Safety Implications of the Transition to CAVs: KY CAV Crash Savings Demonstrator

Reg Souleyrette Univ. of Kentucky

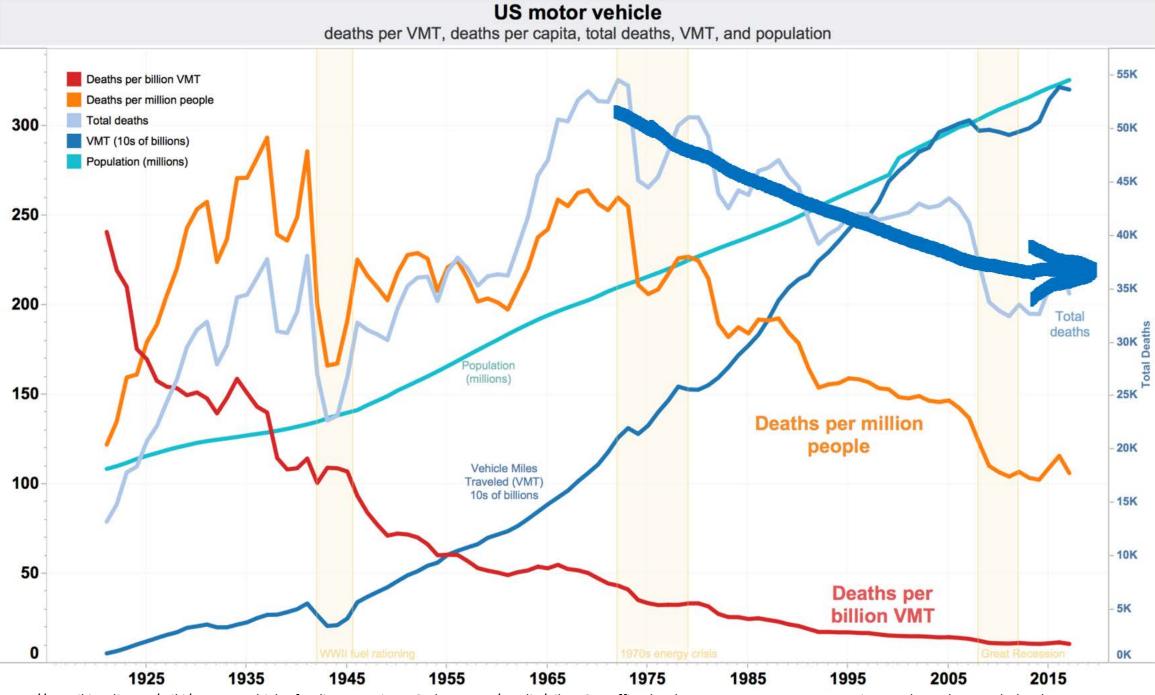




Outline

- Motivation (for studying CAV safety and transition)
- Opportunity (potential for crash reduction)
- Underlying trends/components
- Predicting Safety
- Predicting Safety for CAVs
- KY CAV Crash Savings Demonstrator
- Next Steps

Motivation



https://en.wikipedia.org/wiki/Motor vehicle fatality rate in U.S. by year#/media/File:US traffic deaths per VMT, VMT, per capita, and total annual deaths.png



Examining accident reports involving autonomous vehicles in California

Francesca M. Favarò 🖾, Nazanin Nader, Sky O. Eurich, Michelle Tripp, Naresh Varadaraju

