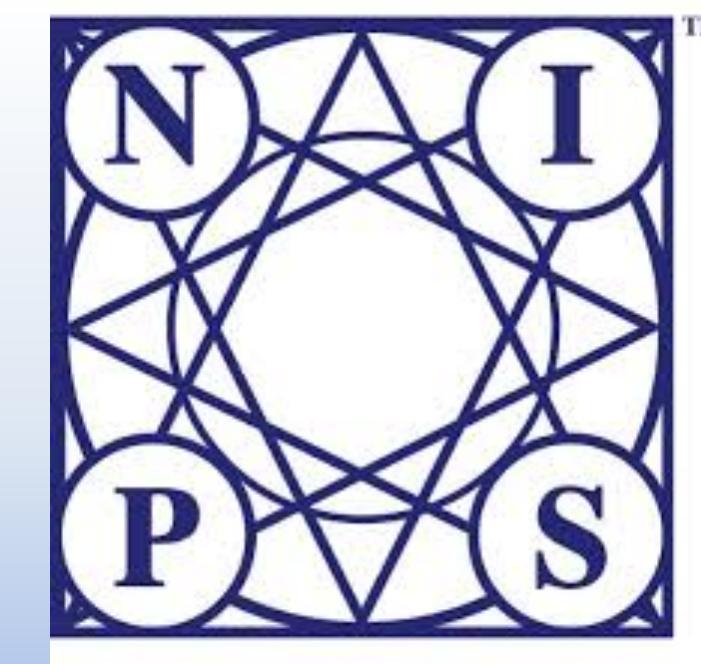


# CURE-TSR : Challenging Real and Unreal Environments for Traffic Sign Recognition



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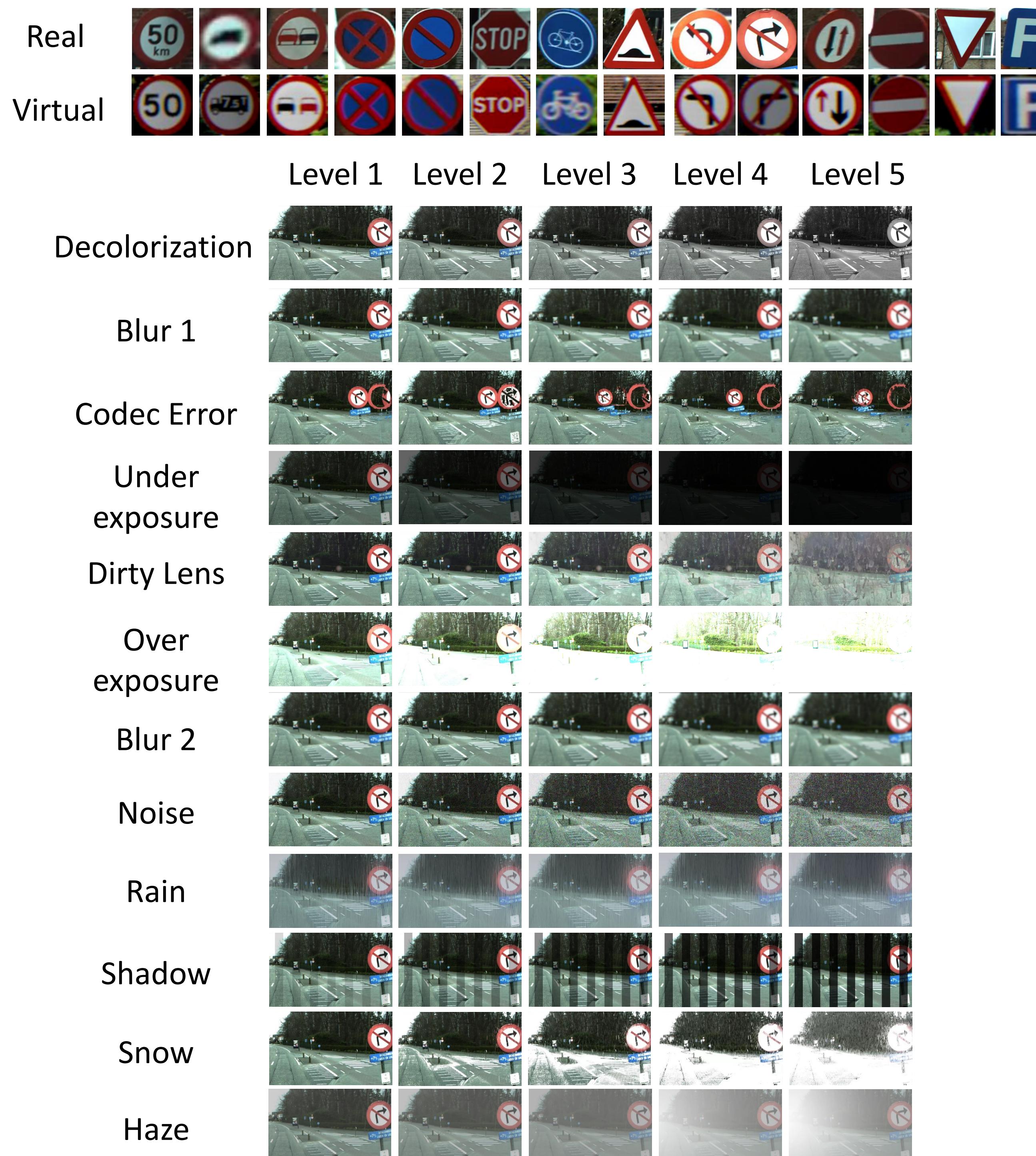


