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# **Modeling human-AV interactions for safety and acceptance of automated vehicles**

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An aerial, slightly hazy photograph of a busy urban intersection. The scene is filled with various vehicles including cars, buses, and taxis, along with many pedestrians crossing the streets. The layout shows multiple lanes and crosswalks, typical of a major city center.

**How to make AVs that can successfully coexist with humans?**  
→ By developing high-fidelity models of human road user behaviour

**What kinds of models?**  
→ Combination of data-driven and neurocognitive models





AITECH

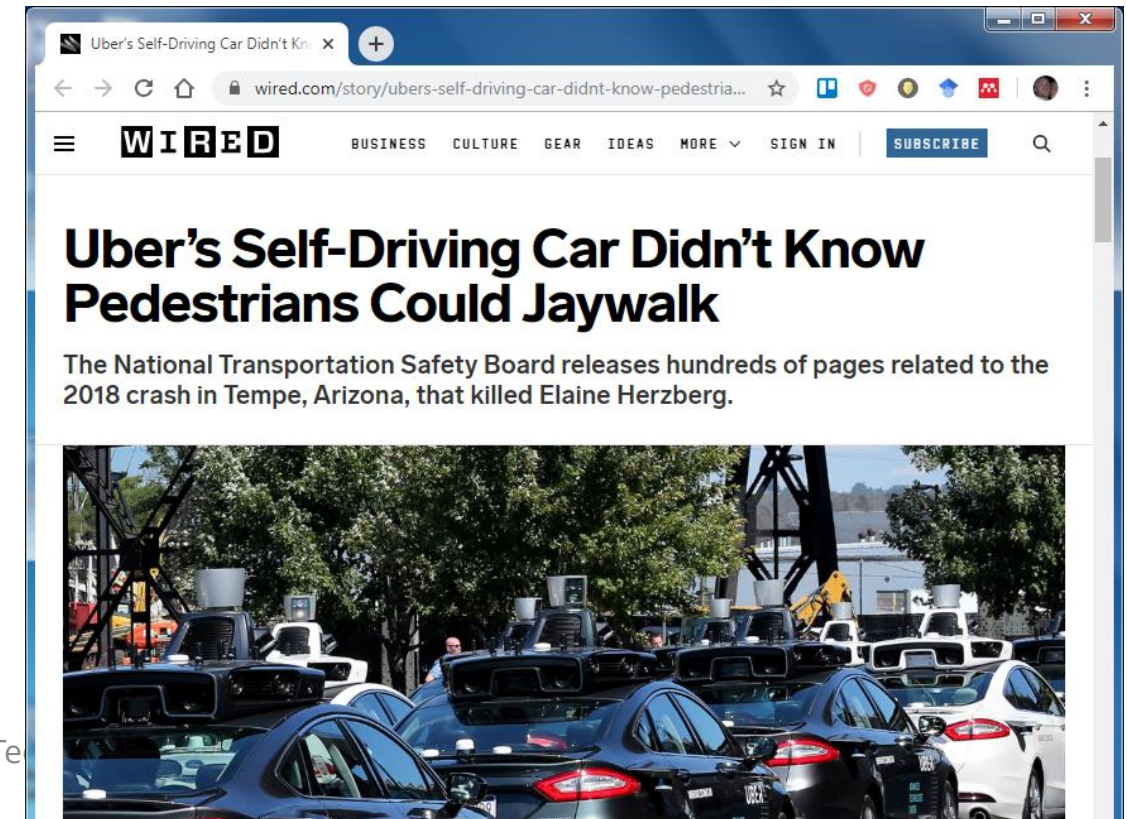
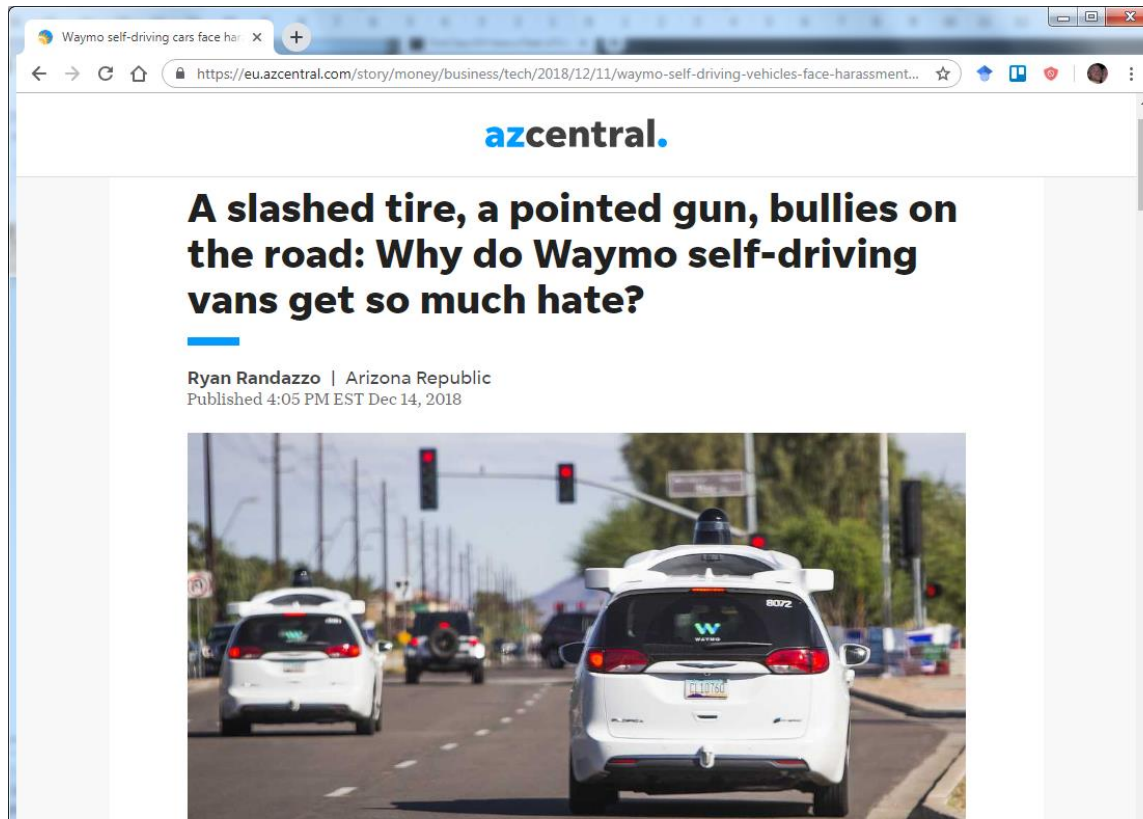
To what extent are these situations examples of **meaningful human control of automation?**

- A road user effectively interacting with an AV, transparently affecting the behaviour of the AV with their own behaviour
- An engineer studying and adjusting how an AV will interact with humans, using computer simulations across wide ranges of scenarios



# AV deployment: two main risks

- Human frustration ← subtleties of local interactions near-crashes
- Human injury ← crashes

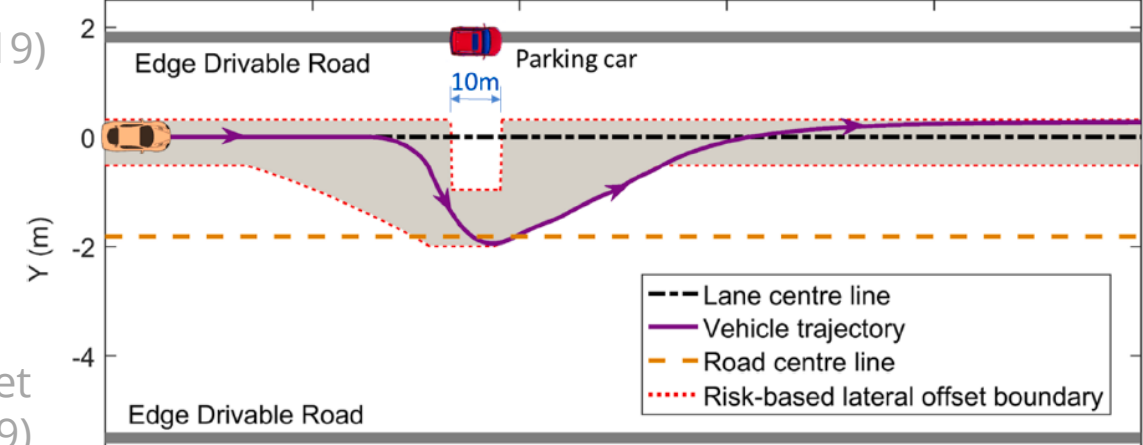


# Why high-fidelity models of human behaviour?

To make...

