

MathWorks Simulation Challenge Year 3



Mark Corless and Lauren Tabolinsky October 10, 2020





Agenda

- Meet the MathWorks Team
- MathWorks offerings for teams
- Simulation Challenge
- New Products
- Questions



Team Mentors

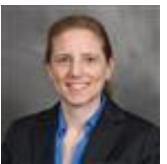
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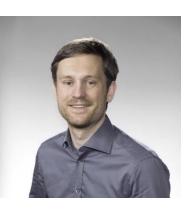




The MathWorks AutoDrive Team

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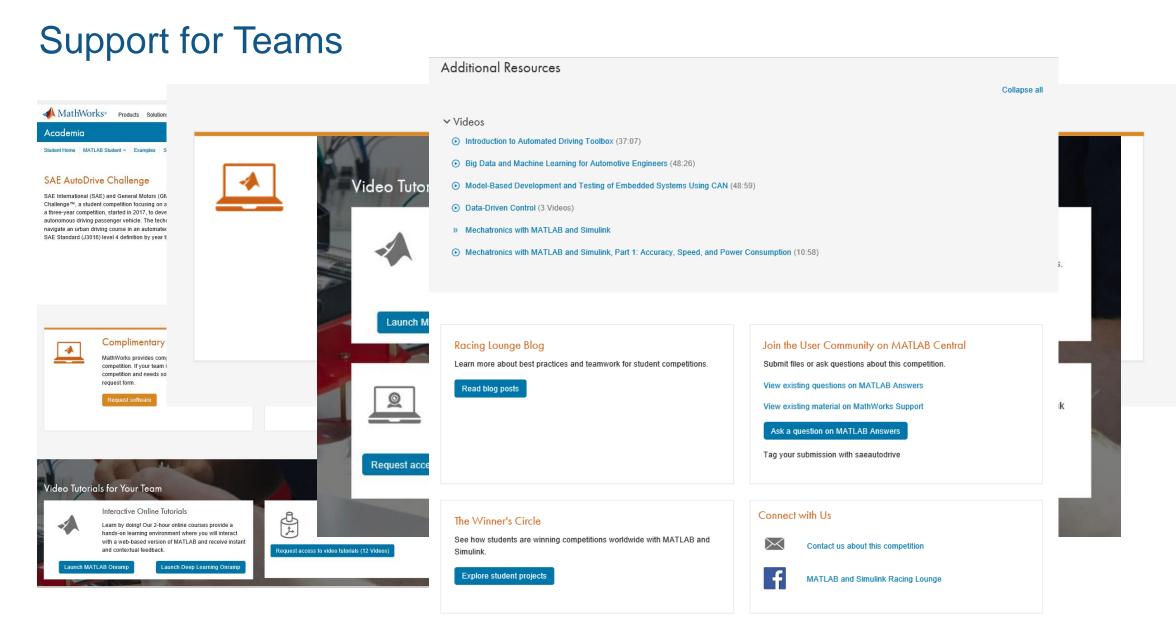


Mark Corless

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AutoDrive Webpage



November Workshop

- MathWorks Workshop (SAE International HQ; Warrendale, PA)
- Dates: Nov. 14-15, 2019
- Agenda is being finalized and registration will open up next week

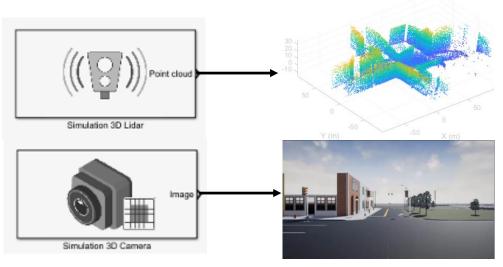




Simulation Challenge

- Last year's challenge was great!
 - 1st customers to see sensors for Unreal engine
 - Highlights
 - **U Toronto**: Best integration between real-world car development and simulation challenge
 - Michigan Tech: Best use of lidar sensor, IPCV and ROS for object detection
 - Texas A+M: IPCV and deep learning using MW tools
- We decided to make the challenge more open this year
 - Points (50 available) will be awarded
 - How has your team applied the core concepts using MathWorks' tools to help achieve the overall competition objectives?







MathWorks Simulation Challenge

for SAE AutoDrive year 3 competition

| # | Task | Points |
|---|--|--------|
| 1 | Synthesize data to test open loop perception algorithm | 10 |
| 2 | Synthesize data to test closed loop controls algorithm | 10 |
| 3 | Generate code from controls algorithm | 10 |
| 4 | Innovate | 15 |
| 5 | Reflect | 5 |



MathWorks Simulation Challenge

for SAE AutoDrive year 3 competition

| # | Task | Points | | | | |
|---|---|--------|--|--|--|--|
| 1 | Synthesize data to test open loop perception algorithm a. What <u>algorithm</u> did you test and why did you choose it? (i.e. object detection, drivable path, localization, sensor fusion) b. How did you <u>synthesize scenario data</u>? (i.e. Unreal Engine, Driving Scenario Designer, sensor models, customizations) c. How did you <u>assess correctness</u> of the algorithm? (i.e. specify truth, assess metrics, automate testing) | 10 | | | | |
| 2 | Synthesize data to test closed loop controls algorithm | 10 | | | | |
| 3 | Generate code from controls algorithm | | | | | |
| 4 | Innovate | 15 | | | | |
| 5 | Reflect | 5 | | | | |



How can I design with virtual driving scenarios?

| Scenes | Cuboid | | | |
|--|---|--|--|--|
| Testing | Controls, sensor fusion, planning | | | |
| Authoring | Driving Scenario Designer App Programmatic API (drivingScenario) | | | |
| Sensing Probabilistic radar (detection list) Probabilistic vision (detection list) Probabilistic lane (detection list) | | | | |



How can I design with virtual driving scenarios?

| Scenes | Cuboid | 3D Simulation (Unreal Engine) | | | |
|-----------|--|--|--|--|--|
| | Ego-Centric View Sconecis Carneer | | | | |
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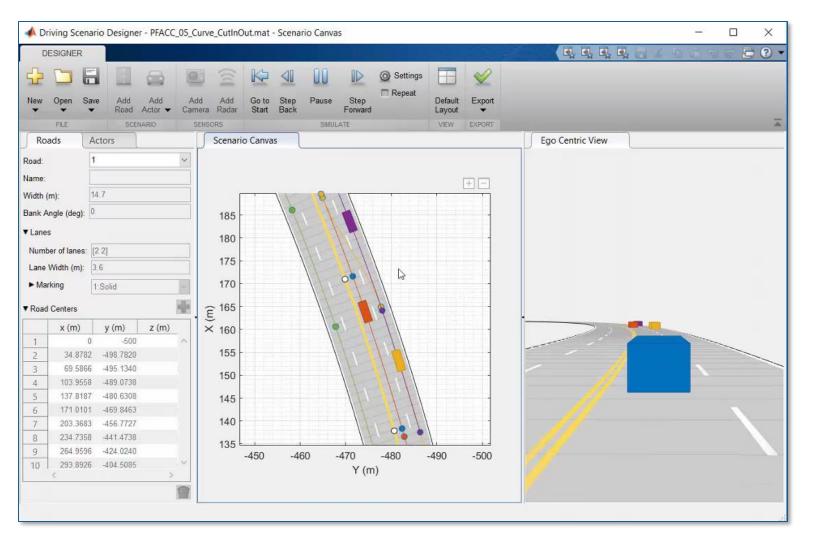