Published: September 20, 2017 • https://doi.org/10.1371/journal.pone.0184952

The Rand Study

Why Waiting for Perfect Autonomous Vehicles May Cost Lives

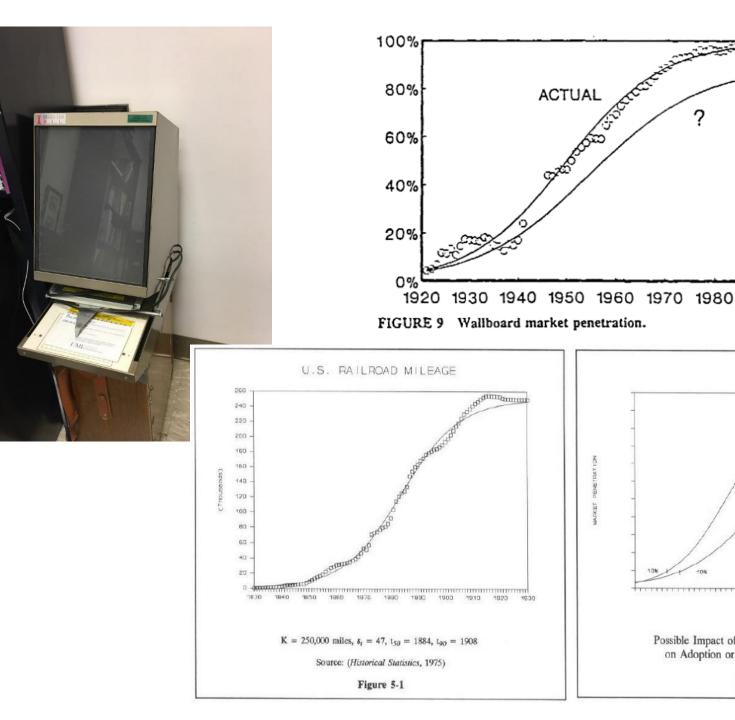


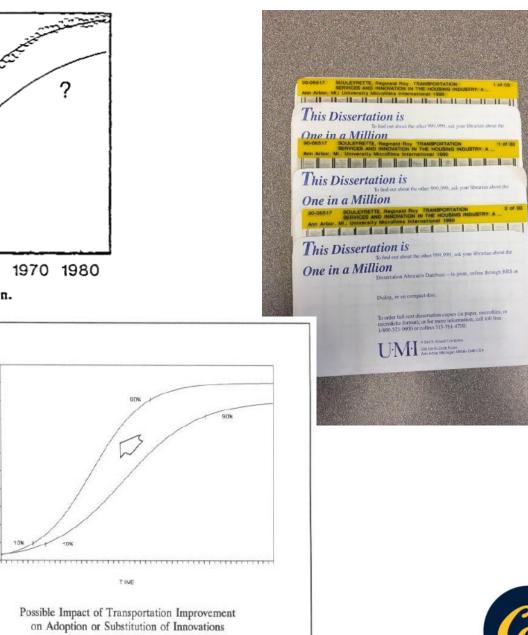
video

https://www.youtube.com/watch?v=Ybr41Sy7K3g



November 7, 2017





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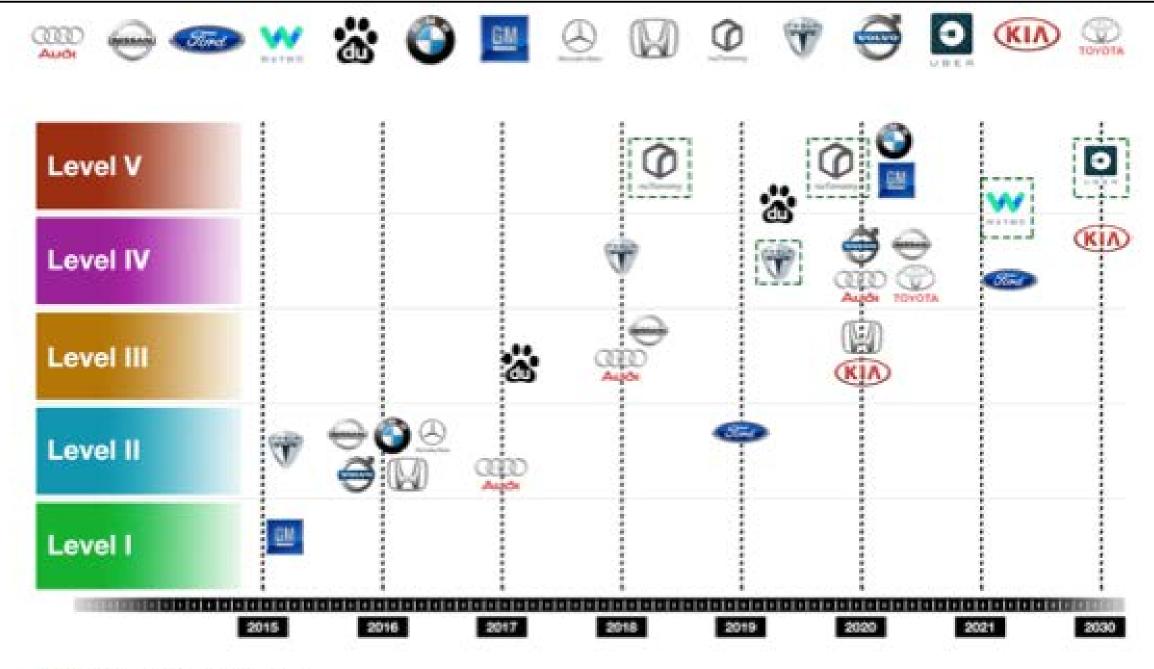
Figure 4-1



Opportunity

So, now we've got this cool, new technology





Proposed for shared riding use

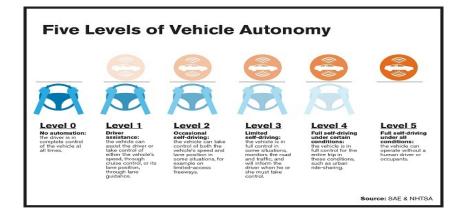
http://journals.plos.org/plosone/article?id=10.1371/journal.pone.0184952

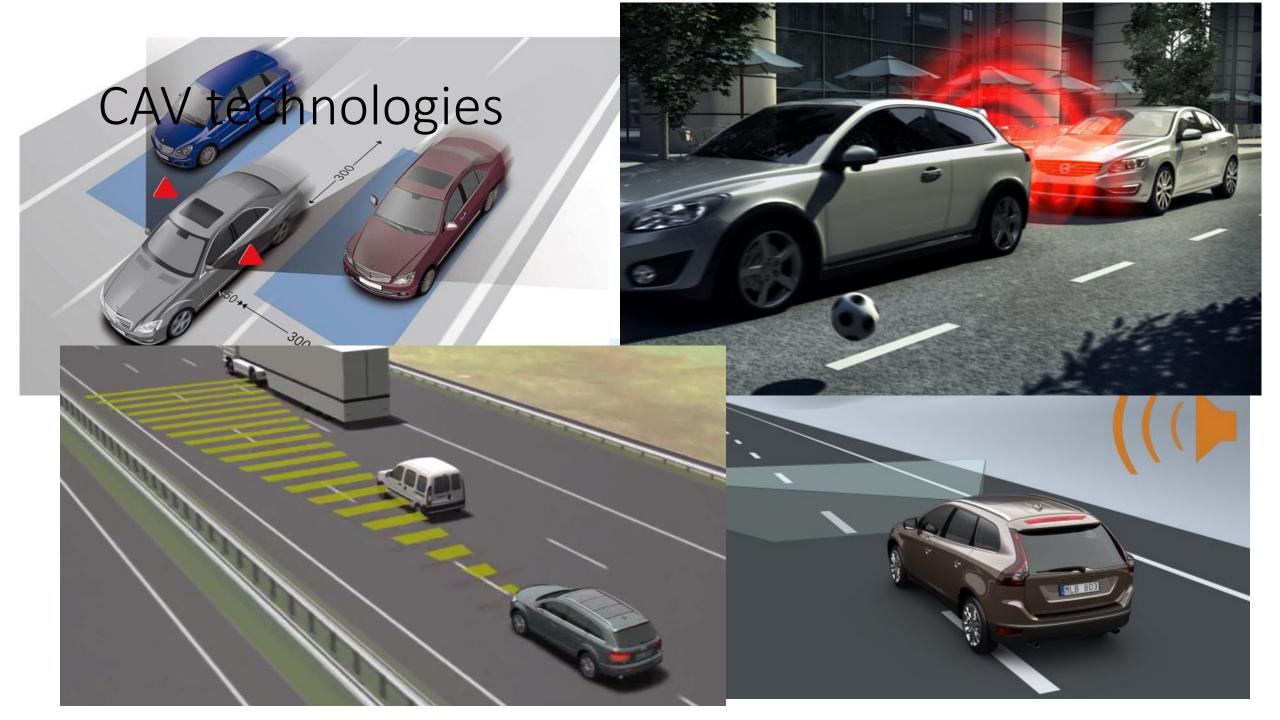
CAV technologies

• L0

- Blind Spot Monitoring
- Lane Departure Warning
- Traffic Sign Recognition
- Left-Turn Assist
- Adaptive Headlights
- L1
 - Adaptive Cruise Control
 - Cooperative Adaptive Cruise Control
 - Automatic Emergency Braking
 - Lane Keeping (Page 10, good for striping)
 - Electronic Stability Control
 - Parental Control

- L2
 - Traffic Jam Assist
 - High Speed Automation
 - Automated Assistance in Roadwork and Congestion
- L3
 - On-Highway Platooning
 - Automated Operation for Military Applications
- L4
 - Google's Driverless Car (Not tested in bad weather)
 - Kill Switch
- L5
 - Fully Autonomous





Safety the problem, CAVs the solution?

