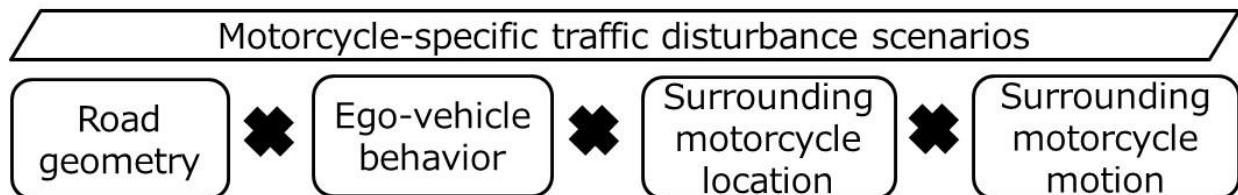


Figure B-1. Structural concept of traffic disturbance scenario for motorcycles.

B.1 Classification of surrounding motorcycle location and motion

When defining scenarios for general vehicles, we defined the location of surrounding vehicles in eight directions around the vehicle. In motorcycle scenarios, in addition to this, we defined right and left of the vehicle as unique locations for motorcycles to build scenarios.

As shown on the left side of Figure B-2, locations unique to motorcycles [L] and [R] are on both sides of the vehicle within the same lane. Motorcycles can move to [L] or [R] by decelerating from 1 in front (a), accelerating from 2 behind (b), or by changing the lane from surrounding locations, 3,4, 5, 6, 7, or 8 (c) (centre in Figure B-2). As shown on the right side of Figure B-2, a motorcycle can move from [L] and [R], where it may approach the vehicle laterally (d), move forward (e), move backward (f), or be parallel to the vehicle (g).

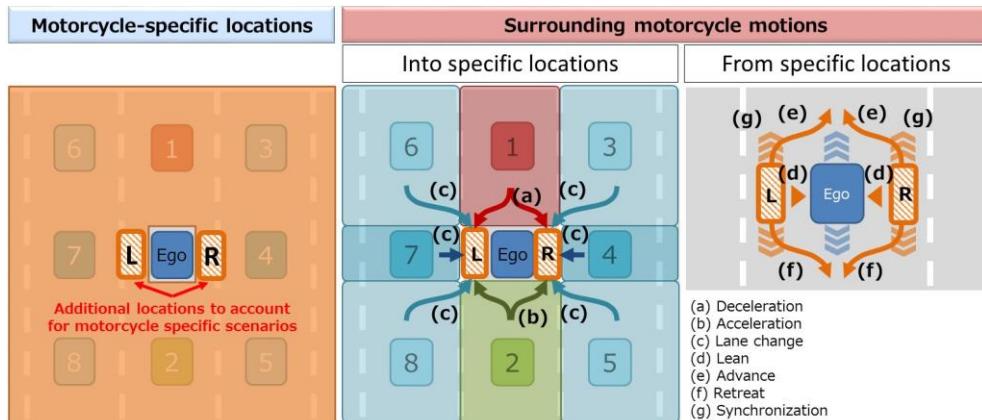


Figure B-2. Locations and motion of motorcycles that could prevent motion of a vehicle (left)

B.2 Traffic disturbance scenario unique to motorcycles

The structure of motorcycle scenarios is expressed by a matrix that includes 56 possible combinations. In a lane change scenario for vehicle, only synchronized motions are targeted. This is because lane change for the vehicle is physically impossible if there are vehicles in locations unique to motorcycles: [L] and [R]. This leaves 18 scenarios that are actually achievable in the real traffic, which are incorporated in the safety assessment (Figure B- 3).

 : EGO  : Motorcycle

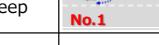
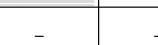
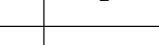
		Surrounding motorcycle position and motion						
Road geometry	Ego-vehicle behavior	Into specific locations			From specific locations			
		Deceleration to side	Acceleration to side	Lane change to side	Lateral approach	Advance	Retreat	Synchronization
Main road	Lane keep							-
	Lane change	-	-	-	-	-	-	
Merging zone	Lane keep	-	-		-	-	-	-
	Lane change	-	-	-	-	-	-	
Departure zone	Lane keep	-	-		-	-	-	-
	Lane change	-	-	-	-	-	-	
Ramp	Lane keep							-
	Lane change	-	-	-	-	-	-	

Figure B- 3. Traffic disturbance scenario for motorcycles.

Annex C

Approach for complex scenarios of traffic disturbance

In an actual traffic environment, multiple traffic participants can take multiple actions at various times. In this Section, we examine scenarios including multiple traffic participants based on the developed concept for the traffic flow scenario.

C.1 Concept of avoidance motion scenario

When surrounding vehicles make sudden dangerous moves, the ego vehicle must react to avoid such action. Such danger can take place during lane keep and lane change. The latter refers to situations when surrounding vehicles are trying to move into the same space as the ego vehicle as they try to change lanes. Action to avoid these vehicles is called avoidance motion, which is a secondary motion by the ego vehicle. Thus, avoidance motion scenarios aim to assess the safety of such secondary behaviour by the ego vehicle.

C.2 Traffic flow scenarios

To understand scenarios created by avoiding dangerous movements of surrounding vehicles, we present a stepwise sequence. This sequence begins with a sudden approach by surrounding vehicles, such as a dangerous approach by surrounding vehicles to the ego vehicle driving while keeping the lane, or when the ego vehicle tries to change the lane (Figure C-1). This is the starting point of avoidance motion by the ego vehicle. Before executing this avoidance motion, the ego vehicle must determine the range wherein it is able to execute the avoidance motion. This range is called the “avoidance area”. For example, when a preceding vehicle suddenly decelerates, creating a potentially dangerous scenario (avoidance trigger), the ego vehicle must judge if there is a space immediately behind (avoidance area), and then must decelerate as the avoidance motion. However, when determining avoidance area, the ego vehicle must consider cut-in vehicles that might enter the same area. When considering these aspects and the environment of the road the vehicle is driving on (e.g., main road, merge lane, etc.), different traffic flow scenarios can be created.

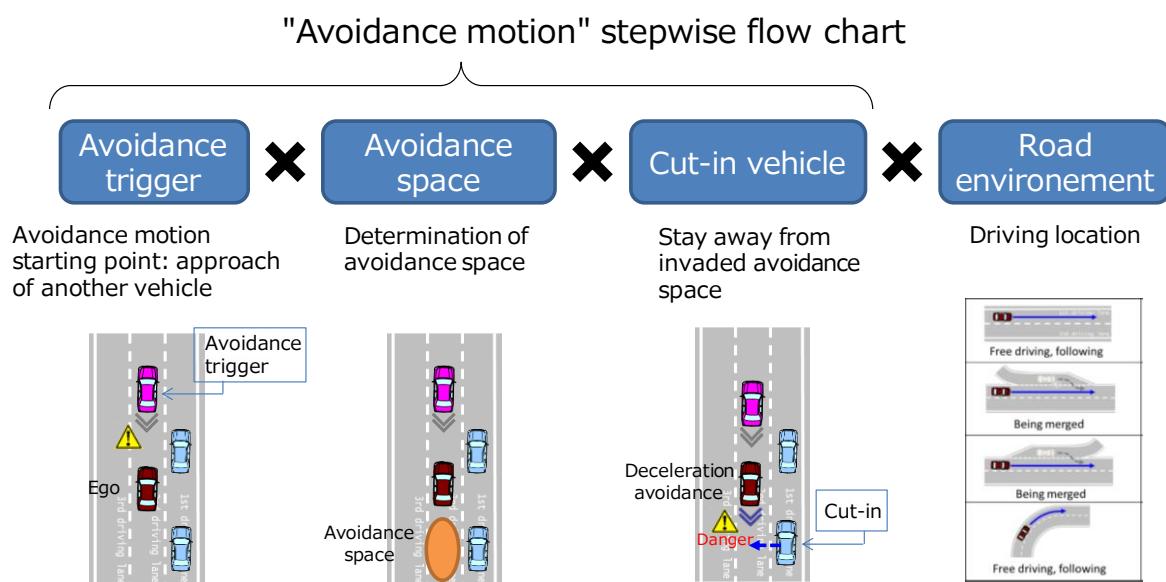


Figure C-1. Steps from the start and finish of an avoidance motion.

C.2.1 Avoidance trigger

Ego-vehicle motion	Avoidance trigger types	Pattern diagram
Lane keep	<p>a) Approach from the front and sides Deceleration of the lead vehicle, cut-in by the lead side vehicles. ※Approach from the back (rear-end collision) is not considered.</p>	<p>a) Approach from the front and sides</p>
Lane change	<p>c) Approach to the lane change destination Cut-ins from directions other than that of the ego-vehicle ※Consideration of lane change to the left isn't necessary because of symmetry</p>	<p>b) Approach to the lane change destination</p>

Figure C-2. Driving situation of the ego vehicle in avoidance motion scenarios.

C.2.2 Avoidance space

Avoidance space is defined as a range wherein the ego vehicle can take an avoidance motion. When approached by surrounding vehicles, the avoidance trigger begins, and the ego vehicle must determine the avoidance space. For safety, the avoidance space is not in the direction where the trigger vehicle is approaching from. Figure C-3 emphasizes the avoidance space for both lane keep scenarios and lane change scenarios.

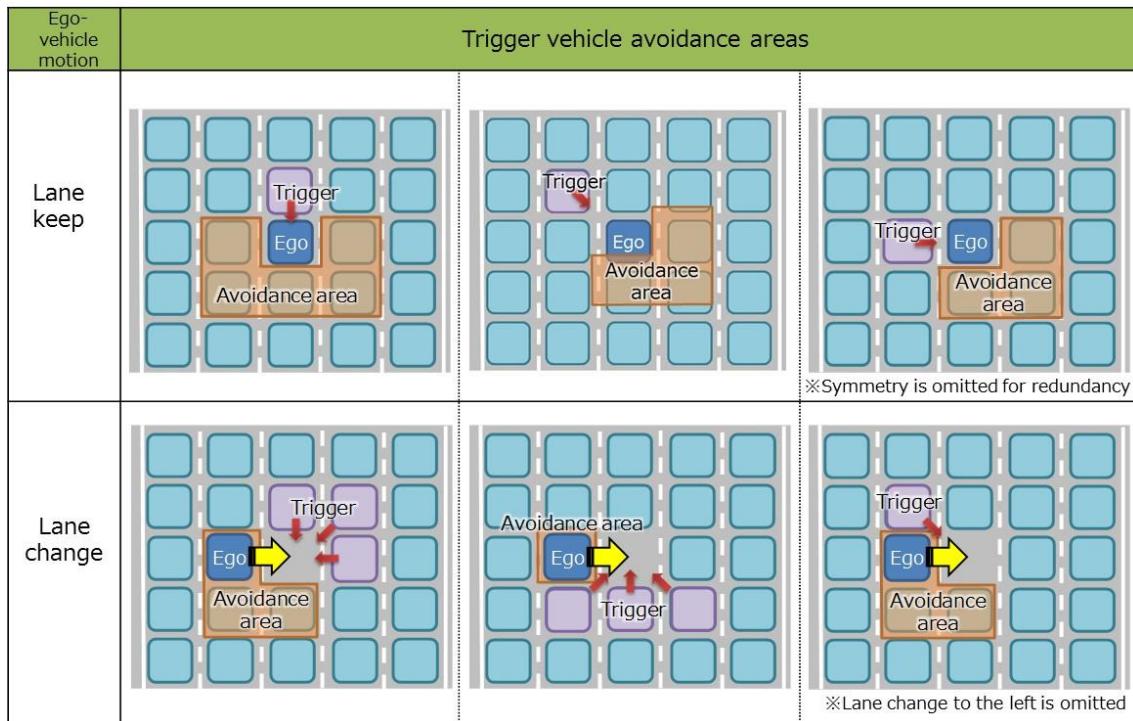


Figure C-3. Avoidance areas for each trigger vehicle for lane keep (top) and lane change (bottom).

In a case of lane keep (top half of Figure C-3), the trigger vehicle approaches from in front of the ego vehicle [L(1)], from front and the side of the ego vehicle [Pl-f (6), Pl-f (3)], or from the side of the ego vehicle [Pl-s (7), Pl-s (4)]. The areas highlighted in red are the avoidance areas (lateral symmetry is omitted). The lower half of Figure C-3 shows a scenario in which the ego vehicle changes lanes (lateral symmetry is omitted). In this case, vehicles in the lane change destination for the ego vehicle become trigger vehicles. Areas highlighted in red are the avoidance areas.

After determining the avoidance area, the pattern of vehicles in the avoidance area must be determined. For example, if deceleration by the preceding vehicle is the trigger, combinations of patterns of vehicles in each cell of the avoidance area becomes $2^5 = 32$ (Figure C-4).

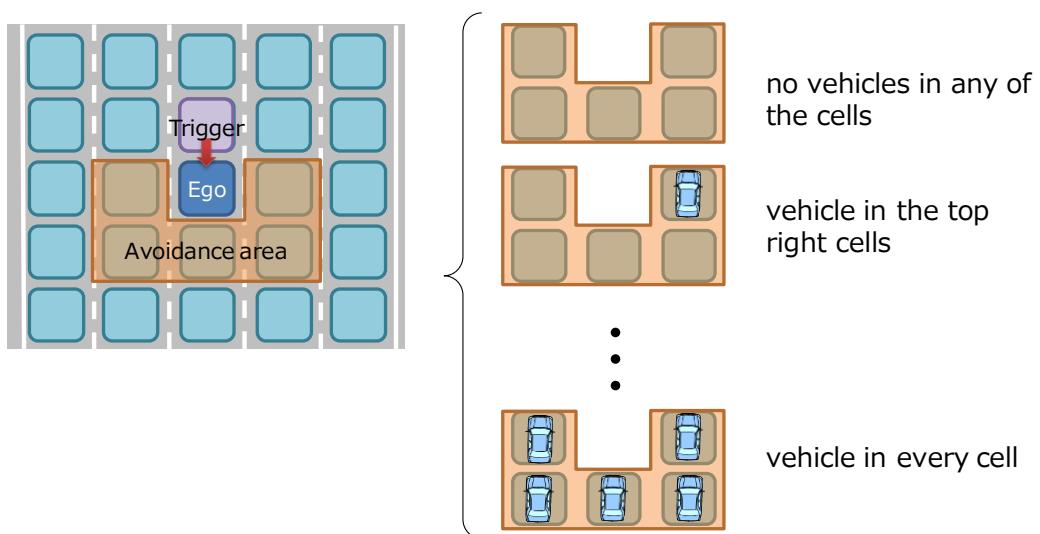


Figure C-4. Patterns of vehicles in each cell in the avoidance area.

C.2.3 Cut-in vehicles into the avoidance area

After confirming whether there are vehicles in the avoidance area (how many and which cell), vehicles that could cut into the avoidance area from adjacent spaces must be identified. Ranges from where cut-in into the avoidance area is possible are shown in Figure C-5.

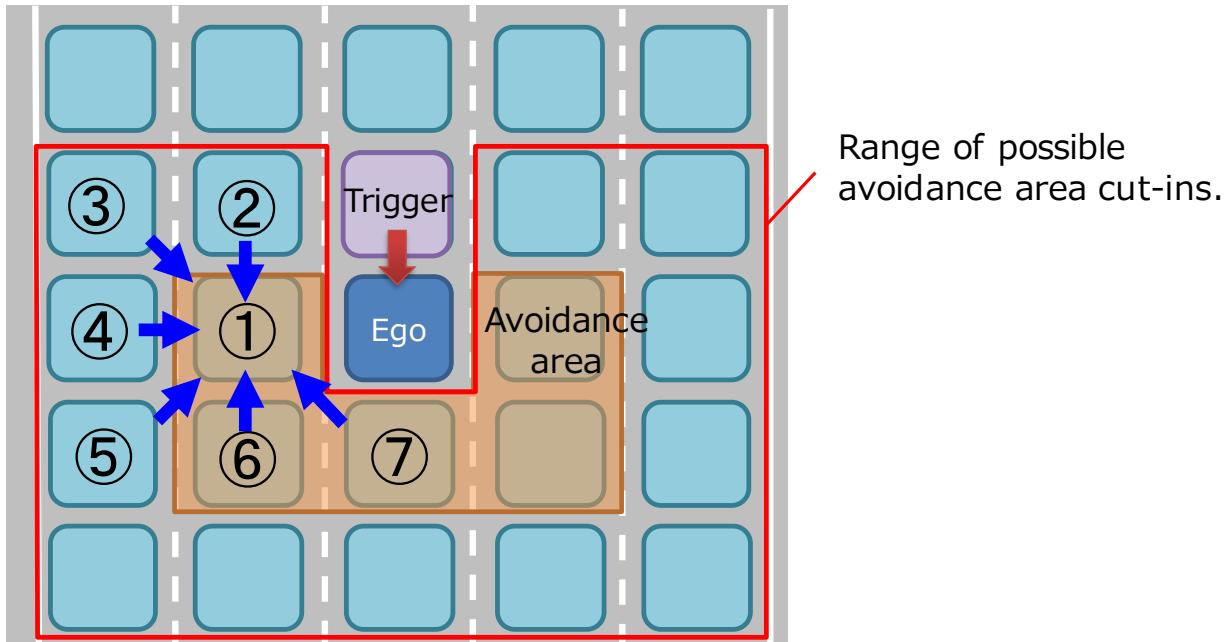


Figure C-5. Range where cut-in into the avoidance area is possible.

The avoidance area is highlighted in red. Considering a case where the ego vehicle moves into cell ① to avoid the trigger vehicle, possible cut-in by vehicles in locations ⑥ and ⑦ in the avoidance area and in adjacent locations ②, ③, ④, and ⑤, must be considered.

C.2.4 Road environment

The road environment is a combination of road geometry and the ego vehicle location, which are two factors that impact the avoidance motion. “Road geometry” is classified into the main road, merge lane, departure lane, and ramp. Ego vehicle locations are defined by the shape of the avoidance area and number of lanes in each road geometry.

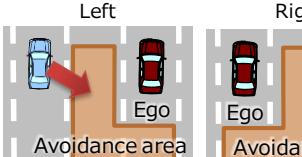
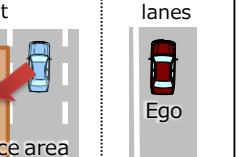
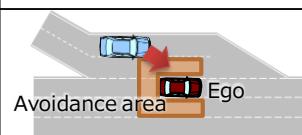
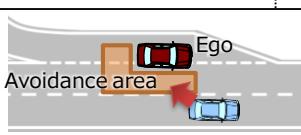
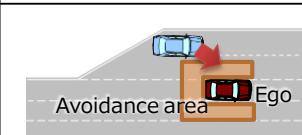
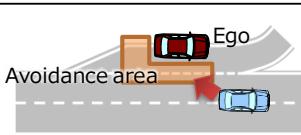
a.Road shape	b.Ego-vehicle position			※走行車線外の有無
	Ego-vehicle position	Adjacent lanes on both sides	One adjacent lane Left Right	No adjacent lanes
Main road	Ego-vehicle position			
	Number of lanes required	Lane 5	Lane 3*	Lane 1*
Merging lane	Ego-vehicle position			
	Number of lanes required	Lane 5	Lane 3*	
Departure lane	Ego-vehicle position			
	Number of lanes required	Lane 5	Lane 3*	
Ramp		Omitted for equivalence with main road (Lane 1, 2*)		

Figure C-6. Classification of road environment in avoidance motion scenarios.

Annex D

Verifying the completeness of scenario database based on accident data

There are two cases to explain how completeness of scenario database is verified based on accident data.

D.1 German In-Depth Accident Study (GIDAS) data

Verification of the completeness of the traffic flow scenario system is possible. For example, one can assess if accidents reported in the German In-Depth Accident Study (GIDAS) database (Otte, Krettek, Brunner, & Zwipp, 2003) are covered. As an assumption, all possible scenarios in the German traffic environment must be presented in the accident classification system of GIDAS.

GIDAS classifies traffic accidents according to the pre-defined rules related to accident characteristics. We related and compared the accident classification system defined by GIDAS (GIDAS code) and traffic flow scenario system.

The table to the upper left of Figure D-1 shows the number of GIDAS accident codes classified after correlation. Categories A, B, and C represent 78 codes and 7,567 accidents included in the analysed database. The verification result of these accident data showed that 33 codes and 6,787 accidents can be analysed under the traffic flow scenario system. The traffic flow scenario system possibly covers 90% of all highway accidents reported in Germany.

Category B comprises eight codes and 49 accidents (0.006% of all highway accidents) related to road characteristics that are not covered by the scenario matrix. Road geometry data used to prepare the list of scenarios is based on Japanese Road Structure Regulations (Japan Road Association, 2004), but it may not cover some characteristics of German highways. To cover the remaining eight signs, adaptation to German road characteristics may be necessary.

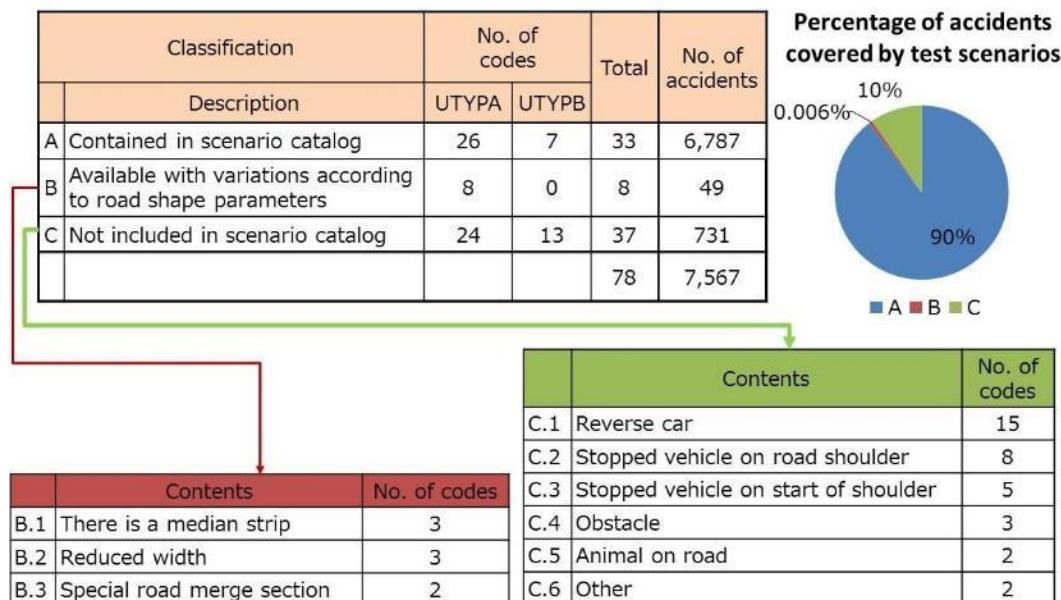


Figure D-1. Scenario database and number of cases (per road and ego-driving).

Category C includes 37 codes and 731 accidents (10% of total) that are not covered by the proposed safety method. Further analysis of codes indicates that three code subcategories (total of 28 codes) were unlawful operations such as driving in the wrong direction on a highway or unlawful parking on the shoulder (C1–C3). Seven remaining codes include obstacles on the road, animals, and other unknowns (C4–C6). Prevention of collision in this category (C) is difficult for AD engineers. For example, an auxiliary approach such as tighter regulations is necessary.

D.2 Pre-crash scenario typology for crash avoidance research (NHTSA)

The NHTSA Pre-Crash Scenario Typology for Crash Avoidance Research defines pre-crash scenario typology for crash avoidance research based on the NHTSA general estimate system crush database. This typology comprises pre-crash scenarios that present vehicle motion, dynamics, and important phenomena immediately before a crash (Najm, Smith, & Yanagisawa, 2007). By applying the same methodology to the GIDAS data, a comparison can be made between typology and the list of scenarios developed in the present report. This typology includes 27 pre-crash scenario categories, 16 of which are about highway accidents. By comparing the scenario database developed from these categories, the completeness of the scenario database can be verified (Figure D- 2). This comparison shows that 6 out of 16 categories are subject to the traffic scenario database. The remaining 10 codes belong to categories that include unlawful or unpreventable actions. For complete coverage, an auxiliary approach for vehicle engineering may be necessary.

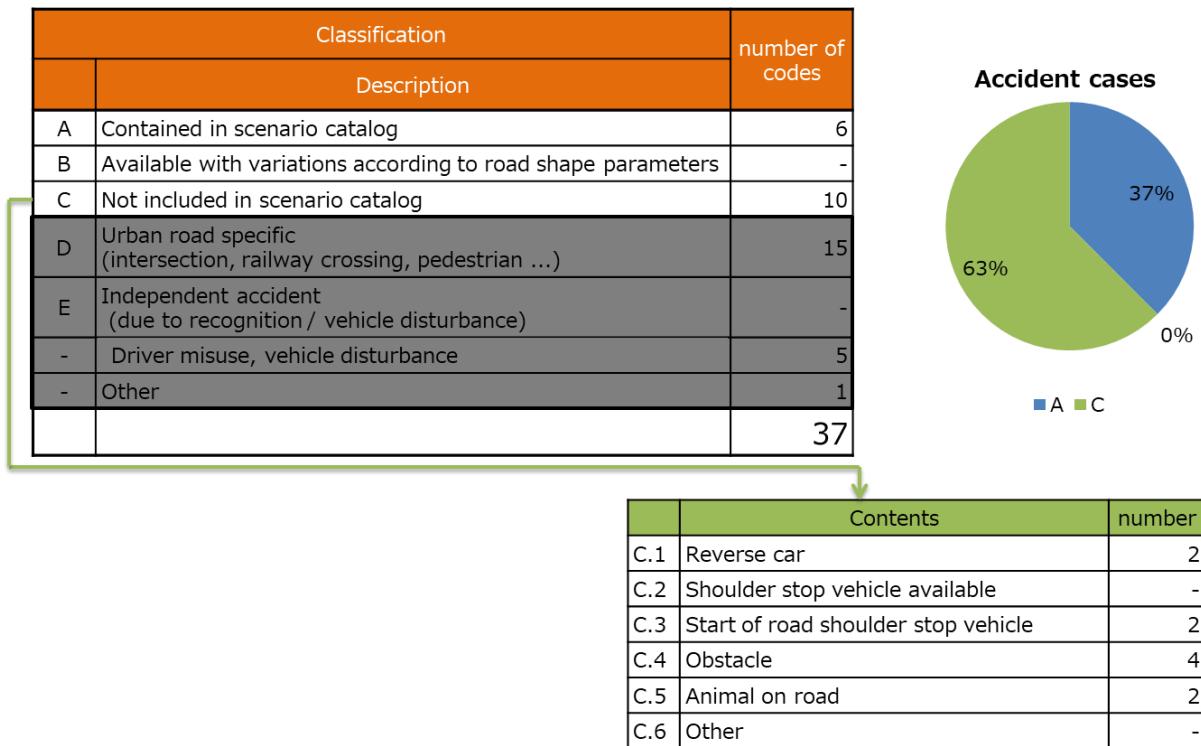


Figure D- 2. Comparison of traffic scenario database and NHTSA pre-crash categories

Annex E

Principle models and evaluation scenarios of perception disturbances

As described in 4.2.1, the principle models of each sensor should be understood and the parameters with their ranges which characterize the models should be defined, in order to derive the perception disturbance scenarios based on sensors' principles. The principle models, parameters with their ranges and the representative of evaluation scenarios for perception disturbances generated in sensors of mmWave Radar, LiDAR and Camera are written up below.

E.1 The processes of principle models description and evaluation scenario derivation

Principle models and evaluation scenarios of perception disturbances are derived according to the following procedure.

- Describe a phenomenon which occurs as a perception disturbance and identify phenomenon parameters
- Make out the model (= principle model) which describes the phenomenon above and identify principle parameters
- List up causal factors and their parameters which contribute to changes of the principle parameters
- Identify a range of each causal factor parameter
- Describe the perception disturbance as change of the causal factor parameters, and define an evaluation scenario with the combination of parameter changing and a traffic flow scenario

Here, any causal factors can be selected for an evaluation scenario in the case that these are described in the same principle model, while the range of causal factor parameters should cover the range of ODD of a system.

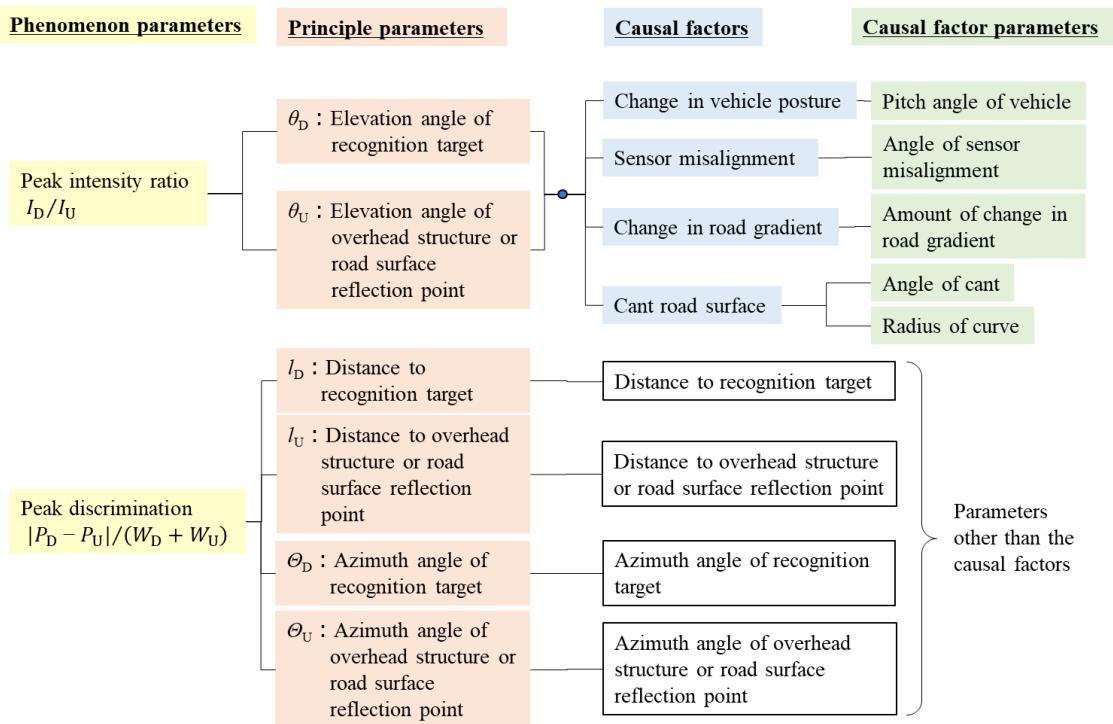


Figure E- 1. Example of a relationship between phenomenon parameters, principle parameters, causal factors and causal factor parameters of a perception disturbance

E.2 The principle models and evaluation scenarios of mmWave Radar

As examples for mmWave Radar, following 4 of principle models and evaluation scenarios of perception disturbances are described.

- Large difference of signal (S) (recognition target)
- Low D/U (road surface multipath)
- Low D/U (change of angle)
- Low S/N (direction of a vehicle)

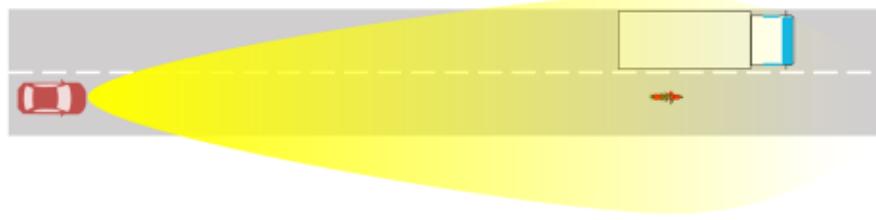
E.2.1 [mmWave Radar] Large difference of signal (S) (recognition target)

E.2.1.1 The Phenomenon and Principle

Large difference of signals
(recognition target)

E.2.1.1.1 The Phenomenon

When pedestrians or motorcycles, etc., that have relatively weak reflection pass by the side of a recognition target with an intense reflection (such as a truck), the reflection signals from the motorcycles, etc., then become buried in the intense reflection signals from the truck, resulting in a false negative.



Phenomenon Parameters

Reflection point cloud

Region	Degree/amount	Duration
Full area of the recognition target	Completely unobtainable	Cannot be obtained for a continuous period of time

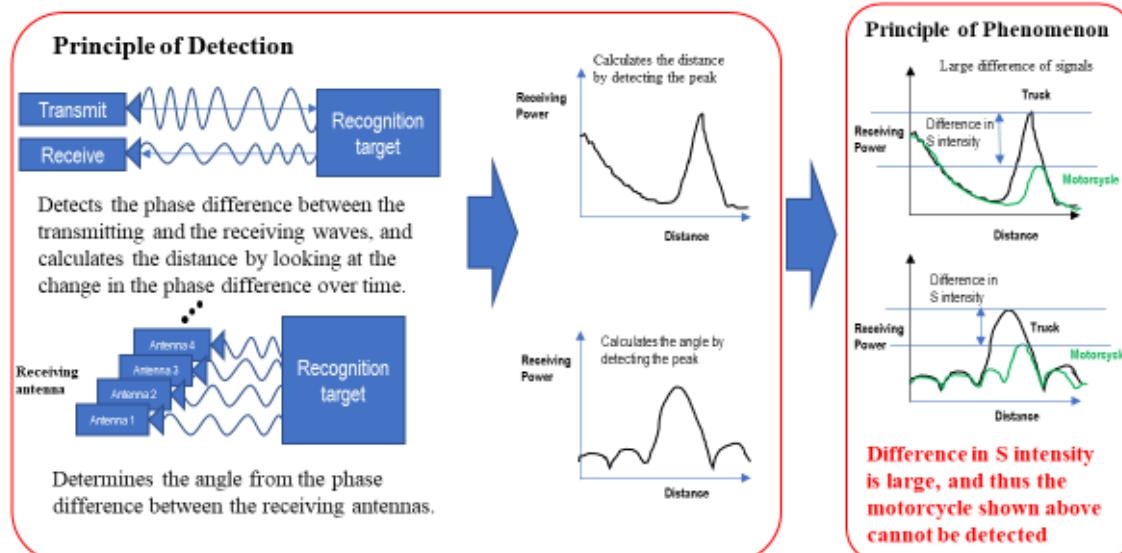
Phenomenon Mode

False negative; a target exists but is not detected.

Large difference of signals
(recognition target)

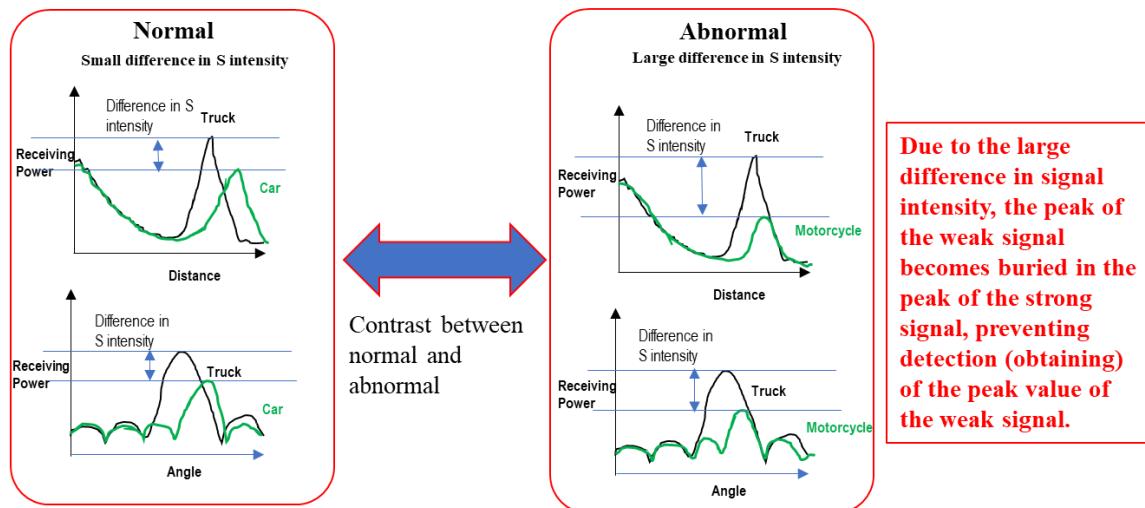
E.2.1.1.2 Outline of the Principle

Due to the large difference in the reflection intensity of the targets, the small signal becomes buried in the large signal, resulting in the recognition target with the weaker reflection going undetected.



Large difference of signals
(recognition target)

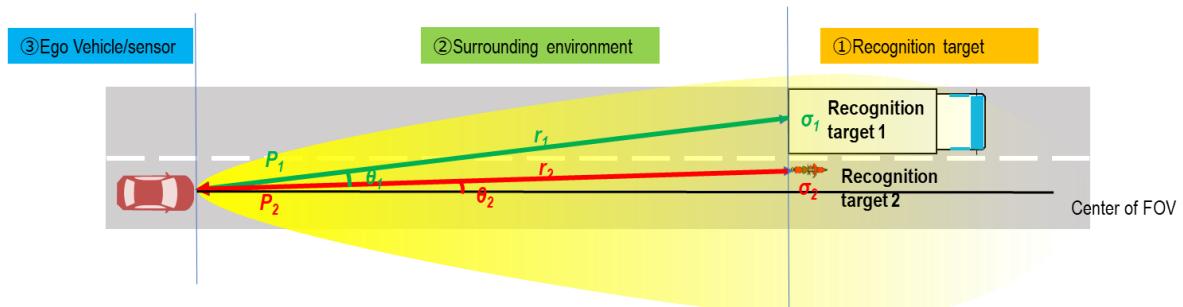
Contrast Between Normal and Abnormal



E.2.1.1 The Phenomenon and Principle

Large difference of signals
(recognition target)

E.2.1.1.3 Principle Model



Model Description
S intensity is the power value of the reflection signal.
Focus on a model describing power

P_t : Power of transmitted waves
 P_n : Power of reflected waves from the recognition target (n)
 λ : Wavelength of radio waves
 $G(\theta)$: Antenna gain
 σ_n : RCS (radar cross section) of the recognition target (n)

※If there are n numbers of recognition target, insert the reference number of each recognition target as a subscript to P , r , and σ . E.g. (n=1,2,3,...)

Power by the reflection from the recognition target (n)

$$P_n = \frac{\lambda^2 \{G(\theta_n)\}^2 P_t}{(4\pi)^3 r_n^4} \sigma_n \quad \rightarrow \quad \sigma = \lim_{r \rightarrow \infty} 4\pi r^2 \left| \frac{E_s(\vartheta, \varphi)}{E_i} \right|^2$$

E_s : Scattered electric field from the recognition target
 E_i : Incidence electric field into the recognition target

In the case of 2 recognition targets, the power from reflection is:

$$P_1 = \frac{\lambda^2 \{G(\theta_1)\}^2 P_t}{(4\pi)^3 r_1^4} \sigma_1 \quad P_2 = \frac{\lambda^2 \{G(\theta_2)\}^2 P_t}{(4\pi)^3 r_2^4} \sigma_2$$

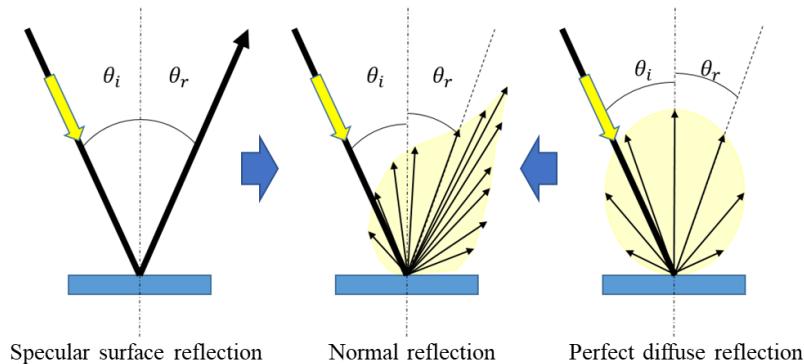
Large difference of signals
(recognition target)

$$\text{When focusing on RCS} \quad \sigma = \lim_{r \rightarrow \infty} 4\pi r^2 \left| \frac{E_s(\vartheta, \varphi)}{E_i} \right|^2$$

E_s : Scattered electric field from the recognition target
 E_i : Incidence electric field into the recognition target

The radar cross-section (RCS) of the recognition target is expressed as the product of the **projected area, reflectance and the directivity of the scattered waves**.

The **directivity of the scattered waves** referred to here, is normally a combination of a specular surface reflection and a perfect diffuse reflection.



Large difference of signals
(recognition target)

Reflectance is...

In the case of vertical polarization:

$$R_p = \frac{\left| \varepsilon_2 \cos \psi_0 - \sqrt{\varepsilon_1 (\varepsilon_2 - \varepsilon_1 \sin^2 \psi_0)} \right|^2}{\left| \varepsilon_2 \cos \psi_0 + \sqrt{\varepsilon_1 (\varepsilon_2 - \varepsilon_1 \sin^2 \psi_0)} \right|^2}$$

In the case of horizontal polarization:

$$R_s = \frac{\left| \sqrt{\varepsilon_1} \cos \psi_0 - \sqrt{\varepsilon_2 - \varepsilon_1 \sin^2 \psi_0} \right|^2}{\left| \sqrt{\varepsilon_1} \cos \psi_0 + \sqrt{\varepsilon_2 - \varepsilon_1 \sin^2 \psi_0} \right|^2}$$

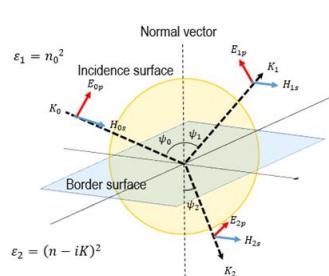
In the case the reflector is a metal, permittivity (ε_r) is:

$$\varepsilon_r = 1 - \frac{\omega_p^2}{\omega^2}$$

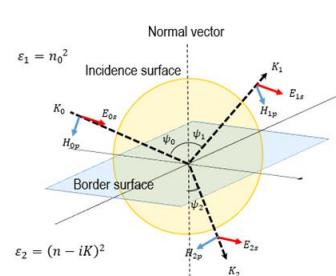
R_p : Reflectance with horizontal polarization
 R_s : Reflectance with vertical polarization

ε_1 : Permittivity of air ε_2 : Permittivity of reflector
 ψ_0 : Incidence angle of the wave ε_r : Permittivity of metal
 ω : Radio wave frequency ω_p : Plasma frequency

(a) In the case of vertical polarization:



(b) In the case of horizontal polarization:



※ The relationship between permittivity and relative permittivity ⇒
Relative permittivity = permittivity of the medium / permittivity in the vacuum

In terms of 'Reflectance,' the parameters are **permittivity and incidence angle, and permittivity is a parameter related to the material.**

Further, **projected area** refers to the reflective surface area of the recognition target, and this will vary depending on the shape, orientation, size and relative position of the recognition target.

Large difference of signals
(recognition target)

Let's summarize what we have learnt so far...

The intensity of the reflected signal (S) from the recognition target depends on the power value (P).

The power value will depend on the positional relationship of the radar to the recognition target, and the RCS of the recognition target.

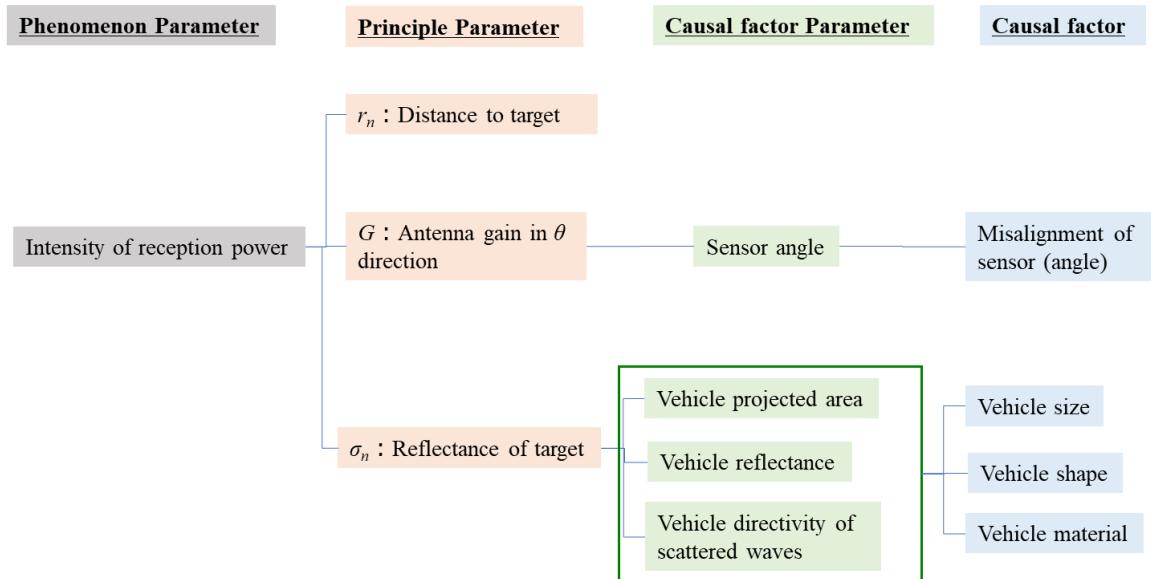
Therefore, all we need to do is understand the distance and angle between the radar and the recognition target, as well as the RCS.

E.2.1.2 Relationship between Principle & Causal Factors of Perception Disturbance

Large difference of signals
(recognition target)

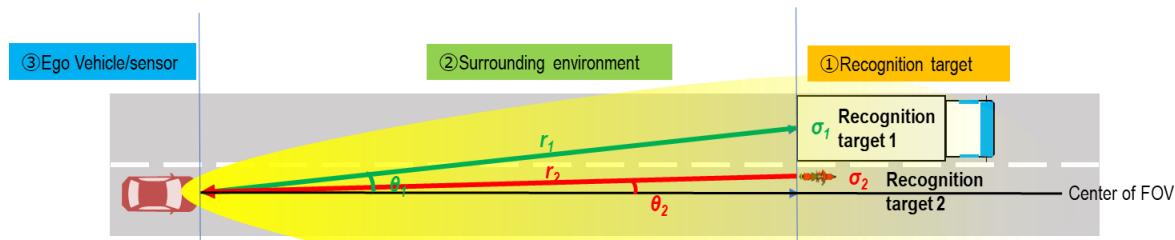
E.2.1.2.1 Disturbance based on principle

The below is a summary of the relationship so far, between the phenomenon parameter, the principle parameter, the causal factor parameter and the causal factor.



Large difference of signals
(recognition target)

E.2.1.2.1 Causal factors of Perception Disturbance based on Principle



※ If there are n numbers of recognition targets, insert the reference number of each recognition target as a subscript to P, r, and σ. E.g. (n=1,2,3,...)

Phenomenon Parameter	Principle Parameter	Causal factor Parameter	Causal factor		
			①Target	②Surrounding environment	③Ego vehicle/sensor
Signal Intensity	Target distance	-	-	-	-
	Antenna gain	-	-	-	-
		Sensor angle	-	-	Sensor misalignment
	Retroreflectivity RCS value (σ_n)	Shape of recognition target	3D shape of subject of target	-	-
		Shape of recognition target	Size	-	-
		Vehicle material (permittivity)	Color	-	-
			Material	-	-
	Combination of recognition targets	↔	-	-	-

Large difference of signals
(recognition target)

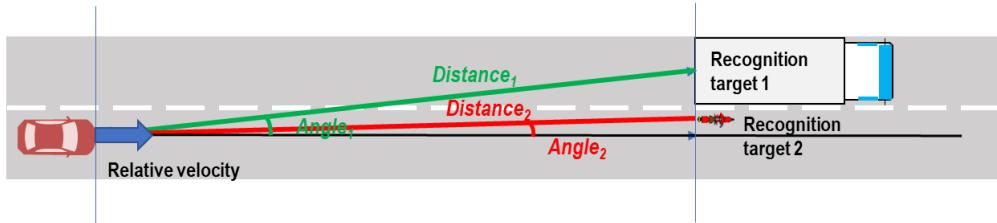
E.2.1.2.2 Parameter Range

Phenomenon Parameter	Principle Parameter	Causal factor Parameter	Causal factor	Parameter Range	Explanation
Signal Intensity	Distance to target	↔	↔	Distance to target (r_n) minimum detectable distance to maximum detectable distance	To evaluate the perceptual device of the radar, test using the range determined by the given radar's specs
	Antenna gain	↔	↔	Within target angle (θ_n) FOV range	Evaluate by varying the parameter within the FOV range determined by the radar's specs
		Sensor angle	Sensor misalignment	Misalignment angle 0 to $\pm X$ deg	Minimum angle where auto-misalignment detection will activate
	Retroreflectivity RCS value (σ_n)	Shape of Recognition target	Shape of Recognition target (3D)	Recognition targets are persons or motor vehicles as classified in the Road Traffic Act First step is large-sized motor vehicles and ordinary two-wheeled motor vehicles	Take into account vehicles which can travel on express ways + persons walking by the side of a stationary vehicle stopped for an emergency
		Size of Recognition target	Size of Recognition target	Vehicle: Motorized bicycle (equivalent) to large-sized motor vehicle (equivalent) Person: ---	Take into account vehicles which can travel on express ways + persons walking by the side of a stationary vehicle stopped for an emergency
		Vehicle material	Color	Define using data on reflectance/transmittance in millimeter waveband	Require database as there is on correlation between detectable colors and physical property values in millimeter wave band
			Material	Define using data on physical property values in millimeter waveband	Require database for physical property values in millimeter wave band
	Combination of Recognition targets	↔	↔	Recognition targets are persons or motor vehicles as classified in the Road Traffic Act	Take into account vehicles which can travel on express ways + persons walking by the side of a stationary vehicle stopped for an emergency

Large difference of signals
(recognition target)

E.2.1.2.3 Evaluation Scenario

- Evaluation is conducted using a recognition target with a strong reflection and a recognition target with a weak reflection
- Evaluation is conducted with the recognition target having a relative speed to the ego vehicle in the direction of approach
- Evaluation is done by varying the positional relationship of the vehicles or motorcycles, etc. (the recognition target)



Parameter Item	Range		Explanation
Distance to recognition targets	Variable	Min to max detectable range	Validate by varying the distance between the min and max detectable distance of the sensor
Angle to recognition targets	Variable	Within FOV	Validate by varying the angles within the radar FOV
Number of recognition targets	Fixed	2	Decide on a recognition target with a weak and one with strong reflection (1 each)
Relative velocity to recognition targets	Fixed	Min to max detectable speed	Validate within the radars min and max detectable speed
Type and combination of recognition targets	Fixed	Vehicle, motorcycle, pedestrian	Representative traffic participant/one with high reflectivity and one with low

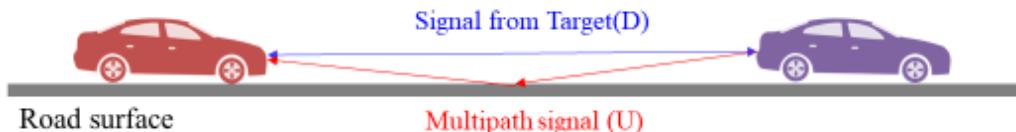
E.2.2 [mmWave Radar] Low D/U (road surface multipath)

E.2.2.1 The Phenomenon and Principle

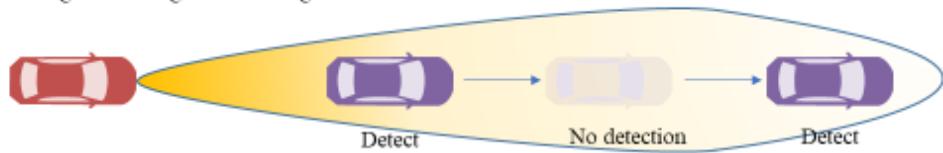
Low D/U
(Road surface multipath)

E.2.2.1.1 The Phenomenon

When there is interference between the signal from the recognition target (D: Desired-Signal) and the signal from the indirect path via the road surface (U: Undesired-Signal), the “signal intensity” of the received signal by the sensor from the recognition target becomes smaller, resulting in a false negative.



At certain distances within the detection area, the “signal intensity” decrease, resulting in the recognition target becoming ‘lost’.



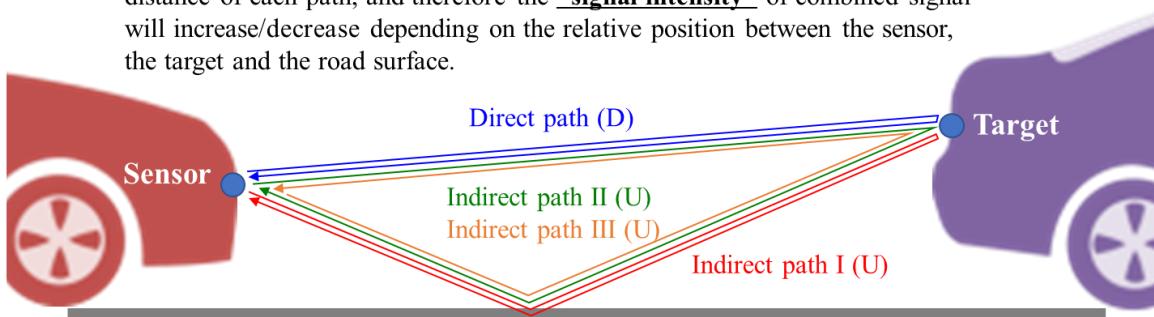
Low D/U
(Road surface multipath)

E.2.2.1.2 The Principle

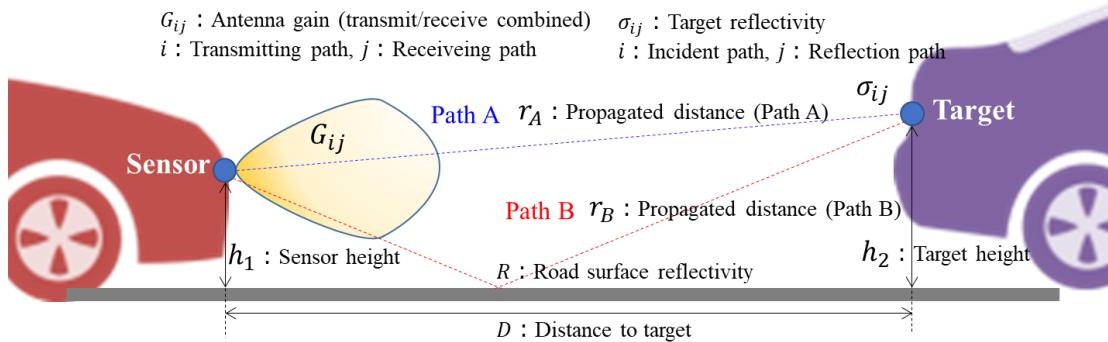
The propagation path, when the signal transmitted from the sensor are reflected by the target and received by the sensor, are categorized into the following four paths:

D/U	Signal path	Propagation path
D:Desired-Signal	Direct path	Sensor → Target → Sensor
U:Undesired-Signal	Indirect path via the road surface I	Sensor → Road surface → Target → Road surface → Sensor
	Indirect path via the road surface II	Sensor → Target → Road surface → Sensor
	Indirect path via the road surface III	Sensor → Road surface → Target → Sensor

The signal received by the sensor are combined of the above signals. The amplitude/phase of each signal will depend on the reflectivity and propagated distance of each path, and therefore the “signal intensity” of combined signal will increase/decrease depending on the relative position between the sensor, the target and the road surface.



Low D/U
(Road surface multipath)

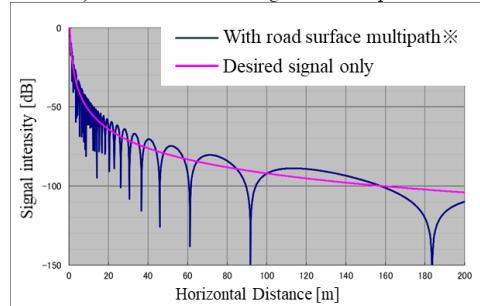


The signal received through each path is calculated as below:

Path	Signal amplitude	Signal phase
Direct path Path A→Path A	$\frac{P_{tx}\lambda^2}{(4\pi)^3} \cdot \frac{G_{AA}\sigma_{AA}}{r_A^4}$	$\phi_0 + \frac{2\pi}{\lambda} \cdot (2r_A)$
Indirect path I Path B→Path B	$\frac{P_{tx}\lambda^2}{(4\pi)^3} \cdot \frac{G_{BB}\sigma_{BB}R^2}{r_B^4}$	$\phi_0 + \frac{2\pi}{\lambda} \cdot (2r_B) + 2\pi$
Indirect path II Path A→Path B	$\frac{P_{tx}\lambda^2}{(4\pi)^3} \cdot \frac{G_{AB}\sigma_{AB}R}{r_A^2 r_B^2}$	$\phi_0 + \frac{2\pi}{\lambda} \cdot (r_A + r_B) + \pi$
Indirect path III Path B→Path A	$\frac{P_{tx}\lambda^2}{(4\pi)^3} \cdot \frac{G_{BA}\sigma_{BA}R}{r_B^2 r_A^2}$	$\phi_0 + \frac{2\pi}{\lambda} \cdot (r_A + r_B) + \pi$

The propagated distance (r_A, r_B), transmitting path (i) and receiving path (j) are determined by the sensor height (h_1), target height (h_2) and distance to the target (D). $(\lambda : \text{wavelength})$

The “signal intensity” decrease at certain distances (determined by conditions) due to the combined signals of each path.



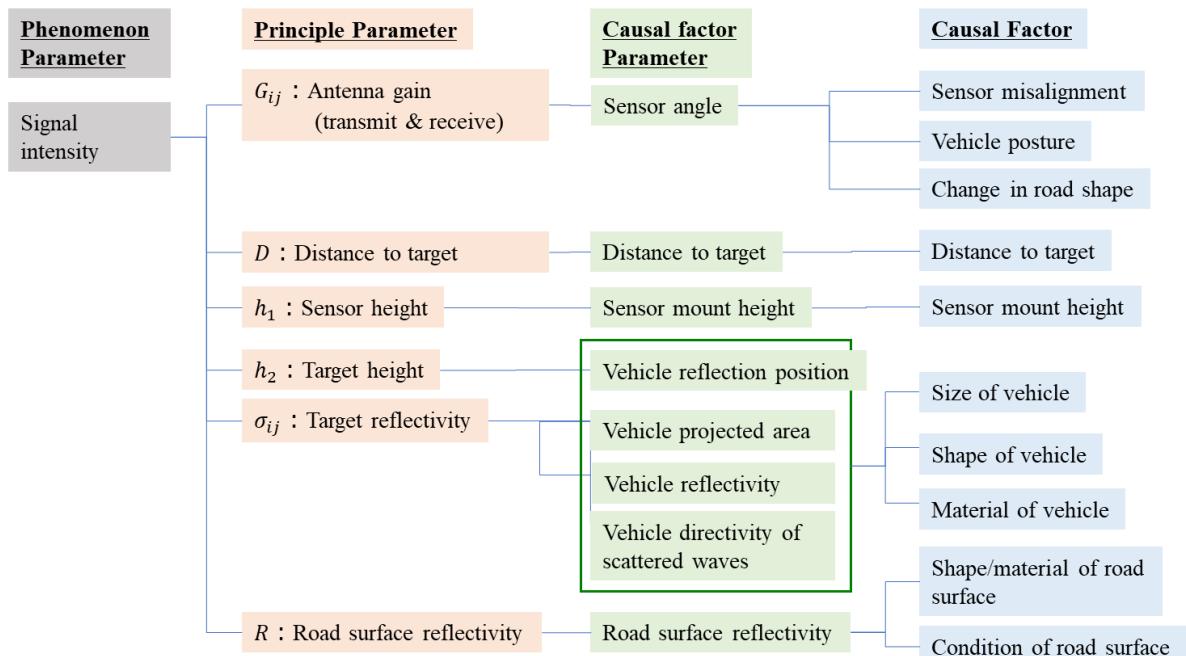
※Example of the direct path and an indirect path

E.2.2.2 The Relationship Between Principle and Causal factor

Low D/U
(Road surface multipath)

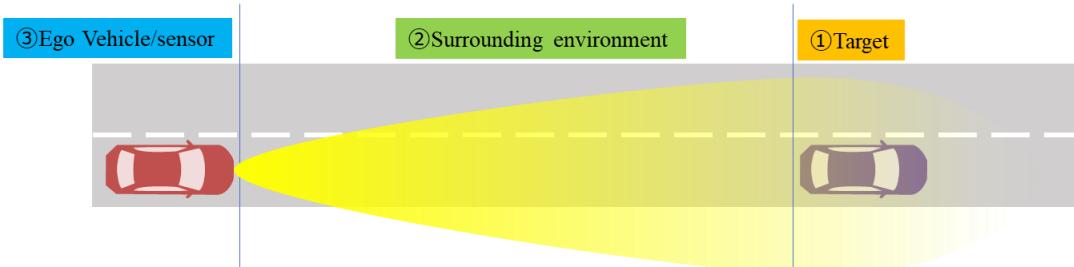
E.2.2.2.1 Causal factor based on Principle

The below is a summary of the relationship between the Principle parameter, causal factor parameter and causal factor which contribute to the “signal intensity” (phenomenon parameter).