Graphically author driving scenarios

Driving Scenario Designer

- Create roads and lane markings
- Add actors and trajectories
- Specify actor size and radar cross-section (RCS)
- Explore pre-built scenarios
- Import OpenDRIVE roads

Automated Driving Toolbox[™] R2018a



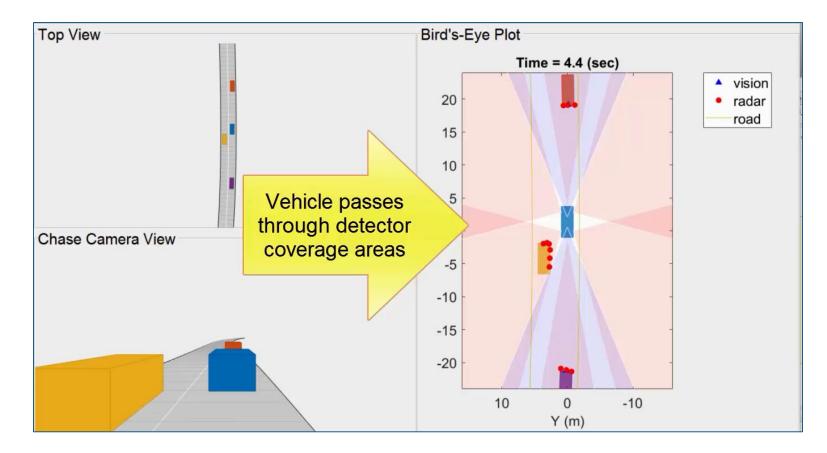


Synthesize scenarios to test sensor fusion algorithms

Sensor Fusion Using Synthetic Radar and Vision Data

- Synthesize road and vehicles
- Add probabilistic vision and radar detection sensors
- Fuse and track detections
- Visualize sensor coverage areas, detections, and tracks

Automated Driving Toolbox[™] R2017a



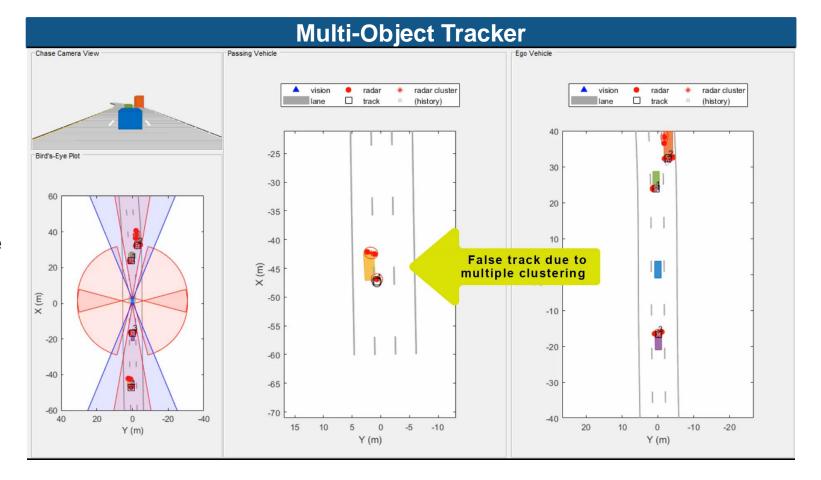
MathWorks

Design multi-object trackers

Extended Object Tracking

- Design multi-object tracker
- Design extended object trackers
- Evaluate tracking metrics
- Evaluate error metrics
- Evaluate desktop execution time

Sensor Fusion and Tracking ToolboxTM Automated Driving ToolboxTM Updated **R2019b**



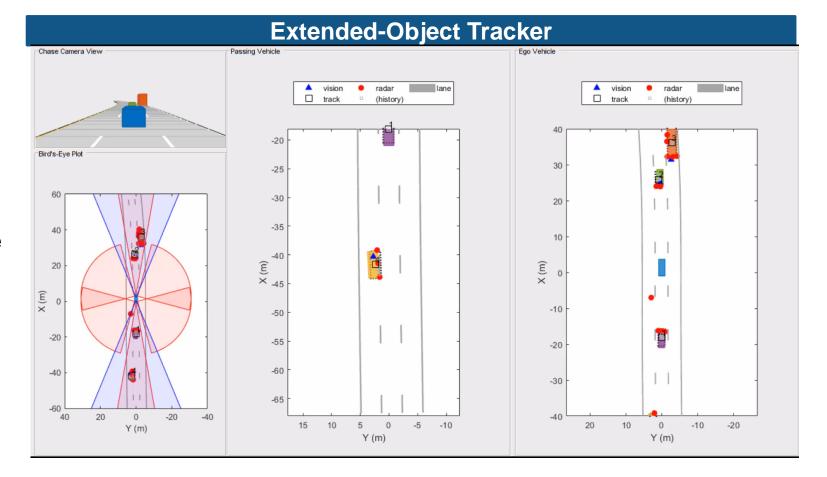
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Sensor Fusion and Tracking Toolbox[™] Automated Driving Toolbox[™] Updated **R2019b**



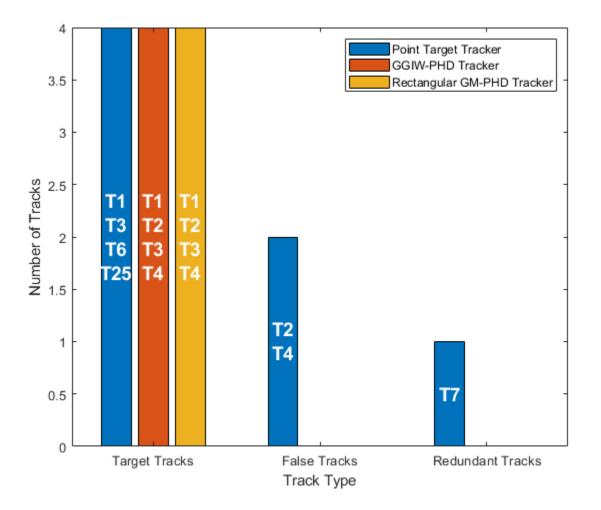


Evaluate tracking performance

Extended Object Tracking

- Design multi-object tracker
- Design extended object trackers
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- Evaluate error metrics
- Evaluate desktop execution time

Sensor Fusion and Tracking ToolboxTM Automated Driving ToolboxTM Updated **R2019b**



Multi-object tracker

GGIW-Probability Hypothesis Density tracker Extended object (size and orientation) tracker

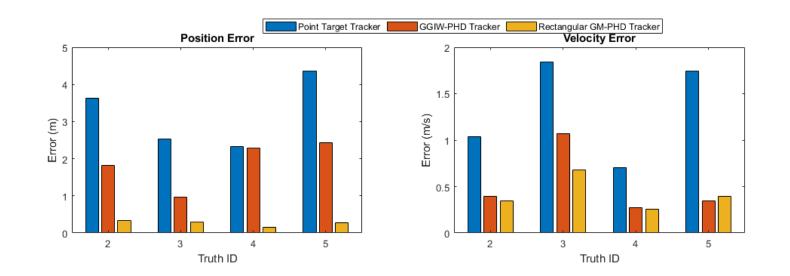


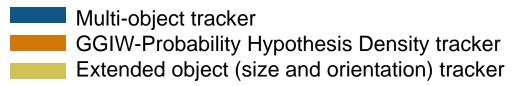
Evaluate error metrics

Extended Object Tracking

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Sensor Fusion and Tracking ToolboxTM Automated Driving ToolboxTM Updated R2019b





