

Otonom Araçlar

Savaş Takan

Otonom Araçlar

Five Levels of Vehicle Autonomy



Karsan'ın Bursa'da ürettiği sürücüsüz otonom otobüs

Çalışan profili:

- Boğaziçi, Odtu, Bilkent
- Lisans, yüksek lisans



Kullandıkları teknolojiler: NVIDIA, ROS, autoware, gezobo ...

Üretim Yeri: Bursa

Türkiye'nin Otomobili Girişim Grubu Sanayi ve Ticaret A.Ş.

3. seviye **otonom** sürüş

Üretim Yeri: Bursa



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Etik



İletişim



Psikolojik



Hukuk



Sosyolojik

Toplumsal Kabul

Otonom Araçlar

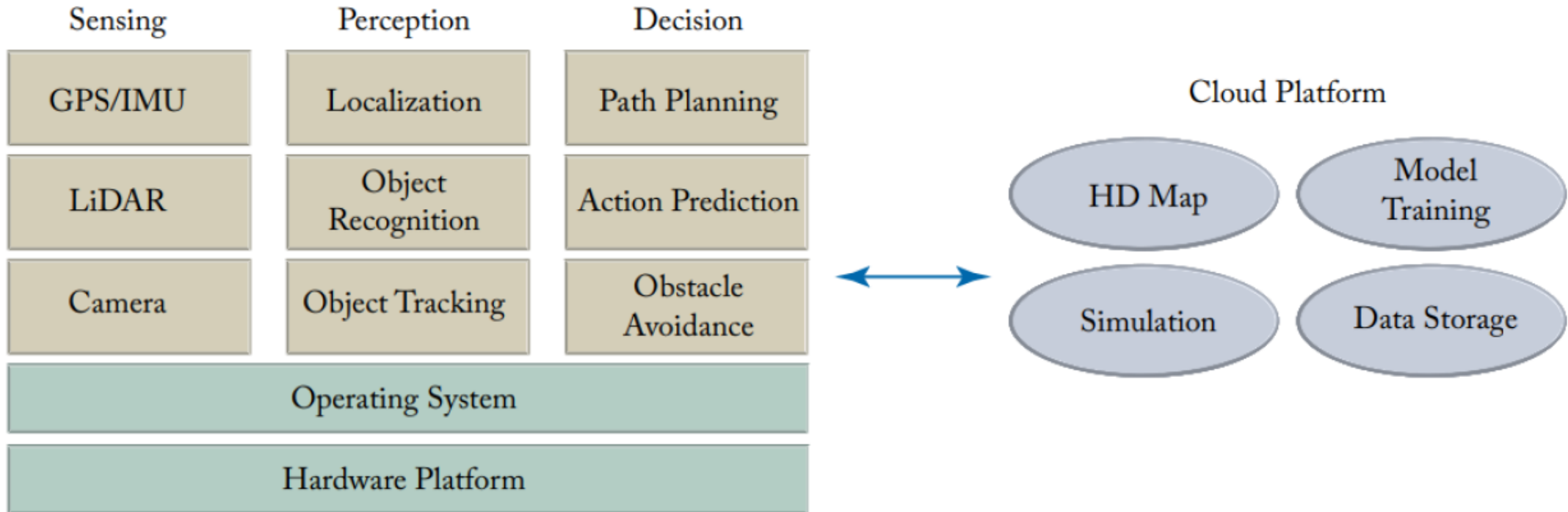
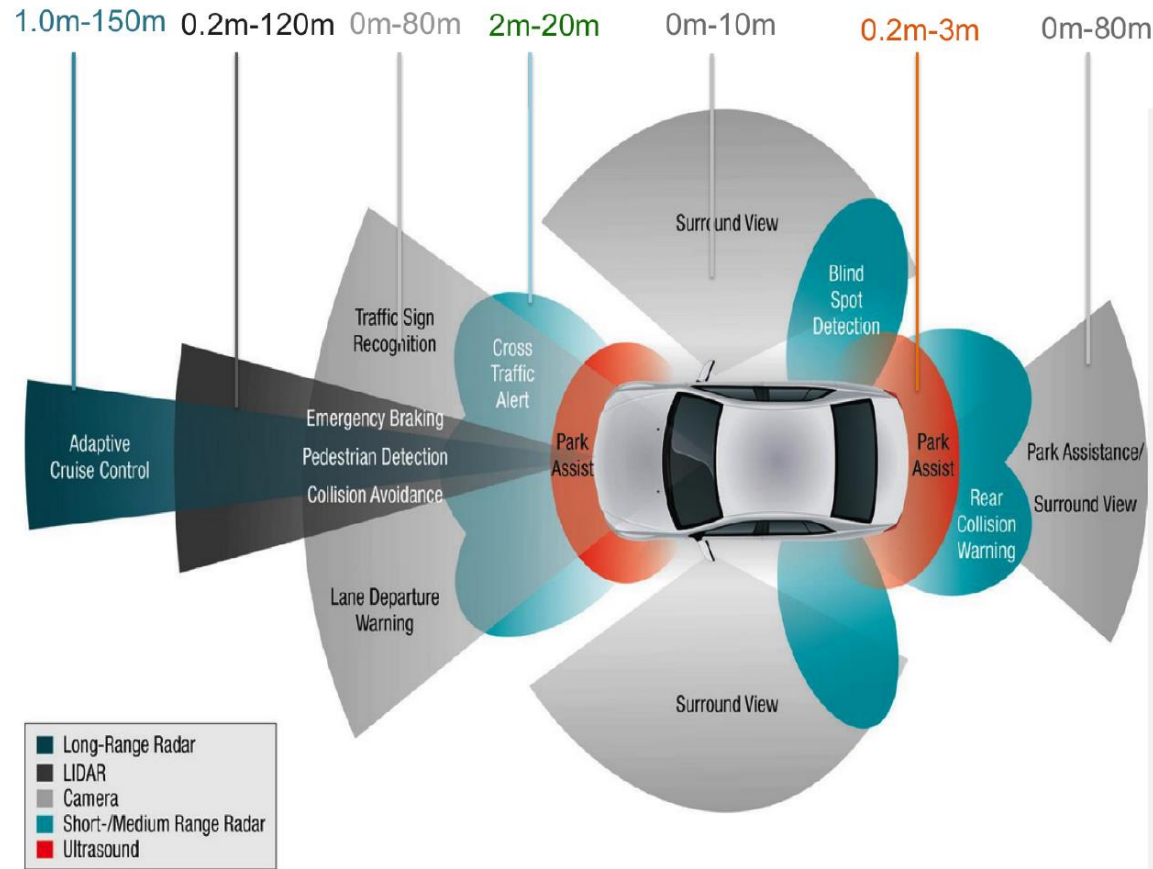


Figure 3.2: Autonomous driving system architecture overview from [140].

Otonom Araçlar



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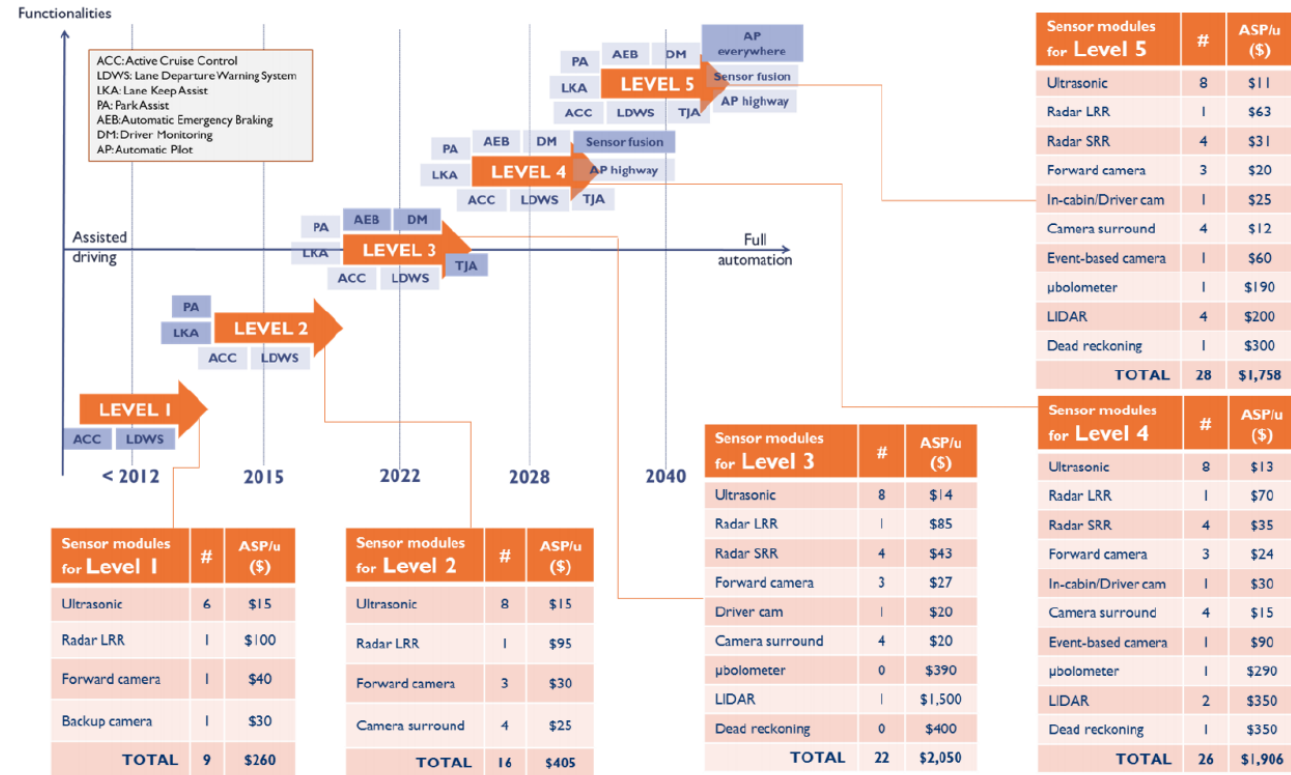


Figure 1.2: Sensor modules for each automation level from [251].

