

AI-Driven Predictive Sonification for Human-on-the-Loop Ground Vehicle Autonomy

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Approved for public release; distribution is unlimited. OPSEC10875.

The Human-on-the-Loop Bottleneck

APPLIED ARTIFICIAL INTELLIGENCE

THE SUPERVISORY CHALLENGE

WHY OVERSIGHT IS STILL REQUIRED

- Near-term ground vehicle autonomy still depends on human oversight for safe operation and timely intervention.
- A single supervisor may need to monitor multiple vehicles in teaming configurations.

WHERE VISUAL INTERFACES FALL SHORT

- Telemetry dashboards, alerts, and map products saturate visual attention.
- Subtle, safety-critical trends are recognized late, shrinking the time available to act.

1 : N

One supervisor, an entire platoon

EASILY MISSED EARLY TRENDS

- Decreasing inter-vehicle spacing
- Developing control oscillation
- Lane-tracking drift

Less time to detect is less time to prevent a violation and recover stable behavior.



Sound as a Second Channel

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WHY SOUND

WHAT SONIFICATION IS

The use of non-speech sound to represent data, so an operator can monitor by ear instead of by eye.

Familiar example: a parking sensor that beeps faster as you approach an obstacle.

WHY A SECOND MODALITY

- Visual attention is already saturated; additional visual alerts compete for the same channel.
- Auditory cues run in parallel, letting supervisors monitor time-varying behavior while keeping eyes on mission and navigation displays.

WHAT THE EVIDENCE SHOWS

- Parameter-mapping sonification and auditory displays suit monitoring without continuous visual fixation.
- Auditory alerting improves event memory and supervisory control of multi-vehicle systems.

DESIGN PRINCIPLES

What effective auditory cueing requires:

- Silent by default
- Encode risk magnitude and trajectory
- Alert before thresholds are violated
- Low-intrusion, never a startle alarm



BACKGROUND & POSITIONING

Foundations

- Situation awareness and out-of-the-loop performance frame supervised autonomy.
Endsley; Kaber and Endsley
- Multimodal interfaces improve awareness and reduce overload in automated vehicles.
Wang et al.
- Auditory displays and parameter-mapping sonification enable monitoring without visual fixation.
Nees and Walker; Hermann et al.; Grond and Berger
- Auditory decision aiding improves supervisory control of multi-vehicle systems.
Donmez et al.

THE GAP

Predictive sonification tailored to human-on-the-loop ground vehicle autonomy remains limited.

OUR CONTRIBUTION

Couple short-horizon forecasting with structured auditory cueing to alert the supervisor before control thresholds are violated, as a modular telemetry overlay that leaves the autonomy stack unchanged.



A Predictive, Risk-Based Auditory Overlay

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OUR CONTRIBUTION

1

Forecast-Driven

Short-horizon forecasting of platoon behavior, rather than reaction to the current state alone.

2

Anticipatory

Alerts the supervisor to emerging instability before control thresholds are violated.

3

Magnitude & Trajectory

Conveys how large the risk is and how it is evolving, not a single binary alarm.