- They say 94% of crashes are due to human error
- but ...can CAVs address all human factors?



"The human error components of walking, biking, motorcycle use will not be completely mitigated even with perfect automated vehicles. Also there is a residual category of tree falls on car, sinkhole, washed away in flood, suicide by auto, that won't be mitigated by smart cars" – S. Polzin, personal correspondence



Safety the problem, AVs the solution?

- Initially, mixing in AVs may make things worse (for some crash types)
 - Money spent on AVs could be spent on less-expensive "safer" cars
 - Money spent on making the infrastructure work with AVs and CAV-tech enabled cars could be spent making roads safer for non-CAVs*
 - Interactions between AVs and non-CAVs may be more dangerous than a driver-operated system
 - Effectiveness requires proper use, can be a distraction, users compensate for risk
- In the long run even, some things might be "worse" ...

* some improvements may help both types of cars



Self-Driving Cars Will Make Organ Shortages Even Worse

We need to prepare for that now.

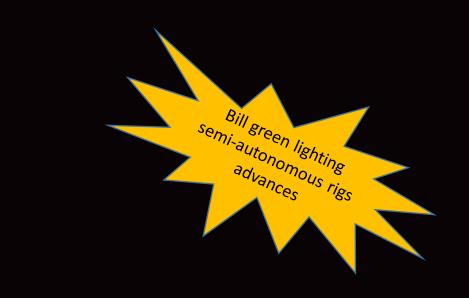
By Ian Adams and Anne Hobson



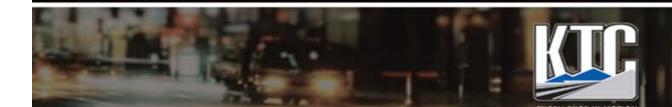
A pilot model Uber self-driving car on display on Sept. 13 in Pittsburgh.

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16% of all organ
donations come from
motor vehicle accidents
- U.S. Department of
Health & Human
Services.



AV safety performance, so far

- 22 of 26 reported AV accidents, AV not at fault (CA crash study)
 - Of the remaining four, two were in manual mode
 - The other two were at speeds <10 mph

Type of Vehicle	Total Number of Vehicles	Percentage of Fleet	Percentage of Total Reported Accidents	Total Miles Travelled	Accident Frequency	Miles per Accident
Google Prototype	37	61.7%	46%	403,226	2.4e-5	40,322
Retrofitted Lexus	23	38.3%	54%	649,841	1.8e-5	54,153

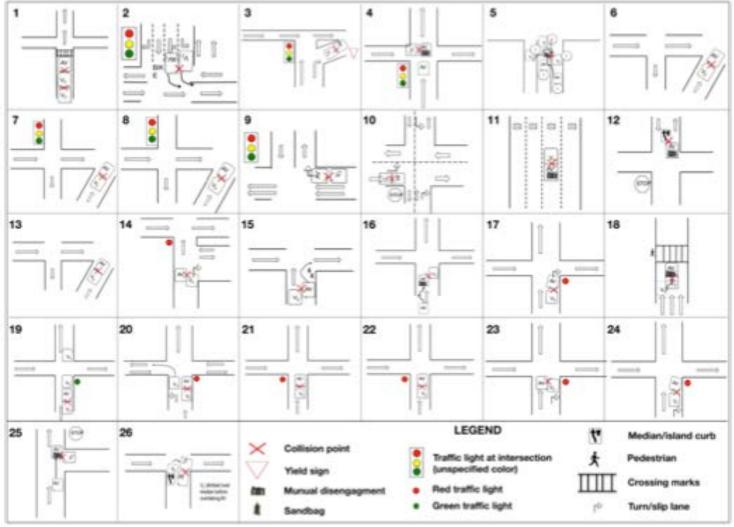
https://doi.org/10.1371/journal.pone.0184952.t002





Autonomous car traffic accidents in California by speed, 2014-2017 Autone. ar was stopped Autone ar was stopped Autone Auton Autonomous mode 🗖 Conventional mode Autonomous 11+ mph car at fault https://www.axios.com/humans-cause-most-self-driving-car-accidents-1513304490-02cdaf3d-551f-46e6-ad98-637e6ef2c0b9.html

CAV Crash Modalities ... useful to our work



http://journals.plos.org/plosone/article?id=10.1371/journal.pone.0184952

Underlying trends/components

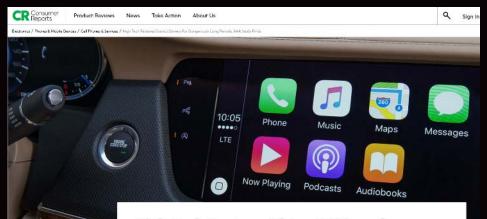
2002 Überlingen mid-air collision



Interactions with

non-CAVs

Tesla fatality



High-Tech Features Distract Drivers for Dangerously Long Periods, AAA Study Finds