Otonom Araçlar

Performance aspect	Human	AV			CV	CAV
		<i>Radar</i>	<i>Lidar</i>	<i>Camera</i>	<i>DSRC</i>	<i>CV+AV</i>
Object detection	Good	Good	Good	Fair	n/a	Good
Object classification	Good	Poor	Fair	Good	n/a	Good
Distance estimation	Fair	Good	Good	Fair	Good	Good
Edge detection	Good	Poor	Good	Good	n/a	Good
Lane tracking	Good	Poor	Poor	Good	n/a	Good
Visibility range	Good	Good	Fair	Fair	Good	Good
Poor weather performance	Fair	Good	Fair	Poor	Good	Good
Dark or low illumination performance	Poor	Good	Good	Fair	n/a	Good
Ability to communicate with other traffic and infrastructure	Poor	n/a	n/a	n/a	Good	Good

Dedicated short-range communications (DSRC)

Otonom Araçlar

CAR AUTOMATION SENSORS & DATA VOLUMES

Sensor type	Quantity	Data generated
Radar	4–6	0.1–15 Mbit/s
LIDAR	1–5	20–100 Mbit/s
Camera	6–12	500–3,500 Mbit/s
Ultrasonic	8–16	<0.01 Mbit/s
Vehicle motion, GNSS, IMU	-	<0.1 Mbit/s

TOTAL ESTIMATED BANDWIDTH

3 Gbit/s (~1.4TB/h) to 40 Gbit/s (~19 TB/h)

Dataset

Table 3.2: Driving datasets from [254].

Dataset	Image	LIDAR	2D annotation	3D annotation	ego signals	Naturalistic	POV	Multi trip	all weathers	day & night
Cityscapes [45]	✓		✓				Vehicle			
Berkley DeepDrive [252]	✓		✓				Vehicle		✓	✓
Mapillary [166]	✓		✓				Vehicle		✓	✓
Oxford RobotCar [148]	✓	✓					Vehicle	✓	✓	✓
KITTI [67]	✓	✓	✓	✓			Vehicle			
H3D [177]	✓	✓		✓			Vehicle			
ApolloScape [95]	✓	✓	✓	✓			Vehicle			
nuScenes [28]	✓	✓	✓	✓			Vehicle		✓	✓
Udacity [219]	✓	✓	✓	✓			Vehicle			
DDD17 [21]	✓		✓		✓		Vehicle		✓	✓
Comma2k19 [199]	✓				✓		Vehicle			✓
LiVi-Set [39]	✓	✓			✓		Vehicle			
NU-drive [213]	✓				✓	Semi	Vehicle	✓		
SHRP2 [218]	✓				✓	✓	Vehicle			
100-Car [119]	✓				✓	✓	Vehicle		✓	✓
euroFOT [20]	✓				✓	✓	Vehicle			
TorontoCity [231]	✓	✓	✓	✓			Vehicle, Aerial, Panorama			
KAIST multi-spectral [41]	✓	✓	✓				Vehicle			✓

Autonomous Driving

- Stanley (Level 5 SAE ADS) – (Stanford University)
- DAVE2 – (NVIDIA)
- Waymo – (Google)

OS

- Real-time operating systems (RTOS)
- General-purpose operating systems (GPOS)



RTOS and GPOS Differences

RTOS

Real-Time Operating System

- Deterministic: no random execution pattern
- Predictable Response Times
- Time Bound
- Preemptive Kernel

Examples:

Contiki source code, FreeRTOS™, Zephyr™ Project

Use Case:

Embedded Computing

GPOS

General-Purpose Operating System

- Dynamic memory mapping
- Random Execution Pattern
- Response Times not Guaranteed

Examples:

Microsoft® Windows® operating system, Apple® macOS® operating system, Red Hat® Enterprise Linux® operating system

Use Case:

Desktop, Laptop, Tablet computers

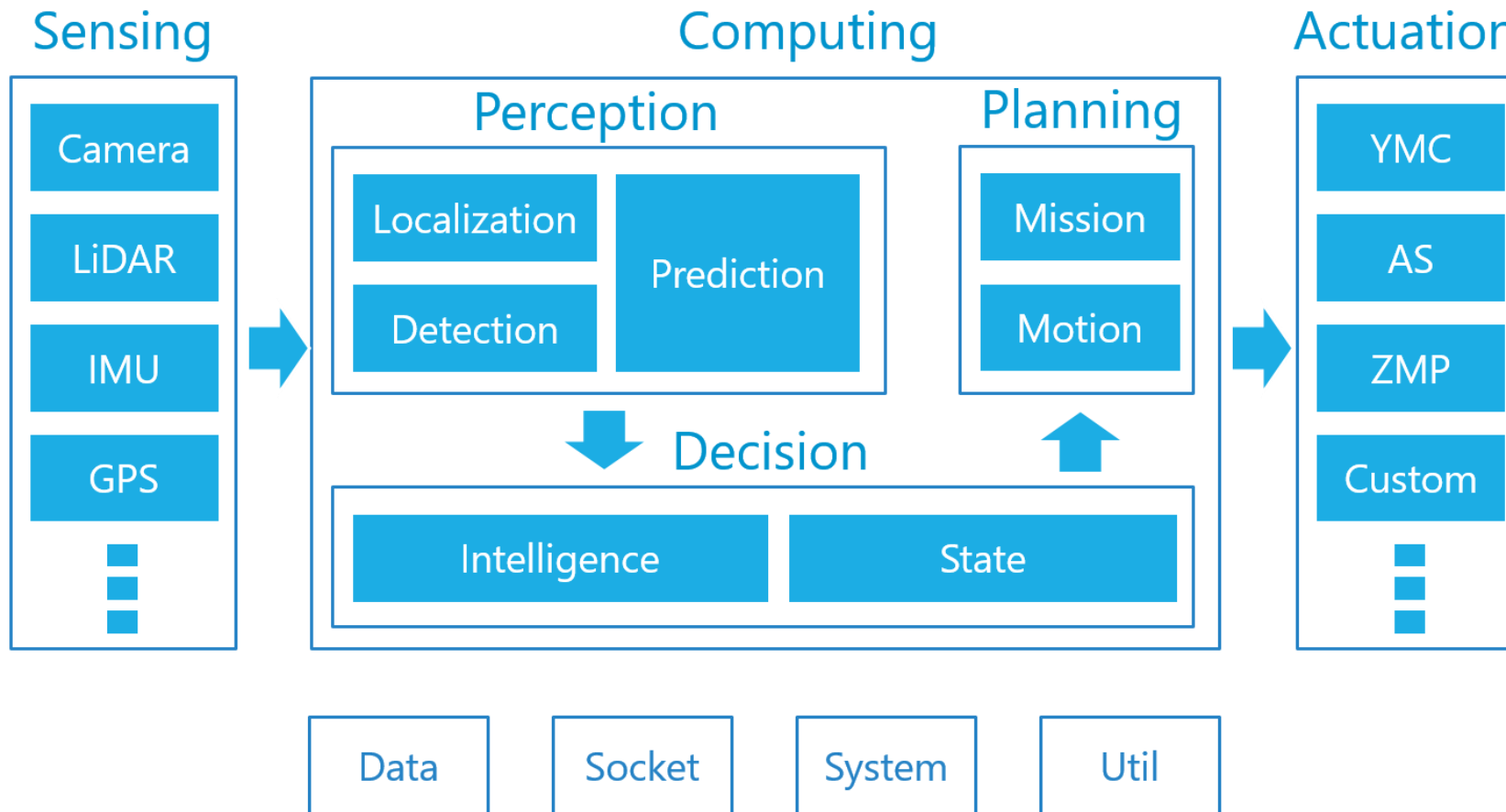
OS

- QNX Neutrino (BlackBerry)
 - Ford, Acura, VW, BMW, and Audi
- WindRiver VxWorks
 - Toshiba, Bosch, BMW, Ford, VW
- Green Hills INTEGRITY
- NVIDIA DRIVE™ OS
 - Tesla, VW, Mercedes-Benz, Audi, Veoneer, and Bosch
- Mentor Nucleus® OS

OS

- **Automotivelinux**
 - BMW, GM, VW, Toyota, Chevrolet, Honda, Mercedes, Tesla, Lyft, Baidu
- Android Automotive OS
 - Renault, Nissan, Mitsubishi
- Apple CarPlay
 - Audi, BMW, Toyota, Nissan, Ford, Honda, Mazda, and Mercedes-Benz.
- **ROS (Robotic Operating System)**
 - General Motors, BMW, Ford, Bosch, Karsan
- Microsoft

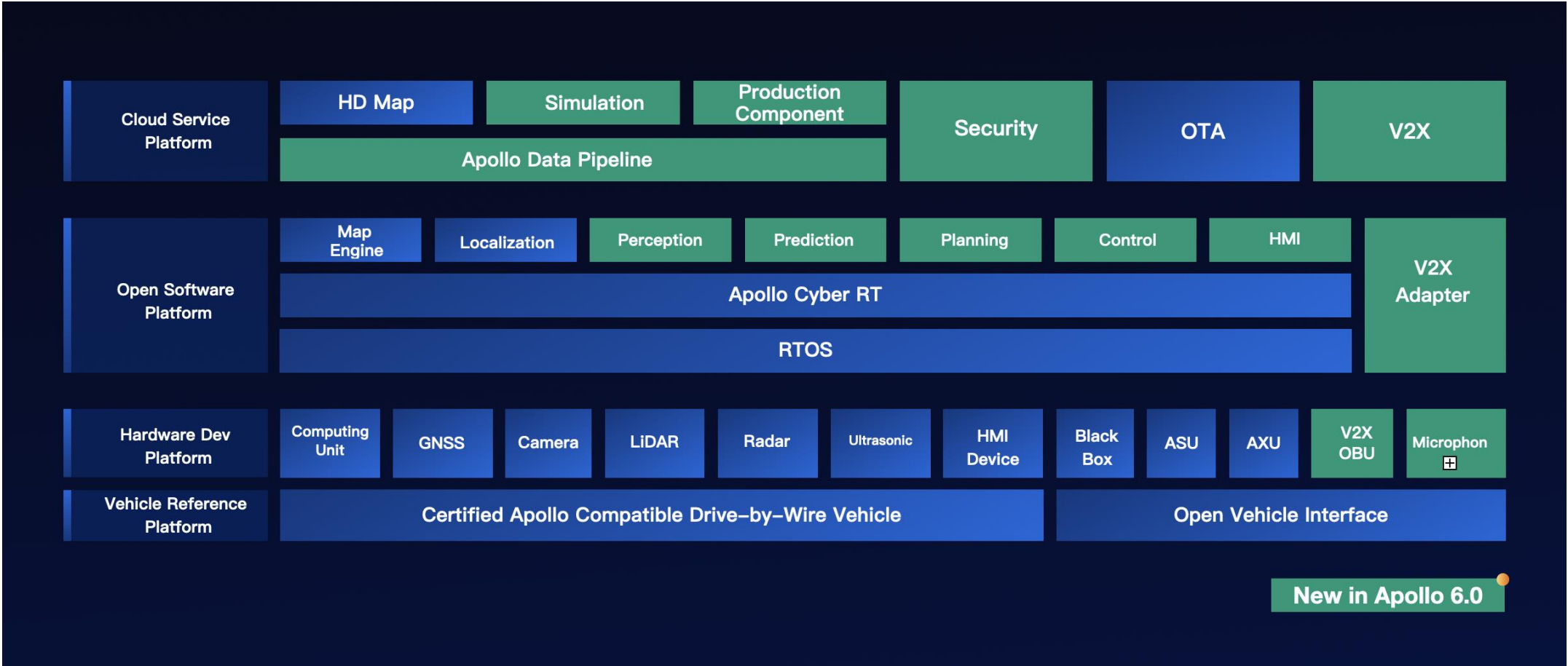
All in One: autoware.ai (ROS 1), autoware.auto (ROS 2)