## Introduction

- In this paper, we investigate the **robustness** of traffic sign recognition algorithms under **challenging conditions**.
- Existing datasets are **limited** in terms of their **size and challenging condition coverage**, which motivated us to generate the Challenging Unreal and Real Environments for Traffic Sign Recognition (CURE-TSR) dataset. It includes more than **two million** traffic sign images that are based on real-world and simulator data.
- We **benchmark** the performance of existing solutions in real-world scenarios and analyze the **performance variation** with respect to **challenging conditions**.
- We show that challenging conditions can **decrease** the performance of baseline methods significantly.
- We also investigate the effect of **data augmentation** and show that utilization of virtual data along with real-world data enhances the average recognition performance.

## Dataset Generation and Visualization

- In order to create **realistic challenging scenarios**, we generate challenging condition types and levels **for entire scenes**.
- The traffic signs are then **cropped** from such scenes.
- Each row in the figure below corresponds to a challenging condition and each column corresponds to a certain level of the condition.



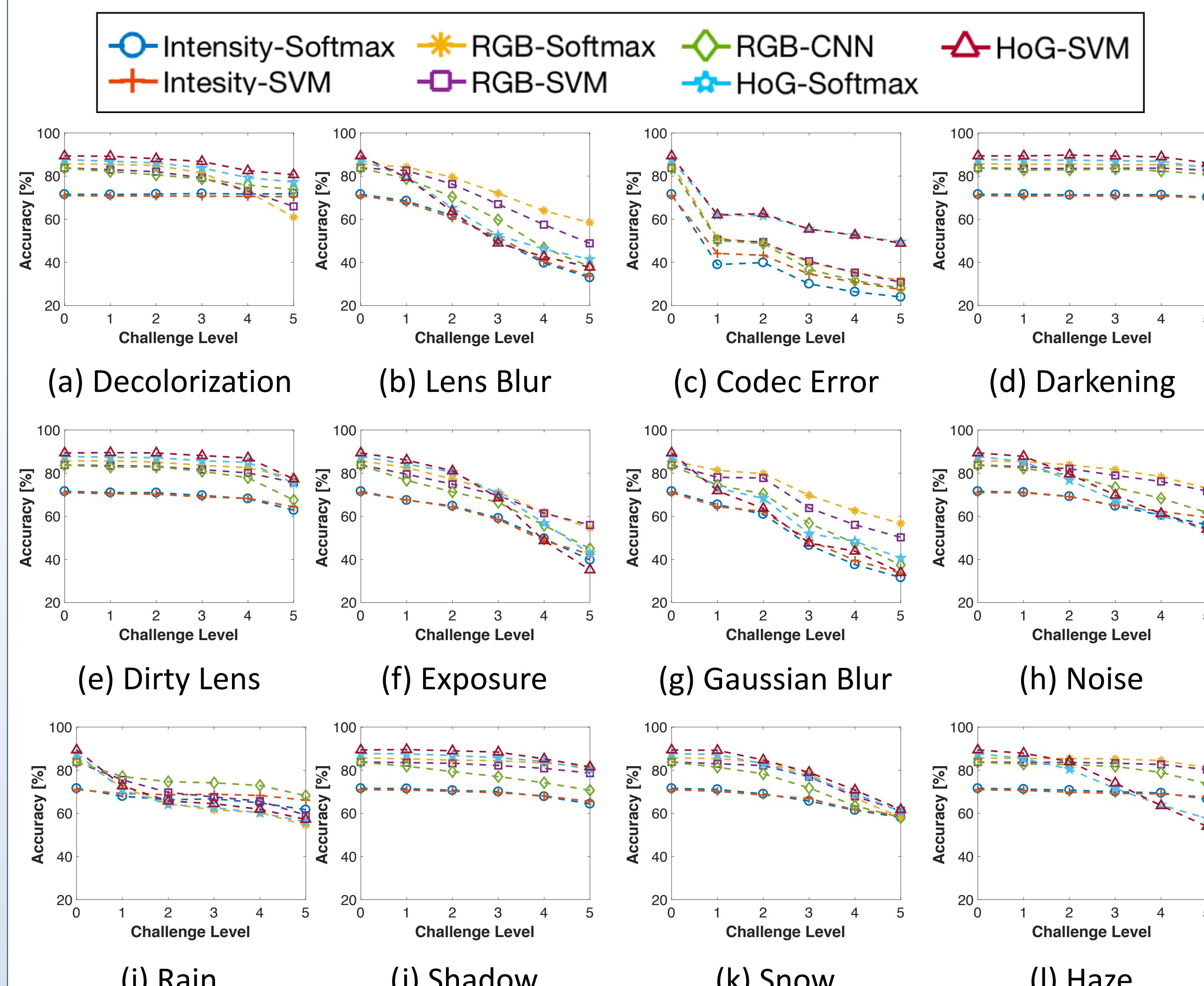
## Dataset – Number of Images

Image Types	Training	Testing
Real Challenge-Free	7,292	3,334
Real Challenge	437,520	200,040
Virtual Challenge-Free	19,610	8,210
Virtual Challenge	1,078,550	451,550

## Dataset - Usage

- To study traffic sign recognition under challenging conditions
- Studying **data augmentation** techniques when traditional augmentation techniques fail.
- Studying **regularization** techniques when data is imperfect.
- Domain adaptation between virtual and real-world images.

