

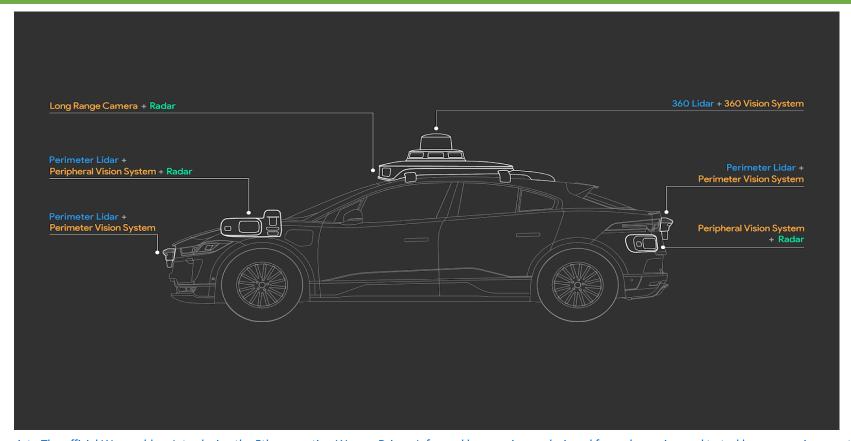
### Storyline

- Intelligent Car Model
- Auto Safety Standard
  - Safety Targets vs. Accident Metrics
- Testability
  - DL Accuracy vs. Safety
  - Systematic Faults & Validation
  - Transient & Permanent Faults
- Diverse Redundancy
  - Reliability Models
  - Need for Diversity

     Systematic Faults

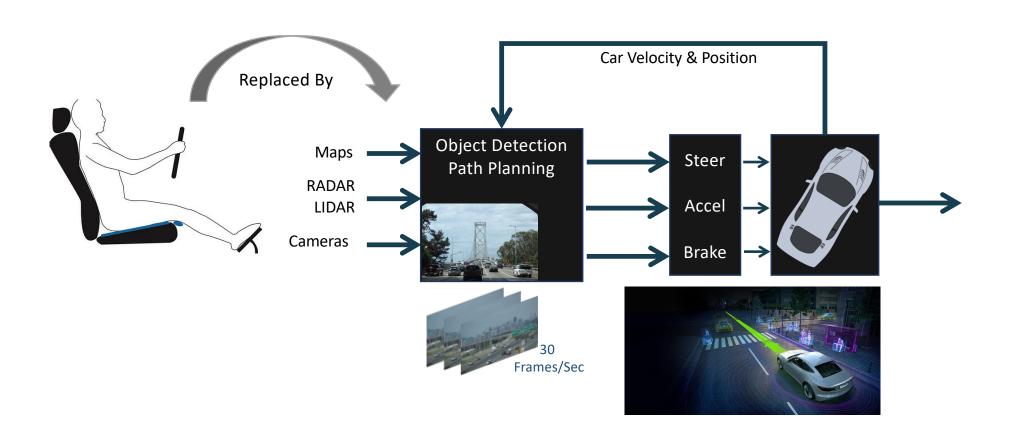
N. Saxena

# Cameras & Sensors in an Intelligent Vehicle

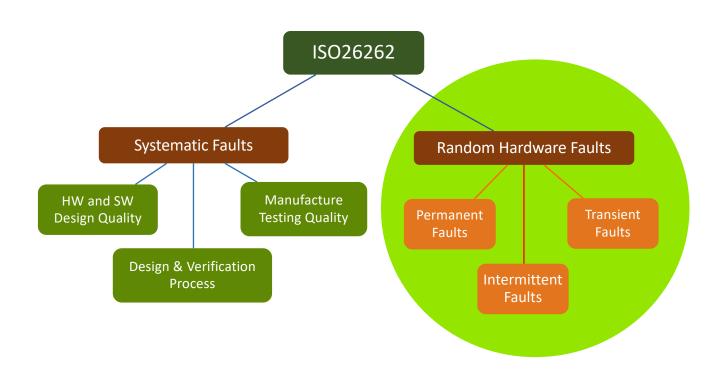


Source: Waypoint - The official Waymo blog: Introducing the 5th-generation Waymo Driver: Informed by experience, designed for scale, engineered to tackle more environments