All in One: Apollo (Baidu)

Hello Apollo	Apollo 1.0	Apollo 1.5	Apollo 2.0	Apollo 2.5	Apollo 3.0	Apollo 3.5	Apollo 5.0	Apollo 5.5	Apollo 6.0
Apollo Platform Announced	Closed Venue AD	Fixed Lane AD	AD on Simple Urban Road	Geo-fenced Highway AD	Production- level Closed Venue AD	City Urban Road AD	AD Empowering Production	Curb-to-Curb Urban Road AD	Towards Driverless Driving
2017.4	2017.7	2017.10	2018.1	2018.4	2018.7	2019.1	2019.7	2019.12	2020.9

All in One: Apollo (Baidu)



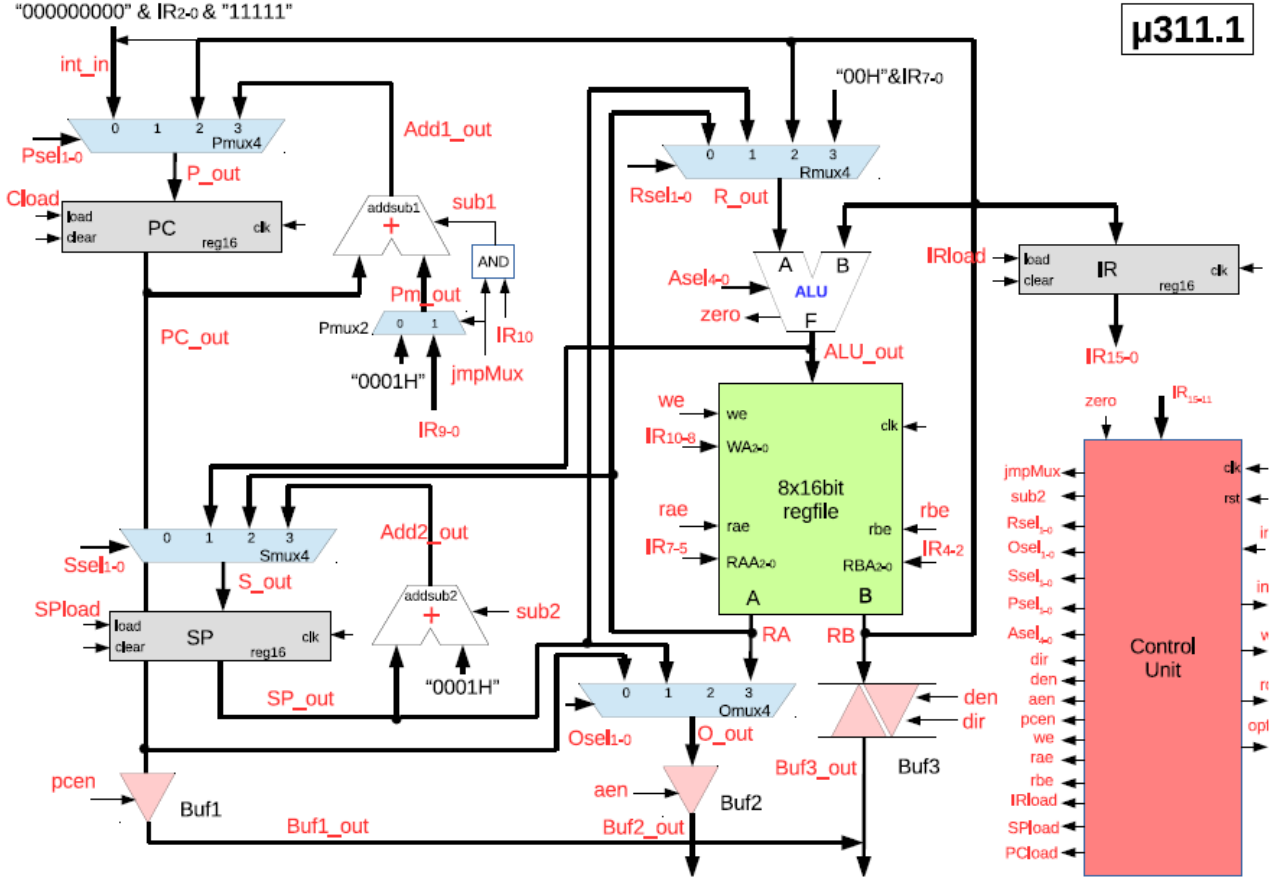
Simulatörler

	Licence	Documentation				Engine	Environment				Sensors					Output Training Labels		
		Installation	Environments	Sensor config	Output labels		For Driving		Actors		Cameras			Others		Semantic Segmentation	2D bbox	3D bbox
							Urban	Off-road	Humans	Cars	RGB	Depth	Thermal / IR	Lidar	Radar			
CARLA	MIT	+	+	+	+	UE	T	-	+	+	+	+	-	+	-	+	+	+
AirSim	MIT	+	+	+	+	UE	T C	F M G	-	+	+	+	+	+	-	+	.*	.*
Deepdrive	MIT	+	-	+	-	UE	R	-	-	+	+	+	-	-	-	.*	.*	+
LGSVL	Propr	+	+	+	+	Unity	C	-	+	+	+	+	-	+	+	+	+	+
Sim4CV	Propr	+	-	-	-	UE	T	D G	+	+	+	+	-	-	-	+	+	+
SynCity	Propr	NA ⁴⁵	NA	NA	NA	Unity	T C H	D F G U M	+	+	+	+	+	+	+	+	+	+
Unikie	Propr	NA	NA	NA	NA	NA	C	U H	+	+	+	-	-	+	-	NA	NA	NA ⁴⁶
rFpro	Propr	NA	NA	NA	NA	NA	T C R	-	+	+	+	+	NA	+	+	+	NA	NA
Cognata	Propr	NA	NA	NA	NA	Custo	C	-	+	+	+	NA	NA	+	+	+	NA	NA
SCANer Studio	Propr	NA	NA	NA	NA	NA	C R	-	NA	+	+	NA	NA	+	+	+	NA	NA
Highwai	Propr	NA	NA	NA	NA	Unity	C	-	+	+	+	+	+	+	NA	NA ⁴⁷	+	+
NVIDIA Drive	Propr	NA	NA	NA	NA	UE	C H	-	NA	NA	+	NA	NA	+	+	+	+	NA

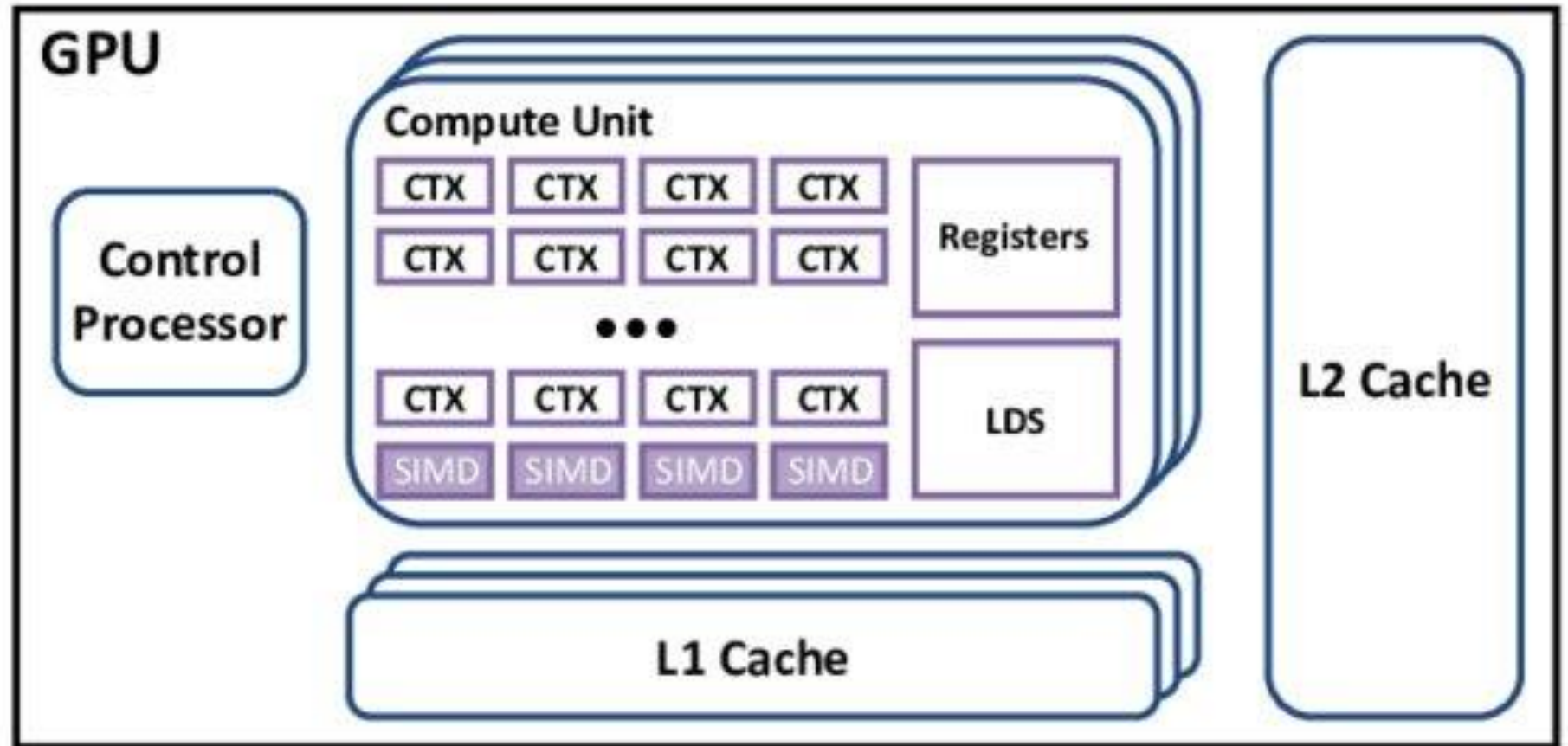
Donanim: heterogeneous computing

- «Mikrokontroller»
- CPU
- GPU
- FPGA

Donanım: CPU ve «Mikrokontroller»

