

# DET RF100 circuit-elements

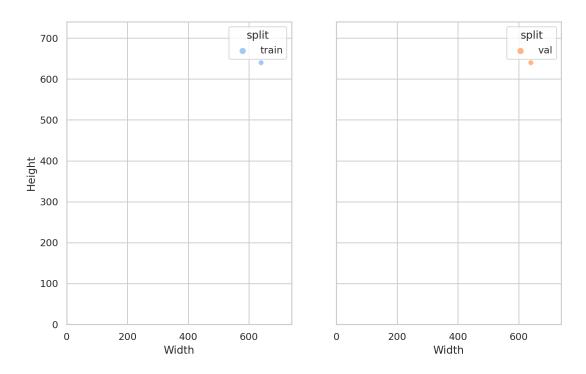
07:13 July 25, 2023

# 1. Selected features

## 1.1. General Statistics

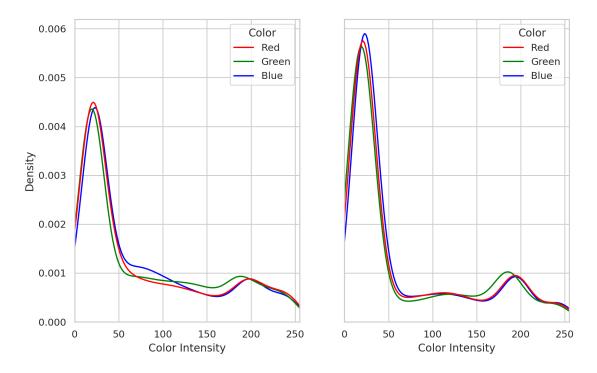
	Train	Validation
Images	672	64
Classes	32	32
Classes in use	31	22
Annotations	192008	23653
Annotations per images	285.73	369.58
Images with no annotations	0	0
Median image resolution	640x640	640x640
Smallest annotation	0	0
Largest annotation	195363	11615
Most annotations in an image	1168	894
Least annotations in an image	3	17

### 1.2. Image Width and Height Distribution



These histograms depict the distributions of image height and width. It's important to note that if certain images have been rescaled or padded, the histograms will represent the size after these operations.

#### 1.3. Color Distribution

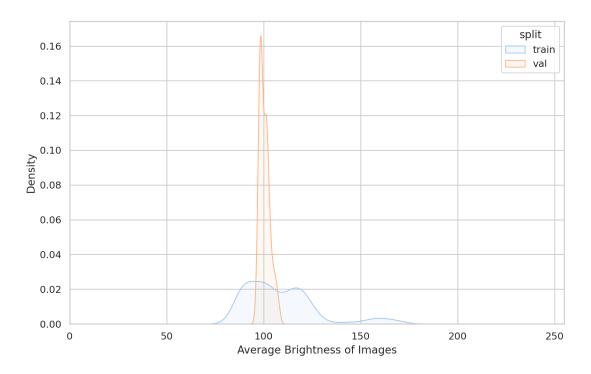


Here's a comparison of RGB or grayscale intensity intensity (0-255) distributions across the entire dataset, assuming RGB channel ordering.

It can reveal discrepancies in the image characteristics between the two datasets, as well as potential flaws in the augmentation process.

E.g., a notable difference in the mean value of a specific color between the two datasets may indicate an issue with the augmentation process.

#### 1.4. Image Brightness Distribution



This graph shows the distribution of the brightness levels across all images.

This may for instance uncover differences between the training and validation sets, such as the presence of exclusively daytime images in the training set and nighttime images in the validation set.

#### 1.5. Image Duplicates

Train duplicated images:

There are 5 duplicated images appearing 10 times across the dataset.

#### Validation duplicated images:

There are 13 duplicated images appearing 26 times across the dataset.

There are 0 duplicates between train and validation.

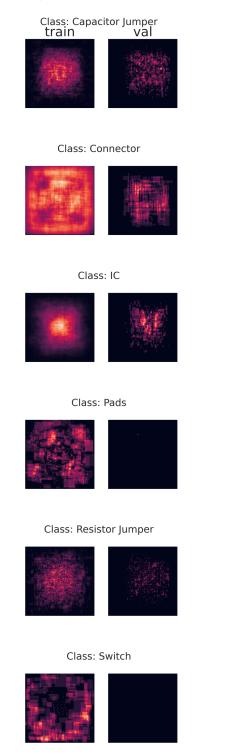
#### 1.6. Visualization of Samples

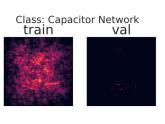


The sample visualization feature provides a visual representation of images and labels. This visualization aids in understanding of the composition of the dataset.

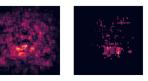
**Notice:** Only 12 random samples are shown. You can increase the number of images by changing `n\_cols` and `n\_rows` in the configuration file.

