Autonomous Vehicles: A CPS or AI Grand Challenge?

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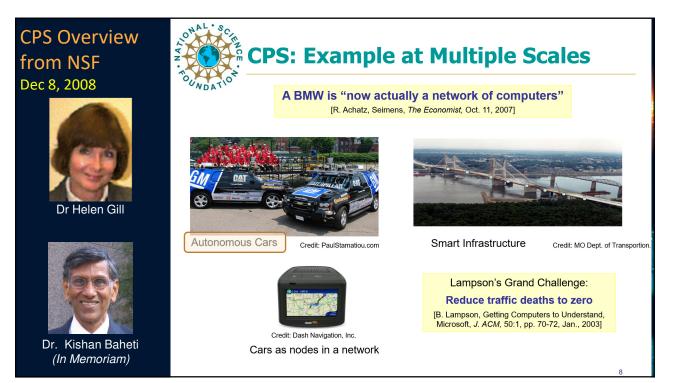
Keynote at NSF CPS PI Meeting, 2021

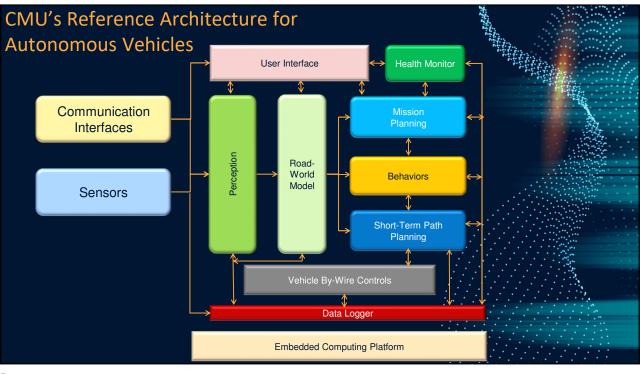


Motivation for Self-Driving Vehicles

- 1.3 million people die every year in automotive crashes
 - >94% of these crashes due to human error
- Traffic delays are very expensive
- The elderly, legally blind and physically-disabled people depend on others for transportation

Grand	 Possible Grand Challenges Zero automotive traffic fatalities, injuries minimized,
Challenges	and significantly reduced traffic congestion and delays Blackout-free electricity generation and distribution Reduce testing and integration time and costs of
for	complex CPS systems (e.g. avionics) by one to two
Cyber-Physical	orders of magnitude Perpetual life assistants for busy, older or disabled
Systems	people Extreme-yield agriculture Energy-aware buildings Location-independent access to world-class medicine Physical critical infrastructure that calls for preventive
June 25, 2008	maintenance Self-correcting and self-certifying cyber-physical
	 Self-correcting and self-certifying cyber-physical systems for "one-off" applications





	Challenges
AV Challenges 2013	Exogenous: The complexities & uncertainties of the real world
	 Weather, lighting, and road conditions; construction; accidents; obsolete information, loss of GPS.
	Endogenous: Online and safe recovery from failures of sensors, actuators, computing or communications.
	Mis-calibration, wear and tear, failures. Verification: How to verify and validate correctness?
	 Interactions: Vehicular Networks Reliability: cost and maintenance, customer acceptance
	 Cyber-Physical Security: thwart connectivity portal attacks Human factors
	Incremental deployment
	2 Legal and regulatory implications
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The AV Hype

2007: DARPA Urban Challenge "No longer science fiction"

2008: Google self-driving vehicles project and Waymo launched in 2016

2013: Ottomatika Inc gets founded

2015:

- Uber Advanced Tech Center in Pittsburgh
- First cross-country autonomous drive in the US with Ottomatika technology
- Ottomatika gets acquired by Delphi (→ Aptiv → Motional)
- TuSimple founded to build driverless trucks

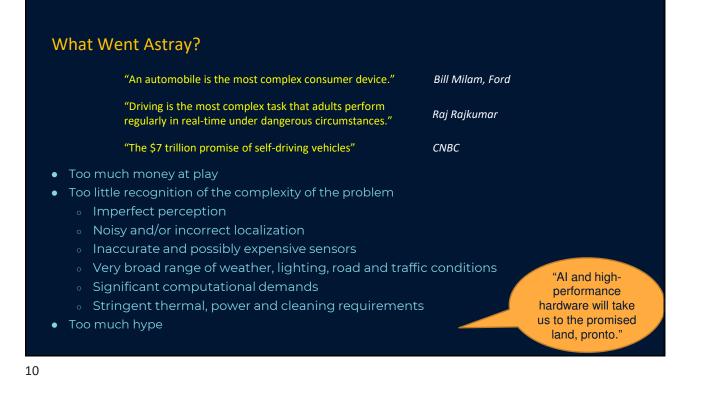
2015 -: Tesla "Basic problems have been solved - we'll have full autonomy in 2 year

