

DET RF100 valentines-chocolate

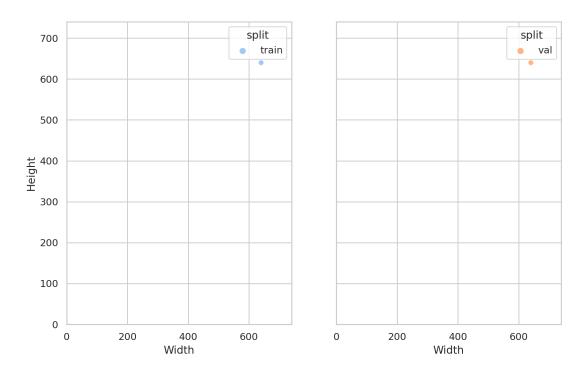
07:12 July 25, 2023

1. Selected features

1.1. General Statistics

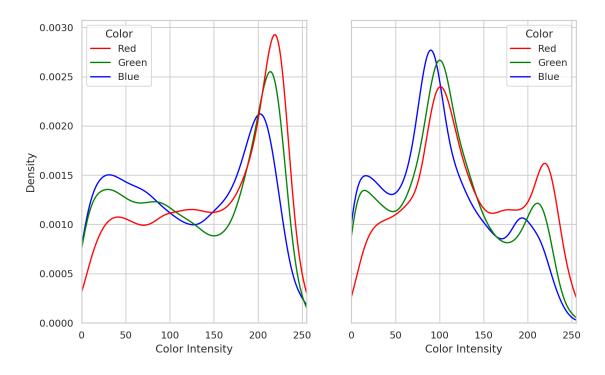
	Train	Validation	
Images	68	13	
Classes	23	23	
Classes in use	22	22	
Annotations	1296	287	
Annotations per images	19.06	22.08	
Images with no annotations	0	0	
Median image resolution	640x640	640x640	
Smallest annotation	1258	2695	
Largest annotation	27650	27588	
Most annotations in an image	25	23	
Least annotations in an image	4	16	

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

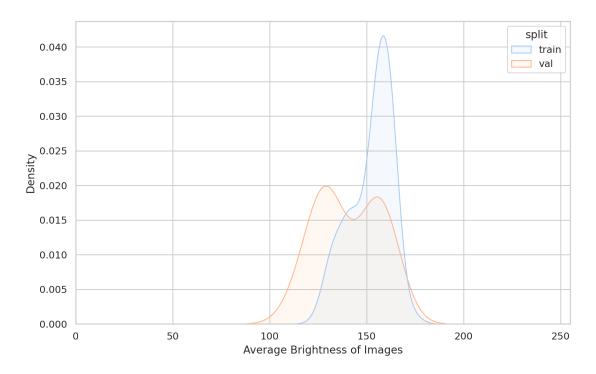


Here's a comparison of RGB or grayscale intensity intensity (0-255) distributions across the entire dataset, assuming RGB channel ordering.

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.

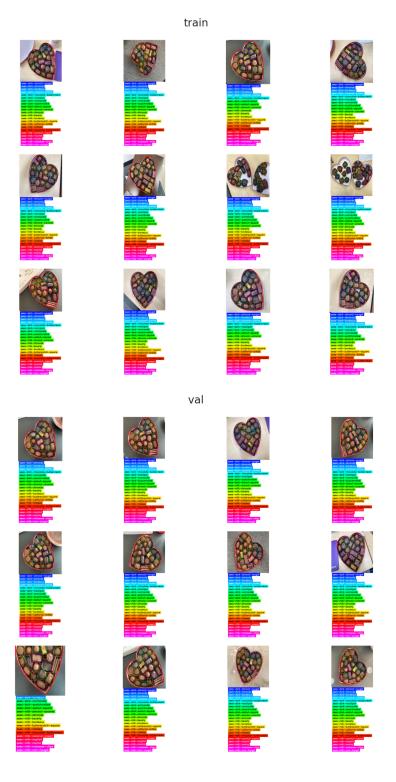
This may for instance uncover differences between the training and validation sets, such as the presence of exclusively daytime images in the training set and nighttime images in the validation set.

1.5. Image Duplicates

Train duplicated images: There are 0 duplicated images. Validation duplicated images: There are 0 duplicated images.

There are 0 duplicates between train and validation.

