

DDAW System monitoring strategy and warning delay performance

Clare Anderson, PhD.

Professor of Sleep and Circadian Science

Centre for Human Brain Health, School of Psychological Sciences
University of Birmingham, UK

Adjunct Professor (Research)

Monash University Accident Research Centre (MUARC), Australia
School of Psychology, Monash University, Australia



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DDAW system monitoring strategy and warning delay performance

PURPOSE: Support the development of future DDAW systems that accurately provide a warning to the driver at an excessive or unsafe level of drowsiness



1. Should an activated DDAW system continuously monitor the driver after providing the first warning, and subsequently continue to warn the driver, even though drowsiness is not typically a transient state (unlike distraction) [sic]?
2. Would the implementation of a continuous monitoring strategy require considerations of factors such as likelihood of false positive warnings, factors influencing likely self-assessment by the driver including elapsed time driving when the warning is first given, the possibility that the driver may have been changed during a single trip?
3. How long is too long for the delay between a driver becoming drowsy ($KSS = 8$) and the system detecting this drowsiness? For example, should the drowsiness regulation consider any data on the time-based safety risk increase for having an accident under continuous driving at a nominal drowsiness level?

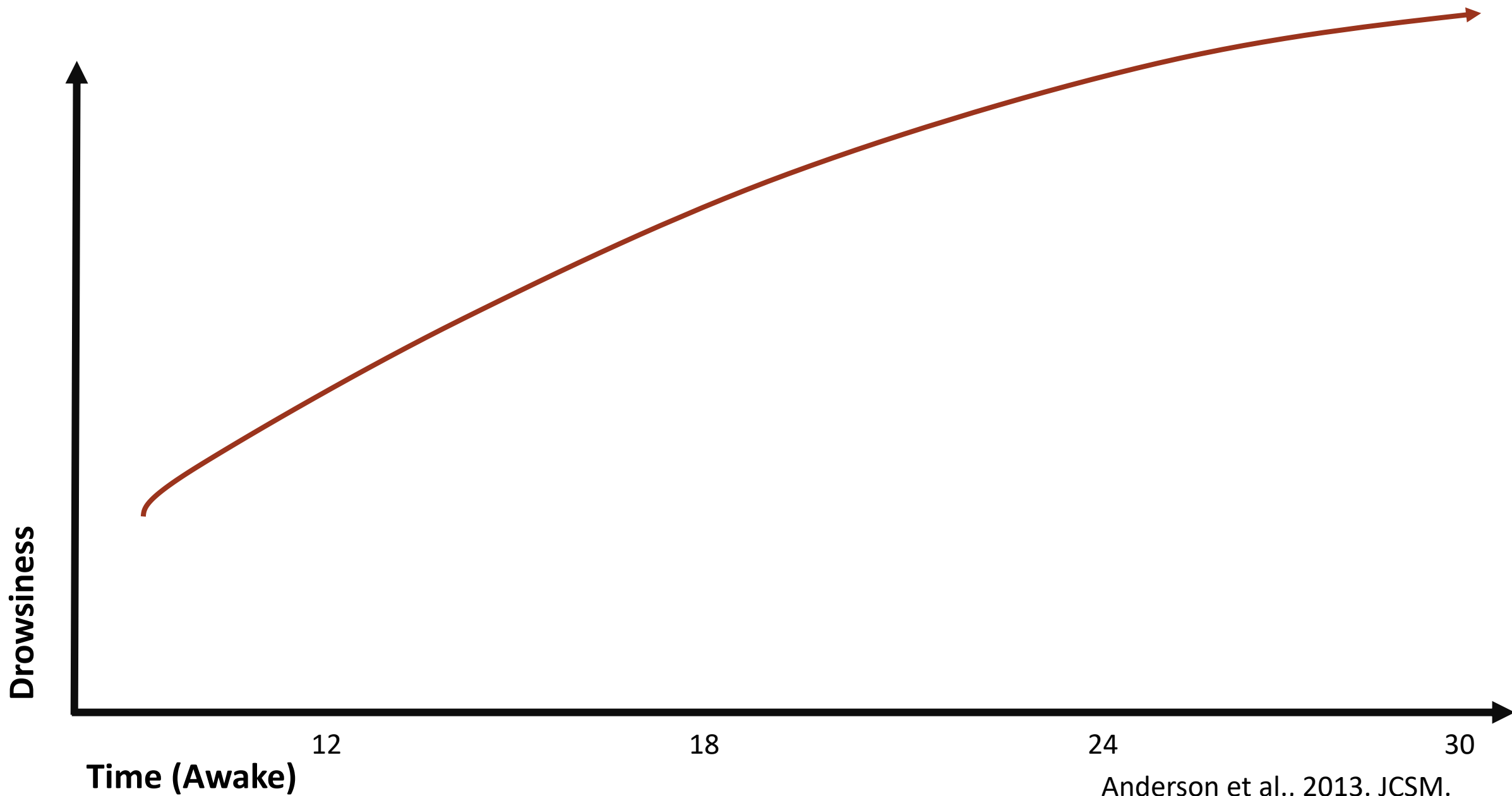




Topics for Discussion

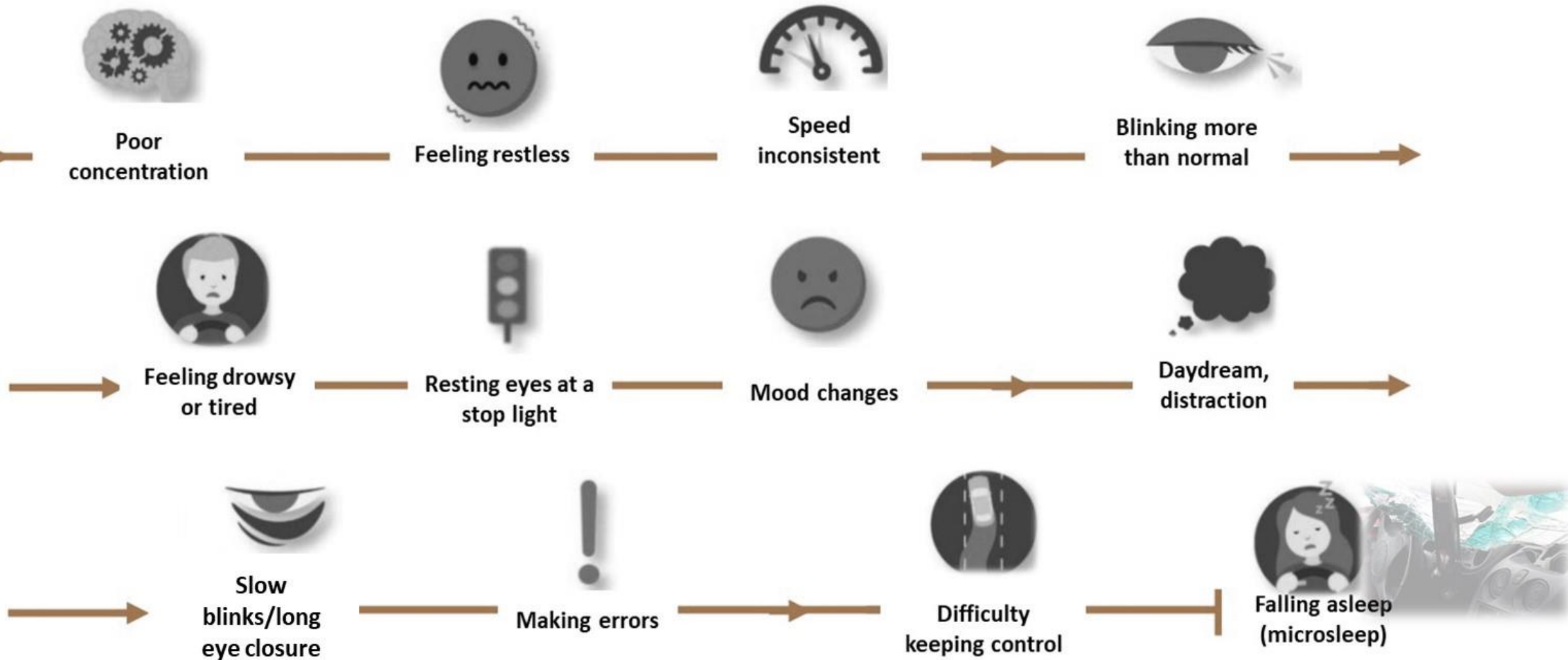
1. Characterise interplay between subjective sleepiness, physiological drowsiness and driving impairment
2. Warnings: To continue to stop after the first event
 - Issue of False (positive) Alarms
 - Impact of Drive Duration
 - Dealing with driver changes
3. Delay between alarms and drowsiness: how long is too long?