Low D/U (Road surface multipath)

The below table shows the relationship between the phenomenon parameter, principle parameter, causal factor parameter and the causal factor.



Phenomenon Parameter	Principle Parameter	Causal factor Parameter	Causal factor		
			①Target	②Surrounding environment	③Ego vehicle/sensor
Signal intensity	Antenna gain	Sensor angle	—	Change in road shape Change in vehicle posture	Sensor misalignment Change in vehicle posture
	Target distance	←	Distance to target	—	—
	Sensor height	Sensor mount height	—	—	Sensor mount height
	target height	Vehicle reflection position	Size of vehicle Shape of vehicle Material of vehicle	—	—
		Vehicle projected area		—	—
		Vehicle reflectivity		—	—
		Vehicle directivity of scattered waves		—	—
	Road surface reflectivity	←	—	Shape/material of road surface Condition of road surface	—

Low D/U (Road surface multipath)

E.2.2.2.2 Parameter Range

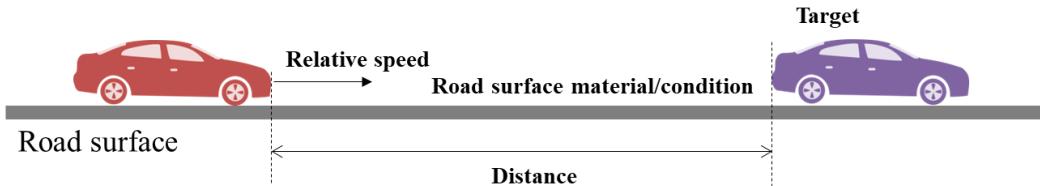
Listing the range of causal factor parameter.

Phenomenon Parameter	Principle Parameter	Causal factor Parameter	Causal factor	Parameter Range	Explanation
Signal intensity	Antenna gain	Sensor angle	Sensor misalignment	Offset angle: 0 to ±X deg	Minimum angle where auto-misalignment detection will activate
			Change in vehicle posture	Pitch angle : 0 to ±X deg	Max angle possible by the vehicle
			Change in road incline	Vertical gradient : -9 – 9%	Article 20 of the Road Construction Ordinance
	Distance to target	←	←	Distance to target : X to Y m	Min to max range detectable by the sensor
	Sensor height	Sensor mount height	←	Mount height : X to Y m	Range of imaginable mounting positions
	Target height	Vehicle reflection position	Size of vehicle Shape of vehicle Material of vehicle	Target classified as motor vehicles under the Road Traffic Act First step is to select three representative types Large-sized vehicle (height : high) Normal vehicle (height : med) Small-sized vehicle (height : low)	The size, shape and material of a vehicle each have complex impacts on each cause parameter. We need to measure the representative examples (large-sized, normal and small-sized vehicles, etc.) and study the impact on each cause parameter.
		Vehicle projected area			
		Vehicle reflectivity			
		Vehicle directivity of scattered waves			
	Road surface reflectivity	←	Shape/material of road surface	All imaginable tracks Asphalt, concrete, gravel, sand, cobblestone...	We need to measure and study the impacts of materials and road surface conditions which affect reflectivity.
			Condition of road surface	All imaginable road surface conditions Wet, ice burn, road repair remains, snow buildup, rut...	

Low D/U (Road surface multipath)

E.2.2.2.3 Evaluation Scenario

- The ego vehicle approaches the target(stationary vehicle) up ahead in its path.



Parameter Item	Variable/Fixed	Range	Explanation
Distance to target	Variable	Min to Max detection Range	Min to max range detectable by the sensor
Relative speed	Fixed	Max speed within ODD	
Target type	Fixed	Large-sized vehicle (height : high) Normal vehicle (height : medium) Small-sized vehicle (height : low)	Three levels of representative examples such as large-sized vehicles, normal vehicles, and small-sized vehicles
Road surface material	Fixed	Asphalt / Metal plate(TBD)	Typical road surface material / highly reflective road surface material
Road surface condition	Fixed	Dry / Wet	Normal road surface condition / highly reflective road surface condition

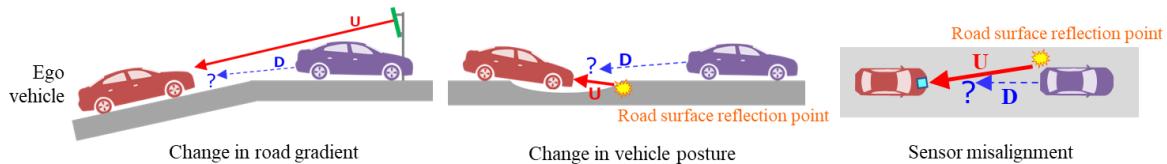
E.2.3 [mmWave Radar] Low D/U (change of angle)

Low D/U
(Change of the angle)

E.2.3.1. Phenomenon and Principle of Perception Disturbance

E.2.3.1.1. Phenomenon

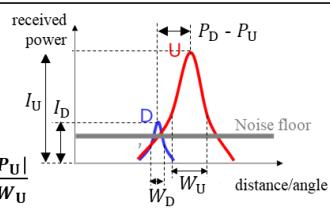
When the radar's central axis of FOV and the road's surface/traveling direction are not parallel due to the road's gradient/cant, the vehicle's posture or due to sensor misalignment, etc., then the reflected signal from the recognition target becomes relatively smaller than the undesired signals from surrounding structures, thus causing it to become buried and resulting in a false negative.



Phenomenon Parameters

$$\text{Peak intensity ratio} \quad \frac{I_D}{I_U}$$

$$\text{Peak discrimination} \quad \frac{|P_D - P_U|}{W_D + W_U}$$



I_D, I_U : Peak intensity
 P_D, P_U : Peak position
 W_D, W_U : HMFW

Phenomenon Mode :

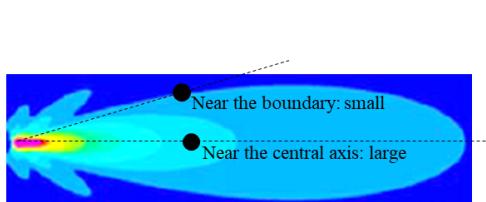
False negative (situation where although a recognition target exists, it is not detected)

Low D/U
(Change of the angle)

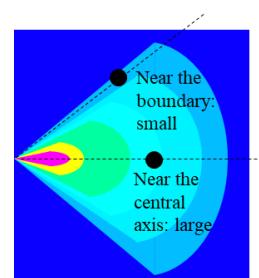
E.2.3.1.2. Principle of Perception Disturbance

- When taking a cross-section of the intensity distribution of the radar's transmission waves in the vertical and horizontal directions, the intensity becomes relatively smaller as the angle moves further away from the central axis of the FOV. (see below)
- The intensity of the receiving wave will also vary depending on whether the reflective object is placed near the central axis or the boundary.

<Distribution of the Radio Field Intensity from a mmWave Radar>



Vertical section

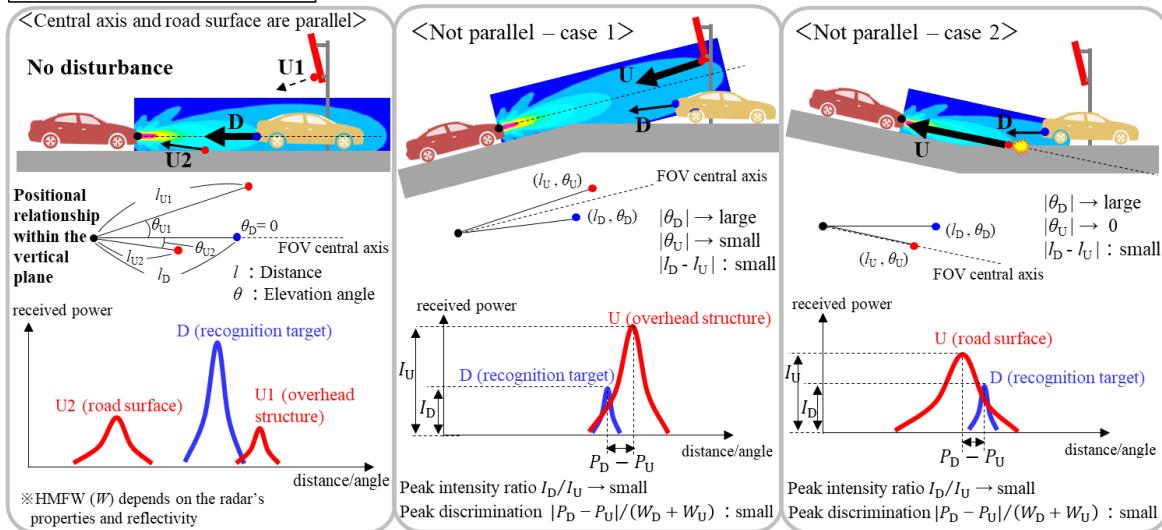


Horizontal section

Low D/U
(Change of the angle)

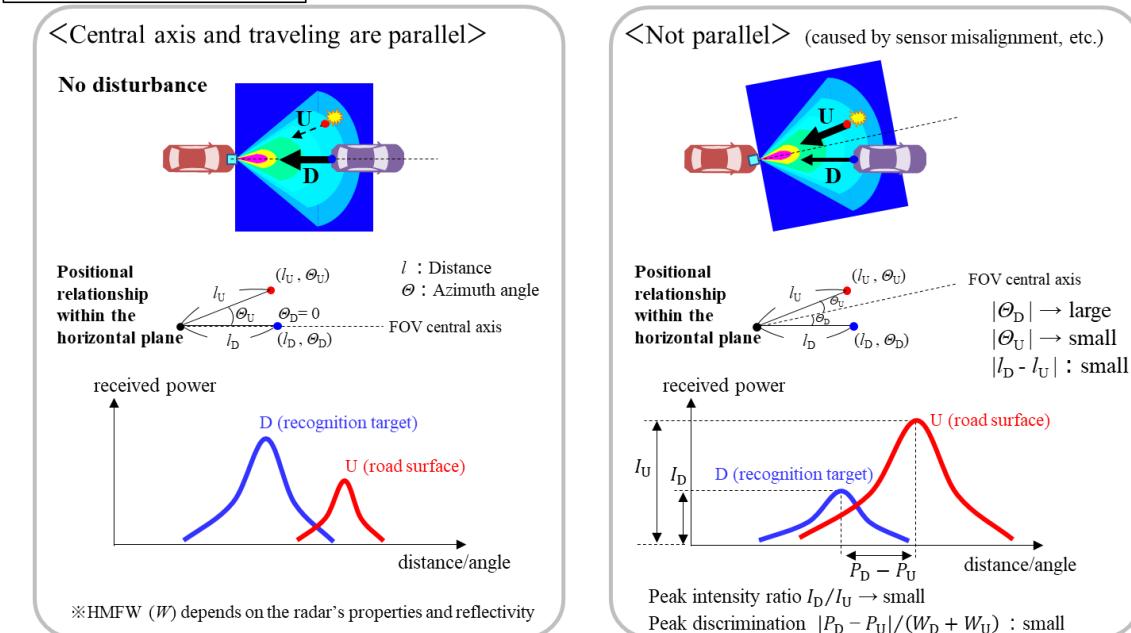
- When the central axis of the radar's FOV is not parallel to the road surface/traveling direction, surrounding structures will move closer to the central axis, and the recognition target will move closer to the boundary. ($|\theta_U| \rightarrow \text{small}$, $|\theta_D| \rightarrow \text{large}$)
- As a result, the intensity of the undesired signal from surrounding structures (I_U) becomes relatively larger than the intensity of the recognition target's signal (I_D).
- When another condition is added which makes the peak discrimination low ($|P_D - P_U| : \text{small}$ or $|W_D + W_U| : \text{large}$), signal D becomes buried in U, and is therefore not detected.

Change in Elevation Angle



Low D/U
(Change of the angle)

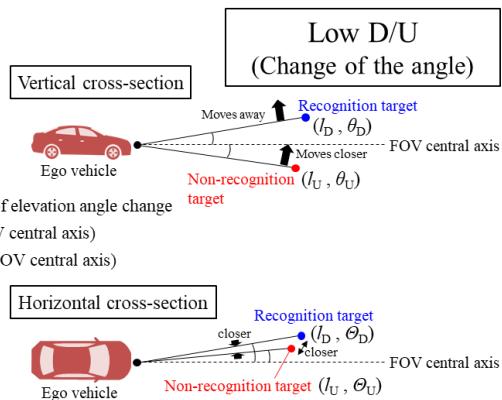
Change in Azimuth Angle



Only the parameters which represent Low D/U phenomenon caused by change of elevation angle are described below
because the phenomena in cases of elevation angle change and azimuth angle change are the same essentially.

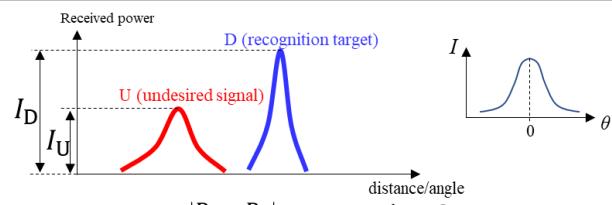
<Parameter conditions for where D becomes buried in U> ※in case of elevation angle change

- Elevation angle $|\theta_D| \rightarrow$ large (recognition target moves away from the FOV central axis)
- Elevation angle $|\theta_U| \rightarrow$ small (non-recognition target moves closer to the FOV central axis)
- Distance to recognition target $l_D \approx l_U$
- Azimuth angle of recognition target $\theta_D \approx \theta_U$
- The sum of HMFW $W_D + W_U \rightarrow$ large
(※Dependent on the radar's properties and reflectivity)



Principle Parameters:

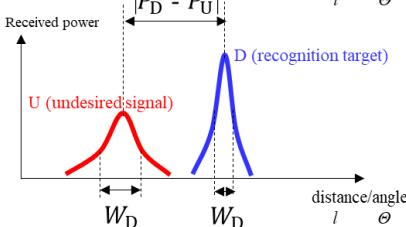
Elevation angle θ_D, θ_U (variable)
⇒ affects I_D/I_U value



**Distance l_D, l_U
Azimuth angle θ_D, θ_U**
⇒ affects $|P_D - P_U|/(W_D + W_U)$ value



※ $\theta_D, \theta_U, \theta_D, \theta_U$ are the angles in relation to the FOV central axis

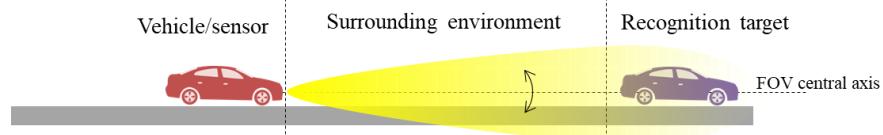


E.2.3.2. Relationship between Principle & Causal Factors of Perception Disturbance

Low D/U
(Change of the angle)

E.2.3.2.1. Causal factors of Perception Disturbance based on Principle

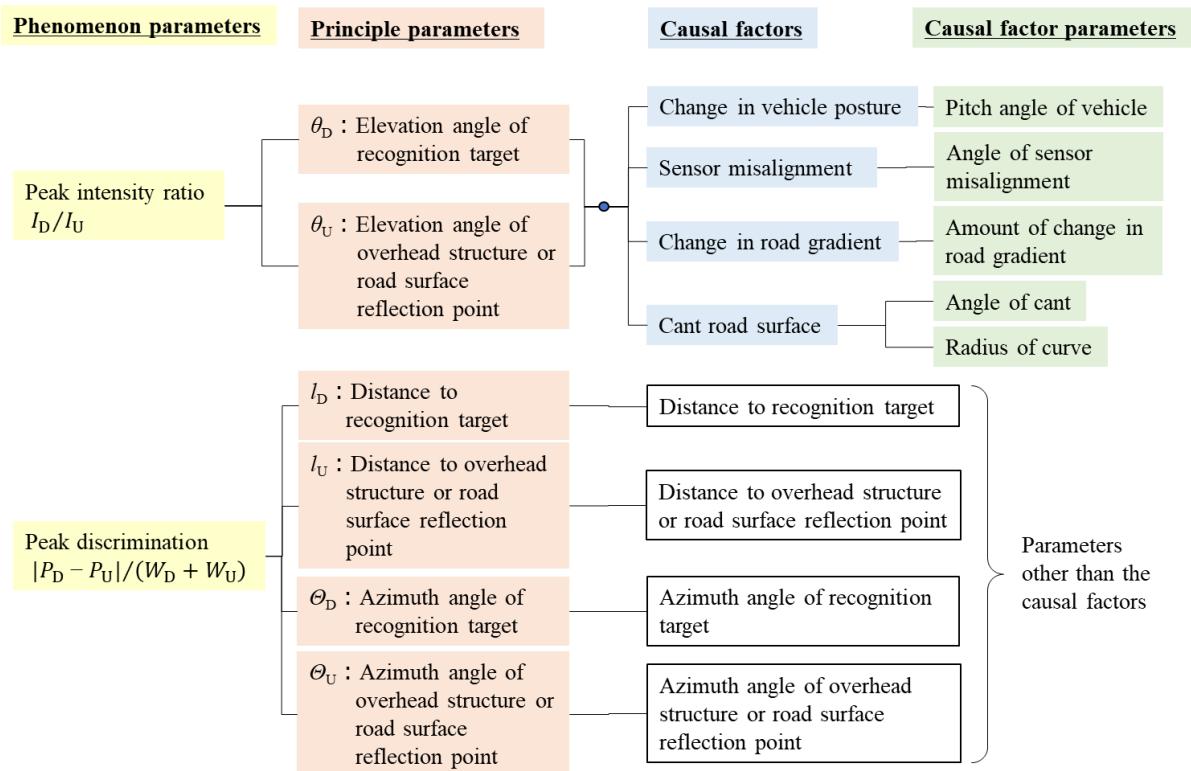
Of the principle parameters, the below are the causal factors of the perception disturbance which give rise to a change in elevation angle $|\theta_D| \rightarrow$ large and $|\theta_U| \rightarrow$ small, being the cause for this phenomenon.



Variable Principle Parameters	Causal factors of the disturbance		
	Vehicle/sensor	Surrounding environment	Recognition target
Elevation angle θ_D, θ_U	<ul style="list-style-type: none"> Change of the vehicle posture 	<ul style="list-style-type: none"> Change in road gradient 	N/A
	<ul style="list-style-type: none"> Misalignment of the sensor 	<ul style="list-style-type: none"> Cant road surface 	

Below shows the relationship between the phenomenon parameters, principle parameters and causal factor parameters.

Low D/U
(Change of the angle)



Low D/U
(Change of the angle)

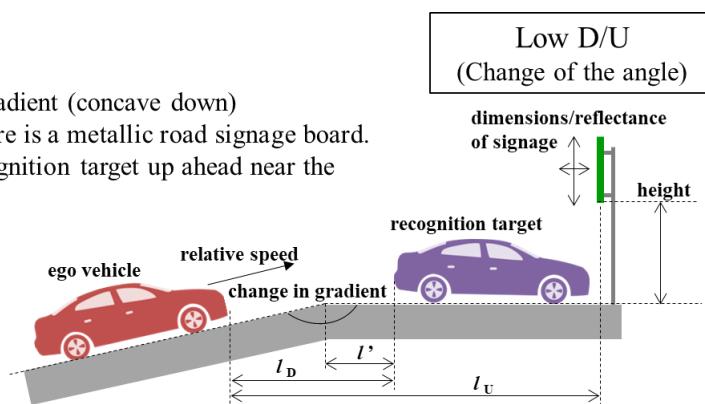
E.2.3.2.2. Parameter Range

Phenomenon Parameters	Principle Parameters	Contributing Causal Factors	Causal Factor Parameters	Range of Causal Factor Parameters	Explanation
Peak intensity ratio I_D/I_U	Elevation angles θ_D, θ_U (variable parameters)	Change in road gradient	Amount of change in road gradient	0 to 18 % (according to Article 20 of the Road Construction Ordinance, elevation angle - 9 to +9 %)	Evaluation range is the maximum angle possible for the sensor, based on a combination of any one or more factors.
		Cant road surface	Angle of cant	0 to 10 % (according to Article 16 of the Road Construction Ordinance)	
			Radius of curve	∞ to 82 m (according to Article 15 of the Road Construction Ordinance)	
		Sensor misalignment	Angle of sensor misalignment	0 to min. angle where auto-misalignment detection will activate	
		Change in vehicle posture	Pitch angle of vehicle	0 to \pm (vehicle's max. possible angle)	
Peak discrimination $ P_D - P_U /(W_D + W_U)$	Distance to objects l_D, l_U	(Not a causal factor)	Distance to recognition target	0 to min. distance required to avoid collision	
			Distance to non-recognition target	0 to min. distance required to avoid collision	
	Azimuth angles θ_D, θ_U	(Not a causal factor)	Angle of recognition target	0 to \pm (max. angle of the sensor's FOV)	
			Angle of non-recognition target	0 to \pm (max. angle of the sensor's FOV)	

E.2.3.2.3. Evaluation Scenario

- Traveling a road with a change in gradient (concave down)
- Ahead of the change in gradient, there is a metallic road signage board.
- The ego vehicle approaches the recognition target up ahead near the signage board in its path.

※The situation with a gradient change (concave down) is selected as the representative scenario because of the higher probability of large reflective intensity from a metallic overhead structure than the road surface.



Parameters		Parameter Range		Explanation
Causal factor	Change in the road gradient	Variable	0 to 18 % equivalent	Use a road which is concave down as a representative
Other than the causal factor	Initial distance to recognition target l_D	Fixed	Distance required to avoid collision	
	Distance to recognition target from the inflection point l'	Variable	0 to l_D	
	Lateral position of recognition target	Fixed	0°	Fixed on the same lane
	Initial distance to signage board l_U	Variable	$l_D - 5$ to $l_D + 5$ (m)	
	Lateral position of signage board	Variable	-3.5 to +3.5 (m)	assume the object within the neighboring lanes
	Height of signage board (to bottom edge)	Fixed	4.5m (above road)/1.5m (roadside)	According the Traffic Sign Installation Standard
	Dimensions of the signage board	Fixed	2.7 × 3.5 (m)	Guidance signage on highways
	Reflectance of the signage board	Fixed	Measured value of the real board	
	Relative speed	Fixed	Max. speed within ODD	
Type of the recognition target		Fixed	Passenger vehicle/Pedestrian	Representative traffic participant/low reflectance

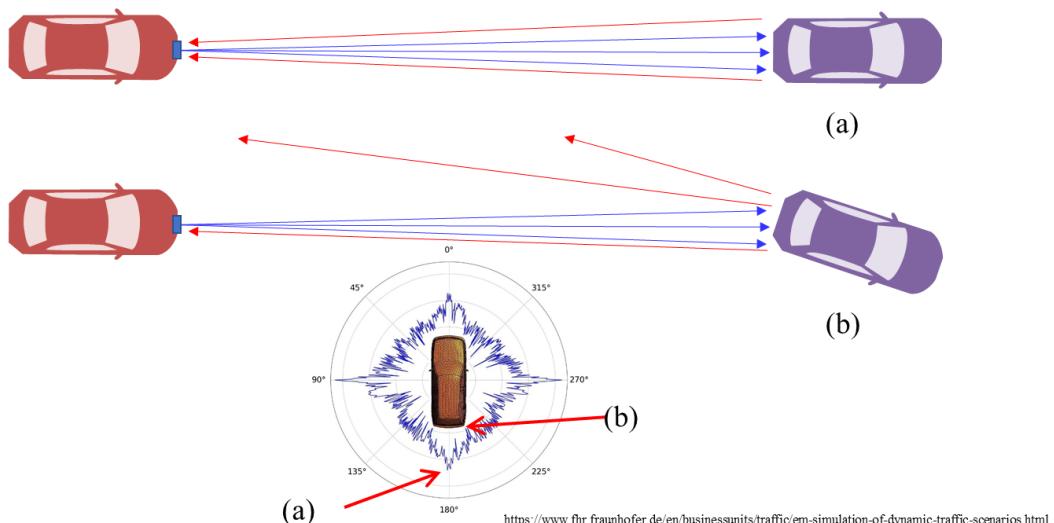
E.2.4 [mmWave Radar] Low S/N (direction of a vehicle)

Low S/N
(Orientation of the vehicle)

E.2.4.1. The Phenomenon and Principle of Perception Disturbance

E.2.4.1.1 The Phenomenon

Electromagnetic waves are transmitted from the radar, and the intensity of the reflected electromagnetic waves which return in the direction of the radar, will depend on the projected area, reflectance and orientation of the target's surface. If the same vehicle is on a different angle, this can cause the reflection to become extremely weak, thus the vehicle, although it may be within the FOV, may go undetected.



<https://www.fir.fraunhofer.de/en/businessunits/traffic/em-simulation-of-dynamic-traffic-scenarios.html>

Low S/N
(Orientation of the vehicle)

E.2.4.1.2 Outline of the Principle

When reflected waves from the target are received by the radar, the intensity of the signal (S) received by the radar will depend on the receiving power (P_r) as determined by the below radar equation.

$$P_r = \frac{\lambda^2 \cdot P_t \cdot G_t(\theta) \cdot G_r(\theta) \cdot \sigma}{(4\pi)^3 \cdot R^4}$$

In this equation, P_t is the transmitting power, $G_t(\theta)$ is the transmitting antenna gain, $G_r(\theta)$ is the receiving antenna gain, σ is the target's radar cross-section, λ is the wavelength, and R is the range between the radar and the target.

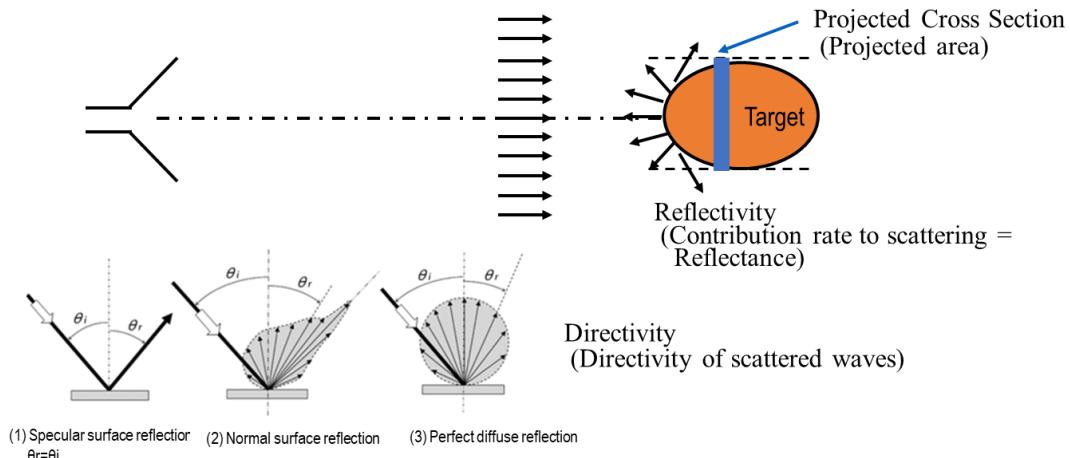
As evident by this radar equation, the orientation of the vehicle contributes to low S/N (the radar cross-section (σ) of the target, or in other words the vehicle, will depend and vary according to the orientation).

Low S/N
(Orientation of the vehicle)

The radar cross-section (σ) is expressed as a product of (a) the target's projected area, (b) the contribution rate to scattering, and (c) the directivity of scattered waves. If an object uses the same material, then the area with high directivity (in other words the points facing the radar) will have stronger reflection.

Further, the contribution rate to scattering (= Reflectance) is, “metal $\doteq 1$ ” and “ $0 \leq \text{non-metal} < 1$ ”.

$$\text{Radar cross-section } (\sigma) = \text{Projected Cross Section} \times \text{Reflectivity} \times \text{Directivity } (\text{m}^2)$$



E.2.4.2. Relationship between Principle & Causal Factors of Perception Disturbance

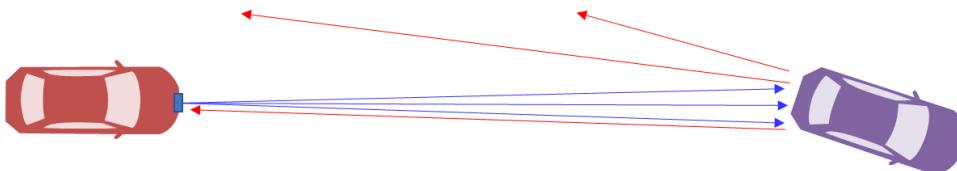
Low S/N
(Orientation of the vehicle)

E.2.4.2.1 Causal factors of Perception Disturbance based on Principle

Phenomenon Parameter	Principle Parameter	Causal Factor Parameter	Causal Factor
	$r : \text{Distance to target}$	Distance to target	Distance to recognition target when viewed from the radar
Signal intensity	$G : \text{Antenna gain in } \theta \text{ direction}$	Angle of target when viewed from the center line of the radar	Position (angle) of recognition target when viewed from the radar
	$\sigma : \text{Target's radar cross-section}$	Projected area	Size of vehicle
		Contribution rate to scattering (Reflectance)	Material of vehicle
		Directivity of scattered waves	Shape of vehicle

Low S/N
(Orientation of the vehicle)

As explained in E.2.4.1, the radar cross-section will vary depending on the orientation of the vehicle. Even if the size, shape and material of the vehicle remain the same, depending on the angle at which it is viewed, the projected area, contribution rate to scattering, and the directivity of the scattered waves will differ, thus making them causal factor parameters.



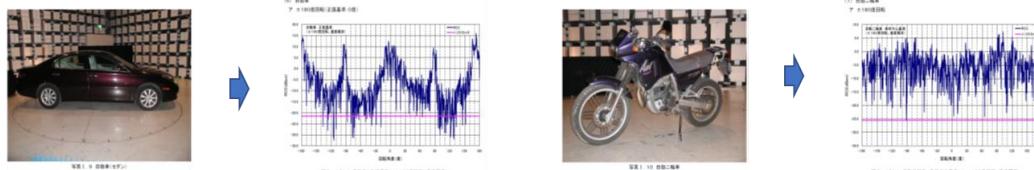
Phenomenon Parameter	Principle Parameter	Causal Factor Parameter	Causal Factors contributing to the change in principle parameter		
			①Recognition target	②Surrounding environment	③Vehicle/sensor
Signal intensity	Radar cross-section	Projected area	Size of vehicle	-	-
		Contribution rate to scattering (Reflectance)	Material of vehicle	-	-
		Directivity of scattered waves	Shape of vehicle	-	-

Low S/N
(Orientation of the vehicle)

E.2.4.2.2 Parameter Range

If the target is something of a complex shape, such as a vehicle, the relationship between the projected area, reflectance, and directivity will be complex. Thus, the radar cross-section (σ) (large, medium and small) has been selected based on previous research, etc.

Examples of Past Research (Measurement Results)



Source) JARI report (J-GLOBAL ID : 200909086392246974), 2004

Phenomenon Parameter	Principle Parameter	Causal factor Parameter	Causal Factor	Parameter Range	Explanation
Signal intensity	Cross-section area of radar reflection	Projected area	Size of vehicle	3 representative models	Stipulate with the sizes of existing vehicles in the world using 3 rep models (large, medium and small)
		Contribution rate to scattering (Reflectance)	Material of vehicle	↑	Stipulate with the materials of existing vehicles in the world using 3 rep models (large, medium and small)
		Direction of scattered waves	Shape of vehicle	↑	Stipulate with the shape of existing vehicles in the world using 3 rep models (large, medium and small)

Low S/N
(Orientation of the vehicle)

E.2.4.2.3 Evaluation Scenario

A scenario whereby the ego vehicle approaches the recognition target (stationary vehicle) up ahead in the ego vehicle's lane, at a constant speed.



Parameter Item	Variable/ Fixed	Range	Explanation
Type of recognition target	Variable	<ul style="list-style-type: none"> Projected area (large/mid/small) Contribution rate to scattering= Reflectance (heavy use of metal / heavy use of non- metal / in-between) Directivity of scattered waves (uniform/biased) 	<ul style="list-style-type: none"> 3 levels of projected area generally 3 levels (no vehicle has zero metal used) 3 levels (relying on concentration of normal vectors in microparts of the vehicle)
Orientation of the target	Variable	0 to 30 deg.	According to the line of the road (curve R)
Distance to the target	Variable	5 to 150 m	
Relative speed	Fixed	20 km/h and below	constant

E.3 The principle models and evaluation scenarios of LiDAR

As examples for LiDAR, following 2 of principle models and evaluation scenarios of perception disturbances are described.

- Attenuation of signal (recognition target)
- Noise

E.3.1 [LiDAR] Attenuation of signal (S) (recognition target)

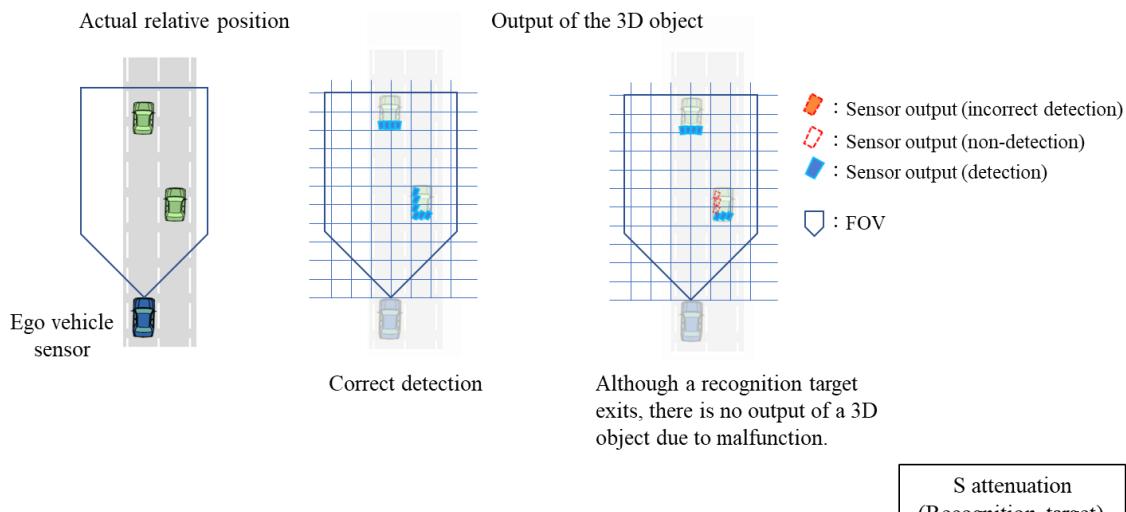
E.3.1.1 The Phenomenon and Principle

S attenuation
(Recognition target)

E.3.1.1.1 The Phenomenon

Explaining the Phenomenon (1/2)

The situation whereby the target cannot be detected as a three-dimensional object unless it is closer in range than what is the assumed detectable range (false negative).



Explaining the Phenomenon (2/2)

The phenomenon occurs when the reflection points attached to a recognition target, are continuously not output to the point cloud.

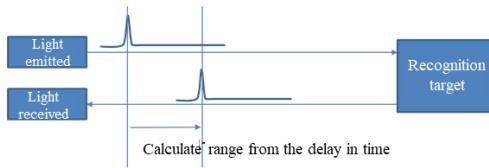
			A		B		C	
			1	2	1	2	1	2
Phenomenon Parameter	Degree	Amount of attenuation	Reflection points which should exist are not output to the point cloud					
		Region of attenuation	Full area within frame		Attached to the subject of recognition		Attached to the frame	
	Time	Duration of attenuation	Continuous	Temporary	Continuous	Temporary	Continuous	Temporary
Phenomenon Mode			— S attenuation is limited to the recognition target, therefore, it is not the full FOV.		The reflection points attached to a recognition target, are continuously not output to the point cloud \Rightarrow not output as a 3D object		← This malfunction is more severe if ‘continuous’ rather than ‘temporary’; therefore, the ‘continuous case’ is used as the representative case.	
							— S attenuation is limited to the recognition target; therefore, the range outside of the recognition target is not considered.	

E.3.1.1.2 Outline of the Principle

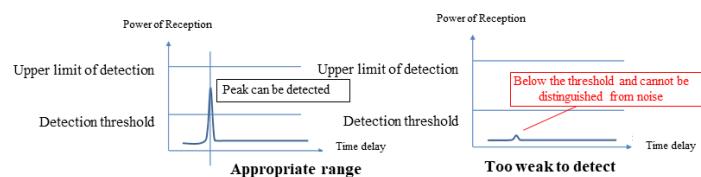
S attenuation
(Recognition target)

The reflection from the target is too weak and thus the peak cannot be detected at the assumed detectable range, leading to the target not being detected.

Detect the peak of the received signal, and calculate the range from the delay in time.



When the reflection is too weak and does not meet the detection threshold, then the S is too small and therefore cannot be detected.

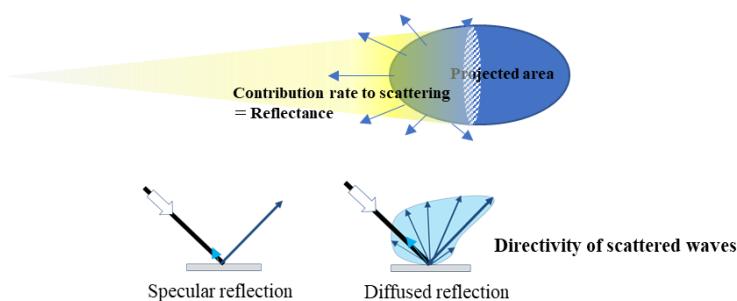


E.3.1.1.3 Principle Model

S attenuation
(Recognition target)

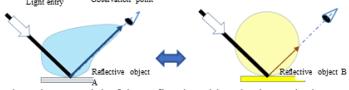
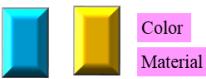
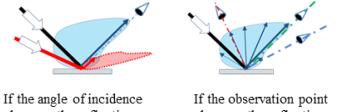
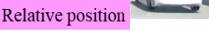
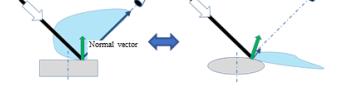
The reflection intensity of the target, the same as for the Millimeter wave, is believed to be expressed as the product of the projected area of the target, the contribution rate to scattering and the directivity of scattered waves, and an object with the same surface material will have a more intense reflection in the directivity (in other words the points facing the LiDAR).

$$\text{Reflection intensity} = [\text{Projected area of the target}] * [\text{Contribution rate to scattering}] * [\text{Directivity of scattered waves}]$$



S attenuation
(Recognition target)

Derive the disturbance of impact, from the principle of the targets' reflection.

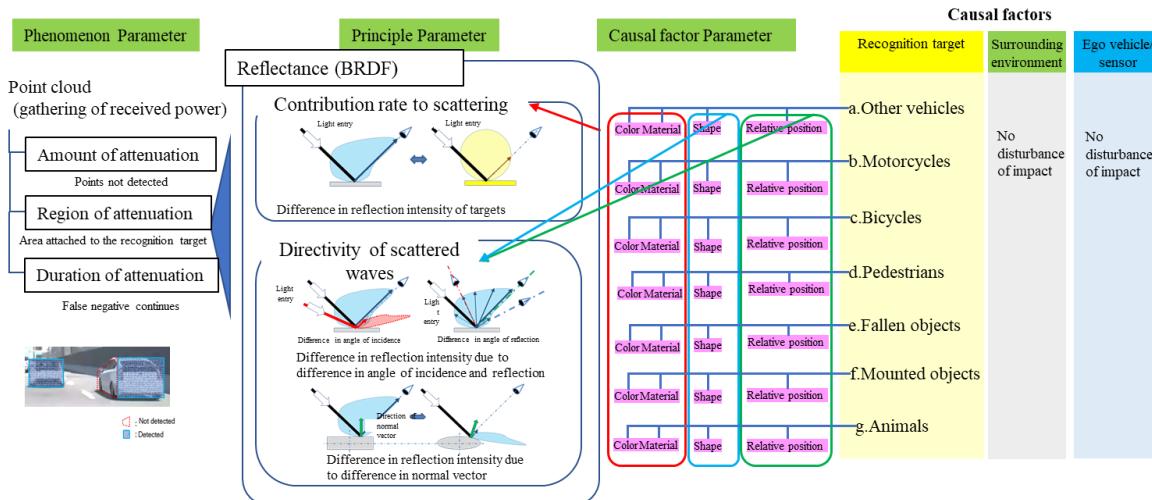
Principle		Causal factors of disturbance
<p>Contribution rate to scattering</p> <p>Difference in the reflection intensity of targets</p>	 <p>When the material of the reflective object is changed, the reflection intensity will change even if the angle of incidence and the observation point remain the same.</p>	<p>The factors that instigate a change in the reflection intensity include different colors and materials used for painted surfaces, clothing, etc.</p>  <p>Picture showing a change in reflection due to different coating used.</p>
<p>Directivity of scattered waves</p> <p>Projected area of the target</p> <p>Difference in the reflection intensity due to a difference in the angle of incidence/reflection</p>	 <p>If the angle of incidence changes, the reflection intensity will change</p> <p>If the observation point changes, the reflection intensity will change</p>	<p>The factors that instigate a change in the reflection intensity when the angle of incidence/reflection changes, include different colors and materials used for painted surfaces, clothing, etc.</p>  <p>Picture showing a change in the reflectance caused by a difference in angle.</p> <p>A factor which instigates a change in the angle of incidence or the observation point, includes the relative position of the ego vehicle to the target.</p> 
Difference in the reflection intensity due to the difference in normal vector.	 <p>If the normal vector of the target's surface changes, the reflection intensity will change.</p>	<p>The factors that instigate a change in the reflection intensity when the angle of incidence/reflection changes, include different colors and materials used for painted surfaces, clothing, etc.</p>  <p>Picture showing a change in the reflection caused by a difference in shape.</p> <p>A factor that instigates a change in the normal vector of a target's surface, includes the shape of the target.</p>  <p>Difference in shape according to model</p>

E.3.1.2 The Relationship Between Principle and Causal Factors of Perception Disturbance

S attenuation
(Recognition target)

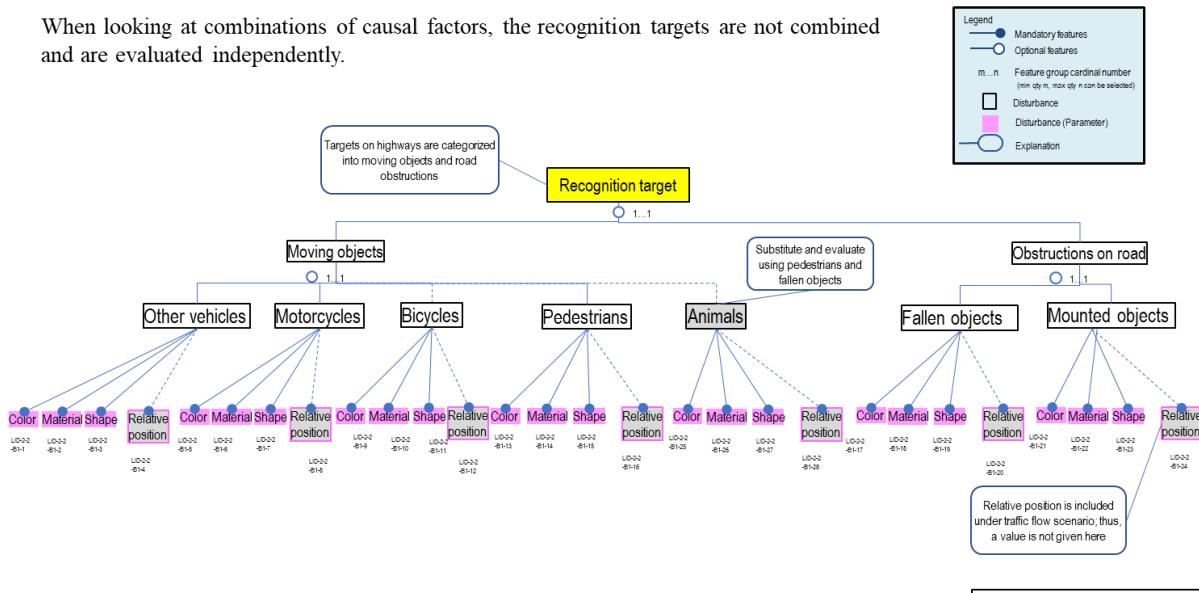
E.3.1.2.1 Causal factors based on Principle

The causal factor parameters are derived from the causal factors of perception disturbance related to reflection from the recognition target based on the principle parameters.



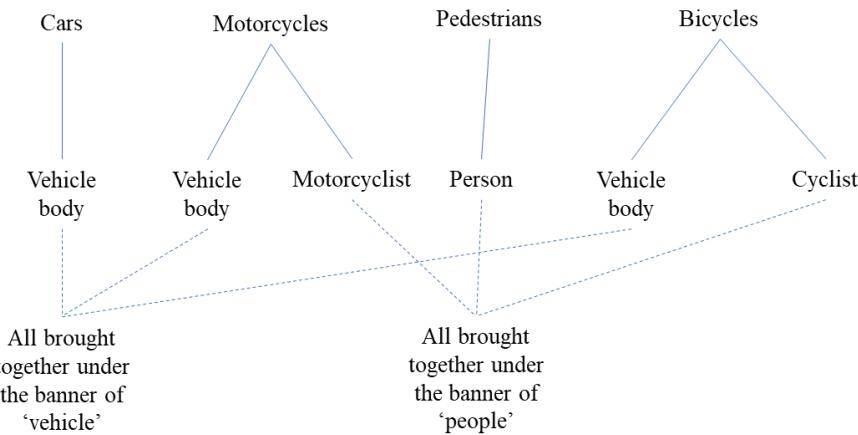
S attenuation
(Recognition target)

When looking at combinations of causal factors, the recognition targets are not combined and are evaluated independently.



E.3.1.2.2 Parameter Range

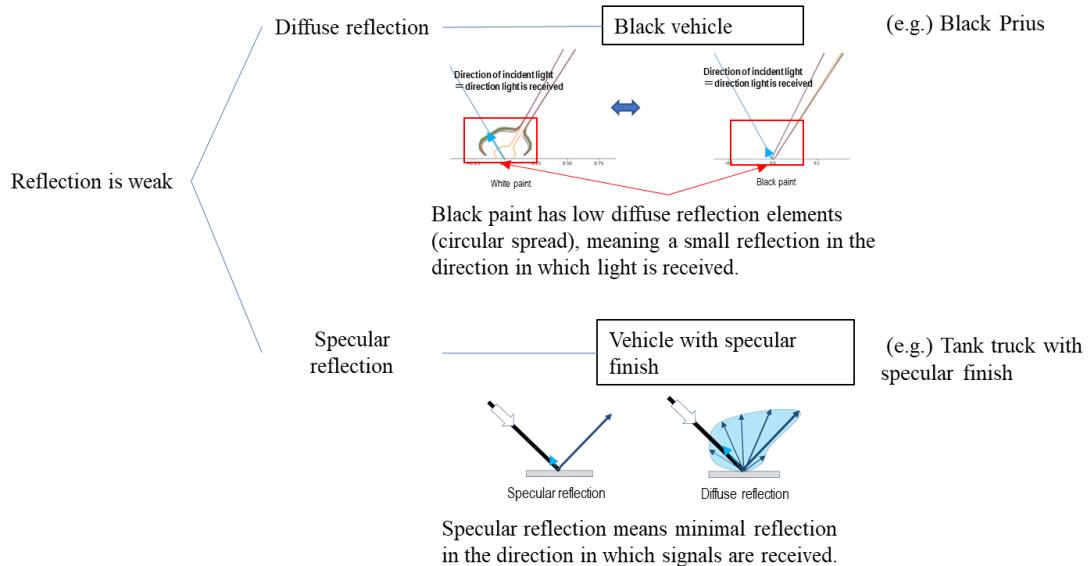
Moving objects are considered in terms of 'vehicles' and 'people'.



S attenuation
(Recognition target)

Vehicle Color Material

The color and the material of a vehicle are looked at in terms of the reflective properties of the surface coating. A black vehicle has been selected for its low diffuse reflection, alongside a vehicle with specular reflection.

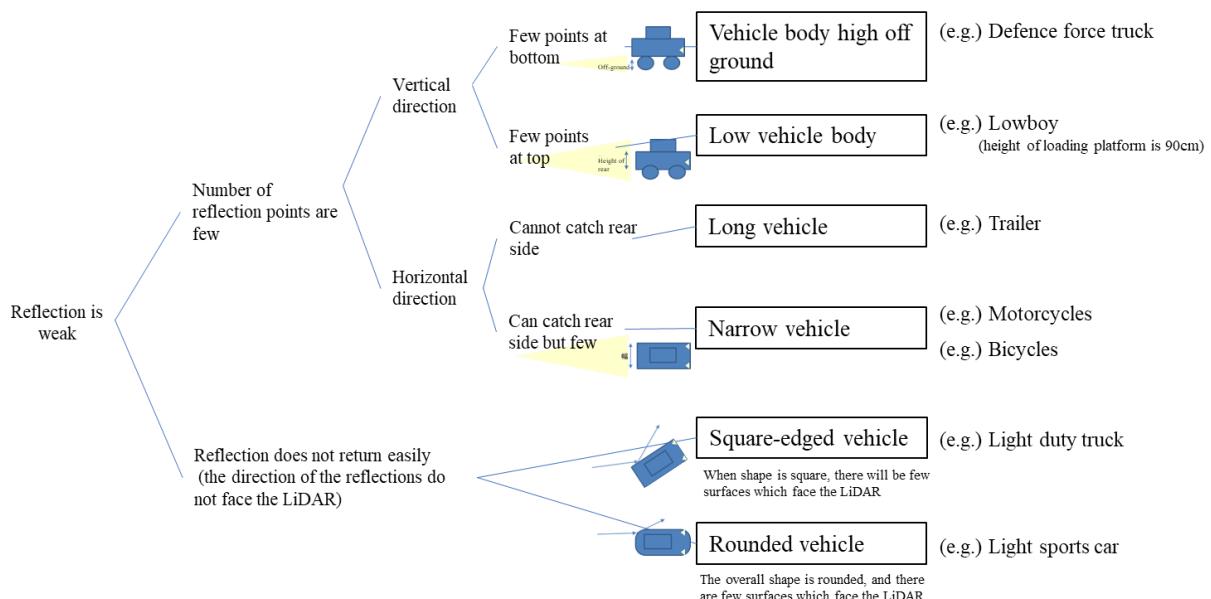


S attenuation
(Recognition target)

Vehicle Shape

The shape of a vehicle is considered in terms of how it hits the LiDAR.

A shape for which the number of reflection points the LiDAR hits is few and a shape for which reflection does not return easily, have been selected.



S attenuation
(Recognition target)

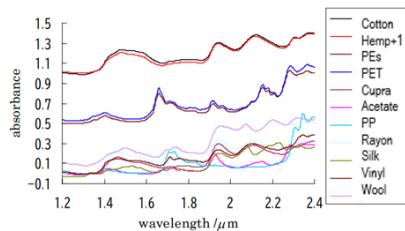
People Color Material

The difference in reflection intensity of pedestrians, motorcyclists (motorcycles) and cyclists (bicycles):

The reflection intensity will differ depending on the person's clothing, luggage, color of skin and hair, helmet, etc.
The parameter range to be considered here is reflection from the clothing only, as this occupies the biggest area.

Select from plant-based (cotton), animal-based (leather) and artificial (chemical fibers, reflective material).

Example of the Different Near Infrared Reflections by Material



http://molsci.center.ims.ac.jp/discussion_past/2003/BK2003/Abs/4pp/4Pp063.pdf

S attenuation
(Recognition target)

People Shape

When considering the difference in the shape of pedestrians we look at their size and posture.

Smaller people are more difficult to detect; therefore, we will consider the build of a Japanese person (who is relatively small) as the worst case scenario.

① Difference in Size (Standing)

	<p>The size of pedestrians is expressed by the frame that surrounds the body. Height, width and thickness correspond to height (B1), shoulder width (D2) and thickness at chest diameter (E2).</p> <table border="1"> <tr> <td>Average Japanese adult male</td> </tr> <tr> <td>Height (B1) : 171.4 cm</td> </tr> <tr> <td>Shoulder width (D2) : 45.6 cm</td> </tr> <tr> <td>Diameter at chest (E2) : 21.1 cm</td> </tr> </table> <p>※ Digital Human Research Center of the National Institute of Advanced Industrial Science and Technology AIST Human Body Measurements Database 1991 - 1992</p> <table border="1"> <tr> <td>Average Japanese adult female</td> </tr> <tr> <td>Height (B1) : 159.1 cm</td> </tr> <tr> <td>Shoulder width (D2) : 40.7 cm</td> </tr> <tr> <td>Diameter at chest (E2) : 21.1 cm</td> </tr> </table> <p>※ Digital Human Research Center of the National Institute of Advanced Industrial Science and Technology AIST Human Body Measurements Database 1991 - 1992</p> <table border="1"> <tr> <td>Average Japanese 3 y/o boy</td> </tr> <tr> <td>Height (B1) : 95.1 cm</td> </tr> <tr> <td>Shoulder width (D2) :</td> </tr> <tr> <td>Diameter at chest (E2) :</td> </tr> </table> <p>※ 2010 Survey by the Ministry of Health, Labour and Welfare</p>	Average Japanese adult male	Height (B1) : 171.4 cm	Shoulder width (D2) : 45.6 cm	Diameter at chest (E2) : 21.1 cm	Average Japanese adult female	Height (B1) : 159.1 cm	Shoulder width (D2) : 40.7 cm	Diameter at chest (E2) : 21.1 cm	Average Japanese 3 y/o boy	Height (B1) : 95.1 cm	Shoulder width (D2) :	Diameter at chest (E2) :
Average Japanese adult male													
Height (B1) : 171.4 cm													
Shoulder width (D2) : 45.6 cm													
Diameter at chest (E2) : 21.1 cm													
Average Japanese adult female													
Height (B1) : 159.1 cm													
Shoulder width (D2) : 40.7 cm													
Diameter at chest (E2) : 21.1 cm													
Average Japanese 3 y/o boy													
Height (B1) : 95.1 cm													
Shoulder width (D2) :													
Diameter at chest (E2) :													

② Difference in Posture

The height from the road will differ depending on posture.
'Posture' is to be considered as a parameter.

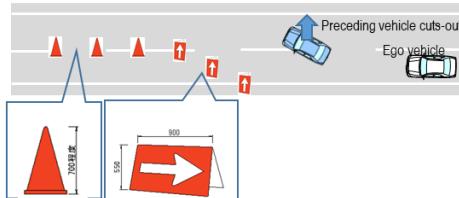
	<p>Sitting height (I1)</p> <p>Average Japanese adult male : 92.6 cm Average Japanese adult female : 86.7 cm</p> <p>※ Digital Human Research Center of the National Institute of Advanced Industrial Science and Technology AIST Human Body Measurements Database 1991 - 1992</p>
	<p>Head length (A1)</p> <p>Average Japanese adult male : 18.9 cm Average Japanese adult female : 18.0 cm, Diameter at chest (E2) Average Japanese adult male : 21.1 cm Average Japanese adult female : 21.1 cm</p> <p>※ Digital Human Research Center of the National Institute of Advanced Industrial Science and Technology AIST Human Body Measurements Database 1991 - 1992</p>
	<p>Riding</p> <p>For riding posture on motorcycles, bicycles, etc., when the rider comes to a stop and places their foot on the ground they will take the shape of standing; thus, we use the same height as a 'standing' person</p>

S attenuation
(Recognition target)

Mounted Objects, Fallen Objects Shape

Mounted objects

For now this includes arrow boards and safety cones which appear at the boarder of driving lanes.



Fallen objects

For now, this will include tires, included under car parts, which ranked one of the highest for fallen objects based on occurrences.

2nd	3rd
Tires	Wood
(outer circumference 503 mm 165/60R12 for light vehicles)	

Object was selected in reference to the ranking in NEXCO's (Central Nippon Expressway Company Limited) study.

- 1st Plastics (hard/soft), fabrics (blankets, bedsheets, etc.): 25,400 occurrences
⇒ × not tall, little impact when driven over
- 2nd Car parts (tires, automobile accessories, etc.): 8,900 occurrences
⇒ ○ some are more than 15cm, can be hard (made of metals, etc.)
- 3rd Wood (square lumber, veneer boards, etc.): 6,900 occurrences
⇒ △ square lumber can be more than 15 cm and hard. To be considered at next stage.
- 4th Road kill (animal corpses): 6,900 occurrences
⇒ × in Japan this is considered to be mainly small animals such as raccoons.
- Other: 17,400 occurrences

S attenuation
(Recognition target)

The below list summarizes the parameter ranges

Principle parameter	Causal factor	Causal factor parameter	Parameter Range	Explanation
Reflectance (BRDF)	Vehicle	Shape	High off-ground vehicle body Low vehicle Motorcycles, bicycles Square-edged vehicles Rounded vehicles	Clears bottom of body and only reflects tires It is difficult for the top layer of beams to hit the loading platform There are few reflection points in the horizontal direction Depending on orientation it is difficult for the normal vector to face the LiDAR It is difficult for the normal vector to face the LiDAR
		Color, Material	Black paint Specular reflection	Has few diffuse reflection elements Depending on the orientation, specular reflection will occur and not return
	Pedestrians	Shape	Big, small Standing, sitting, lying	Evaluate the variations of body build and posture
		Color, Material	Black leather clothing	Of all clothing types, this is assumed to have particularly low reflection
	Mounted objects	Shape	Triangular cones, arrow signs	Appear on tracks as a way of bordering lanes
		Color, Material	Color and material of the above mounted objects	The difference in variations is assumed to be minimal. Low priority.
	Fallen objects	Shape	Tires Wood	Low lying, and difficult for the normal vector to face the LiDAR Low lying, depending on orientation, difficult for the normal vector to face the LiDAR
		Color, Material	Color and material of the above fallen objects	The difference in variations is assumed to be minimal. Low priority.

E.3.1.2.3 Evaluation Scenario

S attenuation
(Recognition target)

Scenario F-1

Evaluate based on “a vehicle cuts-in on a straight road” scenario.

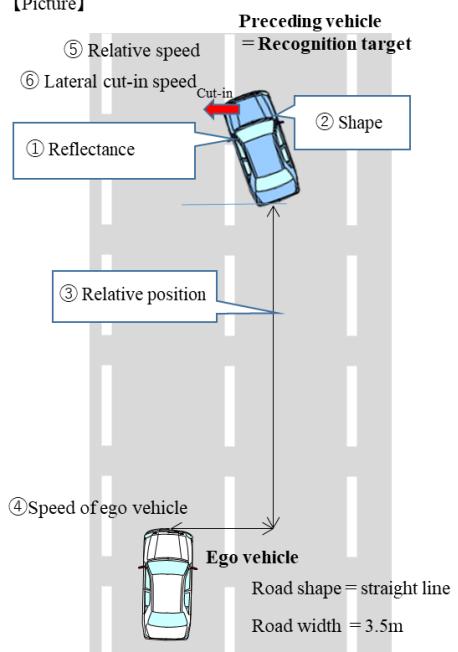
【Outline】

- This evaluation looks at the impact of changes to the shape and reflectance of the cut-in vehicle (recognition target).