### Control System Model– Intelligent Car



# ISO26262 Auto Safety Specification

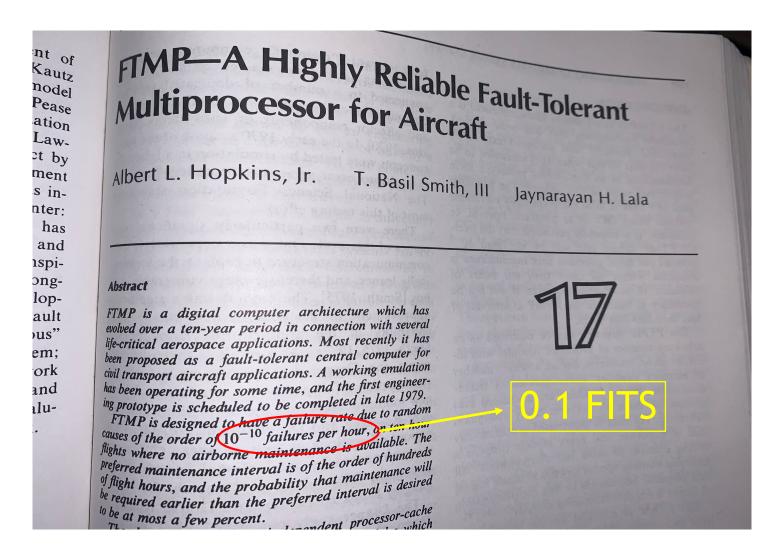


# Random Hardware Faults Targets

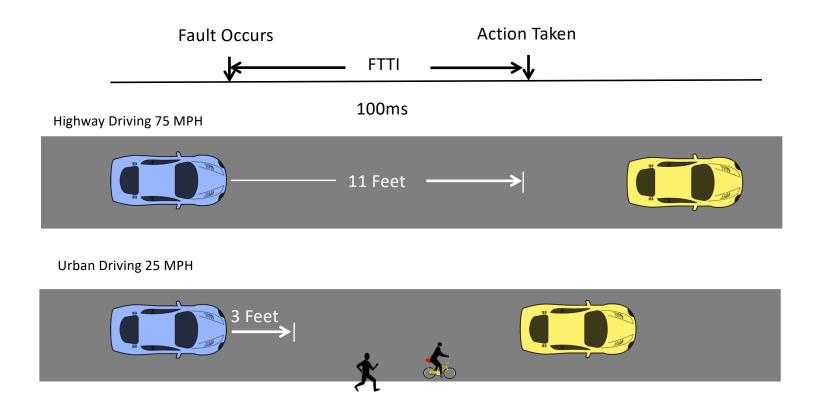
Hardware Random Fault Metrics	ASIL B	ASIL C	ASIL D
Permanent Fault Coverage (SPFM)	90%	97%	99%
Transient Fault Coverage (SPFM)	90%	97%	99%
Latent Fault Coverage (LFM)	60%	80%	90%
Hardware Failure Probability (PMHF)	$100 FIT$ $\leq 10^{-7}/hr$	$100 FIT$ $\leq 10^{-7}/hr$	$10 FIT \\ \leq 10^{-8}/hr$

### FIT = Failures in Time, Time = $10^9$ Hours. 1 FIT = $10^{-9}$ *failures/hour*

ASIL	Automotive Safety Integrity Level
SPFM	Single Point Fault Metric
LFM	Latent Fault Metric
PMHF	Probabilistic Metric for Hardware Failures



## Fault Tolerant Time Interval (FTTI)



### **Accident Statistics— US**

Reference: National Highway Traffic Safety Administration (NHTSA): www.nhtsa.gov

Description	2013 Statistics	2015 Statistics	
Fatal Crashes	30,057	35,092	
<b>Driver Related Fatal Crashes</b>	10,076	10,265	
Non-Fatal Crashes	5,657,000	6,263,834	
Number of Registered Vehicles	269,294,000	281,312,446	
Licensed Drivers	212,160,000	218,084,465	
Vehicle Miles Travelled	2,988,000,000,000	3,095,373,000,000	
Fatal Crash Rate in FITs	250 – 500	283 - 566	
Non-Fatal Crash Rate in FITs	46K <b>-</b> 92K	51K – 102K	
ASIL D 10 FITs is ~ 50x Improvement over Fatal Crash Rate & 4 Orders of Improvement in Non-Fatal CR FITs			