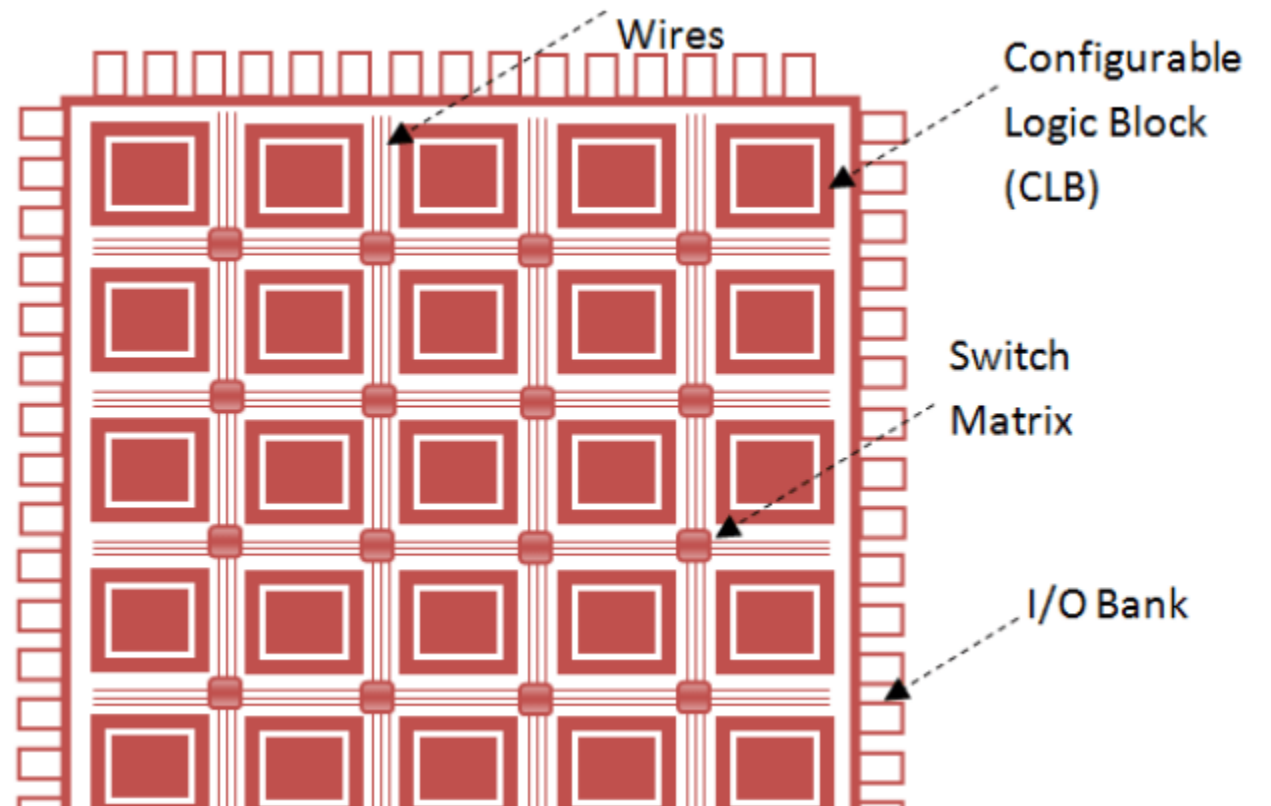


Donanim: GPU

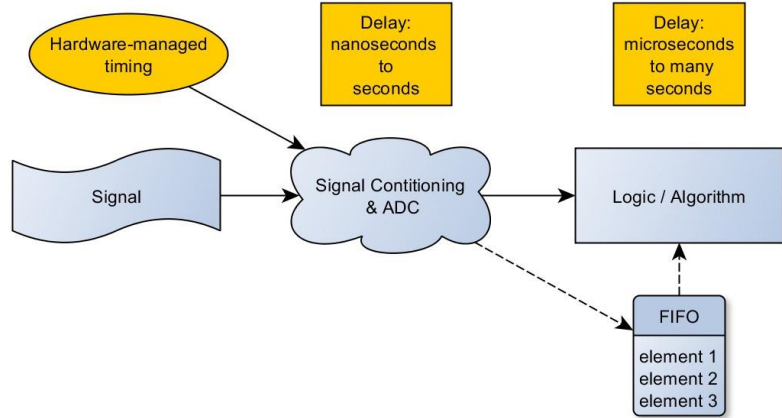
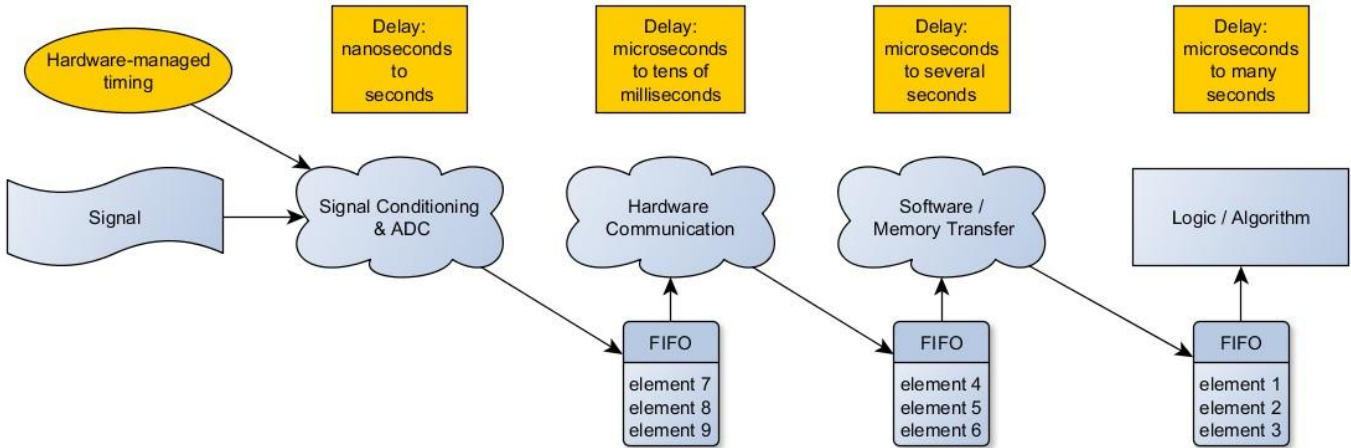


Donanim:FPGA

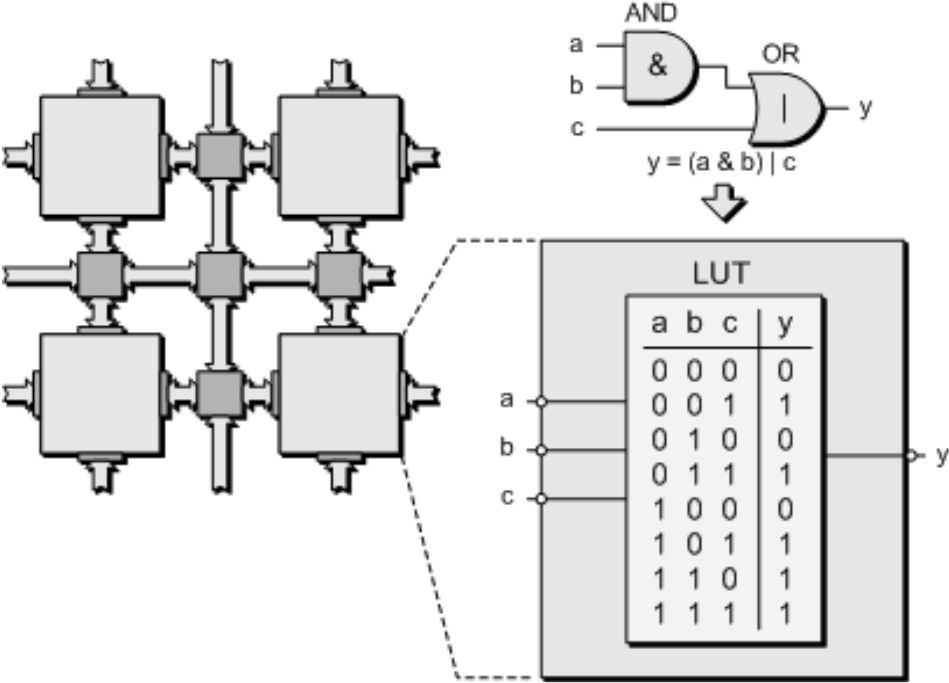
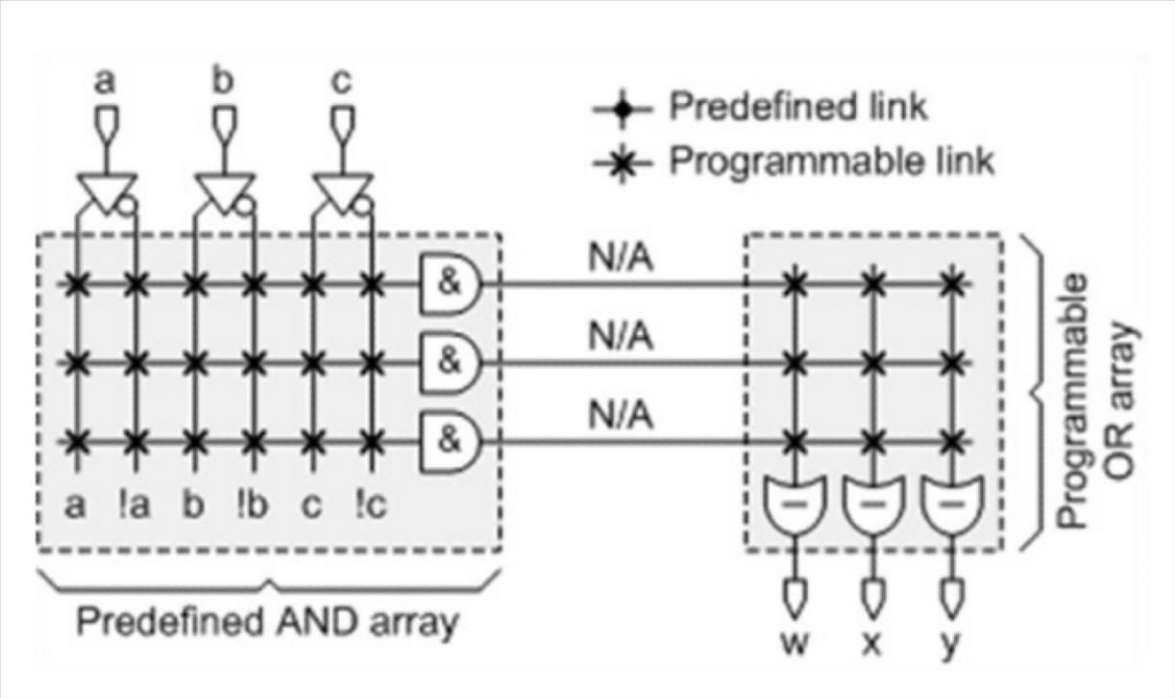
A = 2
B = 3
C = A + B
D = C + A
A = A + 2