【Parameters】

Causal factor parameter Parameters required for evaluation	① Reflectance (directivity)	Coating material = black, specular surface
	② Shape	Vehicle = e.g.) Defence force truck, lowboy, trailer, motorcycle, light duty truck, light sports car
	③ Relative position	This is defined under the traffic flow scenario, thus is not determined here.
	④ Speed of ego vehicle	This is defined under the traffic flow scenario, thus is not determined here.
	⑤ Relative speed	
	⑥ Lateral cut-in speed	

【Picture】



S attenuation
(Recognition target)

Scenario F-2

Evaluate based on “a vehicle cuts-out on a straight road” scenario.

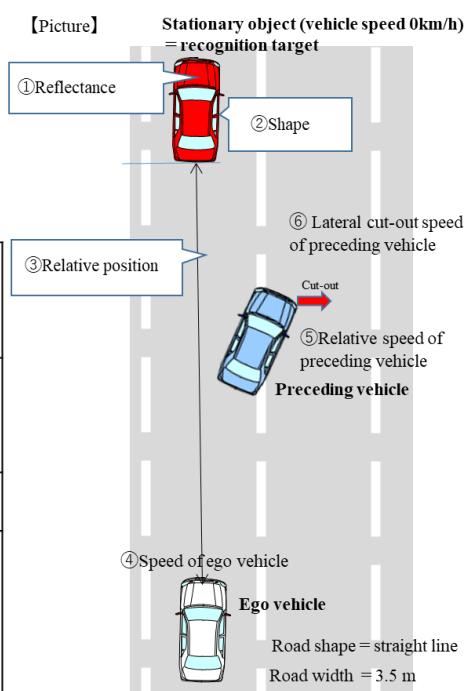
【Outline】

- This evaluation looks at the impact of changes to the shape and reflectance of the stationary object in front of the preceding vehicle which cuts-out (recognition target).

【Parameters】

Causal factor parameter Parameters required for evaluation	① Reflectance (directivity)	Vehicle : Coating material = black, specular surface Person : clothing = leather, chemical fibres, cotton, reflective material Mounted/fallen objects : the reflectance of each target
	② Shape	Vehicle = evaluate using deceleration scenario Person = standing, sitting, lying, traffic controllers, bicycles Mounted objects = safety cones, arrow signs Fallen objects = tires, wood
	③ Relative position	This is defined under the traffic flow scenario; thus, it is not determined here.
	④ Speed of ego vehicle	This is defined under the traffic flow scenario; thus, it is not determined here.
	⑤ Relative speed of preceding vehicle	
	⑥ Lateral cut-out speed of preceding vehicle	

【Picture】



S attenuation
(Recognition target)

Scenario F-3

Evaluate based on “a vehicle decelerates on a straight road” scenario

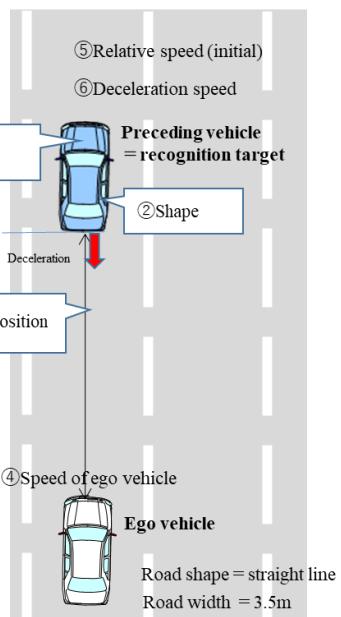
【Outline】

- This evaluation looks at the impact of changes to the shape and reflectance of the decelerating vehicle (recognition target).

【Parameters】

Parameters required for evaluation	① Reflectance (directivity)	Coating material = black, specular surface
	② Shape	Vehicle = e.g.) Defence force truck, lowboy, trailer, motorcycle, light duty truck, light sports car
	③ Relative position	This is defined under the traffic flow scenario; thus, it is not determined here.
	④ Speed of ego vehicle	This is defined under the traffic flow scenario; thus, it is not determined here.
	⑤ Relative speed (initial)	
	⑥ Deceleration speed	

【Picture】



E.3.2 [LiDAR] Noise

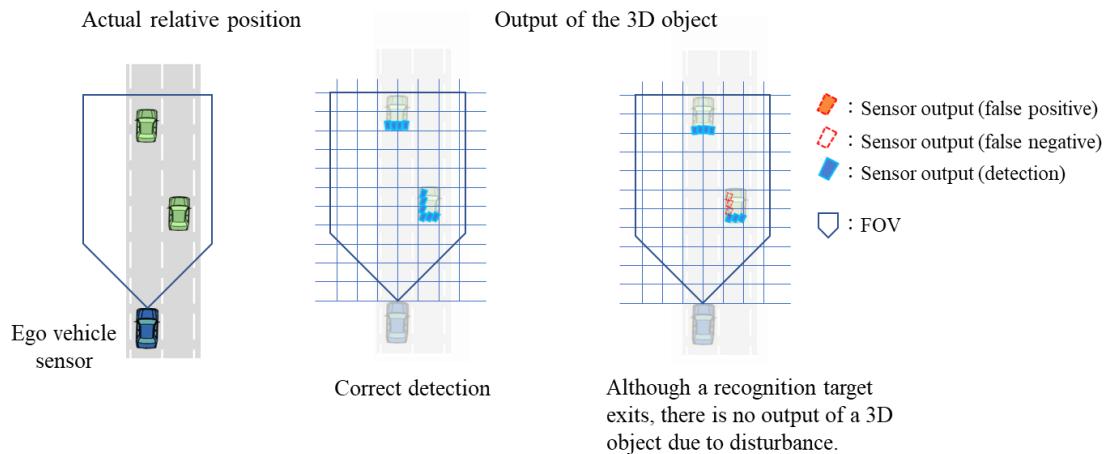
E.3.2.1 The Phenomenon and Principle

Noise

E.3.2.1.1 The Phenomenon

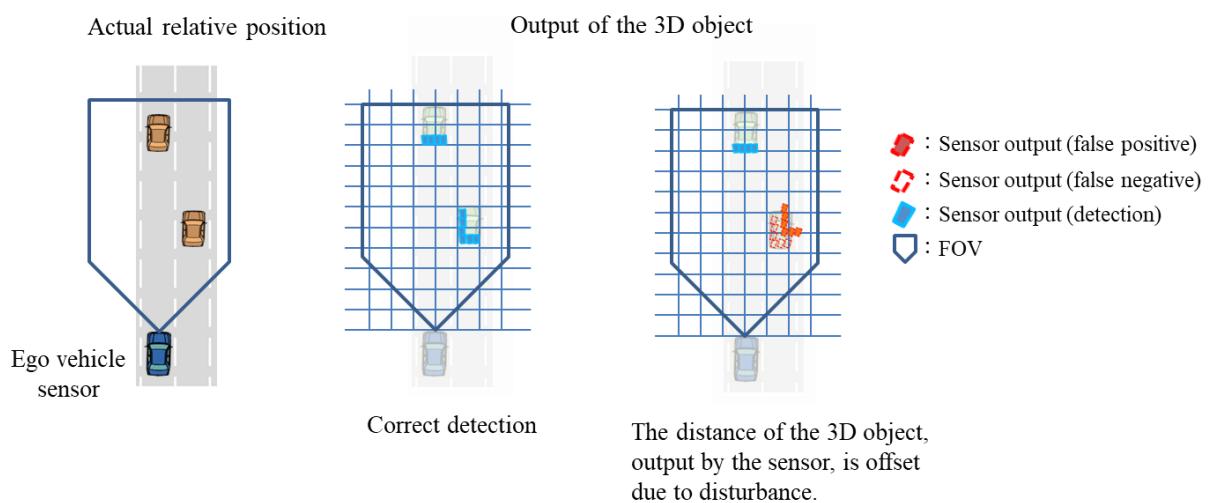
Explaining the Phenomenon

The target cannot be detected as a three-dimensional object (false negative).



Explaining the Phenomenon

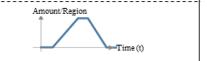
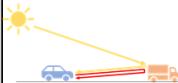
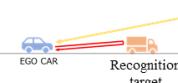
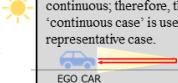
The target is detected in a position which is not the true position (false positive).



Noise

Explaining the Phenomenon

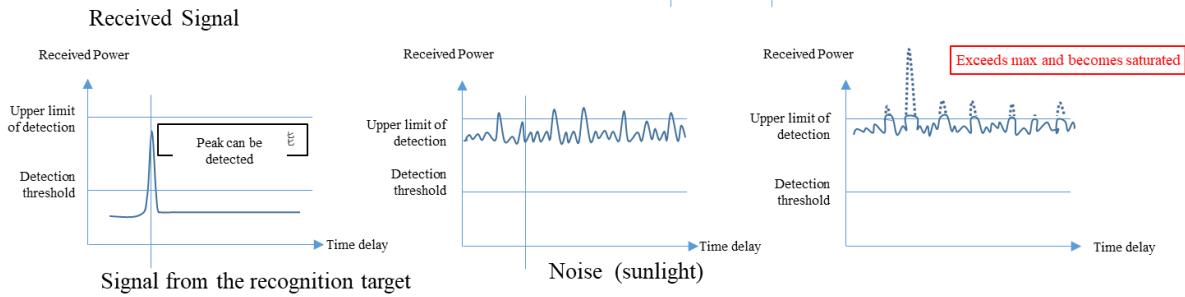
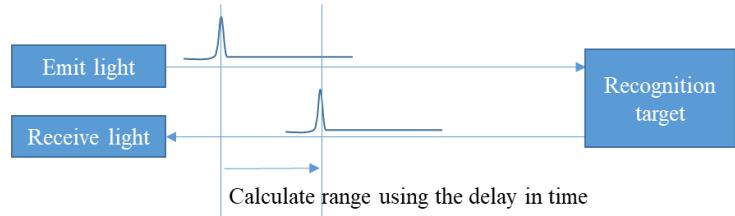
The phenomenon occurs when the reflection points attached to a recognition target, are continuously not output to the point cloud.

			A		B		C	
			1	2	1	2	1	2
Phenomenon Parameter	Degree	Amount of noise	Reflection points which should exist are not output to the point cloud due to noise					
		Region of noise	Full area within frame	Attached to the recognition target	Attached to the frame			
	Time	Duration of attenuation	Continuous	Temporary	Continuous	Temporary	Continuous	Temporary
								
Phenomenon Mode			The scenario whereby the sun reflects off the preceding vehicle (recognition target), creating continual noise. 	← This malfunction is more severe if 'continuous' rather than 'temporary'; therefore, the 'continuous case' is used as the representative case.	The scenario whereby the light from the sun directly enters, creating continual noise. 	← It is possible that the lights or LiDAR from the oncoming vehicle can enter and become noise. However this is temporary and the error is more severe when continuous; therefore, the 'continuous case' is used as the representative case. 		

Noise

E.3.2.1.2 Outline of the Principle

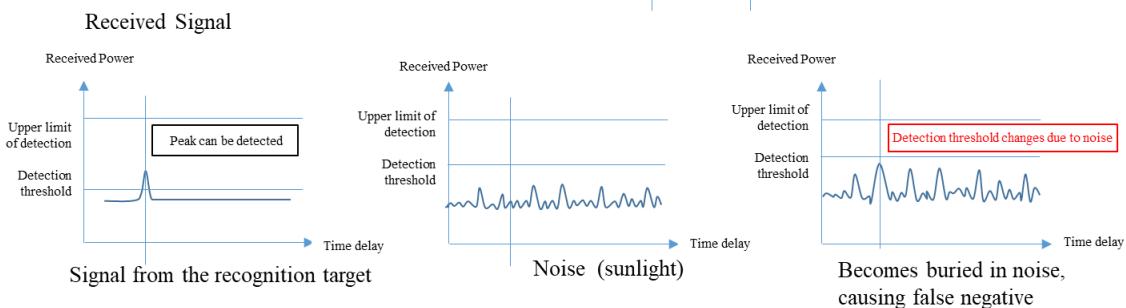
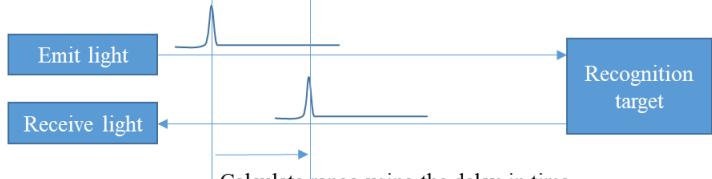
The peak of the received signal is detected, and the range is calculated to determine the delay in time.



If an infrared light such as sunlight, which occurs routinely, enters the light receiver as 'noise', this noise and the reflection from the recognition target as a total becomes saturated, preventing proper detection.

If a powerful light that occurs routinely, such as sunlight, enters the signal receiver, this causes saturation and ultimately malfunction.

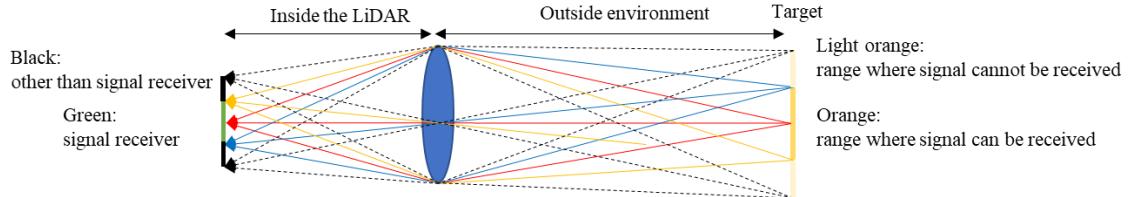
The peak of the received signal is detected, and the range is calculated to determine the delay in time.



If an infrared light such as sunlight, which occurs routinely, enters the light receiver as ‘noise’, and the reflection from the recognition target (which has weak reflection) becomes mixed with the noise, this can prevent detection.

If a powerful light that occurs routinely, such as sunlight, enters the signal receiver, this can cause malfunction.

Noise Due to Background Light



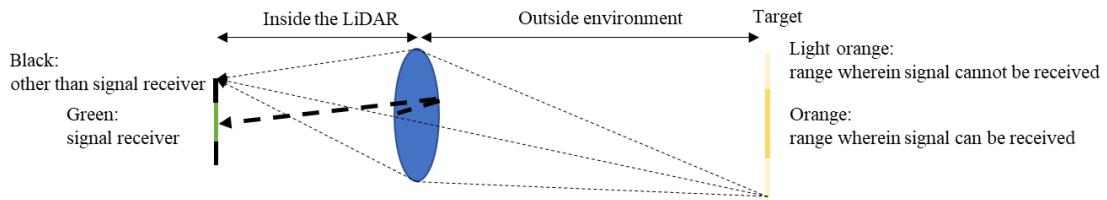
The lights that pass through the lens from the range where signals can be received include ① scattered light from the one that LiDAR sent, ② scattered lights from other lights and ③ self-emitting lights within the range. These cannot be distinguished from each other, and thus they will all pass through the same optical path, and all be received as signals.*

All lights other than ① become noise components.

*As the wavelength sent by the LiDAR itself is recognized, it will normally have a filter that cuts out light from any other wavelength. Lights that become noise components are what fall within the wavelength range used for LiDAR transmission.

Noise

Noise Due to Ghost



Lights that fall outside of the range wherein signals can be received will travel along a normal light path to somewhere other than the signal receiver; thus, they will not be received. However, there are times when lights from outside of the signal receiving range may be received due to internal reflection, etc. (thick line).

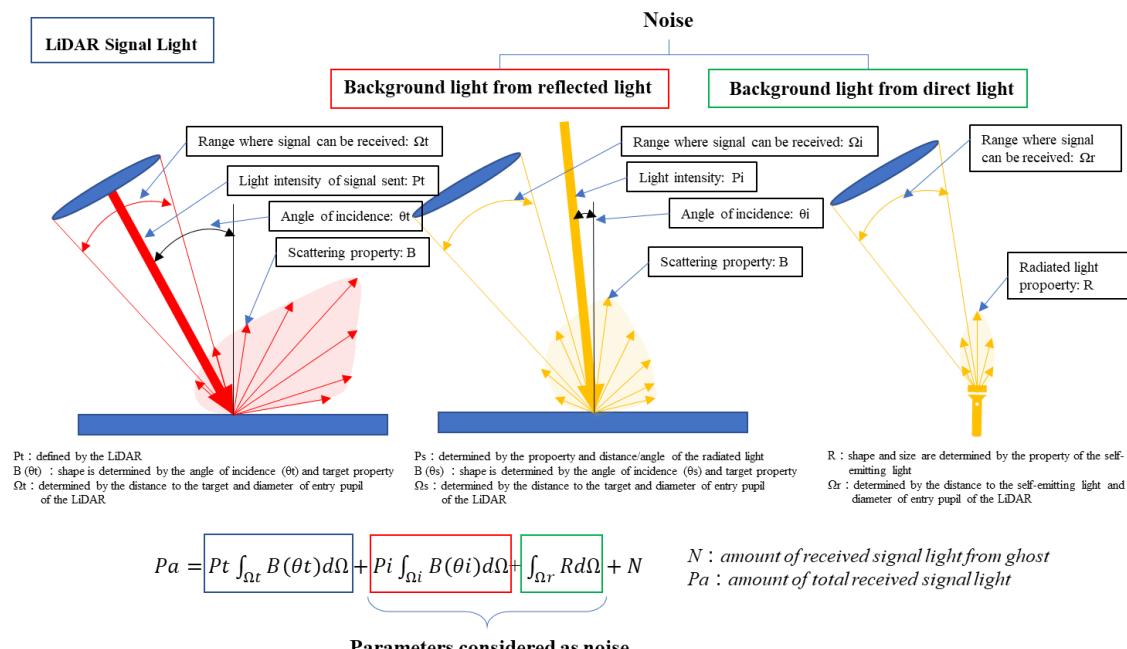
Normally, antireflection film, etc., would be used to suppress internal reflection; therefore, this might occur when there is a strong incident light (such as sunlight, headlamps, signals from other manufacturers' LiDAR, etc.)

Noise caused by internal reflection has been deemed low priority; thus, it will not be dealt with here.

E.3.2.1.3 Principle Model

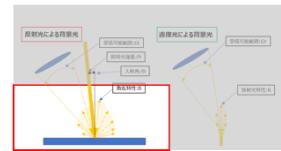
Noise

When considering as 'noise', we suppose background light arising from reflected light and direct light.



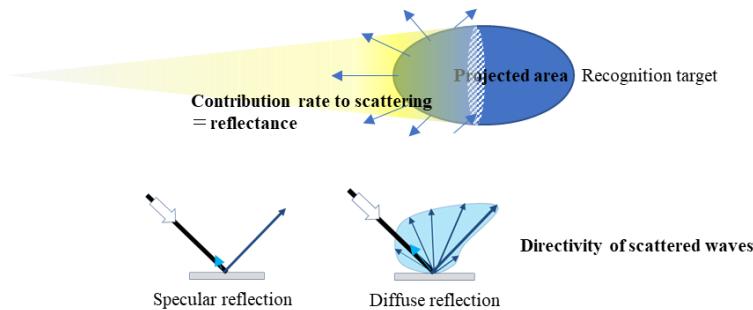
Noise

Scattering property



The reflection intensity of the target, the same as for the Millimeter wave, is believed to be expressed as the product of the projected area of the target, the contribution rate to scattering and the directivity of scattered waves, and an object with the same surface material will have a stronger reflection in the directivity (i.e., the points facing the LiDAR).

$$\text{Reflection intensity} = [\text{Projected area of the target}] * [\text{Contribution rate to scattering}] * [\text{Directivity of scattered waves}]$$



Derive each causal factor of disturbance, from the principle of the target's reflection.

Noise

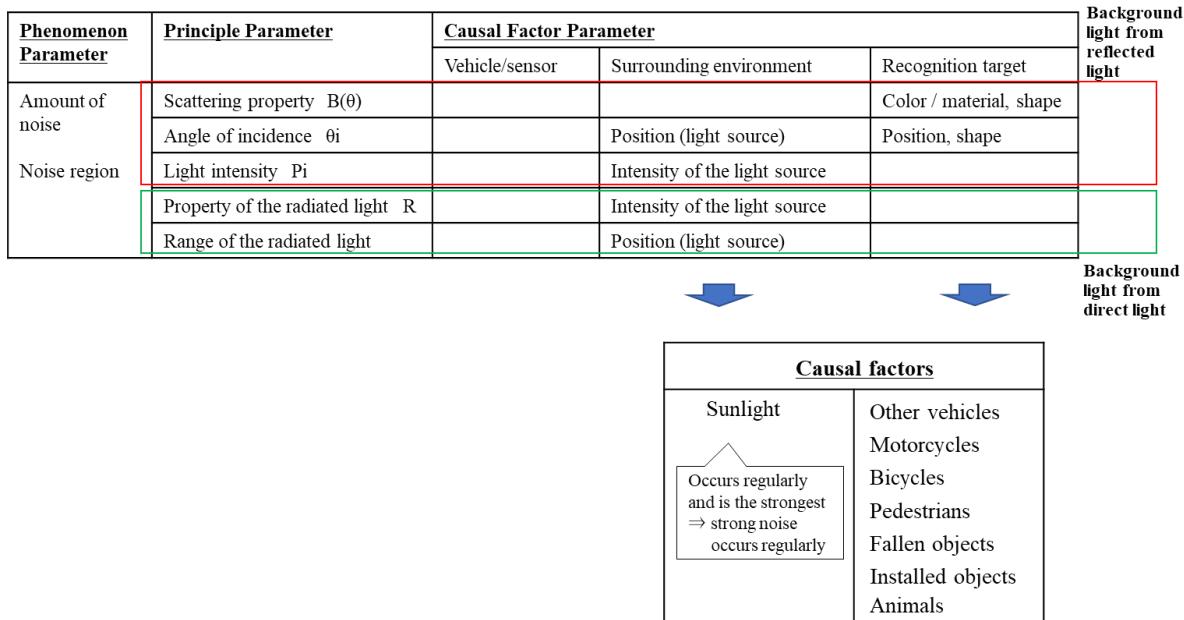
Principle	Causal factors
Scattering property Contribution rate to scattering Difference in the strength of reflection of targets 	The factors that instigate a change in the reflection intensity include different colors and materials used for painted surfaces, clothing, etc. Picture showing a change in reflection due to different coating used.
Direction of scattered waves Projected area of the target 	The factors that instigate a change in the strength of reflection when the angle of incidence/reflection changes, include different colors and materials used for painted surfaces, clothing, etc. Picture showing a change in the rate of reflection caused by a difference in angle.
Difference in the reflection intensity due to the difference in the normal vector. 	The factors that instigate a change in the reflection intensity when the angle of incidence/reflection changes, include different colors and materials used for painted surfaces, clothing, etc. Picture showing a change in the reflectance caused by a difference in shape.
Light intensity Properties of the radiated light Angle of incidence Range of the radiated light	Intensity of the light source (cd/m^2)

E.3.2.2 The Relationship Between Principle and Causal Factors of Perception Disturbances

Noise

E.3.2.2.1 Causal Factors based on Principle

Derive causal factors from the causal factor parameters.



E.3.2.2.2 Parameter Range

Noise

The below is a list summarizing the parameter ranges.

Causal factors			Causal factor parameter	Range	Explanation (or reason)
Environment	Light source	Sunlight	Altitude	0 to 90 deg	With the horizon being 0 deg, and the sky being 90 deg
			Azimuth	0 to 359 deg	True North is 0 deg, and going clockwise East is 90 deg, South is 180 deg and West is 270 deg
			Brightness	0 lx to 100,000 lx	Koyomi handbook (http://photon.sci-museum.kita.osaka.jp/publish/text/koyomi/66.html) Brightness of the sun in the middle of summer

E.3.2.2.3 Evaluation Scenario

Noise

Scenario F-1

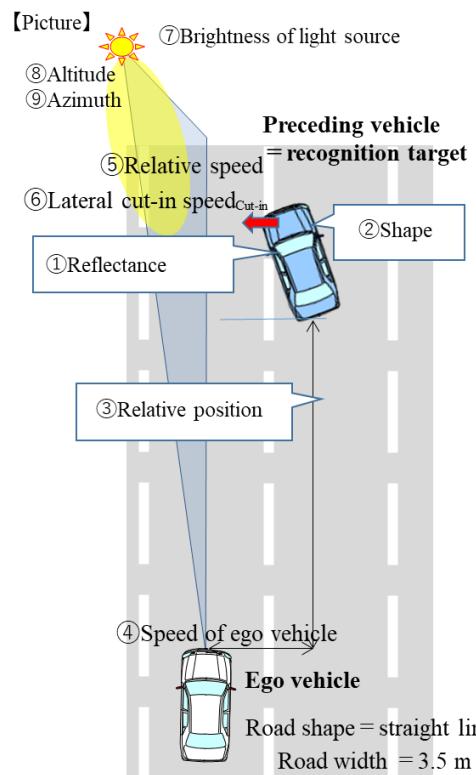
Evaluate based on “a vehicle cuts-in on a straight road” scenario.

【Outline】

- This evaluation looks at the impact of changes to the shape and reflectance of the cut-in vehicle (recognition target).
- Adjust the position of the light source (sun) so that it is in the direct path (on the straight line) of the recognition target.

【Parameters】

Causal factor parameter Parameters required for evaluation	Recognition target	①Reflectance	Coating material = black, specular surface
		②Shape	Vehicle = e.g.) Defence force truck, lowboy, trailer, motorcycle, light duty truck, light sports car
		③Relative position	This is defined under the traffic flow scenario, thus is not determined here.
	Sun	⑦Brightness of light source (lux)	0 to 100,000 lux
		⑧Altitude	0 to 90 deg
		⑨Azimuth	0 to 359 deg
	④Speed of ego vehicle ⑤Relative speed ⑥Lateral cut-in speed	④Speed of ego vehicle	This is defined under the traffic flow scenario, thus is not determined here.
		⑤Relative speed	
		⑥Lateral cut-in speed	



Scenario F-2

Noise

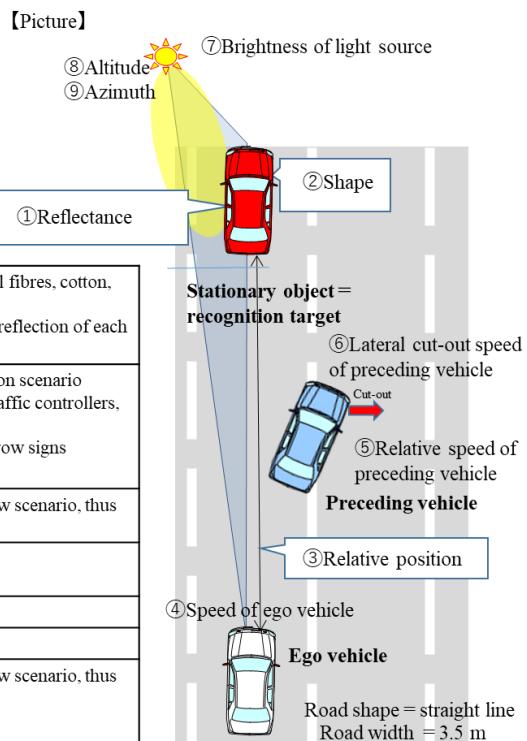
Evaluate based on “a vehicle cuts-out on a straight road” scenario.

【Outline】

- This evaluation looks at the impact of changes to the shape and reflectance of the stationary object in front of the preceding vehicle which cuts-out (recognition target).
- Adjust the position of the light source (sun) so that it is in the direct path (on the straight line) of the recognition target

【Parameters】

Causal factor parameter Parameters required for evaluation	Recognition target	①Reflectance	Person: clothing = leather, chemical fibres, cotton, reflective material Mounted/fallen objects: the rate of reflection of each object
		②Shape	Vehicle = evaluate using deceleration scenario Person = standing, sitting, lying, traffic controllers, bicycles Mounted objects = safety cones, arrow signs Fallen objects = tires, wood
		③Relative position	This is defined under the traffic flow scenario, thus is not determined here.
	Sun	⑦Brightness of light source	0 to 100,000 lux
		⑧Altitude	0 to 90 deg
		⑨Azimuth	0 to 359 deg
	④Speed of ego vehicle ⑤Relative speed of preceding vehicle ⑥Lateral cut-out speed of preceding vehicle	④Speed of ego vehicle	This is defined under the traffic flow scenario, thus is not determined here.
		⑤Relative speed of preceding vehicle	
		⑥Lateral cut-out speed of preceding vehicle	



Scenario F-3

Noise

Evaluate based on “a vehicle decelerates on a straight road” scenario

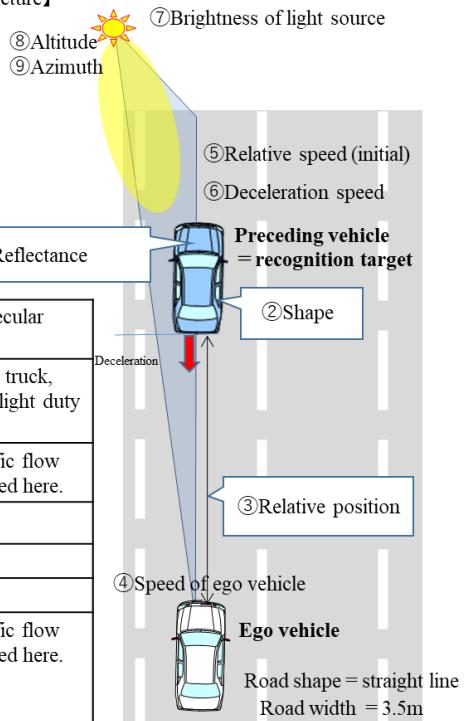
【Outline】

- This evaluation looks at the impact of changes to the shape and reflectance of the decelerating vehicle (recognition target).
- Adjust the position of the light source (sun) so that it is in the direct path (on the straight line) of the recognition target

【Parameters】

Causal factor parameter Parameters required for evaluation	Recognition target	①Reflectance	Coating material = black, specular surface
		②Shape	Vehicle = e.g.) Defence force truck, lowboy, trailer, motorcycles, light duty truck, light sports car
		③Relative position	This is defined under the traffic flow scenario, thus is not determined here.
	Sun	⑦Brightness of light source	0 to 100,000 lux
		⑧Altitude	0 to 90 deg
		⑨Azimuth	0 to 359 deg
	④Speed of ego vehicle ⑤Relative speed (initial) ⑥Deceleration speed	④Speed of ego vehicle	This is defined under the traffic flow scenario, thus is not determined here.
		⑤Relative speed (initial)	
		⑥Deceleration speed	

【Picture】



E.4 The principle models and evaluation scenarios of Camera

As examples for Camera, following 3 of principle models and evaluation scenarios of perception disturbances are described.

- Shielding (image cut off)
- Low spatial frequency / Low contrast (caused by spatial obstruction)
- Excessive (saturation), Whiteout

E.4.1 [Camera] Shielding (image cut off)

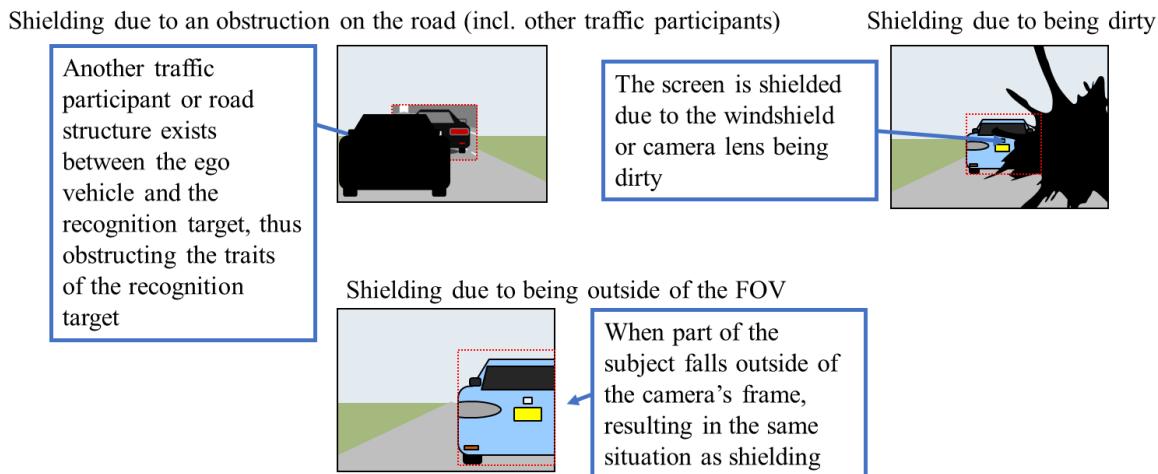
E.4.1.1 The Phenomenon and Principle

Shielding
(Image Cut Off)

E.4.1.1.1 The Phenomenon

The recognition target is partially or fully cut off due to shielding by an object or due to moving out of the FOV, leading to a loss of information required for extracting features. It leads False Negative or position error.

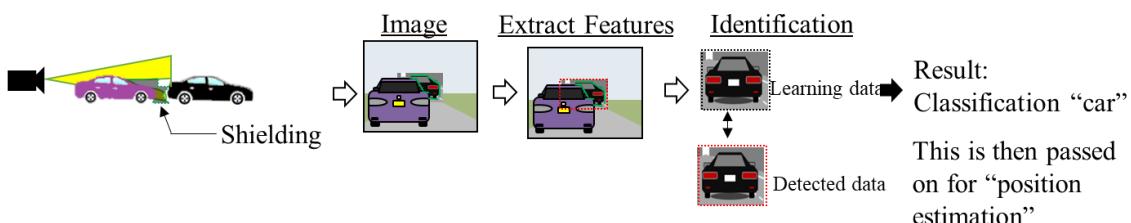
Example



E.4.1.1.2 Outline of Principle

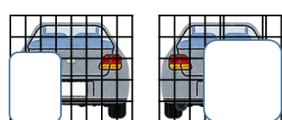
Shielding
(Image Cut Off)

When the recognition target is partially shielded, the camera's recognition function may not be able to properly extract features. Even if features can be extracted, the identification function will not be able to match the learning data, resulting in an error in recognition (non-detection or incorrect classification).

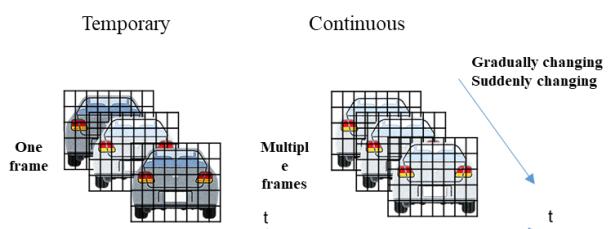


Region & Amount

The position and the number of pixels in which shielding is occurring within the frame of the recognition target

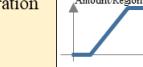
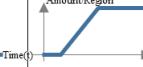


Time



Shielding
(Image Cut Off)

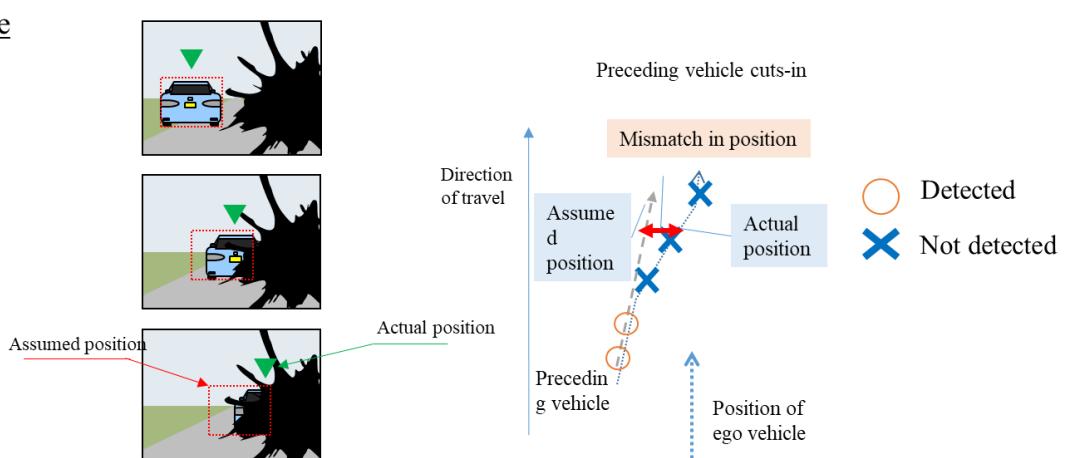
Disturbances based on Principle

Phenomenon Mode		A		B		C	
		1	2	1	2	1	2
Degree	Amount	Defined by the principle for each sensor					
	Region	Full area within frame 		Attached to the recognition target 		Attached to the frame 	
Time	Amount of change per unit of time	Range of change defined by the principle/causal factor					
	Duration	Continuous 		Temporary 		Continuous 	
Disturbance and Evaluation Scenario		The full area of the frame is continuously shielded, and is not removed		The full area of the frame is continuously shielded, but is then removed by the wiper blades, etc.		A foreign object has adhered to the surface of the recognition target	
		The change point between the state of being shielded and the state of not being shielded		The frame is partially dirty, and this is not removed		The frame is partially dirty, but this is then removed by the wiper blades, etc.	

Shielding
(Image Cut Off)

If correct feature extraction is not achieved, the size, orientation, and position of the object cannot be detected correctly. In addition, if the orientation and position are not detected correctly, errors will occur in tracking, causing recognition errors in estimated position and velocity.

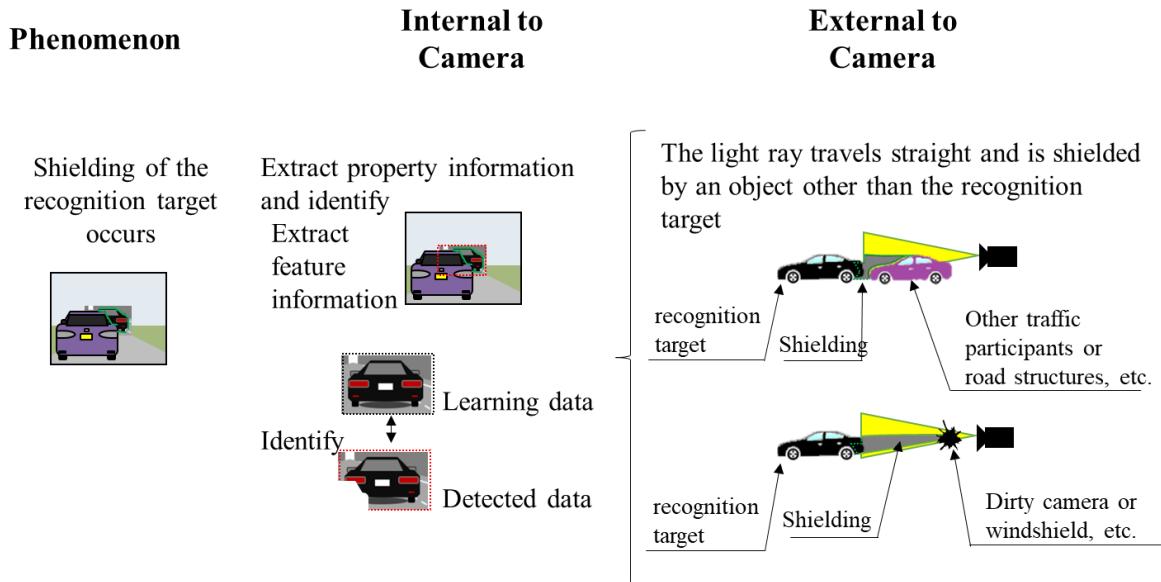
Example



E.4.1.1.3 Principle Model

Shielding
(Image Cut Off)

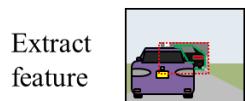
The relationship between internal and external models



Shielding
(Image Cut Off)

Internal to Camera

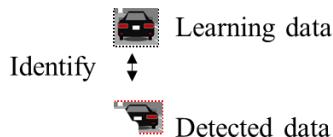
Several techniques exist for extracting trait information and for identification, and therefore cannot be specified. However, here we provide examples of some classic techniques



【Detect Shape】
The differentiation operator is approximated to extract the trait points, corner points, and edges, and then unique analysis, extreme value search, etc. are conducted.
E.g. edge detection, corner detection, blob detection, etc.

【Detect Figure】
Straight line detection, curve detection (Hough transform)

【Detect Region】
Divides the area of the image/cuts out the area of the target and distinguishes it from the remaining area.

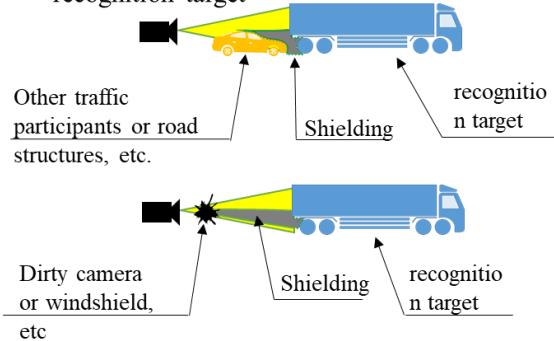


The process for screening points which are highly similar to the learning data.
E.g. template matching, detection based on color, detection using edges, matching of trait information.

Shielding
(Image Cut Off)

External to Camera

The light ray travels straight and is shielded by an object other than the recognition target



<Model directly related to recognition error >

- Use a model where light emitted from the object (including reflected light) travels straight through a constant medium.

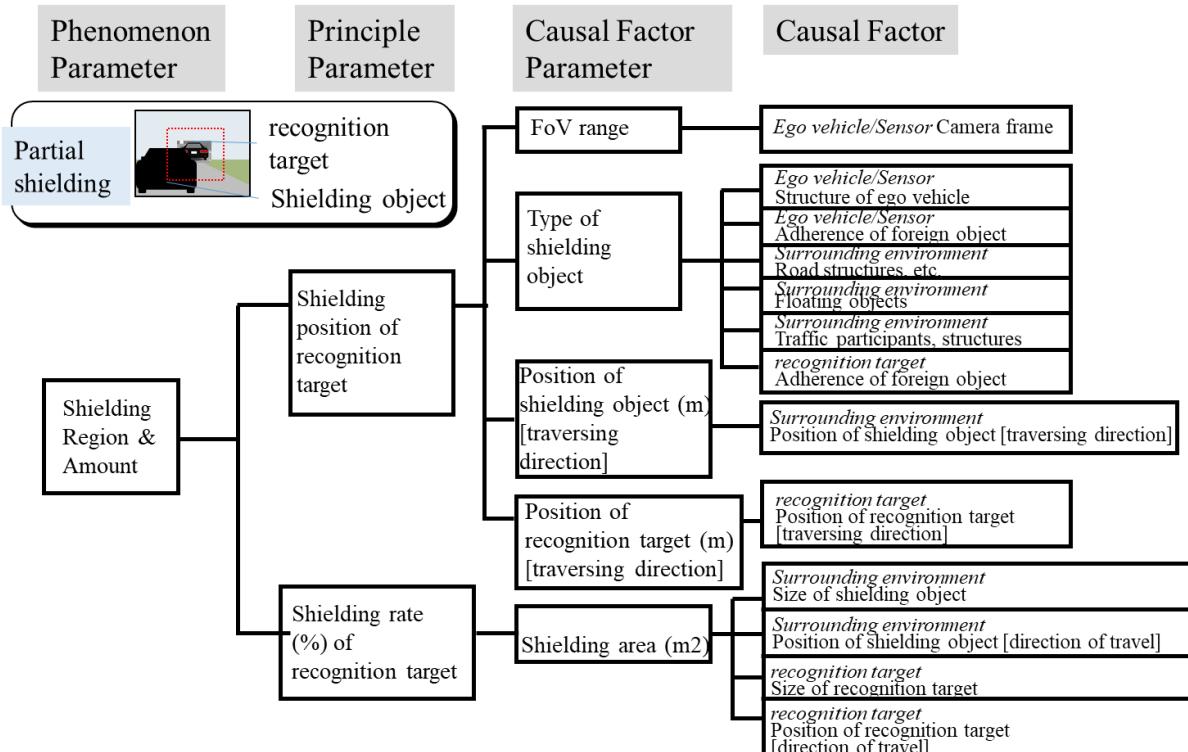
<Model related to recognition error, however impact is minimal (for reference) >

- Refraction occurs on the border of a medium, such as a glass surface or rainwater, etc.
- A phenomenon known for electromagnetic waves (incl. visible light). Strictly speaking there is diffraction, however, based on the degree of impact it is not considered as a problem causing shielding error

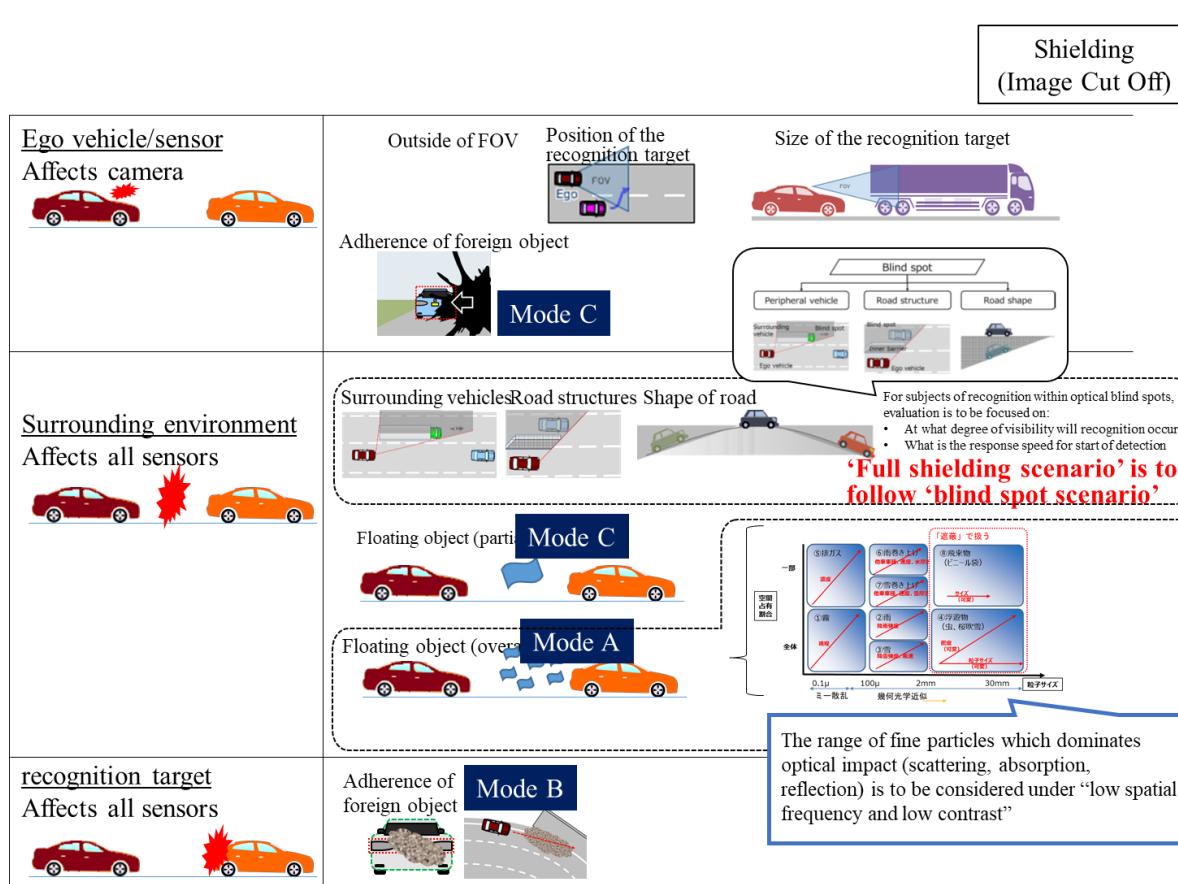
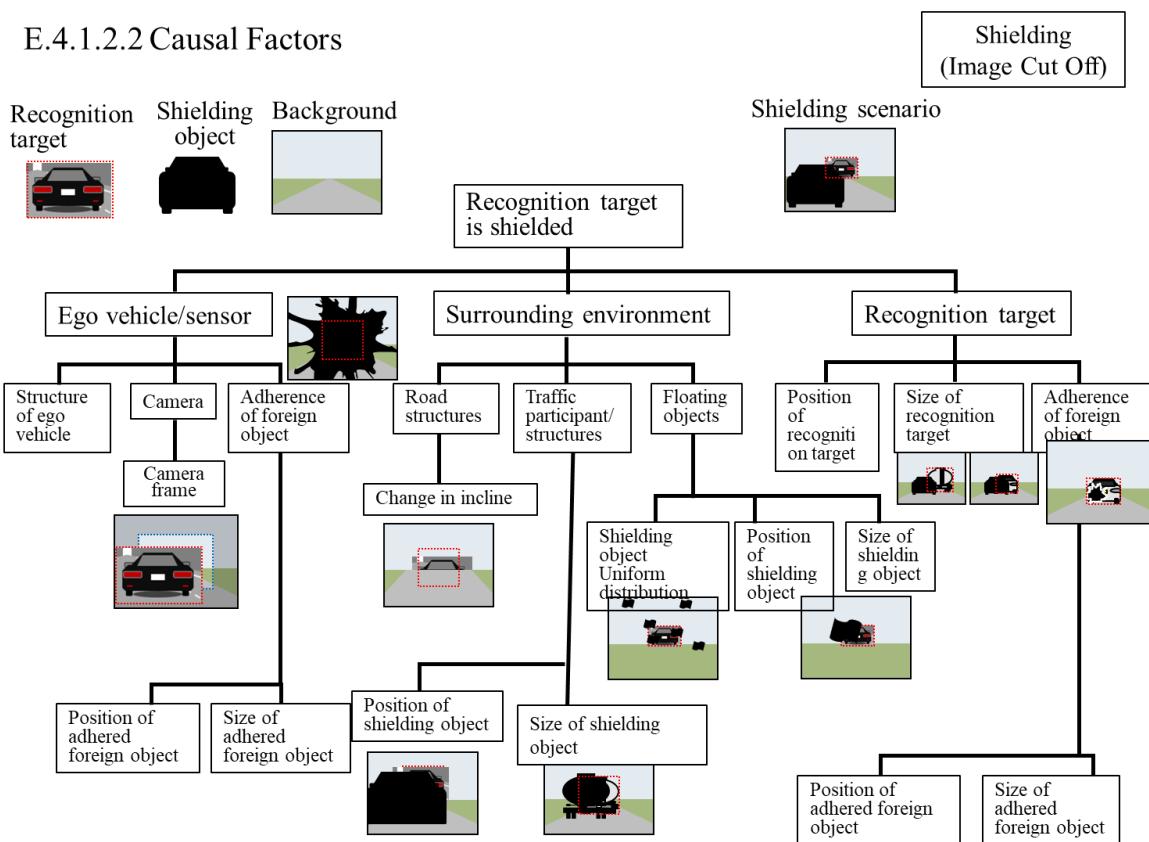
E.4.1.2 Causal Factors based on Principle

Shielding
(Image Cut Off)

E.4.1.2.1 The Relationship Between Principle and Causal Factors



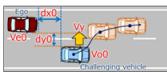
E.4.1.2.2 Causal Factors



Shielding



Shielding
(Image Cut Off)

Functional Scenario	ALKS Scenario	Lane			Traffic Information	Moving Object	Obstruction on Road	Environment
		Lane markings	Structures	Edge of road				
F-1	Cut-in		○		○ ○ ○		○	
F-2	Cut-out		○		○ ○ ○ ○ ○ ○		○	
F-3	Deceleration		○		○ ○ ○		○	
F-B1-14	Lane-keeping		○ ○ ○ ○				○ ○	
F-4	Blind-spot (Vertical)		○		○ ○		○ ○	

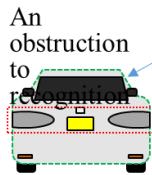
E.4.1.2.3 Parameter Range

Shielding
(Image Cut Off)

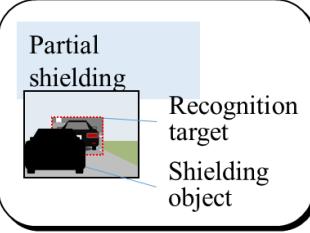
Phenomenon parameter	Principle Parameter	Causal Factor Parameter	Causal Factor	Parameter Range	Conditions	
					STEP 1	STEP 2
Amount / Region	Shielding position of the recognition target	FoV range	Ego vehicle/sensor Camera frame	Depends on camera used by test subject		
		Type of shielding object	Ego vehicle/sensor Structure of ego vehicle	Wiper blades, bonnet		
			Ego vehicle/sensor Adherence of foreign object	Dirty windshield 0 to 100%	Dirty but wiped off with wiper blades	Edges of the image are dirty
			Surrounding environment Floating objects	Uniform distribution Single		
			Surrounding environment Road structures	Road's vertical incline 6%		
			Surrounding environment Traffic participants, structures	Traffic participant : vehicle Structure : side wall		
		Position of shielding object (m) [traversing direction]	Surrounding environment Position of shielding object [traversing direction]	Ratio of wrapping 0 to 100%	Rate of shielding of recognition target approx. 25%, position in traversing direction	Rate of shielding of recognition target approx. 50%, position in traversing direction
		Position of recognition target (m) [traversing direction]	recognition target Adherence of foreign object on recognition target	Dirt		
			recognition target Position of recognition target [traversing direction]	Position relative to the shielding object		
		Rate of shielding (%) of the recognition target	Shielding area (m2)	Size of two-wheeled motor vehicles to large-sized truck	Shielding by a light vehicle	Shielding by a large-sized truck

The Shielding Position

Shielding
(Image Cut Off)



1. An obstruction to recognizing the profile of the subject. The difference in contrast to the background disappears. The profile is shielded
2. An obstruction to recognizing the features of the subject. The traits identifying the recognition target are shielded (in the case on the left this would be the tail lamps and the license plate)



Impact according to the type of shielding object

An obstruction to recognizing the profile of the subject

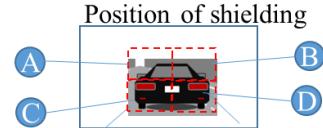
	None	Obstructed
None		
Obstructed		

Mandatory evaluation scenario

Evaluation of C D

which shields features is a must

Note) Including shielding of C&D at the same time

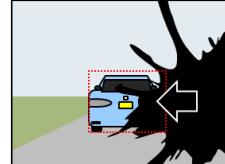
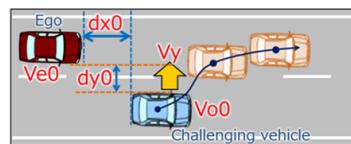


E.4.1.2.4 Evaluation Scenario

Shielding
(Image Cut Off)

E.4.1.2.4.1 Cut-in

The object to be recognized enters the front of the own lane at a constant lateral speed from a position where the field of view is restricted by an attached object.

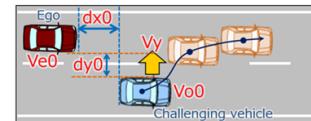
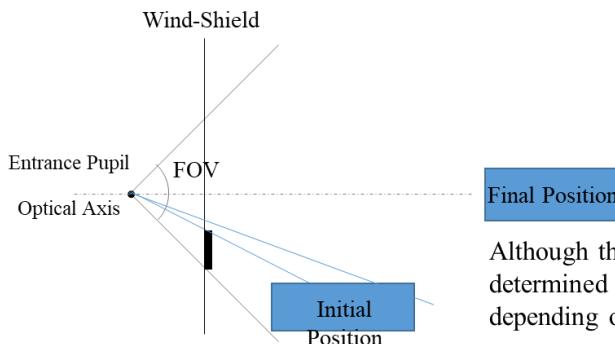


Parameter	Variable/Fixed	Range	Explanation
Distance to the target	Variable	Longitudinal position dx0 [m]	Cut-in distance at the slowest velocity to the maximum sensing distance of the sensor.
		Lateral position y0 : 3.5m	
Relative velocity to the target	Variable	Longitudinal velocity Vo0-Ve0 [kph] Lateral velocity Vy [kph]	Fastest and slowest velocities with respect to the preventable criteria in ALKS.
Type of the target	Fixed	Shape: sedan Color: White	Since the scenario is specified by the shielding ratio, it does not depend on the size and shape of the object part. Select a standard object.
Degree of shielding of the detection-target due to adherence of foreign object	Variable	In relation to the bounding box of the detection-target ① Initial50% → Final0% ② Initial100% → Final50%*	*In case of "initial 100%", the final value depends on the scenario (as a rule of thumb, determine the size of the shielding to be close to 50%)

Supplementary of Cut-in Scenario

Shielding
(Image Cut Off)

In a cut-in scenario, consider a situation where a target that is partially or fully shielded becomes an ACC target by entering your lane.



Whether or not the position and distance after the cut-in is correctly output will determine whether or not it is safely controlled.

Although the shielding rate of the final position is determined by the scenario, it has a certain range depending on the placement of the shielding.

Since the distance between the entrance pupil and the windshield is constant, once the initial target position and the shielding rate are determined, the limits on the size of the shielding are determined. Since the position of the shielding object is arbitrary, the size is not uniquely determined (it can be larger than the viewing angle to the target).

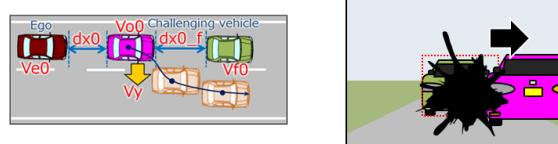


It is desirable to set the size and position as close as possible to the final shielding ratio while adhering to the initial shielding ratio.

E.4.1.2.4.2 Cut-out

Shielding
(Image Cut Off)

The recognition target cuts out from the position where it is shielded. The recognition target in the foreground was partially visible, while the recognition target in the background, which exists farther away, is more shielded.

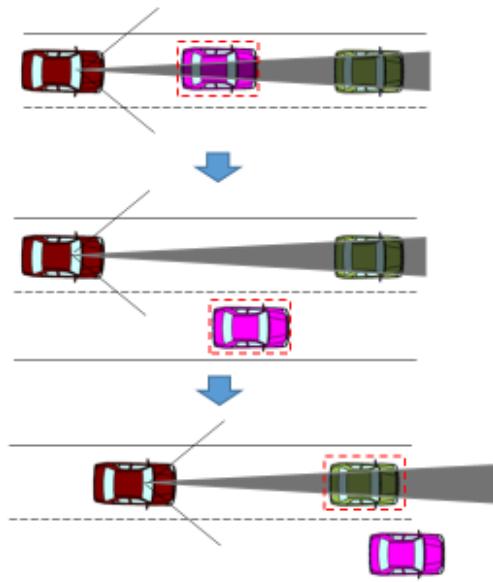


Parameter	Variable/Fixed	Range	Explanation
Distance to the target	Variable	Longitudinal position dx_0 [m]	Cut-out distance at the slowest speed to the maximum sensing distance of the sensor.
		Longitudinal position dx_0_f [m]	
Relative velocity to the target	Variable	Longitudinal velocity $V_{o0}-V_{e0}$ [kph]	The fastest and slowest speeds against the preventable criteria in the cut-out scenario.
		Longitudinal velocity $V_{o0}-V_{f0}$ [kph]	
		Lateral velocity V_y [kph]	
Type of the target	Fixed	Shape: sedan Color: White	Since the scenario is specified by the shielding ratio, it does not depend on the size and shape of the object part. A standard object is selected.
Degree of shielding of the detection-target due to adherence of foreign object	Variable	In relation to the bounding box of the detection-target ① Initial50% → Final0%	The shielding ratio should be set for the vehicle ahead.

Supplementary of Cut-out Scenario

Shielding
(Image Cut Off)

In the cut-out scenario, when the vehicle in front that is partially shielded moves to the adjacent lane and the previous vehicle becomes the ACC target, it is evaluated that the partially shielded condition does not lead to a dangerous event.