Economic Cost of Traffic Crashes (2010) \$242 Billion

**Published AV Non-Fatal Crash FIT Rate = 150K** 

### Object Detection & Path Planning—Contextual Accuracy



# Object Detection, Path Planning & Other Al Functions Need Enormous Computational Power



https://www.anandtech.com/show/11913/nvidia-announces-drive-px-pegasus-at-gtc-europe-2017-feat-nextgen-gpus

# Compute Workload : Perception

Perception Challenge: Achieve "perfect" Object Detection Accuracy

Deep Learning = State of the Art Method

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### Detection Accuracy & Systematic Faults (SW Bugs)

- When does Detection Accuracy Matter?
  - Traffic Light Detection: Red, Green & Orange (100%)
  - Objects in and around Path Plan (100%)
  - Distant Objects Not in Path Plan (0%)
- Validation of SW & Drive System Software Stack
  - Augmented Virtual Reality
  - Evaluate Millions of Scenarios
  - Simulate Millions-of-Miles-Traveled in a Day
    - Use Massively Parallel Super Computers
  - Dangerous Scenarios with No Physical Harm
  - Compute for Safety



Nvidia DRIVE Constellation in Datacenters

# Transient Fault Injection

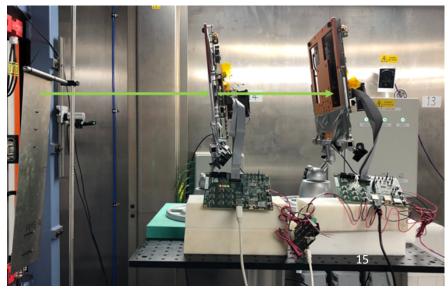
### **Accelerated Neutron Beam Testing**

- Radiation experiments beam testing campaigns
  - Weapons Neutrons Research @ LANSCE
  - ChipIR microelectronics @ Rutherford Appleton Laboratory
- 2000 years of exposure to terrestrial neutron flux