Donanim:FPGA

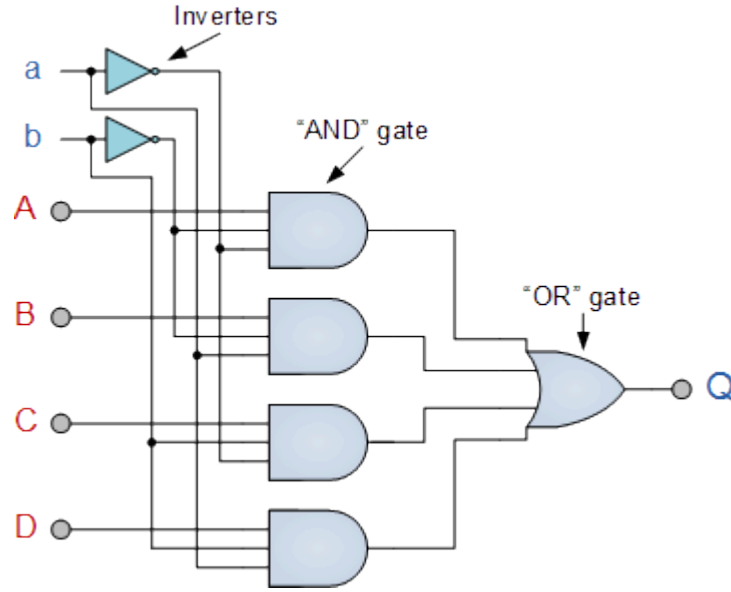
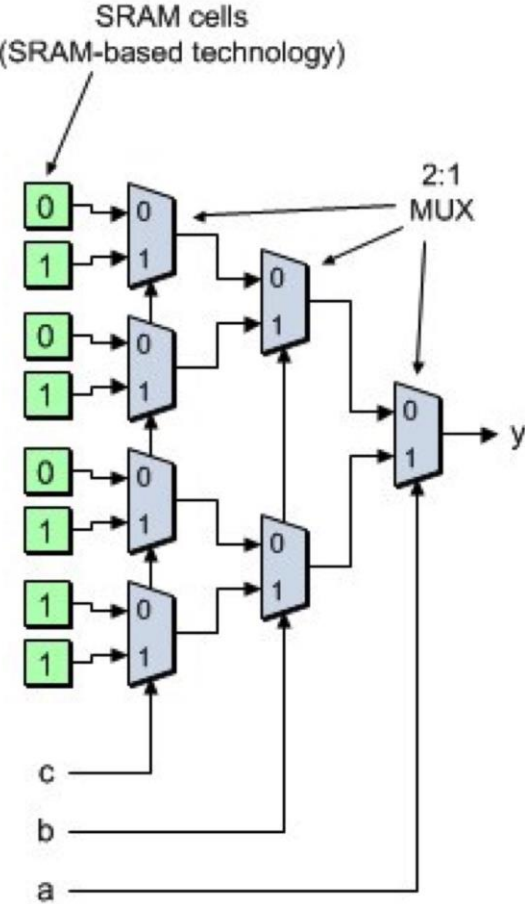
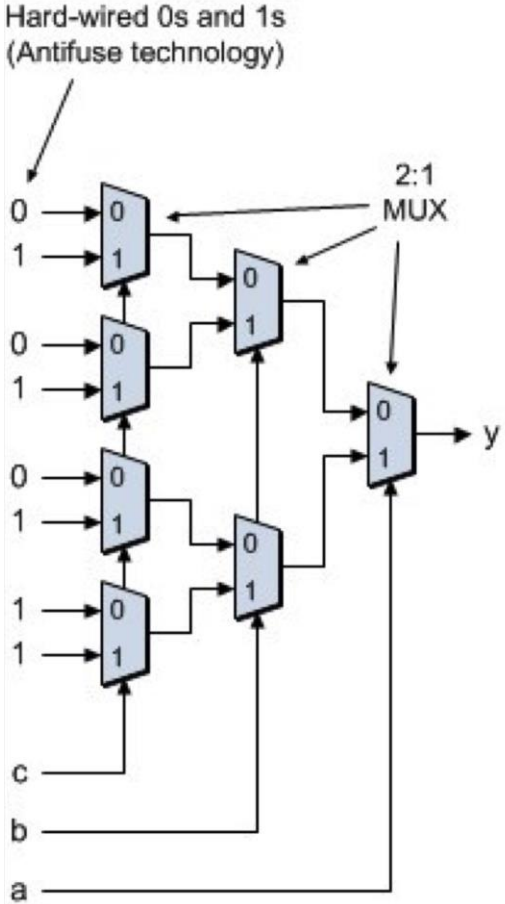