1.6. Visualization of Samples



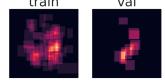
The sample visualization feature provides a visual representation of images and labels. This visualization aids in understanding of the composition of the dataset.

Notice: Only 12 random samples are shown.

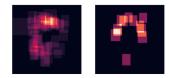
You can increase the number of images by changing `n_cols` and `n_rows` in the configuration file.

1.7. Bounding Box Density

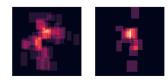
Class: sees-dark-almond-nougat train val



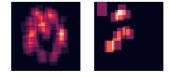
Class: sees-dark-caramel-patties



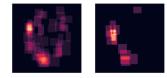
Class: sees-dark-walnut-square



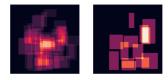
Class: sees-milk-bordeaux

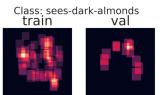


Class: sees-milk-coconut-cream

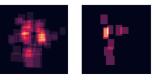


Class: sees-milk-molasses-chips

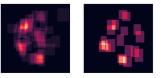




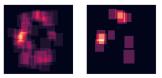
Class: sees-dark-marzipan



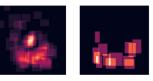
Class: sees-milk-almonds



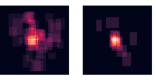
Class: sees-milk-california-brittle



Class: sees-milk-mocha



Class: sees-milk-rum-nougat

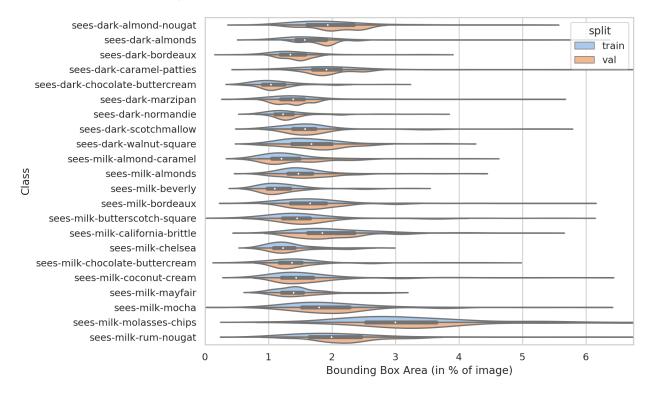


The heatmap represents areas of high object density within the images, providing insights into the spatial distribution of objects. By examining the heatmap, you can quickly detect whether objects are predominantly concentrated in specific regions or if they are evenly distributed throughout the scene.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



Notice: Only the 12 classes with highest density are shown. You can increase the number of classes by changing `n_cols` and `n_rows` in the configuration file.

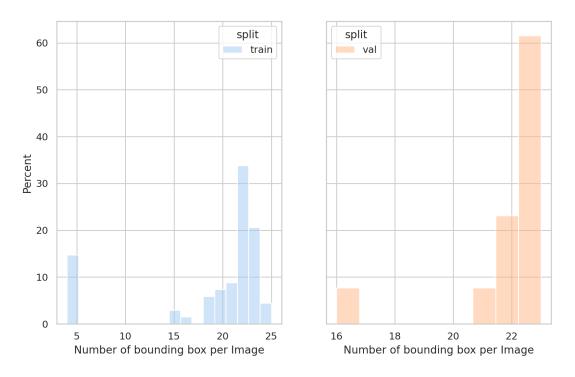
1.8. Distribution of Bounding Box Area



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.

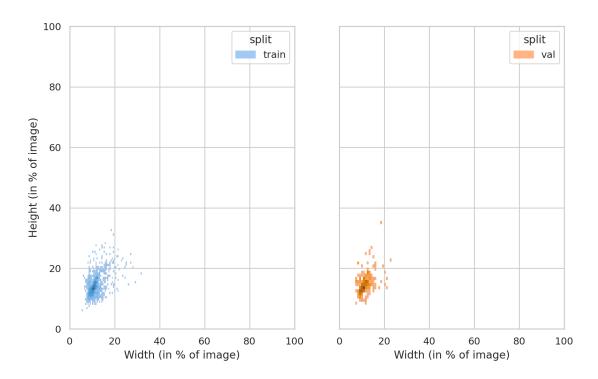
1.9. Distribution of Bounding Box per image



These graphs shows how many bounding boxes appear in images.