Drowsiness over 36 hours

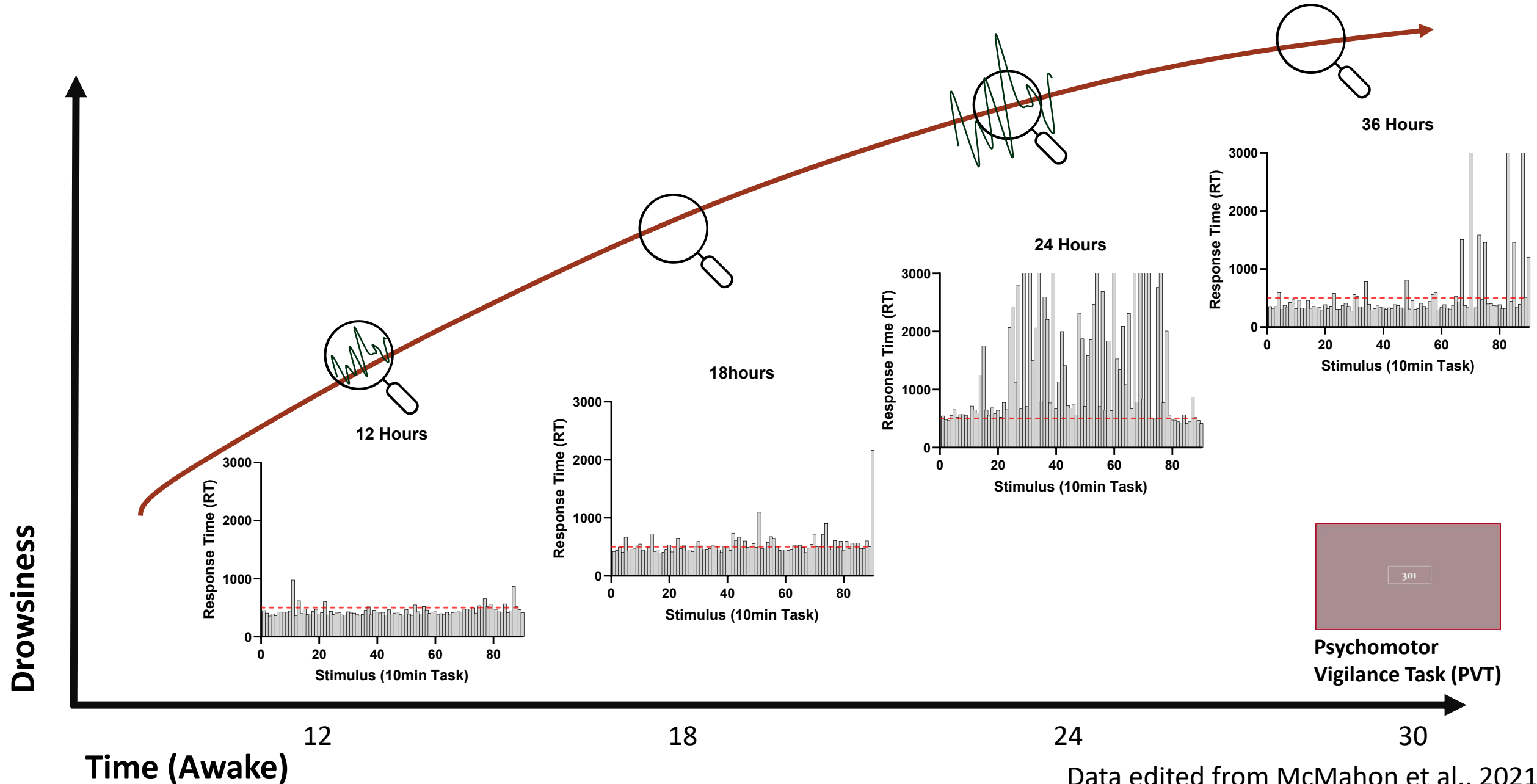


Drowsiness over hours

Insufficient
Sleep (and
other factors)

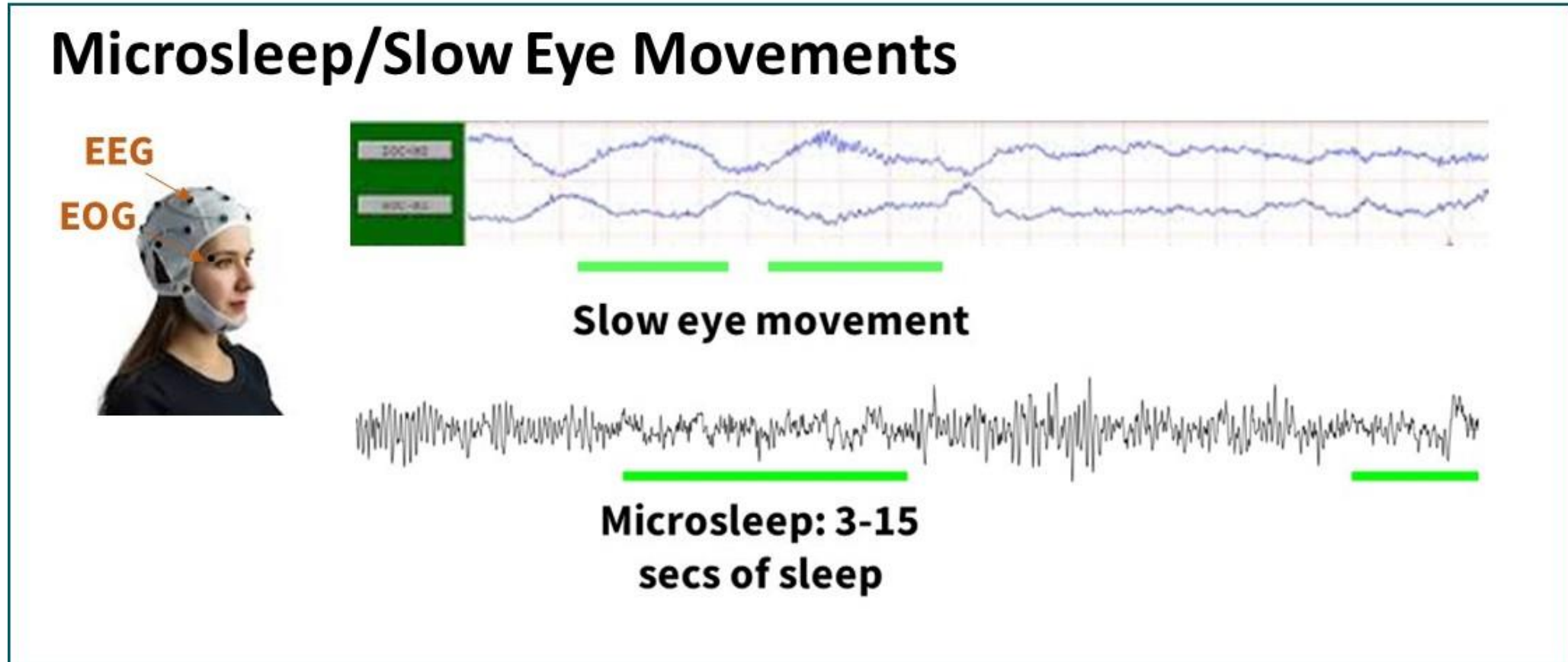


Drowsiness over minutes and seconds



Drowsiness and the process of falling asleep

Drowsiness is the physiological process of falling asleep.....