Donanim:FPGA



Programmable Logic Arrays

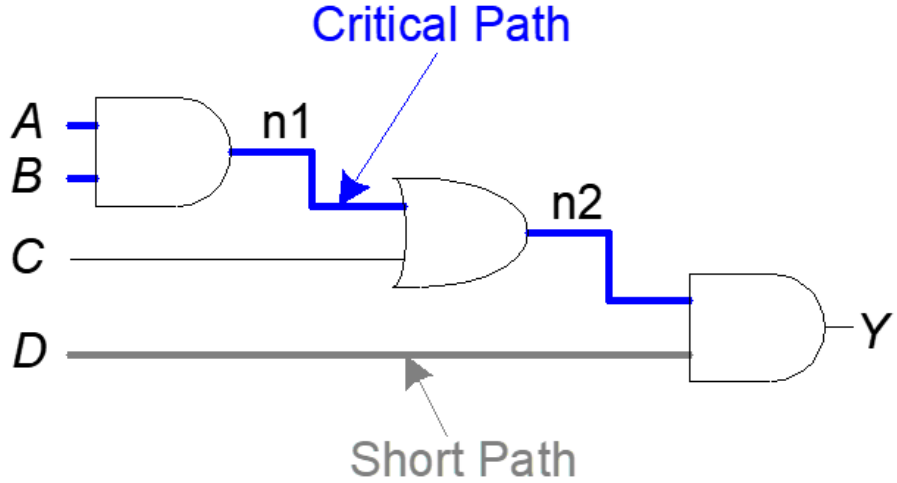
Donanim:FPGA



Donanim:FPGA

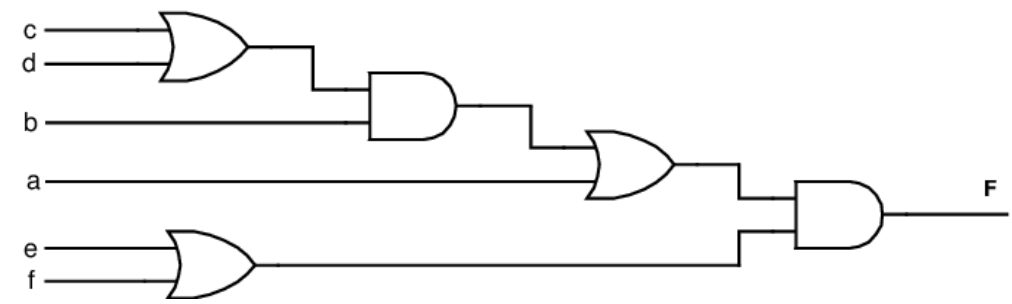
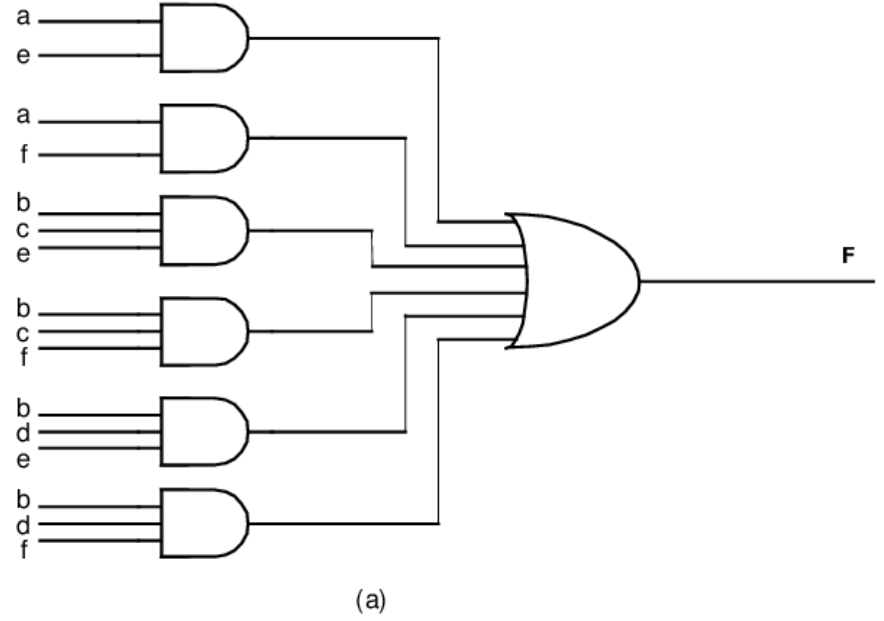
Row number	x_1	x_2	x_3	Minterm	Maxterm
0	0	0	0	$m_0 = \bar{x}_1\bar{x}_2\bar{x}_3$	$M_0 = x_1 + x_2 + x_3$
1	0	0	1	$m_1 = \bar{x}_1\bar{x}_2x_3$	$M_1 = x_1 + x_2 + \bar{x}_3$
2	0	1	0	$m_2 = \bar{x}_1x_2\bar{x}_3$	$M_2 = x_1 + \bar{x}_2 + x_3$
3	0	1	1	$m_3 = \bar{x}_1x_2x_3$	$M_3 = x_1 + \bar{x}_2 + \bar{x}_3$
4	1	0	0	$m_4 = x_1\bar{x}_2\bar{x}_3$	$M_4 = \bar{x}_1 + x_2 + x_3$
5	1	0	1	$m_5 = x_1\bar{x}_2x_3$	$M_5 = \bar{x}_1 + x_2 + \bar{x}_3$
6	1	1	0	$m_6 = x_1x_2\bar{x}_3$	$M_6 = \bar{x}_1 + \bar{x}_2 + x_3$
7	1	1	1	$m_7 = x_1x_2x_3$	$M_7 = \bar{x}_1 + \bar{x}_2 + \bar{x}_3$

Donanim:FPGA



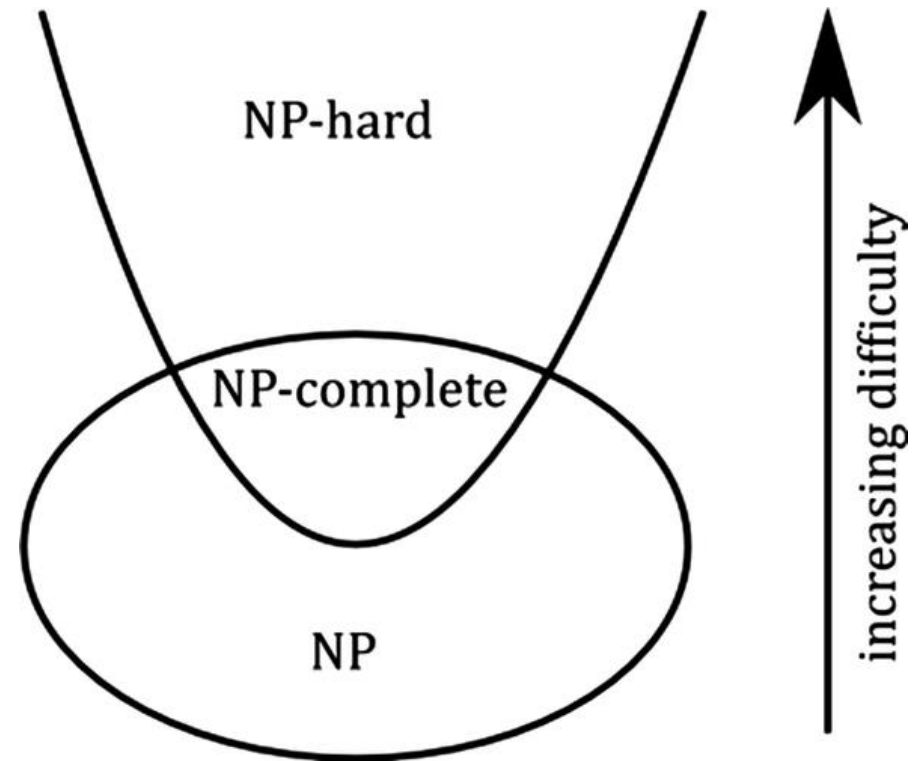
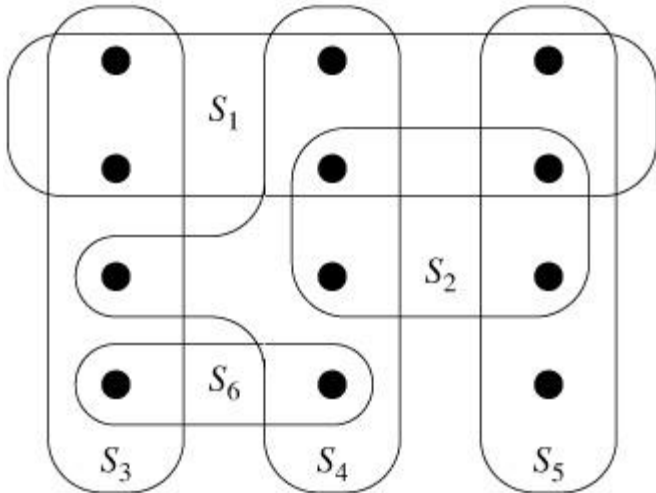
Critical (Long) Path: $t_{pd} = 2t_{pd_AND} + t_{pd_OR}$

Short Path: $t_{cd} = t_{cd_AND}$



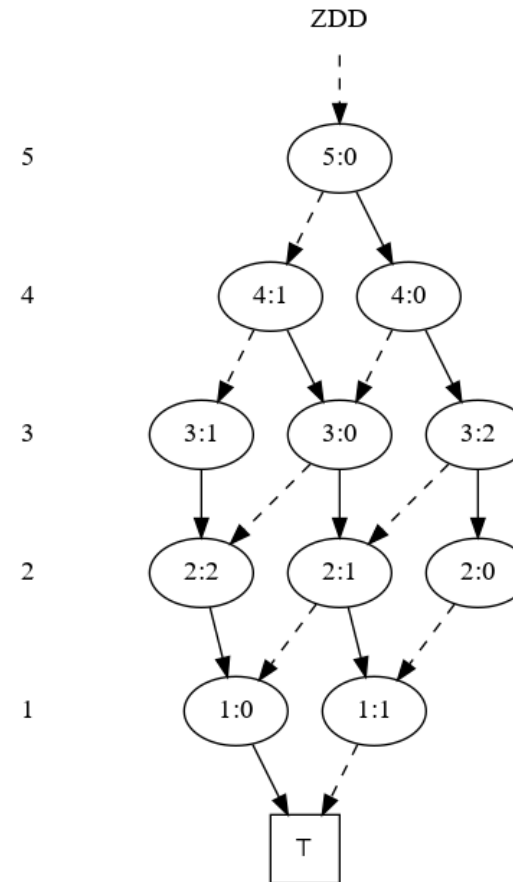
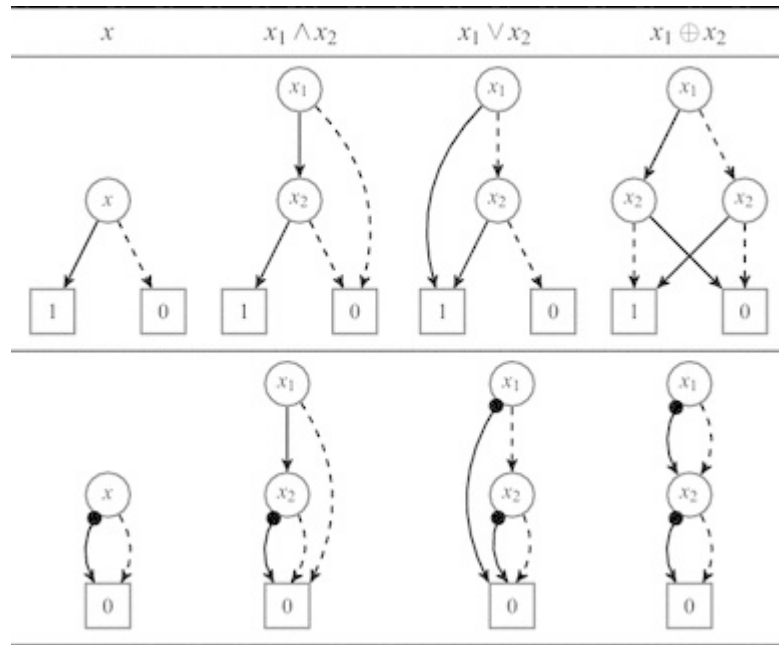
Donanim:FPGA