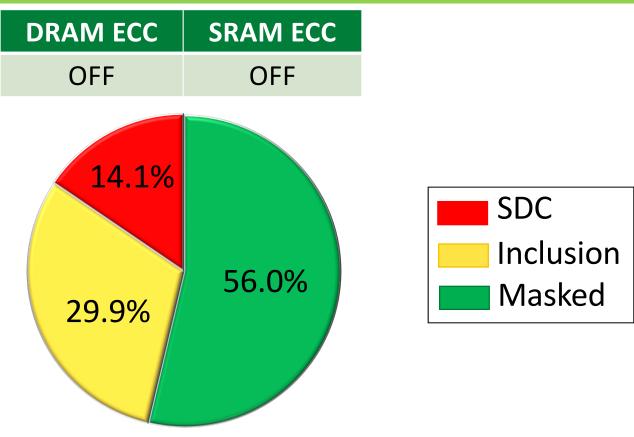
Flight path of neutron beam

Experiment Design

DRAM ECC	SRAM ECC	
OFF	OFF	
ON	OFF	
ON	ON	

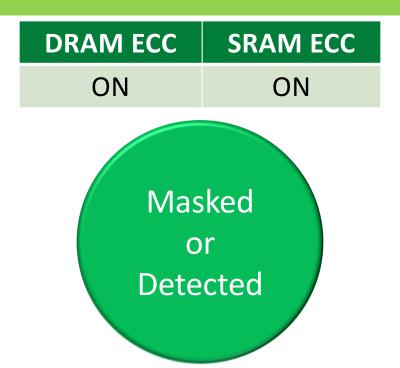


### **Accelerated Beam Testing Results**



SDC: Silent Data Corruption

### **Accelerated Beam Testing Results**



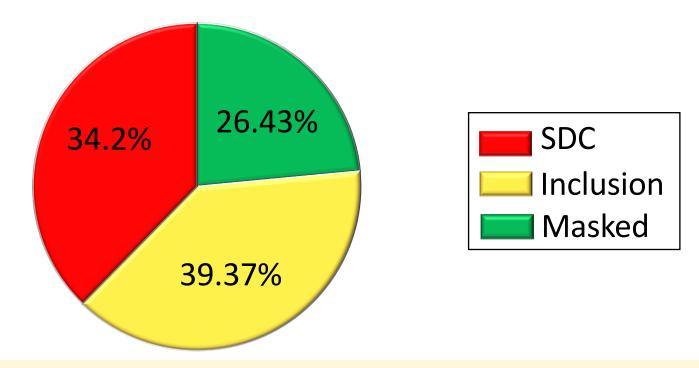
**Zero SDC Events** 

# Permanent Fault Injection

### **Permanent Fault Injection Results**

Faults in input batches: SDC (+ inclusion) < 1.8%</li>

Faults in weights:



Object detection networks are vulnerable to permanent faults

### **Object Detection Conclusion**

- Without protection—object detection networks show high SDC rate
  - Unlike classification networks that show resilience to transient errors
- Zero SDC with chip-level protections
  - For transient faults
- Not all permanent fault are detected by ECC/Parity:
  - Raw permanent FIT rate (hundreds) vs raw transient FIT rate (tens of thousands)
    - Offline structural tests during key-off and key-on events,
    - Online periodic tests (meeting FTTI requirement)

# Road to Resiliency

### Markov Chain Analysis – Need Physical Redundancy

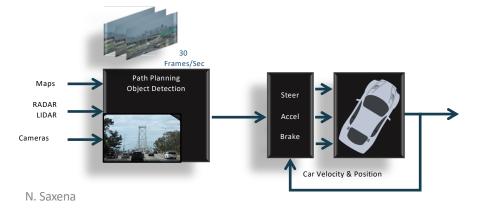
#### Availability is Important Here

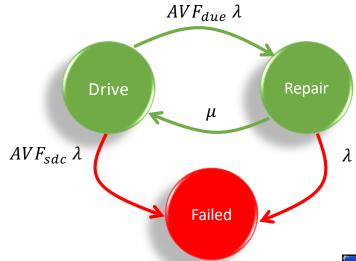
For Driverless Car

### Loss of Frames => Loss of Life

For 3 Frame-Tolerance, Need

$$\frac{1}{\mu}$$
 < 100ms





PROBABILITY
& STATISTICS
WITH
RELIABILITY
QUEUING,
AND
COMPUTER
SCIENCE
APPLICATIONS

KISHOR'S TRIVEDI

22

### Dual Redundant System

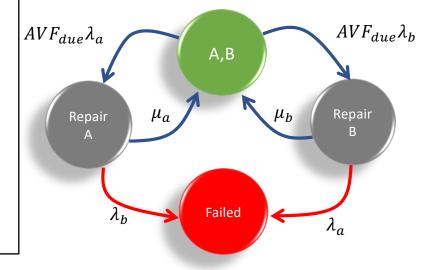
Relaxed Constraints on Repair Rate

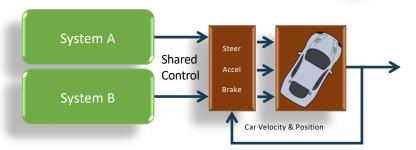
$$\frac{1}{\mu_a} < \frac{1}{\lambda_b}$$

$$\frac{1}{\mu_b} < \frac{1}{\lambda_a}$$

 $\frac{1}{\lambda_a}$  or  $\frac{1}{\lambda_b}$  in the order 1000's of hours

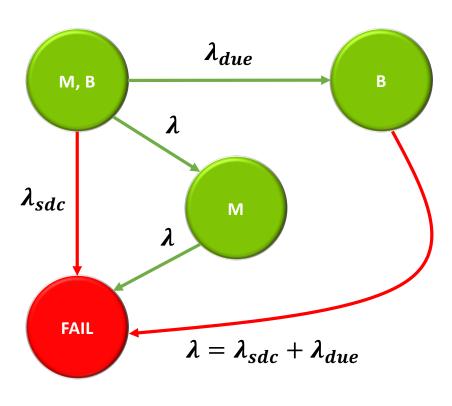
Repair can wait till the next Key-Off Event