- ... AVs drive like humans?
- ... online AV predictions about human behaviour
- ... agents for virtual environments, for simulated AV testing

(Wei et al., 2019)



(Anderson et al., 2019)



(Waymo Safety Report, 2018)

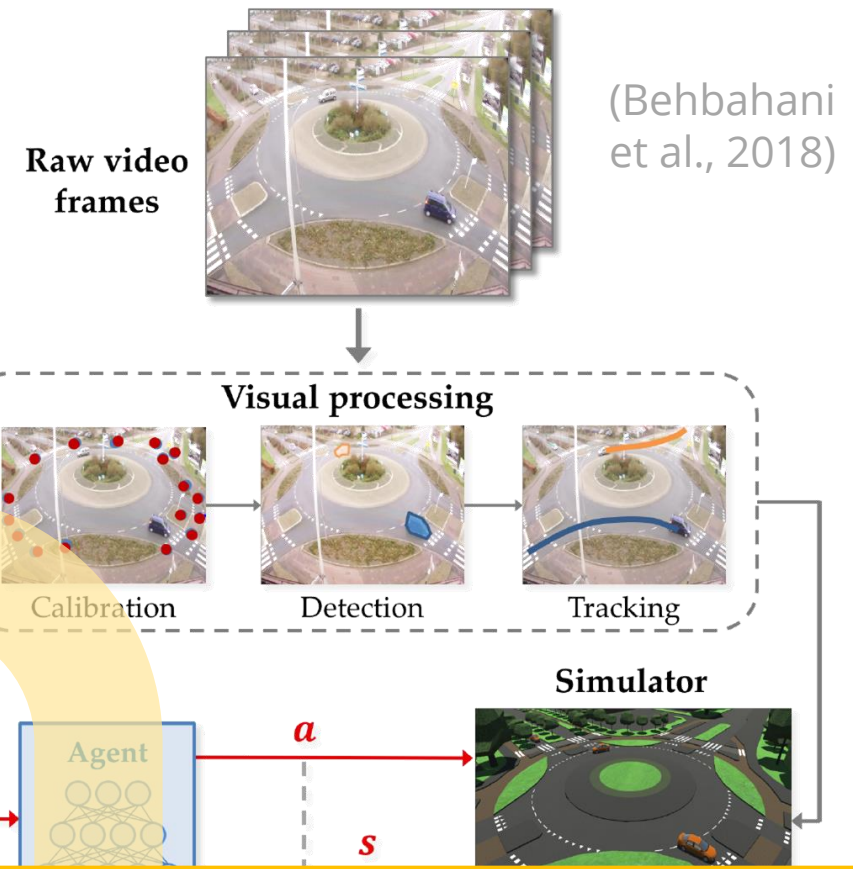


No humans here... ?

# Data-driven models

- Achieve realistic-looking routine traffic
- Challenges in relation to "main risks":
  - Human behaviour in (near-)crashes  
**Very rare in any real-traffic dataset**
  - Human behaviour in local interactions  
**How do we know models are capturing the important subtleties?**

→ **Complement with white-box neurocognitive models**



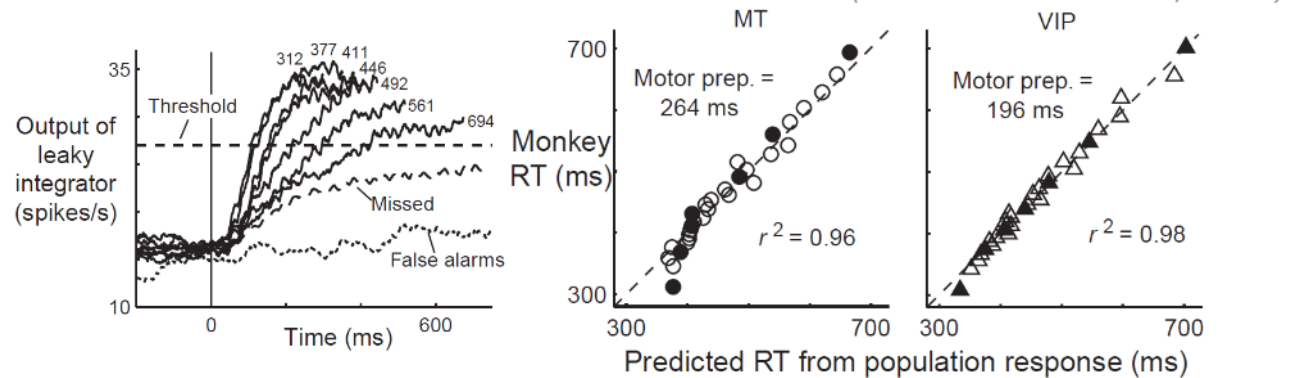
Insight into how mechanisms generalise





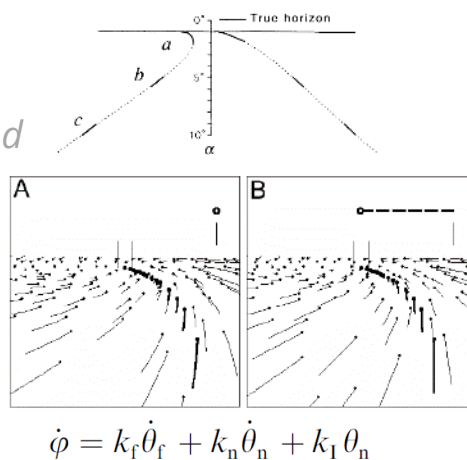
# Framework for routine and (near-)crash driving

## Evidence accumulation

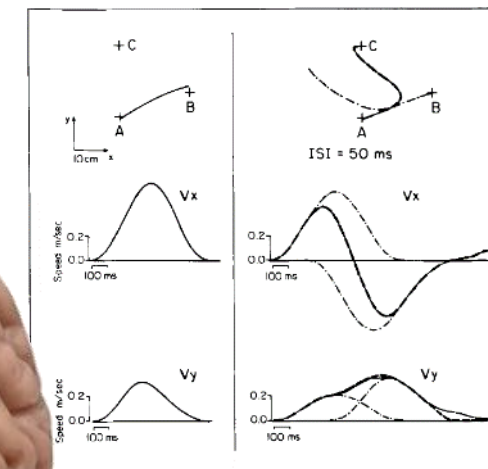


## Perceptual heuristics

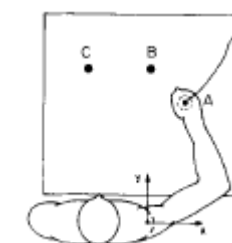
(Land and Horwood, 1995; Wann and Wilkie, 2004; Salvucci and Gray, 2004)



## Motor primitives



(Flash and Henis, 1991)