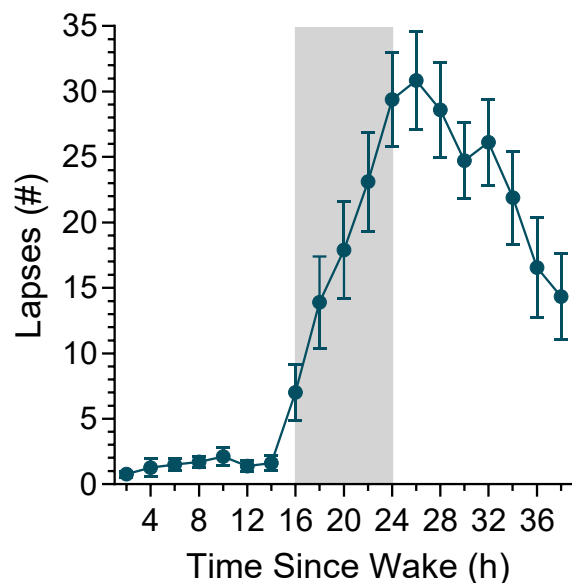
Drowsiness is highly dynamic

..... and microsleep are discrete safety critical events

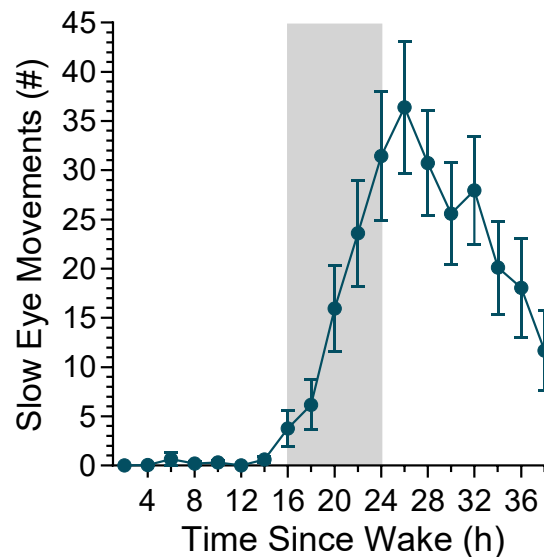


Drowsiness and the 'ground truth'

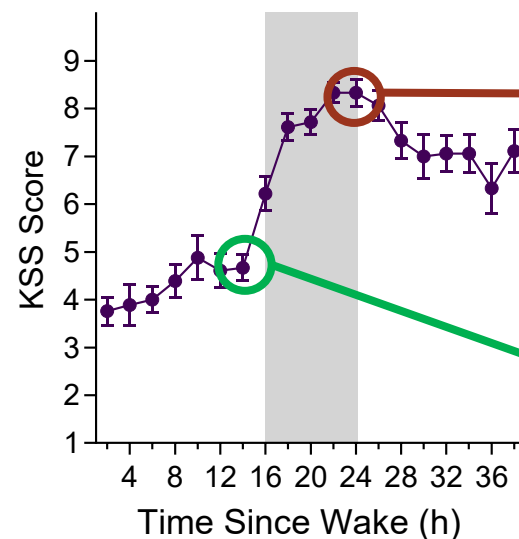
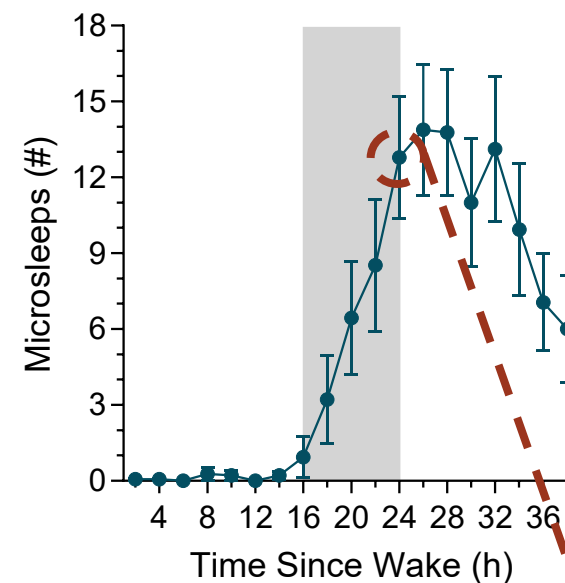
What is the behavioural state?



What is the ocular state?

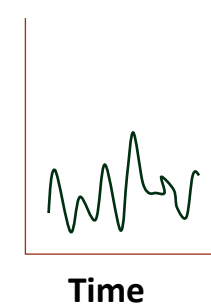


What is the brain state?



KSS 5

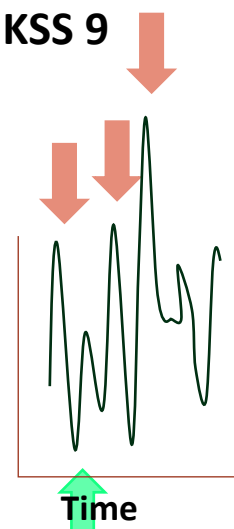
Drowsiness



KSS 9

Drowsiness

Time



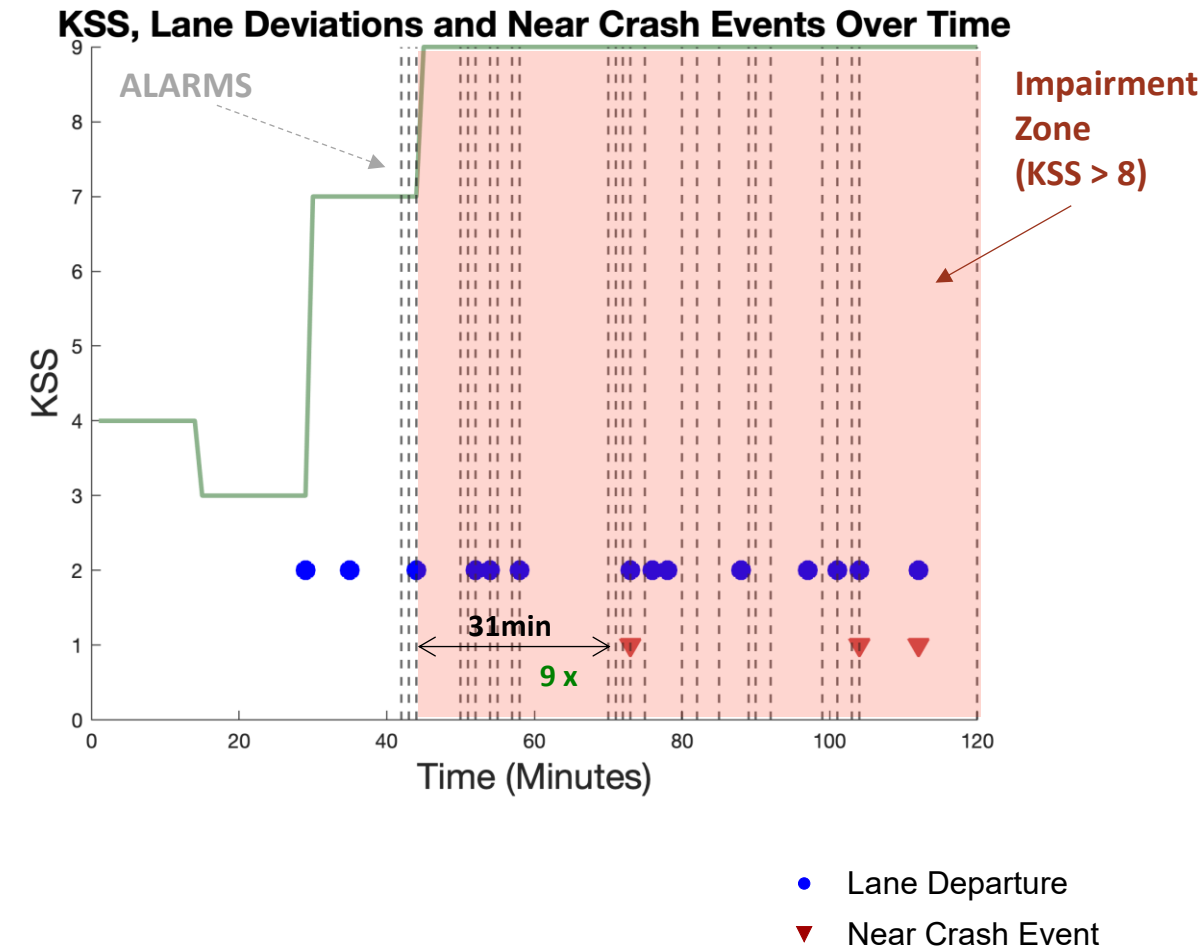
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Switching off the system after the first alarm: a case study