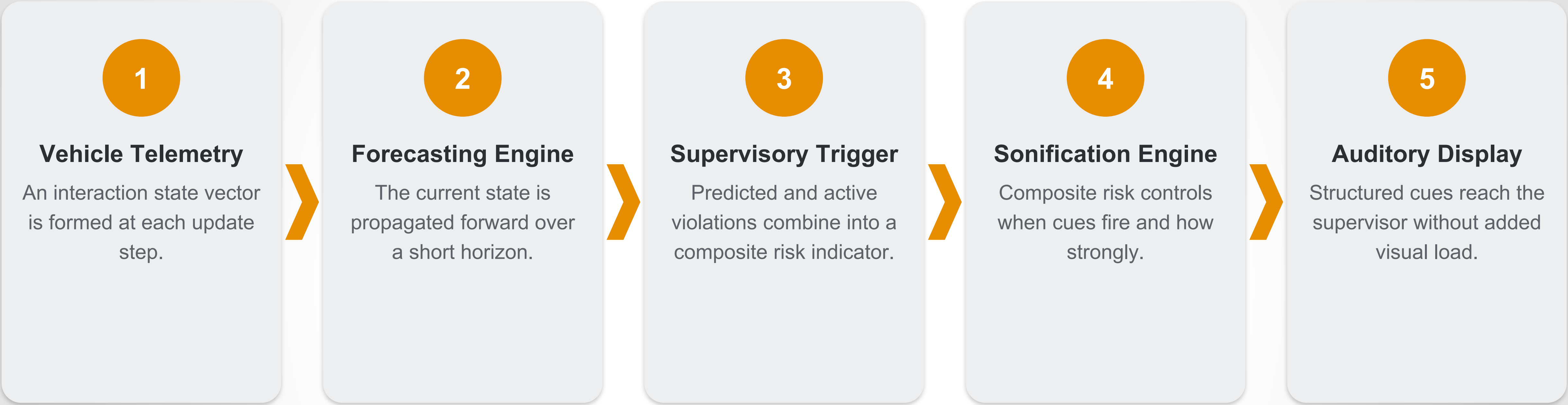
4

Modular Overlay

Consumes existing telemetry and requires no modification to the autonomy stack.



SYSTEM ARCHITECTURE



Silent by default. During nominal conditions, the system stays quiet to preserve auditory bandwidth; structured cues activate only when risk thresholds are exceeded.



Forecasting Platoon Behavior