# Framework for routine and (near-)crash driving



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#1. Intermittent (motor primitive) adjustments

#2. Accumulation of evidence of various kinds

#3. Magnitude heuristics tuned to routine driving

#4. Sensory predictions

**Routine driving**

**Near-crash driving**

Closed-loop

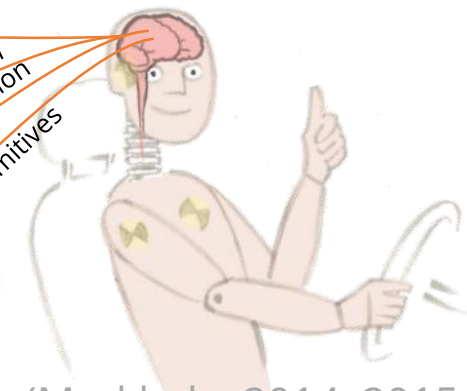
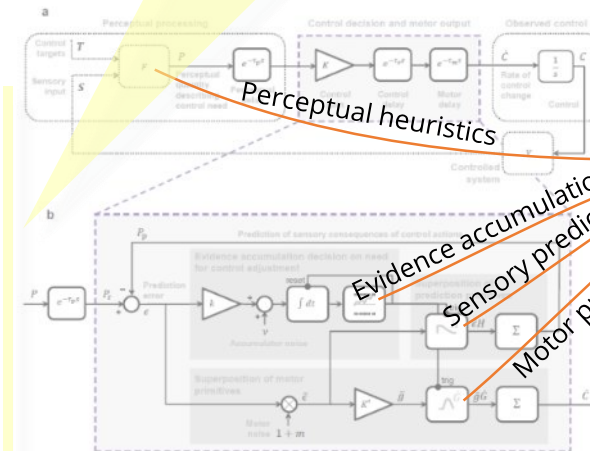
Open-loop

Short delays

Long, random delays

Well-adjusted control

Under- and overreactions



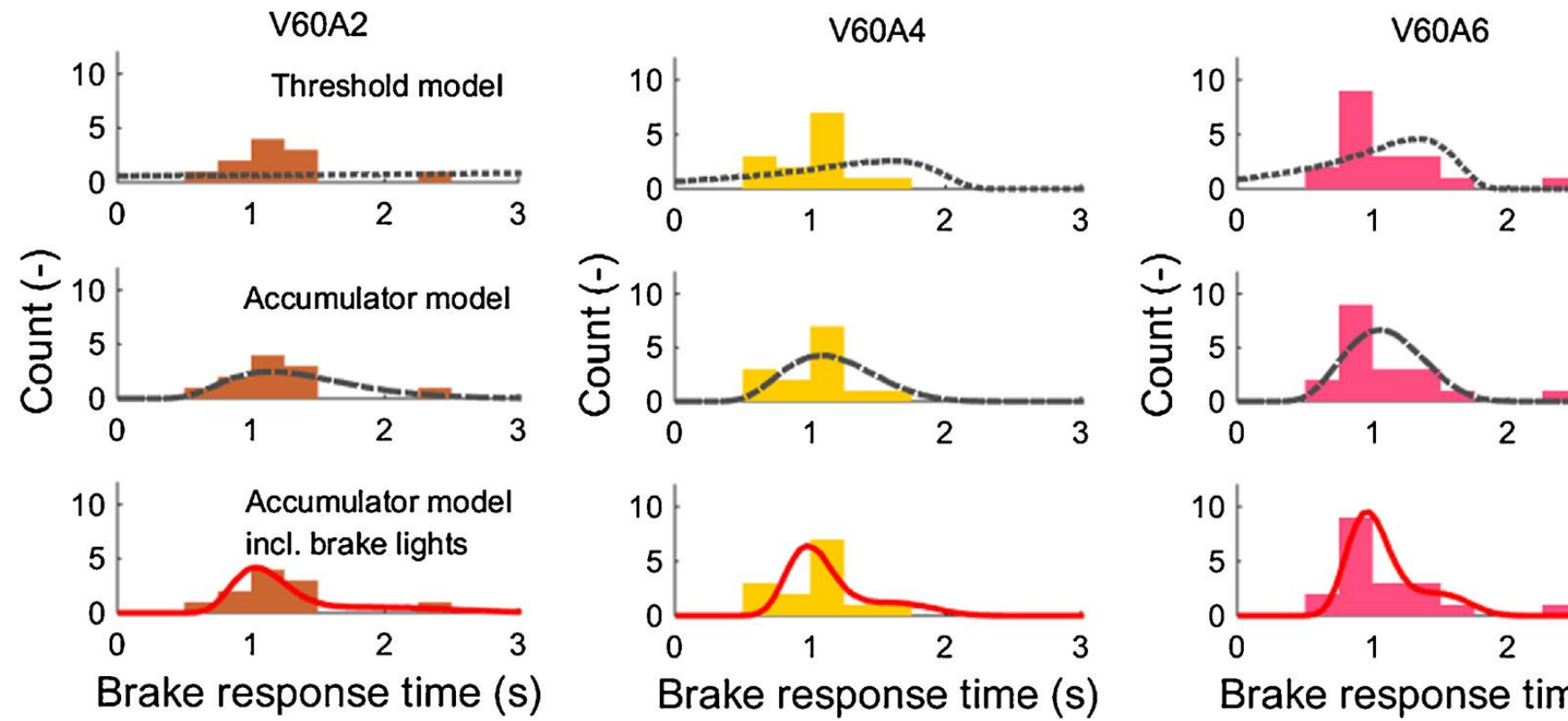
(Markkula, 2014, 2015;  
Markkula et al, 2018)



# ... Explains routine and (near-)crash braking



(Xue et al., 2018)



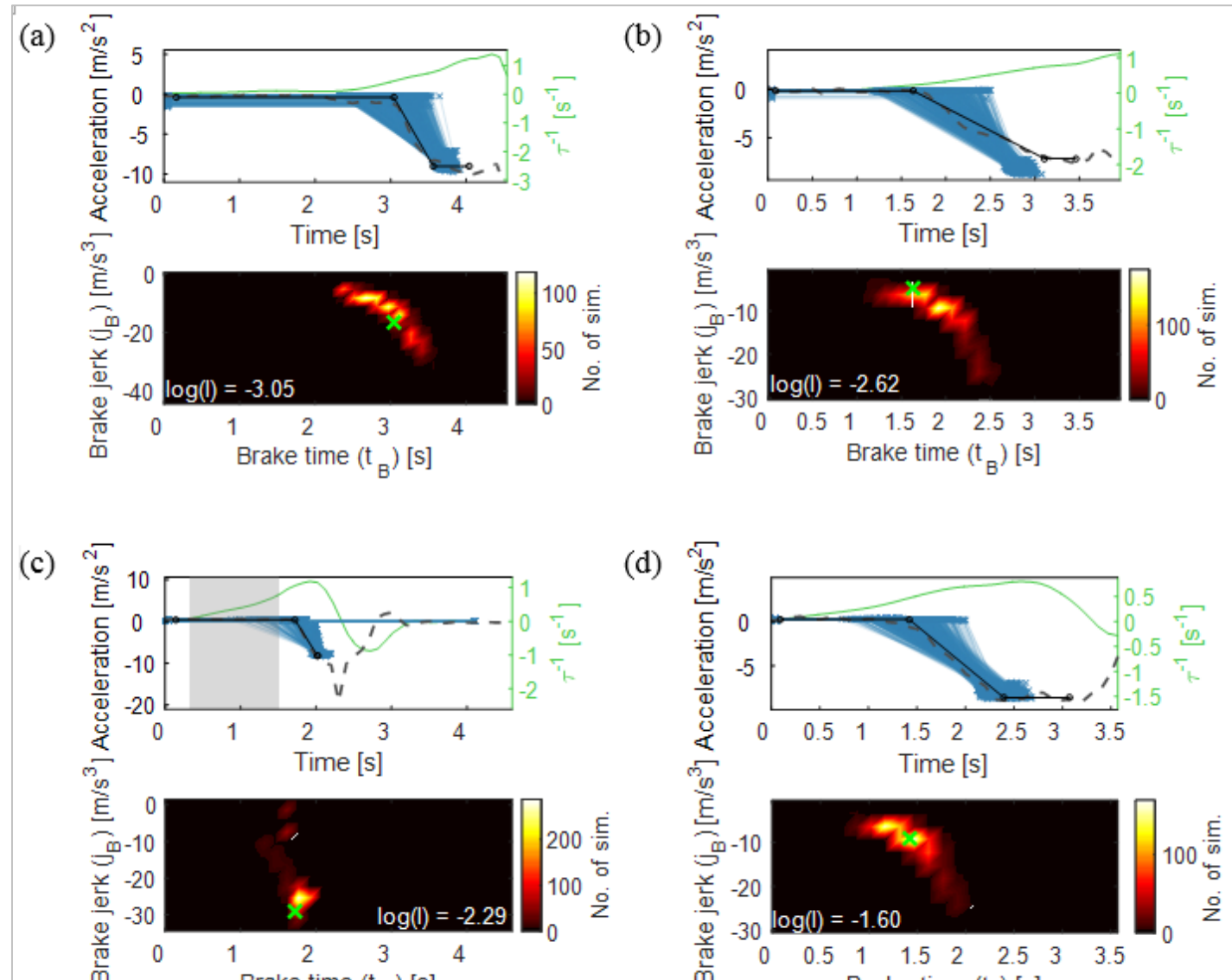
# ... Explains routine and (near-)crash braking