- Boolean satisfiability problem
- Set Cover Problem
- Quine-McCluskey method

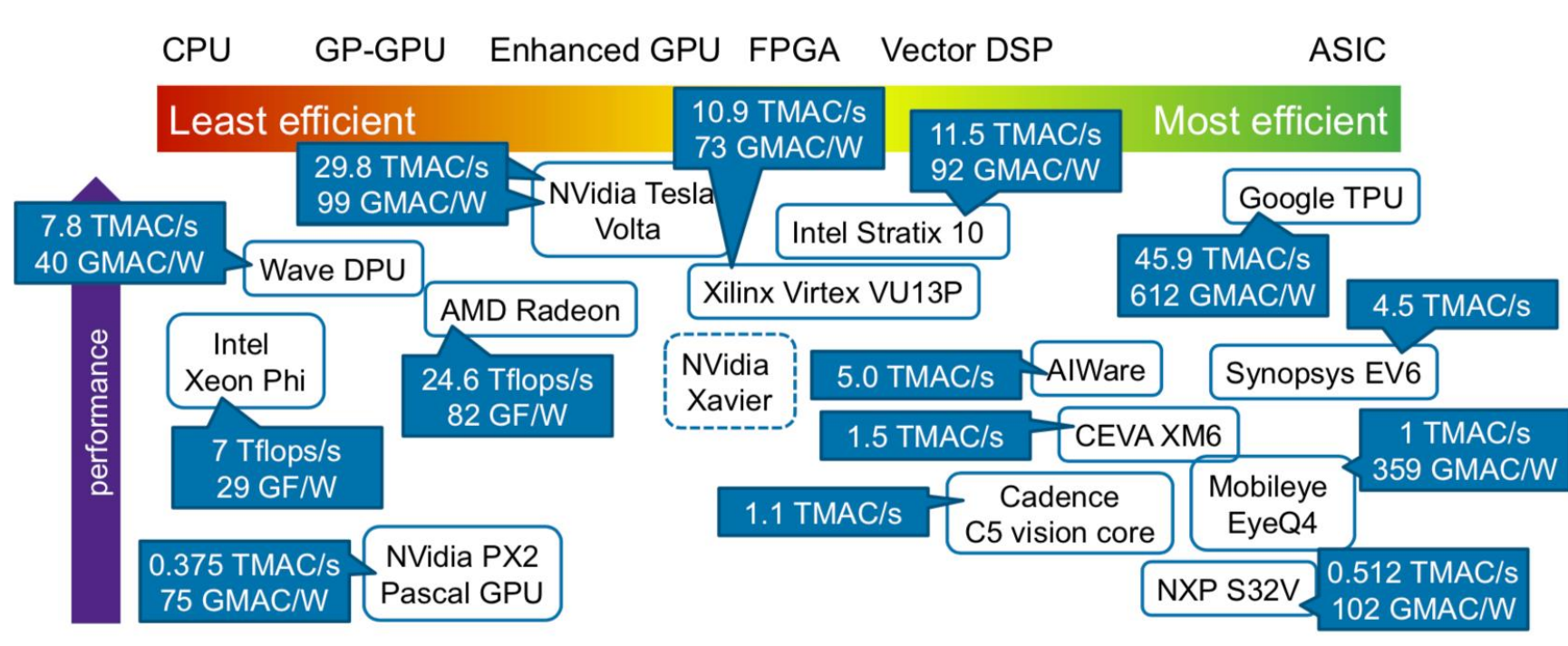


Unate and Binate functions

- BDDs
- ZDDs



Otonom Araçlar



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Test cases	Intensiveness	Best case FPGA		Best case GPU		Time ratio	Energy ratio
		Time	Energy	Time	Energy		
KNN Algorithm 1	Memory-access	4.24ms	0.43mJ	3.04ms	1.2mJ	1.4	0.36
KNN Algorithm 2		1.23ms	0.003J	0.93s	84J	0.0013	3.57E-05
Black Scholes Model European Option	Floating-point computation	0.0788ns	1.67nJ	0.164ns	14.76nJ	0.48	0.11
Black Scholes Model Asian Option		0.0815ns	1.96nJ	0.168ns	15.12nJ	0.45	0.13
Heston Model European Option		0.157ns	3.33nJ	0.604ns	48.32nJ	0.26	0.069
Heston Model European Barrier Option		0.158ns	4.917nJ	0.813ns	65.04nJ	0.19	0.076
Bitonic Sort without HLS Optimizations	Memory-access	152ms	760mJ	16ms	480mJ	9.5	1.58
Bitonic Sort with Optimizations		17ms	272mJ	16ms	480mJ	1.06	0.57

Otonom Araçlar

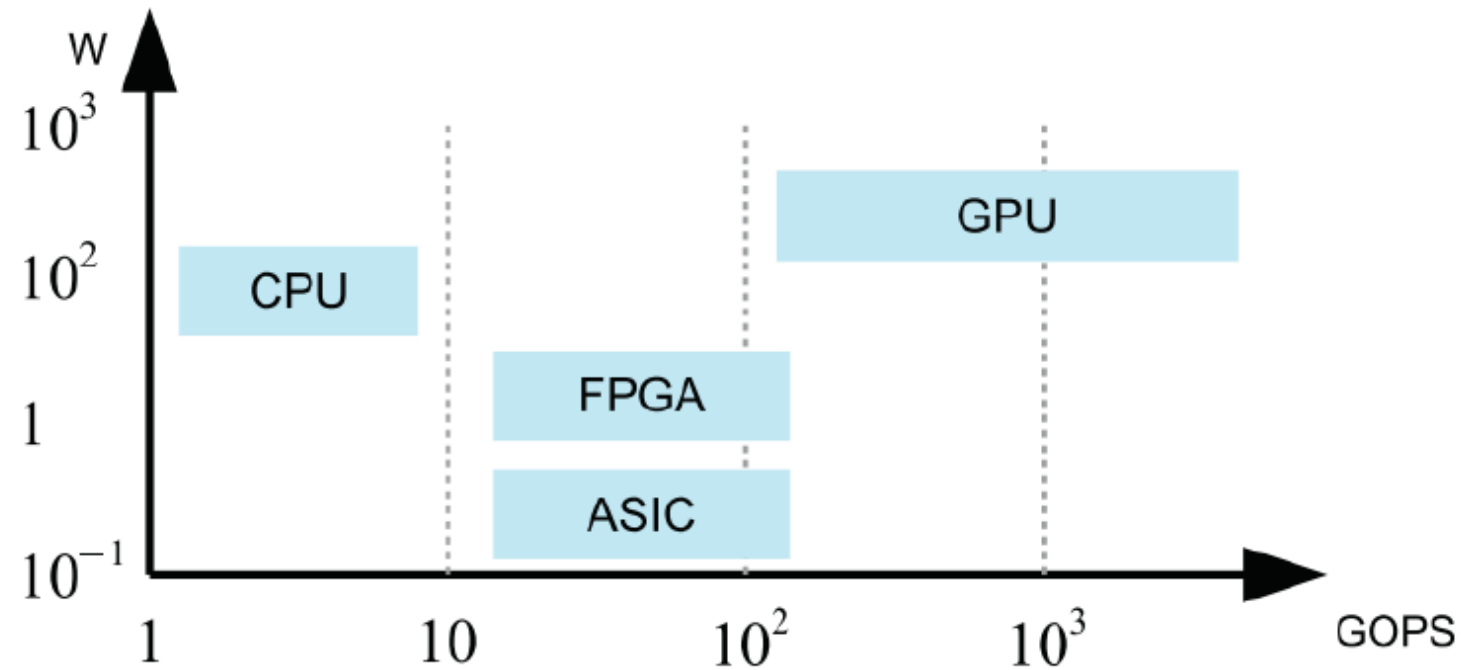
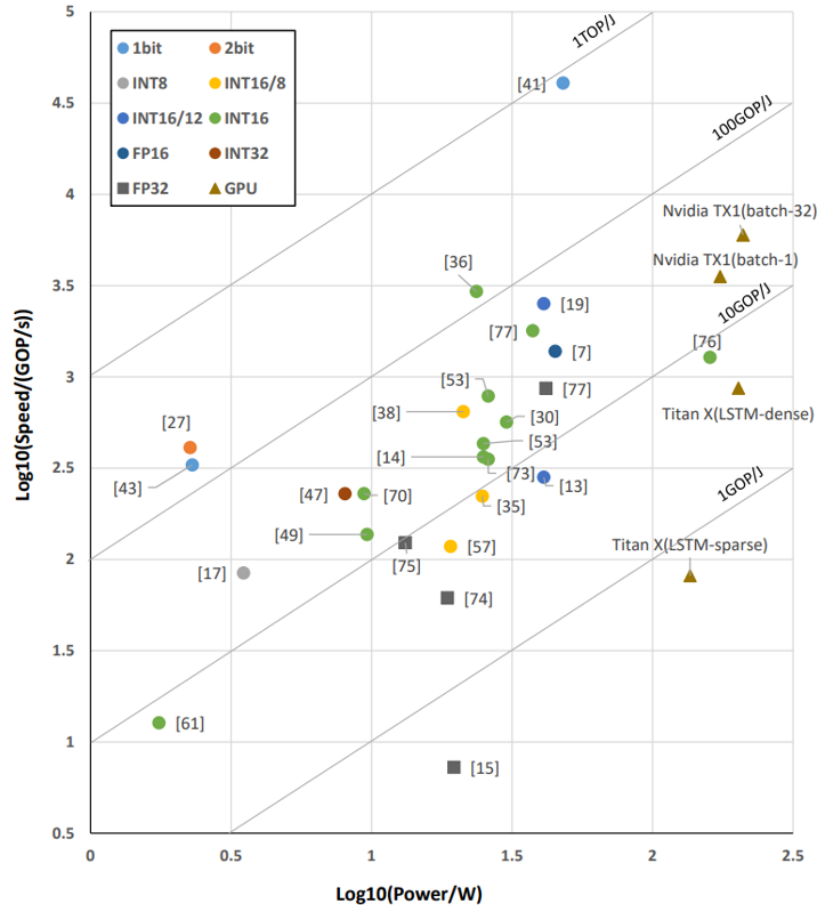


Figure 3.8: Comparison between different platforms.