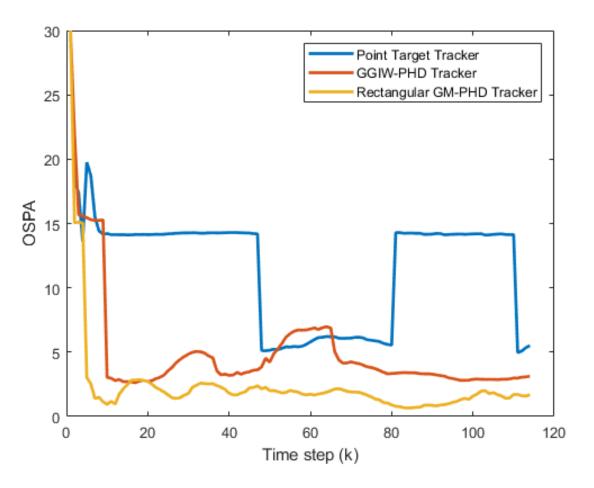


Evaluate OSPA metrics

Extended Object Tracking

- Design multi-object tracker
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Sensor Fusion and Tracking ToolboxTM Automated Driving ToolboxTM Updated **R2019b**



Multi-object tracker GGIW-Probability Hypothesis Density tracker Extended object (size and orientation) tracker

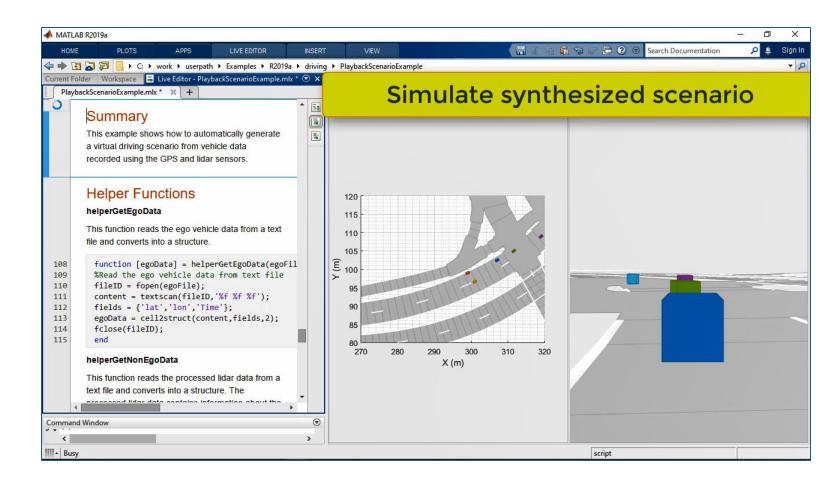


Synthesize driving scenarios from recorded data

Scenario Generation from Recorded Vehicle Data

- Visualize video
- Import OpenDRIVE roads
- Import GPS
- Import object lists

Automated Driving Toolbox[™] R2019a





How can I design with virtual driving scenarios?

| Scenes | Cuboid |
|-----------|--|
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3D Simulation (Unreal Engine)



Controls, sensor fusion, planning, perception

Unreal Engine Editor

Probabilistic radar (detection list) Monocular camera (image, labels, depth) Fisheye camera (image) Lidar (point cloud)

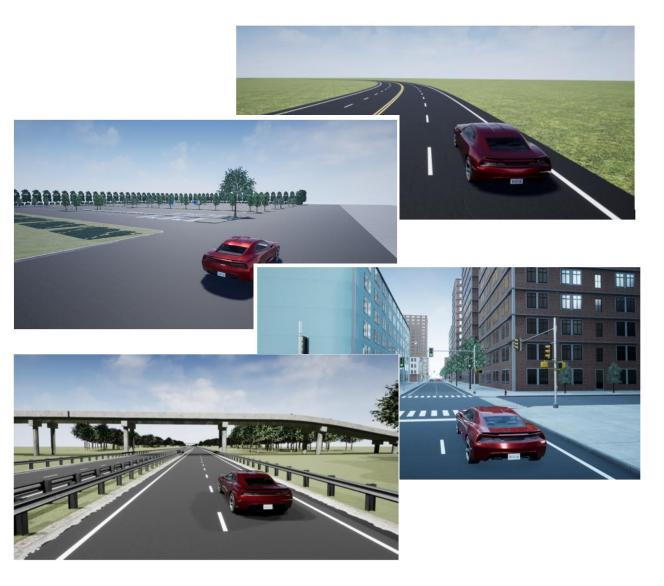


Select from prebuilt 3D simulation scenes

3D Simulation for Automated Driving

- Straight road
- Curved road
- Parking lot
- Double lane change
- Open surface
- US city block
- US highway
- Virtual Mcity

Automated Driving Toolbox[™] R2019b

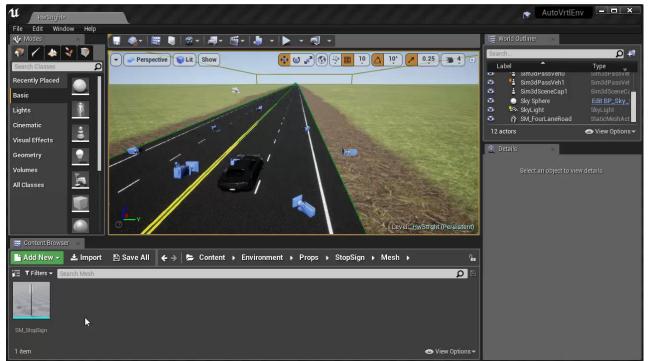




Customize 3D simulation scenes

Support Package for Customizing Scenes

- Install Unreal Engine
- Set up environment and open Unreal Editor
- Configure configuration Block for Unreal Editor co-simulation
- Use Unreal Editor to customize scenes
- Create an Unreal Engine project executable file