(Victor et al, 2015)

(Svård et al., 2020,  
<https://psyarxiv.com/6nkgv>)

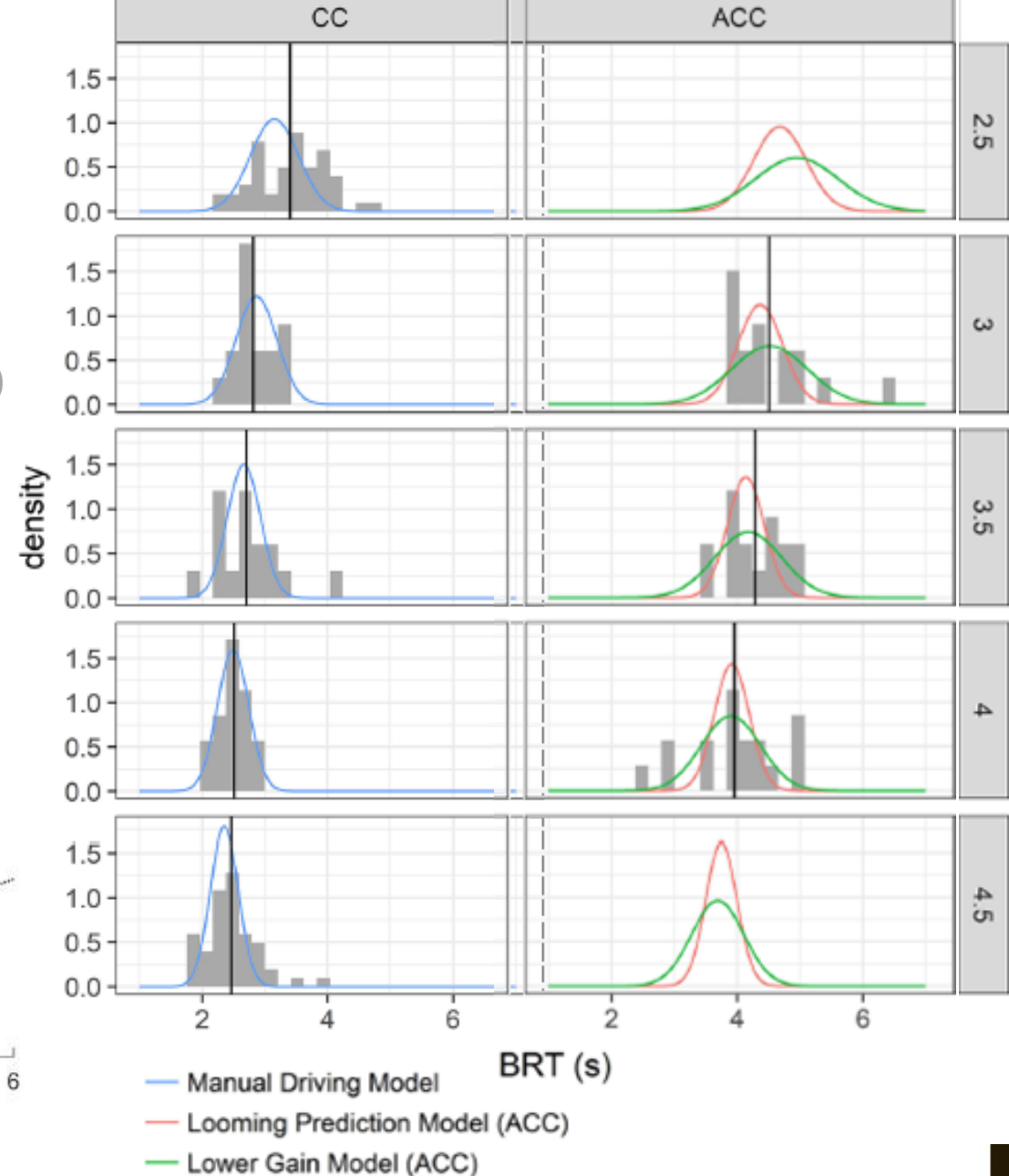
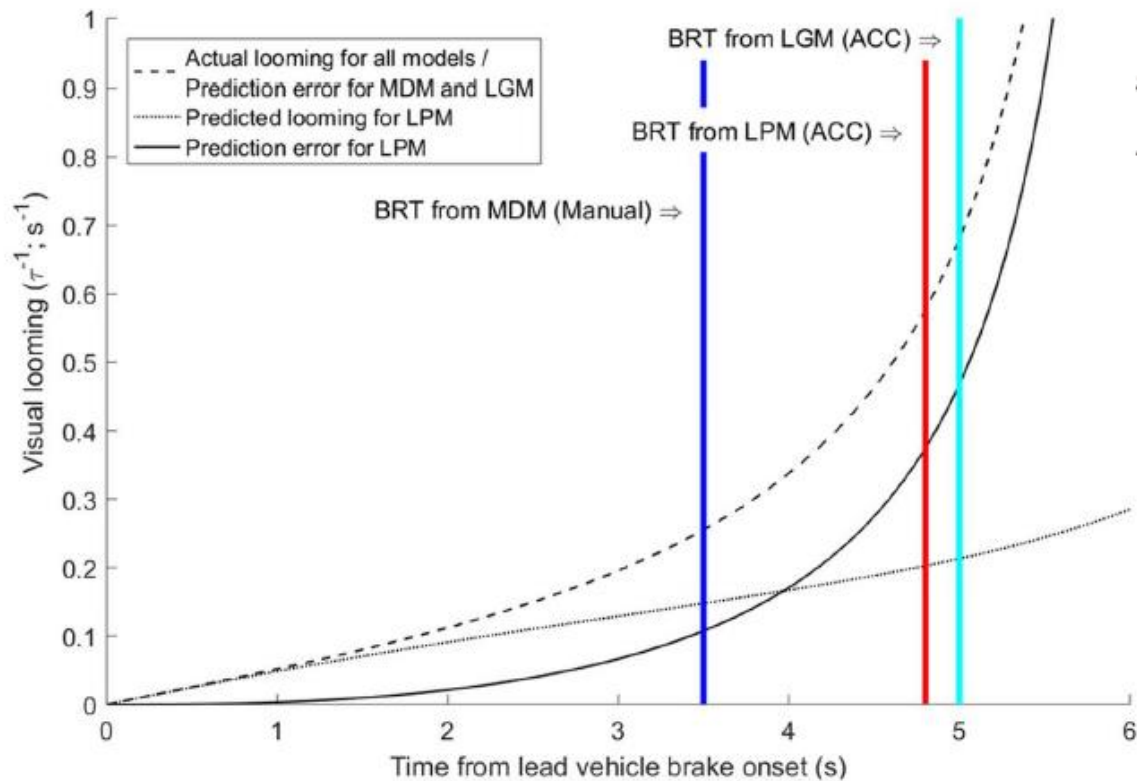
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# "Predictive processing" extension explains response to automation failures

