

# Assessing image data quality for a vehicle damage detection algorithm

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# Our Goal

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*Vehicle Damage  
detection model  $M$*

*Raw Image Data  $Ds$*

$I_M : \{Image \mid Image \in$   
 $True\ positives \cup$   
 $True\ negatives\}$

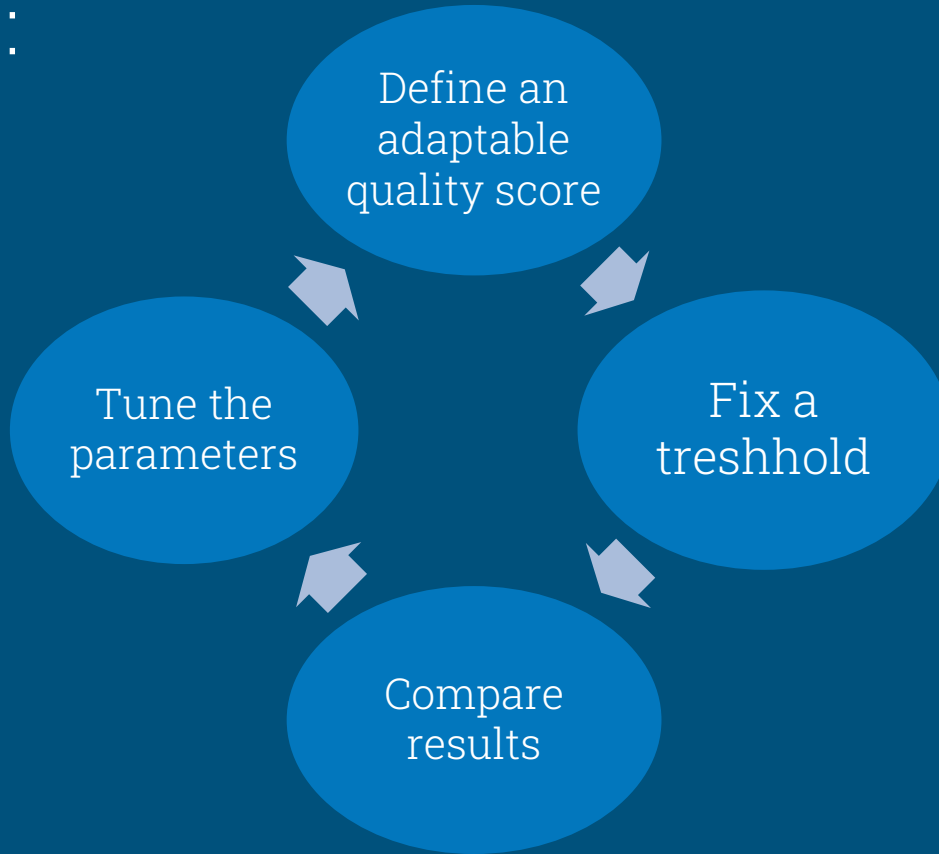
*Our Model  $M'$*

*Raw Image Data  $Ds$*

*Predicting  $Ds - I_M$*

# Approach :

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# An Object detection solution