## Dataset - Performance Benchmarks

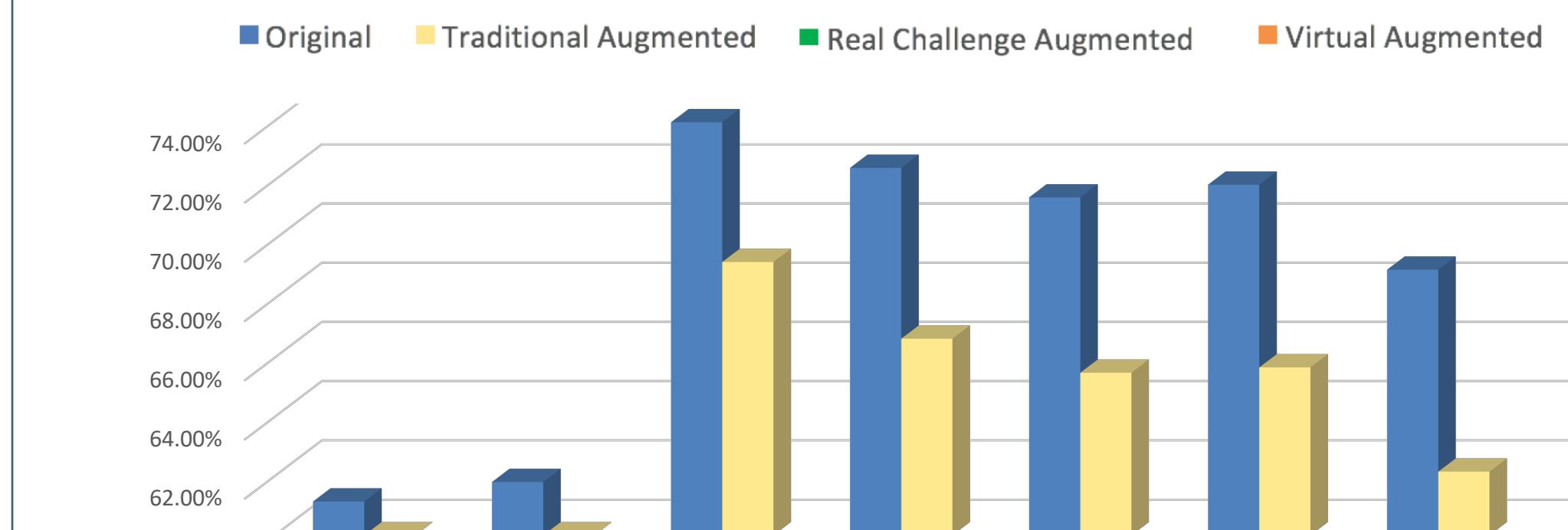


## Dataset – Performance Benchmark Observations

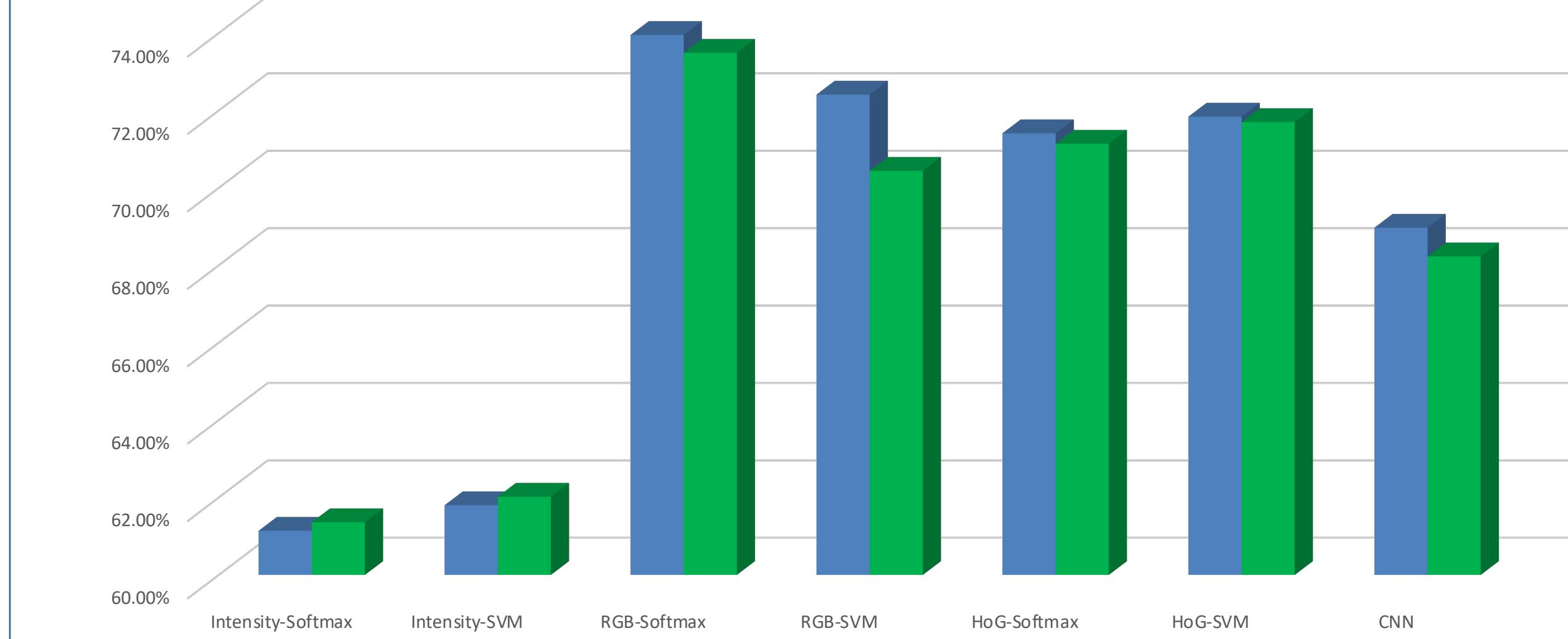
- Challenging conditions** perturb original representation space to **deceive** classifier.
- Different challenges** have their **different** and characteristic degradation **slopes**.
- Decolorization, Darkening and Shadow have relatively **consistent** performance across challenge levels and algorithms.
- Lens blur, Codec error, Exposure and Gaussian blur show **severe** performance degradation.