(Piccinini et al, 2019)



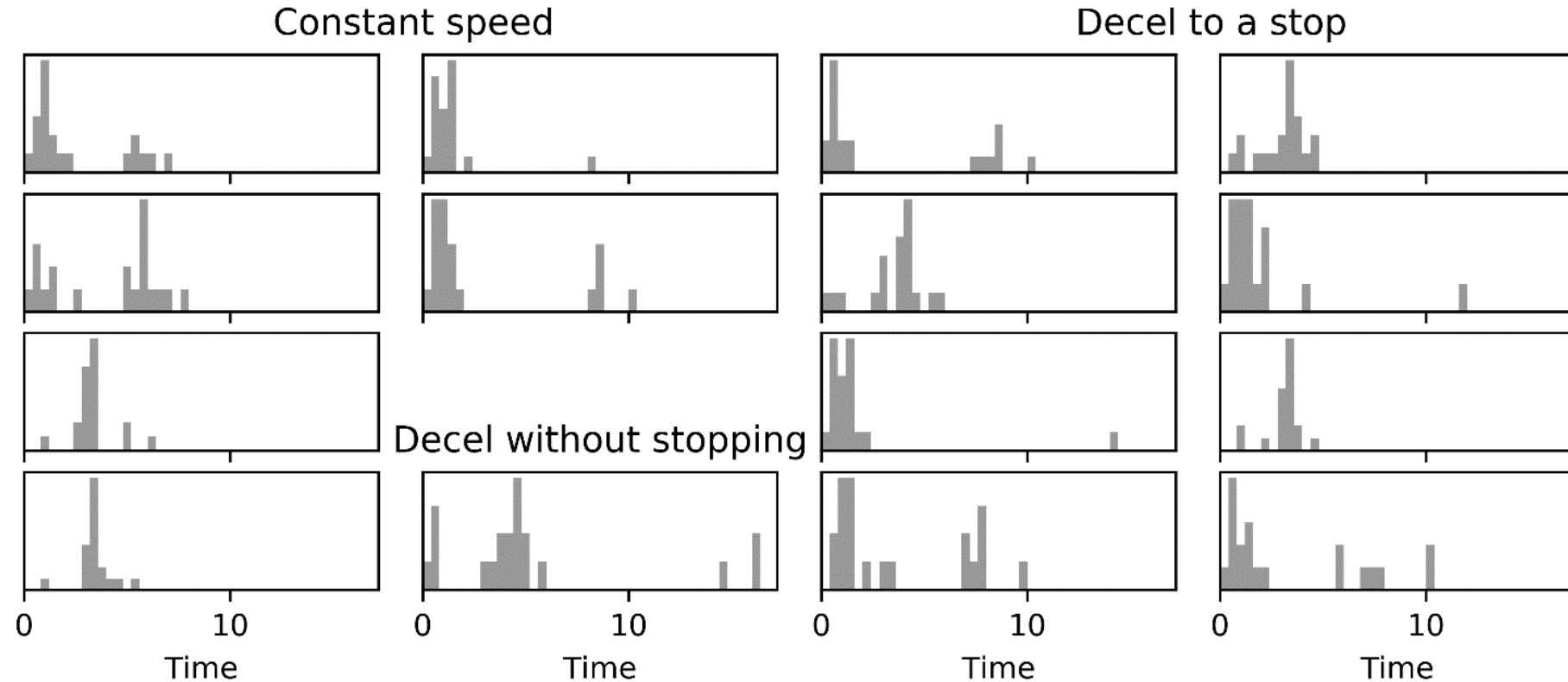
# Using EEG to peek into the decision process

(unpublished data removed)



# Generalising to road crossing interactions

(Giles et al., 2019)



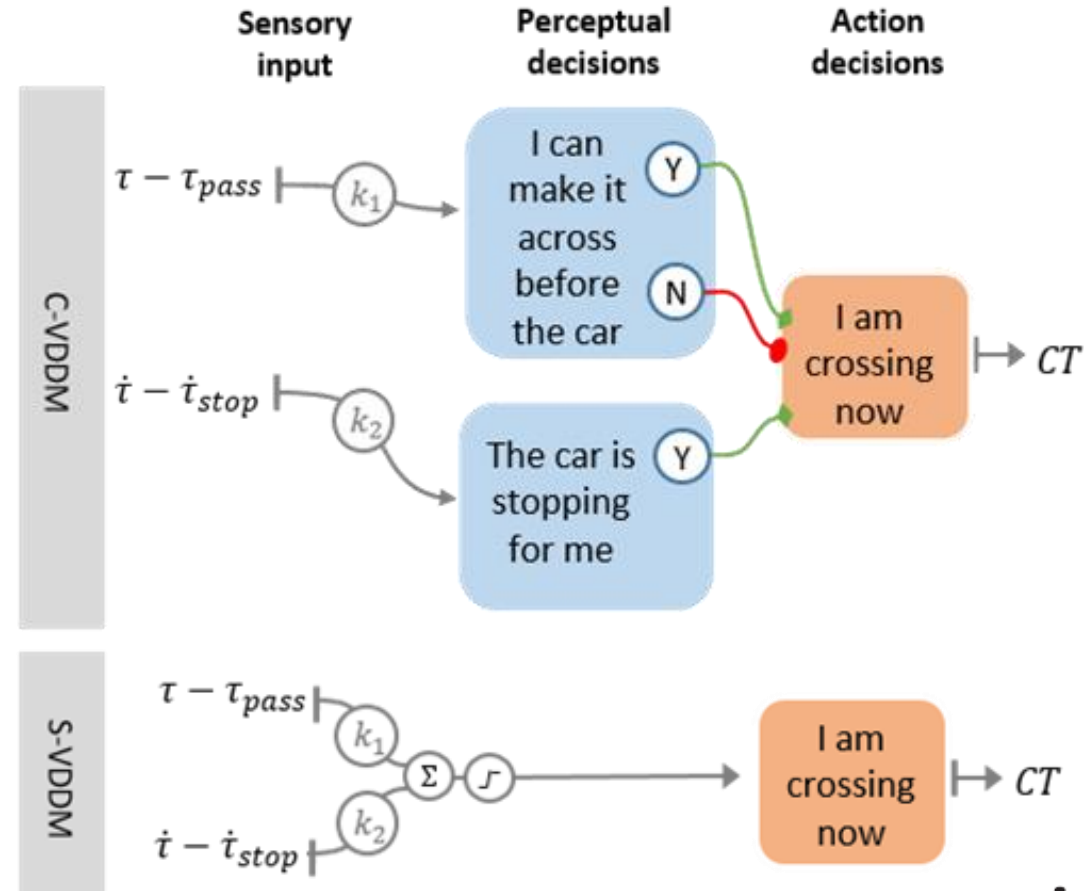
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# Generalising to road crossing interactions

