

Turning an Automated System into an Autonomous system using Model-Based Design

Autonomous Tech Conference 2018

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Autonomous System

A system that functions **independently** or in a **supervised** manner and operates under conditions of uncertainty, in an unknown and unpredictable dynamic environment



Platform

Sense

Model-Based Design (MBD)

Systematic use of models throughout the development process

Scope: Electronic systems that interact with the physical world





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Automated Driving - Sensors Definition





Multi-object tracker to develop sensor fusion algorithms



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Fuse detections with multi-object tracker



Perception

What is Deep Learning ?

Deep learning is a type of machine learning that performs end-to-end learning by learning tasks directly from images, text, and sound.

1. Train a Deep Neural Network from Scratch



2. Fine-tune a pre-trained model (transfer learning)









Network Analysis

Network Analyzer

		Deep Learning Network Analyzer				- 🗆 🗙
et nalysis date: 22-Apr-2018 11:40:40				144 i layers	0 🛕 warning	o I s errors
• pool1-3	ANALYSIS RESULT					
• pool1-n	Ť	NAME	TYPE	ACTIVATIONS	LEARNABLES	
o conv2-3	18	inception_3a-5x5_reduce 16 1x1x192 convolutions with stride [1 1] and padding [0 0 0 0]	Convolution	28×28×16	Weights 3 Bias 3	×1×192×16 L×1×16
conv2-r	19	inception_3a-relu_5x5_reduce	ReLU	28×28×16	-	
conv2-3x3	20	inception_3a-5x5 32 5x5x16 convolutions with stride [1 1] and padding [2 2 2 2]	Convolution	28×28×32	Weights S Bias	5×5×16×32 L×1×32
conv2-r	21	inception_3a-relu_5x5 ReLU	ReLU	28×28×32	-	
conv2-n	22	inception_3a-pool 3x3 max pooling with stride [1 1] and padding [1 1 1 1]	Max Pooling	28×28×192	-	
pool2-3	23	inception_3a-pool_proj 32 1x1x192 convolutions with stride [1 1] and padding [0 0 0 0]	Convolution	28×28×32	Weights : Bias :	L×1×192×32 L×1×32
inceptio	24	inception_3a-relu_pool_proj ReLU	ReLU	28×28×32	-	
inceptio inceptio	25	inception_3a-output Depth concatenation of 4 inputs	Depth concatenation	28×28×256	-	
inceptio	28	inception_3b-1x1 128 1x1x256 convolutions with stride [1 1] and padding [0 0 0 0]	Convolution	28×28×128	Weights Bias	L×1×256×128 L×1×128
e inceptio	27	inception_3b-relu_1x1 ReLU	ReLU	28×28×128	-	
Inceptio inceptio inceptio	28	inception_3b-3x3_reduce 128 1x1x256 convolutions with stride [1 1] and padding [0 0 0 0]	Convolution	28×28×128	Weights : Bias	l×1×256×128 l×1×128
) inceptio• inceptio• inceptio• inceptio	29	inception_3b-relu_3x3_reduce ReLU	ReLU	28×28×128	-	
• inceptio• inceptio• inceptio	30	inception_3b-3x3 192 3x3x128 convolutions with stride [1 1] and padding [1 1 1 1]	Convolution	28×28×192	Weights Bias	3×3×128×192 L×1×192
• inceptio	31	inception_3b-relu_3x3 ReLU	ReLU	28×28×192	-	
▼ 	32	inception_3b-5x5_reduce 32 1x1x256 convolutions with stride [1 1] and padding [0 0 0 0]	Convolution	28×28×32	Weights Bias	L×1×256×32

Confusion Matrix

Confusion (plotconfusion) – 🗆 🗙													
<u>F</u> ile	<u>E</u> dit <u>\</u>	<u>/</u> iew <u>I</u> nsert	<u>T</u> ools <u>D</u> esk	top <u>W</u> indow	<u>H</u> elp			Y					
				Confusio	on Matrix								
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	Cars	0 0.0%	861 29.8%	12 0.4%	3 0.1%	0 0.0%	98.3% 1.7%						
t Class	Suvs	1 0.0%	32 1.1%	1037 35.9%	16 0.6%	3 0.1%	95.2% 4.8%						
Outpu	Trucks	8 0.3%	4 0.1%	7 0.2%	393 13.6%	5 0.2%	94.2% 5.8%						
	Vans	5 0.2%	1 0.0%	1 0.0%	2 0.1%	235 8.1%	96.3% 3.7%						
		94.8% 5.2%	95.9% 4.1%	98.1% 1.9%	94.2% 5.8%	94.0% 6.0%	96.2% 3.8%						
		Bighucks	Cars	SUNS	THUNS	Vans							
Target Class													
	Output Class	Bigtrucks Cars Suvs Trucks Vans	File Edit View Insert Bigtrucks 255 8.8% Cars 0 Cars 0 Suvs 1 0.0% 3% Vans 5 94.8% 5.2% \$\$0,2% \$\$0,2%	Image: Surve of the second s	Image: Confusion (plot File Edit View Insert Iools Desktop Window Bigtrucks 255 0	Image: control contro control contro control control control control control control co	Image: Confusion (plotconfusion) File Edit Yiew Insert Tools Desktop Window Help Confusion Matrix Bigtrucks 255 0 0 3 7 Cars 0 88% 0.0% 0.0% 0.1% 0.2% Cars 0 0 3 7 0.2% 0.0% 0.0% 0.1% 0.0% 0.2% Suvs 1 32 1037 16 3 0.0% 0.1% 0.1% 0.0% 0.1% 0.1% 0.0% 0.1% 0.1% 0.1% 0.1% 0.0% 0.1% 0.1% 0.1% 0.1% 0.0% 0.1% 0.1% 0.2%	Image: Confusion (plotconfusion) - <					

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R-CNN Vs. Semantic Segmentation





Regions with Convolutional Neural Network Features (R-CNN)

Semantic Segmentation using SegNet



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Path Planning



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Path Control



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Vehicle Modeling



Path Control – Explore Built-In Algorithms



% Compute control inputs for pose

PUREPURSUITBASE methods:

 $[v, w] = pp([0 \ 0 \ 0]);$

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ty output will be e output velocities

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Model-Based Design

- Closest point to a line
- Lookahead instance

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Multi-domain Simulation



Robotics Algorithm Low-Level Control Robot Simulation

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Connect

Physical Modeling Environment



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Sign Recognition with Collision Avoidance



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Vehicle Dynamics

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Automotive



Aerospace



Simulink 3D Animation

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Extensive Hardware Support & Customization



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