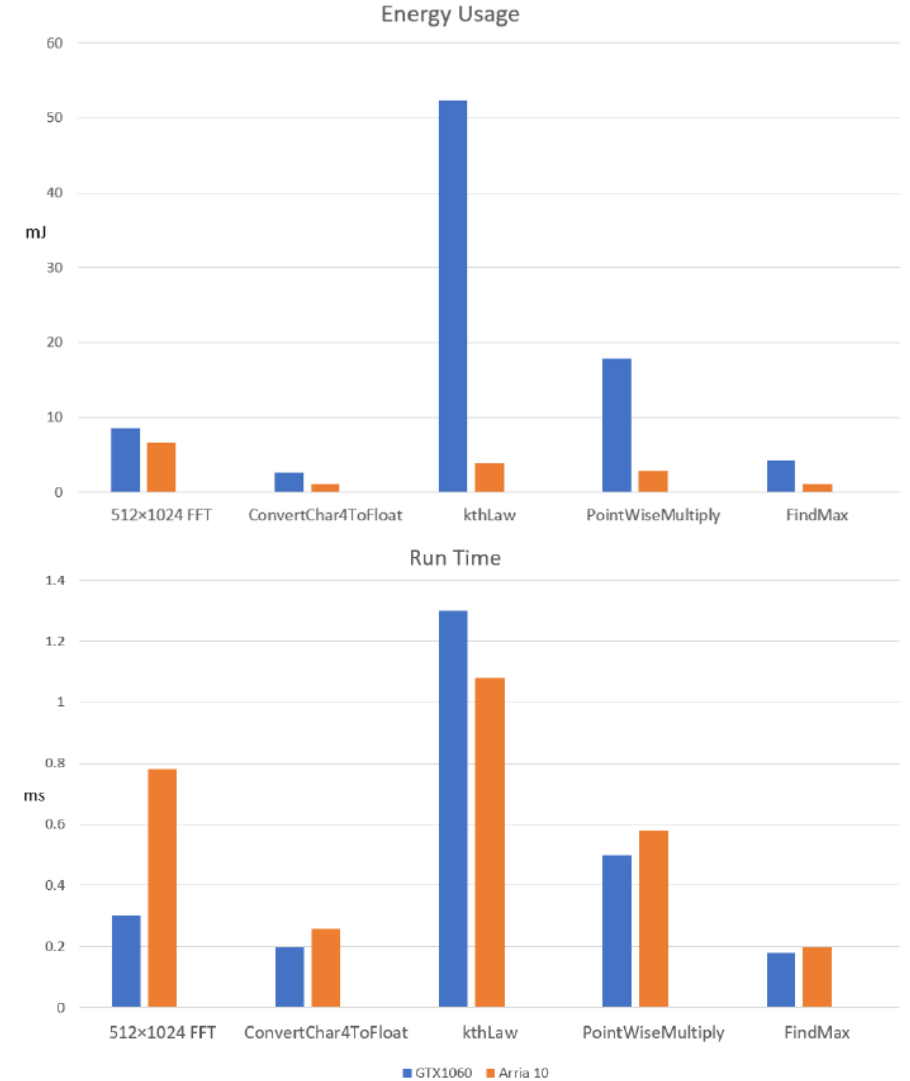
Otonom Araçlar



	Data Format	Speed (GOP/s)	Power (W)	Eff. (GOP/J)	Resource(%)			FPGA chip
					DSP	logic	BRAM	
[43]	1bit	329.47	2.3	143.2	1	34	11	Zynq XC7Z020
[41]	1bit	40770	48	849.38	-	-	-	GX1155
[27]	2bit	410.22	2.26	181.51	41	83	38	Zynq XC7Z020
[17]	INT8	84.3	3.5	24.1	87	84	89	XC7Z020
[57]	INT16/8	117.8	19.1	6.2	13	22	65	5SGSD8
[35]	INT16/8	222.1	24.8	8.96	40	27	40	XC7VX690T
[38]	INT16/8	645.25	21.2	30.43	100	38	70	GX1150
[19]	INT16/12	2520	41	61.5	54	89	88	XCKU060
[61]	INT16	12.73	1.75	7.27	95	67	6	XC7Z020
[49]	INT16	136.97	9.63	14.22	89	84	87	XC7Z045
[70]	INT16	229.5	9.4	24.42	92	71	83	XC7Z045
[73]	INT16	354	26	13.6	78	81	42	XC7VX690T
[14]	INT16	364.4	25	14.6	65	25	46	5SGSMD5
[30]	INT16	565.94	30.2	22.15	60	63	65	XC7VX690T
		431	25	17.1	42	56	52	XC7VX690T
[53]	INT16	785	26	30.2	53	8.3	30	XC7Z020+
								XC7VX690T×6
[76]	INT16	1280.3	160	8	-	-	-	GX1150
[77]	INT16	1790	37.46	47.8	91	43	53	ZCU102
[36]	INT16	2940.7	23.6	124.6	-	-	-	GX1150
[7]	FP16	1382	45	30.7	97	58	92	Stratix V
[47]	INT32	229	8.04	28.5	100	84	18	XC7VX485T
[15]	FP32	7.26	19.63	0.37	42	65	52	XC7VX485T
[74]	FP32	61.62	18.61	3.3	80	61	50	Stratix V
[75]	FP32	123.5	13.18	9.37	88	85	64	GX1150
[77]	FP32	866	41.73	20.75	87	-	46	

Donanım:FPGA vs GPU vs CPU

- FPGA
 - Domain specific hardware
 - high performance-per-watt
 - high-throughput
 - real-time processing
 - flexibility
 - low latency
 - Custom I/O
- GPU
 - Kolay öğrenim
 - Yazılım desteği
 - Geniş kullanım alanı
 - Daha ucuz
- CPU ve Mikrokontroller
 - En kolay öğrenim
 - Yazılım desteği
 - Geniş kullanım alanı
 - En ucuz



Heterogeneous computing

Automotive tracks – Audi A8 Level 3: Aptiv zFAS controller

(Source: www.reverse-costing.com, System Plus Consulting)

➤ NVIDIA Tegra K1

Traffic sign recognition
Pedestrian detection
collision avoidance warning
Light detection
Lane recognition

➤ MobilEye EyeQ3

Traffic sign recognition
360° camera images & processing

Functions: Courtesy of Aptiv



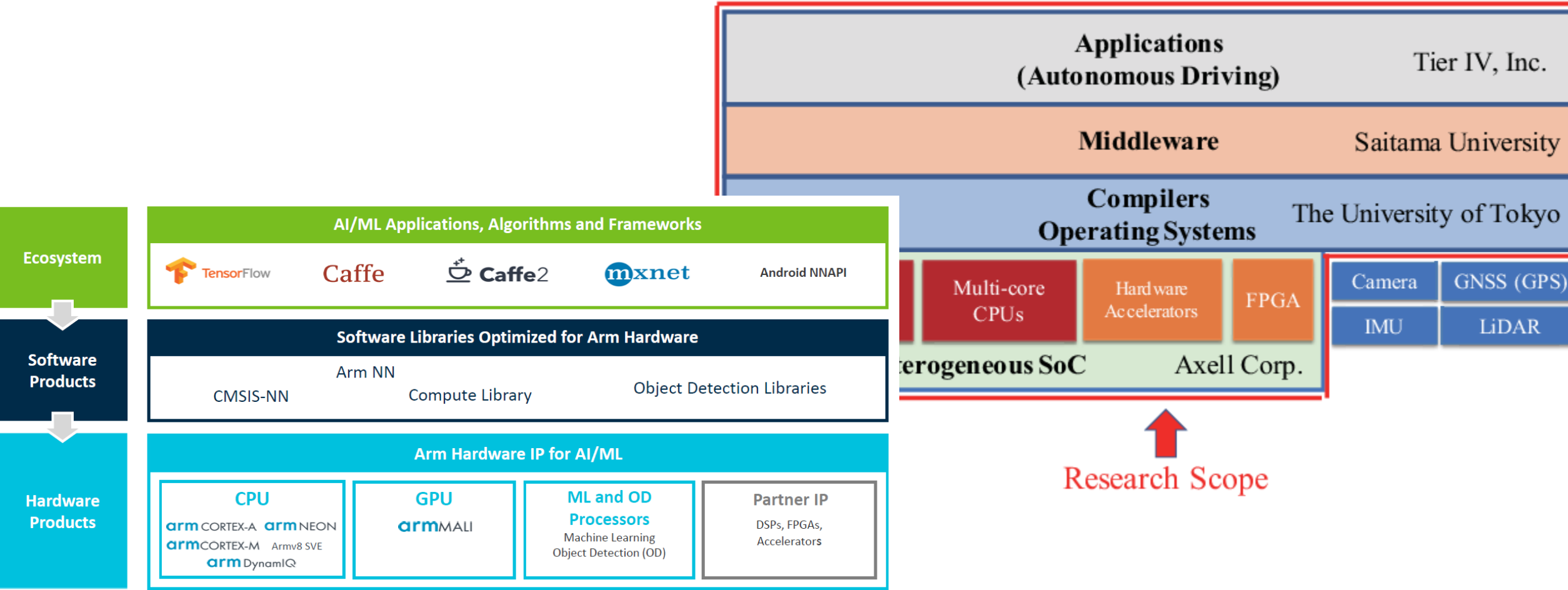
➤ Altera Cyclone

Object fusion
Map fusion
Parking pilot
Pre-crash
Sensor data pre-processing

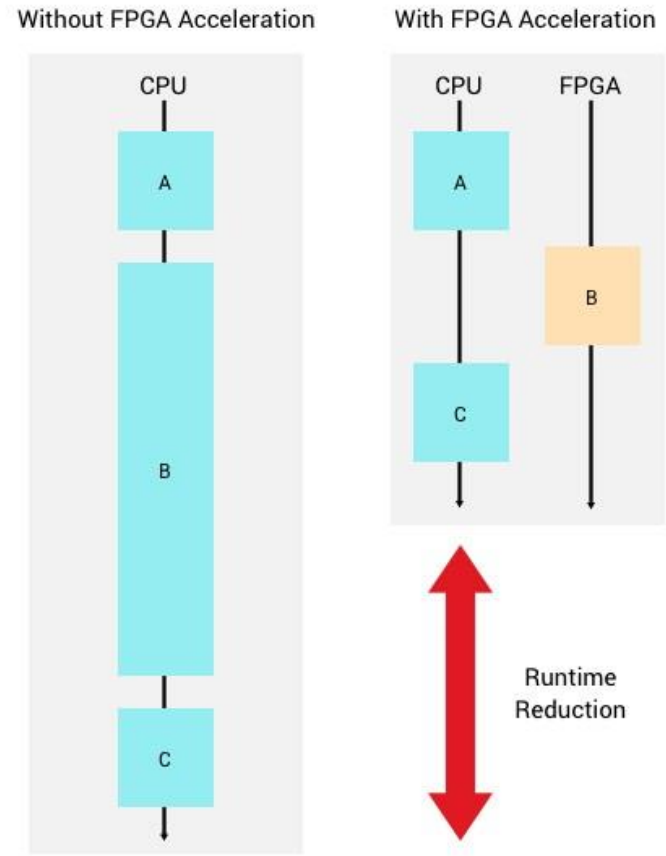
➤ Infineon Aurix Tricore

Traffic jam pilot
Assistance systems
Matrix beam
Road graph

Heterogeneous computing



Heterogeneous computing



Teşekkürler