## Data Augmentation

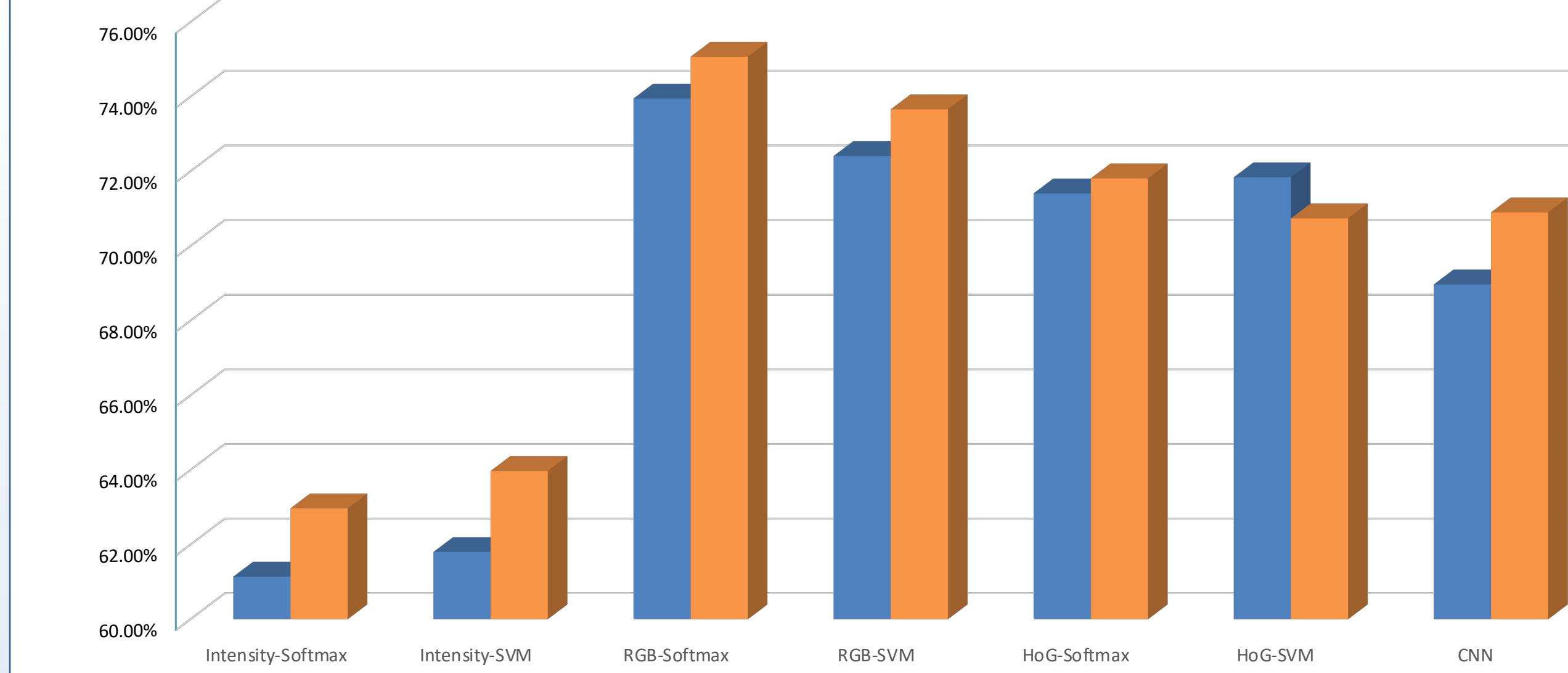
### Traditional Strategies



### Real Challenging data



### Virtual data



## Contributions

- We introduce the most **comprehensive** publicly-available traffic sign recognition dataset with **controlled challenging conditions**.
- We provide a **detailed analysis** of the benchmarked algorithms in terms of their recognition performance under challenging conditions thereby **identifying the vulnerabilities** of algorithms.
- We provide images that originate from captured sequences as well as synthesized sequences, that lead to a better understanding of the relationship between real-world and virtual data in terms of algorithmic performance. This understanding can be utilized to **generate algorithmically invariant virtual datasets** and minimize the need for real-world data collection that require significant resources.
- We use diverse **data augmentation** methods and show that utilization of **limited virtual images** along with real-world data can enhance the recognition performance even when the domain difference is not addressed.