

# **FACT SHEET**

# 2018 Research **Symposium**

## **Traffic Signs Detection & Geospatial Localization**

## & STIC Annual Meeting

#### RESEARCH PROJECT TITLE

Traffic Signs Detection & Geospatial Localization

#### STUDY TIMELINE

January 2018 - June 2019

#### **INVESTIGATORS**

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#### **VTRANS CONTACTS**

Stephen Smith, Data Administration Director

#### MORE INFORMATION

VTrans Research will add link to the final report and other materials on VTrans website

This fact sheet was prepared for the 2018 VTrans Research and Innovation Symposium & STIC Annual Meeting held at the State House in Montpelier, VT, on September 12, 2018 from 8:00 am- 1:00 pm.

Fact sheets can be found for additional projects featured at the 2018 Symposium at

http://vtrans.vermont.gov/planning/res earch/2018symposium

Additional information about the VTrans Research Program can be found at

http://vtrans.vermont.gov/planning/res earch

Additional information about the VTrans STIC Program can be found at <a href="http://vtrans.vermont.gov/boards-">http://vtrans.vermont.gov/boards-</a> councils/stic

#### Introduction

This research aims to develop a deepLearning-based system to process a stream of road-images in order to detect and classify traffic signs, then localize them on a map by estimating their GPS coordinates. Furthermore, we introduce a new and

E1-1: 0.97 Predictions **Ground-Truth** 

large-scale dataset to serve as one of the very few benchmarks for US.

traffic signs recognition (TSR).

Figure 1: The white color indicates predictions by our network, and the green color refers to the manually labeled ground-truth.

### Methodology

We are investigating some of most novel deep learning algorithms and computer

vision techniques for object detection/classification and geospatial localization to find the most suitable approach for our application. In terms of dataset creation, we adopted an open-source tool called labellmg to serve as an easy, yet crossplatform tool for data annotation. Furthermore, our baseline-system uses a heavily modified version of the state-of-the-art object-detector known as RetinaNet.

#### Results

Our dataset covers 294 different types of traffic signs containing more than 51,000 images and over 27,000 annotations. Our baseline-system, on the other hand, which was tested on a subset of the dataset (mainly highway images captured in Vermont) achieved 66% mean average precision (mAP) in terms of detection and classification of 51 different labels (sign-types). More importantly, the system scored an average of (±) 5.24 meters' geospatial margin of error, which was very surprising considering that depth perception is certainly not an easy task when applied on single-vision road images without any sensory information.

### **Potential Impacts**

First and foremost, our research will help to accurately identify and localize road assets in order to make better assessments/maintenance plans. Additionally, we plan on reaching approximately 40K annotations by the end of our research introducing the largest dataset for TSR, and first ever dataset to feature objectrelated GPS information to support future research in this field.