Vehicle Dynamics BlocksetTM

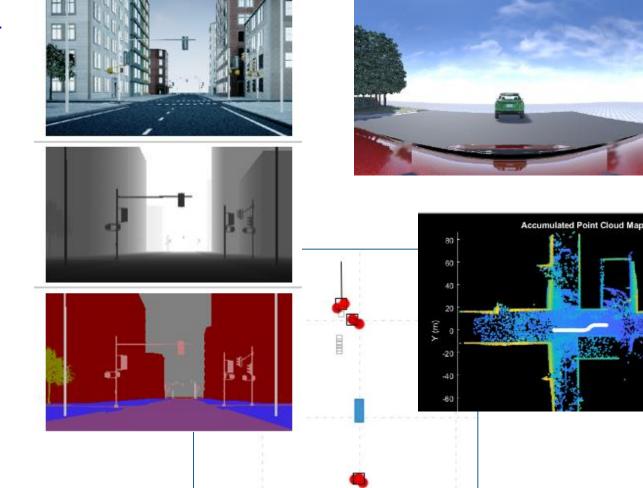




Model sensors in 3D simulation environment

3D Simulation for Automated Driving

- Monocular camera
- Fisheye camera
- Lidar
- Probabilistic radar









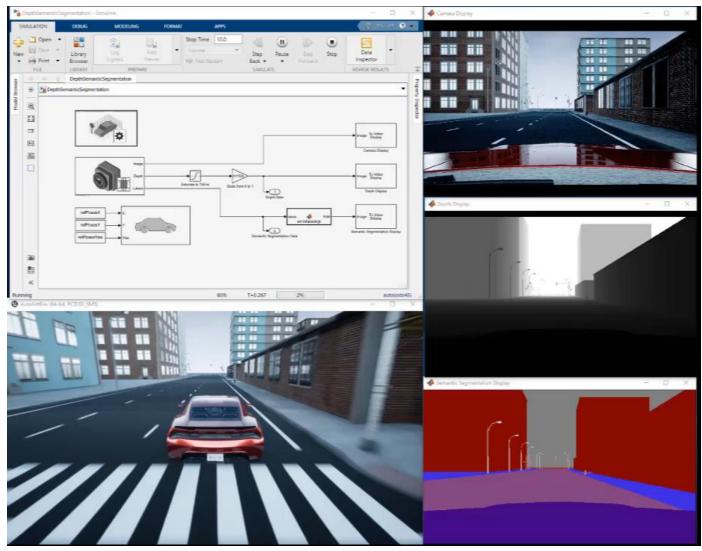
Synthesize monocular camera sensor data

Visualize Depth and Semantic Segmentation Data in 3D Environment

- Synthesize RGB image
- Synthesize depth map
- Synthesize sematic segmentation

Automated Driving ToolboxTM



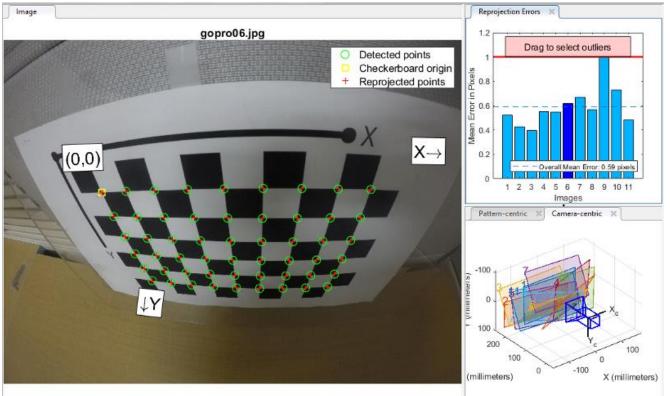




Calibrate monocular camera model

Single Camera Calibrator App

- Prepare the Pattern, Camera, and Images
- Add Images and Select Camera Model
- Calibrate
- Evaluate Calibration Results



Computer Vision Toolbox[™] R2013b



Synthesize fisheye camera sensor data

Simulate a Simple Driving Scenario and Sensor in 3D Environment

- Explore camera model (Scaramuzza)
- Configure distortion center, image size and mapping coefficients
- Visualize results



Automated Driving Toolbox[™] R2019b

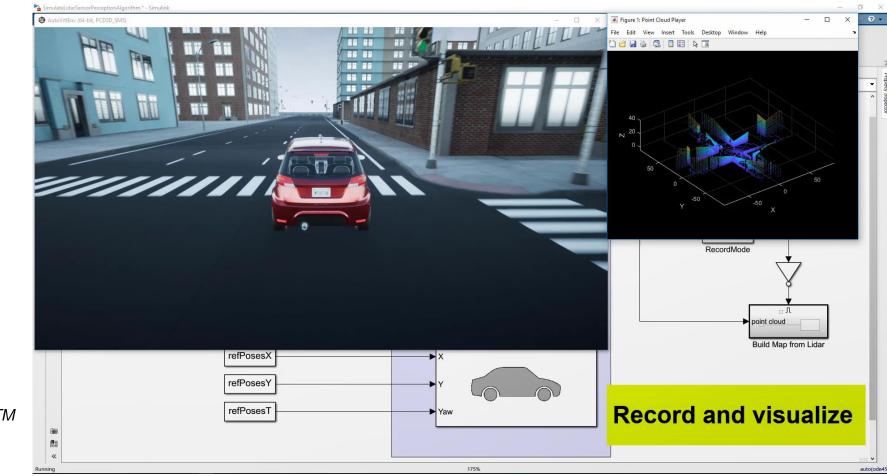


Synthesize lidar sensor data

Simulate Lidar Sensor Perception Algorithm

- Record and visualize
- Develop algorithm
- Build a 3D map
- Use algorithm within simulation environment

Automated Driving Toolbox[™] R2019b



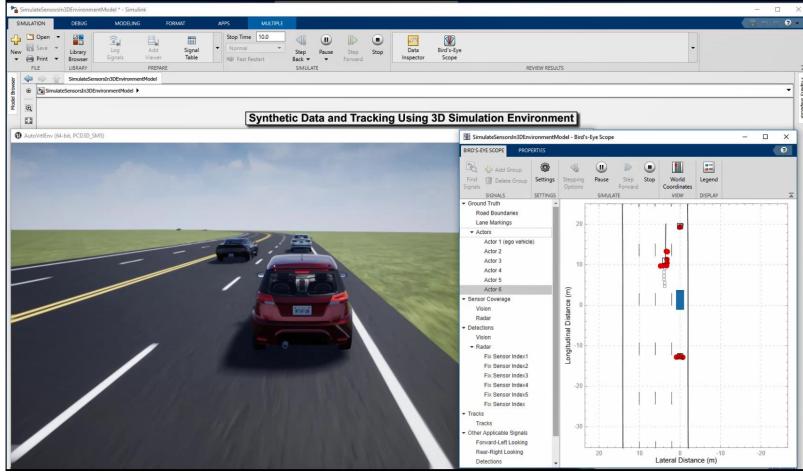


Synthesize radar sensor data

Simulate Radar Sensors in 3D Environment

- Extract the center locations
- Use center location for road creation using driving scenario
- Define multiple moving vehicles
- Export trajectories from app
- Configure multiple probabilistic radar models
- Calculate confirmed track

Automated Driving Toolbox[™] R2019b





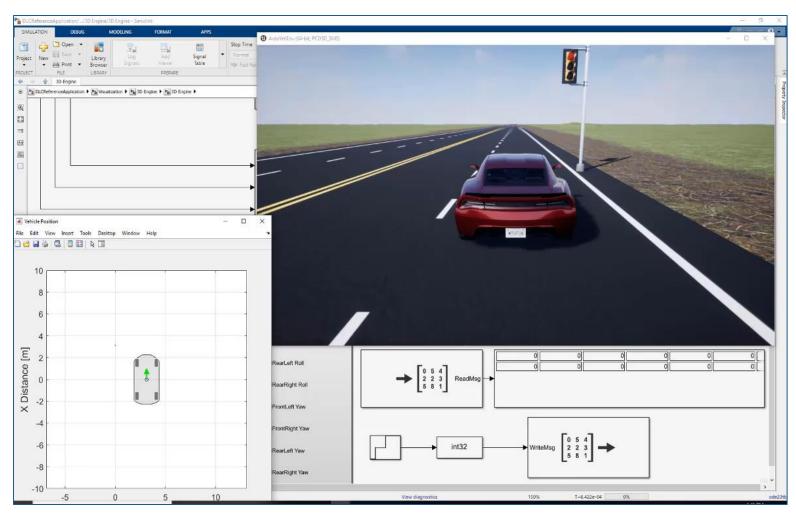
Communicate with the 3D simulation environment