(Giles et al., 2019)



interACT



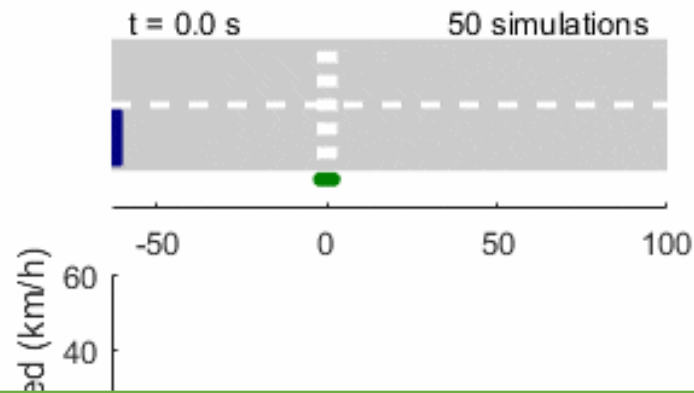
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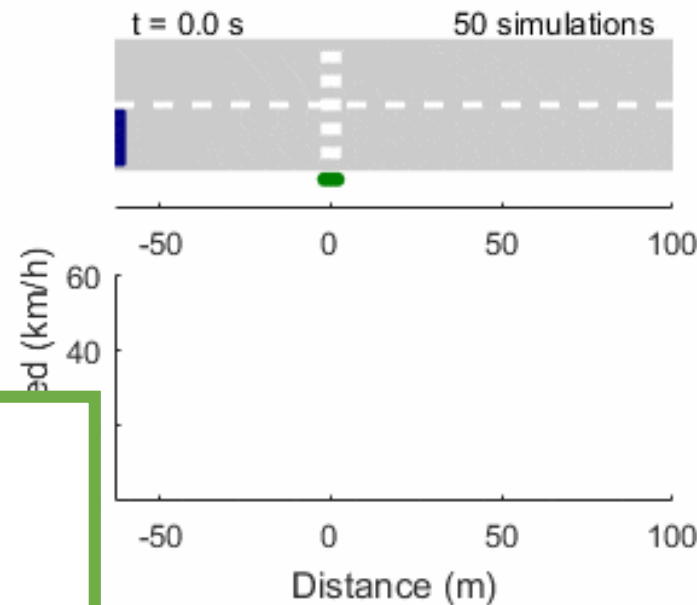


# Using models to optimise AV behaviour

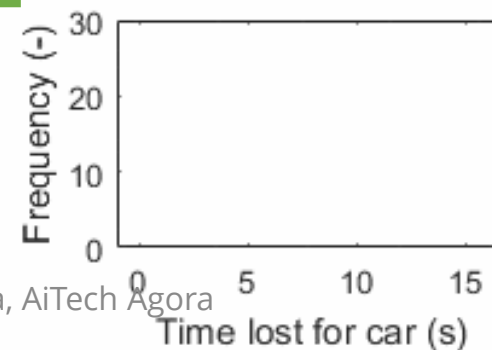
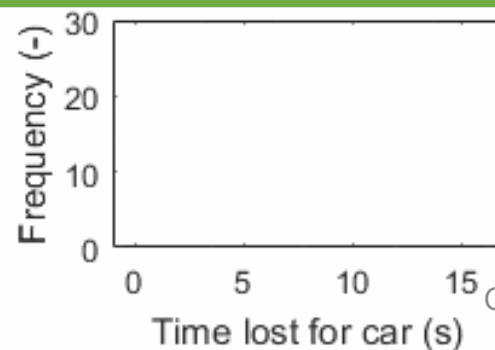
Braking as if to stop exactly at crossing  
( $1.7 \text{ m/s}^2$ )



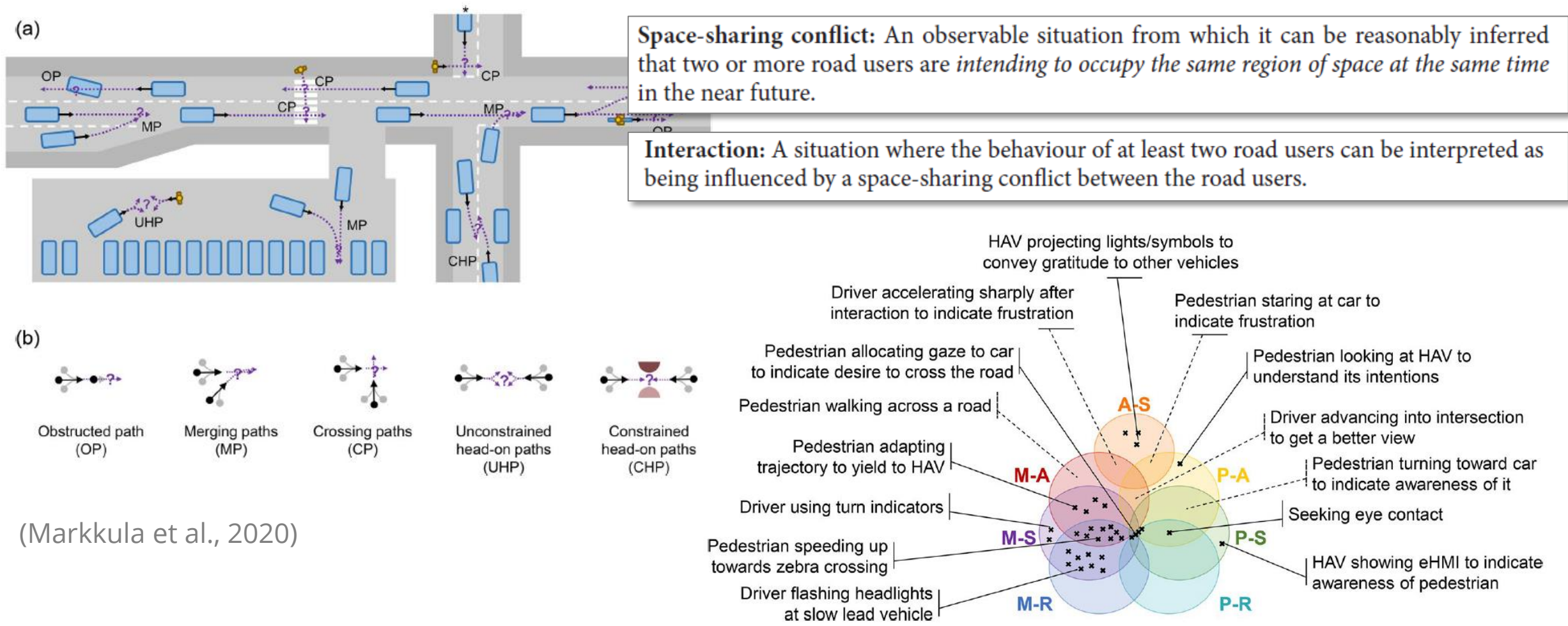
Braking just slightly harder  
( $2.2 \text{ m/s}^2$ )