- Dual control car, with instructor on a closed loop track
- KSS ratings every 15 minutes
 - End 2h or after an intervention procedures

- Subjective Sleepiness (KSS)
- Driving Impairment (Lane Departures)
- Critical Driving Impairment (Emergency Braking)
- ALARMS (Seeing Machines)



Single alarm relies on immediate action



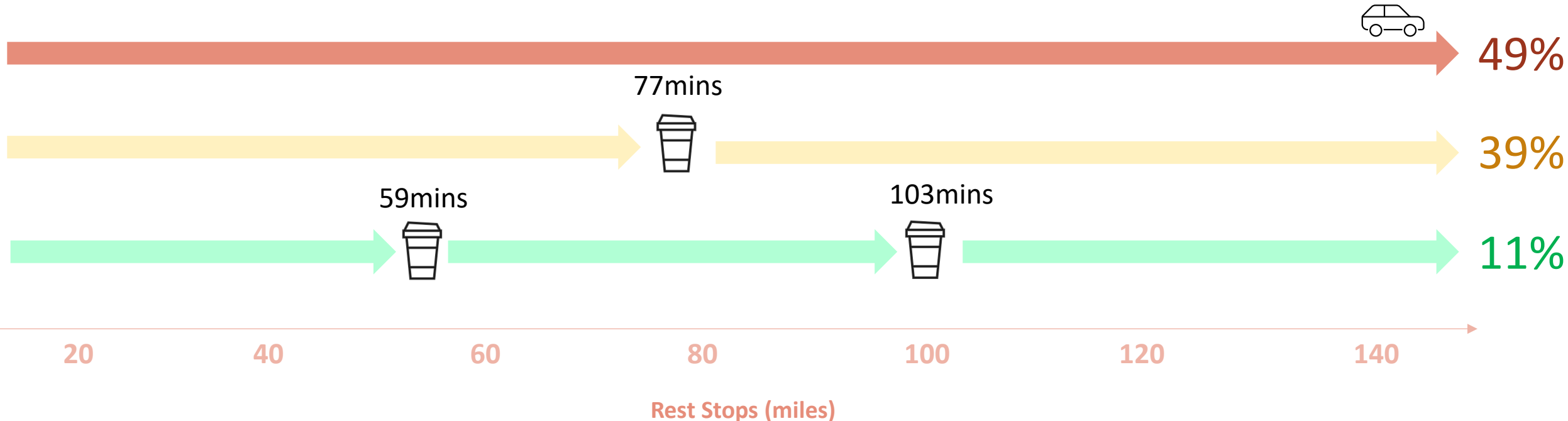
Drivers already have an in-built warning system:
how does it affect their decision to stop?

86% of drivers continued to drive with a KSS of 8

76% of drivers continued to drive with a KSS of 9

TASK: Drive 150 miles.
Rest stops every
20miles

INCENTIVE: \$1 for every
minute under 3hours
Forfeit if crash



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Drivers don't stop for many reasons...



Denial

- They don't believe the alarm



Over-confidence

- They believe they can 'push through'



Safer option

- They are concerned for their safety



Impaired judgment

- Fail to recognise drowsiness

Can we reduce through effective use of alarms... ?



Validity is essential. Long-term 'learning' leads to enhanced trust

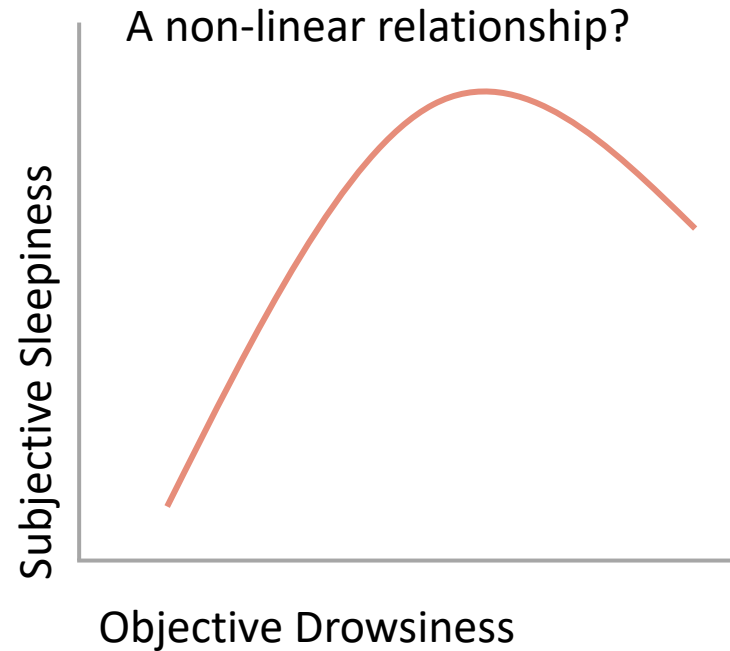


Escalating alarms indicate that 'pushing through' is not working



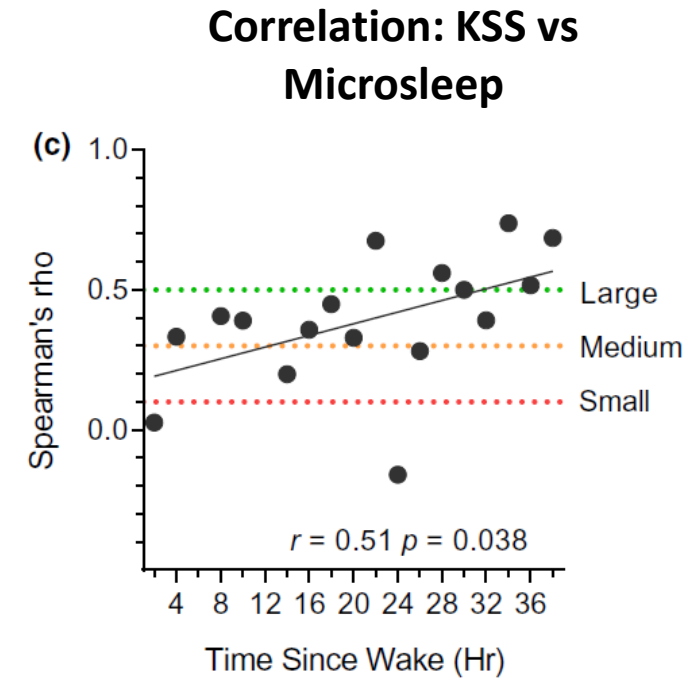
Earlier alarms enable better planning to find a safe space to take action and avoid late-stage drowsiness failures.