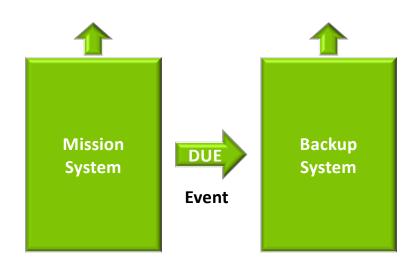




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## Backup Standby Model – Markov Chain





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### Probability of Backup Markov Chain States

Probability of being in M, B state,  $P_{m,b}(t) = e^{-2\lambda t}$ 

Probability of being in B state, 
$$P_b(t) = \frac{\lambda_{due}}{\lambda} (e^{-\lambda t} - e^{-2\lambda t})$$

Probability of being in M state,  $P_m(t) = e^{-\lambda t} - e^{-2\lambda t}$ 

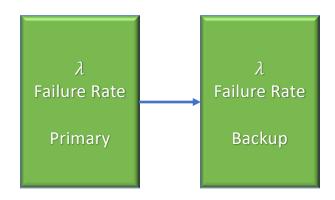
Probability of being in Fail State, 
$$F(t) = 1 - \left(\frac{\lambda + \lambda_{due}}{\lambda}\right)e^{-\lambda t} + \frac{\lambda_{due}}{\lambda}e^{-2\lambda t}$$

$$MTTF = \int_0^\infty t \frac{dF(t)}{dt} dt = \frac{1}{\lambda} + \frac{\lambda_{due}}{2\lambda^2} \ asymtotically \ approaches \ \frac{3}{2\lambda} \ (when \ \lambda_{sdc} = 0)$$

1.5x Gain in MTTF over Simplex or 1.5x Reduction in Effective Failure Rate over an infinite drive time N. Saxena

### Is MTTF Sufficient to Distinguish Two Systems?

### **Duplex System**



$$Duplex MTTF = \frac{3}{2}\lambda$$

#### Simplex System



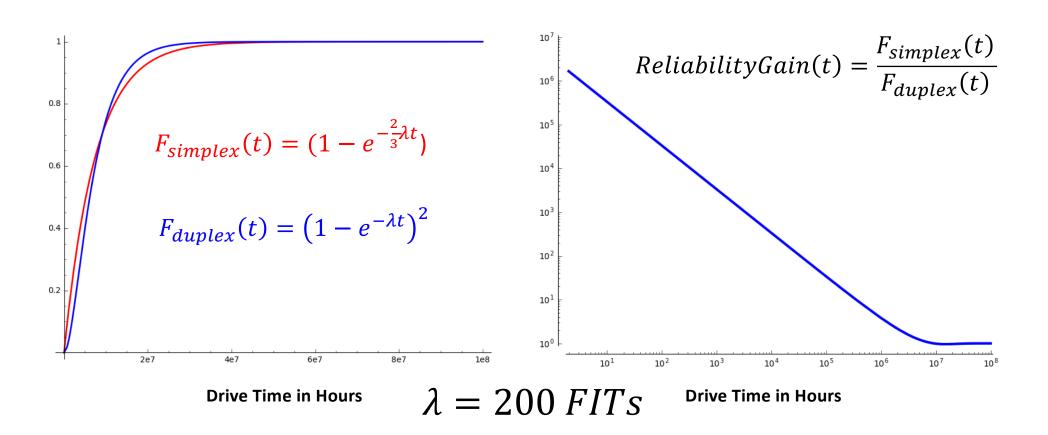
$$Simplex MTTF = \frac{3}{2}\lambda$$

Failure Probability Reduction metric as a function of mission time distinguishes various redundant systems [Mitra, Saxena, McCluskey 2004].