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PREDICTIVE MODEL & FORECASTING ENGINE

STATE VECTOR (per update step)

$$x(t) = [d(t), v(t), a(t), j(t), y(t)]^T$$

$d(t)$	Inter-vehicle distance
$v(t)$	Relative velocity
$a(t)$	Longitudinal acceleration
$j(t)$	Jerk
$y(t)$	Lateral deviation

FORECASTING ENGINE

Predict, then evaluate

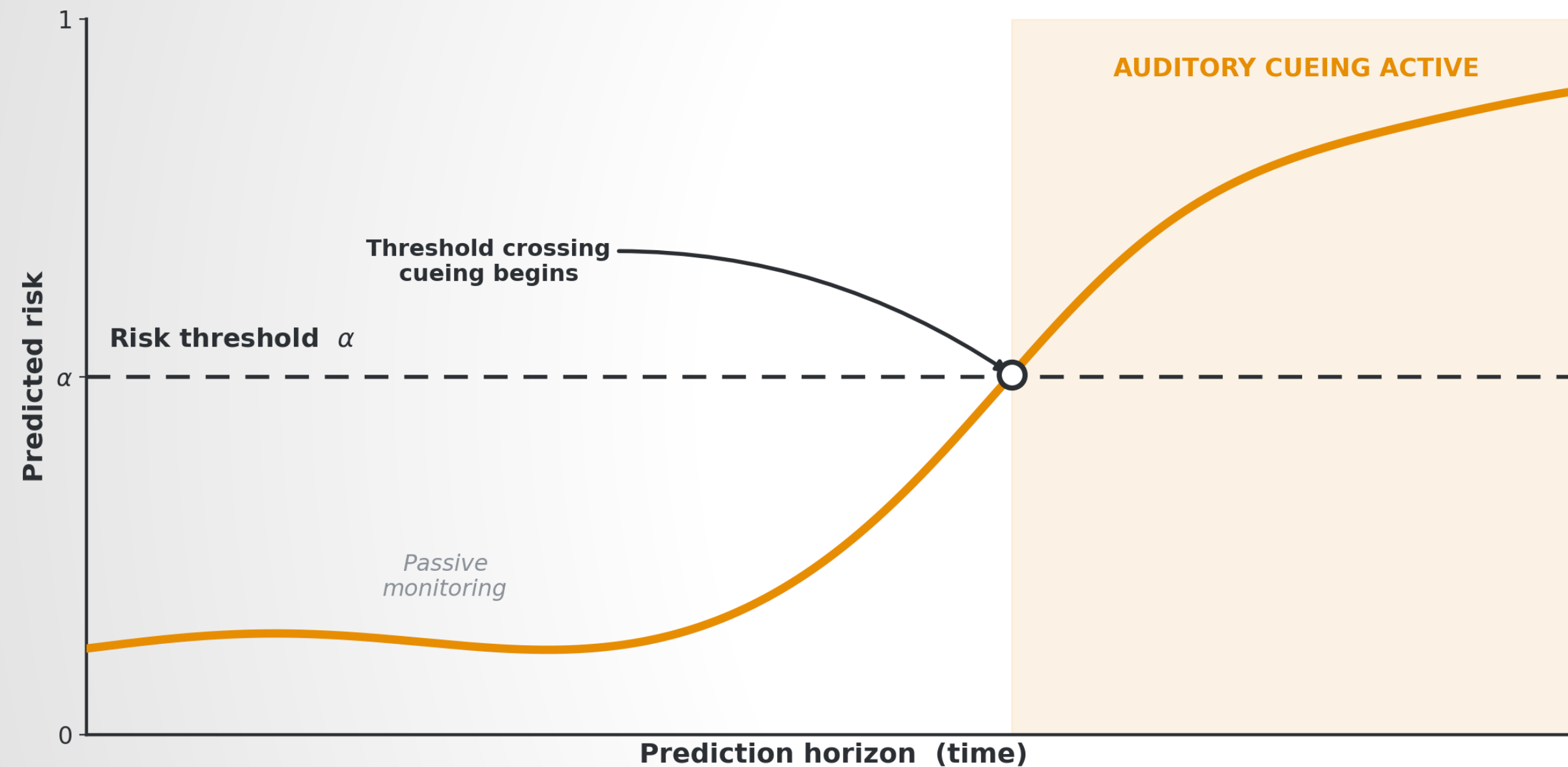
$$\hat{x}(t + \tau) = f_{\theta}(x(t))$$