Continuous alarms allow for changing awareness with increasing drowsiness



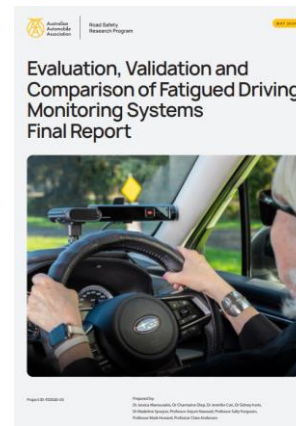
KSS

	Low (1-4)	Moderate (5-7)	High (8-9)
PERCLOS Low (<0.125)	25%	17%	7%
Moderate ($0.125-0.25$)	63%	65%	63%
High (>0.25)	13%	18%	30%



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Active Monitoring Systems: Effectiveness of alarms



The D-TECH Program: 2020-2024

Consortium Lead/Principal Investigator: Clare Anderson

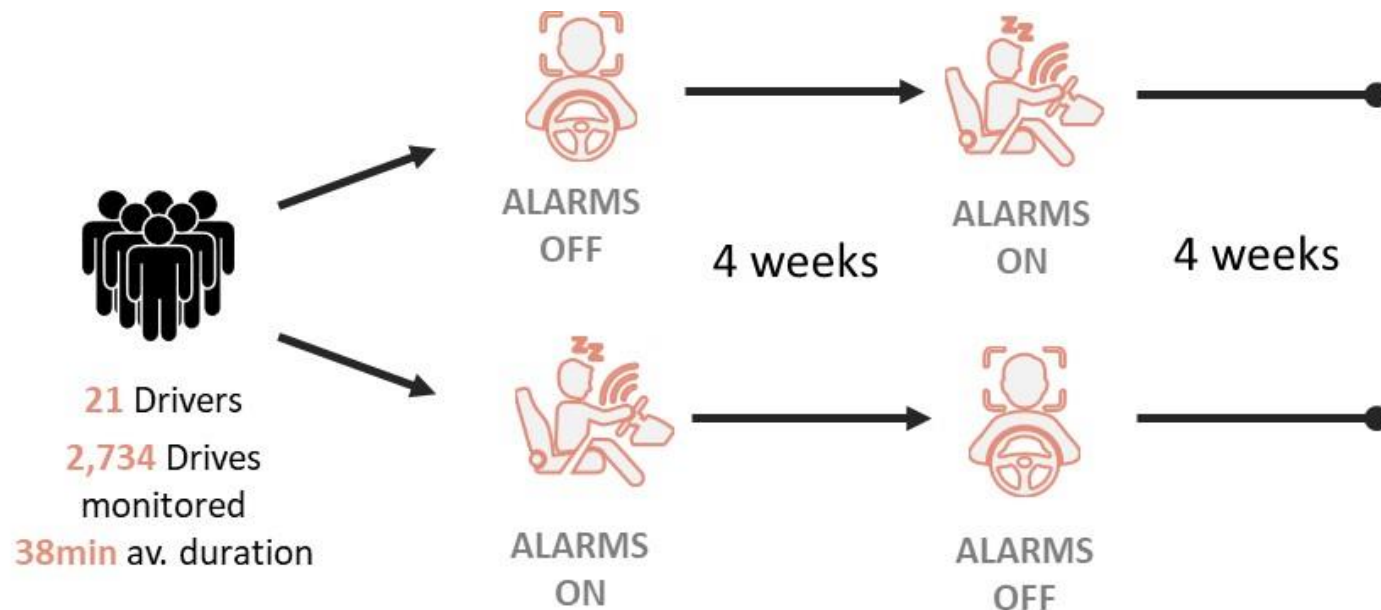
Consortium Leadership Team: Jessica Manousakis, Mark Howard, Sally Fergusson



STUDY 2A:

Naturalistic Study, with randomised cross-over design

Project Lead: Jen Cori



21 Drivers
2,734 Drives
monitored
38min av. duration

55% reduction validated fatigue events

48% reduction hard braking

34% reduction hard cornering

Reduced end of shift KSS

Increased caffeine consumption

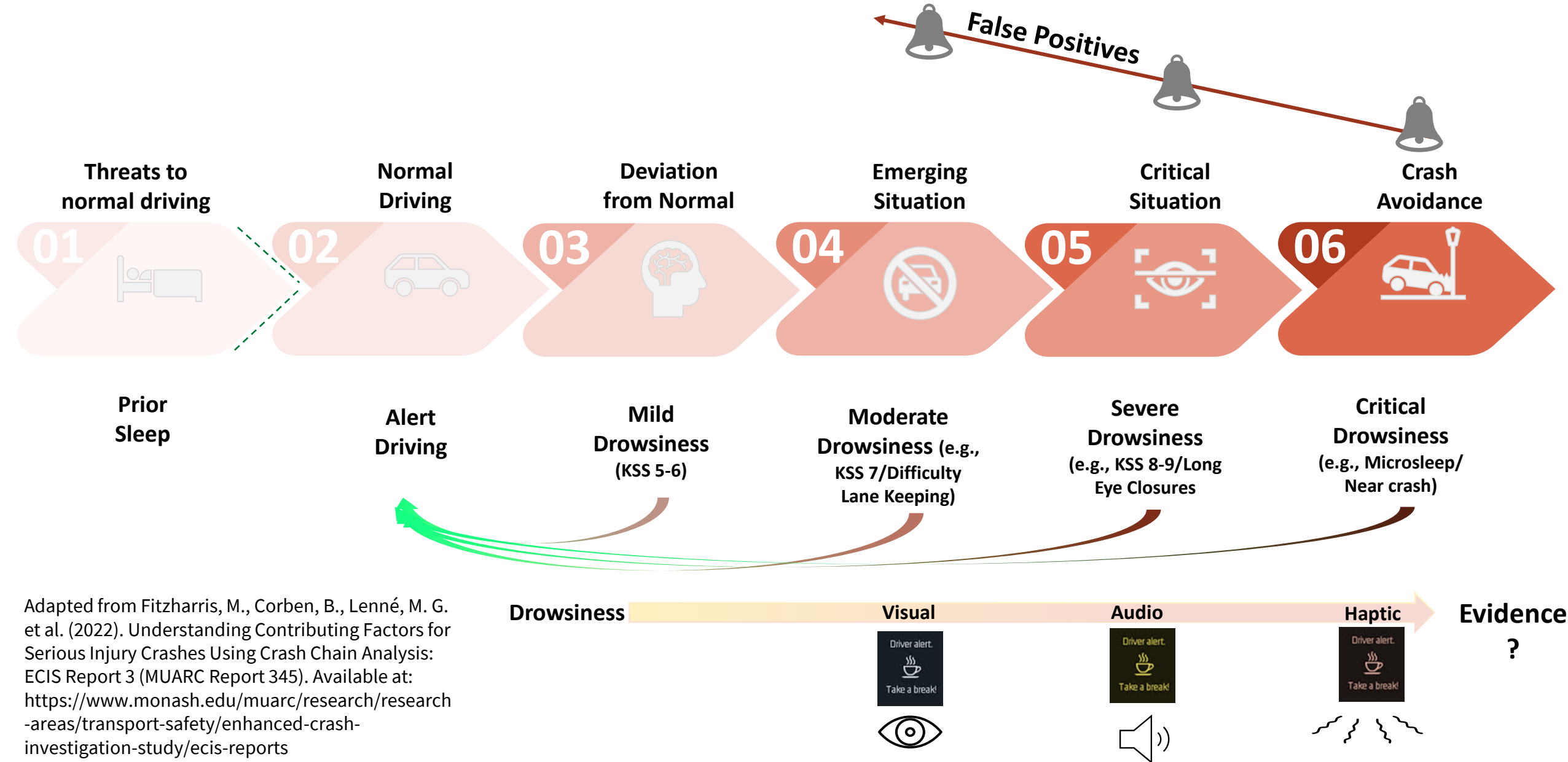
52% improved recognition of fatigue

90% better equipped to handle fatigue

81% felt the alarms were not annoying

100% felt the alarms did not interfere with work

How and when alarms are activated matters



False Positive Warnings

		Predicted Drowsiness	
		NO	YES
Actual Drowsiness	NO	TRUE NEGATIVE	FALSE POSITIVE
	YES	FALSE NEGATIVE	TRUE POSITIVE

A contingency table* organizes the frequencies of two categorical variables, showing how often they occur together.