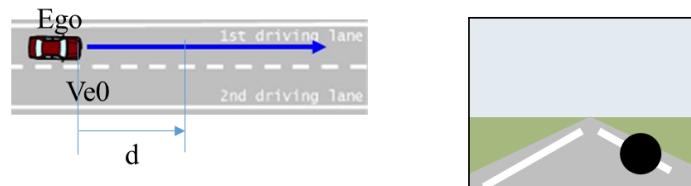


If the vehicle ahead is completely occluded, camera recognition is not possible, so this is another case such as the occlusion detection function. Since the problem here is the error in the output result of the position and distance to the recognition target due to the occlusion, the complete occlusion of the vehicle ahead is not included.

E.4.1.2.4.3 Lane-Keep

Shielding
(Image Cut Off)

Drive at a constant speed along your lane in a shielded situation.

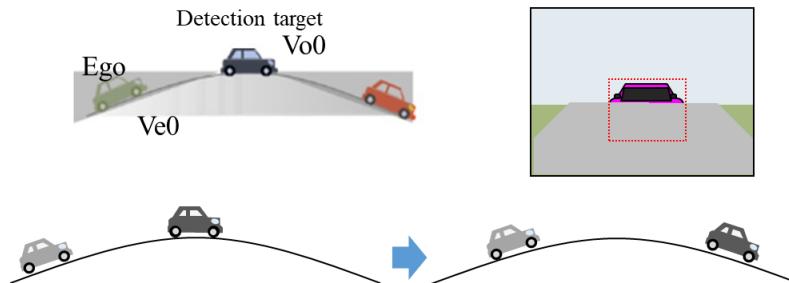


Parameter	Variable/Fixed	Range	Explanation
Velocity of ego vehicle	Fixed	Ve0: 120 kph	The maximum speed limit for the high way in Japan
Width of driving lane	Fixed	3.5m	Typical lane width of the high way in Japan
Curvature of lane	Fixed	R380	
Type of the target	Variable	Shape: solid line, dotted line Color: white, yellow	
Amount which the ego vehicle's driving lane marking lines are shielded due to the adherence of a foreign object (disturbance)	Fixed	Degree of shielding : 50%	
Longitudinal position according to the center of the adherence of a foreign object	Fixed	d: 20m/ 60m/ 100m	

E.4.1.2.4.4 Blind-spot (vertical)

Shielding
(Image Cut Off)

While driving on a sloped road surface (convex shape), approaching the recognition target in front of own lane at a constant speed.



Parameter	Variable/Fixed	Range	Explanation
Distance to the target	Variable	Longitudinal position dx0 [m]	From the limit where the ground surface of the recognition target is visible to the limit where the top of the recognition target is visible.
Relative velocity to the target	Fixed	Longitudinal velocity Vo0-Ve0 [kph]	Follow the traffic flow scenario to be combined.
Type of the target	Fixed	Shape: sedan Color: white	It does not depend on the size and shape of the object because the scenario is defined by the shielding rate. Select a standard object.
Road structure vertical incline	Fixed	Vertical cross sectional incline: 6%	The most severe value with reference to the Road Structure Ordinance.

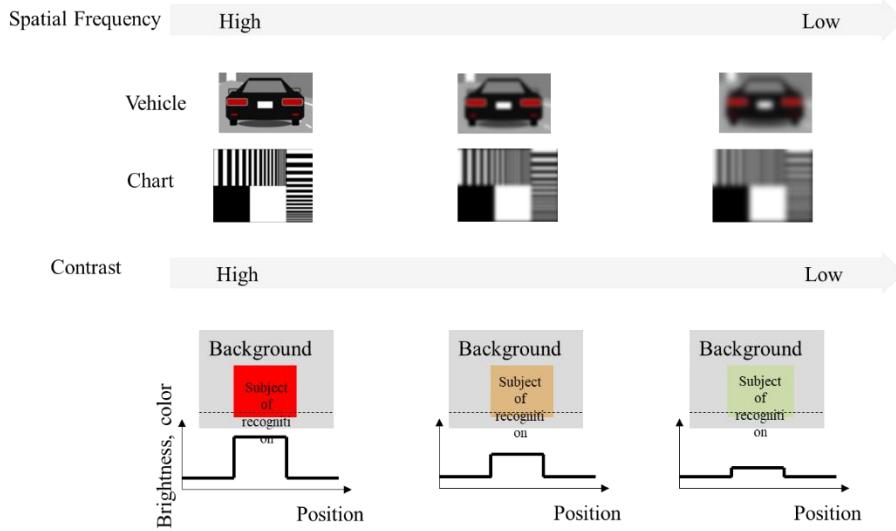
E.4.2 [Camera] Low spatial frequency / Low contrast (caused by spatial obstruction)

E.4.2.1 The phenomenon and principle

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

E.4.2.1.1 The Phenomenon

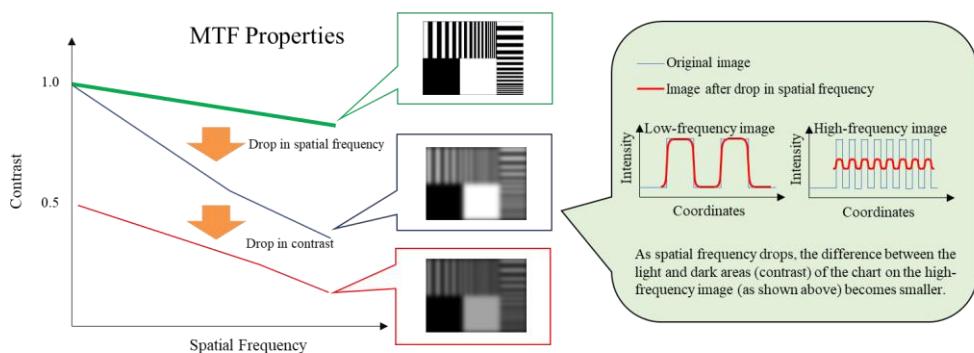
Rain, snow, and fog cause blurring of the contours of objects (decrease in spatial frequency).
At the same time, the contrast of the image is reduced.



The degree of drop in spatial frequency and contrast of the image can be expressed with MTF.

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

A drop in spatial frequency can be expressed as a drop in the high frequency MTF (equivalent to blurring). Contrast is the difference in brightness or chromaticity between the object and the background, and the overall contrast reduction in the image is expressed by the MTF reduction in all frequency bands.



MTF (Modulation Transfer Function)

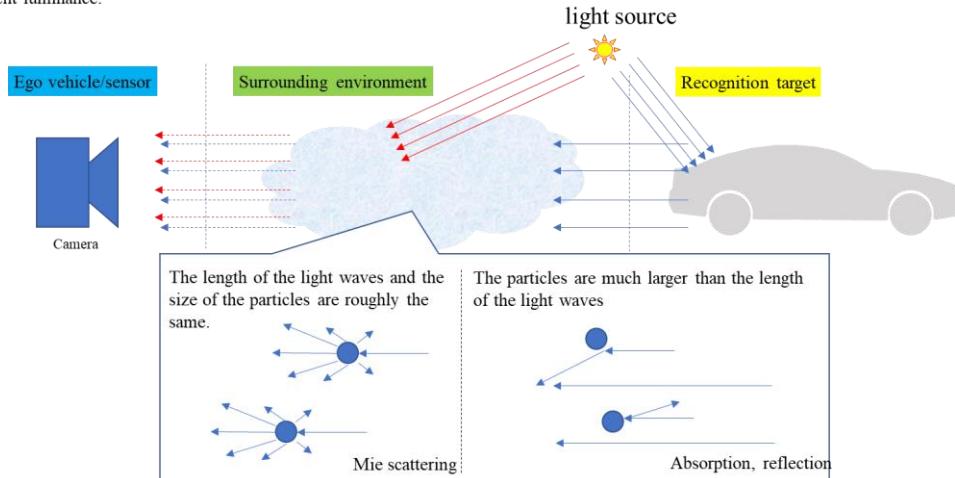
MTF is the amplitude ratio of the frequency and the input/output waves, when sine waves are input into the system. It is the value representing the degree of gathering of light from a certain area of the object, at the corresponding position in the image. It allows to quantitatively test the performance of the lens, and allows testing of the image formation and contrast at the same time.

Low Spatial Frequency
 Low Contrast
 (Caused by Spatial Obstruction)

E.4.2.1.2 The outline of principle

If there is an obstruction within the space of concern then the lights reflected from the recognition target can hit that obstruction (particles) within the space, causing scattering, absorption and reflection, resulting in attenuation prior to reaching the camera. (The degree of scattering, absorption and reflection, will depend on the size and concentration of the particles).

The luminance scattered by obstacles in the space due to the direct luminance from the light source is added to the attenuated luminance to become the camera incident luminance.



Phenomenon Parameter

Low Spatial Frequency
 Low Contrast
 (Caused by Spatial Obstruction)

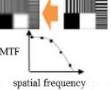
Amount	Spatial frequency 
Contrast	 Difference in brightness & color between the target and the background
Region	Covers the whole frame or is only localized and accompanies the recognition target.
Amount of change over time	Spatial frequency and contrast drop over time (gradually or suddenly)
Duration	Temporary or continuous



Exhibit: Pxhere.com: CC0 License

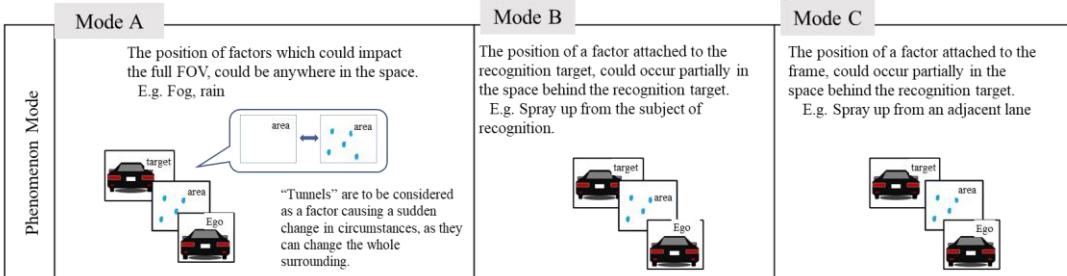
How it Affects Recognition/Controls

These can become a factor for error or prevent detection of the recognition target, when converting the image coordinates into positional information in the 3D space.

Classified into three modes according to the range of occurrence within the angle of view.

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

Phenomenon Parameter	Mode	A				B				C			
		1	2	3	4	1	2	3	4	1	2	3	4
Degree	Amount	Full area within frame				Attached to the recognition target				Attached to the frame			
	Region												
Time	Amount of time per unit of time	Gradually changes		Suddenly changes		Gradually changes		Suddenly changes		Gradually changes		Suddenly changes	
	Duration	Continuous	Temporary	Continuous	Temporary	Continuous	Temporary	Continuous	Temporary	Continuous	Temporary	Continuous	Temporary

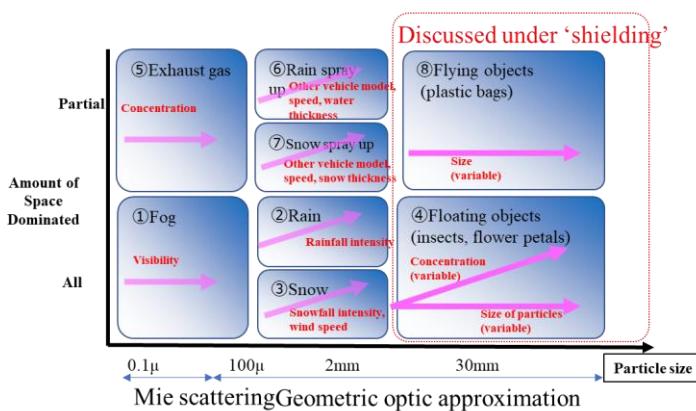


Visibility drops due to spatial obstructions in accordance with the principle explained thus far.

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

There are various types of spatial obstructions. Here we have categorized them by the size of their particles according to previously mentioned principle. This is further categorized into "region", which is dealt with in the phenomenon mode (i.e. all or partial).

(Of these categories, ④ floating objects and ⑧ flying objects are looked at under the error mode "shielding").

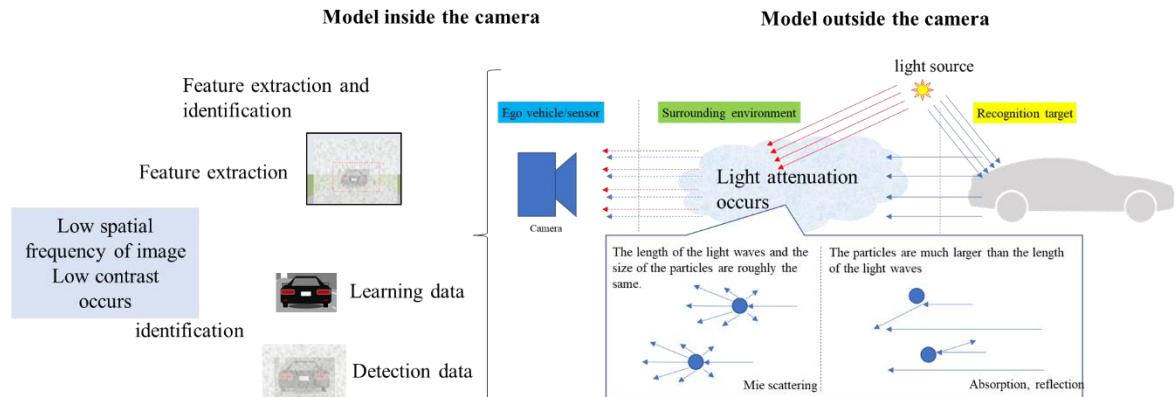


E.4.2.1.3 The Principle Model

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

Relationship between internal and external models

phenomenon



The wavelength of the light depends on the spectral characteristics of the light source.

Internal to Camera

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

Several techniques exist for extracting trait information and for identification, and therefore cannot be specified. However, here we provide examples of some classic techniques.

Extract trait information



【Detect Shape】

The differentiation operator is approximated to extract the trait points, corner points, and edges, and then unique analysis, extreme value search, etc. are conducted.
E.g. Edge detection, corner detection, blob detection, etc.

【Detect Figure】

Straight line detection, curve detection (Hough transform)

【Detect Region】

Divides the area of the image/cuts out the area of the target and distinguishes it from the remaining area.

Identify



Learning data
↓
Detected data

The process for screening points which are highly similar to the learning data.

E.g. Template matching, detection based on color, detection using edges, matching of trait information.

External to Camera

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

According to Koschmieder's intensity attenuation model (below formula), when the attenuation coefficient (σ) (\approx visibility) and distance (d) increase, the apparent intensity of the target becomes closer to the intensity of the surrounding environment (intensity of the background (sky)).

$$L = L_0 e^{-\sigma d} + L_f (1 - e^{-\sigma d})$$

L : apparent intensity of the subject

L_0 : target's intensity without scattering

L_f : brightness of surrounding environment (intensity of the sky)

Source: Mori, "Fog Density Recognition by In-vehicle Camera and Millimeter Wave Radar", IJICIC vol.3, Num.5 Oct 2007

Due to the attenuation of light in the atmosphere, the contrast C of an object when viewed from a distance d is as follows.

$$C = e^{-\sigma d} \quad \sigma: \text{Light attenuation coefficient (dissipation coefficient)} = \text{the rate at which light intensity decays with distance}$$

If the contrast identification limit is ε_0 , then the below can be expressed (Koschmieder's Law), with ε_0 set at 0.02 or 0.05, etc. based on experience.

$$V = \frac{1}{\sigma} \ln \left(\frac{1}{\varepsilon_0} \right)$$

V: Visibility

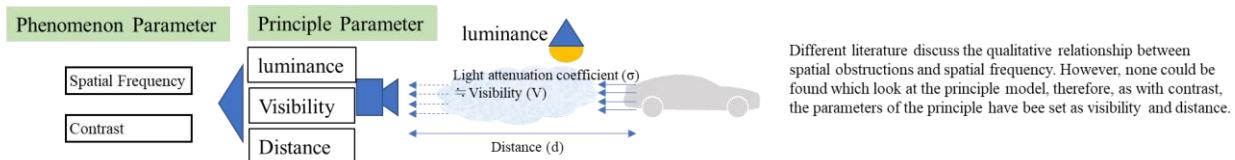
Source: Takata, 2004, 'Measuring Visibility', Japan Society of Snow Engineering Journal, Vol. 3 (20)

Takeuchi, 1991, 'Snow Particles in Space and Visibility', Journal of Geography, 100 (2). 264-272

$$V = \begin{cases} 3.912/\sigma & (\varepsilon_0 = 0.02) \\ 2.996/\sigma & (\varepsilon_0 = 0.05) \end{cases}$$

General visibility meters measure meteorological optical range (MOR) defined by the World Meteorological Organization (WMO).
MOR: the distance at which luminous flux in the collimated beam from an incandescent lamp of 2700K is reduced back to 5% of its original value

Therefore, the contrast is affected by the luminance of the light source, the light attenuation coefficient σ , and the distance d.
Since σ and the viewing distance V are uniquely corresponding values, and the viewing distance is generally used to measure the viewing distance, the principle parameters are as follows.



Mode A Uniform Spatial Obstructions

Causal Factor Parameters (underlined)

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

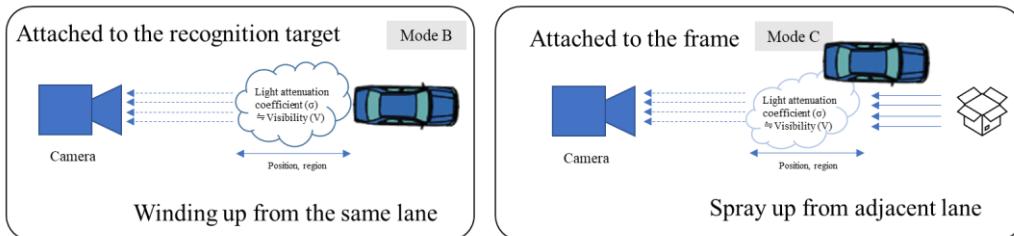
① Fog	The density of fog (attenuation coefficient) is generally expressed in terms of <u>visibility</u> , therefore the causal factor parameter will similarly be 'visibility'.
② Rain	E.g. $V = 8807.1e^{-0.1R}$ R: Intensity of rainfall [mm/10min] Source : Nishimura, 2015, 'Relationship Between Muddy Water Causing Evacuation and Optical Distance', International Journal of Erosion Control Engineering, Tsukuba University.
③ Snow	There are several literature, however visibility will change according to the <u>intensity of snowfall, wind speed</u> , etc. E.g. $V = 1150 \cdot \left(\frac{5}{3} R \right)^{-0.76}$ $V = 10^{-0.77 \log(M_f) + 2.85}$ R: Intensity of snowfall [mm/h] M_f : Blizzard rate (according to intensity of snowfall and wind speed) Source : Saito, 1971, 'Intensity of Snow Fall and Visibility', Report of the National Research Center for Disaster Prevention. Source : Matsuzawa, 2007, 'Study Related to Improvement of Methods for Estimating Visibility in Blizzards', Bulletin of Glaciological Research.

Mode B Mode C

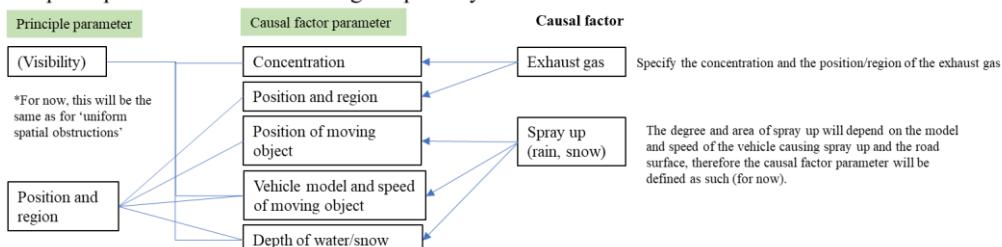
Localized Spatial Obstructions

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

The principle is the same as "uniform spatial obstructions," differing only in the limited scope.

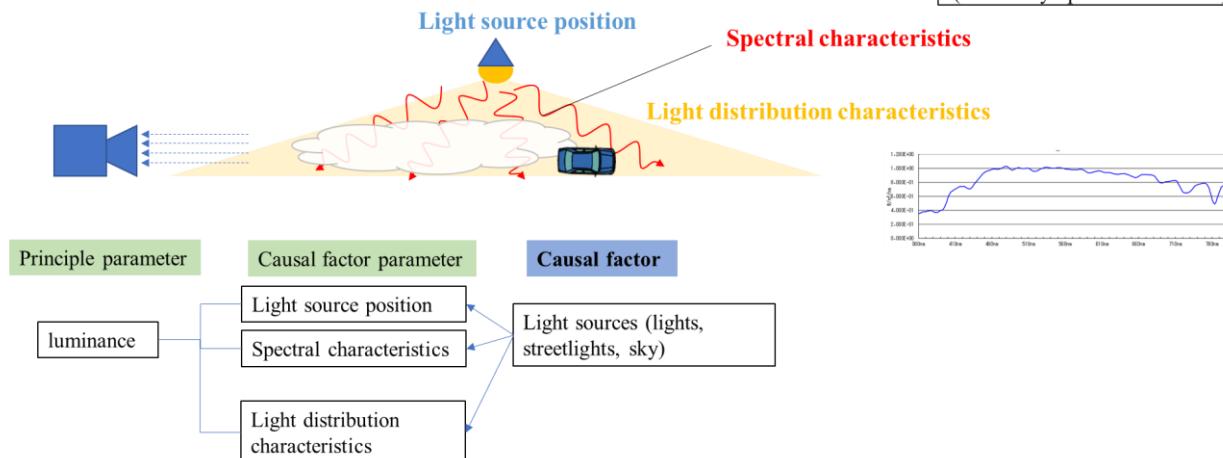


Exhaust gas and spray up (of rain or snow) have been listed as 'causal factors', with the parameters for the principle and causal factor being temporarily defined as below:



The luminance of a light source is determined by the spectral characteristics of the light source, the position of the light source, and the light distribution characteristics.

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)



Impact on Visibility of Light Source

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

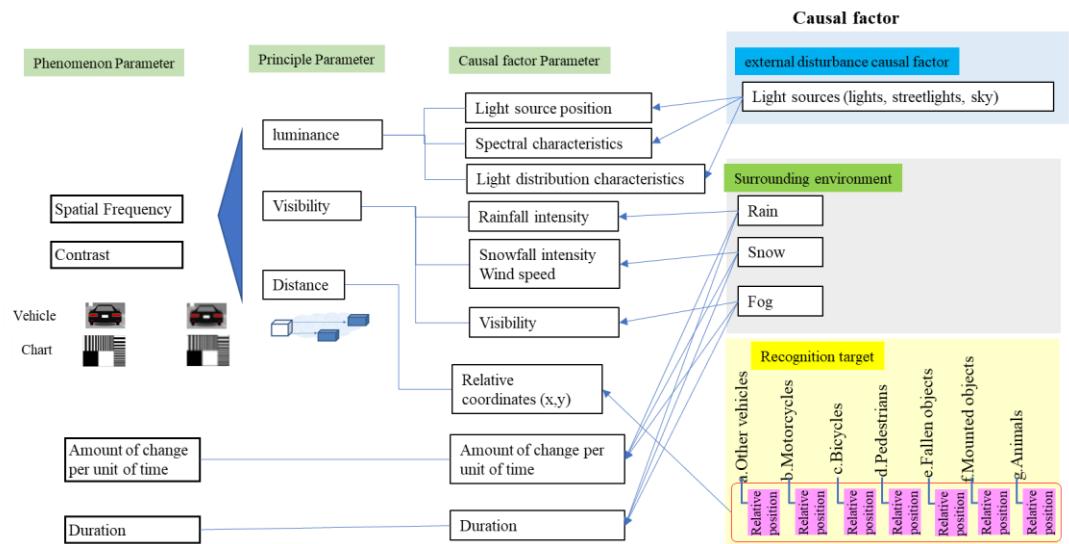
The light source from a target (tail lamps, etc.), can become the trait information used to detect the target (especially at night). Spatial obstructions such as fog, etc. can cause veiling of the lights, causing the intensity distribution of the light veiling to become superimposed onto the intensity distribution of the actual light source. (The intensity will gradually decrease as it extends out away from the light source). The intensity ratio between the veiling and the actual light source, will change in constant relationship with fog density. Further, if the difference in intensity between the light source and the background is large, the veiling will appear more prominent.

E4.2.2 Hierarchy of disturbance causal factor

E4.2.2.1 Causal Factors based on the Principle

Full frame Mode A

The below shows the relationship between disturbances (spatial obstructions) and their parameters.



Attached to the recognition target

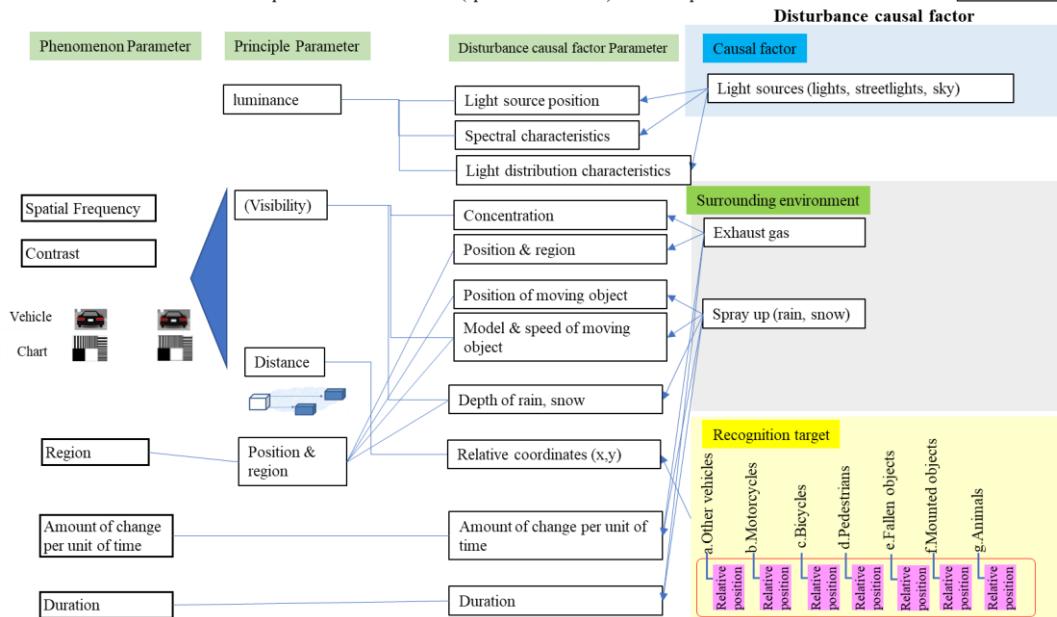
Mode B

Attached to the frame

Mode C

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

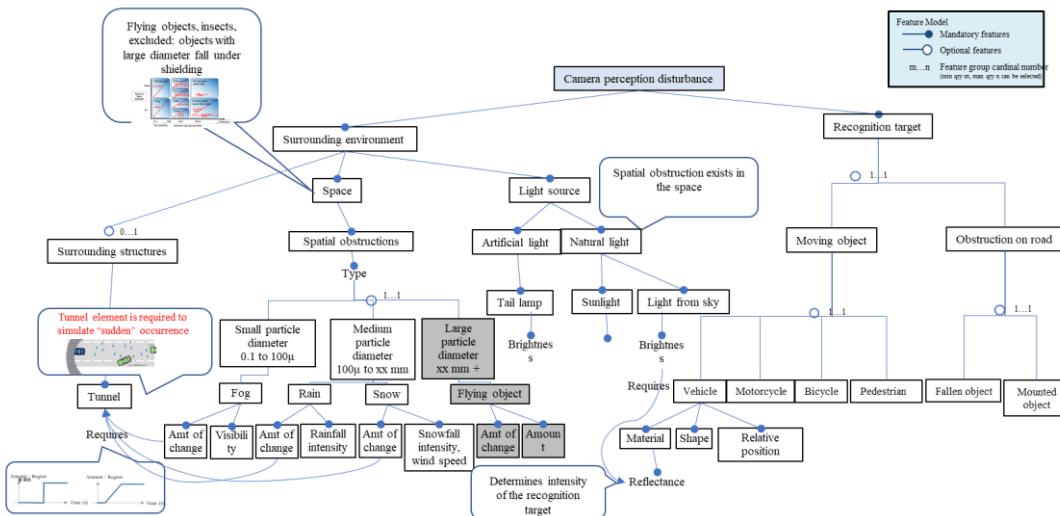
The below shows the relationship between causal factors (spatial obstructions) and their parameters.



Hierarchy of disturbance causal factors

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

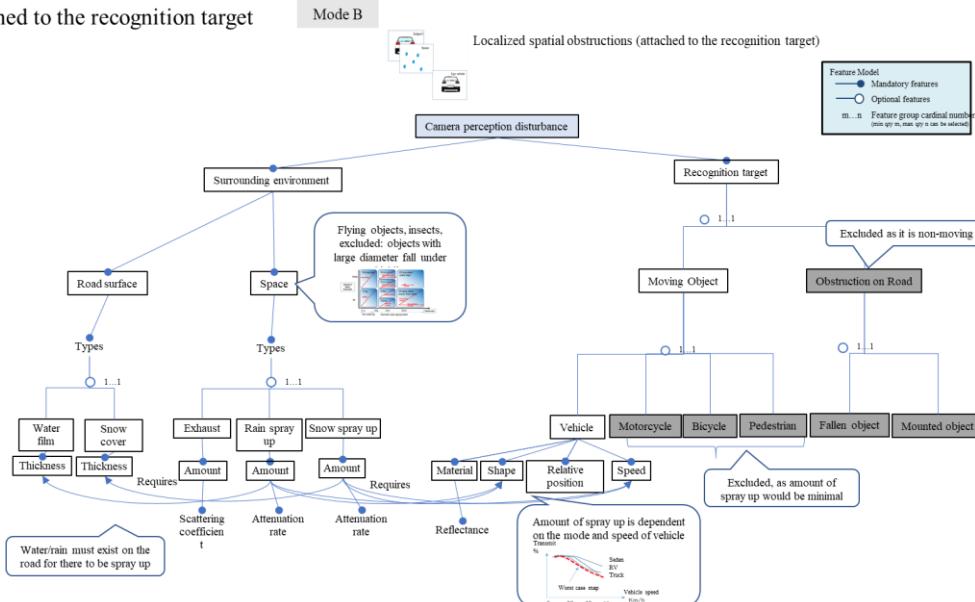
Full frame Mode A



Hierarchy of disturbance causal factors

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

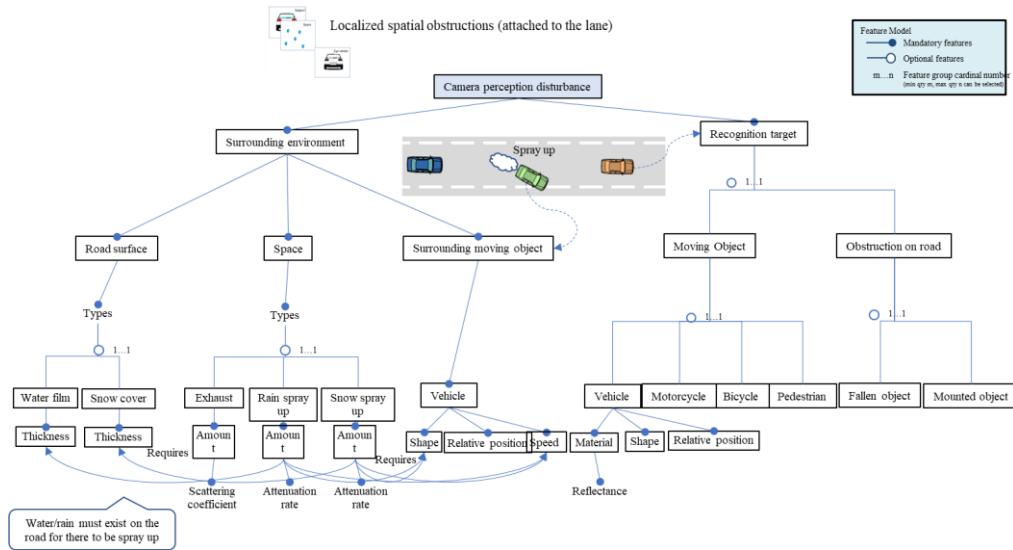
Attached to the recognition target Mode B



Hierarchy of disturbance causal factors

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

Attached to the frame Mode C



E4.2.2.2 Parameter Range

Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

Phenomenon Parameter	Principle Parameter	Disturbance causal factor Parameter	Disturbance causal factor	Parameter Range	Conditions		Basis
					STEP 1	STEP 2	
Spatial frequency Contrast	Visibility	Visibility	Fog	limit of ODD[m]~∞ [m]			
		Rainfall intensity	Rain	0~limit of ODD [mm/h] (50[mm/10min])	30, 50, 80mm/h		
		Snowfall intensity Wind speed	Snow	0~limit of ODD [mm/h] 0~limit of ODD [m/s]			Standard until traffic regulations apply
	Distance	Relative coordinates	Recognition target : relative position	Refer to traffic flow scenario			
				∞ *Assume the exit point of a tunnel or sudden change in weather			Difficult to define worst case (realistic) scenario (max amount of change), thus for now is ∞
				Continuous			'Continuous' is more severe

Low Spatial Frequency Low Contrast (Caused by Spatial Obstruction)
--

Explaining recognition target

Targets that are similar in colour to their background, have lower contrast, thus are more unfavourable (refer to the 'low contrast' error mode). Set the background as asphalt, concrete (black, gray) and snow (white), and select recognition target that are the same in colour.

Types	Parameters
Vehicles	【Colour (body colour) Black, Gray, White
Motorcycles	【Colour (body colour, motorcyclist wear) Black, Gray, White
Bicycles	【Color (motorcyclist wear) Black, Gray, White
Pedestrians	【Color (clothing) Black, Gray, White
Mounted objects	Generally these are highly visible and thus are not usually low in contrast. Use arrow signs and safety cones which are often found bordering lanes.
Fallen objects	【Color】 Here we use tires (car component) which are over 15cm in height and ranked one of the highest when looking at occurrences of fallen objects Black (tire)
Animals	Road kill to be included under fallen objects.

Low Spatial Frequency Low Contrast (Caused by Spatial Obstruction)
--

Derive the functional scenario by linking the ALKS scenario and the causal factors

Functional Scenario	ALKS scenario	Lane		Traffic Information		Moving Object		Obstruction on Road		Explanation		
		Lane markings	Structures	Edge of road	Traffic lights	Road signs	Road surface signs	Other vehicles	Motorcycles	Bicycles	Pedestrians	Fallen objects
F-1	Cut-in							<input type="radio"/>	<input type="radio"/>			
F-2	Cut-out							<input type="radio"/>				
F-3	Deceleration							<input type="radio"/>	<input type="radio"/>			
F-B1-14	Lane Keep	<input type="radio"/>										

E4.2.2.3. Evaluation Scenario

In the real world, disturbances can occur in combinations. Below are the combinations taken from the feature model.

Rain drop adherence to the front of the sensor during rain, leads to 'refraction' error, however it is included here as it may occur in conjunction with another.

*Multiple spatial obstructions could occur simultaneously, however they have been excluded for now.

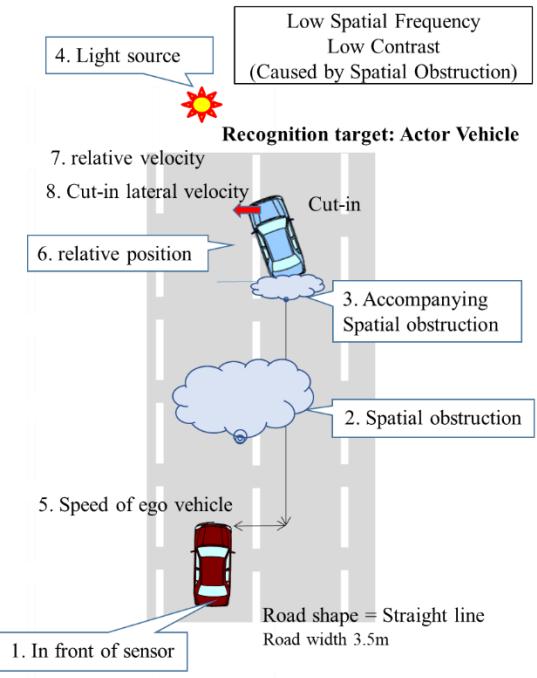
Low Spatial Frequency
Low Contrast
(Caused by Spatial Obstruction)

Scenario No.	Vehicle/Sensor		Surrounding environment						Notes	Mode
	In front of sensor		Spatial obstruction			Accompanying target		Light source		
	Rain drops (Refraction)	Snow (Shielding)	Fog	Rain	Snow	Spray up	Exhaust			
01	x	x	○	x	x	x	x		Day	A
02	x	x	○	x	x	x	x		Night	A
03	x	x	○	x	x	○	x		Day	A,B
04	x	x	○	x	x	○	x		Night	A,B
05	○	x	x	○	x	○	x		Day	A,B,(C)
06	○	x	x	○	x	○	x		Night	A,B,(C)
07	○	x	x	○	x	x	x		Day	05 is harsher
08	○	x	x	○	x	x	x		Night	06 is harsher
09	x	○	x	x	○	○	x		Day	A,B,(C)
10	x	○	x	x	○	○	x		Night	A,B,(C)
11	x	○	x	x	○	x	x		Day	09 is harsher
12	x	○	x	x	○	x	x		Night	10 is harsher
13	x	x	x	x	x	○Rain	x		Day	B
14	x	x	x	x	x	○Rain	x		Night	B
15	x	x	x	x	x	○Snow	x		Day	B
16	x	x	x	x	x	○Snow	x		Night	B
17	x	x	x	x	x	x	○		Day	B
18	x	x	x	x	x	x	○		Night	B

Scenario F-1

Evaluation based on a 'Cut-in on a straight line' scenario

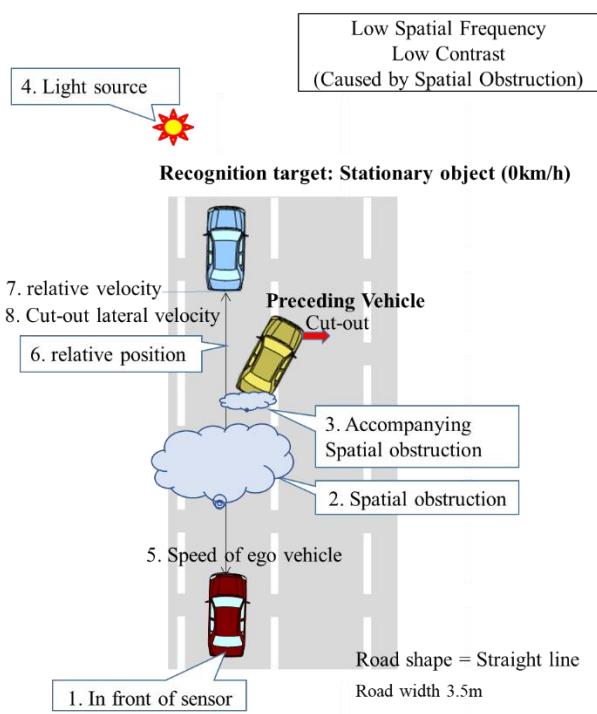
Disturbance causal factor parameter	1. In front of sensor	Rain drops	
		Snow	
	2. Spatial obstruction	Fog	Visibility 10m~1km
		Rain	Rainfall intensity 0~limit of ODD
		Snow	Snowfall intensity 0~limit of ODD Wind speed 0~limit of ODD
	3. Accompanying Spatial obstruction	spray up	
		Exhaust gas	
	4 .Light source	Day	
		Night	
	5. Speed of ego vehicle	Not decided here because of the scope of definition in the traffic flow scenario.	
Parameters required for evaluation	6. relative position		
	7. relative velocity		
	8. Cut-in lateral velocity		



Scenario F-2

Evaluation based on a 'Cut-out on a straight line' scenario

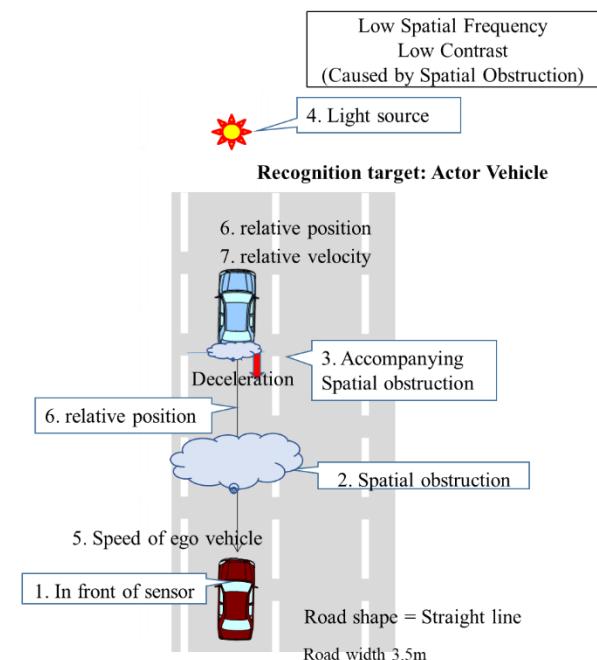
Parameters required for evaluation	Disturbance causal factor parameter
1. In front of sensor	Rain drops Snow
2. Spatial obstruction	Fog Rain Snow
3. Accompanying Spatial obstruction	spray up Exhaust gas
4. Light source	Day Night
5. Speed of ego vehicle	Not decided here because of the scope of definition in the traffic flow scenario.
6. relative position	
7. relative velocity	
8. Cut-out lateral velocity	



Scenario F-3

Evaluation based on a 'Deceleration on a straight-line scenario'

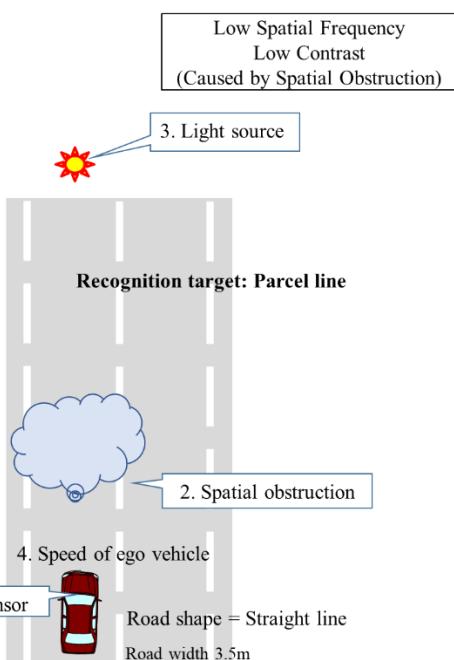
Parameters required for evaluation	Disturbance causal factor parameter
1. In front of sensor	Rain drops Snow
2. Spatial obstruction	Fog Rain Snow
3. Accompanying Spatial obstruction	spray up Exhaust gas
4. Light source	Day/Night
5. Speed of ego vehicle	Not decided here because of the scope of definition in the traffic flow scenario.
6. relative position	
7. relative velocity	



Scenario F-B1-14

Evaluation based on a ‘Lane Keep on a straight line’ scenario

Disturbance causal factor parameter	1. In front of sensor	Rain drops Snow
	2. Spatial obstruction	Fog Rain Snow
	3. Light source	Day/Night
Parameters required for evaluation	4. Speed of ego vehicle	Not decided here because of the scope of definition in the traffic flow scenario.



E.4.3 [Camera] Excessive (saturation), Whiteout

E.4.3.1 The Phenomenon and Principle

Excessive (saturation)
Whiteout

E.4.3.1.1 The Phenomenon

When the bright area within a frame exceeds the upper limit of intensity (upper limit of the dynamic range) which the camera can express, the camera will no longer be able to express the difference in intensity (tone) causing a deficiency in information and ultimately a non-detection.

Example

Whiteout of the vehicle ahead due to the reflection of the test vehicles headlamps



The background shines white making it difficult to make out profiles such as license plates

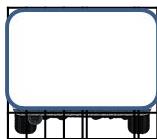
Whiteout caused by the sunlight at the exit point of a tunnel



Excessive (saturation)
Whiteout

Region

The number of pixels in which whiteout is occurring within the frame of the recognition target



Amount

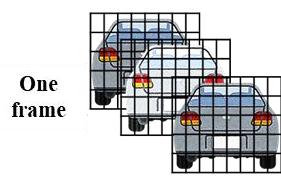
If over threshold then whiteout



No occurrence Occurrence

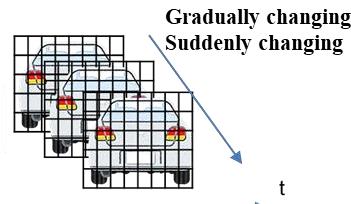
Time

Temporary



Multiple frames

Continuous



E.4.3.1.2 Outline of the Principle

Excessive (saturation)
Whiteout

Causal factors of the disturbance based on the principle

Phenomenon Mode		A		B		C		
	Amount	1	2	1	2	1	2	
Degree	Amount	Define by each principle of the sensor						
	Region	Full range within frame 		Attached to the recognition target Partial whiteout in the frame of the recognition target 		Attached to the frame Recognition target enters the frame where whiteout is occurring 		
	Time	Amount per unit of time	Define the range of the amount of change by principle/causal factors					
Causal factors and Evaluation Scenario	Duration	Continuous	Temporary	Continuous	Temporary	Continuous	Temporary	
	Out of scope In order for whiteout of the full frame to occur continuously, the full frame will need to be covered by the size of a spot light. This has been excluded as it is deemed not possible for this to occur in the scenario where the ego vehicle is following the preceding vehicle.		Out of scope Scenario whereby an intensity which could cause whiteout of the full frame, spreads, and exposure control cannot keep up. Exposure control cannot keep up at the exit point of a tunnel		Scenario whereby a strong intensity continues and is attached to the recognition target Recognition target is the source of the strong intensity light Reflection from the recognition target due to a strong light source		Scenario whereby exposure control cannot keep up with transient intensity. Recognition target is the light source (comes from shielding object, etc.) Reflection from surrounding light source (low⇒high, etc.)	
While overexposed Case where recognition target enters 								

E.4.3.1.3 The Principle model

Excessive (saturation)
Whiteout

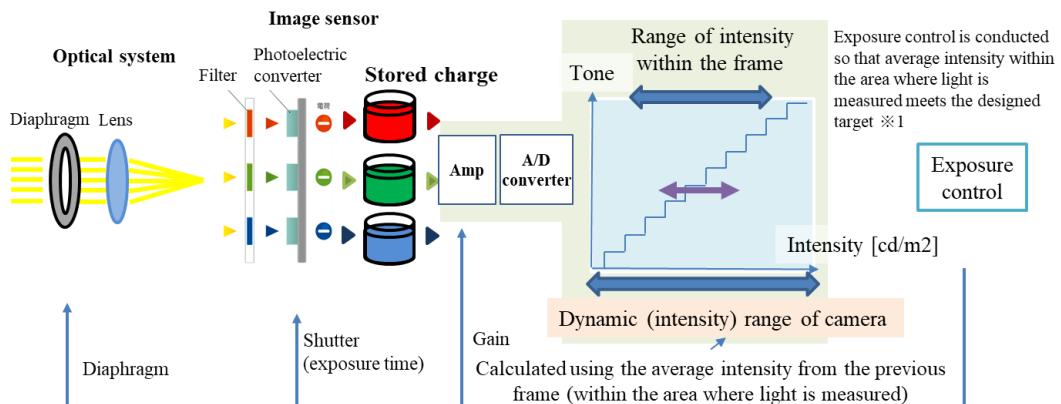
Outline of the principle

Whiteout is a phenomenon which occurs when the 'range of intensity in the scenery' is greater than the 'camera's dynamic (intensity) range' as adjusted by the exposure control.

Basic Principle

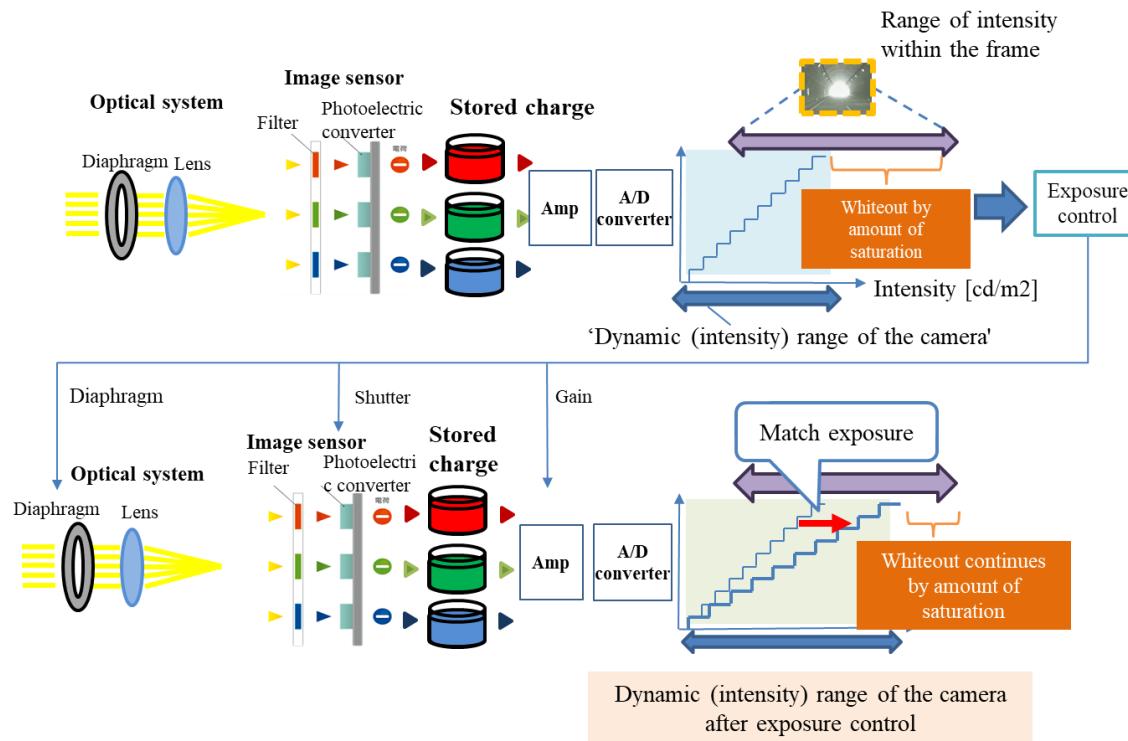
Exposure control identifies the difference in the average intensity of the previous frame (within the area where light is measured) to the target intensity, and determines the intensity range.

The intensity range within the frame is allocated to the above intensity range in order to express color.



※1 Details of control differs depending on manufacturer

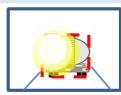
Excessive (saturation)
Whiteout



Excessive (saturation)
Whiteout

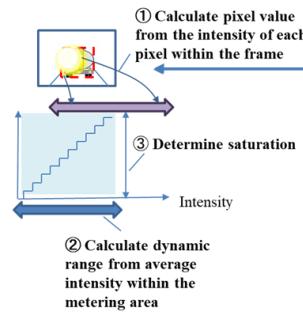
Phenomenon

Whiteout is emerging at a recognition target

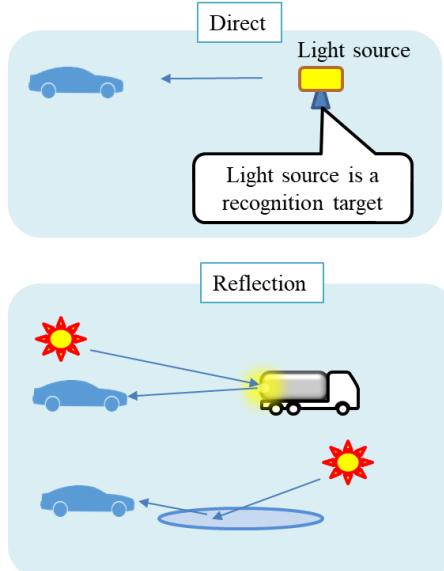


Internal model of a camera

The difference between the saturation value (calculated by the dynamic range) and the pixel value (calculated by intensity of each pixel)

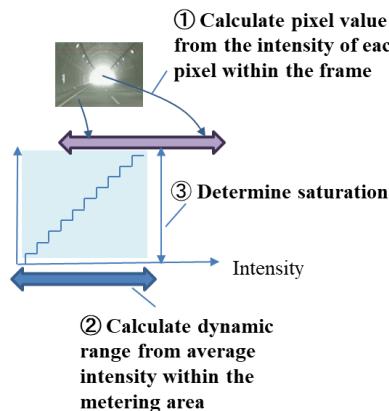


External model of a camera



Internal to Camera

Determines saturation by calculating pixel value and dynamic range



① Calculating pixel value

Pixel value (R,G,B) =

Intensity of each pixel [cd/m²] x Rt[ms] x G x Wg x E x K

Rt : Exposure time [ms] Wg : White gain (fixed value)

G : Gain z

E : Photoelectric converter coefficient (fixed value)

K : Color filter transmittance

Excessive (saturation)
Whiteout

② Calculating dynamic range

Dynamic range is determined by the average intensity of the previous frame (within the area where light is measured).

Average intensity within the frame (area where light is measured)
 $= \sum \text{pixel value} / \text{Number of pixels}$

③ Determining saturation

Brighter pixels than the saturation value (intensity) of each dynamic range will lead whiteout. ※using a table of the saturation value for the model

Bright pixel in the frame [cd/m²] > Saturation value

Principle parameter

Average intensity in the metering area among the image angle [cd/m²]

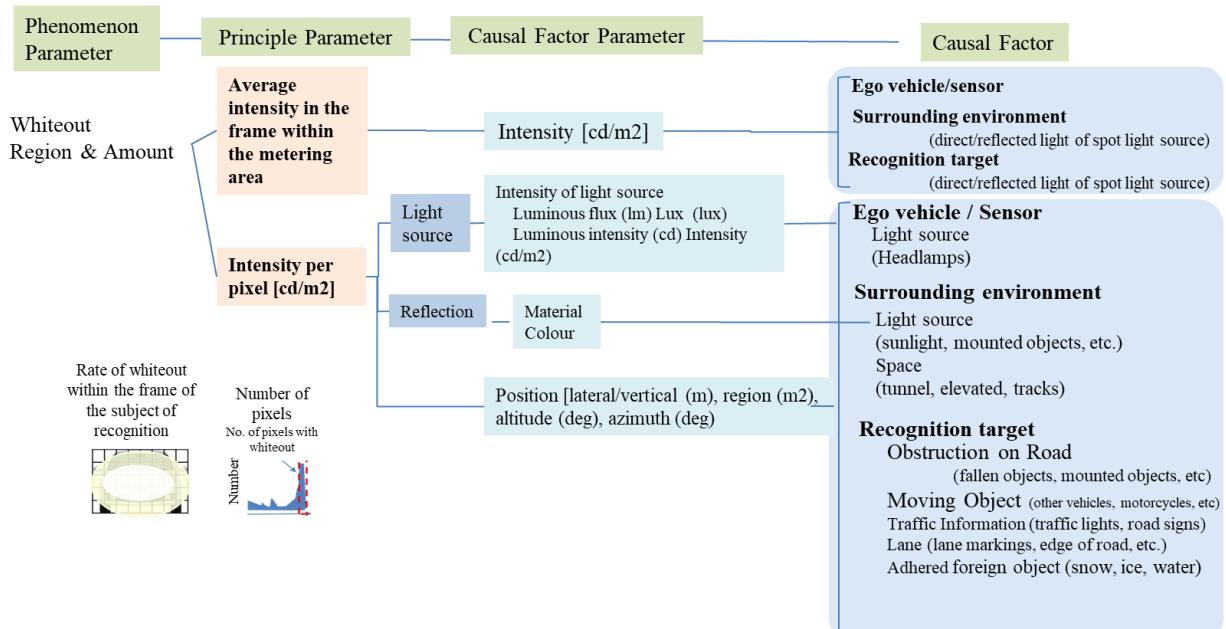


Bright pixel in the image angle [cd/m²]

E.4.3.2 The Causal Factors based on the Principle

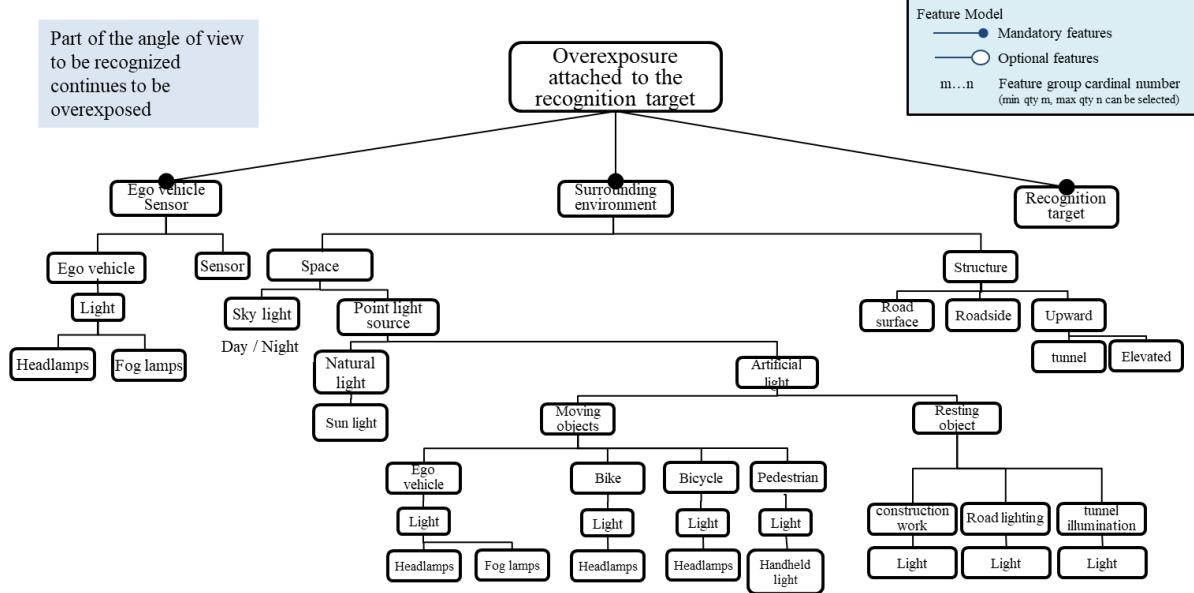
Excessive (saturation)
Whiteout

E.4.3.2.1 The Relationship between the Principle and Causal Factors

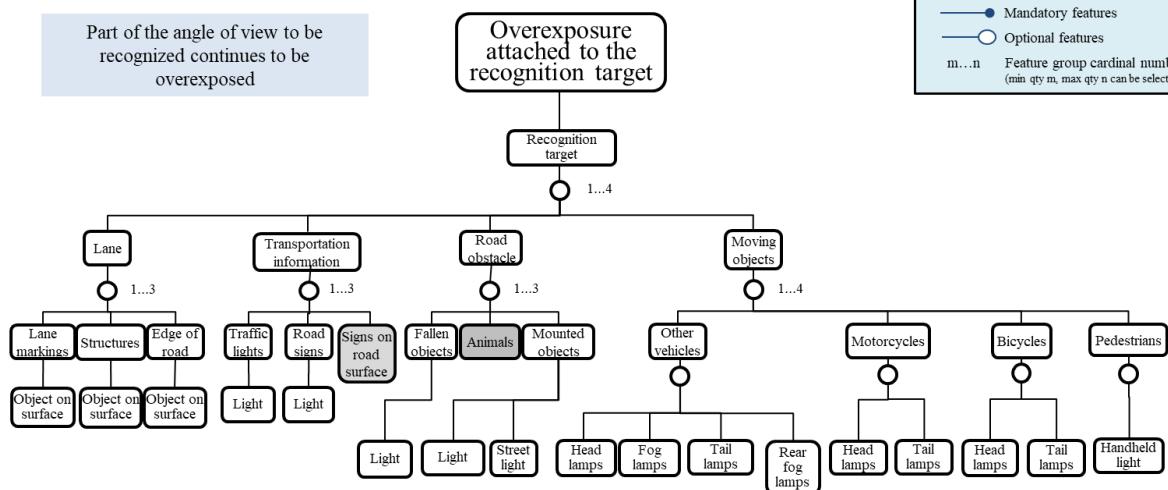


E.4.3.2.2 Causal Factor

Excessive (saturation)
Whiteout



Excessive (saturation)
Whiteout



Excessive (saturation)
Whiteout

Deriving the Functional Scenarios by connection between ALKS scenarios and causal factors

Functional Scenario	ALKS Scenario	Lane			Traffic Information		Moving Object		Obstruction on Road		Environment		Explanation				
		Lane markings	Structures	Edge of road	Traffic lights	Road signs	Road surface signs	Other vehicles	Motorcycles	Bicycles	Pedestrians	Fallen objects	Mounted objects	Animals	Sunlight	Road surface	Up above/tunnel
F-1	Cut-in							○	○						○	○	○
F-2	Cut-out							○	○	○	○	○	○		○	○	○
F-3	Deceleration							○	○						○	○	○
F-B1-14	Lane-keeping		○	○	○	○	○	○							○	○	○

E.4.3.2.3 Parameter Range

Excessive (saturation)
Whiteout

Ego vehicle /Sensor

Causal factor				Causal factor parameter	Range	Remarks
Ego vehicle / Sensor	Parts	Light	Headlamps	Brightness	2-lamp type Min Low 6400cd ~ High 15000cd~ Max Total ~430,000cd 4-lamp type Min Low 6400cd ~ High 12000cd ~ Max Total ~430,000 cd	Refer to the regulations of each country
			Fog lamps	Brightness	10000cd~	Refer to the regulations of each country

Excessive (saturation)
Whiteout

Surrounding environment

Causal factor			Causal factor parameter	Range	Remarks
Space	Light	Sky light			Light Determined by the brightness of sunlight and the brightness of light
		Point light source	Natural light	Sun light	Altitude 0~90degrees
					Up to the maximum altitude just below the equator
		Artificial light	Direction	0~359degrees	Up to the maximum azimuth
			Brightness	0lx~100000lx	Brightness of the midsummer sun
Structure	Light	Moving objects	Brightness	Same as own car headlights	
		Resting object	Brightness	~110k[lm]	struction lights
Structure	Light	Tunnel / Elevated			— Refer to the regulations of each country

Excessive (saturation)
Whiteout

Surrounding environment / Recognition target

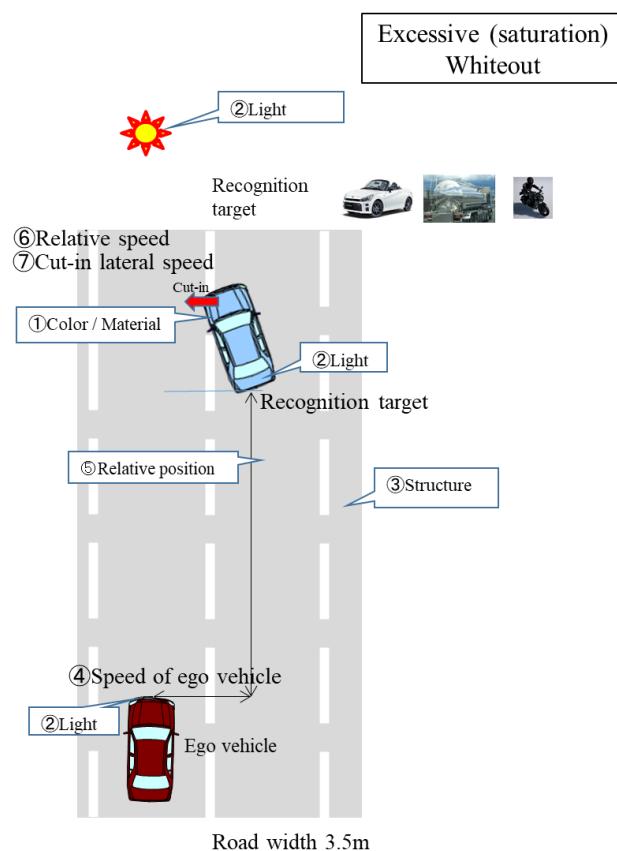
Causal factor	Causal factor parameter	Range	Remarks
Other vehicles	Color / Material	Color(White)	Colors that are prone to overexposure
	Light source	Tail lamps (300cd), Brake lamps (600cd), Hazard lamps (600cd), Rear fog lamps (345cd)	Refer to the regulations of each country
Vehicle with specular reflection	Color / Material	Material (Aluminum)	Material that is prone to overexposure
Motorcycles Bicycles	Color / Material	Color(White)	Colors that are prone to overexposure
	Light source	Tail lamps (300cd), Brake lamps (600cd), Hazard lamps (600cd)	Refer to the regulations of each country
Pedestrians	Color / Material	Color(White)	Colors that are prone to overexposure
	Light source	Handheld light (20~800lm)	Brightness of handheld lights for sale
Animal	Color / Material		

E.4.3.2.4 Evaluation Scenario

E.4.3.2.4.1 Cut-in

Evaluated in a Cut-in scenario on a straight road.