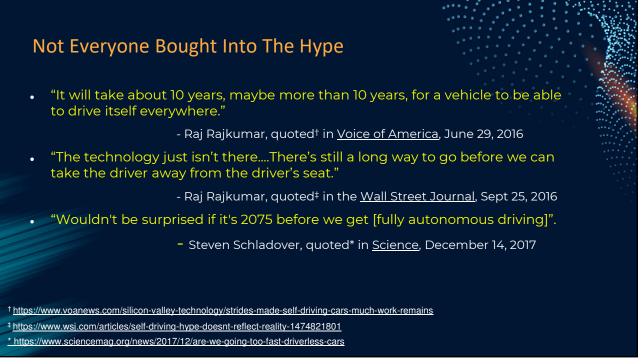
- 2018: "Fully autonomous cross-country drive"
- 2019: Robotaxi Teslas will earn \$30K per year for the owners
- $_{\circ}$ $\,$ 2020: Only regulatory considerations stand in the way

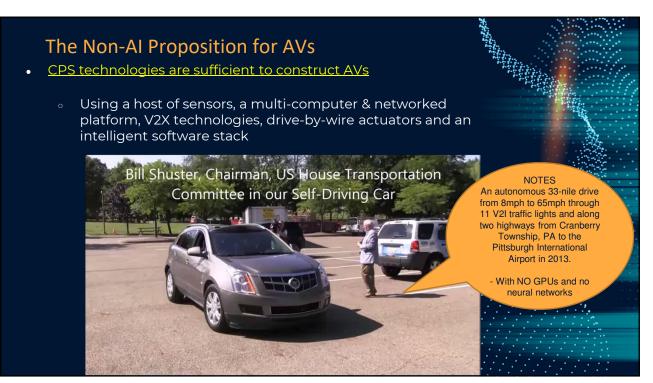
2016: Nuro founded to build driverless delivery vehicles

2017: Lyft "Most Lyft vehicles in 2021 will be fully autonomous, car ownership will go.away by 2025."

... and Ensuing Disillusionment VISIBILITY 2019-: Tesla Peak of Inflated Expectations 2021: "Full Service Driving Beta" being sold – actual functionality is driver assist Plateau of Productivity Multiple fatalities from "AutoPilot" in US and China Slope of Enlightenment 2020: Waymo valuation drops from \$175B to \$30B 2020: Uber ATG sold to a startup company Trough of Disillusionment 2021: Lyft Level-5 unit sold to Toyota **Technology Trigger** TIME







The Non-Al Proposition for AVs

- <u>CPS technologies are sufficient to construct AVs</u>
 - Using a host of sensors (e.g. high-res and FMCW lidars), a multi-computer & networked platform, V2X technologies, drive-by-wire actuators and an intelligent software stack
- The infrastructure, vehicles or people communicate using V2X while on or near roadways.
- Possible but unfortunately *not* entirely practical in the foreseeable future.
- With AI support, fewer and less expensive sensors can be used without pervasive infrastructure support.
 - Al is therefore necessary in practice to construct AVs.
 - Al is integral for AVs to "see" <u>the world designed for human</u> <u>drivers</u>
 - But end-to-end AI is NOT the answer either.

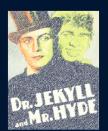




The Two Sides of Humans

Super Intelligence

- The Vision system
 - Human eye: very high resolution with depth info, iris/aperture control, focus
- The (Re)cognition system
 - Near-instant detection and identification of objects and organisms
 - Learn from a few examples
- (Unreasonable?) reasoning power
 Logical reasoning, generalization and
 - Logical reasoning, generalization and specialization
 - Spatial and temporal intelligence
 - Intuition and experience
 - Adaptability (to new vehicles, cultures, rules)
- Coordinated sensing & actuation
 - Interactions: eye contact, social exchanges, game theory