Researchers find navigation and texting systems are the biggest offenders



Proper Utilization

Distractions

Why Car Safety Features Can Be Dangerous

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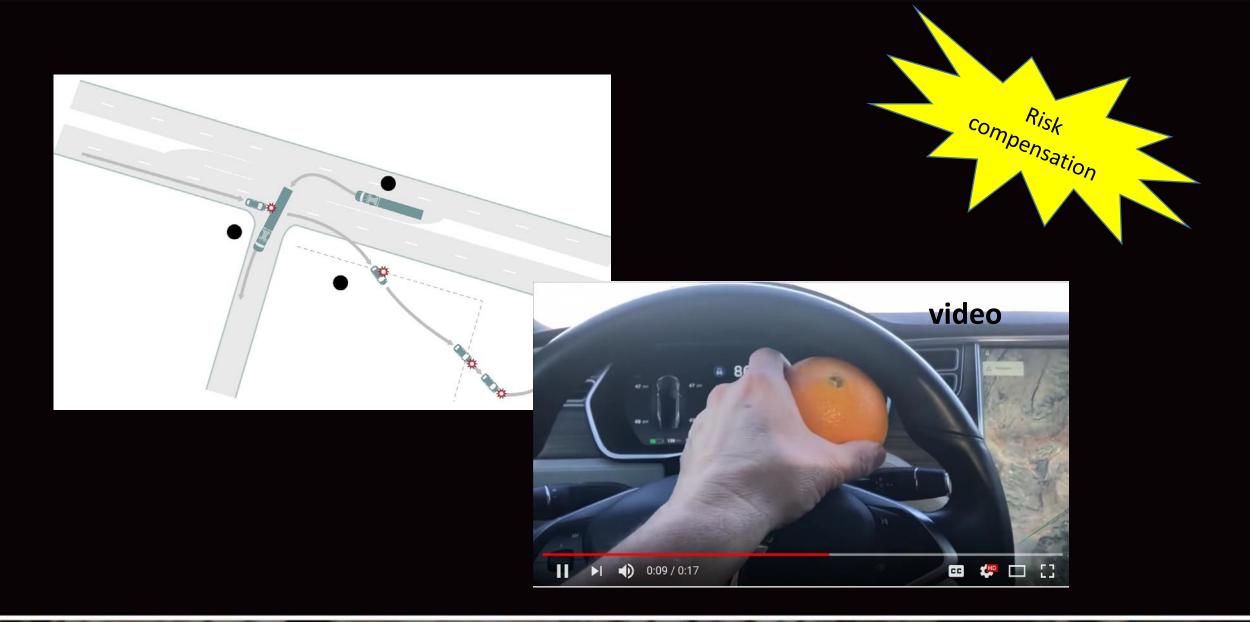


Joann Muller, FORBES STAFF *I write about industrial innovation and the global auto indus*

Car safety is on the minds of just about everyone th proposing new vehicle-safety legislation that would recorders and backup-brake technology.

If too many warning signals are triggered at once, drivers could become confused about how to respond, he says. Or they might ignore the warnings altogether, instead relying on the car to take evasive action.

"Technology is a two-edged sword," acknowledges Preuss. "It could be the thing that kills you, or the thing that absolutely saves you."





Predicting Safety

Conventional Safety Data Analytics



- State of the art is Highway Safety Manual
- Finally getting folks to accept Empirical Bayes (not everyone)
- Some researchers pointing out limitations (exposure, temporal effects)
- All assume static technology* not even time series...

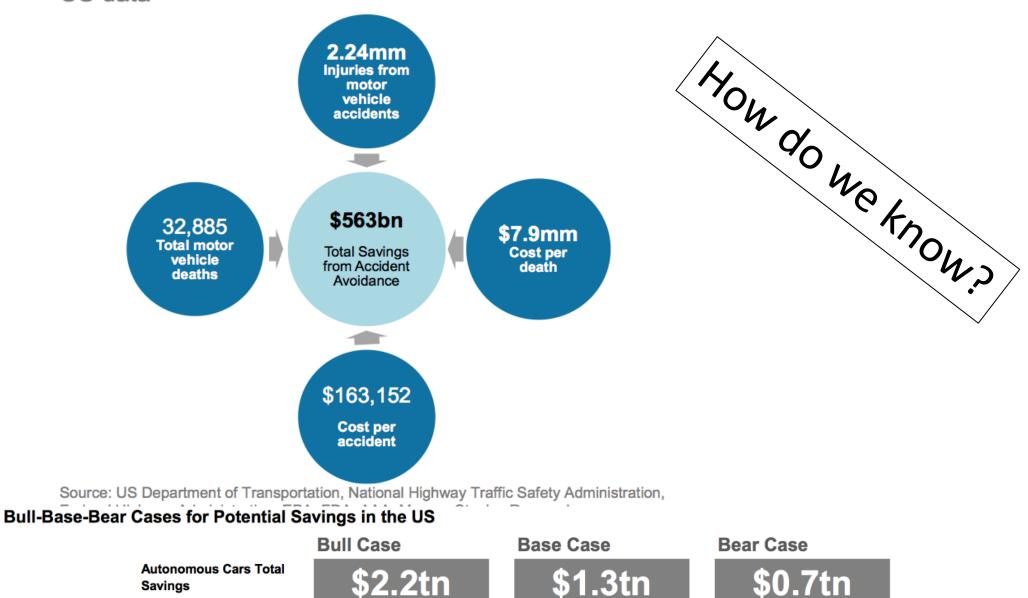
Sort of like estimating vehicle safety using past performance of the horse?

* Also a problem for design standards, e.g., Green Book



Cost of Motor Vehicles-related Fatal and Non-fatal Injuries

US data



Savings

Let's look back ... Issues in 1900:

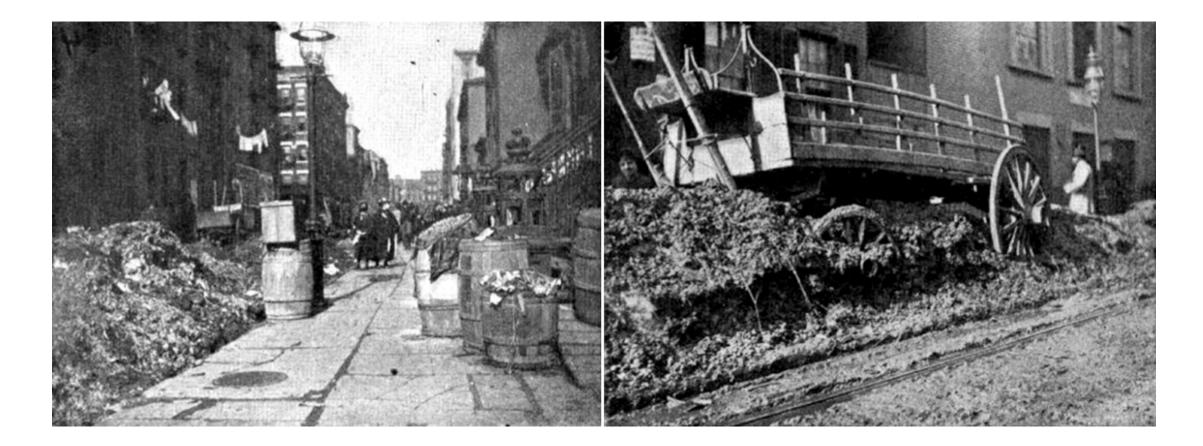




You can tell that cars and electric trolleys are winning in 1911 by the near-absence of mounds of horse poop on the streets. 15 years before this film was shot, 2.5 million pounds of horse manure (and 60k gallons of horse urine) were left on the streets of NYC every day, and it was very evident:



It's real hard to think beyond something like this ..



http://www.hhhistory.com/2017/08/the-great-manure-crisis-of-1894.html

JUNE 12 2012 6:30 AM

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FROM SLATE, NEW AMERICA, AND ASU

I Think, Therefore I Yam

When farmland is scarce, will we all eat roots and tubers?