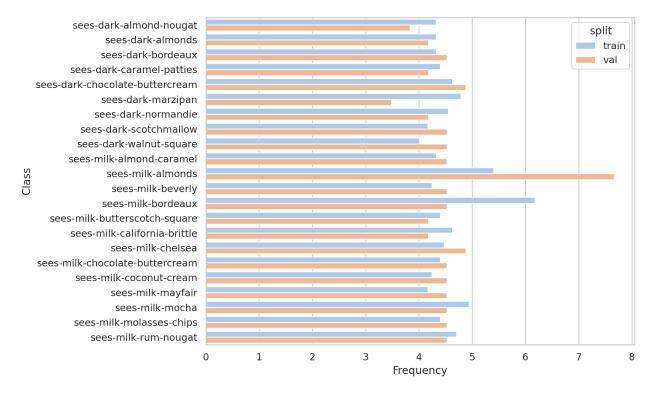
This can typically be valuable to know when you observe a very high number of bounding boxes per image, as some models include a parameter to filter the top k results.

1.10. Distribution of Bounding Box Width and Height



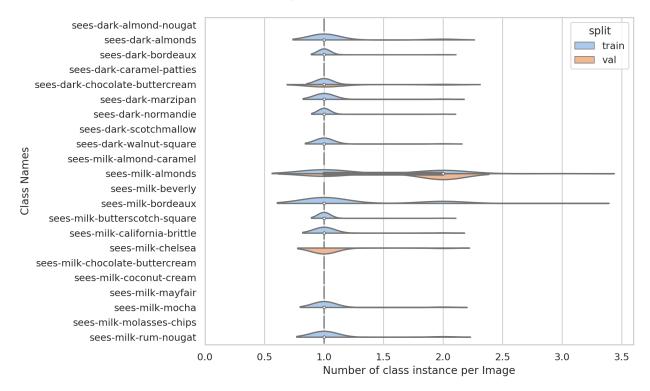
These heat maps illustrate the distribution of bounding box width and height per class. Large variations in object size can affect the model's ability to accurately recognize objects.

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.

1.12. Distribution of Class Frequency per Image



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		
sees-dark-almond-nougat	0300000000	sees-dark-almond-nougat	0300000000	
sees-dark-almonds	0600000000	sees-dark-almonds	0 8 0 0 0 0 0 0 0 0	
sees-dark-bordeaux		sees-dark-bordeaux	0300000000	
sees-dark-caramel-patties	0200000000	sees-dark-caramel-patties	00000000000	
sees-dark-chocolate-buttercream	000000000000	sees-dark-chocolate-buttercream	00000000000	
sees-dark-marzipan	0300000000	sees-dark-marzipan	00000000000	- 100
sees-dark-normandie	02000000000	sees-dark-normandie		100
sees-dark-scotchmallow	0100000000	sees-dark-scotchmallow	0000000000	- 80
sees-dark-walnut-square	011000000000	sees-dark-walnut-square	01500000000	- 80
sees-milk-almond-caramel		sees-milk-almond-caramel	00000000000	
رم sees-milk-almonds		sees-milk-almonds	00000000000	- 60
ം sees-milk-almonds ഇ sees-milk-beverly O sees-milk-bordeaux				
		sees-milk-bordeaux	00000000000	- 40
sees-milk-butterscotch-square	0100000000	sees-milk-butterscotch-square	00000000000	
sees-milk-california-brittle	0 6 0 0 0 0 0 0 0 0	sees-milk-california-brittle	09000000000	- 20
sees-milk-chelsea		sees-milk-chelsea	00000000000	20
sees-milk-chocolate-buttercream		sees-milk-chocolate-buttercream		
sees-milk-coconut-cream	000000000000	sees-milk-coconut-cream	00000000000	- 0
sees-milk-mayfair		sees-milk-mayfair	0300000000	
sees-milk-mocha	0 5 0 0 0 0 0 0 0 0	sees-milk-mocha	007000000000	
sees-milk-molasses-chips	014000000000	sees-milk-molasses-chips	011000000000	
	0800000000		0600000000	
All classes	0400000000	All classes	03000000000	
	10 20 50 80 80 00		10 20 20 20 20 20 20 20 20 20 20 20 20 20	
	0.200.000000000000000000000000000000000		0.200.500.3	
	loU range		loU range	

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.