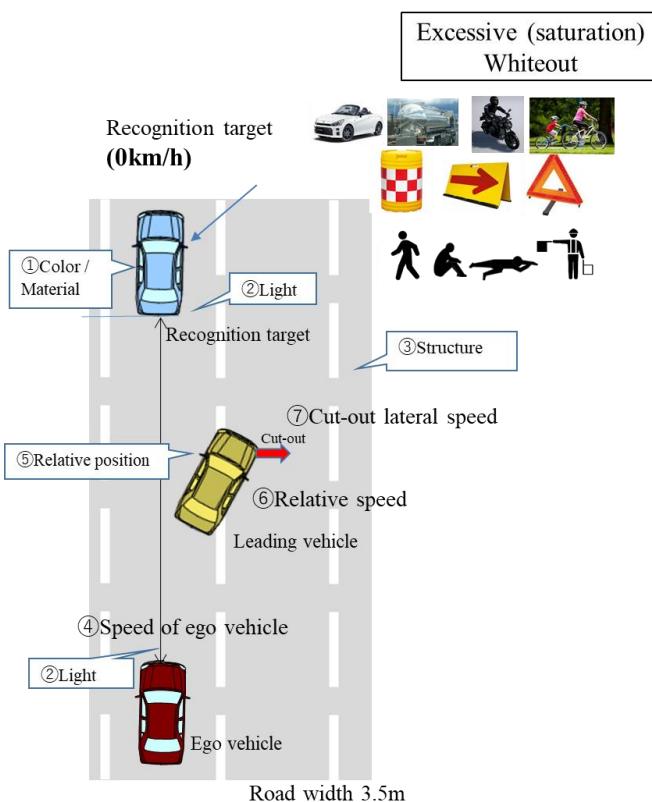
Parameters required for evaluation	①Color / Material	Color: White Material: Aluminum
	②Light	Natural light Sun light Artificial light Headlamps Tail lamps Rear fog lamps
	③Structure	Tunnel
	④Speed of ego vehicle	The range is defined by the traffic flow scenario, thus not defined here
	⑤Relative position	
	⑥Relative speed	
	⑦Cut-in lateral speed	



E.4.3.2.4.3 Cut-out

Evaluated in a Cut-out scenario on a straight road.

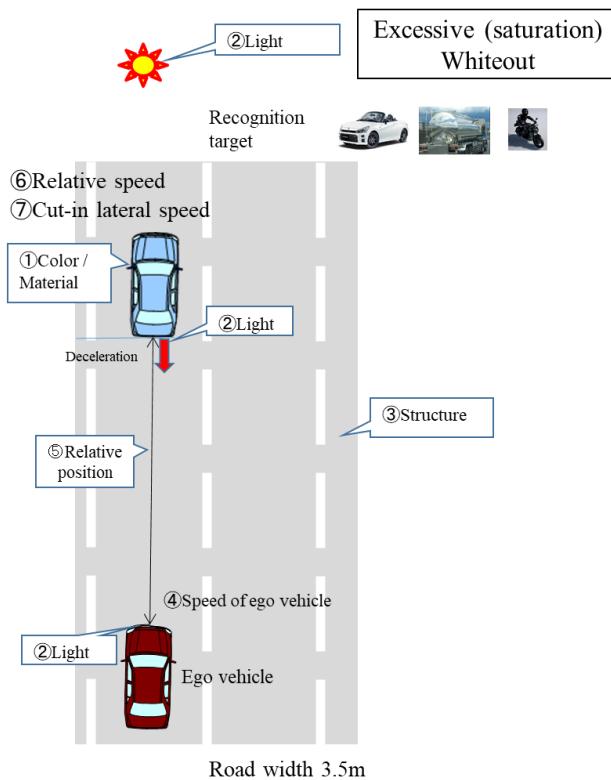
Parameters required for evaluation	①Color / Material	Color: White Material: Aluminum
	②Light	Artificial light Headlamps Tail lamps Rear fog lamps
	③Structure	Tunnel
	④Speed of ego vehicle	The range is defined by the traffic flow scenario, thus not defined here
	⑤Relative position	
	⑥Relative speed	
	⑦Cut-out lateral speed	



E.4.3.2.4.3 Deceleration

Evaluated in the Deceleration scenario on a straight road.

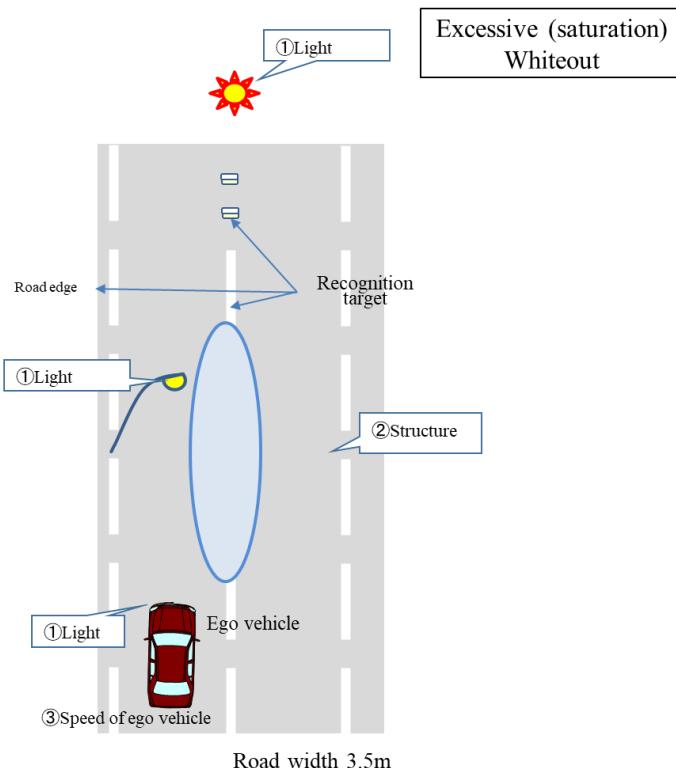
Parameters required for evaluation	①Color / Material	Color: White Material: Aluminum
	②Light	Natural light Sun light Artificial light Headlamps Tail lamps Rear fog lamps
	③Structure	Tunnel
	④Speed of ego vehicle	The range is defined by the traffic flow scenario, thus not defined here
	⑤Relative position	
	⑥Relative speed	
	⑦Deceleration	



E.4.3.2.4.4 Lane Keep

Evaluated in the Lane Keep scenario on a straight road.

Parameters required for evaluation	①Light	Natural light Sun light Artificial light Headlamps
	②Structure	Tunnel
	③Speed of ego vehicle	The range is defined by the traffic flow scenario, thus not defined here



Annex F

Guideline for validation of virtual environment with perception disturbance

Generally, environment in which not only automated vehicle but also human drive vehicle will run is not limited to clear and good condition, that means bad weather like rain and fog situation should be considered. These condition may cause recognition failure because sensor should receive perception disturbance. Safety evaluation of automated vehicle needs validation to consider these kind of disturbance.

Simulation technology, that is remarkably progress especially in physical modelling, is a method to evaluate perception performance with disturbance. Validation in virtual environment is high convenience to apply but validity of virtual environment should be discussed.

This annex will clarify the requirement to be confirmed when principle of perception disturbance for each sensor(camera, millimeter radar, LiDAR) discussed in Annex.E will be reproduced in virtual environment. Additionally a method to validate developed environment will meet each requirement or not will be proposed.

The points to be discussed in this Annex are shown in fig.F-1.

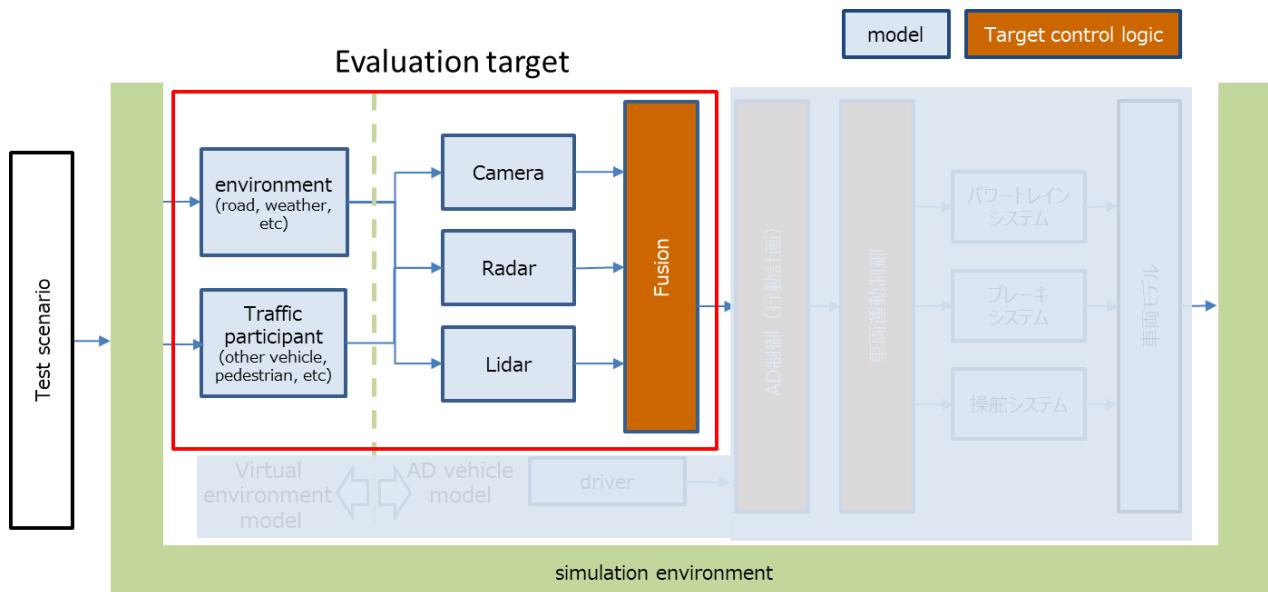


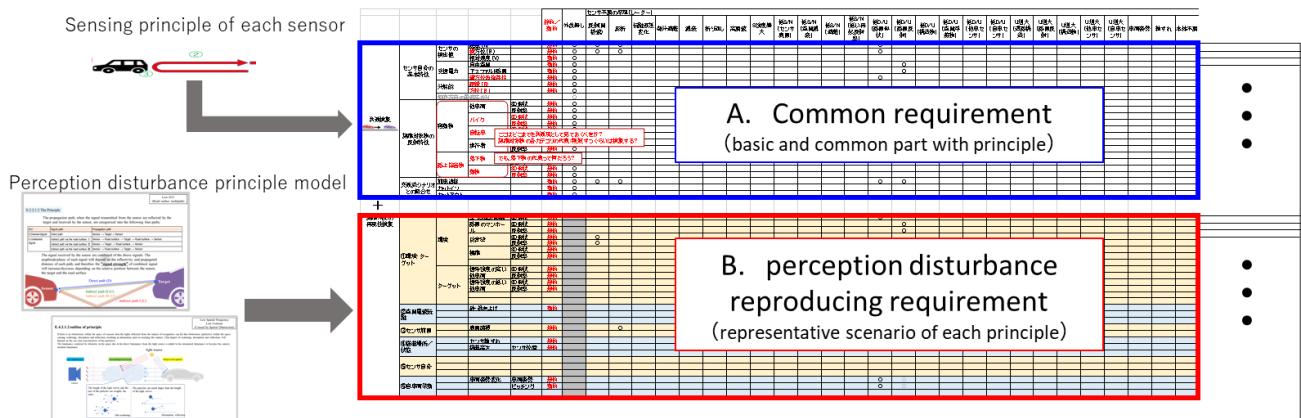
Fig.F-1) area of this Annex

F.1 overview of requirements defined in this Annex

To judge whether perception performance evaluation in virtual environment will work well or not, it is necessary for relatives to have common understanding about how models and environment deployment will be validated. Final target would be to realize that evaluation result in virtual environment and real condition will be matched, thus we propose the definition of validation method in ideal condition (without perception disturbance) in advance to validation with perception disturbance. This means we can easily analyze the root cause of unmatch with disturbance (this is final target) by establishing validation method in ideal condition

We define requirement of validation in ideal condition as “A. Common requirement” and requirement of validation with perception disturbance as “B. perception disturbance reproducing requirement”(fig.F-2). Additionally we propose each validation method about “A. Common requirement” and “B. perception disturbance reproducing requirement”

Requirement for virtual environment



Validation method

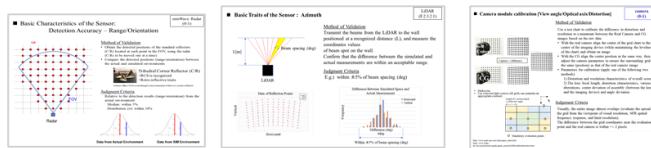


Fig.F-2) overview of this Annex

A. Common requirement

- Define requirement to be confirmed in ideal condition (without perception disturbance) from each sensors' principle

B. perception disturbance reproducing requirement

- Define requirement to be confirmed with perception disturbance
- Clarify necessary principle parameter for reproducing disturbance and disturbance causal factor parameter by classifying various disturbance based on the principle and describe it as a model about each disturbance principle

F.2 Common requirement and reproductivity validation method

In this section items to be confirmed as common requirement and validation method are clarified. As a first step, clarifying the way of thinking about what items should be done as common requirement is shown. After that clarifying validation method for each sensor based on this way of thinking. This method is defined based on each sensors' principle thus it is necessary to clarify method when validating another principle's sensor following the way of thinking. This validation method shown below can be replaced by another method that can verify the same contents.

F.2.1 the way of thinking about common requirement

This section clarifies the way of thinking about the items to be set as common requirement. Component of object detection are defined as below elements as ①sensor/vehicle itself, ②space where the signal propagates ③recognition target(fig.F-3), and items to be validated and their criteria without perception disturbance for each elements are clarified. Additionally the method to validate that recognition target can be detected under basic traffic disturbance scenario is defined to confirm this totally.

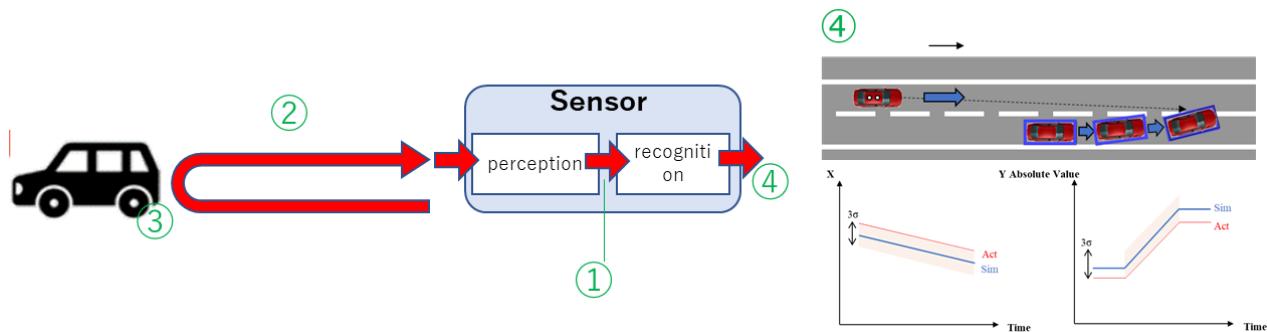


fig.F-3 element of common requirement

① sensor/vehicle basic characteristics

To confirm basic perception results like distance, direction, relative speed, signal intensity(items and condition differ in sensor principle) in ideal condition (without perception disturbance) as a sensor basic characteristics.

② characteristics of propagation, optical characteristics and so on

To confirm signal propagation from perception target to sensor in ideal condition would be reproduced.

③ reflection characteristics of perception target and so on

To confirm perception result would be reproduced. This is not only for perception result but also recognition result.

④ target recognition under traffic scenario

To confirm recognition result of the target under basic traffic scenario (following, cut-in, cut-out) would be reproduced.

F.2.2 The way of thinking about common requirement for each sensor

F.2.2.1 the way of thinking about common requirement for millimeter wave Radar

In accordance with the principles of the Radar perception, validates whether physical amount of distance, direction, relative speed and received wave intensity are reproduced (fig.F-4).

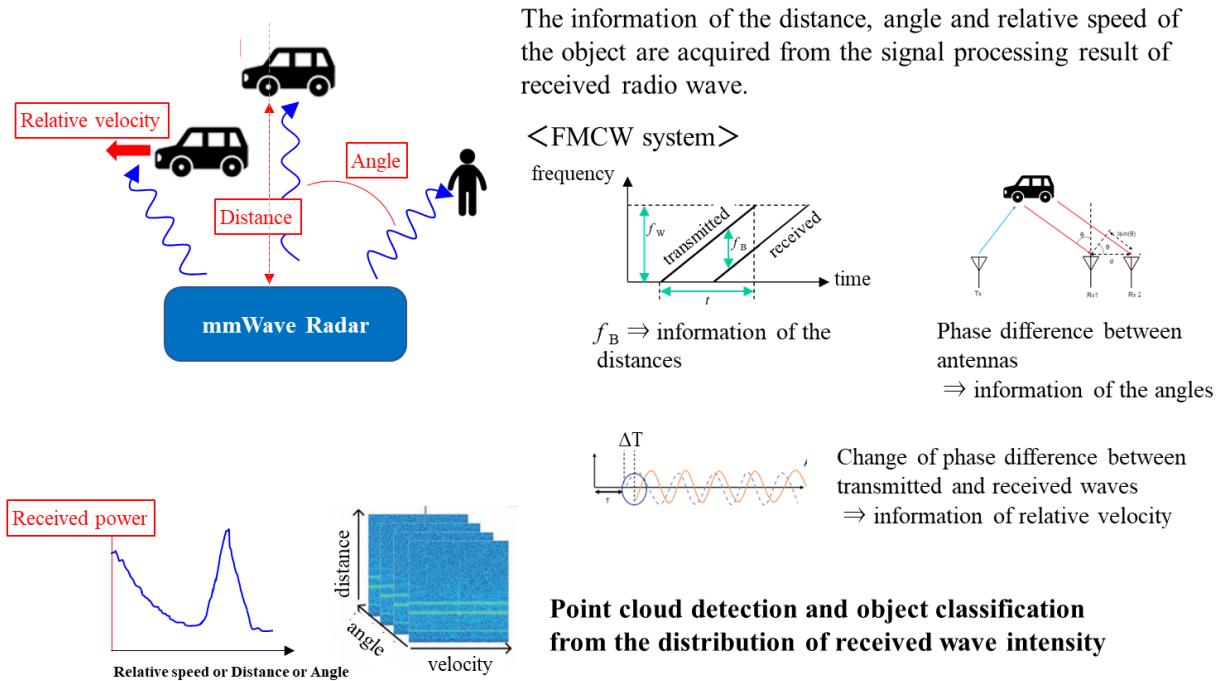


Fig.F-4. the way of thinking about common requirement for millimeter wave Radar

Based on this way of thinking, list of actual requirement shown in table.F-1 is clarified.

										Perception process												Recognition process								
										Signal from perception target (S)						Signal from others						Processing ability	Processing performance							
										Phase			Strength			Noise (N)			Low S/N			Low D/U			Increasing of U					
										Frequency	Reflection (indirect wave)	Refraction	Change of propagation delay	No signal (partial)	Aliasing	Harmonic	Large difference of signal	Low S/N (change of angle)	Low S/N (attenuation at the sensor surface)	Low S/N (attenuation in space)	Low S/N (low retroreflection)	Low D/U (change of angle)	Low D/U (road surface reflection)	Low D/U (surrounding structures)	Low D/U (floating objects in space)	Low D/U (sensors on other cars)	Increasing of U (road surface reflection)	Lack of points to be processed	False detection of undesired signal	No detection of required signal
Basic Characteristics of the Sensor	Detection Accuracy	Items	Paramters	Requirements	Method of Validation No.	not a disturbance	0-1	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Range (R)	Distance	Detecting position of C/R is equivalent to the actual environment.				○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	
		Orientation (azimuth) (θ)	Azimuth angle					○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	
	Resolution	Orientation (elevation) (φ)	Elevation angle					○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	
		Relative speed (V)	Distance	Detecting relative speed of C/R is equivalent to the actual environment.				○	○																					
		Received power (P)	Azimuth angle	Received power of the reflection wave from C/R is equivalent, and the side lobe is				○	○					○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	
	Discrimination	Range (R)	Distance	The minimum resolution when two C/R are closely apposed is equivalent to the actual environment.				○	○					○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	
		Orientation (azimuth) (θ)	Azimuth angle					○	○					○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	
		Orientation (elevation) (φ)	Elevation angle					○	○					○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	
Common Requirement	Properties of radio wave	Relative speed (V)	Relative speed	The minimum resolution when two C/R are moved in different speed is equivalent to the actual environment.			0-5	○	○																					
		Range (R)	Distance	The minimum discrimination when two C/R are closely apposed is equivalent to the actual environment.				○	○					○																
	Vehicle (Passenger Vehicle)	Orientation (azimuth) (θ)	Azimuth angle					○	○					○																
		Orientation (elevation) (φ)	Elevation angle					○	○					○																
		RCS	Angle	RCS of a vehicle is equivalent to the actual environment in all directions.				1-1	○																					
	Vehicle (Large-Sized Vehicle)	Reflection Points	Angle	Reflection peak intensity from a vehicle is equivalent to the actual environment.				1-2	○		○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Reflection Points	Distance	Reflection peak intensity from a vehicle is equivalent to the actual environment.				1-2	○		○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		RCS	Angle	RCS of a large-sized vehicle is equivalent to the actual environment in all directions.				1-1	○																					
	Pedestrian	Reflection Points	Angle	Reflection peak intensity from a large-sized vehicle is equivalent to the actual environment.				1-2	○		○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Reflection Points	Distance	Reflection peak intensity from a large-sized vehicle is equivalent to the actual environment.				1-2	○		○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		RCS	Angle	RCS of the dummy is equivalent to the actual environment in all directions.				1-3	○																					
Basic Traffic Flow Scenario	CCRs	Received Power	Distance	Received power from a vehicle is equivalent to the actual environment.			2-1	○																						
		Detecting Position (Distance/Angle)	Time	Detecting position of the signal from a vehicle is equivalent to the actual environment.				2-2	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Detecting Speed	Time	Detecting speed of the signal from a vehicle is equivalent to the actual environment.				2-3	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Object Detecting Position (Distance/Angle)	Time	Object detecting position of a vehicle is equivalent to the actual environment.				2-4	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Object Detecting Speed	Time	Object detecting speed of a vehicle is equivalent to the actual environment.				2-5	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
	Cut-in (Passenger vehicle)	Object Detecting Position (Distance/Angle)	Time	Object detecting position of a vehicle is equivalent to the actual environment.				2-6	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Object Detecting Speed	Time	Object detecting position of a vehicle is equivalent to the actual environment.				2-7	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
	Cut-in (Large-sized trailer)	Object Detecting Position (Distance/Angle)	Time	Object detecting position of a trailer is equivalent to the actual environment.				2-6	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Object Detecting Speed	Time	Object detecting speed of a trailer is equivalent to the actual environment.				2-7	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
	Cut-out	Object Detecting Position (Distance/Angle)	Time	Object detecting position of a vehicle is equivalent to the actual environment.				2-8	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○
		Object Detecting Speed	Time	Object detecting speed of a vehicle is equivalent to the actual environment.				2-9	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○	○

Appropriate for all the items

Table F-1. List of common

F.2.2.2 the way of thinking about common requirement for LiDAR

In accordance with the principles of the LiDAR perception, validates whether physical quantities like azimuth, range, strength, number of detection points and size are reproduced(fig.F-5).

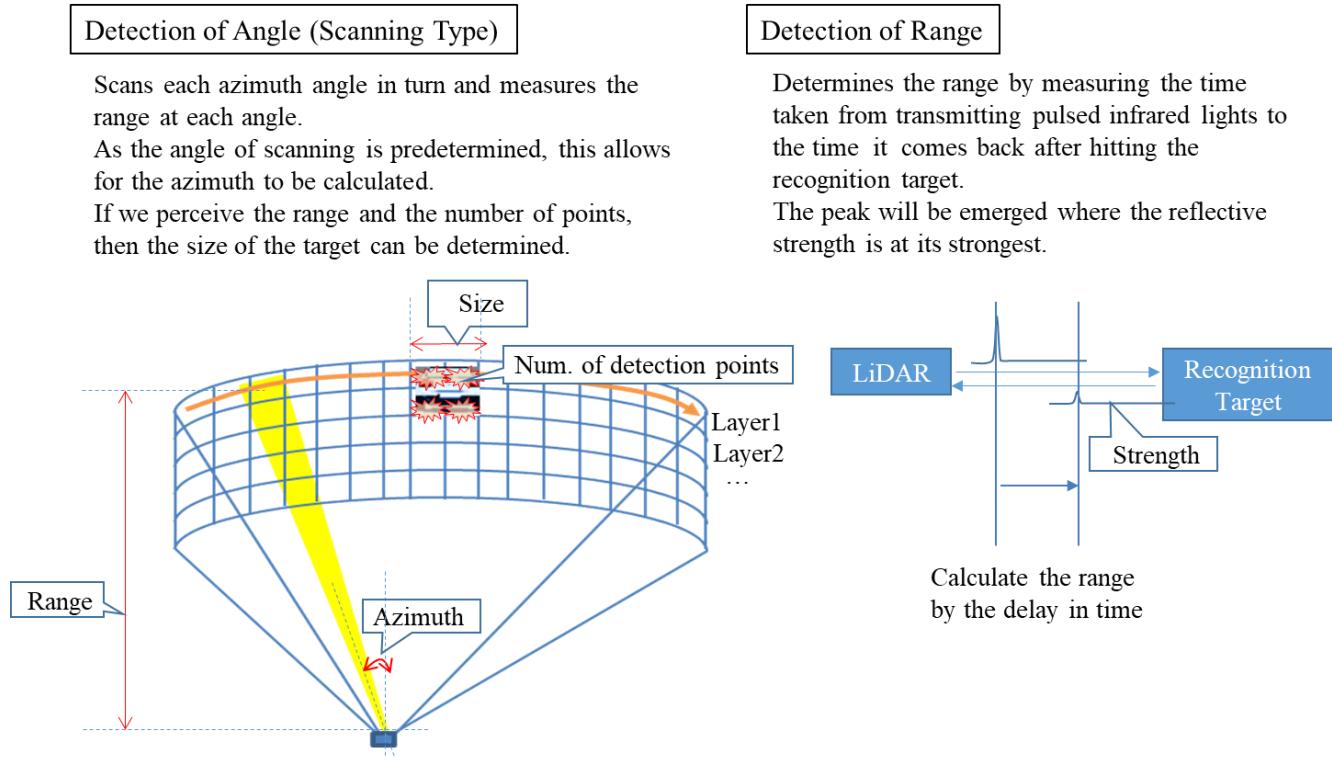


Figure F-5. LiDAR detection principle matrix

Based on this way of thinking, list of actual requirement shown in table.F-2 is clarified.

										No disturbance	Perceptual part																
											Scan timing		Signal from recognition target (S)				S strength			S Propagation direction		S speed	N factor		U factor		
											Misalignment of overall spatial position	Misalignment of position of recognition target	Saturation of S	Attenuation of S				No S due to occlusion	Reflection	Refraction	Arrival time of S	Pulsed noise	DC noise	Multiple reflections	Signal from non-recognition target (Reflection)	Signal from non-recognition target (Refraction)	
											Low reflection of the recognition target	Adhesion to the recognition target	Rain/Snow/Fog	Exhaust gas/Hoisting	Adhesion to the sensor												
common requirement	Verification perspective		Explanation	target	item	Parameters	request		Validation Method No.		[Point cloud data] Direction, Distance Reflectance Shape	Direction	The direction of the reflector can be detected in the same way as in the real environment.	F.2.3.2.1	O	O	O						O				
	Basic characteristics of the sensor itself		In order to verify whether the basic performance of LiDAR can be reproduced, the direction, distance, detection probability, intensity, number of detection points, and size are compared by actual measurement and simulation using a standard reflector with known reflection.	Standard reflector	[Point cloud data] Direction, Distance Reflectance Shape	Distance	The distance of the reflector can be detected in the same way as in the real environment.	F.2.3.2.1.1	O			O	O	O	O	O	O	O	O	O	O	O	O				
						Detection probability	The detection probability of the reflector can be detected in the same way as in the actual environment.	F.2.3.2.1.2	O			O	O	O	O	O	O	O			O	O					
						Strength	The reflection strength of the reflector can be detected in the same way as in the actual environment.	F.2.3.2.1.3	O			O	O	O	O	O	O	O			O	O					
	Reflection characteristics of the object to be recognized	Static verification	After verifying the basic performance of LiDAR, verify whether the target can be reproduced. As a premise, since the reflection of a target depends on the shape, color, and material, the measured reflectance (BRDF) of the shape and paint should be applied. Comparison of direction, distance, detection probability, intensity, number of detection points, and size with actual measurement and simulation with a stationary target.	Vehicles set for basic verification (passenger car + large vehicle)	[Point cloud data] Direction, Distance, Reflectance (BRDF) Shape	Number of detection points	The number of vehicle detection points can be detected in the same way as in the actual environment.	F.2.3.2.2.1	O			O	O	O	O	O	O				O	O	There is no item marked with O because it is a reflection from outside the target and does not affect the appearance to the recognition target. .				
						Size	The size of the vehicle can be detected in the same way as in the actual environment.	F.2.3.2.2.2	O			O	O	O	O	O	O			O	O						
		Dynamic verification (CCRs)	Approach a stationary vehicle and compare changes in direction, distance, number of detection points, and size over time by actual measurement and simulation.	Vehicles set for basic verification (passenger car + large vehicle)	[Point cloud data] Temporal changes in direction and distance	Distance	Being able to detect changes in vehicle distance in the same way as in the actual environment	F.2.3.2.2.3	O			O	O	O	O	O	O	O	O	O	O	O	O				
						Number of detection points	The change in the number of detected points of the vehicle can be detected in the same way as in the actual environment	F.2.3.2.2.3	O			O	O	O	O	O	O			O	O	O	O				
						Size	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.2.3	O			O	O	O	O	O	O	O		O	O	O	O				
	common requirement	Cut-in (Standard-sized car)	Comparison of changes in time direction, distance, number of detection points, and size of vehicles that have been cut-in by actual measurement and simulation	Vehicles set for basic verification (passenger car + large vehicle)	[Point cloud data] Temporal changes in direction and distance	Distance	Being able to detect changes in vehicle distance in the same way as in the actual environment	F.2.3.2.3.1	O																		
						Number of detection points	The change in the number of detected points of the vehicle can be detected in the same way as in the actual environment	F.2.3.2.3.1	O																		
						Size	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.3.1	O																		
						[object]	Being able to detect changes in vehicle distance in the same way as in the actual environment	F.2.3.2.3.2	O																		
						Distance	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.3.2	O																		
		Basic traffic flow scenario	Comparison of changes in time direction, distance, number of detection points, and size of vehicles that have been cut-in by actual measurement and simulation	Vehicle with a long vehicle length	[Point cloud data] Temporal changes in direction and distance	Distance	Being able to detect changes in vehicle distance in the same way as in the actual environment	F.2.3.2.3.3	O																		
						Number of detection points	The change in the number of detected points of the vehicle can be detected in the same way as in the actual environment	F.2.3.2.3.3	O																		
						Size	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.3.3	O		All items to be confirmed are applicable.																
						[object]	Being able to detect changes in vehicle distance in the same way as in the actual environment	F.2.3.2.3.4	O																		
						Distance	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.3.4	O																		
	Cut-out	After the preceding vehicle cuts out, approach the stopped vehicle and compare the changes in the temporal direction, distance, number of detection points, and size by actual measurement and simulation.	Vehicles set for basic verification (white Prius, etc. Both preceding and stopped vehicles)	[Point cloud data] Temporal changes in direction and distance		Distance	Being able to detect changes in vehicle distance in the same way as in the actual environment	F.2.3.2.3.5	O																		
						Number of detection points	The change in the number of detected points of the vehicle can be detected in the same way as in the actual environment	F.2.3.2.3.5	O																		
						Size	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.3.5	O																		
						[object]	Being able to detect changes in vehicle distance in the same way as in the actual environment	F.2.3.2.3.6	O																		
						Distance	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.3.6	O																		
						Size	Being able to detect changes in the size of the vehicle in the same way as in the actual environment	F.2.3.2.3.6	O																		

Table F-2. List of common requirement for LiDAR

F.2.2.3 the way of thinking about common requirement for Camera

Camera sensor is different about perception principle from Radar and LiDAR, those are active type sensors and Camera is passive sensor which does not use signal from the sensor and uses surrounding light information, so that possible information differ from those 2 active sensors(fig.F-6). Camera can use colour information while active type sensors can detect distance information and camera cannot detect it in perception block. This comes from perception principle, that camera sensor uses flat plate light detecting sensor, so that this characteristics is very important to validate reproducibility.

Understanding the Different between Perception Devices

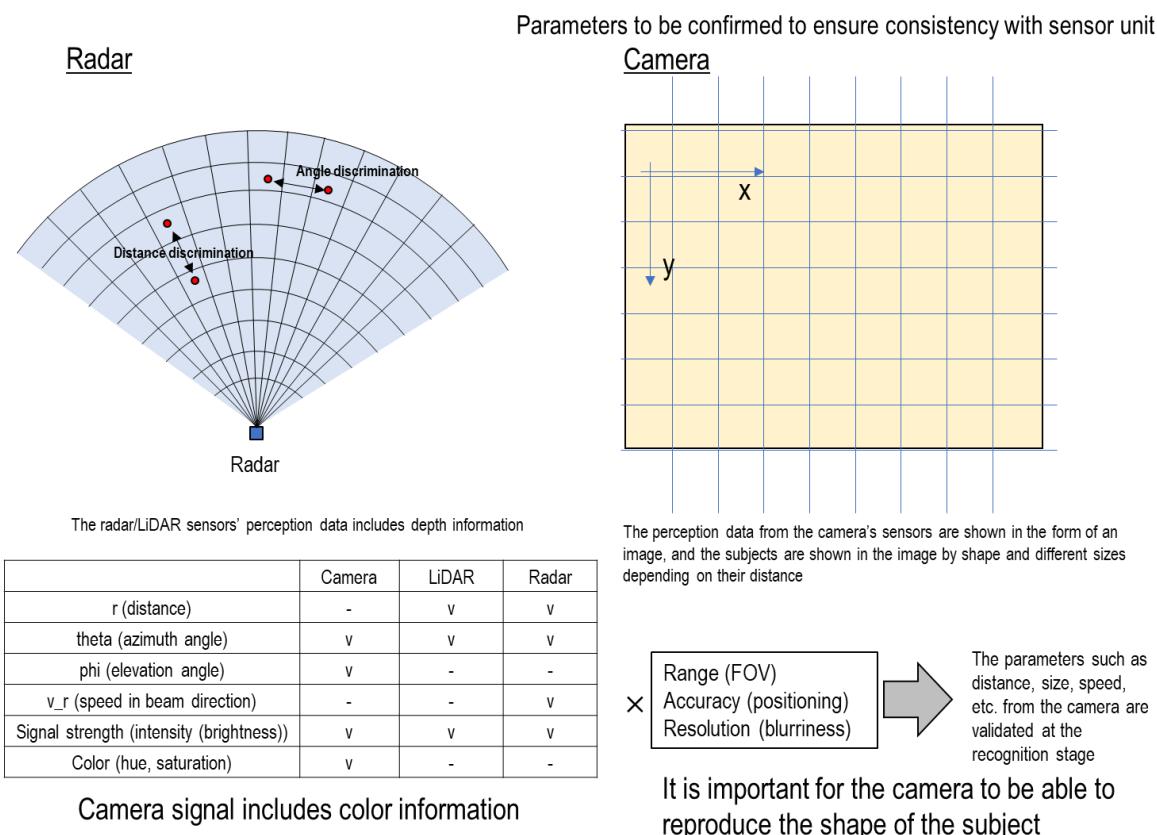


Fig.F-6 comparison between active and passive sensor(camera)

Considering these characteristics, common requirement of camera perception process shown in below(fig.F-7) is clarified.

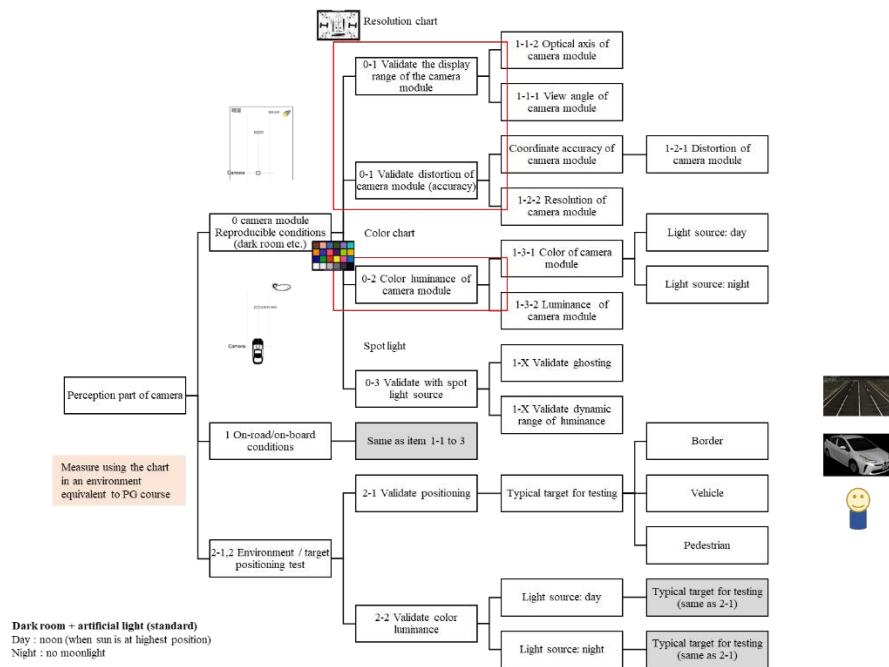


Fig.F-7 common requirement of camera perception process

Camera perception process will be validated about sensor itself, resolution/colour chart under on-vehicle condition, environment/target position reproducibility.

common requirement of camera recognition process shown in below(fig.F-8) is clarified.

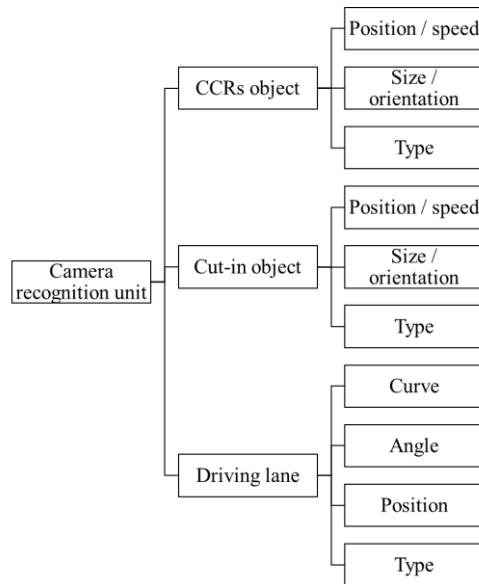


Fig.F-8. common requirement of camera recognition process

Based on this way of thinking, list of actual requirement shown in table.F-3 is clarified.

Verification	Items/ Target Parts	Measurement Items	Parameters	Requirement	Requirement ID	Normal Condition(Day)	Normal Condition(Night)	Perception Part								Recognition Part																									
								Optics				Imager				Image Processing		Feature extraction		Detection, Classification		Positioning		Tracking																	
								Blur	Blur Position shift,	Refraction	Reflection	Deformation	Vignetting	Scattering	Diffraction	Flare	Absorption	Noise	Color Filter	Exposure Time	Flicker	Exposure period	Time range for Exposure	OverExposure	UnderExposure	Crushed Shadows	Lack of Gradation	Brightness	Hue	Chroma	Low spatial frequency	Low contrast	No classification	Detection or classification error	Base position error	Target position error	Tracking error	Velocity error			
Stand-alone camera (Lens)(CMOS)	Adjusting the camera module	Angle of view / Optical axis / Distortion	Imaging range Image center position Distortion, focus	Using test chart, minimize the value gaps of evaluation parameters between RAW images captured by the real camera and created by virtual environment.	0-1	-	✓	◎	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-					
		Color Brilliance	luminance, hue, color	Using test chart, minimize the value gaps of evaluation parameters between RAW images captured by the real camera and created by virtual environment.	0-2	-	✓	○	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-				
		dynamic Range	photoelectric conversion characteristics	Measure photoelectric conversion characteristic of the real sensor. Then minimize the gap between the characteristics of real and virtual sensors.	0-3	-	✓	-	-	-	-	-	-	-	-	-	-	◎	○	○	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
Vehicle-mounted camera (installed Windshield)	Verification in front of the camera	optical axis (mounting position/direction)	image center position optical axis	Using real camera, minimize image position differences between targets placed at different distances on reference optical axis.	1-0	✓	-	(◎)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-				
		distortion	shape, size	Check distortion characteristics caused by WS. Boundary lines of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 2-1 or 2-2)	1-1	(✓)	-	(◎)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-				
		color luminance verification	luminance, hue, color	Check similarity on SxS points on whole image (RAW format) between real and virtual environment with WS under known lighting conditions. (except geometric view point) (Omnimake by substituting 2-2 or 2-5)	1-2	(✓)	-	-	-	-	-	-	-	-	-	-	-	(◎)	(◎)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-			
	(Headlight distribution for own car)	color luminance verification	luminance, hue, color	Check differences of evaluation parameter values between images of real and virtual environment at observation point.	1-3	-	✓	-	-	-	-	-	-	-	-	-	-	◎	○	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
Recognition target and environment (asset)	Fixed point verification (Pedestrians) (Passenger cars: Prius) (Passenger car: NCAP dummy car) (Large vehicles) (Boundaries: white line, solid line, dashed line)	placement verification (Vehicle)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-1)	2-1	(✓)	(✓)	◎	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-				
		color luminance verification(Vehicle)	luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-2)	2-2	(✓)	(✓)	-	○	-	-	-	-	-	-	-	-	-	◎	○	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
		Recognition result (Vehicle)	relative distance	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-3-1)	2-3-1	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	○	○	○	-	-	-	-	-	-	-			
	(Road surface: straight, asphalt)	size, direction		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-3-2)	2-3-2	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-			
		relative velocity		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-3-3)	2-3-3	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-			
		classification		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-3-4)	2-3-4	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	○	○	○	-	-	○	-	-	-	-	-			
Common Requirement	Placement verification(boundary line)	shape, size		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-4)	2-4	(✓)	(✓)	◎	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-			
		color luminance verification(boundary line)	luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-5)	2-5	(✓)	(✓)	-	○	-	-	-	-	-	-	-	-	-	○	○	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
		Recognition result (boundary line)	curvature	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-6-1)	2-6-1	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	○	-	-	○	-	-	-	-	-	-	-		
	(Surface: curved, asphalt) (Boundaries: white line, solid line, dashed line)	direction		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-6-2)	2-6-2	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
		lateral position		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-6-3)	2-6-3	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
		classification		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment. (Omnimake by substituting 3-6-4)	2-6-4	(✓)	(✓)	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
Combination of recognition target and environment (asset)	Low-speed movement verification (approach, separation) (Passenger car: Prius)	Placement verification (Vehicle)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	3-1	✓	✓	◎	○	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-		
		color luminance verification(Vehicle)	luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	3-2	✓	✓	-	○	-	-	-	-	-	-	-	-	-	-	○	○	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-
		Recognition result (Vehicle)	relative distance	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	3-3-1	✓	✓	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	○	○	○	-	-	-	-	-	-	-	-	-	-
	(Road surface: straight, asphalt) (Surface: curved, asphalt) (Boundaries: white line, solid line, dashed line)	size, direction		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	3-3-2	✓	✓	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	
		relative velocity		Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	3-3-3	✓	✓	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	

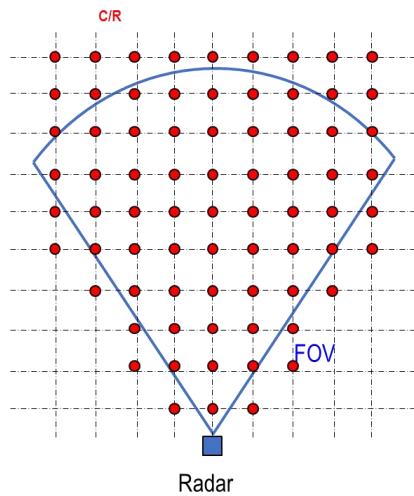
F.2.3 Validation method of common requirement

Validation method of each requirement for each sensor defined in section F.2.2 is shown in this section.

F.2.3.1 Validation method of common requirement of millimeter wave radar

■ Basic Characteristics of the Sensor: Detection Accuracy – Range/Orientation

mmWave Radar
(0-1)



Method of Validation

- Obtain the detected positions of the standard reflectors (C/R) located at each point in the FOV, using the radar (C/Rs to be moved one at a time)
- Compare the detected positions (range/orientation) between the actual and simulated environments



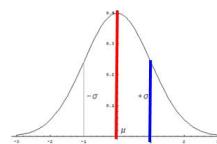
Trihedral Corner Reflector (C/R)
• RCS is recognized
• Retro reflective traits

Source: <https://www.everythingrf.com/community/what-is-a-corner-reflector>

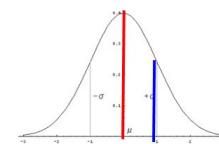
Judgment Criteria

Relative to the detection results (range/orientation) from the actual environment:

- Median: within 5%
Distribution (σ): within 10%



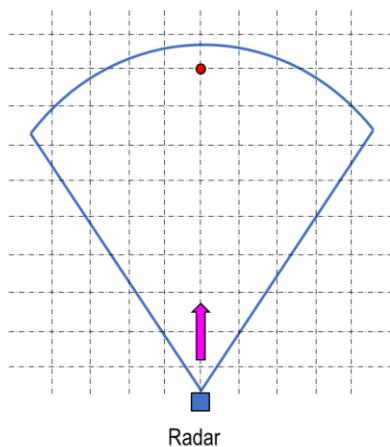
Data from Actual Environment



Data from SIM Environment

■ Basic Characteristics of the Sensor: Detection Accuracy – Relative Speed

mmWave Radar
(0-2)

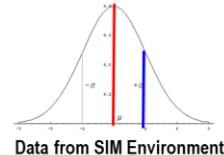
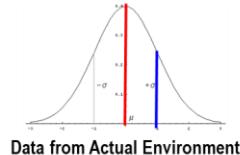
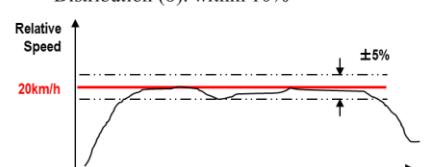


Method of Validation

- Place the C/R directly in front of the radar, and move the radar at a constant speed toward the C/R
- Compare the detected speed between the actual and simulated environments

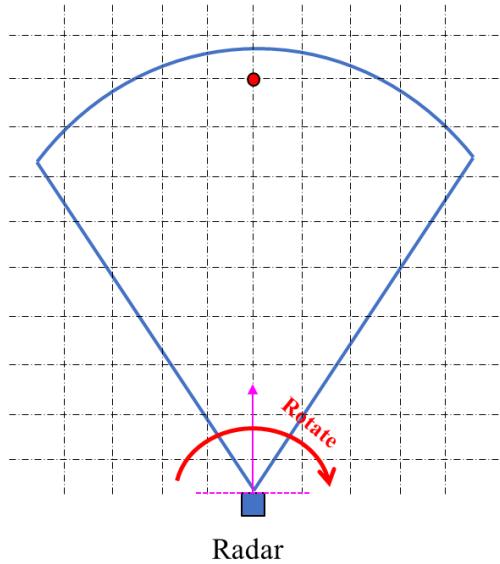
Judgment Criteria

Relative to the detection results from the actual environment:
Median: within 5%
Distribution (σ): within 10%



■ Basic Characteristics of the Sensor: Detection Accuracy - Received Power

mmWave Radar
(0-3)

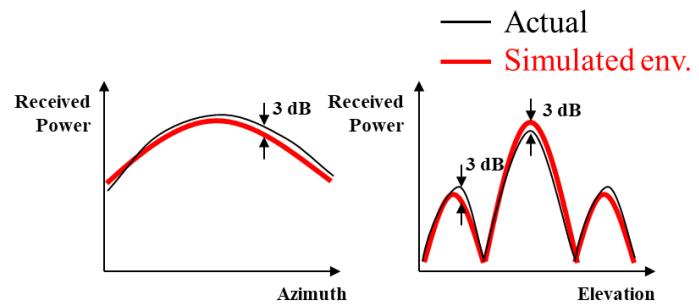


Method of Validation

- Place the C/R directly in front of the radar, and rotate the radar in the azimuth(horizontal) and elevation(vertical) angles, and measure the received power

Judgment Criteria

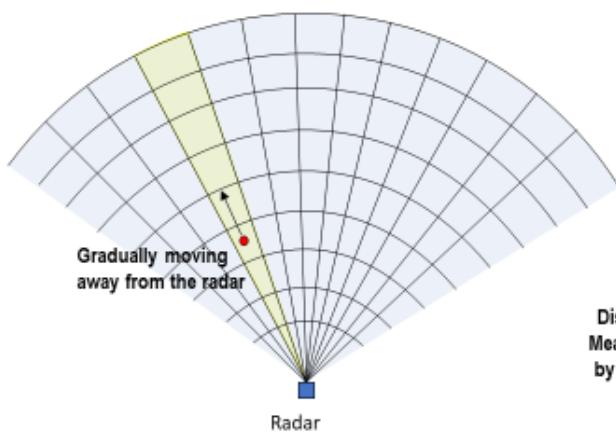
- Difference in the received power : 3 dB or less



■ Basic Characteristics of the Sensor: Resolution - Range/Orientation

mmWave radar
(0-4)

<Resolution Range>



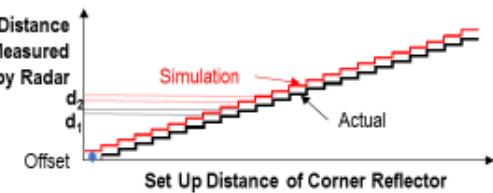
Method of Validation

➤ Resolution Range

Validate the resolution range by placing the C/R within the radar's FOV, and gradually varying the distance away from the radar along the normal vector.

Judgment Criteria

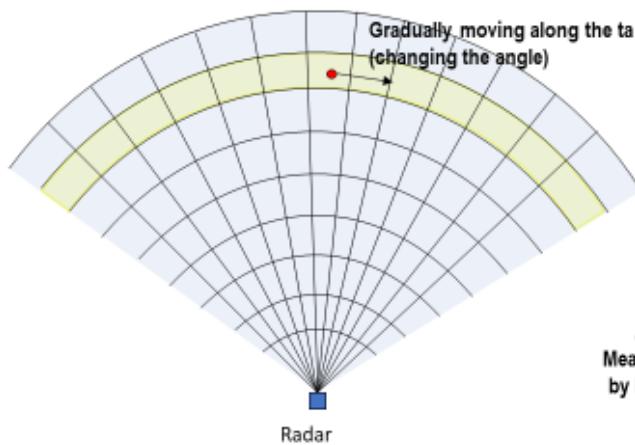
- ◆ The number of steps in the stairs derived by the relationship between the set-up distance of the corner reflector and the distance measured by the radar, is to be the same for both the actual and simulated environments
- ◆ The size of the steps d_1, d_2 shall not differ in size by more than 15% (provisional)



■ Basic Characteristics of the Sensor: Resolution - Range/Orientation

mmWave radar
(0-4)

<Azimuth Resolution >



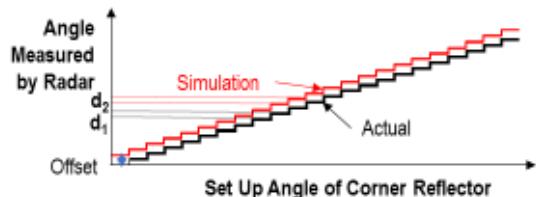
Method of Validation

➤ Azimuth Angle Resolution

Place the C/R within the radar's FOV and gradually move it in the tangential direction. Validate the azimuth angle resolution by comparing the angle measured by the radar to the set-up angle of the C/R.

Judgment Criteria

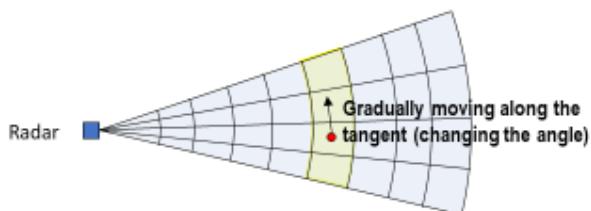
- ◆ The number of steps in the stairs derived by the relationship between the angle measured by the radar and the set-up angle of the C/R, is to be the same for both the actual and simulated environments.
- ◆ The size of the steps d_1, d_2 shall not differ in size by more than 15% (provisional)



■ Basic Characteristics of the Sensor: Resolution - Range/Orientation

mmWave radar
(0-4)

<Elevation Resolution >



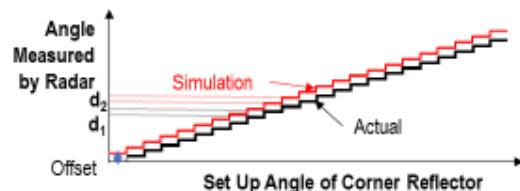
Method of Validation

➤ Elevation Angle Resolution

Place the C/R within the radar's FOV and gradually move it in the tangential direction. Validate the elevation angle resolution by comparing the angle measured by the radar to the set-up angle of the C/R.

Judgment Criteria

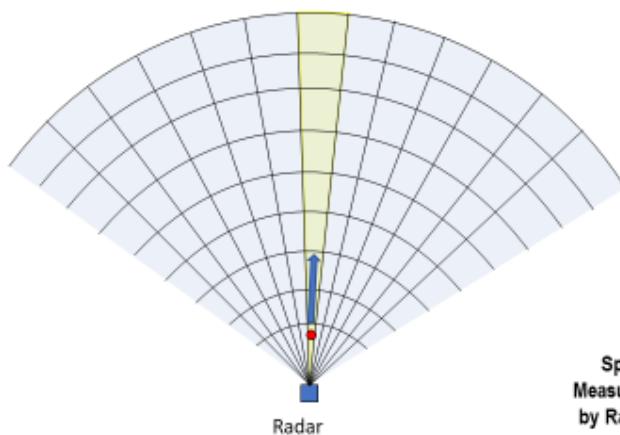
- ◆ The number of steps in the stairs derived by the relationship between the angle measured by the radar and the set-up angle of the C/R, is to be the same for both the actual and simulated environments.
- ◆ The size of the steps d_1, d_2 shall not differ in size by more than 15% (provisional)



■ Basic Characteristics of the Sensor:
Resolution - Relative Speed

mmWave radar
(0-5)

<Speed Resolution>



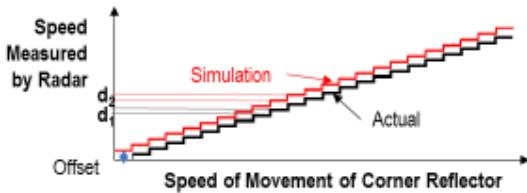
Method of Validation

- Speed Resolution

Place the C/R within the radar's FOV, and evaluate multiple times by keeping the speed of the reflector constant, and then changing the speed.