How FPs impact safety

- Driver may disengage/turn off system
- Driver may lose trust and ignore warnings
- Driver may be distracted (Anderson & Horne, 2013)

How to reduce FPs

- Have single alarm/fewer alarms
- Have a 'late warning' system

How to reduce FPs safely

- Ensure the threshold for detection is reliably high
- Have escalating alarms

* Also referred to as a confusion matrix or cross-tabulation

Is sensitivity enough?

		Predicted Drowsiness		
		NO	YES	
Actual Drowsiness	NO	TRUE NEGATIVE	FALSE POSITIVE	Specificity
	YES	FALSE NEGATIVE	TRUE POSITIVE	Sensitivity
		PPV		

Sensitivity

Proportion of true positives correctly identified

$$\text{Sensitivity} = \text{True Positives} \div (\text{True Positives} + \text{False Negatives})$$

Specificity

Proportion of true negatives correctly identified

$$\text{Specificity} = \text{True negatives} \div (\text{True Negatives} + \text{False Positives})$$

Positive Predictive Validity (PPV)

Proportion of predicted drowsiness events (e.g., Alarms) that are true.

$$\text{PPV} = \text{True Positives} \div (\text{True Positives} + \text{False Positives})$$

DDAW System monitoring strategy and warning delay performance

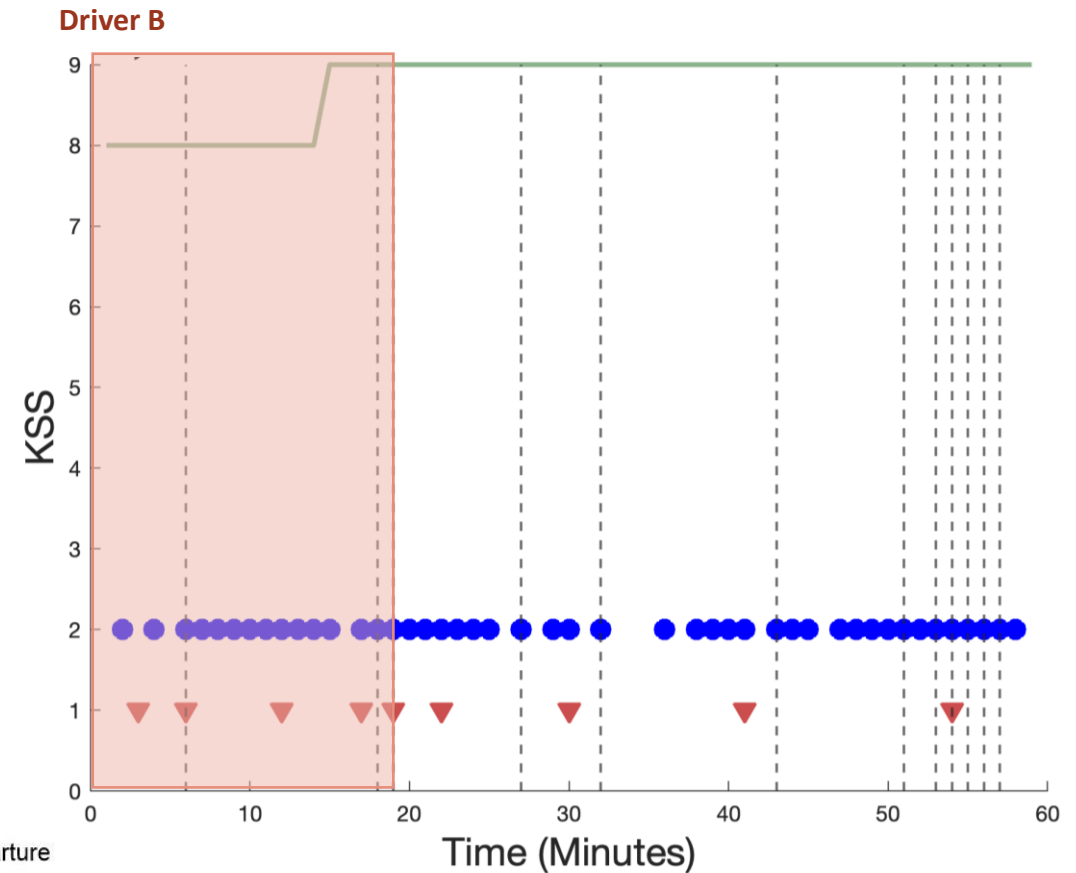
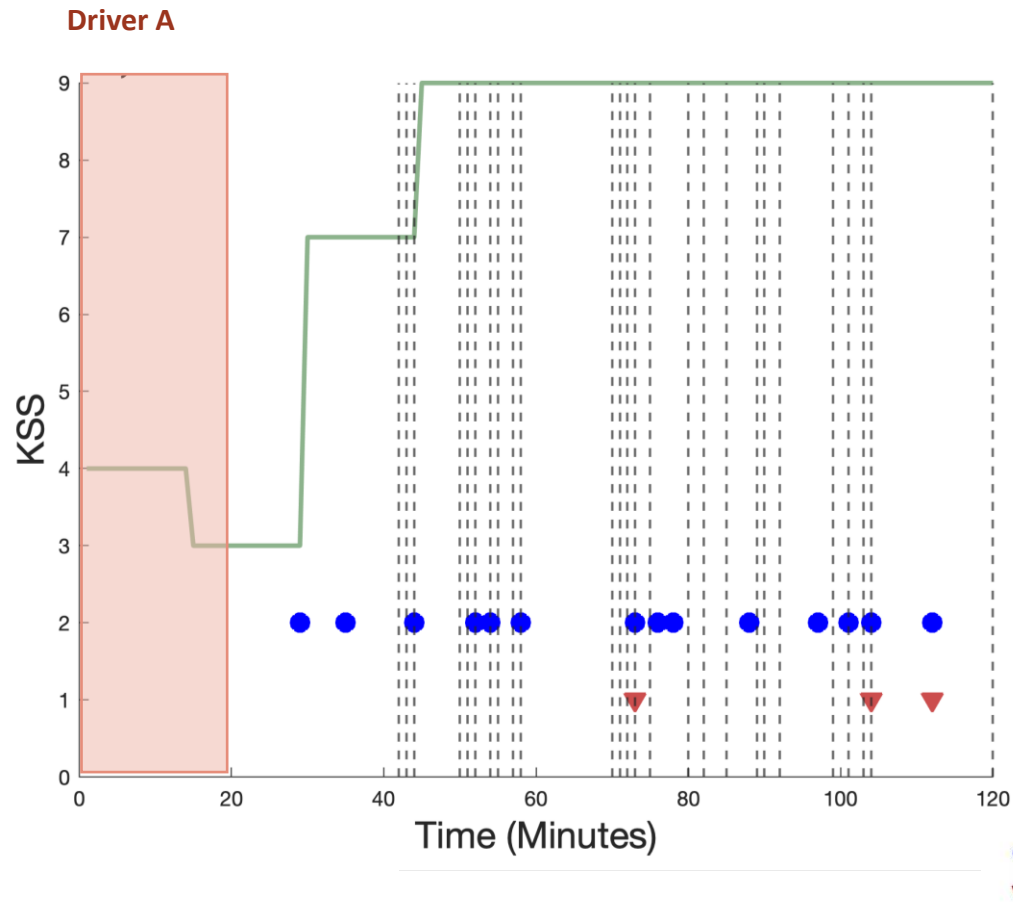
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Continued alarms increase potential for drivers to intervene short-term and long-term improving safety
2. Would the implementation of a continuous monitoring strategy require considerations of factors such as likelihood of **false positive warnings**, factors influencing self-assessment by the driver including **elapsed time driving** when the warning is first given, the possibility that the **driver may have been changed** during a single trip?
3. How long is too long for the delay between a driver becoming drowsy ($KSS = 8$) and the system detecting this drowsiness? For example, should the drowsiness regulation consider any data on the time-based safety risk increase for having an accident under continuous driving at a nominal drowsiness level?



