

## We Need Robust Neural Networks

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### 1. Background: Object Detection and Convolutional Neural Networks (CNN)

Coordinates:  $x, y, w, h$   
Class:  $S_c$

- Our work focuses on understanding and improving the stability of deep convolutional object detection algorithms. Object detection algorithms have two roles (1) localizing objects and (2) classifying objects.

### 2. Problem: Object Detectors are Inconsistent

- Each pair of images are one frame apart
- Images in the left column are correctly detected
- Images in the right column have incorrect detections

\*Detections made by YOLOv3

### 3. Explaining Why: Measuring the Sensitivity of Learned Features

Average activations  $A_i^j$  at layer  $j$  of image  $i$

Sensitivity =

- Small scale tests have shown that particular features learned by the model are hyper-sensitive to small input changes. This seems to be correlated with detection stability.

### 4. Security Implications: Adversarial Examples are Created to Attack Models



- An adversarial attack (glasses) causes the model to misclassify the lady on the left as the man on the right. (Sharif et al, 2016)



- The stop sign on the left is correctly detected. The stop sign on the right is incorrectly detected as a 45 mph speed limit sign. (Eykholt et al, 2018)

### 5. Solution: How to Improve Networks to be More Robust

- Early results show certain features our model learned are sensitive (left)
- We can learn smooth feature representations of the input space (right) with gradient regularization
- Sensitive Function
- Similar but less sensitive function

### 6. Looking Ahead: Research Questions We Must Address

\*Faster-RCNN

\*YOLOv3

- 1. Are two-stage detectors (left) just as inconsistent as one-stage detectors (right)? Why?
- 2. What types of learned features (left) are most sensitive? Color, Texture, Shapes?