Judgment Criteria

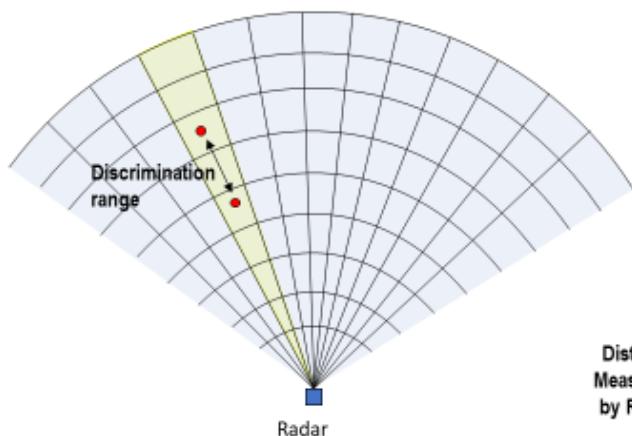
- ◆ The number of steps in the stairs derived by the relationship between the speed measured by the radar and the moving speed of the C/R, is to be the same for both the actual and simulated environments.
- ◆ The size of the steps d_1, d_2 shall not differ in size by more than 15% (provisional)



■ Basic Characteristics of the Sensor:
Discrimination - Range/Orientation

mmWave radar
(0-6)

<Discrimination Range>



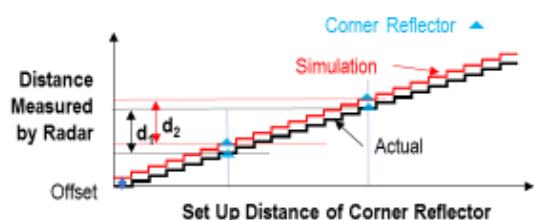
Method of Validation

- Resolution Range

Place 2 C/R's (of same spec) within the radar's FOV in close proximity to each other and validate the discrimination of each target. Then change the distance between the 2 C/R's and take further multiple measurements.

Judgment Criteria

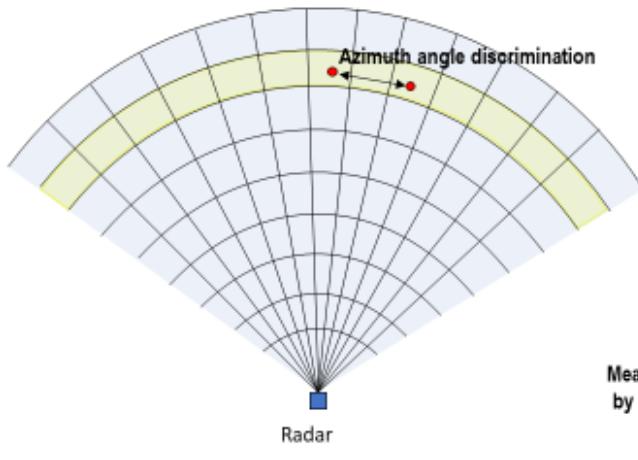
The calculated average distance of discrimination based on multiple measurements for the actual (d_1) and simulated (d_2) environments shall not differ in size by more than 15% (provisional)



■ Basic Characteristics of the Sensor:
Resolution - Relative Speed

mmWave radar
(0-6)

<Azimuth Discrimination>



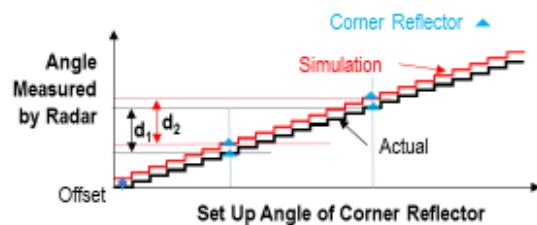
Method of Validation

➤ Azimuth Angle Resolution

Place 2 C/R's (of same spec) within the radar's FOV at the same distance and in close lateral proximity to each other. Validate the discrimination of each target. Then change the set-up angle of the 2 C/R's and take further multiple measurements.

Judgment Criteria

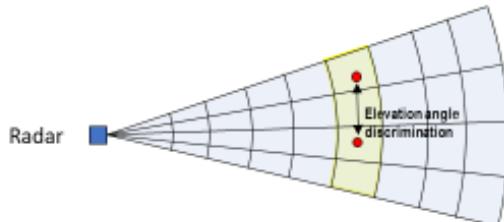
The calculated average angle of discrimination based on multiple measurements for the actual (d_1) and simulated (d_2) environments shall not differ in size by more than 15% (provisional)



■ Basic Characteristics of the Sensor:
Resolution - Relative Speed

mmWave radar
(0-6)

<Elevation Discrimination>



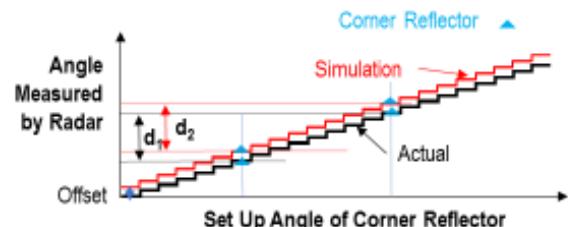
Method of Validation

➤ Elevation Angle Resolution

Place 2 C/R's (of same spec) within the radar's FOV at the same distance and in close vertical proximity to each other. Validate the discrimination of each target. Then change the set-up angle of the 2 C/R's and take further multiple measurements.

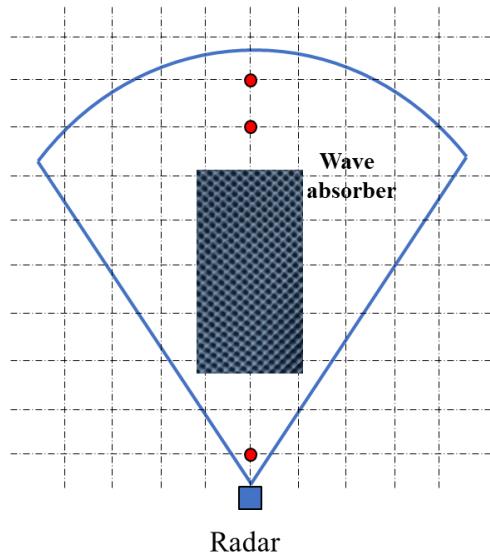
Judgment Criteria

The calculated average angle of discrimination based on multiple measurements for the actual (d_1) and simulated (d_2) environments shall not differ in size by more than 15% (provisional)



■ Properties of Radio Wave Propagation: Free-Space - Received Power

mmWave Radar
(0-7)

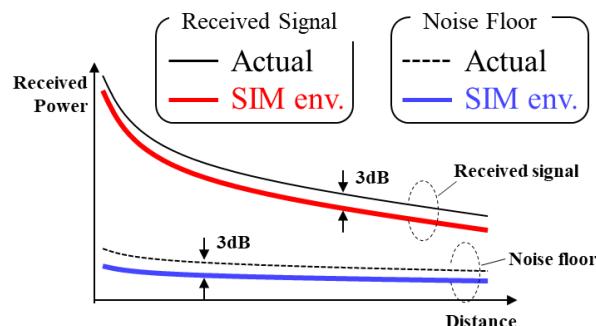


Method of Validation

- Place the C/R directly in front of the radar, then vary the set-up distance of the C/R and measure the received power at each distance
(in order to eliminate the road surface reflection waves, set-up a wave absorber around the point where the road reflects)

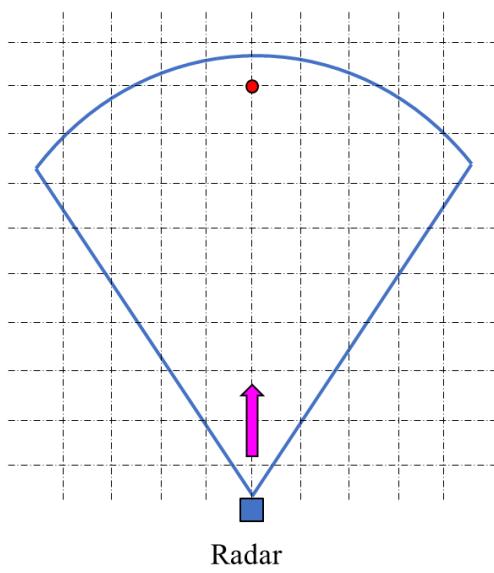
Judgment Criteria

- Difference in received power: 3 dB or less
- Difference in power of noise floor: 3 dB or less
- Difference in SNR: 3 dB or less



■ Properties of Radio Wave Propagation: Road Surface - Received Power

mmWave Radar
(0-8)

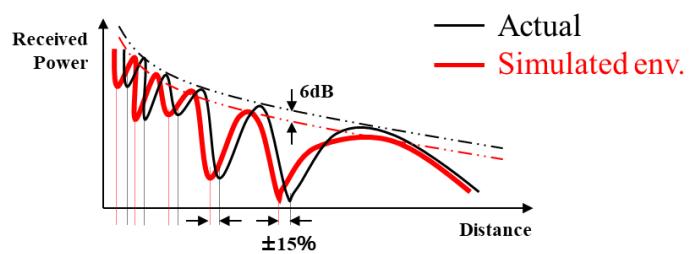


Method of Validation

- Place the C/R directly in front of the radar, and move the radar at a constant speed toward the C/R

Judgment Criteria

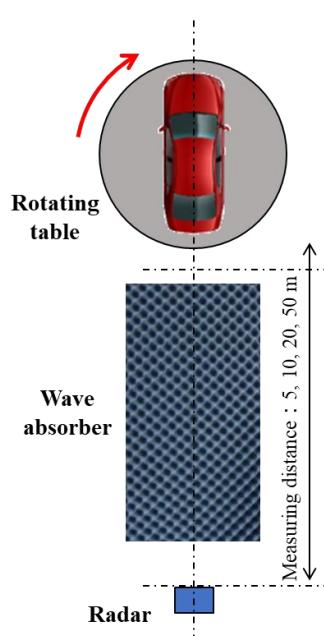
- Difference in envelope: 6 dB or less
- Difference in null point distance: $\pm 15\%$ or less



■ Reflective Properties of the Recognition Target:

mmWave Radar
(1-1)

Vehicle (Passenger Vehicle/Large-Sized Vehicle) RCS

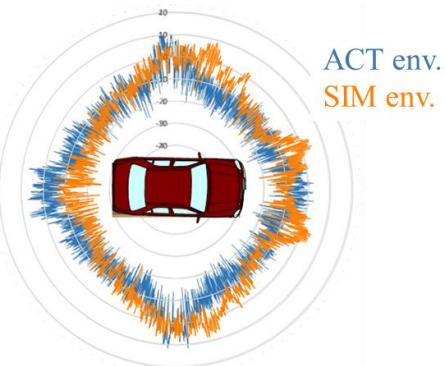


Method of Validation

- Direct the radio wave toward the vehicle on the rotating table.
- Rotate the passenger vehicle and plot the change in received power by the rotated angle.
- Measuring distance: 5, 10, 20, 50 (m)
- Type of vehicle: passenger vehicle, large-sized trailer
- Compare the received power between the actual and simulated environments

Judgment Criteria

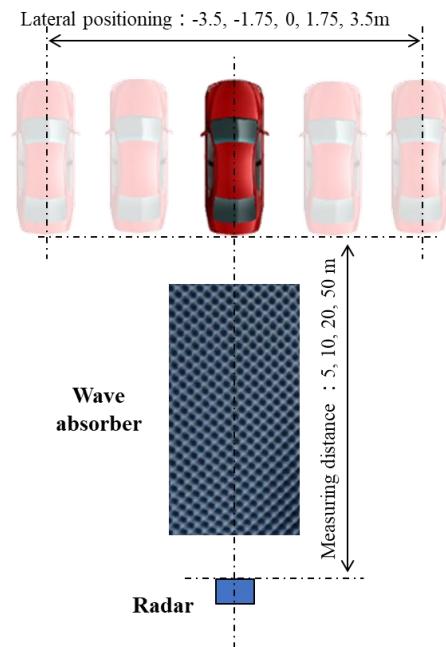
- Difference in received power: 3 dB or less in all directions



■ Reflective Properties of the Recognition Target:

mmWave Radar
(1-2)

Vehicle (Passenger Vehicle/Large-Sized Vehicle) Reflection Points

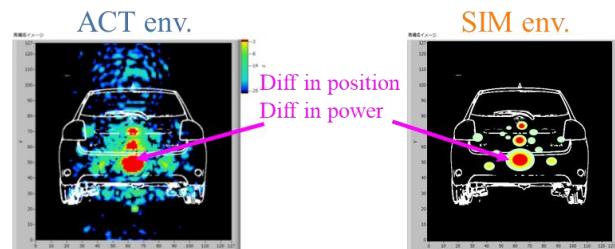


Method of Validation

- Direct the radio wave toward the vehicle from behind, and generate a radar image.
- Measuring distance: 5, 10, 20, 50 (m)
- Type of vehicle: passenger vehicle, large-sized trailer
- Lateral position of vehicle: 0, ± 1.75 , ± 3.5 (m)
(to simulate “in ego vehicle lane”, “on lane marking” and “in adjacent lane”)
- Compare the distributions of reflection intensity between the actual and simulated environment.

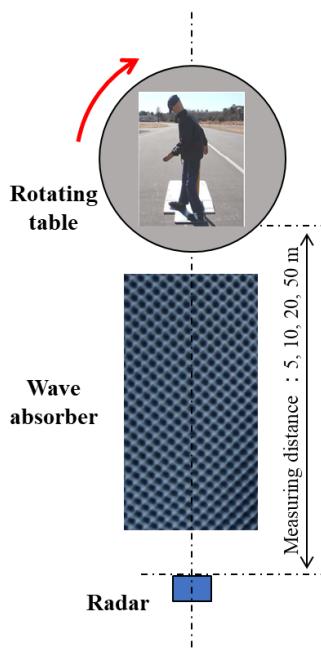
Judgment Criteria

- Difference in peak point of reflection intensity: within 3° of viewing angle
- Difference in received power at peak point: 3 dB or less



■ Reflective Properties of the Recognition Target: Pedestrian RCS

mmWave Radar
(1-3)

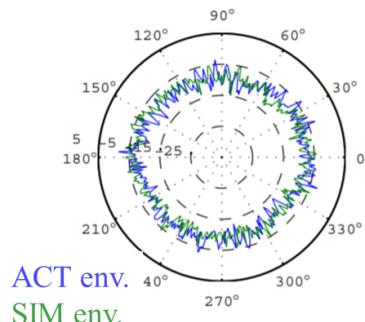


Method of Validation

- Direct the radio wave toward the pedestrian dummy on the rotating table.
- Rotate the dummy and plot the change in the received power by the rotated angle.
- Measuring distance: 5, 10, 20, 50 (m)
- Compare the received power between the actual and simulated environments.

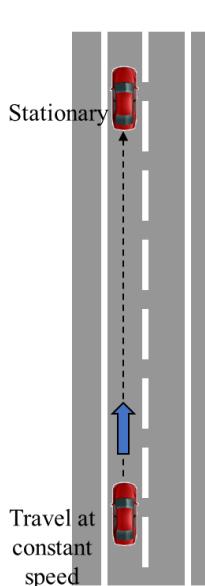
Judgment Criteria

- Difference in received power: 3 dB or less in all directions



■ Basic Traffic Flow Scenario: CCRs Received Power

mmWave Radar
(2-1)

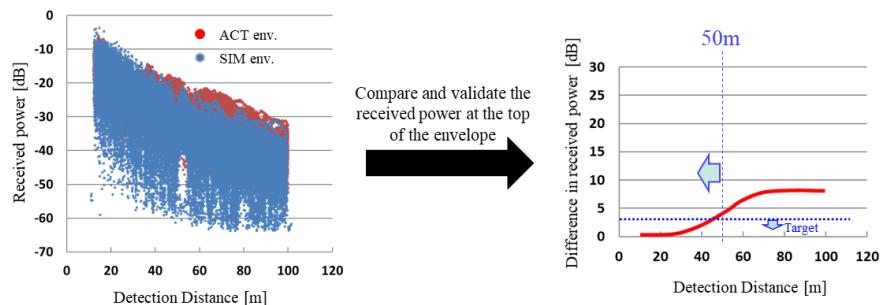


Method of Validation

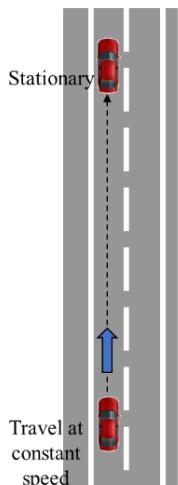
- Simulate NCAP CCRs scenario
- Speed of ego vehicle: 2 points between 5 and 60 km/h (High/Low)
- Measure the received power from the stationary target vehicle ahead and plot the envelope line at the top of the curve.
- Compare the change in envelope lines between the actual and simulated environments.

Judgment Criteria

- Difference in the envelope line of the received power at a relative distance of 50 m and below: 3 dB or less



■ Basic Traffic Flow Scenario: CCRs Detecting Position (2-2)

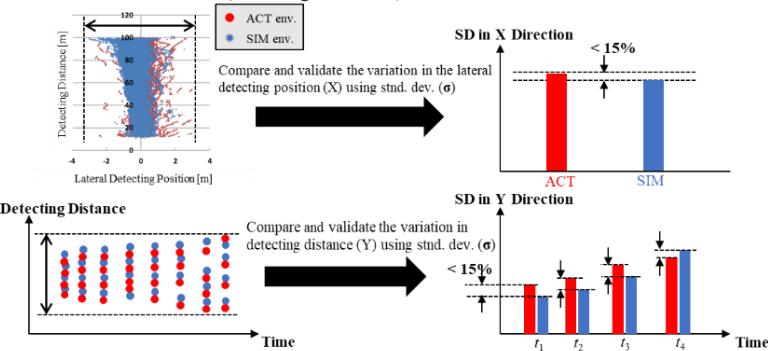


Method of Validation

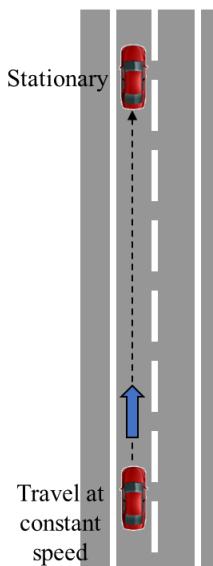
- Simulate NCAP CCRs scenario
- Speed of ego vehicle: 2 points between 5 and 60 km/h (High/Low)
- Measure the position (X, Y) at which the reflection of the stationary target vehicle ahead is detected
- Compare the degree of variation in detecting positions between the actual and simulated environments

Judgment Criteria

- Difference in standard deviation at a relative distance of 50 m and below :
 - 15% or less in X (lateral direction)
 - 15% or less in Y (traveling direction)



■ Basic Traffic Flow Scenario: CCRs Detecting Speed (2-3)

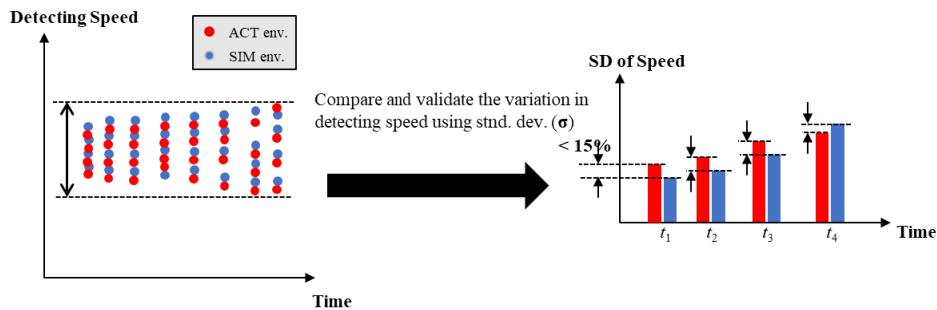


Method of Validation

- Simulate NCAP CCRs scenario
- Speed of ego vehicle: 2 points between 5 and 60 km/h (High/Low)
- Measure the speed at which the reflection of the stationary target vehicle ahead is detected (Y direction)
- Compare the degree of variation in detecting speeds between the actual and simulated environments

Judgment Criteria

- Difference in standard deviation at a relative distance of 50 m and below:
 - 15% or less



■ Basic Traffic Flow Scenario: CCRs

mmWave Radar
(2-4)

Object Detecting Position (Distance/Angle)

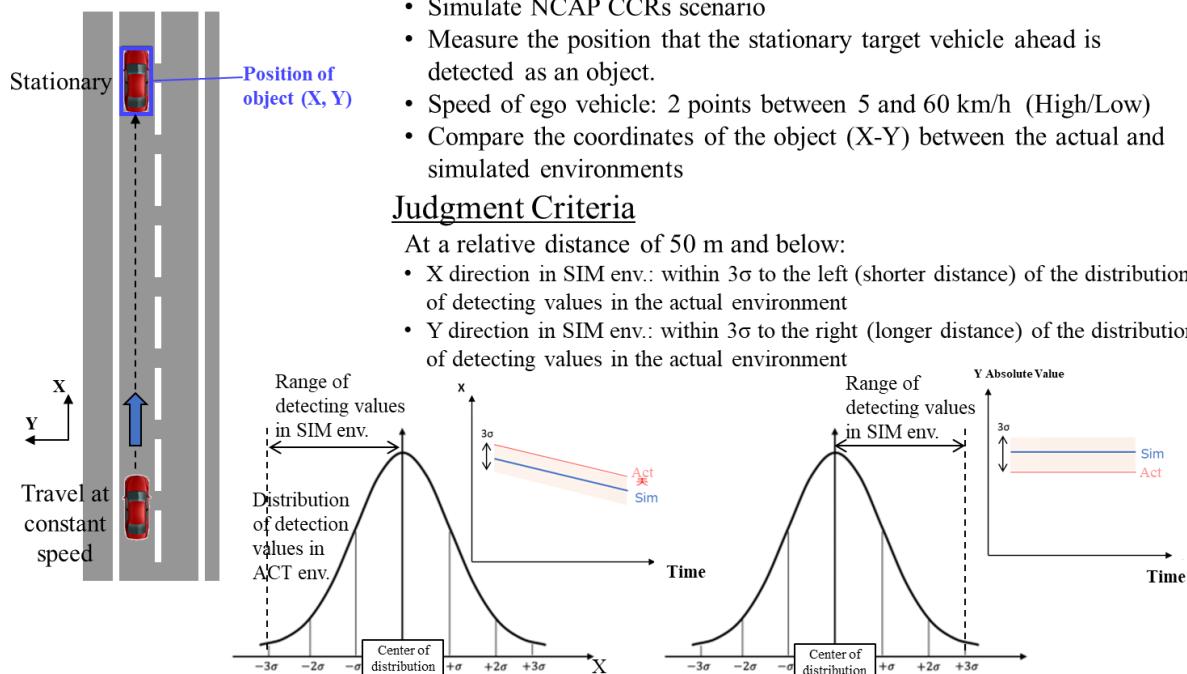
Method of Validation

- Simulate NCAP CCRs scenario
- Measure the position that the stationary target vehicle ahead is detected as an object.
- Speed of ego vehicle: 2 points between 5 and 60 km/h (High/Low)
- Compare the coordinates of the object (X-Y) between the actual and simulated environments

Judgment Criteria

At a relative distance of 50 m and below:

- X direction in SIM env.: within 3σ to the left (shorter distance) of the distribution of detecting values in the actual environment
- Y direction in SIM env.: within 3σ to the right (longer distance) of the distribution of detecting values in the actual environment



■ Basic Traffic Flow Scenario: CCRs

mmWave Radar
(2-5)

Object Detecting Speed

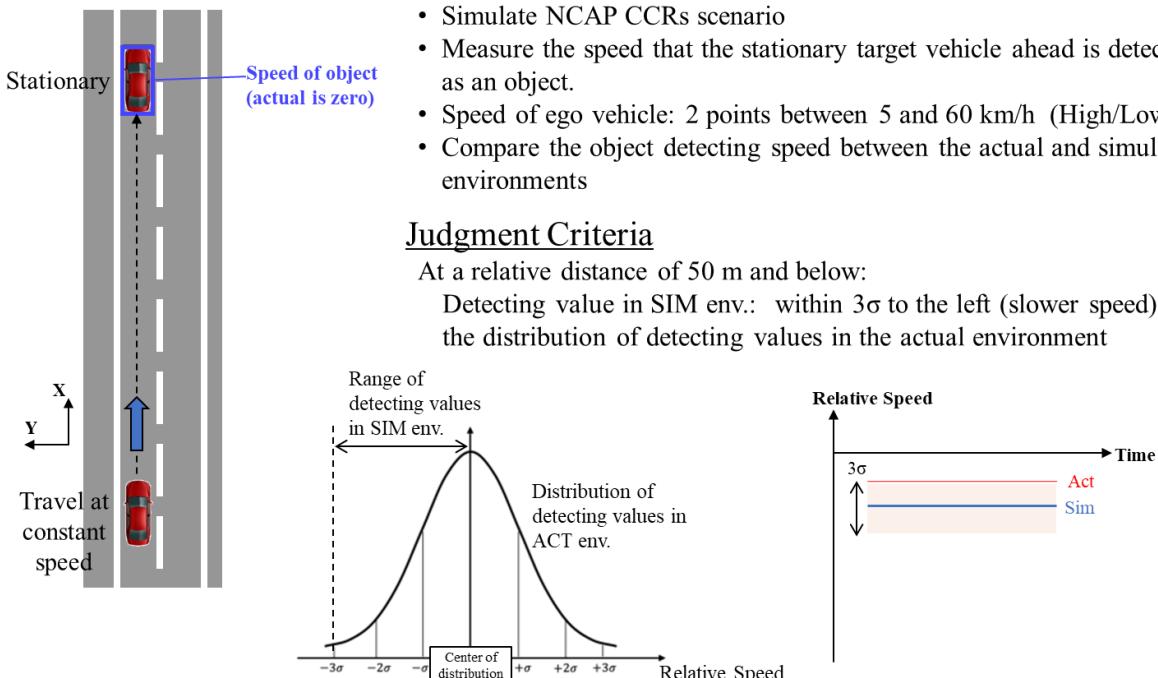
Method of Validation

- Simulate NCAP CCRs scenario
- Measure the speed that the stationary target vehicle ahead is detected as an object.
- Speed of ego vehicle: 2 points between 5 and 60 km/h (High/Low)
- Compare the object detecting speed between the actual and simulated environments

Judgment Criteria

At a relative distance of 50 m and below:

Detecting value in SIM env.: within 3σ to the left (slower speed) of the distribution of detecting values in the actual environment



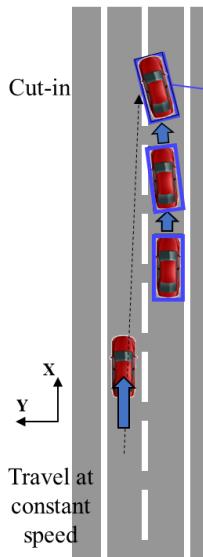
■ Basic Traffic Flow Scenario: Cut-in

mmWave Radar
(2-6)

Object Detecting Position (Distance/Angle)

Method of Validation

- Plot the position at which the target vehicle is detected as an object, in the scenario whereby the vehicle cuts-in in front of the ego vehicle.
- Speed of ego vehicle: constant speed (e.g. 60 km/h)
- Target vehicle speed: constant in traveling direction (e.g. 40 km/h),
3 points between 0.2 and 2.0 m/s in lateral direction
- Type of vehicle: passenger vehicle, large-sized trailer
- Compare the detecting position between the actual and simulated environments

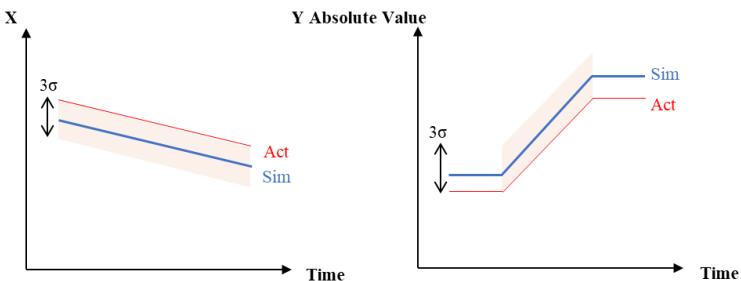


Object position
(X, Y)

Judgment Criteria

At a relative distance of 50 m and below:

- X direction in SIM env.: within 3σ to the left (shorter distance) of the distribution of detecting values in the actual environment
- Y direction in SIM env.: within 3σ to the right (longer distance) of the distribution of detecting values in the actual environment



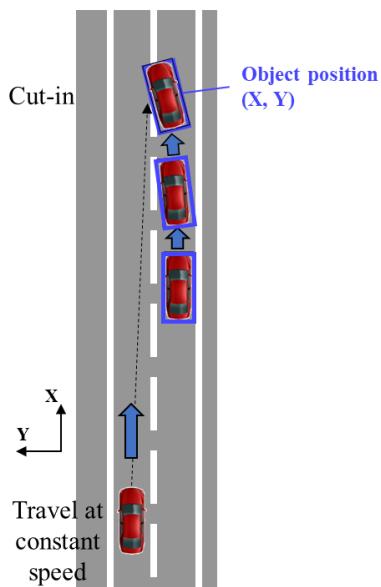
■ Basic Traffic Flow Scenario: Cut-in

mmWave Radar
(2-7)

Object Detecting Speed

Method of Validation

- Plot the speed at which the target vehicle is detected as an object, in the scenario whereby the vehicle cuts-in in front of the ego vehicle.
- Speed of ego vehicle: constant speed (e.g. 60 km/h)
- Object vehicle speed: constant in traveling direction (e.g. 40 km/h),
3 points between 0.2 and 2.0 m/s in lateral direction
- Type of vehicle: passenger vehicle, large-sized trailer
- Compare the detecting speed between the actual and simulated environments

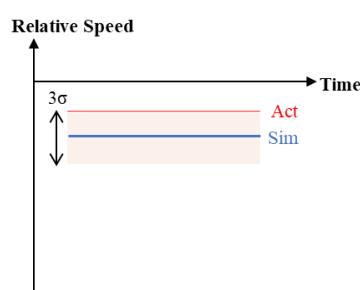


Object position
(X, Y)

Judgment Criteria

At a relative distance of 50 m and below:

Detecting value in SIM env.: within 3σ to the left (slower speed) of the distribution of detecting values in the actual environment

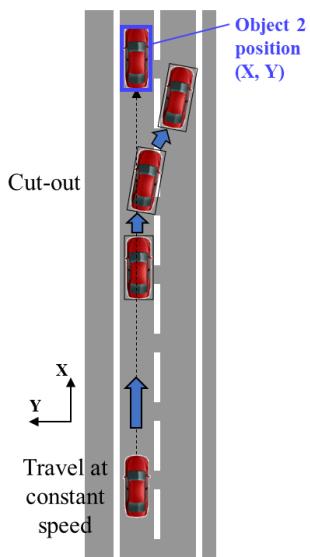


■ Basic Traffic Flow Scenario: Cut-out

mmWave Radar
(2-8)

Object Detecting Position (Distance/Angle)

Method of Validation

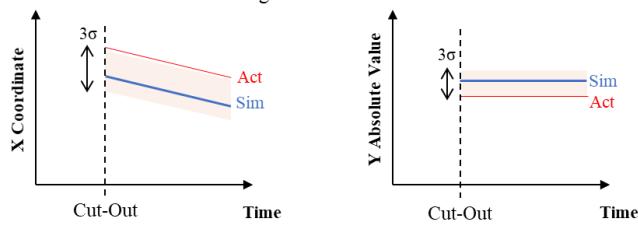


- Plot the position at which the stationary vehicle (object 2) is detected as an object, in the scenario whereby there is a stationary vehicle in front of the preceding vehicle which cuts-out
- Speed of ego vehicle: constant speed (e.g. 60 km/h)
- Object 1 vehicle speed: constant in traveling direction (e.g. 40 km/h), nearly 1.0 m/s in lateral direction
- Object 1 type: passenger vehicle (vehicle which cuts out)
- Object 2 type: passenger vehicle (stationary vehicle)
- Compare the position at which object 2 is detected between the actual and simulated environments

Judgment Criteria

Detecting position after cutting out:

- X direction in SIM env. : within 3σ to the left (shorter distance) of the distribution of detecting values in the actual environment
- Y direction in SIM env. : within 3σ to the right (longer distance) of the distribution of detecting values in the actual environment

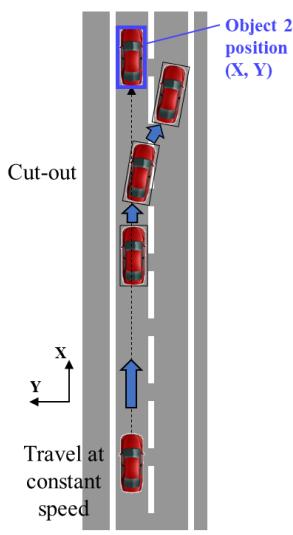


■ Basic Traffic Flow Scenario: Cut-out

mmWave Radar
(2-9)

Object Detecting Speed

Method of Validation

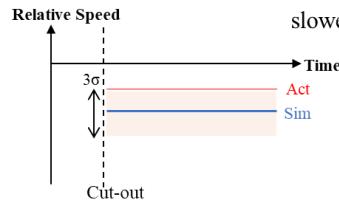


- Plot the speed at which the stationary vehicle (object 2) is detected as an object, in the scenario whereby there is a stationary vehicle in front of the preceding vehicle which cuts-out
- Speed of ego vehicle: constant speed (e.g. 60 km/h)
- Object 1 vehicle speed: constant in traveling direction (e.g. 40 km/h), nearly 1.0 m/s in lateral direction
- Object 1 type: passenger vehicle (vehicle which cuts out)
- Object 2 type: passenger vehicle (stationary vehicle)
- Compare the speed at which object 2 is detected between the actual and simulated environments

Judgment Criteria

After cutting out:

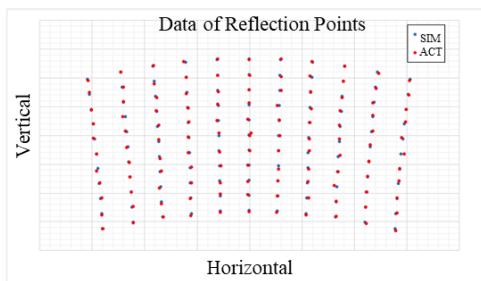
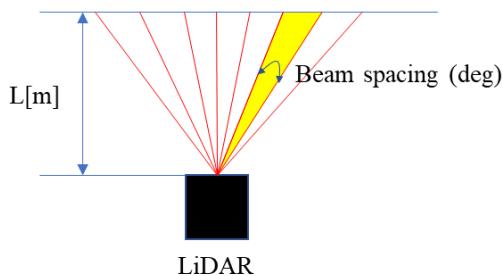
Detecting value in SIM env.: within 3σ of the distribution of detecting values in an actual environment (toward slower speed)



F.2.3.2 Validation method of common requirement of LiDAR

■ Basic Traits of the Sensor : Azimuth

LiDAR
(F.2.3.2.1)



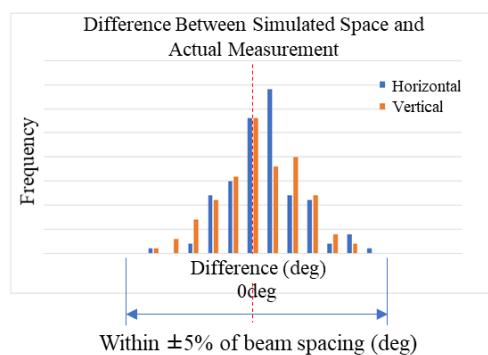
Method of Validation

Transmit the beams from the LiDAR to the wall positioned at a recognized distance (L), and measure the coordinates values of beam spot on the wall.

Confirm that the difference between the simulated and actual measurements are within an acceptable range.

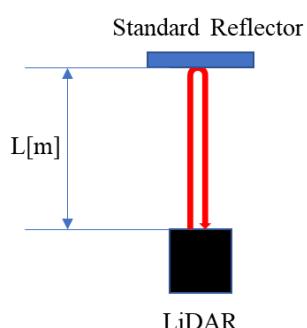
Judgment Criteria

E.g.) within $\pm 5\%$ of beam spacing (deg)



■ Basic Traits of the Sensor : Range

LiDAR
(F.2.3.2.1.1)



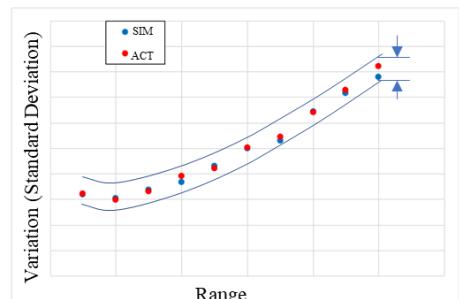
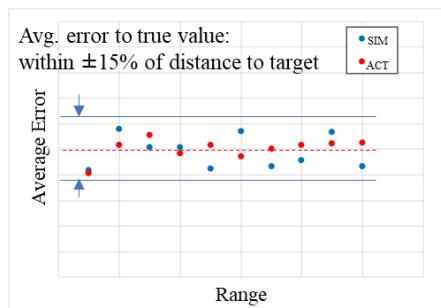
Method of Validation

Vary the distance between the LiDAR and the standard reflector and measure the distance error and variation. Confirm that they are within an acceptable range.

Judgment Criteria

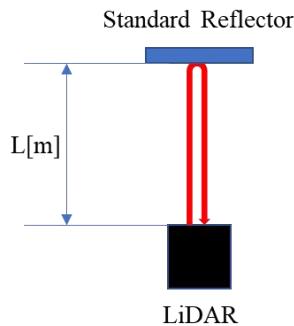
E.g.) Avg. error to true value: within $\pm 15\%$ of distance to target
Variation difference to actual measurement: within $\pm 15\%$

Variation difference to actual measurement: within $\pm 15\%$



■ Basic Traits of the Sensor : Strength / Detection rate

LiDAR
(F.2.3.2.1.2)
(F.2.3.2.1.3)

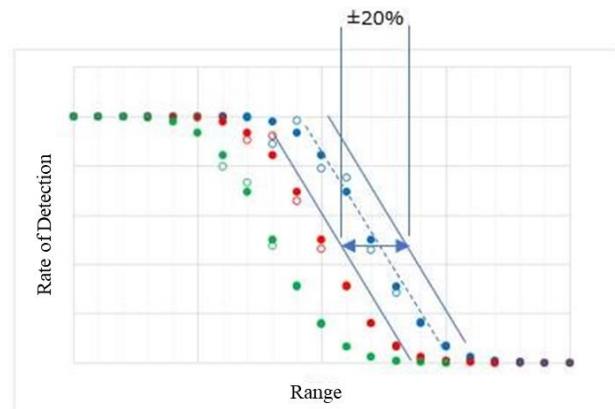


Method of Validation

Vary the distance between the LiDAR and the standard reflector, and measure the strength of reception and rate of detection. Confirm that they are within an acceptable range.

Judgment Criteria

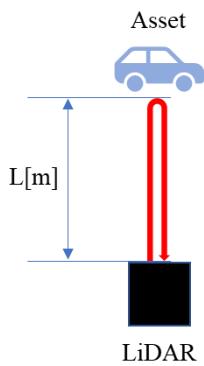
E.g.) Intensity error in relation to actual measurement value: within $\pm 20\%$



※Blue: reflectivity xx% Red: reflectivity xx% Green: reflectivity xx%
※ ●Actual ○Simulation

■ Reflective Traits of the Recognition Target : Size

LiDAR
(F.2.3.2.2.2)

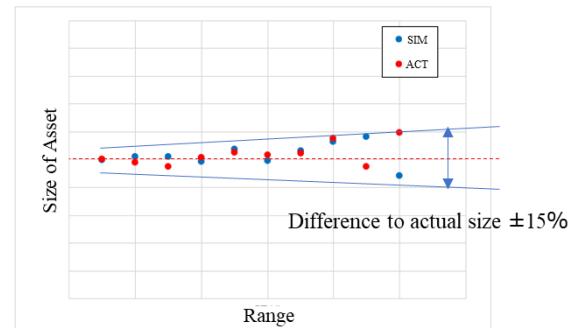
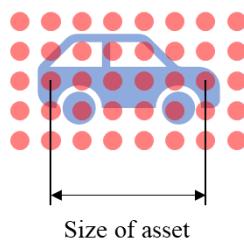


Method of Validation

Vary the distance between the LiDAR and the asset and measure the size of asset. Confirm that the difference in size is within an acceptable range.

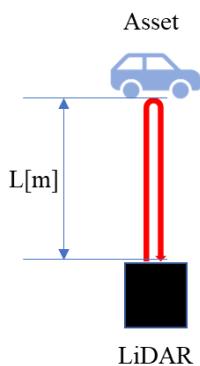
Judgment Criteria

E.g.) Difference to actual size within 15%



■ Reflective Traits of the Recognition Target : Number of Detection Points

LiDAR
(F.2.3.2.2.1)

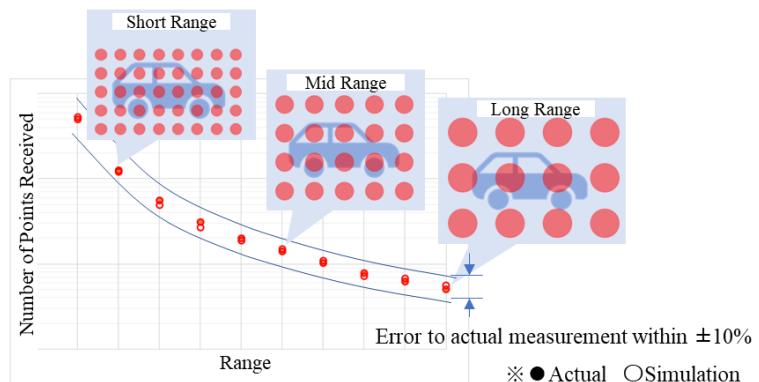


Method of Validation

Vary the distance between the LiDAR and the asset and measure.
Confirm that the difference in the number of detected points is within an acceptable range.

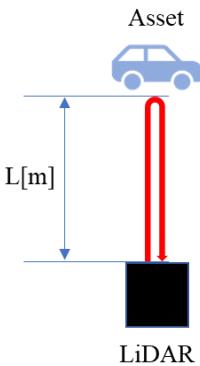
Judgment Criteria

E.g.) Error to number of actual points to be within $\pm 15\%$
(do not include large distances where the number of detected points reduces)



■ Reflective Traits of the Recognition Target : Size

LiDAR
(F.2.3.2.2.2)

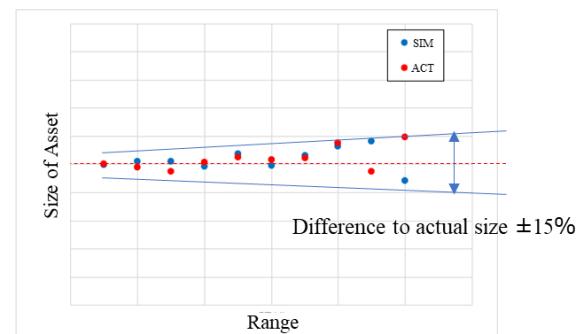
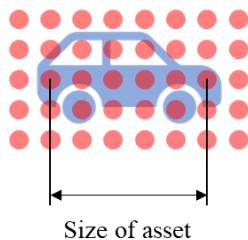


Method of Validation

Vary the distance between the LiDAR and the asset and measure the size of asset.
Confirm that the difference in size is within an acceptable range.

Judgment Criteria

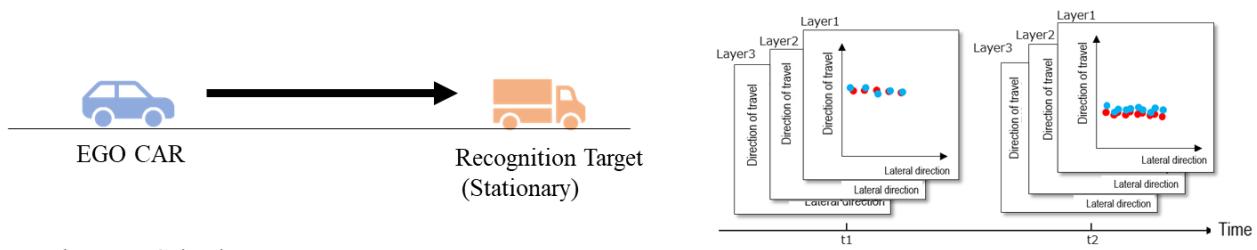
E.g.) Difference to actual size within 15%



■ Reflective Traits of the Recognition Target : Dynamic Validation [Pointcloud Data]

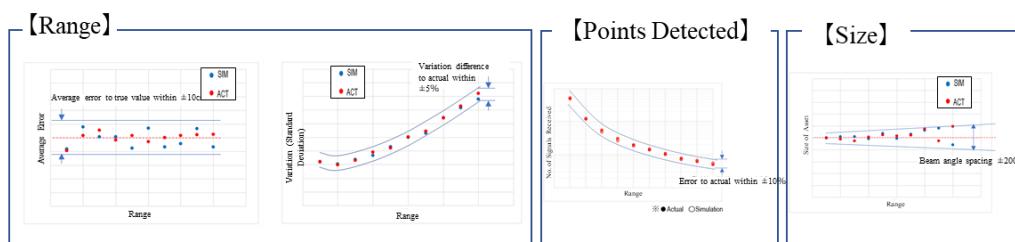
Method of Validation

Approach the stationary vehicle, and compare the change in azimuth, range, number of detection points, and size over time between the actual and the simulation. (E.g. approach the recognition target at 40 km/h, and apply the brakes when distance to the vehicle is 20 m and come to a stop)



Judgment Criteria

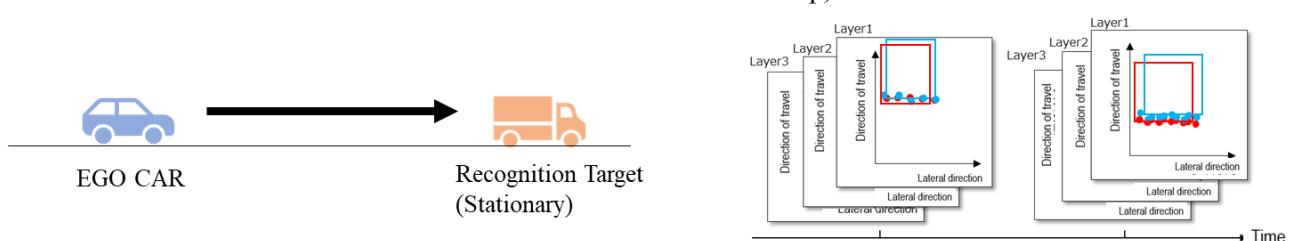
E.g.) The range, number of detection points and size over time, are to satisfy previously mentioned criteria (4.2, 4.4, 4.5)



■ Reflective Traits of the Recognition Target : Dynamic Validation [Object]

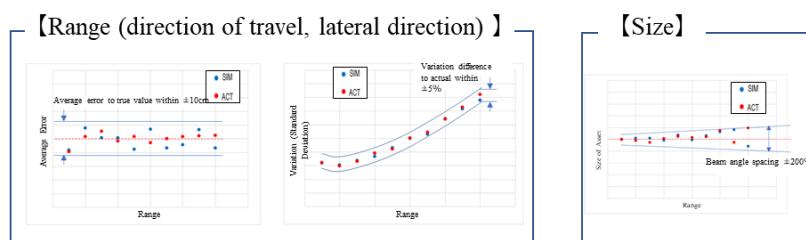
Method of Validation

Approach the stationary vehicle, and compare the change in the object's range and size over time between the actual and the simulation. (E.g. approach the recognition target at 40 km/h, and apply the brakes when distance to the vehicle is 20 m and come to a stop).



Judgment Criteria

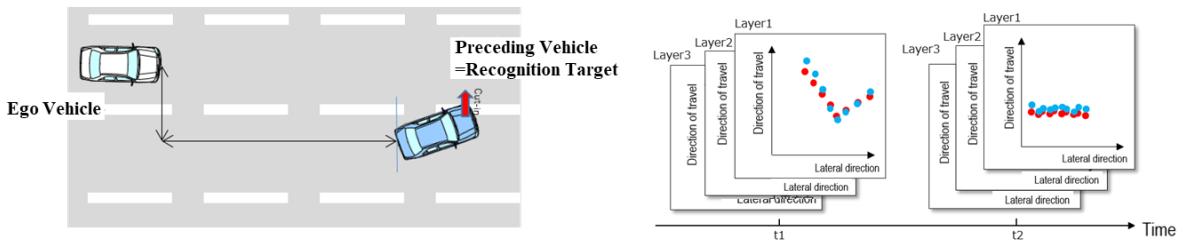
E.g.) The range (in the direction of travel and lateral direction) and size over time, are to satisfy previously mentioned criteria (4.2, 4.5)



■ Basic Traffic Flow Scenario : Cut-In Scenario (Normal Vehicle) 【 Pointcloud Data 】

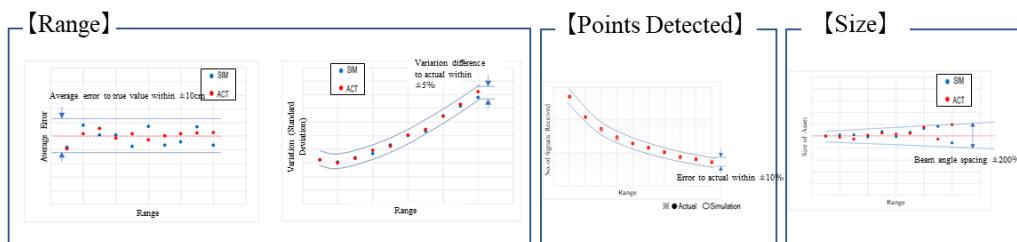
Method of Validation

Compare the change in azimuth, range, number of detection points and size over time between the actual and the simulation, for when a normal vehicle cuts in. (E.g. speed of ego vehicle 60 km/h and preceding vehicle to cut in at 40 km/h in the direction of travel and 1.0m/s in the lateral direction)



Judgment Criteria

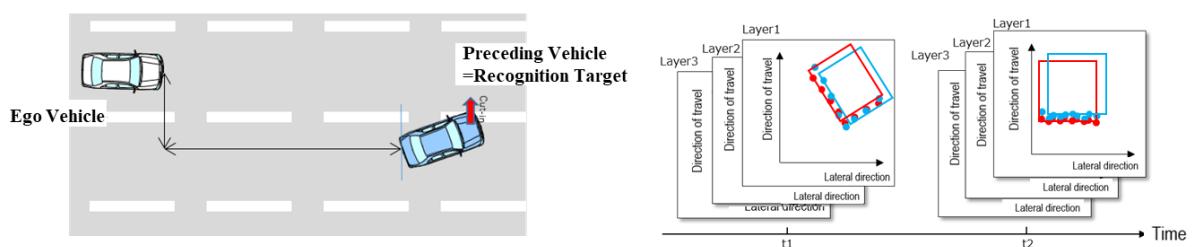
E.g.) The range, number of detection points and size over time, are to satisfy previously mentioned criteria (4.2, 4.4, 4.5)



■ Basic Traffic Flow Scenario : Cut-In Scenario (Normal Vehicle) 【 Object 】

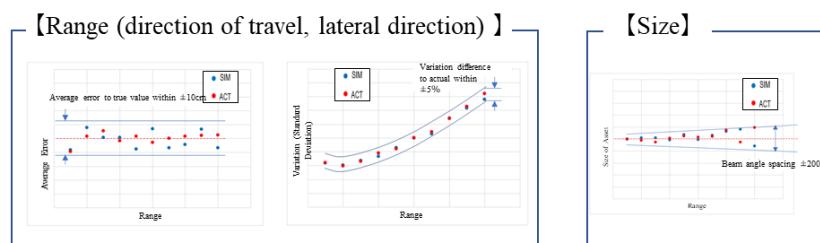
Method of Validation

Compare the change in the object's range and size over time between the actual and the simulation for when a normal vehicle cuts in. (E.g. speed of ego vehicle 60 km/h and preceding vehicle to cut in at 40km/h in the direction of travel and 1.0m/s in the lateral direction)



Judgment Criteria

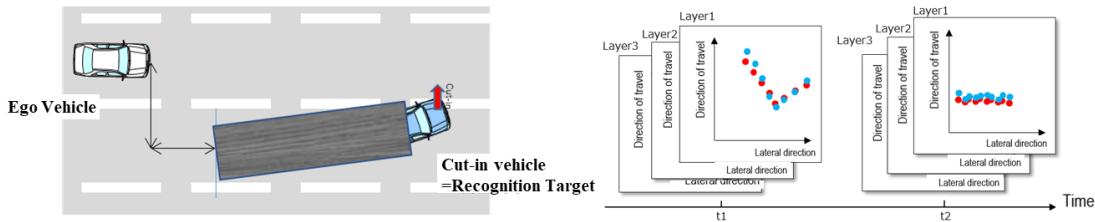
E.g.) The range (in the direction of travel and lateral direction) and size over time, are to satisfy previously mentioned criteria (4.2, 4.5)



■ Basic Traffic Flow Scenario : Cut-In Scenario (Large-Sized Vehicle) 【 Pointcloud Data 】

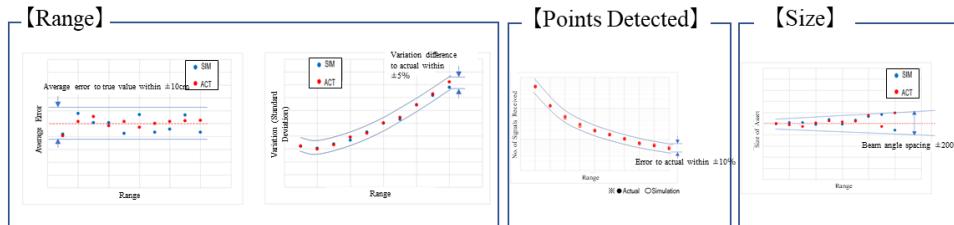
Method of Validation

Compare the change in azimuth, range, number of detection points and size over time between the actual and the simulation, for when a large-sized vehicle cuts in. (E.g. speed of ego vehicle 60 km/h and preceding vehicle to cut in at 40 km/h in the direction of travel and 1.0 m/s in the lateral direction)



Judgment Criteria

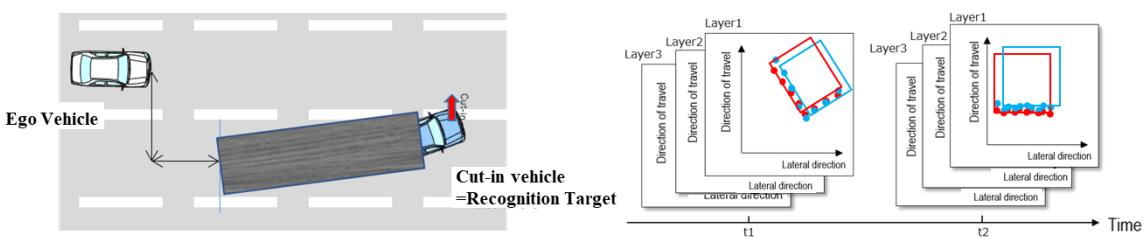
E.g.) The range, number of detection points and size over time, are to satisfy previously mentioned criteria (4.2, 4.4, 4.5)



■ Basic Traffic Flow Scenario : Cut-In Scenario (Large-Sized Vehicle) 【 Object 】

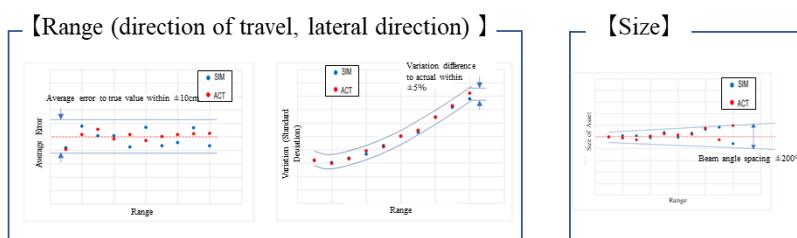
Method of Validation

Compare the change in the object's range and size over time between the actual and the simulation, for when a large-sized vehicle cuts in. (E.g. speed of ego vehicle 60 km/h and preceding vehicle to cut in at 40 km/h in the direction of travel and 1.0 m/s in the lateral direction)



Judgment Criteria

E.g.) The range (in the direction of travel and lateral direction) and size over time, are to satisfy previously mentioned criteria (4.2, 4.5)

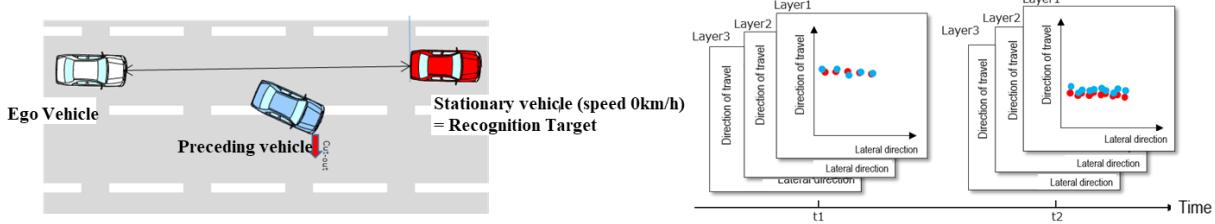


■ Basic Traffic Flow Scenario : Cut-Out Scenario 【 Pointcloud Data 】

LiDAR
(F.2.3.2.3.5)

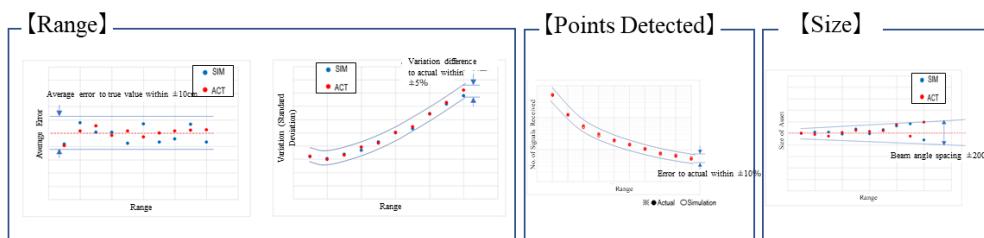
Method of Validation

Compare the change in azimuth, range, number of detection points and size over time between the actual and the simulation, for when the preceding vehicle cuts out, resulting in the approach toward a stationary vehicle. (E.g. speed of ego vehicle while traveling behind the preceding vehicle is 40 km/h as it cuts out, then ego vehicle approaches the recognition target)



Judgment Criteria

E.g.) The range, number of detection points and size over time, are to satisfy previously mentioned criteria (4.2, 4.4, 4.5)

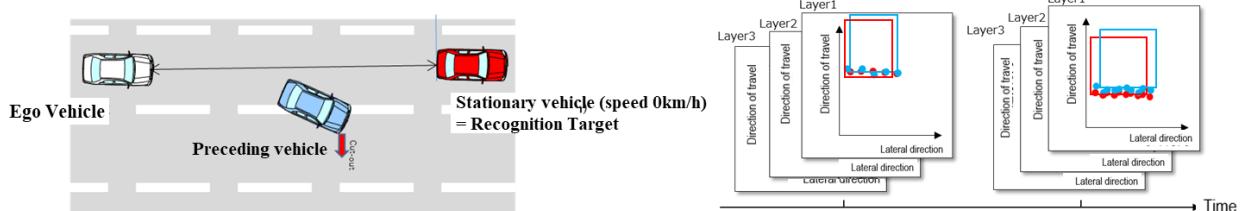


■ Basic Traffic Flow Scenario : Cut-Out Scenario 【 Object 】

LiDAR
(F.2.3.2.3.6)

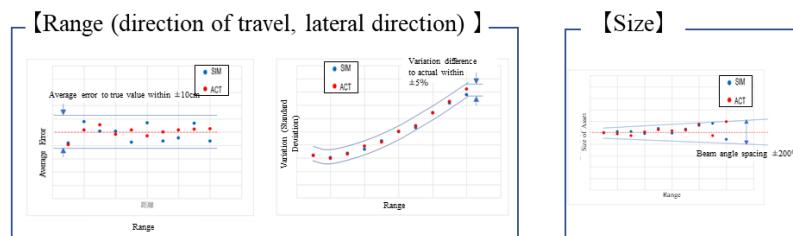
Method of Validation

Compare the change in range and size over time between the actual and the simulation, for when the preceding vehicle cuts out, resulting in the approach toward a stationary vehicle. (E.g. speed of ego vehicle while traveling behind the preceding vehicle is 40 km/h as it cuts out, then ego vehicle approaches the recognition target)



Judgment Criteria

E.g.) The range (in the direction of travel and lateral direction) and size over time, are to satisfy previously mentioned criteria (4.2, 4.5)



F.2.3.3 Validation method of common requirement of Camera

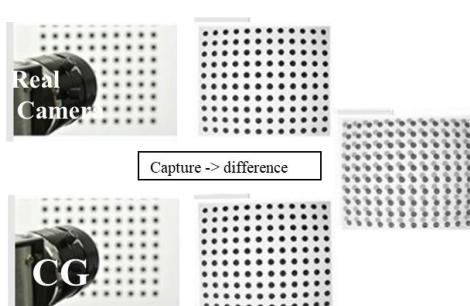
■ Camera module calibration [View angle/Optical axis/Distortion]

camera
(0-1)

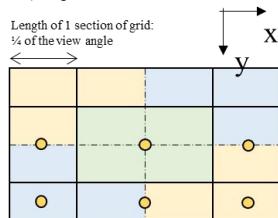
Method of Validation

Use a test chart to calibrate the difference in distortion and resolution to a minimum between the Real Camera and CG images based on the raw data

- With the real camera align the center of the grid chart to the center of the imaging device (while maintaining the levelness of the chart) and obtain an image
- With the CG align the center position in the same way, then adjust the camera parameters to ensure the surrounding grid is the same (position) as that of the real camera image
- Parameters for calibration (apply one of the following two methods)
 - Distortion and resolution characteristics of overall screen
 - The lens' focal length, distortion characteristics, various aberrations, center deviation of assembly (between the lens and the imaging device) and angle deviation



- Darkroom
- Use a known light source (All grids can maintain an appropriate contrast)



● Mandatory evaluation points

https://www.pearl-opt.com/chart/paper_chart.html
https://www.seikadi.com/measurement/qanda/qanda_material/Effectoflensdistortion.html

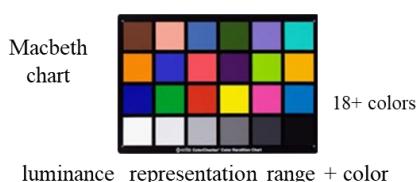
■ Camera module calibration [Color/Luminance]

camera
(0-2)

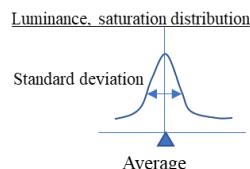
Method of Validation

Use a test chart to calibrate the difference in luminance representation and color to a minimum between the real camera and CG images based on the raw data

- Measure with each color block in the chart
- Measure statistics of 9 pixels or more at the target image position (Brightness, average value again, standard deviation)
- Evaluate by luminance representation and color reproducibility of general camera performance



- Darkroom
- Use a known light source



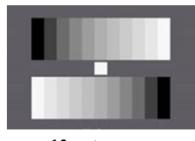
Judgement Criteria

For each item, the difference from the real camera is within +/- 5%

■ Camera module calibration [Dynamic range]

camera
(0-3)

Gray chart

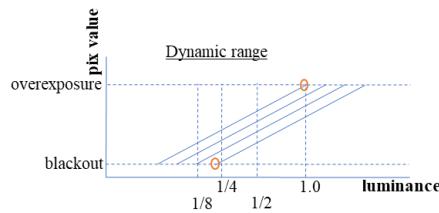


13+ steps



Validate dynamic range

- Darkroom
- Use a known light source



Method of Validation

The dynamic range of the camera with respect to the luminance is measured by gradually changing the luminance of the known lighting. Minimize the gap between the real camera and CG images based on raw data.

