

# Universal Vehicle Protocols vs. Local Realities:

Systematic Gaps and Safety Risks in AV  
Deployment across Developing Regions

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# The “Reality Gap” in Design



**The Training Ground (Ideal)**

Standardized, predictable, rule-based.



**The Deployment Reality (LMIC)**

High entropy, mixed traffic, chaotic.

# Research Framework



## Research Questions

**Q1:** Does safety performance significantly degrade in LMIC environments compared to ideal testbeds?

**Q2:** Which local infrastructure, human, or policy factors have the greatest impact on AV safety and adoption outcomes?

**Q3:** What design, regulatory, or educational changes are required for more equitable and effective AV deployment in underdeveloped countries?



## Scope (PICO)

**P**opulation: Autonomous Vehicles (L3–L5)

**I**ntervention: Operation in LMIC/Chaotic roads

**C**omparator: Operation in Ideal/High-Income roads

**O**utcome: Safety metrics (Crash/Disengagement rates)

# The Human-Machine Mismatch



## Universal Protocol

Algorithms assume "universal order"—clear lanes, static signs, and compliant behavior.



## Environmental Entropy





LMIC roads present high variability, perceptual noise, and informal human behavior.

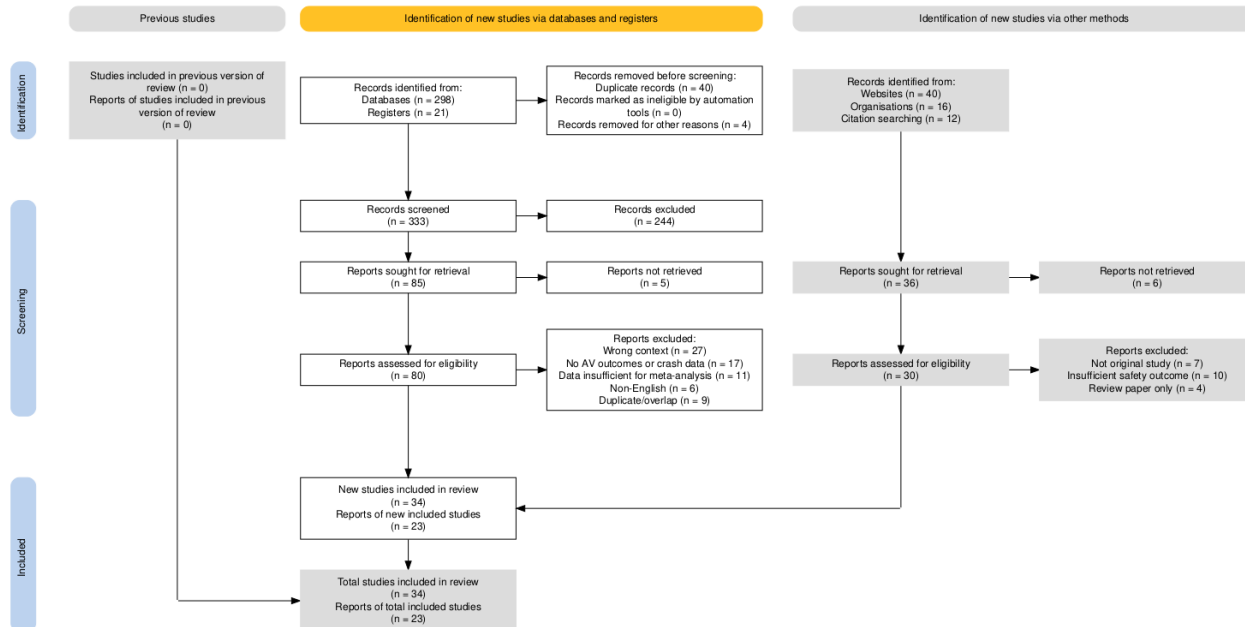


## Perceptual Breakdown

The mismatch leads to sensor confusion, system disengagement, and safety failures.

# Systematic Review (PRISMA)

-  **Rigorous Filtering:** Started with 500+ records from IEEE, Scopus, and TRID.
-  **Final Sample:** N = 34 high-quality studies included for quantitative synthesis.
-  **Quality Control:** Studies screened for quantitative safety data in non-ideal environments.
-  **Method:** Random-Effects Meta-Analysis (REML) with HKSJ adjustments.



# Meta-Analysis

## Pooled Random-Effects Meta-Analysis:

- Pooled log RR ( $\mu$ ): 0.366
- 95% CI (log): -0.552 to 1.284 (HKSJ)
- $\tau^2$  (REML): 1.399
- $I^2$ : 98.5%
- 95% PI (log): -2.516 to 3.248