Send and Receive Double-Lane Change Scene Data

- Simulation 3D Message Set
 - Send data to Unreal Engine
 - Traffic light color
- Simulation 3D Message Get
 - Retrieve data from Unreal Engine
 - Number of cones hit

Vehicle Dynamics BlocksetTM







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for SAE AutoDrive year 3 competition

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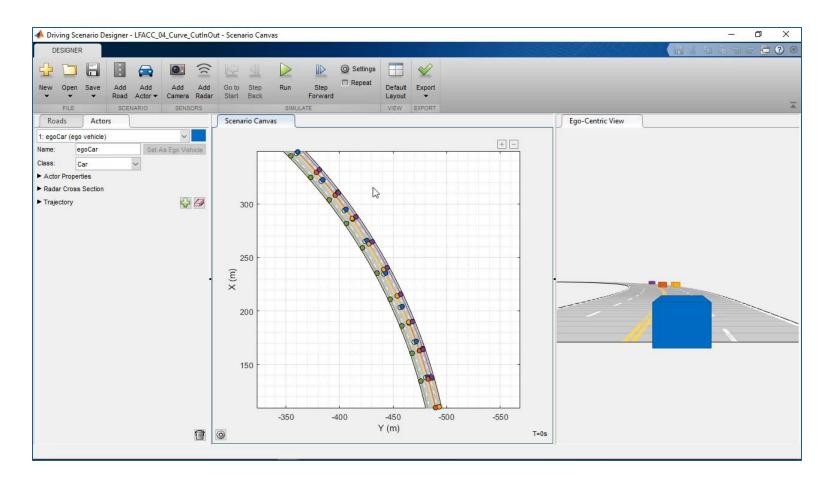
Integrate driving scenario into closed loop simulation

Lane Following Control with Sensor Fusion

- Integrate scenario into system
- Design lateral (lane keeping) and longitudinal (lane spacing) model predictive controllers
- Visualize sensors and tracks
- Generate C/C++ code
- Test with software in the loop (SIL) simulation

Model Predictive Control Toolbox[™] Automated Driving Toolbox[™] Embedded Coder[®]







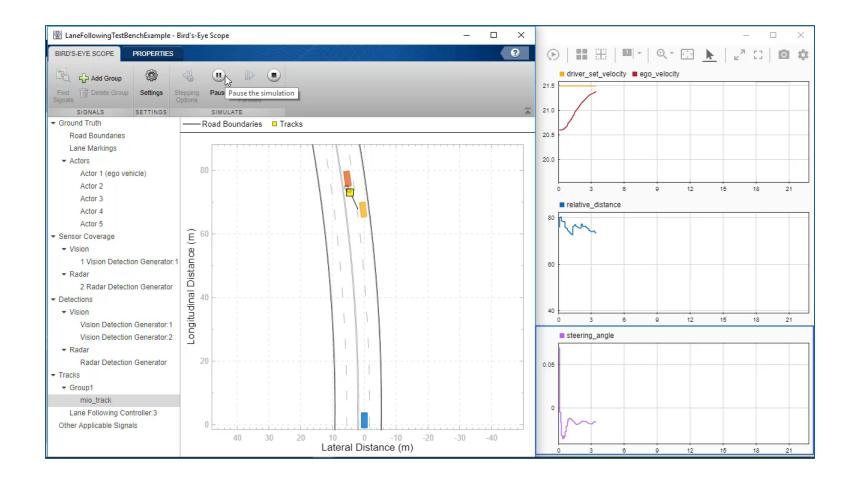
Design lateral and longitudinal controls

Lane Following Control with Sensor Fusion

- Integrate scenario into system
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Model Predictive Control Toolbox[™] Automated Driving Toolbox[™] Embedded Coder[®]







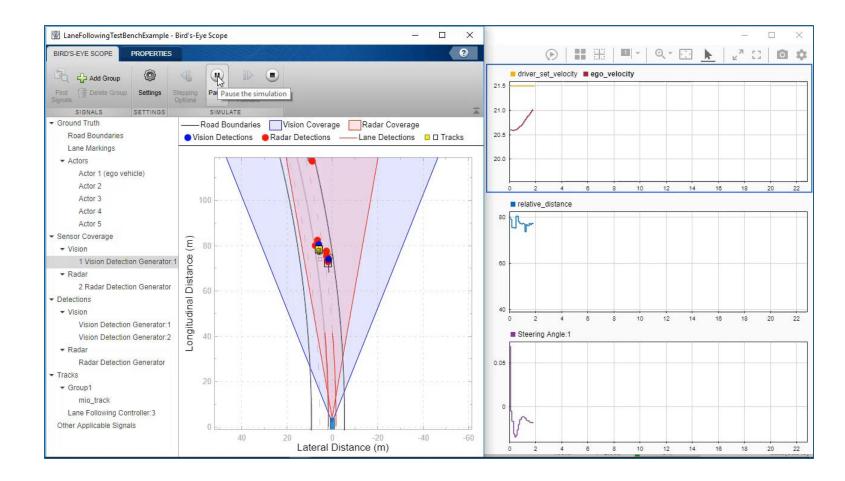
Visualize sensor detections and tracks

Lane Following Control with Sensor Fusion

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Model Predictive Control Toolbox[™] Automated Driving Toolbox[™] Embedded Coder[®]







Automate testing against driving scenarios

Lo

Testing a Lane Following Controller with Simulink Test

- Author high level requirements
- Synthesize driving scenarios
- Specify assessment criteria
- Run interactive simulation
- Automate regression testing
- Review verification status

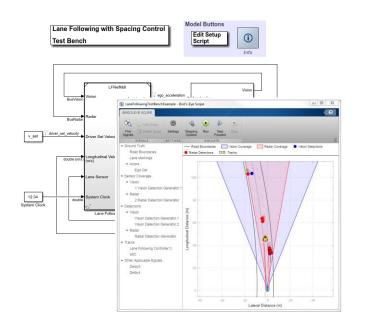
Simulink TestTM Automated Driving ToolboxTM Model Predictive Control ToolboxTM

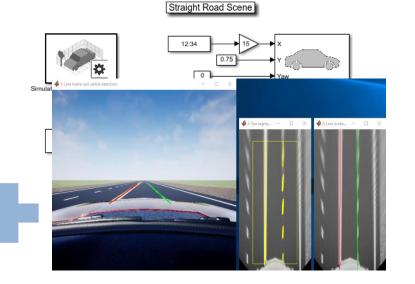


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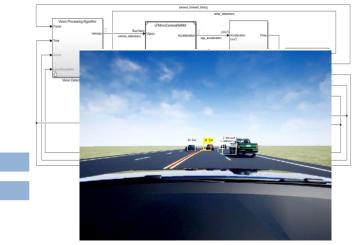


Synthesize scenarios to test your design





Lane Following with Mono Camera Detector Test Bench



Lane Following Control with Sensor Fusion

Model Predictive Control Toolbox[™] Automated Driving Toolbox[™] Embedded Coder[®]



Design of Lane Marker Detector in 3D Simulation Environment

Automated Driving Toolbox[™] R2019b

Lane-Following Control with Monocular Camera Perception

Model Predictive Control Toolbox[™] Automated Driving Toolbox[™] Vehicle Dynamics Blockset[™]