Model code released:  
<https://osf.io/49awh/>

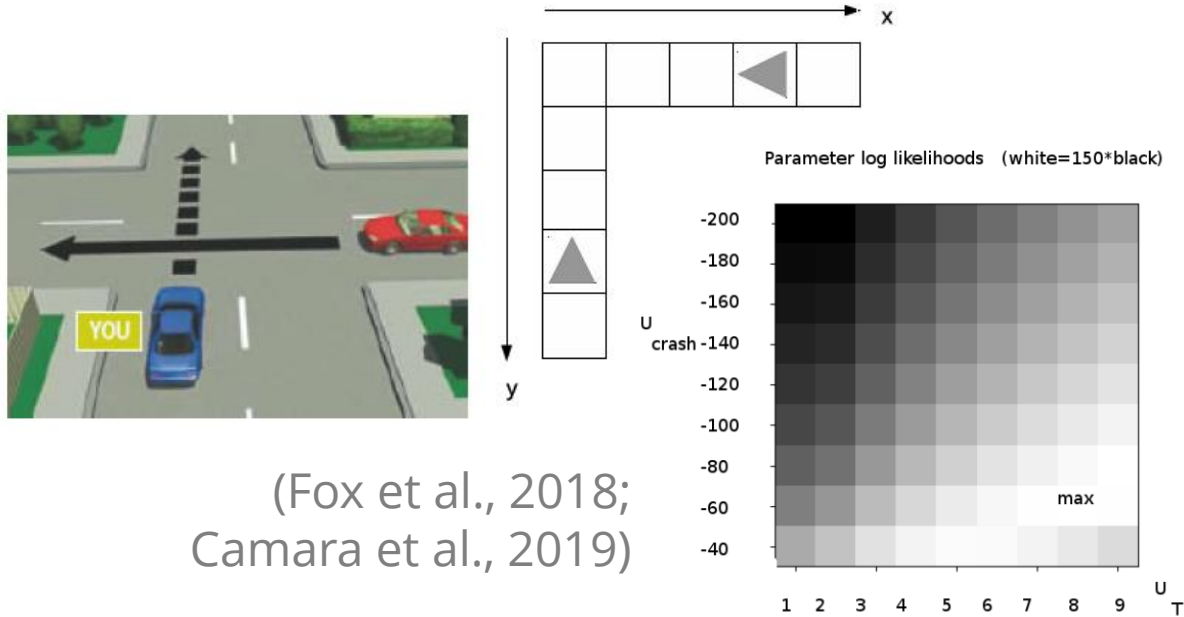


# But there is lots more to interactions in traffic...





# Strategic (/game theoretic) behaviour



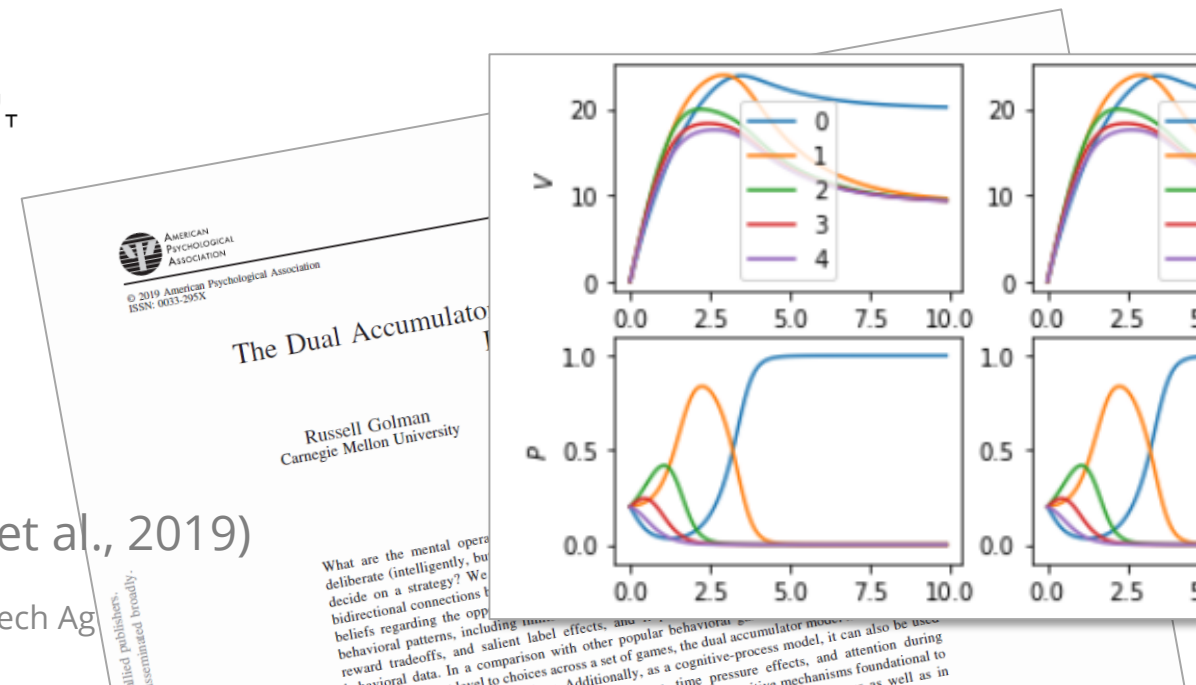
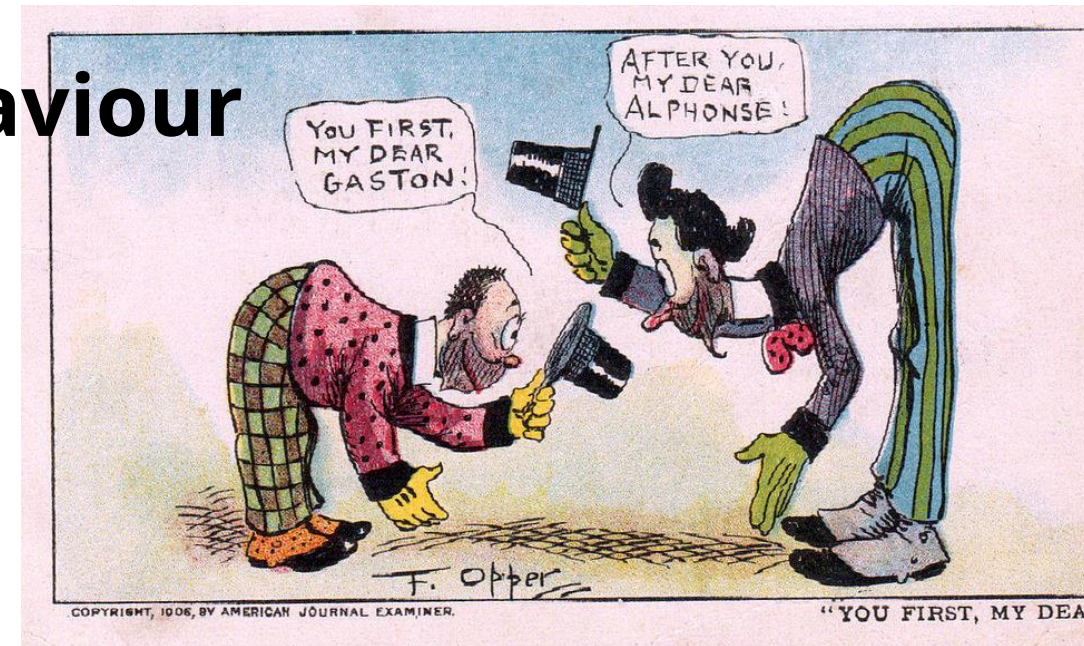
(Fox et al., 2018;  
Camara et al., 2019)

Fig. 7: Pedestrian behaviour preference

- Complication: human behaviour is often not game-theoretically optimal
- And humans value strange things

(Golman et al., 2019)