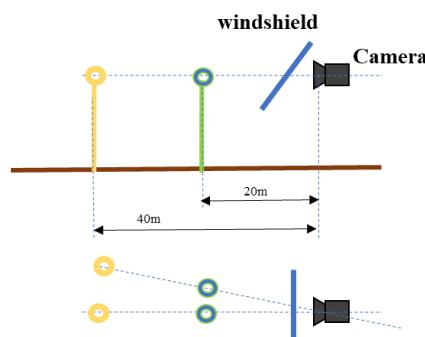
- Luminance vs. pixel value until overexposure
- The luminance of overexposure is set to 1, and the luminance until blackout is compared in 6 steps or more.
- It is desirable that the measurement points have a geometric progression. (1/2, 1/4, 1/8, etc.)

Judgement Criteria

- The difference between the luminance at the time of overexposure and the real camera is within +/- 5%.
- The difference between the luminance at the time of blackout and the real camera is within +/- 5%.

■ on-board Camera front calibration [Optical axis]

camera
(1-0)



Calibration the mounting to minimize deviation from the real camera using two target points at different distances on the reference optical axis.

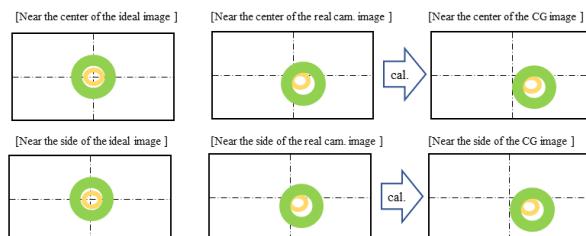
- On-board
- Sky light (Sun light source)
- PG equivalent asphalt road surface, horizontal plane

Method of Validation

- In a real camera, the camera is mounted and calibrated so that the target points with different distances on the optical axis are the center points of the imaging device.
- Calibration the related parameters so that they are (positionally) equivalent even in CG.
- Calibration parameters
 - 1) Height and orientation of camera mounting
 - 2) Curvature, inclination, and refractive index of the windshield

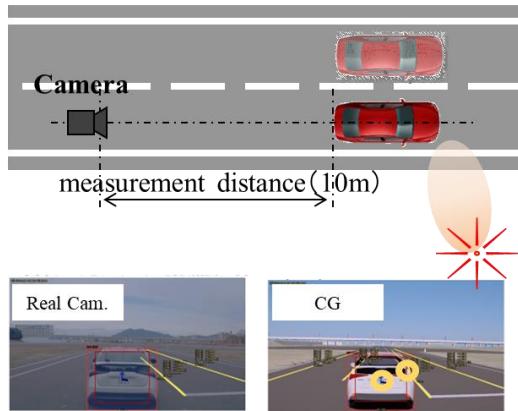
Judgement Criteria

At the position of the target point on the image, the difference from the real camera is within the range of +/- 2 pixels.



- on-board Camera front calibration(vehicle) [Distortion]
- Asset (ego-vehicle stopped) recognition target (vehicle) [Position]
- Scenario recognition target (vehicle) [Position]

camera
(1-1)
(2-1/3-1)
(4-1/5-1)



- Use high-precision GPS to maintain a vehicle distance of about 10m and acquire real camera images in the ego lane and adjacent lanes.
- Create image with CG in the same scenario
- Measure the positions of several singular points on the vehicle (such as gaps in the body) on the image.
- Comparison between real environment and simulation environment

For each item, the difference from the real camera depends on the position on the image.

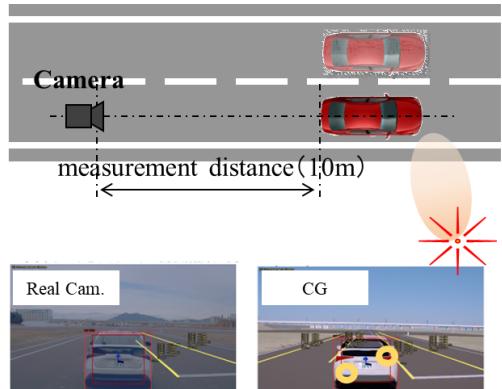
- Near the center: +/- within 5 pixels
- Near the surrounding: +/- within 10 pixels

※The measurement error of the vehicle distance is 1%.

- On-board
- Sky light (Sun light source)
- PG equivalent asphalt road surface, horizontal plane

- on-board Camera front calibration(vehicle) [Color/Luminance]
- Asset (ego-vehicle stopped/low-speed) recognition target (vehicle) [Color/Luminance]
- Scenario recognition target (vehicle) [Color/Luminance]

camera
(1-2)
(2-2/3-2)
(4-2/5-2)



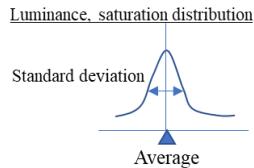
- On-board
- Sky light (Sun light source)
- PG equivalent asphalt road surface, horizontal plane

Method of Validation

- Use high-precision GPS to maintain a vehicle distance of about 10m and acquire real camera images in the ego lane and adjacent lanes.
- Create image with CG in the same scenario
- Measure the luminance and color expressions of the body, bumper, and preferably the tail lamp, when the sun is shining and when there is a shadow.
- Measure statistics (brightness, average saturation, standard deviation) of 9 pixels or more at the target image position.
- Comparison between real environment and simulation environment

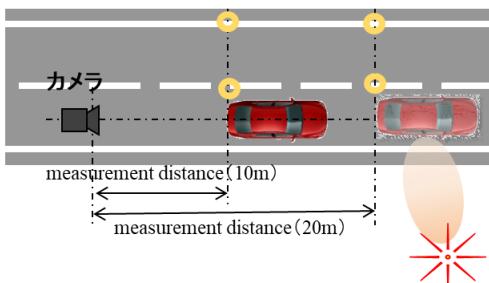
Judgement Criteria

For each item, the difference from the real camera is within +/- 10%



- on-board Camera front calibration(boundary) [Distortion]
- Asset (ego-vehicle stopped/low-speed) recognition target (boundary) [Position]
- Scenario recognition target (boundary) [Position]

camera
(1-1)
(2-4/3-4)
(5-4)



Method of Validation

- Use high-precision GPS to identify the comparison position (white lines and roads 10m and 20m away) and acquire real camera images.
- Create image with CG in the same scenario
- Measure the position on the image of the comparison position on the boundary
- Compare the shape of the lane maker on the image with a real camera and CG

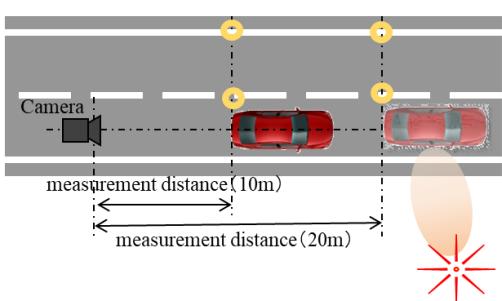
Judgement Criteria

For each item, the difference from the real camera is within +/- 10%

- On-board
- Sky light (Sun light source)
- PG equivalent asphalt road surface, horizontal plane

- on-board Camera front calibration(boundary) [Color/Luminance]
- Asset (ego-vehicle stopped/low-speed) recognition target (boundary) [Color/Luminance]
- Scenario recognition target (boundary) [Color/Luminance]

camera
(1-2)
(2-5/3-5)
(5-5)

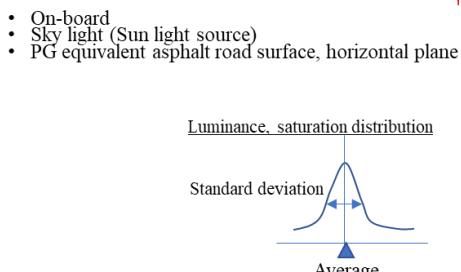


Method of Validation

- Use high-precision GPS to identify the comparison position (white lines and roads 10m and 20m away) and acquire real camera images.
- Create image with CG in the same scenario
- Measure the luminance and color expressions of the lane marker and road, when the sun is shining and when there is a shadow.
- Measure statistics (brightness, average saturation, standard deviation) of 9 pixels or more at the target image position.
- Comparison between real environment and simulation environment

Judgement Criteria

For each item, the difference from the real camera is within +/- 10%



- Asset (ego-vehicle stopped) recognition target (vehicle)
[Distance/Speed]
- Scenario recognition target (vehicle) [Distance/Speed]

camera
(2-3-1/2-3-3/3-3-
1/3-3-3)
(4-3-1/4-3-3/5-3-
1/5-3-3)



Method of Validation

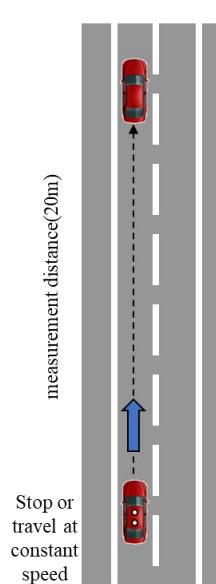
- Speed of ego vehicle : stopped or about 5 km/h
- Travel on known roads (straight lines, steady circles (eg R100))
- Maintain a distance of about 20m from the target vehicle
- Measure the time-series data of the distance (position) and relative speed to the target vehicle in front of the ego lane.
- Target vehicles are passenger vehicles and large-sized vehicles
- Comparison between real environment and simulation

Judgement Criteria

- The distance is within +/- 5% of the difference from the real camera.
- The speed is within +/- 10% of the difference from the real camera.

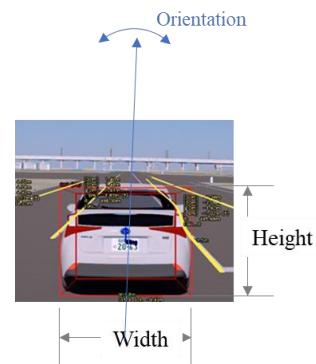
- Asset (ego-vehicle stopped) recognition target (vehicle)
[Size/Orientation]
- Scenario recognition target (vehicle) [Size/Orientation]

camera
(2-3-2/3-3-2)
(4-3-2/5-3-2)



Method of Validation

- Speed of ego vehicle : stopped or about 5 km/h
- Travel on known roads (straight lines, steady circles (eg R100))
- Maintain a distance of about 20m from the target vehicle
- Measure the time-series data of the height, width and orientation to the target vehicle in front of the ego lane.
- Target vehicles are passenger vehicles and large-sized vehicles
- Comparison between real environment and simulation

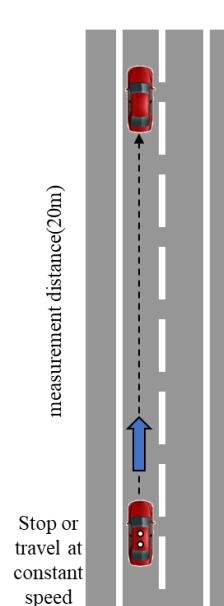


Judgement Criteria

- The size is within +/- 5% of the difference from the real camera.
- The orientation is within +/- 5% of the difference from the real camera.

- Asset (ego-vehicle stopped) recognition target (vehicle) [Type]
- Scenario recognition target (vehicle) [Type]

camera
(2-3-4/3-3-4)
(4-3-4/5-3-4)



Method of Validation

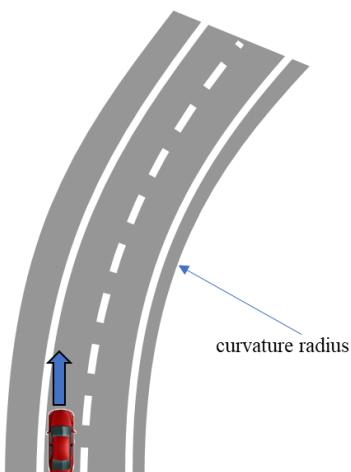
- Speed of ego vehicle : stopped or about 5 km/h
- Travel on known roads (straight lines, steady circles (eg R100))
- Maintain a distance of about 20m from the target vehicle
- Measure the time-series data of the type to the target vehicle in front of the ego lane.
- Target vehicles are passenger vehicles and large-sized vehicles
- Comparison between real environment and simulation environment

Judgement Criteria

- Target type is matching

- Asset (ego-vehicle stopped/low-speed) recognition target (boundary) [Curvature]
- Scenario recognition target (boundary) [Curvature]

camera
(2-6-1/3-6-1)
(5-6-1)



Method of Validation

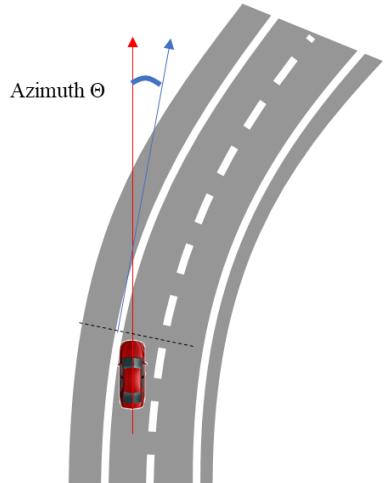
- Speed of ego vehicle : stopped or about 5 km/h
- Travel on known roads (straight lines, steady circles (eg R100))
- Measure time-series data of the radius of curvature of a traveling steady circle lane marking
- Comparison between real environment and simulation

Judgement Criteria

- The curvature is within +/- 5% of the difference from the real camera.

- Asset (ego-vehicle stopped/low-speed) recognition target (boundary) [Azimuth]
- Scenario recognition target (boundary) [Azimuth]

camera
(2-6-2/3-6-2)
(5-6-2)



Method of Validation

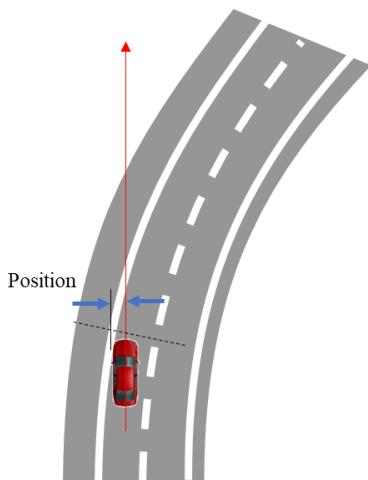
- Speed of ego vehicle : stopped or about 5 km/h
- Travel on known roads (straight lines, steady circles (eg R100))
- Measure the time-series data of the azimuth of the lane marking that travels based on the own vehicle
- Comparison between real environment and simulation

Judgement Criteria

- The azimuth is within +/- 5% of the difference from the real camera.

- Asset (ego-vehicle stopped/low-speed) recognition target (boundary) [Horizontal position]
- Scenario recognition target (boundary) [Horizontal position]

camera
(2-6-3/3-6-3)
(5-6-3)



Method of Validation

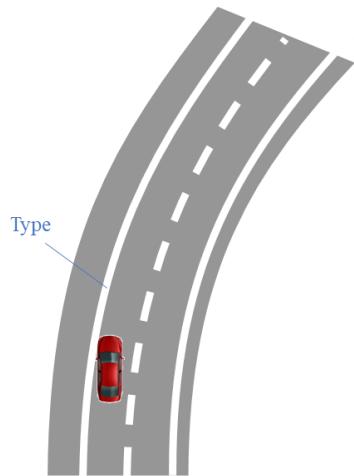
- Speed of ego vehicle : stopped or about 5 km/h
- Travel on known roads (straight lines, steady circles (eg R100))
- Measure the time-series data of the position of the lane marking that travels based on the own vehicle
- Target the left and right lane markings of ego lane
- Comparison between real environment and simulation

Judgement Criteria

- The Horizontal position is within +/- 5% of the difference from the real camera.

- Asset (ego-vehicle stopped/low-speed) recognition target (boundary) [Type]
- Scenario recognition target (boundary) [Type]

camera
(2-6-4/3-6-4)
(5-6-4)



Method of Validation

- Speed of ego vehicle : stopped or about 5 km/h
- Travel on known roads (straight lines, steady circles (eg R100))
- Measure the time-series data of the type of each lane marking
- Types are dashed lines, solid lines, colors, and other output types defined by each recognition process.
- Comparison between real environment and simulation

Judgement Criteria

- Target type is matching

F.3 perception disturbance reproducing requirement and reproductivity validation method

In this section items to be confirmed as perception disturbance reproducing requirement and validation method are clarified. The way of study is the same as common requirement.

As a first step, clarifying the way of thinking about what items should be done as perception disturbance reproducing requirement is shown. After that clarifying validation method for each sensor based on this way of thinking. This method is defined based on each sensors' principle thus it is necessary to clarify method when validating another principle's sensor following the way of thinking. This validation method shown below can be replaced by another method that can verify the same contents.

F.3.1 the way of thinking about perception disturbance reproducing requirement

This section clarifies the way of thinking about the items to be set as perception disturbance reproducing requirement. Doing same process of common requirement, component of object detection are defined as below elements as ①sensor/vehicle itself, ②space where the signal propagates ③recognition target(fig.F-3), and items to be validated and their criteria without perception disturbance for each elements are clarified. Additionally the method to validate that recognition target can be detected under basic traffic disturbance scenario is defined to confirm this totally.

F.3.2 The way of thinking about perception disturbance reproducing requirement for each sensor

F.3.2.1 The way of thinking about perception disturbance reproducing requirement for millimeter ware Radar

In accordance with the principles of the Radar perception, validates whether physical amount of distance, direction, relative speed and received wave intensity are reproduced (fig.F-4)

Based on this way of thinking, list of actual requirement shown in table.F-4 is clarified.

										Perception process															Recognition process									
										Signal from perception target (S)										Signal from others					Processing ability	Processing performance								
										Phase		Strength		Noise (N)			Low S/N		Low D/U			Increasing of U												
										Change of DOA		High intensity		No signal (partial)	Aliasing	Harmonic	Large difference of signal	Low S/N (change of angle)	Low S/N (attenuation at the sensor surface)	Low S/N (attenuation in space)	Low S/N (low retroreflection)	Low D/U (change of angle)	Low D/U (road surface reflection)	Low D/U (surrounding structures)	Low D/U (floating objects in space)	Low D/U (sensors on other cars)	Low D/U (sensors on ego cars)	Increasing of U (road surface reflection)	Detection (grouping of reflected points)	Clustering (tracking of target)	Tracking (classification of target)			
										Frequency	Reflection (indirect wave)	Refraction	Change of propagation delay	No signal (partial)	Aliasing	Harmonic	Large difference of signal	Low S/N (change of angle)	Low S/N (attenuation at the sensor surface)	Low S/N (attenuation in space)	Low S/N (low retroreflection)	Low D/U (change of angle)	Low D/U (road surface reflection)	Low D/U (surrounding structures)	Low D/U (floating objects in space)	Low D/U (sensors on other cars)	Low D/U (sensors on ego cars)	Increasing of U (road surface reflection)	Lack of points to be processed	False detection of undesired signal	No detection of required signal	Unexpected distribution of point cloud	Unexpected movements (between frames)	Unexpected objects
Validation of Disturbance Reproducibility	Simulating Large Difference of Signals	Signal Intensity Ratio	Distance/Angle	Signal intensity ratio of target 1 and 2 is equivalent to the actual environment.			3-1											○																
		HMFW Ratio	Distance/Angle	HMFW ratio of target 1 and 2 is equivalent to the actual environment.			3-2											○																
		Buried Signals	Distance/Angle	The signal from a motorcycle is obscured by the signal of a large vehicle in the same way as the actual environment.			3-3											○																
	Simulating Low D/U due to Road Surface Multipath	Simulating the Disturbance Phenomena	Received Power	Distance	Envelope line in received power is equivalent to the actual environment.			4-1																										
		Road Surface Material/Road Surface Condition	Received Power	Distance	Envelope in received power of the reflected wave from C/R is equivalent to the actual environment.			4-2																										
		Null points	Distance	Null points distances in received power of the reflected wave from C/R is equivalent to the actual environment.			4-3																											
	Simulating Low D/U Due to Change of the Angle	Simulating the Disturbance Phenomena	Buried Signals	Distance/Angle	The phenomenon, whereby the signal from the recognition target becomes buried in the signal from the signage board, occurs in the same way as the actual environment.			5-1																										
		Reflective Properties of Overhead Structures	Signal Intensity Ratio	Distance/Angle	Signal intensity ratio of the target and signage board is equivalent to the actual environment.			5-2																										
		HMFW Ratio	Distance/Angle	HMFW ratio of the target and signage board is equivalent to the actual environment.			5-3																											
	Simulating Low S/N Due to Vehicle Orientation	Simulating the Disturbance Phenomena	Cumulative distribution of the received power	Vehicle orientation	The cumulative distribution of the received power within a certain distance range is equivalent to the actual environment.			6-1																										

Table.F-4: perception disturbance of millimeter radar, reproducibility validation and disturbance principle

F.3.2.2 The way of thinking about perception disturbance reproducing requirement for LiDAR

In accordance with the principles of the LiDAR perception, validates whether physical quantities like azimuth, range, strength, number of detection points and size are reproduced(fig.F-5).

Based on this way of thinking, list of actual requirement shown in table.F-5 is clarified.

						No disturbance	Perceptual part																
							Scan timing		Signal from recognition target (S)			S strength			S Propagation direction		S speed	N factor		U factor			
							Misalignment of overall spatial position	Misalignment of position of recognition target	Saturation of S	Attenuation of S				No S due to occlusion	Reflection	Refraction	Arrival time of S	Pulsed noise	DC noise	Multiple reflections	Signal from non-recognition target (Reflection)	Signal from non-recognition target (Refraction)	
Disturbance Reproducibility verification	Noise	Error mean and variation	A standard reflector is installed in front of LiDAR, the error average and variance are measured by changing the distance, and it is verified that the difference from the actual measurement is within the judgment criteria.	Standard reflector	light source	Altitude	Being able to detect after changing the position from 0 to 90 degrees	F.2.4.2.1										O	O				
						Direction	Being able to detect after changing the position from 0 to 360 degrees	F.2.4.2.1										O	O				
		Reception strength and detection probability				Brightness	The brightness can be detected by changing the brightness from 0 to XX mW/mm ² . (Since the wavelength range differs depending on LiDAR, set it within the range that can be taken according to the wavelength that Lidar emits.)	F.2.4.2.1										O	O				
		A standard reflector is installed in front of LiDAR, the reception intensity and detection probability are measured by changing the distance, and it is verified that it is within the judgment criteria.	Standard reflector	light source	Altitude	Being able to detect after changing the position from 0 to 90 degrees	F.2.4.2.2	O										O					
					Direction	Being able to detect after changing the position from 0 to 360 degrees	F.2.4.2.2	O										O					
					Brightness	The brightness can be detected by changing the brightness from 0 to XX mW/mm ² . (Since the wavelength range differs depending on LiDAR, set it within the range that can be taken according to the wavelength that Lidar emits.)	F.2.4.2.2	O										O					
	Attenuation of S	Install the asset in front of LiDAR and change the distance to verify the difference in the number of received points.	Asset (Vehicles, Motorcycles, People, Installations, Falling objects)	Vehicle	light source	Altitude	Being able to detect after changing the position from 0 to 90 degrees	F.2.4.2.3										O	O				
						Direction	Being able to detect after changing the position from 0 to 360 degrees	F.2.4.2.3										O	O				
						Brightness	The brightness can be detected by changing the brightness from 0 to XX mW/mm ² . (Since the wavelength range differs depending on LiDAR, set it within the range that can be taken according to the wavelength that Lidar emits.)	F.2.4.2.3										O	O				
		Reproducibility of cognitive disturbance	Install the asset in front of LiDAR and change the distance to verify the difference in the number of received points.			Reflector	Shape	It can be detected by vehicles with high ground clearance, vehicles with low vehicle height, motorcycles, bicycles, angular vehicles, and rounded vehicles.	F.2.4.2.4									O					
						Mirror reflector	Color, Material		F.2.4.2.4									O					

Table.F-5: perception disturbance of LiDAR, reproducibility validation and disturbance principle

F.3.2.3 The way of thinking about perception disturbance reproducing requirement for Camera

As shown in the section about common requirement, camera can use colour information while active type sensors can detect distance information and camera cannot detect it in perception block, so that this characteristics is very important to validate reproductivity under perception disturbance.

Based on this way of thinking, list of actual requirement shown in table.F-6 is clarified.

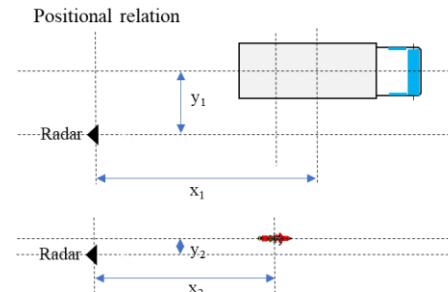
Verification	Items/ Target Parts	Measurement Items	Parameters	Requirement	Perception Part												Recognition Part															
					Optics				Imager				Image Processing				Feature extaction		Detection, Classification		Positioning		Tracking									
					Requirement ID	Normal Condition(Sight)	Normal Condition(Day)	Normal Condition(Night)	Blue Position shift,	Refraction	Reflection	Flare	Flare, Ghost, Double image	Reflected image	Scattering	Diffraction	Absorption	Noise	Color Filter	Motion Blur	Exposure Time	Flicker	Exposure period	Brightness	Hue	Chroma	Hidden	Low spatial frequency	Low contrast	No classification	(False positive detection)	Detection or classification error
perception disturbance reproducing requirement	(Shelter: mud, water droplets) (passenger cars, large vehicles) (boundary lines: white, solid, dashed) (Road surface: curved, asphalt)	Placement verification (shield)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-1-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
			luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-1-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—
		Placement verification (landmarks)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-2-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—
			luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-2-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—
		Recognition result (Object)	relative distance	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-3-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—
			size, direction	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-3-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—
			relative velocity	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-3-3	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—
			classification	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-3-4	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
		Placement verification(boundary line)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-4	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
			luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-5	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—
		Recognition result (boundary line)	curvature	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-6-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
			direction	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-6-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
			lateral position	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-6-3	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
			classification	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-1-6-4	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
		Placement verification(tunnel)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-1-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
			luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-1-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
		Placement verification(landmarks)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-2-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
			luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-2-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
		Recognition result (Object)	relative distance	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-3-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
			size, direction	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-3-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
			relative velocity	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-3-3	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
			classification	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-3-4	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
		Placement verification(boundary line)	shape, size	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-4	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
			luminance, hue, color	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-5	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	
		Recognition result (boundary line)	curvature	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-6-1	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—
			direction	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-6-2	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—
			lateral position	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-6-3	✓	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	—	○	—	
			classification	Recognized parameter values of the recognition target in virtual environment are similar to ones in real environment.	B-2-6-4	✓	—	—	—	—	—</																					

F.3.3 Validation method of perception disturbance reproducing requirement

Validation method of each requirement for each sensor defined in section F.3.2 is shown in this section.

F.3.3.1 Validation method of perception disturbance reproducing requirement of millimeter wave Radar

■ Simulating Large Difference of Signals: Simulating the Disturbance Phenomena – Signal Intensity Ratio / HMFW Ratio



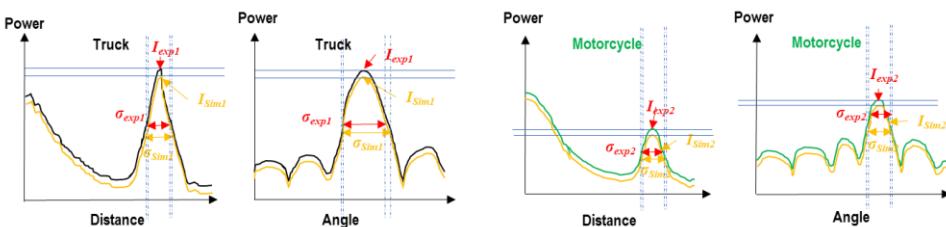
Method of Validation

- Evaluate using targets of differing degrees of reflectance (large-sized motor vehicle such as a truck (target 1) and a motorcycle (target 2))
- Evaluate in a stationary state at a speed (0km/h) with the recognition target on a horizontal straight road
- Evaluate recognition target 1 and 2 separately
- Position of recognition targets
 $x_1:140(m)$ $x_2:150(m)$ $y_1:3.5(m)$ $y_2:1(m)$
- Compare to the signal intensity ratio $I_{exp1}/I_{exp2} I_{sim1}/I_{sim2}$ and the HMFW ratio $\sigma_{exp1}/\sigma_{exp2} \sigma_{sim1}/\sigma_{sim2}$ of the truck and the motorcycle in the actual and simulated environment

mmWave radar
(3-1/3-2)

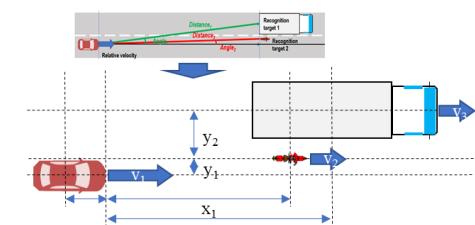
Judgement Criteria

The signal intensity ratio $I_{exp1}/I_{exp2} I_{sim1}/I_{sim2}$ and the HMFW ratio $\sigma_{exp1}/\sigma_{exp2} \sigma_{sim1}/\sigma_{sim2}$ between the actual and simulated environment must be $\pm 5\%$ or less.



■ Simulating Large Difference of Signals: Simulating the Disturbance Phenomena – Buried Signals

mmWave radar
(3-3)



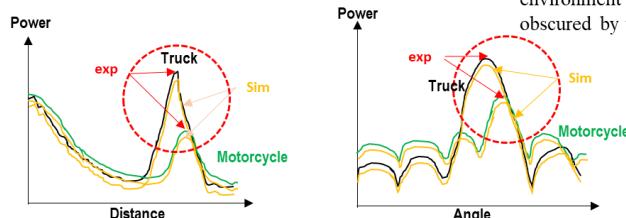
Method of Validation

- Using targets of different degrees of reflectance: large-sized vehicle such as a truck (target 1) and a motorcycle (target 2)
- Driving on a straight road
- Target 1 and 2 traveling parallel in adjacent lanes, and in the direction of approach of the ego vehicle
- Evaluate two scenarios (one is where the relative velocities of target 1 and 2 are the same as that of the ego vehicle, the other is where relative velocities differ)
- Initial position and velocity
 $x_1:140(m)$ $x_2:150(m)$ $y_1:1(m)$ $y_2:2.5(m)$
 $v_1:\text{approx. }100(\text{km/h})$ $v_2:\text{approx. }90(\text{km/h})$
 $v_3:\text{approx. }90(\text{km/h})$
- Evaluate whether motorcycles with low intensity are buried by the signals of large vehicles in actual and simulated environments

Phenomenon	Code Parameter	Definition	Parameter Range	Description	Case Change
Distance beyond detection range	-	-	Detects objects at longer detection distance than standard distance	To increase the perception distance in the rear, but it is necessary to increase the power of the radar to do so.	Case change the distance
Attenuation	-	-	Within the detection range (RFID range)	Setting the range of the sensor when the RF range is set to the detection range.	
Sensor angle	Sensor management	Set the angle of the sensor to the target angle.	When the angle of the sensor is set to the target angle, the sensor can detect the target angle.		
Strength difference (DCW, DCL, DCV)	Shape of object recognition (DCW)	Shape of object recognition (DCW)	Objects which are detected by the sensor which can be detected by the sensor.	When the sensor detects an object which can be detected by the sensor, it is detected by the sensor.	Case change the detection
	Size of object recognition (DCV)	Size of object recognition (DCV)	Objects which are detected by the sensor which can be detected by the sensor.	When the sensor detects an object which can be detected by the sensor, it is detected by the sensor.	
	Vehicle	Color	Define using data on size of reflectance in millimeter waveband.	When the sensor detects an object which can be detected by the sensor, it is detected by the sensor.	
		Material	Define using data on physical property values in millimeter waveband.	When the sensor detects an object which can be detected by the sensor, it is detected by the sensor.	
Controllability of objects in recognition	-	-	Objects which are detected by the sensor which can be detected by the sensor.	When the sensor detects an object which can be detected by the sensor, it is detected by the sensor.	

Judgement Criteria

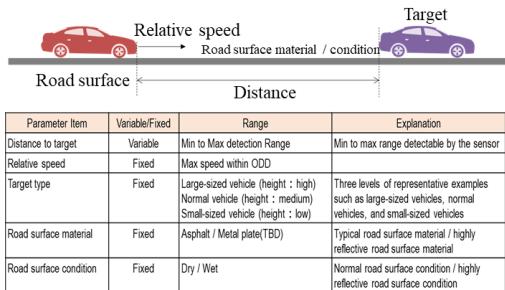
The reproducibility of the phenomenon in simulated environment where a motorcycle with a low intensity is obscured by the signal of a large vehicle with a high intensity.



The red frame is an example of this situation, where the signal of motorcycles is buried in the signal of heavy vehicles.

■ Validation the Low D/U due to Road Surface Multipath:
Validation the Phenomenon - Received Power

mmWave Radar
(4-1)

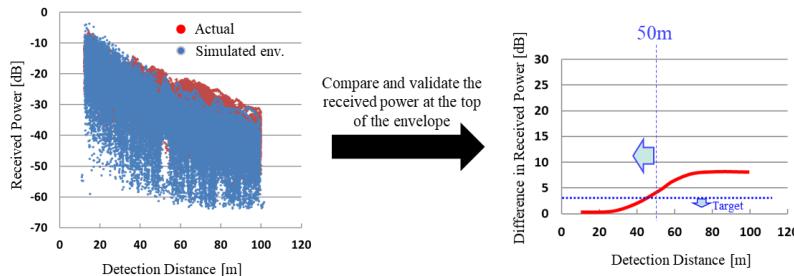


Method

- Simulate evaluation scenario for “low D/U (road surface multipath)” :
 - Approach the recognition target (stationary vehicle) ahead in the same lane as the ego vehicle
- Distance to subject :
 - Sensor min detectable distance to max detectable distance
- Relative speed : constant speed(ex. About 20km/h)
- Type of target : passenger vehicle, large-sized trailer
- Road surface material : asphalt, metal plate (TBD)
- Road surface condition : dry, wet

Judgment Criteria

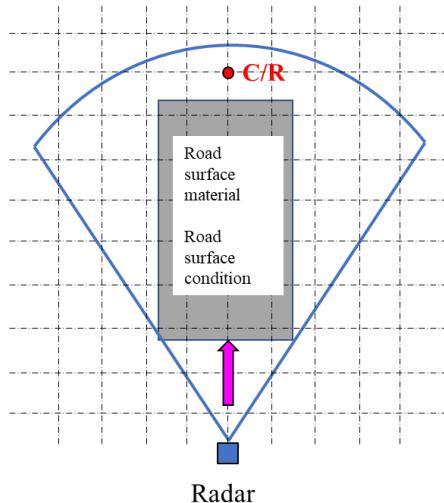
- Difference in the envelope line of the received power at a relative distance of 50 m and below: 3 dB or less



■ Validation the Low D/U due to Road Surface Multipath:

Road Surface Material/Road Surface Condition - Received Power/Null points

mmWave Radar
(4-2/4-3)



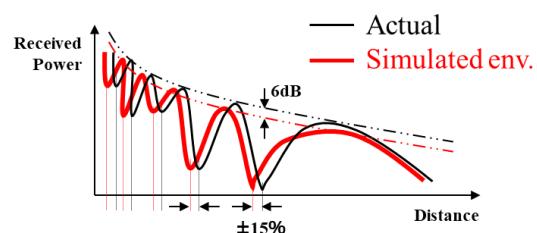
Method

- Place C/R in front of radar, and move radar toward C/R
- Road surface material : asphalt, metal plate (TBD)
- Road surface condition : dry, wet

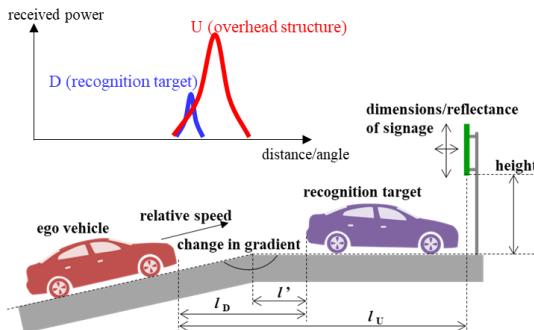
Judgement Criteria

Plot the received power of the maximum reflection point at each distance

- Difference in envelope : 6 dB or less
- Difference in null point distance : $\pm 15\%$ or less



■ Simulating Low D/U Due to Change of the Angle: Simulating the Disturbance Phenomenon – Buried Signals



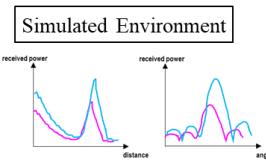
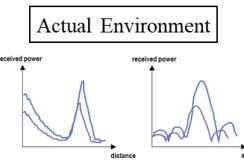
Method of Validation

- Simulate the scenario “Low D/U due to change of the angle”:
 - Traveling a road with a change in gradient (concave down)
 - A metallic signage board ahead after the inflection point
 - The ego vehicle is to approach the stationary vehicle stopped nearby the signage board ahead.
- Change in gradient : 2 points between 3 and 10 ($^{\circ}$)
- I_D : 5 (m) fixed
- I_U initial value : 15 (m)
- I_U initial value : 20 (m)
- Type of the recognition target : a passenger vehicle

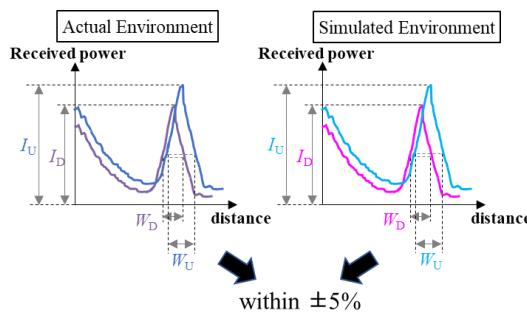
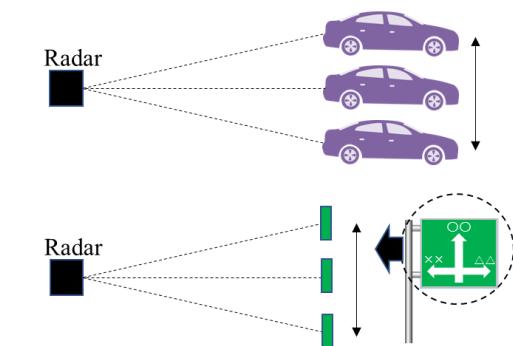
	Parameters	Parameter Range	Explanation
Other than the causal factor	Change in the road gradient	Variable 0 to 18 % equivalent	Use a road which is concave down as a representative
	Initial distance to recognition target I_D	Fixed Distance required to avoid collision	
	Distance to recognition target from the inflection point I'	Variable 0 to I_D	
	Initial position of recognition target	Fixed 0°	Fixed on the same lane
	Initial distance to signage board I_U	Variable $I_D - 5$ to $I_D + 5$ (m)	assume the object within the neighboring lanes
	Lateral position of signage board	Variable -3.5 to +3.5 (m)	
	Height of signage board (to bottom edge)	Fixed 4.5 m (above road)/1.5 m (roadside)	According the Traffic Sign Installation Standard
	Dimensions of the signage board	Fixed 2.7×3.5 (m)	Guidance signage on highways
	Reflectance of the signage board	Fixed Measured value of the real board	
	Relative speed	Fixed Max. speed within QDD	
	Type of the recognition target	Fixed Passenger vehicle/Pedestrian	Representative traffic participant/low reflectance

Judgment Criteria

The phenomenon, whereby the signal from the recognition target becomes buried in the signal from the signage board, occurs in the same way in both the actual and the simulated environments.



■ Simulating Low D/U Due to Change of the Angle: Reflective Properties of Overhead Structures – Signal Intensity Ratio / HMFW

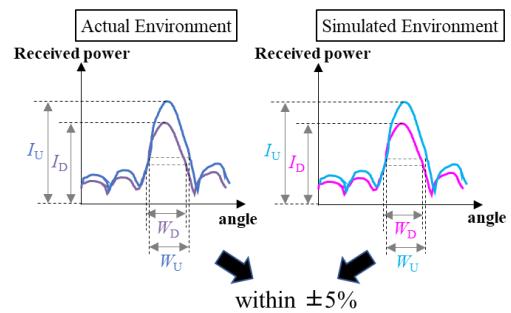


Method of Validation

- Direct the radio wave toward the recognition target (passenger vehicle) and the signage board (flat area)
- Vary the angle of the vehicle/board and the radar within the vertical plane
- Measurement angles : $0^{\circ}, \pm 5^{\circ}, \pm 10^{\circ}$
- Measurement distance : 15, 20 (m)
- Compare the ratio of peak intensity I_D/I_U and the HMFW W_D/W_U between the actual and simulated environments

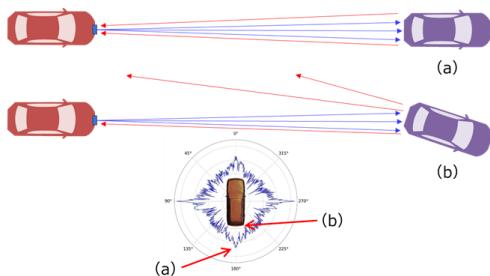
Judgment Criteria

Difference in I_D/I_U , W_D/W_U between actual and simulated environments : within $\pm 5\%$



■ Simulating Low S/N Due to Vehicle Orientation: Simulating the Disturbance Phenomenon

mmWave Radar
(6-1)



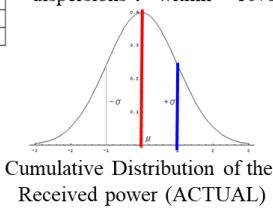
Parameter Item	Variable/ Fixed	Range	Explanation
Type of recognition target	Variable	<ul style="list-style-type: none"> Projected area (large/mid/small) Contribution rate to scattering= Reflectance (heavy use of metal / heavy use of non-metal / in-between) Directivity of scattered waves (uniform/biased) 	<ul style="list-style-type: none"> 3 levels of projected area generally 3 levels (no vehicle has zero metal used) 3 levels (relying on concentration of normal vectors in microparts of the vehicle)
Orientation of the target	Variable	0 to 30 deg.	According to the line of the road (curve R)
Distance to the target	Variable	5 to 150 m	
Relative speed	Fixed	20 km/h and below	constant

Method of Validation

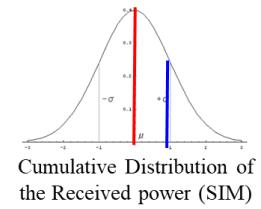
- Simulate the scenario for evaluating “low S/N due to vehicle orientation” :
 - Place a stationary vehicle up ahead on a straight road, and approach it at a slow speed
 - Change the orientation of the vehicle ahead and travel
- Vehicle angle : 0 to 30 deg. (constant)
- Initial distance between vehicles : 150 m
- Vehicle speed : 20 km/h and below (constant)
- Type of recognition target : passenger vehicle
- Record the received power in both the actual and simulated environments

Judgment Criteria

Show the **cumulative distribution of the received power (dBm)** within a certain distance range between vehicles (e.g. 10 to 20 m), and compare the averages and dispersions : within $\pm 10\%$



Cumulative Distribution of the Received power (ACTUAL)



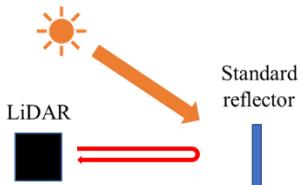
Cumulative Distribution of the Received power (SIM)

F.3.3.2

Validation method of perception disturbance reproducing requirement of LiDAR

■ Noise : Average error and Standard deviation

LiDAR
(F.2.4.2.1)



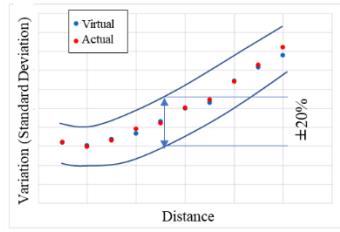
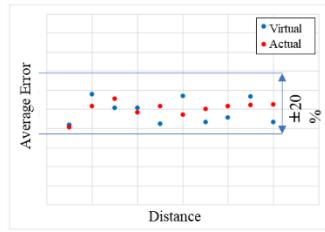
Method for Validation

Place a standard reflector in front of the LiDAR and vary the distance to measure the average error and standard deviation, to ensure the difference to the actual measurement falls within the judgment criteria.

Judgment Criteria

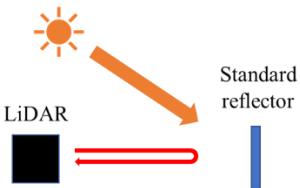
Average error : within $\pm 20\%$ of distance to subject
 σ : within $\pm 20\%$

Disturbance Factors		Disturbance Factor Parameter	Range	Basis (or reasons)
Type of space	Light source	Halogen lights, etc.	Altitude	0~90 degrees
			Azimuth	0~360 degrees
			Brightness	0~XX mW/mm ²
				Wavelength range may differ depending on LiDAR, thus set based on possible range of each



■ Noise : Strength of signal and Detection rate

LiDAR
(F.2.4.2.2)



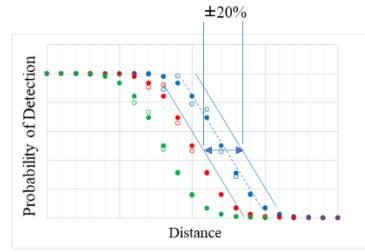
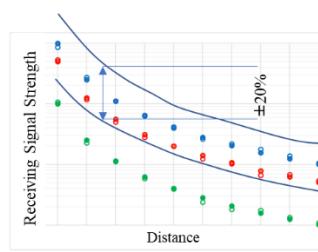
Method for Validation

Place a standard reflector in front of the LiDAR and vary the distance to measure the strength of the signal and detection rate, to ensure the difference to the actual measurement falls within the judgment criteria.

Judgment Criteria

Intensity error in relation to actual measurement value: within $\pm 20\%$
 Difference in range to actual measurement @ 90% detection: within $\pm 20\%$
 Difference in range to actual measurement @ 50% detection: within $\pm 20\%$
 Difference in range to actual measurement @ 10% detection: within $\pm 20\%$

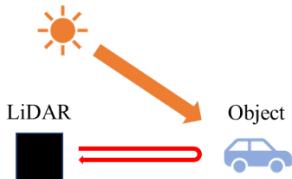
Disturbance Factors		Disturbance Factor Parameter	Range	Basis (or reasons)
Type of space	Light source	Halogen lights, etc.	Altitude	0~90 degrees
			Azimuth	0~360 degrees
			Brightness	0~XX mW/mm ²
				Wavelength range may differ depending on LiDAR, thus set based on possible range of each



※Blue: reflectivity xx% Red: reflectivity xx% Green: reflectivity xx%
 ※● Actual ○ Virtual

■ Noise : Number of Detection Points

LiDAR
(F.2.4.2.3)



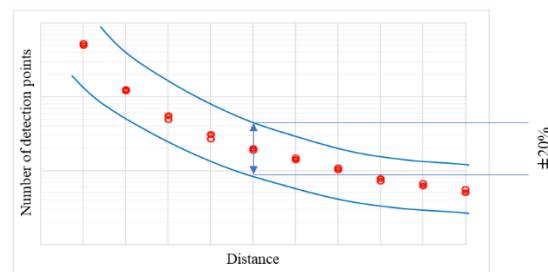
Method for Validation

Place the object in front of the LiDAR and vary the distance to validate the difference in number of detection points

Judgment Criteria

Error in relation to number of detection points in actual measurement to be within $\pm 20\%$
(Do not include large distances where the number of detected points decreases)

Disturbance Factors		Disturbance Factor Parameter	Range	Basis (or reasons)
Type of space	Light source	Halogen lights, etc.		
		Altitude	0–90 degrees	Possible range
		Azimuth	0–360 degrees	Possible range
		Brightness	0–XX mW/mm ²	Wavelength range may differ depending on LiDAR, thus set based on possible range of each



■ Attenuation : Reproduction of attenuation recognition target

Attenuation of S
(F.2.4.2.4)



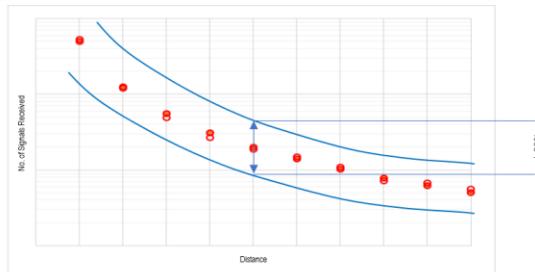
Method

Place an object in front of the LiDAR and vary the distance, to verify the difference in the number of signals received

Disturbance Factor	Disturbance Factor Parameter	Range	Basis (or reason)
Vehicle	Shape	Vehicles with high ground clearance or low vehicle height Motorcycles, bicycles Angular vehicles, Rounded vehicles	Clears bottom of body and only receives reflection from tires The top layer of the beam has difficulty hitting the roof rack Number of reflection points in horizontal direction is minimal Depending on orientation, it may be difficult for the direction of the normal vector to align with the LiDAR It may be difficult for the direction of the normal vector to align with the LiDAR
	Color, material properties	Black paint Specular reflection	Does not diffuse reflection well Depending on the orientation, the specular reflection will occur and not return

Judgment Criteria

E.g.) Error in relation to actual points measured to be within $\pm 20\%$
(Do not include large distances where number of signals reduces)

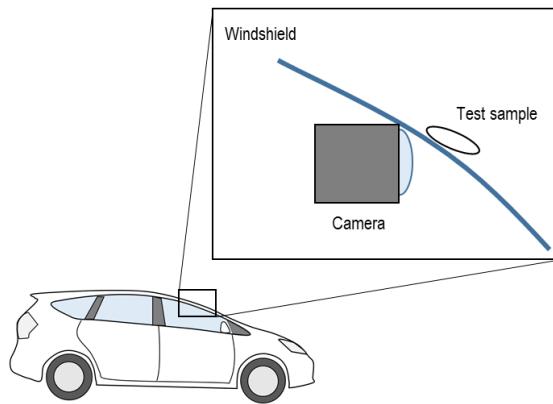


F.3.3.3 Validation method of perception disturbance reproducing requirement of Camera

■ Shielding Placement verification (shield)/ color luminance verification(shield)

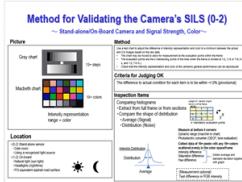
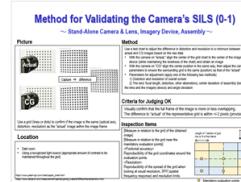
- Object adheres to front of sensor: consistency of perception device

Camera
(B-1-1-1
B-1-1-2)



Method

A test sample is applied to the windshield of the vehicle (while in a stationary state), and validation is conducted by evaluating the strength of the signal and color.



Judgment Criteria

Difference in pixilation:

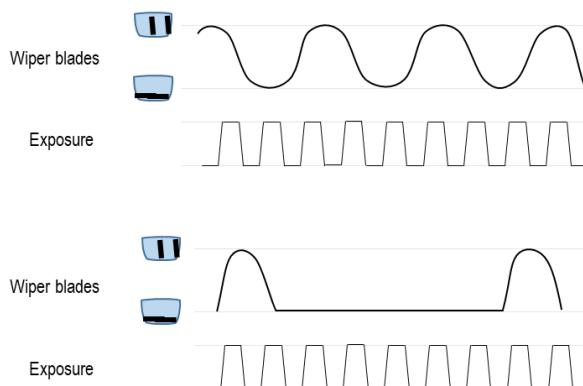
- within +/-5% to theoretical value or +/-5 pixels
- within +/-2% to actual or +/-2 pixels

■ Shielding Placement verification (shield)/ color luminance verification(shield)

- Obstruction in front of sensor: consistency of perception device

Camera
(B-1-1-1
B-1-1-2)

Wiper blades : intermittent, continuous



Method

Study the simulation of the image of wiper blades, as an element which can change over time as opposed to the adherence of a foreign object as seen in 1.

Evaluate the items that can change over time such as the movement of the wiper blades and the shutter speed.

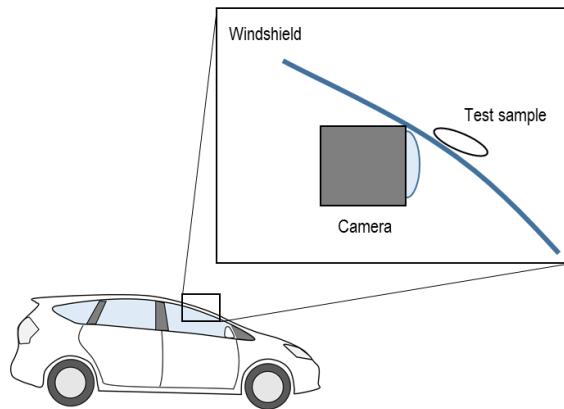
Judgment Criteria

- The changes match in regard to the exposure over time and wiper blade movement
- Difference in pixel value with an image of the same condition: 5% or less

■ Shielding Placement verification (Object)/ color luminance verification(Object)

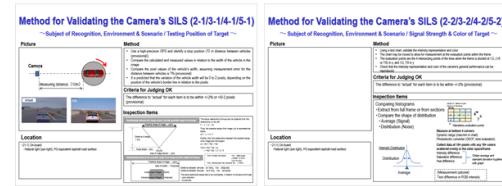
1. Object adheres to front of sensor: consistency of perception device

Camera
(B-1-2-1
B-1-2-2)



Method

A test sample is applied to the windshield of the vehicle (while in a stationary state), and validation is conducted by evaluating the strength of the signal and color.



Judgment Criteria

Difference in pixilation:

- within +/-5% to theoretical value or +/-5 pixels
- within +/-2% to actual or +/-2 pixels

■ Shielding Recognition result (Object)

Camera
(B-1-3)

Recognition Disturbance			Outline of Error	Extracting Traits			
		Categorization of Causes		Shielding	(Cannot see subject)		
Modeling Categories				Details of Error	Cannot see due to cause on vehicle side	Cannot see due to cause on subject side	Blind spot
Vehicle / Sensor	In front of sensor	① Adherence of foreign object / obstruction ② Shielding: wiper blades	Shielding: dirt, dust, etc.	Adherence of dirt, dust, etc. (lose full picture)	3	—	—
			Shielding: snow, ice, etc.	Adherence of snow, ice, etc. (lose full picture)	3	—	—
			Shielding: water, etc.	Adherence of water, etc. (lose full picture)	2	—	—
			Shielding: insect, bird poo, etc.	Adherence of insect, bird dropping, etc. (lose full picture)	2	—	—
Surrounding Environment	Surrounding structures / Road surface	③ Shape ④ Shielding	Incline	Difference in position/incline of road in picture			3
			Non-transparent	Parked vehicle, surrounding trees, flying objects			3
	Surrounding objects (moving)						
Subject of Recognition	On track	Lane lines	Dirty, scraped	Fallen leaves, shielding due to snow pile, dirty, scraped, re-painted		3	—
	Moving	⑤ Other vehicles	Adherence of foreign object Color Shape Range	Standard color of adhering foreign object (similar or different to vehicle color)		1	—
				Various shapes of foreign objects adhere (shape/pattern of dirt, stickers, etc.)		1	—
				Area adhered (one location to full vehicle)		3	—

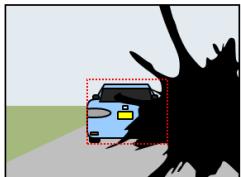
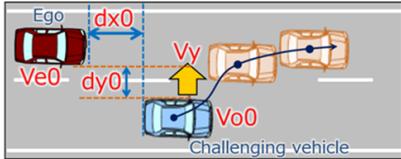
(Taken from Phenomenon and Cause Matrix)

Validation method to be identified for each of the above 5 categories

■ Shielding Recognition result (Object)

Camera
(B-1-3)

1. Object adheres to front of sensor Cut-in scenario



Method

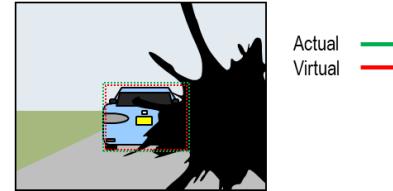
The recognition target cuts into the lane of the ego vehicle (in front) at a constant side-way speed, whilst vision is impaired due to the adherence of a foreign object.