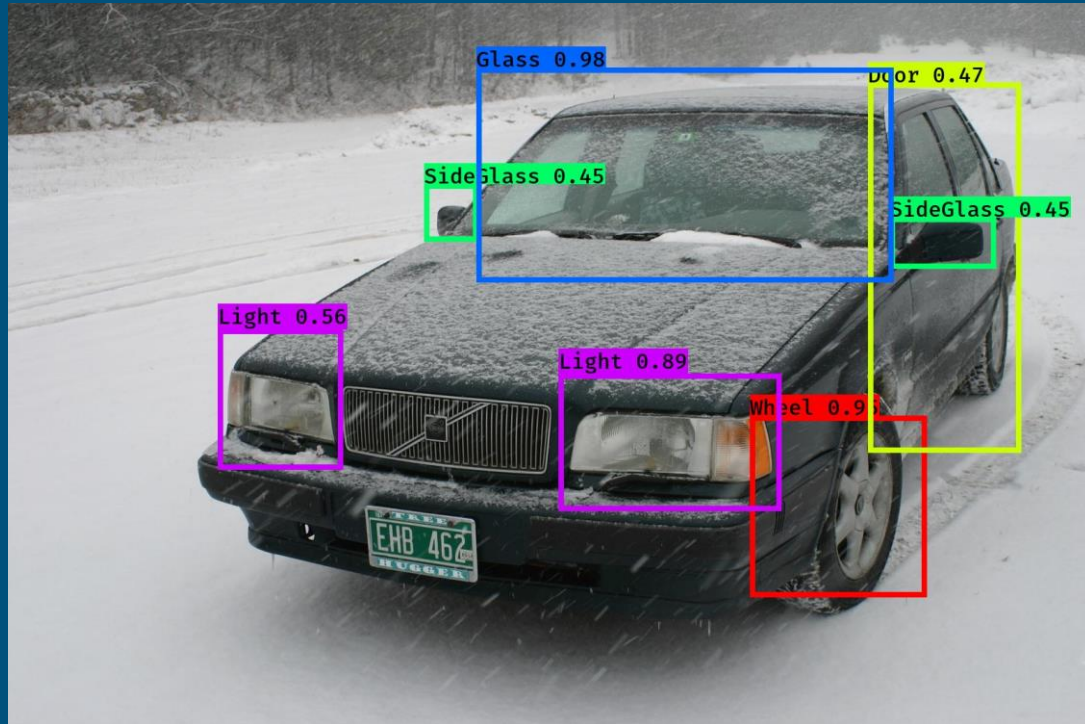
Sliding window  
object detection

R CNN

Fast R CNN

Faster R CNN

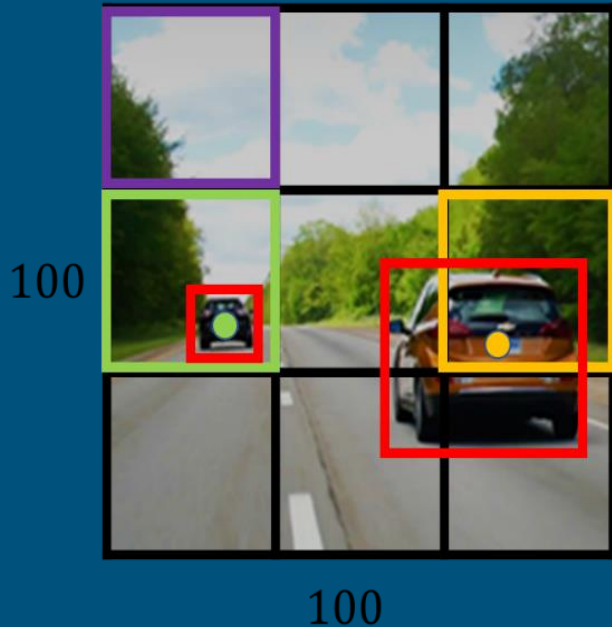
Yolo



Vehicle parts detection solution using YOLOv5

# Data : Training a custom yolo model

Labels for training for  
each grid cell:



$$y = \begin{bmatrix} p_c \\ b_x \\ b_y \\ b_h \\ b_w \\ c_1 \\ c_2 \\ c_3 \end{bmatrix}$$

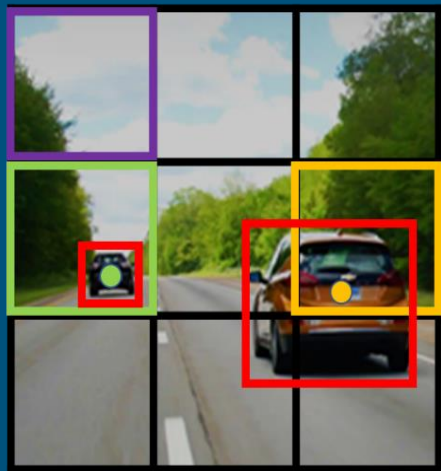
$$\begin{bmatrix} 0 \\ ? \\ ? \\ ? \\ ? \\ ? \\ ? \\ ? \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ b_x \\ b_y \\ b_h \\ b_w \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ b_x \\ b_y \\ b_h \\ b_w \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