Gustav Markkula, AiTech Ag



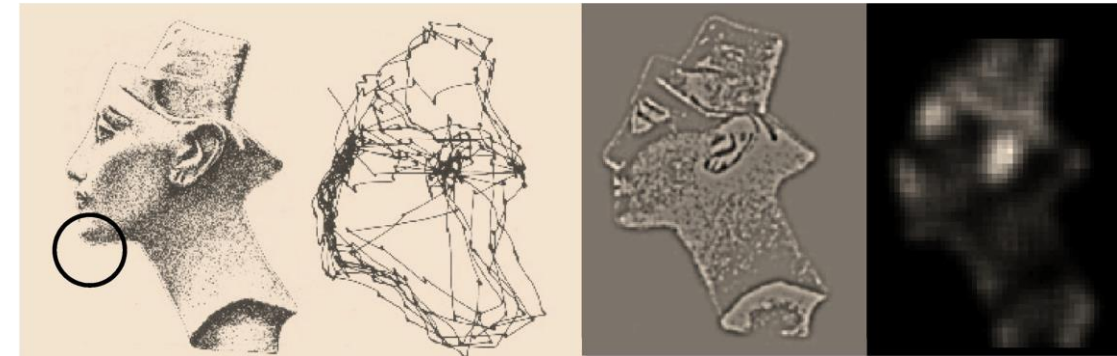
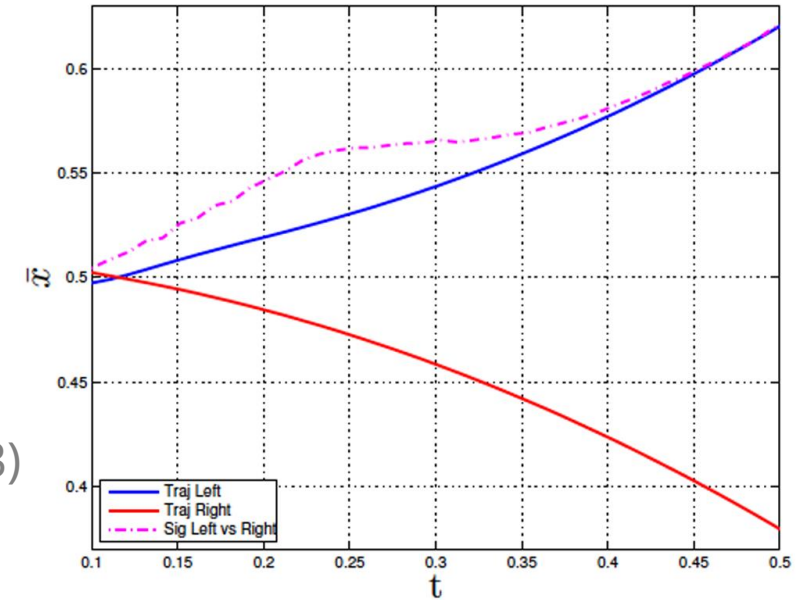
# Other important areas for further model development

Human...

- ... recognition of actions/intentions
- ... communication
- ... attention/gaze allocation

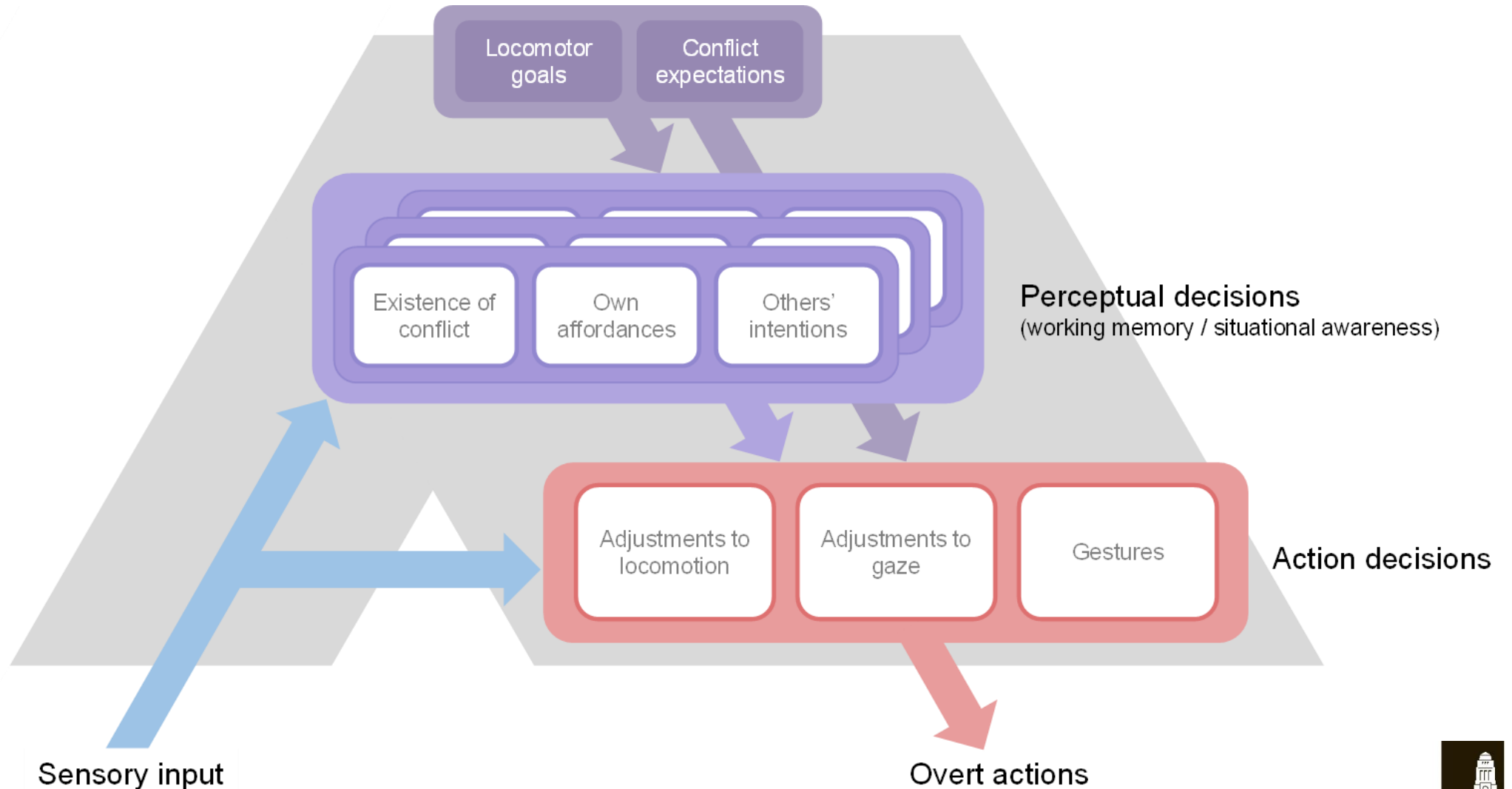
Overall: contemporary computational cognitive (neuro)science sort of provides the needed components...

(Pezzulo et al., 2013)



(Friston et al., 2012)

# Towards a neurocognitive modelling framework





# COMMOTIONS

## Computational Models of Traffic Interactions for Testing of Automated Vehicles

- 2019-2023, £1.4M UK project
- More complete neurocognitive models of interactions
- Investigate complementarity with data-driven models



"Green paper" inviting input:  
<https://osf.io/vbcaz>

An aerial, slightly hazy view of a busy urban intersection. The scene is filled with various vehicles including cars, buses, and taxis. Pedestrians are visible crossing the streets and walking along the sidewalks. The layout of the roads, including crosswalks and lane markings, is clearly visible. The overall atmosphere is one of a bustling, modern city environment.

**Safe and acceptable AVs require complementing data-driven models of human behaviour with neurocognitive models**

**We (and others) are working on this challenge  
– input and discussion more than welcome!**



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# Thanks!

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"Green paper" inviting input:  
<https://osf.io/vbcaz>



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