Simulate controls with perception

Lane-Following Control with Monocular Camera Perception

- Author target vehicle trajectories
- Synthesize monocular camera and probabilistic radar sensors
- Model lane following and spacing control in Simulink
- Model lane boundary and vehicle detectors in MATLAB code

Model Predictive Control Toolbox[™] Automated Driving Toolbox[™] Vehicle Dynamics Blockset[™] Updated **R2019**





Visualize logged simulation detection and camera data

Lane-Following Control with Monocular Camera Perception

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Model Predictive Control Toolbox[™] Automated Driving Toolbox[™] Vehicle Dynamics Blockset[™] Updated R2019b

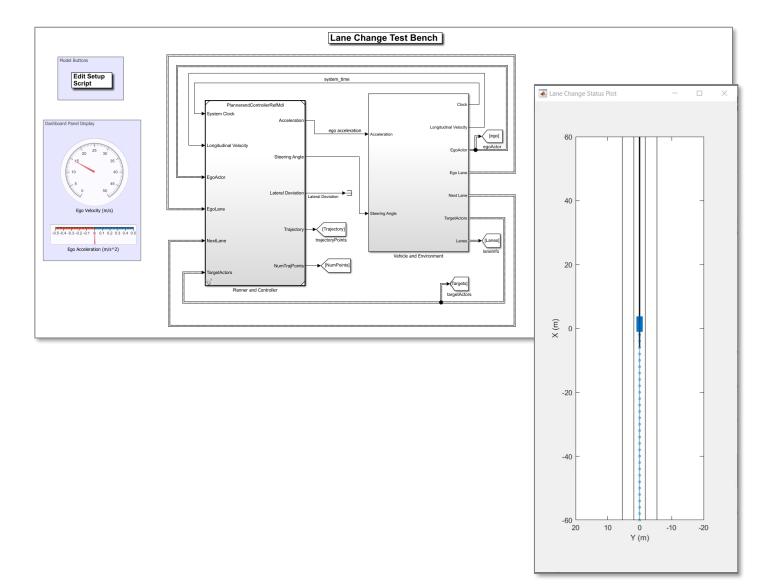




Design highway automated lane change maneuver

Lane Change for Highway Driving

- Find most important objects
- Generate optimal trajectory for collision-free lane change
- Extract path from trajectory
- Follow path with Model Predictive Control (MPC)



Navigation ToolboxTM Model Predictive Control ToolboxTM Automated Driving ToolboxTM





for SAE AutoDrive year 3 competition

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| 4 | Innovate | 15 |
| 5 | Reflect | 5 |



Generate C/C++ code for path planner and controller

<u>Code Generation for Path</u> <u>Planning and Vehicle Control</u>

- Simulate system
- Configure for code generation
- Generate C/C++ code
- Test using Software-In-the-Loop
- Measure execution time of generated code

Automated Driving Toolbox[™] Embedded Coder R2019c

| 187 | // model step function | |
|-----|---|-------------------------------|
| 188 | void step0(); | p0 = 50 msec rate |
| 189 | | $p_1 = 100 \text{ msec rate}$ |
| 190 | // model step function | |
| 191 | <pre>void step1();</pre> | |
| 192 | | |
| 193 | <pre>// model terminate function</pre> | |
| 194 | <pre>void terminate();</pre> | |
| 195 | | |
| 196 | // Constructor | |
| 197 | AutomatedParkingValetModelClass(); | |
| 198 | | |
| 199 | // Destructor | |
| 200 | ~AutomatedParkingValetModelClass(); | 5 |
| 201 | | |
| 202 | <pre>// Root inport: '<u><root>/Costmap</root></u>' set metho</pre> | d |
| 203 | <pre>void setCostmap(costmapBus localArgInput);</pre> | Methods to access |
| 204 | | |
| 205 | // Root inport: ' <u><root>/GoalPose</root></u> ' set meth | inputs and outputs |
| 206 | <pre>void setGoalPose(real_T localArgInput[3]);</pre> | |

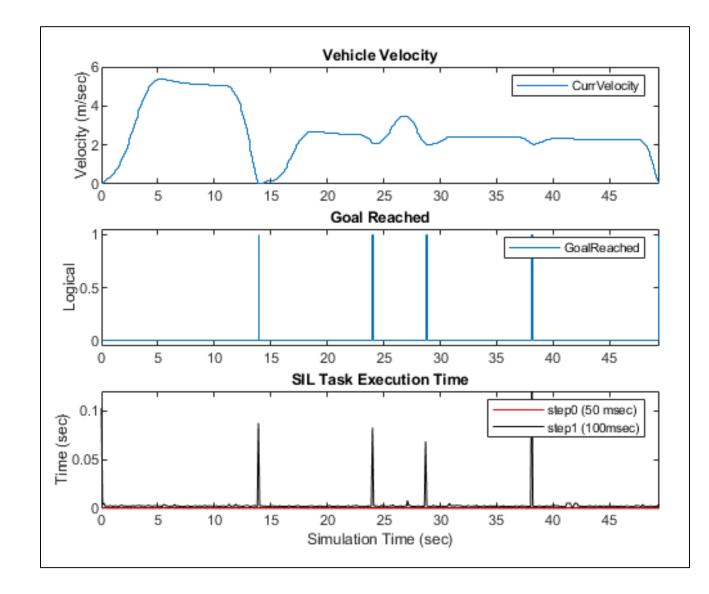


Measure execution time of generated code

<u>Code Generation for Path</u> <u>Planning and Vehicle Control</u>

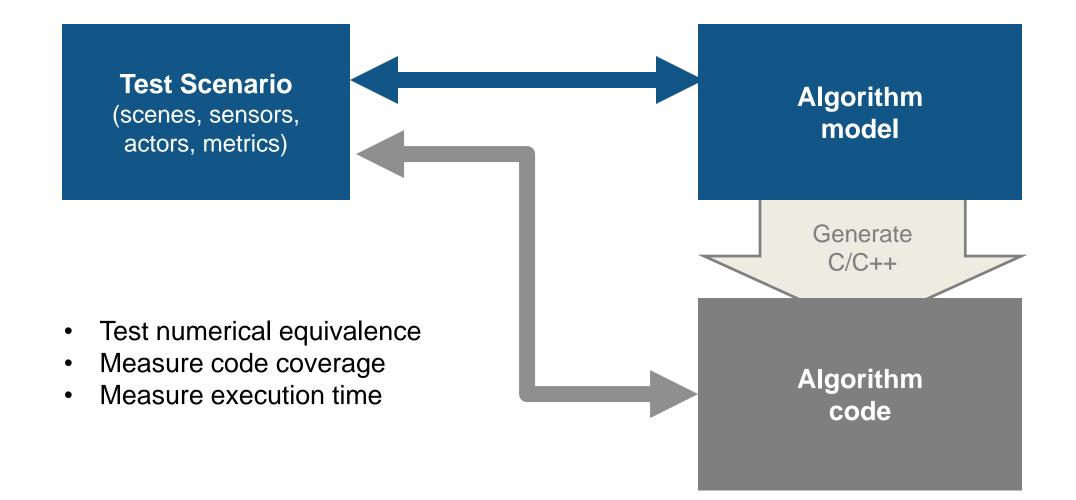
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Automated Driving Toolbox[™] Embedded Coder R2019c