- Distractability
 - Dozing off, conversations, diversion of interest, intoxication

Imperfect sensing

- Drivers with poor vision, age impairments,
- Eyes at the front only

Non-ideal reasoning

- Incorrect understanding of vehicle dynamics
- Poor multitasking
- Overconfidence
- Instincts and intuition can be *wrong* at critical moments

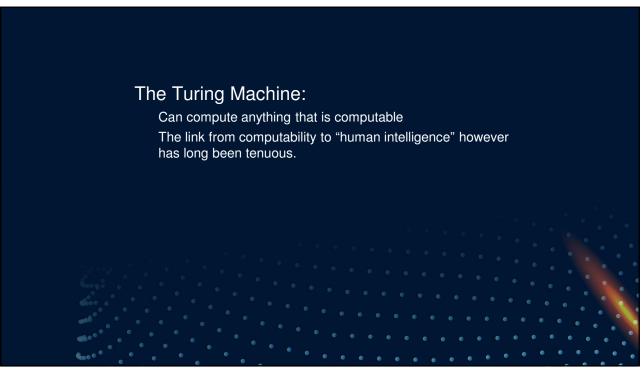
Potentially slow reactions

Poor reflexes due to age or health

Questions to be Asked of Today's AI

- What does a deepnet really learn?
 - Appeal of deepnets: "no programming"; specify only inputs and outputs
 - Moves the onus of feeding ALL the right inputs to the "programmer"
 - Garbage in \rightarrow Garbage out
 - Biased inputs \rightarrow Biased outputs
- What happens when an input not resembling training data is provided to a deepnet?
 - <u>No one</u> knows for sure
- Does a deepnet really learn any generalization or specialization beyond becoming a high-quality pattern matching scheme?
 - With high likelihood, no.





A Turing Test for Automated Vehicles

If a self-driving vehicle cannot be distinguished from a human-driven vehicle by humans outside the vehicle, it is said to have passed the *Turing Test for Automated Vehicles*.

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In Reality, AVs will be compared to human drivers More conservative much of the time, More aggressive some of the time, May get stuck occasionally, and May take "counter-intuitive" actions that surprise and/or upset humans around the AV

Counterexample #1: Feedback Control Systems

- Stability, controllability and observability properties
- Ability to deal with disturbances
- Latency implications
- Decidable vs Undecidable
 problems

Example:

- Inverted Pendulum
 - Yes, one can use machine learning to mimic how a human balances a stick on one's palm – BUT WHY?

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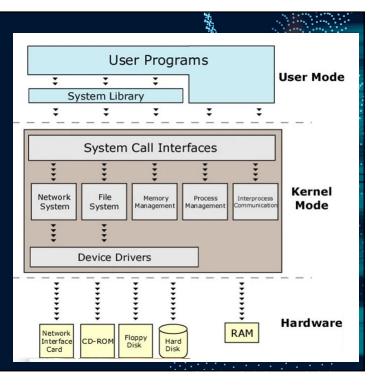
Counterexample #2: Our Global Computing and Network Fabric

- Human innovations that enable general-purpose computations
- Hardware accelerators that perform efficient special-purpose computations
- Wired networks and "network protocol stacks" that led to the creation of the internet and WWW
- Wireless networks that removed the need for a 'leash'

All these are the **products of ingenious human minds** leveraging the power of computing and communications (as serfs).

Domain #3: Operating Systems

- Linux, Windows, MacOS, Android, iOS, ...
- Brilliant human minds design an architecture comprising of multiple interacting modules
 - "Divide and conquer" for coding, debugging, testing and deployment
 - Yield usability, upgradability and generality.
- <u>Al is very far from being mature</u> enough to design and realize very general and very usable software platforms.



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Domain #4: Real-Time Systems

- Sound mathematical foundations
 - Worst-case schedulability:
 - Rate-monotonic and earliestdeadline-first scheduling policies for guaranteed timing behavior
 - Provable properties under practical constraints:
 - Synchronization, periodics/aperiodics, mixed criticality, multiprocessors, endto-end latencies, jitter management, ...
 - Has evolved into a science of real-time systems



Domain #5: Software Apps

- Complex software applications:
 - Web browsers (https, TCP/IP, GUI, clientserver)
 - Web servers (cloud, client-server, cybersecurity)
 - MS Office (easy-to-use WYSIWYG for nontechnical people, tight integration, inexpensive hardware)
 - Compilers (parsing, lexical analysis, code optimization, code generation, platform independence)
 - **"Zoom"** (high-bandwidth low-latency communications, audio/video sync)
- Yes, the zoom "virtual background" uses Al
 - But... it also chops off the shoulder, parts of the head, etc. at times
 - Would you rely on it for a safety-critical system? Should you?









