

## A. Papers

This is a list of papers that tackle the DL model and attacks and defenses against self-driving cars. A summary of only a part of them can be found here, in the Autonomous Cars folder: <https://github.abudhabi.nyu.edu/ha59/HK-CyberSecLab/tree/master/AutonomousCarSecurity>. If you don't have access to the repo, kindly ask the professor to give you access.

### Attacks and Defenses

- Vrizlynn et al. [Autonomous Vehicle Security: A Taxonomy of Attacks and Defences](#)
- Papernot et al. [The Limitations of Deep Learning in Adversarial Settings](#)
- Papernot et al. [Distillation as a Defense to Adversarial Perturbations against Deep Neural Networks](#)
- Papernot et al. [Practical Black-Box Attacks against Machine Learning](#)
- Eykholt et al. [Robust Physical-World Attacks on Deep Learning Visual Classification](#)
- Petit et al. [Potential Cyberattacks on Automated Vehicles](#)
- Petit et al. [Remote Attacks on Automated Vehicles Sensors: Experiments on Camera and LiDAR](#)
- Szegedy et al. [Intriguing properties of neural networks](#)
- Goodfellow et al. [Explaining and Harnessing Adversarial Examples](#)
- Papernot et al. [Transferability in Machine Learning: from Phenomena to Black-Box Attacks using Adversarial Samples](#)
- Grosse et al. [Adversarial Perturbations Against Deep Neural Networks for Malware Classification](#)
- [Adversarial Sample code here](#); Nguyen et al. [Deep Neural Networks are Easily Fooled: High Confidence Predictions for Unrecognizable Images](#)
- Xu et al. [Automatically Evading Classifiers: A Case Study on PDF Malware Classifiers](#)
- Kantchelian et al. [Evasion and Hardening of Tree Ensemble Classifiers](#)
- Biggio et al. [Support Vector Machines Under Adversarial Label Noise](#)
- Biggio et al. [Poisoning Attacks against Support Vector Machines](#)
- Ororbia II et al. [Unifying Adversarial Training Algorithms with Flexible Deep Data Gradient Regularization](#)
- Jin et al. [Robust Convolutional Neural Networks under Adversarial Noise](#)
- Goodfellow et al. [Deep Learning Adversarial Examples – Clarifying Misconceptions](#)
- \*Tian et al. [DeepTest: Automated Testing of Deep-Neural-Network-driven Autonomous Cars](#) - <https://deeplearningtest.github.io/deepTest/>
- \*Rauber et al. Foolbox: [A Python toolbox to benchmark the robustness of machine learning models](#) - full documentation: <https://media.readthedocs.org/pdf/foolbox/stable/foolbox.pdf>
- Su, J. et al. [One pixel attack for fooling deep neural networks](#)

## Other Useful Links

- <https://medium.freecodecamp.org/hacking-cars-a-guide-tutorial-on-how-to-hack-a-car-5eafcfbbb7ec>
- <http://blog.davidsingleton.org/nnrccar/>
- <https://github.com/udacity/self-driving-car/tree/master/datasets>
- <https://www.linkedin.com/pulse/teaching-car-how-drive-using-deep-learning-muhieddine-el-kaissi/>
- <https://nicolovaligi.com/reading-list-udacity-self-driving-challenge-3.html>
- NIPS: <https://nips.cc/Conferences/2018/CompetitionTrack>

## Capstone Project Code

- [https://www.youtube.com/redirect?q=https%3A%2F%2Fgithub.com%2FKairos-Automotive%2Fcarla-brain&redir\\_token=kPP2arRKn7FmCATm-OKHeMCxO6d8MTUyODE3Nzc5MkAxNTI4MDkxMzky&v=956Q7wU0-IE&event=video\\_description](https://www.youtube.com/redirect?q=https%3A%2F%2Fgithub.com%2FKairos-Automotive%2Fcarla-brain&redir_token=kPP2arRKn7FmCATm-OKHeMCxO6d8MTUyODE3Nzc5MkAxNTI4MDkxMzky&v=956Q7wU0-IE&event=video_description)

Interesting resources to look into: **driving assistants** – MobilEye C2-270, ibo LUX 3

## B. Simulators

### 1. Udacity

This is the first simulator I would recommend you to start with, as it is fairly simple, as you will see.

No operating system preferred.

#### Installation

- Github repository: <https://github.com/udacity/self-driving-car-sim>
- Link to a repository that explains the installation steps: [https://github.com/IIISourcell/How\\_to\\_simulate\\_a\\_self\\_driving\\_car](https://github.com/IIISourcell/How_to_simulate_a_self_driving_car)
- Installation video: <https://www.youtube.com/watch?v=EaY5QiZwSP4&index=3&list=PLSRMSHOzuuvAJJEx49s7-lzR7L0fOGHSv&t=0s>
- More resources on how the deep learning model works: <https://github.com/naokishibuya/car-behavioral-cloning>
- Paper for the NVIDIA model (that the simulator is based on): <https://arxiv.org/pdf/1604.07316v1.pdf>

### 2. Apollo Auto

You need Linux to run this simulator.

Although it is more complex than Udacity, I didn't find it particularly useful. However, the GPS architecture is worth looking into.

Official website: <http://apollo.auto>

#### Installation

- Github repository: <https://github.com/ApolloAuto/apollo>

### 3. Carla

I recommend Linux to run this simulator.

By far, the most complex simulator, it looks very much like a video game that has both training and autonomous modules.

#### Installation

- Github repository: <https://github.com/carla-simulator/carla>
- How to build on Linux: [http://carla.readthedocs.io/en/latest/how\\_to\\_build\\_on\\_linux/](http://carla.readthedocs.io/en/latest/how_to_build_on_linux/)
- How to build on Windows:  
[http://carla.readthedocs.io/en/latest/how\\_to\\_build\\_on\\_windows/](http://carla.readthedocs.io/en/latest/how_to_build_on_windows/)
- Paper: <http://carla.org>  
<http://proceedings.mlr.press/v78/dosovitskiy17a/dosovitskiy17a.pdf>
- To run the autonomous mode: first run PythonClient and then run the server

### C. Attack Models

- Camera spoofing
- GPS spoofing

### D. Courses

- Coursera, Machine Learning: <https://www.coursera.org/learn/machine-learning> (it has just started on the 11th, it's a good opportunity for you to enroll now)
- Google, Deep Learning <https://ae.udacity.com/course/deep-learning--ud730>
- MIT, Autonomous Cars course: <http://selfdrivingcars.mit.edu> (very good class that I recommend)  
Berkeley: <https://deepdrive.berkeley.edu/node/107>

### E. Presentations & Reports

If you don't have access to the repo, ask the professor to grant you access.

- <https://github.abudhabi.nyu.edu/ha59/HK-CyberSecLab/tree/master/AutonomousCarSecurity>