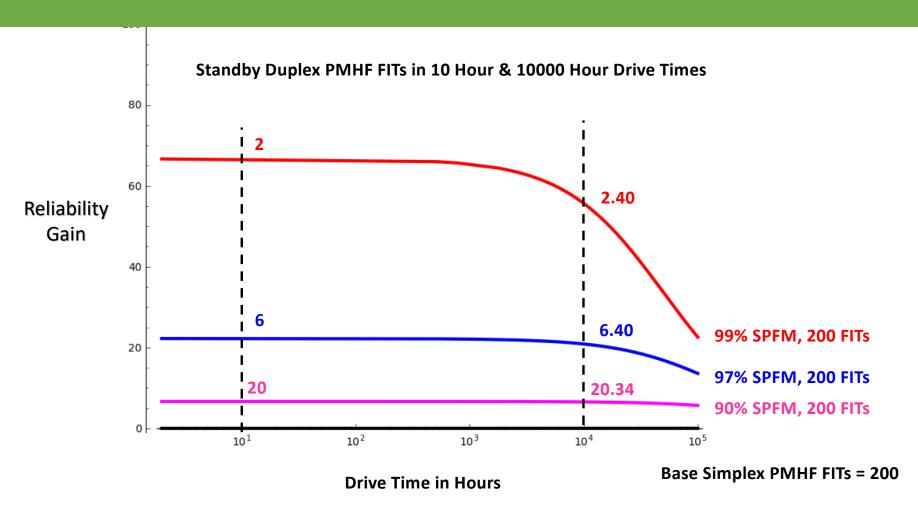
S. Mitra, N.R. Saxena, and E.J. McCluskey, "Efficient Design Diversity Estimation for Combinational Circuits," IEEE Trans. Comp., Vol. 53, Issue 11, pp. 1,483-1,492, Nov. 2004

S. Mitra, N.R. Saxena and E.J. McCluskey, "Common-Mode Failures in Redundant VLSI Systems: A Survey," *IEEE Trans. Reliability*, Special Issue on Fault-Tolerant VLSI Systems, Vol. 49, Issue 3, pp. 285-295, Sept. 2000.

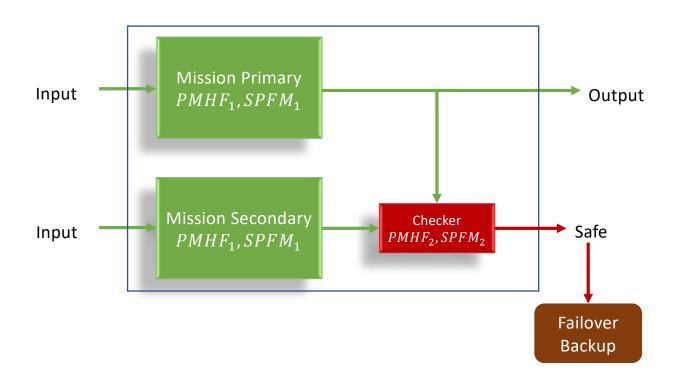
# Reliability Gain with Perfect Duplex $\times 10^6$ in 2 Hour Drive Time



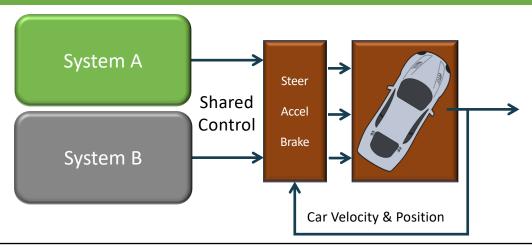
### Back-Up Standby Model—SPFM Sensitivity



# Duplex System with Decoupled Checker



### Design Diversity



#### Coping with Systematic Hardware and Software Design Errors

- [Siewiorek et. al. 1978] (byte reversal copies C.mmp processor)
- [Sedmak and Liebergot 1980] (complementary function diversity in VLSI)
- [Chen and Avizienis 1978] (N-version programming, SIFT software implemented fault-tolerance)
- [Horning et. al 1974] (Recovery Blocks) [Patel] RESO Technique
- [Amman and Knight 1987] (Data Diversity)
- [McCluskey, Saxena, Mitra 1998] Diversity for Reconfigurable Logic & Quantifying Diversity

### Conclusions

Road to Resiliency ⇒ Dual Redundancy or Graceful Degradation

- Mitigates Permanent Fault Testing
- Higher Availability During Mission Critical Time (Drive Time)

### **Systematic Faults**

- Rigorous Testing and Validation
   Need 3-to-4 Orders of Improvement
- Physical Redundancy with Design Diversity

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