→ bounding box

# A confidence vector



*$S \times S$  grid, each grid cell predicts  $B$  bounding boxes and  $N$  conditional class probabilities.*

$$Pr(Class\ i|Object)*Pr(Object)*IoU = Pr(Class\ i)*IoU.$$

$$\textbf{Confidence} = \textbf{Pr(object)} * \textbf{IoU}$$

*IoU: Intersection over Union between the predicted box and the ground truth.*

*The final predictions are encoded as an  $S \times S \times (B*5 + N)$  tensor*

*If no object exists in a cell, its confidence score should be zero.*

# A first criterion of data quality

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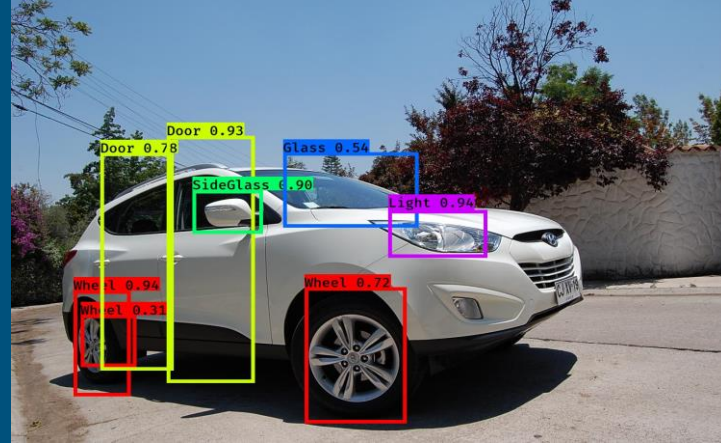
*The confidence threshold allowed us to extract two different relevance classes from our dataset*

***Relevant Data<sub>cy</sub>*** :  $\{\text{Image } I \mid \min(C_i(I)) > \text{Threshold}_1\}$

***Irrelevant Data<sub>cy</sub>*** :  $\{\text{Image } I \mid \max(C_i(I)) < \text{Threshold}_1\}$

*$C_i(I)$  is the confidence score of class  $i$  for an image  $I$*

- We trained the yolov5 object detection model to recognize 5 classes of vehicle parts. We collected images of vehicles of different types and states using web scraping, then we set a confidence score threshold  $Threshold_1 = 0.30$ .



Initial image: 1504 x 1000  
Horizontal resolution: 300 dpi  
Vertical resolution: 300 dpi  
Color depth: 24

$$\min(C_i(I)) > Threshold_1$$



# Example

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- The image on the right belongs to the irrelevant dataset, with  $\max(C_i(I)) = 0.19 < Threshold_1 = 0.30$
- Based on our Model, this image is irrelevant. We will predict that the image is non usable by the model M.

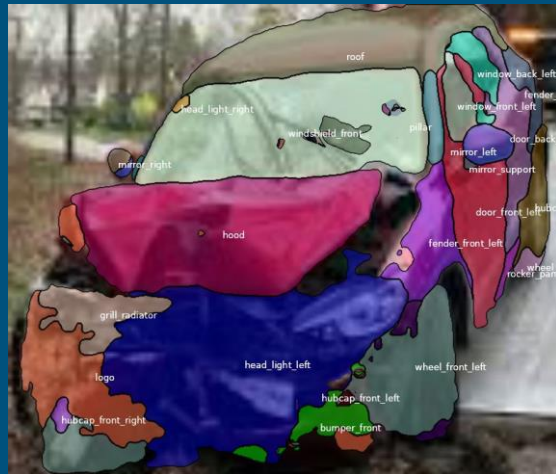


# Example

We were able to test this prediction with the demo available on the company's website.



Initial image: 130 x 111  
Horizontal resolution: 96 dpi  
Vertical resolution: 96 dpi  
Color depth: 24



Monk's website demo : damage  
detection results

# Points to explore

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- Learning more about the vehicle damage detection algorithm
- Using the demo to test image quality scores.
- Effects of inter-class variations: camera position/ sensors, light, internal parameters.
- Image preprocessing pipelines.
- A quality score to predict false predictions ?