Time factor should be carefully considered: case study

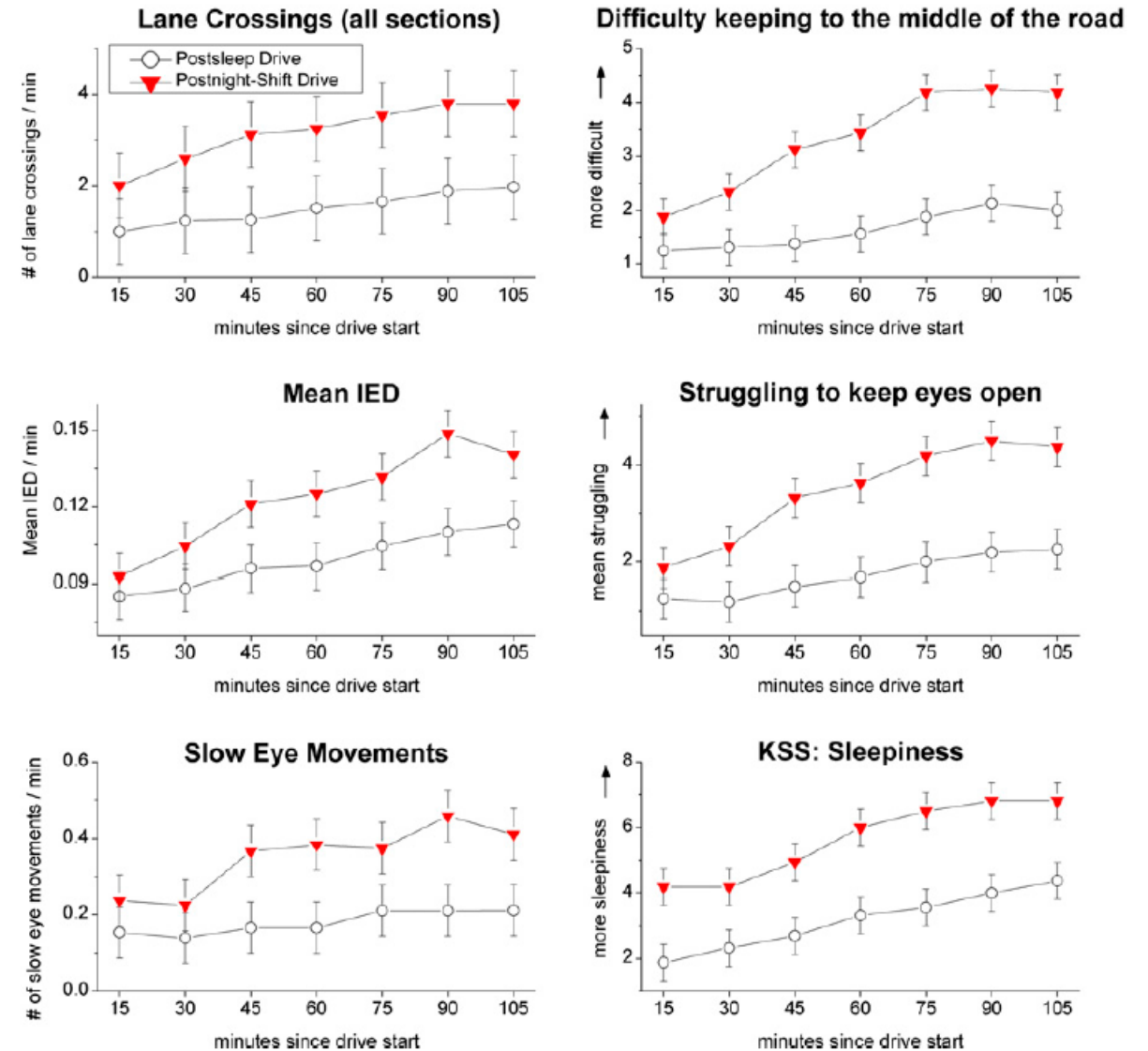
What's the potential impact on ignoring the first 20 minutes (as an example) to reduce FPs:



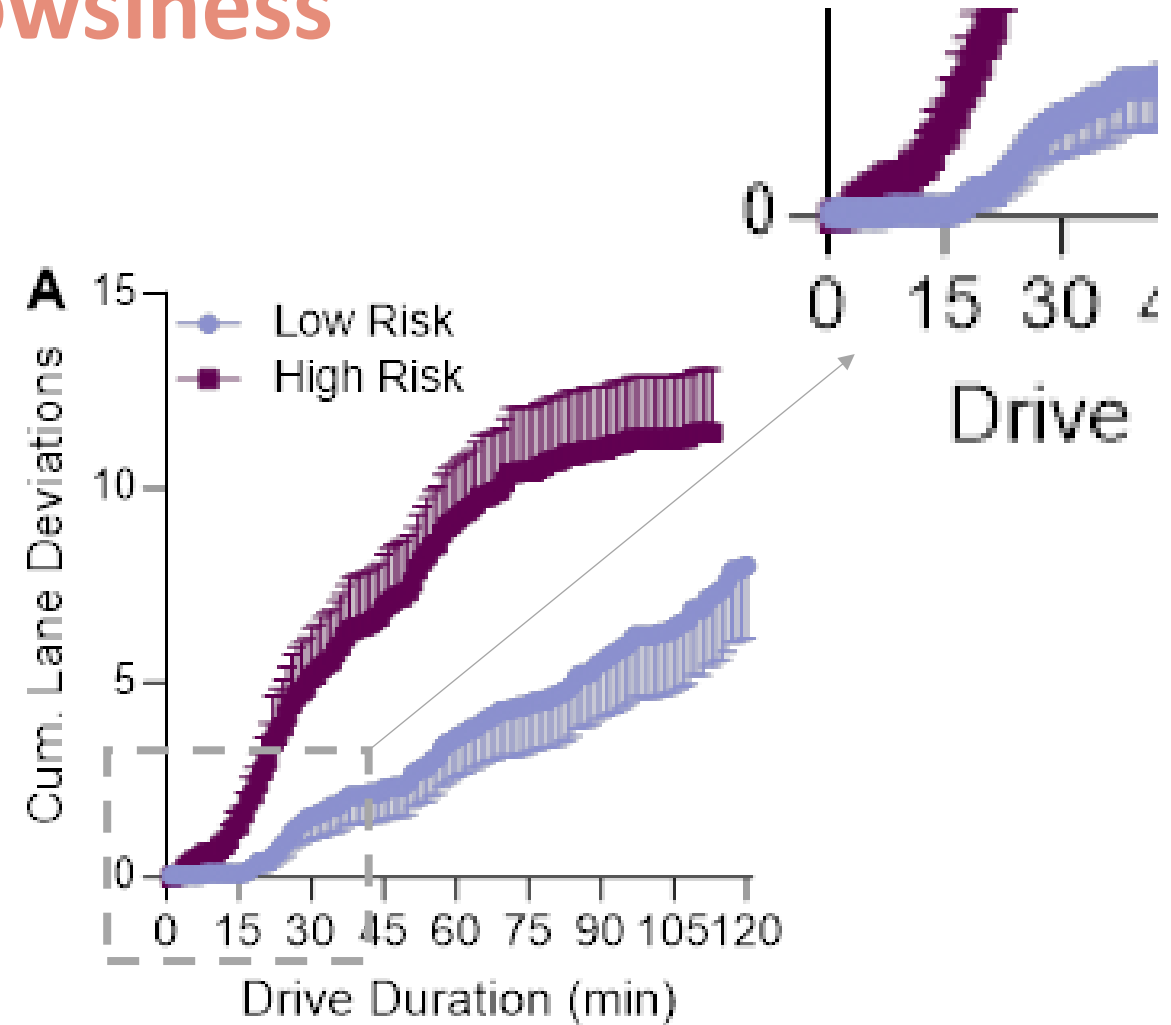
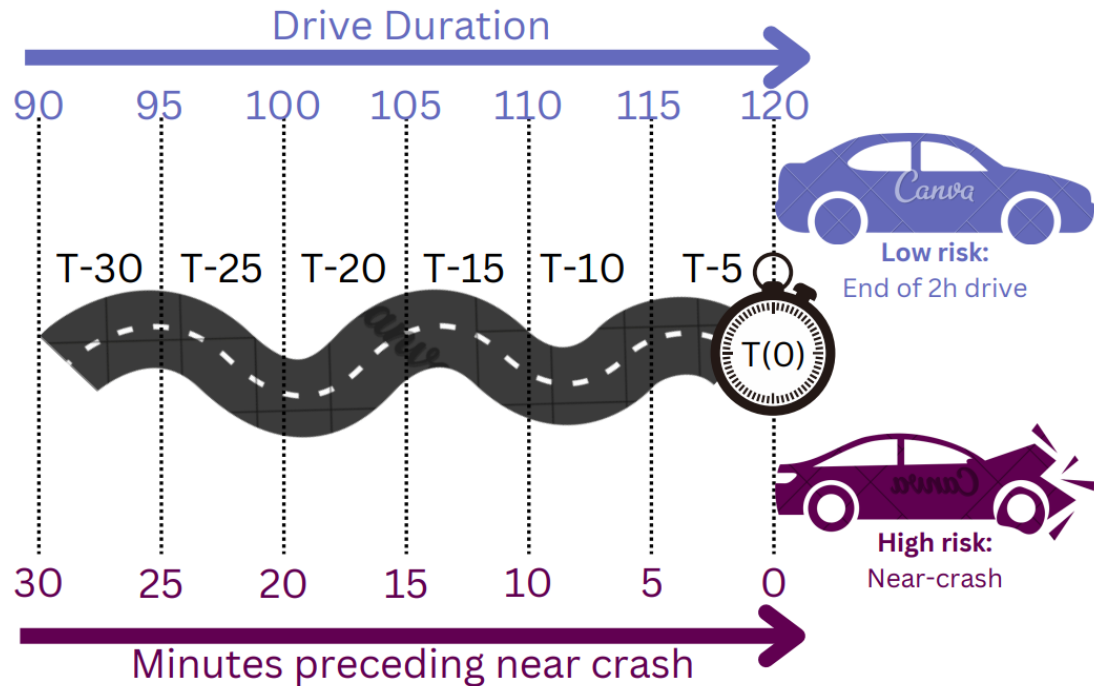
Switching off the system after the first alarm