By Will Oremus



Sacrilege

Horses are most commonly infected with bovine **TB**, which may still carry a zoonotic risk; this means it may be passed to **humans**. ... Strains of this disease **can** affect mammals of all species, including **humans**, **horse**, cattle, cats, and dogs.



Tuberculosis in Horses - Symptoms, Causes, Diagnosis, Treatment ... https://wagwalking.com/horse/condition/tuberculosis



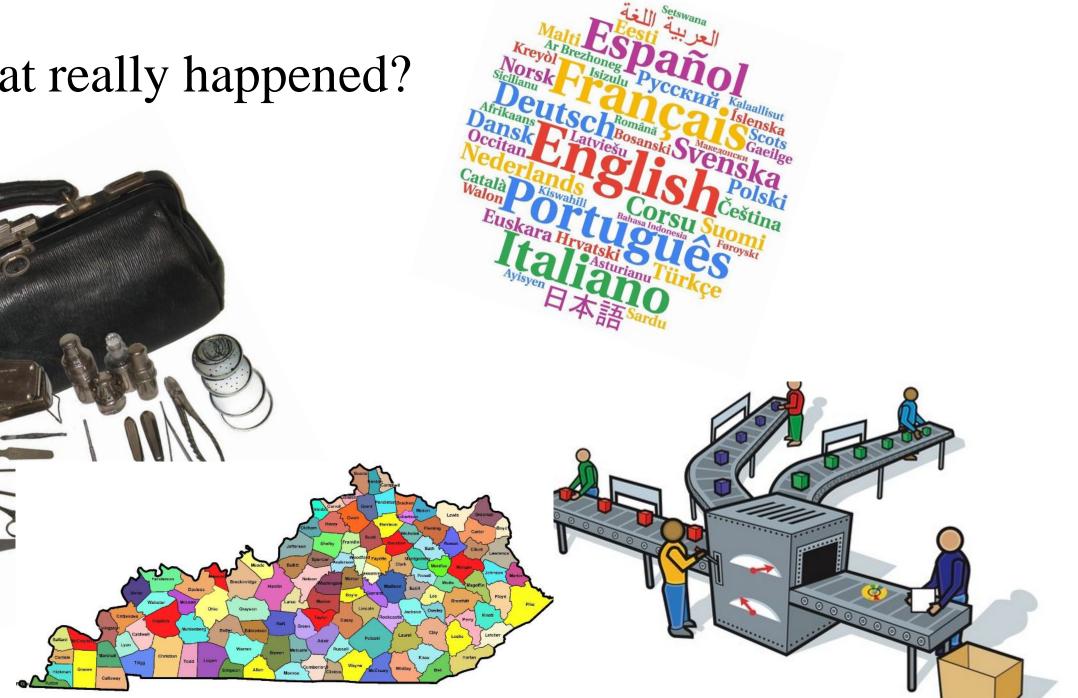
A mean horse and a gentle car...

"If there are faults with cars, only time is wanted to make them disappear ... there is no mechanism more inoffensive, no means of transportation more sure and safe" -- Scientific American, 1900

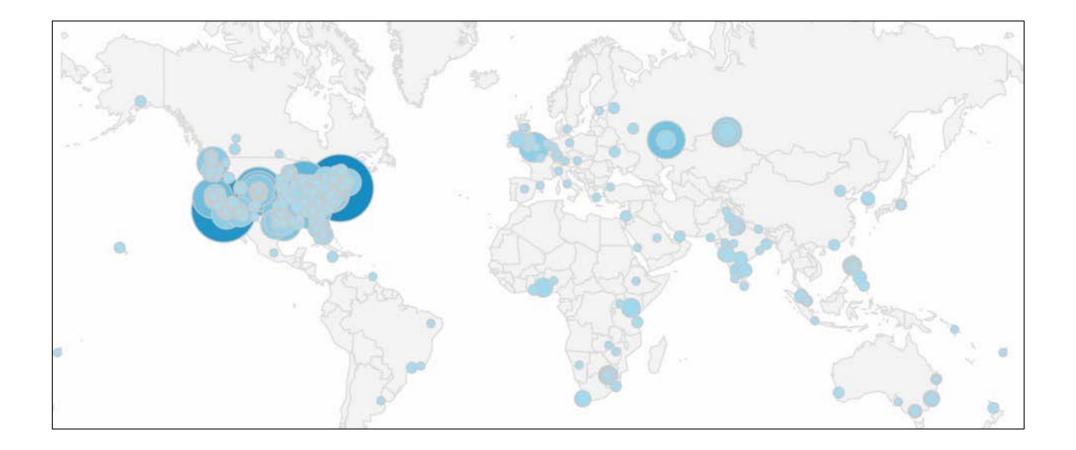


What really happened?

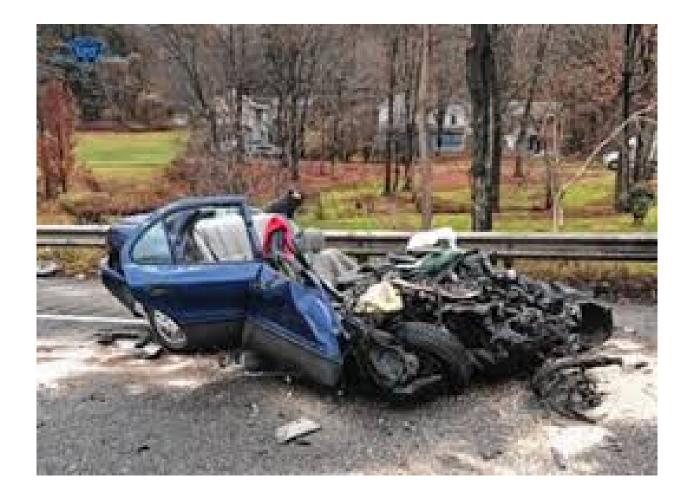
Collectible



What really happened?

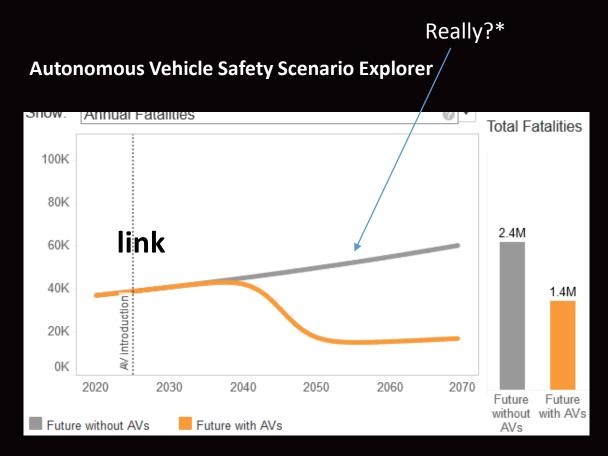


What really happened?