- Dealing with the uncertainties of an unconstrained road-world operating environment
- Highways, urban, suburban, ex-urban and rural roads
- Crashes, Road debris, Emergency Vehicles
- School zones, kids, the elderly, strollers, wheelchairs, jaywalkers and absent-minded crossing
- Weather conditions
- Road conditions
- Lighting conditions
- Construction zones
- Wild life



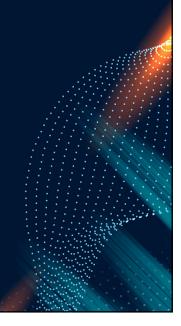
What Does Being "Safe" Mean?

Strong safety:

Zero crashes, zero injuries and zero fatalities with the AV.

Weak (practical) safety:

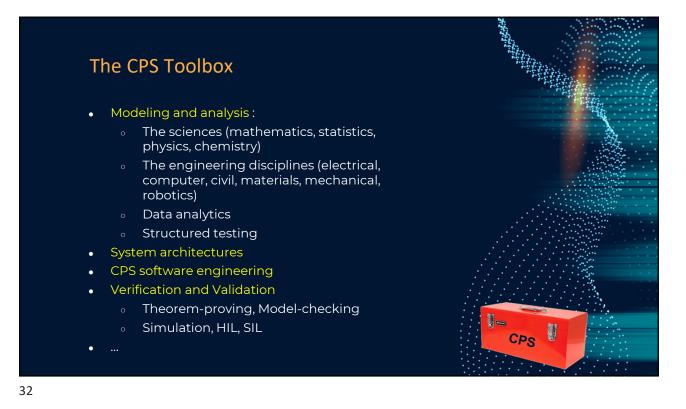
The AV is **not "at fault"** when a crash occurs, per accepted legal precedents and insurance practices.

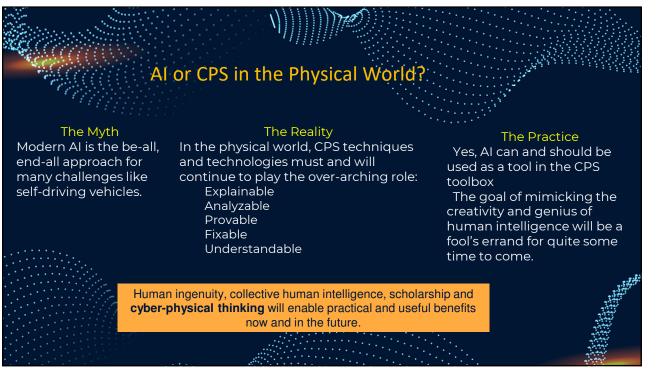


Other CPS AV Challenges

- Sensor Fusion
 - minimizing false positives and false negatives
- Generalization and Adaptation
 - Knowledge Encoding
 - Customizability to Different Countries & Driving Cultures
- Behavioral
 - Co-Existence with Other Vehicles and Humans
 - Human Expectation and Comfort
 - Safe behaviors in the presence of faults
 - "Ethical" decision-making
 - Dealing with Pranksters
- Cyber-Physical Security

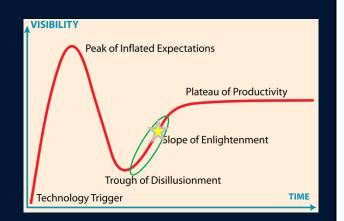






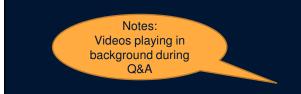
Where Do We Stand vis-à-vis AVs?

- Multi-sensor fusion with AI as <u>one</u> key enabler
- Aggregation of well-designed subsystems
- Composability of behaviors
- Safe path planning
- Controllability and observability



LEVEL 2 HIGHWAY DRIVING 25 MILES @ 65MPH

Level-2 Highway Driving @ 65mph



Level-4 Driving from 15 mph to 65mph

LEVEL 4 DRIVING FROM 15MPH TO 65MPH