Judgment Criteria

- Difference in longitudinal distance: 5% or less
- Difference in lateral distance: 5% or less
- Difference in longitudinal relative speed: 10% or less
- Difference in lateral relative speed: 10% or less
- Difference in width/height: 5pix or less

Reference: Parameters of Validation Scenario

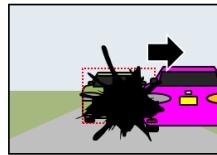
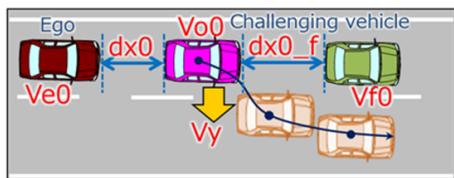
Parameter	Variable/Fixed	Range
Distance to the target	Variable	Longitudinal position dx0 [m] Lateral position y0 : 3.5m
Relative velocity to the target	Variable	Longitudinal velocity Vo0-Ve0 [kph] Lateral velocity Vy [kph]
Type of the target	Fixed	Shape: sedan Color: White
Degree of shielding of the detection-target due to adherence of foreign object	Variable	In relation to the bounding box of the detection-target ① Initial50% → Final0% ② Initial100% → Final50%*



■ Shielding Recognition result (Object)

Camera
(B-1-3)

1. Object adheres to front of sensor Cut-out scenario



Method

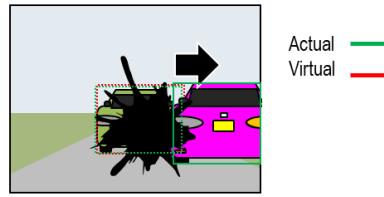
The vehicle traveling in front cuts-out from the shielded position. The vehicle traveling in front, and the vehicle in front of that, are both the recognition targets.

Judgment Criteria

- Difference in longitudinal distance: 5% or less
- Difference in lateral distance: 5% or less
- Difference in longitudinal relative speed: 10% or less
- Difference in lateral relative speed: 10% or less
- Difference in width/height: 5pix or less

Reference: Parameters of Validation Scenario

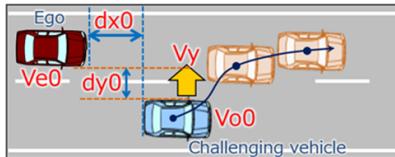
Parameter	Variable/Fixed	Range
Distance to the target	Variable	Longitudinal position dx0 [m] Longitudinal position dx0_f [m]
Relative velocity to the target	Variable	Longitudinal velocity Vo0-Ve0 [kph] Longitudinal velocity Vo0-Vf0 [kph] Lateral velocity Vy [kph]
Type of the target	Fixed	Shape: sedan Color: White
Degree of shielding of the detection-target due to adherence of foreign object	Variable	In relation to the bounding box of the detection-target ① Initial50% → Final0%



■ Shielding Recognition result (Object)

Camera
(B-1-3)

2. Obstruction in front of sensor Cut-in scenario



Method

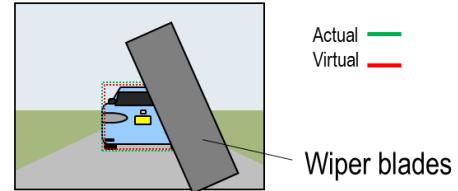
The recognition target cuts into the lane of the ego vehicle (in front) at a constant side-way speed, whilst wiper blades are in motion.

Judgment Criteria

- Difference in longitudinal distance: 5% or less
- Difference in lateral distance: 5% or less
- Difference in longitudinal relative speed: 10% or less
- Difference in lateral relative speed: 10% or less
- Difference in width/height: 5pix or less

Reference: Parameters of Validation Scenario

Parameter	Variable/Fixed	Range
Distance to the target	Variable	Longitudinal position dx_0 [m]
		Lateral position y_0 : 3.5m
Relative velocity to the target	Variable	Longitudinal velocity $Vo_0 - Ve_0$ [kph] Lateral velocity Vy [kph]
Type of the target	Fixed	Shape: sedan Color: White
Wiper blade movement	Fixed	1. Intermittent 2. Continuous



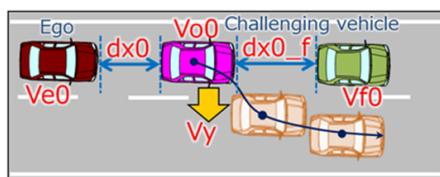
Actual
Virtual

Wiper blades

■ Shielding Recognition result (Object)

Camera
(B-1-3)

2. Obstruction in front of sensor Cut-out scenario



Method

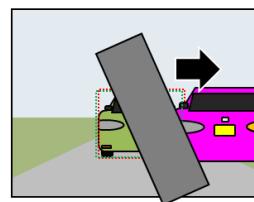
The recognition target cuts out while wiper blades are in motion.

Judgment Criteria

- Difference in longitudinal distance: 5% or less
- Difference in lateral distance: 5% or less
- Difference in longitudinal relative speed: 10% or less
- Difference in lateral relative speed: 10% or less
- Difference in width/height: 5pix or less

Reference: Parameters of Validation Scenario

Parameter	Variable/Fixed	Range
Distance to the target	Variable	Longitudinal position dx_0 [m]
		Longitudinal position $dx_{0,f}$ [m]
Relative velocity to the target	Variable	Longitudinal velocity $Vo_0 - Ve_0$ [kph] Longitudinal velocity $Vo_0 - Vf_0$ [kph] Lateral velocity Vy [kph]
Type of the target	Fixed	Shape: sedan Color: White
Wiper blade movement	Fixed	1. Intermittent 2. Continuous

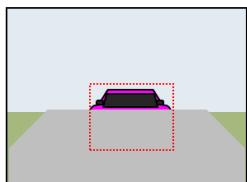
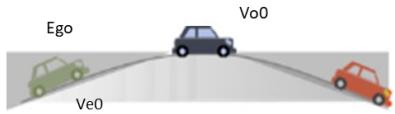


Actual
Virtual

■ Shielding Recognition result (Object)

Camera
(B-1-3)

3. Road surface shape incline Blind-spot (vertical) scenario



Method

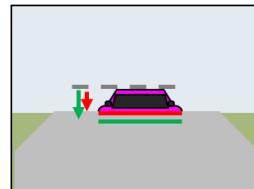
The ego vehicle travels along a road with a vertical incline (hump shape), and approaches the recognition target up ahead (in the ego-vehicle's driving lane), at a constant speed.

Judgment Criteria

- Difference in longitudinal distance: 5% or less
- Difference in lateral distance: 5% or less
- Difference in longitudinal relative speed: 10% or less
- Difference in lateral relative speed: 10% or less
- Difference in width/height: 5pix or less

Reference: Parameters of Validation Scenario

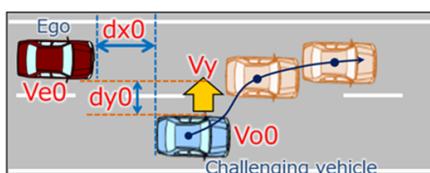
Parameter	Variable/Fixed	Range
Distance to the target	Variable	Longitudinal position dx_0 [m]
Relative velocity to the target	Variable	Longitudinal velocity $Vo_0 - Ve_0$ [kph]
Type of the target	Fixed	Shape: sedan Color: White
Road structure vertical incline	Fixed	Vertical cross sectional incline: 6%



Camera
(B-1-3)

■ Shielding Recognition result (Object)

4. Shielding by nearby moving objects (flying objects) Cut-in scenario



Method

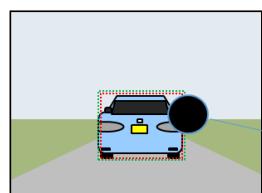
The recognition target cuts into the lane of the ego vehicle (in front) at a constant side-way speed, whilst a flying object crosses in front of the ego vehicle.

Judgment Criteria

- Difference in longitudinal distance: 5% or less
- Difference in lateral distance: 5% or less
- Difference in longitudinal relative speed: 10% or less
- Difference in lateral relative speed: 10% or less
- Difference in width/height: 5pix or less

Reference: Parameters of Validation Scenario

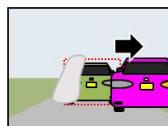
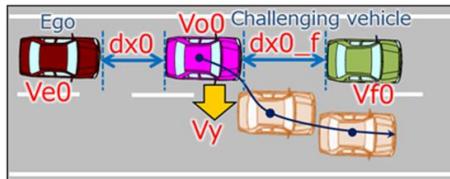
Parameter	Variable/Fixed	Range
Distance to the target	Variable	Longitudinal position dx_0 : 00~△△m
		Lateral position y_0 : 3.5m
Relative velocity to the target	Variable	Longitudinal velocity $Vo_0 - Ve_0$: 00~△△kph Lateral velocity Vy : 00kph
Type of the target	Fixed	Shape: sedan Color: White
Flying object	Variable	Size (diameter) : 00 to △△ cm Sideways velocity: 00kph



■ Shielding Recognition result (Object)

5. Adherence of foreign object to other vehicle Cut-out scenario

Camera
(B-1-3)



Reference: Parameters of Validation Scenario

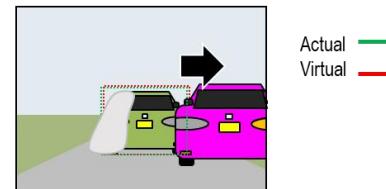
Parameter	Variable/Fixed	Range
Distance to the target	Variable	Longitudinal position dx_0 : ○○~△△m
		Longitudinal position dx_0_f : ○○~△△m
Relative velocity to the target	Variable	Longitudinal velocity Vo_0-Ve_0 : ○○~△△kph Longitudinal velocity Vo_0-Vf_0 : ○○~△△kph Lateral velocity Vy : ○○kph
Type of the target	Fixed	Shape: sedan Color: White
Degree of shielding of the detection-target due to partial shielding by a cover	Variable	30% to 70% shielding in relation to the vehicle width of the detection-target

Method

A vehicle traveling behind a recognition target which is covered by a cover, cuts out.

Judgment Criteria

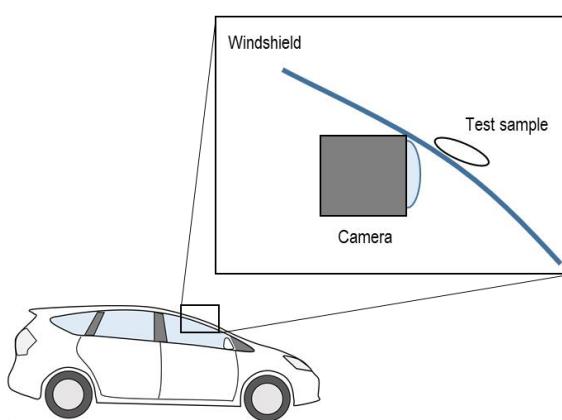
- Difference in longitudinal distance: 5% or less
- Difference in lateral distance: 5% or less
- Difference in longitudinal relative speed: 10% or less
- Difference in lateral relative speed: 10% or less
- Difference in width/height: 5pix or less



■ Shielding Placement verification (boundary line)/ color luminance verification (boundary line)

1. Object adheres to front of sensor: consistency of perception device

Camera
(B-1-4
B-1-5)



Method

A test sample is applied to the windshield of the vehicle (while in a stationary state), and validation is conducted by evaluating the strength of the signal and color.



Judgment Criteria

Difference in pixilation:

- within +/-5% to theoretical value or +/-5 pixels
- within +/-2% to actual or +/-2 pixels

■ Shielding Recognition result (boundary line)

Camera
(B-1-6)

Recognition Disturbance			Extracting Traits Shielding		
Modeling Categories	Categorization of Causes		Outline of Error Cause (Specific Examples)	(Cannot see subject)	
				Details of Error	Cannot see due to cause on vehicle side
Vehicle / Sensor	In front of sensor	① Adherence of foreign object / obstruction	Shielding: dirt, dust, etc.	Adherence of dirt, dust, etc. (lose full picture)	3
			Shielding: snow, ice, etc.	Adherence of snow, ice, etc. (lose full picture)	3
			Shielding: water, etc.	Adherence of water, etc. (lose full picture)	2
			Shielding: insect, bird poo, etc.	Adherence of insect, bird dropping, etc. (lose full picture)	2
Surrounding Environment	Surrounding structures	Shape	② Shielding: wiper blades	Wiper blades moving (lose full picture)	1
	Road surface			Difference in position/incline of road in picture	3
	Surrounding objects (moving)	Shielding	Non-transparent	Parked vehicle, surrounding trees, flying objects	3
	On track	Lane lines	Dirty, scraped	Fallen leaves, shielding due to snow pile, dirty, scraped, re-painted	3
Subject of Recognition	Moving	③ Other vehicles	Color	Standard color of adhering foreign object (similar or different to vehicle color)	1
				Various shapes of foreign objects adhere (shape/pattern of dirt, stickers, etc.)	—
			Shape		1
			Range	Area adhered (one location to full vehicle)	3

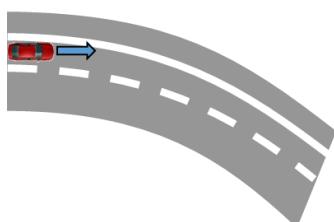
(Taken from Phenomenon and Cause Matrix)

Validation method to be identified for each of the above 3 categories

■ Shielding Recognition result (boundary line)

Camera
(B-1-6)

1. Object adheres to front of sensor: Lane-keeping scenario



Method

The ego vehicle travels at a constant speed keeping to its driving lane, whilst vision is impaired due to the adherence of a foreign object.

Judgment Criteria

- Difference in radius of curvature : 5% or less
- Difference in orientation : 5% or less
- Difference in position : 5% or less
- Target type is matching

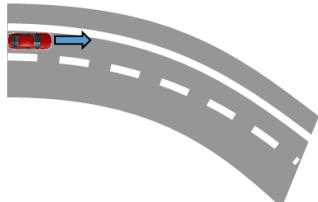
Reference: Parameters of Validation Scenario

Parameter	Variable/Fixed	Range
Velocity of ego vehicle	Fixed	Ve0 : 120kph
Width of driving lane	Fixed	3.5m
Curvature of lane	Fixed	ROO
Type of the target	Variable	Shape: solid line, dotted line Color: white, yellow
Amount which the ego vehicle's driving lane marking lines are shielded due to the adherence of a foreign object (disturbance)	Fixed	Amount of shielding: 50%

■ Shielding Recognition result (boundary line)

Camera
(B-1-6)

2. Obstruction in front of sensor : Lane-keeping scenario



Method

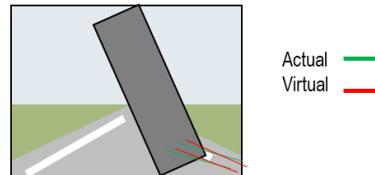
The ego vehicle travels at a constant speed, staying in its driving lane, whilst the wiper blades are in motion.

Judgment Criteria

- Difference in radius of curvature : 5% or less
- Difference in orientation : 5% or less
- Difference in position : 5% or less
- Target type is matching

Reference: Parameters of Validation Scenario

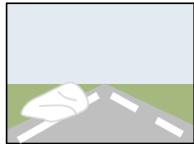
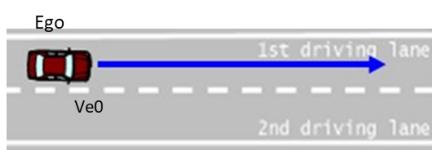
Parameter	Variable/Fixed	Range
Velocity of ego vehicle	Fixed	Ve0 : 120kph
Width of driving lane	Fixed	3.5m
Curvature of lane	Fixed	ROO
Type of the target	Variable	Shape: solid line, dotted line Color: white, yellow
Wiper blade movement	Fixed	1. Intermittent 2. Continuous



■ Shielding Recognition result (boundary line)

Camera
(B-1-6)

3. Dirty marking line: Lane-keeping scenario



Method

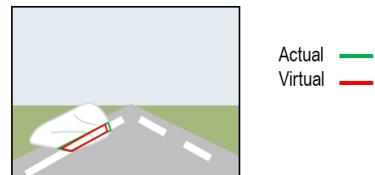
The ego vehicle travels at a constant speed keeping to its driving lane, whilst sections of the lane marking line are shielded by fallen leaves on the road, piles of snow, etc.

Judgment Criteria

- Difference in amount of shielding of marking line : pixels 5% or less
- Difference in radius of curvature : 5% or less
- Difference in orientation : 5% or less
- Difference in position : 5% or less
- Target type is matching

Reference: Parameters of Validation Scenario

Parameter	Variable/Fixed	Range
Velocity of ego vehicle	Fixed	Ve0 : 120kph
Width of driving lane	Fixed	3.5m
Curvature of lane	Fixed	ROO
Type of the target	Variable	Shape: solid line, dotted line Color: white, yellow
Amount which the ego vehicle's driving lane marking lines are shielding due to adherence of a foreign object (disturbance)	Fixed	Amount of shielding: 50%

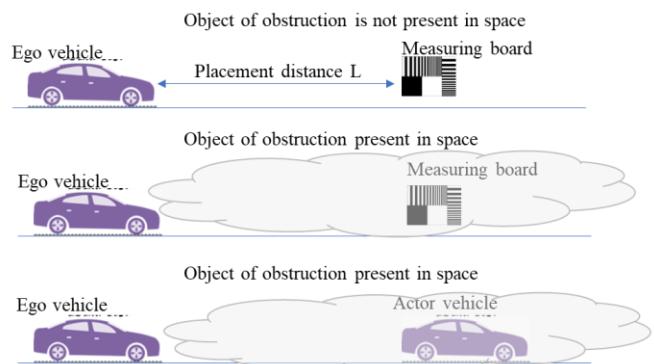


■ Camera: Frequency within Space, Low Contrast – Simulating the Disturbance Phenomenon

Camera
(B-2)

Validation Method

- The evaluation scenario assumes validation is conducted with varied disturbance parameters to create conditions such as performance limitation(s)
- The tool is a combination of disturbance parameters in the scenario that can be supported (=can be measured in actual conditions) and confirmation the disturbances are reproduced



(Validation 1) From Perception Device

- Spatial frequency and contrast are verified by placing a dedicated measurement board and swinging a disturbance factor parameter
*Confirmed by checking the RAW data for brightness

(Validation 2) From Recognition Unit

- The ego vehicle approaches the stationary actor vehicle, and the recognition results are checked and swinging a disturbance factor parameter

Judgement Criteria

(Validation 1) Difference in spatial frequency and contrast between SIM & Actual to be within $\pm 5\%$ deviation

Measuring conditions: average measured value and theoretical value to fall within $\pm 5\%$ deviation

(Validation 2) Difference in distance at which the vehicle can “recognize” between SIM & actual to be within $\pm 5\%$ deviation

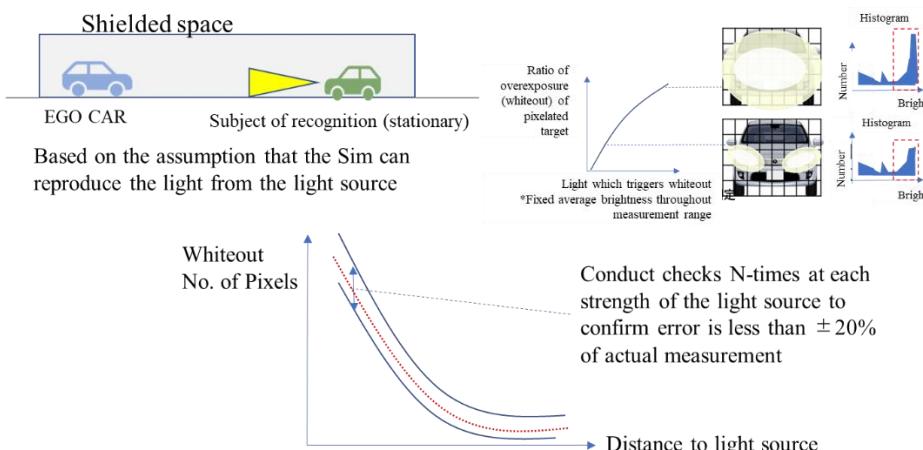
■ Camera: Excessive (saturation) Whiteout

Camera
(B-3)

1. Dynamic Test (Perceptual Device)

Validation Method ① Confirming Dynamic Range

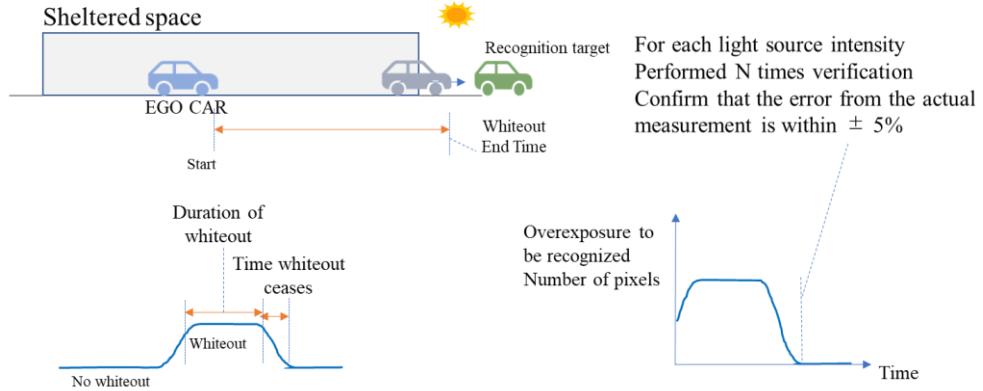
Use a histogram to confirm the change in whiteout (brightness) in relation to the degree of brightness and changes in brightness of a stationary light source in a shielded space.



Validation Method ② Exposure Control: Confirming Feedback Speed

Confirm the time taken until whiteout occurs in a shielded space, to the time it stops

Check the time until the end of overexposure in the shielded space Place the stopped vehicle outside the shielded space, and the EGO vehicle approaches the stopped vehicle at a constant speed. Measure the time from the exit to the end of overexposure



Annex G

Validation of Simulation Tools and Simulation Test Methods Related to UN Regulation No. 157

G.1 Purpose and Scope

To summarize the concepts behind the validation technique for simulation tools and simulation test methods used in compliance testing for the traffic disturbance scenario defined in UN Regulation No. 157 (low-velocity ALKS). Note that errors in the perception unit are not taken into account (100% recognition is assumed), with the subjects of evaluation being the main AD control system (Planer) and vehicle motion control system (Fig. G-1).

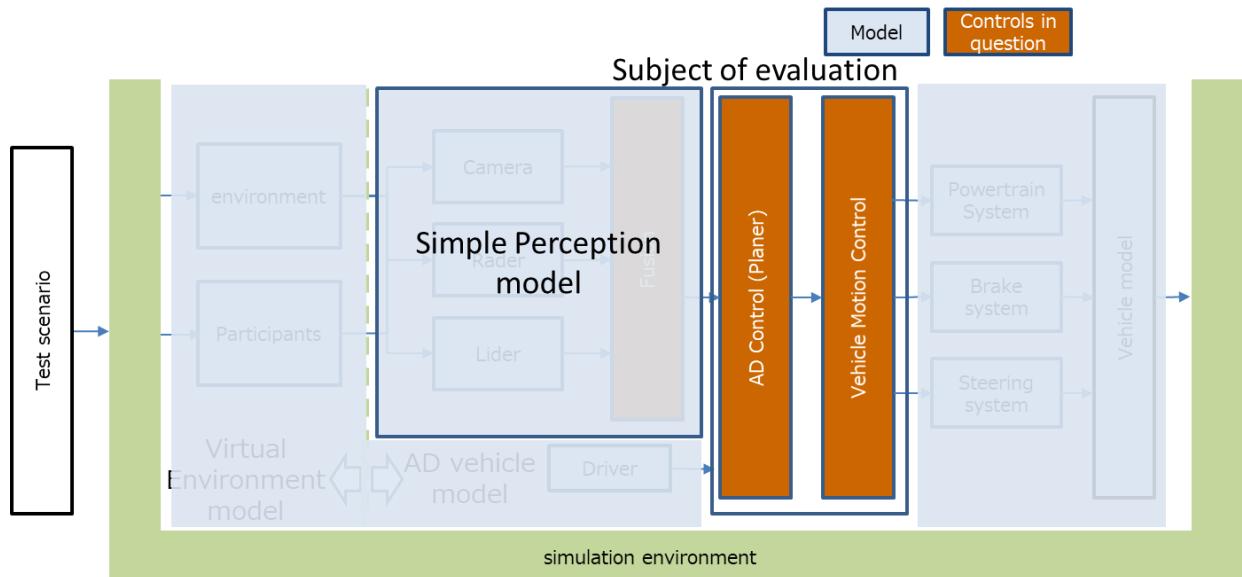


Figure G-1. Control Systems (Subject of Evaluation) in the Traffic Disturbance Scenario

G.2 Terminology

Following are the definitions of the terminology used in this chapter.

(A) **Automated Driving System (ADS)**

A system that has the function to perform a part or all of the driving required by the driver on behalf of the driver by performing a part or all of a dynamic driving task (DDT) by automatically identifying driving conditions, making decisions, and controlling the steering.

(B) **Parameters**

Physical quantities (e.g., vehicle velocity and distance) used for measuring data, conducting simulations, etc.

(C) **Calculated Value**

Value determined from the results of calculations performed using the simulation tool.

(D) **Provided Value**

Value provided by the scenario.

(E) **Scenario**

A scene that incorporates one (or more) ADS and one (or more) target vehicle while performing a specified DDT and the narrative of the subsequent interactions that arise thereafter.

In this section, this is the narrative formed by the evaluation conditions when conducting actual tests and simulations, including the initial conditions of the ego and other vehicles (vehicle velocity, longitudinal distance, etc.), behavior of other vehicles (cut-in, etc.), and road conditions (number of driving lanes, road width, etc.).

(F) Preventable Threshold

The threshold between “no collision” and “everything other than no collision (collision, etc.)” shown by the graphs under “5. Reference” in Appendix 3 Guidance on Traffic disturbance critical scenarios for ALKS of the UNR-157.

G.3 Method for Validating the Simulation Tool

G.3.1 Purpose of This Chapter

This chapter describes the process and requirements for determining whether the simulation tool can accurately reflect an actual test. Before running a simulation test, this confirmation must be completed.

G.3.2 Validation Method and Criteria

Following parts describe the method and criteria used for validating simulation tools, along with justification.

Validation Method

Apply the same environmental information from the actual test for the selected scenario to the simulation and then compare the relative distance to other vehicles (hereinafter “longitudinal distance”).

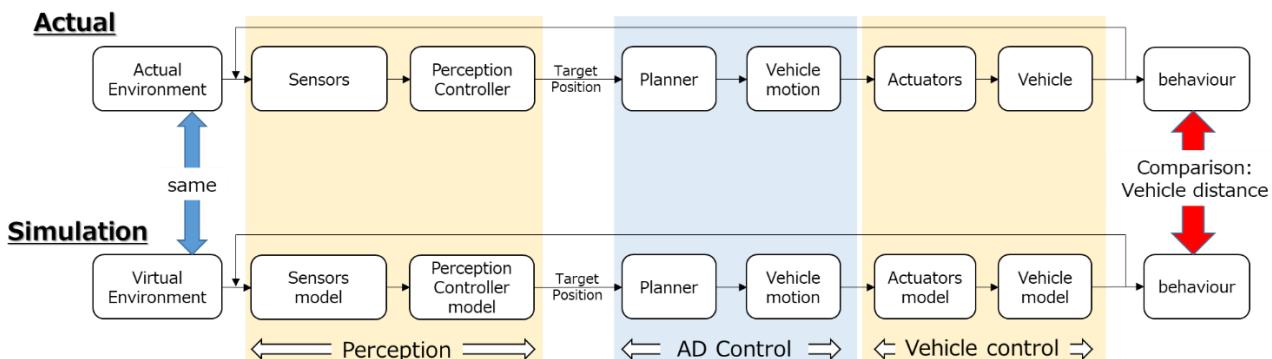


Figure G-2. ADS Structure

Justification of the Concept:

The compliance test in question determines whether the ego vehicle will (or will not) collide with other vehicle. Therefore, the simulation tool must be capable of accurately simulating longitudinal distances (physical quantification for determining whether a collision has or has not occurred). Furthermore, the ambient circumstances that compose the “inputs,” such as the location of the preceding vehicle, must be equal to accurately compare the outcomes of the acutual test and simulation. Based on this, it is possible to conclude that the aforementioned validation method can demonstrate the simulation tool’s suitability for this purpose.

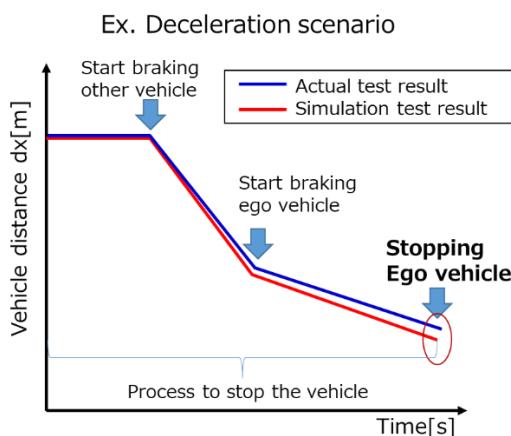
Criteria

When the ego vehicle reaches a stationary or steady state¹, the resulting longitudinal distance² between the ego and target vehicles for which the collision is being avoided must be greater in the actual test than that in the simulation tool. Here, we compare the “no-collision (preventable)” territory and process leading up to the ego vehicle reaching a stationary or steady state to be used as a reference.

Furthermore, to demonstrate that the above criteria have been met, the simulation tool itself must first satisfy “3.3 Simulation Tool Requirements.”

¹ “Steady state” refers to the state where there is no longer a difference in velocity between the ego and target vehicles as a result of the ego vehicle’s collision avoidance behavior.

² Longitudinal distance refers to the length of the perpendicular distance line created from the front end of the ego vehicle to the rear end of the target vehicle.



Justification of the Concept:

To “confirm that the test results for collision/non-collision by the ADS are always superior to the results of the criteria for collision/non-collision (i.e., the purpose of the compliance test),” the success or failure of actual avoidance performance for a particular test scenario can be demonstrated by showing that results calculated using the simulation tool are always superior to the criteria, as long as the simulation test results show that the actual test results will always perform better.

G.3.3 Simulation Tool Requirements

The simulation tool must conform to the following two requirements to be valid.

Requirement 1: The simulation tool must calculate and output the parameters that influence the determination of whether a collision occurred.

(For the parameters that contribute to each scenario, refer to “Attachment 1. Scenario-Specific Parameters of Impact”)

Requirement 2: To be able to compare calculation results, it must be proved that “a correlation exists¹” between the parameters calculated and those assessed via actual tests.

¹ “A correlation exists” does not mean that the calculated parameter values perfectly match, but rather that the changes in the parameters vary in a similar way.

G.4 Procedure for Validating the Simulation Tool

G.4.1 Purpose of This Chapter

This chapter describes the steps that lead to validating the simulation tools using the technique described in the previous chapter.

G.4.2 Procedure for Validating the Simulation Tool

① Choose the Scenario and Parameters that will be Used to Confirm the Validity

From the list of scenarios necessary for the compliance test, select the scenario(s) and parameters to be used to confirm the validity (refer to G.5 ADS Safety Performance Evaluation Simulation Method).

INPUT: the scenario and parameter range listed under “G.5 ADS Safety Performance Evaluation Simulation Method.”

OUTPUT: chosen scenario and parameters for validation

NOTE: For low-velocity ALKS, ADS avoidance behavior is limited to “deceleration” (avoidance by steering does not occur); therefore, a scenario and characteristics that demonstrate the correlation in ADS deceleration performance between the actual and simulation tests should be chosen. The maximum deceleration by ADS “G” should ideally be included in the range of deceleration performance to be compared.

② Preliminary Actual Test

Perform an actual test before conducting a validation test to measure each parameter required to be input/adjusted in the simulation tool.

INPUT: selected performance characteristics that impact the results of the simulation tool

OUTPUT: actual test data to be used for adjusting the characteristics of the vehicle model

③ Input and Adjust the Settings for the Simulation Tool and Environment

Input and adjust the settings (e.g., braking performance) based on the specifications of the target vehicle to be used in the simulation (e.g., vehicle weight) and the data obtained from “② Preliminary Actual Test.”

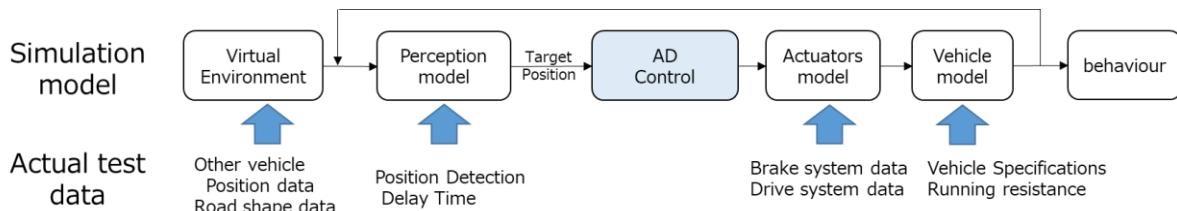
INPUT: actual test data to be used for adjusting the characteristics of the vehicle model

OUTPUT: the simulation tool and environment where the settings have been input and adjusted

NOTE: Adjusting the simulation tool refers to adjusting the perception and vehicle unit models from the preadjusted state to the state where they are aligned with the actual conditions to satisfy the criteria and simulation tool requirements for validation described in Chapter 3.

An example of inputting and adjusting settings

Input and adjust the settings of the perception and vehicle unit models using the measurement data obtained from “② Preliminary Actual Test.”



NOTE: If the perception unit's responsiveness is based on a "time delay," correlation (validity) must be confirmed by matching the increase/decrease in the timing of the longitudinal/lateral position and velocity of the target vehicle as recognized by the actual perception unit to the actual physical measurements of the position and velocity of the target vehicle.

④ Actual Test for Confirming Validity

Conduct an actual test based on the scenario selected in “① Choose the Scenario to be Used to Confirm Validity” above.

INPUT: test scenario and parameters (test conditions) under which the actual test will be performed

OUTPUT: actual measurement data to be used for confirming the validity of the respective test scenario

⑤ Simulation for Confirming Validity

Conduct simulation based on the scenario selected in “① Choose the Scenario to be Used to Confirm Validity” above.

INPUT: actual test measurement parameters for respective test scenarios, simulation, and environment for which settings have been input and adjusted

OUTPUT: simulation data to be used for confirming the validity of respective test scenarios

NOTE: Information on other vehicles to be input into the simulation can be created based on the position data of each test vehicle positioned using GNSS, for example, during the actual test conducted in ④.

⑥ Confirming the Validity of the Simulation Environment

Compare the results from ④ and ⑤ to confirm the validity of the simulation environment.

INPUT: actual measurement and simulation data for confirming the validity of each respective test scenario

OUTPUT: the result of confirming the validity of the simulation environment

NOTE: The procedure does not necessarily proceed in order from ① to ⑤, but rather it may repeat from ② to ⑤ until the judgment criteria are satisfied.

G.5 ADS Safety Performance Evaluation Simulation Method

G.5.1 Purpose of This Chapter

To discuss the simulation test method used to ensure that the compliance test's pass/fail criteria are met (i.e., confirm that the test results for collision/non-collision by the ADS are always superior to the results of the criteria for collision/non-collision) using the validated simulation tool.

G.5.2 Test Method

Adopt the environment described in “Simulation tools and implementation environment (G.6 Submission Documents-3),” with the simulation input comprising a combination of the following two items:

1. The scenario, in other words, the allocation and behavior of the ADS-equipped ego vehicle (hereinafter “ego vehicle”) and surrounding vehicles (hereinafter “other vehicles”).

Following are the eligible scenarios:

(a) Cut-in scenario [No.1]

(b) Cut-out scenario [No.2]

(c) Deceleration scenario [No.4]

*The number within the [] corresponds to the numbers within the Figures in Attachment 2 “Hazardous Scenarios.”

2. The parameters of the ego and other vehicles within the scenario

- ① The velocity of the ego and other vehicles
- ② Acceleration/deceleration velocity of the ego and other vehicles
- ③ Distance between the ego and other vehicles

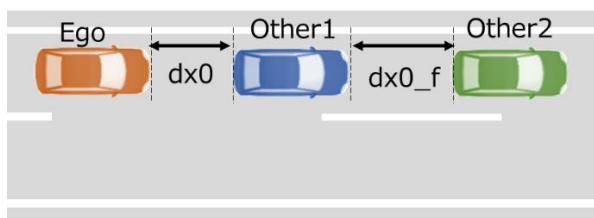
Next are the definitions of each scenario used above and parameters of the ego and other vehicles within the given scenarios.

G.5.3 Definition of the Parameters of the Ego and Other Vehicles

- ① Basic Definition of Initial Longitudinal Distance (dx_0)

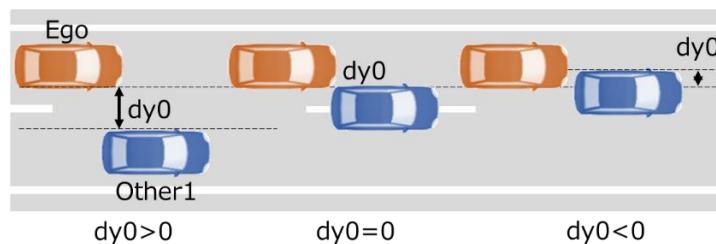
The initial longitudinal distance is the length of the perpendicular distance line created from the front end of one vehicle to the rear end of another.

The distance between the ego vehicle and the vehicle in front of the ego vehicle (“other vehicle 1”) is shown as dx_0 (m), with the distance between “other vehicle 1” and the vehicle in front (“other vehicle 2”) shown as dx_{0_f} (m).



- ② Basic Definition of Initial Lateral Distance (dy_0)

Lateral distance is the length between the edge lines of the adjacent sides of two vehicles. The sign preceding the value will be “plus” if the “other vehicle 1” does not overlap with the ego vehicle and “minus” if there is an overlap. Thus, if the value is “0,” the two perpendicular distance lines perfectly overlap.



- ③ Basic Definition of Initial Velocity

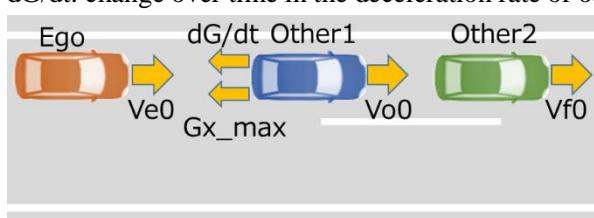
Ve_0 (km/h): initial velocity of the ego vehicle

Vo_0 (km/h): initial velocity of the preceding vehicle (other vehicles 1) in the ego lane or adjacent lane

Vf_0 (km/h): initial velocity of other vehicles 2

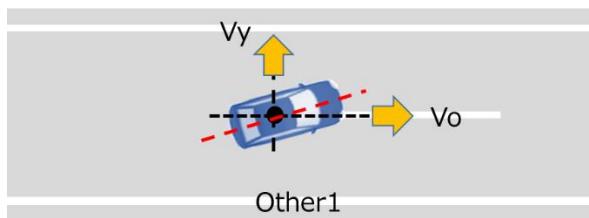
$G_x \max$ (G): deceleration rate of other vehicles 1

dG/dt : change over time in the deceleration rate of other vehicles 1



④ Basic Definition of Lateral Velocity

V_y (m/s): lateral velocity of other vehicle 1 and the velocity perpendicular to the lane line

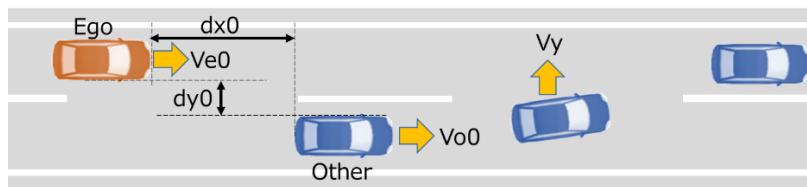


*Refer to Attachment 3, "Definition of the Behavior of Other Vehicles," for more details

G.5.4 Definition of Each Scenario

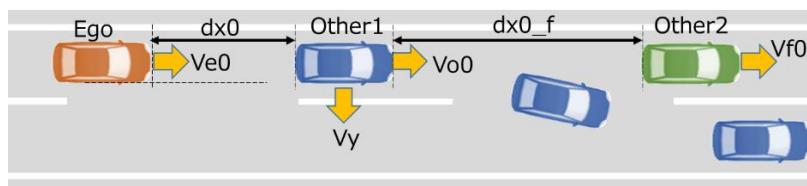
(a) Cut-in Scenario

The "parameters of the ego and other vehicles" as defined in G.4.2 are used as follows:



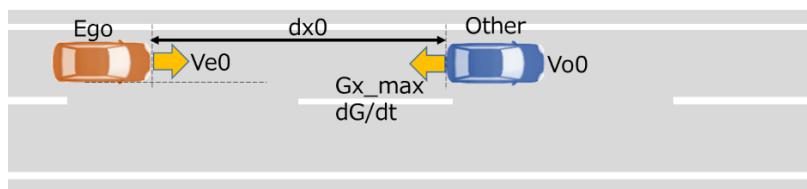
(b) Cut-out scenario

In this scenario, the "parameters of the ego vehicle, other vehicles 1, and other vehicles 2" as defined in G.4.2 are used as follows:



(c) Deceleration scenario

In this scenario, the "parameters of the ego and other vehicles" as defined in G.4.2 are used as follows:



G.5.5 Criteria for Pass or Fail

The collision must not occur within the preventable range (no-collision territory) as defined in "5. Reference" in Appendix 3 Guidance on Traffic disturbance critical scenarios for the ALKS of UNR-157.

G.5.6 Parameter Range for Simulations

- ① Parameter Values and Ranges Common Across Scenarios
 (1) Road parameter values

Road Parameters	Value	Unit
Number of lanes	2	-
Road width	3.5	m
Road friction coefficient	1.0	μ
Horizontal gradient	0	%
Vertical gradient	0	%
Curve radius	∞	%

- (2) Vehicle parameters

Vehicle Parameters	Ego Vehicle	Other Vehicle 1	Other Vehicle 2
Vehicle width	(According to the application vehicle)	1.9 m	1.9 m
Vehicle length	(According to the application vehicle)	5.3 m	5.3 m
Shape	Rectangular	Rectangular	Rectangular
Position of travel	Middle of the lane	Middle of the lane	Stationary in the middle of the lane

② Scenario-Specific Parameter Ranges

The parameter ranges listed under (1) to (3) below form the basic parameter ranges. However, this can be individually set based on the applicant's driving environment conditions, etc.

(1) Parameter ranges for cut-in scenario

Parameter	Range
Ve0 [Initial velocity of ego vehicle]	$20 \leq Ve0 \leq [60] \text{ km/h}$
Ve0 – Vo0 [Relative velocity]	$0 \leq Ve0 - Vo0 \leq 40 \text{ km/h}$ * ¹
dx0 [Initial longitudinal distance]	$0 \leq dx0 \leq 60 \text{ m}$
dy0 [Initial lateral distance]	$\{(3.5\text{-ego vehicle width})/2 + 0.8 \text{ (other vehicle side)}\} \text{ m}$
Vy [Lateral velocity]	$0 < Vy \leq 3.0 \text{ m/s}$

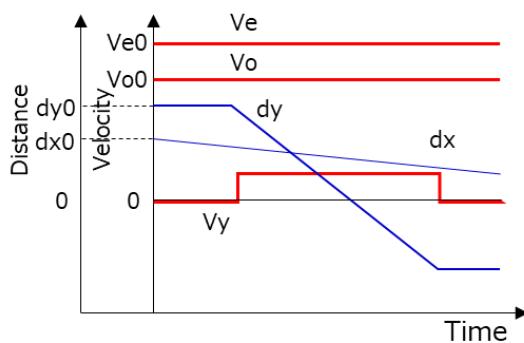
The value given in [] is the maximum designed velocity of the ego vehicle

*¹ Do not include cases where the velocity of the cut-in vehicle is greater than the velocity of the ego vehicle.

Note: When the cut-in vehicle's velocity is slower, do not include lateral velocity values, which are physically impossible. (For example, a combination such as "vehicle velocity 10 km/h (2.78 m/s) wherein the lateral velocity is 3 m/s.")

Note: When the range of movement of autonomous driving (the subject of application) is limited to only when the ego vehicle is tracking the vehicle in front, do not include the combinations of the lateral velocity and longitudinal distance of the cut-in vehicle for which cut-in would occur in front of the preceding vehicle or "into" the preceding vehicle (collision).

E.g., the change in cut-in parameters over time.



*Refer to Attachment 1(a) for the parameters over a time series

(2) Parameter ranges for cut-out scenario

Parameter	Range
Ve0 [Initial velocity of ego vehicle]	$10 \leq Ve0 \leq [60] \text{ km/h}$
Vo0 [Velocity of preceding vehicle]	$10 \leq Vo0 \leq [60] \text{ km/h}^{*2}$
Vf0 [Initial velocity of other vehicle]	0 km/h
dx0_f [Initial longitudinal distance]	$0 < dx0_f \leq 100 \text{ m}$
Vy [Lateral velocity]	$0 < Vy \leq 3.0 \text{ m/s}$

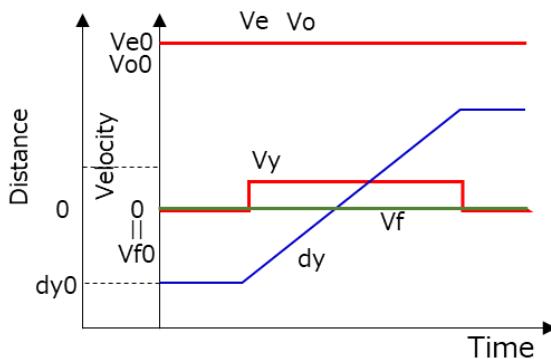
The value given in [] is the maximum designed velocity of the ego vehicle

*2 Velocity of the preceding vehicle = velocity of the ego vehicle

Note: When the velocity of the cut-out vehicle is slower, do not include lateral velocity values that are physically impossible. (For example, a combination such as “vehicle velocity 10 km/h (2.78 m/s) wherein the lateral velocity is 3 m/s.”)

Note: When considering the “longitudinal distance,” do not include conditions where the cut-out vehicle collides with the stationary vehicle.

E.g., the change in cut-out parameters over time.



*Refer to Attachment 1(b) for the parameter over a time series

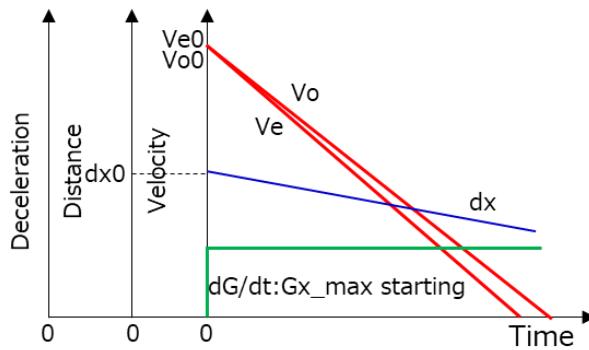
(3) Parameter ranges for deceleration scenario

Parameter	Range
Ve0 [Initial velocity of ego vehicle]	$10 \leq Ve0 \leq [60] \text{ km/h}$
Vo0 [Velocity of preceding vehicle]	$10 \leq Vo0 \leq [60] \text{ km/h}^{*3}$
Gx_max [Deceleration velocity of preceding vehicle]	$0 < Gx_{\max} \leq 1.0G$
dG/dt [Rate of change in the deceleration velocity of other vehicles]	Limitless

The value given in [] is the maximum designed velocity of the ego vehicle

*3 Velocity of the preceding vehicle = velocity of the ego vehicle

E.g., the change in deceleration parameters over time.



*Refer to Attachment 1(c) for the parameter over a time series

G.5.7 Conducting Simulation

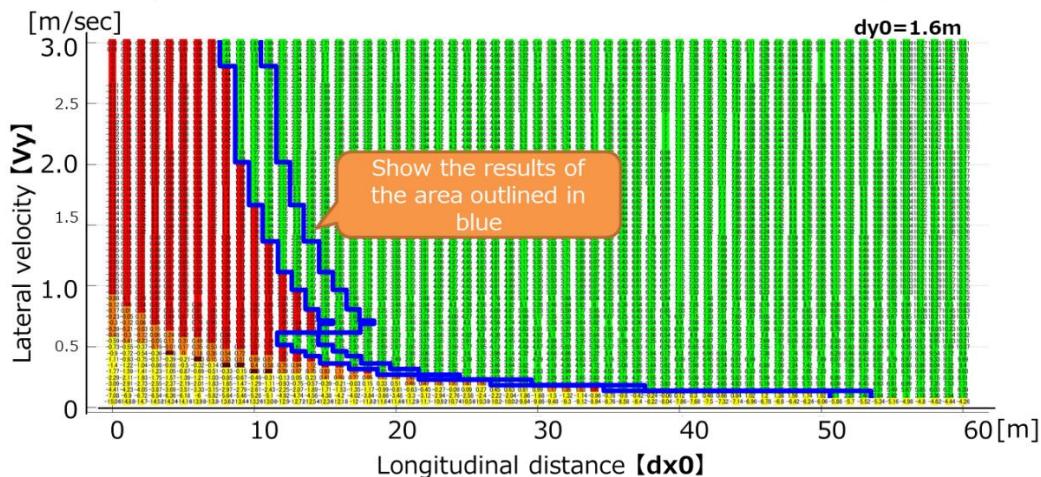
Conduct simulations based on the following ranges.

(1) Close to the Preventable/Unpreventable Threshold

Concerning the cut-in and cut-out scenarios, confirmation is to be conducted at +1 and +2 m from the threshold line from the borderline of pass/fail toward the direction in which the longitudinal distance becomes greater to confirm a broader range of collision/avoidance (not only limited to nearby the threshold line).

NOTE: The minimum increment of the lateral velocity is 0.1 m/s intervals.

Example of cut-in: ego vehicle velocity (Ve_0) = 30 km/h, other vehicle velocity (Vo_0) = 10 km/h.



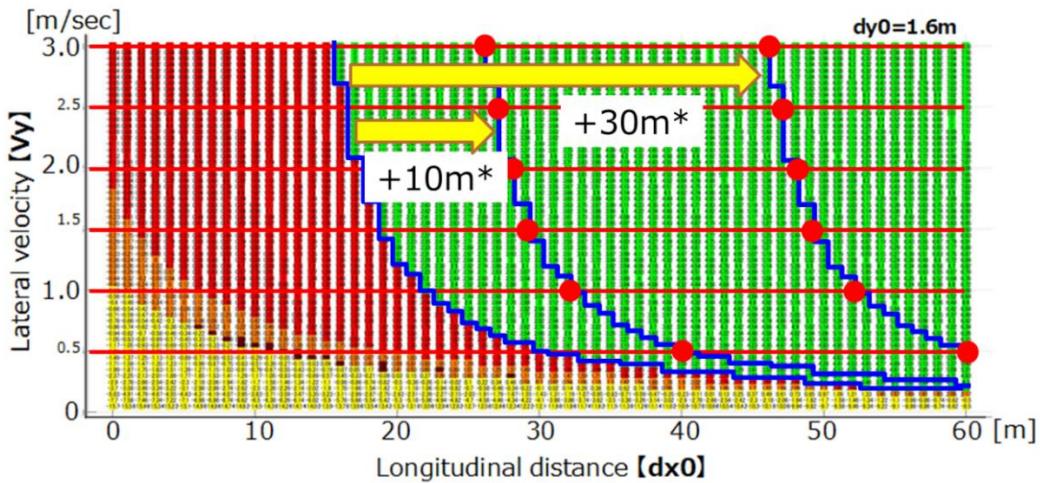
(2) Preventable Territory

Concerning the cut-in and cut-out scenarios, to also confirm that collision will not occur at random points within the preventable territory other than solely near the threshold line of preventable and unpreventable (i.e., to ensure a complete result), confirmation is to be additionally conducted at expanding intervals from the threshold line between unpreventable and preventable (pass/fail criteria) at +10 and +30 m. The reason for selecting "+10 m" and "+30 m" is to ensure that confirmation is not only in a limited number of points close to the center of the preventable range but also points at which the distance between vehicles is large.

Furthermore, the ego vehicle velocity and relative velocity combination cover the full range of combinations possible within the ODD range.

NOTE: Lateral velocity is to be at the increments of 0.5 m/s; if these increments are impossible, conduct based on possible increments.

Example of cut-in: ego vehicle velocity (V_{e0}) = 60 km/h, other vehicle velocity (V_{o0}) = 30 km/h.



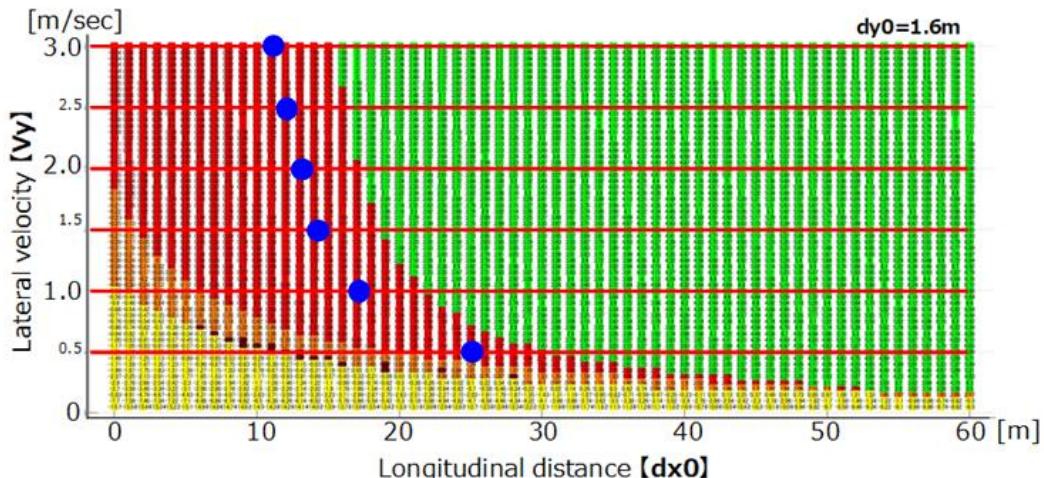
(3) Unpreventable (Collision) Territory

Confirm (for cut-in only) that best effort (=controls for collision avoidance are not stopped) within the unpreventable territory. The points to be used for the distance between vehicles within the unpreventable territory are up to each company's discretion.

Further, the ego vehicle velocity and relative velocity combination is to cover the full range of combinations possible within the ODD range.

NOTE: Lateral velocity is to be at the increments of 0.5 m/s. Avoidance is allowed.

Example of cut-in: ego vehicle velocity (V_{e0}) = 60 km/h, other vehicle velocity (V_{o0}) = 30 km/h.



In this example, if considerably distant from the preventable/unpreventable threshold, the higher is the likelihood that side collision or collision before deceleration will occur; thus, the points selected here for the distance between vehicles are, beginning from the threshold line, uniformly shortened at 5 m increments based on the average vehicle length of 5 m.

G.6 Submission Documents

The following documents must be submitted when conducting the compliance test.

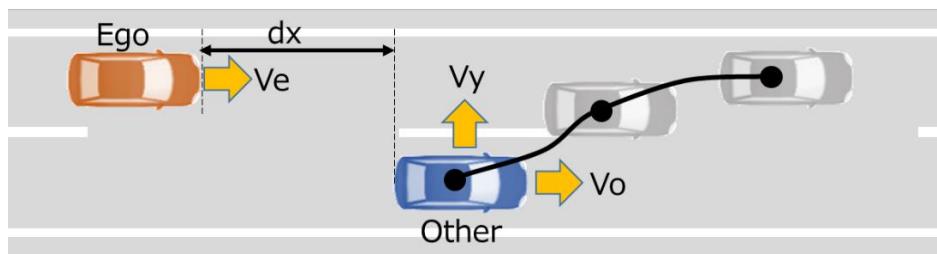
1. Test results confirming the validity of the simulation tool (Chapter G.4)
2. Simulation test and judgment results related to the ADS safety evaluation (Chapter G.5.7)
3. Simulation tools and implementation environment
Structure of the hardware and software and structure of the simulation test tool and model

NOTE: Detailed information related to the test vehicle is explained under TRIAS 48-J122-01, TRIAS 48-R157-01 Appendix 1 “1. Test Vehicle and Test Conditions.”

Attachment 1. Scenario-Specific Parameters of Impact

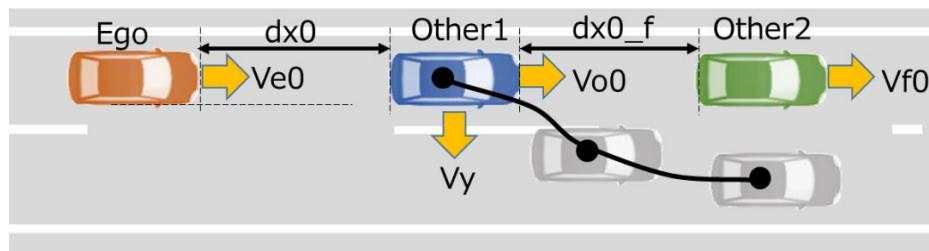
(a) Cut-in Scenario

Parameter	Attribute
Ego vehicle velocity [Ve]	Calculated value
Longitudinal distance between the ego and other vehicles [dx]	Calculated value
Other vehicle lateral velocity [Vy]	Provided value
Other vehicle velocity [Vo]	Provided value



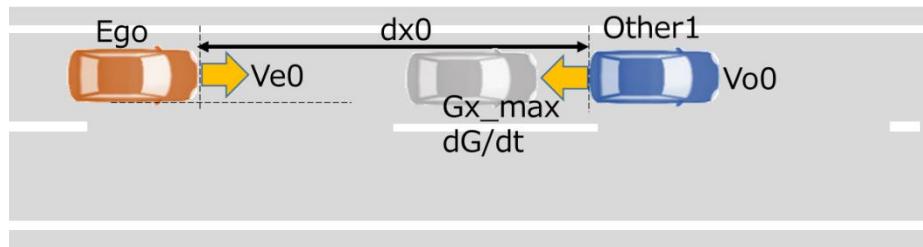
(b) Cut-Out Scenario

Parameter	Attribute
Longitudinal distance between the ego vehicle and other vehicle 1 [dx]	Calculated value
Longitudinal distance between the other vehicle 1 and 2 [dx_f]	Calculated value
Ego vehicle velocity [Ve]	Calculated value
Other vehicle 1 lateral velocity [Vy]	Provided value
Other vehicle 1 velocity [Vo]	Provided value



(c) Deceleration Scenario

Parameter	Attribute
Longitudinal distance between the ego and other vehicles [dx]	Calculated value
Ego vehicle velocity [Ve0]	Calculated value
Other vehicle deceleration velocity [Gx_max]	Provided value
Other vehicle velocity [Vo0]	Provided value



The tool must be equipped with the simulation elements required to calculate and output the above

Attachment 2. Hazardous Scenarios

		Surrounding Traffic Participants' Position and Behavior				
		Cut in	Cut out	Acceleration	Deceleration (Stop)	
Road geometry	Ego-vehicle behavior					
Main roadway	Lane keep					
	Lane change					
Merge	Lane keep					
	Lane change					
Branch	Lane keep					
	Lane change					

Attachment 3. Definition of the Behavior of Other Vehicles

This evaluation compares the ADS and preventable and unpreventable criteria when dealing with the behavior of other vehicles that obstruct the ego vehicle. Thus, the behavior of the other vehicles must be applied under the same conditions. The following defines the model and behavior of the other vehicle(s) to align them with the graph as shown in the “5. Reference” of Appendix 3 of UN-R157.

- “Other vehicle(s)” are to be mass models
- Lateral speed for cut-in and cut-out is applied using a step function
- Initial velocity (V_{o0}) is to be maintained for the longitudinal velocity during cut-in and cut-out
- Deceleration rate in the deceleration scenario is to be applied using the step function (jerk [dG/dt] is ∞)
- The direction of travel (the orientation of the composite vector formed by V_o and V_y) is to be taken as the orientation of the vehicle during cut-in and cut-out