Predicting Safety for CAVs

Back to The Rand Study



https://www.rand.org/pubs/tools/TL279/tool.html

*even with no full AV, the safety curve would probably bend down (blind spot warnings, stability control, automatic braking, etc.)



How bout some more dials?



What will cause crash reduction/increase, how, when and how much?

- Technology A reduces crash type 3 by x%
- Works best on facility type y (representing how many crashes?)
- Market penetration ... How much of this technology do we have:
 - Now?
 - In 5 years?
 - Ultimately?



$F=F_{0} \times G_{VMT} \times (1-MP_{AV}) \times (1-CRF_{AV}) \times MP_{CAV \text{ tech}} \times (1-CRF_{CAV1}) \times (1-CRF_{CAV2}) \times ...$ Warning! Do not use

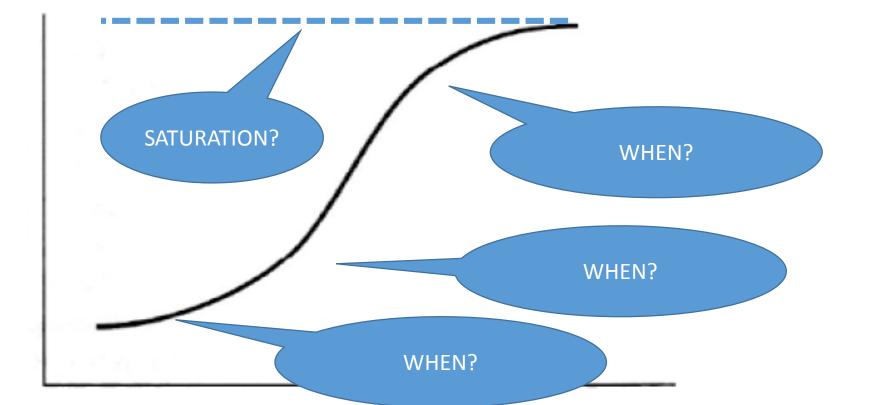
• Market Penetration = f (cost, regulation, ...)

Warning! Do not use this equation for anything

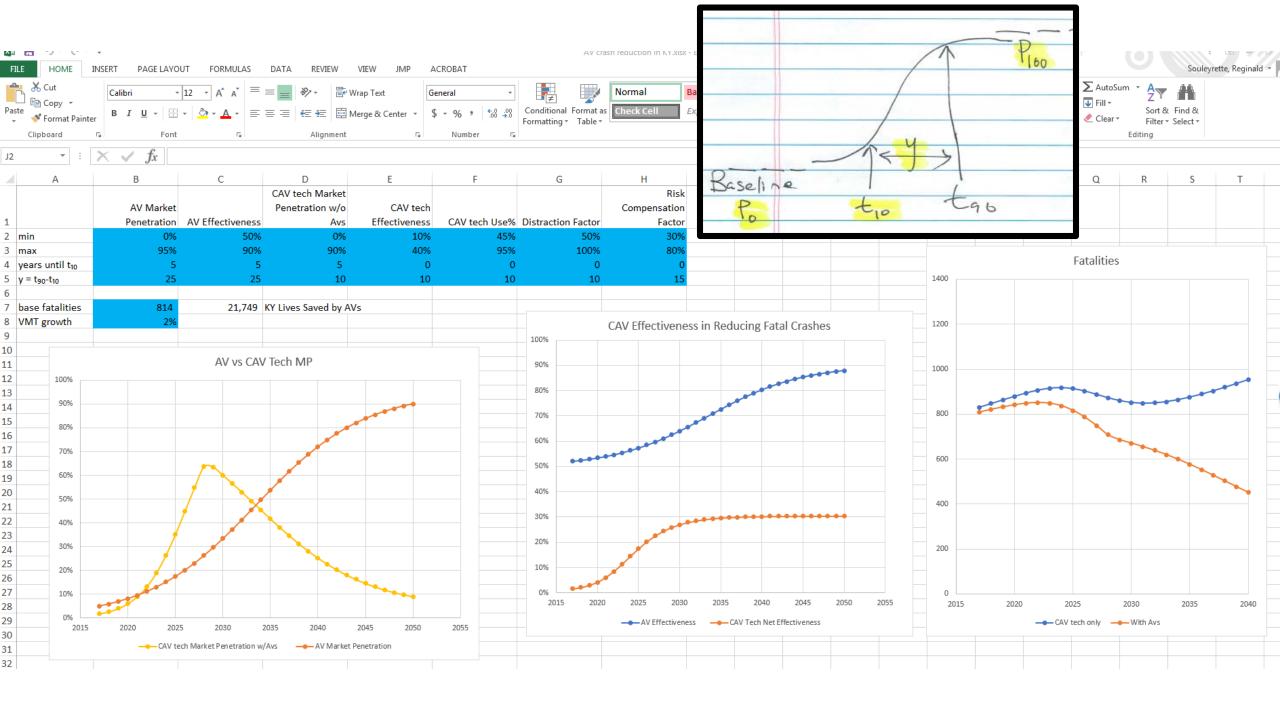
- Effectiveness = f(technology, road system, crash type, less fatigue, ...)
- Proper Use = f(behavioral, quality, effectiveness, ease...)
- Risk Compensation = f(behavioral, other)
- Distraction = f(behavioral, familiarity, user interface quality/maturity, ...)
- Subject to some limits, interactions, ...

- S curves can represent
- Market Penetration
- Effectiveness ration
- Various monotonic changes over time





More Dials



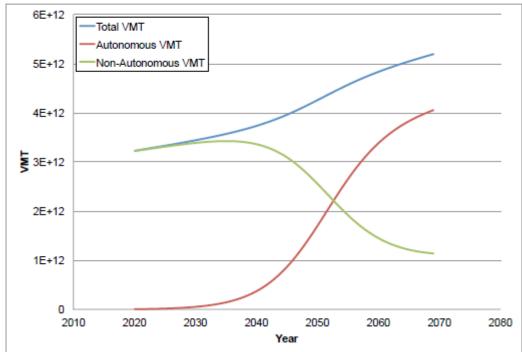
What's next?