Evaluate generated code with software-in-the-loop (SIL) simulation

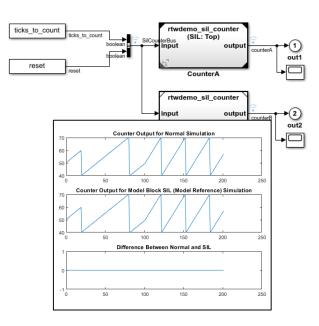


Evaluate generated code with software-in-the-loop (SIL) simulation

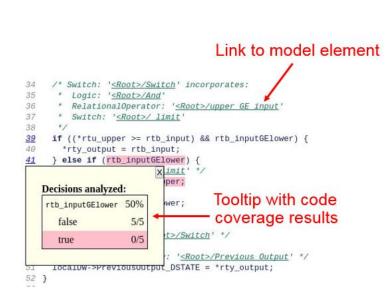
Measure

code coverage

Test numerical equivalence



Software-in-the-Loop Simulation Embedded Coder®

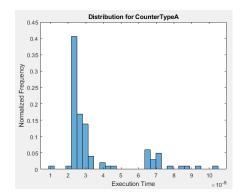


Code Coverage for Models in Software-in-the-Loop (SIL) Mode Embedded Coder®

Profile code execution time

2. Profiled Sections of Code

| Section | Maximum Execution Time in ns | Average Execution Time in ns |
|--|------------------------------------|------------------------------------|
| [+] <u>rtwdemo_sil_topmodel_initialize</u> | 80 | 80 |
| [+] rtwdemo_sil_topmodel_step [0.1 0] | 358 | 129 |



View and Compare Code Execution Times Embedded Coder®



Deploy to ROS node

Generate standalone ROS node

| Configuration Parameters: RobotController/Configuration (Active) | | | | | |
|--|--|--|--|--|--|
| ★ Commonly Used Parameters | ≡ All Parameters | | | | |
| Select: Solver Data Import/Export Diagnostics Hardware Implementation Model Referencing Simulation Target D Code Generation Simulink Coverage HDL Code Generation | Hardware board: Robot Operating System (ROS) Code Generation system target file: ert.tlc Device vendor: Generic Device details Hardware board settings Operating system options Base rate task priority: 40 | | | | |
| | Target Hardware Resources | | | | |

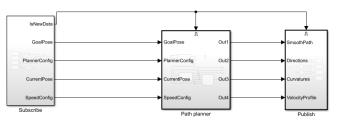
Generate a Standalone ROS Node from Simulink ROS Toolbox™

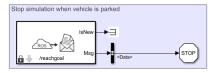
Embedded Coder®

R2019**b**

Generate ROS nodes for parking valet

Automated Parking Valet: ROS node for Path Planner





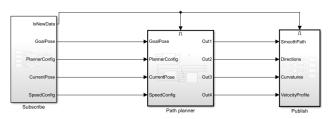
Automated Parking Valet with ROS in Simulink

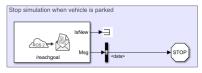
ROS Toolbox[™] Embedded Coder[®]



Generate ROS 2.0 nodes for parking valet

Automated Parking Valet: ROS 2 node for Path Planner





Automated Parking Valet with ROS 2 in Simulink ROS ToolboxTM Embedded Coder[®]



for SAE AutoDrive year 3 competition

| # | Task | Points |
|---|---|--------|
| 1 | Synthesize data to test open loop perception algorithm | 10 |
| 2 | Synthesize data to test closed loop controls algorithm | 10 |
| 3 | Generate code from controls algorithm | 10 |
| 4 | a. What did you do with MathWorks tools that <u>differentiates</u> you from other teams? (i.e. Analyze recorded ROS/CAN data, label recorded data, train deep learning network, build a custom App, share work through projects with revision control) b. What <u>insight</u> did you gain while doing this? | 15 |
| 5 | Reflect | 5 |



Connect to CAN and CAN-FD data

CAN platform support

- Connect to live data
- Read logged data

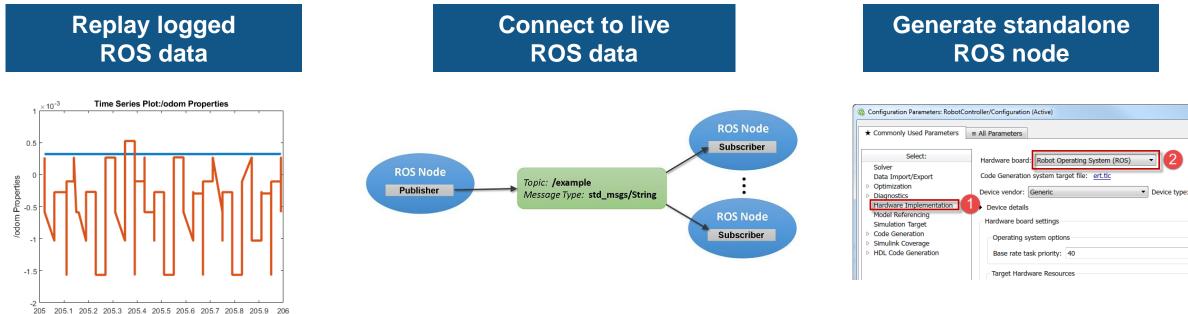
Vehicle Network Toolbox[™] Updated **R2019b**

| Vendor | Windows® | Linux® |
|---|--------------|--------------|
| MathWorks [®] virtual channels | \checkmark | \checkmark |
| Vector | \checkmark | |
| PEAK-System | \checkmark | \checkmark |
| Kvaser | \checkmark | \checkmark |
| National Instruments® | \checkmark | |

| File Format | Windows | Linux |
|-------------|--------------|--------------|
| BLF | \checkmark | |
| CDF | \checkmark | \checkmark |
| MDF | \checkmark | |



Integrate with ROS 1.0 and ROS 2.0



05.1 205.2 205.3 205.4 205.5 205.6 205.7 205.8 Time (seconds)

Work with rosbag Logfiles ROS ToolboxTM

R2019b

Exchange Data with ROS Publishers and Subscribers ROS Toolbox[™]



Generate a Standalone ROS 2 Node from Simulink ROS Toolbox™

Simulink Coder™



Read point cloud from Velodyne log file

fillo

MathWorks[®]

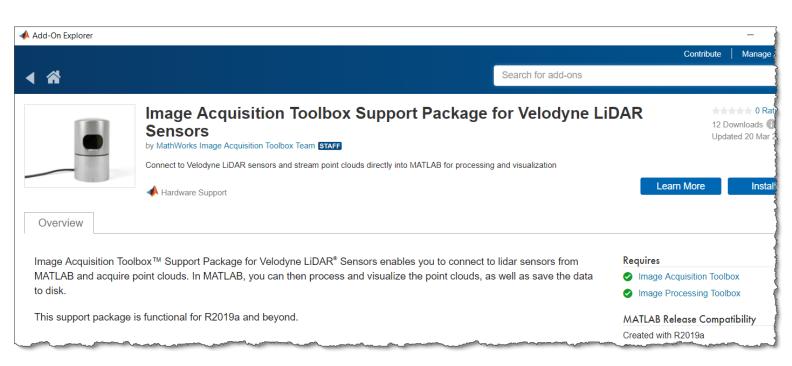
- Read point cloud data from Velodyne packet capture (PCAP) file
- The reader supports the following Velodyne LiDAR models:
 - VLP-16, Puck LITE, Puck Hi-Res, VLP-32C, HDL-32E, and HDL-64E
 - VLS-128 support package is available per request
- User can provide device specific calibration XML file

veloReader = velodyneFileReader(fileName,deviceModel,'CalibrationFile',calibFile);