Naturalistic/track study of drowsy driving in shift workers

After a night on-shift (limited sleep), evidence of impairment was observed within the first 15 minutes, and driving impairment, ocular metrics, and EEG microsleep all increased with drive duration.



Impairment can be observed very early, especially for those most affected by drowsiness



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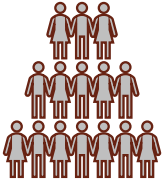
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A reliable, valid system accurately assessed against an evidence-based ground truth should counter these factors.

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Time Delay KSS versus Driving Outcomes: A KSS of 8 is late in the drowsiness trajectory



N=21 (12 m)

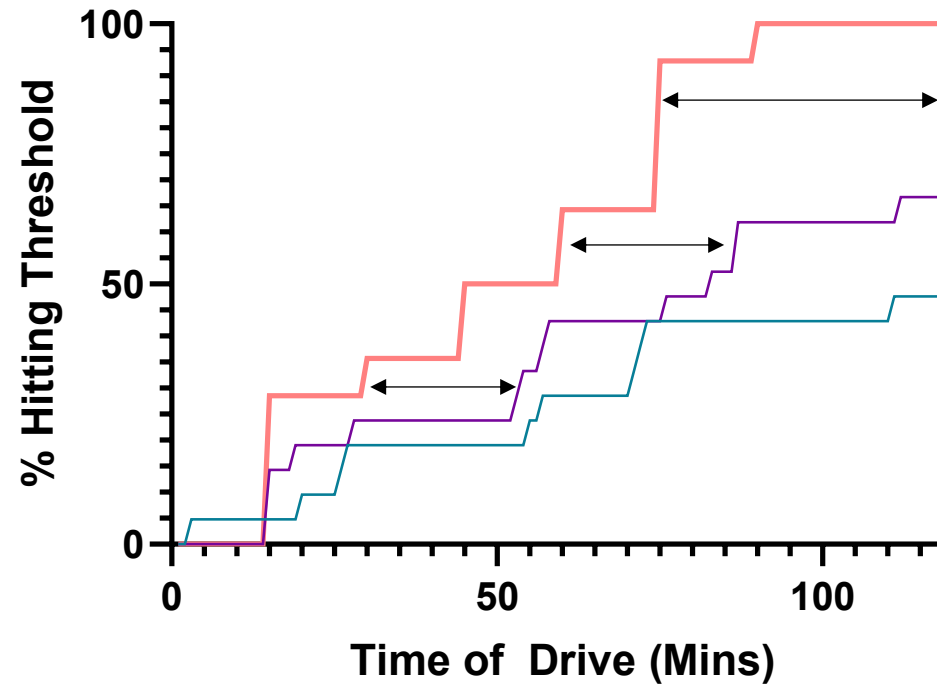
2520 mins driving
(without sleep)



Closed loop track trial

- Subjective Sleepiness (KSS)
- Driving Impairment (Lane Departures)
- Critical Driving Impairment (Emergency Braking)
- ALARMS (Seeing Machines)

— NCE (Near Crash Event)
— LD4 (4LD per 15min)
— KSS8 (KSS>8)



On average, KSS 8 is reached:

23mins before a near-crash

6min before a lane departure
drowsiness index

PPV = > 80% for LD and NC

Data adapted from Manousakis et al. Evaluation, Validation and Comparison of Fatigued Driving Monitoring Systems Final Report. (<https://www.aaa.asn.au/library/evaluation-validation-and-comparison-of-fatigued-driving-monitoring-technologies/>, 2024).

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A KSS of 8 is late in the trajectory of drowsiness.

References

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Acknowledgments

COLLABORATORS AND RESEARCH SUPPORT

Project Leaders/Post Doctoral Fellows

Jessica Manousakis, Ph.D. *

Sophie Mason, Ph.D. *

Katy Jeppe, Ph.D.

Suzanne Ftouni, Ph.D.

Anna Cai, Ph.D.

Charmaine Diep, Ph.D.

Madelaine Sprajcer, Ph.D.

Michael Lee, Ph.D.

Jennifer Cori, Ph.D.

Collaborators

Mark Howard, M.D., Ph.D.

Jim Horne, Ph.D.

Sally Fergusson, Ph.D.

Anjam Naweed, Ph.D.

Charles Czeisler, Ph.D.

Bill Horrey, Ph.D.

Mike Lenne, Ph.D.

Shantha Rajaratnam, Ph.D.

All students and research support staff in the Division of Sleep Medicine (Brigham and Women's Hospital), the Monash Sleep and Circadian Laboratory, and the METEC Driver Training Facility.

All the participants who make the work possible.



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* Contributed to data analysis contained here

FUNDING SUPPORT



Department
of Transport
and Planning



Australian Government
Australian Research Council



Business
Cooperative Research
Centres Program
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Australian
Automobile
Association



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University of Birmingham, UK

Adjunct Professor (Research)

Monash University Accident Research Centre (MUARC), Australia



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