A learned function $f_{\theta}(\cdot)$ propagates the current state over a short horizon τ to a predicted state estimate, supporting detection of unsafe behavior before control thresholds are violated.



Turning Forecasts into a Trigger

RISK ESTIMATION & SUPERVISORY TRIGGER



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RISK ESTIMATE

$$r(t + \tau) = g(\hat{x}(t + \tau))$$

$g(\cdot)$ maps the predicted state $\hat{x}(t + \tau)$ to a normalized risk value $r(t + \tau)$.

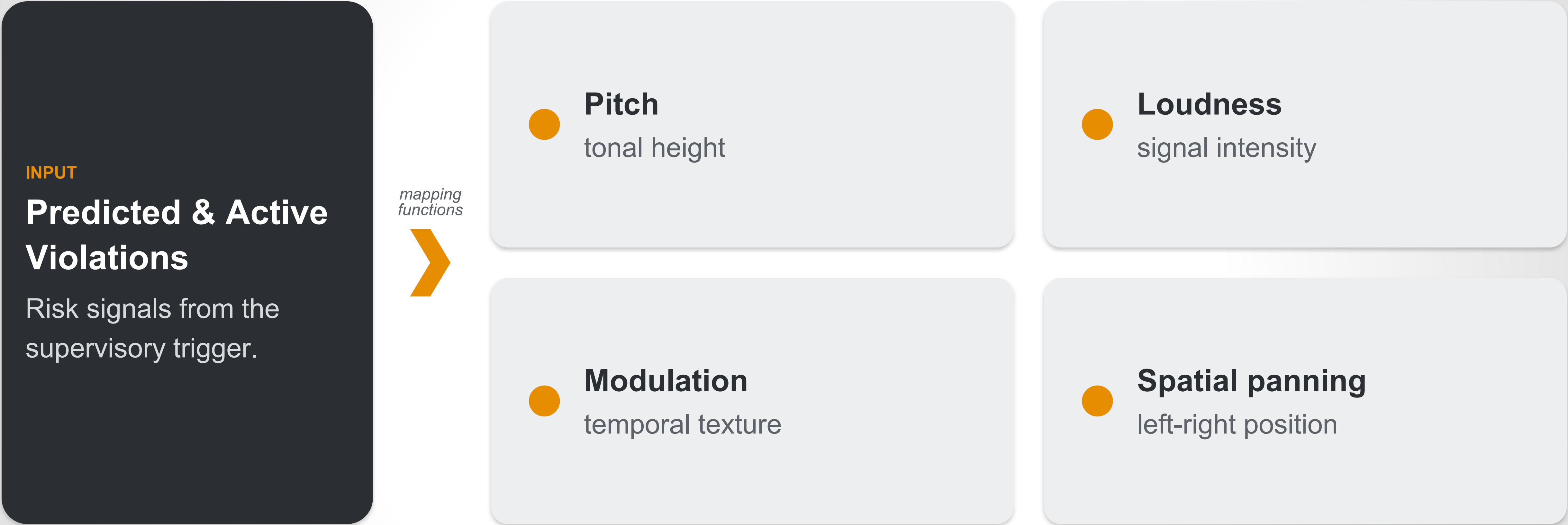
ALERT CONDITION

$$\text{Alert} = \begin{cases} 1, & r(t + \tau) \geq \alpha \\ 0, & r(t + \tau) < \alpha \end{cases}$$

Cueing is enabled when predicted risk exceeds a tunable threshold α before an active violation occurs.



SONIFICATION DESIGN FRAMEWORK



EXAMPLE PARAMETER MAPPING

Inter-vehicle spacing error maps to pitch; cue intensity rises as predicted risk grows.

$$f_{\text{pitch}}(t) = f_0 + k_p e_d(t)$$



Three Supervisory States

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OPERATIONAL USE CASES

1

Normal Operation

SYSTEM SILENT

Forecast predicts stable behavior with no predicted or active violations. The system stays silent and the supervisor uses standard mission displays.

2

Anomaly

CUES BUILD

The forecast predicts a follower will breach a threshold within the horizon. Low-intrusion cues begin and grow with risk, giving early, intuitive awareness.

3

Intervention

HIGHER URGENCY

An active violation is detected in real time. Output shifts to a higher-urgency profile; the supervisor adjusts platoon behavior, modifies mission parameters, or assumes manual control.



PLANNED EXPERIMENTAL EVALUATION

Study Design

- Participants act as platoon supervisors in a simulated four-vehicle convoy.
- Within-subjects, counterbalanced conditions: visual-only and visual plus predictive sonification.
- Scenarios span nominal operation, a developing anomaly, and a control violation requiring response.
- The simulator presents realistic vehicle dynamics, communications delays, and mission tasking.

Metrics

OBJECTIVE MEASURES

- Time to detect predicted violations
- Time to initiate corrective action
- Frequency of missed or late interventions
- Stability recovery time after intervention

SUBJECTIVE & COGNITIVE MEASURES

- NASA-TLX workload
- SART situation awareness
- Usability and trust of the interface



Fielding the Overlay

INTEGRATION PATH FOR GROUND VEHICLE AUTONOMY

HOW IT INTEGRATES

- A modular, non-intrusive overlay that consumes existing telemetry and communication streams, leaving core control unchanged.
- Standardized V2V and V2I interfaces, including DSRC and C-V2X message sets.
- Leverages common data products: vehicle state estimates, health-monitoring signals, and autonomy-supervisor outputs.

WHY IT DROPS IN CLEANLY

- External to the primary autonomy stack, so it suits both legacy and next-generation Army systems.
- Insertable into experimental platforms, technology demonstrators, and test assets without disrupting safety architectures.
- Supports incremental deployment, exercising partial functionality during early integration.

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INTEGRATION PATHWAYS



Leader-follower convoys

Platoon spacing and stability monitoring



Autonomous logistics vehicles

Sustained route supervision



Teleoperated systems

Low-visual-burden alerting channel



Optionally manned combat vehicles

Supervisory oversight when unmanned



CONCLUSION & FUTURE WORK

Key result. Coupling short-horizon forecasting with structured auditory cueing enables earlier detection of emerging instability while reducing reliance on visually intensive supervisory interfaces.

Conclusion

- Integrates predictive modeling, threshold-based supervisory logic, and intuitive sonification into a single modular architecture.
- Supports nominal supervision, early anomaly awareness, and time-critical intervention without unnecessary cognitive burden.

Future Work

- Complete the planned human-subject experimental evaluation.
- Integrate the architecture with operational autonomy platforms.
- Explore adaptive mapping, task-dependent cueing, and multi-platform formations.



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