#### 1.7. Bounding Box Density

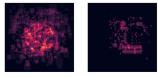




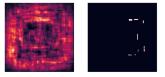
#### Class: Electrolytic Capacitor



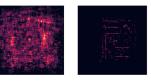
Class: Inductor



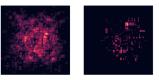
#### Class: Pins



#### Class: Resistor Network



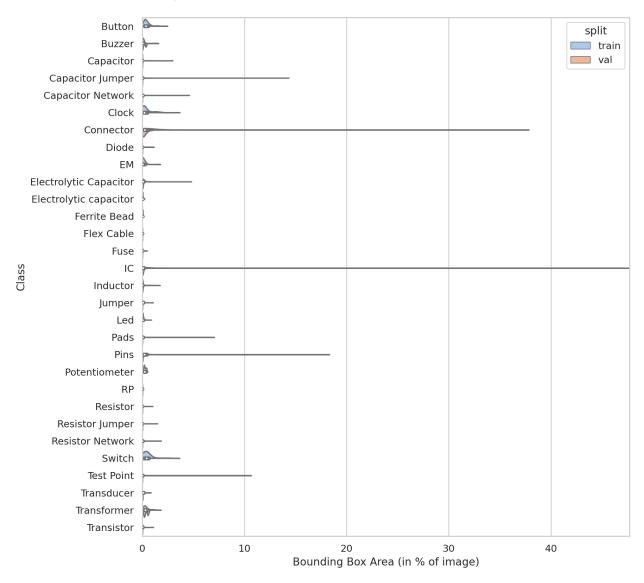
#### Class: Transistor



The heatmap represents areas of high object density within the images, providing insights into the spatial distribution of objects. By examining the heatmap, you can quickly detect whether objects are predominantly concentrated in specific regions or if they are evenly distributed throughout the scene. This information can serve as a heuristic to assess if the objects are positioned appropriately within the expected areas of interest.



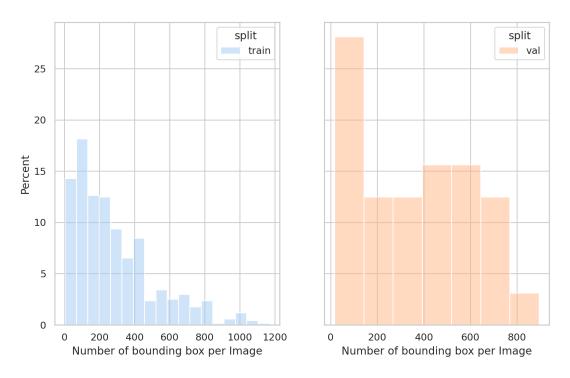
#### 1.8. Distribution of Bounding Box Area



This graph shows the frequency of each class's appearance in the dataset. This can highlight distribution gap in object size between the training and validation splits, which can harm the model's performance.

Another thing to keep in mind is that having too many very small objects may indicate that your are downsizing your original image to a low resolution that is not appropriate for your objects.

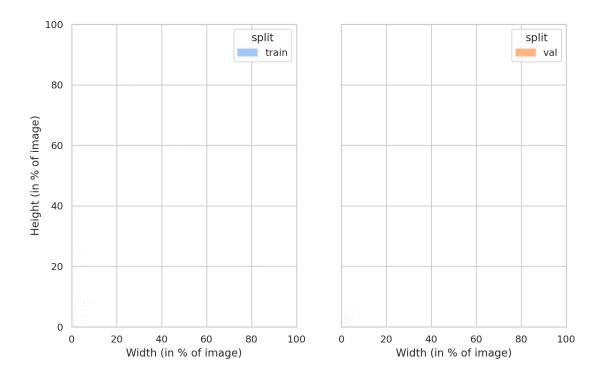
#### 1.9. Distribution of Bounding Box per image



These graphs shows how many bounding boxes appear in images.

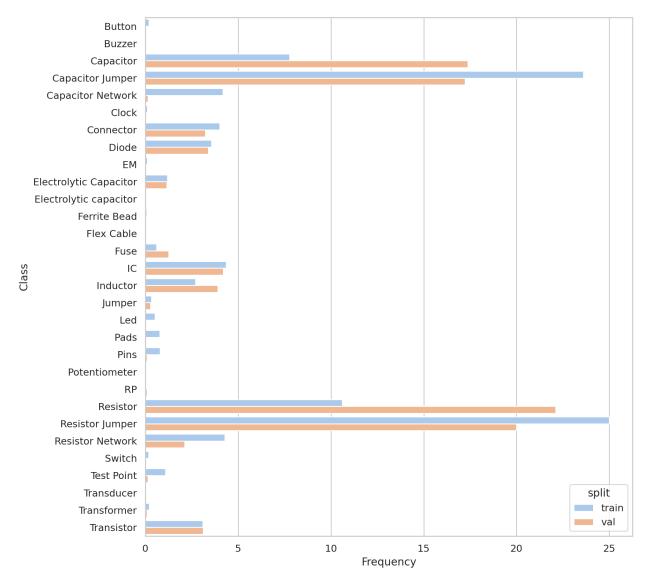
This can typically be valuable to know when you observe a very high number of bounding boxes per image, as some models include a parameter to filter the top k results.