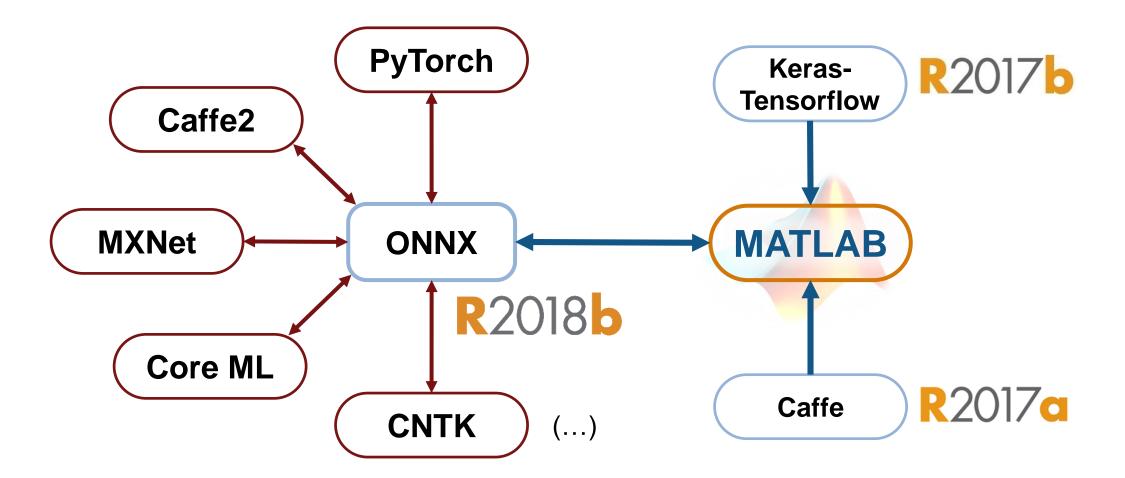
Stream live Velodyne point cloud data

- You can connect to and stream point clouds from the following Velodyne LiDAR models:
 - HDL-32E sensor
 - VLP-32C Ultra Puck sensor
 - VLP-16 Puck sensor
 - VLP-16 Puck Lite sensor
 - VLP-16 Puck Hi-Res sensor





Interoperate with neural network frameworks



Open Neural Network Exchange



Design vision perception systems

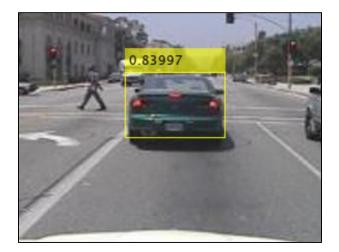
Label recorded data



Get Started with the Ground <u>Truth Labeler</u> Automated Driving ToolboxTM Computer Vision ToolboxTM



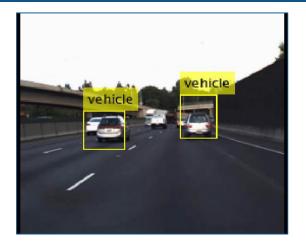
Train deep learning networks



Object Detection Using YOLO v2 Deep Learning Computer Vision ToolboxTM Deep Learning ToolboxTM



Generate code



Generate C++ Code for Object Detection Using YOLO v2 and Intel MKL-DNN Deep Learning Toolbox[™] MATLAB Coder

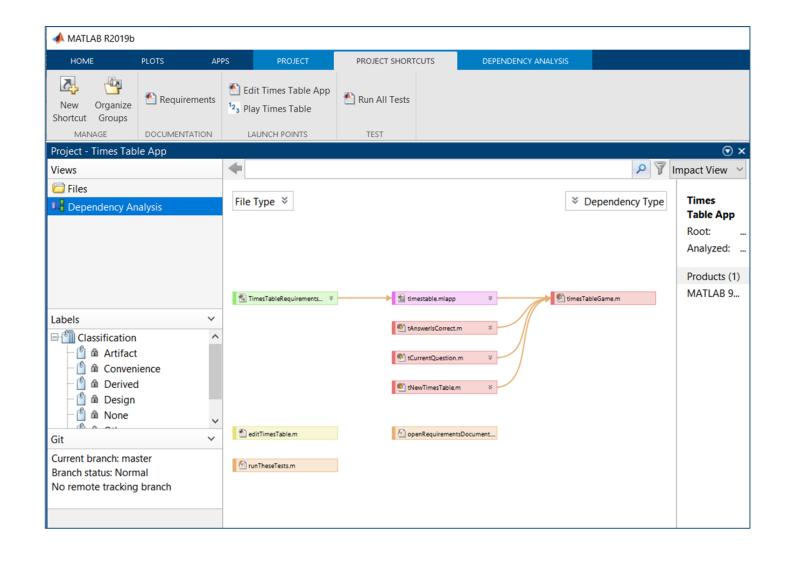
R2019a

Organize your work and collaborate with projects

Explore an example project

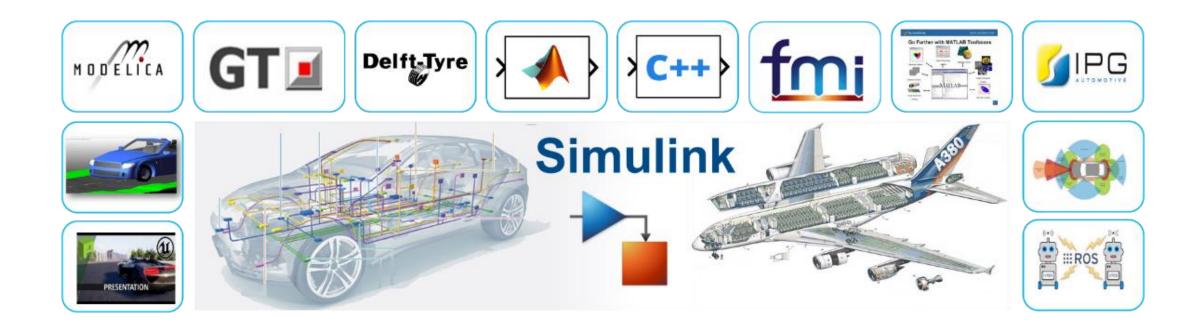
- View, search, and sort files
- Run frequently used files
- Integrate with source control
- Review changes
- Analyze dependencies
- Commit modified files

MATLAB® **R2019**

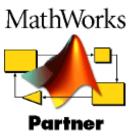




Connect to third party tools



152 Interfaces to 3rd Party Modeling and Simulation Tools (as of March 2019)





for SAE AutoDrive year 3 competition

| # | Task | Points |
|---|--|--------|
| 1 | Synthesize data to test open loop perception algorithm | 10 |
| 2 | Synthesize data to test closed loop controls algorithm | 10 |
| 3 | Generate code from controls algorithm | 10 |
| 4 | Innovate | 15 |
| 5 | Reflect a. Would you do something <u>different next time</u>? b. Is there anything <u>missing</u> from the tools that would have helped you? | 5 |



for SAE AutoDrive year 3 competition

| # | Task | Points |
|---|--|--------|
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| 3 | Generate code from controls algorithm | 10 |
| 4 | Innovate | 15 |
| 5 | Reflect | 5 |

Additional clarification of tasks and scoring will be provided at November training