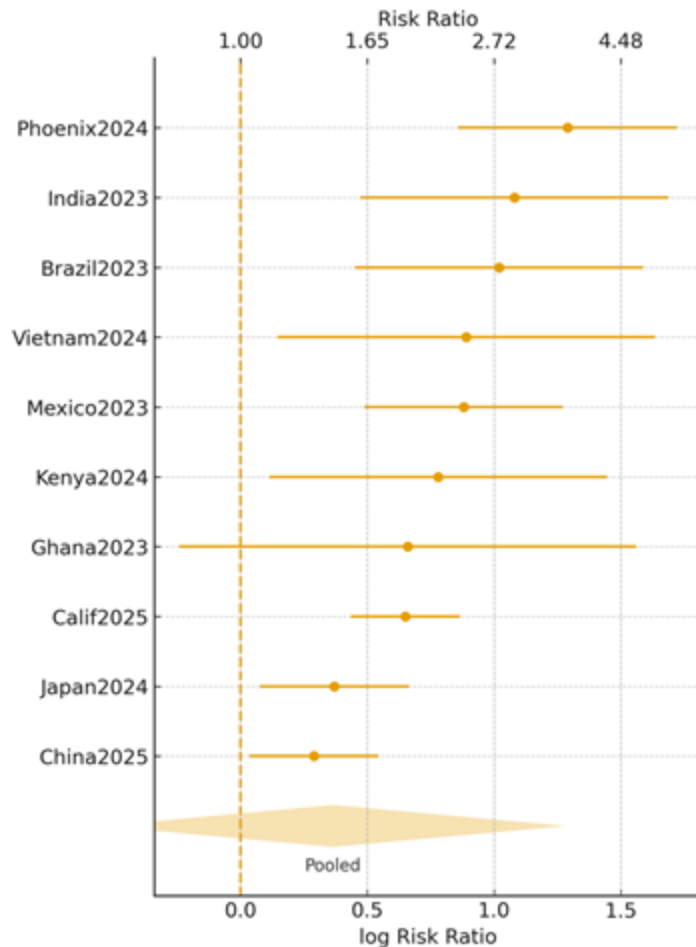
- **High Heterogeneity ( $I^2=98.5\%$ ):** The vast inconsistency in safety results is not random; it is driven by environmental factors.
- **The "Variance" Drop:** In meta-regression, when controlling for **Road Quality** and **Detection Failures**, the between-study variance ( $\tau^2$ ) dropped to nearly **zero** ( $\approx 0$ ).
- *Interpretation:* Local infrastructure quality explains almost *all* the performance difference between studies.

# Environment as Moderator

Study Location	Road Quality	Lane Markings	Detection Failure Rate	Risk Outcome
Phoenix, USA	Good	Good	0.00%	Baseline
Mexico	Poor	Patchy	20.00%	High Variance
India	Poor	Very Poor	20.00%	High Risk
Vietnam	Poor	Patchy	30.00%	High Risk

- **Infrastructure as a Predictor:** "Worse" road quality and "Poor" lane markings are statistically associated with higher safety risks and increased intervention rates.
- **The Reality Gap:** Universal protocols perform as intended in "Good" environments (e.g., Phoenix) but degrade significantly when introduced to "Patchy" or "Poor" infrastructure.

# Results – The “Context Penalty”



## Key Findings

➤ **Increased Risk:** Pooled Risk Ratio of **1.44** (95% CI: 0.58–3.61) suggests ~44% higher risk in LMIC environments.

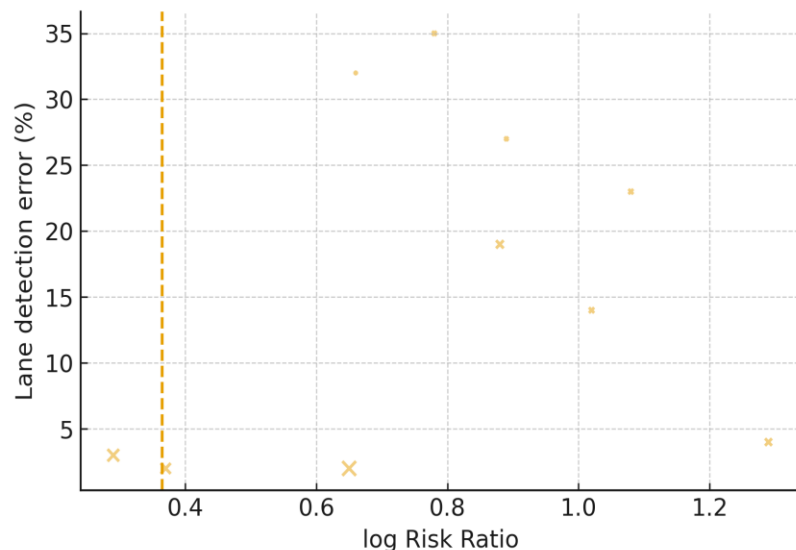
✂ **High Inconsistency:**  $I^2 = 98.5\%$  indicates massive variability in performance.

📊 **Wide Prediction Interval:** Safety is not stable. A new deployment could result in vastly different outcomes depending on local context.



# Why it Fails: Perceptual Overload

## Regression Analysis (Detection Failures)



### The Driver of Risk:

- Meta-regression of Log Risk Ratio vs. Detection Failure Rate (%).
- Slope ( $\beta$ ) = 0.0435 ( $p < 0.001$ ).



### The Trend:

- For every 1% increase in sensor detection failures, the safety risk score increases significantly.
- As sensor noise increases (due to environment), safety risk spikes linearly.



**Conclusion:** Infrastructure degradation directly causes perceptual breakdown.

# HF/E Design Implications

## For Engineers

### Design for Entropy:

- Shift from "Ideal World" training to "Entropy-First" design.
- Develop perception stacks robust to 50% occlusion.
- Systems must handle informal behaviors (e.g., pedestrian jaywalking, un-laned traffic) as standard, not "edge cases."

## For Policymakers

### Context-Aware Certification:

- Reject "copy-paste" safety ratings from high-income nations.
- Mandate **local validation testing**. (In-situ validation).
- Adopt **Context-Aware Certification**. (A 5-star rating in Phoenix  $\neq$  Safe in Nairobi.)

# Safety is Context-Dependent

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The "Universal AV" is a myth. To bridge the gap, we must design for the messy, complex human environments of the real world.

Questions?