### 1.10. Distribution of Bounding Box Width and Height

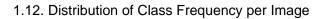


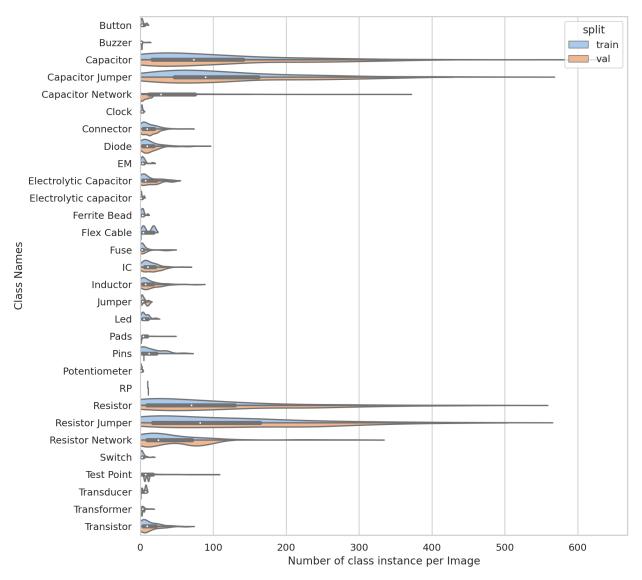
These heat maps illustrate the distribution of bounding box width and height per class. Large variations in object size can affect the model's ability to accurately recognize objects.

#### 1.11. Class Frequency



Frequency of appearance of each class. This may highlight class distribution gap between training and validation splits. For instance, if one of the class only appears in the validation set, you know in advance that your model won't be able to learn to predict that class.





This graph shows how many times each class appears in an image. It highlights whether each class has a constant number of appearances per image, or whether there is variability in the number of appearances from image to image.

#### 1.13. Intersection of Bounding Boxes

	train	val
Button		Button 0 0 0 0 0 0 0 0 0 0 0
Buzzer	0 0 0 0 0 0 0 0 0	Buzzer 0025 0 0 0 0 0 0 0 0 0
Capacitor	0032178431000	Capacitor 00322412753210
Capacitor Jumper	00362010650000	Capacitor Jumper 0042 26 14 7 5 1 1 0 0
Capacitor Network	0023710000000	Capacitor Network 008333333333333333333333
Clock	0070000000000	Clock 0 0 0 0 0 0 0 0 0 0 0
Connector	005210000000	Connector 00 3 1 1 1 1 1 0 0 0
Diode	003311 4 2 1 0 0 0 0	Diode 00361610655210
EM	004000000000	EM 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
Electrolytic Capacitor	007100000000	Electrolytic Capacitor 007 1 1 1 0 0 0 0 0
Electrolytic capacitor	002500000000000	Electrolytic capacitor 0 0 0 0 0 0 0 0 0 0 0
Ferrite Bead	0032 4 0 0 0 0 0 0 0	Ferrite Bead         0 <t< td=""></t<>
Flex Cable	007350149900000	Flex Cable         0
Fuse	0036145320000	Fuse 0048 34 12 0 0 0 0 0 0
IC	0030000000000	IC 00 5 2 2 2 2 2 2 0 0 - 60
ගී Inductor	0031147430000	හ Inductor 0022145542000
S Inductor	00321120000000	S         Inductor         Inductor <thinductor< th=""> <thinductor< th=""> <thinduc< td=""></thinduc<></thinductor<></thinductor<>
Led	0012100000000	Led 0 0 0 0 0 0 0 0 0 0 0 - 40
Pads	0019720000000	Pads 0 0 0 0 0 0 0 0 0 0 0
Pins	0090000000000	Pins 00 0 0 0 0 0 0 0 0 0 0
Potentiometer	0000000000000	Potentiometer 0 0 0 0 0 0 0 0 0 0 0 0 0 0 - 20
RP	0040 0 0 0 0 0 0 0 0	RP 0000 0 0 0 0 0 0 0 0 0 0 0
Resistor	003014 6 3 2 0 0 0 0	Resistor 0054321812932100
Resistor Jumper		Resistor Jumper 10067 44 28 15 10 2 1 0 0
Resistor Network	0024620000000	Resistor Network 0019 1 0 0 0 0 0 0 0 0
Switch	<b>100 19 3 1 0 0 0 0 0 0</b>	Switch 0 0 0 0 0 0 0 0 0 0 0
Test Point	<b>12</b> 6 2 1 1 0 0 0 0	Test Point 00161616161600000
Transducer	001000000000	Transducer         0
Transformer		Transformer 0000 0000 0000 00000000000000000000
Transistor		Transistor 003616151413111093
Unknown Unlabeled		Unknown Unlabeled 0 0 0 0 0 0 0 0 0 0 0
All classes	<b>.00</b> 26146430000	All classes 0082314 8 6 2 1 0 0
	IoU         < 0.10	0.10 0.10 0.10 0.10 0.10 0.10 0.10 0.10

The distribution of the box Intersection over Union (IoU) with respect to other boxes in the sample. The heatmap shows the percentage of boxes that overlap with IoU in range [0..T] for each class. Intersection of all boxes is considered (Regardless of classes of corresponding bboxes).

**Notice:** To better understand how to tackle the data issues highlighted in this report, explore our comprehensive <u>course</u> on analyzing computer vision datasets.