Directional analysis	Toda	y	CAV effi	icacy	Total Collisions	Crash Data %	· · ·	/ Injury	%	w/ Fatality	%	w/ Property Damage	Total fatalities	Total Injuries	
Collision with Pedestrian/Bike/Animal/Fixed Object	15%	i	1009		20269	12.58		3720	2.3103%	204	0.1267%	16345	210	4596	
Left Turn Collision			1009		4739	2.943		1325	0.8229%	19	0.0118%	3395	21	2274	
Right Turn Collision			100%	%	1096	0.680	07%	115	0.0714%	0	0.0000%	981	0	165	
Both Vehicles Moving Straight			1009	%	17054	10.59	14%	2861	1.7768%	26	0.0161%	14167	32	4564	
Rear End Collision			1009	%	21845	13.56	68%	3174	1.9712%	25	0.0155%	18646	32	4889	
Sideswipe	1000		S-curve	Graphing	10105	- 100				~ ~ ~	0.01000	11000	7	1700	
Head-on Collision	2020	2025	2030	2035	2040	2050	Penetration	n		Estima	ted Collisior	n w/ Bike/Pede	strian/Animal/	Fixed Object Re	duction by
Vehicle Backing							Factor						CAV's		
Ran Off Roadway	0.02	0.05	0.76	0.94	0.97	0.98		6	0.00	2023	20.28	20.33	2038	20.43	2048
Vehicle Moving in Wrong Direction	0.00	0.00	0.00	0.00	0.00	0.00			0.00	-					
Crossover Collision w/ Median	0.00	0.00	0.00	0.00	0.00	0.00			0.10						
Overturning	0.00	0.00	0.00	0.00	0.00	0.00			0.20						
	0.00	0.00	0.00	0.00	0.00	0.00									
Avoidance	0.00	0.00	0.00	0.00	0.00	0.00			0.30						
Steering	0.11	0.18	0.18	0.74	0.89	0.93		5	0.40						
Braking	0.00	0.00	0.00	0.00	0.00	0.00		~							
Steering and Braking	0.05	0.11	0.40	0.96	0.97	0.98		6	0.50						
No avoidance	0.00	0.00	0.00	0.00	0.00	0.00			0.60						
	0.00	0.00	0.00	0.00	0.00	0.00						\ Fig	gure 4.2	2. Scena	rio of I
Environmental									0.70			$\mathbf{\nabla}$			
Glare									0.80			\rightarrow			
Construction Zones									0.90				6E+12		
Slick Surfaces													02412	Total	VMT
Shoulder Drop-offs/Slides									1.00						
														-Autor	nomous V
Air bags													5E+12	Non-A	Autonomo
Deployment													02.12		
Not installed			Run off	Road Colisio	n Reduction						Hea	id-on Co			
Switch on		2018	2023	20.2B	2033	20.38	3 20	48	20.48	2053 2018	20.2				
Switch off		0.00								0.00			4E+12	,	
No switch		0.10								0.10			46.12		
Driver age		0.20								0.20		-			
Driver age		0.30								0.30			5 3E+12	, 📖	
16-24										6.87			5 2011	·	
-															
16-24		0.40								0.40					
16-24 25-40 41-64															
16-24 25-40		0.40								0.40			25+12	,	
16-24 25-40 41-64 65+													2E+12	2	
16-24 25-40 41-64		0.50								0.50			2E+12	2	
16-24 25-40 41-64 65+		0.50								0.50			2E+12	2	
16-24 25-40 41-64 65+		0.50								0.50					
16-24 25-40 41-64 65+		0.50								0.50			2E+12 1E+12		
16-24 25-40 41-64 65+		0.50								0.50					

Learn more from others

igure 4.2. Scenario of Highly Automated and Non–Highly Automated VMT

20.58



Analyze *these* technologies:

• L0

- Blind Spot Monitoring
- Lane Departure Warning
- Traffic Sign Recognition
- Left-Turn Assist
- Adaptive Headlights
- L1
 - Adaptive Cruise Control
 - Cooperative Adaptive Cruise Control
 - Automatic Emergency Braking
 - Lane Keeping (Page 10, good for striping)
 - Electronic Stability Control
 - Parental Control

- L2
 - Traffic Jam Assist
 - High Speed Automation
 - Automated Assistance in Roadwork and Congestion
- L3
 - On-Highway Platooning
 - Automated Operation for Military Applications
- L4
 - Google's Driverless Car (Not tested in bad weather)
 - Kill Switch
- L5
 - Fully Autonomous

2006 – 2010 Collision History

And these crash types:

- Directional Analysis
 - Collision w/ peds/bikes/fixed objects
 - LT Collision
 - RT collision
 - Rear End
 - Sideswipe
 - Head-on
 - Crossover Collision w/ median
 - Ran-off roadway
 - Overturning
 - Wrong Direction
- Driver Distraction
 - Cell-phone
 - Other inside Vehicle
 - Outside Vehicle
- Environmental
 - Glare
 - Construction Zones
 - Slick Surfaces
 - Drop-offs & Slides

- Directional Analysis
 - Collision w/ peds/bikes/fixed objects
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- Environmental
 - Glare
 - Construction Zones
 - Slick Surfaces
 - Drop-offs & Slides

	2006 -	- 2010 Collision History					
2006 – 2010 Co	ollision History	Fatal and Serious Crashes – Thurston County Percentage	Fata C				
Overall Numbers	Total number of collisions	3					
By Collision Type	Hit fixed object	48					
By Collision Type	Overturn	10					
By Collision Type	Angle (left turn)	9					
By Collision Type	Head on	7					
By Light Condition	Daylight	52					
By Light Condition	Dark – no street lights	33					
By Junction Relationship	Intersection-related	19					
By Junction Relationship	Driveway-related	5					
By Junction Relationship	Non-Intersection	77					
Hit Fixed Object Crashes Only – By Fixed Object Hit	Tree/stump (stationary)	14					
Hit Fixed Object Crashes Only – By Fixed Object Hit	Roadway ditch	7					
Hit Fixed Object Crashes Only – By Fixed Object Hit	Utility pole	7					
By Roadway Curvature	Straight and level	42					
By Roadway Curvature	Horizontal curve (all)	45					
By Speed Limit (Number of Drivers)	35 mph	28					
By Speed Limit (Number of Drivers)	50 mph	69					
By Contributing Circumstance Number of Drivers)	Exceeding safe/stated speed	48					
By Contributing Circumstance Number of Drivers)	Under influence of alcohol/drugs	42					
By Contributing Circumstance Number of Drivers)	Over centerline	16					
By Contributing Circumstance Number of Drivers)	Inattention/distraction	11					
By Driver Age Group	Ages 16-20	26					
By Driver Age Group	Ages 41-50	28					
By Seat Belt/Car Seat Use (Number of Occupants)	No restraint	33					

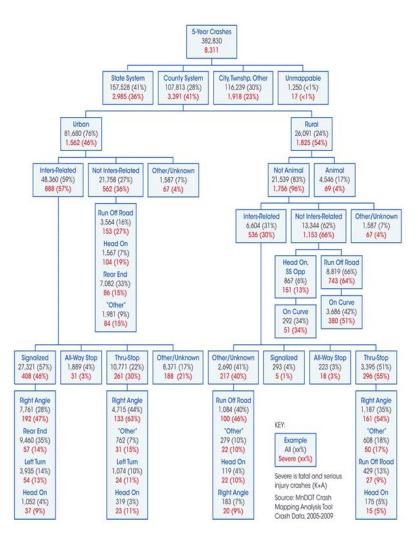
Crash type mitigation levels (effectiveness)

- Many crash types (e.g., ROR, speeding with parental control;) can be effectively mitigated by level 2 or 3 CAV technology
- Some crash types (e.g., Speeding) might be mitigated somewhat by level 2 or 3 (e.g., with speed limit sign recognition and reminders, IF the driver does not wish to intentionally speed)
- Others (e.g., drunk driving) can only be effectively mitigated by truly self driving cars?

In the context of *these* facility characteristics:

- Access Control
 - Full Access
 - Partial Access
 - Permitted Access
- Auxiliary Lanes
 - Truck Climbing
 - Parking
 - Turning
 - Merging
 - Cycling
- Bike/Peds
 - Cycling lane
 - Sidewalk
 - Crosswalk
 - Multi-use path
 - Shared lanes

- Interchanges
 - Diamond
 - Double diamond crossover
 - Partial
 - Trumpet
 - Cloverleaf
 - Displaced LT
 - Pavement
 - Unimproved/primitive
 - Graded & Drained
 - Soil/Gravel/Stone
 - Highly Flexible
 - Concrete
 - Composite
- Operation Type
 - One-way
 - Two-way



Research typology

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A	В		D	E	F	G	Н	I	J	к	L	М	N	0
					Level 0					Connect Lev	ed and Autonom el 1	nous Vehicle Fe	atures	·
			Blind Spot Monitoring	Lane Departure Warning	Traffic Sign Recognition		Adaptive Headlights	Adaptive Cruis Control	e Cooperative Adaptive Cruise Control	Automatic Emergency Braking	Lane Keeping	Electronic Stability Contr	Parental Contro	Traffic J Assis
		Deployment		_					Control	braking				
	Airbags	Not Installed												-
		Switch (on/off)												+
		Steering Braking		+		-		1				-	-	+
	Avoidance	Steering & Braking						1				+		+
		No avoidance		1		1		1				1	1	+
		Collision with Pedestrian/Bike/Animal/Fixed												
		Object Left Turn Collision							-			-		+
		Right Turn Collision												
		Both Vehicles Moving Straight												-
		Rear End Collision												
	Directional Analysis	Sideswipe												
		Head-on Collision												
Oracle Date		Vehicle Backing Ran Off Roadway							-			-		
Crash Data		Vehicle Moving in Wrong				-			-					+
		Direction												_
		Crossover Collision w/ Median Overturning on Ramps							_					
		Cell Phone												+
	Driver Distraction	Inside Vehicle												+
		Outside Vehicle												
	Ejection From Vehicle													
		Glare (sun)				_								_
	Environmental Factors	Construction Zones							_					
	ŀ	Slick Surfaces Shoulder Drop-offs/Slides										+	+	+
		Vehicle Damage						1				1		+
		Driver Injury												
	Severity	Passenger Injury												
		Environmental Damage Pedestrian Injury										+		+
		Full										+		+
	Access Control	Partial					1							+
		Permit										1		+
		Truck Climbing												
	[Parking												+
	Auxiliary Lanes	Turning										+		+
		Merging Bike												+
		Bike Lane												+
		Sidewalk						1				1	1	+
	Bike/Ped Facilities	Crosswalk												
		Multi-use Path												
		Shared Bike lane												
		Diamond												
		Partial												+
	Interchanges	Trumpet						+						

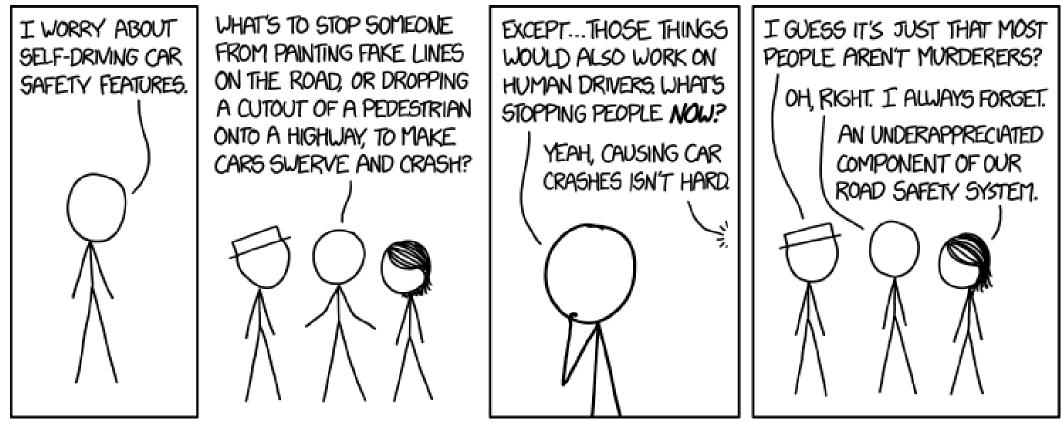
Specific component s-curves

		Level 0				
		Blind Spot Monitoring	Lane Departure Warning	Traffic Sign Recognition	Left-Turn As	
	Deployment					
Airbags	Not Installed					
	Switch (on/off)					
	Steering					
Avoidance	Braking					
Avoluance	Steering & Braking					
	No avoidance					
	Collision with Pedestrian/Bike/Animal/Fixed Object					
	Left Turn Collision					
	Right Turn Collision					

Questions/suggestions?

• Thank you, and Go Cats! 🙂





https://xkcd.com/1958/